

# An Optimization of Novel Hybrid Bilateral Filter for Isotropic and Anisotropic of 2D-Images in All kinds of Transmissions

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**Abstract:** *In this paper, a new deblurring and denoising algorithm is proposed for isotropy and anisotropy reconstruction for 2D-images using Novel Hybrid Bilateral Filter (NHBF) in low level to high level noise values. To overcome the drawback of traditional denoise filter mechanism, we used NHBF which will reconstruction an image from low abstracted to high abstracted pixel values. Experiment analysis reveals that our algorithm is efficient than other traditional algorithms like Mean Filter (MF) and Bilateral Filter (BF). This proposed method can be utilized in Satellite Image Processing, Atmospheric research Laboratories and Scientific Image Processing.*

**Keywords:** Reconstruction, Bilateral Filter, Satellite Image Processing, Atmospheric research Laboratories and Scientific Image Processing.

## 1. Introduction

Generally images are of two types, one is Gray Scale Images and other one is RGB images. Gray scale image has only one channel and RGB image have three channels. Noise may occur due to heat of camera sensors, analogue to digital conversion and communication channels. This type of noises cause to degrade the image quality, therefore it is very important to remove the positive and negative impulses by using filtering technique. Most probably the Linear Filtering technique will consider the two filtering techniques like Mean Filter (MF) and Least Mean Square (LMS). Median Filter considers under non-linear filtering technique. Eliminating the noise process may contaminate the nature of an image in order to either improve its original visualization. Pixels corrupt at two intensity levels that is high or low intensity values. To identify the neighboring pixels Standard Median Filter (SMF) proposed in [1]. The SM filter was the first effective impulse noise removal filtering technique.

Speckle reducing anisotropic diffusion (SRAD) approach is derived based on partial differential equation (PDE) in [1] which will allows the generation of an image based on shape of the image and size of the window. The specific class of splitting-based finite, various theoretical and adequate methods, and additive operator splitting (AOS) approach are discussed in [1]. A two stage fuzzy filter approach for reduce heavy noise is presented in [2], firstly; computes a fuzzy derivative for eight direction. In second; fuzzy derivatives to smoothing by weighting the neighboring pixels using pixel values. Impulse noise detectors and removal method is propose for Grayscale and Color Images are presented in [3]. A improved decision-base algorithm (DBA) is proposed in [4] which efficiently removes the salt and pepper noise in gray scale and color image preserving in detail. The DBA utilized previously processed neighboring pixel values to get

better denoised in gray and color images. A new impulse noise removal approach for color image is presented in [5] work base on noise detection algorithm and weighted vector median filter. Additive white Gaussian noise (AWGN) process that is independent of the actual RGB values. Such approaches cannot effectively remove color noise produced by today's CCD digital camera. For this a new approach two phase unified framework for color noise which will work based on single piecewise smooth and noise level function (NLF) model is presented in [6]. Wavelet transforms are very powerful to remove noise suspicions in various wavelet based images. A wavelet based denoise method is presented in [7]. Edge detection is a process that detects the presence and location of edges constituted by sharp changes in color, intensity of an image. The image brightness is depending up on the depth of discontinuities of an image, surface orientation, different material properties and variations in scene. So it is a difficult task to remove the noise without eliminating the sharp characteristics of the image, such as edges, corners and other sharp structures during the denoising process. Edge detection is susceptible to noise. This is due to the fact that the edge detection algorithms are designed to respond to sharp changes, which are caused by noisy pixels. A new fuzzy filtering techniques for color image works for de-fuzzification is presented in [8] which preserves any type of edges in any direction. Due to Haze in images, resulting contrast reduction and fading of color. To remove haze in images a new two stage approach is proposed in [9]. In first, denoise the image prior to dehazing this treats haze and noise separately, and second approach is denoise using dehaze of an iterative approach using adaptive and non-parametric method.

