



# GGIF and Bilateral Filtering based Deep Learning Technique for Image Dehazing

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**Abstract:** Fog occurs as a result of natural environmental conditions that reduce image quality. Perceptual fog density (PFD) is used to calculate the fog density parameter. In this paper, a globally guided filtering technique using a bilateral filter and a deep neural network is used to create the new image approach. This new method is implemented on the MATLAB programming environment, the proposed methods are obtained through PFD computation to show the results of the current and proposed technology with optimized images. The proposed technique preserves the fine texture, image color density and eliminates haze compared to the existing techniques.

**Keywords:** GGIF, Deep Learning, Dehazing, Bilateral filtering

## I. INTRODUCTION

Haze mainly occurs when some pollutants like sulphur dioxide, nitrogen dioxide, ozone, carbon monoxide etc accumulate in dry air. It also contains fine particles such as smoke and dust. The bad weather situation such as fog, smoke, rain, smog etc are often destroy clarity of outdoor images scenes. The light absorbed and scattered before reaching the camera because of the atmospheric suspended particles such as aerosols, water droplets, fog etc. Due to this the images captured are resulted in poor visibility, blurred, decreased contrast and faded surface colors images that are in bad weather. This effect may be degraded and also can ruin the value of under-water and aerial photography for industrial and creative photographers. It may also be used for satellite imagery, including geological and cloud-based mapping, land planning and architectural design. The result is a cumulative loss of picture contrast and an additive term because of this uniform light. The debased picture is deciphered in this model as a sum of two segments: the commitment of air light and the unsure brilliance of the ground. The transmission rate, a scalar that sets the perceivability at every pixel, consolidates the two three-channel shading vectors mathematically. The three surface shading esteems and the exchange an incentive for every pixel must be resolved so as to recuperate a without haze picture. [1] Poor visibility in bad weather, including mist or fog, is a major challenge for many computer vision applications. Highly necessary for the quality of the graphical algorithm is the elimination of haze.

[2] Haze negatively degrades the value of a picture in outdoor scenes, thereby impacting its esthetic attractiveness and exposure. Due to its unplaced nature, single image dehazing is particularly challenging. [3] Haze reduce perceptibility and lower the identifying feature of visible objects. Corruption for every pixel is distinctive and relies on the scene point distance from the lens.



Figure 1: Dehazed images and comparisons. [4]

## II. PREVIOUS WORK

[1] Zheqi Lin et al., A new quick dehazing approach for video and image processing in real time. An improved guidance sorting scheme calculates the transmitting chart seamlessly and values the underlying picture profound data. The results demonstrate that both the dehazing and the real-time output of the system are great. The algorithm can be used for beneficial preprocessing in many applications, including security, smart vehicles and remote sensing because of its pace and visual enhancement performance. The results show that both dehazing and the system's real-time efficiency were good. Thanks to their pace and visual enhancement, the device algorithm can be used for useful preprocessing on many applications including security, smart vehicles and remotely sensing.

[2] He Zhang et al., Many current studies, including the latest convolutionary neural network (DNN) derived approaches, depend on the traditional computational paradigm in which the hazy picture is represented as the superposition of attenuated scene radiance and ambient illumination.

[3] Dana Berman et al., The transfer coefficients that regulate the scene constriction and calculate the cloudiness of each pixel reflects this dependency. Past approaches struggle with the single image dehazing issue by using different set priors. Then again, an earlier, non-nearby estimate relies on another. This calculation was based on the assumption that a nebulous object's colors in RGB space are very close to a few hundred recognizable colors. Our key assumption is that pixels are usually not positioned in a given group, i. e. they are scattered across the whole object plane and are placed at different distances from the camera.

[4] Mihai Budescu et al., Every matrix component containing color details for one pixel can be depicted by an object as a vector. The matrix is used in the neural network as input data. The tiny object measurements, which make learning simple and fast, specify the vector size and number of input vectors. A Sigmoid characteristic is the transport function used. The analysis frequency contains values between  $[0,1]$  and the deviation under  $0,1$ . A wide variety of operational fields such as geotechnical engineering, civil engineering, robotics, industrial control, safety, automation and transportations have been successfully implemented in photo recognition (ANN).

[5] Jiahao Pang et al., The dehazing of a single image is complicated because it is very ugly. The elimination of Haze based on a dark channel prior is successful but it is computationally expensive to refine the transmission map. Recent work has shown that it is possible to use a directed filter to simplify the map. Through merging the dark channel with the controlled object filtering in depth, we are producing single image dehazing. An active method of adjusting criteria, by evaluating the trade offs of the solution. Our system produces satisfactorily dehazed tests at low approximation, as shown in experiments and correlations.

