

WAVELET-BASED OUTLIER DETECTION AND DENOISING
OF AIRBORNE LASER SCANNING DATA

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OF AIRBORNE LASER SCANNER DATA**

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ABSTRACT

WAVELET-BASED OUTLIER DETECTION AND DENOISING OF AIRBORNE LASER SCANNING DATA

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The method of airborne laser scanning – also named as LIDAR – has recently turned out to be an efficient way for generating high quality digital surface and elevation models. In this work, wavelet-based outlier detection and different wavelet thresholding (wavelet shrinkage) methods for denoising of airborne laser scanning data are discussed. The task is to investigate the effect of wavelet-based outlier detection and find out which wavelet thresholding methods provide best denoising results for post-processing. Data and results are analyzed and visualized by using a MATLAB program which was developed during this work.

Keywords: Outlier detection, wavelet thresholding, image denoising, LIDAR, DEM

ÖZ

HAVADAN LAZER TARAMA VERİSİNDE DALGACIK TABANLI AYKIRI DEĞER BELİRLEME VE GÜRÜLTÜ TEMİZLEME

AKYAY, Tolga

Yüksek Lisans, Jeodezi ve Coğrafi Bilgi Teknolojileri Bölümü

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Havadan lazer tarama veya LIDAR olarak da adlandırılan yöntem son yıllarda yüksek kalitede sayısal yüzey ve yükseklik modeli oluşturmada hızlı ve verimli olmuştur. Bu çalışmada, havadan lazer tarama verisinde dalgacık tabanlı aykırı değerlerin belirlenmesi ve değişik dalgacık sıkıştırma yöntemleri kullanılarak havadan lazer tarama verisinde gürültü temizlenmesi tartışılmıştır. Amaç, ileri aşamalardaki çalışmalarda fayda sağlaması için dalgacık tabanlı aykırı değer belirlemenin etkisini araştırmak ve hangi dalgacık sıkıştırma yönteminin gürültü temizlenmesinde daha iyi sonuçlar verdiğini bulmaktır. Veri ve sonuçlar bu çalışmada geliştirilen bir MATLAB uygulaması tarafından analiz edilmiş ve görüntülenmiştir.

Anahtar kelimeler: Aykırı değer belirleme, dalgacık sıkıştırma, görüntülerde gürültü temizleme, LIDAR, DEM

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CHAPTER 1

INTRODUCTION

Airborne laser scanning – also named as LIDAR – technology offers one of the most accurate, expedient and cost-effective ways to capture wide-area elevation information to produce highly detailed digital surface models (DSM) and digital elevation models (DEM). LIDAR data sets contain vast amount of information. Buildings, trees and power lines are individually discernible features. This data are digital and are directly processed to produce detailed bare earth DEMs at vertical accuracies of 0.15 meters to 1 meter. Derived products include contour maps, slope/aspect, 3-D topographic images, virtual reality visualizations and more. LIDAR data can be seamlessly integrated with other data sets, including orthophotos, multispectral, hyperspectral and panchromatic imagery. LIDAR is combined with GIS data and other surveying information to generate complex geomorphic-structure mapping products, building renderings, advanced 3-D modeling/earthworks and many more high quality mapping products.

Despite LIDAR signal contains detailed information, like any other signal, it may also contain outliers and random noise. Outliers should be removed and noise should be filtered out in order to obtain good quality final products. Bilen and Huzubazar proposes a general outlier detection method based on

wavelets [22]. Many denoising algorithms have been developed in the spatial and the Fourier domain. In the past two decades, the focus was shifted to wavelet domain due to its properties such as sparsity and multi-resolution structure. Various wavelet thresholding (wavelet shrinkage) methods have been published by researchers. Donoho and Johnstone who introduced the hard and soft thresholding functions proposed a universal method and an adaptive method based on SURE principle [9]. Kaur, Gupta and Chauhan proposed an adaptive method based on generalized Gaussian distribution [17]. The famous non-linear Bayesian estimator was applied to wavelet domain by Abramovich, Sapatinas and Silverman [19]. Ogden and Parzen established thresholding techniques based on statistical tests of hypotheses [20]. Nason suggests a cross-validation algorithm for wavelet shrinkage [21]. Tuncer investigated the performance of various wavelet thresholding methods on digital and SAR images and compared them with non-wavelet processes [28]. These wavelet thresholding methods are tested on LIDAR data for DSM generation.

Filtering of the LIDAR data is challenging due to the complex distribution of objects on Earth's surface. This problem has been investigated so far with a variety of algorithms. A filtering based on mathematical morphology called Progressive Morphological Filtering has been developed to enable extraction of ground points from LIDAR measurements with minimum human interaction [26]. Vu and Tokunaga proposed a new method based on multi-resolution analysis enabling the classification of objects based on their size [27].

In this work, the algorithm based on wavelets proposed by Bilen and Huzurbazar is used for outlier detection and removal. For denoising, one universal and five adaptive wavelet thresholding methods are implemented.

Universal method is VisuShrink and adaptive methods include NormalShrink, SureShrink, BayesShrink, Hypthesis Testing and Generalized Cross-validation. Progressive morphological filtering is used for DEM generation. The effect of outlier detection and denoising prior to progressive morphological filtering is also investigated. The program used for analyzing and visualizing the data and results is developed using MATLAB.

This thesis is organized as follows. Chapter 2 and 3 gives brief information about LIDAR and DEMs respectively. In Chapter 4, the concept of image denoising using wavelets is discussed. Chapter 5 covers progressive morphological filtering. Chapter 6 includes information about the test data used in this work. In Chapter 7, the methodology is presented and each step of the algorithm is explained in details. In Chapter 8, the results of wavelet-based outlier detection and denoising is given and discussed. Finally, Chapter 9 covers the conclusions achieved in the scope of this work.

CHAPTER 2

LIDAR

2.1 General Description

LIDAR (**L**ight **D**etection **A**nd **R**anging) is an optical remote sensing technology that measures properties of scattered light to find range and/or other information of a distant target. The well-known method to determine distance to an object or surface is to use laser pulses. Similar to radar technology, the range to an object is determined by measuring the time delay between transmission of a pulse and detection of the reflected signal [1].

$$dist = \frac{t_{2way} \times c}{2} \quad (2.1)$$

where $dist$ is the distance measured, t_{2way} is the time elapsed for laser to travel from the sensor to the object and back to the sensor, and c is the speed of light.

The primary difference between LIDAR and radar is that LIDAR uses much shorter wavelengths of the electromagnetic spectrum, typically in the ultraviolet, visible, or near infrared. A laser has a very narrow beam which

allows the mapping of physical features with very high resolution compared to radar. In addition, many chemical compounds interact more strongly at visible wavelengths than at microwaves, resulting in a stronger image of these materials [1].

In early stages, LIDAR has been used extensively for atmospheric research and meteorology. With the deployment of the **Global Positioning System (GPS)** in the 1980's, precision positioning of aircraft became possible. GPS based surveying technology has made airborne surveying and mapping applications possible and practical. Many have been developed, using downward-looking LIDAR instruments mounted in aircraft or satellites. By merging LIDAR, GPS and inertial attitude technologies, it has become possible to directly measure the shape (elevation) of the earth's surface beneath the aircraft's flight path. This elevation data is generated at the rate of thousands of points per second, with absolute vertical accuracies of up to 15 cm.

When LIDAR data is collected, the instrument also measures the intensity of the returned light which can be used for land-cover classification. The resulting image is a gray-scale measurement of the reflectance amplitude of the light pulse. Intensity data is not concerned in this work. The main focus will be on the elevation data.

2.2 System Components

In general, there are two types of LIDAR systems: **micropulse** LIDAR systems and **high energy** systems. Micropulse systems have developed as a

result of the ever increasing amount of computer power available combined with advances in laser technology. They use considerably less energy in the laser, typically on the order of one microjoule, and are often “eye-safe”, meaning they can be used without safety precautions. High-power systems are common in atmospheric research [1].

There are several major components in a LIDAR system:

- **Laser:** 600-1000 nm lasers are most common for non-scientific applications. They are inexpensive but, since they can be focused and easily absorbed by the eye, the maximum power is limited by the need to make them eye-safe. A common alternative 1550 nm lasers are eye-safe at much higher power levels since this wavelength is not focused by the eye, but the detector technology is less advanced and so these wavelengths are generally used at longer ranges and lower accuracies. They are also used for military applications as 1550 nm is not visible in night vision goggles unlike the shorter 1000 nm infrared laser. Airborne topographic mapping LIDAR systems generally use 1064 nm diode pumped YAG (Yttrium Aluminum Garnet; $\text{Y}_3\text{Al}_5\text{O}_{12}$) lasers, while bathymetric systems generally use 532 nm frequency doubled diode pumped YAG lasers because 532 nm penetrates water with much less attenuation than does 1064 nm. Laser settings include the laser repetition rate (which controls the data collection speed). Pulse length is generally an attribute of the laser cavity length, the number of passes required through the gain material (YAG, YLF, etc.), and Q-switch speed. Better target resolution is achieved with shorter pulses, provided the LIDAR receiver detectors and electronics have sufficient bandwidth [1].

- **Scanner and optics:** How fast images can be developed is also affected by the speed at which it can be scanned into the system. There are several options to scan the azimuth and elevation, including dual oscillating plane mirrors, a combination with a polygon mirror, a dual axis scanner. Optic choices affect the angular resolution and range that can be detected. A hole mirror or a beam splitter are options to collect a return signal.
- **Receiver and receiver electronics:** Receivers are made out of several different materials. Two common ones are Silicon (Si) and Indium Gallium Arsenide (InGaA). They are made in either PIN diode or Avalanche photodiode configurations. The sensitivity of the receiver is another parameter that has to be balanced in a LIDAR design.
- **Position and navigation systems:** LIDAR sensors that are mounted on mobile platforms such as airplanes or satellites require instrumentation to determine the absolute position and orientation of the sensor. Such devices generally include a GPS receiver and an Inertial Measurement Unit (IMU).

The following figure demonstrates typical Airborne LIDAR System (ALS) components:

- Laser Transmitter, Scanner and Receiver
- Aircraft Positioning – Differential GPS (with post-processing)
- Aircraft Attitude – Pitch, Roll, Yaw – Inertial Measurement Unit (GPS-Aided)

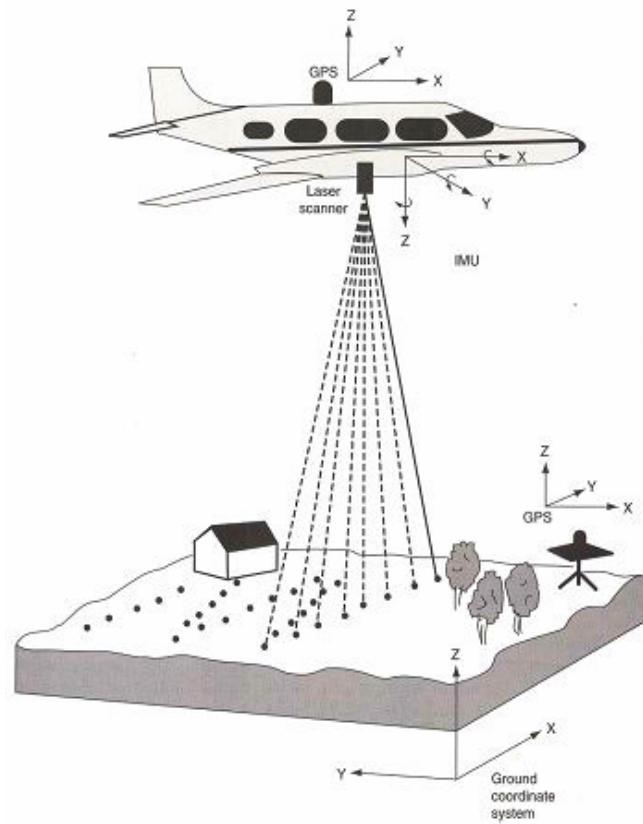


Figure 2.1: Airborne LIDAR System components [2]

2.3 Products

Primary products derived from the raw LIDAR data include:

- Digital Elevation Models (DEM)
- Triangular Irregular Networks (TIN)
- 3-D Models
- Intensity Images

CHAPTER 3

DIGITAL ELEVATION MODEL

3.1 General Description

A **digital elevation model (DEM)** is a digital representation of ground surface topography or terrain. A DEM can be represented as a raster (a grid of squares, or an image) or as a **triangular irregular network (TIN)**. A TIN is a vector based representation of the physical land surface or sea bottom, made up of irregularly distributed nodes and lines with three dimensional coordinates (x, y, z) that are arranged in a network of nonoverlapping triangles.

DEMs are commonly built using remote sensing techniques, however, they may also be built from land surveying. DEMs are used often in geographic information systems, and are the most common basis for digitally-produced relief maps.

A DEM generally refers to a representation of the earth's surface (or subset of this), excluding features such as vegetation, buildings, bridges, etc. The DEM often comprises much of the raw dataset, which may have been acquired

through techniques such as photogrammetry, LIDAR, Interferometric Synthetic Aperture Radar (IFSAR), land surveying, etc. A **digital surface model (DSM)** on the other hand includes buildings, vegetation, and roads, as well as natural terrain features. The DEM provides a so-called bare-earth model, devoid of landscape features. While a DSM may be useful for landscape modeling, city modeling and visualization applications, a DEM is often required for flood or drainage modeling, land-use studies, geological applications, and much more.

3.2 Production

DEMs may be prepared in a number of ways, but they are frequently obtained by remote sensing rather than direct survey.

One powerful technique for generating DEMs is using airborne active sensors. These technologies include LIDAR and IFSAR – also named as INSAR.

LIDAR uses a rapidly pulsed laser rangefinder mounted in a light aircraft with accompanying differential GPS and INS (inertial navigation system) to locate and orient the aircraft. Ten thousand to 30.000 points (x, y, z) are surveyed per second with a spatial precision of a few decimeters, at a cost of a few hundredths of a cent per point. Flying height is on the order of 1.000 meters and airspeeds are on the order of 100 knots. The surveyed points include ground, vegetation, and structures, and require extensive filtering to extract usable terrain models. In heavily vegetated areas LIDAR systems can be challenged to locate the ground surface. Most LIDAR systems use a near-infrared laser that does not penetrate fog or rain [3].

IFSAR bypasses some of these problems. Flying higher and faster than most LIDAR systems, an insar-equipped plane can cover large areas. Longer (cm to m) wavelengths penetrate fog and rain. Extensive development work has resulted in systems that process data almost as rapidly as it is acquired. Unfortunately, there are wave-length related tradeoffs: longer, ~0.5 m (P-band) wavelengths with the potential to penetrate vegetation to the ground of necessity lack high spatial resolution; wavelengths of a few cm (X-band) provide better spatial resolution but image the tree tops [4].

Another powerful technique for generating a DEM is using the digital image correlation (stereo photogrammetry) method. It implies two optical images acquired with different angles taken from an airplane or satellite.

Older methods of generating DEMs often involve interpolating digital contour maps that may have been produced by direct survey of the land surface. This method is still used in mountain areas where airborne scanning is not always satisfactory. The contour data or any other sampled elevation datasets (by GPS or ground survey) are not DEMs, but may be considered DTMs. A DEM implies that elevation is available continuously at each location in the study area.

The quality of a DEM is a measure of how accurate elevation is at each pixel (absolute accuracy) and how accurately is the morphology presented (relative accuracy). Several factors play an important role for quality of DEM-derived products:

- terrain roughness
- sampling density (elevation data collection method)

- grid resolution or pixel size
- interpolation algorithm
- vertical resolution
- terrain analysis algorithm

3.3 Uses

Common uses of DEMs include:

- extracting terrain parameters
- modeling water flow or mass movement (for example avalanches)
- creation of relief maps
- rendering of 3D visualizations
- creation of physical models (including raised-relief maps)
- rectification of aerial photography or satellite imagery
- reduction (terrain correction) of gravity measurements (gravimetry, physical geodesy)
- terrain analyses in geomorphology and physical geography

CHAPTER 4

IMAGE DENOISING USING WAVELETS

4.1 Introduction

Digital images play an important role both in daily applications such as satellite television, magnetic resonance imaging, computer tomography, as well as in research and technology such as geographical information systems and astronomy. Generally, datasets collected by image sensors are contaminated by noise. Imperfect instruments, problems with the data acquisition, and interfering natural phenomena can all degrade the data of interest. Furthermore, noise can be introduced by transmission errors and compression.

There are many different cases of distortions. One of the most prevalent cases is distortion due to additive white Gaussian noise which can be caused by poor image acquisition or by transferring the image data in noisy communication channels. Gaussian noise removal can be effectively done by linear filtering methods. Other types of noises include impulse and speckle noises. Impulse noise is caused by malfunctioning pixels in camera sensors, faulty memory locations in hardware, or transmission in a noisy channel. Two

common types of impulse noise are the salt-and-pepper noise and the random-valued noise. For images corrupted by salt-and-pepper noise, the noisy pixels can take only the maximum and the minimum values while in the case of random-valued noise, they can take any random value in the dynamic range.

Speckle is a random, deterministic, interference pattern in an image formed with coherent radiation of a medium containing many sub-resolution scatterers. The texture of the observed speckle pattern does not correspond to underlying structure. The local brightness of the speckle pattern, however, does reflect the local echogenicity of the underlying scatterers.

Therefore, denoising is often a necessary and the first step to be taken before the image data is analyzed. It is necessary to apply an efficient denoising technique to compensate for such data corruption. Image denoising still remains a challenge for researchers because noise removal introduces artifacts and causes blurring of the images. Denoising of electronically distorted images is an old but also still a relevant industrial problem. In the past two decades, many methods for denoising have been developed and reported in the literature. There are many works on the restoration of images corrupted by impulse noise. The goal of impulse noise removal is to suppress the noise while preserving the integrity of edge and detail information.

There are two basic approaches to image denoising, spatial filtering methods and transform domain filtering methods [5]. A traditional way to remove noise from image data is to employ spatial filters. Spatial filters employ a low pass filtering on groups of pixels with the assumption that the noise occupies

the higher region of frequency spectrum. Low-pass filters will not only smooth away noise but also blur edges in signals and images while the high-pass filters can make edges even sharper and improve the spatial resolution but will also amplify the noisy background [6].

Spatial filters can be classified into non-linear and linear filters. With **non-linear filters**, the noise is removed without any attempts to explicitly identify it. Many non-linear filters fall into the category of order statistic neighbour operators. This means that the local neighbours are sorted into ascending order (according to their value) and this list is processed to give an estimate of the underlying image brightness. The simplest order statistic operator is the median [7], where the central value in the ordered list is used for the new value of the brightness. The median is good at reducing impulse noise (for example salt-and-pepper noise) where the noisy pixels contain no information about their original values. However, when the noise level is over 50%, some details and edges of the original image are smeared by the filter [8]. A mean or average filter is the optimal **linear filter** for Gaussian noise removal in the sense of mean square error. Linear filters too tend to blur sharp edges, destroy lines and other fine image details, and perform poorly in the presence of signal-dependent noise.

Wavelet Analysis, a new form of signal analysis is far more efficient than Fourier analysis wherever a signal is dominated by transient behavior or discontinuities. Engineers and Scientists are now using wavelets to compress digital signals and images, to speed up fundamental scientific problem and to get rid of noise from digital signals [8]. Several investigations have been made into additive noise suppression in signals and images using wavelet transforms. Much of the early work on wavelet noise removal based on

thresholding the **Discrete Wavelet Transform (DWT)** coefficients of an image and then reconstructing it [9, 10].

This section does not go through the detailed mathematics of the Wavelet Transform. If the reader wishes to get a glance of Wavelet Transform, a suggested reading is *A Really Friendly Guide to Wavelets* by Valens [11].

4.2 Wavelet Domain Filtering

Working in the wavelet domain is advantageous because the DWT tends to concentrate the energy of the desired signal in a small number of coefficients, hence, the DWT of the noisy image consists of a small number of coefficients with high Signal to Noise Ratio (SNR) and a large number of coefficients with low SNR. After discarding the coefficients with low SNR (i.e. noisy coefficients), the image is reconstructed using inverse DWT. As a result, noise is removed or filtered from the observations.

4.3 Discrete Wavelet Transform

Wavelets are functions generated from one single function ψ by dilations and translations [12], where dilation means scaling the wavelet and translation means shifting the wavelet. The basic idea of the wavelet transform is to represent any arbitrary function as a superposition of wavelets. Any such superposition decomposes the given function into different scale levels where each level is further decomposed with a resolution adapted to that level [12].

The DWT is identical to a hierarchical subband system where the subbands are logarithmically spaced in frequency and represent octave-band decomposition. By applying DWT, the image is actually divided i.e. decomposed into four subbands and critically sub sampled as shown in Figure 4.1(a). These four subbands arise from separable applications of vertical and horizontal analysis filters for wavelet decomposition as shown in Figure 4.2(a) [13].

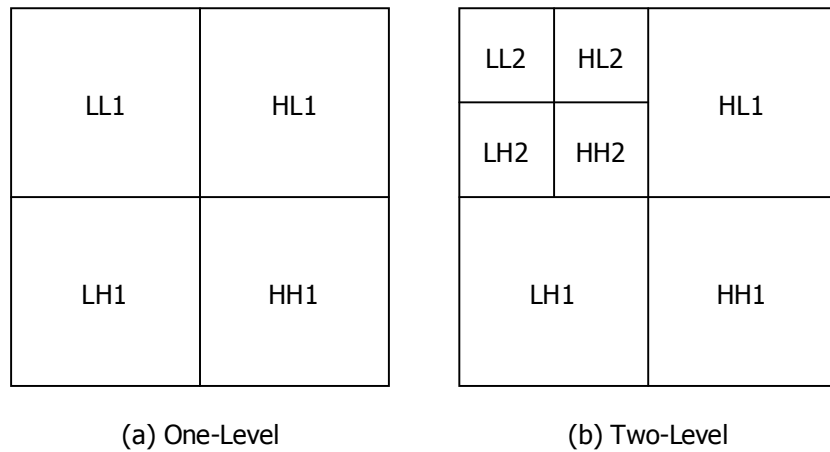


Figure 4.1: Image Decomposition

The filters H_0 and G_0 shown in Figure 4.2(a) are one-dimensional Low Pass Filter (LPF) and High Pass Filter (HPF), respectively. Thus, decomposition provides subbands corresponding to different resolution levels and orientation. These subbands labeled LH1, HL1 and HH1 represent the finest scale wavelet coefficients, i.e. detail images while the subband LL1 corresponds to coarse level coefficients, i.e. approximation image. To obtain the next coarse level of wavelet coefficients, the subband LL1 alone is further decomposed and critically sampled using similar filter bank shown in Figure

4.2(a). This results in two-level wavelet decomposition as shown in Figure 4.1(b).

Similarly, to obtain further decomposition, LL2 will be used. This process continues until some final scale is reached [11]. The decomposed image can be reconstructed using a reconstruction (i.e. Inverse DWT) or synthesis filter as shown in Figure 4.2(b). Here, the filters H_1 and G_1 represents low pass and high pass reconstruction filters respectively.

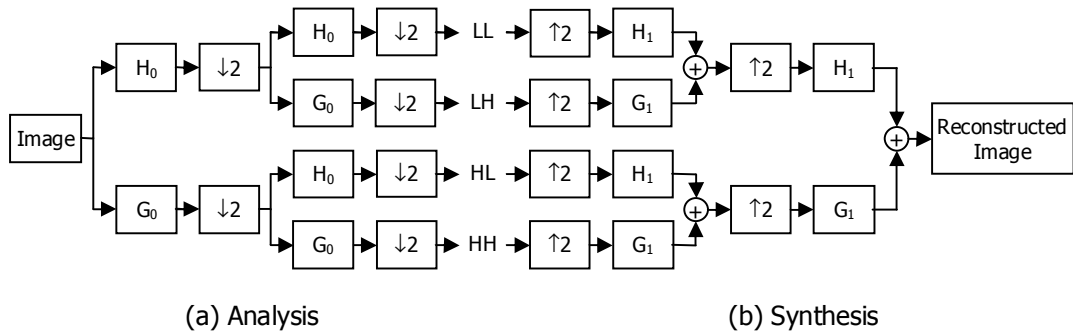


Figure 4.2: One level filter bank for computation of 2-D DWT and Inverse DWT

4.4 Wavelet-Based Denoising

Wavelets give a superior performance in image denoising due to properties such as sparsity and multi-resolution structure. With Wavelet Transform gaining popularity in the last two decades, the focus was shifted from the Spatial and Fourier domain to the Wavelet transform domain and various algorithms for denoising in wavelet domain were introduced.

The basic block diagram of wavelet based image denoising system is shown in Fig. 4.3. Wavelet Based Denoising method relies on the fact that noise commonly manifests itself as fine-grained structure in the image and DWT provides a scale based decomposition. Thus, most of the noise tends to be represented by wavelet coefficients at the finer scales. Discarding these coefficients would result in a natural filtering of the noise on the basis of scale. Because the coefficients at such scales also tend to be the primary carriers of edge information, this method sets the DWT coefficients to zero if their values are below a threshold. These coefficients are mostly those corresponding to noise. The edge relating coefficients on the other hand, are usually above the threshold. The Inverse DWT of the thresholded coefficients is the denoised image. It has been found that such denoising is effective in that although noise is suppressed, edge features are retained without much damage [14].

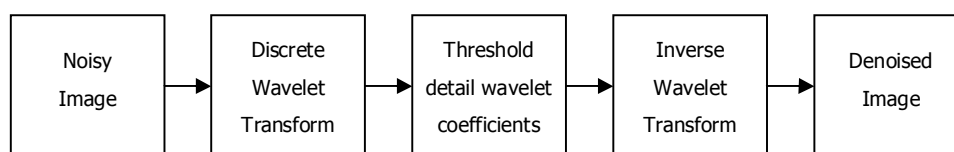


Figure 4.3: Block diagram of wavelet based image denoising system

Researchers published different ways to compute the parameters for the thresholding of wavelet coefficients. Data adaptive thresholds were introduced to achieve optimum value of the threshold. Later efforts found that substantial improvements in perceptual quality could be obtained by translation invariant methods based on thresholding of an Undecimated Wavelet Transform [15]. These thresholding techniques were applied to the

non-orthogonal wavelet coefficients to reduce artifacts. Multi-wavelets were also used to achieve similar results.

Wavelet thresholding [9, 10, 16], is a signal estimation technique that exploits the capabilities of wavelet transform for signal denoising. As one may observe, threshold selection is an important question when denoising. A small threshold may yield a result close to the input, but the result may still be noisy. A large threshold on the other hand, produces a signal with a large number of zero coefficients. This leads to a smooth signal. Paying too much attention to smoothness, however, destroys details and in image processing may cause blur and artifacts.

The detail wavelet coefficients (horizontal, vertical and diagonal) are thresholded by the two common thresholding functions:

- **Hard thresholding**

$$f_h(x) = \begin{cases} x & \text{if } |x| \geq \lambda \\ 0 & \text{otherwise} \end{cases} \quad (4.1)$$

- **Soft thresholding**

$$f_s(x) = \begin{cases} x - \lambda & \text{if } x \geq \lambda \\ 0 & \text{if } |x| < \lambda \\ x + \lambda & \text{if } x \leq -\lambda \end{cases} \quad (4.2)$$

where x is the detail wavelet coefficient and λ is threshold.

In the scope of this work, six threshold estimation methods are implemented. These are explained in detail in the following sub-sections.

4.4.1 VisuShrink

VisuShrink is a non-data dependent method done by applying the Universal Threshold (UT) proposed by Donoho and Johnstone [9]. This threshold is given by

$$T_U = \hat{\sigma} \sqrt{2 \log M} \quad (4.3)$$

where $\hat{\sigma}$ is standard deviation and M is the number of pixels in the image.

In practice, we do not know the exact standard deviation. Noise variance is estimated by the robust median estimator:

$$\hat{\sigma} = \frac{\text{median}(|Y_{ij}|)}{0.6745}, \quad Y_{ij} \in \text{subband HH}_1 \quad (4.4)$$

It is proved that the maximum of any M values are independent and identically distributed as $N(0, \hat{\sigma}^2)$ will be smaller than the UT with high probability. The probability approaches 1 as M increases. Thus, with high probability, a pure noise signal is estimated as being identically zero. However, for denoising images, VisuShrink is found to yield an overly smoothed estimate. This is because the UT is derived under the constraint that with high probability, the estimate should be at least as smooth as the signal. So the UT tends to be high for large values of M , killing many signal coefficients along with the noise. Thus, the threshold does not adapt well to discontinuities in the signal.

4.4.2 NormalShrink

NormalShrink is an adaptive threshold estimation method based on the generalized Gaussian distribution (GGD) modeling of subband coefficients. It is computationally more efficient and adaptive because the parameters required for estimating the threshold value T_N depend on subband data [17].

$$T_N = \frac{\beta \hat{\sigma}^2}{\hat{\sigma}_y} \quad (4.5)$$

where $\hat{\sigma}^2$ is noise variance calculated by (4.4) from subband HH1, $\hat{\sigma}_y$ is the standard deviation of the subband and β is the scale parameter computed for each scale:

$$\beta = \sqrt{\log\left(\frac{L_k}{J}\right)} \quad (4.6)$$

where, L_k is the length of the subband at k^{th} scale and J is the total number of decompositions.

4.4.3 SureShrink

SureShrink is a thresholding by applying subband adaptive threshold. A separate threshold is computed for each subband based on SURE (Stein's unbiased risk estimator) which is a method for estimating the loss $\|\hat{\mu} - \mu\|^2$ in an unbiased fashion. Let wavelet coefficients in the j^{th} subband be

$\{X_i : i = 1, \dots, d\}$, $\hat{\mu}$ is the soft threshold estimator $\hat{X}_i = \eta_i(X_i)$, we apply Stein's result [18] to get an unbiased estimate of the risk $E\|\hat{\mu}^{(t)}(x) - \mu\|^2$:

$$\text{SURE}(t; X) = d - 2 \cdot \#\{i : |X_i| \leq t\} + \sum_{i=1}^d \min(|X_i|, t)^2 \quad (4.7)$$

where $\#$ denotes the cardinality of the set.

For an observed vector x (set of noisy wavelet coefficients in a subband), we can find the threshold T_s that minimizes $\text{SURE}(t; X)$:

$$T_s = \operatorname{argmin}_{c \leq t \leq \sqrt{2 \log d}} \text{SURE}(t; X) \quad (4.8)$$

Such a method can reasonably be expected to do well in terms of minimizing the risk, since for large sample sizes the Law of Large Numbers will guarantee that the SURE criterion is close to the true risk.

4.4.4 BayesShrink

The BayesShrink rule uses a Bayesian mathematical framework for images to derive subband dependent thresholds that are nearly optimal for soft thresholding [19].

Based on the observation that the wavelet coefficients in a subband of a natural image can be summarized adequately by a generalized Gaussian distribution (GGD). This observation is well-accepted in the image processing community.

$$t_B = \frac{\hat{\sigma}^2}{\hat{\sigma}_X} \quad (4.9)$$

$$\hat{\sigma}_X = \sqrt{\max(\hat{\sigma}_Y^2) - \hat{\sigma}^2, 0} \quad (4.10)$$

$$\hat{\sigma}_Y^2 = \frac{1}{N_s} \sum_{n=1}^{N_s} w_n^2 \quad (4.11)$$

where $\hat{\sigma}^2$ is the noise variance calculated by (4.4).

4.4.5 Hypothesis Testing

The aim of this approach is to use the data values themselves to separate wavelet coefficients into large coefficients (significant signal) and small coefficients (random noise). This method also operates on a level-by-level basis. At any given level, let $X_1, \dots, X_d$ be independent with $X_i \sim N(\mu_i, 1)$, $i = 1, \dots, d$, where d is 2^j for level j . Let I_d represent a non-empty subset of the indices $\{1, \dots, d\}$ with I_d' denoting its complement. Then we may consider testing the following hypotheses:

$$\begin{aligned} H_0 : \mu_1 = \dots = \mu_d = 0 \\ H_a : \mu_i = 0 \text{ for } i \in I_d'; \mu_i \neq 0 \text{ for } i \in I_d \end{aligned} \quad (4.12)$$

where H_0 is the null hypothesis and H_a is the alternative hypothesis.

Using the hypotheses in (4.12), we can establish simple thresholding techniques based on statistical tests of hypotheses [20].

If the cardinality of I_d is unknown, the likelihood ratio test reduces simply to:

$$T_{d,*}(X) = \sum_{i=1}^d X_i^2 \quad (4.13)$$

which is distributed as a Chi-Square (χ^2) random variable with d degrees of freedom under the null hypothesis.

If the cardinality of I_d in (4.12) is known, say m , the likelihood ratio test is:

$$T_{d,m}(X) = \sum_{i=d-m+1}^d (X^2)_{(i)} \quad (4.14)$$

where $(X^2)_{(i)}$ is the i^{th} largest value of $\{X_1^2, \dots, X_d^2\}$. According to the null hypothesis, the test statistic in (4.14) is a sum of order statistics from an iid sample of χ_1^2 random variables because of the normality of the X_i 's.

Testing (4.12) for only one element in I_d , the null distribution becomes:

$$P[T_{d,1} < x] = P\left[\max_{i=1,\dots,d} |X_i| < \sqrt{x}\right] = [\Phi(\sqrt{x}) - \Phi(-\sqrt{x})]^d \quad (4.15)$$

Straightforward calculation gives the $(1 - \alpha)$ critical point of the distribution in (4.15):

$$x_d^\alpha = \left\{ \Phi^{-1} \left[\frac{1}{2} \left((1 - \alpha)^{1/d} + 1 \right) \right] \right\}^2 \quad (4.16)$$

The special case of the hypotheses in (4.12) where I_d has only one element is expressed as:

$$\begin{aligned} H_0 : \mu_1 = \dots = \mu_d = 0 \\ H_a : \exists i \in \{1, \dots, d\} \text{ such that } \mu_i \neq 0 \end{aligned} \tag{4.17}$$

The proposed recursive algorithm is as follows:

1. Compute the test statistic $T_{d,1}$ and compare it with the critical value x_d^α .
2. If the test in step 1 is significant, remove the observation with the greatest value from the dataset, set d to $d-1$, and go to step 1.
3. If the test in step 1 is not significant, set the threshold value to the observation with the greatest absolute value in the current dataset.

The primary goal of this algorithm is to throw out large coefficients until the remaining set of coefficients is believed to be pure noise. Then, all the noise coefficients are shrunk to zero.

The level α of the test of hypotheses can be thought of as being a smoothing parameter. A smaller α value makes it more difficult to include coefficients in the reconstruction and results in a smoother estimator. Since α represents the usual significance level of ordinary test of hypotheses, standard values are $\alpha = .05$ and $\alpha = .01$.

4.4.6 Generalized Cross-validation

Cross-validation is widely used as an automatic procedure to choose the smoothing parameter in many statistical applications. The generalized cross-validation (GCV) method is performed by systematically expelling a data point from the construction of an estimate, predicting what the removed value would be and then comparing the prediction with the value of the expelled point [21].

The aim of function estimation in this method is minimizing the mean integrated square error (MISE) between wavelet shrinkage estimator $\hat{f}_t(x)$ and the true function $f(x)$. The threshold t should minimize:

$$M(t) = E \int \left\{ \hat{f}_t(x) - f(x) \right\}^2 dx \quad (4.18)$$

In practice, the function f is not known and so M has to be estimated.

Nason suggests a two-fold cross-validation algorithm that can be used to automatically select a threshold for a wavelet shrinkage estimator based on 2^M points. Let $g_1, \dots, g_n$ represent given data where $n = 2^M$, remove all the odd-indexed g_i from the set. This leaves 2^{M-1} evenly indexed g_j which are reindexed from $j = 1, \dots, 2^{M-1}$. A function estimate $\hat{f}_t^E$ is then constructed using a particular threshold t from the re-indexed g_j . In order to compare the function estimator with the left-out noisy data an interpolated version of $\hat{f}_t^E$ is formed:

$$\bar{f}_{t,j}^E = \frac{1}{2}(\hat{f}_{t,j+1}^E + \hat{f}_{t,j}^E), \quad j=1, \dots, n/2 \quad (4.19)$$

$\hat{f}_{t,n/2+1}^E = \hat{f}_{t,1}^E$ because f is assumed to be periodic. The $\hat{f}_t^O$ is also computed for the odd indexed points and the interpolant $\bar{f}_t^O$ is computed using (4.19). The full estimate for $M(t)$ compares the interpolated wavelet estimators and left-out points:

$$\hat{M}(t) = \sum_{j=1}^{n/2} \left\{ (\bar{f}_{t,j}^E - g_{2j+1})^2 + (\bar{f}_{t,j}^O - g_{2j})^2 \right\} \quad (4.20)$$

The estimate $\hat{M}$ relies on two estimates of f_t based upon $n/2$ data points. Donoho and Johnstone's [10] universal threshold (4.3) can be used to supply a heuristic method for obtaining a cross-validated threshold for n data points. If the threshold for n points is $T_{UV}(n)$ then the threshold for $n/2$ points will be $T_{UV}(n/2)$ and the relationship between them can be written as:

$$T_{UV}(n) \approx \left(1 - \frac{\log 2}{\log n} \right)^{-1/2} T_{UV}(n/2) \quad (4.21)$$

After the estimate $\hat{M}(t)$ has been minimized, the correction (4.21) is applied to obtain final cross-validated threshold.

This correction can be extended to a 2^k -fold cross-validation procedure, where 2^k estimates $\bar{f}$ are obtained, each based on $n/2^k$ data points selected in a regular way from the original data. Each estimated is

interpolated to the original grid and validated by comparing to the remaining $n(1 - 2^{-k})$ data points. The correction term (4.21) is easily extended to this case and will give the correction factor:

$$\left(1 - \frac{k \log 2}{\log n}\right)^{-1/2} \quad (4.22)$$

4.5 Wavelet-Based Outlier Detection

Outliers are commonly encountered in data and their presence can seriously distort model identification, parameter estimation and forecasting. Outliers are described as observations which are unusual, but not necessarily errors [22].

Wavelets are fairly new family of basis functions that are used to express and approximate other functions. They combine powerful properties such as orthonormality, different degrees of smoothness, localization in time and scale, compact support and fast implementation. Wavelet coefficients are capable of revealing aspects of the data that other techniques might miss, aspects such as changes in variance, level changes, and discontinuities in functions. Thus, wavelet analysis is well suited for outlier detection [22].

The wavelet-based procedure proposed by Bilen and Huzurbazar [22] is as follows:

1. Apply discrete wavelet transform using Haar wavelet to the observed series $\{Z_t\}$ to obtain the first level of wavelet coefficients $D(J-1)$.
2. Estimate $\hat{\sigma}_1$ from the data by taking the mean absolute deviations (AD) from the median of the wavelet coefficients:

$$\hat{\sigma}_1 = AD(D(J-1)) = \frac{1}{n_1} \sum_{k=1}^{n_1} |d_k(1) - M_1| \quad (4.23)$$

where $d_k(1)$ denotes the wavelet coefficient at location k , in the first level of the decomposition and M_1 is the median of the level 1 coefficients, $D(J-1)$.

3. Calculate the threshold limit τ_1 using the estimated $\hat{\sigma}_1$:

$$\tau_1 = \hat{\sigma}_1 \sqrt{2 \log(n)} \quad (4.24)$$

4. Find $S = \{s_1, \dots, s_m\}$, the set of indices such that $|d_k(1)| > \tau_1$.
5. Use the set of indices S to locate the exact position of outliers: Let (s_k) be the location of the wavelet coefficient such that $|d_k(1)| > \tau_1$. Then the outlier is either in location $(2s_k)$ or $(2s_k - 1)$. To find the exact location, compute the sample mean of the original series $\{Z_t; t = 1, \dots, n\}$ without observations at locations $(2s_k)$ and $(2s_k - 1)$:

$$\bar{Z}^* = \frac{1}{n-2} \sum_{t \neq 2s_k, 2s_k-1} Z_t \quad (4.25)$$

The location of the outlier is $(2s_k)$ if $|Z_{2s_k} - \bar{Z}^*| > |Z_{2s_k-1} - \bar{Z}^*|$; otherwise, the location of the outlier is $(2s_k - 1)$.

CHAPTER 5

PROGRESSIVE MORPHOLOGICAL FILTERING

5.1 Introduction

High-resolution DEMs are essential for many GIS related analysis and visualization. The airborne LIDAR technology is revolutionizing our way to acquire a high-resolution DEM by allowing rapid and inexpensive measurements of topography over a large area. Airborne LIDAR systems usually obtain measurements for the horizontal coordinates (x , y) and elevation (z) of the reflective objects scanned by the laser beneath the flight path. These measurements generate a 3-dimensional cloud of points with irregular spacing. The laser-scanned objects include buildings, vehicles, vegetation (canopy and understory), and "bare ground". To generate a DEM, measurements from ground and non-ground features have to be identified and classified. Removing non-ground points from LIDAR data sets has proven to be a challenging task.

5.2 Morphological Filters

Mathematical morphology composes operations based on set theory to extract features from an image. Two fundamental operations, **dilation** and **erosion**, are commonly employed to enlarge (dilate) or reduce (erode) the size of features in binary images. Dilation and erosion operations may be combined to produce **opening** and **closing** operations. The concept of erosion and dilation has been extended to multi-level (grayscale) images and corresponds to finding the minimum or maximum of the combinations of pixel values and the kernel function, respectively, within a specified neighborhood of each raster [23].

These concepts can also be extended to the analysis of a continuous surface such as a digital surface model as measured by LIDAR data. For a LIDAR measurement $p(x, y, z)$, the dilation of elevation z at x and y is defined as:

$$d_p = \max_{(x_p, y_p) \in w} (z_p) \quad (5.1)$$

where points (x_p, y_p, z_p) represent p 's neighbors (coordinates) within a window, w . The window can be a onedimensional line or two-dimensional rectangle or other shapes. The dilation output is the maximum elevation value in the neighborhood of p . Erosion is a counterpart of dilation and is defined as:

$$e_p = \min_{(x_p, y_p) \in w} (z_p) \quad (5.2)$$

The combination of erosion and dilation generates opening and closing operations that are employed to filter LIDAR data. The opening operation is achieved by performing an erosion of the data set followed by a dilation, while the closing operation is accomplished by carrying out a dilation first and then an erosion. An erosion operation removes objects of sizes smaller than the window size such as trees, while the dilation restores the shapes of large objects such as buildings. The ability of an opening operation to preserve features larger than the window size is very useful in some applications. For example, the measurements of cliff edges can be preserved if the morphological filters are applied to the LIDAR measurements for rocky coasts.

5.3 Progressive Morphological Filter

It has been shown that morphological filters can remove measurements for buildings and trees from LIDAR data, but it is difficult to detect all non-ground objects of various sizes using a fixed filtering window size. This problem can be solved by increasing window sizes of morphological filters gradually.

The detailed steps to use the progressive morphological filter to construct the DEMs are given as follows:

1. The irregularly spaced (x, y, z) LIDAR measurements are loaded. A regularly spaced minimum surface grid is constructed by selecting the minimum elevation in each grid cell. Point coordinates (x, y, z) are

stored in each grid cell. If a cell contains no measurements, it is assigned the value of nearest point measurement.

