

# Improved the Performance of Image Denoising Technique Using Wavelet and SOM Neural Network Model

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## ABSTRACT

Image denoising is necessary and the initial step to be taken prior to the image data is analyzed. It is essential to apply an efficient denoising technique, to compensate such data corruption. The effort of image denoising is to improve an image that is cleaner than its noisy observation. Therefore, a substantial technology in image analysis is noise reduction and the initial step to be taken prior to images is analyzed. In this paper proposed a hybrid method for image denoising for improvement of gray scale image. The process of gray scale image gets the high component value of noise in environment. For the reduction of these noise used wavelet transform domain method. The wavelet transform method is well recognized method for noise reduction. In wavelet transform method the local noise component value are not considered. Then after the denoising process noise are remain in gray image. For these low components value collection used multiple sequences. And finally used self-organized map network.

**Keywords: - Image, Denoising, Wavelet, SOM.**

## INTRODUCTION

Image denoising is very important phase of image preprocessing technique. The preprocessing of image used for the minimization and removal of noise quantity in a given image for the processing of application. For the prediction and analysis of critical disease in medical science used computer vision technique[1,2]. The computer vision technique produces two types of image gray scale image and RGB image. The gray scale image just like a black & white image. The gray scale image consumes more value of distorted signal and damage the quality of image in terms of noise. Due to maximum value of noise the actual content of image are not visualized in clear manner. Now need for image denoising process for better prediction and analysis of medical image. Now days' various authors and scientists used different types of filter for denoising purpose of medical image. The transform based function is used for the purpose of image denoising. The wavelet transform function well knows technique used for image denoising process. The wavelet transform function denoise the image into different

terms of layer is called. Low pass filter and high pass filter. Some authors used multiple sequence of transform function for the minimization of noise value. In this dissertation modified the multiple sequence technique method using SOM neural network model [3,4]. Traditional wavelet transform de-noising has multi-scale properties and can deal with high-frequency noise effectively unlike time domain analysis. However, this method uses the same threshold value, and it does not meet the uneven distribution of medical images noise characteristics, which result in the loss of important information, excessive image smoothing and so on. Median filter is a kind of nonlinear filter, and it has good abilities of keeping edge [6,7]. The filter can overcome the shortcoming of image blur which probably resulted from minimum mean square filtering and mean filter. The rest of paper discuss as in section 2 discuss the related work. in section 3. Discuss the methodology. in section 4 discuss experimental result analysis and finally discuss conclusion & future work in section 5.

## 2. RELATED WORK

Arundhati Misra, B Kartikeyan and S Garg Et al. [1] According to analysts, Coherent frameworks, for example, Synthetic Aperture Radar(SAR), Ultra Sonography(USG), or MRI (Magnetic Resonance Imaging) are inalienably influenced by a grainy sort of commotion called the spot clamor. The nearness of such commotion debases the radiometry of the pictures and renders such picture troublesome for picture investigation or elucidation. The applications including arrangement, division or surface investigation of the abundancy information from SAR constantly require a spot lessened picture for appropriate information examination. It is likewise observed that the request of the mother wavelet picked is imperative in denoising of such territories. Albeit higher request wavelets have the ability to relate well with more honed elements, they likewise tend to hold high recurrence commotion. The ideal decision in this manner relies on upon the sort of the specific information. Here Daub4 and Daub20 gave great outcomes for both the sorts of scenes.

Yatong Xu, Xin Jin and Qionghai Dai Et al. [2] They examine a novel Kinect profundity de-noising calculation, in which the profundity guide is partitioned into two sections with surface edges' help. At that point channel the edge locale

in spatial-worldly space mutually in view of the investigation of profundity order along the surface edges. In this way, both intra outline and bury outline data are abused to stifle the obscure and vacillation of the profundity limits. Explore comes about have shown that the prepared profundity gets enhanced with more steady and precise edge arrangements.

