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Efficient Sandstorm Image Color Correction Using Rank-Based Singular Value Recombination

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Abstract: Sandstorm images have a color cast that is reddish or yellowish due to the attenuation of the color channel. When light propagates through sand particles, it is scattered. Additionally, if some sand particles have a certain color, the obtained image in this circumstance experiences a color shift because some color channels are attenuated. Therefore, sandstorm images have a symmetrically distributed color cast throughout the image. There have been many studies aiming to enhance sandstorm images. In many studies, to enhance color-casted images, researchers have used the various methods as with gamma correction. However, artificial color shift occurs in an enhanced image because these methods do not reflect the image's adaptive feature. This paper proposes a sandstorm image enhancement method using singular value recombination based on rank. The singular value of an image reflects the image's characteristics adaptively, and improved images using the singular values have no artificial color because the balanced image eliminates the degraded color cast. Because the balanced image has hazy or dusty features, the singular value ratio can be used to enhance the image. The enhanced images produced using the proposed method are superior compared to state-of-the-art methods, objectively and subjectively.

Keywords: sandstorm image enhancement; singular value decomposition; rank-based recombined singular value



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1. Introduction

Images that are obtained in sandstorm circumstances have a color cast that is reddish or yellowish. When the light propagates in sand particles, if the sand particles have a certain color, then the obtained image has an artificial color in accordance with the particle's color. Because the distribution of color shift in sandstorm images is symmetrical, to apply sandstorm images in the computer vision and image recognition field, a compensation procedure for the degraded color channel is needed. There have been many studies aiming to enhance degraded sandstorm images. However, the existing methods cause an artificial color cast in the improved image. To enhance sandstorm images naturally, a color compensation procedure is needed. Because sandstorm images and hazy or dusty images have the same obtaining procedure, to enhance the degraded image, there are two categories of methods, namely, model-based methods and model-free improvement methods. Model-based methods use the haze image model [1–5], and model-free methods use image processing methods.

There are many studies that have aimed to enhance sandstorm images based on model-free methods. Cheng et al. [6] enhanced degraded sandstorm images using robust gray world [7] and guided image filter [8]. Ameen's [9] method improves color-degraded sandstorm images using triple measures of images; however, the enhanced images produced using this method have an artificial color shift because the constant triple measurements do not adequately reflect the image's features. Shi et al. enhanced sandstorm images using the mean shift of color components and adaptive CLAHE [10]. Although Shi et al.'s [10] method is able to correct the color, the enhanced images using this method have a new

color in some cases. This method is able to improve sandstorm images; however, artificial color can appear in the enhanced image.

The model-based sandstorm image enhancement methods use the hazy image model [1–5]. Hazy images have the same obtaining procedure as dusty images. Therefore, hazy image enhancement methods are used in the dusty image enhancement area. He et al. [1] enhanced hazy images. This method enhances the hazy image using dark channel prior, which estimates the darkest region of image. However, because the constant kernel is used to estimate the darkest region of the image, a ringing effect occurs, and to compensate for this phenomenon, this method uses the guided image filter [8]. Meng et al. [11] enhanced dusty images using the refined transmission map. This method improves the dusty image using the refined transmission map which estimates the adaptive boundary region better than DCP [11]. Shi et al. [12] enhanced sandstorm images using the mean shift of color components and the transmission map. Shi et al.'s [12] method has a color correction operator; however, some of the enhanced images using this method have an artificial color shift because the color correction operator does not reflect the image's features adaptively. Gao et al.'s method [13] enhances sandstorm images using the reversal of the color channel and an adaptive transmission map. Gao et al.'s method [13] also enhances sandstorm images; however, the enhanced images produced by this method seem hazy, even though the dehazing procedure is applied. T. Naseeba et al. enhanced sandstorm images using the merging of three techniques, namely the depth estimation module (DEM), the color analysis module (CAM), and the visibility restoration module (VRM) [14]. DEM uses the median filter and adaptive gamma correction technique. CAM uses the gray world assumption and analyzes the color features of hazy images. VRM applies the adjusted transmission map and color corrected image [14]. Dhara et al. enhanced hazy images using adaptive airlight refinement and nonlinear color balancing [15]. This method enhances the hazy image, even though the hazy image has a color cast. However, an artificial color cast appears in the enhanced image. Lee enhanced sandstorm images using the normalized eigenvalue and adaptive dark channel prior [16].

Recently, machine learning-based dehazing methods have been studied. Ren et al. enhanced hazy images using a convolutional neural network (CNN) [17]. This method uses two types of networks, one which predicts the transmission map and another which refines the result. This method enhances the images; however, the weak point is that in the case of nighttime images, a color shift occurs due to the color of lamplight. Wang et al. proposed a dehazing method using atmospheric illumination prior [18]. Zhu et al. improved hazy images using a color attenuation prior [19]. This method enhances hazy images by modeling their scene depth. However, the weak point of this method is that it needs a more flexible model to estimate the scattering coefficient [19]. Zhang et al. improved hazy images using multiscale CNN [20]. This method enhances hazy images; however, the weak point of this method is that the estimated transmission map has an incorrect region, and this causes the artificial effect [20].

Sandstorm images have a distorted color channel because of scattered light. Moreover, because degraded images have attenuated color channels, in order to enhance color-casted sandstorm images, a color correction procedure is needed. This paper proposes a sandstorm image enhancement method using rank-based singular value recombination. The corrected image produced using the proposed method has no color cast and hazy features. Because the color-corrected sandstorm images have hazy features, this paper applies dehazing methods to enhance the corrected image. There are many methods to enhance hazy images using dehazing algorithms. However, these methods cause an artificial effect in the enhanced image, such as the ringing effect and color shift. Additionally, the guided image filter [8] is able to provide an enhanced image. Cheng et al.'s method [6] uses the guided image filter [8] to enhance the image naturally. Therefore, this paper uses the guided image filter [8] to enhance balanced sandstorm images using the singular value ratio. The enhanced images produced using the proposed method seem to naturally be without color shift and ringing effect. Additionally, through comparison with state-of-the-art methods,

the performance of the proposed method is shown to be superior to that of other methods both subjectively and objectively.

2. Proposed Method

Sandstorm images have a color cast that is reddish or yellowish due to the scattering of light by the color of sand particles. Additionally, images in sandstorm circumstances have attenuated color channels. Therefore, when a sandstorm image is enhanced, if the color correction procedure is skipped, then the improved image has an artificial color cast. Therefore, this paper proposes the color correction method using singular value recombination based on rank. Tripathi et al. used singular value decomposition to analyze images [21]. Li et al. denoised images using the images' singular value [22]. The singular value of the color channel reflects the image's features, such as contrast. If an image has low contrast, the singular value of the first order has a lower value than that of an image with high contrast, because the first-ranked singular value reflects the image's background. Therefore, the singular value is able to show the features of an image, such as contrast.

This paper corrects the degraded color channel using the rank-based singular value, and this is described as:

$$[U^c, \Sigma^c, V^c] = svd(I^c), \tag{1}$$

where $svd(\cdot)$ is the operator of singular value decomposition, I^c is the input image, Σ^c is the diagonal singular value matrix of color channel, $c \in \{r, g, b\}$, and U^c and V^c are orthogonal matrices. In Equation (1), the input image is decomposed into singular values. The singular value of each color channel shows the channel's features based on rank. The singular value therefore shows the image's features. If an image is degraded more, the maximum singular value is lower than that of a less degraded channel. Therefore, if the distribution of the singular value is uniform, the color-casted channel is able to be corrected. Figure 1 shows a dusty image and a color-distorted image, and the singular value of each color channel. As shown in Figure 1, the distribution of the singular value of a dusty channel is uniform and the color channels are not attenuated. However, the color-distorted sandstorm image does not have uniform color channels, and their singular values are not uniform; the singular value of the red channel is the highest, and the singular value of the blue channel is the lowest because the blue channel is the most attenuated and has low contrast in comparison with the red channel. Therefore, the singular value of the image is able to reflect the image's condition.

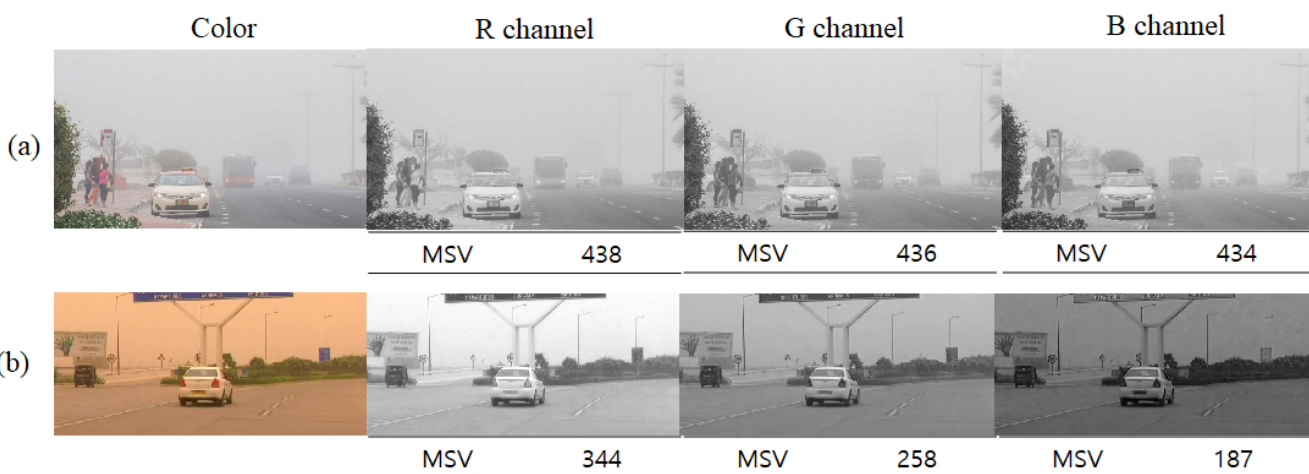


Figure 1. The distribution of the singular value on each color channel of a color distortion image and dusty image (the tables below the image show the singular value of each color channel, and MSV is the maximum singular value): (a) dusty input image and each color channel; (b) color-distorted image and each color channel.

As shown in Figure 1, because the singular value of the image reflects the image's characteristics, this paper uses the recombined singular value based on rank to correct the image's color. Figure 2 shows the diagram of the proposed method. As shown in Figure 2, this paper, based on the model-free sandstorm image enhancement method, improves a degraded sandstorm image using the image's singular value, and to enhance the corrected image, the guided image filter is used.

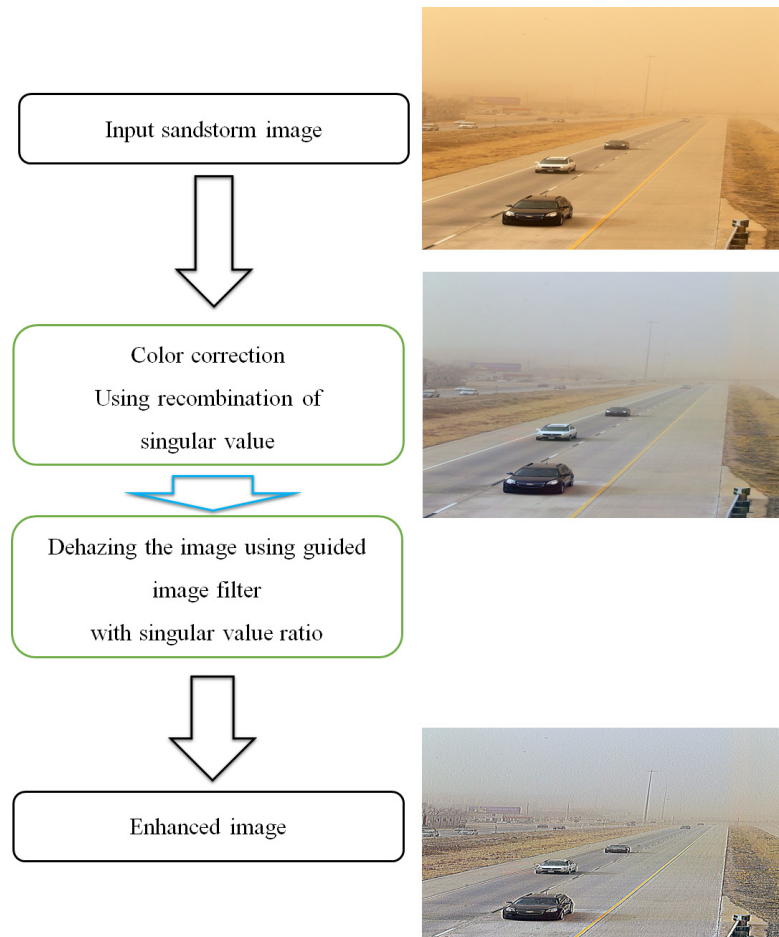


Figure 2. The diagram of the proposed method.

