CONTENT-BASED COMPRESSION IN AERIAL IMAGING USING INTER-COLOR CORRELATION

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ABSTRACT:

The high regional correlation between the R, G, B components of color aerial imaging is used as a basis for a new coding technique. In this work we approximate subordinate colors in smooth segmented regions of an image as a function of one of its base colors. This way, only a reduced number of parameters are required for coding a significant part of the color information. The encoded images with localized high correlation, such as aerial imaging, also introduce an option of progressive transmission, which could be used for slower communication channels. Our conclusion is that the use of correlation in color images compression could be superior to the traditional decorrelation techniques, in particular for results are superior to those obtained by other compression techniques that are based on decorrelated color spaces, as in JPEG. We also introduce an option of progressive transmission, which could be used for slower communication channels. Our conclusion is that the use of correlation in color images compression could be superior to the traditional decorrelation techniques, in particular for images with localized high correlation, such as aerial imaging.

1. BACKGROUND

Most aerial images are characterized by high correlation between their R, G, B color components. Many compression techniques take advantage of this redundancy and transform the color primaries into a decorrelated color space such as YIQ or YUV. The JPEG compression algorithm is a typical example of this approach where the I and Q information of the YIQ representation are encoded separately at lower rates than the Y. In recent years, however, a new approach has been developed, where instead of decorrelating the color components, the correlation is used to approximate two of the components as functions of the third component (Goffman and Porat, 2002; Kotera and Kanamori, 1990). Since color images also have spatial or regional color correlation, the color information could be further compressed based on its spatial correlation. A promising approach that segments gray images into regions and achieves a high compression ratio is based on the Region Growing algorithm and was introduced in (Kunt et al. 1985), regarded as a ‘second generation’ compression method.

In our work we extend the principles of Kunt et al. and of Goffman and Porat to account for these two types of correlation - localized spatial correlation and the inter-color correlation. We will show that the combined approach of the ‘second generation’ tools for image compression improves significantly the results reported in previous color compression algorithms (Goffman and Porat, 2002; Kotera and Kanamori, 1990). In Region Growing we refer to the standard image processing technique that describes an image as a set of adjacent regions. The pixels inside the regions represent the texture information of the image, and the pixels of the region boundaries represent the contour details. In our proposed algorithm, contour and texture information are coded separately.

2. STAGES OF THE ALGORITHM

The proposed algorithm is described in detail in this section. The flowchart of the encoder is described in Figure 1.

Component Separation – based on the RGB image, the Y component is calculated to represent the luminance of the image. This component is likely to best represent edges in the image, regardless of their actual color, R, G, or B.

Pre-processing – reduction of noise in the image while preserving the information of the contour. If the segmentation process is done on the original image, the noise that appears as small variation of the gray level will produce undesired contours. In this work the inverse gradient filter (Wang and Vaganucci 1981) is used due to its two main qualities:

1. The inverse gradient filter reduces the variance, of each region, while preserving the average.
2. The inverse gradient filter has little effect on the sharpness of edges in the picture.

Segmentation – The image is then segmented into contour pixels and texture pixels. This procedure partitions the image into regions in which the brightness is continuous and smooth. The segmentation starts with a seed pixel and examines the properties of neighbouring pixels. If the neighbouring pixels have similar properties, they are added to the region. The procedure stops when no additional pixels satisfy the criteria to be added to the region, or when the algorithm reaches the image boundary. In order to achieve a high compression ratio, the number of regions is reduced before coding the information. Two heuristics are used in this paper to decrease the number of regions obtained by region growing: elimination of the small regions and merging weakly contrasted adjacent regions.

Contour Coding – The contour pixels are coded using chain code (Freeman 1961). In this scheme, the location of the first contour pixel in a chain has to be transmitted. Then the direction of each neighbouring contour pixel is transmitted sequentially. To improve the compression, the chain is represented by the changes in the direction rather than the actual directions, using Huffman coding. This work assumes 8
adjacent neighbours, i.e., allowing 8 possible directions from each pixel. Once a complete chain (region) is transmitted, the starting point of the next region has to be coded. Here to save bits, only the distance (vertically and horizontally) between the new starting point and the previous one are transmitted instead. Since the number of bits allocated to this part of the algorithm is significant (typically some 50% of the bit budget) it is of interest to consider the factors that affect this part of the coding.

The number of bits needed to code each direction is estimated as follows:

\[ b_{pd} = \sum p_i \cdot L_i \]  \tag{1}

where \( b_{pd} \) is the number of bits per direction, \( p_i \) is the frequency of each change in direction, \( L_i \) is the length of each code word; \( i=0, 1, 2, ..7 \).

Then,

\[ b_{sc} = \text{ceiling}(\log_2(\text{max_dist})) \]  \tag{2}

where \( b_{sc} \) is the number of bits per each starting point, \( \text{max_dist} \) is the maximum distance between consecutive two starting points of chains.

From (1) and (2) we can calculate the number of bits per contour \( (b_{pc}) \):

\[ b_{pc} = b_{pd} + \frac{[(N_{chain} - 1) \times b_{sc} + 2 \times fsb]}{N_{contour}} \]  \tag{3}

where \( N_{chain} \) is the number of chains, \( N_{contour} \) is the number of contour points, and \( fsb \) is the number of bits needed to code the first starting point.