Zayed M. Ramadan Et al. [3] they described, two indiscreet clamor models connected on a few pictures of various attributes, and an extensive variety of commotion densities were considered in this paper. Equivalent and unequal measures of pepper and salt densities are both considered and analyzed in this paper. Little sizes of square filtering windows utilized as a part of the talked about strategy diminish the obscuring issue while safeguarding picture points of interest. Broad reenactment comes about demonstrate the better execution of the examined technique over other existing cutting edge strategies as far as picture rebuilding quality and conservation of picture fine points of interest and sharp edges.

Tae Hyun Kim and Kyoung Mu Lee Et al. [4] they address the issue of the customary coarse-to-fine deblurring outline work, which offers ascend to curios while reestablishing little structures with particular movement. they in this manner examine a novel portion re-introduction technique which lessens the mistake of movement flow spread from a coarser level. They have displayed an efficient dynamic scene deblurring strategy that does not require precise movement division with the guide of vigorous TV-L1 based model.

Mostafa Karbasia, Sara Bilala, Reza Aghababaeyanb, Abdolvahab Ehsani Radc, Zeeshan Bhattia and Asadullah Shaha Et al. [5] In this paper, the Kinect gadget, its components, noising sorts were talked about. An extraordinary procedure of taking out clamor and foundation from a picture; got through the Kinect gadget, is examined. The iterative strategy displayed, is a novel technique for foundation subtraction. The iterative procedure was utilized for de-noising information, with the outcomes uncovering that the foundation of the scene and shadow have been expelled totally. Through this procedure, the profundity information acquired from Kinect gadgets can without much of a stretch be translated in various applications, with more exactness of recognizing the frontal subject. Assist, the nature of profundity guide has enhanced and was improved after de-noising process.

Linlin Xu, Graduate Student Jonathan Li, Yuanming Shu and Junhuan Peng Et al. [6] specialists characterized extend this line of study to the denoising of engineered opening radar (SAR) pictures in light of grouping the boisterous picture into disjoint nearby areas with comparable spatial structure and denoising every locale by the straight least mean-square mistake (LMMSE) filtering in foremost segment examination (PCA) space. This grouping methodology can be dealt with as the unsupervised partner of the usually received piece coordinating methodology. It requires less calculation. Besides, it is prepared to do adaptively distinguishing "comparable" fixes by considering the closeness to various bunch focuses. In the denoising stage, keeping in mind the end goal to dodge the constraints of the homomorphic

approach, they have constructed their denoising plan on ASDN and determined a PCA-based LMMSE denoising model for multiplicative commotion. their approach is the first to assemble the PCA-construct denoising technique in light of the ASDN display for SAR picture denoising. Other than SAR pictures, it is likewise appropriate to another flag subordinate clamor. The denoised patches of all bunches were finally used to reproduce the clamor free picture.

Ladan Ebadi, Helmi Z. M. Shafri, Shattri B. Mansor and Ravshan Ashurov Et al. [7] as indicated by creators, to accept that the quantity of perceptions is to the force of two, the flag under scrutiny is developed bi-infinitely and its limit is shut. To apply first-era wavelets, the flag under scrutiny ought to be amplified. Flag expansion techniques result in antiquities and limit impacts. Number wavelet trans-frames performed by lifting technique outline to whole numbers, so they are invertible through finite exactness math. A lifting plan takes into account the set-up execution of wavelet changes and decreases the calculation time and memory necessities.

Charles-Alban Deledalle, Loic Denis, Florence Tupin, Andreas Reigber and Marc Jager Et al. [8] they depict a general technique, NL-SAR, that constructs developed non-nearby neighborhoods for denoising abundancy, polarimetric or potentially interferometric SAR pictures. They give the source code together with this paper depicting their philosophy. The code is sensibly quick and can be connected on vast pictures utilizing parallel models. A key element that ought to facilitate the wide use of the technique is the completely programmed tuning of all parameters. The strategies acquainted with weight likenesses and to join distinctive evaluations could give helpful building pieces to plan techniques for combination of a few radar pictures, or to distinguish movement and change in radar pictures.

