A Novel Method for Fog Removal in Images based on Turbulence Mitigation and Contrast Enhancement

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ABSTRACT—In a real time world photo graphic and video surveillance were playing a vital role. Humans are using this photo for video surveillance either for security or for entertainment purpose. The general problems were occurred while imaging the atmosphere is most appearance of fog and the appearance of atmospheric turbulence in the images. There were many researchers have provided a recovery for either the fog removal or turbulence mitigation from the image. Eventually, they failed to recover both the problem at the same time. This paper provided an analysis that incorporates both model such as Fog removal and Turbulence mitigation. This paper suggest CETM (Contrast Enhancement Turbulence Mitigation) method whichis more efficient so that it can operate in a fractions of second for the real-time applications while imaging.

KEYWORDS—Single Image Defogging ,Image alignment, Dehazing, Debluring Turbulace Mitigatin,Contrast enhancement

I. INTRODUCTION

An image definedin the 'real world' is considered to be a function of two real variables such as $a(x,y)$ where a is the amplitude (e.g. brightness) of the image at the *real* coordinate position $(x,y$)In a highly advanced image processing scheme it should be likely to apply specific image processing operations to selected regions. To suppress motion blurone fragment of an image (region) might be administered while another part might be processed to improve color rendition [1].

Vision is the most advanced of our sense, so it's not surprising that Figure play the single most major role in human insight. However, unlike humans, who are limited to the visual band of the electromagnetic (EM) spectrum ,imaging machine cover almost the entire EM spectrum, ranging from gamma to radio waves .They can operate on generated by source that humans are not accustomed to associating [2].These include ultrasound electron microscopy and computer generated. Thus digital image processing encompasses a wide and varied field of application.

A common problem in the atmosphere is fog and atmospheric turbulence. Over the years, many researchers have suggested perception into the physics of either the turbulence [3] or fog turbulence but not both. Most recently, researchers have proposed methods to remove fog in fast enough for real-time processing. Additionally, methods have been proposed by other researchers that address the atmospheric turbulence problem. This work, provide an analysis that incorporates both physics model. i) Fog and ii) Turbulence

To observe the contrast enhancement (fog removal) can affect image alignment and image averaging. This paper proposed, a new joint contrast enhancement and turbulence mitigation (CETM) method that utilizes estimation from the contrast enhancement algorithm to improve the turbulence removal algorithm [4]. This paper contribute a new turbulence mitigation object metric that measure temporal consistency. Finally, designed the CETM to be efficient such that it can operate in fraction of a second for near real-time applications. Imaging system positioned near the ocean often suffer in performance, perceptually and objectively, because of atmospheric turbulence, fog ,sun-glare, camera movement from wind battering and several other adverse weather condition. The environment itself will have fog or mist, airstream and heat that reasons eddy currents which is perceived as turbulence in an imaging system [5].

II. EXISTING METHOD

Already in existing methods there are several approaches are available to remove fog substance and some more methods are available to remove turbulence from the image. But it can remove either turbulence or fog, but not able to remove both at a time. Atmospheric turbulence caused by variation of refractive index along the optical transmission path can toughly distress the performance of larger-distance imaging systems [6]. It produces geometric distortion, motion blur (if not sufficiently short the exposure time), and occasionally out-of-focus blur when the turbulence is violent [7].

III. OBJECTIVE

The general problems were occurred while imaging in the atmosphere is the appearance of fog and also the appearance of atmospheric noise and turbulent in the images. This article work is carried out for both models such as fog removal and Turbulence mitigation and also to verify that contrast enhancement and turbulence mitigation algorithm is more efficient than all other existing algorithm. Fog removal can be operate in fractions of second for the real-time applications.

IV. PROPOSED METHOD

A. Turbulence Mitigation and Contrast Enhancement

Atmospheric turbulence can severely degrade the quality of Figure produced by long range observation systems, rendering the Figure unsuitable for vision applications such as surveillance or scene assessment. The major visual special effects affected by atmospheric turbulence are geometric distortion and space-time-varying blur. The distortion is primarily generated by optical turbulence and scattering and absorption by particulates aerosols, for example, diffuse light will also cause blur. Several approaches have been used to restore Figure, including adaptive optics techniques and pure image processing-based methods. Due to random fluctuations in turbulence, calculating a sensible approximation of atmospheric modulation transfer function (MTF) is extremely difficult. However, this function is of dangerous significance for optics-based restoration approaches. Therefore, this article, focused only on image processing to manage the degradation instigated by turbulence. Assuming that the scene and the image sensor are both static, the mathematical model has been used to interpret imaging process.

