



An Evaluation of a Novel Technique to Suppress Corruptive Artifacts for Example-Based Color Transfer Framework

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ABSTRACT: Color manipulation is one of the most common tasks in image editing. How to transfer the colors of the given reference to the target effectively is a challenging problem and is significant in color transfer. Automatic color appearance adjustment is still of high demand, owing to the inherent difficulties to handle complex structures ubiquitous in natural images. The main purpose of the project is to develop a novel technique for Example-based color transfer framework which aims at copying the color appearance from a given “example” to a target image.

A novel technique to suppress corruptive artifacts for example based color transfer framework is proposed a technique which colorize the image, detect the corruptive artifact and suppress it. It also preserves edges. Time performance is improved in this system. In our system grain suppression, color fidelity and information loss are satisfied by reducing noise, increasing color fidelity and reducing information loss. We have presented a framework that allows a user to easily and quickly create a color transformed image and grain effect in the process of transfer, but also achieves the effect of detail preserving or enhancing.

KEYWORDS: Color Transfer, Suppression, Corruptive Artifacts, Example-based, Color fidelity, Noise.

I. INTRODUCTION

Color manipulation is one of the most common tasks in image editing. While artists resort to photo editing tools to manually adjust color appearance, automatic color appearance adjustment is still of high demand, owing to the inherent difficulties to handle complex structures ubiquitous in natural images. Arguably, example-based color transfer, which aims at copying the color appearance from a given “example” to a target grayscale or color image, is the most effective way to tackle the problem. Taking Fig. 1 as an example, due to the big difference in the intensity distribution between the reference and the target, an unsatisfactory transferred result was produced, with remarkable artifacts as follows[1].

Color distortion: Some disharmonious or unexpected colors appear which are not included in the reference image.

Grain effect: A phenomenon appears due to enhancing the noise level of the picture under the stretched mapping. Commonly, it looks like some noises or irregular blocks.

Loss of details: The fine-level details in the target image are missed after the color transfer. Ideally, color transfer between reference and target images should satisfy the following goals.

Color fidelity: The color distribution of the target should be close to that of the reference image.

Grain suppression: No visual artifacts (grain/blocky artifacts) should be generated in the target image.

Detail preservation: Details in the original target should be preserved after the transfer.

International Journal of Innovative Research in Computer and Communication Engineering

(An ISO 3297: 2007 Certified Organization)

Vol. 3, Issue 9, September 2015

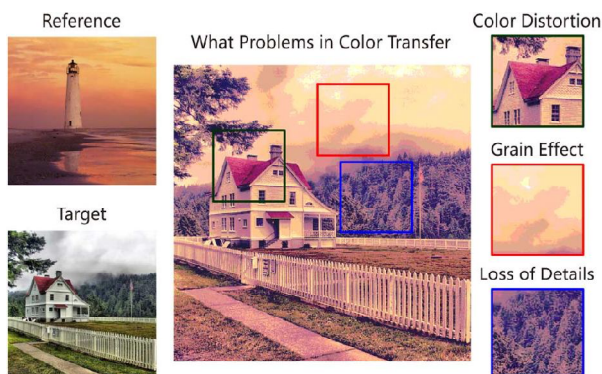


Fig..1 Grain effect, color distortion and loss of details appear in the color transfer method

Color Transfer

Color transfer is an image processing technique which can produce a new image combining one source image's contents with another image's color style. This article describes a method for a more general form of color correction that borrows one image's color characteristics from another. We can imagine many methods for applying the colors of one image to another. Our goal is to do so with a simple algorithm, and our core strategy is to choose a suitable color space and then to apply simple operations there. When a typical three channel image is represented in any of the most well-known color spaces, there will be correlations between the different channels' values. For example, in RGB space, most pixels will have large values for the red and green channel if the blue channel is large. This implies that if we want to change the appearance of a pixel's color in a coherent way, we must modify all color channels in tandem. This complicates any color modification process. What we want is an orthogonal color space without correlations between the axes. Color can be added to greyscale images in order to increase the visual appeal of images such as old black and white photos, classic movies or scientific illustrations. Color transfer refers to the category of methods designed to change the color appearance of an image according to the color content of another image. The group of color transfer approaches contains two major categories of colorizing and recoloring. The Color transfer of the image is shown in the following figure. In this the reference color is transferred to the target image without changing the objects of target image.