Mario Mastriani, and Alberto E. Giraldez Et al. [9] They made new sort design to fill in as the shrinkage capacity of DWT-2D. Dissimilar to the standard neural shrinkage strategies, the novel has a lower computational cost. The general ideal execution of NS is investigated and contrasted with the direct dot diminishment strategy. It is demonstrated that the shrinkage dot diminishment techniques are more compelling than direct strategies when the flag vitality focuses on couple of coefficients in the change area. In addition, significantly expanded diversion proportion emphatically demonstrates change in location execution. At long last, the technique is computationally productive and can essentially lessen the dot while saving the determination of the first picture, and maintaining a strategic distance from a few levels of decay and piece impact.

Asoke Nath Et al. [10] The creators have talked about various denoising strategies, for example, separating approach, wavelet based approach, and multifractal approach and a near investigation of every one of these techniques. The LMS versatile channel turns out to be superior to anything the mean channel however has additional time multifaceted nature. The middle channel has no commotion exhibit in it and is near the fantastic picture. The sharpness of the picture is held not at all like on account of straight sifting. For the

situation where a picture is undermined with Gaussian clamor, the wavelet shrinkage denoising has turned out to be about ideal. It makes a decent showing with regards to denoising pictures that are profoundly unpredictable and are defiled with clamor that has a perplexing nature.

Mario Mastriani Et al. [11] They have built up a Projection onto Approximation Coefficients system for picture sifting and pressure inside wavelet area. The reproductions demonstrate that the POAC have preferred execution over the most generally utilized thresholding method for pressure and denoising which incorporate Soft-Thresholding and Hard-Thresholding. In addition, the novel showed to be effective to evacuate increased commotion, and all uncle of clamor in the undecimated wavelet area. At long last, cleaner pictures propose potential enhancements for grouping and acknowledgment.

Kanika Gupta and S.K Gupta Et al. [12] as indicated by specialists, Removal of commotion from the first flag is still a bottleneck for analysts. There are a few strategies and procedures distributed and every strategy has its own particular favorable circumstances, inconveniences and suspicions. In this paper, various measures of Image Denoising Techniques are talked about and it may be conceivable to get mistaken for every one of the strategies, so it is vital to outline those to recover the full substance of the paper. The determination of Denoising system relies on upon what sort of denoising is required. Assist, it relies on upon what sort of data is required. Couple of illustrations, in view of writing survey done in this paper.

Mario Mastriani Et al. [13] The reenactments demonstrate that the FuzzyThresh have preferable execution over the most com-monly utilized channels for SAR symbolism (for the examined seat check parameters) which incorporate factual channels, VisuShrink (ST, HT and SST) with Daubechies wavelet premise 15 and 1 level of decay, SureShrink, Oracle-Shrink, BayesShrink, OracleThresh and an adaptation of TNN. This perception has guided me to figure another versatile edge-safeguarding use of the FuzzyThresh custom-made to spot defiled symbolism. The ideal parameters of this connected calculation to improve dot lessening relies on upon picture attributes, and insights of commotion. In certainties, the novel exhibited to be productive to evacuate increased clamor, and all uncle of commotion in the undecimated wavelet space.

Mario Mastriani Et al. [14] creators depict, the reenactments demonstrate that the POSAShrink have preferable execution over the most usually utilized channels for SAR symbolism which incorporate measurable channels and a few wavelets procedures as far as smoothing uniform locales and safeguarding edges and components. The successfulness of the method empowers the likelihood of utilizing the approach in various ultrasound and radar applications. Indeed, cleaner pictures discuss potential enhancements for grouping and acknowledgment. Plus, significantly expanded avoidance proportion unequivocally demonstrates change in location execution. Fabrizio Argenti, Alessandro Lapini, Tiziano Bianchi and Luciano Alparone Et al. [15] A preprocessing

venture of point focuses on that must hold their radiometry in the wake of despeckling and a portioned approach, in which test measurements are computed on homogeneous sections, finish Bayesian despeckling in wavelet area. New skylines are without a doubt toward com-squeezed detecting, of which denoising is by all accounts a standout amongst the most encouraging application, despite target troubles originate from the flag subordinate, and subsequently no stationary, commotion display.

