

## Multi-Stage Enhancement Approach for Image Dehazing

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### ABSTRACT

*Over the past decades, huge efforts have been devoted for image enhancement under uncontrolled scene such as fog and haze. This work proposes Multi-stage de-hazing approach for improving the quality of hazy images. Four main stages are introduced, in our approach, to achieve an automated, efficient and robust de-hazy processing. The first two stages are utilized to diminish the blurring noise and enhance the contrast using Wiener filter and contrast scattering in the RGB color, respectively. In contrast, the last two stages are utilized for luminance and quality enhancement using luminance spreading and color correction. It is obvious from the experimental that the proposed approach significantly improves the prominence of the hazy images and outperforms the performance of conventional methods, such as multi-scale fusion and histogram equalization. In addition, it is also found that our approach exhibits low complexity compared to existing works.*

## 1. Introduction

Nowadays, hazy weather appears more and more frequently in different areas of the world as a result of pollution. Under hazy weather conditions, suspended particles in atmosphere for example fog, murk, the mist, dust causes poor visibility image and distorts the colors of the scene. Images taken in such conditions represent a major issue for numerous outdoor applications of vision community including surveillance, object recognition, target tracking, which require high-quality images. Figure 1 shows an example of hazy and de-hazy images taken from [1].

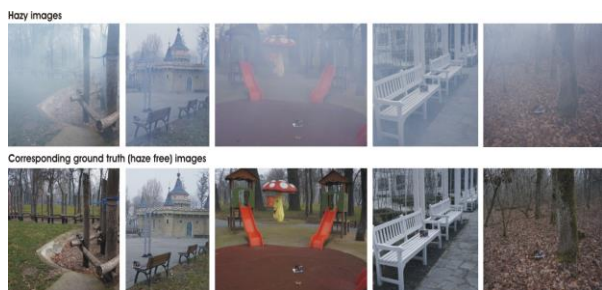


Figure 1: Sample Hazy and De-hazy Images

Images under hazy weather exhibit very poor quality and blurred even in a small dynamic range of color intensities [2]. The main

factors causing hazy images might be represented as the mixture of scene vivacity, air-light and communication [3,4]. The study of haze is related to the works on scattering of light in the atmosphere (see Figure 2). In imaging system, the radiance captured by the camera along the sightline is declined due to atmospheric light and it is replaced by scattered light that leading to lose contrast and colors in captured images.

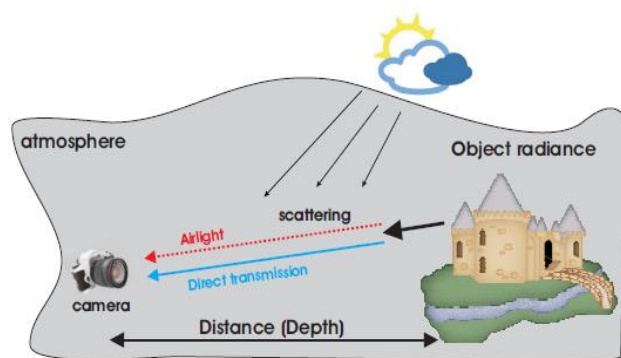


Figure 2: Optical model against diverse weather situations [5].

In general, the following equation has been widely employed to improve the quality of the haze images:

$$IM(x) = F(v)t(v) + L(1 - t(v)) \quad (1)$$

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where  $F$ ,  $L$  and  $I$  represent haze image, free scene in the haze image and the atmospheric light, respectively. In contrast,  $v$  is denoted as the location of the image component while  $t$  is denoted as the intermediate of broadcast to describe the percentage of the bright, which is not dispersed by the camera. The haze reduction aims to recuperate  $F$ ,  $L$ , and  $t$  from  $IM$ .

## 2. Literature Review

In the recent years, different types of techniques have been proposed for de-hazing images, which further are classified into two main categories with respect to the amount of images utilized for de-hazing process [6,7]. The first category relies on multiple images taken from a hazy scene with different densities at the same point [8, 9, 10]. This category provided impressive results by assuming the scene depth can be estimated from multiple images. However, this requires specific inputs (fixed scene under different weather condition) making this type of method impractical for real-time applications.

To overcome the need of changed weather conditions [11], polarization based approach has been introduced to simulate changed weather conditions. This simulation is obtained by applying various filters on different images. However, this approach still is not appropriate for real-time applications as the static scenes cannot be considered unless the polarization filter based approaches are utilized.

The second category relies on a single image to achieve haze removal [12–13]. This type of methods is capable effectively to improve both the contrast and the visibility of hazy images. However, it may result in vanishing some detailed information and increasing the computational burden.

He et al. [14] developed a Dark Channel Prior (DCP) method based on the measurements about the haze-free images. The proposed method utilized dark pixels to estimate haze transmission map. These dark pixels are categorized through low intensity worth of as a minimum one-color parameter. Meng et al. [15] proposed an enhancement to dark channel prior through estimating transmission map. The prior map was estimated via enhancing intrinsic edge limit with biased L1 standard relative regularization. In contrast, Zhu et al [16] developed a color attenuation prior (CAP) that defines a undeviating model on local priors which aims to recover depth information. Some other works tried to optimize the dark channel by combining it with other methods. For example, Xie et al (2010) introduced a model that utilizing dark channel and Multi-Scale Retinex [17]. This technique implemented on the luminance element in YCbCr space. Wiener filtering also has been implemented with dark channel prior to have well enhancement [18]

Over the past decades, Retinex theory [19] has been considered a millstone in many image enhancement and de-hazy methods. This theory is basically based on lightness and color constancy and it has several advantages including dynamic range compression, color independence and color and lightness rendition. Therefore, Retinex theory has been extensively employed in image processing

tasks. For example, the center/surround Retinex method has gained a huge interest as it offers low computational cost and no calibration for scenes.