2.1. Color Correction

As shown in Figure 1, because degraded sandstorm images have a color cast due to attenuated color channels, to correct sandstorm images, a color balancing procedure is needed. This paper uses the image's singular value to balance the image, reflecting the image's features. The obtaining procedure of the image's singular value is shown in Equation (1). Because the image is attenuated more, the singular value of the image is lower, and to enhance the degraded image, recombination of the singular value is needed. The recombination of the singular value is described as:

$$\Sigma_B^c = \Sigma_{ind}^c + \alpha^c \cdot \{m(\Sigma_{ind}^c) - \Sigma_{ind}^c\}, \quad (2)$$

$$\alpha^c = \frac{\Sigma_1^c}{\Sigma_1^c + \Sigma_2^c}, \quad (3)$$

where Σ_B^c is the recombined singular value based on rank, Σ_{ind}^c is the rank of each color channel's singular value, $m(\cdot)$ is the average operation on each color channel based on rank, and α^c is the normalized singular value of the first ranked singular value via sum of

first and second ranked singular value to prohibit all singular values being the same. Using Equations (2) and (3), the color-casted image is corrected and has no artificial color.

Figure 3 shows the input image and color-corrected image produced by Equations (2) and (3) and the distribution of the singular value of each color channel. As shown in Figure 3, the corrected image has no color cast and seems to have hazy features, and the distribution of the singular value is not the same but uniform in comparison with the distribution of the input image's singular values. By decreasing the singular value of the red color channel, the contrast of the red channel is decreased. If the red channel's contrast is maintained and the other channels such as green and blue are enhanced, a color shift can occur in the enhanced image because the intensity value of the red channel is superior to that of the other channels, and in some cases, the intensity value of the red channel is close to one. Therefore, to enhance sandstorm images naturally, the adjusting procedure of the red channel is also needed. As shown in Figure 3, the color correction performance of the proposed method is competitive.

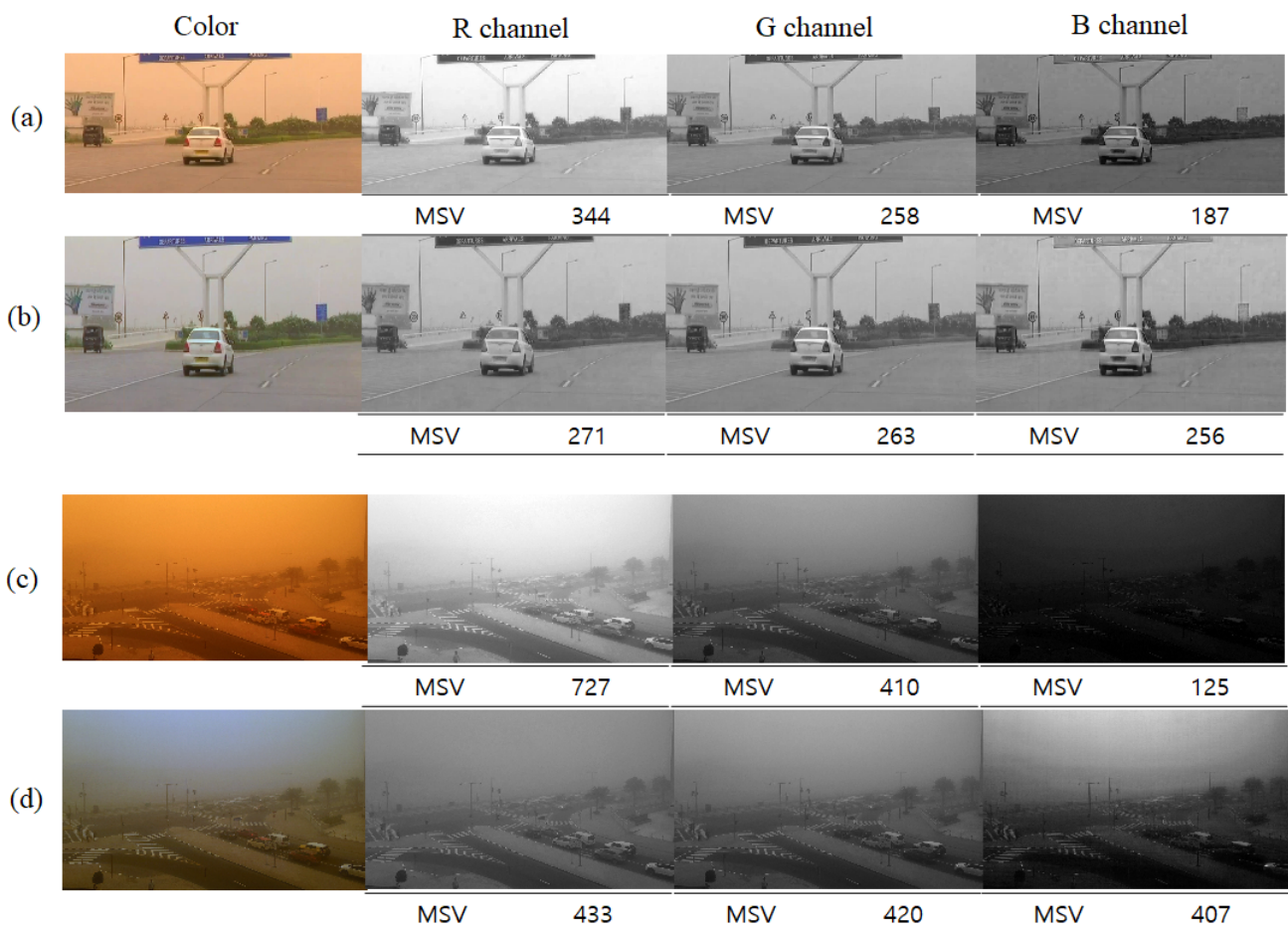


Figure 3. The distribution of singular value each color channel of input images and enhanced images by the proposed method (the tables below the image show the singular value of each color channel and MSV is the maximum singular value): (a,c) input images and each color channel; (b,d) enhanced images and each color channel.

2.2. Dehazing

The color-corrected sandstorm image seems hazy or dusty. Dusty images and hazy images look dimmed and unclear. Therefore, to enhance dimmed images, a dehazing procedure is needed. There are many methods that have been studied to improve hazy or dusty images. The dark channel prior (DCP) method [1] is frequently used to enhance

hazy images. The DCP method estimates an image’s darkest region, and using this a transmission map is estimated. The estimation procedure of DCP is described as:

$$I_{dark}(x) = \min_c \left(\min_{y \in \Omega(x)} \left(\frac{I^c(y)}{A^c} \right) \right), \tag{4}$$

where $I_{dark}(x)$ is the estimated dark channel $c \in \{r, g, b\}$, $\Omega(x)$ is the patch region, A^c is the backscattered light of each color channel [1], and x is the location of the pixel. He et al.’s method [1] estimates the darkest region in an image using constant patch, and because of this, the estimated image has a squared effect. Additionally, the transmission map which shows the propagation path of light also has a ringing effect due to the transmission map being estimated by reversing DCP; to compensate for this point, a guided image filter is used [8]. The description of the estimation of the transmission map is as follows:

$$t(x) = 1 - \omega \cdot I_{dark}(x), \tag{5}$$

where $t(x)$ is the transmission map, ω is a constant value to show the ‘aerial perspective’ [1,23,24], and is set to 0.95 in He et al.’s method [1]. As shown in Equation (5), the transmission map is estimated by reversing the dark channel. The dehazing method by He et al. [1] is useful. However, the constant patch value causes an artificial effect, such as a square effect. Additionally, this causes new distortion in the enhanced image. Cheng et al. [6] enhanced sandstorm images using the guided image filter [8]. The procedure of applying guided image filtering [8] in the dehazing area is described as:

$$I_G^c(x) = G_f(I_B^c(x), k, eps), \tag{6}$$

where $I_G^c(x)$ is the guided filtered image, $I_B^c(x)$ is the color-balanced image, k is a kernel and is set to 2, eps is set to 0.4^2 , and x is the location of the pixel. In Equation (6), the color-balanced image is blurred, and it acts as the initial procedure of dehazing using the guided image filter [8]. The dehazing procedure using the guided image filter [8] is described as:

$$I_E^c(x) = (I_B^c(x) - I_G^c(x)) \cdot w^c + I_G^c(x), \tag{7}$$

where $I_E^c(x)$ is the enhanced image and w^c is the controlling factor to produce the naturally enhanced image. In Equation (7), the color-balanced image is enhanced, and the image is clearly not dimmed. In Equation (7), the enhanced image is controlled by factor w^c . To obtain the naturally enhanced image, the controlling factor is also applied to the image adaptively. As shown in the color balancing procedure, the singular value of the image reflects the image’s condition, such as contrast. If the image is enhanced, then the dusty components are also increased. Additionally, this is also reflected through the variation of the singular value. Therefore, this paper uses the variation of the singular value of the color channel to obtain the controlling factor, w^c . The obtaining procedure of controlling factor w^c is described as:

$$w^r = \frac{1}{2} \cdot \left(\frac{\max(\Sigma_{B,1}^r, \Sigma_1^r)}{\min(\Sigma_{B,1}^r, \Sigma_1^r)} + w_0 \right) \tag{8}$$

$$w^{c'} = \frac{1}{2} \cdot \left(\frac{\Sigma_{B,1}^{c'}}{\Sigma_1^{c'}} + w_0 \right), \tag{9}$$

where w^r is the controlling factor of the red channel, $\Sigma_{B,1}^r$ is the first ranked singular value of the balanced red channel, Σ_1^r is the first ranked singular value of the input red channel, $c' \in \{g, b\}$, and w_0 is the initial controlling factor and is set to 25. If the first ranked singular value of the balanced red channel is lower than the input image’s first-ranked singular value, then the controlling factor is lower than 1. Although the red

channel is not as attenuated as other color channels, because of dusty components, the red channel also has many dusty particles. However, the controlling measure is lower than one because the effect of enhancement is not vivid, and therefore, through the comparison of the maximum and minimum singular value, the red channel's controlling measure is obtained. In Equations (8) and (9), the balanced image is enhanced naturally by reflecting the image's condition.

Figure 4 shows the comparison between the enhanced images produced using the DCP method [1] and the proposed method. Figure 4c shows the enhanced images produced using the DCP method [1]. As shown in Figure 4c, the enhanced images produced using the DCP method [1] demonstrate an artificial effect in the sky region. Figure 4d shows the enhanced images produced using the proposed method. As shown in Figure 4d, the enhanced images produced using the proposed method display no artificial effects or color distortion, and seem natural.