### 3. PROPOSED ALGORITHM

this section, we discuss image denosing methodology based on SOM neural network model. The image features are extracted from the image using wavelet transform function. SOM acts as a clustering mechanism that projects N-dimensional features from the WT function into an M-dimensional feature space. The resulting vectors are fed into an SOM that categorizes them onto one of the relearned noise classes. The proposed scheme is work along with MS. The MS process the collection task of local intensity of medical image data. The collected noise value combined with high intensity image value and generates vector value for the process. They mapped features from each frame of the word onto the SOM output to form a trajectory of winner nodes for a given word. The SOM learns this trajectory for each denosing constraints value is comprised of a hierarchical organization of SOM and SOM. SOM receives inputs from the WT function bank and maps onto an M-dimensional space where M is the dimensionality of the SOM output node distribution. The transformed feature vectors are fed into the SOM, which classifies them. We call the feature space generated from the WT function output as primary feature space and M-dimensional feature space from SOM output as secondary feature space. The vectors from the secondary feature space are called secondary feature vectors.

#### Processing of proposed Algorithm

Step1. Initially input image passes through WT function and decomposed into two layers different value.

Step2.the layers value different higher and lower part.

Step3. The collection of lower intensity value used MS (multiple sequences)

step4. MS collects the local noise value after that combined with high intensity value.

Step5. After collecting total noise value convert into feature vector image data passes through self organized map network

Step6. In phase of feature mapping in feature space of SOM network create a fixed cluster according to threshold of details of image part.

Step7. Here show steps of processing of SOM network

- 1) Initialize each node's weights.
- 2) Choose a random vector from training data and present it to the SOM.
- 3) Every node is examined to find the Best Matching Unit (BMU).
- 4) The radius of the neighborhood around the BMU is calculated. The size of the neighborhood decreases with each iteration.
- 5) Each node in the BMU's neighborhood has its weights adjusted to become more like the BMU.

Nodes closest to the BMU are altered more than the nodes furthest away in the neighborhood.

- 6) Repeat from step 2 for enough iteration for convergence.
- 7) Calculating the BMU is done according to the Euclidean distance among the node's weights ( $W_1, W_2, \dots, W_{in}$ ) and the input vector's values ( $V_1, V_2, \dots, V_{an}$ ).
  - 1) This gives a good measurement of how similar the two sets of data are to each other.
- 8) The new weight for a node is the old weight, plus a fraction ( $L$ ) of the difference between the old weight and the input vector... adjusted ( $\theta$ ) based on distance from the BMU.
- 9) The learning rate,  $L$ , is also an exponential decay function.
  - 1) This ensures that the SOM will converge.
- 10) The  $\lambda$  represents a time constant, and  $t$  is the time step

Steps 8. After processing of SOM network out data of image  
 Step 9. Finally gets denoise image and calculate the value of PSNR value.

