



AUTOMATED APPROACH FOR COLOR IMAGE GENERATION

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ABSTRACT

We introduce a general technique for “generating” images by automatic transferring luminance and color between a source and a target images both can be in black/ white, grayscale and color images at any time. Although the general problem of adding luminance and chromatic values to a target image was not exact, so, the current approach attempts to provide a method to help minimize the amount of human labor required for this task.

This method perform the image generating rely on luminance matching techniques between the source and target pixel neighborhood, and transfer the best pixels of neighborhood matching, i.e., luminance and chromatic information from source to target. The results demonstrate this technique introduce natural-looking generating in entertaining manner and work surprisingly well for a large variety of images, no more time of frustration attempting to get just the look you want without human intervention. It is compatible with such popular image processing software as “Adobe PhotoShop”.

Although this study find a new method focus only on the homogeneous images, future work will solve the problems of non-homogeneous images.

الخلاصة

قُدمت تقنية عامة لتوليد الصور بواسطة نقل أوتوماتيكي لشدة الإضاءة و الألوان بين صور المصدر و الهدف و كلاهما ممكن أن يكونا صور أبيض و أسود أو رمادية أو الملونة في أي وقت على الرغم من أن المشكلة العامة لإضافة قيم الإضاءة و الألوان إلى المصدر غير مضبوطة، فهذه الطريقة هي محاولة تساعد على تقليل كمية الجهد المطلوبة من قبل الإنسان لتنفيذ هذه المهمة.

هذه الطريقة تعمل على توليد الصور بالاعتماد على تقنية التماثل بالإضاءة بين مجاورات عنصر الصورة للمصدر و الهدف و نقل أفضل عنصر صورة من بين المجاورات المتماثلة (شدة الإضاءة و الألوان) من المصدر إلى الهدف. أثبتت النتائج بأن هذه التقنية لتوليد الصور ذات نظرة طبيعية وبطريقة ممتعة و عمل مدهش لأنواع مختلفة من الصور، وليس هناك أوقات كثيرة لمحاولة الحصول على الأشكال التي نريدها و بدون تدخل من قبل الإنسان. و هذه الطريقة ممكن استخدامها كبرنامج عام لمعالجة الصور. على الرغم من أن هذه الدراسة تقترح طريقة جديدة تركز على الصور المتجانسة لكن مستقبلها ممكن حل المشاكل المرتبطة بالصور الغير متجانسة.

KEYWORDS

Image Processing, Generation, luminance and chromatic.

INTRODUCTION

Although various image processing concepts are in common use in the computer graphics [6], an application of image generation to this domain is a recent field of research.

One wishes to generate an image from another for many reasons: choose the colors and intensity which increase the visual appeal of an image such as an old black and white photo; they make an old movie nicer, and help to make a scientific illustration more attractive. Two images are converted into three dimensional (YIQ) pixel values, such that dealing the target image as an image whose elements (pixels) are characterized only by one feature (luminance or intensity).

Our method is based on a simple premise: neighboring pixels in a square block that have similar intensities between source and target image blocks, should have the luminance and the chromatic information of a pixel from a source image with the best matching. This matching is formalized using a cost function and an optimization problem that can be solved efficiently using standard techniques.

To perform the matching, the average pixel luminance within its neighborhood and the standard deviation of the luminance in a pixel neighborhood with a size of $(N * N)$ pixels are used. The matching is performed on a pixel with its neighborhood of the source image using a sequential search.

In this study, we examine the luminance values in the neighborhood of each pixel in the target image and add to it the luminance and the chromatic information of a pixel from a source image with the best neighborhood matching. The procedure of evaluation works by minimizing the matching error (in terms of luminance) between pairs of source and target pixel neighborhoods. Then, transfer luminance and chromatic from the best evolved source neighborhood matching to the target [1].

Our concept of transferring chromatic values from one image to another is inspired by work by Welsh et al and Reinhard et al, [2][3], in which chromatic values are transferred between two images. In their work, chromatic values from the source image are transferred to a second image using a simple but surprisingly successful procedure. In preceding work, it transferred chromatic values as well as the luminance value from source to target in order to satisfy the meaning of the generation process. In the proposed approach, we would like to satisfy the following:

1. Quality; it should produce pleasing results.
2. Generality; it should work remarkably well for a wide range of images.
3. User-friendly; the number of tunable input parameters should be minimal.
4. Simplicity; the algorithm implementation is simple.
5. Efficiency; computational cost, time required by this technique.

