



# Algorithms for Global Image Denoising

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**Abstract**— visual information transmitted in the form of digital images is becoming a major method of communication in the modern age. But the main drawback in digital images is inheritance of noise while their acquisition or transmission. Removing noise from digital images is a big challenge for researchers. Several noise removal algorithms have been proposed till date. Choice of denoising algorithm is application dependent and depends upon the type of noise present in the image. Every algorithm has its own assumptions, advantages and limitations. This paper presents a comparative analysis of various noise suppression algorithms.

**Index Terms**— image denoising, noise removal algorithms, image processing, etc.

## I. INTRODUCTION

The search for efficient image denoising methods is still a valid challenge at the crossing of functional analysis and statistics. Most of algorithms have not yet attained a desirable level of applicability. Project aims of removing noise from digital images. denoising algorithm is application dependent and depends upon type of noise present in the image. Enhance the visual appearance of images and improve the manipulation of datasets. Enhancement techniques can emphasize image artefacts or even lead to loss of information if not correctly used. The main application is quantitative diagnostic image. surgery planning analysis diagnostic analysis mainly CT scan of brain. surgical planning mainly diagnostic aortic aneurism.

## II. PROBLEM STATEMENT

The distortions of images by noise are common during its acquisition, processing, compression, transmission, and reproduction. Images may contain various types of noises like Salt and Pepper noise, Speckle noise and Poisson noise. Usually the real and imaginary parts of image are considered corrupted by additive white Gaussian noise (AWGN). So denoising of images corrupted by additive white Gaussian noise is a classical problem in image processing and it has become a promising and very challenging research area in recent years. Shrinkage methods are often used for suppressing additive white Gaussian noise, where thresholding is used to retain the larger wavelet coefficients alone. Shrinkage Algorithms fail to retain the edges, corners and flat regions of

the image being processed. Therefore, it is necessary to suppress noise while producing sharp images without loss of finer details.

## III. PROPOSED SYSTEM

In the proposed system architecture uses an image to denoising. Before denoising, the pre-processing is carried out for an image to reduce image size and color space conversion. Nyström algorithm is used for applying a pre-filter on the noisy image, a small fraction of the pixels are sampled. After that Eigen decomposition is carried out for sampled image the Eigen values and eigenvectors is obtained. The final estimate of the image is constructed by means of shrinkage of the filter Eigen values.

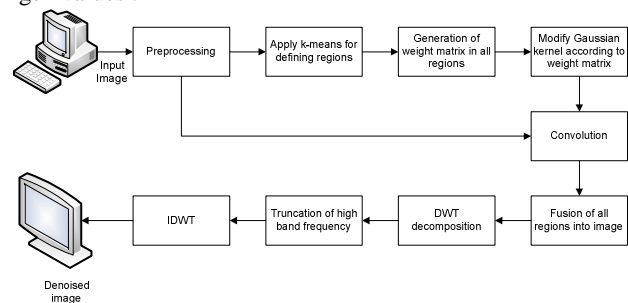


Figure 1: Proposed System Block Diagram.

## IV. METHODOLOGY

Figure 1 shows the block diagram of the overall flow of the proposed system for global image denoising. The total system is divided into five modules i.e. preprocessing module, clustering module, modify Gaussian kernel module, fusion module and Filtering module respectively as shown in figure. From the block diagram it is evident that Denoising Module is the core for this system. The query input image is preprocessed and apply the k-means clustering algorithm. Later the image processing techniques are used to extract the Gaussian kernel features of the obtained block. Finally apply the filtering module or denoising module for manipulation of the image data to produce a visually high quality image.

1. Pre-Processing
  - a. Conversion of RGB Format to Gray scale
2. K-Means Clustering
3. Modify Gaussian Kernel Module
4. Wavelet Based Filtering Module
5. Discrete wavelet transforms (DWT) Feature

## V. DATA FLOW DIAGRAM

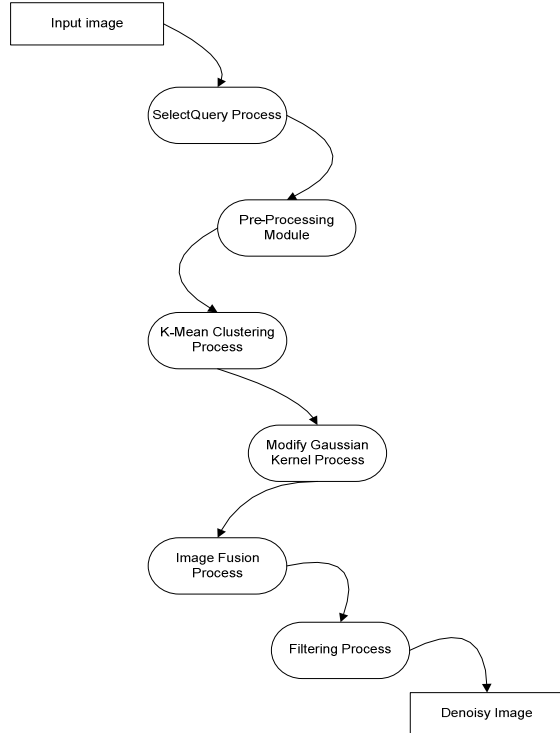


Figure 2: DFD of the proposed system

## VI. RESULT AND ANALYSIS

The detailed description for running a program and results obtained in each stage is explained briefly below. Initially when the run icon is clicked a menu dialog box is displayed as shown in Figure 16. This consists of different menus like Select Query, Clustering, Image Fusion, Filtering, and exit. When the first menu is selected Select query program is executed. Where input samples as shown and selected and feature is extracted from all the samples.