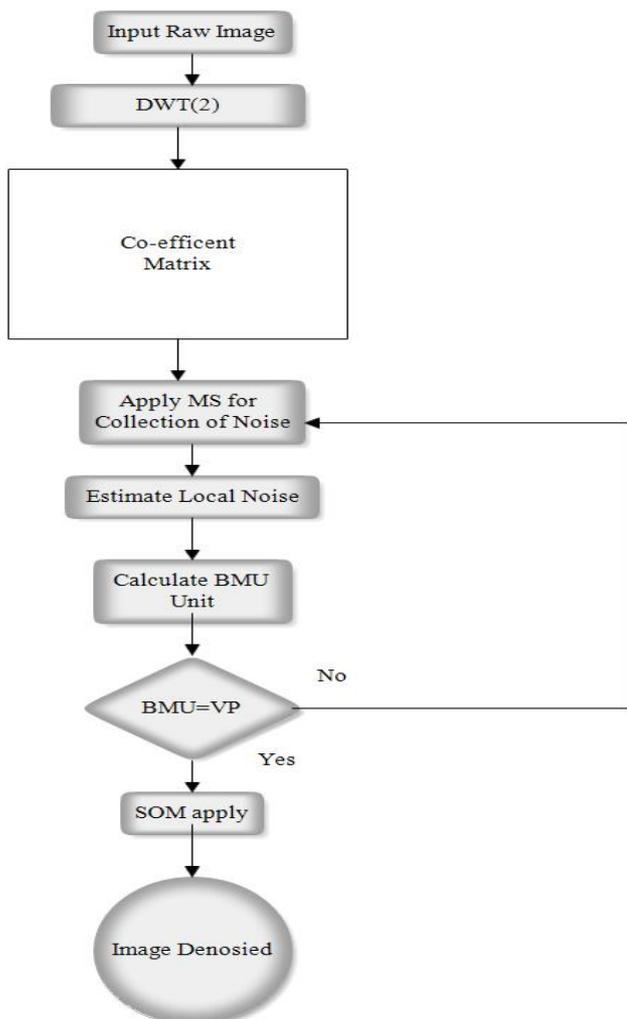


Figure 1 proposed model for image denoising based on wavelet and SOM neural network model.

#### 4. EXPERIMENTAL RESULT

To investigate the effectiveness of the proposed method for image denoising and image filtration. We perform some experimental task; all these tasks perform in matlab 7.8.0 software and well famous image data set such as cameraman, baboon and Leena image. The PSNR value measure for the evaluation of performance.

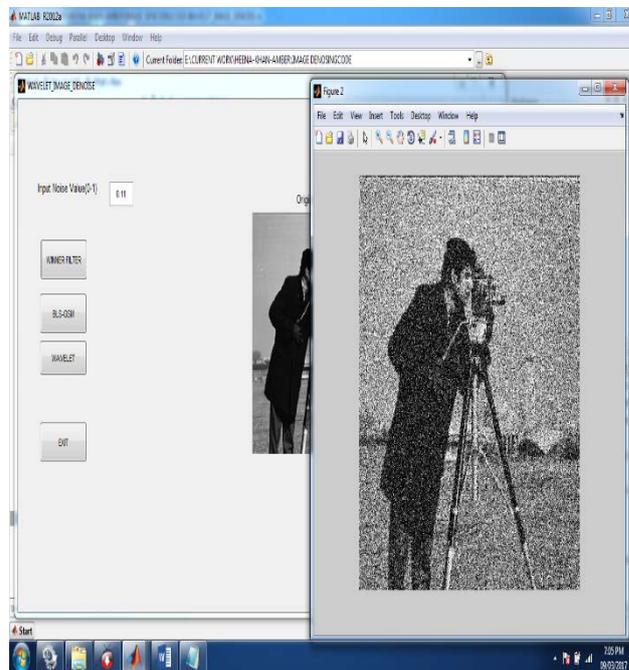


Figure2 : window show that proceeding output after the importing a 1.jpg image with input noise value is 0.11 using winner filter method in our implementation.

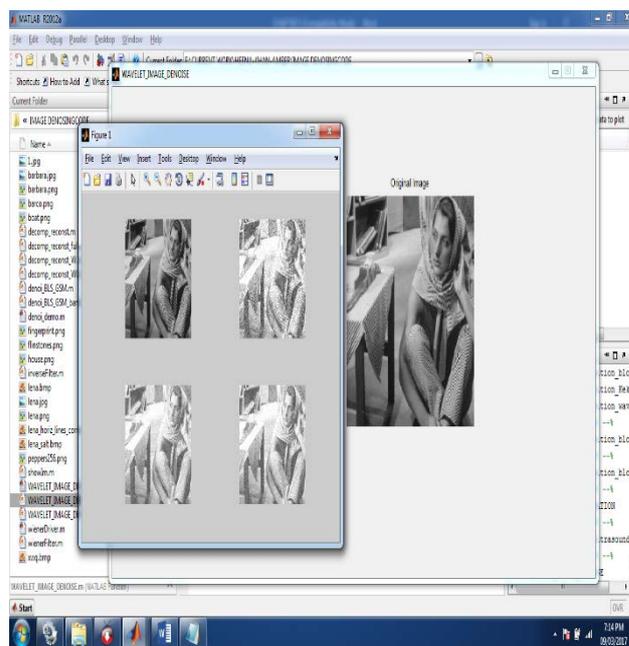


Figure 3: window show that proceeding output after the importing a barbara.jpg image with input noise value is 0.28 using Wavelet method in our implementation.