2. The progressive morphological filter whose major component is an opening operation is applied to the grid surface. At the first iteration, the minimum elevation surface together with an initial filtering window size provide the inputs for the filter. In the following iterations, the filtered surface obtained from the previous iteration and an increased window size from Step 3 are used as input for the filter. The output of this step include:
 - a. The further smoothed surface from the morphological filter
 - b. The detected non-ground points based on the elevation difference threshold
3. The size of the filter window is increased and the elevation difference threshold is calculated. Steps 2 to 3 are repeated until the size of the filter window is greater than a predefined maximum value. This value is usually set to be slightly larger than the maximum building size.
4. The last step is to generate the DEMs based on the data set after non-ground measurements have been removed. Note that each cell of the minimum surface grid generated in Step 1 contains an original or interpolated LIDAR point with elevation representing the cell value. Any filtering operation performed to the grid is actually applied to points in cells. Therefore, the progressive morphological filter classifies the LIDAR measurements at the point level.

5.4 Parameters for Progressive Morphological Filter

The selection of the window size and elevation difference threshold is critical to achieve good results when applying the morphological filter. For window size selection, one straight-forward choice is to increase the window size linearly by the following formula:

$$w_k = 2kb + 1 \quad (5.3)$$

where w_k is the full window size, $k = 1, 2, \dots, M$ and b is the initial window size. The maximum window size (number of cells) is equal to $2Mb + 1$. Taking $2Mb + 1$ as the window size guarantees that the filter window is symmetric around the central point so that the programming of the opening operation is simplified. The advantage of increasing the window size linearly is that gradually changing topographic features are well preserved. However, considerable computing time is needed for an area with large non-ground objects.

Alternatively, the window size can be increased exponentially to reduce the number of iterations used in the filter.

$$w_k = 2b^k + 1 \quad (5.4)$$

where b is the base of an exponential function, $k = 1, 2, \dots, M$, and again $2b^M + 1$ is equal to the maximum window size.

The elevation difference threshold can be determined based on the slope of topography in the study area. There is a relationship between the maximum elevation difference $dh_{\max(t),k}$ for the terrain, window size w_k , and the terrain slope s assuming that the slope is constant.

$$s = \frac{dh_{\max(t),k}}{(w_k - w_{k-1})/2} \quad (5.5)$$

Therefore, the elevation threshold $dh_{T,k}$ is given by

$$dh_{T,k} = \begin{cases} dh_0 & \text{if } w_k \leq 3 \\ s(w_k - w_{k-1})c + dh_0 & \text{if } w_k > 3 \\ dh_{\max} & \text{if } dh_{T,k} > dh_{\max} \end{cases} \quad (5.6)$$

where dh_0 is the initial elevation difference threshold, s is the slope, c is the cell size, and $dh_{\max}$ is the maximum elevation difference threshold.

In urban areas, primary non-ground objects include cars, trees, and buildings. The sizes of individual cars and trees are much less than that of the buildings, so most of them are often removed in the first several iterations, while the large buildings will be removed last. The maximum elevation difference threshold $dh_{\max}$ can be set to a fixed height (e.g., the lowest building height) to ensure that building complexes are identified. The optimum s is usually achieved by iteratively comparing the filtered and unfiltered data. On the other hand, the non-ground objects in the mountainous areas are primarily vegetation (trees). There is no need to set

up a fixed maximum elevation difference threshold to remove trees, and $dh_{\max}$ is usually set up as the largest elevation difference in the study area.

CHAPTER 6

TEST DATA

The test data is downloaded from ISPRS Test on Extracting DEMs from Point Clouds [24]. Eight sites (Vaihingen/Enz test field and Stuttgart city center) have been chosen for comparing the performance of filters. The areas are originally scanned with an Optech laser scanner. Four represent urban landscapes (City Sites) and the other four represent rural landscapes (Forest Sites). The point density for the City and Forest sites are roughly 0.67 and 0.18 points per square meter, respectively. Each Site is associated with one text file. Each text file represents first and last pulse returns.

Each text file is a space delimited point list. Every record represents two points (first and last return) and four data items are provided for each point in the record:

```
[X1 - First Return] - x coordinate (Easting)
[Y1 - First Return] - y coordinate (Northing)
[Z1 - First Return] - z coordinate (Height)
[I1 - First Return] - Intensity of returned pulse
[X2 - Last Return] - x coordinate (Easting)
[Y2 - Last Return] - y coordinate (Northing)
[Z2 - Last Return] - z coordinate (Height)
[I2 - Last Return] - Intensity of returned pulse
```

Sample:

```

...
...
...
495246.77 5420160.76 251.33 192 495246.77 5420160.78 251.25 192
495246.73 5420163.65 251.26 193 495246.73 5420163.66 251.20 193
495246.69 5420166.65 251.31 290 495246.69 5420166.65 251.31 290
495246.65 5420169.67 251.32 192 495246.65 5420169.68 251.27 192
495246.61 5420172.70 251.29 290 495246.61 5420172.71 251.23 290
495246.57 5420175.73 251.27 290 495246.57 5420175.73 251.23 290
495246.53 5420178.63 251.22 290 495246.53 5420178.64 251.15 290
495246.49 5420181.52 251.23 319 495246.49 5420181.53 251.18 319
495246.45 5420184.68 251.33 348 495246.45 5420184.68 251.34 348
495246.41 5420187.73 251.31 290 495246.41 5420187.73 251.29 290
...
...
...

```

Table 6.1: Summary of sites

Location	Site	Planimetric Resolution
City	Site 1	0.67 points per square meter (point spacing: 1.0 – 1.5 m)
	Site 2	
	Site 3	
	Site 4	
Forest	Site 5	0.18 points per square meter (point spacing: 2.0 – 3.5 m)
	Site 6	
	Site 7	
	Site 8	

Table 6.2: City sites

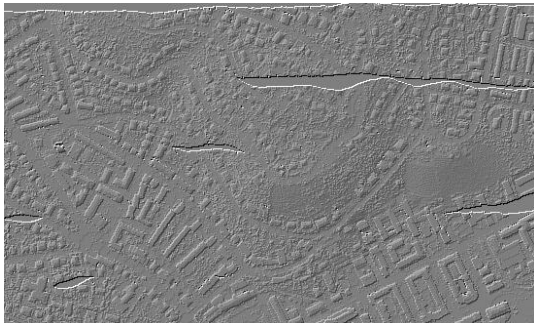
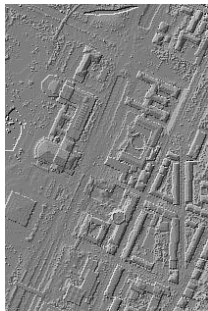
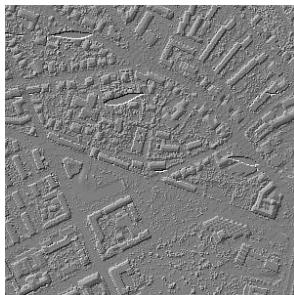
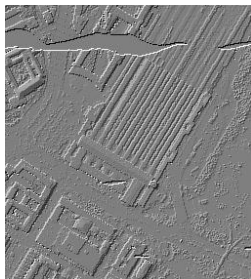
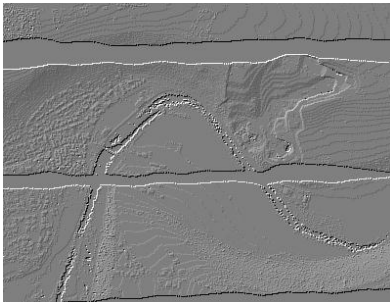
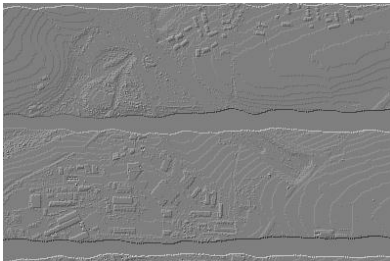
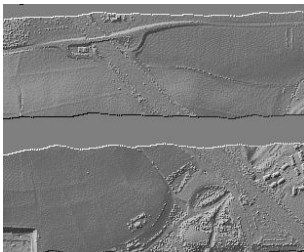
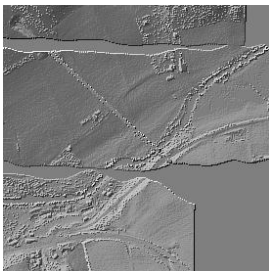
Site	Data Bounds		3-D View
Site 1	X min	512050	
	Y min	5403440	
	X max	513150	
	Y max	5404100	
	Features of interest: • Steep slopes • Mixture of vegetation and buildings on hillside • Buildings on hillside		
Site 2	X min	513450	
	Y min	5402650	
	X max	513870	
	Y max	5403280	
	Features of interest: • Large buildings • Irregularly shaped buildings • Road with bridge and small tunnel		
Site 3	X min	512023	
	Y min	5403120	
	X max	512549	
	Y max	5403500	
	Features of interest: • Densely packed buildings with vegetation between them • Building with eccentric roof (bottom left corner) • Open space with mixture of low and high features		
Site 4	X min	513120	
	Y min	5403190	
	X max	513630	
	Y max	5403760	
	Features of interest: • Railway station with trains (low density of terrain points)		

Table 6.3: Forest sites

Site	Data Bounds		3-D View
Site 5	X min	493650	
	Y min	5419770	
	X max	492250	
	Y max	5421000	
	Features of interest: - Steep slopes with vegetation - Quarry - Vegetation on river bank		
Site 6	X min	496350	
	Y min	5420750	
	X max	497950	
	Y max	5421817	
	Features of interest: - Large buildings - Road with embankment		
Site 7	X min	495950	
	Y min	5421350	
	X max	497200	
	Y max	5422380	
	Features of interest: - Bridge - Underpass - Road with embankments		
Site 8	X min	499450	
	Y min	5418330	
	X max	500550	
	Y max	5419430	
	Features of interest: - High bridge (diagonal left to right) - Breakline (Bottom center) - Vegetation on river bank		

CHAPTER 7

METHODOLOGY

7.1 Introduction

In this chapter, the wavelet-based outlier detection procedure, obtaining an elevation image from laser points through interpolation, wavelet-based denoising of this elevation image using six wavelet thresholding methods and finally obtaining a DSM or DEM is explained. The complete procedure of the algorithm is shown as a flowchart in Figure 7.1. The step-by-step details of the complete procedure are described in the sub-sections of this chapter. The results are presented in Appendix A and they are discussed in Chapter 8.

The program for analyzing and visualizing the data and results is developed using MATLAB. The key MATLAB functions used and implemented in the algorithm are listed in Appendix B.

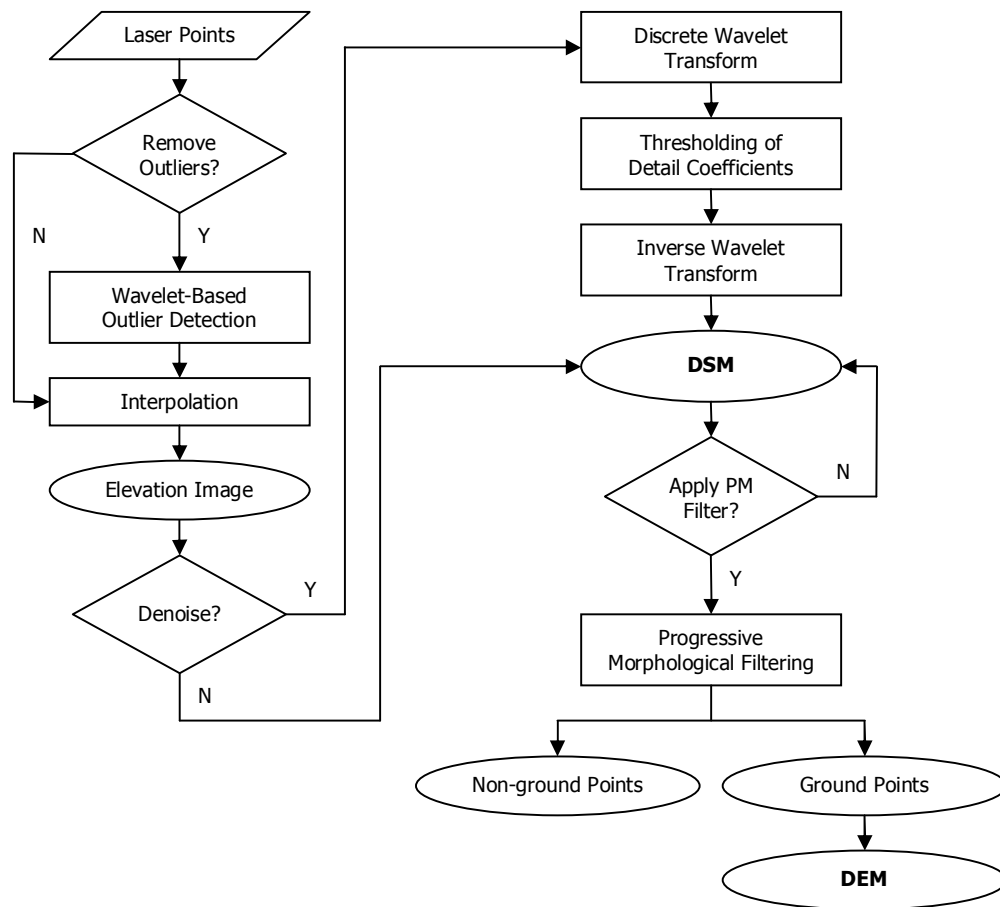


Figure 7.1: Flowchart of algorithm

7.2 Wavelet-Based Outlier Detection

The laser points may contain outliers such as spikes, artifacts. These observations should be removed to obtain a smoother elevation image. For this task, the wavelet-based procedure proposed by Bilen and Huzurbazar

[22] as described in Section 4.5 is used. In this procedure, the laser points are considered as time series.

The results for both DSM and DEM generation are tabulated in Tables A.1 and A.2.

7.3 Interpolation

In order to apply wavelet-based denoising and progressive morphological filtering on the test data, the nonuniformly spaced laser points must be converted to a uniformly spaced elevation image. There are numerous interpolation methods and determination of the best and suitable interpolation method is beyond the scope of this work. Linear interpolation on the TIN gives the most accurate interpolated image. First return points are used for DSM generation and last return points are used for DEM generation.

TIN is obtained from scattered laser points using **Delaunay triangulation**. In mathematics and computational geometry, a Delaunay triangulation for a set $\mathbf{P}$ of points in the plane is a triangulation $DT(\mathbf{P})$ such that no point in $\mathbf{P}$ is inside the circumcircle of any triangle in $DT(\mathbf{P})$ [25]. TIN is then gridded into a uniformly spaced elevation image through linear interpolation.

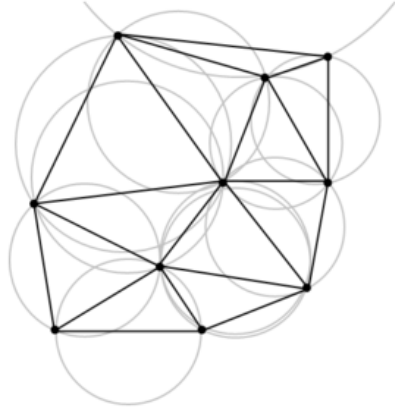


Figure 7.2: Delaunay triangulation in the plane with circumcircles shown

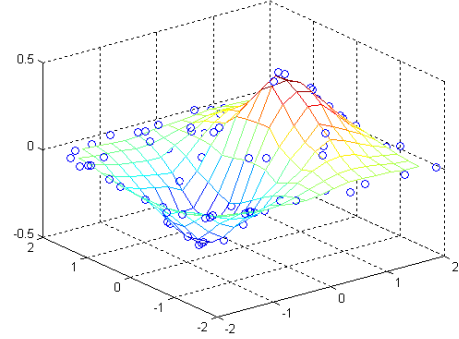


Figure 7.3: Sample gridded data

7.4 Wavelet-Based Denoising

After obtaining an elevation image free from outliers, wavelet-based denoising procedure as described in Section 4.4 can be applied. Horizontal, vertical and diagonal detail coefficients are obtained for each desired level using Discrete Wavelet Transform (DWT) with various wavelet models. The wavelet models used in this work are listed in Appendix C. Threshold values for each set of detail coefficients are calculated using six threshold estimation methods and each set is thresholded by both hard (4.1) and soft (4.2) thresholding functions. Finally, inverse wavelet transform is applied using the thresholded detail coefficients to obtain the denoised image.

The following six threshold estimation methods are implemented in this work:

- VisuShrink
- NormalShrink
- SureShrink
- BayesShrink
- Hypothesis Testing
- Generalized Cross-validation

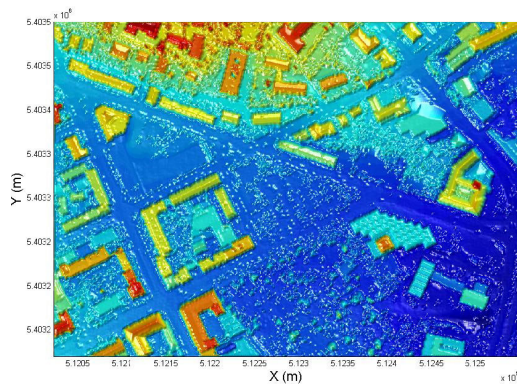
Among these six threshold estimation methods, VisuShrink is a universal method and the other remaining are adaptive methods. The results of each wavelet thresholding method can be found in Tables A.3 to A.50

7.5 Progressive Morphological Filtering

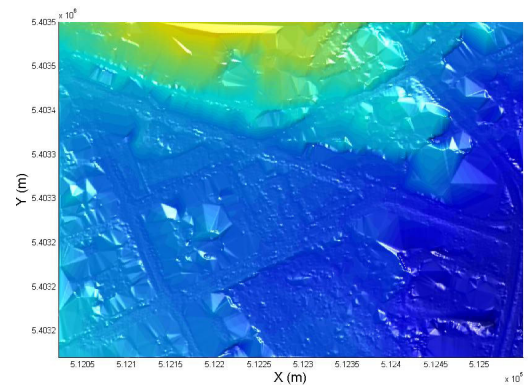
After obtaining a DSM, it may be further decomposed into ground and non-ground points for DEM generation. For this task, progressive morphological filtering explained in Section 5.3 is used. Filter parameters used in this work are listed in Table 7.1 [26]. The plots of City Site 3 with and without filtering are illustrated in Figure 7.4 to demonstrate the effect of progressive morphological filtering.

Table 7.1: Progressive morphological filter parameters

Location	City Sites	Forest Sites
Cell size (c in (5.6))	1	1
Base of exponential window (b in (5.4))	2	2
Increment step for windows (k in (5.4))	0,1,2,...,8	0,1,2,...,5
Slope (s in (5.6))	0.08	1.2
Initial threshold (dh_0 in (5.6))	0.25	0.2
Maximum threshold ($dh_{\max}$ in (5.6))	2.5	210



(a) without filtering



(b) with filtering

Figure 7.4: Plots of City Site 3 with and without progressive morphological filtering

CHAPTER 8

RESULTS AND DISCUSSION

8.1 Introduction

Test results are obtained according to the methodology described in the previous chapter. In this chapter, the results of wavelet-based outlier detection and denoising are discussed.

The symbols used in the test results are:

μ Mean

σ Standard Deviation

8.2 Wavelet-Based Outlier Detection

Outlier detection procedures are usually based on an exact or approximate likelihood ratio test which requires that an initial model fitted to the data. This is a difficult procedure and further complicated in the presence of outliers. The detection procedure based on wavelets proposed by Bilen and Huzurbazar [22] described in Section 4.5 does not require any model fitting.

It is simple, automatable, efficient and practical for large datasets. Here, unusual or inconsistent values (spikes, artifacts) are removed from the dataset in order to obtain a smoother image for better denoising results. The results of wavelet-based outlier detection are given in Table A.1 and A.2.

The benefit of the wavelet-based outlier detection on DSM generation can be observed in Table A.1. σ decreases while preserving μ . Figure 8.1 clearly visualizes the benefit on DSM generation.

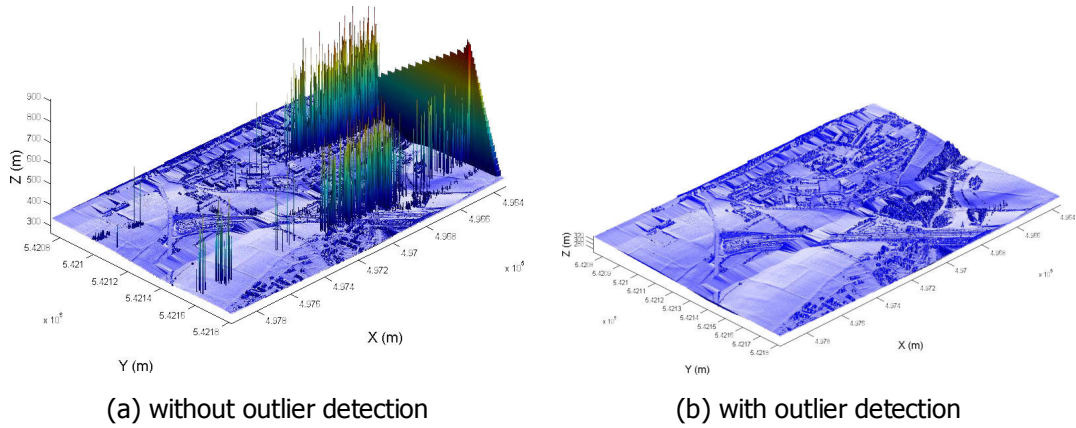


Figure 8.1: Effect of outlier detection on DSM generation in Forest Site 6

Analyzing Table A.2, the expected benefit is not observed on DEM generation. One reason may be the fact that progressive morphological filtering is already filtering out the outliers. However, in the presence of large artifacts, we observe the improvement as shown in Figure 8.2.

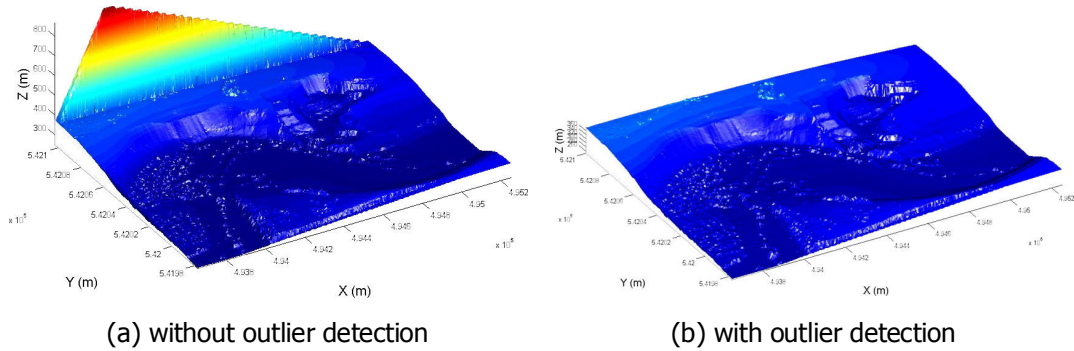


Figure 8.2: Effect of outlier detection on DEM generation in Forest Site 5

8.2 Wavelet-Based Denoising

Wavelet thresholding has been proven to be an effective way of image denoising by many researchers in the literature. It has been found out that adaptive thresholding performs better than universal thresholding. In order to prove these findings in this work, six wavelet thresholding methods, which are described in Section 4.4, were tested on the test data.

The performance of the wavelet thresholding methods are evaluated in Tables A.3 to A.50. μ and σ values are calculated for each method decomposing up to four levels. For better denoising results, one may expect to see decrease in σ while μ remaining constant. Tables 8.1 and 8.2 summarizes the results of wavelet-based denoising.

Table 8.1: Summary of wavelet-based denoising results for DSM generation

Method	Hard Thresholding				Soft Thresholding			
	including outliers		excluding outliers		including outliers		excluding outliers	
	σ	wavelet	σ	wavelet	σ	wavelet	σ	wavelet
VisuShrink on City Site 1, $\sigma = 31.1700$	31.1599	db20	31.0675	db20	31.0829	rbio3.9	30.9951	rbio3.9
NormalShrink on City Site 2, $\sigma = 11.7044$	11.7038	db20	11.5640	db20	11.6856	db20	11.5503	rbio3.9
SureShrink on City Site 3, $\sigma = 9.1404$	9.1323	bior2.2	9.0300	db20	9.0620	bior5.5	8.9610	haar bior1.1
BayesShrink on City Site 4, $\sigma = 10.5357$	10.5357	N/A	10.4378	N/A	10.5296	db20	10.4328	rbio3.9
Hypothesis Testing on City Site 2, $\sigma = 11.7044$	11.4874	db20	11.3555	bior5.5	10.8861	haar rbio1.1	10.7619	haar bior1.1
GCV on City Site 3, $\sigma = 9.1404$	8.3856	rbio1.3	8.3117	bior2.2	8.3645	rbio1.3	8.2803	rbio1.3

Table 8.2: Summary of wavelet-based denoising results for DEM generation

Method	Hard Thresholding				Soft Thresholding			
	including outliers		excluding outliers		including outliers		excluding outliers	
	σ	wavelet	σ	wavelet	σ	wavelet	σ	wavelet
VisuShrink on City Site 1, $\sigma = 25.6302$	22.9175	rbio3.1	22.4519	rbio3.1	23.1710	rbio3.1	23.5240	rbio3.1
NormalShrink on City Site 2, $\sigma = 3.5970$	3.5625	db20	3.6184	db20	3.5305	rbio3.1	3.5900	rbio3.9
SureShrink on City Site 3, $\sigma = 4.4727$	4.3547	db6	4.3456	bior3.1	4.3747	bior3.1	4.3781	bior3.1
BayesShrink on City Site 4, $\sigma = 3.5185$	3.5147	sym20	3.5596	db10	3.5098	rbio2.8	3.5540	rbio2.8
Hypothesis Testing on City Site 2, $\sigma = 3.5970$	1.9447	rbio3.5	1.9625	rbio3.5	2.1014	rbio3.3	2.0568	rbio3.3
GCV on City Site 3, $\sigma = 4.4727$	4.1013	bior1.5	4.1379	bior1.5	4.1071	bior1.5	4.1365	bior1.5

According to Table 8.1, soft thresholding always performs better than hard thresholding for DSM generation. σ always decreases as the level of decomposition increases, i.e. always smoothing the image. However, in some cases, hard thresholding performed better than soft thresholding for DEM

generation according to Table 8.2,. This conflict may be due to the mechanism of progressive morphological filtering.

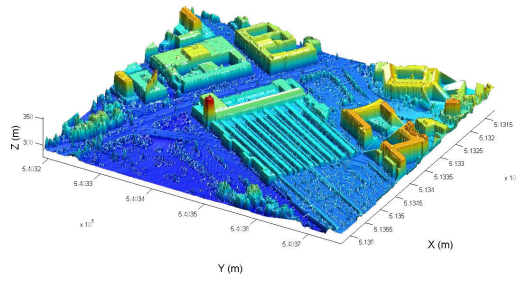
In order to find out the best wavelet thresholding method for DSM generation, each method is applied to City Site 4 with soft thresholding using Daubechies 20 wavelet decomposing up to four levels both including and excluding outliers. The statistical results are presented in Table 8.3 and 8.4 and for comparison, the 3-D DSM plots with two levels of decomposition are illustrated in Figure 8.3.

Table 8.3: Overall denoising performance analysis for DSM generation on City Site 4 including outliers

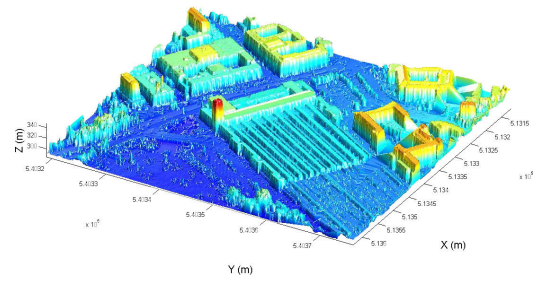
City Site 4, $\mu = 299.6017, \sigma = 10.5357$								
Method	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
VisuShrink	299.6019	10.5006	299.6018	10.4587	299.6020	10.4172	299.6022	10.3828
NormalShrink	299.6017	10.5238	299.6017	10.5199	299.6017	10.5189	299.6017	10.5187
SureShrink	299.6018	10.5072	299.6018	10.4906	299.6018	10.4864	299.6018	10.4853
BayesShrink	299.6017	10.5314	299.6017	10.5301	299.6017	10.5297	299.6017	10.5296
Hypothesis Testing	299.6018	10.4876	299.6011	10.4052	299.6018	10.2409	299.6038	9.9553
GCV	299.6018	10.5094	299.6015	10.4648	299.6021	10.3685	299.6037	10.1806

Table 8.4: Overall denoising performance analysis for DSM generation on City Site 4 excluding outliers

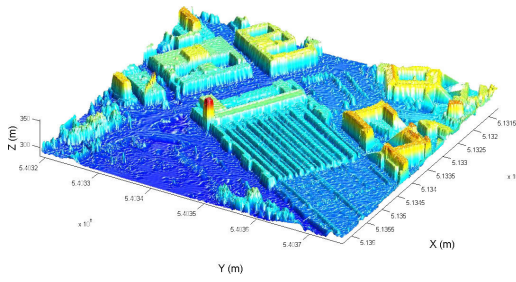
City Site 4, $\mu = 299.5510, \sigma = 10.4379$								
Method	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
VisuShrink	299.5511	10.4101	299.5511	10.3740	299.5512	10.3374	299.5513	10.3073
NormalShrink	299.5510	10.4286	299.5510	10.4256	299.5510	10.4248	299.5510	10.4247
SureShrink	299.5511	10.4135	299.5511	10.3969	299.5511	10.3920	299.5511	10.3910
BayesShrink	299.5510	10.4345	299.5510	10.4335	299.5510	10.4332	299.5510	10.4331
Hypothesis Testing	299.5511	10.3991	299.5503	10.3235	299.5513	10.1622	299.5534	9.8755
GCV	299.5511	10.4178	299.5508	10.3794	299.5512	10.2884	299.5533	10.1001



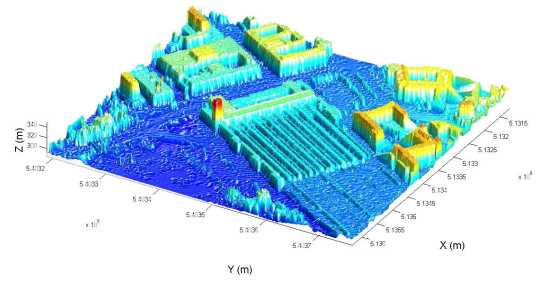
(a) Original including outliers



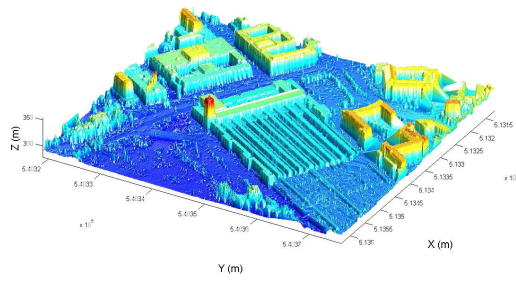
(b) Original excluding outliers



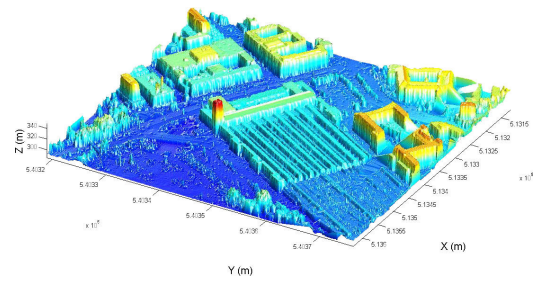
(c) VisuShrink including outliers



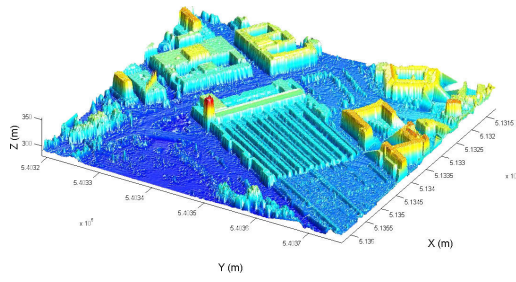
(d) VisuShrink excluding outliers



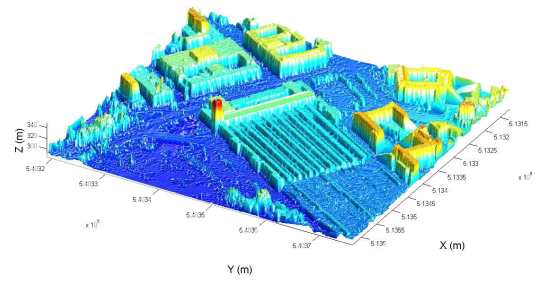
(e) NormalShrink including outliers



(f) NormalShrink excluding outliers

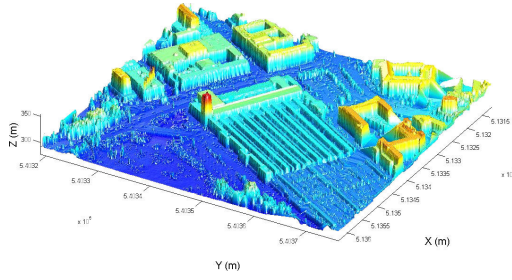


(g) SureShrink including outliers

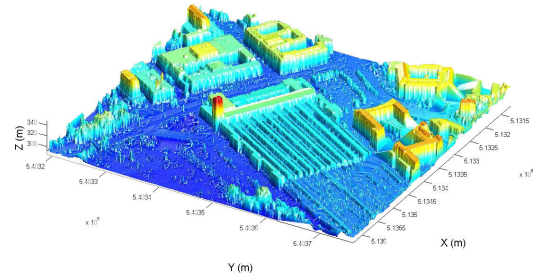


(h) SureShrink excluding outliers

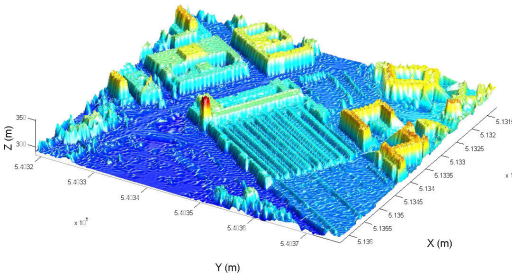
Figure 8.3: 3-D DSM plots of City Site 4



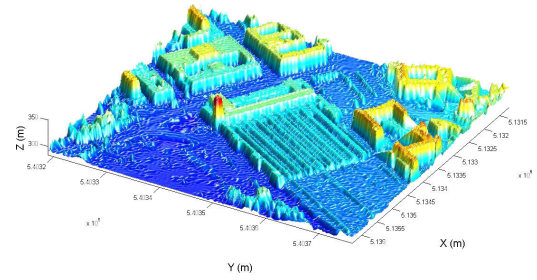
(i) BayesShrink including outliers



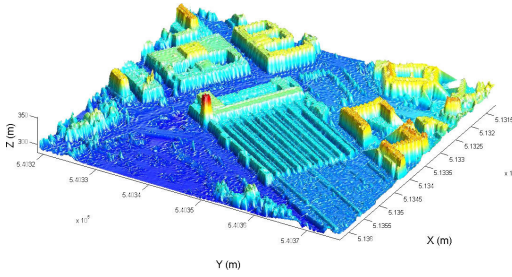
(j) BayesShrink excluding outliers



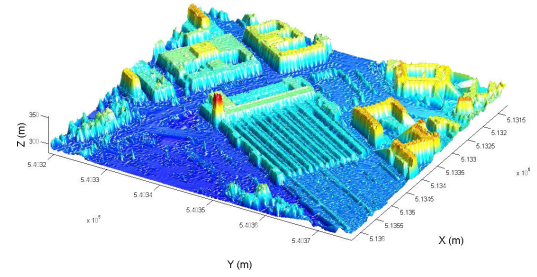
(k) Hypothesis Testing including outliers



(l) Hypothesis Testing excluding outliers



(m) GCV including outliers



(n) GCV excluding outliers

Figure 8.3 (continued)

NormalShrink and BayesShrink are both adaptive thresholding methods based on generalized Gaussian distribution (GGD). Although these two methods are well-accepted in the image processing community, they performed poor in our case which can be verified in Table 8.3 and 8.4. The poor performance may be related to the noise characteristics in the signal. SureShrink, which is an adaptive thresholding method, did not perform as well as the universal

method VisuShrink. However, the remaining adaptive methods Hypothesis Testing and Generalized Cross-validation got the best scores, Hypothesis Testing being the winner.

The choice of wavelet model is an important parameter for denoising. Tables 8.1 and 8.2 shows the best performing wavelet models for each case. However, while making the choice, it has to be verified visually. Some wavelets such as Haar and Biorthogonal 1.1 affect the structure of the signal. In our case, Daubechies 20 got reasonable results.

The level decomposition is also an important parameter for denoising. If it is low, the desired denoising performance is not reached. If it is too high, the structure of the actual signal is lost. As an example, for Hypothesis Testing, the original signal is lost after two levels of decomposition.

In order to investigate the effect of wavelet-based denoising on DEM generation, each method is applied to City Site 3 with hard thresholding using Daubechies 20 wavelet and decomposing up to four levels both including and excluding outliers. The statistical results are given in Table 8.5 and 8.6 and the 3-D DEM plots with one level of decomposition are illustrated in Figure 8.4.

Table 8.5: Overall denoising performance analysis for DEM generation on City Site 3 including outliers

City Site 3, $\mu = 310.7064$, $\sigma = 4.4727$								
Method	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
VisuShrink	310.4581	4.3536	310.4155	4.4541	310.4283	4.4501	310.4339	4.4559
NormalShrink	310.6566	4.4470	310.6595	4.4482	310.6586	4.4479	310.6585	4.4480
SureShrink	310.5053	4.3692	310.5099	4.3705	310.5087	4.3686	310.5084	4.3684
BayesShrink	310.7130	4.4703	310.7155	4.4718	310.7154	4.4718	310.7154	4.4718
Hypothesis Testing	310.4251	4.3561	310.3158	4.4035	310.0583	4.3888	309.7669	4.5203
GCV	310.4813	4.3655	310.3510	4.4060	310.0916	4.3644	309.7193	4.3316

Table 8.6: Overall denoising performance analysis for DEM generation on City Site 3 excluding outliers

City Site 3, $\mu = 310.7514$, $\sigma = 4.4699$								
Method	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
VisuShrink	310.5955	4.3632	310.5296	4.3639	310.5647	4.3772	310.5706	4.3791
NormalShrink	310.7469	4.4730	310.7429	4.4621	310.7428	4.4621	310.7428	4.4621
SureShrink	310.5889	4.3676	310.6144	4.3945	310.6124	4.3941	310.6135	4.3950
BayesShrink	310.7531	4.4653	310.7526	4.4652	310.7526	4.4652	310.7526	4.4652
Hypothesis Testing	310.5401	4.3788	310.4285	4.3993	310.1308	4.3724	309.9498	4.5067
GCV	310.5778	4.3800	310.4203	4.3679	310.1719	4.3737	309.7653	4.3038

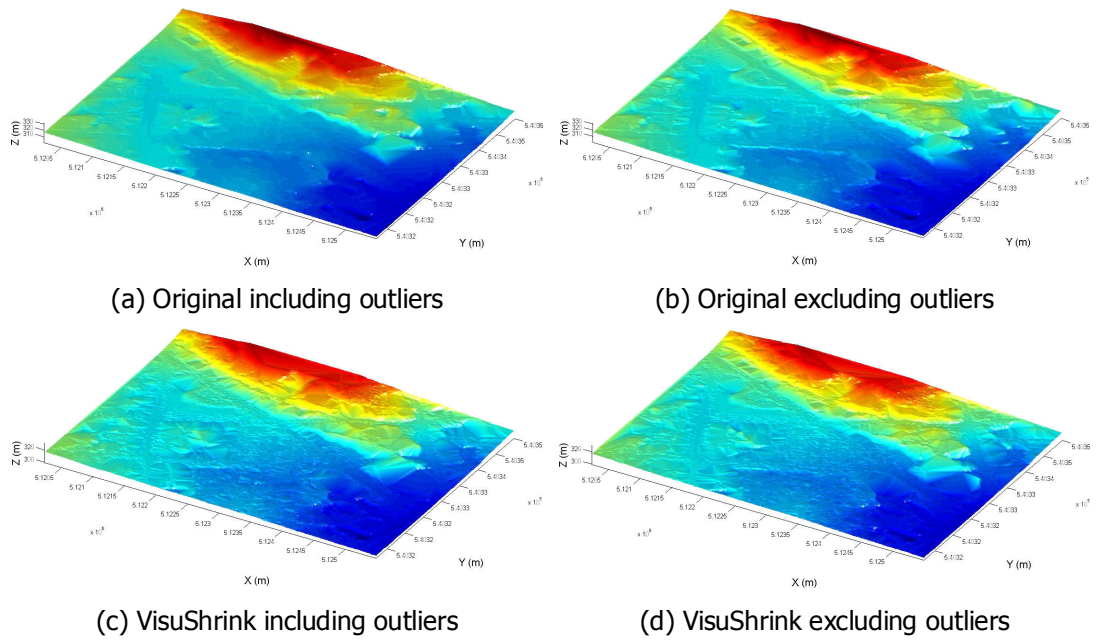


Figure 8.4: 3-D DEM plots of City Site 3

Although the σ values seem to be better after denoising, the plots seem to be noisy compared with originals. The success of wavelet-based denoising on DEM generation with progressive morphological filtering is indeterminate.

CHAPTER 9

CONCLUSIONS

The method of airborne laser scanning or LIDAR is one of the most popular methods to produce highly detailed DSMs and DEMs. The primary goal of this thesis is to obtain better final products using wavelet-based techniques. A wavelet-based outlier detection procedure was used to remove outliers. The performance of one universal (VisuShrink) and five adaptive (NormalShrink, SureShrink, BayesShrink, Hypothesis Testing, Generalized Cross-validation) wavelet shrinkage methods were tested for denoising. The conclusions are as follows:

Wavelet-based outlier detection has positive effect on DSM generation. It removes the spikes and artifacts. Although the expected benefit was not observed for DEM generation in every case, it works well in the presence of large artifacts.

Wavelet thresholding is an effective way of image denoising. Soft thresholding performs better than hard thresholding in wavelet-based denoising for DSM generation. Subband adaptive methods are more suitable than the universal method. Hypothesis Testing is a good choice for threshold estimation. The wavelet model used for denoising must be chosen carefully both verifying

statistically and visually. The decomposition level is an important parameter in wavelet-based denoising, since desired denoising performance is not reached when low and the actual signal is lost when too high. The effect of wavelet-based denoising on DEM generation is indeterminate. It must be further investigated considering the mechanism of progressive morphological filtering.

Although this study gives a deep understanding of the proposed method, further research is necessary to validate the applicability of the method. The final DSMs and DEMs obtained in this work must be compared with actual ones using known ground control points. This was not possible in this work, since the actual approved DEMs of the test sites were not available.