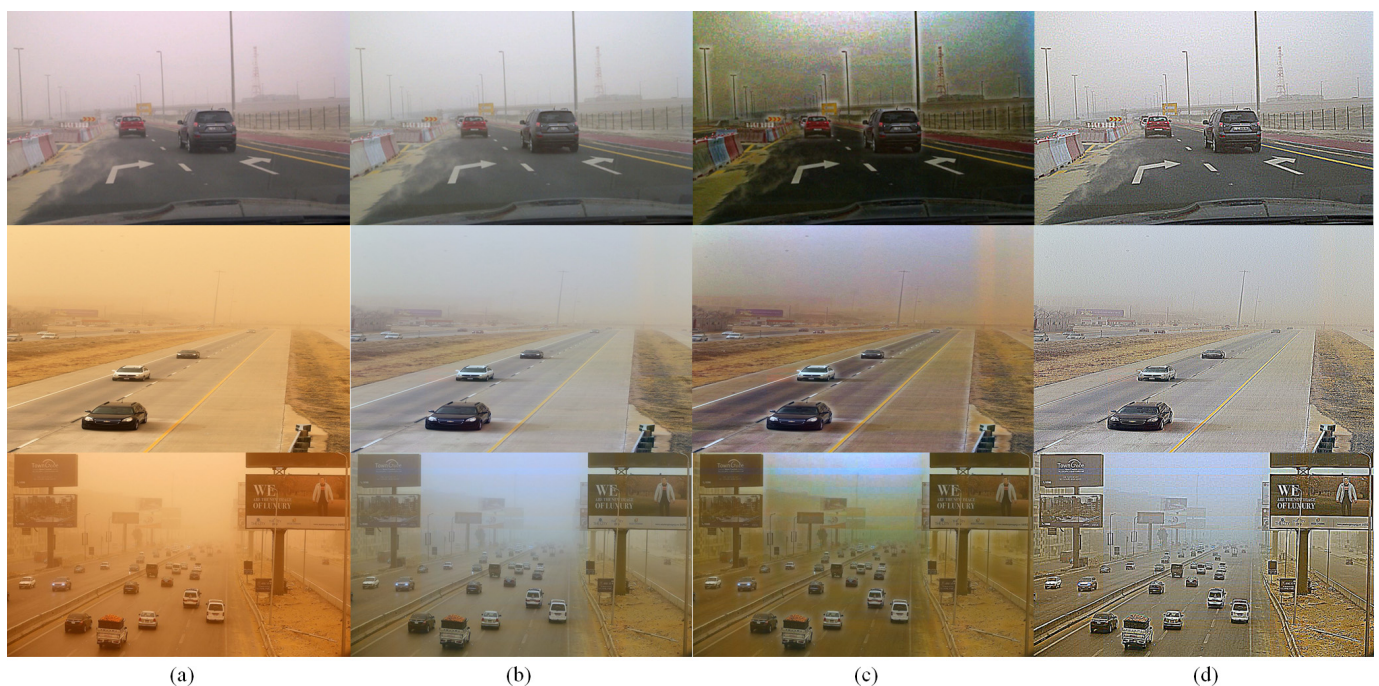


Figure 4. Comparison between the enhanced images produced using the DCP method [1] and the proposed method: (a) input images; (b) color-balanced images; (c) enhanced images produced using the DCP method [1]; (d) enhanced images produced using the proposed method.

As shown in Figures 3 and 4, to enhance sandstorm images naturally, color balancing and the image adaptive dehazing procedure are needed. Additionally, the enhanced images produced using the proposed method have no color shift or distorted areas. Therefore, the proposed method has sufficient performance to enhance the sandstorm image enhancement field.

3. Experimental Results and Discussion

Distorted sandstorm images demonstrate a color shift due to attenuated color channels. To enhance a degraded sandstorm image, the imbalanced color channel is compensated. The proposed method performs suitable balancing of the color components using the recombined singular value of each color channel. The balanced image produced using the proposed method seems natural, without any color distortion. This section shows the performance of the proposed method through comparison with state-of-the-art methods. To compare this method with state-of-the-art methods subjectively, this paper composes two categories. One is a comparison of the color-balanced result, and the other is a comparison of the enhanced image with sandstorm images from various environments in the Detection

in Adverse Weather Nature (DAWN) dataset [25] and the Weather Phenomenon Database (WEAPD) [26]. Additionally, to assess the enhanced sandstorm images objectively using the proposed method and state-of-the-art methods, three metrics are used.

3.1. Color Correction

Color-degraded sandstorm images seem reddish and yellowish. To enhance sandstorm images naturally, a color compensation procedure is needed. If not, the enhanced image has an artificial color shift. This section presents the comparison of color-balanced images produced using the proposed method and state-of-the-art methods subjectively. Shi et al.'s [12] method enhances sandstorm images using the mean shift of color components. Shi et al. [10] improved the color components of degraded sandstorm images using mean shift and gamma correction. Al Ameen's method [9] enhances color-degraded sandstorm images using the gamma correction. Lee enhanced degraded sandstorm images using the normalized eigenvalue [16].

Figures 5–7 show color-balanced images produced using the proposed method and state-of-the-art methods. Figure 5 shows a variously degraded sandstorm image. Shi et al.'s [12] method enhances sandstorm images using the mean shift of color components. This method operates efficiently to balance the image's color. However, artificial color shift can occur in the enhanced image because this method uses just the mean shift of color components. Shi et al. [10] improves the color components of degraded sandstorm images using mean shift and gamma correction. This method improves the distorted sandstorm image, but a new color shift can occur in the enhanced image because of the mean shift of color components. Al Ameen's method [9] enhances color-degraded sandstorm images using the gamma correction. This method improves distorted sandstorm images; however, the weak point of this method is using a constant value to enhance sandstorm images, and it does not reflect images' features adaptively. As shown in Figure 5, the color correction on the lightly degraded sandstorm image is not a tough task for Shi et al.'s [12] and Shi et al.'s [10] methods. The Al Ameen method [9] uses a constant value to enhance sandstorm images, and the enhanced image's artificial color shift occurs because the constant value does not reflect the image's feature adaptively. The enhanced images produced using the Lee method [16] seem bright, because to enhance the degraded sandstorm images, Lee's method [16] uses the normalized eigenvalue, and due to this, the enhanced images have a bright region; the eigenvalue of the red channel is more abundant than that of other color channels. However, the enhanced images produced using the proposed method enhance the color components of degraded sandstorm images naturally without any color shift.

Figure 6 shows lightly degraded sandstorm images and severely distorted sandstorm images, along with enhanced images produced using the proposed method and state-of-the-art methods. Shi et al. [12] improved the degraded sandstorm images using the mean shift of color components. This method enhances the sandstorm images naturally, but in the case of greatly degraded sandstorm images, the enhanced images produced using this method have a blueish artificial color shift. The Shi et al. [10] method enhances the lightly degraded sandstorm images naturally. However, in the case of the much-distorted sandstorm images, the enhanced images demonstrate a new blueish color shift. The Shi et al. methods [10,12] enhance the sandstorm images in the case of light degradation, but in the case of the greatly degraded sandstorm images, an artificial color shift can occur in the enhanced images due to these methods not reflecting the image's features adaptively. Al Ameen's method [9] enhances the sandstorm images; however, the enhanced images using this method have an artificial color shift due to using a constant value to enhance the distorted color, and it does not reflect the image's features. The improved images produced using the Lee method [16] have a bright region due to the abundant red channel. Meanwhile, the enhanced images produced using the proposed method have no artificial color shift.

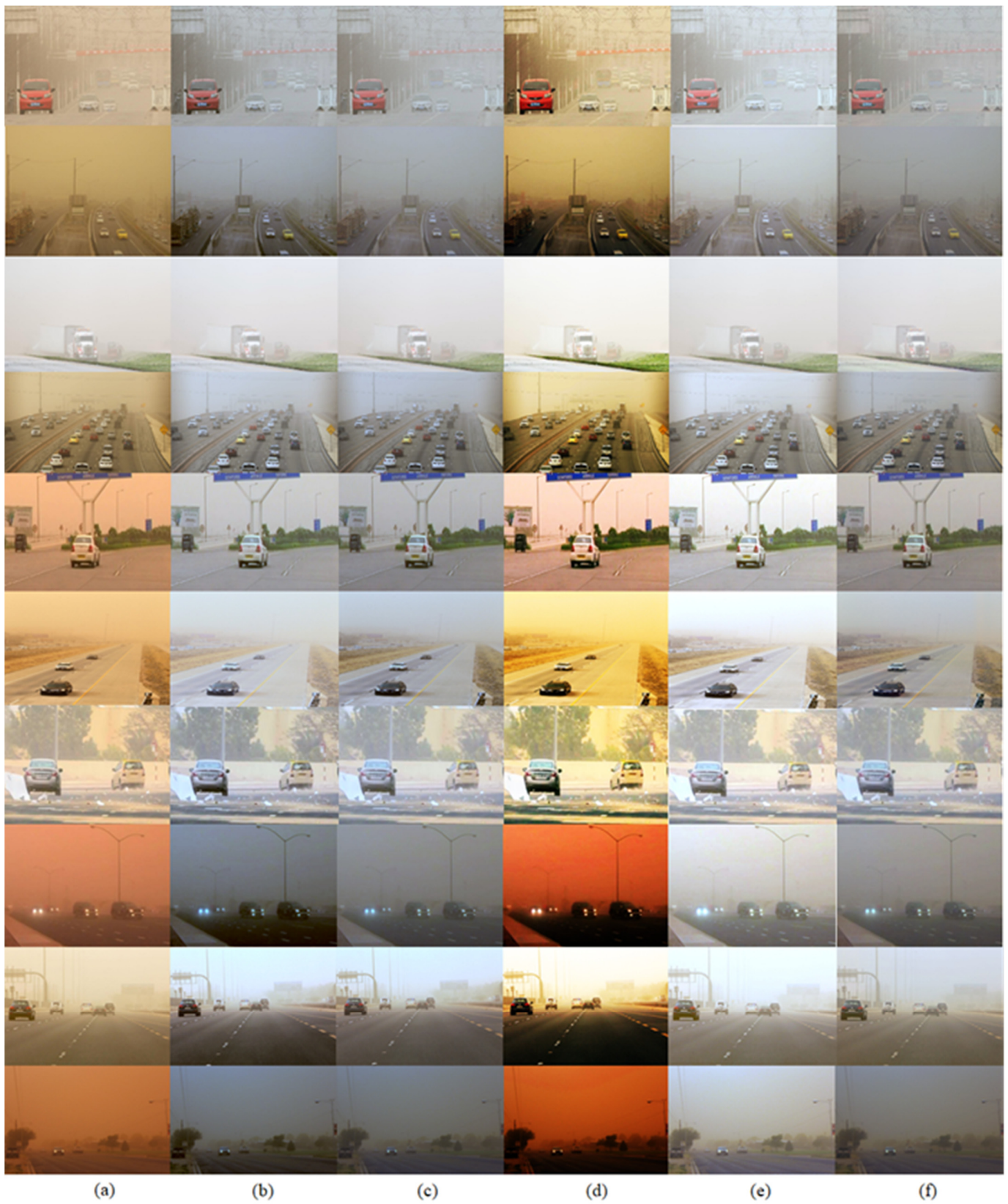


Figure 5. The comparison of color correction using existing methods and the proposed method using the DAWN dataset [25]: (a) input; (b) Shi et al. [10]; (c) Shi et al. [12]; (d) Al Ameen [9]; (e) Lee [16]; (f) Proposed method.



Figure 6. The comparison of color correction using existing methods and the proposed method using the DAWN dataset [25], and WEAPD [26]: (a) input; (b) Shi et al. [10]; (c) Shi et al. [12]; (d) Al Ameen [9]; (e) Lee [16]; (f) Proposed method.



Figure 7. The comparison of color correction using existing methods and the proposed method using WEAPD [26]: (a) input; (b) Shi et al. [10]; (c) Shi et al. [12]; (d) Al Ameen [9]; (e) Lee [16]; (f) Proposed method.

Figure 7 shows variously degraded sandstorm images. Shi et al.'s [12] method enhances the sandstorm images; however, the enhanced images have a bluish color shift in

the greatly degraded sandstorm images. Shi et al.'s [10] method enhances the sandstorm images. However, the enhanced images using this method have a blueish artificial color shift in some cases. The Al Ameen method [9] enhances the sandstorm images in cases of light degradation. However, in the case of the greatly degraded sandstorm images, the enhanced images have a yellowish artificial color shift. The enhanced images produced by means of Lee's method [16] have a bright area due to the abundant red channel and rare color components. Meanwhile, the enhanced sandstorm images produced using the proposed method seem natural and without any color shift.