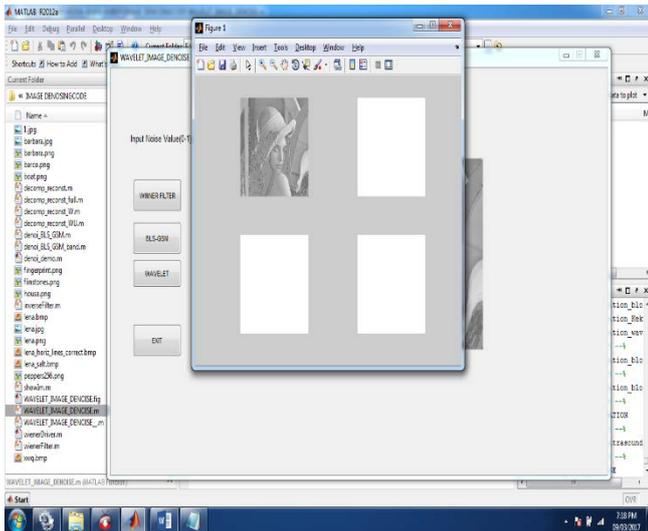


Figure 4 : window show that final output after the importing a lena.jpg image with input noise value is 0.87 using BLS-GSM method in our implementation.

COMPARATIVE PERFORMANCE EVALUATION

Method	PSNR (DB)	Used Image	Input Noise Value
WINNER FILTER METHOD	9.7256	barbara.jpg	0.28
BLS-GSM METHOD	7.1090		
WAVELET METHOD	12.0829		

Table 1: Shows the Comparative output PSNR value using Winner filter, BLS-GSM and Wavelet method applied on barbara.jpg Image and input noise value 0.28.

Method	PSNR (DB)	Used Image	Input Noise Value
WINNER FILTER METHOD	7.9953	lena.jpg	0.87
BLS-GSM METHOD	3.7570		
WAVELET METHOD	8.7572		

Table 2: Shows the Comparative output PSNR value using Winner filter, BLS-GSM and Wavelet method applied on lena.jpg Image and input noise value 0.87.

COMPARATIVE PERFORMANCE ANALYSES

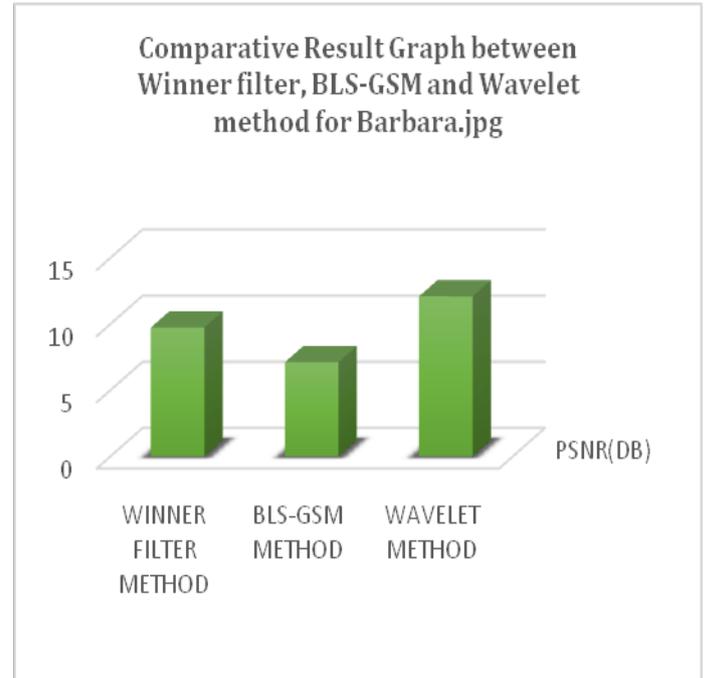


Figure 5 : Comparative result graph for the Barbara image using all image Denosing (Winner filter, BLS-GSM and Wavelet method) methods and here we get resultant values PSNR for input noise value 0.28.

