

AN UNIQUE EDGE PRESERVING NOISE FILTERING TECHNIQUE FOR IMPULSE NOISE REMOVAL

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Abstract

Image de-noising is the technique to reduce noises from corrupted images. The aim of the image de-noising is to improve the contrast of the image or perception of information in images for human viewers or to provide better output for other automated image processing techniques. This paper presents a new approach for color image de-noising with Fuzzy Filtering techniques using centroid method for defuzzification. It preserves any type of edges (including tiny edges) in any direction. The experimental result shows the effectiveness of the proposed method.

Keywords

Impulse noise; Median filter; Color image de-noising; Defuzzification; Edge Preservation;

1. INTRODUCTION

Color provides a significant portion of visual information to human beings and enhances their ability of object detection. In the RGB color model, the three primaries are Red, Green, and Blue. They can be combined, two at a time to create the secondary hues. Magenta is (R+B), cyan is (B+G), and yellow is (R+G). There are two reasons for using the red, green, and blue colors as primaries: (1) the cones in the eye are very sensitive to these colors and (2) adding red, green, and blue can produce many colors although not all colors. For various reasons digital images are often contaminated with noise at the time of acquisition or transmission. The noise introduces itself into an image by replacing some of the pixels of the original image by new pixels having luminance values near or equal to the minimum or maximum of the allowable dynamic luminance range. Pre-processing of an image is conducted with a view to adjusting the image for further classification and segmentation. In the process, however, image features should not be destroyed. This is a difficult task in any image processing system. For this purpose various types of filters are used. Among those median filter is an important class of filters. In the present paper we shall discuss some of the median filters.

1.1 Different median filters:

Some important median filters are discussed below in short.

The standard median filter [28,24,8,17,11] : In this filter luminance values in a window are arranged in an order and the median value is selected. This filter reduces noise reasonably well, but in the process some information is also lost at low noise densities.

The weighted median filter [30,19,11,10] and the centre-weighted median filter [16,13,25,6,11] have been proposed to avoid the inherent drawbacks of the standard median filter by controlling the trade off between the noise suppression and detail preservation.

The switching median filter [31,11] is a type of median filter with an impulse detector. It is so designed that if the centre pixel is identified by the detector as a corrupted pixel, then it is replaced with the output of the median filter, otherwise, it is left unchanged. The tri state median filter [7] and the multi-state median filter (MSMF) [8] are two modification of switching median filter.

Two other types of switching median filters worth mentioning are progressive switching median filter (PSMF) [29] and Signal-dependent rank-ordered mean filter (SDROMF) [1,2].

The progressive switching median filter (PSMF) [29] is a derivative of the basic switching median filter. In this filtering approach, detection and removal of impulse noise are iteratively done in two separate stages. Despite its improved filtering performance it has a very high computational complexity due to its iterative nature. Signal-dependent rank-ordered mean filter (SDROMF) [1,2] uses rank-ordered mean filter.

Adaptive centre weighted median (ACWM) [19,11] filter avoids the drawbacks of the CWM filters and switching median filter. Input data will be clustered by scalar quantization (SQ) method, this results in fixed threshold for all of images.

Fuzzy-Median filters of different types have gained lot of importance in image processing. Some such filters [3,18,27,5,22,15] are worth mentioning.

2. PROPOSED METHOD FOR DE-NOISING

2.1 Phase I

Any given digital RGB color image can be represented by a two dimensional array of size $M \times N$ for each Red, Green and Blue primaries. When we are capturing or transmitting the images, then there will be some noises due to improper opening and closing of the shutter, atmospheric turbulence, misfocus of lens, relative motion between camera and objects [12]. For removing the noise from the image we can pursue the following steps:

Step 1: Let the input and output images of median filter are $X(i,j)$ & $y_1(i,j)$ respectively. The median filter uniformly replaces the central pixel of a window (W) by the median of the pixels bounded by predefined window (W) size $w \times w$. The output of the median filter [26,10,32] is given by

$$y_1(i,j) = \text{median} \{ X(i-s; j-t) : (s,t) \in W \} \quad (1)$$

$$\text{and } W = \{ (s, t) : -n \leq s, t \leq n \} \quad (2)$$

Here we consider $n=1$

Step 2: The pixel values of the neighborhood pixels of (i,j) of original image are sorted in some specific order.