Based on the Retinex theory, this work proposes a novel Multi-stage de-hazing method for improving the superiority of hazy images. In the proposed approach, the first two stages are utilized to diminish the blurring noise and enhance the contrast, while the last two stages are utilized for luminance and quality enhancement. More details about the proposed approach is given in the next part.

## 3. Proposed Approach

This approach involves four stages to enhance the quality of hazy images. These stages as follows

- i. Wiener filter.
- ii. Contrast scattering
- iii. Luminance spreading
- iv. Color correcting.

The workflow of the proposed approach is described in Figure 3, in which every stage is explained in the next parts.

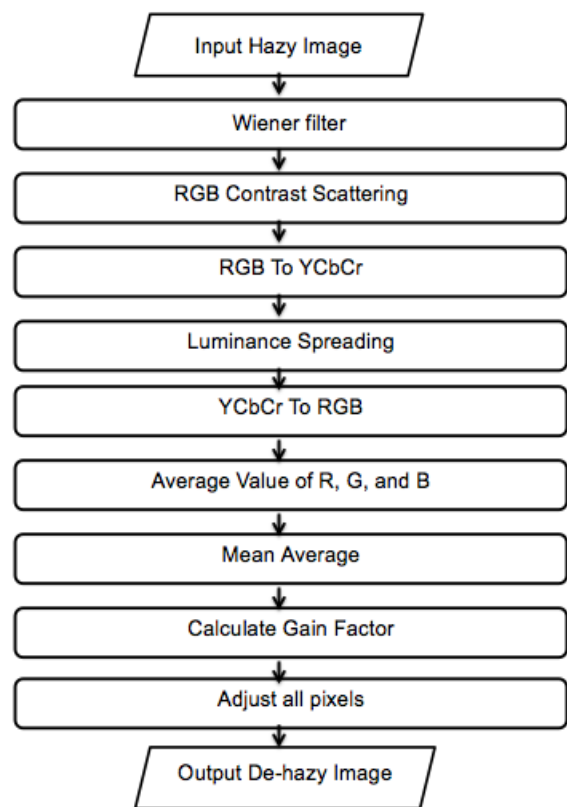


Figure 3: Workflow of the proposed approach.

### 3.1 Wiener filter

The exciting camera devices still have some problems specifically with images taken form haze environment. Haze environment greatly affects quality of image captured by the camera. As mention above a noise occurs in the captured image. To overcome this problem, the well-known method have been used to reduce blur in images due to linear motion is the Wiener filter. This filter

was proposed by Norbert Wiener during the 1940s and was published in 1949 [20]. It reduces the noise impact in frequency domain. The main goal of this filter is to obtain noisy image  $\hat{f}$  of the original image  $f$  and reduce mean square error  $e^2$  between the two images [21]. The next formula is used to measure the error where E represents the expected value.

Following three conditions are assumed:

- The images have zero mean, i.e., the images will be noise free.
- Noise and the image are uncorrelated.
- Grey levels in the restored image are linear functions of the gray levels in the noisy image.

Using the above conditions, the best value of the error function in frequency domain is given by next formulas

$$\begin{aligned} \hat{F}(u,v) &= \left[ \frac{H^*(u,v)S_f(u,v)}{S_f(u,v)|H(u,v)|^2 + S_\eta(u,v)} \right] G(u,v) \\ &= \left[ \frac{H^*(u,v)}{|H(u,v)|^2 + S_\eta(u,v)/S_f(u,v)} \right] G(u,v) \\ &= \left[ \frac{1}{|H(u,v)|^2 + S_\eta(u,v)/S_f(u,v)} \right] G(u,v) \end{aligned} \quad (2)$$

where

$H(u,v)$  = Degradation function

$H^*(u,v)$  = complex conjugate of  $H(u,v)$   $|H(u,v)|^2 = H^*(u,v)H(u,v)$ ,  $S_h(u,v)$  = Power spectrum of noise,  $S_f(u,v)$  = Power spectrum of the un-degraded image.

In haze environment the original image is not typically available. Thus, the power spectrum ratio is substituted by a parameter K that is experimentally determined. Practically, the value of K is chosen to be 0.00025 in order for achieving the best visual result. The following formula is used to decrease noise from hazy images.

$$F^\wedge(u,v) = \left[ \frac{1}{|H(u,v)|^2 + K} \right] G(u,v) \quad (3)$$

### 3.2 Contrast Scattering

In the second stage, the contrast scattering technique is used so as to improve the hazy images contrast. This technique selects smallest and highest values of the each RGB color model. Then, all the color parameter values are scattered among the smallest and highest values, as shown in Figure 4. Next formula is used for contrast scattering:

$$O_o = (O_i - sml) \frac{(\max - \min)}{(\text{lag} - sml)} + \min \quad (4)$$

where  $O_o$  is is denoted as the novel pixel value and  $O_i$  is denoted as the current pixel value. min , max, sml, and lag represent he minimum, maximum, least, and highest pixel value in each color parameter respectively.