**Texture coding** – since each region is relatively smooth, the color components in each region is described using a smooth two-dimensional polynomial function. The order of the polynomial function is determined according to the approximation error and the number of bits needed to code the region information. In our algorithm, we use one of the primary colors of the image (R,G,B) to represent the brightness. In most cases we select the Green, which is closely related to the gray level \( (Y) \) information (Goffman and Porat, 2002).

The coefficients of each polynomial are calculated so that the error between the approximation and the original region is minimized.

For a zero order polynomial function the coefficient is the average value. For a first order polynomial function the approximation and the error are:

\[ Y_{\text{approx}} = a_1 + a_2 X_1 + a_3 X_2 \]  \tag{4}

\[ SE = \sum \left( y_i - a_1 - a_2 x_{i1} - a_3 x_{i2} \right)^2 \]

where \( Y_{\text{approx}} \) is the approximation of \( Y \), \( X_1, X_2 \) are axes of the polynomial function, \( a_1, a_2, a_3 \) are the approximation coefficients, SE represents the Squared Error.

The coefficients that minimize the SE are found based on

\[ \begin{bmatrix} n \sum x_i^0 & \sum x_i x_{i1} & \sum x_i x_{i2} & \sum x_i^2 & \sum x_i x_{i1} & \sum x_i x_{i2} & \sum x_i^2 & \sum y_i x_{i1} & \sum y_i x_{i2} \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} \sum y_i \sum y_i x_{i1} \sum y_i x_{i2} \end{bmatrix} \]  \tag{5}

For a second order polynomial function the approximation is

\[ Y_{\text{approx}} = a_1 + a_2 X_1 + a_3 X_2 + a_4 X_1^2 + a_5 X_2^2, \]  \tag{6}

and the coefficients that minimize the SE are found using

\[ \begin{bmatrix} n \sum x_i^0 & \sum x_i x_{i1} & \sum x_i x_{i2} & \sum x_i^2 & \sum x_i x_{i1} & \sum x_i x_{i2} & \sum x_i^2 & \sum y_i x_{i1} & \sum y_i x_{i2} \end{bmatrix} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{bmatrix} \]  \tag{7}

**Color coding** - The two additional subordinate colors are coded using a polynomial expansion of the base color as follows:

\[ C_1 = \sum_{k=0}^{n} P(C_0) = a_1 C_0^k + a_2 C_0^{k+1} + ... + a_n C_0 + a_0 \]  \tag{8}

\[ C_2 = \sum_{k=0}^{n} P(C_0) = b_1 C_0^k + b_2 C_0^{k+1} + ... + b_n C_0 + b_0 \]

where \( C_0 \) is the base color, \( C_{1,2} \) are the interpolated subordinate colors, \( k \) is the order of the polynomial expansion, and \( a_k, b_k \) are the coefficients of the polynomial expansion.

**Quantization** - the coefficients of the color polynomial expansion and the polynomial coefficient of each region are quantized and transmitted to the receiver along with contour information.

The decoder reverses the order of the above procedure.
3. PROGRESSIVE TRANSMISSION

For slow communication channels, such as aerial wireless links, it is possible to improve the efficiency of the transmission if the information is sent progressively. In such a process we reduce the size of the regions gradually. We start with the transmission of large regions and then subdivide the regions to improve the resolution of the received image. The first compressed image has only large regions. In each consecutive stage, new and smaller regions are added to produce a more refined image. It should be noted that the statistics of the contour’s chains may be different in each stage. The length of the chains decreases in later steps, so that the code for end-of-chain appears more frequently. This implies that each stage may need different Huffman codes. In this work the selected code is the one that minimizes the number of bit per direction. From the gathered statistics, 16 different Huffman codes were found. Thus only the index of the code is transmitted to the receiver.

4. COMPRESSION RESULTS

The algorithm for aerial imaging was implemented with the Green as a base-color. Each region of the base color was approximated using a second order polynomial function. The subordinate colors were approximated using a first order polynomial expansion of the base color. For comparison, images compressed by the proposed method and by the JPEG algorithm are shown in Figures 2 and 4. Results of progressive transmission are shown in Figures 3 and 5.

5. CONCLUSIONS

This technique achieves a high compression ratio with higher image quality compared to the JPEG algorithm, especially for images of aerial mapping, where the blockiness effect of JPEG is very noticeable. Unlike the arbitrary 8x8 blocks of JPEG, here the regions are based on the natural segmentation of the mapping, providing an efficient tool for high quality content-based compression. The quality of the image can be controlled using progressive transmission – in each stage a more refined image is obtained. Results of the compression for various images have been shown, and specific properties of the algorithm have been analyzed and discussed. Our conclusion is that a regional color correlation approach could be superior to the traditional decorrelation methods, especially for aerial imaging.

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REFERENCES

Figure 2. (a) Original image, (b) image compressed with the proposed algorithm with compression ratio of 102 and PSNR=25.93db, (c) image compressed with jpeg algorithm with compression ratio of 73 and PSNR=24.69db.

Figure 3. Stages in progressive transmission (a) the first progressive image, (b), (c), (d) are the results of the next stages of the progressive transmission with compression ratio of 67.3, 54.6, 39.8 and 24 respectively.
Figure 4. (a) Original image, (b) image compressed with the proposed algorithm with compression ratio of 84 and PSNR=22.36db, (c) image compressed with jpeg algorithm with compression ratio of 83 and PSNR=22.22db

Figure 5. Stages in progressive transmission (a) the first progressive image, (b), (c), (d) are the results of the next stages of the progressive transmission with compression ratio of 48.6, 43.1, 33.4 and 21.8 respectively