As shown in Figures 5–7, the enhanced images produced using state-of-the-art methods have an artificial color shift in various types of sandstorm images. However, the enhanced images produced using the proposed method have no color shift and seem natural. Therefore, the color balancing algorithm of the proposed method is superior to the state-of-the-art method.

3.2. Enhanced Image

The color-imbalanced sandstorm images are corrected using the proposed method, and the balanced images seem natural. Because the balanced sandstorm images seem hazy, to enhance the image, a dehazing procedure is needed. To enhance the hazy image, dehazing algorithms are used frequently. He et al. enhanced hazy images using dark channel prior and a transmission map [1]. Meng et al. improved hazy images using a boundary-refined transmission map [11]. Ren et al. enhanced hazy images using a convolutional neural network (CNN) [17]. Gao et al. improved sandstorm images using the autoreversing blue channel compensation method and an adaptive transmission map [13]. Shi et al. enhanced sandstorm images using the mean shift of color components and an adjustable transmission map [12]. Al Ameen [9] improved sandstorm images using the gamma correction method. Shi et al. enhanced sandstorm images using the mean shift of color components and adaptive CLAHE [10]. Lee improved degraded sandstorm images using normalized eigenvalue and adaptive DCP [16]. Dhara et al. improved hazy images using adaptive airlight refinement and nonlinear color balancing [15]. Because sandstorm images resemble dusty or hazy images, to compare the enhanced images, existing dehazing methods were used.

Figures 8–13 show sandstorm images and enhanced sandstorm images using the proposed method and state-of-the-art methods.

Figures 8 and 9 show lightly degraded sandstorm images and greatly degraded sandstorm images, as well as enhanced images produced using the proposed method and state-of-the-art methods. He et al.'s [1] method enhances the sandstorm images in the case of lightly degraded sandstorm images, because this method has no color balancing procedure and in the enhanced image, an artificial color shift occurs. Meng et al.'s [11] method improves the sandstorm images in the lightly distorted case. However, in the most degraded sandstorm images, the enhanced images produced using this method have an artificial color because this method has no color correction procedure. Ren et al.'s method [17] enhanced the lightly degraded sandstorm images because the lightly degraded sandstorm images resemble dusty images. However, in the most degraded sandstorm images, the enhanced images produced using this method have a new color shift. Gao et al.'s [13] method enhances the sandstorm images without color shift, and the improved images seem naturally. Shi et al.'s [12] method enhances the sandstorm images, but the enhanced images have color shift and a ringing effect because of the transmission map. When estimating the transmission map, a kernel of a certain size is used. Additionally, because of this, the enhanced images demonstrate a color shift and a ringing effect. Al Ameen's [9] method enhances the degraded sandstorm images in the lightly degraded case. However, in the greatly degraded sandstorm images, the enhanced images have an artificial color shift. Because this method uses a constant value regardless of the image's features to enhance sandstorm images, the enhanced images have a distorted region. The enhanced images produced using Dhara et al.'s method [15] have an artificial color in some cases, as although

this method has a color balancing procedure, it does not sufficiently reflect the sandstorm images' features. Lee's method [16] improves the sandstorm images; however, due to the abundant red channel, a bright region is shown. Shi et al.'s method [10] improves the sandstorm images; however, a bluish color appears in some of the images due to the mean shift of color components. Meanwhile, the enhanced images produced using the proposed method have no color shift or ringing effect in both the lightly and greatly degraded sandstorm images.

Figures 10 and 11 show the lightly degraded sandstorm images and greatly distorted sandstorm images, as well as the improved images produced using the proposed method and state-of-the-art methods. He et al.'s method [1] improves the lightly degraded sandstorm images; however, in the case of the greatly degraded sandstorm images, the enhanced images demonstrate a color shift and an artificial effect. The enhanced images produced using Meng et al.'s [11] method have an artificial effect as well as a color shift and a ringing effect. Ren et al.'s [17] method improves the sandstorm images in the case of light degradation. However, in most of the degraded sandstorm images, the enhanced images have a greenish color shift and an artificial effect. Gao et al.'s method [13] enhances both the lightly and greatly degraded sandstorm images. Shi et al.'s [12] method enhances the sandstorm images; however, the enhanced images have a color shift and an artificial region due to the dehazing procedure as well as the transmission map. Al Ameen's [9] method enhances the sandstorm images in case of lightly degraded sandstorm images. However, in the case of greatly degraded sandstorm images, the enhanced images have a color shift because this method uses a constant value to correct the color, and this causes the new color-distorted region. The enhanced images produced using Dhara et al.'s method [15] have an artificial color in some images owing to this method not sufficiently reflecting sandstorm images' features. The enhanced images produced by Lee's method [16] have a bright area in some cases due to the red channel being abundant and that of the eigenvalue also being more abundant than others. The enhanced images produced using Shi et al.'s method [10] have a bluish color in some images due to the mean shift of color components. Meanwhile, the enhanced images produced using the proposed method have no color shift or artificial regions.

Figures 12 and 13 show the degraded sandstorm images and enhanced images produced using the proposed method and state-of-the-art methods. The enhanced images produced using the He et al. [1] method have a color shift and ringing effect, and the degraded color is still present because this method has no color balancing procedure. Meng et al.'s [11] method enhances sandstorm images in the case of lightly degraded sandstorm images. In the case of greatly distorted sandstorm images, the enhanced images produced using Meng et al.'s [11] method have an artificial color shift and a ringing effect because this method lacks a color correction procedure. The enhanced images produced using Ren et al.'s [17] method have a color shift and an artificial ringing effect, and color degradation occurs still because this method has no color balancing procedure. The Gao et al. [13] method enhances sandstorm images naturally without a color shift. However, the enhanced images display bright regions. The Shi et al. [12] method enhances the distorted sandstorm images. However, the ringing effect and distorted areas occur due to the dehazing procedure as well as the transmission map. The Al Ameen [9] method enhances sandstorm images in case of lightly degraded sandstorm images, because this method has no image-adaptive color correction procedure and uses a constant value to correct the image's color. The Shi et al. method [10] improves sandstorm images naturally; however, in some images, a blueish artificial color is present. Lee's method [16] enhances the degraded sandstorm images naturally; however, a bright region is present and it looks unnatural. Dhara et al.'s method [15] enhances the sandstorm images; however, degraded color is present in some images due to this method not being able to reflect the sandstorm images' features.

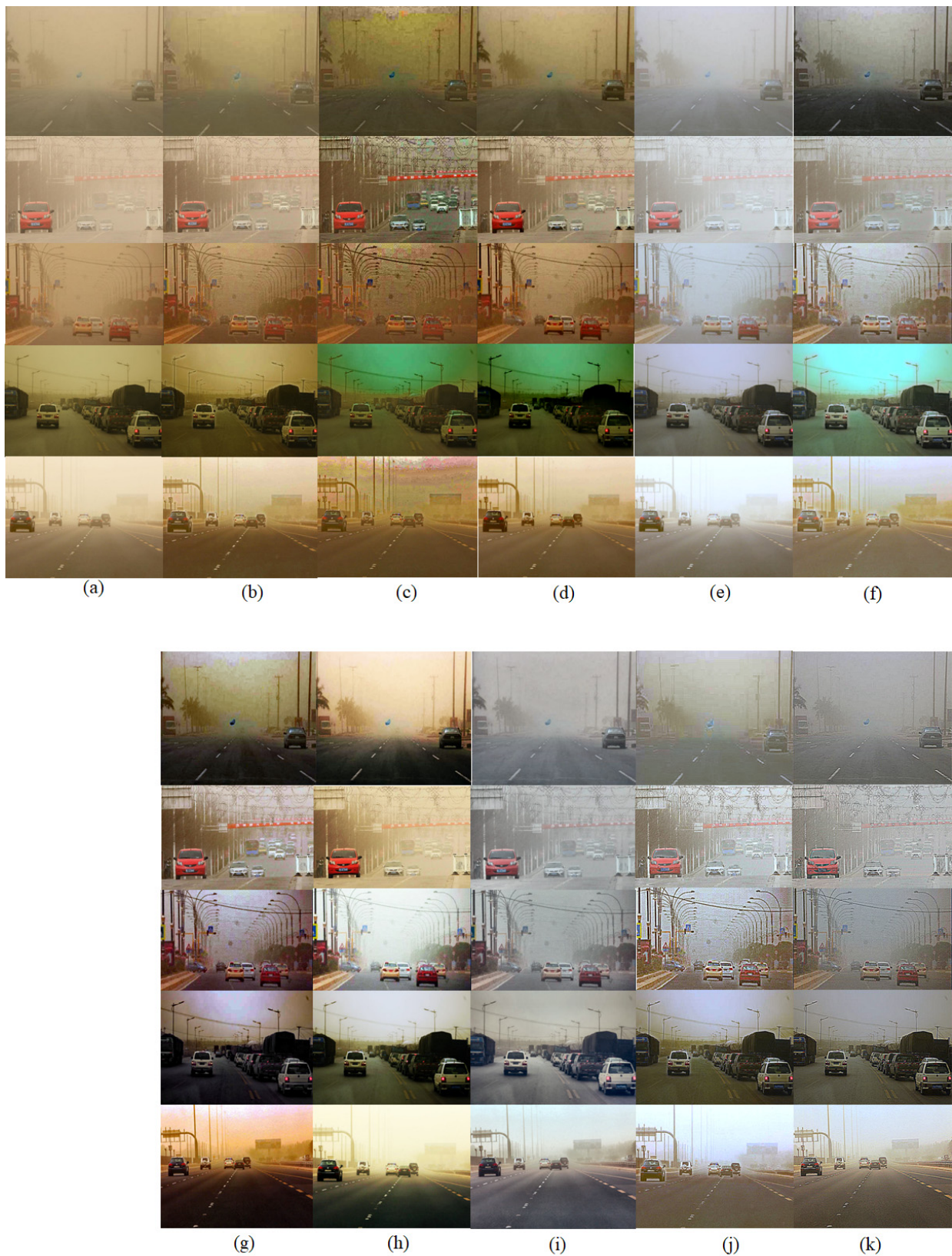


Figure 8. Comparison of enhanced images produced using existing methods and the proposed method using the DAWN dataset [25]: (a) input; (b) He et al. [1]; (c) Meng et al. [11]; (d) Ren et al. [17]; (e) Gao et al. [13]; (f) Shi et al. [12]; (g) Dhara et al. [15]; (h) Al Ameen [9]; (i) Shi et al. [10]; (j) Lee [16]; (k) Proposed method.



Figure 9. The comparison of enhanced images using existing methods and the proposed method using the DAWN dataset [25]: (a) input; (b) He et al. [1]; (c) Meng et al. [11]; (d) Ren et al. [17]; (e) Gao et al. [13]; (f) Shi et al. [12]; (g) Dhara et al. [15]; (h) Al Ameen [9]; (i) Shi et al. [10]; (j) Lee [16]; (k) Proposed method.