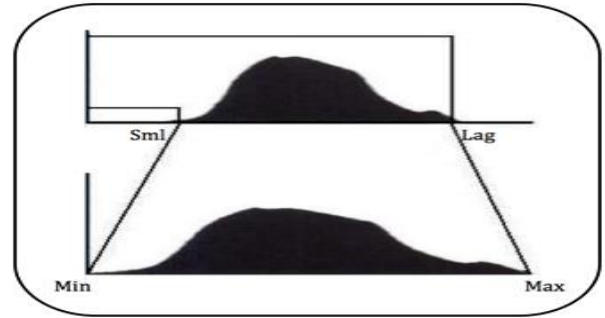


Figure 4: Contrast Scattering

### 3.3 Luminance Spreading

Subsequently, in order to achieve a true Luminance; an improvement method is applied. The Luminance in YCbCr plays a major role in image quality. The YCbCr color model is used to separate chrominance and luminance, in which, Y denotes the luminance parameter; while, Cb and Cr are respectively the chrominance parameters. On the other hand, RGB color model mutually make luminance. the following formulas are used to convert RGB color model to YCbCr color model:

$$Y = 0.587 * R + 0.114 * B + 0.299 * R \quad (5)$$

$$Cb = -0.299 * R - 0.587 * G + 0.886 * B = B - Y \quad (6)$$

$$Cr = 0.701 * R - 0.587 * G - 0.114 * B = R - Y \quad (7)$$

Consequently, the luminance parameter is adjusted to reach true colors and illumination. The proposed approach spans the luminance toward both directions 0 and 255 as shown in figure 5. The next formula is used for luminance spreading [22].

$$O_o = \frac{(O_i - \min) \times 255}{(\max - \min)} \quad (8)$$

where  $O_o$  and  $O_i$  are novel and current luminance value correspondingly. min denotes the minimum value, while max denotes the maximum value of luminance. Subsequently, following formulas are used to transfer image from the YCbCr to the RGB color model.

$$R = Cr + Y \quad (9)$$

$$G = Y - 0.194 * Cb - 0.509 * Cr \quad (10)$$

$$B = Cb + Y \quad (11)$$

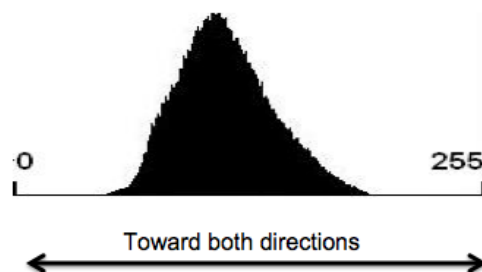


Figure 5: Luminance Spreading

### 3.4 Color Correcting

Color correcting method plays a major role to produce great quality images, which need balance number for pixels values. In this approach to get true colors, following formulas are used to calculate the average value [23]:

$$R_{avg} = \frac{1}{MN} \sum_{x=1}^M \sum_{y=1}^N I_{Red}(x, y) \quad (12)$$

$$G_{avg} = \frac{1}{MN} \sum_{x=1}^M \sum_{y=1}^N I_{Green}(x, y) \quad (13)$$

$$B_{avg} = \frac{1}{MN} \sum_{x=1}^M \sum_{y=1}^N I_{Blue}(x, y) \quad (14)$$

The proposed approach utilizes the mean average value of all colors. The mean average value is employed to increase the other values due to which the entire color is balanced. Mean value is used to overcome the over-saturation matter and keeps both the illumination and color at same levels [24]. This value is calculated for de-haze image using next formulas:

$$Avg = \frac{Red_{avg} + Green_{avg} + Blue_{avg}}{3} \quad (15)$$

$$Gain = \frac{Avg}{Blue_{avg}} \quad (16)$$

Then a diagonal model [25] is used to adjust all pixels values the image. Following formulas are utilized to recover the superiority of the hazy images.

$$R = Gain \times Red \quad (17)$$

$$G = Gain \times Green \quad (18)$$

$$B = Gain \times Blue \quad (19)$$

where Red, Green, and Blue are the values of the pixels in the hazy images respectively; while,  $R$ ,  $G$ , and  $B$  denote the balanced values of the pixels.

### 4. Experimental Evaluation

The proposed methods were assessed based on detached techniques such as PSNR (peak signal to noise ratio), SSIM (structural similarity index), and RMSE (root mean square error). RMSE is well known method used for result evaluation. RMSE is an accumulative aligned fault among de-hazy image and a free haze image by using next formula:

$$RMSE = \sqrt{\frac{1}{IJ} \sum_{x=1}^I \sum_{y=1}^J [F(x, y) - H(x, y)]^2} \quad (20)$$

when RMSE is very small and close to zero that means a high quality de-haze image.

The second approach is PSNR. In order to get an excellent quality of de-haze image, the value of PSNR should be high. However, to find MSE among the new and de-haze images respectively.

Furthermore, we need to find the MSE between free haze and de-haze images in order to calculate PSNR. Next formulas were used to find MSE and PSNR

$$MSE = \frac{1}{IJ} \sum_{x=1}^I \sum_{y=1}^J [F(x, y) - H(x, y)]^2 \quad (21)$$

$$PSNR = 10 * \log_{10} \left( \frac{255^2}{MSE} \right) \quad (22)$$

The third approach is the SSIM (structural similarity index). SSIM measures the visual impact of brightness, dissimilarity and construction. The value of SSIM is among [0, 1] when free haze and de-haze images are almost identical; the SSIM is close to 1.

$$SSIM(z, v) = [L(z, v)]^\alpha \cdot [T(z, v)]^\beta \cdot [s(z, v)]^\gamma \quad (23)$$

where  $\alpha$ ,  $\beta$  and  $\gamma$  are factors to adjust the relative importance of the components[26]. Next formulas were used to compute SSIM

$$L(z, v) = \frac{2\mu_z\mu_v + T_1}{\mu_z^2 + \mu_v^2 + T_1} \quad (24)$$

$$T(z, v) = \frac{2\sigma_z\sigma_v + T_2}{\sigma_z^2 + \sigma_v^2 + T_2} \quad (25)$$

$$s(z, v) = \frac{\sigma_{zv} + T_3}{\sigma_z\sigma_v + T_3} \quad (26)$$

$$SSIM(z, v) = \frac{(2\mu_z\mu_v + T_1)(2\sigma_{zv} + T_2)}{(\mu_z^2 + \mu_v^2 + T_1)(\sigma_z^2 + \sigma_v^2 + T_2)} \quad (27)$$

where  $\mu_z$ ,  $\mu_v$ ,  $\sigma_z$ ,  $\sigma_v$ , and  $\sigma_{zv}$  are the mean, standard deviation, and covariance for images  $z$ ,  $v$ .  $T_1 = (0.01 * L)^2$ , where  $L$  is 255.

$$T_2 = (0.03 * L)^2 \quad (28)$$

$$T_3 = T_2/2 \quad (29)$$

### 5. Experimental Setup

In this work, we employed two kinds of open-source datasets namely I-Haze [2], and generalized dataset for validating the proposed approach performance. All the datasets are explained below.