Speckle noise in images is causing difficulties for image understanding. Reduction of speckle noise is one of the most essential tasks in image preprocessing. An innovative

approach to reduce speckle noise is presented in [10]. Image segmentation will play a major role to process segregating an image into different essential parts of object or regions. These regions share common characteristics based on color, intensity, texture, etc. segmentation includes edge detection, region growing methods, Clustering methods, Compression based methods, Histogram-based methods, etc. In image processing edge detection techniques are used to locate the borders of homogeneous region in an image depending on characteristics such as texture and intensity of image. To recover the original edges of an image an efficient approach called reconstructed color image segmentation using edge detection and threshold method is presented in [11]. A new method called YCoCg color Image Edge detection presented in [12]. Gray scale and color images impulse noise removal algorithm is presented in [13]. In the process of remove noise the image does not deteriorate. Reconstruction process is called the reconstruction of an image using decomposition or analysis of applied noised image. A new Image Decomposition Using Wavelet Transform is proposed in [14].

## 2. Proposed Algorithm

In this section we proposed Another filter computation, which will identify noise and after that displace noise corrupted pixel. Results demonstrate that the proposed channel produces fabulous results at various commotion levels.

**Step 1:** Read the input image and convert it into GRAY scale image.

**Step 2:** Select 2-D window of size 5x5 and assume the window size as W=3

**Step 3:** Calculating the histogram bins (P) and cumulative sum of histogram bins(Ω)

**Step 4:** Calculate mean (μx) by using the following equaiton  
 Mean (μ)= cumulative sum of histogram bins(Ω) \* number of bins in histogram of image.

**Step 5:** Calculate the sigma\*mena by using the following equation

$$\sigma \text{ or } \sigma_x \mu \text{ "mu" or } \mu_x = \frac{((\mu \cdot \Omega - \mu)^2)}{\Omega \cdot 1 - \Omega} \quad \text{E.q... (1)}$$

$$\sigma_x * \mu_x = \frac{((\mu \cdot \Omega - \mu)^2)}{\Omega \cdot 1 - \Omega} \quad \text{E.q ... (2)}$$

where index=mean (maximum value of (σx \* μx))

σ= (index-L)/(number of bins in histogram of image-L)

L is minimum to maximum pixel limits of input image

**Step 6:** Convert the image from RGB imge to LAB Image

**Step 7:** Apply the novel bilateral filter F=H\*G (minimum :: maximum limits)

Where H =exponential(-((dl)2+(da)2+(db)2)/(2\*(sigma)2);

Here l is representing Lightness & a and b are the color dimensions

G=Gaussian domain weight age values

**Step 8:** Apply the soft threshold for given image and convert it from LAB image to RGB image to get Novel Bilateral Filtered image.

**Step 9:** apply noval canny edge deteciton based on renyi entropy for 2D images to get better edge detection

$$Q=(1/(1-\alpha))*\log_2(\text{sum}(P,2)+\text{eps}) \quad \text{E.q ... (3)}$$

Where p=The value for thresh is relative to the highest value of the gradient magnitude of the image

Alpha = Shape Parameter

Eps = Floating Point Accuracy

**Step 10:** Round off these Q values we are getting renyi entropy values

**Step 11:** renyi entropy maximum values are taken as threshold for canny edge detection and stop the procedure

## 3. Simulation Results & Discussion

The simulation results are compared with traditional filters and proposed NHBF and shown in the Table 1, NHBF technique generating better results in high densities also. In table 2, generally we have shown the result based on PSNR values with general filter techniques. Figure1 shown the input image, figure 2 shows the MF simulation result, figure 3 shows the WF simulation result, figure 4 shows the proposed NHBF simulation result and figure 5 shows the performance of different filtering technique plot.