Step 3: The Fuzzy membership [14,22,20,17] value is assigned for each pixel in a window of $w \times w$ size using some suitable membership functions. The membership function used here is given below.

- (i) A triangular shaped membership function is used.
- (ii) The highest and lowest gray values get the membership value zero.
- (iii) Assign the membership value 1 to the mean value of the gray level of the window of size $w \times w$.

The triangular membership function [14,20,17], also called bell-shaped function with straight lines, can be defined as follows:

$$\begin{aligned} \mu(x;\alpha,\beta,\gamma) &= 0 && \text{if } x \leq \alpha \\ &= (x-\alpha)/(\beta-\alpha+1) && \text{if } \alpha < x \leq \beta \\ &= (\alpha-x)/(\beta-\alpha+1) && \text{if } \beta < x \leq \gamma \end{aligned} \quad (3)$$

One typical plot of the triangular membership is given in the following figure.

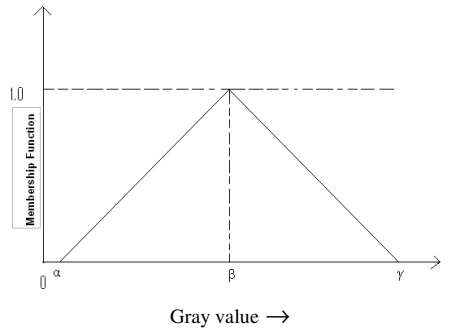


Figure 1: Triangular membership function

Step 4: Now de-fuzzyfy [14,20,21,32] the membership value using the Centroid method by the following formula, and select the output for that window. Let it be $y_2(i,j)$ at the point (i,j) in the window of size $w \times w$

$$y_2(i, j) = \frac{\sum_x x \mu(i, j)}{\sum_x \mu(i, j)} \quad (4)$$

2.2 Phase II

If we apply the phase I on the noisy images, noise would be removed. But the problem is, though the noises are represented as 0 and 255, so all 0 and 255 will also be removed; even of those were not noises. For preserving the actual data we pursues the following steps

Step 1: For the noisy pixel at the point (i,j) , we compute $p(i,j)$ and $q(i,j)$ by the following formulas [21]:

$$p(i,j) = |f(i,j) - \text{median}\{L f(i,j)\}|.$$

Here $L(f(i,j))$ is the 8-neighborhood of point (i,j) .

$$q(i,j) = \frac{|f(i,j) - f_{c_1}(i,j)| + |f(i,j) - f_{c_2}(i,j)|}{2} \quad (5)$$

Here $f_{c_1}(i,j)$, $f_{c_2}(i,j)$ are the closest values of $f(i,j)$ in the filter window of size $w \times w$.

Then rearrange $p(i,j)$ and $q(i,j)$ in ascending order for all $i,j = 0,1,2$.

Step2: Compute $w(i,j) = F(p(i,j), q(i,j))$ such that

$$w(i,j) = 1 - \frac{p(i,j) + k_1}{p(i,j) + q(i,j) + k_2} \quad (6)$$

Here k_1 and k_2 are real quantity, which depends on the quality of the input images. Select the maximum $w(i,j)$.

Step3: If $y_1(i,j) \neq y_2(i,j)$
 then $y_3(i,j) = w(i,j)y_1(i,j) + (1 - w(i,j))y_2(i,j)$ (7)
 Else $y_3(i,j) = y_1(i,j)$

Step 4: Continue this process for all the pixels which are noisy.
 The computational procedure is illustrated below.
 Example: Consider a color image of size 3×3 , as follows,

Table 1: color image with size 3×3

163	193	151		83	113	71		55	81	39
179	0	138		99	0	58		68	0	29
192	172	127		112	92	47		85	63	21

Red

Green

Blue

For the Green component, we calculate membership values as follows

Original value: 0 Mean value: 75 Median value $y_1(i,j)$: 83

Table 2 : Membership values for pixels

Sorted order	0	47	58	71	83	92	99	112	113
Membership value	0.00	0.26	0.55	0.89	0.79	0.55	0.37	0.26	0.00