#### 5.1 Used Datasets

In this study, we utilized two datasets to show the efficacy of the proposed approach. These datasets are defined as below.

Generalized Dataset: Hazy images were downloaded from different free websites such as personal web pages. The collected images served as inputs to the program and process in order to produce the enhanced images. The generalized dataset consisted of total of 50 images with different haze effect.

I-Haze Dataset: C. O. Ancuti et al have proposed I-Haze dataset, which includes 35 of hazy and conforming interior images of haze-free. Haze images have been produced under the same illumination conditions includes a MacBeth color checker.

#### 5.2 Setup

To validate the effectiveness of defined method, the following set of experiments were executed using Matlab.

- The first experiment was conducted on generalized dataset to show the efficacy of the defined method on real hazy images against the state-of-the-art methods namely multi-scale fusion[13], Dark channel prior [14], Gray world, White patch and Histogram equalization.
- The second experiment was conducted on I-Haze dataset in order to show the performance of the proposed approach against the state-of-the-art methods. Here, RMSE, PSNR, and SSIM have been calculated to compare proposed results with state-of-the-art methods.
- Finally in the third experiment, a inclusive set of the comparisons was performed using two datasets. In the first experiment, the defined technique is matched against the well-known methods like multi-scale fusion [13], dark channel prior [14], gray world, white patch and histogram equalization. While, in the second set of experiment, the comparison has been performed against the latest methods such as RMSE, PSNR, and SSIM.

## 6. Results and Discussion

### 6.1 A First Set of Experiments

As described before, this experiment validates the efficacy of the defined approach on real dataset. A few sample images are shown in Figure 6. As observed from figure 6., images produced using the proposed approach have better clarity and no blur noise matched against latest methods.

### 6.2 Second Set of Experiments

As described before, the second experiment was conducted to show the importance of the defined method using I-Haze dataset. For this purpose, this study has used the most known methods for image enhancement and restoration. The unbiased valuations are described in Table 1. The proposed approach was able to provide smaller RMSE values and higher PSNR values. In addition, the values of SSIM are close to 1 compared to the result produced by latest methods. It is clear that the defined approach has provided better results s shown in Figure 6.

### 6.3 Third Set of Experiments

In the third set of experiments, a strong comparison has been performed under the two datasets. The comparison results are presented in Figures 7 – 11. It is obvious that the defined method showed significant performance than of the existing methods. The proposed approach can acquire the resulting image without color distortion. Nevertheless, the histogram equalization creates unwanted color and some artefacts in images. The. The results produced by Multi-scale fusion and dark channel prior were acceptable, but new colours were created in some images, which changed their appearances. Therefore, the proposed approach accurately and robustly enhance different kind of images degraded by haze.

## 7. Conclusions

De-hazing approach have become a need for many computer applications. This work has proposed a Multi-stage approach for image de-hazing. Four main stages have been utilized in our

approach. The first stage is defined to diminish the blurring noise using Wiener filter. The second stage is defined to enhance the luminance, where each RGB color pixel was spread among the least and highest values. IN the next stage, the luminance adjustments is achieved by transforming the resultant images from the RGB model to YCbCr model. Finally, the last stage is utilized to overcome colorcast by color correcting. The proposed approach has showed high superior performance under challenging conditions including distorting images, incomplete variety, and little contrast. The approach also keeps the properties and characteristics of the images through de-hazing the images. In addition, unlike other approaches, our approach exhibit low complexity makes applicable for online scenarios

## Conflict of Interest

The authors declare no conflict of interest.