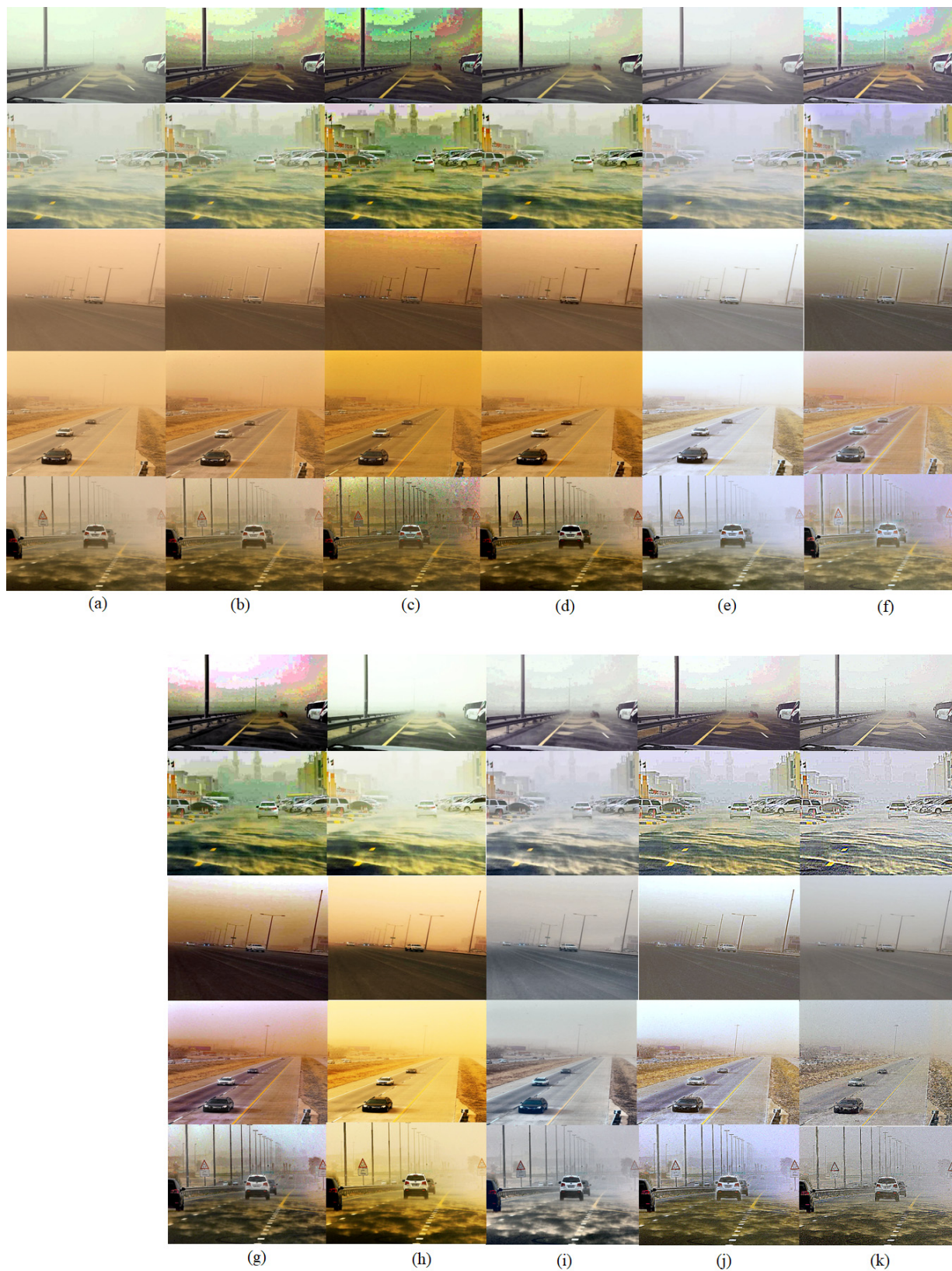


Figure 10. The comparison of enhanced images using existing methods and the proposed method using the DAWN dataset [25]: (a) input; (b) He et al. [1]; (c) Meng et al. [11]; (d) Ren et al. [17]; (e) Gao et al. [13]; (f) Shi et al. [12]; (g) Dhara et al. [15]; (h) Al Ameen [9]; (i) Shi et al. [10]; (j) Lee [16]; (k) Proposed method.

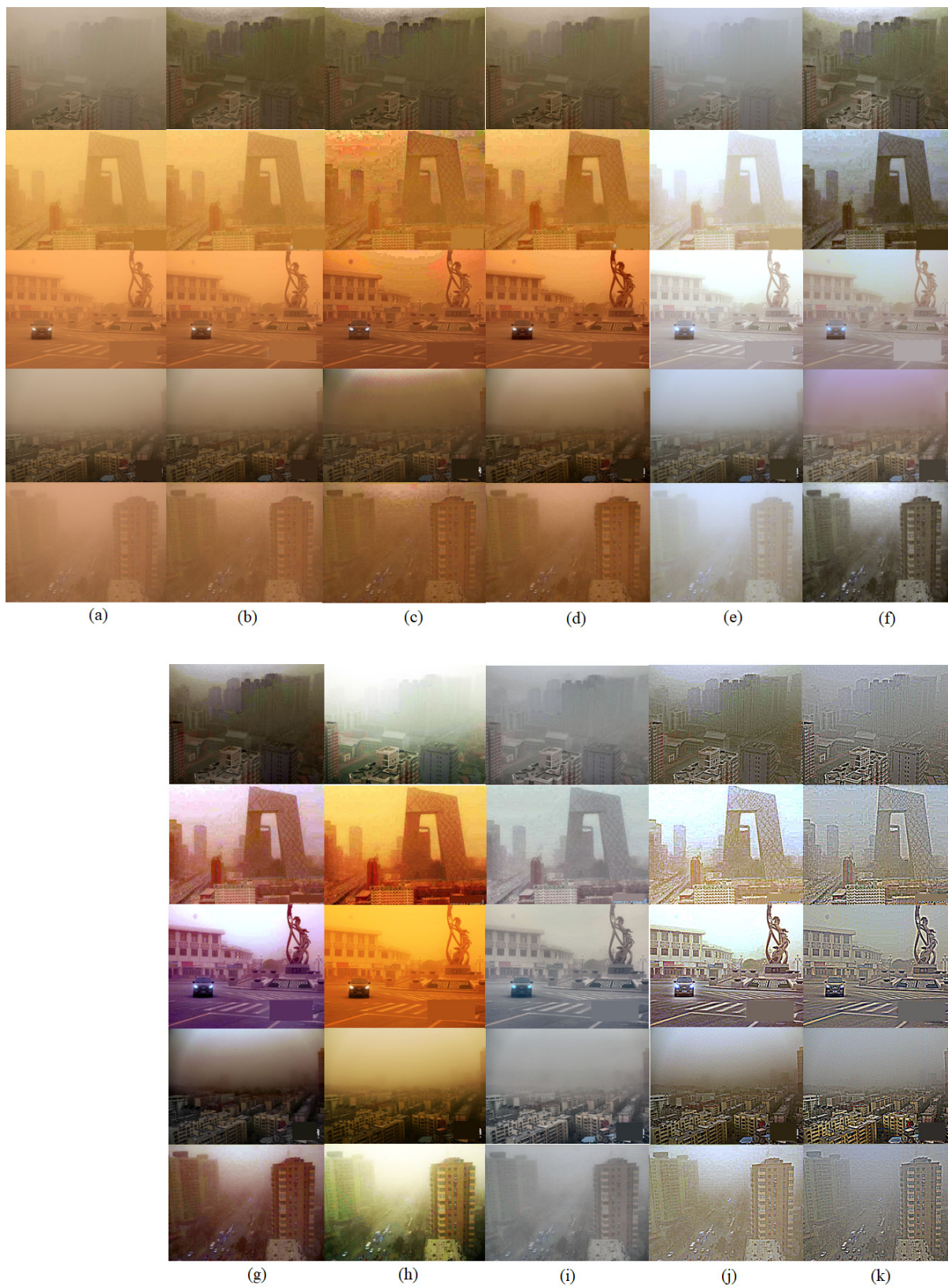


Figure 11. The comparison of enhanced images using existing methods and the proposed method using WEAPD [26]: (a) input; (b) He et al. [1]; (c) Meng et al. [11]; (d) Ren et al. [17]; (e) Gao et al. [13]; (f) Shi et al. [12]; (g) Dhara et al. [15]; (h) Al Ameen [9]; (i) Shi et al. [10]; (j) Lee [16]; (k) Proposed method.

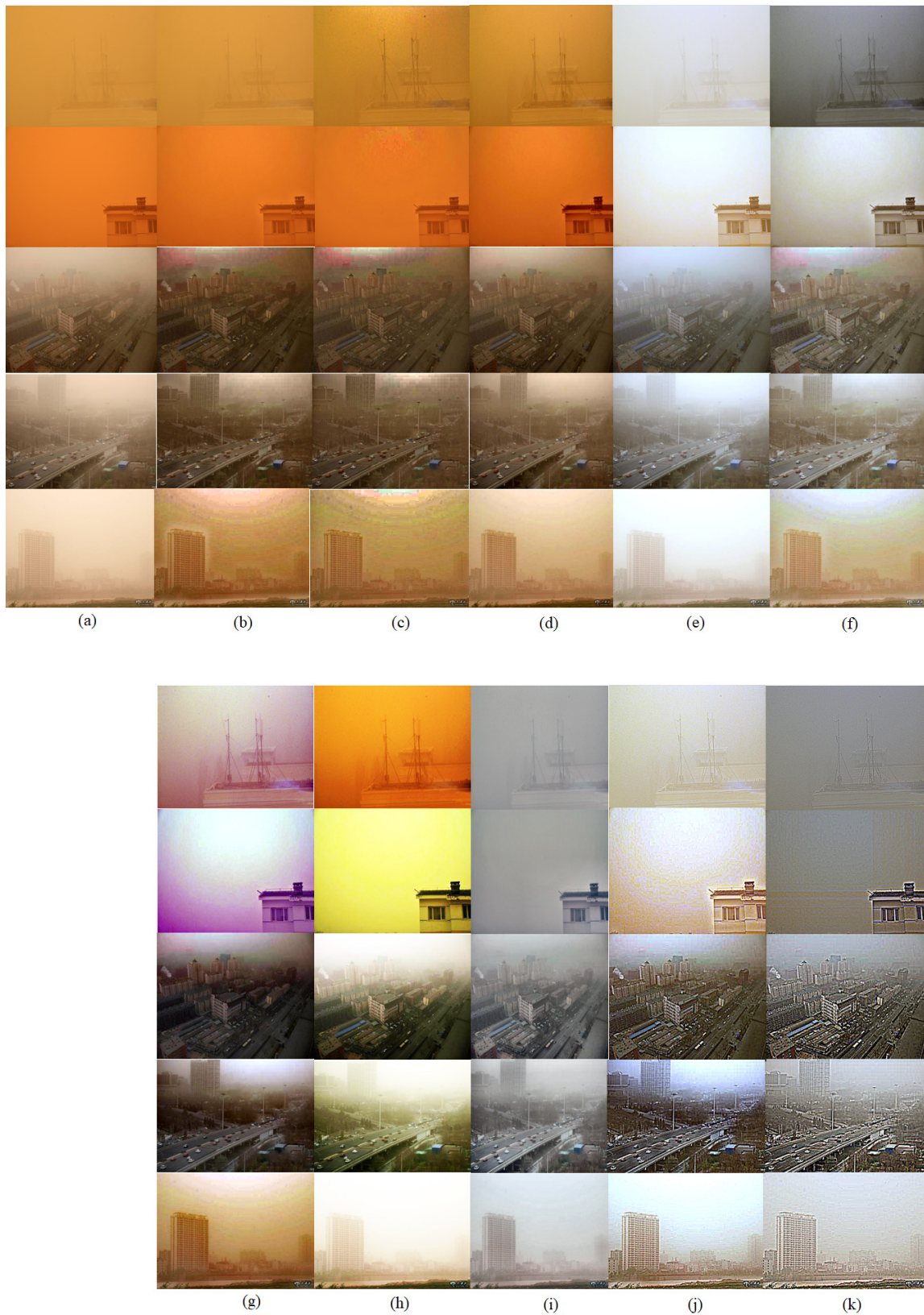


Figure 12. The comparison of enhanced images using existing methods and the proposed method using WEAPD [26]: (a) input; (b) He et al. [1]; (c) Meng et al. [11]; (d) Ren et al. [17]; (e) Gao et al. [13]; (f) Shi et al. [12]; (g) Dhara et al. [15]; (h) Al Ameen [9]; (i) Shi et al. [10]; (j) Lee [16]; (k) Proposed method.