**Table 1:** The Mean, Wiener and NHBF comparison in terms of MSE, RMSE and PSNR

Noise Level %	Filter	MSE	RMSE	PSNR
10	Mean	1.3906	0.75	22.6503
	Wiener	0.938	0.3062	34.3627
	NHBF	0.0351	0.185	<b>62.6768</b>
20	Mean	2.668	1.0698	19.8206
	Wiener	0.938	0.3062	34.3627
	NHBF	0.1066	0.2603	<b>57.8515</b>
30	Mean	8.375	1.6417	14.8526
	Wiener	0.938	0.3062	34.3627
	NHBF	0.2645	0.3173	<b>53.9066</b>
40	Mean	6.3711	1.6238	16.0403
	Wiener	0.938	0.3062	34.3627
	NHBF	0.5943	0.3699	<b>50.3908</b>
50	Mean	19.4102	2.3494	11.2021
	Wiener	0.938	0.3062	34.3627
	NHBF	0.9391	0.4107	<b>48.4037</b>
60	Mean	14.1875	2.183	12.5633
	Wiener	0.938	0.3062	34.3627
	NHBF	1.3109	0.4454	<b>46.9552</b>
70	Mean	31.6992	2.6693	9.0719
	Wiener	0.938	0.3062	34.3627
	NHBF	1.9035	0.4836	<b>45.3354</b>
80	Mean	34.0234	2.5981	8.7646
	Wiener	0.938	0.3062	34.3627
	NHBF	2.56	0.5155	<b>44.0483</b>
90	Mean	41.2031	2.7049	7.9331
	Wiener	0.938	0.3062	34.3627
	NHBF	3.2803	0.5466	<b>42.9717</b>

**Table 2:** Result & Comparison of mean, Wiener and proposed NHBF Filtering technique

Noise Level %	Mean Filter - PSNR	Wiener Filter - PSNR	Proposed - NHBF PSNR	Noise Level %
10	22.6503	34.3627	62.6768	10
20	19.8206	34.3627	57.8515	20
30	14.8526	34.3627	53.9066	30
40	16.0403	34.3627	50.3908	40
50	11.2021	34.3627	48.4037	50
60	12.5633	34.3627	46.9552	60
70	9.0719	34.3627	45.3354	70
80	8.7646	34.3627	44.0483	80
90	7.9331	34.3627	42.9717	90

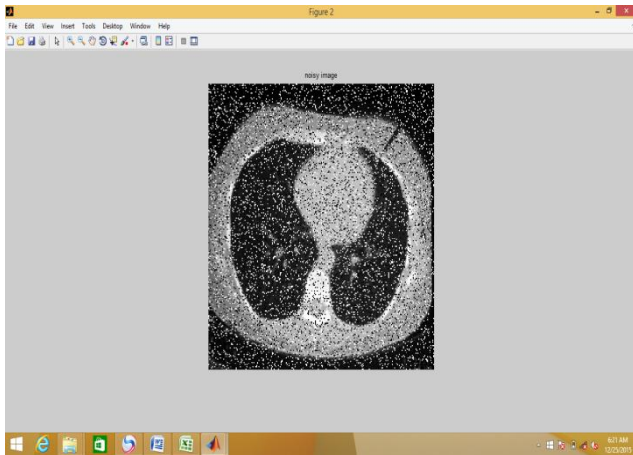


Figure 1: Noise Image (it differs at each noise level)