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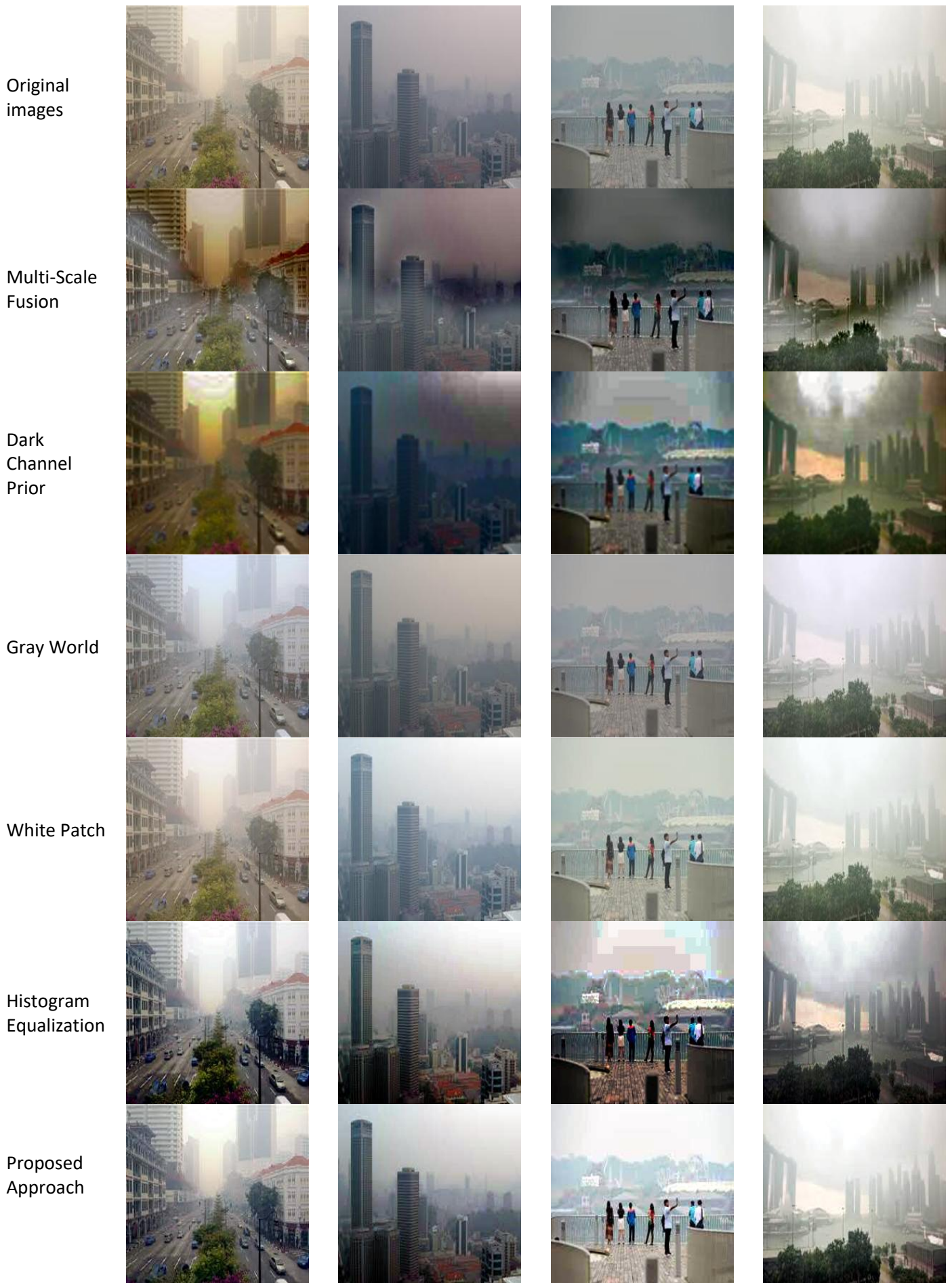
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Table 1: The average value of RMSE, PSNR, and SSIM of 30 images

30 Images Average value of	Proposed Approach	Multi-scale fusion	Dark channel prior	Gray World	White Patch	Histogram Equalization
<b>RMSE</b>	<b>21.8</b>	<b>29.6</b>	<b>34.4</b>	<b>26.7</b>	<b>27.9</b>	<b>22.5</b>
<b>PSNR</b>	<b>21.9</b>	<b>18.9</b>	<b>17.7</b>	<b>20.1</b>	<b>20.7</b>	<b>21.5</b>
<b>SSIM</b>	<b>0.75</b>	<b>0.56</b>	<b>0.61</b>	<b>0.68</b>	<b>0.66</b>	<b>0.73</b>