Contrast enhancements improve the perceptibility of objects in the background by improving the brightness variance between objects and their backgrounds. Contrast enhancements are classically executed as a contrast stretch trailed by a tonal enhancement, although these could both be performed in single step. A contrast stretch improves the brightness differences uniformly across the dynamic range of the image, although tonal enhancements advance the intensity differences in the shadow (dark), midtone (gray), or highlight (bright) regions at the expense of the brightness differences in the other regions.

B. Recommended Block Diagram With Description

The function of the image denoise block is the process of removing noise from the image. Symmetric filtering method is used here to remove the unwanted noise in the input image. The Defogging block is used to remove fog effectively and efficiently for each frame. Fast Fourier transform defogging method is used here for the defogging process. Turbulence in the specified image has been mitigated by means of the turbulence mitigation metric block. This turbulence problem can be initiated due to bad weather or disturbed atmospheric situations. In this process, different Figure is matched with one another by means of Image Alignment block. The method of matching one image called template with another image is called Image alignment. There are numerous applications for image alignment like tracking objects on video, motion analysis, and many other tasks of computer vision. To remove the blur caused by the atmosphere and image alignment errors, Image Deblur block is preferred. De-convolution method is used for deblurring here.
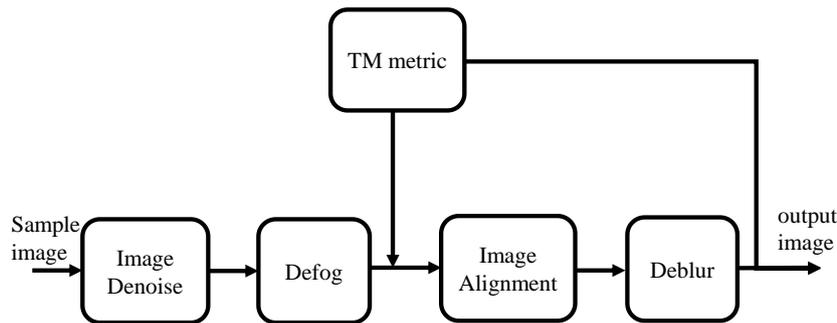
Fig 1. Proposed Block Diagram

V. METHDOLOGY

A. Removal of Noise From Turbulance

It is shown that contrast enhancement does not improve the image alignment performance when image noise is present. The result in terms of error is the same if enhancement is before or after global tracking [8]. The analysis does show, however, reducing image noise can improve the image alignment performance. Additionally, it is observed that removing noise before fog removal is an important approach in order to improve transmission and air light estimation, therefore this work proposed how to remove noise, enhance the contrast and then estimate the global motion.

In order to diminish the complication of our procedure such that the processing speed is near real-time, a two dimensional median filter is preferred for each color channel for fast single image denoising.

a.   Median Filter

A median filter belongs to the class of nonlinear filters distinct the mean filter [9-10]. The median filter also tracks the moving window method analogous to the mean filter. It is necessary to scan a 3×3, 5×5, or 7×7 grain of pixels over pixel matrix of the entire image. The computed median value should be placed in the center pixel value in the window. Median filtering can be done by sorting out all the pixel values from the surrounding neighborhood into numerical order and then substituting the pixel being deliberated with the middle pixel value. Note that the median value must be inscribed to a distinct array or buffer so that the outcomes are not corrupted as the process is performed. The median is supplementary robust likened to the mean. Thus, a single very unrepresentative pixel in a neighborhood will not affect the median value meaningfully. Subsequently the median value must really be the value of one of the pixels in the neighborhood, the median filter do not generate new impractical pixel values when the filter straddles an edge. For this reason the median filter is considerable superior at conserving piercing edges than the mean filter. These advantages aid median filters in denoising uniform noise as well from an image.
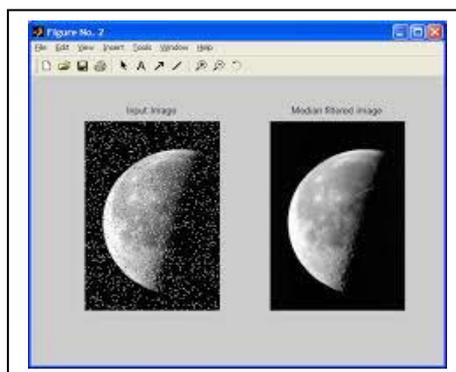

