

# Image Restoration in Non-Linear Filtering Domain using MDB approach

S. K. Satpathy, S. Panda, K. K. Nagwanshi and C. Ardil

**Abstract**—This paper proposes a new technique based on non-linear Minmax Detector Based (MDB) filter for image restoration. The aim of image enhancement is to reconstruct the true image from the corrupted image. The process of image acquisition frequently leads to degradation and the quality of the digitized image becomes inferior to the original image. Image degradation can be due to the addition of different types of noise in the original image. Image noise can be modeled of many types and impulse noise is one of them. Impulse noise generates pixels with gray value not consistent with their local neighborhood. It appears as a sprinkle of both light and dark or only light spots in the image. Filtering is a technique for enhancing the image. Linear filter is the filtering in which the value of an output pixel is a linear combination of neighborhood values, which can produce blur in the image. Thus a variety of smoothing techniques have been developed that are non linear. Median filter is the one of the most popular non-linear filter. When considering a small neighborhood it is highly efficient but for large window and in case of high noise it gives rise to more blurring to image. The Centre Weighted Mean (CWM) filter has got a better average performance over the median filter. However the original pixel corrupted and noise reduction is substantial under high noise condition. Hence this technique has also blurring affect on the image. To illustrate the superiority of the proposed approach, the proposed new scheme has been simulated along with the standard ones and various restored performance measures have been compared.

**Keywords**—Filtering, Minmax Detector Based (MDB), noise, centre weighted mean filter, PSNR, restoration.

## I. INTRODUCTION

IMAGE processing algorithms are designed to handle different problem domains. Some of the broad area includes (i) Image Representation and Modeling Image Transforms, (ii) Image Enhancement Image Restoration, (iii) Image Analysis Image Reconstruction, and (iv) Image Data Compression. Many image-processing algorithms are designed to deal with the above-mentioned problems. Image restoration is principally refers to the removal or minimization of known

degradation in an image *i.e.* the recovery of an image that has been degraded by some prior knowledge of the degradation phenomenon. Consequently the restoration techniques are oriented towards modeling the degradation and applying the inverse process in order to recover the original image. Instead of using traditional frequency domain concept, here some algebraic methods have been used for the system, which has the advantage of computational complexity over the frequency domain restoration technique. When one would like to remove the noise, it follows certain filtering operation where the signal has to be passed through a filter and the filter in turn removes the undesirable components. Frequently linear techniques are used because linear filters are easy to implement and design. Further they are optimal among the classes of all filtering operations when the noise is additive and Gaussian [1].

### A. Image Noise Model

When it has been discussed on noise it can be getting introduced in the image, either at the time of image generation (e.g. when we use camera and photographic films to capture an image) or at the time of image transmission. Accordingly different categories of noise having certain characteristics. In photographic films; the recording noise is mainly due to the silver grains that precipitate during the film exposure. They behave randomly during both film exposure and development. They are also randomly located on the films. This kind of noise, which is due to silver grains, is called film grain noise. This is a Poisson process and becomes a Gaussian process in its limit. The film grain noise does not any statistical correlation for distance between samples greater than the grain size. Therefore film grain noise is a white noise two-dimensional random process. In photo electronic detectors, two kind of noise appears: (a) Thermal Noise: Its source is the various electronic circuits. It's a two-dimensional additive white zero-mean Gaussian noise, and (b) Photoelectron Noise: it is produced by random fluctuation of the number of photons on the light sensitive surface of the detector [2]. If its level is low it can be treated as a poisson distributed noise otherwise it can be treated as a Gaussian distributed noise. Thus it can be said as signal dependent noise. Another kind of noise that is present during the image transmission is Salt-Pepper noise [3]. It appears as black and/or white impulse of the image. Its source is usually man-made or atmospheric noise, which appears as impulsive noise. It has following form: where  $z(k,j)$  denotes the impulse and  $i(k,j)$  denotes the original image intensity at the pixel  $(k, j)$ . In case of the-CCD cameras, the main form of noise is the transfer loss noise.