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APPENDIX A

TEST RESULTS

Table A.1: Results of outlier detection for DSM generation

Site	without outlier detection		with outlier detection	
	μ	σ	μ	σ
City Site 1	348.3725	31.1700	348.4279	31.0761
City Site 2	303.4698	11.7044	303.4662	11.5645
City Site 3	318.7184	9.1404	318.6986	9.0395
City Site 4	299.6017	10.5357	299.5510	10.4379
Forest Site 5	289.3199	36.2734	289.2938	35.9261
Forest Site 6	306.0719	33.7566	302.7327	13.2528
Forest Site 7	299.8622	40.9171	298.6160	33.9746
Forest Site 8	263.0109	14.9046	262.7298	14.4656

Table A.2: Results of outlier detection for DEM generation

Site	without outlier detection		with outlier detection	
	μ	σ	μ	σ
City Site 1	331.3947	25.6302	331.5556	25.7231
City Site 2	292.9100	3.5970	292.9549	3.6377
City Site 3	310.7064	4.4727	310.7514	4.4699
City Site 4	291.9491	3.5185	292.0253	3.5621
Forest Site 5	296.1060	57.3415	289.6428	36.6865
Forest Site 6	303.0317	13.0421	302.9940	13.0441
Forest Site 7	294.9100	9.8161	294.9317	9.8140
Forest Site 8	262.1757	15.1136	261.9629	14.7175

Table A.3: Results of VisuShrink with hard thresholding for DSM generation on City Site 1 including outliers

City Site 1, $\mu = 348.3725$, $\sigma = 31.1700$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	348.3725	31.1685	348.3725	31.1681	348.3725	31.1681	348.3725	31.1681
db2	348.3725	31.1675	348.3725	31.1667	348.3726	31.1665	348.3725	31.1666
db4	348.3725	31.1670	348.3725	31.1660	348.3725	31.1659	348.3724	31.1659
db6	348.3726	31.1661	348.3726	31.1649	348.3726	31.1647	348.3726	31.1647
db8	348.3724	31.1657	348.3725	31.1642	348.3725	31.1641	348.3725	31.1641
db10	348.3725	31.1652	348.3724	31.1635	348.3724	31.1632	348.3724	31.1632
db12	348.3725	31.1646	348.3725	31.1626	348.3725	31.1624	348.3725	31.1624
db20	348.3724	31.1630	348.3723	31.1603	348.3724	31.1599	348.3724	31.1599
sym2	348.3725	31.1675	348.3725	31.1667	348.3726	31.1665	348.3725	31.1666
sym4	348.3724	31.1674	348.3724	31.1666	348.3724	31.1665	348.3724	31.1665
sym6	348.3724	31.1672	348.3724	31.1663	348.3724	31.1662	348.3724	31.1661
sym8	348.3724	31.1669	348.3725	31.1660	348.3725	31.1659	348.3724	31.1659
sym10	348.3724	31.1668	348.3725	31.1657	348.3725	31.1656	348.3725	31.1656
sym12	348.3725	31.1667	348.3725	31.1657	348.3725	31.1656	348.3725	31.1656
sym20	348.3725	31.1663	348.3724	31.1651	348.3724	31.1649	348.3724	31.1649
coif1	348.3726	31.1674	348.3726	31.1668	348.3725	31.1667	348.3726	31.1667
coif2	348.3724	31.1675	348.3724	31.1667	348.3724	31.1667	348.3724	31.1666
coif3	348.3724	31.1672	348.3724	31.1664	348.3724	31.1664	348.3724	31.1664
coif4	348.3724	31.1670	348.3724	31.1662	348.3724	31.1660	348.3724	31.1660
coif5	348.3724	31.1669	348.3725	31.1659	348.3724	31.1658	348.3724	31.1658
bior1.1	348.3725	31.1685	348.3725	31.1681	348.3725	31.1681	348.3725	31.1681
bior1.3	348.3725	31.1703	348.3725	31.1703	348.3725	31.1703	348.3725	31.1703
bior1.5	348.3725	31.1706	348.3725	31.1706	348.3725	31.1706	348.3725	31.1706
bior2.2	348.3725	31.1683	348.3724	31.1681	348.3725	31.1681	348.3725	31.1681
bior2.4	348.3725	31.1688	348.3725	31.1687	348.3725	31.1687	348.3725	31.1687
bior2.6	348.3725	31.1690	348.3725	31.1689	348.3725	31.1689	348.3725	31.1689
bior2.8	348.3725	31.1691	348.3725	31.1691	348.3725	31.1691	348.3725	31.1691
bior3.1	348.3725	31.1700	348.3725	31.1701	348.3725	31.1701	348.3725	31.1701
bior3.3	348.3725	31.1692	348.3725	31.1693	348.3725	31.1693	348.3725	31.1693
bior3.5	348.3725	31.1691	348.3725	31.1691	348.3725	31.1691	348.3725	31.1691
bior3.7	348.3725	31.1691	348.3725	31.1692	348.3725	31.1692	348.3725	31.1692
bior3.9	348.3725	31.1692	348.3725	31.1692	348.3725	31.1692	348.3725	31.1692
bior4.4	348.3724	31.1674	348.3725	31.1666	348.3725	31.1665	348.3725	31.1664
bior5.5	348.3725	31.1662	348.3725	31.1634	348.3724	31.1628	348.3724	31.1628
bior6.8	348.3724	31.1674	348.3724	31.1666	348.3724	31.1665	348.3724	31.1665
rbio1.1	348.3725	31.1685	348.3725	31.1681	348.3725	31.1681	348.3725	31.1681
rbio1.3	348.3725	31.1678	348.3724	31.1672	348.3725	31.1671	348.3725	31.1671
rbio1.5	348.3725	31.1679	348.3725	31.1672	348.3724	31.1672	348.3725	31.1672
rbio2.2	348.3726	31.1684	348.3727	31.1705	348.3727	31.1712	348.3727	31.1713
rbio2.4	348.3724	31.1666	348.3725	31.1666	348.3726	31.1670	348.3726	31.1670
rbio2.6	348.3724	31.1663	348.3724	31.1657	348.3723	31.1660	348.3723	31.1660
rbio2.8	348.3724	31.1661	348.3724	31.1652	348.3724	31.1655	348.3724	31.1655
rbio3.1	348.3725	31.2194	348.3726	31.4082	348.3723	31.6127	348.3723	31.7357
rbio3.3	348.3725	31.1720	348.3725	31.2042	348.3726	31.2229	348.3726	31.2263
rbio3.5	348.3725	31.1674	348.3725	31.1828	348.3726	31.1938	348.3725	31.1961
rbio3.7	348.3725	31.1660	348.3724	31.1760	348.3725	31.1852	348.3726	31.1868
rbio3.9	348.3725	31.1654	348.3725	31.1733	348.3727	31.1820	348.3726	31.1832
rbio4.4	348.3724	31.1679	348.3725	31.1676	348.3725	31.1676	348.3725	31.1676
rbio5.5	348.3725	31.1688	348.3725	31.1687	348.3725	31.1687	348.3725	31.1687
rbio6.8	348.3724	31.1670	348.3724	31.1662	348.3725	31.1661	348.3725	31.1661
dmey	348.3725	31.1660	348.3725	31.1646	348.3725	31.1645	348.3725	31.1646

Table A.4: Results of VisuShrink with hard thresholding for DEM generation on City Site 1 including outliers

City Site 1, $\mu = 331.3947$, $\sigma = 25.6302$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	330.9192	25.3069	330.7289	25.2312	330.6863	25.1986	330.6730	25.1895
db2	331.0187	25.4384	330.6986	25.2140	330.6446	25.1826	330.6510	25.1671
db4	330.8901	25.3479	330.8893	25.2405	330.8916	25.2683	330.9160	25.2756
db6	331.1053	25.4398	331.0663	25.3099	331.1285	25.3243	331.1223	25.2932
db8	331.0896	25.4199	331.2276	25.2677	331.2633	25.2014	331.2493	25.1781
db10	331.3109	25.4740	331.4933	25.2515	331.4932	25.2344	331.4541	25.2396
db12	331.3456	25.4609	331.5063	25.3027	331.5345	25.3305	331.5699	25.3736
db20	331.7777	25.4669	332.3115	25.2541	332.2805	25.1845	332.2619	25.1999
sym2	331.0187	25.4384	330.6986	25.2140	330.6446	25.1826	330.6510	25.1671
sym4	331.1614	25.5016	331.1100	25.4235	331.0741	25.4207	331.0697	25.4219
sym6	331.1840	25.5113	331.1090	25.3858	331.1239	25.3869	331.1024	25.3719
sym8	331.2071	25.4839	331.2473	25.4405	331.2648	25.3873	331.2973	25.4397
sym10	331.2586	25.5183	331.2413	25.4137	331.2540	25.4104	331.2613	25.4261
sym12	331.1906	25.4921	331.2888	25.4035	331.2872	25.3700	331.2879	25.3607
sym20	331.2223	25.4765	331.4447	25.3384	331.5070	25.3219	331.5055	25.3181
coif1	331.0453	25.5215	330.8188	25.4020	330.7698	25.4017	330.7300	25.3935
coif2	331.1848	25.4997	331.0736	25.4089	331.0428	25.3620	331.0487	25.3701
coif3	331.1363	25.5040	331.0927	25.3650	331.0815	25.3683	331.0905	25.3813
coif4	331.2107	25.5109	331.1822	25.3729	331.2434	25.3797	331.2359	25.3638
coif5	331.2198	25.5137	331.1959	25.3746	331.1688	25.3916	331.1675	25.3987
bior1.1	330.9192	25.3069	330.7289	25.2312	330.6863	25.1986	330.6730	25.1895
bior1.3	331.0050	25.3226	330.8637	25.2045	330.8689	25.2148	330.8652	25.2165
bior1.5	330.9570	25.3344	330.8599	25.2465	330.8777	25.2334	330.8850	25.2386
bior2.2	331.2381	25.5628	331.0744	25.4767	331.0549	25.4549	331.0503	25.4587
bior2.4	331.1799	25.5565	331.0561	25.4998	331.0720	25.5080	331.0734	25.5074
bior2.6	331.1583	25.5434	331.0554	25.4884	331.0336	25.4966	331.0390	25.4976
bior2.8	331.1484	25.5378	331.0401	25.4278	331.0265	25.4401	331.0145	25.4337
bior3.1	331.0584	25.5261	330.9992	25.4805	330.9877	25.4762	330.9891	25.4707
bior3.3	331.0690	25.4742	331.0046	25.4408	330.9859	25.4391	330.9804	25.4387
bior3.5	331.0312	25.4836	330.9622	25.4177	330.9532	25.4130	330.9528	25.4126
bior3.7	331.0426	25.4570	330.9899	25.4442	330.9852	25.4420	330.9867	25.4420
bior3.9	331.0574	25.4216	330.9572	25.3699	330.9493	25.3685	330.9520	25.3694
bior4.4	331.1385	25.5062	331.0586	25.4052	330.9922	25.3828	330.9788	25.3834
bior5.5	331.3303	25.5559	331.3735	25.3783	331.6096	25.4283	331.6332	25.4352
bior6.8	331.1518	25.5199	331.1104	25.3616	331.0835	25.3390	331.0801	25.3451
rbio1.1	330.9192	25.3069	330.7289	25.2312	330.6863	25.1986	330.6730	25.1895
rbio1.3	331.1101	25.4896	330.8320	25.4246	330.8127	25.4133	330.8169	25.4130
rbio1.5	331.0685	25.4747	330.9781	25.3171	331.0388	25.3376	331.0197	25.3317
rbio2.2	330.5512	25.2897	330.0230	24.9823	330.0827	24.8183	330.0626	24.7798
rbio2.4	331.2928	25.5172	331.2078	25.3145	331.3317	25.2851	331.3270	25.2324
rbio2.6	331.3800	25.5475	331.6699	25.4421	331.5562	25.2265	331.5617	25.2155
rbio2.8	331.4453	25.5766	331.7716	25.4043	331.9640	25.3466	331.9364	25.2801
rbio3.1	329.9312	25.3845	327.2936	24.1906	328.8109	22.8526	327.9401	22.9175
rbio3.3	331.6650	25.6827	332.3452	25.1546	332.7569	24.1134	333.4308	24.1909
rbio3.5	331.5466	25.6643	332.6614	25.4091	333.0631	24.9295	333.1868	24.8064
rbio3.7	331.6354	25.6281	332.9714	25.6260	333.5057	25.0652	333.9415	25.1541
rbio3.9	331.7824	25.6530	333.0065	25.4828	333.5559	24.7981	333.6234	24.5606
rbio4.4	331.1374	25.5108	331.0128	25.3799	331.0251	25.3649	331.0322	25.3637
rbio5.5	331.1905	25.4622	331.0803	25.3474	331.0672	25.3446	331.0819	25.3530
rbio6.8	331.1708	25.4930	331.1750	25.3916	331.1825	25.3725	331.1784	25.3639
dmey	331.5105	25.5410	331.5353	25.3883	331.5339	25.3256	331.5501	25.3316

Table A.5: Results of VisuShrink with hard thresholding for DSM generation on City Site 1
excluding outliers

City Site 1, $\mu = 348.4279$, $\sigma = 31.0761$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	348.4279	31.0745	348.4279	31.0741	348.4279	31.0741	348.4279	31.0741
db2	348.4279	31.0736	348.4279	31.0729	348.4280	31.0728	348.4280	31.0728
db4	348.4279	31.0732	348.4279	31.0724	348.4279	31.0723	348.4279	31.0723
db6	348.4279	31.0726	348.4279	31.0715	348.4279	31.0714	348.4279	31.0714
db8	348.4279	31.0723	348.4279	31.0710	348.4279	31.0709	348.4279	31.0709
db10	348.4279	31.0718	348.4279	31.0703	348.4279	31.0701	348.4279	31.0701
db12	348.4279	31.0713	348.4280	31.0696	348.4279	31.0694	348.4279	31.0694
db20	348.4278	31.0701	348.4278	31.0678	348.4278	31.0675	348.4278	31.0675
sym2	348.4279	31.0736	348.4279	31.0729	348.4280	31.0728	348.4280	31.0728
sym4	348.4279	31.0736	348.4278	31.0729	348.4279	31.0728	348.4279	31.0728
sym6	348.4279	31.0735	348.4278	31.0727	348.4278	31.0726	348.4278	31.0726
sym8	348.4279	31.0733	348.4279	31.0725	348.4279	31.0724	348.4279	31.0724
sym10	348.4279	31.0732	348.4279	31.0723	348.4279	31.0722	348.4279	31.0722
sym12	348.4279	31.0731	348.4279	31.0722	348.4279	31.0721	348.4279	31.0720
sym20	348.4280	31.0728	348.4279	31.0718	348.4279	31.0716	348.4278	31.0716
coif1	348.4280	31.0736	348.4280	31.0730	348.4279	31.0729	348.4280	31.0729
coif2	348.4279	31.0737	348.4278	31.0730	348.4278	31.0730	348.4278	31.0730
coif3	348.4279	31.0735	348.4279	31.0728	348.4279	31.0727	348.4279	31.0727
coif4	348.4278	31.0734	348.4278	31.0726	348.4278	31.0725	348.4278	31.0725
coif5	348.4278	31.0732	348.4279	31.0724	348.4278	31.0723	348.4278	31.0723
bior1.1	348.4279	31.0745	348.4279	31.0741	348.4279	31.0741	348.4279	31.0741
bior1.3	348.4279	31.0763	348.4279	31.0763	348.4279	31.0763	348.4279	31.0763
bior1.5	348.4279	31.0766	348.4279	31.0767	348.4280	31.0767	348.4280	31.0767
bior2.2	348.4279	31.0743	348.4279	31.0742	348.4279	31.0742	348.4279	31.0742
bior2.4	348.4279	31.0749	348.4279	31.0748	348.4279	31.0747	348.4279	31.0747
bior2.6	348.4279	31.0751	348.4279	31.0750	348.4279	31.0750	348.4279	31.0750
bior2.8	348.4279	31.0752	348.4279	31.0751	348.4279	31.0751	348.4279	31.0751
bior3.1	348.4279	31.0759	348.4279	31.0760	348.4279	31.0760	348.4279	31.0760
bior3.3	348.4279	31.0752	348.4279	31.0752	348.4279	31.0752	348.4279	31.0752
bior3.5	348.4279	31.0751	348.4279	31.0752	348.4279	31.0752	348.4279	31.0752
bior3.7	348.4279	31.0752	348.4279	31.0752	348.4279	31.0752	348.4279	31.0752
bior3.9	348.4279	31.0752	348.4279	31.0752	348.4279	31.0752	348.4279	31.0752
bior4.4	348.4279	31.0736	348.4279	31.0729	348.4279	31.0728	348.4279	31.0728
bior5.5	348.4280	31.0725	348.4279	31.0701	348.4279	31.0695	348.4279	31.0695
bior6.8	348.4279	31.0736	348.4279	31.0729	348.4279	31.0729	348.4279	31.0729
rbio1.1	348.4279	31.0745	348.4279	31.0741	348.4279	31.0741	348.4279	31.0741
rbio1.3	348.4279	31.0739	348.4279	31.0733	348.4279	31.0732	348.4279	31.0732
rbio1.5	348.4279	31.0740	348.4279	31.0734	348.4279	31.0733	348.4279	31.0733
rbio2.2	348.4280	31.0746	348.4281	31.0766	348.4281	31.0772	348.4282	31.0773
rbio2.4	348.4279	31.0729	348.4279	31.0730	348.4280	31.0734	348.4280	31.0734
rbio2.6	348.4279	31.0726	348.4279	31.0722	348.4277	31.0724	348.4278	31.0724
rbio2.8	348.4279	31.0725	348.4278	31.0719	348.4278	31.0721	348.4278	31.0721
rbio3.1	348.4279	31.1248	348.4280	31.3104	348.4277	31.5110	348.4278	31.6329
rbio3.3	348.4279	31.0780	348.4279	31.1083	348.4280	31.1254	348.4280	31.1286
rbio3.5	348.4279	31.0737	348.4280	31.0882	348.4282	31.0977	348.4281	31.0996
rbio3.7	348.4279	31.0725	348.4278	31.0820	348.4278	31.0903	348.4280	31.0916
rbio3.9	348.4279	31.0719	348.4279	31.0796	348.4281	31.0872	348.4280	31.0885
rbio4.4	348.4279	31.0742	348.4279	31.0739	348.4280	31.0739	348.4279	31.0739
rbio5.5	348.4279	31.0750	348.4279	31.0748	348.4279	31.0748	348.4279	31.0748
rbio6.8	348.4279	31.0733	348.4279	31.0727	348.4279	31.0726	348.4279	31.0726
dmey	348.4279	31.0726	348.4279	31.0715	348.4279	31.0714	348.4279	31.0715

Table A.6: Results of VisuShrink with hard thresholding for DEM generation on City Site 1 excluding outliers

City Site 1, $\mu = 331.5556$, $\sigma = 25.7231$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	331.1057	25.3922	330.9123	25.3185	330.8794	25.3041	330.8733	25.2990
db2	331.1994	25.6017	330.9841	25.3731	330.9296	25.3472	330.9116	25.3440
db4	331.1462	25.4487	331.1355	25.3360	331.1161	25.3229	331.1395	25.3202
db6	331.3036	25.4830	331.4507	25.3844	331.4933	25.4383	331.5048	25.4337
db8	331.3329	25.5721	331.4706	25.4485	331.4087	25.3882	331.3953	25.3949
db10	331.4993	25.5383	331.8096	25.4385	331.8448	25.3477	331.8543	25.3343
db12	331.5753	25.5364	331.7041	25.3417	331.7386	25.3516	331.7363	25.3186
db20	331.9425	25.5092	332.4208	25.3179	332.4991	25.3112	332.4536	25.2709
sym2	331.1994	25.6017	330.9841	25.3731	330.9296	25.3472	330.9116	25.3440
sym4	331.4580	25.6181	331.3527	25.5133	331.3028	25.4600	331.3125	25.4659
sym6	331.4032	25.5755	331.3010	25.4509	331.3511	25.4509	331.3528	25.4499
sym8	331.4360	25.5932	331.4611	25.5123	331.4427	25.5216	331.4630	25.5333
sym10	331.4360	25.5840	331.4347	25.4396	331.4344	25.4212	331.4290	25.4156
sym12	331.4622	25.6060	331.5028	25.4510	331.4891	25.4513	331.5025	25.4541
sym20	331.5031	25.5766	331.5976	25.4126	331.6228	25.3848	331.6102	25.3694
coif1	331.1982	25.5889	330.9572	25.4197	330.9559	25.4326	330.9377	25.4145
coif2	331.4519	25.6447	331.3787	25.6047	331.3635	25.5602	331.3566	25.5563
coif3	331.3813	25.6176	331.4006	25.5135	331.4364	25.5138	331.4497	25.5007
coif4	331.3811	25.5929	331.4676	25.4799	331.4815	25.4550	331.4958	25.4680
coif5	331.3852	25.5837	331.5005	25.4500	331.5152	25.4468	331.4946	25.4469
bior1.1	331.1057	25.3922	330.9123	25.3185	330.8794	25.3041	330.8733	25.2990
bior1.3	331.1611	25.4100	331.0049	25.2990	331.0108	25.2757	330.9887	25.2692
bior1.5	331.1507	25.4434	331.0245	25.3207	331.0661	25.3160	331.0672	25.3164
bior2.2	331.4383	25.6762	331.3239	25.6051	331.3313	25.6046	331.3313	25.6084
bior2.4	331.3831	25.6619	331.3009	25.6243	331.3033	25.6274	331.3036	25.6232
bior2.6	331.3599	25.6406	331.2607	25.6218	331.2502	25.6083	331.2483	25.6120
bior2.8	331.3695	25.6431	331.2767	25.5745	331.2600	25.5602	331.2622	25.5602
bior3.1	331.3330	25.6589	331.2893	25.6187	331.2950	25.6159	331.2904	25.6172
bior3.3	331.3068	25.6087	331.2591	25.5777	331.2429	25.5695	331.2425	25.5664
bior3.5	331.2849	25.6024	331.2060	25.5772	331.1885	25.5658	331.1871	25.5621
bior3.7	331.3258	25.6076	331.2613	25.5742	331.2628	25.5843	331.2613	25.5834
bior3.9	331.2926	25.5794	331.2599	25.5839	331.2548	25.5710	331.2534	25.5715
bior4.4	331.4022	25.6082	331.2631	25.4697	331.2364	25.4565	331.2384	25.4543
bior5.5	331.5063	25.6174	331.5760	25.4914	331.7330	25.5502	331.6529	25.5073
bior6.8	331.3830	25.6191	331.3143	25.5259	331.3254	25.5373	331.3175	25.5389
rbio1.1	331.1057	25.3922	330.9123	25.3185	330.8794	25.3041	330.8733	25.2990
rbio1.3	331.3552	25.6393	331.1572	25.4897	331.1537	25.4677	331.1452	25.4634
rbio1.5	331.3046	25.6133	331.3215	25.5497	331.2969	25.5439	331.2834	25.5458
rbio2.2	330.8154	25.4227	330.3057	25.1660	330.4656	25.0859	330.4582	25.0506
rbio2.4	331.4765	25.6030	331.6128	25.4834	331.6519	25.4204	331.6745	25.4137
rbio2.6	331.6229	25.6192	331.9528	25.5181	332.0344	25.4029	332.0692	25.3769
rbio2.8	331.6834	25.6213	332.0493	25.4635	332.2411	25.3494	332.2747	25.3095
rbio3.1	330.1956	25.7200	327.7843	24.0520	328.6844	22.3341	328.7546	22.4519
rbio3.3	332.0012	25.7665	332.7597	25.2089	333.2353	24.5432	333.6295	24.4226
rbio3.5	331.8024	25.7136	332.8834	25.4035	333.2038	24.9171	333.6252	24.9005
rbio3.7	331.9021	25.7105	333.1675	25.4100	334.1558	25.2868	334.3885	25.2231
rbio3.9	332.0018	25.6989	333.2911	25.3701	333.5324	24.7972	333.6871	24.6369
rbio4.4	331.3944	25.6207	331.2566	25.4896	331.2936	25.4777	331.2891	25.4563
rbio5.5	331.4003	25.5603	331.2960	25.4850	331.3053	25.4831	331.3000	25.4730
rbio6.8	331.4689	25.6207	331.4937	25.5074	331.4430	25.4604	331.4648	25.4760
dmey	331.6077	25.5550	331.7676	25.3830	331.7547	25.3459	331.7525	25.3263

Table A.7: Results of VisuShrink with soft thresholding for DSM generation on City Site 1 including outliers

City Site 1, $\mu = 348.3725$, $\sigma = 31.1700$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	348.3725	31.1551	348.3725	31.1403	348.3724	31.1274	348.3722	31.1182
db2	348.3724	31.1575	348.3726	31.1425	348.3733	31.1280	348.3736	31.1168
db4	348.3726	31.1585	348.3725	31.1444	348.3725	31.1298	348.3724	31.1180
db6	348.3726	31.1582	348.3727	31.1430	348.3729	31.1267	348.3730	31.1134
db8	348.3724	31.1583	348.3727	31.1421	348.3728	31.1250	348.3731	31.1114
db10	348.3724	31.1579	348.3722	31.1412	348.3724	31.1232	348.3725	31.1082
db12	348.3724	31.1575	348.3728	31.1399	348.3726	31.1210	348.3728	31.1051
db20	348.3723	31.1570	348.3722	31.1379	348.3725	31.1161	348.3732	31.0984
sym2	348.3724	31.1575	348.3726	31.1425	348.3733	31.1280	348.3736	31.1168
sym4	348.3724	31.1592	348.3722	31.1462	348.3725	31.1323	348.3727	31.1213
sym6	348.3724	31.1594	348.3725	31.1459	348.3726	31.1323	348.3727	31.1208
sym8	348.3724	31.1593	348.3722	31.1458	348.3724	31.1316	348.3726	31.1199
sym10	348.3724	31.1593	348.3726	31.1454	348.3726	31.1309	348.3726	31.1191
sym12	348.3726	31.1592	348.3724	31.1453	348.3724	31.1302	348.3727	31.1187
sym20	348.3726	31.1590	348.3724	31.1445	348.3725	31.1289	348.3726	31.1166
coif1	348.3727	31.1575	348.3726	31.1432	348.3726	31.1287	348.3732	31.1179
coif2	348.3724	31.1594	348.3723	31.1466	348.3724	31.1336	348.3725	31.1227
coif3	348.3724	31.1595	348.3722	31.1464	348.3724	31.1326	348.3724	31.1219
coif4	348.3724	31.1595	348.3722	31.1461	348.3722	31.1321	348.3725	31.1211
coif5	348.3724	31.1594	348.3724	31.1458	348.3723	31.1312	348.3726	31.1200
bior1.1	348.3725	31.1551	348.3725	31.1403	348.3724	31.1274	348.3722	31.1182
bior1.3	348.3725	31.1642	348.3725	31.1567	348.3725	31.1484	348.3726	31.1418
bior1.5	348.3725	31.1658	348.3725	31.1596	348.3724	31.1526	348.3724	31.1463
bior2.2	348.3724	31.1593	348.3723	31.1506	348.3724	31.1431	348.3727	31.1381
bior2.4	348.3725	31.1613	348.3727	31.1538	348.3727	31.1475	348.3729	31.1427
bior2.6	348.3725	31.1621	348.3725	31.1553	348.3726	31.1495	348.3726	31.1450
bior2.8	348.3726	31.1626	348.3726	31.1560	348.3727	31.1506	348.3727	31.1463
bior3.1	348.3725	31.1625	348.3725	31.1588	348.3724	31.1563	348.3723	31.1544
bior3.3	348.3725	31.1620	348.3726	31.1579	348.3726	31.1551	348.3727	31.1529
bior3.5	348.3725	31.1624	348.3724	31.1582	348.3724	31.1551	348.3725	31.1530
bior3.7	348.3725	31.1626	348.3726	31.1584	348.3725	31.1552	348.3726	31.1531
bior3.9	348.3725	31.1628	348.3724	31.1587	348.3724	31.1555	348.3725	31.1533
bior4.4	348.3723	31.1593	348.3726	31.1461	348.3728	31.1323	348.3731	31.1208
bior5.5	348.3727	31.1578	348.3726	31.1380	348.3728	31.1124	348.3729	31.0885
bior6.8	348.3724	31.1598	348.3726	31.1475	348.3726	31.1353	348.3727	31.1256
rbio1.1	348.3725	31.1551	348.3725	31.1403	348.3724	31.1274	348.3722	31.1182
rbio1.3	348.3725	31.1577	348.3722	31.1431	348.3725	31.1283	348.3724	31.1159
rbio1.5	348.3725	31.1590	348.3725	31.1447	348.3724	31.1291	348.3725	31.1159
rbio2.2	348.3729	31.1577	348.3729	31.1436	348.3730	31.1219	348.3741	31.1022
rbio2.4	348.3725	31.1568	348.3728	31.1386	348.3730	31.1166	348.3734	31.0963
rbio2.6	348.3723	31.1572	348.3720	31.1386	348.3721	31.1158	348.3721	31.0941
rbio2.8	348.3722	31.1575	348.3728	31.1382	348.3727	31.1141	348.3725	31.0915
rbio3.1	348.3725	31.2258	348.3724	31.4841	348.3710	32.0676	348.3707	32.9853
rbio3.3	348.3725	31.1612	348.3727	31.1679	348.3736	31.1766	348.3730	31.1657
rbio3.5	348.3725	31.1576	348.3722	31.1445	348.3727	31.1264	348.3725	31.1036
rbio3.7	348.3725	31.1569	348.3727	31.1385	348.3726	31.1138	348.3737	31.0903
rbio3.9	348.3725	31.1567	348.3721	31.1368	348.3715	31.1101	348.3726	31.0829
rbio4.4	348.3725	31.1595	348.3728	31.1469	348.3729	31.1343	348.3731	31.1239
rbio5.5	348.3724	31.1617	348.3724	31.1534	348.3725	31.1466	348.3726	31.1414
rbio6.8	348.3724	31.1591	348.3727	31.1450	348.3726	31.1299	348.3726	31.1172
dmey	348.3724	31.1590	348.3724	31.1440	348.3724	31.1277	348.3726	31.1150

Table A.8: Results of VisuShrink with soft thresholding for DEM generation on City Site 1 including outliers

City Site 1, $\mu = 331.3947, \sigma = 25.6302$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	331.1620	25.3770	331.2460	25.3927	331.4116	25.3883	331.4478	25.4321
db2	331.1018	25.5559	331.5190	25.5941	331.8801	25.7036	331.9928	25.6964
db4	331.2895	25.5475	331.9569	25.6416	332.2851	25.7685	332.4853	25.8405
db6	331.3700	25.5439	332.3088	25.7129	332.7735	25.8286	332.9361	25.8923
db8	331.3224	25.4571	332.4774	25.6665	332.9954	25.8307	333.2184	25.8669
db10	331.5414	25.5817	332.6936	25.6962	333.1989	25.8026	333.3478	25.7920
db12	331.5338	25.5717	332.7144	25.7585	333.3852	25.8735	333.5660	25.9509
db20	331.8167	25.5499	333.2881	25.5567	333.8669	25.7530	334.1658	25.8048
sym2	331.1018	25.5559	331.5190	25.5941	331.8801	25.7036	331.9928	25.6964
sym4	331.3863	25.5239	331.9940	25.6756	332.2479	25.7243	332.4724	25.7681
sym6	331.4260	25.4990	332.1335	25.7197	332.4858	25.7904	332.6569	25.8372
sym8	331.4018	25.4564	332.2663	25.7240	332.7012	25.7959	332.7639	25.8626
sym10	331.4626	25.4905	332.3194	25.7116	332.6979	25.8420	332.7971	25.8180
sym12	331.4818	25.5752	332.4694	25.7368	332.7176	25.7732	332.9277	25.8433
sym20	331.5671	25.5499	332.6542	25.7545	332.9614	25.7496	333.1413	25.8589
coif1	331.1841	25.5278	331.4992	25.5963	331.8850	25.6702	332.0504	25.7256
coif2	331.3976	25.5638	331.8962	25.6346	332.3007	25.7910	332.4316	25.8363
coif3	331.3993	25.5444	332.0696	25.6527	332.4721	25.8242	332.5371	25.8392
coif4	331.3791	25.5072	332.1919	25.6893	332.6975	25.8186	332.7654	25.8342
coif5	331.4189	25.5089	332.3498	25.7388	332.7244	25.8626	332.8824	25.9795
bior1.1	331.1620	25.3770	331.2460	25.3927	331.4116	25.3883	331.4478	25.4321
bior1.3	331.1701	25.3465	331.3118	25.3087	331.4237	25.3171	331.5247	25.3727
bior1.5	331.1820	25.3523	331.3925	25.3627	331.6140	25.3946	331.6096	25.3851
bior2.2	331.3864	25.5967	331.5434	25.5969	331.6403	25.6118	331.7054	25.6518
bior2.4	331.2944	25.5228	331.4803	25.5653	331.5782	25.6240	331.5876	25.5907
bior2.6	331.2799	25.5282	331.4334	25.5488	331.5759	25.5701	331.5543	25.5518
bior2.8	331.2572	25.5142	331.4232	25.5379	331.5055	25.5349	331.5298	25.5434
bior3.1	331.2531	25.5721	331.3101	25.5405	331.3357	25.5413	331.3674	25.5650
bior3.3	331.2497	25.5299	331.4130	25.4959	331.4236	25.4948	331.4885	25.5264
bior3.5	331.2162	25.5159	331.3620	25.5348	331.3534	25.5234	331.4121	25.5590
bior3.7	331.2126	25.5138	331.3924	25.5276	331.4351	25.5601	331.4469	25.5546
bior3.9	331.1874	25.4916	331.3546	25.5324	331.3175	25.5281	331.3645	25.5545
bior4.4	331.4638	25.5632	331.8705	25.7113	332.1896	25.7833	332.3592	25.8264
bior5.5	331.4975	25.5705	332.5129	25.7942	333.4813	26.1098	333.8139	26.2063
bior6.8	331.3680	25.5713	331.9686	25.6858	332.2312	25.7566	332.3684	25.7922
rbio1.1	331.1620	25.3770	331.2460	25.3927	331.4116	25.3883	331.4478	25.4321
rbio1.3	331.4976	25.5894	332.0384	25.7040	332.3421	25.8409	332.4838	25.8288
rbio1.5	331.3884	25.5744	332.0664	25.7221	332.5998	25.8945	332.6456	25.9258
rbio2.2	330.9737	25.4069	331.5858	25.5249	333.6025	25.8801	334.2578	25.9202
rbio2.4	331.5045	25.5773	332.5418	25.8202	333.5444	25.9934	333.9156	26.1819
rbio2.6	331.5977	25.5856	332.7887	25.8196	333.8059	26.1513	334.1549	26.2379
rbio2.8	331.6664	25.5979	332.9662	25.8749	334.0950	26.0409	334.2945	26.1420
rbio3.1	330.4510	25.7467	330.9478	24.7275	329.8882	24.9842	326.4562	23.1710
rbio3.3	332.0752	25.7365	334.1818	25.6901	336.4569	25.3704	337.2574	25.2185
rbio3.5	331.8410	25.6836	333.5941	25.8464	335.6477	26.2928	336.4555	26.2072
rbio3.7	331.7296	25.6141	333.5182	25.7755	335.6495	26.1364	336.3138	26.2420
rbio3.9	331.6960	25.6514	333.8651	25.9569	335.5442	26.0126	336.4774	26.1351
rbio4.4	331.3487	25.5864	331.9648	25.6396	332.4424	25.7655	332.5892	25.8170
rbio5.5	331.3865	25.6009	331.7395	25.5982	331.9549	25.6387	332.0150	25.6531
rbio6.8	331.4311	25.5434	332.3749	25.7718	332.8324	25.8227	332.9605	25.8872
dmey	331.6989	25.4810	332.9402	25.7231	333.2973	25.8183	333.5366	25.9365

Table A.9: Results of VisuShrink with soft thresholding for DSM generation on City Site 1
excluding outliers

City Site 1, $\mu = 348.4279$, $\sigma = 31.0761$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	348.4279	31.0619	348.4279	31.0473	348.4278	31.0344	348.4277	31.0252
db2	348.4278	31.0649	348.4280	31.0507	348.4286	31.0367	348.4289	31.0259
db4	348.4280	31.0661	348.4279	31.0529	348.4279	31.0390	348.4279	31.0277
db6	348.4280	31.0660	348.4281	31.0519	348.4283	31.0366	348.4284	31.0242
db8	348.4278	31.0661	348.4281	31.0513	348.4282	31.0352	348.4285	31.0226
db10	348.4278	31.0658	348.4277	31.0504	348.4278	31.0334	348.4280	31.0193
db12	348.4279	31.0655	348.4282	31.0494	348.4280	31.0317	348.4282	31.0169
db20	348.4278	31.0652	348.4277	31.0477	348.4279	31.0275	348.4286	31.0112
sym2	348.4278	31.0649	348.4280	31.0507	348.4286	31.0367	348.4289	31.0259
sym4	348.4278	31.0667	348.4277	31.0545	348.4279	31.0414	348.4281	31.0310
sym6	348.4278	31.0669	348.4279	31.0544	348.4280	31.0415	348.4281	31.0307
sym8	348.4278	31.0670	348.4277	31.0545	348.4279	31.0411	348.4280	31.0301
sym10	348.4278	31.0670	348.4280	31.0542	348.4280	31.0406	348.4281	31.0296
sym12	348.4280	31.0669	348.4279	31.0540	348.4278	31.0398	348.4281	31.0290
sym20	348.4280	31.0668	348.4279	31.0535	348.4279	31.0388	348.4280	31.0274
coif1	348.4281	31.0649	348.4280	31.0514	348.4281	31.0375	348.4286	31.0270
coif2	348.4278	31.0669	348.4278	31.0549	348.4279	31.0426	348.4279	31.0323
coif3	348.4278	31.0670	348.4277	31.0548	348.4278	31.0418	348.4278	31.0317
coif4	348.4278	31.0671	348.4277	31.0546	348.4277	31.0415	348.4279	31.0312
coif5	348.4278	31.0670	348.4278	31.0545	348.4278	31.0408	348.4280	31.0302
bior1.1	348.4279	31.0619	348.4279	31.0473	348.4278	31.0344	348.4277	31.0252
bior1.3	348.4279	31.0711	348.4280	31.0638	348.4279	31.0556	348.4280	31.0490
bior1.5	348.4279	31.0727	348.4279	31.0667	348.4279	31.0598	348.4279	31.0535
bior2.2	348.4279	31.0663	348.4278	31.0578	348.4279	31.0505	348.4281	31.0456
bior2.4	348.4279	31.0683	348.4281	31.0610	348.4281	31.0550	348.4283	31.0503
bior2.6	348.4280	31.0691	348.4279	31.0626	348.4280	31.0569	348.4280	31.0525
bior2.8	348.4280	31.0696	348.4280	31.0633	348.4281	31.0579	348.4281	31.0538
bior3.1	348.4279	31.0689	348.4279	31.0653	348.4279	31.0627	348.4278	31.0608
bior3.3	348.4279	31.0688	348.4281	31.0648	348.4281	31.0621	348.4281	31.0601
bior3.5	348.4279	31.0692	348.4278	31.0651	348.4279	31.0621	348.4279	31.0602
bior3.7	348.4279	31.0695	348.4280	31.0654	348.4279	31.0624	348.4280	31.0604
bior3.9	348.4279	31.0697	348.4278	31.0657	348.4278	31.0626	348.4279	31.0604
bior4.4	348.4278	31.0668	348.4280	31.0544	348.4282	31.0414	348.4285	31.0304
bior5.5	348.4281	31.0656	348.4280	31.0472	348.4282	31.0229	348.4283	31.0003
bior6.8	348.4278	31.0673	348.4280	31.0559	348.4280	31.0444	348.4281	31.0352
rbio1.1	348.4279	31.0619	348.4279	31.0473	348.4278	31.0344	348.4277	31.0252
rbio1.3	348.4279	31.0650	348.4277	31.0512	348.4281	31.0371	348.4280	31.0252
rbio1.5	348.4279	31.0664	348.4279	31.0531	348.4279	31.0384	348.4279	31.0260
rbio2.2	348.4283	31.0655	348.4284	31.0523	348.4284	31.0314	348.4294	31.0126
rbio2.4	348.4279	31.0647	348.4282	31.0478	348.4283	31.0269	348.4288	31.0077
rbio2.6	348.4277	31.0651	348.4275	31.0478	348.4276	31.0264	348.4276	31.0059
rbio2.8	348.4277	31.0654	348.4281	31.0476	348.4281	31.0250	348.4279	31.0037
rbio3.1	348.4279	31.1327	348.4278	31.3877	348.4265	31.9654	348.4260	32.8670
rbio3.3	348.4279	31.0689	348.4280	31.0756	348.4290	31.0841	348.4283	31.0732
rbio3.5	348.4279	31.0656	348.4278	31.0534	348.4282	31.0359	348.4278	31.0139
rbio3.7	348.4279	31.0650	348.4280	31.0479	348.4280	31.0246	348.4289	31.0020
rbio3.9	348.4279	31.0648	348.4276	31.0464	348.4271	31.0209	348.4280	30.9951
rbio4.4	348.4279	31.0671	348.4281	31.0554	348.4283	31.0435	348.4285	31.0336
rbio5.5	348.4279	31.0689	348.4278	31.0611	348.4280	31.0546	348.4280	31.0496
rbio6.8	348.4278	31.0668	348.4281	31.0538	348.4281	31.0395	348.4280	31.0274
dmey	348.4278	31.0668	348.4279	31.0532	348.4279	31.0381	348.4280	31.0265

Table A.10: Results of VisuShrink with soft thresholding for DEM generation on City Site 1
excluding outliers