[6] Luuk Spreeuwiers et al., In this study, neural networks were introduced as object filters and the evaluation of these image filters. Neural networks are a relatively new approach for the storage of data and information. In this study, neural networks were introduced as object filters and the evaluation of these image filters. Neural networks are a relatively new approach for the storage of data and information. Many methods for the representation of objects and patterns using neural networks exist, but most of them concentrate on target detection, character identification, associative memories that remember images and retina simulation. The study discussed in this paper involves the use of neural networks for object processing, i.e. strategies to recover objects, optimize images and refine and remove features. Again all of these processes generate an object that is better suited for human and artificial vision systems analysis.

[7] Nicholas Carlevaris-Bianco et al., The mechanism moves through the illumination as the energy is transferred from the object to the eyewitness. The influence of dispersal can drastically reduce the quality of pictures in media of massive suspended particles such as mist or turbid liquid. A measurement is provided in underwater pictures of removal of light dispersing impacts referred to as dehazing. Our key commitment is to a specific, yet amazing, earlier approach that uses the solid distinction between the three colored water channels to assess the depth of the scene. Using this marker to reduce the fluctuating spatial effect of the fog on the image at that level. We use a single image to operate with our technique without any special hardware or earlier scene training. A current detail chart of the scene is given as a part of the dehazing process. Current findings on different true objects, as well as on a controlled test array, in which objective distinction and actual reality are identified.

[8] F. Farnood Ahmadi et al., The findings (road networks) were organized in a completely computerized (CAD) model vector which could be used with limited editing in a Geographical Information System (GIS). Image analysis and the use of CAD-based equipment are anticipated as the technique.

[9] Yan Hu et al., Active DNN-based approach to recover the underwater image consists of the T-network (T-network) propagation projections and the global environmental light estimates (A-network). Through learning about the connection between subsea and its corresponding black channel transmission map and global ambient light, the underwater photos are retrieved and improved by an optical model of the underwater.

[10] Shweta K. Narnaware et al., Pictures and photographs are ways for viewing and processing data. When an object is transformed, e.g. digitalization, imaging, distribution, processing, etc., To some point, the production image becomes diminished. Therefore, it is necessary at undergo a process called an image enhancement for an output image. Literature reveals that there are still some limitations to many solutions to enhanced picture up to now, including Histogram Pairing, Spatial Averaging, Median Filter, Un Clear Masking & High Boost Filtering etc.

[11] Birendra Kumar Patel et al., The use of data compression in digital pictures is image compression. A computer model which is induced by an observable mechanism in neural networks of organic neurons in the brain is a major role of the image compression system used to reduce the number of pixels and to effectively reduce display and transmitting costs and an Deep Learning Network. Image compression based on the leven berg-marquardt neural back propagation network and performed through the image is divided up into several blocs and the single neural network is selected based on its difficulty for each node.

[12] Jun Cheng et al., Approximate retinal objects may decay as a result of attenuation and dispersal of human lens. A modern directed retinal filters (SGRIF) structural protection assures a restore of pictures based on the designs of attenuation and dispersion. Thanks to fogs, dung and mist, until entering the screen, spread in the atmosphere. Disposal or airlight is the product of the creation of fog. The reliability of the retinal picture is often decreased, one of the problems in the study. Of example, the retinal picture is reduced through a cataract in a human eye like a blurry camera lens which decreases imaging performance. The specifics of retinal images are often blurred and difficulties raised in the collection and interpretation of retinal objects.

[13] Hongteng Xu et al., The photo gradient can then be called a representation of the fractal array if the pixels of a real picture are viewed as a fractal set. Based on the invariance (a special case of a bi-lippschitz invariance), the gradient of a high-resolution picture can be determined from the gradient of a low resolution photo. Moreover, by maintaining the local fractal length of the gradient during the up-sampling process you can further increase the high-resolution image. A regularization concept centered on the fractal dimension and length scale in variances may be useful in extracting high-resolution image data. The interaction and discrepancy between the technique and some other approaches for state-of - the-art interpolation was analyzed. Experimental results show that the approach is preferable to other methods, with better resolution and development performance.

[14] A new globally guided image filtering is introduced in this paper. The proposed filter can be applied to produce sharper images and preserves details in regions of fine structure visibly better than the existing locally guided image filtering. It is applied to study single image haze removal. Experimental results demonstrate that the proposed haze removal algorithm indeed improves visual quality of dehazed images.

[15] Two methods of poisson noise removal from images were introduced in this paper, the bilateral filter method and the FDCT method. The intrinsic features of the image are better preserved using FCDT method which has a higher SNR. But the execution time of bilateral filter is lesser than FDCT method. These two methods can be switched in applications according to the prevailing situations.

### III. IMPLEMENTATION

In this research, we proposed an improved image dehazing method using deep learning and globally guided image filtering, and this proposed model improves the effectiveness and robustness of images. The two phases are combined with improved work. In the initial stage GGIF the algorithms are implemented and in the next stage DNN the methods are implemented.