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### B. Digital Filters

Elimination of noise is one of the major works to be done in computer vision and image processing, as noise leads to the error in the image. Presence of noise is manifested by undesirable information, which is not at all related to the image under study, but in turn disturbs the information present in the image. It is translated into values, which are getting added or subtracted to the true gray level values on a gray level pixel. These unwanted noise information can be introduced because of so many reasons like: acquisition process due to cameras quality and restoration, acquisition condition, such as illumination level, calibration and positioning or it can be a function of the scene environment. Noise elimination is a main concern in computer vision and image processing. A digital filter [1][3] is used to remove noise from the degraded image. As any noise in the image can be result in serious errors. Noise is an unwanted signal, which is manifested by undesirable information. Thus the image, which gets contaminated by the noise, is the degraded image and using different filters can filter this noise. Thus filter is an important subsystem of any signal processing systems. Thus filters are used for image enhancement, as it removes undesirable signal components from the signal of interest. Filters are of different type i.e. linear filters or nonlinear filters [20]. In early times, as the signals handled were analog, filters used are of analog. Gradually digital filters were took over the analog systems because of their flexibility, low cost, programmability, reliability, etc. for these reasons digital filters are designed which works with digital signals. The design-of digital filters involves three basic steps: (i) the specification of the desired properties of the system, (ii) the approximation of these specifications using a causal discrete-time system, and (iii) the realization of the system using finite precision arithmetic.

The paper has been organized in the following manner; section II proposes problem formulation and solution methodology, section III describes the result and discussion, section IV gives concluding remarks and further works and finally section V incorporates all the references been made for completion of this work.

## II. PROBLEM FORMULATION AND SOLUTION METHODOLOGY

As it had seen that noise elimination is a main concern in computer vision and image processing. For example in many applications where operators based on computing image derivatives is applied, any noise in the image can results in serious errors. Noise presence is manifested by undesirable information, not related to the scene under study, which perturbs the information relative to the form observable in the image. It is translated to more or less severe values, which are added or subtracted to the original values on a number of pixels. Noise is of many types. Thus image noise can be Gaussian, Uniform or impulsive distribution. Here we will discuss about the, impulse noise. Impulse noise generates pixels with gray level values not consistent with their local neighborhood values. The impulse noise appears as sprinkle of

light or dark spots in the image. This impulse noise can be eliminated or the degraded image can be enhanced by the use of advance filter. Due to certain disadvantages of linear filters, nonlinear method of filtering has been proposed in this paper. The filter cannot be modeled by convolutions fall into the category of nonlinear filter. Nonlinear filter can be very effective in removing the impulse noise. The median filter replaces the middle pixel value with the median value of the window. The most popular non-linear filter is the median filter. When considering a small neighborhood it is highly efficient and has proved to be very effective in removing noise of an impulsive nature despite its simple definition. Nevertheless, the median filter often fails to provide sufficient smoothing of non-impulsive noise and its result is sometimes unpredictable. Several techniques have been proposed which try to take the advantage of the average performance of the median filter, either to evaluate noise density, set up parameters or to guide the filtering process.

The weighted median (WM) filter [9] is an extension of the median filter, which -gives more weight to some values within the window. This WM filter allows a degree of control of the smoothing behavior through the weights that can be set, and therefore, it is a promising image enhancement technique. A special case of WM filter called center weighted median (CWM) filter [10]. This filter gives more weight only to the central value of a window, and thus it is easier to design and implement than general WM filters. These approaches involve a preliminary identification of corrupted pixels in an effort to prevent alteration of true pixel values.

A nonlinear smoothing filter namely the Peak-and-Valley filter is proposed to reduce impulsive-like noise while modifying the gray levels of the image as little as possible, resulting in a maximum preservation of the original information [11]. It could be used as a basis for the design of more sophisticated impulsive noise elimination filters, replacing the conventional median filter in performing the preliminary processing. In this paper thrust has been made to devise a filtering scheme to remove impulse noise from images such that the scheme should work at high noise conditions and should perform superior to the existing schemes in terms of noise rejection and retention of original image properties. The detection scheme is devised keeping the CWM filter in mind whereas the median filter is used for the filtering operation for detected noisy pixels. Extensive simulation has been carried out to compare the performance of the proposed filter with other standard schemes.

### A. Solution Methodology

In any noise removal schemes, attention is given to remove noise from images in addition to keeping as much original properties as possible. However, since both the objectives are contradicting in nature, it is not possible for any nonlinear scheme to fulfill both the objectives. It has been observed through the simulation of the existing schemes [10,11,20] that they also fail in providing satisfactory results under high noise conditions. Since impulse noise is not uniformly distributed across the image, it is desirable to replace the corrupted ones

through a suitable filter. For this purpose, a preprocessing is required to detect the corrupted location prior to filtering.