City Site 1, $\mu = 331.5556$, $\sigma = 25.7231$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	331.3289	25.4801	331.4728	25.5182	331.6406	25.5498	331.7219	25.5882
db2	331.3612	25.6632	331.8500	25.6928	332.1597	25.7923	332.3612	25.8147
db4	331.5888	25.6357	332.2741	25.7151	332.6272	25.8804	332.8051	25.9554
db6	331.6911	25.6448	332.5100	25.6990	333.0460	25.8828	333.1173	25.9105
db8	331.6270	25.6119	332.6003	25.7596	333.1420	25.8856	333.3245	25.9152
db10	331.7261	25.6405	332.9982	25.7931	333.5497	25.8789	333.7052	25.9751
db12	331.7968	25.6194	332.9806	25.7449	333.5389	25.9024	333.7014	25.9123
db20	332.0583	25.5798	333.5385	25.7183	334.0381	25.8690	334.2613	25.9110
sym2	331.3612	25.6632	331.8500	25.6928	332.1597	25.7923	332.3612	25.8147
sym4	331.5690	25.6321	332.2494	25.7754	332.5414	25.8220	332.6938	25.8491
sym6	331.6121	25.6168	332.4200	25.8236	332.7430	25.8820	332.9186	25.9487
sym8	331.6779	25.6094	332.5994	25.8022	332.9638	25.8989	333.0486	25.9460
sym10	331.7045	25.6367	332.5913	25.7578	333.0082	25.8728	333.0804	25.8381
sym12	331.6589	25.6557	332.7941	25.8545	332.9449	25.8192	333.1309	25.8842
sym20	331.7399	25.6556	332.9674	25.8432	333.2433	25.8377	333.3207	25.8973
coif1	331.4052	25.5900	331.7997	25.7008	332.1441	25.7999	332.2945	25.8390
coif2	331.6574	25.6637	332.2142	25.7527	332.5849	25.8872	332.6645	25.8971
coif3	331.6855	25.6600	332.4144	25.7757	332.7184	25.8868	332.7777	25.9246
coif4	331.7381	25.6604	332.5354	25.7903	332.9338	25.9159	333.1040	25.9363
coif5	331.7246	25.6329	332.6690	25.8010	332.9155	25.9323	333.1409	26.0544
bior1.1	331.3289	25.4801	331.4728	25.5182	331.6406	25.5498	331.7219	25.5882
bior1.3	331.3276	25.4348	331.5103	25.3866	331.6421	25.4563	331.7128	25.4903
bior1.5	331.3037	25.4532	331.5056	25.4149	331.7542	25.5005	331.7920	25.5170
bior2.2	331.5638	25.7037	331.7640	25.7111	331.8628	25.7660	331.9363	25.7766
bior2.4	331.4961	25.6602	331.6722	25.6763	331.7949	25.7040	331.8038	25.6998
bior2.6	331.4815	25.6580	331.6361	25.6436	331.7793	25.7182	331.7921	25.6983
bior2.8	331.4666	25.6404	331.6221	25.6377	331.7634	25.6788	331.8130	25.6940
bior3.1	331.4853	25.6751	331.6107	25.6497	331.6952	25.7081	331.6995	25.7139
bior3.3	331.5432	25.6519	331.6609	25.6361	331.7080	25.6621	331.7147	25.6453
bior3.5	331.4895	25.6571	331.6165	25.6746	331.6367	25.6891	331.6903	25.6840
bior3.7	331.4888	25.6439	331.6252	25.6336	331.6580	25.6523	331.6849	25.6326
bior3.9	331.4949	25.6352	331.5693	25.6357	331.5840	25.6378	331.6188	25.6327
bior4.4	331.6941	25.6723	332.1749	25.7692	332.5026	25.9356	332.6424	25.9433
bior5.5	331.7083	25.6878	332.8789	25.8297	333.7189	26.2496	334.0506	26.3744
bior6.8	331.6936	25.6631	332.2811	25.7657	332.5870	25.8986	332.6876	25.8735
rbio1.1	331.3289	25.4801	331.4728	25.5182	331.6406	25.5498	331.7219	25.5882
rbio1.3	331.6724	25.6871	332.2485	25.8038	332.6396	25.9386	332.8078	25.9141
rbio1.5	331.6121	25.7143	332.3865	25.8503	332.7729	26.0064	332.9017	26.0180
rbio2.2	331.2530	25.5273	331.9855	25.6974	334.0624	26.0355	334.7370	26.1028
rbio2.4	331.7530	25.6332	332.8848	25.8658	333.7552	26.0787	334.0895	26.2235
rbio2.6	331.9140	25.6850	333.0953	25.8793	334.1297	26.1576	334.3849	26.2344
rbio2.8	331.9417	25.6837	333.2100	25.8392	334.3561	26.1508	334.5328	26.2288
rbio3.1	330.4222	25.7691	331.3314	24.9340	329.2715	25.3244	327.8846	23.5240
rbio3.3	332.4219	25.7195	334.4707	25.8318	336.6149	25.6123	337.4513	25.5741
rbio3.5	331.9931	25.7080	333.9525	26.0187	336.0786	26.3272	336.7471	26.3583
rbio3.7	331.9506	25.6733	333.9858	25.9322	335.9888	26.2981	336.7832	26.4458
rbio3.9	331.9444	25.6875	334.1808	26.0341	335.8788	26.0038	336.5254	26.0801
rbio4.4	331.5982	25.6360	332.2954	25.7416	332.7506	25.9231	332.8550	25.9357
rbio5.5	331.6321	25.6787	332.0596	25.6371	332.2240	25.6922	332.2958	25.7071
rbio6.8	331.7736	25.6413	332.7148	25.8011	333.2478	25.9451	333.3316	25.9467
dmey	331.9355	25.6141	333.1977	25.8084	333.4661	25.8879	333.7343	26.0006

Table A.11: Results of NormalShrink with hard thresholding for DSM generation on City Site 2 including outliers

City Site 2, $\mu = 303.4698$, $\sigma = 11.7044$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
db2	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
db4	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
db6	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
db8	303.4698	11.7043	303.4698	11.7043	303.4698	11.7043	303.4698	11.7043
db10	303.4698	11.7043	303.4698	11.7043	303.4698	11.7043	303.4698	11.7043
db12	303.4698	11.7042	303.4698	11.7042	303.4698	11.7042	303.4698	11.7042
db20	303.4698	11.7038	303.4698	11.7038	303.4698	11.7038	303.4698	11.7038
sym2	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
sym4	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
sym6	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
sym8	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
sym10	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
sym12	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
sym20	303.4698	11.7043	303.4698	11.7043	303.4698	11.7043	303.4698	11.7043
coif1	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
coif2	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
coif3	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
coif4	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
coif5	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior1.1	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior1.3	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior1.5	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior2.2	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior2.4	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior2.6	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior2.8	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior3.1	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior3.3	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior3.5	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior3.7	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior3.9	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior4.4	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior5.5	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
bior6.8	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio1.1	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio1.3	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio1.5	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio2.2	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio2.4	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio2.6	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio2.8	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio3.1	303.4698	11.7047	303.4698	11.7047	303.4698	11.7047	303.4698	11.7047
rbio3.3	303.4698	11.7045	303.4698	11.7046	303.4698	11.7046	303.4698	11.7046
rbio3.5	303.4698	11.7046	303.4698	11.7046	303.4698	11.7046	303.4698	11.7046
rbio3.7	303.4698	11.7046	303.4697	11.7047	303.4697	11.7047	303.4697	11.7047
rbio3.9	303.4698	11.7046	303.4698	11.7047	303.4698	11.7047	303.4698	11.7047
rbio4.4	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio5.5	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
rbio6.8	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044	303.4698	11.7044
dmey	303.4698	11.7043	303.4698	11.7043	303.4698	11.7044	303.4697	11.7044

Table A.12: Results of NormalShrink with hard thresholding for DEM generation on City Site 2 including outliers

City Site 2, $\mu = 292.9100$, $\sigma = 3.5970$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.9100	3.5966	292.9100	3.5966	292.9100	3.5966	292.9100	3.5966
db2	292.9083	3.5952	292.9083	3.5952	292.9083	3.5952	292.9083	3.5952
db4	292.9100	3.5958	292.9103	3.5956	292.9103	3.5956	292.9103	3.5956
db6	292.9112	3.5973	292.9103	3.5965	292.9103	3.5965	292.9103	3.5965
db8	292.9075	3.5949	292.9067	3.5950	292.9067	3.5950	292.9067	3.5950
db10	292.9051	3.5915	292.9046	3.5906	292.9045	3.5906	292.9045	3.5906
db12	292.9025	3.5942	292.9024	3.5941	292.9025	3.5941	292.9025	3.5941
db20	292.8798	3.5635	292.8784	3.5628	292.8785	3.5625	292.8785	3.5625
sym2	292.9083	3.5952	292.9083	3.5952	292.9083	3.5952	292.9083	3.5952
sym4	292.9094	3.5957	292.9093	3.5955	292.9093	3.5955	292.9093	3.5955
sym6	292.9094	3.5962	292.9092	3.5959	292.9092	3.5959	292.9092	3.5959
sym8	292.9098	3.5961	292.9100	3.5959	292.9100	3.5959	292.9100	3.5959
sym10	292.9103	3.5973	292.9105	3.5974	292.9105	3.5974	292.9105	3.5974
sym12	292.9121	3.6049	292.9107	3.6030	292.9107	3.6030	292.9107	3.6030
sym20	292.9072	3.5961	292.9076	3.5967	292.9077	3.5967	292.9077	3.5967
coif1	292.9096	3.5972	292.9096	3.5972	292.9096	3.5972	292.9096	3.5972
coif2	292.9108	3.5972	292.9107	3.5971	292.9107	3.5971	292.9107	3.5971
coif3	292.9095	3.5955	292.9095	3.5955	292.9095	3.5955	292.9095	3.5955
coif4	292.9080	3.5946	292.9081	3.5946	292.9081	3.5946	292.9081	3.5946
coif5	292.9097	3.5967	292.9099	3.5968	292.9100	3.5968	292.9100	3.5968
bior1.1	292.9100	3.5966	292.9100	3.5966	292.9100	3.5966	292.9100	3.5966
bior1.3	292.9102	3.5971	292.9102	3.5971	292.9102	3.5971	292.9102	3.5971
bior1.5	292.9089	3.5954	292.9089	3.5954	292.9089	3.5954	292.9089	3.5954
bior2.2	292.9095	3.5958	292.9095	3.5958	292.9095	3.5958	292.9095	3.5958
bior2.4	292.9091	3.5961	292.9091	3.5961	292.9091	3.5961	292.9091	3.5961
bior2.6	292.9092	3.5961	292.9092	3.5961	292.9092	3.5961	292.9092	3.5961
bior2.8	292.9094	3.5964	292.9094	3.5964	292.9094	3.5964	292.9094	3.5964
bior3.1	292.9092	3.5964	292.9092	3.5964	292.9092	3.5964	292.9092	3.5964
bior3.3	292.9073	3.5935	292.9073	3.5935	292.9073	3.5935	292.9073	3.5935
bior3.5	292.9080	3.5944	292.9080	3.5944	292.9080	3.5944	292.9080	3.5944
bior3.7	292.9092	3.5955	292.9092	3.5955	292.9092	3.5955	292.9092	3.5955
bior3.9	292.9080	3.5939	292.9080	3.5939	292.9080	3.5939	292.9080	3.5939
bior4.4	292.9079	3.5941	292.9079	3.5941	292.9079	3.5941	292.9079	3.5941
bior5.5	292.9095	3.5965	292.9098	3.5968	292.9097	3.5971	292.9097	3.5971
bior6.8	292.9111	3.5984	292.9113	3.5987	292.9113	3.5987	292.9113	3.5987
rbio1.1	292.9100	3.5966	292.9100	3.5966	292.9100	3.5966	292.9100	3.5966
rbio1.3	292.9101	3.5966	292.9101	3.5966	292.9101	3.5966	292.9101	3.5966
rbio1.5	292.9076	3.5936	292.9078	3.5941	292.9078	3.5941	292.9078	3.5941
rbio2.2	292.9082	3.5939	292.9085	3.5941	292.9085	3.5941	292.9085	3.5941
rbio2.4	292.9088	3.5961	292.9089	3.5954	292.9089	3.5954	292.9089	3.5954
rbio2.6	292.9081	3.5939	292.9088	3.5957	292.9088	3.5957	292.9088	3.5957
rbio2.8	292.9148	3.6048	292.9143	3.6045	292.9140	3.6044	292.9140	3.6044
rbio3.1	292.8969	3.5900	292.8980	3.5910	292.8980	3.5908	292.8980	3.5908
rbio3.3	292.9049	3.6002	292.9045	3.6004	292.9045	3.6006	292.9045	3.6006
rbio3.5	292.9062	3.6071	292.9053	3.6006	292.9053	3.6005	292.9053	3.6005
rbio3.7	292.8922	3.5823	292.8906	3.5813	292.8911	3.5813	292.8911	3.5813
rbio3.9	292.8889	3.5812	292.8860	3.5790	292.8862	3.5785	292.8861	3.5785
rbio4.4	292.9102	3.5976	292.9101	3.5977	292.9101	3.5977	292.9101	3.5977
rbio5.5	292.9099	3.5963	292.9098	3.5963	292.9098	3.5963	292.9098	3.5963
rbio6.8	292.9089	3.5952	292.9090	3.5959	292.9090	3.5959	292.9090	3.5959
dmey	292.9132	3.6010	292.9137	3.6025	292.9137	3.6025	292.9137	3.6026

Table A.13: Results of NormalShrink with hard thresholding for DSM generation on City Site 2 excluding outliers

City Site 2, $\mu = 303.4662$, $\sigma = 11.5645$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
db2	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
db4	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
db6	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
db8	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
db10	303.4662	11.5644	303.4662	11.5644	303.4662	11.5644	303.4662	11.5644
db12	303.4662	11.5644	303.4662	11.5644	303.4662	11.5644	303.4662	11.5644
db20	303.4662	11.5640	303.4662	11.5640	303.4662	11.5640	303.4662	11.5640
sym2	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
sym4	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
sym6	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
sym8	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
sym10	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
sym12	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
sym20	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
coif1	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
coif2	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
coif3	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
coif4	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
coif5	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior1.1	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior1.3	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior1.5	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior2.2	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior2.4	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior2.6	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior2.8	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior3.1	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior3.3	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior3.5	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior3.7	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior3.9	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior4.4	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior5.5	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
bior6.8	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
rbio1.1	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
rbio1.3	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
rbio1.5	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
rbio2.2	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
rbio2.4	303.4662	11.5646	303.4662	11.5646	303.4662	11.5646	303.4662	11.5646
rbio2.6	303.4662	11.5646	303.4662	11.5646	303.4662	11.5646	303.4662	11.5646
rbio2.8	303.4662	11.5646	303.4662	11.5646	303.4662	11.5646	303.4662	11.5646
rbio3.1	303.4662	11.5648	303.4662	11.5649	303.4662	11.5649	303.4662	11.5649
rbio3.3	303.4662	11.5647	303.4662	11.5647	303.4662	11.5647	303.4662	11.5647
rbio3.5	303.4662	11.5647	303.4662	11.5648	303.4662	11.5648	303.4662	11.5648
rbio3.7	303.4662	11.5647	303.4662	11.5648	303.4662	11.5648	303.4662	11.5648
rbio3.9	303.4662	11.5647	303.4662	11.5648	303.4662	11.5648	303.4662	11.5648
rbio4.4	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
rbio5.5	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
rbio6.8	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645
dmey	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645	303.4662	11.5645

Table A.14: Results of NormalShrink with hard thresholding for DEM generation on City Site 2 excluding outliers

City Site 2, $\mu = 292.9549$, $\sigma = 3.6377$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.9549	3.6377	292.9549	3.6377	292.9549	3.6377	292.9549	3.6377
db2	292.9544	3.6389	292.9543	3.6389	292.9543	3.6389	292.9543	3.6389
db4	292.9550	3.6409	292.9551	3.6409	292.9551	3.6409	292.9551	3.6409
db6	292.9556	3.6396	292.9553	3.6391	292.9553	3.6391	292.9553	3.6391
db8	292.9595	3.6501	292.9601	3.6504	292.9600	3.6504	292.9600	3.6504
db10	292.9562	3.6429	292.9565	3.6432	292.9565	3.6432	292.9565	3.6432
db12	292.9530	3.6350	292.9538	3.6365	292.9538	3.6365	292.9538	3.6365
db20	292.9335	3.6172	292.9319	3.6177	292.9319	3.6184	292.9319	3.6184
sym2	292.9544	3.6389	292.9543	3.6389	292.9543	3.6389	292.9543	3.6389
sym4	292.9569	3.6404	292.9565	3.6406	292.9565	3.6406	292.9565	3.6406
sym6	292.9556	3.6396	292.9554	3.6395	292.9555	3.6396	292.9555	3.6396
sym8	292.9557	3.6401	292.9558	3.6403	292.9558	3.6402	292.9558	3.6402
sym10	292.9588	3.6438	292.9590	3.6439	292.9590	3.6439	292.9590	3.6439
sym12	292.9588	3.6440	292.9589	3.6435	292.9589	3.6435	292.9589	3.6435
sym20	292.9550	3.6419	292.9559	3.6425	292.9559	3.6425	292.9559	3.6425
coif1	292.9534	3.6375	292.9533	3.6376	292.9533	3.6376	292.9533	3.6376
coif2	292.9557	3.6402	292.9557	3.6401	292.9557	3.6401	292.9557	3.6401
coif3	292.9566	3.6399	292.9568	3.6397	292.9568	3.6397	292.9568	3.6397
coif4	292.9574	3.6440	292.9575	3.6443	292.9575	3.6443	292.9575	3.6443
coif5	292.9570	3.6439	292.9571	3.6443	292.9571	3.6443	292.9571	3.6443
bior1.1	292.9549	3.6377	292.9549	3.6377	292.9549	3.6377	292.9549	3.6377
bior1.3	292.9548	3.6378	292.9548	3.6378	292.9548	3.6378	292.9548	3.6378
bior1.5	292.9549	3.6379	292.9549	3.6379	292.9549	3.6379	292.9549	3.6379
bior2.2	292.9528	3.6377	292.9529	3.6376	292.9529	3.6376	292.9529	3.6376
bior2.4	292.9541	3.6371	292.9541	3.6371	292.9541	3.6371	292.9541	3.6371
bior2.6	292.9541	3.6371	292.9541	3.6371	292.9541	3.6371	292.9541	3.6371
bior2.8	292.9537	3.6379	292.9537	3.6379	292.9537	3.6379	292.9537	3.6379
bior3.1	292.9552	3.6395	292.9552	3.6395	292.9552	3.6395	292.9552	3.6395
bior3.3	292.9555	3.6399	292.9555	3.6399	292.9555	3.6399	292.9555	3.6399
bior3.5	292.9549	3.6387	292.9549	3.6387	292.9549	3.6387	292.9549	3.6387
bior3.7	292.9551	3.6390	292.9551	3.6390	292.9551	3.6390	292.9551	3.6390
bior3.9	292.9552	3.6387	292.9552	3.6387	292.9552	3.6387	292.9552	3.6387
bior4.4	292.9546	3.6379	292.9546	3.6381	292.9546	3.6381	292.9546	3.6381
bior5.5	292.9558	3.6418	292.9553	3.6410	292.9553	3.6410	292.9553	3.6410
bior6.8	292.9573	3.6403	292.9562	3.6401	292.9562	3.6401	292.9562	3.6401
rbio1.1	292.9549	3.6377	292.9549	3.6377	292.9549	3.6377	292.9549	3.6377
rbio1.3	292.9557	3.6395	292.9559	3.6398	292.9559	3.6398	292.9559	3.6398
rbio1.5	292.9548	3.6388	292.9549	3.6391	292.9548	3.6391	292.9548	3.6391
rbio2.2	292.9542	3.6365	292.9542	3.6363	292.9542	3.6363	292.9542	3.6363
rbio2.4	292.9571	3.6421	292.9575	3.6422	292.9575	3.6422	292.9575	3.6422
rbio2.6	292.9567	3.6418	292.9569	3.6422	292.9569	3.6421	292.9569	3.6421
rbio2.8	292.9588	3.6474	292.9589	3.6458	292.9589	3.6457	292.9589	3.6457
rbio3.1	292.9456	3.6352	292.9458	3.6363	292.9457	3.6363	292.9457	3.6363
rbio3.3	292.9515	3.6419	292.9502	3.6398	292.9502	3.6398	292.9502	3.6398
rbio3.5	292.9509	3.6414	292.9478	3.6391	292.9476	3.6394	292.9476	3.6394
rbio3.7	292.9445	3.6374	292.9421	3.6368	292.9423	3.6373	292.9423	3.6373
rbio3.9	292.9430	3.6321	292.9366	3.6274	292.9366	3.6276	292.9366	3.6276
rbio4.4	292.9560	3.6419	292.9558	3.6418	292.9558	3.6418	292.9558	3.6418
rbio5.5	292.9565	3.6393	292.9565	3.6393	292.9565	3.6393	292.9565	3.6393
rbio6.8	292.9564	3.6409	292.9562	3.6409	292.9562	3.6409	292.9562	3.6409
dmey	292.9569	3.6411	292.9566	3.6410	292.9566	3.6410	292.9566	3.6410

Table A.15: Results of NormalShrink with soft thresholding for DSM generation on City Site 2 including outliers

City Site 2, $\mu = 303.4698$, $\sigma = 11.7044$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4698	11.7041	303.4698	11.7040	303.4698	11.7039	303.4698	11.7039
db2	303.4697	11.7029	303.4697	11.7024	303.4697	11.7023	303.4697	11.7023
db4	303.4698	11.7017	303.4698	11.7009	303.4698	11.7007	303.4698	11.7007
db6	303.4698	11.7006	303.4698	11.6994	303.4698	11.6991	303.4698	11.6990
db8	303.4698	11.6991	303.4697	11.6975	303.4697	11.6971	303.4697	11.6970
db10	303.4697	11.6978	303.4697	11.6958	303.4697	11.6952	303.4697	11.6951
db12	303.4698	11.6965	303.4698	11.6940	303.4698	11.6933	303.4698	11.6932
db20	303.4698	11.6914	303.4698	11.6870	303.4698	11.6859	303.4698	11.6856
sym2	303.4697	11.7029	303.4697	11.7024	303.4697	11.7023	303.4697	11.7023
sym4	303.4698	11.7024	303.4698	11.7018	303.4698	11.7017	303.4698	11.7016
sym6	303.4698	11.7019	303.4698	11.7012	303.4698	11.7010	303.4698	11.7010
sym8	303.4698	11.7014	303.4698	11.7006	303.4698	11.7004	303.4698	11.7003
sym10	303.4698	11.7011	303.4698	11.7001	303.4698	11.6998	303.4698	11.6998
sym12	303.4697	11.7006	303.4698	11.6996	303.4698	11.6993	303.4698	11.6992
sym20	303.4697	11.6995	303.4697	11.6981	303.4697	11.6977	303.4697	11.6976
coif1	303.4698	11.7028	303.4698	11.7024	303.4698	11.7022	303.4698	11.7022
coif2	303.4698	11.7024	303.4698	11.7018	303.4698	11.7016	303.4698	11.7016
coif3	303.4698	11.7018	303.4698	11.7011	303.4698	11.7009	303.4698	11.7008
coif4	303.4698	11.7013	303.4698	11.7004	303.4698	11.7002	303.4698	11.7001
coif5	303.4698	11.7009	303.4698	11.6998	303.4698	11.6996	303.4698	11.6995
bior1.1	303.4698	11.7041	303.4698	11.7040	303.4698	11.7039	303.4698	11.7039
bior1.3	303.4698	11.7042	303.4698	11.7042	303.4698	11.7042	303.4698	11.7042
bior1.5	303.4698	11.7042	303.4698	11.7042	303.4698	11.7042	303.4698	11.7042
bior2.2	303.4698	11.7038	303.4698	11.7037	303.4698	11.7037	303.4698	11.7037
bior2.4	303.4698	11.7039	303.4698	11.7038	303.4698	11.7038	303.4698	11.7038
bior2.6	303.4698	11.7039	303.4698	11.7038	303.4698	11.7038	303.4698	11.7038
bior2.8	303.4698	11.7039	303.4698	11.7038	303.4698	11.7038	303.4698	11.7038
bior3.1	303.4698	11.7038	303.4698	11.7038	303.4698	11.7038	303.4698	11.7038
bior3.3	303.4698	11.7038	303.4698	11.7037	303.4698	11.7037	303.4698	11.7037
bior3.5	303.4698	11.7038	303.4698	11.7037	303.4698	11.7037	303.4698	11.7037
bior3.7	303.4698	11.7037	303.4698	11.7037	303.4698	11.7037	303.4698	11.7037
bior3.9	303.4698	11.7037	303.4698	11.7037	303.4698	11.7037	303.4698	11.7037
bior4.4	303.4698	11.7025	303.4698	11.7020	303.4698	11.7018	303.4698	11.7018
bior5.5	303.4698	11.7004	303.4697	11.6983	303.4697	11.6975	303.4697	11.6972
bior6.8	303.4698	11.7022	303.4698	11.7016	303.4698	11.7014	303.4698	11.7014
rbio1.1	303.4698	11.7041	303.4698	11.7040	303.4698	11.7039	303.4698	11.7039
rbio1.3	303.4698	11.7026	303.4698	11.7020	303.4698	11.7018	303.4698	11.7018
rbio1.5	303.4698	11.7021	303.4697	11.7014	303.4698	11.7011	303.4697	11.7010
rbio2.2	303.4698	11.7011	303.4698	11.6997	303.4698	11.6993	303.4698	11.6992
rbio2.4	303.4698	11.7004	303.4698	11.6987	303.4698	11.6981	303.4698	11.6979
rbio2.6	303.4698	11.6997	303.4697	11.6974	303.4697	11.6967	303.4697	11.6964
rbio2.8	303.4698	11.6991	303.4698	11.6964	303.4698	11.6955	303.4698	11.6952
rbio3.1	303.4698	11.7036	303.4697	11.7032	303.4697	11.7030	303.4697	11.7028
rbio3.3	303.4698	11.6991	303.4697	11.6970	303.4696	11.6963	303.4697	11.6961
rbio3.5	303.4698	11.6974	303.4698	11.6941	303.4697	11.6929	303.4697	11.6925
rbio3.7	303.4698	11.6960	303.4698	11.6910	303.4698	11.6894	303.4698	11.6889
rbio3.9	303.4698	11.6950	303.4698	11.6897	303.4698	11.6877	303.4698	11.6871
rbio4.4	303.4698	11.7021	303.4698	11.7015	303.4698	11.7013	303.4698	11.7013
rbio5.5	303.4698	11.7030	303.4698	11.7027	303.4698	11.7027	303.4698	11.7027
rbio6.8	303.4698	11.7011	303.4698	11.7001	303.4698	11.6998	303.4698	11.6997
dmey	303.4698	11.6985	303.4698	11.6966	303.4698	11.6961	303.4698	11.6960

Table A.16: Results of NormalShrink with soft thresholding for DEM generation on City Site 2 including outliers

City Site 2, $\mu = 292.9100$, $\sigma = 3.5970$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.9093	3.5958	292.9098	3.5959	292.9099	3.5962	292.9099	3.5962
db2	292.9117	3.5967	292.9121	3.5966	292.9122	3.5964	292.9121	3.5964
db4	292.9150	3.6000	292.9154	3.5997	292.9155	3.6003	292.9155	3.5999
db6	292.9204	3.6124	292.9255	3.6165	292.9258	3.6167	292.9257	3.6168
db8	292.9145	3.6086	292.9172	3.6092	292.9171	3.6094	292.9170	3.6094
db10	292.9066	3.5895	292.9074	3.5884	292.9065	3.5875	292.9065	3.5875
db12	292.9059	3.5992	292.9050	3.5972	292.9057	3.5971	292.9063	3.5977
db20	292.8462	3.5204	292.8480	3.5172	292.8546	3.5309	292.8552	3.5312
sym2	292.9117	3.5967	292.9121	3.5966	292.9122	3.5964	292.9121	3.5964
sym4	292.9152	3.6023	292.9167	3.6004	292.9159	3.6000	292.9163	3.6001
sym6	292.9139	3.5994	292.9148	3.5992	292.9145	3.5983	292.9146	3.5987
sym8	292.9154	3.6068	292.9168	3.6069	292.9166	3.6064	292.9170	3.6070
sym10	292.9149	3.6058	292.9187	3.6129	292.9186	3.6116	292.9186	3.6116
sym12	292.9172	3.6066	292.9183	3.6057	292.9185	3.6056	292.9183	3.6054
sym20	292.9065	3.5910	292.9146	3.6019	292.9141	3.6005	292.9142	3.6009
coif1	292.9134	3.5992	292.9132	3.5982	292.9140	3.5982	292.9140	3.5982
coif2	292.9139	3.5994	292.9141	3.5993	292.9144	3.5990	292.9145	3.5990
coif3	292.9128	3.5979	292.9136	3.5972	292.9138	3.5977	292.9136	3.5975
coif4	292.9144	3.6077	292.9167	3.6073	292.9169	3.6076	292.9168	3.6075
coif5	292.9183	3.6139	292.9205	3.6141	292.9201	3.6136	292.9202	3.6136
bior1.1	292.9093	3.5958	292.9098	3.5959	292.9099	3.5962	292.9099	3.5962
bior1.3	292.9103	3.5978	292.9094	3.5961	292.9094	3.5961	292.9094	3.5961
bior1.5	292.9096	3.5980	292.9109	3.5979	292.9109	3.5979	292.9109	3.5979
bior2.2	292.9098	3.5981	292.9101	3.5989	292.9103	3.5989	292.9103	3.5989
bior2.4	292.9104	3.5985	292.9104	3.5980	292.9104	3.5981	292.9104	3.5981
bior2.6	292.9111	3.5993	292.9113	3.5998	292.9113	3.5998	292.9113	3.5998
bior2.8	292.9116	3.6003	292.9118	3.6003	292.9118	3.6003	292.9118	3.6003
bior3.1	292.9108	3.5973	292.9116	3.5983	292.9116	3.5983	292.9116	3.5983
bior3.3	292.9122	3.5996	292.9123	3.5996	292.9123	3.5996	292.9123	3.5996
bior3.5	292.9123	3.5996	292.9125	3.5997	292.9123	3.5997	292.9126	3.5997
bior3.7	292.9130	3.6004	292.9126	3.5999	292.9128	3.5999	292.9128	3.5999
bior3.9	292.9129	3.6001	292.9128	3.5999	292.9128	3.5999	292.9128	3.5999
bior4.4	292.9129	3.5989	292.9128	3.5970	292.9130	3.5968	292.9130	3.5968
bior5.5	292.9255	3.6155	292.9301	3.6161	292.9295	3.6144	292.9288	3.6135
bior6.8	292.9138	3.6000	292.9172	3.6057	292.9172	3.6049	292.9172	3.6049
rbio1.1	292.9093	3.5958	292.9098	3.5959	292.9099	3.5962	292.9099	3.5962
rbio1.3	292.9124	3.5982	292.9152	3.5993	292.9146	3.5987	292.9147	3.5987
rbio1.5	292.9169	3.6072	292.9182	3.6079	292.9184	3.6087	292.9186	3.6086
rbio2.2	292.9150	3.5962	292.9163	3.5969	292.9176	3.5969	292.9180	3.5969
rbio2.4	292.9188	3.5979	292.9211	3.6022	292.9200	3.6000	292.9204	3.6003
rbio2.6	292.9166	3.5980	292.9188	3.5992	292.9174	3.5978	292.9167	3.5963
rbio2.8	292.9133	3.5979	292.9216	3.6043	292.9222	3.6039	292.9227	3.6036
rbio3.1	292.8606	3.5418	292.8351	3.5267	292.8426	3.5311	292.8414	3.5305
rbio3.3	292.8976	3.5727	292.8887	3.5764	292.8902	3.5762	292.8897	3.5754
rbio3.5	292.9046	3.5947	292.8883	3.5692	292.8884	3.5685	292.8886	3.5679
rbio3.7	292.9043	3.5985	292.8918	3.5752	292.8928	3.5728	292.8943	3.5743
rbio3.9	292.8890	3.5720	292.8783	3.5535	292.8804	3.5586	292.8809	3.5581
rbio4.4	292.9179	3.6062	292.9190	3.6070	292.9191	3.6072	292.9192	3.6073
rbio5.5	292.9110	3.5967	292.9143	3.6031	292.9142	3.6031	292.9142	3.6031
rbio6.8	292.9157	3.6038	292.9191	3.6064	292.9189	3.6052	292.9187	3.6050
dmey	292.8938	3.5741	292.8949	3.5694	292.8945	3.5692	292.8946	3.5695

Table A.17: Results of NormalShrink with soft thresholding for DSM generation on City Site 2 excluding outliers

City Site 2, $\mu = 303.4662$, $\sigma = 11.5645$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4662	11.5642	303.4662	11.5640	303.4662	11.5640	303.4662	11.5640
db2	303.4662	11.5630	303.4662	11.5626	303.4662	11.5625	303.4662	11.5625
db4	303.4662	11.5622	303.4662	11.5615	303.4662	11.5613	303.4662	11.5613
db6	303.4662	11.5613	303.4662	11.5603	303.4662	11.5600	303.4662	11.5600
db8	303.4662	11.5601	303.4662	11.5588	303.4662	11.5585	303.4662	11.5584
db10	303.4661	11.5591	303.4661	11.5574	303.4661	11.5570	303.4661	11.5569
db12	303.4662	11.5583	303.4662	11.5562	303.4662	11.5557	303.4662	11.5556
db20	303.4662	11.5550	303.4662	11.5517	303.4662	11.5509	303.4662	11.5507
sym2	303.4662	11.5630	303.4662	11.5626	303.4662	11.5625	303.4662	11.5625
sym4	303.4662	11.5628	303.4662	11.5623	303.4662	11.5621	303.4662	11.5621
sym6	303.4662	11.5624	303.4662	11.5618	303.4662	11.5616	303.4662	11.5616
sym8	303.4662	11.5621	303.4662	11.5614	303.4662	11.5612	303.4662	11.5611
sym10	303.4662	11.5617	303.4662	11.5609	303.4662	11.5607	303.4662	11.5607
sym12	303.4662	11.5613	303.4662	11.5604	303.4662	11.5601	303.4662	11.5601
sym20	303.4662	11.5605	303.4662	11.5593	303.4662	11.5590	303.4662	11.5589
coif1	303.4662	11.5631	303.4662	11.5627	303.4662	11.5625	303.4662	11.5625
coif2	303.4662	11.5628	303.4662	11.5623	303.4662	11.5621	303.4662	11.5621
coif3	303.4662	11.5623	303.4662	11.5617	303.4662	11.5615	303.4662	11.5615
coif4	303.4662	11.5619	303.4662	11.5612	303.4662	11.5610	303.4662	11.5609
coif5	303.4662	11.5616	303.4662	11.5607	303.4662	11.5605	303.4662	11.5604
bior1.1	303.4662	11.5642	303.4662	11.5640	303.4662	11.5640	303.4662	11.5640
bior1.3	303.4662	11.5644	303.4662	11.5643	303.4662	11.5643	303.4662	11.5643
bior1.5	303.4662	11.5644	303.4662	11.5643	303.4662	11.5643	303.4662	11.5643
bior2.2	303.4662	11.5639	303.4662	11.5638	303.4662	11.5638	303.4662	11.5638
bior2.4	303.4662	11.5640	303.4662	11.5639	303.4662	11.5639	303.4662	11.5639
bior2.6	303.4662	11.5640	303.4662	11.5639	303.4662	11.5639	303.4662	11.5639
bior2.8	303.4662	11.5640	303.4662	11.5640	303.4662	11.5639	303.4662	11.5639
bior3.1	303.4662	11.5639	303.4662	11.5639	303.4662	11.5639	303.4662	11.5639
bior3.3	303.4662	11.5639	303.4662	11.5639	303.4662	11.5638	303.4662	11.5638
bior3.5	303.4662	11.5639	303.4662	11.5639	303.4662	11.5638	303.4662	11.5638
bior3.7	303.4662	11.5639	303.4662	11.5639	303.4662	11.5639	303.4662	11.5639
bior3.9	303.4662	11.5639	303.4662	11.5639	303.4662	11.5639	303.4662	11.5638
bior4.4	303.4662	11.5629	303.4662	11.5624	303.4662	11.5623	303.4662	11.5623
bior5.5	303.4662	11.5610	303.4662	11.5592	303.4662	11.5585	303.4662	11.5582
bior6.8	303.4662	11.5626	303.4662	11.5621	303.4662	11.5619	303.4662	11.5619
rbio1.1	303.4662	11.5642	303.4662	11.5640	303.4662	11.5640	303.4662	11.5640
rbio1.3	303.4662	11.5629	303.4662	11.5623	303.4662	11.5622	303.4662	11.5621
rbio1.5	303.4662	11.5625	303.4662	11.5618	303.4662	11.5616	303.4662	11.5615
rbio2.2	303.4662	11.5616	303.4662	11.5604	303.4662	11.5600	303.4662	11.5599
rbio2.4	303.4662	11.5612	303.4662	11.5597	303.4662	11.5592	303.4662	11.5591
rbio2.6	303.4662	11.5606	303.4662	11.5587	303.4662	11.5581	303.4662	11.5579
rbio2.8	303.4662	11.5602	303.4662	11.5579	303.4662	11.5571	303.4662	11.5569
rbio3.1	303.4662	11.5642	303.4661	11.5639	303.4661	11.5637	303.4661	11.5636
rbio3.3	303.4662	11.5601	303.4661	11.5582	303.4661	11.5576	303.4661	11.5574
rbio3.5	303.4662	11.5588	303.4662	11.5560	303.4662	11.5549	303.4662	11.5546
rbio3.7	303.4662	11.5577	303.4662	11.5534	303.4662	11.5521	303.4662	11.5516
rbio3.9	303.4662	11.5569	303.4662	11.5525	303.4662	11.5509	303.4662	11.5503
rbio4.4	303.4662	11.5626	303.4662	11.5621	303.4662	11.5619	303.4662	11.5619
rbio5.5	303.4662	11.5632	303.4662	11.5630	303.4662	11.5630	303.4662	11.5630
rbio6.8	303.4662	11.5618	303.4662	11.5610	303.4662	11.5607	303.4662	11.5607
dmey	303.4662	11.5598	303.4662	11.5583	303.4662	11.5579	303.4662	11.5579

Table A.18: Results of NormalShrink with soft thresholding for DEM generation on City Site 2 excluding outliers

City Site 2, $\mu = 292.9549$, $\sigma = 3.6377$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.9558	3.6406	292.9567	3.6413	292.9565	3.6414	292.9566	3.6414
db2	292.9630	3.6495	292.9638	3.6491	292.9636	3.6492	292.9636	3.6492
db4	292.9668	3.6489	292.9670	3.6486	292.9670	3.6487	292.9670	3.6487
db6	292.9627	3.6461	292.9647	3.6433	292.9671	3.6468	292.9671	3.6468
db8	292.9609	3.6485	292.9601	3.6518	292.9600	3.6510	292.9602	3.6512
db10	292.9477	3.6288	292.9423	3.6144	292.9448	3.6175	292.9447	3.6175
db12	292.9423	3.6213	292.9472	3.6248	292.9454	3.6213	292.9457	3.6206
db20	292.9165	3.5936	292.9241	3.5982	292.9241	3.5968	292.9239	3.5966
sym2	292.9630	3.6495	292.9638	3.6491	292.9636	3.6492	292.9636	3.6492
sym4	292.9637	3.6497	292.9642	3.6497	292.9644	3.6495	292.9645	3.6496
sym6	292.9647	3.6501	292.9673	3.6524	292.9673	3.6520	292.9674	3.6519
sym8	292.9632	3.6482	292.9680	3.6511	292.9677	3.6504	292.9680	3.6504
sym10	292.9676	3.6539	292.9681	3.6514	292.9683	3.6510	292.9682	3.6509
sym12	292.9620	3.6497	292.9641	3.6522	292.9646	3.6518	292.9646	3.6518
sym20	292.9480	3.6244	292.9489	3.6222	292.9493	3.6224	292.9493	3.6223
coif1	292.9621	3.6455	292.9632	3.6483	292.9636	3.6485	292.9637	3.6486
coif2	292.9628	3.6462	292.9643	3.6493	292.9644	3.6489	292.9644	3.6489
coif3	292.9657	3.6494	292.9673	3.6517	292.9682	3.6517	292.9682	3.6517
coif4	292.9646	3.6480	292.9664	3.6527	292.9667	3.6523	292.9667	3.6522
coif5	292.9648	3.6493	292.9648	3.6497	292.9651	3.6500	292.9650	3.6497
bior1.1	292.9558	3.6406	292.9567	3.6413	292.9565	3.6414	292.9566	3.6414
bior1.3	292.9566	3.6414	292.9555	3.6395	292.9555	3.6394	292.9554	3.6394
bior1.5	292.9555	3.6410	292.9567	3.6412	292.9569	3.6413	292.9569	3.6413
bior2.2	292.9582	3.6422	292.9584	3.6427	292.9584	3.6427	292.9584	3.6427
bior2.4	292.9585	3.6432	292.9584	3.6426	292.9584	3.6426	292.9584	3.6426
bior2.6	292.9575	3.6410	292.9571	3.6405	292.9571	3.6405	292.9571	3.6405
bior2.8	292.9575	3.6409	292.9576	3.6411	292.9576	3.6411	292.9576	3.6411
bior3.1	292.9589	3.6433	292.9593	3.6436	292.9592	3.6436	292.9592	3.6436
bior3.3	292.9580	3.6409	292.9579	3.6414	292.9578	3.6414	292.9578	3.6414
bior3.5	292.9588	3.6425	292.9590	3.6426	292.9589	3.6426	292.9587	3.6427
bior3.7	292.9589	3.6420	292.9589	3.6418	292.9589	3.6418	292.9589	3.6418
bior3.9	292.9591	3.6422	292.9593	3.6421	292.9593	3.6421	292.9593	3.6421
bior4.4	292.9637	3.6468	292.9639	3.6483	292.9640	3.6482	292.9641	3.6482
bior5.5	292.9736	3.6641	292.9797	3.6679	292.9789	3.6660	292.9792	3.6658
bior6.8	292.9643	3.6505	292.9668	3.6517	292.9674	3.6522	292.9675	3.6523
rbio1.1	292.9558	3.6406	292.9567	3.6413	292.9565	3.6414	292.9566	3.6414
rbio1.3	292.9637	3.6511	292.9695	3.6586	292.9696	3.6589	292.9697	3.6589
rbio1.5	292.9648	3.6528	292.9658	3.6522	292.9660	3.6520	292.9661	3.6521
rbio2.2	292.9670	3.6472	292.9674	3.6467	292.9670	3.6466	292.9671	3.6465
rbio2.4	292.9710	3.6582	292.9729	3.6605	292.9760	3.6643	292.9760	3.6643
rbio2.6	292.9690	3.6553	292.9701	3.6546	292.9685	3.6544	292.9689	3.6547
rbio2.8	292.9640	3.6506	292.9682	3.6480	292.9695	3.6491	292.9697	3.6491
rbio3.1	292.9231	3.6101	292.8970	3.5927	292.9022	3.5891	292.9013	3.5922
rbio3.3	292.9522	3.6389	292.9409	3.6261	292.9428	3.6303	292.9432	3.6313
rbio3.5	292.9463	3.6261	292.9403	3.6201	292.9416	3.6189	292.9415	3.6187
rbio3.7	292.9454	3.6195	292.9382	3.6178	292.9380	3.6173	292.9384	3.6182
rbio3.9	292.9481	3.6321	292.9278	3.5988	292.9215	3.5896	292.9218	3.5900
rbio4.4	292.9669	3.6545	292.9669	3.6542	292.9671	3.6542	292.9670	3.6542
rbio5.5	292.9668	3.6533	292.9622	3.6434	292.9622	3.6434	292.9622	3.6434
rbio6.8	292.9673	3.6494	292.9661	3.6449	292.9667	3.6453	292.9665	3.6453
dmey	292.9469	3.6262	292.9482	3.6258	292.9485	3.6258	292.9483	3.6261

Table A.19: Results of SureShrink with hard thresholding for DSM generation on City Site 3 including outliers