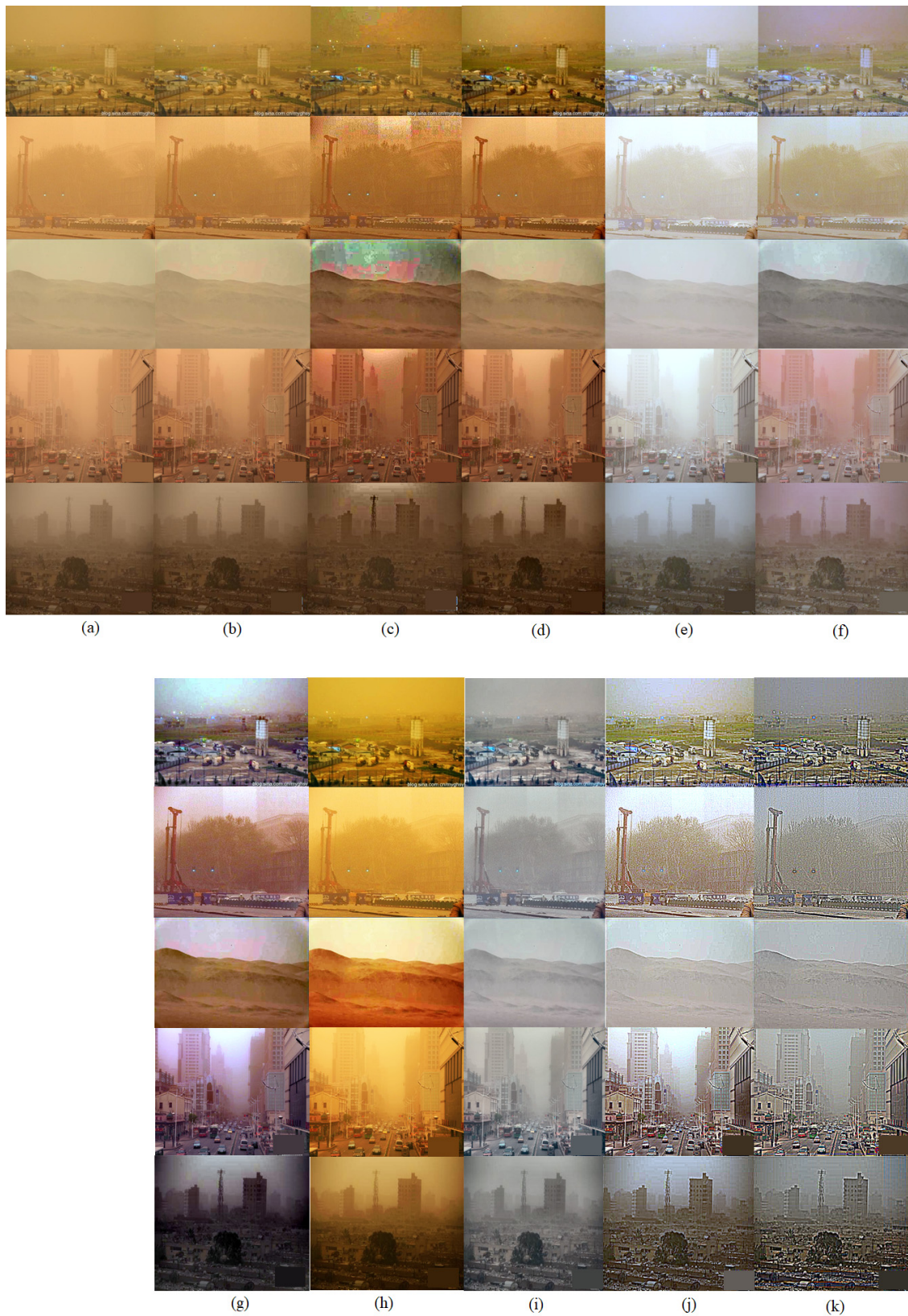


Figure 13. The comparison of color correction using existing methods and the proposed method using WEAPD [26]: (a) input; (b) He et al. [1]; (c) Meng et al. [11]; (d) Ren et al. [17]; (e) Gao et al. [13]; (f) Shi et al. [12]; (g) Dhara et al. [15]; (h) Al Ameen [9]; (i) Shi et al. [10]; (j) Lee [16]; (k) Proposed method.

As shown in Figures 8–13, to enhance the degraded sandstorm images naturally, the image adaptive color correction procedure and the dehazing procedure are needed. The proposed method enhances the degraded sandstorm images naturally in both lightly degraded sandstorm images and greatly degraded sandstorm images without any color shift or artificial regions. Therefore, the improved performance of the proposed method in the sandstorm image enhancement field is superior compared to that of state-of-the-art methods subjectively.

3.3. Objective Comparison

The enhanced sandstorm images are compared in Figures 5–13. As shown in Figures 5–13, the proposed method has good subjective performance in the sandstorm image enhancement field. This section presents the objective comparison of the enhanced images produced using the proposed method and state-of-the-art methods. To assess the performance, this paper uses three metrics, namely UIQM [27], NIQE [28], and FADE [29]. The UIQM [27] measure is used in the enhanced underwater image assessment area. Sandstorm and underwater images have similar features, such as color distortion. Underwater images have a blueish or greenish color distortion due to light attenuation. Additionally, sandstorm images have reddish or yellowish color degradation owing to color channel attenuation because of light being scattered by sand particles. Because underwater images have similar features to sandstorm images, for this reason, this paper assessed the enhanced sandstorm images using the UIQM measure [27]. The UIQM [27] measure indicates the image's sharpness, contrast, and colorfulness. If the score is high, then the enhanced image has high quality. The NIQE [28] measure indicates how natural an image is using the image's statistical features such as the Gaussian total variation. If the enhanced image has good quality, then the NIQE [28] has a low score. The FADE [29] score reflects how hazy the enhanced image is. If the score is lower, then the enhanced image is less hazy, and vice versa.

Tables 1–3 show the NIQE [28] scores for Figures 8 and 9. Table 1 shows the NIQE [28] scores for Figures 8 and 9. He et al.'s [1] method has a lower NIQE score than Gao et al.'s [13] method, although the enhanced images produced, using He et al.'s [1] method, have a color shift and ringing effect; however, the enhanced images are less hazy than those produced using Gao et al.'s method [13], and therefore, the NIQE score of He et al.'s [1] method is less than that of Gao et al.'s method [13]. Gao et al.'s method [13] has a higher NIQE score than that of Al Ameen's [9] method. Although the enhanced images produced using Gao et al.'s [13] method have no color shift, they are hazier than those produced using Al Ameen's method [9]. Therefore, the NIQE score of Gao et al.'s [13] method is higher than that of Al Ameen's method [9]. Meng et al.'s method [11] has a lower NIQE score than that of Shi et al.'s method [10], as although the enhanced images produced using Meng et al.'s [11] method have a distorted color, they are less hazy than those produced using Shi et al.'s method [12]. Shi et al.'s [12] method has a higher NIQE score than that of Ren et al.'s method [17], although the enhanced images produced using Ren et al.'s method [17] have a color shift. Lee's method [16] has a lower NIQE score than that of Shi et al.'s [10] method, although the enhanced images produced using the Lee method [16] have a bright region, because the enhanced images have no artificial color cast. Dhara et al.'s method [15] has a lower NIQE score than that of Gao et al.'s [13] method because the enhanced images produced using Dhara et al.'s method [15] are less hazy than those produced using Gao et al.'s [13] method. Shi et al.'s [10] method has a higher NIQE score than that of Meng et al.'s [11] method, although the enhanced images produced using Shi et al.'s method [10] have less color cast than those produced using Meng et al.'s [11] method, because the improved images produced using Meng et al.'s [11] method are less hazy than those produced using Shi et al.'s [10] method. Meanwhile, the enhanced images produced using the proposed method have a lower NIQE score than those produced using the existing methods because the enhanced images produced by the proposed method have no color shift or ringing effect.

Table 1. The comparison of NIQE scores [28] for Figures 8 and 9 (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	19.768	19.789	19.694	19.798	19.884	19.788	19.668	19.863	19.532	19.370
	20.271	20.163	16.766	19.470	20.658	19.293	20.343	19.377	15.064	13.787
	18.595	18.193	16.992	18.074	18.954	17.772	18.163	18.220	16.216	16.222
	19.800	19.884	19.849	19.948	19.871	19.747	19.539	19.568	18.425	16.931
	20.302	20.308	20.229	20.364	20.453	20.223	20.268	20.242	18.333	16.987
	19.778	19.751	19.451	19.799	19.871	19.610	19.699	19.780	19.497	19.339
	20.596	20.150	20.516	20.587	20.594	20.251	20.457	20.348	20.375	20.343
	20.043	20.073	19.853	20.151	20.476	19.949	20.263	20.171	18.119	16.591
	19.797	19.607	18.729	19.701	19.700	19.424	19.652	19.653	17.313	17.080
	19.624	19.410	19.723	19.865	19.825	19.550	19.674	19.542	17.260	16.001
AVG	19.857	19.733	19.180	19.776	20.029	19.561	19.773	19.676	18.013	17.265

Table 2. The comparison of NIQE scores [28] for Figures 10 and 11 (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	20.960	21.582	20.787	21.488	21.786	21.157	21.129	21.365	19.218	17.678
	21.703	21.532	20.755	21.361	22.343	20.834	21.485	21.198	16.797	15.275
	19.848	19.912	19.909	19.862	19.848	19.742	19.768	19.833	19.209	18.501
	19.027	18.811	18.976	19.038	18.994	18.675	19.219	18.448	15.548	15.184
	18.969	19.084	16.380	18.808	19.165	18.091	18.994	18.602	15.195	14.932
	20.302	20.893	20.981	20.617	21.403	20.343	19.812	21.580	18.852	19.316
	21.705	21.003	22.047	21.690	21.433	20.920	21.651	21.414	18.128	17.838
	22.250	22.290	22.041	21.398	21.140	20.395	21.801	21.568	15.854	16.064
	21.484	20.984	21.506	21.218	21.167	21.065	21.525	21.611	20.858	21.001
	20.113	19.624	20.777	19.988	20.209	19.661	19.649	20.099	18.609	17.822
AVG	20.636	20.572	20.416	20.547	20.749	20.088	20.503	20.572	17.827	17.361

Table 3. The comparison of NIQE scores [28] for Figures 12 and 13 (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	19.652	19.245	19.458	19.508	19.612	19.191	19.381	19.525	18.540	18.611
	20.393	20.098	20.712	20.443	20.427	20.376	20.307	20.277	19.796	19.487
	22.435	21.981	22.640	22.391	22.438	21.703	21.182	21.318	17.567	15.464
	22.483	21.274	24.697	21.605	21.418	21.044	21.008	21.088	17.432	16.564
	20.263	20.848	20.243	20.185	20.423	20.006	20.177	20.481	18.728	18.308
	22.044	21.743	20.567	20.872	22.614	19.089	20.906	21.390	16.553	16.324
	19.668	19.706	18.677	19.457	19.710	19.127	19.650	19.500	16.080	16.156
	22.450	21.379	22.899	22.892	23.057	22.365	22.441	22.525	21.755	21.450
	22.198	22.799	20.548	21.418	22.791	20.600	21.876	21.427	14.919	15.078
	22.114	21.898	21.205	21.481	22.507	20.981	21.429	22.779	18.391	16.662
AVG	21.370	21.097	21.165	21.025	21.500	20.448	20.836	21.031	17.976	17.410

Table 2 shows the NIQE [28] scores for Figures 10 and 11. He et al.'s method [1] has a lower NIQE score than that of the Gao et al. [13] method. Although the enhanced images produced using the He et al. [1] method have a color shift and ringing effect, because they are less hazy than those produced using Gao et al.'s [13] method, the NIQE score is lower than that of Gao et al.'s method [13]. Gao et al.'s [13] method has a higher NIQE score than that of Al Ameen's method, although the enhanced images produced using Al Ameen's method [9] have a color shift. Because the enhanced images produced by Al Ameen's method [9] are less hazy than those produced using Gao et al.'s method [13], this

is reflected in the NIQE score. The enhanced images produced using Meng et al.'s [11] method have a lower NIQE score than those produced using Shi et al.'s method [12]. Although the enhanced images using Meng et al.'s method [11] have a color shift, because the enhanced images produced using Meng et al.'s [11] method are less hazy than those produced using Shi et al.'s method [12], the NIQE score is lower than that of Shi et al.'s method [12]. The Shi et al. method [12] has a lower NIQE score than that of Ren et al.'s [17] method in some image. The enhanced images produced using Ren et al.'s [17] method have a lower NIQE score than those produced using Gao et al.'s method [13] in some image, although the enhanced images produced using the Ren et al. method [17] have a color shift, because the enhanced images are less hazy, and so the NIQE score is less than that of Gao et al.'s method [13] in some image. Lee's [16] method has a lower NIQE score than that of Gao et al.'s [13] method because the enhanced images produced using Lee's method [16] are less hazy than those produced using Gao et al.'s method [13]. Shi et al.'s [10] method has a higher NIQE score than that of Meng et al.'s method [11] in some image, although the enhanced images produced using Shi et al.'s [10] method have a less artificial color cast than those produced using Meng et al.'s method [11], because the enhanced images are hazier than those produced using Meng et al.'s method [11]. Meanwhile, the enhanced images produced using the proposed method have a lower NIQE score than those produced using other methods, because the enhanced images produced using the proposed method have no color shift and ringing effect.