### B. Mathematical Analysis

To assess the performance of the proposed filters for removal of impulse noise and to evaluate their comparative performance, different standard performance indices have been used in the thesis. These are defined as follows:

**Peak Signal to Noise Ratio (PSNR):** It is measured in decibel (dB) and for gray scale image it is defined as:

$$\text{PSNR (dB)} = 10 \log_{10} \left[ \frac{\sum_i \sum_j 255^2}{\sum_i \sum_j (S_{i,j} - \hat{S}_{i,j})^2} \right] \quad (1)$$

Where  $S_{i,j}$  and  $\hat{S}_{i,j}$  are the original and restored image pixels respectively. The higher the PSNR in the restored image, the better is its quality.

**Signal to Noise Ratio Improvement (SNRI):** SNRI in dB is defined as the difference between the Signal to Noise Ratio (SNR) of the restored image in dB and SNR of restored image in dB i.e.

$$\text{SNRI (dB)} = \text{SNR of restored image in dB} - \text{SNR of noisy image in dB} \quad (2)$$

Where,

SNR of restored image dB=

$$10 \log_{10} \left[ \frac{\sum_i \sum_j S_{i,j}^2}{\sum_i \sum_j (S_{i,j} - \hat{S}_{i,j})^2} \right] \quad (3)$$

SNR of Noisy image in dB =

$$10 \log_{10} \left[ \frac{\sum_i \sum_j S_{i,j}^2}{\sum_i \sum_j (S_{i,j} - X_{i,j})^2} \right] \quad (4)$$

Where,  $X_{i,j}$  is Noisy image pixel

The higher value of SNRI reflects the better visual and restoration performance.

**Percentage of Spoiled Pixels (POSP):** It may be defined as the number of unaffected original pixels replaced with a different gray value after filtering i.e.

POSP =

$$\left[ \frac{\text{Number of original pixels change their grayscale}}{\text{Number of non - noisy pixels}} \right] \times 100 \quad (5)$$

**Percentage of Noise Attenuated (PONA):** It may be defined as the number of pixels getting improved after being filtered.

PONA =

$$\left[ \frac{\text{Number of noisy pixels getting improved}}{\text{Total number of noisy pixels}} \right] \times 100 \quad (6)$$

This parameter reflects the capability of the impulse noise detector used prior to filtering. Small value of POSP shows the improved performance of the detector. High value of POSP leads to removal of original image properties and edge jitters in restored images. The more the percentage the better will be the attenuation characteristic of the filter filtering.

### C. Proposed Algorithms

#### Algorithm: MDB Filter

- Step 1. Take corrupted image (X).
- Step 2. Take a 3 x 3 window (W).  
Let the center pixel be the test pixel.
- Step 3. Shift the window row wise then column wise to cover the entire pixel in the image and repeat Step4 and Step5.
- Step 4. If (test pixel < min (rest of the pixel in W) OR (test pixel > max (rest of the pixel in W) then the test pixel is corrupted.
- Step 5. If the test pixel is corrupted apply median filter to the test filter in the window W.
- Step 6. Stop.

### III. RESULT AND DISCUSSION

The quantitative results has been given in table [Table I – Table VI] for the standard LENA image, for different percentage of noise, starting from 5% to 30% with a step of 5%, and the comparative analysis has been presented in figure [fig. 3 and fig. 4] for both LENA and CAMMAN image showing the performance of CWM filters for gain =1 and 2 respectively from 5% to 30% along with Salt and Pepper noise. Along with the figures and tables some graphs, shown in figure [graph1 – graph 2], has also been given, for all the quantitative measures, for both LENA and CAMMAN image to have a quick insight into the comparative performance of the existing filters along with the proposed one i.e. MDB filter.

TABLE I  
% NOISE ATTENUATED FOR LENA.TIF

% Impulse Noise	CWM for K=1	CWM for K=2	MDB
5	98.7562	97.6801	99.757
10	98.8724	96.4177	99.8234
15	97.572	92.7032	99.7087
20	97.9097	85.4656	99.5008
25	96.1055	80.0837	99.2512
30	95.5977	75.0516	99.2028

TABLE II  
% IMAGE SPOIL FOR LENA.TIF

% Impulse Noise	CWM for K=1	CWM for K=2	MDB
5	47.3299	27.2792	21.6439
10	46.5746	25.7312	19.4961
15	46.5171	23.633	18.0531
20	46.1974	22.6867	16.9588
25	44.5388	20.9869	14.7011
30	44.4892	20.2045	13.566

TABLE III  
% SNR RESTORED FOR LENA.TIF

% Impulse Noise	CWM for K=1	CWM for K=2	MDB
5	21.9903	23.3947	22.9579
10	21.101	20.5031	22.2465
15	19.4544	17.3634	21.8329
20	18.5534	14.5687	20.7446
25	16.7922	13.0016	19.4695
30	15.9019	11.3396	18.8912