City Site 3, $\mu = 318.7184$, $\sigma = 9.1404$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.7184	9.1382	318.7184	9.1381	318.7184	9.1381	318.7184	9.1381
db2	318.7184	9.1369	318.7184	9.1366	318.7184	9.1366	318.7184	9.1366
db4	318.7184	9.1358	318.7184	9.1355	318.7184	9.1355	318.7184	9.1355
db6	318.7183	9.1352	318.7184	9.1350	318.7183	9.1350	318.7183	9.1350
db8	318.7184	9.1345	318.7184	9.1343	318.7184	9.1343	318.7184	9.1343
db10	318.7183	9.1338	318.7183	9.1336	318.7183	9.1336	318.7183	9.1336
db12	318.7184	9.1340	318.7184	9.1338	318.7184	9.1338	318.7184	9.1338
db20	318.7183	9.1329	318.7183	9.1327	318.7183	9.1327	318.7183	9.1327
sym2	318.7184	9.1369	318.7184	9.1366	318.7184	9.1366	318.7184	9.1366
sym4	318.7184	9.1364	318.7184	9.1361	318.7184	9.1361	318.7184	9.1361
sym6	318.7184	9.1358	318.7184	9.1355	318.7184	9.1355	318.7184	9.1355
sym8	318.7184	9.1355	318.7184	9.1352	318.7184	9.1352	318.7184	9.1352
sym10	318.7184	9.1353	318.7184	9.1350	318.7184	9.1350	318.7184	9.1350
sym12	318.7184	9.1349	318.7184	9.1346	318.7184	9.1346	318.7184	9.1346
sym20	318.7184	9.1341	318.7184	9.1338	318.7184	9.1338	318.7184	9.1338
coif1	318.7185	9.1371	318.7185	9.1368	318.7185	9.1368	318.7185	9.1368
coif2	318.7183	9.1358	318.7183	9.1356	318.7183	9.1355	318.7183	9.1355
coif3	318.7184	9.1358	318.7184	9.1354	318.7184	9.1354	318.7184	9.1354
coif4	318.7184	9.1351	318.7184	9.1348	318.7184	9.1348	318.7184	9.1348
coif5	318.7184	9.1351	318.7184	9.1348	318.7184	9.1348	318.7184	9.1348
bior1.1	318.7184	9.1382	318.7184	9.1381	318.7184	9.1381	318.7184	9.1381
bior1.3	318.7184	9.1407	318.7184	9.1406	318.7184	9.1406	318.7184	9.1406
bior1.5	318.7184	9.1411	318.7184	9.1411	318.7184	9.1411	318.7184	9.1411
bior2.2	318.7184	9.1324	318.7184	9.1323	318.7184	9.1323	318.7184	9.1323
bior2.4	318.7184	9.1340	318.7184	9.1339	318.7184	9.1339	318.7184	9.1339
bior2.6	318.7184	9.1349	318.7184	9.1348	318.7184	9.1348	318.7184	9.1348
bior2.8	318.7184	9.1355	318.7184	9.1355	318.7184	9.1355	318.7184	9.1355
bior3.1	318.7184	9.1396	318.7184	9.1397	318.7184	9.1397	318.7184	9.1397
bior3.3	318.7184	9.1332	318.7184	9.1333	318.7184	9.1333	318.7184	9.1333
bior3.5	318.7184	9.1340	318.7184	9.1341	318.7184	9.1341	318.7184	9.1341
bior3.7	318.7184	9.1340	318.7184	9.1340	318.7184	9.1340	318.7184	9.1340
bior3.9	318.7184	9.1337	318.7184	9.1338	318.7184	9.1338	318.7184	9.1338
bior4.4	318.7183	9.1351	318.7183	9.1347	318.7183	9.1347	318.7183	9.1347
bior5.5	318.7184	9.1380	318.7184	9.1376	318.7184	9.1376	318.7184	9.1376
bior6.8	318.7184	9.1348	318.7184	9.1344	318.7184	9.1344	318.7184	9.1344
rbio1.1	318.7184	9.1382	318.7184	9.1381	318.7184	9.1381	318.7184	9.1381
rbio1.3	318.7184	9.1377	318.7184	9.1375	318.7184	9.1375	318.7184	9.1375
rbio1.5	318.7184	9.1383	318.7184	9.1381	318.7184	9.1381	318.7184	9.1381
rbio2.2	318.7184	9.1405	318.7184	9.1406	318.7184	9.1406	318.7184	9.1406
rbio2.4	318.7184	9.1397	318.7184	9.1399	318.7184	9.1399	318.7184	9.1399
rbio2.6	318.7184	9.1396	318.7184	9.1397	318.7184	9.1397	318.7184	9.1397
rbio2.8	318.7184	9.1397	318.7184	9.1398	318.7184	9.1398	318.7184	9.1398
rbio3.1	318.7184	9.1409	318.7184	9.1411	318.7184	9.1411	318.7184	9.1411
rbio3.3	318.7184	9.1413	318.7184	9.1417	318.7184	9.1417	318.7184	9.1417
rbio3.5	318.7184	9.1412	318.7184	9.1417	318.7184	9.1417	318.7184	9.1417
rbio3.7	318.7184	9.1411	318.7184	9.1416	318.7184	9.1416	318.7184	9.1417
rbio3.9	318.7184	9.1411	318.7184	9.1417	318.7184	9.1417	318.7184	9.1417
rbio4.4	318.7183	9.1382	318.7183	9.1382	318.7183	9.1382	318.7183	9.1382
rbio5.5	318.7186	9.1362	318.7186	9.1362	318.7186	9.1362	318.7186	9.1362
rbio6.8	318.7184	9.1370	318.7184	9.1369	318.7184	9.1369	318.7184	9.1369
dmey	318.7184	9.1338	318.7184	9.1335	318.7184	9.1335	318.7184	9.1335

Table A.20: Results of SureShrink with hard thresholding for DEM generation on City Site 3 including outliers

City Site 3, $\mu = 310.7064$, $\sigma = 4.4727$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.7066	4.4400	310.7275	4.4415	310.7294	4.4446	310.7300	4.4439
db2	310.6330	4.3917	310.6603	4.4045	310.6661	4.4145	310.6666	4.4138
db4	310.6474	4.4423	310.6318	4.4103	310.6372	4.4078	310.6370	4.4079
db6	310.5857	4.3930	310.5698	4.3563	310.5698	4.3547	310.5696	4.3547
db8	310.5885	4.3969	310.5933	4.4045	310.5942	4.4051	310.5941	4.4049
db10	310.5831	4.3948	310.5761	4.3997	310.5768	4.3973	310.5761	4.3971
db12	310.5903	4.4110	310.5791	4.4171	310.5786	4.4183	310.5788	4.4183
db20	310.5053	4.3692	310.5099	4.3705	310.5087	4.3686	310.5084	4.3684
sym2	310.6330	4.3917	310.6603	4.4045	310.6661	4.4145	310.6666	4.4138
sym4	310.6429	4.4222	310.6444	4.3970	310.6448	4.3958	310.6439	4.3947
sym6	310.6324	4.4211	310.6290	4.4125	310.6293	4.4134	310.6297	4.4134
sym8	310.6024	4.3850	310.6161	4.3918	310.6192	4.3985	310.6202	4.3987
sym10	310.6033	4.4040	310.6062	4.3938	310.6027	4.3918	310.6016	4.3906
sym12	310.5680	4.3918	310.5763	4.4068	310.5741	4.4042	310.5740	4.4041
sym20	310.5412	4.3777	310.5625	4.4042	310.5686	4.4013	310.5681	4.4011
coif1	310.6204	4.3839	310.6395	4.3822	310.6430	4.3858	310.6429	4.3860
coif2	310.6466	4.4394	310.6332	4.4002	310.6357	4.4028	310.6359	4.4028
coif3	310.6244	4.4201	310.6330	4.4276	310.6316	4.4223	310.6318	4.4222
coif4	310.5848	4.3761	310.5968	4.3758	310.5924	4.3737	310.5930	4.3737
coif5	310.5588	4.3735	310.5928	4.3975	310.5908	4.3970	310.5908	4.3969
bior1.1	310.7066	4.4400	310.7275	4.4415	310.7294	4.4446	310.7300	4.4439
bior1.3	310.6720	4.4279	310.6724	4.4167	310.6752	4.4244	310.6749	4.4248
bior1.5	310.6542	4.4163	310.6626	4.4378	310.6655	4.4380	310.6648	4.4376
bior2.2	310.7110	4.4784	310.6907	4.4386	310.6947	4.4446	310.6917	4.4446
bior2.4	310.7079	4.4872	310.7049	4.4687	310.7030	4.4681	310.7028	4.4681
bior2.6	310.6708	4.4494	310.6614	4.4206	310.6639	4.4244	310.6639	4.4248
bior2.8	310.6570	4.4466	310.6772	4.4431	310.6741	4.4442	310.6740	4.4444
bior3.1	310.5993	4.3919	310.5937	4.3840	310.5928	4.3833	310.5928	4.3833
bior3.3	310.6154	4.3978	310.6167	4.4060	310.6173	4.4060	310.6173	4.4061
bior3.5	310.6298	4.4228	310.6128	4.4089	310.6127	4.4096	310.6127	4.4095
bior3.7	310.6127	4.3975	310.6131	4.3866	310.6130	4.3863	310.6129	4.3863
bior3.9	310.6266	4.4193	310.6323	4.4308	310.6331	4.4312	310.6331	4.4312
bior4.4	310.6475	4.4291	310.6502	4.4149	310.6569	4.4211	310.6585	4.4219
bior5.5	310.6272	4.4540	310.6276	4.4188	310.6240	4.4212	310.6257	4.4217
bior6.8	310.6012	4.4099	310.6099	4.4052	310.6076	4.4046	310.6072	4.4044
rbio1.1	310.7066	4.4400	310.7275	4.4415	310.7294	4.4446	310.7300	4.4439
rbio1.3	310.6554	4.4149	310.6436	4.4270	310.6483	4.4261	310.6477	4.4260
rbio1.5	310.6558	4.4482	310.6426	4.4317	310.6406	4.4295	310.6418	4.4301
rbio2.2	310.6125	4.3948	310.6057	4.3671	310.6099	4.3658	310.6076	4.3672
rbio2.4	310.6019	4.3960	310.6144	4.4346	310.6200	4.4340	310.6202	4.4357
rbio2.6	310.5881	4.3711	310.5880	4.3721	310.5846	4.3701	310.5873	4.3714
rbio2.8	310.5939	4.3815	310.6093	4.3974	310.6047	4.3879	310.6041	4.3890
rbio3.1	310.6024	4.4000	310.6733	4.4615	310.6871	4.4667	310.6871	4.4667
rbio3.3	310.6177	4.4242	310.6155	4.5236	310.6188	4.5233	310.6231	4.5241
rbio3.5	310.6399	4.4191	310.6213	4.4471	310.6263	4.4544	310.6218	4.4515
rbio3.7	310.6028	4.3931	310.5816	4.4258	310.5918	4.4386	310.5906	4.4393
rbio3.9	310.6017	4.4047	310.5915	4.4249	310.5922	4.4297	310.5922	4.4290
rbio4.4	310.6171	4.4214	310.6160	4.4114	310.6167	4.4103	310.6163	4.4109
rbio5.5	310.5859	4.3930	310.5908	4.3874	310.5882	4.3872	310.5876	4.3876
rbio6.8	310.6021	4.3954	310.5974	4.3887	310.6037	4.3904	310.6035	4.3908
dmey	310.5357	4.3695	310.5287	4.3719	310.5274	4.3660	310.5275	4.3661

Table A.21: Results of SureShrink with hard thresholding for DSM generation on City Site 3 excluding outliers

City Site 3, $\mu = 318.6986$, $\sigma = 9.0395$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.6986	9.0364	318.6986	9.0362	318.6986	9.0362	318.6986	9.0362
db2	318.6986	9.0352	318.6986	9.0350	318.6986	9.0350	318.6986	9.0350
db4	318.6986	9.0335	318.6986	9.0332	318.6987	9.0332	318.6986	9.0332
db6	318.6985	9.0330	318.6985	9.0327	318.6985	9.0327	318.6985	9.0327
db8	318.6986	9.0325	318.6987	9.0323	318.6987	9.0322	318.6987	9.0322
db10	318.6985	9.0318	318.6985	9.0316	318.6985	9.0316	318.6985	9.0316
db12	318.6986	9.0320	318.6986	9.0317	318.6986	9.0317	318.6986	9.0317
db20	318.6985	9.0301	318.6985	9.0300	318.6985	9.0300	318.6985	9.0300
sym2	318.6986	9.0352	318.6986	9.0350	318.6986	9.0350	318.6986	9.0350
sym4	318.6986	9.0345	318.6986	9.0341	318.6986	9.0341	318.6986	9.0341
sym6	318.6986	9.0340	318.6986	9.0338	318.6986	9.0338	318.6986	9.0338
sym8	318.6986	9.0338	318.6986	9.0334	318.6986	9.0334	318.6986	9.0334
sym10	318.6985	9.0334	318.6986	9.0332	318.6986	9.0332	318.6986	9.0332
sym12	318.6986	9.0328	318.6986	9.0325	318.6986	9.0325	318.6986	9.0325
sym20	318.6986	9.0323	318.6986	9.0320	318.6986	9.0320	318.6986	9.0320
coif1	318.6986	9.0356	318.6986	9.0353	318.6986	9.0353	318.6986	9.0353
coif2	318.6985	9.0347	318.6986	9.0344	318.6986	9.0344	318.6986	9.0344
coif3	318.6986	9.0342	318.6986	9.0338	318.6986	9.0338	318.6986	9.0338
coif4	318.6986	9.0337	318.6986	9.0334	318.6985	9.0333	318.6985	9.0333
coif5	318.6986	9.0334	318.6986	9.0330	318.6986	9.0330	318.6986	9.0330
bior1.1	318.6986	9.0364	318.6986	9.0362	318.6986	9.0362	318.6986	9.0362
bior1.3	318.6986	9.0397	318.6986	9.0396	318.6986	9.0396	318.6986	9.0396
bior1.5	318.6986	9.0401	318.6986	9.0401	318.6986	9.0401	318.6986	9.0401
bior2.2	318.6986	9.0303	318.6986	9.0303	318.6986	9.0302	318.6986	9.0302
bior2.4	318.6986	9.0323	318.6986	9.0322	318.6986	9.0322	318.6986	9.0322
bior2.6	318.6986	9.0334	318.6986	9.0334	318.6986	9.0334	318.6986	9.0334
bior2.8	318.6986	9.0341	318.6986	9.0340	318.6986	9.0340	318.6986	9.0340
bior3.1	318.6986	9.0361	318.6986	9.0362	318.6986	9.0362	318.6986	9.0362
bior3.3	318.6986	9.0314	318.6986	9.0315	318.6986	9.0315	318.6986	9.0315
bior3.5	318.6986	9.0311	318.6986	9.0312	318.6986	9.0312	318.6986	9.0312
bior3.7	318.6986	9.0317	318.6986	9.0318	318.6986	9.0318	318.6986	9.0318
bior3.9	318.6986	9.0319	318.6986	9.0320	318.6986	9.0320	318.6986	9.0320
bior4.4	318.6985	9.0325	318.6985	9.0322	318.6985	9.0322	318.6985	9.0322
bior5.5	318.6986	9.0361	318.6986	9.0357	318.6986	9.0357	318.6986	9.0357
bior6.8	318.6985	9.0324	318.6985	9.0320	318.6985	9.0320	318.6985	9.0320
rbio1.1	318.6986	9.0364	318.6986	9.0362	318.6986	9.0362	318.6986	9.0362
rbio1.3	318.6986	9.0356	318.6986	9.0354	318.6986	9.0354	318.6986	9.0354
rbio1.5	318.6986	9.0360	318.6986	9.0358	318.6986	9.0358	318.6986	9.0358
rbio2.2	318.6986	9.0396	318.6986	9.0397	318.6986	9.0397	318.6986	9.0397
rbio2.4	318.6986	9.0386	318.6986	9.0388	318.6986	9.0388	318.6986	9.0388
rbio2.6	318.6986	9.0385	318.6986	9.0386	318.6986	9.0386	318.6986	9.0386
rbio2.8	318.6986	9.0385	318.6986	9.0387	318.6986	9.0387	318.6986	9.0387
rbio3.1	318.6986	9.0401	318.6986	9.0403	318.6986	9.0403	318.6986	9.0403
rbio3.3	318.6986	9.0406	318.6986	9.0409	318.6986	9.0409	318.6986	9.0409
rbio3.5	318.6986	9.0403	318.6986	9.0408	318.6986	9.0408	318.6986	9.0408
rbio3.7	318.6986	9.0403	318.6986	9.0408	318.6986	9.0408	318.6986	9.0408
rbio3.9	318.6986	9.0403	318.6986	9.0409	318.6986	9.0410	318.6986	9.0410
rbio4.4	318.6985	9.0368	318.6985	9.0368	318.6985	9.0368	318.6985	9.0368
rbio5.5	318.6987	9.0349	318.6987	9.0349	318.6987	9.0349	318.6987	9.0349
rbio6.8	318.6986	9.0357	318.6986	9.0355	318.6986	9.0355	318.6986	9.0355
dmey	318.6986	9.0319	318.6986	9.0316	318.6986	9.0316	318.6986	9.0317

Table A.22: Results of SureShrink with hard thresholding for DEM generation on City Site 3 excluding outliers

City Site 3, $\mu = 310.7514$, $\sigma = 4.4699$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.7710	4.4665	310.7698	4.4491	310.7714	4.4518	310.7713	4.4520
db2	310.7452	4.4547	310.7442	4.4322	310.7459	4.4317	310.7456	4.4312
db4	310.7166	4.4155	310.7197	4.4252	310.7211	4.4249	310.7213	4.4248
db6	310.6540	4.3761	310.6825	4.3911	310.6823	4.3915	310.6815	4.3910
db8	310.6653	4.4057	310.6650	4.4067	310.6663	4.4071	310.6665	4.4076
db10	310.6186	4.3915	310.6427	4.4217	310.6346	4.4190	310.6347	4.4185
db12	310.6277	4.4046	310.6444	4.4209	310.6437	4.4235	310.6432	4.4235
db20	310.5889	4.3676	310.6144	4.3945	310.6124	4.3941	310.6135	4.3950
sym2	310.7452	4.4547	310.7442	4.4322	310.7459	4.4317	310.7456	4.4312
sym4	310.6757	4.4181	310.6922	4.3929	310.6964	4.3933	310.6966	4.3933
sym6	310.6662	4.3876	310.6867	4.3924	310.6898	4.3941	310.6910	4.3936
sym8	310.6581	4.3898	310.6723	4.3793	310.6709	4.3773	310.6710	4.3771
sym10	310.6596	4.3928	310.6804	4.4007	310.6797	4.3976	310.6781	4.3974
sym12	310.6564	4.4100	310.6817	4.4183	310.6844	4.4250	310.6841	4.4243
sym20	310.6301	4.3988	310.6493	4.4265	310.6514	4.4276	310.6508	4.4275
coif1	310.7063	4.4325	310.7257	4.4251	310.7311	4.4315	310.7319	4.4325
coif2	310.6829	4.4189	310.7072	4.4139	310.7076	4.4135	310.7081	4.4127
coif3	310.6700	4.4012	310.6816	4.4012	310.6858	4.4005	310.6860	4.4001
coif4	310.6591	4.3791	310.6622	4.3929	310.6615	4.3936	310.6615	4.3939
coif5	310.6550	4.3844	310.6515	4.3923	310.6577	4.3914	310.6574	4.3916
bior1.1	310.7710	4.4665	310.7698	4.4491	310.7714	4.4518	310.7713	4.4520
bior1.3	310.7305	4.4407	310.7280	4.4293	310.7354	4.4294	310.7355	4.4295
bior1.5	310.7230	4.4488	310.7156	4.4488	310.7156	4.4456	310.7159	4.4454
bior2.2	310.7328	4.4251	310.7268	4.4181	310.7283	4.4159	310.7285	4.4160
bior2.4	310.7181	4.4279	310.7107	4.4126	310.7128	4.4159	310.7125	4.4163
bior2.6	310.7090	4.4300	310.7192	4.4446	310.7252	4.4456	310.7225	4.4451
bior2.8	310.7253	4.4448	310.7038	4.4147	310.7014	4.4142	310.7011	4.4144
bior3.1	310.6692	4.3656	310.6505	4.3455	310.6511	4.3456	310.6511	4.3456
bior3.3	310.6659	4.3663	310.6717	4.3784	310.6711	4.3759	310.6711	4.3759
bior3.5	310.6900	4.4012	310.6937	4.4100	310.6908	4.4087	310.6908	4.4087
bior3.7	310.6689	4.3683	310.6757	4.3702	310.6757	4.3702	310.6757	4.3702
bior3.9	310.6890	4.4123	310.6921	4.4142	310.6942	4.4170	310.6943	4.4170
bior4.4	310.6606	4.3860	310.6775	4.3793	310.6799	4.3837	310.6786	4.3825
bior5.5	310.6917	4.4423	310.7285	4.4419	310.7295	4.4290	310.7303	4.4296
bior6.8	310.6711	4.4008	310.6737	4.3959	310.6769	4.3958	310.6779	4.3965
rbio1.1	310.7710	4.4665	310.7698	4.4491	310.7714	4.4518	310.7713	4.4520
rbio1.3	310.7183	4.4050	310.7145	4.4019	310.7192	4.4093	310.7188	4.4091
rbio1.5	310.7058	4.4138	310.7005	4.4134	310.6916	4.4090	310.6927	4.4090
rbio2.2	310.6548	4.3641	310.6769	4.3824	310.6782	4.3896	310.6798	4.3897
rbio2.4	310.6904	4.4233	310.6879	4.4111	310.6896	4.4143	310.6903	4.4142
rbio2.6	310.6552	4.3816	310.6659	4.3746	310.6714	4.3776	310.6698	4.3757
rbio2.8	310.6323	4.3619	310.6360	4.3475	310.6354	4.3502	310.6353	4.3530
rbio3.1	310.6728	4.4231	310.7120	4.4538	310.7116	4.4603	310.7112	4.4604
rbio3.3	310.6456	4.4027	310.6519	4.4783	310.6608	4.4927	310.6616	4.4911
rbio3.5	310.6681	4.4273	310.6802	4.4579	310.6964	4.4706	310.6934	4.4632
rbio3.7	310.6938	4.4489	310.6587	4.4794	310.6617	4.4798	310.6626	4.4816
rbio3.9	310.6769	4.4239	310.6804	4.4147	310.6887	4.4152	310.6907	4.4173
rbio4.4	310.6884	4.4216	310.7030	4.4375	310.7089	4.4383	310.7083	4.4360
rbio5.5	310.6612	4.4205	310.6574	4.4133	310.6583	4.4126	310.6579	4.4128
rbio6.8	310.6389	4.3733	310.6394	4.3663	310.6444	4.3702	310.6407	4.3692
Dmey	310.6285	4.3795	310.6412	4.3924	310.6436	4.3931	310.6432	4.3931

Table A.23: Results of SureShrink with soft thresholding for DSM generation on City Site 3 including outliers

City Site 3, $\mu = 318.7184$, $\sigma = 9.1404$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.7184	9.0984	318.7184	9.0776	318.7184	9.0656	318.7184	9.0639
db2	318.7183	9.1037	318.7183	9.0833	318.7184	9.0754	318.7184	9.0739
db4	318.7186	9.1049	318.7185	9.0837	318.7185	9.0767	318.7185	9.0726
db6	318.7181	9.1034	318.7182	9.0838	318.7182	9.0789	318.7182	9.0780
db8	318.7184	9.1029	318.7186	9.0823	318.7186	9.0764	318.7186	9.0745
db10	318.7184	9.1026	318.7183	9.0830	318.7183	9.0795	318.7183	9.0785
db12	318.7183	9.1023	318.7184	9.0846	318.7184	9.0803	318.7184	9.0791
db20	318.7182	9.1003	318.7182	9.0821	318.7182	9.0785	318.7182	9.0773
sym2	318.7183	9.1037	318.7183	9.0833	318.7184	9.0754	318.7184	9.0739
sym4	318.7183	9.1055	318.7179	9.0837	318.7180	9.0747	318.7180	9.0721
sym6	318.7182	9.1053	318.7185	9.0852	318.7184	9.0773	318.7185	9.0741
sym8	318.7183	9.1049	318.7180	9.0850	318.7180	9.0781	318.7180	9.0761
sym10	318.7183	9.1052	318.7185	9.0843	318.7186	9.0765	318.7185	9.0746
sym12	318.7185	9.1049	318.7183	9.0840	318.7183	9.0760	318.7183	9.0746
sym20	318.7185	9.1038	318.7184	9.0833	318.7184	9.0780	318.7184	9.0763
coif1	318.7186	9.1040	318.7183	9.0825	318.7184	9.0733	318.7185	9.0687
coif2	318.7182	9.1043	318.7179	9.0823	318.7178	9.0737	318.7178	9.0701
coif3	318.7182	9.1052	318.7179	9.0833	318.7180	9.0744	318.7180	9.0724
coif4	318.7182	9.1041	318.7180	9.0839	318.7179	9.0776	318.7179	9.0759
coif5	318.7182	9.1040	318.7180	9.0830	318.7180	9.0761	318.7180	9.0743
bior1.1	318.7184	9.0984	318.7184	9.0776	318.7184	9.0656	318.7184	9.0639
bior1.3	318.7184	9.1192	318.7183	9.1053	318.7183	9.1005	318.7183	9.0997
bior1.5	318.7184	9.1236	318.7184	9.1117	318.7183	9.1068	318.7183	9.1044
bior2.2	318.7184	9.0942	318.7182	9.0724	318.7182	9.0644	318.7182	9.0633
bior2.4	318.7184	9.1011	318.7187	9.0848	318.7187	9.0802	318.7187	9.0782
bior2.6	318.7184	9.1045	318.7182	9.0889	318.7182	9.0840	318.7182	9.0819
bior2.8	318.7184	9.1069	318.7186	9.0940	318.7186	9.0903	318.7186	9.0891
bior3.1	318.7184	9.1048	318.7184	9.0969	318.7184	9.0955	318.7184	9.0953
bior3.3	318.7184	9.1000	318.7184	9.0912	318.7184	9.0892	318.7184	9.0889
bior3.5	318.7184	9.1020	318.7183	9.0928	318.7184	9.0915	318.7183	9.0906
bior3.7	318.7184	9.1026	318.7185	9.0943	318.7185	9.0927	318.7185	9.0920
bior3.9	318.7184	9.1024	318.7183	9.0937	318.7183	9.0924	318.7183	9.0919
bior4.4	318.7181	9.1022	318.7185	9.0779	318.7185	9.0673	318.7185	9.0653
bior5.5	318.7186	9.1089	318.7189	9.0811	318.7188	9.0683	318.7188	9.0620
bior6.8	318.7181	9.1042	318.7184	9.0824	318.7184	9.0749	318.7184	9.0724
rbio1.1	318.7184	9.0984	318.7184	9.0776	318.7184	9.0656	318.7184	9.0639
rbio1.3	318.7184	9.1026	318.7179	9.0789	318.7178	9.0689	318.7178	9.0671
rbio1.5	318.7184	9.1076	318.7185	9.0847	318.7183	9.0753	318.7183	9.0711
rbio2.2	318.7186	9.1164	318.7184	9.1005	318.7185	9.0949	318.7185	9.0918
rbio2.4	318.7182	9.1106	318.7184	9.0923	318.7184	9.0852	318.7185	9.0807
rbio2.6	318.7182	9.1108	318.7180	9.0884	318.7179	9.0805	318.7179	9.0743
rbio2.8	318.7182	9.1114	318.7185	9.0909	318.7184	9.0825	318.7184	9.0749
rbio3.1	318.7184	9.1389	318.7184	9.1385	318.7183	9.1378	318.7183	9.1374
rbio3.3	318.7184	9.1263	318.7184	9.1195	318.7184	9.1152	318.7184	9.1147
rbio3.5	318.7184	9.1223	318.7183	9.1104	318.7183	9.1063	318.7183	9.1032
rbio3.7	318.7184	9.1205	318.7186	9.1068	318.7186	9.1010	318.7186	9.0980
rbio3.9	318.7184	9.1198	318.7182	9.1042	318.7182	9.0968	318.7182	9.0944
rbio4.4	318.7182	9.1096	318.7184	9.0935	318.7184	9.0878	318.7184	9.0850
rbio5.5	318.7186	9.1085	318.7187	9.0955	318.7187	9.0930	318.7187	9.0903
rbio6.8	318.7182	9.1071	318.7184	9.0886	318.7184	9.0827	318.7184	9.0782
dmey	318.7183	9.1028	318.7181	9.0815	318.7181	9.0763	318.7181	9.0751

Table A.24: Results of SureShrink with soft thresholding for DEM generation on City Site 3 including outliers

City Site 3, $\mu = 310.7064$, $\sigma = 4.4727$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.7818	4.5063	310.8721	4.5024	310.8814	4.5014	310.8959	4.5157
db2	310.6728	4.4292	310.7617	4.4678	310.7592	4.4460	310.7578	4.4400
db4	310.6228	4.4172	310.7000	4.4426	310.7082	4.4577	310.7228	4.4693
db6	310.6185	4.4291	310.6996	4.4596	310.7112	4.4640	310.7107	4.4608
db8	310.5983	4.4265	310.6689	4.4514	310.6705	4.4538	310.6774	4.4576
db10	310.5331	4.3537	310.6471	4.4562	310.6519	4.4490	310.6532	4.4535
db12	310.5620	4.3973	310.6269	4.4661	310.6458	4.4742	310.6476	4.4756
db20	310.5329	4.4003	310.5939	4.4698	310.5950	4.4619	310.5960	4.4575
sym2	310.6728	4.4292	310.7617	4.4678	310.7592	4.4460	310.7578	4.4400
sym4	310.6269	4.3925	310.7174	4.4671	310.7314	4.4751	310.7303	4.4715
sym6	310.6156	4.4059	310.6893	4.4282	310.7216	4.4574	310.7248	4.4548
sym8	310.6076	4.4041	310.6887	4.4484	310.7076	4.4434	310.7120	4.4425
sym10	310.5870	4.3887	310.6744	4.4475	310.7073	4.4805	310.7000	4.4652
sym12	310.5620	4.3891	310.6226	4.4013	310.6387	4.4080	310.6461	4.4139
sym20	310.5446	4.3845	310.6158	4.4046	310.6233	4.4080	310.6255	4.4093
coif1	310.6389	4.3832	310.7263	4.4565	310.7444	4.4571	310.7354	4.4514
coif2	310.6453	4.4206	310.7209	4.4717	310.7285	4.4764	310.7379	4.4864
coif3	310.6128	4.3965	310.6961	4.4625	310.7052	4.4631	310.7109	4.4613
coif4	310.5852	4.3890	310.6698	4.4427	310.6977	4.4709	310.6992	4.4731
coif5	310.5660	4.3725	310.6856	4.4612	310.6866	4.4743	310.6896	4.4735
bior1.1	310.7818	4.5063	310.8721	4.5024	310.8814	4.5014	310.8959	4.5157
bior1.3	310.7203	4.4728	310.7759	4.4768	310.7954	4.4749	310.7929	4.4670
bior1.5	310.6974	4.4565	310.7733	4.4950	310.7999	4.5046	310.7896	4.4838
bior2.2	310.6874	4.4291	310.7753	4.5043	310.7787	4.4895	310.7851	4.4933
bior2.4	310.6719	4.4291	310.7416	4.4388	310.7541	4.4458	310.7493	4.4349
bior2.6	310.6542	4.4111	310.7188	4.4570	310.7223	4.4612	310.7277	4.4635
bior2.8	310.6406	4.3981	310.7138	4.4315	310.7136	4.4303	310.7169	4.4306
bior3.1	310.5995	4.3930	310.6254	4.3765	310.6240	4.3749	310.6226	4.3747
bior3.3	310.5837	4.3732	310.6484	4.4013	310.6499	4.4077	310.6498	4.4085
bior3.5	310.6000	4.4003	310.6555	4.4201	310.6538	4.4188	310.6504	4.4145
bior3.7	310.5996	4.3930	310.6329	4.3943	310.6319	4.3961	310.6339	4.3998
bior3.9	310.6012	4.3939	310.6420	4.4131	310.6510	4.4156	310.6472	4.4134
bior4.4	310.6403	4.4106	310.7279	4.4464	310.7424	4.4515	310.7501	4.4589
bior5.5	310.5827	4.3997	310.6850	4.4231	310.7216	4.4556	310.7390	4.4729
bior6.8	310.6023	4.4035	310.6902	4.4393	310.7021	4.4423	310.7101	4.4467
rbio1.1	310.7818	4.5063	310.8721	4.5024	310.8814	4.5014	310.8959	4.5157
rbio1.3	310.6966	4.4529	310.7616	4.4870	310.7697	4.4786	310.7810	4.4844
rbio1.5	310.6834	4.4543	310.7398	4.4813	310.7574	4.4924	310.7597	4.4857
rbio2.2	310.5945	4.3538	310.6234	4.3761	310.6240	4.3862	310.6166	4.3809
rbio2.4	310.6399	4.4226	310.6880	4.4410	310.7226	4.4606	310.7239	4.4584
rbio2.6	310.6186	4.4070	310.7029	4.4614	310.6984	4.4730	310.7126	4.4881
rbio2.8	310.5978	4.3899	310.6783	4.4518	310.6985	4.4546	310.7100	4.4664
rbio3.1	310.5627	4.3920	310.4950	4.3539	310.4871	4.4137	310.4887	4.4149
rbio3.3	310.6247	4.4339	310.5686	4.4504	310.5940	4.5147	310.6019	4.5144
rbio3.5	310.6225	4.4025	310.5927	4.4052	310.5991	4.4269	310.6064	4.4317
rbio3.7	310.6324	4.4242	310.6113	4.4224	310.6239	4.4558	310.6322	4.4559
rbio3.9	310.6243	4.4199	310.6074	4.3998	310.6096	4.4192	310.6075	4.4186
rbio4.4	310.6346	4.4137	310.6927	4.4393	310.7066	4.4432	310.7051	4.4354
rbio5.5	310.6011	4.3908	310.6545	4.4275	310.6619	4.4287	310.6625	4.4242
rbio6.8	310.5991	4.3962	310.6890	4.4516	310.6996	4.4542	310.7068	4.4591
dmey	310.5404	4.3886	310.6010	4.4244	310.5994	4.4162	310.6057	4.4150

Table A.25: Results of SureShrink with soft thresholding for DSM generation on City Site 3 excluding outliers

City Site 3, $\mu = 318.6986$, $\sigma = 9.0395$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.6986	8.9980	318.6986	8.9759	318.6986	8.9625	318.6986	8.9610
db2	318.6985	9.0047	318.6985	8.9841	318.6985	8.9757	318.6986	8.9734
db4	318.6987	9.0063	318.6987	8.9852	318.6987	8.9783	318.6987	8.9748
db6	318.6983	9.0056	318.6984	8.9841	318.6984	8.9778	318.6984	8.9757
db8	318.6986	9.0057	318.6988	8.9864	318.6987	8.9783	318.6987	8.9774
db10	318.6986	9.0058	318.6985	8.9858	318.6985	8.9806	318.6984	8.9784
db12	318.6984	9.0059	318.6985	8.9882	318.6985	8.9825	318.6985	8.9805
db20	318.6984	9.0035	318.6984	8.9855	318.6984	8.9806	318.6984	8.9796
sym2	318.6985	9.0047	318.6985	8.9841	318.6985	8.9757	318.6986	8.9734
sym4	318.6984	9.0072	318.6982	8.9876	318.6982	8.9788	318.6983	8.9753
sym6	318.6984	9.0076	318.6987	8.9887	318.6987	8.9820	318.6987	8.9784
sym8	318.6985	9.0074	318.6983	8.9863	318.6982	8.9779	318.6982	8.9757
sym10	318.6984	9.0076	318.6987	8.9884	318.6987	8.9802	318.6987	8.9781
sym12	318.6987	9.0077	318.6986	8.9878	318.6986	8.9793	318.6986	8.9764
sym20	318.6987	9.0071	318.6986	8.9872	318.6986	8.9795	318.6986	8.9776
coif1	318.6986	9.0060	318.6984	8.9829	318.6985	8.9736	318.6985	8.9708
coif2	318.6983	9.0080	318.6982	8.9878	318.6981	8.9798	318.6981	8.9746
coif3	318.6984	9.0081	318.6981	8.9876	318.6982	8.9792	318.6982	8.9773
coif4	318.6984	9.0076	318.6982	8.9874	318.6981	8.9793	318.6981	8.9756
coif5	318.6984	9.0074	318.6982	8.9857	318.6982	8.9783	318.6982	8.9757
bior1.1	318.6986	8.9980	318.6986	8.9759	318.6986	8.9625	318.6986	8.9610
bior1.3	318.6986	9.0213	318.6985	9.0078	318.6985	9.0014	318.6985	9.0006
bior1.5	318.6986	9.0255	318.6986	9.0143	318.6985	9.0091	318.6985	9.0068
bior2.2	318.6986	8.9967	318.6985	8.9761	318.6986	8.9691	318.6986	8.9668
bior2.4	318.6986	9.0041	318.6988	8.9873	318.6988	8.9830	318.6989	8.9804
bior2.6	318.6986	9.0078	318.6984	8.9923	318.6983	8.9872	318.6984	8.9855
bior2.8	318.6986	9.0098	318.6988	8.9951	318.6988	8.9910	318.6987	8.9895
bior3.1	318.6986	9.0052	318.6986	8.9974	318.6986	8.9960	318.6986	8.9958
bior3.3	318.6986	9.0022	318.6986	8.9934	318.6986	8.9914	318.6986	8.9913
bior3.5	318.6986	9.0036	318.6985	8.9941	318.6986	8.9927	318.6986	8.9926
bior3.7	318.6986	9.0054	318.6987	8.9969	318.6987	8.9947	318.6987	8.9937
bior3.9	318.6986	9.0063	318.6985	8.9977	318.6985	8.9961	318.6985	8.9959
bior4.4	318.6983	9.0038	318.6986	8.9812	318.6986	8.9725	318.6986	8.9681
bior5.5	318.6988	9.0097	318.6990	8.9825	318.6989	8.9694	318.6990	8.9623
bior6.8	318.6983	9.0056	318.6985	8.9853	318.6986	8.9774	318.6986	8.9744
rbio1.1	318.6986	8.9980	318.6986	8.9759	318.6986	8.9625	318.6986	8.9610
rbio1.3	318.6986	9.0022	318.6982	8.9790	318.6981	8.9680	318.6980	8.9654
rbio1.5	318.6986	9.0071	318.6987	8.9841	318.6985	8.9746	318.6985	8.9709
rbio2.2	318.6987	9.0177	318.6986	9.0019	318.6986	8.9938	318.6987	8.9911
rbio2.4	318.6984	9.0130	318.6986	8.9937	318.6986	8.9859	318.6986	8.9818
rbio2.6	318.6984	9.0135	318.6982	8.9918	318.6981	8.9829	318.6981	8.9769
rbio2.8	318.6984	9.0136	318.6986	8.9924	318.6986	8.9829	318.6986	8.9777
rbio3.1	318.6986	9.0390	318.6986	9.0388	318.6986	9.0381	318.6986	9.0381
rbio3.3	318.6986	9.0267	318.6986	9.0202	318.6986	9.0164	318.6986	9.0163
rbio3.5	318.6986	9.0227	318.6985	9.0111	318.6985	9.0065	318.6985	9.0039
rbio3.7	318.6986	9.0212	318.6987	9.0087	318.6988	9.0033	318.6987	8.9993
rbio3.9	318.6986	9.0205	318.6985	9.0053	318.6985	8.9999	318.6985	8.9963
rbio4.4	318.6984	9.0112	318.6986	8.9952	318.6986	8.9913	318.6986	8.9881
rbio5.5	318.6988	9.0117	318.6989	8.9988	318.6989	8.9961	318.6989	8.9942
rbio6.8	318.6984	9.0101	318.6986	8.9916	318.6986	8.9847	318.6986	8.9810
dmey	318.6984	9.0068	318.6983	8.9871	318.6983	8.9825	318.6983	8.9814

Table A.26: Results of SureShrink with soft thresholding for DEM generation on City Site 3 excluding outliers

City Site 3, $\mu = 310.7514$, $\sigma = 4.4699$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.8209	4.5014	310.9333	4.5234	310.9500	4.5097	310.9656	4.5229
db2	310.7306	4.4382	310.7953	4.4542	310.8280	4.4591	310.8283	4.4554
db4	310.6778	4.4113	310.7536	4.4601	310.8019	4.5011	310.8031	4.4910
db6	310.6770	4.4328	310.7901	4.4803	310.8102	4.4968	310.8142	4.4997
db8	310.6253	4.3681	310.7275	4.4305	310.7497	4.4500	310.7546	4.4480
db10	310.6090	4.3859	310.7304	4.4696	310.7614	4.4794	310.7621	4.4790
db12	310.6614	4.4259	310.6912	4.4365	310.7064	4.4455	310.7013	4.4380
db20	310.6226	4.4095	310.6543	4.4384	310.6827	4.4668	310.6834	4.4664
sym2	310.7306	4.4382	310.7953	4.4542	310.8280	4.4591	310.8283	4.4554
sym4	310.7165	4.4463	310.8037	4.4768	310.8124	4.4800	310.8181	4.4890
sym6	310.6923	4.4339	310.7616	4.4528	310.7863	4.4598	310.7926	4.4581
sym8	310.6724	4.4198	310.7701	4.4589	310.7841	4.4797	310.7931	4.4812
sym10	310.6668	4.4166	310.7364	4.4524	310.7502	4.4613	310.7539	4.4572
sym12	310.6299	4.3990	310.7080	4.4192	310.7290	4.4379	310.7340	4.4418
sym20	310.6097	4.3907	310.6914	4.4114	310.6960	4.4142	310.6965	4.4147
coif1	310.6885	4.3907	310.7949	4.4648	310.8220	4.4812	310.8259	4.4854
coif2	310.7217	4.4578	310.7915	4.4771	310.8055	4.4862	310.8183	4.4985
coif3	310.6842	4.4330	310.7622	4.4522	310.7653	4.4591	310.7645	4.4594
coif4	310.6791	4.4320	310.7558	4.4576	310.7636	4.4679	310.7642	4.4713
coif5	310.6758	4.4345	310.7317	4.4417	310.7561	4.4529	310.7540	4.4525
bior1.1	310.8209	4.5014	310.9333	4.5234	310.9500	4.5097	310.9656	4.5229
bior1.3	310.7791	4.4820	310.8436	4.4947	310.8597	4.4761	310.8661	4.4780
bior1.5	310.7593	4.4690	310.8404	4.5079	310.8394	4.5008	310.8425	4.4995
bior2.2	310.7527	4.4491	310.8409	4.4995	310.8545	4.5026	310.8530	4.5022
bior2.4	310.7067	4.4263	310.8050	4.4585	310.8170	4.4664	310.8218	4.4700
bior2.6	310.6986	4.4211	310.7884	4.4653	310.7968	4.4673	310.7991	4.4679
bior2.8	310.6861	4.4104	310.7775	4.4422	310.7819	4.4340	310.7781	4.4316
bior3.1	310.6494	4.3565	310.6863	4.3866	310.6821	4.3781	310.6844	4.3781
bior3.3	310.6627	4.3821	310.7059	4.4081	310.6991	4.3954	310.6989	4.3959
bior3.5	310.6409	4.3747	310.7061	4.4004	310.7179	4.4185	310.7187	4.4182
bior3.7	310.6521	4.3941	310.7029	4.4193	310.7021	4.4077	310.7010	4.4045
bior3.9	310.6409	4.3909	310.7112	4.4107	310.7094	4.4096	310.7112	4.4133
bior4.4	310.7079	4.4430	310.7975	4.4642	310.7995	4.4647	310.8171	4.4778
bior5.5	310.6409	4.3814	310.7595	4.4463	310.7960	4.4748	310.8084	4.4842
bior6.8	310.6748	4.4237	310.7557	4.4470	310.7678	4.4575	310.7738	4.4567
rbio1.1	310.8209	4.5014	310.9333	4.5234	310.9500	4.5097	310.9656	4.5229
rbio1.3	310.7393	4.4436	310.8344	4.5123	310.8502	4.5141	310.8627	4.5110
rbio1.5	310.7090	4.4216	310.7915	4.4724	310.8088	4.4708	310.8126	4.4618
rbio2.2	310.6630	4.3608	310.7010	4.3788	310.6936	4.3671	310.7009	4.3879
rbio2.4	310.7476	4.4683	310.7780	4.4675	310.7914	4.4573	310.8013	4.4574
rbio2.6	310.6927	4.4259	310.7554	4.4394	310.7686	4.4682	310.7891	4.4869
rbio2.8	310.6816	4.4259	310.7398	4.4419	310.7620	4.4660	310.7752	4.4647
rbio3.1	310.6098	4.3710	310.5421	4.3344	310.5283	4.3717	310.5445	4.3889
rbio3.3	310.6670	4.4053	310.6259	4.4364	310.6462	4.5063	310.6331	4.4881
rbio3.5	310.6697	4.3993	310.6772	4.4337	310.6945	4.4682	310.6927	4.4765
rbio3.7	310.6849	4.4251	310.7035	4.4425	310.6820	4.4166	310.6914	4.4099
rbio3.9	310.6788	4.4227	310.6926	4.4132	310.6991	4.4251	310.7131	4.4366
rbio4.4	310.7005	4.4253	310.7581	4.4480	310.7882	4.4729	310.7864	4.4758
rbio5.5	310.6575	4.3909	310.7041	4.3944	310.7106	4.3936	310.7093	4.3917
rbio6.8	310.6842	4.4463	310.7461	4.4579	310.7678	4.4800	310.7644	4.4709
dmey	310.6207	4.4110	310.6992	4.4570	310.7019	4.4553	310.7155	4.4637