According to Table 2 the graph is as follows –

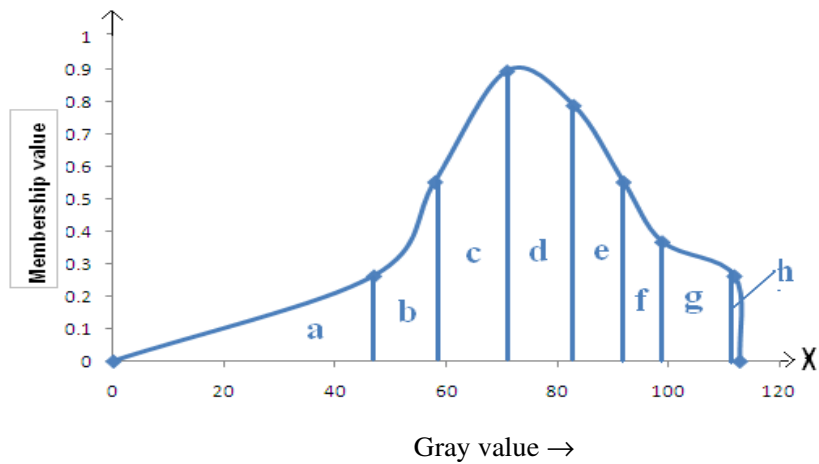


Figure 2

Now we calculate $y_2(i,j)$ using centroid method

Table 3

Sector	Area of the sector	Centroid of the sector
a	6.184211	23
b	4.486842	52
c	9.407894	64
d	10.105263	77
e	6.039474	87
f	3.223684	95
g	2.565789	105
h	0.013158	112

So, the calculated value for $\sum_x x\mu(i, j) = 2858.328857$

And the calculated value for $\sum_x \mu(i, j) = 42.02631$

Therefore, $y_2(i, j) = 68$ [Selected value in phase I]

Rearranged $p(i,j)$ and $q(i,j)$ as follows

Table-4: Different values of p, q and w.

$p(i,j)$	$q(i,j)$	$w(i,j)$
0	7	0.956522
9	7.5	0.207792
12	8	0.509804
16	10	0.388889
25	10.5	0.388278
29	12	0.328947
30	12.5	0.202703
36	17.5	0.47222
83	52.5	0.330275

Therefore $y_3(i,j) = 82$

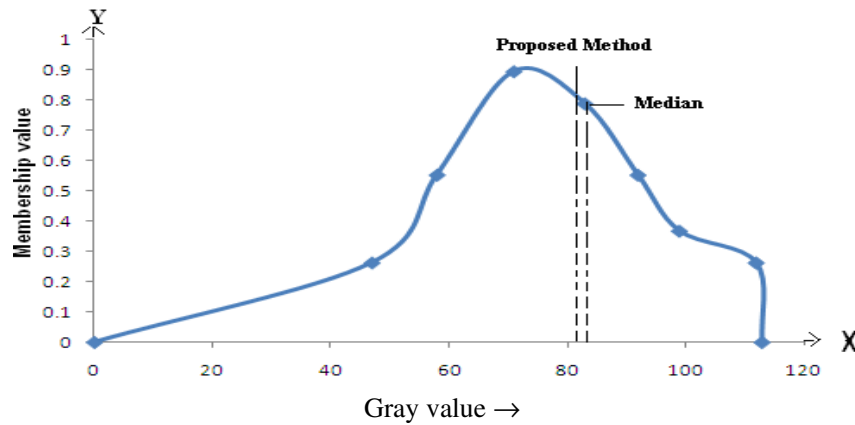


Figure 3

The figure 3 shows the membership function for the pixel values of Green component of RGB image. And also indicate the selected value in the proposed method of Green component.

Now, for the Red component of RGB image, the selected values are as follows

Original value: 0

Mean value: 146

Median value $y_1(i, j)$: 163

$$y_2(i, j) = 108 \text{ [Selected value in phase I]}$$

And $y_3(i, j) = 160$ [in Phase II]

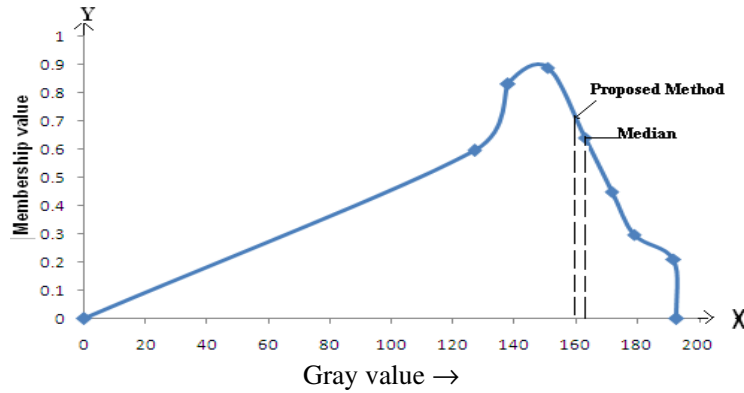


Figure 4 : Membership function for Red component

The Figure 4 shows the membership function for the pixel values of Red component of RGB image. And also indicate the selected value in proposed method for RED component.