Table 3 shows the NIQE scores for Figures 12 and 13. He et al.'s method [1] has a lower NIQE score than that of the Gao et al. method [13] because the enhanced images produced, using He et al.'s method [1], are less hazy than those produced using the Gao et al. method [13]. The Al Ameen method [9] has a higher NIQE score than that of the Meng et al. [11] method in some image, although the enhanced images produced using the Meng et al. method [11] have a color shift. The Meng et al. method [11] has a lower NIQE score than that of the Shi et al. [12] method in some image, although the enhanced images using the Meng et al. [11] method have a color shift, because the enhanced images produced using the Meng et al. [11] method are less hazy than those produced using the Shi et al. method [12]. The enhanced images produced using the Ren et al. method [17] have a lower NIQE score than those produced using the Gao et al. method [13] because the enhanced images produced using the Gao et al. method [13] are hazier than those produced using the Ren et al. method [17]. Lee's method [16] has a lower NIQE score than that of the Shi et al. method [10]. Shi et al.'s method [10] has a lower NIQE score than that of the Gao et al. method [13] because the enhanced images produced using the Shi et al. method [10] are less hazy than those produced using the Gao et al. [13] method. Dhara et al.'s method [15] has a lower NIQE score than that of the Al Ameen method [9] because the enhanced images produced using Dhara et al.'s [15] method have less color casting than those produced using the Al Ameen method [9]. Meanwhile, the enhanced images produced using the proposed method have a lower NIQE score than those of other methods because the enhanced images produced by the proposed method have no color shift and ringing effect.

As shown in Tables 1–3, to enhance sandstorm images naturally, image color balancing and hazy components are considered.

Table 4 shows the average NIQE [28] scores for Figures 8–13 and the DAWN dataset [25] and WEAPD [26]. As shown in Table 4, although the existing dehazing methods cause a color shift, because the enhanced images are less hazy than those produced using the sandstorm image enhancement methods, these methods have a lower NIQE score than those of the existing sandstorm image enhancement methods.

Table 4. The comparison of average NIQE scores [28] for Figures 8–13, the DAWN dataset [25], and WEAPD [26] (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
AVG(30)	20.621	20.467	20.254	20.449	20.759	20.032	20.371	20.426	17.939	17.346
AVG(323)	19.833	19.803	19.698	19.892	19.931	19.583	19.714	19.633	17.971	17.047
AVG(692)	21.062	21.322	20.998	21.198	21.290	20.899	20.957	20.647	18.684	17.922

Tables 5–7 show the UIQM [27] scores for Figures 8–13. If the image is enhanced well, the UIQM score is high.

Table 5. The comparison of UIQM scores [27] for Figures 8 and 9 (a higher score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	0.366	0.798	0.527	0.518	0.326	0.816	0.660	0.428	0.752	0.865
	1.037	1.176	1.786	1.358	0.880	1.424	1.031	1.149	1.728	1.798
	1.201	1.313	1.478	1.337	0.961	1.494	1.293	1.233	1.625	1.664
	0.843	1.009	0.823	1.058	0.727	1.196	0.833	0.838	1.398	1.499
	0.696	0.919	0.767	0.696	0.555	1.005	0.668	0.737	1.415	1.554
	0.429	0.845	0.721	0.569	0.351	0.887	0.623	0.480	0.926	1.119
	0.358	0.798	0.542	0.482	0.328	0.961	0.633	0.462	0.737	0.967
	0.867	0.661	0.870	0.732	0.561	1.116	0.792	0.761	1.434	1.509
	0.573	0.582	0.908	0.779	0.603	1.047	0.702	0.785	1.483	1.578
	0.826	0.800	0.795	0.749	0.635	0.972	0.815	0.864	1.394	1.349
AVG	0.720	0.890	0.922	0.828	0.593	1.092	0.805	0.774	1.289	1.390

Table 6. The comparison of UIQM scores [27] for Figures 10 and 11 (a higher score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	0.853	0.745	0.878	0.735	0.565	1.063	0.818	0.765	1.357	1.438
	0.980	1.020	1.350	1.222	0.756	1.333	1.019	1.045	1.598	1.660
	0.481	0.712	0.598	0.556	0.403	1.008	0.586	0.532	1.056	1.141
	0.919	0.929	0.982	0.977	0.848	1.249	0.896	1.065	1.764	1.942
	1.339	1.370	1.628	1.677	1.202	1.645	1.243	1.409	1.863	1.689
	0.984	1.138	1.032	0.885	0.640	1.293	1.195	0.900	1.683	2.021
	0.652	1.069	0.916	0.818	0.694	1.121	1.164	0.854	1.380	1.565
	0.887	1.155	1.098	1.159	0.921	1.399	0.876	1.047	1.687	1.809
	1.081	1.281	1.231	1.187	0.971	1.519	0.996	1.177	2.164	2.258
	0.621	1.143	0.806	0.745	0.558	1.189	1.198	0.736	1.428	1.651
AVG	0.880	1.056	1.052	0.996	0.756	1.282	0.999	0.953	1.560	1.717

Table 5 shows the UIQM scores for Figures 8 and 9. He et al.'s method [1] has a higher UIQM score than that of Gao et al.'s [13] method, although the enhanced images produced using He et al.'s method [1] have a color shift, because the enhanced images are less hazy than those produced using the Gao et al. method [13]. The enhanced images produced using the Al Ameen method [9] have a higher UIQM score than those produced using the Gao et al. method [13], as although the enhanced images produced using the Al Ameen method [9] have color distortion, the hazy effect in the enhanced images is less pronounced than that seen in those produced using the Gao et al. method [13]. Meng et al.'s method [11] has a higher UIQM score than that of the Al Ameen method [9] in some image, as although the enhanced images have a color shift, the hazy effect is less pronounced than that seen in those produced using the Al Ameen method [9]. Shi et al.'s method [12] has a lower

UIQM score than that of Ren et al.'s method [17] in some image, although the enhanced images produced using Ren et al.'s method [17] have a color shift, because the hazy effect is less pronounced than that seen in those produced using Shi et al.'s method [12]. Lee's method [16] has a higher UIQM score than that of Gao et al.'s [13] method because the enhanced images produced using Lee's method [16] are less hazy than those produced using Gao et al.'s [13] method. Shi et al.'s method [10] has a lower UIQM score than that of Ren et al.'s [17] method in some image, although the enhanced images using Ren et al.'s method [17] have an artificial color because the hazy effect is less pronounced than that seen in those produced using Shi et al.'s method [10]. Dhara et al.'s method [15] has a higher UIQM score than that of Meng et al.'s [11] method because Dhara et al.'s method [15] produces enhanced images with a less pronounced color cast than those produced using Meng et al.'s method [11]. Meanwhile, the enhanced images produced using the proposed method have a higher UIQM score than those produced using the other methods because the enhanced images produced by the proposed method have no color shift or ringing effect.

Table 7. The comparison of UIQM scores [27] for Figures 12 and 13 (a higher score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	0.325	0.849	0.812	0.629	0.453	1.037	0.906	0.520	1.401	1.406
	0.517	0.996	0.435	0.805	0.666	1.177	0.796	0.569	1.527	1.909
	1.265	1.263	1.155	1.109	0.945	1.443	1.194	1.178	1.763	1.766
	1.258	1.221	1.260	1.118	0.959	1.476	1.103	1.205	1.875	1.898
	0.900	0.661	0.860	0.696	0.570	0.932	0.771	0.714	1.235	1.399
	0.932	1.151	1.068	1.258	0.981	1.425	0.971	1.143	1.696	1.846
	0.964	1.069	1.317	1.210	0.919	1.504	1.014	1.189	1.827	2.005
	0.320	0.882	0.867	0.590	0.315	0.850	0.635	0.448	0.828	1.063
	1.058	1.132	1.276	1.188	1.006	1.432	1.007	1.201	1.907	1.977
	1.018	1.361	1.144	1.197	0.983	1.353	0.931	1.136	1.665	1.773
AVG	0.856	1.059	1.019	0.980	0.780	1.263	0.933	0.930	1.572	1.704

Table 6 shows the UIQM scores [27] for Figures 10 and 11. He et al.'s [1] method has a higher UIQM score than that of Geo et al.'s method [13], although the enhanced images produced, using He et al.'s [1] method, have color distortion because the hazy effect is less produced than that seen in those produced using Gao et al.'s method [13]. Al Ameen's method [9] has a higher UIQM score than that of Meng et al.'s method [11] in some image because the UIQM score reflects the image's colorfulness, contrast, and sharpness. Meng et al.'s method [11] has a lower UIQM score than that of Al Ameen's method [9] in some image because the enhanced images produced using Meng et al.'s method [11] have a color shift. Shi et al.'s method [12] has a higher UIQM score than that of Meng et al.'s method [11] in some image because the enhanced images using Shi et al.'s method [12] have less color shift than those produced using Meng et al.'s method [11]. Ren et al.'s method [17] has a lower UIQM score than that of Shi et al.'s method [12] in some image because the UIQM score reflects the image's colorfulness, and the enhanced images produced using Ren et al.'s [17] method demonstrate a color shift. Lee's method [16] has a higher UIQM score than that of Dhara et al.'s [15] method because the enhanced images produced using Lee's method [16] have less color cast. The Dhara et al. method [15] has a higher UIQM score than that of Meng et al.'s [11] method because the enhanced images produced using Dhara et al.'s method [15] have less color cast. Shi et al.'s method [10] has a higher UIQM score than that of Gao et al.'s [13] method because the enhanced images using Shi et al.'s method [10] are less hazy than those produced using Gao et al.'s method [13]. Meanwhile, the enhanced images produced by the proposed method have a higher UIQM score than those produced using other methods because the enhanced images produced using the proposed method have no color distortion or ringing effect.

Table 7 shows the UIQM [27] scores for Figures 12 and 13. He et al.'s [1] method has a higher UIQM score than that of Gao et al.'s method [13] because the hazy effect in the enhanced images by He et al. method [1] is less pronounced than that seen in those produced using the Gao et al. method [13]. Al Ameen's method [9] has a higher UIQM score than that of the He et al. [1] and Gao et al. [13] methods. Although the enhanced images produced using the Al Ameen method [9] have color distortion, the hazy effect is less pronounced than that seen in the enhanced images produced using the Gao et al. method [13], and the color distortion is less than that seen in those produced using the He et al. method [1]. Meng et al.'s method [11] has a higher UIQM score than that of the Gao et al. method [13], although the enhanced images produced using Meng et al.'s method [11] have color degradation, because the enhanced images are less hazy than those produced using the Gao et al. method [13]. The enhanced images produced using Shi et al.'s method [12] have a higher UIQM score than those produced using Meng et al.'s method [11] in some image because the enhanced images produced by means of the Shi et al. method [12] have less color shift than those produced using the Meng et al. method [11]. The enhanced images produced by means of the Ren et al. method [17] have a higher UIQM score than those produced using the Gao et al. method [13], although the enhanced images produced using the Ren et al. method [17] have color distortion, because the enhanced images are less hazy than those produced using the Gao et al. method [13]. Lee's method [16] has a higher UIQM score than that of the Gao et al. method [13] because the enhanced images produced using Lee's method [16] are less hazy. Shi et al.'s method [10] has a lower UIQM score than that of Meng et al.'s method [11], although the enhanced images produced using Shi et al.'s method [10] have less color cast than those produced using the Meng et al. method [11], because they are hazier. Dhara et al.'s method [15] has a higher UIQM score than that of Meng et al.'s method [11] because the enhanced images using Dhara et al.'s [15] method have less color cast. Meanwhile, the enhanced images provided by the proposed method have a higher UIQM score than that of the other methods because the enhanced images produced using the proposed method have no color shift or ringing effect, and the enhanced images seem natural.