Fig 2. Input and Output of Median Filter

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As mentioned earlier, the image "moon.tif" is corrupted with salt and pepper noise with the imnoise () function after loading the image using imread ().Fig 2 is the image corrupted with salt and pepper noise and is given to the function medfilt2 () for median filtering. The window specified is of size 3×3.Fig 2 is the output after median filtering. It can be observed that the edges are preserved and the quality of denoising is much better compared to the input image

b.   Defogging of Image

A defogger, demister, or defroster is a system to clear condensation and thaw frost from the windshield, back glass, and/or side windows of a motor vehicle. Let's see the benefits and applications of image defogging [11]. To remove fog effectively and efficiently for each frame and have an estimate of transmission and used the Locally Adaptive Wiener Defogging method [12]. The Wiener Defogging method is 50 to 100 times faster than existing methods and can operate at real time speeds for frames of sizes 720×480 and larger. This paper proposed a modification to the Wiener Defog method in order to automate the defogging process. The window size must be sufficiently large to estimate a smooth transmission map that best reflects the scene depth and also suggest a method for automatically selecting $|\Omega|$ based on the statistics of the frames. Given a foggy color image $\mathbf{i} \in R\psi$, denote the

Number of color channels with ψ.

$$I (x) = t (x) o (x) + (1 − t (x)) a + w(x). \tag{1}$$

The main idea in estimating the transmission with an adaptive local Wiener filter in is that the observation model is composed of the veiling $v = (1 − t)$ and texture n_{txt}. The observation is the dark pixel measurement for each pixel is given by,

$$D(X) = \min i_c (X) \tag{2}$$

Where $Ic(x)$ is the nth color channel of the image $i(x)$. (For grey Figure, $d(x) = i (x) \in R$). The dark pixel measure takes advantage of the atmospheric dichromatic model by assuming

at least one color component is possibly dark therefore exposing the atmospheric veiling. Thus the model at each pixel is

$$D(x) = v(x) + n_{txt} \tag{3}$$

Where v is t min i_c (X) true veiling and n_{txt} is the textured noise. The goal in the Wiener Defog method is to filter out the texture from the observation d by treating the texture n_{txt} as noise but preserve edges from depth discontinuities. For Figure with a large amount of texture (forest floor with leaves), the size of window must be sufficiently large to filter out the texture variations. The choice of arises from Ω the local moment estimators,

$$\hat{\mu}_{.v} (X) = \frac{1}{|\Omega(X)|} \sum_{je\Omega(x)} d(X) \tag{4}$$

Where $|\Omega(X)|$ is the number of pixels within the local sample window positional at pixel location x.

a.   Image Aligment

This work employ the symmetric filter, Fast Fourier Transform (FFT) [13-14] to estimate the global motion parameter g. This filter is fast and efficient and works well with turbulence video because the phase correlation update averages out the zero mean phase perturbations induced by turbulence [15].

The local motion induced by turbulence must also be compensated in order to have a sharper image after frame averaging and also enhanced before optical flow [16] in order to improve the motion estimation. Enhancing before motion compensation reduces the number of possible motion vectors for each pixel. For video coding, intra-frame coding is used more when enhancing first therefore details are preserved whereas inter-frame coding is used more when there is low contrast which results in loss of details.

b.   *Image Deblurring*

Where deblurring is used to remove the blurriness from the image. Image deblurring can be formulated as the process of inverting image blurring. This section first introduces a model of image degradation, and then presents problem definitions of image deblurring with their basic solution strategies and associated difficulties

c.   Turbulance Mitigation Metric

In order to illustrate that the turbulence has been mitigated in time with our proposed method, article a Turbulence Mitigation Metric (TMM) [14]. The goal in turbulence mitigation is to not only recover the true structure (phase) of the scene but also to enforce temporal consistency in each color channel. Instead of using a subjective measure by displaying multiple frames of a video sequence, this work developed an objective measure. Our goal in designing the TMM is to make it an objective measure of turbulence mitigation, simple to construct, and easily accessible such that it can be used in any other turbulence mitigation method. One approach to developing a TMM is to utilize the motion estimated from optical flow and global estimations.