TABLE IV  
% SNR OF NOISY FOR LENA.TIF

% Impulse Noise	CWM for K=1	CWM for K=2	MDB
5	13.5613	13.6655	13.4663
10	10.5288	10.4552	10.3947
15	8.9576	8.872	8.9921
20	7.6403	7.5795	7.6698
25	6.5894	6.7567	6.5552
30	5.8324	5.941	5.926

TABLE V  
% DIFFERENCE IN SNR FOR LENA.TIF

% Impulse Noise	CWM for K=1	CWM for K=2	MDB
5	8.4291	9.7291	9.4916
10	10.5721	10.0479	11.8519
15	10.4968	8.4915	12.8408
20	10.1931	6.9892	13.0949
25	10.2028	6.2449	12.9143
30	10.0695	5.3986	12.9652

TABLE VI  
PEAK SIGNAL TO NOISE RATIO FOR LENA.TIF

% Impulse Noise	CWM for K=1	CWM for K=2	MDB
5	27.4687	28.8749	28.4385
10	26.5805	25.9823	27.7254
15	24.9343	22.8439	27.312
20	24.0316	20.0496	26.2449
25	22.2725	18.4822	24.9492
30	21.3819	16.8204	24.3714

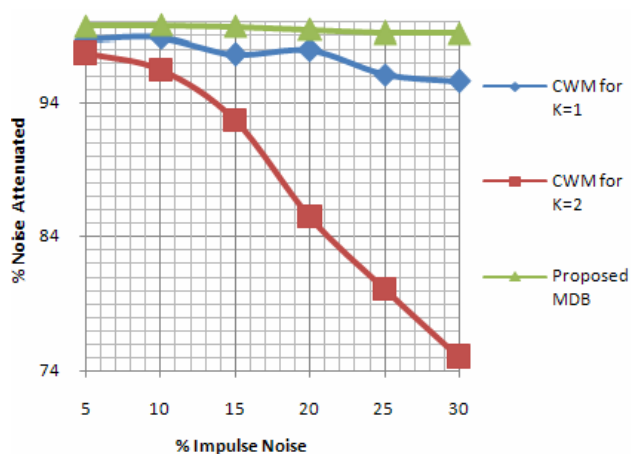


Fig. 1: Comparison of % Impulse Noise vs. % Noise Attenuated

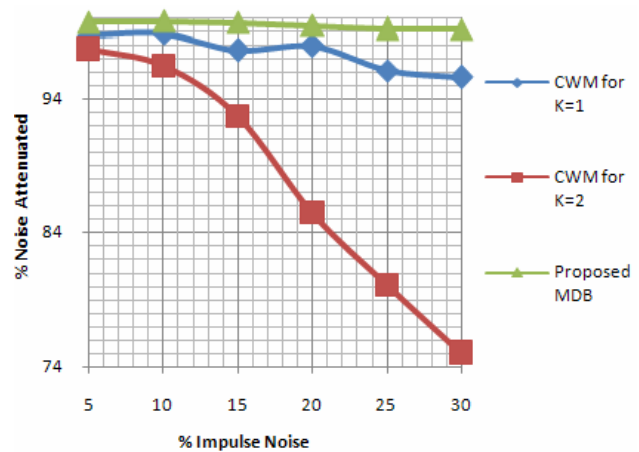


Fig. 2: Comparison of % Impulse Noise vs. PSNR

Given Image      Noisy : Noise Density in %:5



CWM K=1

Proposed MDB



Noisy : Noise Density in %:10      CWM K=1      Proposed MDB



Noisy : Noise Density in %:30      CWM K=1      Proposed MDB



Fig. 3: Result with Cameraman.tif 256x256 image at various noise levels

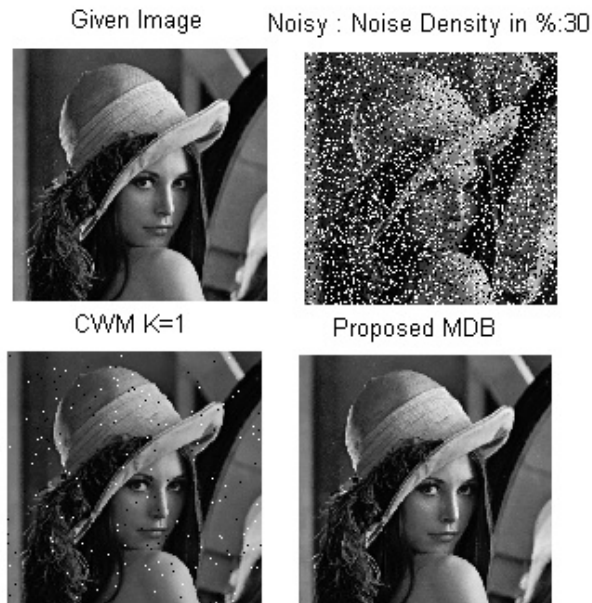


Fig. 3: Result with Lena.tif 256x256 image at 30% Impulse Noise

#### IV. CONCLUSION AND FURTHER ENHANCEMENT

This paper proposed new non-linear filters to remove the impulse noise from the images. To illustrate the efficiency of the proposed MDB schemes, we have simulated the new schemes along with the existing ones and various restored measures have been compared. All the filtering techniques have been simulated in MATLAB 7.1 with Pentium-IV processor. The schemes are simulated using standard images LENA and CAMMAN. The impure contaminations include Salt and Pepper noise. The proposed schemes MDB filter is found to be superior, i.e. better restored results and other parameter for restoration compared to the existing schemes when Salt and Pepper impulse noise is considered. The noise removal schemes can be extended based on other types of detection like Fuzzy and ANN. There is another emerging area of image restoration technique, called Blind De-Convolution technique. In this direction further work can be extended by utilizing soft computing techniques, like FUZZY and ANN, so that the properties of images can be better retained in the restored images.

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#### REFERENCES