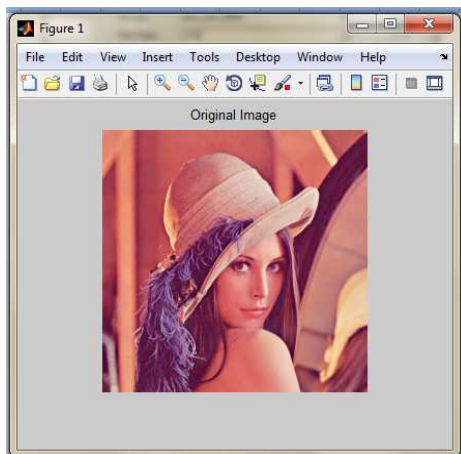


Figure 3: Input image

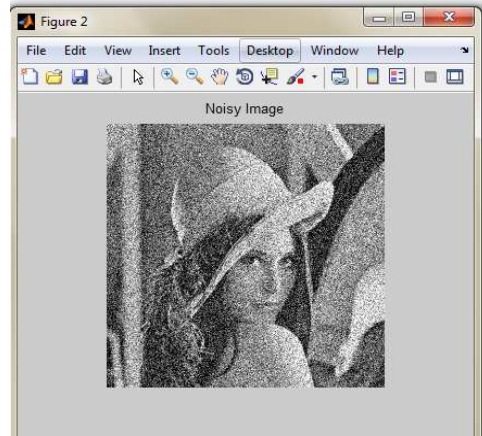


Figure 4: Processed image.

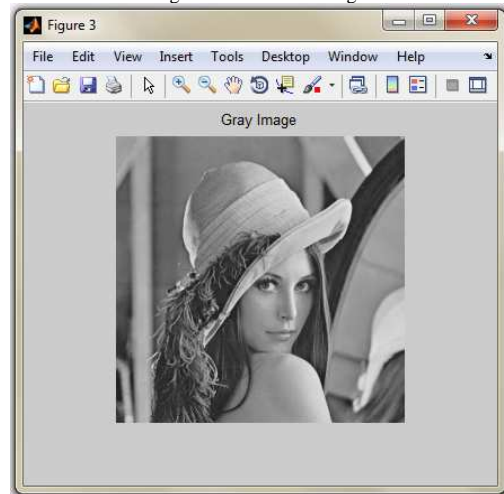


Figure 5: Next processed image

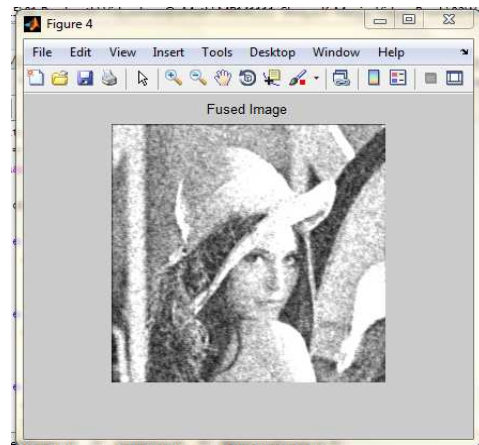


Figure 6: Next Processed image

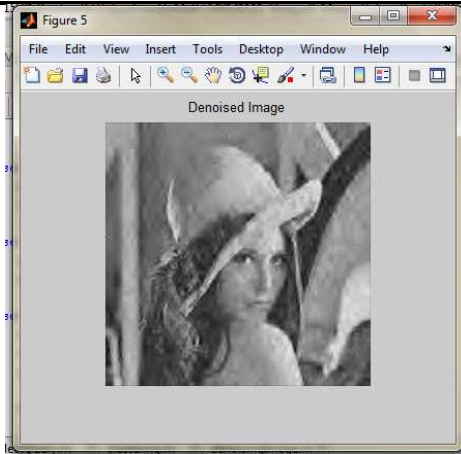


Figure 7: Next Processed Image

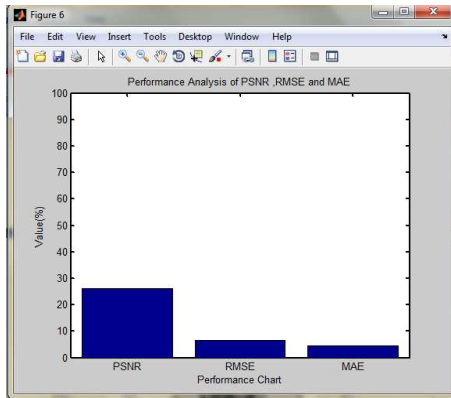


Figure 8: Performance analysis Graph for the proposed System.

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## VII.CONCLUSION

In this project work on proposed a Removing noise from digital images. Denoising algorithm is application dependent and depends upon the type of noise present in the image. The filters perform well on digital images but they have some constraints regarding resolution degradation. These filters operate by smoothing over a fixed window and it produces artifacts around the object and sometimes causes over smoothing thus causing blurring of image.

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