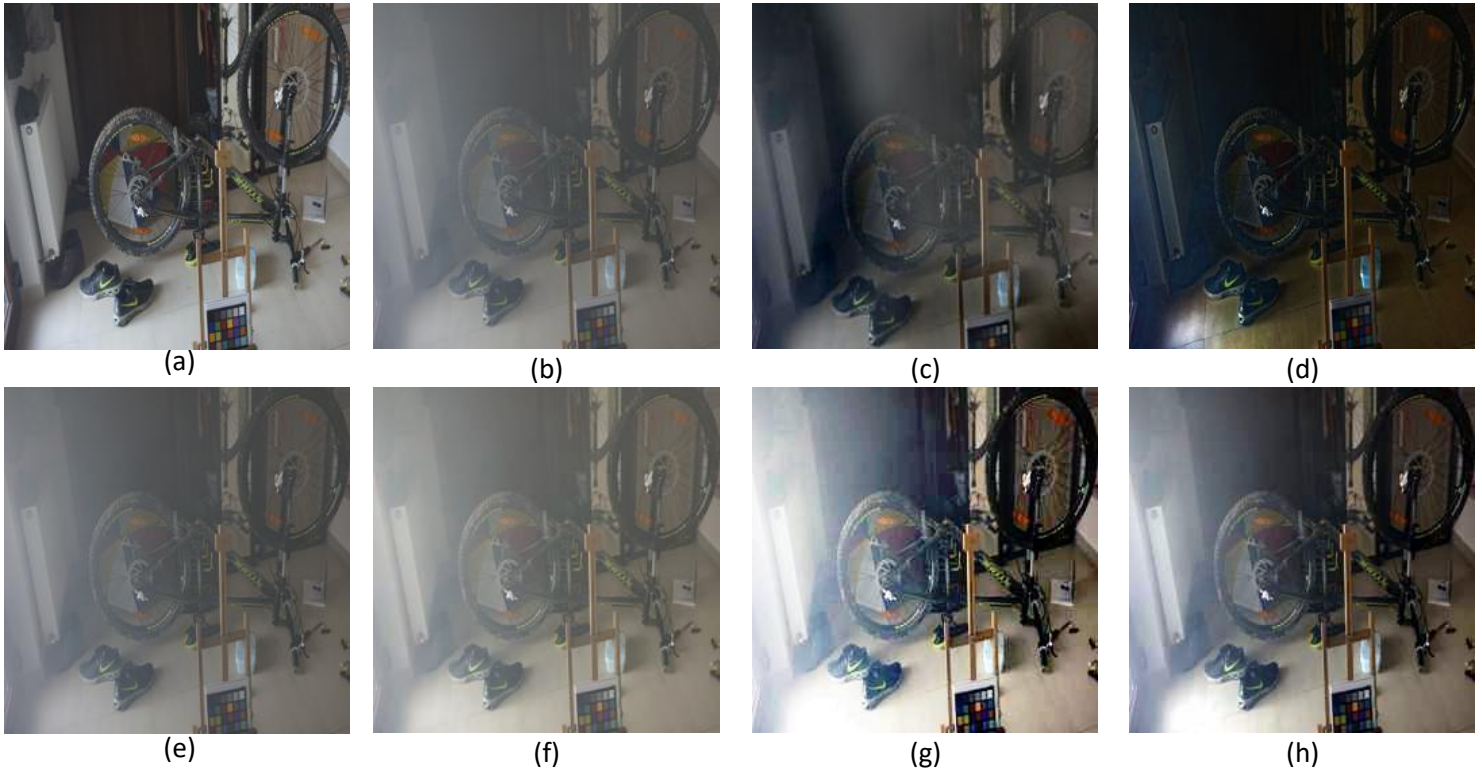


Figure 7. Comparative results for image number 5 from I-Haze dataset. (a) input image, (b) hazy image, (c) multi-scale fusion, (d) Dark channel prior, (e) Gray World, (f) White Patch, (g) Histogram Equalization, and (h) proposed approach

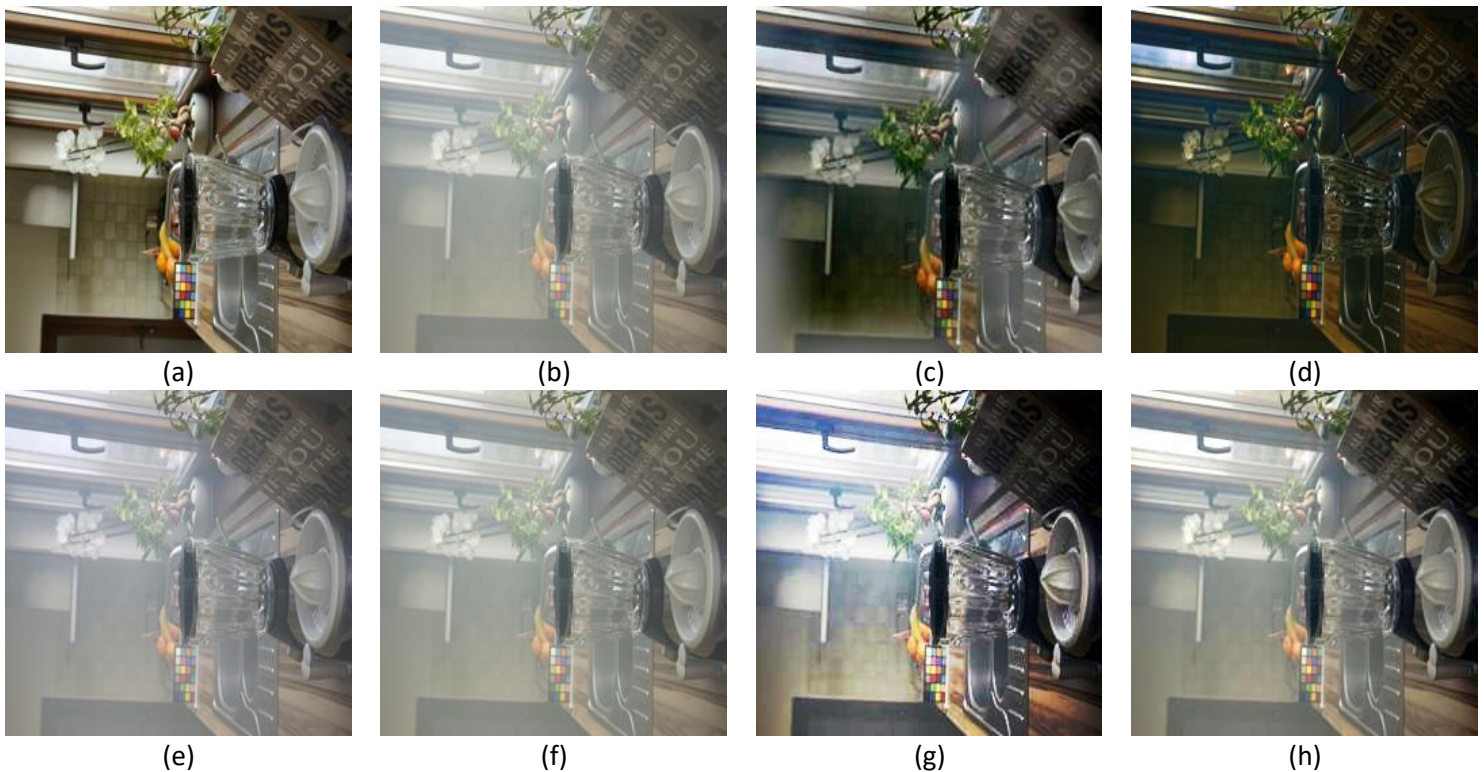


Figure 8. Comparative results for image number 9 from I-Haze dataset. (a) input image , (b) hazy image, (c) multi-scale fusion, (d) Dark channel prior, (e) Gray World, (f) White Patch, (g) Histogram Equalization, and (h) proposed approach