- [1] A.K.Jain, "Fundamentals of Digital Image Processing", Engelwood Cliff, N. J.: Print ice Hall, 2006.
- [2] T.S. Huang, G.J. Yang, And G.Y. Tang, "A Fast Two Dimensional Median Filtering Algorithm", IEEE Trans. On Accustics, Speech, Signal Processing, Vol. ASSP-27, No.1, Feb 1977.
- [3] J.S.Lee, "Digital Image Enhancement and Noise Filtering by use of Local Statistics", IEEE Trans. On Pattern Analysis and. Machine Intelligence, Vol.PAMI-29, March,1980.
- [4] E. Ataman, V.K. Aatre, and K.W. Wong, "Some Statistical Properties of Median Filters", IEEE Trans. On Acoustics, Speech, Signal Processing, Vol. ASSP-29, No.5, October, 1981.
- [5] Pitas, and A.N. Vanetsanopoulos, "Non Linear Mean Filters In Image Processing", IEEE Trans. On Accustics, Spe.ech, Signal Processing, Vol.10, No.4, June 2006.
- [6] G. Panda, B. Majhi and P.K. Dash, "Development of Efficient Filtering Technique for Image Data Corrupted by Impulse Noise", CSI 2004.S.E. Umbaugh, "Computer Vision and Image Processing: A Practical Approach", Englewood Cliffs, Nj Prantice-Hall, 1 998, pp. 151-193.
- [7] H.A. David, "Order Statistics", Newyork Willy 1981.
- [8] D. Brownrigg, "The Weighted Median Filter" Communication Association, pp. 807-993, Sept.1991.
- [9] S.J. Kuo and Y.H. Lee, "Center Weighted Median Filters and Their Applications to. Image Enhancement", IEEE Trans. Circuit. Syst. Vol.38, pp.984-993, Sept.1991.
- [10] Piotr S. Windyaga, "Fast Implusive Noise Removal", IEEE Trans. On Image Processing, Vol.1 0, No. 1, January 2001.
- [11] J.Astola And P.Kuosmanen, "Fundamentals Of Nonlinear Digital Filtering", Boca Raton, FL: CRC, 1997.
- [12] I.Pitas and A.N. Vanetsanopoulos. "Non Linear Digital Filters: and Applications", P.Ch. The Kluwer Academic Publishers.
- [13] J.W. Tukley, "Explanatory Data Analysis", Addison-Wesley, 1977.
- [14] W.K. Pratt, "Digital Image Processing", J.Wiley, 2008.
- [15] H.A David, "Order Statistics", John- Wiley, 1981.
- [16] P.S Hiber, "Robust Statistics", John Willey, 1981.
- [17] L.R RCibiner, RW. Schafer, "Digital Processing Of Speech Signals", Prantice-Hall, 2006.
- [18] H.C. Andrews, B.R. Hunt, "Digital Image Restoration", Prantice-Hall, 1997.
- [19] RC. Gonzalilez, RE. Woods, "Digital Image Processing", Addison Wesley, 2008.
- [20] H.Taub, D.L. Schilling, "Principles of Communication Systems", TMH, 1991.
- [21] B.B. Choudhury, and D.Datta Majumdar, "Two-Tone Image Processing and recognition", Wiley Estern Limited, 1993.

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