THE LUMINANCE/ CHROMINANCE COLOR SYSTEM:

When processing color images we can use the RGB system and process each color plane separately. This is the simplest approach to dealing with color.

To overcome this, we often make use of another color representation known as the luminance/ chrominance color system. Color images are broken down into a luminance component



corresponding to the brightness and two chrominance components representing the color information. The luminance is essentially a monochrome version of the image, whilst the chrominance is calculated from what are known as color differences. However, there are various different ways of calculating these components. For example, in the PAL standard used in European television systems, the luminance(Y) and the chrominance components (I and Q) are defined as:

$$\begin{aligned}Y &= 0.30R + 0.59G + 0.11B \\I &= 0.493(B - Y) = -0.15R - 0.29G + 0.44B \\Q &= 0.877(R - Y) = 0.62R - 0.52G - 0.11B\end{aligned}$$

Where R, G, and B are the values of Red, Green and Blue components respectively [4]. Note that the chrominance components are defined in terms of the difference between the blue and red components and the luminance, and hence are known as color differences. Other standards, such as the NTSC standard used in the USA and the SECAM standard used in Europe, use the same definition for luminance although differ in their definitions of the two chrominance components.

One significant advantage of using the YIQ system is that it allows us to process the color and brightness/ luminance of an image separately, e.g. modifying the brightness without changing the color and vice versa.

In the below, we present the characteristics components of this algorithm.

THE PROPOSED GENERATION APPROACH

In this section, we describe the general algorithm for transferring luminance and chromatic from source to target image. The general procedure for this task requires a simple idea. Both source and target RGB image are converted to a decorrelated color space that minimize the correlation between the three coordinate axes of the color space. The color space provides three decorrelated, example of decorrelated color spaces is YIQ principal channels corresponding to an chromatic luminance channel, and two chromatic channels in which change made on one color channel should minimally effect values in the in other channels. Then we use each pixel with its neighborhood to move from one pixel to another. In each time, we use a neighborhood statistic to determine the best match among them. Then, find the best match for a pixel. Once the best matching pixel is found, the Y, I and Q values are transferred to the target pixel.

Error Evaluation

Next, it is important to define a suitable function to minimize luminance-matching error between two images.

In order to transfer luminance and chromaticity values from the source to the target, Each pixel in the target image must be closest to a pixel within best neighborhood in the source image. The comparison is based on the luminance value which is determined by the Y channel in YIQ space and neighborhood statistics of that pixel.

The neighborhood statistics are precomputed over the image and consist of the standard deviation of the luminance values of the pixel neighborhood. The best matching is selected based on the weighted average of luminance (50 %) and standard deviation (50 %). The comparison is based on the luminance value and neighborhood statistics of that pixel. This step is similar to texture synthesis algorithms, [4][5], in which the distance is used to find texture matches.

$$\text{Error } (N_t, N_s) = \text{match } (N_t, N_s) \quad (1)$$

Where;

N_t : Target pixel neighborhood.

N_s : Source pixel neighborhood.

and the match between two (N_t and N_s) Pixels neighborhood to find suitable pixels from source is defined to be:

$$0.5|\mu_s(i, j) - \mu_t(i, j)| + 0.5|\delta_s(i, j) - \delta_t(i, j)| \quad (2)$$

Where function μ and function δ are represent luminance average and standard deviation of its argument both taken with respect to a N*N source or target neighborhood as referred to by subscript s or t respectively.

i, j : are indices of a pixel in row i and column j .

Once the best matching pixel is found the luminance and chromaticity values are transferred to the target pixel.

Hybrid Image Matching

The search technique imposed by the proposed generation approach has hybrid form: a global and local search. A global search explores the luminance search space of the source image to locate suitable regions to that of luminance regions of the target image. This global search is hybrid with a local search that selected the most suitable pixels luminance to that of the corresponding pixels in the target image.

The Proposed Algorithm

In this section, we go to produce final image by generation process. The general procedure requires a few simple steps.