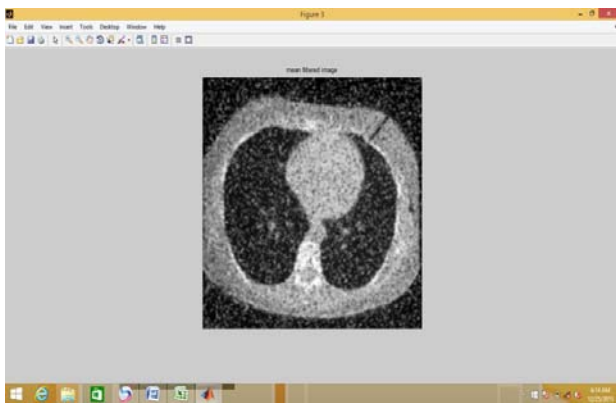


Figure 2: MF simulation result

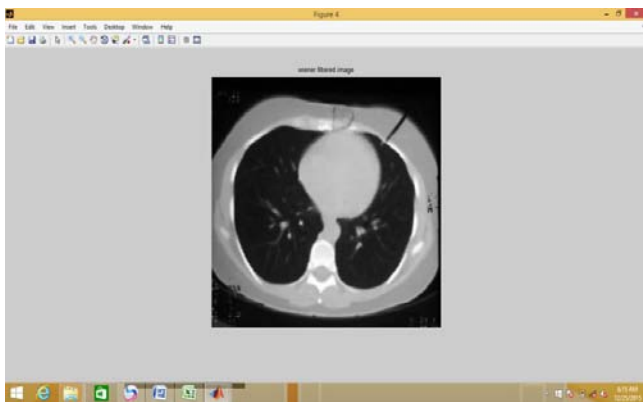


Figure 3: WF simulation result

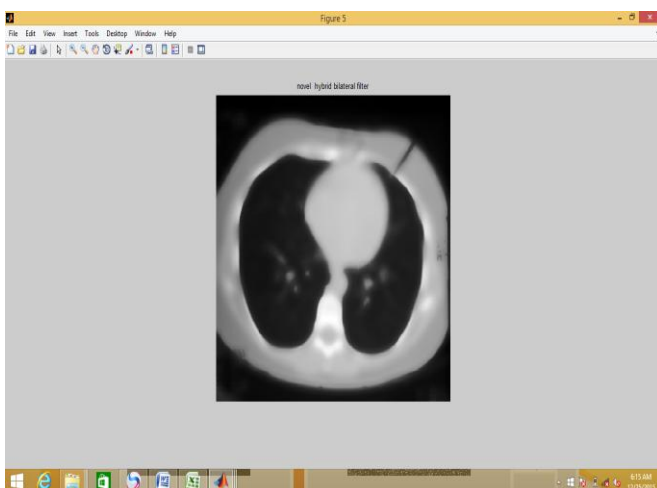


Figure 4: proposed NHBF simulation result

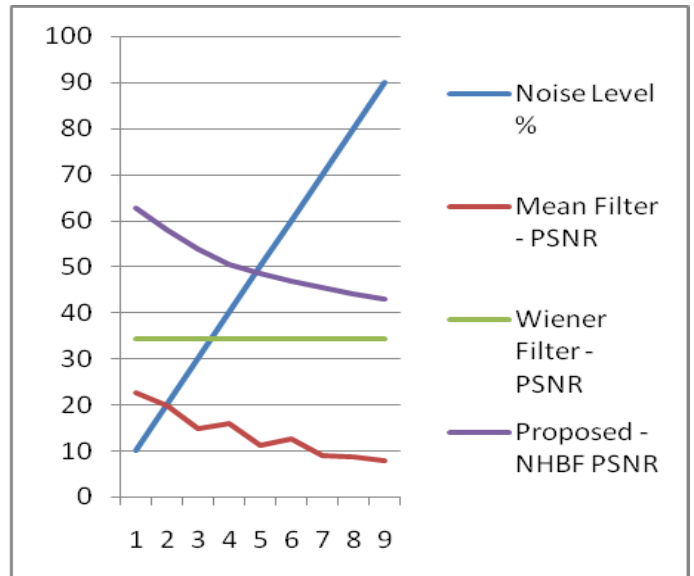


Figure 5: plot for differentiation of MF, WF and NHBF

#### 4. Conclusion

The proposed NHBF, fabricate that an successful lossless images in all noise values for isotropic and anisotropic of 2D-Images. This filtering mechanism will work in different hierarchy of pixels in all dimensions and the results are compared with the traditional filtering techniques like MF and BF with the characteristics of PSNR, MSE and RMSE it proves that NHBF work more robotic than traditional filters at all noise values.

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