Table A.27: Results of BayesShrink with hard thresholding for DSM generation on City Site 4 including outliers

City Site 4, $\mu = 299.6017$, $\sigma = 10.5357$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db2	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db4	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db6	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db8	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db10	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db12	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db20	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
sym2	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
sym4	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
sym6	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
sym8	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
sym10	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
sym12	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
sym20	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
coif1	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
coif2	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
coif3	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
coif4	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
coif5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior1.1	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior1.3	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior1.5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior2.2	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior2.4	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior2.6	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior2.8	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior3.1	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior3.3	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior3.5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior3.7	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior3.9	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior4.4	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior5.5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior6.8	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio1.1	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio1.3	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio1.5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio2.2	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio2.4	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio2.6	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio2.8	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio3.1	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio3.3	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio3.5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio3.7	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio3.9	299.6017	10.5357	299.6017	10.5358	299.6017	10.5358	299.6017	10.5358
rbio4.4	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio5.5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio6.8	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
dmey	299.6017	10.5357	299.6017	10.5358	299.6017	10.5358	299.6017	10.5358

Table A.28: Results of BayesShrink with hard thresholding for DEM generation on City Site 4 including outliers

City Site 4, $\mu = 291.9491$, $\sigma = 3.5185$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185
db2	291.9477	3.5176	291.9477	3.5176	291.9477	3.5176	291.9477	3.5176
db4	291.9478	3.5182	291.9478	3.5182	291.9478	3.5182	291.9478	3.5182
db6	291.9479	3.5172	291.9476	3.5172	291.9476	3.5172	291.9476	3.5172
db8	291.9486	3.5158	291.9488	3.5163	291.9488	3.5163	291.9488	3.5163
db10	291.9523	3.5165	291.9524	3.5165	291.9524	3.5165	291.9524	3.5165
db12	291.9481	3.5149	291.9470	3.5148	291.9470	3.5148	291.9470	3.5148
db20	291.9521	3.5162	291.9520	3.5158	291.9521	3.5157	291.9521	3.5157
sym2	291.9477	3.5176	291.9477	3.5176	291.9477	3.5176	291.9477	3.5176
sym4	291.9494	3.5181	291.9495	3.5181	291.9495	3.5181	291.9495	3.5181
sym6	291.9475	3.5166	291.9476	3.5167	291.9476	3.5167	291.9476	3.5167
sym8	291.9491	3.5173	291.9486	3.5174	291.9486	3.5174	291.9486	3.5174
sym10	291.9481	3.5172	291.9482	3.5170	291.9482	3.5170	291.9482	3.5170
sym12	291.9471	3.5164	291.9473	3.5164	291.9473	3.5164	291.9473	3.5164
sym20	291.9469	3.5148	291.9472	3.5147	291.9472	3.5147	291.9472	3.5147
coif1	291.9488	3.5189	291.9492	3.5191	291.9492	3.5191	291.9492	3.5191
coif2	291.9502	3.5175	291.9501	3.5175	291.9501	3.5175	291.9501	3.5175
coif3	291.9474	3.5173	291.9476	3.5173	291.9476	3.5173	291.9476	3.5173
coif4	291.9484	3.5172	291.9482	3.5173	291.9482	3.5173	291.9482	3.5173
coif5	291.9505	3.5183	291.9504	3.5183	291.9504	3.5183	291.9504	3.5183
bior1.1	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185
bior1.3	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185
bior1.5	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185
bior2.2	291.9488	3.5186	291.9488	3.5186	291.9488	3.5186	291.9488	3.5186
bior2.4	291.9490	3.5186	291.9490	3.5186	291.9490	3.5186	291.9490	3.5186
bior2.6	291.9490	3.5186	291.9490	3.5186	291.9490	3.5186	291.9490	3.5186
bior2.8	291.9490	3.5186	291.9490	3.5186	291.9490	3.5186	291.9490	3.5186
bior3.1	291.9493	3.5191	291.9493	3.5191	291.9493	3.5191	291.9493	3.5191
bior3.3	291.9489	3.5188	291.9489	3.5188	291.9489	3.5188	291.9489	3.5188
bior3.5	291.9485	3.5190	291.9485	3.5190	291.9485	3.5190	291.9485	3.5190
bior3.7	291.9485	3.5191	291.9485	3.5191	291.9485	3.5191	291.9485	3.5191
bior3.9	291.9484	3.5190	291.9484	3.5190	291.9484	3.5190	291.9484	3.5190
bior4.4	291.9489	3.5179	291.9489	3.5179	291.9489	3.5179	291.9489	3.5179
bior5.5	291.9458	3.5153	291.9454	3.5155	291.9454	3.5155	291.9454	3.5155
bior6.8	291.9491	3.5174	291.9492	3.5174	291.9492	3.5174	291.9492	3.5174
rbio1.1	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185	291.9490	3.5185
rbio1.3	291.9496	3.5186	291.9496	3.5186	291.9496	3.5186	291.9496	3.5186
rbio1.5	291.9493	3.5174	291.9497	3.5174	291.9497	3.5174	291.9497	3.5174
rbio2.2	291.9496	3.5181	291.9500	3.5178	291.9500	3.5178	291.9500	3.5178
rbio2.4	291.9476	3.5174	291.9473	3.5174	291.9473	3.5174	291.9473	3.5174
rbio2.6	291.9460	3.5178	291.9451	3.5172	291.9452	3.5172	291.9452	3.5172
rbio2.8	291.9470	3.5179	291.9475	3.5179	291.9474	3.5179	291.9474	3.5179
rbio3.1	291.9508	3.5194	291.9506	3.5195	291.9507	3.5195	291.9507	3.5195
rbio3.3	291.9488	3.5192	291.9494	3.5195	291.9494	3.5195	291.9494	3.5195
rbio3.5	291.9493	3.5189	291.9488	3.5196	291.9489	3.5197	291.9489	3.5197
rbio3.7	291.9502	3.5186	291.9512	3.5176	291.9510	3.5176	291.9513	3.5176
rbio3.9	291.9534	3.5167	291.9546	3.5192	291.9530	3.5195	291.9530	3.5195
rbio4.4	291.9485	3.5172	291.9486	3.5172	291.9486	3.5172	291.9486	3.5172
rbio5.5	291.9482	3.5180	291.9482	3.5180	291.9482	3.5180	291.9482	3.5180
rbio6.8	291.9496	3.5175	291.9489	3.5176	291.9488	3.5176	291.9488	3.5176
dmey	291.9484	3.5161	291.9481	3.5161	291.9481	3.5161	291.9481	3.5161

Table A.29: Results of BayesShrink with hard thresholding for DSM generation on City Site 4 excluding outliers

City Site 4, $\mu = 299.5510$, $\sigma = 10.4379$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
db2	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
db4	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
db6	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
db8	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
db10	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
db12	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
db20	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
sym2	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
sym4	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
sym6	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
sym8	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
sym10	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
sym12	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
sym20	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
coif1	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
coif2	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
coif3	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
coif4	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
coif5	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior1.1	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior1.3	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior1.5	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior2.2	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior2.4	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior2.6	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior2.8	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior3.1	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior3.3	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior3.5	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior3.7	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior3.9	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior4.4	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior5.5	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
bior6.8	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio1.1	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio1.3	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio1.5	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio2.2	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio2.4	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio2.6	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio2.8	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio3.1	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio3.3	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio3.5	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio3.7	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio3.9	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio4.4	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio5.5	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
rbio6.8	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379
dmey	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379	299.5510	10.4379

Table A.30: Results of BayesShrink with hard thresholding for DEM generation on City Site 4 excluding outliers

City Site 4, $\mu = 292.0253$, $\sigma = 3.5621$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621
db2	292.0230	3.5614	292.0229	3.5614	292.0229	3.5614	292.0229	3.5614
db4	292.0216	3.5615	292.0216	3.5616	292.0216	3.5616	292.0216	3.5616
db6	292.0247	3.5599	292.0245	3.5598	292.0245	3.5598	292.0245	3.5598
db8	292.0252	3.5609	292.0252	3.5609	292.0252	3.5609	292.0252	3.5609
db10	292.0294	3.5595	292.0284	3.5600	292.0289	3.5596	292.0289	3.5596
db12	292.0277	3.5616	292.0279	3.5602	292.0279	3.5602	292.0279	3.5602
db20	292.0260	3.5600	292.0263	3.5611	292.0263	3.5611	292.0263	3.5611
sym2	292.0230	3.5614	292.0229	3.5614	292.0229	3.5614	292.0229	3.5614
sym4	292.0257	3.5618	292.0257	3.5618	292.0257	3.5618	292.0257	3.5618
sym6	292.0278	3.5630	292.0277	3.5630	292.0277	3.5630	292.0277	3.5630
sym8	292.0272	3.5619	292.0272	3.5619	292.0272	3.5619	292.0272	3.5619
sym10	292.0258	3.5612	292.0258	3.5614	292.0258	3.5614	292.0258	3.5614
sym12	292.0255	3.5608	292.0257	3.5609	292.0257	3.5609	292.0257	3.5609
sym20	292.0273	3.5613	292.0273	3.5613	292.0273	3.5613	292.0273	3.5613
coif1	292.0248	3.5613	292.0249	3.5613	292.0249	3.5613	292.0249	3.5613
coif2	292.0260	3.5613	292.0260	3.5613	292.0260	3.5613	292.0260	3.5613
coif3	292.0262	3.5611	292.0262	3.5612	292.0262	3.5612	292.0262	3.5612
coif4	292.0253	3.5612	292.0252	3.5610	292.0252	3.5610	292.0252	3.5610
coif5	292.0250	3.5612	292.0251	3.5612	292.0251	3.5612	292.0251	3.5612
bior1.1	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621
bior1.3	292.0254	3.5622	292.0254	3.5622	292.0254	3.5622	292.0254	3.5622
bior1.5	292.0255	3.5621	292.0255	3.5621	292.0255	3.5621	292.0255	3.5621
bior2.2	292.0249	3.5622	292.0249	3.5622	292.0249	3.5622	292.0249	3.5622
bior2.4	292.0252	3.5622	292.0252	3.5622	292.0252	3.5622	292.0252	3.5622
bior2.6	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621
bior2.8	292.0254	3.5622	292.0254	3.5622	292.0254	3.5622	292.0254	3.5622
bior3.1	292.0255	3.5621	292.0255	3.5621	292.0255	3.5621	292.0255	3.5621
bior3.3	292.0251	3.5620	292.0251	3.5620	292.0251	3.5620	292.0251	3.5620
bior3.5	292.0246	3.5623	292.0246	3.5623	292.0246	3.5623	292.0246	3.5623
bior3.7	292.0247	3.5623	292.0247	3.5623	292.0247	3.5623	292.0247	3.5623
bior3.9	292.0246	3.5623	292.0246	3.5623	292.0246	3.5623	292.0246	3.5623
bior4.4	292.0253	3.5619	292.0253	3.5619	292.0253	3.5619	292.0253	3.5619
bior5.5	292.0253	3.5614	292.0254	3.5613	292.0254	3.5613	292.0254	3.5613
bior6.8	292.0250	3.5611	292.0250	3.5612	292.0250	3.5612	292.0250	3.5612
rbio1.1	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621	292.0253	3.5621
rbio1.3	292.0259	3.5624	292.0259	3.5625	292.0259	3.5625	292.0259	3.5625
rbio1.5	292.0240	3.5613	292.0243	3.5613	292.0243	3.5613	292.0243	3.5613
rbio2.2	292.0251	3.5616	292.0252	3.5619	292.0252	3.5619	292.0252	3.5619
rbio2.4	292.0253	3.5613	292.0259	3.5621	292.0259	3.5621	292.0259	3.5621
rbio2.6	292.0262	3.5626	292.0270	3.5629	292.0270	3.5629	292.0270	3.5629
rbio2.8	292.0263	3.5614	292.0260	3.5611	292.0258	3.5612	292.0258	3.5612
rbio3.1	292.0272	3.5652	292.0287	3.5655	292.0287	3.5655	292.0287	3.5655
rbio3.3	292.0234	3.5606	292.0251	3.5608	292.0251	3.5608	292.0251	3.5608
rbio3.5	292.0265	3.5640	292.0264	3.5639	292.0265	3.5639	292.0265	3.5639
rbio3.7	292.0287	3.5634	292.0310	3.5631	292.0310	3.5632	292.0310	3.5632
rbio3.9	292.0302	3.5613	292.0300	3.5602	292.0303	3.5601	292.0303	3.5601
rbio4.4	292.0237	3.5602	292.0234	3.5603	292.0234	3.5603	292.0234	3.5603
rbio5.5	292.0248	3.5619	292.0250	3.5621	292.0250	3.5621	292.0250	3.5621
rbio6.8	292.0264	3.5622	292.0263	3.5622	292.0263	3.5622	292.0263	3.5622
dmey	292.0231	3.5600	292.0229	3.5601	292.0229	3.5601	292.0229	3.5601

Table A.31: Results of BayesShrink with soft thresholding for DSM generation on City Site 4 including outliers

City Site 4, $\mu = 299.6017$, $\sigma = 10.5357$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
db2	299.6017	10.5354	299.6017	10.5353	299.6017	10.5352	299.6017	10.5352
db4	299.6017	10.5350	299.6017	10.5348	299.6017	10.5347	299.6017	10.5347
db6	299.6017	10.5346	299.6017	10.5342	299.6017	10.5341	299.6017	10.5341
db8	299.6017	10.5341	299.6017	10.5336	299.6017	10.5335	299.6017	10.5334
db10	299.6017	10.5336	299.6017	10.5330	299.6017	10.5328	299.6017	10.5327
db12	299.6017	10.5332	299.6017	10.5324	299.6017	10.5322	299.6017	10.5321
db20	299.6017	10.5314	299.6017	10.5301	299.6017	10.5297	299.6017	10.5296
sym2	299.6017	10.5354	299.6017	10.5353	299.6017	10.5352	299.6017	10.5352
sym4	299.6017	10.5352	299.6017	10.5350	299.6017	10.5350	299.6017	10.5350
sym6	299.6017	10.5350	299.6017	10.5348	299.6017	10.5348	299.6017	10.5348
sym8	299.6017	10.5349	299.6017	10.5346	299.6017	10.5346	299.6017	10.5346
sym10	299.6017	10.5348	299.6017	10.5345	299.6017	10.5344	299.6017	10.5344
sym12	299.6017	10.5347	299.6017	10.5343	299.6017	10.5342	299.6017	10.5342
sym20	299.6017	10.5343	299.6017	10.5339	299.6017	10.5338	299.6017	10.5337
coif1	299.6017	10.5354	299.6017	10.5353	299.6017	10.5352	299.6017	10.5352
coif2	299.6017	10.5352	299.6017	10.5351	299.6017	10.5350	299.6017	10.5350
coif3	299.6017	10.5351	299.6017	10.5349	299.6017	10.5348	299.6017	10.5348
coif4	299.6017	10.5349	299.6017	10.5347	299.6017	10.5346	299.6017	10.5346
coif5	299.6017	10.5348	299.6017	10.5345	299.6017	10.5344	299.6017	10.5344
bior1.1	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior1.3	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior1.5	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
bior2.2	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior2.4	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior2.6	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior2.8	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior3.1	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior3.3	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior3.5	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior3.7	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior3.9	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356	299.6017	10.5356
bior4.4	299.6017	10.5353	299.6017	10.5351	299.6017	10.5351	299.6017	10.5351
bior5.5	299.6017	10.5346	299.6017	10.5340	299.6017	10.5337	299.6017	10.5336
bior6.8	299.6017	10.5352	299.6017	10.5350	299.6017	10.5350	299.6017	10.5349
rbio1.1	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357	299.6017	10.5357
rbio1.3	299.6017	10.5352	299.6017	10.5351	299.6017	10.5350	299.6017	10.5350
rbio1.5	299.6017	10.5351	299.6017	10.5349	299.6017	10.5348	299.6017	10.5348
rbio2.2	299.6017	10.5348	299.6017	10.5345	299.6017	10.5343	299.6017	10.5343
rbio2.4	299.6017	10.5347	299.6017	10.5341	299.6017	10.5339	299.6017	10.5339
rbio2.6	299.6017	10.5344	299.6017	10.5337	299.6017	10.5335	299.6017	10.5334
rbio2.8	299.6017	10.5342	299.6017	10.5333	299.6017	10.5330	299.6017	10.5329
rbio3.1	299.6017	10.5352	299.6017	10.5350	299.6017	10.5349	299.6017	10.5349
rbio3.3	299.6017	10.5339	299.6017	10.5332	299.6017	10.5330	299.6017	10.5329
rbio3.5	299.6017	10.5334	299.6017	10.5322	299.6017	10.5318	299.6017	10.5316
rbio3.7	299.6017	10.5329	299.6017	10.5313	299.6017	10.5308	299.6017	10.5306
rbio3.9	299.6017	10.5325	299.6017	10.5305	299.6017	10.5298	299.6017	10.5296
rbio4.4	299.6017	10.5351	299.6017	10.5350	299.6017	10.5349	299.6017	10.5349
rbio5.5	299.6017	10.5354	299.6017	10.5353	299.6017	10.5353	299.6017	10.5353
rbio6.8	299.6017	10.5348	299.6017	10.5345	299.6017	10.5344	299.6017	10.5344
dmey	299.6017	10.5339	299.6017	10.5334	299.6017	10.5333	299.6017	10.5332

Table A.32: Results of BayesShrink with soft thresholding for DEM generation on City Site 4 including outliers

City Site 4, $\mu = 291.9491$, $\sigma = 3.5185$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	291.9498	3.5180	291.9502	3.5179	291.9502	3.5179	291.9502	3.5179
db2	291.9531	3.5161	291.9552	3.5157	291.9553	3.5157	291.9554	3.5157
db4	291.9606	3.5175	291.9617	3.5168	291.9621	3.5167	291.9624	3.5167
db6	291.9572	3.5132	291.9609	3.5132	291.9611	3.5130	291.9608	3.5131
db8	291.9607	3.5173	291.9636	3.5172	291.9638	3.5169	291.9639	3.5170
db10	291.9619	3.5162	291.9660	3.5147	291.9672	3.5146	291.9669	3.5147
db12	291.9638	3.5179	291.9681	3.5166	291.9695	3.5162	291.9693	3.5162
db20	291.9587	3.5126	291.9680	3.5155	291.9697	3.5149	291.9700	3.5148
sym2	291.9531	3.5161	291.9552	3.5157	291.9553	3.5157	291.9554	3.5157
sym4	291.9576	3.5167	291.9601	3.5169	291.9609	3.5169	291.9608	3.5167
sym6	291.9564	3.5147	291.9593	3.5154	291.9593	3.5155	291.9596	3.5155
sym8	291.9577	3.5156	291.9597	3.5152	291.9602	3.5151	291.9604	3.5152
sym10	291.9596	3.5159	291.9633	3.5161	291.9636	3.5156	291.9636	3.5155
sym12	291.9597	3.5169	291.9638	3.5185	291.9637	3.5175	291.9637	3.5177
sym20	291.9605	3.5158	291.9649	3.5159	291.9671	3.5167	291.9670	3.5168
coif1	291.9540	3.5158	291.9555	3.5158	291.9557	3.5158	291.9559	3.5157
coif2	291.9576	3.5174	291.9605	3.5186	291.9605	3.5188	291.9608	3.5187
coif3	291.9565	3.5167	291.9588	3.5163	291.9592	3.5164	291.9592	3.5164
coif4	291.9573	3.5153	291.9593	3.5160	291.9598	3.5157	291.9599	3.5157
coif5	291.9570	3.5157	291.9607	3.5161	291.9613	3.5160	291.9617	3.5160
bior1.1	291.9498	3.5180	291.9502	3.5179	291.9502	3.5179	291.9502	3.5179
bior1.3	291.9478	3.5174	291.9480	3.5174	291.9480	3.5174	291.9480	3.5174
bior1.5	291.9478	3.5174	291.9488	3.5173	291.9489	3.5172	291.9489	3.5172
bior2.2	291.9496	3.5169	291.9497	3.5170	291.9496	3.5170	291.9496	3.5170
bior2.4	291.9509	3.5177	291.9510	3.5174	291.9510	3.5174	291.9510	3.5174
bior2.6	291.9502	3.5176	291.9502	3.5173	291.9502	3.5173	291.9502	3.5173
bior2.8	291.9502	3.5173	291.9503	3.5173	291.9503	3.5173	291.9503	3.5173
bior3.1	291.9515	3.5163	291.9518	3.5163	291.9517	3.5163	291.9517	3.5163
bior3.3	291.9505	3.5163	291.9506	3.5163	291.9506	3.5163	291.9506	3.5163
bior3.5	291.9500	3.5158	291.9501	3.5158	291.9501	3.5158	291.9501	3.5158
bior3.7	291.9503	3.5157	291.9506	3.5156	291.9507	3.5156	291.9507	3.5156
bior3.9	291.9498	3.5152	291.9499	3.5152	291.9499	3.5152	291.9499	3.5152
bior4.4	291.9567	3.5160	291.9561	3.5146	291.9565	3.5146	291.9566	3.5146
bior5.5	291.9633	3.5187	291.9728	3.5180	291.9754	3.5187	291.9750	3.5187
bior6.8	291.9572	3.5169	291.9590	3.5165	291.9583	3.5166	291.9585	3.5165
rbio1.1	291.9498	3.5180	291.9502	3.5179	291.9502	3.5179	291.9502	3.5179
rbio1.3	291.9549	3.5185	291.9556	3.5181	291.9561	3.5179	291.9561	3.5179
rbio1.5	291.9575	3.5171	291.9599	3.5168	291.9603	3.5164	291.9604	3.5164
rbio2.2	291.9611	3.5143	291.9648	3.5153	291.9654	3.5154	291.9653	3.5152
rbio2.4	291.9615	3.5166	291.9665	3.5184	291.9678	3.5176	291.9675	3.5179
rbio2.6	291.9603	3.5140	291.9680	3.5107	291.9683	3.5106	291.9686	3.5106
rbio2.8	291.9616	3.5136	291.9696	3.5119	291.9707	3.5095	291.9708	3.5098
rbio3.1	291.9355	3.5154	291.9385	3.5134	291.9388	3.5134	291.9392	3.5142
rbio3.3	291.9496	3.5180	291.9590	3.5131	291.9597	3.5128	291.9595	3.5127
rbio3.5	291.9657	3.5162	291.9789	3.5151	291.9839	3.5136	291.9831	3.5135
rbio3.7	291.9688	3.5154	291.9859	3.5131	291.9867	3.5114	291.9874	3.5116
rbio3.9	291.9718	3.5158	291.9841	3.5126	291.9890	3.5115	291.9891	3.5117
rbio4.4	291.9584	3.5158	291.9592	3.5162	291.9592	3.5163	291.9592	3.5163
rbio5.5	291.9519	3.5151	291.9516	3.5153	291.9516	3.5153	291.9516	3.5153
rbio6.8	291.9594	3.5165	291.9614	3.5160	291.9620	3.5157	291.9616	3.5158
dmey	291.9603	3.5167	291.9617	3.5160	291.9622	3.5158	291.9618	3.5158

Table A.33: Results of BayesShrink with soft thresholding for DSM generation on City Site 4 excluding outliers

City Site 4, $\mu = 299.5510$, $\sigma = 10.4379$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
db2	299.5510	10.4375	299.5510	10.4374	299.5510	10.4374	299.5510	10.4374
db4	299.5510	10.4372	299.5510	10.4370	299.5510	10.4370	299.5510	10.4370
db6	299.5510	10.4369	299.5510	10.4366	299.5510	10.4365	299.5510	10.4365
db8	299.5510	10.4366	299.5510	10.4361	299.5510	10.4360	299.5510	10.4360
db10	299.5510	10.4362	299.5510	10.4356	299.5510	10.4355	299.5510	10.4354
db12	299.5510	10.4359	299.5510	10.4353	299.5510	10.4351	299.5510	10.4351
db20	299.5510	10.4345	299.5510	10.4335	299.5510	10.4332	299.5510	10.4331
sym2	299.5510	10.4375	299.5510	10.4374	299.5510	10.4374	299.5510	10.4374
sym4	299.5510	10.4374	299.5510	10.4372	299.5510	10.4372	299.5510	10.4372
sym6	299.5510	10.4373	299.5510	10.4371	299.5510	10.4370	299.5510	10.4370
sym8	299.5510	10.4372	299.5510	10.4369	299.5510	10.4369	299.5510	10.4369
sym10	299.5510	10.4371	299.5510	10.4368	299.5510	10.4368	299.5510	10.4367
sym12	299.5510	10.4370	299.5510	10.4367	299.5510	10.4366	299.5510	10.4366
sym20	299.5510	10.4367	299.5510	10.4364	299.5510	10.4363	299.5510	10.4362
coif1	299.5510	10.4375	299.5510	10.4374	299.5510	10.4374	299.5510	10.4374
coif2	299.5510	10.4374	299.5510	10.4373	299.5510	10.4372	299.5510	10.4372
coif3	299.5510	10.4373	299.5510	10.4371	299.5510	10.4370	299.5510	10.4370
coif4	299.5510	10.4372	299.5510	10.4370	299.5510	10.4369	299.5510	10.4369
coif5	299.5510	10.4371	299.5510	10.4368	299.5510	10.4367	299.5510	10.4367
bior1.1	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
bior1.3	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
bior1.5	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
bior2.2	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377
bior2.4	299.5510	10.4378	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377
bior2.6	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
bior2.8	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
bior3.1	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377
bior3.3	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377
bior3.5	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377
bior3.7	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377
bior3.9	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377	299.5510	10.4377
bior4.4	299.5510	10.4375	299.5510	10.4373	299.5510	10.4373	299.5510	10.4373
bior5.5	299.5510	10.4369	299.5510	10.4364	299.5510	10.4361	299.5510	10.4361
bior6.8	299.5510	10.4373	299.5510	10.4372	299.5510	10.4372	299.5510	10.4372
rbio1.1	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378	299.5510	10.4378
rbio1.3	299.5510	10.4374	299.5510	10.4373	299.5510	10.4372	299.5510	10.4372
rbio1.5	299.5510	10.4373	299.5510	10.4371	299.5510	10.4370	299.5510	10.4370
rbio2.2	299.5510	10.4370	299.5510	10.4367	299.5510	10.4365	299.5510	10.4365
rbio2.4	299.5510	10.4369	299.5510	10.4365	299.5510	10.4363	299.5510	10.4363
rbio2.6	299.5510	10.4367	299.5510	10.4361	299.5510	10.4359	299.5510	10.4358
rbio2.8	299.5510	10.4366	299.5510	10.4358	299.5510	10.4356	299.5510	10.4355
rbio3.1	299.5510	10.4374	299.5510	10.4372	299.5510	10.4372	299.5510	10.4371
rbio3.3	299.5510	10.4363	299.5510	10.4357	299.5510	10.4354	299.5510	10.4353
rbio3.5	299.5510	10.4359	299.5510	10.4348	299.5510	10.4345	299.5510	10.4344
rbio3.7	299.5510	10.4356	299.5510	10.4342	299.5510	10.4337	299.5510	10.4336
rbio3.9	299.5510	10.4353	299.5510	10.4336	299.5510	10.4330	299.5510	10.4328
rbio4.4	299.5510	10.4373	299.5510	10.4372	299.5510	10.4372	299.5510	10.4371
rbio5.5	299.5510	10.4375	299.5510	10.4375	299.5510	10.4375	299.5510	10.4375
rbio6.8	299.5510	10.4371	299.5510	10.4368	299.5510	10.4368	299.5510	10.4367
dmey	299.5510	10.4364	299.5510	10.4360	299.5510	10.4359	299.5510	10.4359

Table A.34: Results of BayesShrink with soft thresholding for DEM generation on City Site 4 excluding outliers

City Site 4, $\mu = 292.0253$, $\sigma = 3.5621$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.0258	3.5620	292.0260	3.5619	292.0257	3.5619	292.0257	3.5619
db2	292.0277	3.5599	292.0290	3.5599	292.0291	3.5598	292.0291	3.5598
db4	292.0354	3.5619	292.0364	3.5607	292.0368	3.5606	292.0368	3.5606
db6	292.0330	3.5573	292.0349	3.5565	292.0345	3.5570	292.0345	3.5570
db8	292.0366	3.5576	292.0396	3.5564	292.0394	3.5563	292.0394	3.5563
db10	292.0366	3.5604	292.0413	3.5612	292.0416	3.5610	292.0414	3.5610
db12	292.0354	3.5599	292.0366	3.5565	292.0363	3.5575	292.0362	3.5574
db20	292.0395	3.5615	292.0414	3.5629	292.0415	3.5621	292.0416	3.5621
sym2	292.0277	3.5599	292.0290	3.5599	292.0291	3.5598	292.0291	3.5598
sym4	292.0305	3.5592	292.0314	3.5593	292.0318	3.5592	292.0319	3.5591
sym6	292.0301	3.5575	292.0322	3.5585	292.0324	3.5583	292.0326	3.5583
sym8	292.0313	3.5578	292.0342	3.5577	292.0344	3.5576	292.0344	3.5576
sym10	292.0326	3.5584	292.0376	3.5599	292.0385	3.5598	292.0384	3.5598
sym12	292.0404	3.5615	292.0417	3.5608	292.0421	3.5596	292.0425	3.5607
sym20	292.0407	3.5609	292.0408	3.5601	292.0415	3.5602	292.0412	3.5600
coif1	292.0285	3.5577	292.0310	3.5584	292.0313	3.5585	292.0313	3.5585
coif2	292.0316	3.5597	292.0351	3.5612	292.0347	3.5612	292.0347	3.5612
coif3	292.0325	3.5583	292.0344	3.5583	292.0345	3.5583	292.0346	3.5583
coif4	292.0322	3.5572	292.0363	3.5579	292.0369	3.5577	292.0368	3.5579
coif5	292.0322	3.5578	292.0339	3.5570	292.0342	3.5568	292.0343	3.5570
bior1.1	292.0258	3.5620	292.0260	3.5619	292.0257	3.5619	292.0257	3.5619
bior1.3	292.0247	3.5614	292.0248	3.5614	292.0249	3.5614	292.0249	3.5614
bior1.5	292.0246	3.5616	292.0247	3.5616	292.0247	3.5616	292.0247	3.5616
bior2.2	292.0280	3.5609	292.0280	3.5611	292.0280	3.5611	292.0280	3.5611
bior2.4	292.0270	3.5608	292.0274	3.5604	292.0275	3.5604	292.0275	3.5604
bior2.6	292.0268	3.5603	292.0264	3.5607	292.0264	3.5607	292.0264	3.5607
bior2.8	292.0260	3.5604	292.0266	3.5598	292.0265	3.5598	292.0265	3.5598
bior3.1	292.0273	3.5606	292.0270	3.5599	292.0270	3.5599	292.0270	3.5599
bior3.3	292.0250	3.5591	292.0250	3.5590	292.0250	3.5590	292.0250	3.5590
bior3.5	292.0257	3.5589	292.0259	3.5589	292.0259	3.5589	292.0259	3.5589
bior3.7	292.0272	3.5592	292.0269	3.5593	292.0269	3.5593	292.0269	3.5593
bior3.9	292.0271	3.5592	292.0269	3.5592	292.0269	3.5592	292.0269	3.5592
bior4.4	292.0320	3.5604	292.0314	3.5591	292.0340	3.5608	292.0333	3.5604
bior5.5	292.0407	3.5617	292.0491	3.5596	292.0515	3.5597	292.0517	3.5597
bior6.8	292.0318	3.5586	292.0316	3.5582	292.0337	3.5601	292.0338	3.5601
rbio1.1	292.0258	3.5620	292.0260	3.5619	292.0257	3.5619	292.0257	3.5619
rbio1.3	292.0318	3.5625	292.0332	3.5624	292.0342	3.5622	292.0341	3.5622
rbio1.5	292.0315	3.5602	292.0325	3.5595	292.0330	3.5590	292.0331	3.5590
rbio2.2	292.0338	3.5563	292.0373	3.5583	292.0371	3.5580	292.0371	3.5578
rbio2.4	292.0357	3.5579	292.0422	3.5570	292.0439	3.5587	292.0439	3.5587
rbio2.6	292.0359	3.5561	292.0450	3.5581	292.0448	3.5564	292.0447	3.5564
rbio2.8	292.0354	3.5547	292.0459	3.5546	292.0460	3.5542	292.0467	3.5540
rbio3.1	292.0165	3.5608	292.0171	3.5603	292.0128	3.5605	292.0129	3.5594
rbio3.3	292.0283	3.5608	292.0328	3.5546	292.0327	3.5542	292.0325	3.5542
rbio3.5	292.0392	3.5577	292.0522	3.5580	292.0526	3.5560	292.0528	3.5561
rbio3.7	292.0447	3.5592	292.0561	3.5560	292.0571	3.5553	292.0582	3.5548
rbio3.9	292.0419	3.5553	292.0592	3.5549	292.0622	3.5541	292.0632	3.5543
rbio4.4	292.0297	3.5581	292.0313	3.5575	292.0314	3.5574	292.0314	3.5574
rbio5.5	292.0302	3.5604	292.0299	3.5603	292.0300	3.5600	292.0300	3.5600
rbio6.8	292.0317	3.5570	292.0360	3.5579	292.0370	3.5586	292.0368	3.5586
dmey	292.0409	3.5608	292.0429	3.5610	292.0431	3.5607	292.0433	3.5606

Table A.35: Results of Hypothesis Testing with hard thresholding for DSM generation on City Site 2 including outliers

City Site 2, $\mu = 303.4698$, $\sigma = 11.7044$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4698	11.6535	303.4698	11.5979	303.4695	11.5425	303.4700	11.5084
db2	303.4687	11.6638	303.4694	11.6127	303.4690	11.5581	303.4698	11.5228
db4	303.4695	11.6699	303.4695	11.6205	303.4690	11.5652	303.4703	11.5224
db6	303.4699	11.6707	303.4697	11.6188	303.4700	11.5614	303.4714	11.5151
db8	303.4692	11.6692	303.4689	11.6152	303.4691	11.5537	303.4680	11.5089
db10	303.4693	11.6679	303.4690	11.6116	303.4669	11.5494	303.4675	11.5034
db12	303.4704	11.6683	303.4697	11.6102	303.4684	11.5456	303.4703	11.4989
db20	303.4702	11.6667	303.4699	11.6061	303.4714	11.5382	303.4739	11.4874
sym2	303.4687	11.6638	303.4694	11.6127	303.4690	11.5581	303.4698	11.5228
sym4	303.4703	11.6717	303.4701	11.6239	303.4691	11.5723	303.4688	11.5335
sym6	303.4700	11.6725	303.4701	11.6253	303.4693	11.5731	303.4643	11.5399
sym8	303.4698	11.6724	303.4695	11.6234	303.4687	11.5670	303.4697	11.5299
sym10	303.4698	11.6724	303.4693	11.6246	303.4695	11.5725	303.4700	11.5337
sym12	303.4698	11.6714	303.4704	11.6220	303.4710	11.5697	303.4693	11.5264
sym20	303.4699	11.6714	303.4702	11.6215	303.4707	11.5647	303.4744	11.5370
coif1	303.4706	11.6662	303.4702	11.6136	303.4711	11.5631	303.4748	11.5311
coif2	303.4703	11.6720	303.4699	11.6247	303.4681	11.5719	303.4658	11.5363
coif3	303.4701	11.6726	303.4697	11.6245	303.4681	11.5702	303.4681	11.5294
coif4	303.4700	11.6728	303.4696	11.6243	303.4692	11.5669	303.4678	11.5290
coif5	303.4699	11.6728	303.4696	11.6240	303.4694	11.5704	303.4680	11.5291
bior1.1	303.4698	11.6535	303.4698	11.5979	303.4695	11.5425	303.4700	11.5084
bior1.3	303.4698	11.6999	303.4698	11.6991	303.4680	11.6957	303.4697	11.6900
bior1.5	303.4698	11.7092	303.4694	11.7199	303.4702	11.7257	303.4683	11.7218
bior2.2	303.4706	11.6591	303.4699	11.6033	303.4726	11.5563	303.4749	11.5338
bior2.4	303.4706	11.6694	303.4725	11.6278	303.4729	11.5914	303.4751	11.5698
bior2.6	303.4706	11.6738	303.4702	11.6354	303.4687	11.6002	303.4681	11.5851
bior2.8	303.4706	11.6764	303.4716	11.6417	303.4716	11.6147	303.4718	11.6001
bior3.1	303.4698	11.6765	303.4697	11.7035	303.4682	11.7847	303.4713	11.8301
bior3.3	303.4698	11.6696	303.4702	11.6616	303.4692	11.6903	303.4641	11.7114
bior3.5	303.4698	11.6703	303.4698	11.6529	303.4721	11.6590	303.4706	11.6675
bior3.7	303.4698	11.6712	303.4695	11.6558	303.4702	11.6604	303.4678	11.6693
bior3.9	303.4698	11.6721	303.4698	11.6546	303.4696	11.6607	303.4689	11.6761
bior4.4	303.4702	11.6698	303.4717	11.6199	303.4716	11.5568	303.4744	11.5073
bior5.5	303.4695	11.6714	303.4688	11.6154	303.4671	11.5389	303.4616	11.4879
bior6.8	303.4701	11.6718	303.4704	11.6237	303.4711	11.5699	303.4706	11.5259
rbio1.1	303.4698	11.6535	303.4698	11.5979	303.4695	11.5425	303.4700	11.5084
rbio1.3	303.4698	11.6657	303.4706	11.6147	303.4690	11.5632	303.4692	11.5192
rbio1.5	303.4698	11.6720	303.4689	11.6273	303.4693	11.5800	303.4656	11.5384
rbio2.2	303.4704	11.6826	303.4698	11.6931	303.4713	11.7510	303.4749	11.8095
rbio2.4	303.4702	11.6745	303.4711	11.6420	303.4700	11.6368	303.4732	11.6461
rbio2.6	303.4702	11.6750	303.4698	11.6352	303.4681	11.6048	303.4667	11.5986
rbio2.8	303.4700	11.6756	303.4712	11.6348	303.4703	11.6040	303.4684	11.6004
rbio3.1	303.4698	11.8663	303.4691	13.0826	303.4683	17.6436	303.4682	25.4798
rbio3.3	303.4698	11.7035	303.4675	11.8878	303.4662	12.4345	303.4686	13.0754
rbio3.5	303.4698	11.6877	303.4699	11.7391	303.4647	11.9565	303.4662	12.2065
rbio3.7	303.4698	11.6838	303.4699	11.7003	303.4701	11.8383	303.4646	11.9961
rbio3.9	303.4698	11.6824	303.4699	11.6778	303.4701	11.7726	303.4681	11.8911
rbio4.4	303.4703	11.6765	303.4719	11.6464	303.4712	11.6218	303.4748	11.6008
rbio5.5	303.4691	11.6762	303.4677	11.6439	303.4670	11.6170	303.4652	11.6003
rbio6.8	303.4700	11.6743	303.4711	11.6327	303.4703	11.5886	303.4691	11.5587
dmey	303.4695	11.6721	303.4693	11.6199	303.4684	11.5557	303.4665	11.5142

Table A.36: Results of Hypothesis Testing with hard thresholding for DEM generation on City Site 2 including outliers