1. RGB for both target and source images are converted into YIQ space. This space has been chosen because it promptly provides the luminance value which is a crucial datum for our procedure.
2. Evaluate mean and standard deviation for each pixel luminance and the neighborhood statistics. After the evaluation process is completed.
3. Search the best pixel and neighborhood luminance.
4. Transfer (YIQ) from source image to the target image.
5. Finally, convert YIQ to RGB from the final image. Although, this procedure is very simple and direct the experimental results show that it works very well on a large set of images. Results, comments and discussion are the object of the next section.

THE EXPERIMENTAL RESULTS

In fact, there is no good objective criterion available for measuring the perceived image similarity. However, there are a number of common error measurements [7]. *Mean Square Error* (MSE) measures the average amount of difference between pixels of output image and original target image. If the *MSE* is small, the output image closely resembled the original.

The sum of luminance differences between output image and original image is defined as:




$$MSE = \frac{1}{n_t * m_t} \sum_{i=0}^{n_t-1} \sum_{j=0}^{m_t-1} |lumI_{(i,j)} - lumP_{(i,j)}| \quad (3)$$

Where function *lum* represent a luminance value of its argument, n_t and m_t : are respectively width and depth of target image.

lumI :The original image.

lumP :The generated image.

Table 1: Represent The results For Target Images

Target image	Target Image Size	Source Image size	MSE	Time (sec.)
	144*198	144*198	3.87	1397
	180*135	180*135	3.4	1700
	90*162	135*144	2.63	462
	108*126	72*45	1.5	70
	108*162	72*72	5.48	147
	99*129	72*108	12.13	162
	117*90	90*90	2.5	139
	36*36	81*81	2.7	11
	117*171	90*168	3.585	633
	216*162	216*162	2.97	455
	111*148	123*163	1.83	1001
	165*135	172*141	3.138	798
	162*114	171*114	6.92	643

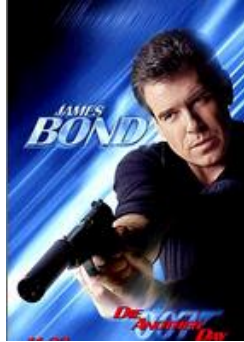
This section reports all the experimental results obtained with the proposed method. Fig 1 shows some examples of generated images obtained with the generation algorithm. Different source or target pixel neighborhood size can be used such that (3*3,5*5,7*7,...,15*15) but the most

suitable size in this method is 5×5 . obviously, decreasing this size results in more acceptable results but at the expense of increasing maximum required more time. Table 1 reports the results of the MSE and running time of the algorithm that range from 11 seconds to 28 minutes on a PentiumIII 900 MHZ CPU. Running time will be depend on the neighborhood size and the size of images. Most images can be generated reasonably well in under a minute.

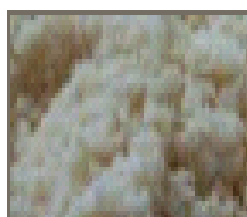
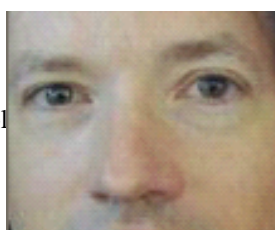
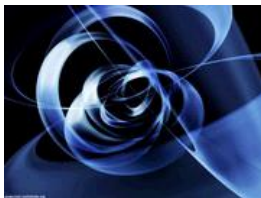
Target Image



Source Image



Generated Image





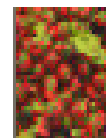
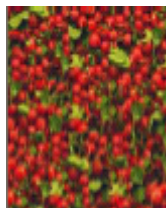
Target Image



Source Image



Generated Image





Target Image



Source Image



Generated Image



Availab

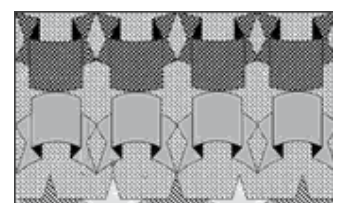
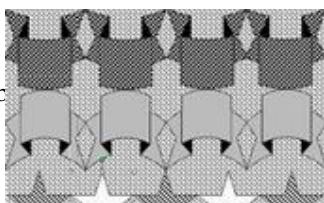




Fig.1: Some examples of generated image obtained with the generated strategy. First column is the target image, second column is the source image and third column is generated image.

CONCLUSION

In this paper, we have formulated a new and general approach to the problem of generating image images without user intervention. We have intentionally kept the basic technique simple and general by not requiring registration between the images or incorporating spatial information. Our technique can be made applicable to a large class of images by automatic search function.

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