Similarly, for the Blue component of RGB image, the selected values are as follows

Original value: 0

Mean value: 49

Median value $y_1(i, j)$: 55

$$y_2(i, j) = 47 \text{ [Selected value in phase I]}$$

and $y_3(i, j) = 54$

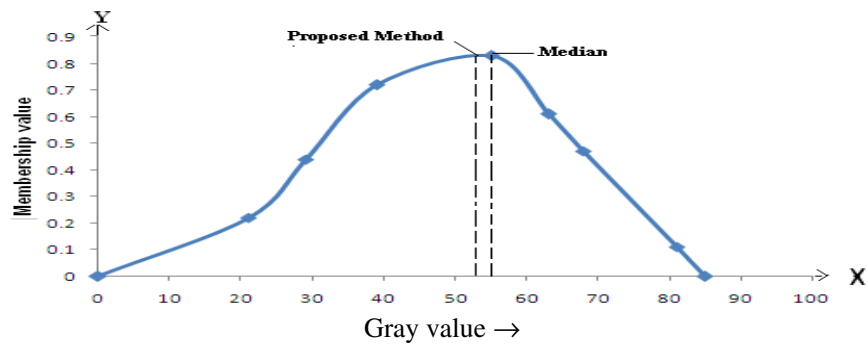


Figure 5: Membership function for Blue component

The Figure 5 shows the membership function for the pixel values of Blue components of RGB image. And also indicate the selected value in proposed method.

3. FLOW CHART

The flow chart of the proposed method is shown below.

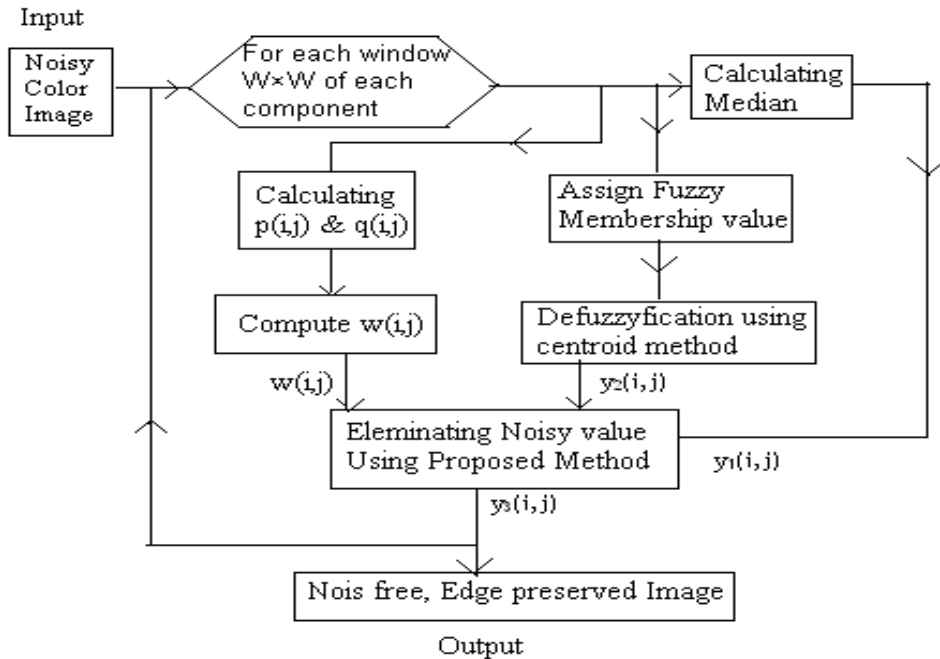


Figure 6

4. EXPERIMENTAL RESULT

The effectiveness of the proposed method is shown experimentally. It eliminates the fixed value impulse noise. And it is tested for different images in different sizes. The peak signal-to-noise ratio (PSNR) [19,22,23,27,29,30] value gives the performance of restoration quantitatively, which is defined as

$$PSNR = 10 \log_{10} \left(\frac{\sum_k 255^2}{\sum_k (d(k) - y(k))^2} \right) \quad (8)$$

Where 255 is the peak pixel value of the image, $d(k)$ represents the value of the desired output, and $y(k)$ represents the value of the physical output.