Fig.2 Color transfer Example

When we attempt to convert a greyscale source into a colored representation using the color information available in a reference image, the method is called colorizing. On the other hand, recoloring refers to the case where we wish to change the color appearance of a colored image to pretend another image's color content [21]

II. RELATED WORK

A novel color transfer framework is used to achieve a unified corruptive artifacts suppression, which is specified in grain suppression, color fidelity and detail manipulation. How to transfer the colors of the given reference to the target effectively is a challenging problem and is significant in color transfer.



International Journal of Innovative Research in Computer and Communication Engineering

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Rapid development has been witnessed in the last decade in the field of color transfer. Representative approaches include classical histogram matching, statistical transfer, N - dimensional probability density function transfer, gradient-preserving transfer non-rigid dense correspondence transfer, progressive transfer, to list a few. Although these approaches are effective in transferring the color information, they would occasionally produce visual artifacts, owing primarily to the contradictive roles of color distribution preservation and image content distribution. In 2001 Erik Reinhard et al[26] provided method for a more general form of color correction that borrows one image's color characteristics from another. In 2002 Tomihisa Welsh et al.[25] has been introduced a general technique for "colorizing" grayscale images by transferring color between a source, color image and a destination, grayscale image. In 2004 G. Petschnigg et al.[24] introduced a Joint bilateral filter (JBF) which is the first guided edge-preserving smoothing approach. The JBF exploits the pixel intensity of the reference which is correlated to the target to improve the filtering effect. However, like the bilateral filter (BLF), JBF cannot avoid the halo artifact and gradient reversal problem. The grain effect can be treated as a special type of noises and it would be removed by linear smoothing. Although the linear smoothing can remove the grains, the over-blurring would destroy the original image details and lower the sharpness of edges. In 2007 Chang *et al.*[19] proposed a color category-based approach that categorized each pixel as one of the basic categories to prevent from the grain effect. Then a convex hull was generated in $L^a\beta$ color space for each category of the pixel set, and the color transformation was applied with each pair of convex hull of the same category. In 2007 for the color distortion Tai *et al.*[20] proposed a modified EM algorithm to segment probabilistically the input images and construct Gaussian Mixture Models (GMMs) for them, and the relationship was constructed by each Gaussian component pairs between the target and the reference under Reinhard's approach. In 2010 Dong et al[14] proposed a dominant color idea for color transfer. When the amount of dominant colors of the target was consistent with that of the reference, the color of the reference would be transferred to obtain a satisfactory result. However, when the amount of dominant colors was not balanced, the unsatisfactory result would be produced. Dong's proposed an improved approach distribution-aware conception to consider the spatial color distribution in the reference image. In 2011[11] Wang *et al.* developed the learning based color transfer methods to train out the proper color mapping relationship. Also tone and color adjustment rules as mappings has been defined, and proposed to approximate complicated spatially varying nonlinear mappings in a piecewise manner.

In 2012 Mrs. Smriti Kumar et.al[8] presented a general technique for "colorizing" grayscale images by transferring color between a source or color image and a destination or target or grayscale image by matching luminance and texture information between the images. This technique of the gray image coloring uses a very simple algorithm that chooses a decorrelated color space and then applies simple operations there. In 2013 Fuzhang Wu, et al.[6] presented a novel content-based method for transferring the color patterns between images. Unlike previous methods that rely on image color statistics, our method puts an emphasis on high-level scene content analysis. This method can also be used to re-color video clips with spatially-varied color effects. In 2013 Rashmi et. al.[4] presented different edge detection method. Different edge detection techniques as Prewitt, Robert, Sobel, Marr Hildrith and Canny operators. On comparing them conclusion is that canny edge detector performs better than all other edge detectors on various aspects such as it is adaptive in nature, performs better for noisy image, gives sharp edges, low probability of detecting false edges etc.

In 2014 Zhuo Su, et al[1] proposed a novel unified color transfer framework with corruptive artifacts suppression, which performs iterative probabilistic color mapping with self-learning filtering scheme and multiscale detail manipulation scheme in minimizing the normalized Kullback-Leibler distance.

III. PROPOSED ALGORITHM

A. Dataset Collection:

In this module dataset of various images would be found out for future evaluation. As it is example-based framework so we can consider any image. Standard dataset is not required.