Fig 3. Example for Weiner deconvolution method a) Blurred image b) Deblurred image

This approach, however, is complex given that not every method uses optical flow and it required an input image sequence and output image sequence.

VI. PROPOSED ALGORITHM

The following algorithm is to be implemented as per the requirement of fog removal

Step 1: START

Step2: Read the input image

Step 3: Compute air light value from the image with fog .if He method followed, following steps are taken

   Step 3a: Read the R, G, B components

   Step 3b: Divide each by 255

   Step 3c: Find min & max value of three values

   Step 3d: Find difference between min & max values

   Step 3e: Assign V component as the max value

   Step 3f: Assign S component as the difference divided by the V component value.

   Step 3g: Compute H component according to weather max value is R, G, and B

   Step 3h: Apply symmetric filter on the HSV image.

   Step 3i: Find all pixels greater than the largest value.

   Step 3j: Take maximum of these values as air light values.

Step 4: if air light computation is by OUR method the following steps are taken.

   Step 4a: Apply symmetric filter on the image

   Step 4b: Return the highest value of the image as air light value.

Step 5: If no method is specified take the first pixel of the image as air light value

Step 6: Derive the boundary condition.

Step 7: using the air light value. Calculate the optimal t value using following steps

International Journal of Engineering Technology Science and Research
IJETSR
www.ijetsr.com
ISSN 2394 – 3386
Volume 4, Issue 8
August 2017

Step 7a: parameter values are set accordingly (including scaling parameter)

Step 7b: Iterate through to obtain optimal transmission value.

Step 8: Divide the R, G and B component respectively to obtain the R, G, B components of input image by the optimal t value.   .

Step 9: add the air light components respectively to obtain the R, G, B components of the dehazed image.

Step 10: Display the input image without fog

Step 11: Stop.

VII. HARDWARE AND SOFTWARE

Our experiments was executed in a 64-bit machine with Intel(R) Core(TM) i7-4500U CPU @ 3.0GHz with 8GB of RAM (critical for large values) and NVIDIA 4 GB Graphics cards. Software were used Matlab and Open CV for fast processing.

VIII. RESULTS AND DISCUSSION

A. Experiment with Real Time Image

To demonstrate the proposed method along with the TMM, Contrast Enhancement Turbulence Mitigation (CETM) algorithm method is applied to a simulated foggy and turbulent image sequence. The PSNR and TMM values are measured with different K values and plot the results in the following figure. The minimum frames to average Kmin = 44 was estimated. The TMM measure starts low and on average increases after each new image. The TMM is 0 at frame K = 10 because the sample size in time was set to 10. The highest TMM is with K = 90 and lowest with K = 10. What is interesting is that there is a significant improvement from 10 to 44 frames but not much improvement when the number of frames to average is almost doubled from 44 to 90. An identical survey can be made with the PSNR values in the figure. The PSNR is increased dramatically at the beginning and the performance reaches a limit at each K value. The performance of the method at each stage of the algorithm is

Illustrated in Fig 4, Fig 5, Fig 6, Fig 7, Fig 8 and Fig 9. The same simulated sequence is used and each measurement is with K = 44. The PSNR value of the contrast enhanced sequence is very low because noise is introduced after enhancement. The optical flow slightly improves the result and a dramatic improvement occurs with frame averaging. Without optical flow the frame averaging performs about 1 dB lower than with optical flow.

B. Simulation Results

The following figure Depict the simulated output of both RGB and for Gray scale image.

   a. Simulation Results for removal of fog from image

(a)   (b)

Fig 4.Example 1 of Proposed Defogging Algorithm for Color Image (a) Foggy Image (b) Defogging Image

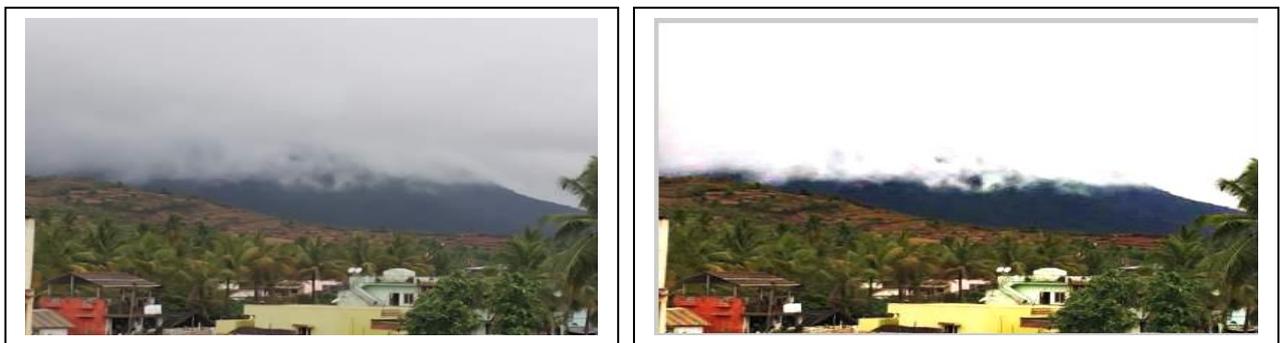(a)   (b)