As shown in Tables 5–7, to enhance a sandstorm image naturally, the image's colorfulness should be considered.

Table 8 shows the average UIQM scores [27] for Figures 8–13, the DAWN dataset [25], and WEAPD [26]. The enhanced images using the Gao et al. method [13] have a lower UIQM score than that of the dehazing methods, although the enhanced images have no color distortion. Meanwhile, the proposed method has a higher UIQM score than that of the other methods because the enhanced images produced using the proposed method have no color distortion or ringing effect.

Table 8. The comparison of UIQM scores [27] for Figures 8–13, the DAWN dataset [25], and WEAPD [26] (a higher score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
AVG(30)	0.818	1.002	0.998	0.935	0.709	1.212	0.912	0.886	1.487	1.604
AVG(323)	0.780	0.938	0.928	0.840	0.671	1.184	0.870	0.816	1.426	1.581
AVG(692)	1.118	1.271	1.222	1.207	0.995	1.405	1.074	1.166	1.827	1.929

As shown in Table 8, to enhance a degraded sandstorm image naturally, the image's colorfulness, sharpness, and contrast should be considered.

Tables 9–12 show the FADE [29] score for Figures 8–13.

Table 9. The comparison of FADE scores [29] for Figures 8 and 9 (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	2.122	2.780	1.215	1.824	8.682	1.731	3.466	6.768	2.483	1.961
	0.968	0.907	0.370	0.749	2.153	0.719	1.305	1.336	0.362	0.298
	0.412	0.846	0.294	0.374	1.384	0.469	0.684	0.831	0.207	0.159
	0.827	1.572	0.897	0.743	2.740	0.914	1.610	2.092	0.643	0.452
	1.227	2.277	0.747	1.419	4.279	1.134	1.837	3.576	0.983	0.699
	1.575	2.530	0.879	1.486	7.286	1.479	2.841	5.170	1.380	1.104
	1.878	1.286	1.002	1.437	6.919	1.732	3.313	5.737	2.150	1.494
	1.452	3.146	0.905	1.937	6.270	1.650	2.884	4.570	1.961	1.051
	0.707	0.629	0.464	0.561	2.701	0.831	1.269	2.500	0.472	0.362
	0.645	2.231	0.666	0.936	3.612	0.925	1.467	2.312	1.189	0.847
AVG	1.181	1.820	0.744	1.147	4.603	1.158	2.068	3.489	1.183	0.843

Table 10. The comparison of FADE scores [29] for Figures 10 and 11 (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	1.712	4.684	1.056	2.979	5.889	1.333	1.922	3.624	1.103	0.969
	1.215	1.475	0.465	0.944	2.657	0.714	1.149	1.751	0.498	0.424
	1.695	1.721	1.101	1.619	8.198	1.682	3.748	5.921	2.089	1.469
	0.876	0.997	0.618	0.709	3.140	0.997	1.371	2.182	0.551	0.359
	0.442	0.479	0.268	0.347	0.925	0.411	0.746	0.742	0.194	0.174
	0.626	1.961	0.614	0.915	3.654	0.709	0.745	2.002	0.342	0.316
	0.769	0.584	0.420	0.594	3.256	0.851	0.844	2.240	0.687	0.356
	0.621	0.490	0.397	0.501	2.489	0.748	1.857	1.379	0.408	0.236
	0.682	0.539	0.445	0.554	1.602	0.434	1.288	1.053	0.215	0.192
	0.749	1.023	0.544	0.666	3.926	0.716	0.811	2.715	0.547	0.399
AVG	0.939	1.395	0.593	0.983	3.574	0.860	1.448	2.361	0.663	0.489

Table 11. The comparison of FADE scores [29] for Figures 12 and 13 (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
	0.927	0.508	0.511	0.641	4.862	1.249	1.802	4.734	0.691	0.645
	0.989	1.455	0.888	0.887	4.583	2.712	3.485	5.547	0.847	0.745
	0.400	0.942	0.477	0.588	1.607	0.534	0.727	1.065	0.190	0.169
	0.447	0.990	0.427	0.640	1.667	0.520	0.983	1.071	0.181	0.199
	0.698	4.744	0.576	1.483	5.743	1.124	1.625	3.739	1.252	0.663
	0.663	0.492	0.388	0.480	1.933	0.808	1.323	1.273	0.299	0.218
	0.506	0.466	0.313	0.405	2.240	0.541	1.046	1.216	0.321	0.224
	2.268	1.134	0.902	1.710	8.752	1.395	3.718	6.046	1.960	1.145
	0.534	0.437	0.322	0.431	1.658	0.551	0.986	0.913	0.205	0.189
	0.648	0.494	0.478	0.517	1.581	0.488	1.178	1.038	0.222	0.160
AVG	0.808	1.166	0.528	0.778	3.463	0.992	1.687	2.664	0.617	0.436

Table 12. The comparison of average FADE scores [29] for Figures 8–13, the DAWN dataset [25], and WEAPD [26] (a lower score represents better enhancement).

	[1]	[9]	[11]	[17]	[13]	[15]	[12]	[10]	[16]	Proposed Method
AVG(30)	0.976	1.461	0.622	0.969	3.880	1.003	1.734	2.838	0.821	0.589
AVG(323)	1.368	1.675	0.798	1.348	3.648	0.913	1.684	2.881	0.794	0.596
AVG(692)	0.670	0.989	0.531	0.704	1.983	0.665	1.085	1.482	0.383	0.294

Table 9 shows the FADE [29] score for Figures 8 and 9. He et al.'s method has a higher FADE score than that of Meng et al.'s method [11] because the Meng et al. method [11] has a more suitable refined transmission map. Meng et al.'s method [11] has a lower FADE score among the existing dehazing methods because this method has a refined transmission map. Ren et al.'s method [17] has a higher FADE score than that of Meng et al.'s method [11] because the enhanced images are hazier. Gao et al.'s method [13] has a higher FADE score than that of other methods because the enhanced images are too hazy. Shi et al.'s method [12] has a lower FADE score than that of Gao et al.'s method [13] because the enhanced images are less hazy. Lee et al.'s method [16] has a lower FADE score than that of Shi et al.'s method [12] because the enhanced images are less hazy. Dhara et al.'s method [15] has a lower FADE score than that of Gao et al.'s method [13] because the enhanced images produced using the Gao et al. method [13] are hazier. Shi et al.'s method [10] has a higher FADE score than that of Shi et al.'s method [12] because the enhanced images are hazier. Al Ameen's method [9] has a lower FADE score than that of Gao et al.'s method [13], although the enhanced images have a color cast, because the hazy effect is less pronounced than that seen in those produced using the Gao et al. method [13]. Meanwhile, the proposed method has a lower FADE score than that of the other methods because the enhanced images produced using the proposed method have no color cast and the enhanced images are less hazy.

Table 10 shows the FADE [29] scores for Figures 10 and 11. He et al. [1] has a lower FADE score than that of Gao et al.'s method [13], although the enhanced images produced using the He et al. method [1] have a color cast, because the enhanced images produced using the He et al. method [1] are less hazy than those produced using the Gao et al. method [13]. Meng et al.'s method [11] has a lower FADE score than that of the He et al. method [1] because the enhanced images are less hazy. Ren et al.'s method [17] has a lower FADE score than that of the Gao et al. method [13], although the enhanced images produced using the Ren et al. method [17] have a color cast, because the hazy effect is less pronounced. Gao et al.'s method [13] has a high FADE score, although the enhanced images have no color cast, because the hazy effect is more pronounced. Lee's method [16] has a lower FADE score than that of the Gao et al. method [13] because the enhanced images are less hazy. Shi et al.'s [12] method has a lower FADE score than that of the Gao et al. method [13] because the enhanced images are less hazy. Shi et al.'s method [10] has a higher FADE score than that of the other dehazing methods, although the enhanced images have no color cast, because the enhanced images are hazier. Dhara et al.'s method [15] has a higher FADE score than that of Meng et al.'s method [11] because the enhanced images are hazier. However, the enhanced images using the proposed method have a lower FADE score than that of the other methods because the enhanced images produced using the proposed method have no color cast and are less hazy.

Table 11 shows the FADE [29] scores for Figures 12 and 13. He et al.'s method [1] has a higher FADE score than that of Meng et al.'s method [11] because the Meng et al. method [11] has a refined transmission map and it provides a less hazy image. Meng et al.'s method [11] has a lower FADE score than that of the Ren et al. method [17] because the enhanced images using Meng et al.'s method [11] are less hazy. Ren et al.'s method [17] has a lower FADE score than that of He et al.'s method [1] because the enhanced images are less hazy. Gao et al.'s method [13] has a higher FADE score than that of the other methods, although the enhanced images have no color cast because the enhanced images have a hazy effect. Al Ameen's method [9] has a lower FADE score than that of Gao et al.'s method [13], although the enhanced images have a color cast, because the hazy effect is less pronounced. Lee's method [16] has a lower FADE score than that of the Gao et al. method [13] because the enhanced images are less hazy. Shi et al.'s method [12] has a higher FADE score than that of Al Ameen's method [9], although the enhanced images have no color cast, because the hazy effect is more pronounced. Shi et al.'s method [10] has a higher FADE score than that of Meng et al.'s method [11], although the enhanced images have no color cast, because the hazy effect is more pronounced. Dhara et al.'s method [15] has a lower FADE score than

that of Gao et al.'s method [13], although the enhanced images have a color cast, because the enhanced images are less hazy. Meanwhile, the enhanced images produced using the proposed method have a lower FADE score than that of the other methods because the enhanced images produced using the proposed method have no color cast and are less hazy.

Table 12 shows the average FADE [29] score for Figures 8–13, the DAWN dataset [25], and WEAPD [26]. The existing dehazing methods have lower FADE scores than those of the sandstorm image enhancement methods, although the enhanced images have a color cast. Meanwhile, the enhanced images produced using the proposed method have a lower FADE score than that of other methods because the enhanced images have no color cast and are less hazy.

As shown in Figures 8–13 and Tables 1–12, the performance of the proposed method in the sandstorm image enhancement field is superior compared to that of other existing methods.

The experimental setup consisted of an Intel® Core™ i7-8700 CPU @ 3.20GHz, 32 GB RAM, GeForce GTX 1650 4 GB GPU.

4. Conclusions

Color-degraded sandstorm images caused by the scattering of light by sand particles are compensated using the proposed method utilizing the recombined singular value based on rank. Additionally, the corrected image seems hazy. Therefore, to enhance the compensated image, this paper apply the dehazing procedure. Because the existing dehazing methods have artificial effects such as the ringing effect, this paper uses the guided image filter with the singular value ratio between the input and balanced images. The enhanced images using the proposed method have no color shift or ringing effect. The contribution of this paper to the sandstorm image enhancement field is that this paper enhances the degraded color channel using the image's unique features such as the singular value, and in this way, the enhanced images can be used adaptively in various fields of degraded image enhancement, not only of sandstorm images but also any distorted image. Moreover, a strong point of this paper is that it can be applied to extensive image enhancement areas. However, its weak point is that in cases of densely dusty images, the dehazing effect is limited. To enhance sandstorm images more naturally and adaptively, future research should perform the estimation of the image adaptive dehazing procedure with transmission maps. Because the existing transmission map struggles to reflect the image's features, such as bright regions and distance, the key point of future research is estimating the image adaptive transmission map, which includes distinguishing dark and bright regions in addition to distance to enhance the image naturally. Moreover, the image adaptive color balancing procedure could be performed to enhance color-degraded images with a reddish, yellowish, or bluish color cast and make these images seem natural.

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