City Site 2, $\mu = 292.9100$, $\sigma = 3.5970$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.8965	3.5635	292.8726	3.4909	292.9627	3.5737	293.0526	3.4500
db2	292.7831	3.4080	292.7675	3.2869	292.7407	3.1920	292.6735	2.9808
db4	292.7222	3.3551	292.6564	3.1830	292.5078	2.9313	292.4836	2.8052
db6	292.7026	3.3310	292.6005	3.1110	292.4917	2.8759	292.4374	2.8167
db8	292.6667	3.2742	292.5206	2.9935	292.3752	2.7326	292.2966	2.6574
db10	292.6845	3.3036	292.5076	2.9913	292.3704	2.7265	292.2463	2.6379
db12	292.6473	3.2542	292.4979	3.0093	292.3194	2.6253	292.1820	2.5936
db20	292.6114	3.2140	292.3783	2.7720	292.2550	2.5935	292.1388	2.5687
sym2	292.7831	3.4080	292.7675	3.2869	292.7407	3.1920	292.6735	2.9808
sym4	292.7601	3.4102	292.6841	3.2482	292.6032	2.9386	292.5633	2.8042
sym6	292.7454	3.3840	292.6331	3.1503	292.5916	2.9747	292.5309	2.7933
sym8	292.7333	3.3718	292.6269	3.1478	292.5327	2.9283	292.5161	2.8305
sym10	292.7142	3.3430	292.5958	3.1029	292.4900	2.8270	292.4533	2.7624
sym12	292.6766	3.3160	292.5791	3.1156	292.4495	2.7561	292.3589	2.6980
sym20	292.6462	3.2654	292.5332	3.0455	292.4111	2.7182	292.3142	2.6028
coif1	292.7645	3.3954	292.7324	3.2823	292.6759	3.1064	292.6269	2.8595
coif2	292.7820	3.4378	292.7274	3.2874	292.6281	3.0419	292.6178	2.8779
coif3	292.7567	3.3946	292.6785	3.2135	292.5349	2.8641	292.4372	2.6854
coif4	292.7263	3.3642	292.6474	3.1744	292.5379	2.9237	292.4488	2.8151
coif5	292.7095	3.3428	292.6153	3.1183	292.4911	2.8163	292.3793	2.5962
bior1.1	292.8965	3.5635	292.8726	3.4909	292.9627	3.5737	293.0526	3.4500
bior1.3	292.8203	3.4918	292.7309	3.3559	292.6346	3.0884	292.6726	3.1817
bior1.5	292.7726	3.4286	292.7108	3.3264	292.5390	3.0596	292.4604	2.8750
bior2.2	292.8645	3.5128	292.9127	3.5001	292.8844	3.3961	292.8375	3.2002
bior2.4	292.8145	3.4471	292.8279	3.4116	292.8063	3.2760	292.7638	3.1481
bior2.6	292.7584	3.3715	292.7843	3.3522	292.7394	3.1759	292.7489	3.0914
bior2.8	292.7381	3.3451	292.7823	3.3409	292.6695	3.0870	292.6657	2.9747
bior3.1	292.7300	3.3614	292.6029	3.1120	292.4114	2.8460	292.2615	2.8954
bior3.3	292.7631	3.4094	292.6995	3.2597	292.5830	2.9145	292.5400	3.0166
bior3.5	292.7456	3.3825	292.6896	3.2188	292.5536	3.0261	292.4778	2.9766
bior3.7	292.7466	3.3846	292.6849	3.2368	292.5182	2.8734	292.3749	2.7354
bior3.9	292.7435	3.3803	292.6458	3.1876	292.5918	2.9401	292.4422	2.7584
bior4.4	292.7957	3.4483	292.7323	3.3059	292.7271	3.1641	292.7048	3.0340
bior5.5	292.7466	3.4070	292.6622	3.1545	292.5123	2.9073	292.4323	2.6875
bior6.8	292.7448	3.3691	292.6637	3.2187	292.6269	3.0413	292.5867	2.8606
rbio1.1	292.8965	3.5635	292.8726	3.4909	292.9627	3.5737	293.0526	3.4500
rbio1.3	292.7407	3.3546	292.6823	3.2271	292.6608	3.0724	292.6358	2.9680
rbio1.5	292.7241	3.3635	292.6693	3.1671	292.5779	2.9715	292.5346	2.8500
rbio2.2	292.6643	3.2935	292.4153	2.9290	292.2901	2.5588	292.0054	2.1158
rbio2.4	292.7708	3.3908	292.5518	3.0241	292.3809	2.6627	292.1911	2.4119
rbio2.6	292.7525	3.3889	292.5876	3.0836	292.3340	2.6198	292.2147	2.4425
rbio2.8	292.7334	3.3640	292.5267	3.0052	292.3433	2.6237	292.1879	2.4930
rbio3.1	292.1987	2.7622	291.4852	2.0921	288.1218	7.3238	285.7713	13.8108
rbio3.3	292.5348	3.0907	292.0263	2.4087	291.4896	2.0606	290.8732	2.0736
rbio3.5	292.6182	3.2440	292.2692	2.7285	291.8590	2.1693	291.4300	1.9447
rbio3.7	292.6642	3.3061	292.3714	2.7493	291.9942	2.1710	291.6138	2.1401
rbio3.9	292.6786	3.3124	292.3879	2.8074	292.0508	2.2694	291.5956	2.0794
rbio4.4	292.7278	3.3368	292.5908	3.1104	292.5231	2.8901	292.5008	2.7502
rbio5.5	292.7410	3.3843	292.6198	3.1215	292.5374	2.8682	292.5135	2.7346
rbio6.8	292.7365	3.3745	292.5833	3.0920	292.4979	2.8935	292.4481	2.7795
dmey	292.6335	3.2538	292.4897	2.9237	292.4326	2.7888	292.3312	2.6561

Table A.37: Results of Hypothesis Testing with hard thresholding for DSM generation on City Site 2 excluding outliers

City Site 2, $\mu = 303.4662$, $\sigma = 11.5645$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4662	11.5179	303.4662	11.4635	303.4661	11.4097	303.4669	11.3750
db2	303.4653	11.5302	303.4661	11.4821	303.4656	11.4270	303.4673	11.3901
db4	303.4660	11.5363	303.4660	11.4900	303.4650	11.4346	303.4666	11.3914
db6	303.4665	11.5377	303.4664	11.4891	303.4665	11.4331	303.4684	11.3877
db8	303.4658	11.5362	303.4656	11.4849	303.4659	11.4239	303.4648	11.3784
db10	303.4657	11.5352	303.4651	11.4819	303.4627	11.4201	303.4636	11.3733
db12	303.4668	11.5358	303.4659	11.4813	303.4647	11.4187	303.4662	11.3699
db20	303.4665	11.5348	303.4661	11.4784	303.4680	11.4112	303.4705	11.3566
sym2	303.4653	11.5302	303.4661	11.4821	303.4656	11.4270	303.4673	11.3901
sym4	303.4667	11.5382	303.4666	11.4940	303.4654	11.4435	303.4650	11.4033
sym6	303.4664	11.5390	303.4665	11.4946	303.4653	11.4413	303.4612	11.4097
sym8	303.4662	11.5392	303.4661	11.4934	303.4657	11.4378	303.4667	11.3988
sym10	303.4662	11.5393	303.4656	11.4936	303.4661	11.4412	303.4663	11.4005
sym12	303.4663	11.5381	303.4666	11.4924	303.4673	11.4405	303.4653	11.3960
sym20	303.4663	11.5382	303.4666	11.4915	303.4670	11.4346	303.4703	11.4075
coif1	303.4669	11.5327	303.4665	11.4831	303.4674	11.4333	303.4708	11.4010
coif2	303.4666	11.5384	303.4663	11.4937	303.4640	11.4413	303.4626	11.4059
coif3	303.4664	11.5393	303.4661	11.4942	303.4647	11.4405	303.4628	11.3999
coif4	303.4663	11.5394	303.4660	11.4939	303.4654	11.4369	303.4634	11.3967
coif5	303.4662	11.5396	303.4660	11.4937	303.4658	11.4401	303.4649	11.3977
bior1.1	303.4662	11.5179	303.4662	11.4635	303.4661	11.4097	303.4669	11.3750
bior1.3	303.4662	11.5652	303.4662	11.5658	303.4646	11.5642	303.4665	11.5576
bior1.5	303.4662	11.5744	303.4660	11.5879	303.4663	11.5940	303.4649	11.5897
bior2.2	303.4668	11.5260	303.4662	11.4717	303.4688	11.4242	303.4704	11.3969
bior2.4	303.4668	11.5361	303.4685	11.4961	303.4688	11.4582	303.4717	11.4349
bior2.6	303.4668	11.5405	303.4665	11.5042	303.4650	11.4685	303.4639	11.4518
bior2.8	303.4668	11.5429	303.4677	11.5099	303.4676	11.4816	303.4673	11.4645
bior3.1	303.4662	11.5390	303.4661	11.5614	303.4644	11.6170	303.4692	11.6654
bior3.3	303.4662	11.5352	303.4664	11.5234	303.4649	11.5447	303.4591	11.5615
bior3.5	303.4662	11.5365	303.4662	11.5188	303.4676	11.5272	303.4668	11.5367
bior3.7	303.4662	11.5377	303.4660	11.5200	303.4666	11.5249	303.4644	11.5375
bior3.9	303.4662	11.5386	303.4661	11.5199	303.4660	11.5251	303.4662	11.5483
bior4.4	303.4665	11.5364	303.4676	11.4880	303.4678	11.4251	303.4705	11.3746
bior5.5	303.4660	11.5373	303.4653	11.4842	303.4636	11.4080	303.4587	11.3555
bior6.8	303.4663	11.5386	303.4668	11.4932	303.4676	11.4395	303.4676	11.3970
rbio1.1	303.4662	11.5179	303.4662	11.4635	303.4661	11.4097	303.4669	11.3750
rbio1.3	303.4662	11.5312	303.4667	11.4814	303.4654	11.4317	303.4649	11.3852
rbio1.5	303.4662	11.5375	303.4655	11.4962	303.4652	11.4489	303.4611	11.4088
rbio2.2	303.4668	11.5477	303.4664	11.5607	303.4673	11.6200	303.4717	11.6795
rbio2.4	303.4666	11.5399	303.4672	11.5093	303.4659	11.5044	303.4685	11.5124
rbio2.6	303.4664	11.5403	303.4663	11.5031	303.4644	11.4736	303.4636	11.4668
rbio2.8	303.4664	11.5411	303.4674	11.5022	303.4665	11.4715	303.4637	11.4687
rbio3.1	303.4662	11.7232	303.4656	12.9197	303.4657	17.4653	303.4648	25.6441
rbio3.3	303.4662	11.5650	303.4640	11.7449	303.4627	12.2920	303.4643	12.9276
rbio3.5	303.4662	11.5505	303.4662	11.5989	303.4609	11.8200	303.4627	12.0753
rbio3.7	303.4662	11.5470	303.4663	11.5617	303.4670	11.6997	303.4615	11.8568
rbio3.9	303.4662	11.5459	303.4662	11.5413	303.4666	11.6403	303.4659	11.7606
rbio4.4	303.4666	11.5424	303.4677	11.5135	303.4669	11.4897	303.4704	11.4679
rbio5.5	303.4656	11.5422	303.4644	11.5119	303.4637	11.4855	303.4645	11.4705
rbio6.8	303.4664	11.5405	303.4674	11.5010	303.4666	11.4575	303.4651	11.4270
dmey	303.4660	11.5395	303.4658	11.4903	303.4649	11.4253	303.4626	11.3851

Table A.38: Results of Hypothesis Testing with hard thresholding for DEM generation on City Site 2 excluding outliers

City Site 2, $\mu = 292.9549$, $\sigma = 3.6377$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.9509	3.6276	292.9235	3.5390	293.0521	3.6321	293.1390	3.5636
db2	292.8669	3.5335	292.8268	3.3242	292.7362	3.1150	292.6964	2.9623
db4	292.7871	3.4104	292.6843	3.2172	292.5367	2.9336	292.5469	2.8513
db6	292.7782	3.4153	292.6667	3.2057	292.5293	2.9116	292.4597	2.8250
db8	292.7231	3.3469	292.6154	3.1176	292.4105	2.7149	292.3659	2.6639
db10	292.7593	3.4060	292.6003	3.0987	292.4107	2.6832	292.2677	2.5558
db12	292.7357	3.3606	292.5542	3.0334	292.3907	2.7159	292.2985	2.6816
db20	292.6883	3.2960	292.5117	2.9624	292.2697	2.5680	292.1683	2.6154
sym2	292.8669	3.5335	292.8268	3.3242	292.7362	3.1150	292.6964	2.9623
sym4	292.8227	3.4675	292.7486	3.3243	292.6667	3.0289	292.6169	2.8461
sym6	292.8098	3.4509	292.6908	3.2167	292.5875	2.9255	292.5543	2.7776
sym8	292.8012	3.4479	292.6749	3.2003	292.5768	3.0226	292.5149	2.7853
sym10	292.7946	3.4465	292.6556	3.1923	292.6138	3.0527	292.5553	2.8776
sym12	292.7306	3.3621	292.6741	3.2355	292.5538	2.9084	292.4290	2.7352
sym20	292.7064	3.3352	292.6278	3.1788	292.4757	2.8212	292.3329	2.5914
coif1	292.8415	3.4865	292.7493	3.2651	292.7253	3.1247	292.6870	2.9090
coif2	292.8347	3.4767	292.7701	3.3310	292.6551	3.0641	292.6926	2.9482
coif3	292.8037	3.4356	292.7148	3.2660	292.6035	2.9433	292.4894	2.7234
coif4	292.7927	3.4222	292.6636	3.1922	292.5803	2.9930	292.4914	2.8300
coif5	292.7820	3.4174	292.6503	3.1682	292.5587	2.8978	292.4283	2.6938
bior1.1	292.9509	3.6276	292.9235	3.5390	293.0521	3.6321	293.1390	3.5636
bior1.3	292.8874	3.5809	292.8274	3.5081	292.7261	3.1824	292.7454	3.1661
bior1.5	292.8538	3.5488	292.7634	3.4066	292.5962	3.2018	292.5341	2.9558
bior2.2	292.9307	3.5942	292.9540	3.5516	292.9069	3.3760	292.8958	3.2237
bior2.4	292.8664	3.4958	292.8950	3.4823	292.8901	3.3845	292.8424	3.2588
bior2.6	292.8272	3.4470	292.8514	3.4150	292.7771	3.2446	292.7874	3.1304
bior2.8	292.8168	3.4225	292.8536	3.4182	292.7524	3.2198	292.7382	3.0171
bior3.1	292.8065	3.4585	292.6930	3.2577	292.5388	3.0484	292.4264	2.9328
bior3.3	292.8155	3.4535	292.7618	3.3117	292.6587	2.9751	292.5267	2.9258
bior3.5	292.8176	3.4626	292.7742	3.3500	292.5664	2.9840	292.4968	2.9360
bior3.7	292.8050	3.4534	292.7669	3.3508	292.6193	3.0027	292.5467	3.0036
bior3.9	292.7958	3.4351	292.7174	3.2658	292.6782	3.0354	292.4716	2.7709
bior4.4	292.8422	3.4859	292.8144	3.3892	292.8169	3.2732	292.6959	2.9923
bior5.5	292.8176	3.4781	292.7373	3.2894	292.5715	2.9459	292.4317	2.6451
bior6.8	292.8086	3.4485	292.7559	3.3309	292.6728	3.1119	292.6461	2.9473
rbio1.1	292.9509	3.6276	292.9235	3.5390	293.0521	3.6321	293.1390	3.5636
rbio1.3	292.7809	3.3957	292.7716	3.3701	292.7390	3.1812	292.6910	3.0269
rbio1.5	292.7630	3.3761	292.6964	3.2123	292.6222	3.0473	292.5746	2.9346
rbio2.2	292.7252	3.3153	292.4886	2.9339	292.3006	2.4171	292.1047	2.2028
rbio2.4	292.8242	3.4653	292.6223	3.1284	292.4248	2.7320	292.2973	2.4880
rbio2.6	292.8021	3.4406	292.5950	3.0829	292.3854	2.7376	292.3279	2.5837
rbio2.8	292.7826	3.4153	292.6119	3.1444	292.4446	2.7756	292.2853	2.5857
rbio3.1	292.2128	2.7552	291.6265	2.1082	289.3127	6.6307	276.2405	32.5903
rbio3.3	292.6169	3.1801	292.1249	2.4885	291.7632	2.0817	291.0589	2.1562
rbio3.5	292.6905	3.2862	292.3214	2.7800	291.8994	2.2053	291.4673	1.9625
rbio3.7	292.7274	3.3317	292.4426	2.8139	292.1211	2.3025	291.7332	2.1045
rbio3.9	292.7539	3.3781	292.4260	2.8932	292.0725	2.2796	291.6382	2.1542
rbio4.4	292.8126	3.4341	292.6448	3.1967	292.5921	2.9391	292.5490	2.7681
rbio5.5	292.7807	3.4500	292.6923	3.2104	292.5981	2.9891	292.5875	2.8396
rbio6.8	292.7811	3.4041	292.6359	3.1415	292.5551	2.9841	292.5508	2.9200
dmey	292.7080	3.3285	292.5663	3.0701	292.4691	2.8483	292.3580	2.6583

Table A.39: Results of Hypothesis Testing with soft thresholding for DSM generation on City Site 2 including outliers

City Site 2, $\mu = 303.4698$, $\sigma = 11.7044$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4698	11.6119	303.4698	11.4660	303.4701	11.2185	303.4850	10.8861
db2	303.4685	11.6438	303.4705	11.5432	303.4767	11.3400	303.4722	11.0025
db4	303.4696	11.6546	303.4681	11.5653	303.4743	11.3858	303.4768	11.0673
db6	303.4702	11.6568	303.4686	11.5686	303.4709	11.3923	303.4781	11.0548
db8	303.4695	11.6564	303.4667	11.5691	303.4666	11.3942	303.4706	11.0512
db10	303.4695	11.6563	303.4699	11.5684	303.4653	11.3951	303.4679	11.0679
db12	303.4701	11.6577	303.4697	11.5707	303.4665	11.3968	303.4747	11.0565
db20	303.4701	11.6579	303.4688	11.5696	303.4668	11.3994	303.4683	11.0473
sym2	303.4685	11.6438	303.4705	11.5432	303.4767	11.3400	303.4722	11.0025
sym4	303.4704	11.6553	303.4704	11.5639	303.4656	11.3908	303.4666	11.0730
sym6	303.4702	11.6573	303.4702	11.5727	303.4693	11.3980	303.4601	11.0881
sym8	303.4701	11.6582	303.4697	11.5707	303.4726	11.3980	303.4651	11.0851
sym10	303.4701	11.6587	303.4700	11.5743	303.4703	11.4093	303.4638	11.0905
sym12	303.4695	11.6578	303.4707	11.5720	303.4695	11.4088	303.4667	11.0837
sym20	303.4695	11.6585	303.4703	11.5732	303.4690	11.4089	303.4755	11.1011
coif1	303.4709	11.6470	303.4689	11.5427	303.4692	11.3494	303.4771	11.0122
coif2	303.4704	11.6556	303.4689	11.5652	303.4692	11.3867	303.4670	11.0731
coif3	303.4703	11.6576	303.4690	11.5696	303.4666	11.4012	303.4630	11.0822
coif4	303.4702	11.6584	303.4692	11.5715	303.4720	11.4004	303.4655	11.0736
coif5	303.4701	11.6589	303.4693	11.5725	303.4693	11.4084	303.4669	11.0810
bior1.1	303.4698	11.6119	303.4698	11.4660	303.4701	11.2185	303.4850	10.8861
bior1.3	303.4698	11.6793	303.4713	11.6263	303.4680	11.4875	303.4754	11.2098
bior1.5	303.4698	11.6933	303.4671	11.6618	303.4692	11.5472	303.4641	11.2747
bior2.2	303.4705	11.6406	303.4678	11.5300	303.4705	11.3371	303.4754	11.0205
bior2.4	303.4707	11.6532	303.4738	11.5627	303.4716	11.3889	303.4761	11.1018
bior2.6	303.4708	11.6588	303.4688	11.5706	303.4717	11.4058	303.4676	11.1320
bior2.8	303.4709	11.6621	303.4726	11.5806	303.4733	11.4260	303.4693	11.1474
bior3.1	303.4698	11.6510	303.4733	11.5882	303.4728	11.5391	303.4933	11.4153
bior3.3	303.4698	11.6512	303.4705	11.5797	303.4743	11.4593	303.4598	11.2751
bior3.5	303.4698	11.6538	303.4702	11.5734	303.4691	11.4514	303.4652	11.2294
bior3.7	303.4698	11.6557	303.4693	11.5820	303.4711	11.4552	303.4689	11.2268
bior3.9	303.4698	11.6571	303.4701	11.5783	303.4698	11.4542	303.4698	11.2078
bior4.4	303.4703	11.6535	303.4734	11.5638	303.4702	11.3781	303.4748	11.0402
bior5.5	303.4694	11.6550	303.4675	11.5668	303.4665	11.3734	303.4595	11.0172
bior6.8	303.4702	11.6574	303.4713	11.5716	303.4725	11.4034	303.4670	11.0859
rbio1.1	303.4698	11.6119	303.4698	11.4660	303.4701	11.2185	303.4850	10.8861
rbio1.3	303.4698	11.6413	303.4731	11.5327	303.4735	11.3300	303.4734	11.0056
rbio1.5	303.4698	11.6507	303.4664	11.5607	303.4728	11.3761	303.4625	11.0537
rbio2.2	303.4711	11.6594	303.4684	11.6043	303.4694	11.5071	303.4803	11.2452
rbio2.4	303.4705	11.6534	303.4736	11.5683	303.4686	11.3953	303.4756	11.0851
rbio2.6	303.4702	11.6548	303.4687	11.5655	303.4700	11.3828	303.4606	11.0739
rbio2.8	303.4701	11.6562	303.4719	11.5692	303.4725	11.3948	303.4625	11.0776
rbio3.1	303.4698	11.8808	303.4640	13.2833	303.4576	18.7554	303.4882	30.9390
rbio3.3	303.4698	11.6726	303.4674	11.6957	303.4654	11.8781	303.4687	12.0860
rbio3.5	303.4698	11.6615	303.4703	11.6023	303.4652	11.5254	303.4603	11.3464
rbio3.7	303.4698	11.6596	303.4696	11.5888	303.4683	11.4610	303.4643	11.2091
rbio3.9	303.4698	11.6593	303.4701	11.5793	303.4711	11.4357	303.4702	11.1565
rbio4.4	303.4707	11.6592	303.4739	11.5802	303.4694	11.4175	303.4747	11.1230
rbio5.5	303.4687	11.6605	303.4667	11.5821	303.4664	11.4270	303.4612	11.1714
rbio6.8	303.4703	11.6586	303.4716	11.5745	303.4723	11.4079	303.4652	11.0981
dmey	303.4699	11.6598	303.4696	11.5741	303.4687	11.4051	303.4696	11.0694

Table A.40: Results of Hypothesis Testing with soft thresholding for DEM generation on City Site 2 including outliers

City Site 2, $\mu = 292.9100$, $\sigma = 3.5970$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.9308	3.6244	292.9611	3.6340	293.1880	3.8402	293.3343	3.8016
db2	292.7699	3.4011	292.7922	3.3170	292.7658	3.2653	292.7456	3.1713
db4	292.7181	3.3513	292.6711	3.2109	292.6225	3.1102	292.6608	3.0171
db6	292.7311	3.3848	292.6606	3.2151	292.6028	3.0575	292.6315	3.0045
db8	292.6623	3.2685	292.5901	3.1097	292.5574	3.0329	292.5907	2.9465
db10	292.6969	3.3332	292.6079	3.1340	292.5712	3.0317	292.5838	2.9823
db12	292.6738	3.3047	292.6059	3.1553	292.5487	2.9541	292.5907	3.0052
db20	292.6558	3.3036	292.5335	2.9995	292.5026	2.9052	292.5097	2.8605
sym2	292.7699	3.4011	292.7922	3.3170	292.7658	3.2653	292.7456	3.1713
sym4	292.7318	3.3653	292.6637	3.2200	292.6878	3.1141	292.7561	3.1073
sym6	292.7418	3.3777	292.6443	3.1750	292.6326	3.0707	292.7066	3.0490
sym8	292.7318	3.3784	292.6546	3.1910	292.6229	3.1138	292.6992	3.0654
sym10	292.7106	3.3464	292.6360	3.1963	292.6433	3.1329	292.7127	3.1150
sym12	292.6768	3.3113	292.6430	3.2262	292.6096	2.9991	292.6442	3.1037
sym20	292.6653	3.3167	292.6022	3.1475	292.5857	2.9898	292.6095	2.9690
coif1	292.7499	3.3828	292.6771	3.2080	292.5944	3.0031	292.5776	2.8667
coif2	292.7825	3.4381	292.7200	3.2712	292.6872	3.1792	292.7707	3.1378
coif3	292.7415	3.3797	292.6885	3.2281	292.6597	3.0918	292.6658	3.0245
coif4	292.7297	3.3811	292.6534	3.1771	292.6276	3.1091	292.6760	3.0838
coif5	292.7188	3.3650	292.6289	3.1372	292.6259	3.0568	292.6055	2.9517
bior1.1	292.9308	3.6244	292.9611	3.6340	293.1880	3.8402	293.3343	3.8016
bior1.3	292.8572	3.5730	292.7872	3.4651	292.8241	3.4116	292.9288	3.4862
bior1.5	292.8000	3.5056	292.7440	3.3982	292.7205	3.3721	292.8242	3.3667
bior2.2	292.8444	3.4937	292.9084	3.5220	292.9316	3.4636	292.9689	3.3301
bior2.4	292.7809	3.4140	292.7722	3.3428	292.9062	3.4352	293.0240	3.4407
bior2.6	292.7389	3.3647	292.7788	3.3418	292.8278	3.3634	292.9531	3.3716
bior2.8	292.7239	3.3539	292.7443	3.3391	292.8109	3.3282	292.8301	3.1500
bior3.1	292.7399	3.3738	292.6528	3.2403	292.5403	3.0323	292.5232	3.1854
bior3.3	292.7676	3.4330	292.7210	3.2498	292.7242	3.1458	292.7899	3.2014
bior3.5	292.7506	3.4086	292.7646	3.3458	292.8512	3.4170	292.8277	3.2906
bior3.7	292.7397	3.3946	292.7223	3.2526	292.7534	3.1935	292.7652	3.1783
bior3.9	292.7337	3.3862	292.7706	3.3471	292.8476	3.3518	292.8199	3.2668
bior4.4	292.7916	3.4394	292.6930	3.2395	292.7365	3.1931	292.7998	3.1386
bior5.5	292.7334	3.3810	292.6720	3.1848	292.6070	3.0389	292.6525	2.9706
bior6.8	292.7520	3.3968	292.6540	3.2027	292.6726	3.1668	292.7918	3.1961
rbio1.1	292.9308	3.6244	292.9611	3.6340	293.1880	3.8402	293.3343	3.8016
rbio1.3	292.7595	3.3846	292.6982	3.2733	292.7076	3.2123	292.8261	3.2677
rbio1.5	292.7509	3.3844	292.6846	3.1899	292.6604	3.1605	292.7199	3.1288
rbio2.2	292.6176	3.2204	292.4256	2.9010	292.1100	2.2309	291.8830	2.1156
rbio2.4	292.7825	3.4104	292.6117	3.1296	292.4961	2.8895	292.5010	2.8520
rbio2.6	292.7867	3.4488	292.6769	3.2196	292.5445	2.9888	292.5118	2.8307
rbio2.8	292.7653	3.4195	292.6117	3.1367	292.5830	3.0493	292.6077	3.0484
rbio3.1	292.1269	2.6589	291.4159	2.0987	290.0385	5.6931	252.6760	68.3568
rbio3.3	292.5644	3.1552	292.2000	2.6807	291.6852	2.2374	291.1953	2.1014
rbio3.5	292.6501	3.2794	292.4385	2.9449	292.0538	2.4256	291.8567	2.2337
rbio3.7	292.6975	3.3430	292.5016	2.9664	292.2827	2.6211	292.0965	2.4853
rbio3.9	292.6971	3.3362	292.5915	3.1642	292.3924	2.8058	292.1805	2.5398
rbio4.4	292.7209	3.3325	292.6137	3.1320	292.5924	3.0282	292.6584	3.0348
rbio5.5	292.7201	3.3535	292.6423	3.1358	292.6326	3.0927	292.7507	3.1362
rbio6.8	292.7430	3.3893	292.6225	3.1628	292.6129	3.0925	292.6730	3.0907
dmey	292.6427	3.2782	292.5399	3.0024	292.5816	3.0389	292.5559	2.9529

Table A.41: Results of Hypothesis Testing with soft thresholding for DSM generation on City Site 2 excluding outliers

City Site 2, $\mu = 303.4662$, $\sigma = 11.5645$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	303.4662	11.4839	303.4662	11.3442	303.4665	11.0979	303.4811	10.7619
db2	303.4650	11.5153	303.4669	11.4217	303.4729	11.2199	303.4688	10.8789
db4	303.4660	11.5257	303.4646	11.4435	303.4702	11.2674	303.4723	10.9459
db6	303.4666	11.5282	303.4651	11.4479	303.4674	11.2749	303.4741	10.9360
db8	303.4659	11.5278	303.4635	11.4484	303.4633	11.2777	303.4672	10.9317
db10	303.4658	11.5277	303.4662	11.4478	303.4616	11.2794	303.4640	10.9486
db12	303.4665	11.5291	303.4661	11.4507	303.4630	11.2818	303.4707	10.9394
db20	303.4665	11.5295	303.4653	11.4498	303.4636	11.2846	303.4651	10.9307
sym2	303.4650	11.5153	303.4669	11.4217	303.4729	11.2199	303.4688	10.8789
sym4	303.4668	11.5267	303.4667	11.4424	303.4622	11.2722	303.4638	10.9514
sym6	303.4667	11.5286	303.4667	11.4512	303.4656	11.2797	303.4572	10.9683
sym8	303.4666	11.5295	303.4661	11.4495	303.4687	11.2803	303.4615	10.9645
sym10	303.4665	11.5299	303.4664	11.4529	303.4667	11.2910	303.4602	10.9684
sym12	303.4659	11.5286	303.4670	11.4507	303.4659	11.2911	303.4629	10.9614
sym20	303.4660	11.5294	303.4666	11.4521	303.4653	11.2910	303.4718	10.9797
coif1	303.4672	11.5187	303.4653	11.4210	303.4658	11.2299	303.4743	10.8892
coif2	303.4668	11.5270	303.4654	11.4437	303.4654	11.2686	303.4636	10.9534
coif3	303.4667	11.5289	303.4655	11.4482	303.4632	11.2831	303.4595	10.9614
coif4	303.4666	11.5297	303.4656	11.4502	303.4681	11.2827	303.4618	10.9516
coif5	303.4666	11.5301	303.4658	11.4513	303.4657	11.2902	303.4636	10.9584
bior1.1	303.4662	11.4839	303.4662	11.3442	303.4665	11.0979	303.4811	10.7619
bior1.3	303.4662	11.5501	303.4677	11.5044	303.4644	11.3686	303.4712	11.0867
bior1.5	303.4662	11.5635	303.4637	11.5397	303.4656	11.4296	303.4608	11.1534
bior2.2	303.4669	11.5127	303.4643	11.4072	303.4671	11.2151	303.4728	10.8946
bior2.4	303.4672	11.5248	303.4701	11.4404	303.4680	11.2689	303.4729	10.9799
bior2.6	303.4672	11.5300	303.4653	11.4483	303.4676	11.2850	303.4639	11.0104
bior2.8	303.4673	11.5330	303.4690	11.4582	303.4697	11.3054	303.4654	11.0243
bior3.1	303.4662	11.5198	303.4694	11.4600	303.4686	11.3993	303.4878	11.2753
bior3.3	303.4662	11.5219	303.4670	11.4550	303.4694	11.3370	303.4551	11.1507
bior3.5	303.4662	11.5247	303.4666	11.4488	303.4659	11.3290	303.4623	11.1048
bior3.7	303.4662	11.5266	303.4659	11.4577	303.4677	11.3329	303.4655	11.0995
bior3.9	303.4662	11.5279	303.4664	11.4543	303.4664	11.3327	303.4667	11.0818
bior4.4	303.4667	11.5249	303.4697	11.4423	303.4665	11.2604	303.4714	10.9193
bior5.5	303.4659	11.5256	303.4642	11.4457	303.4629	11.2563	303.4565	10.8989
bior6.8	303.4667	11.5287	303.4677	11.4503	303.4688	11.2850	303.4633	10.9651
rbio1.1	303.4662	11.4839	303.4662	11.3442	303.4665	11.0979	303.4811	10.7619
rbio1.3	303.4662	11.5126	303.4692	11.4103	303.4693	11.2102	303.4688	10.8831
rbio1.5	303.4662	11.5216	303.4630	11.4382	303.4688	11.2561	303.4586	10.9307
rbio2.2	303.4675	11.5302	303.4649	11.4817	303.4660	11.3871	303.4774	11.1242
rbio2.4	303.4669	11.5246	303.4698	11.4459	303.4650	11.2765	303.4723	10.9634
rbio2.6	303.4666	11.5260	303.4652	11.4437	303.4660	11.2640	303.4574	10.9534
rbio2.8	303.4665	11.5273	303.4682	11.4471	303.4689	11.2754	303.4587	10.9563
rbio3.1	303.4662	11.7460	303.4608	13.1483	303.4542	18.6120	303.4843	30.8285
rbio3.3	303.4662	11.5418	303.4640	11.5685	303.4619	11.7537	303.4649	11.9667
rbio3.5	303.4662	11.5315	303.4666	11.4786	303.4618	11.4052	303.4571	11.2263
rbio3.7	303.4662	11.5298	303.4662	11.4655	303.4650	11.3393	303.4609	11.0848
rbio3.9	303.4662	11.5296	303.4664	11.4566	303.4672	11.3157	303.4667	11.0345
rbio4.4	303.4671	11.5302	303.4701	11.4577	303.4657	11.2987	303.4714	11.0011
rbio5.5	303.4651	11.5309	303.4633	11.4599	303.4626	11.3074	303.4583	11.0504
rbio6.8	303.4667	11.5297	303.4680	11.4528	303.4687	11.2894	303.4615	10.9772
dmey	303.4664	11.5311	303.4660	11.4532	303.4651	11.2879	303.4658	10.9498

Table A.42: Results of Hypothesis Testing with soft thresholding for DEM generation on City Site 2 excluding outliers

City Site 2, $\mu = 292.9549$, $\sigma = 3.6377$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	292.9681	3.6575	292.9708	3.6086	293.1985	3.7922	293.3196	3.7433
db2	292.8281	3.4679	292.8121	3.3213	292.7822	3.2419	292.7572	3.1600
db4	292.7708	3.3691	292.6808	3.2020	292.6373	3.1108	292.6913	3.0444
db6	292.7800	3.4235	292.7139	3.2701	292.6479	3.1226	292.6166	2.9927
db8	292.7277	3.3469	292.6334	3.1273	292.6097	3.1299	292.5962	2.9483
db10	292.7490	3.3671	292.6639	3.2130	292.6015	3.0428	292.6409	3.0469
db12	292.7520	3.3726	292.6147	3.1143	292.5772	2.9995	292.6130	3.0269
db20	292.7118	3.3306	292.5847	3.0796	292.5386	2.9612	292.5276	2.8773
sym2	292.8281	3.4679	292.8121	3.3213	292.7822	3.2419	292.7572	3.1600
sym4	292.8106	3.4542	292.7211	3.3002	292.7254	3.1546	292.7684	3.1107
sym6	292.8077	3.4506	292.6750	3.1972	292.6624	3.1087	292.6813	3.0077
sym8	292.8027	3.4457	292.7030	3.2538	292.6771	3.1763	292.6989	3.0752
sym10	292.7941	3.4345	292.6674	3.2038	292.6916	3.1887	292.6944	3.0615
sym12	292.7407	3.3840	292.6876	3.2637	292.6374	3.0368	292.6728	3.1181
sym20	292.7183	3.3509	292.6439	3.1967	292.6156	3.0425	292.6156	2.9500
coif1	292.8228	3.4881	292.6994	3.2176	292.6406	3.0639	292.5895	2.8574
coif2	292.8391	3.4895	292.7654	3.3331	292.7035	3.1753	292.7809	3.1361
coif3	292.8238	3.4729	292.7160	3.2532	292.6940	3.1335	292.6875	3.0563
coif4	292.8006	3.4366	292.6857	3.2082	292.6521	3.1402	292.6865	3.0915
coif5	292.7872	3.4210	292.6658	3.1790	292.6565	3.0973	292.6122	2.9476
bior1.1	292.9681	3.6575	292.9708	3.6086	293.1985	3.7922	293.3196	3.7433
bior1.3	292.9026	3.6117	292.8802	3.5652	292.8785	3.5052	292.9828	3.5099
bior1.5	292.8351	3.5387	292.7922	3.4354	292.7882	3.4863	292.8694	3.4288
bior2.2	292.8983	3.5527	292.9571	3.5707	292.9728	3.4855	293.0130	3.3655
bior2.4	292.8372	3.4774	292.8235	3.3816	292.9767	3.5168	293.0103	3.3757
bior2.6	292.8023	3.4462	292.8147	3.3693	292.8486	3.3565	292.9724	3.3732
bior2.8	292.7777	3.4093	292.7652	3.3065	292.8488	3.3727	292.8704	3.1880
bior3.1	292.8093	3.4602	292.7148	3.3142	292.6892	3.1779	292.6371	3.2458
bior3.3	292.8068	3.4264	292.7880	3.3514	292.7628	3.1964	292.8119	3.1842
bior3.5	292.8077	3.4305	292.8410	3.4564	292.8417	3.3520	292.8600	3.3053
bior3.7	292.7992	3.4172	292.7895	3.3429	292.8241	3.2886	292.8472	3.3036
bior3.9	292.7910	3.4145	292.8338	3.4326	292.8638	3.3724	292.8178	3.2530
bior4.4	292.8382	3.4770	292.7528	3.2893	292.7790	3.2355	292.8106	3.1338
bior5.5	292.8106	3.4505	292.7269	3.2408	292.6477	3.0738	292.6520	2.9633
bior6.8	292.8157	3.4570	292.7093	3.2506	292.7231	3.2123	292.7983	3.1940
rbio1.1	292.9681	3.6575	292.9708	3.6086	293.1985	3.7922	293.3196	3.7433
rbio1.3	292.7889	3.3998	292.7523	3.3116	292.7572	3.2661	292.8616	3.2914
rbio1.5	292.7972	3.4255	292.7108	3.2290	292.6626	3.1271	292.7515	3.1410
rbio2.2	292.6599	3.2587	292.4574	2.8978	292.1183	2.1923	291.8772	2.0887
rbio2.4	292.8399	3.4918	292.6619	3.1964	292.5646	2.9816	292.5048	2.8381
rbio2.6	292.8335	3.4763	292.7202	3.2747	292.5762	3.0135	292.5165	2.8484
rbio2.8	292.8054	3.4369	292.6817	3.2224	292.6225	3.0703	292.6253	3.0335
rbio3.1	292.1741	2.7057	291.5302	2.1029	290.1170	5.7876	249.0670	71.4461
rbio3.3	292.6467	3.2561	292.2583	2.7069	291.7678	2.2585	291.2041	2.0568
rbio3.5	292.7106	3.3098	292.4946	3.0229	292.0820	2.4576	291.8638	2.2135
rbio3.7	292.7416	3.3547	292.5520	3.0237	292.3263	2.7164	292.1350	2.5202
rbio3.9	292.7646	3.3924	292.6315	3.1853	292.4149	2.8082	292.2000	2.5243
rbio4.4	292.8164	3.4591	292.6458	3.1751	292.6292	3.0623	292.6729	3.0364
rbio5.5	292.7650	3.4149	292.6814	3.1898	292.6712	3.1331	292.7309	3.0853
rbio6.8	292.8086	3.4500	292.6577	3.1836	292.6686	3.1626	292.6948	3.1015
dmey	292.7324	3.3707	292.5904	3.0929	292.5995	3.0551	292.5437	2.9385

Table A.43: Results of GCV with hard thresholding for DSM generation on City Site 3 including outliers

City Site 3, $\mu = 318.7184$, $\sigma = 9.1404$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.7184	9.1402	318.7184	9.0094	318.7184	8.7767	318.7184	8.4141
db2	318.7185	9.0948	318.7204	9.0130	318.7250	8.8158	318.7418	8.4321
db4	318.7186	9.1023	318.7159	9.0242	318.7125	8.8454	318.7113	8.4716
db6	318.7178	9.1030	318.7189	9.0362	318.7163	8.8617	318.7236	8.5149
db8	318.7185	9.1042	318.7208	9.0364	318.7141	8.8599	318.7205	8.6966
db10	318.7189	9.1058	318.7158	9.0327	318.7136	8.8615	318.7109	8.6662
db12	318.7182	9.1056	318.7197	9.0397	318.7146	8.8639	318.7210	8.5118
db20	318.7181	9.1063	318.7176	9.0368	318.7221	8.8765	318.7238	8.5052
sym2	318.7185	9.0948	318.7204	9.0130	318.7250	8.8158	318.7418	8.4321
sym4	318.7180	9.1013	318.7129	9.0208	318.7167	8.8473	318.7315	8.4889
sym6	318.7179	9.1029	318.7243	9.0401	318.7121	8.8554	318.7178	8.5088
sym8	318.7179	9.1037	318.7152	9.0294	318.7082	8.8530	318.7136	8.5054
sym10	318.7179	9.1041	318.7215	9.0410	318.7267	8.8719	318.7115	8.4835
sym12	318.7188	9.1053	318.7159	9.0328	318.7186	8.8636	318.7107	8.4611
sym20	318.7188	9.1058	318.7164	9.0351	318.7204	8.8677	318.7268	8.6030
coif1	318.7182	9.0942	318.7146	9.0022	318.7218	8.8123	318.7432	8.4522
coif2	318.7179	9.1013	318.7146	9.0220	318.7064	8.8440	318.7156	8.4951
coif3	318.7179	9.1034	318.7154	9.0287	318.7239	8.8587	318.7105	8.4669
coif4	318.7179	9.1040	318.7158	9.0306	318.7118	8.8568	318.7139	8.4779
coif5	318.7179	9.1045	318.7160	9.0323	318.7235	8.8665	318.7302	8.5112
bior1.1	318.7184	9.1402	318.7184	9.0094	318.7184	8.7767	318.7184	8.4141
bior1.3	318.7184	9.1235	318.7170	9.0916	318.7164	8.9857	318.7155	8.7014
bior1.5	318.7184	9.1362	318.7193	9.1296	318.7161	9.0521	318.7178	8.7652
bior2.2	318.7179	9.0896	318.7156	8.9860	318.7237	8.7780	318.7491	8.3962
bior2.4	318.7181	9.0997	318.7240	9.0277	318.7163	8.8313	318.7219	8.4742
bior2.6	318.7182	9.1042	318.7150	9.0257	318.7058	8.8448	318.7164	8.4992
bior2.8	318.7183	9.1068	318.7225	9.0452	318.7247	8.8705	318.7124	8.4846
bior3.1	318.7184	9.0973	318.7202	9.0136	318.7224	8.9262	318.7251	8.9295
bior3.3	318.7184	9.0983	318.7205	9.0325	318.7068	8.8538	318.7072	8.5207
bior3.5	318.7184	9.1010	318.7154	9.0216	318.7153	8.8499	318.7126	8.5219
bior3.7	318.7184	9.1029	318.7226	9.0368	318.7276	8.8601	318.7187	8.5513
bior3.9	318.7184	9.1044	318.7147	9.0278	318.7124	8.8507	318.7148	8.5071
bior4.4	318.7178	9.0999	318.7239	9.0314	318.7160	8.8428	318.7225	8.4960
bior5.5	318.7191	9.1030	318.7220	9.0330	318.7099	8.8493	318.7133	8.4970
bior6.8	318.7178	9.1032	318.7213	9.0380	318.7247	8.8628	318.7124	8.4800
rbio1.1	318.7184	9.1402	318.7184	9.0094	318.7184	8.7767	318.7184	8.4141
rbio1.3	318.7184	9.0892	318.7146	8.9932	318.7099	8.7809	318.7061	8.3856
rbio1.5	318.7184	9.0974	318.7188	9.0120	318.7126	8.8222	318.7062	8.4353
rbio2.2	318.7187	9.1041	318.7142	9.0610	318.7233	8.9966	318.7441	8.8259
rbio2.4	318.7179	9.0988	318.7238	9.0359	318.7144	8.8615	318.7239	8.5277
rbio2.6	318.7177	9.1001	318.7148	9.0223	318.7067	8.8438	318.7160	8.5090
rbio2.8	318.7177	9.1016	318.7216	9.0349	318.7242	8.8677	318.7144	8.4919
rbio3.1	318.7184	9.3179	318.7152	10.7668	318.7013	16.9273	318.6637	37.0375
rbio3.3	318.7184	9.1173	318.7195	9.1556	318.7164	9.3645	318.7172	9.8874
rbio3.5	318.7184	9.1066	318.7136	9.0566	318.7164	8.9947	318.7148	9.1102
rbio3.7	318.7184	9.1050	318.7222	9.0494	318.7271	8.9324	318.7146	8.7961
rbio3.9	318.7184	9.1043	318.7147	9.0344	318.7126	8.9095	318.7150	8.6511
rbio4.4	318.7180	9.1044	318.7239	9.0479	318.7138	8.8756	318.7224	8.5527
rbio5.5	318.7187	9.1091	318.7226	9.0514	318.7108	8.8722	318.7149	8.5260
rbio6.8	318.7179	9.1039	318.7213	9.0408	318.7244	8.8720	318.7124	8.4951
dmey	318.7180	9.1059	318.7172	9.0365	318.7214	8.8715	318.7235	8.5190