#### A. Globally guided Filter GGIF

The suggested filter is used to produce sharper and well-preserved images. [14] The G-GIF is composed of a global structure transfer filter and a global edge-preserving smoothing filter. The function of the structure transfer filter is to transfer the predefined structure to the image to be filtered while the function of the smoothing filter is to smooth the transferred image so as to produce the output image.

#### B. Deep Learning Networks

In this work, firstly we create white balance image and then load the dehazing parameter. Parameters pass through NN convolution and npool layers. Multimapping local extremism calculations used for feature extraction. The NN loss layer and softmax layer is applied then it will be proceeding with relu layer and offset layer for neural networks. The final output image is kept for exploring the perceptual fog density (PDF).

#### C. Bilateral Filter

[15] The Bilateral filter was proposed by Tomasi and Manduchi which has the peculiarity of preserving the edges along with Gaussian noise removal. The filter follows non-linear behavior. The bilateral filter is a weighted averaging filter used for removing Gaussian noise. The bilateral filter (BF) can be used for removing noise. It is an effective method for noise removal as universal noise like impulse noise, salt and pepper noise and Gaussian noise. Bilateral filter is much faster and simple to implement.

#### D. Flowchart of proposed algorithm

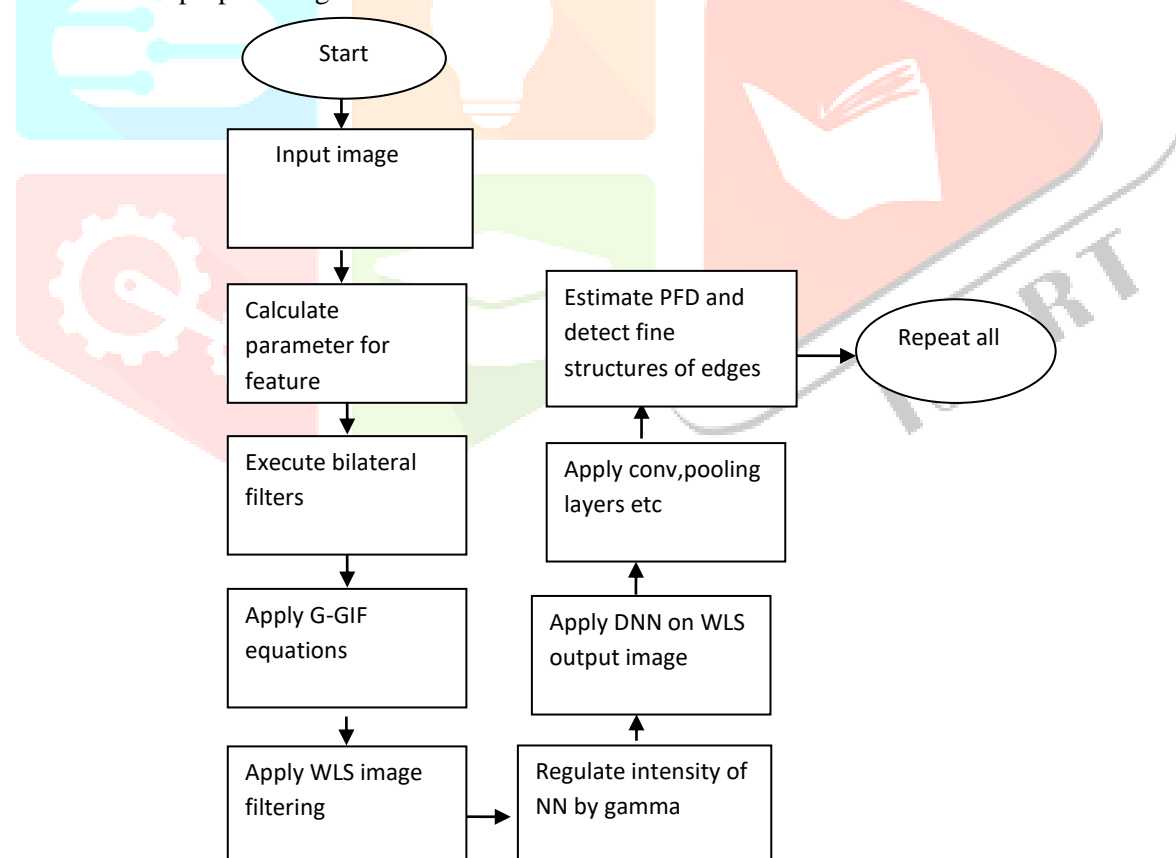


Fig.2: Flowchart of Proposed Algorithm

### IV.RESULT

In this section, the results of all images are seen on the basis of PFD and its edge preserving and fine structure of the colors are observed.