The experimental results were compared with median-filter (MED).

4.1 Original Images



lena



bamboo



track



grass



road



forest



wires



s1



s2

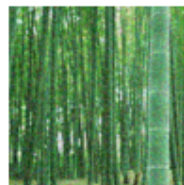


s3

4.2 Noisy Images



n_lena



n_bamboo



n_track



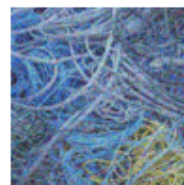
n_grass



n_road



n_forest



n_wires



n_s1



n_s2



n_s3

4.3 Median Filter Output



m_lena



m_bamboo



m_track



m_grass



m_road



m_forest



m_wires



m_s1



m_s2



m_s3

4.4 Proposed Filter Output



p_lena



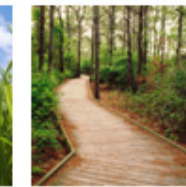
p_bamboo



p_track



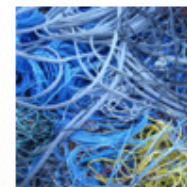
p_grass



p_road



p_forest



p_wires



p_s1



p_s2



p_s3

Table 5: PSNR Comparison Table

Image name	PSNR for Median Filter	PSNR Proposed filter
lena	27.583501	27.588477
bamboo	25.972569	25.980521
track	25.941161	25.949895
grass	24.008344	24.012961
road	25.584527	25.599688
forest	25.729469	25.743479
wires	23.657962	23.666070
s1	25.216643	25.2242
s2	24.469417	24.473686
s3	24.610376	24.618010

Table 6: Edge Count Comparison Table

Image name	Original Image	Median Filter Output	Proposed Filter Output
lena	22384	22267	22383
bamboo	32688	32563	32663
track	34686	34468	34576
grass	27381	27224	27244
road	39945	39668	39830
forest	32062	31650	31754
wires	41625	41337	41425
s1	33467	33522	33339
s2	29437	29370	29415
s3	26717	26577	26695

From the Table 5 we observed that the proposed method is better than the median filter due to better PSNR. From the Table 6 we conclude that more edges [9] be preserved for the proposed method than median filter.

5. CONCLUSION

In this paper, we propose a new filtering approach for noise removal. It based on soft computing techniques which can remove the noise and at the same time tiny edges can be preserved. This enhances the reliability for getting the output image, which is more or less same with original image. From the Table 5 we conclude that the best PSNR can be achieved by using the proposed method. The proposed method is tested more than 1000 pictures, which gives the better result for each test image.

REFERENCE

- [1] Abreu E., Lightstone M., Mitra S. K., Arakawa K., "A new efficient approach for the removal of impulse noise from highly corrupted images," IEEE Trans Image Processing, vol. 5, no. 6, pp. 1012-1025, 1996
- [2] Abreu E., Mitra S.K., "A signal-dependent rank ordered mean (SD-ROM) filter. A new approach for removal of impulse From highly corrupted images", Proceedings of IEEE ICASSP-95, Detroit, MI, 1995, pp. 2371-2374.
- [3] Arakawa K., "Median filters based on fuzzy rules and its application to image restoration", Fuzzy Sets and Systems 77 (1996) 3-13.
- [4] Baudes, A., B. Coll, Morel J, M, 2005. "A non-local algorithm for noise denoising", Computer Vision and Pattern Recognition, CVPR. IEEE Conf., 2: 60-65, 20-25
- [5] Bezdek J. C., Pattern Recognition with Fuzzy Objective Function Algorithms. New York: Plenum, 1981
- [6] Chen T., Wu H. R., "Adaptive impulse detection using center- weighted median filters," IEEE Signal Processing Letters, vol. 8, no. 1, pp. 1-3, 2001.
- [7] Chen T., Ma K. K ., Chen L.-H, "Tri-state median filter for image denoising", IEEE Trans. Image Processing, vol. 8, no. 12, pp. 1834 – 1838, 1999.
- [8] Chen. T., Wu H. R., "Space variant median filters for the restoration of impulse noise corrupted images," IEEE Trans. on Circuits and Systems II: Analog and Digital Signal processing, vol. 48, no. 8, pp. 784-789, 2001.
- [9] Gonzalez. R. C, Digital Image processing. 2nd Edition TMH
- [10] Hamza A., Krim H., "Image denoising A non linear robust statistical approach." IEEE. Trans. Signal Processing 49(2), pp. 3045-3054; 2001
- [11] Hwang H., Haddad R. A., "Adaptive Median filters: New algorithms and results". IEEE transaction on Image processing, 4, pp. 499-502 1995
- [12] Jayaraman S., Esakkirajan, S., Veerakumar T, "Digital Image processing", Tata McGraw Hill Education Private Limited.
- [13] Jeong B. , Lee Y. H., "Design of weighted order statistic filters using the perceptron algorithm," IEEE Trans. Signal Processing, vol. 42, no. 11, pp. 3264-3269, 1994.
- [14] Konar A, Computational Intelligence, Springer, New York, 2004