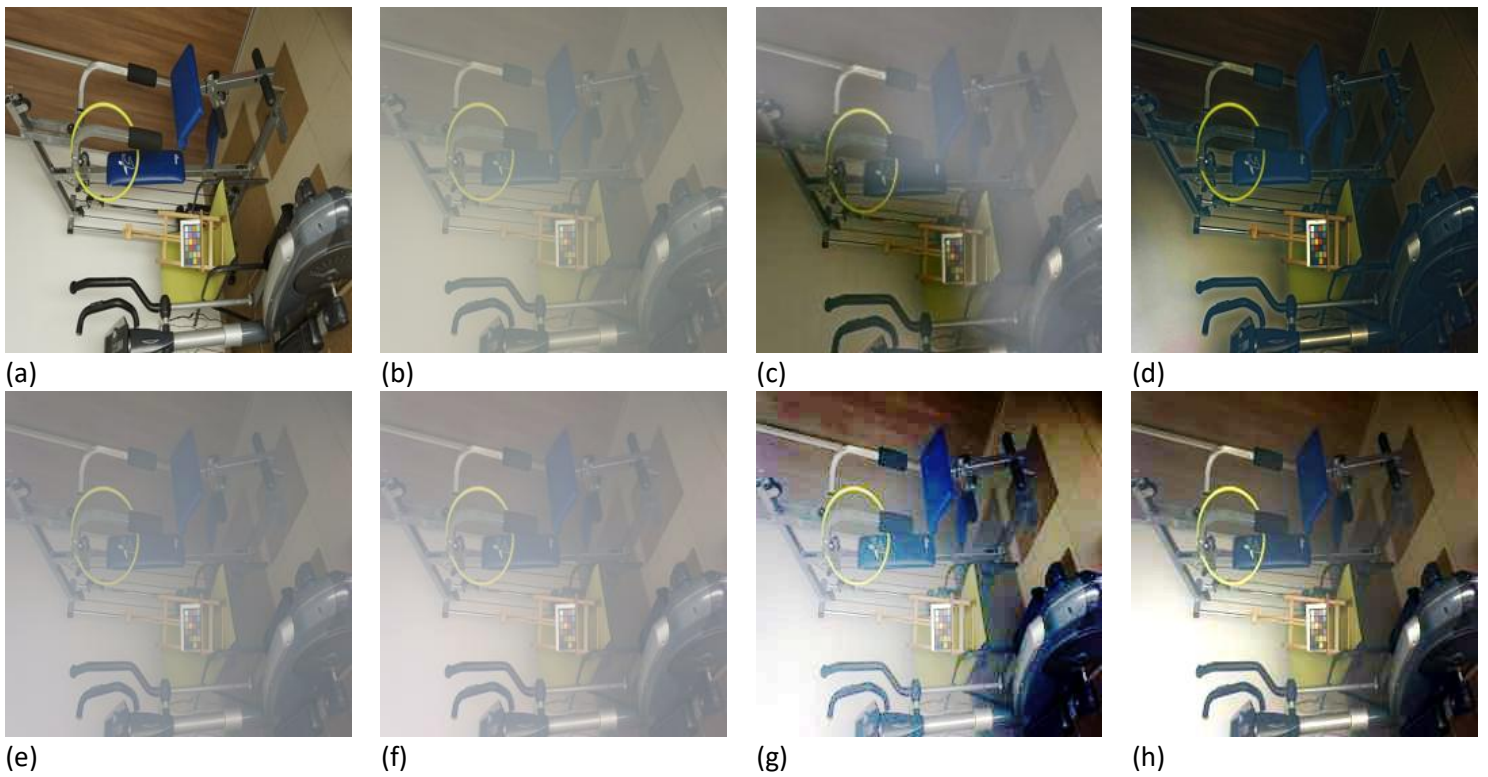


Figure. 9. Comparative results for image number 12 from I-Haze dataset. (a) input image , (b) hazy image, (c) multi-scale fusion, (d) Dark channel prior, (e) Gray World, (f) White Patch, (g) Histogram Equalization, and (h) proposed approach