Table A.44: Results of GCV with hard thresholding for DEM generation on City Site 3 including outliers

City Site 3, $\mu = 310.7064$, $\sigma = 4.4727$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.3806	4.2154	310.3216	4.2734	310.2930	4.1597	310.1231	4.2033
db2	310.5776	4.4141	310.5003	4.3762	310.3258	4.3236	310.2078	4.4595
db4	310.5688	4.4220	310.4459	4.3900	310.1605	4.2799	309.7788	4.3614
db6	310.5126	4.3541	310.3941	4.3103	310.1339	4.2914	309.7674	4.2326
db8	310.5281	4.3729	310.3992	4.3626	310.1032	4.3198	309.9121	4.3897
db10	310.5044	4.3908	310.3703	4.3608	310.0778	4.3255	309.6880	4.3862
db12	310.5159	4.3709	310.3499	4.3185	310.0604	4.3392	309.5500	4.2874
db20	310.4813	4.3655	310.3510	4.4060	310.0916	4.3644	309.7193	4.3316
sym2	310.5776	4.4141	310.5003	4.3762	310.3258	4.3236	310.2078	4.4595
sym4	310.5493	4.3808	310.4420	4.3501	310.2880	4.3483	309.8500	4.2103
sym6	310.5293	4.3616	310.4025	4.3299	310.1658	4.3188	309.7085	4.2191
sym8	310.5111	4.3524	310.3977	4.3524	310.2097	4.3710	309.7668	4.3187
sym10	310.4920	4.3459	310.3604	4.3349	310.1373	4.2557	309.7182	4.2586
sym12	310.4913	4.3683	310.3649	4.3038	310.1669	4.2868	309.7226	4.2896
sym20	310.4754	4.3556	310.3353	4.3046	310.1326	4.2937	309.6879	4.2520
coif1	310.5397	4.3557	310.4971	4.3963	310.3446	4.3727	310.0725	4.4018
coif2	310.5244	4.3556	310.4276	4.3668	310.2384	4.3448	309.8083	4.2808
coif3	310.5102	4.3470	310.4145	4.3781	310.2211	4.3133	309.8678	4.3495
coif4	310.4963	4.3423	310.4152	4.3858	310.1986	4.3610	310.0255	4.4234
coif5	310.4817	4.3273	310.4051	4.3645	310.2090	4.3039	309.7445	4.1979
bior1.1	310.3806	4.2154	310.3216	4.2734	310.2930	4.1597	310.1231	4.2033
bior1.3	310.6115	4.4219	310.4778	4.3590	310.1769	4.2519	309.8689	4.1464
bior1.5	310.5488	4.3760	310.3096	4.2864	309.8271	4.1444	309.5190	4.1013
bior2.2	310.5734	4.3745	310.5071	4.3833	310.3737	4.3703	310.1504	4.3570
bior2.4	310.5355	4.3579	310.4375	4.3435	310.2679	4.3200	309.8767	4.2461
bior2.6	310.5101	4.3320	310.4284	4.3828	310.2649	4.3663	309.8767	4.2893
bior2.8	310.5038	4.3397	310.3821	4.3380	310.2193	4.3378	309.8178	4.2330
bior3.1	310.5248	4.3697	310.2688	4.3297	309.9286	4.3485	309.1803	4.4102
bior3.3	310.5766	4.4123	310.4080	4.3267	310.1651	4.2740	309.7611	4.2895
bior3.5	310.5703	4.4049	310.4485	4.3521	310.2559	4.3416	309.9996	4.4344
bior3.7	310.5552	4.3884	310.4093	4.3117	310.1949	4.2861	309.8955	4.3404
bior3.9	310.5483	4.3912	310.4540	4.3633	310.2641	4.3320	309.8877	4.2994
bior4.4	310.5423	4.3613	310.4216	4.3108	310.2549	4.3215	309.8422	4.2077
bior5.5	310.5409	4.3884	310.4051	4.3405	310.2689	4.3773	309.7991	4.2197
bior6.8	310.5121	4.3518	310.3872	4.3210	310.2318	4.3286	309.7783	4.2906
rbio1.1	310.3806	4.2154	310.3216	4.2734	310.2930	4.1597	310.1231	4.2033
rbio1.3	310.6419	4.4661	310.5419	4.3933	310.4408	4.3924	310.1559	4.3161
rbio1.5	310.5889	4.4223	310.4786	4.3944	310.2468	4.2745	310.0991	4.4160
rbio2.2	310.5067	4.3535	310.4731	4.4227	310.2652	4.4180	309.9006	4.3748
rbio2.4	310.5449	4.3553	310.4182	4.3663	310.2223	4.3858	309.8101	4.3873
rbio2.6	310.5386	4.3568	310.4325	4.3786	310.2313	4.3390	309.8248	4.3117
rbio2.8	310.5258	4.3603	310.4148	4.3389	310.2114	4.3087	309.8172	4.2812
rbio3.1	310.1313	4.2563	308.6466	4.3858	303.2783	10.9602	223.5093	168.1689
rbio3.3	310.5426	4.4236	310.1821	4.3721	309.2593	4.3769	308.3751	4.4574
rbio3.5	310.5457	4.4136	310.3497	4.3528	309.9434	4.3377	309.6900	4.3161
rbio3.7	310.5573	4.4077	310.3338	4.3041	310.1022	4.3290	309.7902	4.4739
rbio3.9	310.5342	4.3783	310.4233	4.3565	310.2114	4.3792	309.7873	4.3217
rbio4.4	310.5140	4.3308	310.3763	4.3522	310.1727	4.3554	309.7857	4.3095
rbio5.5	310.5119	4.3759	310.3793	4.3546	310.1729	4.3765	309.7358	4.2784
rbio6.8	310.5080	4.3496	310.3698	4.3200	310.1707	4.3049	309.7406	4.3198
dmey	310.4705	4.3425	310.3268	4.3379	310.1496	4.2947	309.6759	4.2153

Table A.45: Results of GCV with hard thresholding for DSM generation on City Site 3
excluding outliers

City Site 3, $\mu = 318.6986$, $\sigma = 9.0395$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.6986	9.0419	318.6986	8.9188	318.6986	8.6899	318.6986	8.3209
db2	318.6987	9.0022	318.7002	8.9289	318.7043	8.7362	318.7206	8.3546
db4	318.6988	9.0105	318.6966	8.9384	318.6932	8.7649	318.6918	8.3900
db6	318.6979	9.0106	318.6990	8.9498	318.6966	8.7813	318.7039	8.4315
db8	318.6987	9.0114	318.7006	8.9517	318.6944	8.7803	318.7008	8.6110
db10	318.6992	9.0132	318.6963	8.9468	318.6939	8.7797	318.6914	8.5775
db12	318.6983	9.0136	318.6996	8.9551	318.6946	8.7855	318.7010	8.4265
db20	318.6982	9.0137	318.6978	8.9513	318.7020	8.7972	318.7037	8.4213
sym2	318.6987	9.0022	318.7002	8.9289	318.7043	8.7362	318.7206	8.3546
sym4	318.6981	9.0088	318.6937	8.9346	318.6969	8.7662	318.7107	8.4072
sym6	318.6980	9.0103	318.7038	8.9559	318.6924	8.7751	318.6979	8.4291
sym8	318.6980	9.0110	318.6958	8.9445	318.6890	8.7735	318.6935	8.4136
sym10	318.6980	9.0115	318.7013	8.9565	318.7067	8.7929	318.6916	8.4001
sym12	318.6991	9.0128	318.6964	8.9474	318.6987	8.7850	318.6908	8.3751
sym20	318.6991	9.0135	318.6968	8.9500	318.7005	8.7905	318.7069	8.5243
coif1	318.6984	9.0027	318.6955	8.9187	318.7021	8.7328	318.7223	8.3708
coif2	318.6980	9.0092	318.6952	8.9371	318.6870	8.7615	318.6956	8.4052
coif3	318.6979	9.0106	318.6960	8.9434	318.7037	8.7782	318.6907	8.3740
coif4	318.6979	9.0114	318.6963	8.9459	318.6925	8.7770	318.6940	8.3923
coif5	318.6979	9.0118	318.6965	8.9474	318.7036	8.7874	318.7095	8.4270
bior1.1	318.6986	9.0419	318.6986	8.9188	318.6986	8.6899	318.6986	8.3209
bior1.3	318.6986	9.0299	318.6971	9.0053	318.6961	8.8968	318.6934	8.5836
bior1.5	318.6986	9.0419	318.6993	9.0436	318.6966	8.9747	318.6983	8.6836
bior2.2	318.6982	8.9985	318.6968	8.9017	318.7045	8.6969	318.7282	8.3117
bior2.4	318.6982	9.0077	318.7035	8.9427	318.6963	8.7504	318.7018	8.3938
bior2.6	318.6982	9.0116	318.6956	8.9412	318.6866	8.7620	318.6963	8.4121
bior2.8	318.6982	9.0138	318.7023	8.9602	318.7042	8.7912	318.6917	8.3981
bior3.1	318.6986	9.0039	318.6999	8.9216	318.7034	8.8324	318.7110	8.8365
bior3.3	318.6986	9.0068	318.6994	8.9463	318.6887	8.7707	318.6867	8.4374
bior3.5	318.6986	9.0093	318.6962	8.9362	318.6958	8.7685	318.6932	8.4249
bior3.7	318.6986	9.0110	318.7022	8.9511	318.7071	8.7782	318.6953	8.4191
bior3.9	318.6986	9.0122	318.6954	8.9410	318.6929	8.7677	318.6948	8.4320
bior4.4	318.6979	9.0079	318.7035	8.9468	318.6960	8.7594	318.7024	8.4067
bior5.5	318.6993	9.0103	318.7016	8.9498	318.6903	8.7663	318.6937	8.4089
bior6.8	318.6979	9.0107	318.7012	8.9534	318.7050	8.7836	318.6925	8.3926
rbio1.1	318.6986	9.0419	318.6986	8.9188	318.6986	8.6899	318.6986	8.3209
rbio1.3	318.6986	8.9963	318.6953	8.9070	318.6909	8.6975	318.6881	8.3258
rbio1.5	318.6986	9.0043	318.6989	8.9276	318.6930	8.7400	318.6870	8.3557
rbio2.2	318.6987	9.0118	318.6949	8.9765	318.7033	8.9223	318.7230	8.7499
rbio2.4	318.6981	9.0072	318.7035	8.9517	318.6946	8.7792	318.7037	8.4415
rbio2.6	318.6979	9.0084	318.6956	8.9380	318.6874	8.7624	318.6959	8.4228
rbio2.8	318.6979	9.0096	318.7015	8.9508	318.7043	8.7769	318.6944	8.4084
rbio3.1	318.6986	9.2188	318.6954	10.7404	318.6817	17.3525	318.6444	37.2154
rbio3.3	318.6986	9.0241	318.6994	9.0689	318.6965	9.3068	318.6969	9.7775
rbio3.5	318.6986	9.0143	318.6942	8.9768	318.6966	8.9267	318.6952	9.0426
rbio3.7	318.6986	9.0130	318.7020	8.9672	318.7067	8.8506	318.6949	8.7111
rbio3.9	318.6986	9.0128	318.6953	8.9513	318.6930	8.8427	318.6951	8.6029
rbio4.4	318.6980	9.0116	318.7035	8.9622	318.6940	8.7946	318.7024	8.4485
rbio5.5	318.6991	9.0162	318.7025	8.9654	318.6912	8.7906	318.6951	8.4392
rbio6.8	318.6979	9.0114	318.7012	8.9560	318.7045	8.7901	318.6923	8.4053
dmey	318.6982	9.0131	318.6976	8.9516	318.7014	8.7922	318.7037	8.4352

Table A.46: Results of GCV with hard thresholding for DEM generation on City Site 3
excluding outliers

City Site 3, $\mu = 310.7514$, $\sigma = 4.4699$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.4289	4.1786	310.4549	4.2912	310.4135	4.1922	310.2033	4.2079
db2	310.6425	4.4233	310.5988	4.4266	310.4200	4.3536	310.3141	4.5001
db4	310.6190	4.4129	310.5306	4.3960	310.2545	4.2639	309.8756	4.3389
db6	310.5981	4.3767	310.5010	4.3372	310.2588	4.3340	309.8821	4.3013
db8	310.6016	4.4048	310.4909	4.3795	310.2115	4.3337	310.0185	4.3965
db10	310.5811	4.3912	310.4545	4.3699	310.2313	4.3517	309.7863	4.3959
db12	310.5885	4.3749	310.4435	4.3388	310.1648	4.3355	309.6595	4.3295
db20	310.5778	4.3800	310.4203	4.3679	310.1719	4.3737	309.7653	4.3038
sym2	310.6425	4.4233	310.5988	4.4266	310.4200	4.3536	310.3141	4.5001
sym4	310.6237	4.3959	310.5016	4.3617	310.3653	4.3685	309.9619	4.2671
sym6	310.5941	4.3666	310.4823	4.3322	310.2696	4.3598	309.8513	4.2857
sym8	310.5855	4.3533	310.4816	4.3575	310.3110	4.3750	309.8631	4.3717
sym10	310.6041	4.3775	310.4621	4.3297	310.2596	4.2776	309.8386	4.2949
sym12	310.5808	4.3990	310.4522	4.3137	310.2641	4.3070	309.8163	4.3689
sym20	310.5881	4.3984	310.4219	4.3113	310.2372	4.3011	309.7696	4.2745
coif1	310.6227	4.3819	310.5636	4.4002	310.4421	4.3800	310.1723	4.3817
coif2	310.6017	4.3590	310.5130	4.3889	310.3477	4.3856	309.9386	4.3262
coif3	310.5953	4.3623	310.4844	4.3757	310.3229	4.3411	309.9157	4.3831
coif4	310.5835	4.3597	310.4712	4.3603	310.2828	4.3593	310.0947	4.4247
coif5	310.5837	4.3624	310.4509	4.3502	310.3144	4.3248	309.8498	4.2124
bior1.1	310.4289	4.1786	310.4549	4.2912	310.4135	4.1922	310.2033	4.2079
bior1.3	310.6862	4.4134	310.5424	4.3490	310.2678	4.2484	309.9953	4.1921
bior1.5	310.6452	4.3976	310.4074	4.3108	310.0119	4.1918	309.6184	4.1379
bior2.2	310.6646	4.4182	310.5813	4.3949	310.4635	4.3912	310.1961	4.3092
bior2.4	310.6207	4.3892	310.5074	4.3249	310.3580	4.3497	309.9834	4.2673
bior2.6	310.5945	4.3743	310.5143	4.3992	310.3604	4.3881	309.9179	4.2527
bior2.8	310.5889	4.3754	310.4837	4.3448	310.3198	4.3597	309.8951	4.2507
bior3.1	310.6209	4.4284	310.4185	4.3057	310.1583	4.3184	309.3659	4.1768
bior3.3	310.6406	4.4217	310.5078	4.3543	310.2939	4.3291	309.9739	4.3879
bior3.5	310.6266	4.4182	310.5440	4.3740	310.3416	4.3527	310.0509	4.4579
bior3.7	310.6183	4.4120	310.5100	4.3525	310.3019	4.3116	309.9672	4.3274
bior3.9	310.6258	4.4219	310.5367	4.3772	310.3238	4.3156	309.9730	4.3290
bior4.4	310.6276	4.3945	310.5140	4.3275	310.3419	4.3261	309.9638	4.2617
bior5.5	310.6188	4.4035	310.5294	4.4010	310.3681	4.4154	309.9047	4.2174
bior6.8	310.5975	4.3762	310.4701	4.3234	310.3039	4.3346	309.8873	4.3199
rbio1.1	310.4289	4.1786	310.4549	4.2912	310.4135	4.1922	310.2033	4.2079
rbio1.3	310.6657	4.4296	310.6177	4.4080	310.5227	4.4253	310.3072	4.3779
rbio1.5	310.6337	4.4175	310.5468	4.3971	310.3493	4.2974	310.2229	4.4599
rbio2.2	310.5713	4.3528	310.5321	4.4487	310.3310	4.4328	310.0195	4.4340
rbio2.4	310.6326	4.3958	310.4987	4.3543	310.3077	4.3767	309.9221	4.4071
rbio2.6	310.6126	4.3790	310.5100	4.3836	310.3220	4.3858	309.9209	4.3561
rbio2.8	310.6089	4.3876	310.5054	4.3474	310.3088	4.3492	309.9451	4.3175
rbio3.1	310.2557	4.3153	308.7915	4.3504	302.8888	11.2175	251.6966	144.0014
rbio3.3	310.6041	4.4215	310.3202	4.3756	309.4661	4.3666	308.5255	4.5333
rbio3.5	310.6230	4.4389	310.4225	4.3622	310.0545	4.3960	309.7486	4.3491
rbio3.7	310.6214	4.4390	310.4581	4.3473	310.1667	4.3048	309.8164	4.4293
rbio3.9	310.6110	4.4166	310.4928	4.3947	310.3095	4.3773	309.8764	4.3483
rbio4.4	310.5953	4.3725	310.4789	4.3472	310.2633	4.3605	309.8841	4.3595
rbio5.5	310.6081	4.4175	310.4954	4.3944	310.2367	4.3733	309.8144	4.3008
rbio6.8	310.5974	4.3727	310.4506	4.3163	310.2473	4.3146	309.8620	4.3300
dmey	310.5482	4.3598	310.4002	4.3448	310.2867	4.3466	309.8005	4.2545

Table A.47: Results of GCV with soft thresholding for DSM generation on City Site 3 including outliers

City Site 3, $\mu = 318.7184$, $\sigma = 9.1404$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.7184	9.1402	318.7184	9.0088	318.7184	8.7703	318.7184	8.3790
db2	318.7185	9.0947	318.7204	9.0122	318.7250	8.8098	318.7418	8.4169
db4	318.7186	9.1020	318.7160	9.0229	318.7126	8.8414	318.7113	8.4529
db6	318.7178	9.1028	318.7189	9.0347	318.7163	8.8564	318.7236	8.4960
db8	318.7185	9.1040	318.7208	9.0357	318.7141	8.8568	318.7202	8.6616
db10	318.7189	9.1052	318.7158	9.0314	318.7136	8.8568	318.7109	8.6577
db12	318.7182	9.1052	318.7197	9.0390	318.7145	8.8607	318.7210	8.4996
db20	318.7181	9.1058	318.7175	9.0352	318.7220	8.8679	318.7237	8.4954
sym2	318.7185	9.0947	318.7204	9.0122	318.7250	8.8098	318.7418	8.4169
sym4	318.7180	9.1008	318.7129	9.0195	318.7162	8.8431	318.7311	8.4673
sym6	318.7179	9.1026	318.7243	9.0394	318.7121	8.8510	318.7178	8.4827
sym8	318.7179	9.1035	318.7152	9.0288	318.7082	8.8500	318.7136	8.4816
sym10	318.7179	9.1039	318.7215	9.0405	318.7267	8.8690	318.7115	8.4694
sym12	318.7188	9.1052	318.7159	9.0326	318.7188	8.8629	318.7108	8.4537
sym20	318.7188	9.1056	318.7164	9.0346	318.7205	8.8660	318.7267	8.5908
coif1	318.7182	9.0939	318.7146	9.0015	318.7221	8.8089	318.7435	8.4305
coif2	318.7179	9.1011	318.7146	9.0212	318.7064	8.8387	318.7156	8.4674
coif3	318.7179	9.1029	318.7154	9.0278	318.7239	8.8554	318.7105	8.4458
coif4	318.7179	9.1038	318.7158	9.0298	318.7118	8.8531	318.7139	8.4651
coif5	318.7179	9.1042	318.7160	9.0314	318.7235	8.8642	318.7296	8.4976
bior1.1	318.7184	9.1402	318.7184	9.0088	318.7184	8.7703	318.7184	8.3790
bior1.3	318.7184	9.1232	318.7170	9.0908	318.7164	8.9744	318.7143	8.6458
bior1.5	318.7184	9.1359	318.7193	9.1284	318.7161	9.0453	318.7176	8.7447
bior2.2	318.7179	9.0893	318.7156	8.9851	318.7237	8.7751	318.7491	8.3858
bior2.4	318.7181	9.0993	318.7240	9.0265	318.7164	8.8237	318.7220	8.4454
bior2.6	318.7182	9.1038	318.7150	9.0250	318.7058	8.8399	318.7164	8.4683
bior2.8	318.7183	9.1064	318.7225	9.0444	318.7249	8.8662	318.7126	8.4721
bior3.1	318.7184	9.0960	318.7202	9.0095	318.7201	8.9004	318.7126	8.8171
bior3.3	318.7184	9.0980	318.7205	9.0276	318.7068	8.8433	318.7072	8.4869
bior3.5	318.7184	9.1006	318.7155	9.0184	318.7154	8.8428	318.7126	8.4543
bior3.7	318.7184	9.1025	318.7226	9.0336	318.7276	8.8466	318.7153	8.4662
bior3.9	318.7184	9.1039	318.7147	9.0251	318.7125	8.8451	318.7148	8.4811
bior4.4	318.7178	9.0997	318.7239	9.0309	318.7161	8.8368	318.7226	8.4629
bior5.5	318.7191	9.1028	318.7220	9.0316	318.7100	8.8395	318.7133	8.4731
bior6.8	318.7178	9.1030	318.7213	9.0376	318.7248	8.8593	318.7125	8.4639
rbio1.1	318.7184	9.1402	318.7184	9.0088	318.7184	8.7703	318.7184	8.3790
rbio1.3	318.7184	9.0888	318.7146	8.9922	318.7099	8.7767	318.7061	8.3645
rbio1.5	318.7184	9.0972	318.7188	9.0114	318.7126	8.8175	318.7065	8.4144
rbio2.2	318.7187	9.1036	318.7142	9.0596	318.7235	8.9852	318.7442	8.7514
rbio2.4	318.7179	9.0985	318.7238	9.0343	318.7144	8.8521	318.7239	8.4968
rbio2.6	318.7177	9.0999	318.7148	9.0204	318.7067	8.8362	318.7160	8.4684
rbio2.8	318.7177	9.1012	318.7216	9.0343	318.7242	8.8512	318.7144	8.4519
rbio3.1	318.7184	9.3133	318.7152	10.6897	318.7013	16.0640	318.6637	31.2036
rbio3.3	318.7184	9.1155	318.7195	9.1467	318.7164	9.2922	318.7172	9.4185
rbio3.5	318.7184	9.1056	318.7136	9.0498	318.7164	8.9396	318.7148	8.6327
rbio3.7	318.7184	9.1040	318.7222	9.0465	318.7271	8.8864	318.7146	8.5266
rbio3.9	318.7184	9.1039	318.7147	9.0304	318.7126	8.8627	318.7150	8.5021
rbio4.4	318.7180	9.1038	318.7239	9.0464	318.7139	8.8686	318.7224	8.5110
rbio5.5	318.7187	9.1088	318.7226	9.0486	318.7110	8.8663	318.7151	8.5100
rbio6.8	318.7179	9.1037	318.7213	9.0400	318.7244	8.8649	318.7124	8.4733
dmey	318.7181	9.1057	318.7172	9.0361	318.7217	8.8693	318.7237	8.5149

Table A.48: Results of GCV with soft thresholding for DEM generation on City Site 3 including outliers

City Site 3, $\mu = 310.7064$, $\sigma = 4.4727$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.3806	4.2154	310.3280	4.2709	310.2983	4.1536	310.1407	4.1619
db2	310.5782	4.4139	310.5001	4.3696	310.3254	4.3054	310.2168	4.4534
db4	310.5699	4.4172	310.4475	4.3867	310.1921	4.2863	309.9021	4.3530
db6	310.5149	4.3520	310.3946	4.3116	310.1350	4.2890	309.7873	4.2247
db8	310.5317	4.3685	310.4042	4.3592	310.1312	4.3288	309.8979	4.3605
db10	310.5052	4.3890	310.3754	4.3544	310.0974	4.3146	309.6929	4.3793
db12	310.5184	4.3677	310.3541	4.3112	310.0783	4.3399	309.6009	4.2524
db20	310.4858	4.3522	310.3571	4.3913	310.1844	4.3535	309.7647	4.3131
sym2	310.5782	4.4139	310.5001	4.3696	310.3254	4.3054	310.2168	4.4534
sym4	310.5502	4.3774	310.4479	4.3463	310.2953	4.3394	309.8529	4.1890
sym6	310.5277	4.3564	310.4054	4.3298	310.1712	4.3134	309.7218	4.1948
sym8	310.5082	4.3457	310.4033	4.3474	310.2112	4.3568	309.7785	4.3014
sym10	310.4908	4.3396	310.3618	4.3337	310.1375	4.2523	309.7262	4.2619
sym12	310.4906	4.3662	310.3643	4.3014	310.1695	4.2843	309.7209	4.2889
sym20	310.4759	4.3536	310.3376	4.3033	310.1367	4.2909	309.7035	4.2459
coif1	310.5359	4.3525	310.4981	4.3944	310.3534	4.3668	310.1001	4.3924
coif2	310.5260	4.3487	310.4316	4.3648	310.2351	4.3304	309.8332	4.2473
coif3	310.5113	4.3427	310.4159	4.3770	310.2227	4.3143	309.8766	4.3426
coif4	310.4976	4.3436	310.4152	4.3859	310.1914	4.3513	310.0268	4.4125
coif5	310.4832	4.3277	310.4072	4.3626	310.2113	4.3037	309.7652	4.1934
bior1.1	310.3806	4.2154	310.3280	4.2709	310.2983	4.1536	310.1407	4.1619
bior1.3	310.6124	4.4231	310.4830	4.3592	310.1727	4.2409	309.9048	4.1293
bior1.5	310.5498	4.3773	310.3101	4.2822	309.8235	4.1033	309.5484	4.1071
bior2.2	310.5734	4.3710	310.5090	4.3796	310.3847	4.3662	310.1589	4.3528
bior2.4	310.5366	4.3540	310.4372	4.3424	310.2715	4.3101	309.9029	4.2183
bior2.6	310.5111	4.3277	310.4320	4.3772	310.2622	4.3501	309.8836	4.2697
bior2.8	310.5056	4.3371	310.3827	4.3327	310.2155	4.3290	309.8122	4.2310
bior3.1	310.5237	4.3708	310.2747	4.3224	309.9307	4.3609	309.1725	4.3696
bior3.3	310.5751	4.4110	310.4145	4.3234	310.1890	4.2933	309.8175	4.3420
bior3.5	310.5692	4.4064	310.4458	4.3518	310.2534	4.3367	310.0130	4.4270
bior3.7	310.5542	4.3905	310.4126	4.3109	310.2202	4.2753	309.9304	4.2955
bior3.9	310.5475	4.3938	310.4567	4.3564	310.2664	4.3260	309.9038	4.2796
bior4.4	310.5433	4.3575	310.4215	4.3113	310.2685	4.3143	309.8617	4.1996
bior5.5	310.5401	4.3867	310.4047	4.3383	310.2715	4.3766	309.8275	4.2068
bior6.8	310.5129	4.3493	310.3873	4.3172	310.2284	4.3219	309.7833	4.2661
rbio1.1	310.3806	4.2154	310.3280	4.2709	310.2983	4.1536	310.1407	4.1619
rbio1.3	310.6413	4.4651	310.5415	4.3914	310.4459	4.3834	310.1829	4.2987
rbio1.5	310.5890	4.4206	310.4824	4.3877	310.2856	4.3012	310.1079	4.4108
rbio2.2	310.5043	4.3512	310.4726	4.4183	310.2671	4.4054	309.9229	4.3510
rbio2.4	310.5435	4.3520	310.4180	4.3601	310.2356	4.3645	309.8090	4.3384
rbio2.6	310.5399	4.3553	310.4347	4.3752	310.2356	4.3332	309.8304	4.2642
rbio2.8	310.5262	4.3574	310.4165	4.3325	310.2060	4.2953	309.8494	4.2457
rbio3.1	310.1399	4.2462	308.6776	4.3910	303.4611	10.6199	281.7410	51.1837
rbio3.3	310.5453	4.4234	310.2078	4.3509	309.3167	4.2812	308.5138	4.3969
rbio3.5	310.5471	4.4142	310.3537	4.3409	309.9613	4.3012	309.5969	4.2667
rbio3.7	310.5575	4.4087	310.3470	4.2923	310.0973	4.2933	309.8269	4.3865
rbio3.9	310.5352	4.3788	310.4243	4.3528	310.2085	4.3632	309.8025	4.2815
rbio4.4	310.5119	4.3295	310.3778	4.3474	310.1813	4.3481	309.7908	4.2938
rbio5.5	310.5120	4.3759	310.3761	4.3507	310.1727	4.3642	309.7423	4.2568
rbio6.8	310.5083	4.3494	310.3701	4.3168	310.1698	4.2982	309.7657	4.2922
dmey	310.4714	4.3431	310.3288	4.3346	310.1617	4.2934	309.6745	4.2157

Table A.49: Results of GCV with soft thresholding for DSM generation on City Site 3 excluding outliers

City Site 3, $\mu = 318.6986$, $\sigma = 9.0395$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	318.6986	9.0419	318.6986	8.9183	318.6986	8.6827	318.6986	8.2848
db2	318.6987	9.0021	318.7002	8.9281	318.7043	8.7292	318.7206	8.3366
db4	318.6988	9.0103	318.6966	8.9378	318.6932	8.7602	318.6917	8.3672
db6	318.6979	9.0103	318.6990	8.9492	318.6966	8.7764	318.7039	8.4106
db8	318.6987	9.0111	318.7006	8.9513	318.6944	8.7772	318.7004	8.5743
db10	318.6992	9.0129	318.6963	8.9458	318.6939	8.7769	318.6918	8.5488
db12	318.6983	9.0130	318.6996	8.9541	318.6947	8.7811	318.7011	8.4153
db20	318.6982	9.0133	318.6978	8.9501	318.7020	8.7898	318.7037	8.4135
sym2	318.6987	9.0021	318.7002	8.9281	318.7043	8.7292	318.7206	8.3366
sym4	318.6981	9.0083	318.6937	8.9337	318.6965	8.7627	318.7103	8.3817
sym6	318.6980	9.0100	318.7038	8.9551	318.6924	8.7703	318.6979	8.3968
sym8	318.6980	9.0108	318.6958	8.9434	318.6890	8.7693	318.6935	8.3951
sym10	318.6980	9.0112	318.7013	8.9558	318.7067	8.7915	318.6916	8.3835
sym12	318.6991	9.0125	318.6964	8.9470	318.6989	8.7840	318.6910	8.3681
sym20	318.6991	9.0130	318.6968	8.9491	318.7005	8.7876	318.7067	8.5209
coif1	318.6984	9.0023	318.6955	8.9166	318.7024	8.7284	318.7225	8.3460
coif2	318.6980	9.0089	318.6952	8.9360	318.6870	8.7571	318.6956	8.3806
coif3	318.6979	9.0105	318.6960	8.9426	318.7037	8.7766	318.6907	8.3595
coif4	318.6979	9.0112	318.6963	8.9445	318.6925	8.7733	318.6940	8.3809
coif5	318.6979	9.0116	318.6965	8.9462	318.7036	8.7858	318.7092	8.4151
bior1.1	318.6986	9.0419	318.6986	8.9183	318.6986	8.6827	318.6986	8.2848
bior1.3	318.6986	9.0298	318.6971	9.0046	318.6961	8.8929	318.6934	8.5607
bior1.5	318.6986	9.0417	318.6993	9.0429	318.6966	8.9672	318.6981	8.6663
bior2.2	318.6982	8.9981	318.6968	8.9006	318.7045	8.6936	318.7282	8.3001
bior2.4	318.6982	9.0073	318.7035	8.9418	318.6964	8.7415	318.7019	8.3586
bior2.6	318.6982	9.0113	318.6956	8.9398	318.6866	8.7581	318.6962	8.3810
bior2.8	318.6982	9.0136	318.7023	8.9594	318.7050	8.7875	318.6925	8.3850
bior3.1	318.6986	9.0036	318.7005	8.9146	318.7013	8.8075	318.6986	8.7193
bior3.3	318.6986	9.0065	318.6994	8.9425	318.6887	8.7591	318.6867	8.4043
bior3.5	318.6986	9.0091	318.6962	8.9327	318.6958	8.7613	318.6932	8.3713
bior3.7	318.6986	9.0107	318.7022	8.9493	318.7070	8.7671	318.6952	8.3790
bior3.9	318.6986	9.0119	318.6954	8.9395	318.6929	8.7643	318.6948	8.3963
bior4.4	318.6979	9.0077	318.7035	8.9461	318.6961	8.7551	318.7025	8.3763
bior5.5	318.6993	9.0100	318.7016	8.9471	318.6903	8.7576	318.6937	8.3861
bior6.8	318.6979	9.0106	318.7012	8.9527	318.7050	8.7812	318.6925	8.3772
rbio1.1	318.6986	9.0419	318.6986	8.9183	318.6986	8.6827	318.6986	8.2848
rbio1.3	318.6986	8.9960	318.6953	8.9062	318.6909	8.6934	318.6873	8.2803
rbio1.5	318.6986	9.0039	318.6989	8.9268	318.6930	8.7348	318.6874	8.3310
rbio2.2	318.6987	9.0113	318.6949	8.9739	318.7035	8.9065	318.7232	8.6721
rbio2.4	318.6981	9.0066	318.7035	8.9497	318.6946	8.7707	318.7037	8.4113
rbio2.6	318.6979	9.0079	318.6956	8.9355	318.6874	8.7548	318.6959	8.3817
rbio2.8	318.6979	9.0091	318.7015	8.9495	318.7043	8.7725	318.6944	8.3649
rbio3.1	318.6986	9.2164	318.6954	10.6058	318.6817	16.0351	318.6444	31.2853
rbio3.3	318.6986	9.0226	318.6994	9.0596	318.6965	9.2160	318.6969	9.3464
rbio3.5	318.6986	9.0135	318.6942	8.9634	318.6966	8.8603	318.6952	8.5497
rbio3.7	318.6986	9.0121	318.7020	8.9621	318.7067	8.8086	318.6949	8.4422
rbio3.9	318.6986	9.0120	318.6953	8.9448	318.6930	8.7843	318.6951	8.4200
rbio4.4	318.6980	9.0112	318.7035	8.9613	318.6940	8.7876	318.7025	8.4251
rbio5.5	318.6991	9.0160	318.7025	8.9642	318.6913	8.7856	318.6952	8.4243
rbio6.8	318.6979	9.0111	318.7012	8.9550	318.7045	8.7868	318.6923	8.3868
dmey	318.6982	9.0130	318.6976	8.9509	318.7017	8.7910	318.7040	8.4318

Table A.50: Results of GCV with soft thresholding for DEM generation on City Site 3 excluding outliers

City Site 3, $\mu = 310.7514$, $\sigma = 4.4699$								
Wavelet	Level 1		Level 2		Level 3		Level 4	
	μ	σ	μ	σ	μ	σ	μ	σ
haar	310.4289	4.1786	310.4584	4.2841	310.4201	4.1816	310.2256	4.1611
db2	310.6425	4.4233	310.5958	4.4228	310.4120	4.3277	310.3143	4.4917
db4	310.6189	4.4085	310.5307	4.3950	310.2768	4.2830	309.9798	4.3308
db6	310.5981	4.3767	310.5018	4.3365	310.2604	4.3336	309.8830	4.2683
db8	310.6023	4.4048	310.4876	4.3796	310.2556	4.3562	309.9883	4.3439
db10	310.5821	4.3908	310.4582	4.3646	310.2371	4.3512	309.7921	4.3965
db12	310.5882	4.3725	310.4430	4.3361	310.1778	4.3361	309.7113	4.2843
db20	310.5797	4.3758	310.4207	4.3655	310.2605	4.3687	309.8030	4.2910
sym2	310.6425	4.4233	310.5958	4.4228	310.4120	4.3277	310.3143	4.4917
sym4	310.6223	4.3946	310.5072	4.3578	310.3691	4.3606	309.9547	4.2392
sym6	310.5929	4.3653	310.4853	4.3319	310.2758	4.3578	309.8633	4.2596
sym8	310.5858	4.3531	310.4861	4.3508	310.3078	4.3641	309.8853	4.3358
sym10	310.6045	4.3774	310.4608	4.3284	310.2618	4.2778	309.8709	4.3068
sym12	310.5808	4.3992	310.4520	4.3141	310.2677	4.3069	309.8112	4.3670
sym20	310.5880	4.3983	310.4221	4.3113	310.2440	4.3014	309.7966	4.2627
coif1	310.6217	4.3799	310.5612	4.3978	310.4461	4.3796	310.1747	4.3702
coif2	310.6010	4.3553	310.5119	4.3838	310.3480	4.3719	309.9595	4.3030
coif3	310.5945	4.3587	310.4838	4.3751	310.3242	4.3410	309.9214	4.3827
coif4	310.5828	4.3596	310.4705	4.3600	310.2765	4.3477	310.0924	4.4184
coif5	310.5837	4.3624	310.4512	4.3494	310.3150	4.3243	309.8618	4.2112
bior1.1	310.4289	4.1786	310.4584	4.2841	310.4201	4.1816	310.2256	4.1611
bior1.3	310.6851	4.4134	310.5456	4.3418	310.2664	4.2407	310.0153	4.1676
bior1.5	310.6441	4.3976	310.4049	4.3006	310.0104	4.1801	309.6356	4.1365
bior2.2	310.6630	4.4162	310.5828	4.3926	310.4682	4.3834	310.2115	4.3100
bior2.4	310.6192	4.3872	310.5075	4.3261	310.3617	4.3461	310.0177	4.2338
bior2.6	310.5972	4.3740	310.5119	4.3959	310.3620	4.3788	309.9474	4.2614
bior2.8	310.5916	4.3752	310.4828	4.3451	310.3200	4.3570	309.9236	4.2631
bior3.1	310.6218	4.4231	310.4256	4.3031	310.1334	4.3195	309.3179	4.2218
bior3.3	310.6424	4.4184	310.5197	4.3495	310.3131	4.3277	310.0312	4.4144
bior3.5	310.6287	4.4146	310.5436	4.3716	310.3324	4.3337	310.0796	4.4265
bior3.7	310.6203	4.4085	310.5082	4.3400	310.3042	4.3009	310.0083	4.3188
bior3.9	310.6282	4.4178	310.5374	4.3716	310.3258	4.3080	309.9869	4.3038
bior4.4	310.6263	4.3926	310.5155	4.3271	310.3376	4.3269	309.9838	4.2475
bior5.5	310.6189	4.4036	310.5313	4.4001	310.3721	4.4135	309.9277	4.2060
bior6.8	310.5995	4.3748	310.4699	4.3218	310.2977	4.3229	309.9352	4.3262
rbio1.1	310.4289	4.1786	310.4584	4.2841	310.4201	4.1816	310.2256	4.1611
rbio1.3	310.6657	4.4296	310.6201	4.4032	310.5253	4.4227	310.3200	4.3552
rbio1.5	310.6335	4.4174	310.5476	4.3892	310.3378	4.2878	310.2260	4.4579
rbio2.2	310.5740	4.3510	310.5325	4.4448	310.3376	4.4165	310.0302	4.4082
rbio2.4	310.6356	4.3955	310.4999	4.3514	310.3081	4.3606	309.9237	4.3591
rbio2.6	310.6150	4.3768	310.5071	4.3768	310.3373	4.3584	309.9346	4.3040
rbio2.8	310.6111	4.3855	310.5069	4.3472	310.3169	4.3414	309.9637	4.2852
rbio3.1	310.2669	4.3069	308.8096	4.3463	303.2456	10.7965	286.0772	48.3183
rbio3.3	310.6075	4.4172	310.3192	4.3559	309.4708	4.2861	308.6524	4.4075
rbio3.5	310.6268	4.4312	310.4287	4.3486	310.0840	4.3286	309.7243	4.3149
rbio3.7	310.6246	4.4315	310.4687	4.3293	310.1625	4.2801	309.8429	4.3761
rbio3.9	310.6162	4.4072	310.4956	4.3835	310.3023	4.3541	309.8933	4.2969
rbio4.4	310.5987	4.3724	310.4792	4.3446	310.2685	4.3470	309.8884	4.3339
rbio5.5	310.6091	4.4152	310.4974	4.3943	310.2363	4.3700	309.8480	4.2864
rbio6.8	310.5999	4.3711	310.4514	4.3146	310.2471	4.3030	309.8959	4.2968
dmey	310.5522	4.3579	310.4029	4.3424	310.2941	4.3460	309.8009	4.2492

APPENDIX B

KEY FUNCTIONS USED AND IMPLEMENTED IN MATLAB

Key built-in functions used:

- dwt
- meshgrid
- griddata
- wavedec2
- waverec2
- wthresh
- mean
- std
- surf

Key functions implemented:

- outlier
- denoise2
- pmfilter

APPENDIX C

IMPLEMENTED WAVELET MODELS

The following wavelet models are implemented in this work:

- Haar
- Daubechies (2, 4, 6, 8, 10, 12, 20)
- Symlets (2, 4, 6, 8, 10, 12, 20)
- Coiflets (1-5)
- Biorthogonal (1.1, 1.3, 1.5, 2.2, 2.4, 2.6, 2.8, 3.1, 3.3, 3.5, 3.7, 3.9, 4.4, 5.5, 6.8)
- Reverse Biorthogonal (1.1, 1.3, 1.5, 2.2, 2.4, 2.6, 2.8, 3.1, 3.3, 3.5, 3.7, 3.9, 4.4, 5.5, 6.8)
- Discrete Meyer