- [15] Lakshimiprabha S., A new method of image denoising based on fuzzy logic. *International Journal of Soft computing* 3(1): 74-77, 2008
- [16] Lee Y. H., Ko, . S.-J “Center weighted median filters and their applications to image enhancement,” *IEEE Trans. Circuits and Systems*, vol. 38, no. 9, pp. 984 – 993, 1991.
- [17] Lee, C. S. V H Kuo, P. T Yu. “Weighted Fuzzy mean filters for Image processing”. *Fuzzy Sets Sys.* 89, 157-180; 1997
- [18] Lin T.-C., Yu P.-T., Partition fuzzy median filter based on fuzzy rules for image restoration, *Fuzzy Sets and Systems* 147 (2004) 75 - 97.
- [19] Lin T.C., Yu P.-T, “ A new adaptive center weighted median filter for suppressing impulsive noise in images”, *Information Sciences* 177 (2007) 1073-1087.
- [20] Ross. T. J, *Fuzzy logic with Engineering Applications*, 2nd edition, University of New Mexico, USA, Wiley India.
- [21] Sadoghi H., Yazdi, Homayouni F., “Impulsive Noise Suppression of Images Using Adaptive Median Filter”, *International Journal of Signal Processing, Image Processing and Pattern Recognition* Vol. 3, No. 3, September, 2010
- [22] Sau K., Chanda A., “Image De-Noiseing with A Fuzzy Filtering Technique”, India, 2011
- [23] Sau K., Chanda A., Karmakar P, “Image De-noising with Edge Preservation Using Soft Computing Techniques” *Annual International Conference on Emerging Research Areas(AICERA)*. 2011 in Collaboration with IUPUI United States
- [24] Sonka, M., Hlavac V., Boyle R., *Image Processing, Analysis, and Machine Vision*, PWS Publishing, Pacific Grove, Calif, USA, 1999.
- [25] Sun T. , Neuvo Y, “Detail-preserving median based filters in image processing,” *Pattern Recognition Letters*, vol. 15, no. 4, pp. 341– 347, 1994.
- [26] Somasundram K. Shanmugavadivy P. Adaptive iterative order Statistics Filter, *ICGST-GVIP journal* Vol. 9 issue 4, pp 23-32, 2009
- [27] Smolka B., Chydzinski A., Fast detection and impulsive noise removal in color images, *Real-Time Imaging* 11 (2005) 389 - 402.
- [28] Umbaugh, S. E. *Computer Vision and Image Processing*, Prentice-Hall, Englewood Cliffs, NJ, USA, 1998.
- [29] Wangand Z., Zhang D., “Progressive switching median filter for the removal of impulse noise from highly corrupted images,”*IEEE Trans. on Circuits and Systems II: Analog and Digital Signal Processing*, vol. 46, no. 1, pp. 78–80, 1999.
- [30] Yli-Harja O., Astola J., and Neuvo Y., “Analysis of the properties of median and weighted median filters using threshold logic and stack filter representation,” *IEEE Trans Signal Processing*, vol. 39, no. 2, pp. 395– 410, 1991.
- [31] Zhang S., Karim M. A, “A new impulse detector for switching median filters,” *IEEE Signal Processing Letters*, vol. 9, no. 11, pp. 360–363, 2002
- [32] Zimmermann H. J “Fuzzy set theory- And its applications”, second edition, Allied publishers limited.

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