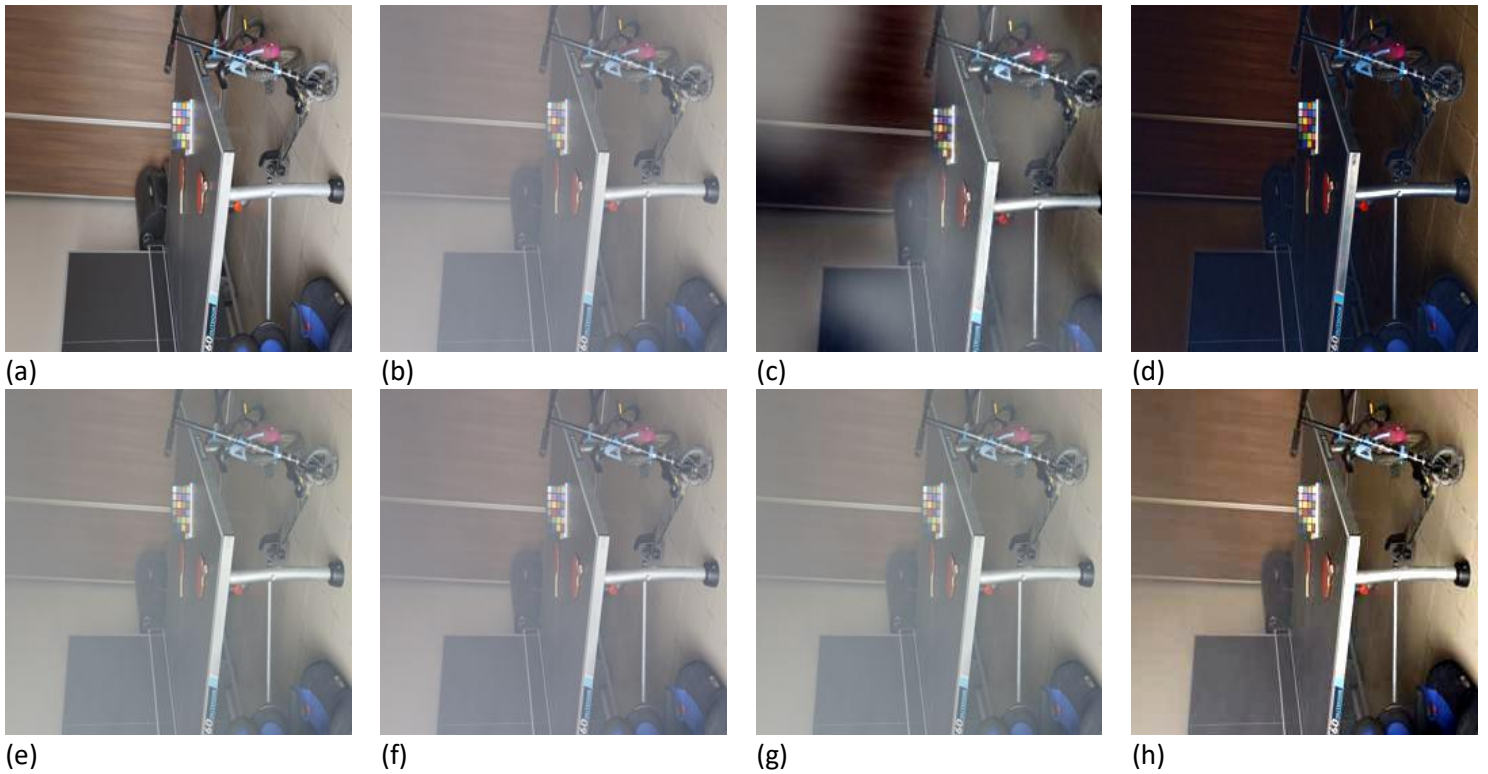


Figure. 10. Comparative results for image number 20 from I-Haze dataset. (a) input image , (b) hazy image, (c) multi-scale fusion, (d) Dark channel prior, (e) Gray World, (f) White Patch, (g) Histogram Equalization, and (h) proposed approach

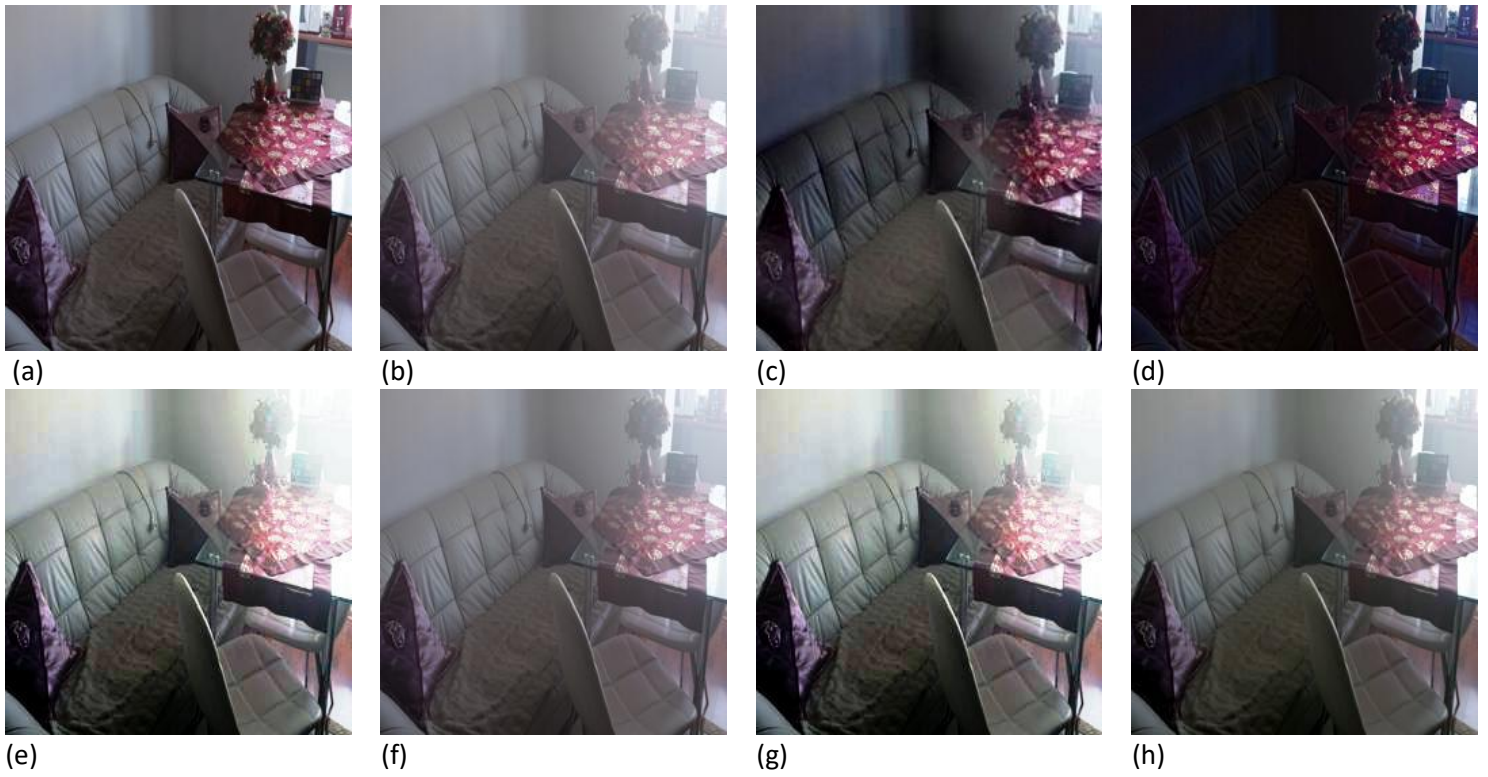


Figure. 11. Comparative results for image number 22 from I-Haze dataset. (a) input image , (b) hazy image, (c) multi-scale fusion, (d) Dark channel prior, (e) Gray World, (f) White Patch, (g) Histogram Equalization, and (h) proposed approach