B. Color Transfer and Detection of Corruptive Artifacts:

Color transfer is nothing but the transformation of color. In this process color from the reference image is transferred to the target image without changing the objects. In this module, we develop an algorithm to check presence of corruptive artifacts, and display the location where the artifacts are corrupted.

International Journal of Innovative Research in Computer and Communication Engineering

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C. Suppression of corruptive artifact and Modification of Example-based color transfer algorithm:

This module would consist of proper color transfer algorithm to suppress corruptive artifacts by using patch based method from the points where they are detected. And we have to achieve grain suppression, color fidelity and detail manipulation by providing novel technique. Time performance would be reduced. And also edge preservation is also find out.

D. Result Evaluation and Optimization:

In this module, result would be evaluated and optimized if required. Result would be in the form of delay, and accuracy of detection.

IV. IMPLEMENTATION

One of the most common tasks in image processing is to alter an image's color. Grain suppression, color fidelity and information loss these are the important parameters to be satisfied. A novel technique has proposed to detect and suppress corruptive artifacts by improving time performance. The implemented methodology has to provide the color transformation and suppression of corruptive artifacts by improving Time performance. The project has gone through the following phases:

A. Colorization of Image

In the color transfer stage, Ycbr technique is used to transfer the color. In the color transfer method the background effects of reference image transfer to the target image. For colorization purpose we convert our source image into a decorrelated color space and en find the minimum difference between pallet and gray image. After finding minimum difference we replace gray (x,y) with pallet (x,y). And after RGB conversion colorizing image is obtained. In the Gray image and Color image is given as a input. Colorization is shown in the fig.4.



Fig.3 Input Gray Image and Color Image



Fig.4 Colorized Image

B. Detection Of Corruptive Artifacts

In the color transfer technique the color is transferred from reference image to the target image. Due to the big difference in the intensity distribution between the reference and the target, an unsatisfactory transferred result was produced, with remarkable artifacts. Artifacts are various types like grain suppression, color fidelity and detail preservation. So after colorization of image detection of artifacts is carried out by applying Laplacian of Gaussian and canny edge detection. To satisfy these requirements Canny used the calculus of variations – a technique which finds the function which optimizes a given functional. The optimal function in Canny's detector is described by the sum of four exponential terms, but it can be approximated by the first derivative of a Gaussian.

Laplacian of Gaussian

Laplacian filters are derivative filters used to find areas of rapid change (edges) in images. Since derivative filters are very sensitive to noise, it is common to smooth the image (e.g., using a Gaussian filter) before applying the Laplacian. This two-step process is call the Laplacian of Gaussian (LoG) operation[33].

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To include a smoothing Gaussian filter, combine the Laplacian and Gaussian functions to obtain a single equation

$$LoG(x, y) = -\frac{1}{\pi\sigma^4} \left[1 - \frac{x^2+y^2}{2\sigma^2} \right] e^{-\frac{x^2+y^2}{2\sigma^2}}$$

where x is the distance from the origin in the horizontal axis, y is the distance from the origin in the vertical axis, and σ is the standard deviation of the Gaussian distribution. Figure 5 shows the output for the detection of corruptive artifact.



Fig. 5 Detection of Corruptive Artifact

C. Suppression of corruptive artifacts

After the detection of artifacts from colorizing image in the form of edges suppression of artifact is carried out. Patch Based method is used for the suppression of corruptive artifacts. Restoring damaged regions of an image and removing undesired objects are termed as image inpainting. The basic idea is to fill in the lost or broken parts of an image using the surrounding information in such a way that the final restored result appears to be natural to a not familiar observer. These specified regions are called inpainting domain or target regions and the undamaged parts, in which information is used to repair the target region, are called source regions. In the example-based algorithms, the unknown blocks of target region are inpainted by the most similar blocks extracted from the source region, using the available information.

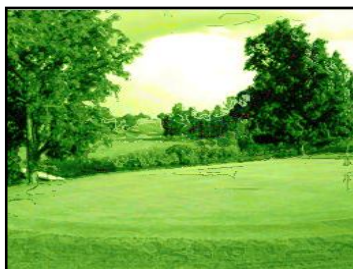


Fig.6 Suppression of Corruptive Artifacts

For the suppression patch based method is used. We divide the image into patches. Next we go to each patch and find the best matching other patch for replacement. Check if noise is present in the patch. Noise is present so find the next patch. Check if a noise free pixel is found and previous patch is noisy, so check the next patch. Next patch also has a noisy pixel in the same area, so we cannot use it. In this way suppression using patch based method is carried out. Suppression of corruptive artifact is shown in fig.6

D. Improvement of algorithm to apply on color image

A novel technique to suppress corruptive artifacts for example based color transfer framework is proposed a technique which colorize the image, detect the corruptive artifact and suppress it. It also preserves edges. Time performance is improved in this system. In our system grain suppression, color fidelity and information loss are satisfied by reducing noise, increasing color fidelity and reducing information loss. This modification is implemented.