In fig. 3, the GUI output of first image is shown below,

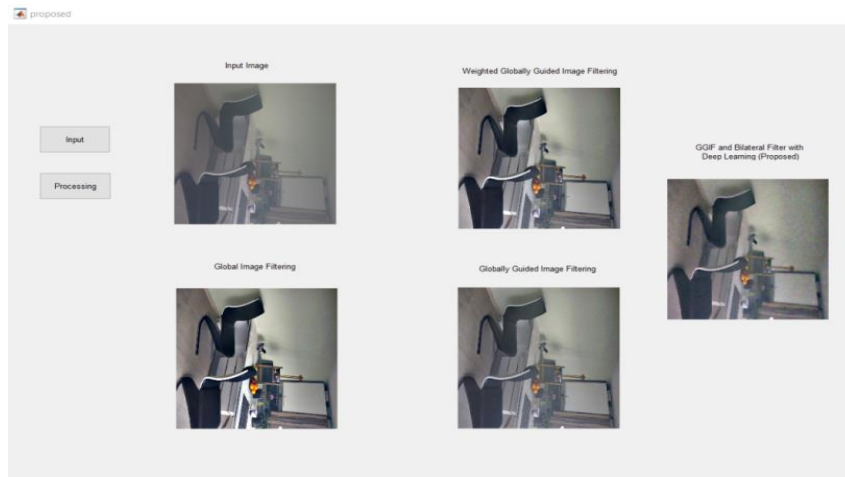


Fig. 3: output of the Input Image 1

In fig. 4, PFD Output of the Images and the minimum PFD is found in proposed method to make the images more clear.

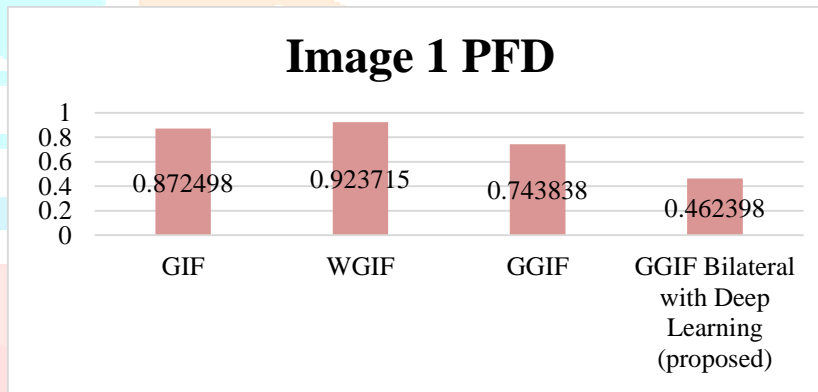


Fig. 4: PFD Output of Input Image 1

In fig. 5, the GUI output of second input image is shown below-

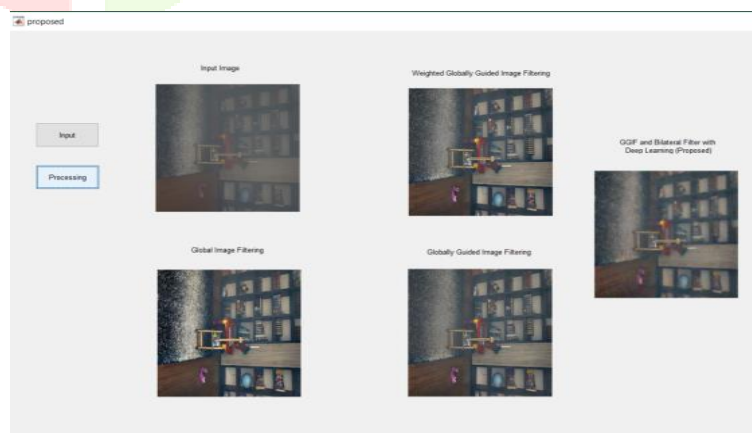


Fig. 5: output of Input Image 2

In fig. 6, PFD Output of the Images and the minimum PFD is found in proposed method to make the images more clear.



Fig. 6: PFD output of Input Image 2

## V. CONCLUSION

A globally guided image filter (GGIF) and DNN techniques are introduced in this paper to analysis the image features enhancement and haze elimination of the images. The results of the approach show that the algorithms can produce images with better optical quality comparable to the other image filters. Deep learning neural network based globally guided image filtering technique using bilateral filter are applied to betterment of the image characteristics such as haze elimination, fine details preservation and realistic color matching. The main parameter which is improved in this work is perceptual fog density (PFD). Perceptual fog density defines the result of the research work. MATLAB environment is used for making this software. MATLAB is capable to run any number of pictures with dissimilar size. It also preserved authenticity and color edifice of the images.

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