**Fig 5.Example 2 of Proposed Defogging Algorithm for Color Image (a) Foggy Image
(b) Defogging Image**

b.   Simulation Results for turbulence image

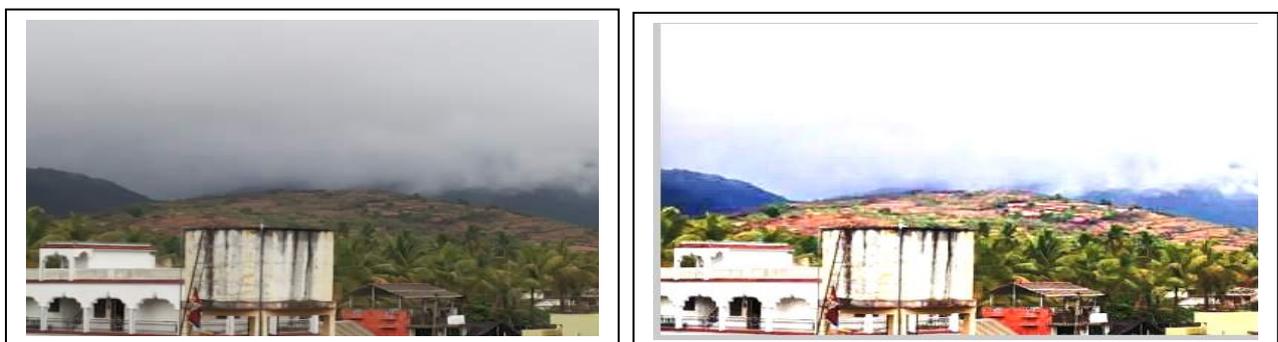
(a)   (b)

**Fig 6. Example 1 of Turbulence Removal Algorithm for Color Image (a) Turbulent Image
(b) Turbulent Mitigated Image**

(a)   (b)

**Fig 7. Example 2 of Turbulence Removal Algorithm for Color Image (a) Turbulent image
(b) Turbulent Mitigated Image**

International Journal of Engineering Technology Science and Research
IJETSR
www.ijetsr.com
ISSN 2394 – 3386
Volume 4, Issue 8
August 2017

c. Simulation results for removal of both fog and turbulence from image

(a)   (b)

Fig 8. Example 1 of Fog and Turbulence Removal from Color Image (a) Image with Fog and Turbulence
(b) Defogged and Turbulent Mitigated Image

(a)   (b)

Fig 9. Example 2 of Fog and Turbulence Removal from Color Image (a) Image with Fog and Turbulence
(b) Defogged and Turbulent Mitigated Image

VIII. CONCLUSION AND FUTURE WORK

Many researchers have provided a method to remove either the fog or turbulence from the images but not both at the same time. Many researchers have proposed methods to remove fog in images fast enough for real-time processing. Additionally, some more methods have been proposed by other researchers that address the atmospheric turbulence problem. This work analyzed fog and turbulence of image by observing how contrast enhancements (fog removal) can affect image alignment and image averaging. A new joint contrast enhancement and turbulence mitigation (CETM) method that utilizes estimations from the contrast enhancement algorithm to improve the turbulence removal algorithm.

The main goal of my work was to develop a joint turbulence mitigation and fog removal method that can recover the object image fast enough for near real-time performance.

In future work of this project is proposed to apply fog removal algorithm in traffic monitoring cameras for real time surveillance of vehicles and also as a driver assistant when travelling in early morning hours at late evening during the foggy days. This is extremely useful in monitoring systems during early morning hours or late evening as conventional cameras would give shady images/videos (image/videos with high noise).

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