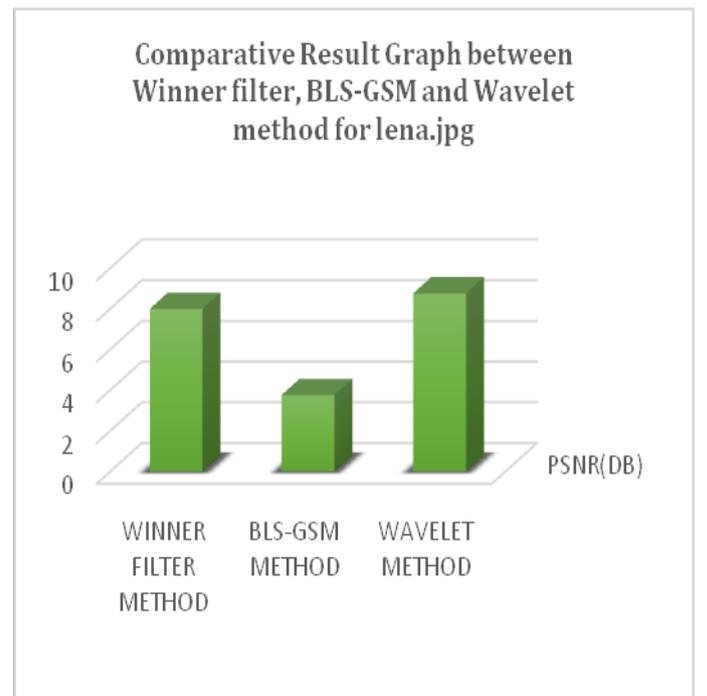


Figure 6: Comparative result graph for the Lena image using all image Denosing (Winner filter, BLS-GSM and Wavelet method) methods and here we get resultant values PSNR for input noise value 0.87.

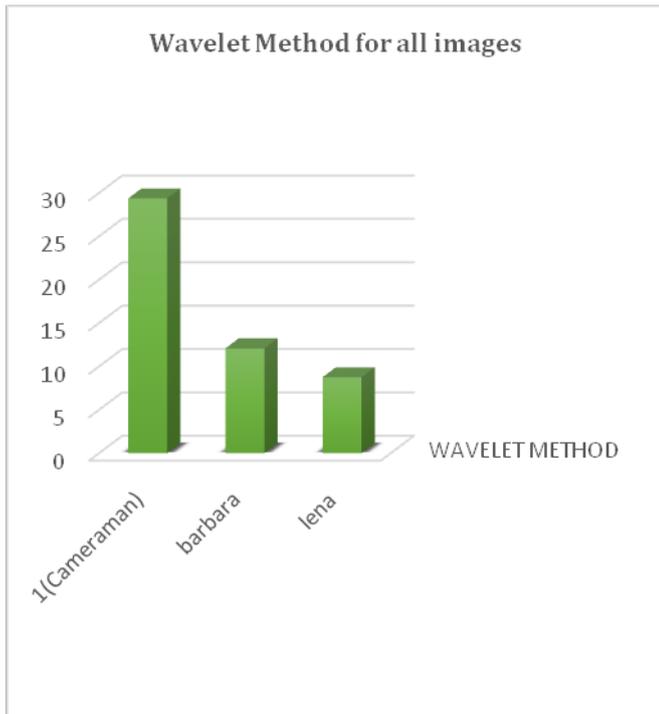


Figure 7: implementation resultant graph on the basis of PSNR values for all images (1.jpg, Barbara and Lena) using Wavelet method.

## 5. CONCLUSION AND FUTURE WORK

In this paper a hybrid of SOM-wavelet method based on wavelet transform function and neural networks is proposed. SOM were used to find correlation between noised and original DWT coefficients and approximation. Experimental results showed capability of proposed method to remove noise in terms of PSNR and visual quality. Different architectures and different activation functions is considered. The experimental results show the mean with the traditional denoising methods, the proposed threshold-based denoising digital image denoising algorithm for mixed digital image denoising is relatively clear, especially in the more noise, more complex cases", can show its good performance. In the denoising process in order to achieve better denoising effect, the system takes more time to pay; the other for color digital image processing has not been a good result. Therefore, focus on late goals and improve the efficiency of color image denoising. However, the algorithm has a disadvantage of needing more computing time when select a larger hybrid generation. This will be a key problem to solve in the following work.

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