V. RESULTS ANALYSIS

In this section some of the parameters of our project are evaluated and comparisons with parameters of various color transfer method are performed to check the efficiency and accuracy of our novel technique.

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Kullback-Leibler Distance Comparison

The Kullback-Leibler distance (K-L) can measure the similarity between two completely determined probability distributions. Here, we apply it to measure the difference between the reference and transferred result in color transfer.

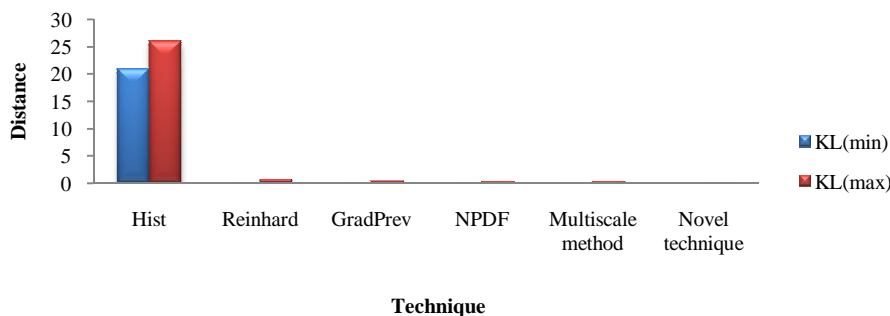


Fig.7 Graphical representation of KL distance

Here we have the comparison of KL distance for different method with our method. KL distance should be minimum as shown in figure 7.

Time Performance Comparison

We adopted the experimental runtime to measure our approach, and compared it with other approaches. We selected different sizes as the tested samples. Histogram matching and Reinhard's approach had an efficient runtime response. However, as mentioned above, both of them were hard to obtain a satisfactory visual performance. Xiao's gradient-preserving approach and Pitié's N-dimensional PDF approach required too much time, because both of them needed to solve a large-scale optimization equation. Especially, if the size was over large, these two approaches would break down. By contrast, our approach had a sound time response and was better than previous approaches in usability. Time is calculated by taking difference of end time and start time.

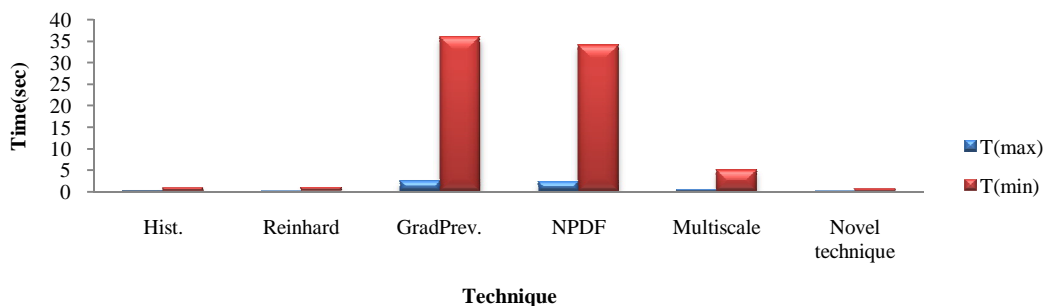


Fig.8 Graphical representation of Time

Time performance of proposed methodology is minimum as compared to Traditional method as shown in fig. 8

VI. CONCLUSION AND FUTURE WORK

Proposed method is applicable to images taken under a variety of conditions, is faster than our previous method. In our system a novel technique is used to colorize the image. During colorization whatever artifacts are produced that is detected and suppressed by using novel technique for example based color transfer framework. It gives the faster colorization of image than previous method by reducing grain suppression and improving color fidelity and detail preservation. In the future, we will extend our framework for video editing. . We also would like to develop a method to handle video with more clarity and faster.



International Journal of Innovative Research in Computer and Communication Engineering

(An ISO 3297: 2007 Certified Organization)

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