Performance, Analysis of Filtering Schedule
Using Deblocking Filter for the Reduction of Block Artifacts from MPEQ Compressed Document Images

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Abstract: Problem statement: The visual effects of blocking artifacts can be reduced by using deblocking filter. Also with out smoothing the natural edges, the perceived quality of video sequence can be enhanced. This study proposes a method to remove blocking artifacts in low bit-rate block based video coding. Approach: The proposed algorithm has two separate filtering modes, which are selected by pixel behavior around the block boundary. In each mode, proper one-dimensional filtering operations are performed across the block boundary along horizontal and vertical directions, respectively. In the first mode corresponding flat regions, a strong filter is applied inside the block as well as on the block boundary, because the flat regions are more sensitive to the Human Visual System (HVS) and the artifacts propagated from the previous frame due to motion compensation are distributed inside the block. In the second mode corresponding to other regions, a sophisticated smoothing filter, which is based on the frequency information around block boundaries, is used to reduce blocking artifacts adaptively without introducing undesired blur. Even though the proposed deblocking filter is quite simple, it improves both subjective and objective image quality for various image features.

Results and Conclusion: Deblocking filter improves the PSNR of about 0.1 dB for video encoded using MPEG-4 and H.264 without using its own in-loop deblocking filter. It has proven to be good in the reduction of the very annoying blocking artifacts caused by video compression.

Key words: H.264 compression, deblocking filters, video coding, discrete cosine transform, filtering, Peak Signal to Noise Ratio (PSNR)

INTRODUCTION

The main topics in the modern research of the field of multimedia are video compression and video coding. As the amount of information is huge on videos, video compression plays a vital role to transmit videos. By considering height, width, number of channels (usually three), color depth (usually minimum 8 bits) and sequence length (expressed in number of frames) are the parameters required for the calculation of amount of information contained in raw video. N bits = n Frames * H * W * n Channels * color Depth.

Many different Encoding strategies were proposed in literature. The main achievement is to obtain the representation of the sequence which is as tiny as possible. Block-based processing technique was used by the most of the video coding standards of the part. But, this kind of processing can cause visible blocking artifacts in the encoded video which are annoying for the user. Thus their effect should be mitigated as much as possible. The main source of blocking artifacts is the block-based integer Discrete Cosine Transform (DCT) in intra- and inter-frame prediction error coding. The second source of blocking artifacts is motion compensated prediction. The interpolated pixel data present in different reference frames can be used to generate motion compensated blocks. Discontinuities on the edge of copied block occur as there is almost never a perfect fit for this data. Video compression leads to blocking which is the annoying visible artifacts. This problem is reduced by the small 4x4 transform size used in H.264/MPEG-4 AVC. Deblocking filter is an advantageous tool to maximize coding performance (List et al., 2003).

Existing system: The compression artifacts arise in JPEG, MPEG and H.264 by quantization of DCT coefficients. The quantization of low frequency coefficients results in blocky noise and the quantization of high frequency coefficients results in mosquito noise.
In data transmission if channel bandwidth is narrow, then data rate will be low and quantization level is dropped. As a result, the compression artifacts are increased. There are many methods which reduces the compression artifacts at the decoder. The Deblocking Edge Filter (DEF) method (Aujol et al., 2005; List et al., 2003; Chambolle, 2004) was reduces the blocky noise by using the noise removal technique called Projection On to Convex Sets (POCS), which is based on an iterative filtering (Dolar et al., 2009). An approach using Wavelet transform was proposed (Goto et al., 2008) for effective blocky noise reduction.

Another effective approach is Total Variation (TV) regularization (Kaup, 1998; Kim et al., 1998; Orchard et al., 1997). Which reduces the noise (Robertson and Stevenson, 2005). By utilizing this method, it is possible to reduce blocky noise. Alter et al. (2005) In this method based on a projected TV regularization especially targeting on DCT noise removal (Zakhoe, 1992). In this method, the total variation is reduced under DCT coefficient quantization constraint. This method will reduce a compression distortion, but minuteness of image is lost due to detetion of texture components. Thus the reduction of blocky noise and mosquito noise became insufficient at low bit rates (Rudin et al., 1992). The goal of a deblocking filter is to reduce blockiness and also preserving the sharpness of the content of the picture. To attain this, the large absolute difference between samples near a block edge is measured, which should be reduced (Choi and Kim, 2000). If magnitude of that difference is large then it cannot be described by coarseness of the quantization used in encoding in which the edge are more likely to reflect the actual behaviour of the source picture and should not be smoothed over.

Deblocking filter is implemented in the encoding loop shown in Fig. 1. In literature, two main approaches for deblocking can be found (List et al., 2003).

The first approach is post processing filter in which the deblocking operation is applied at each frame of the video after encoding/decoding procedure. The second approach is loop filtering in which filtering operation is carried out in encoding loop,which has the advantage of using filtered frames as reference frames leading to a higher quality prediction in motion compensation.

At the same time, the disadvantage is the application of identical filtering for the purpose to stay in synchronization with encoder. The deblocking effect can be improved by usage of post processing deblocking filter and loop filter. Filtering is not required in case of real edges of the video as application of filter will do blurring and it may result in difficulty in distinguishing real edges. so, an additional condition other than non-zero boundary strength (bs) is needed for effective use of
deblocking filter. Block edge samples \((p_2, p_1, p_0, q_0, q_1, q_2)\) are filtered only if they meet the following conditions Eq. 1 and 2:

\[
bs > 0 \quad (1)
\]

\[
|p_0 - q_0| < \alpha \quad \&\&\quad |p_1 - p_0| < \beta \quad \&\&\quad |q_1 - q_0| \leq \beta \quad (2)
\]

where, \(\alpha\) and \(\beta\) are the thresholds defined in the standards. Two types of filters are used: strong filter (5-tap filtering) and normal filter (4-tap filtering).

Filters are applied according to following:

\[
\text{if } ((|p_0 - q_0| < \alpha \&\& |p_1 - p_0| < \beta \&\& |q_1 - q_0| \leq \beta ) \&\& bs == 4) \text{ apply strong filter;}
\]

\[
\text{else if } ((|p_0 - q_0| < \alpha \&\& |p_1 - p_0| < \beta \&\& |q_1 - q_0| \leq \beta ) \&\& 0 < bs < 4) \text{ apply normal filter;}
\]

\[
\text{else no filter;}
\]

**MATERIALS AND METHODS**

The processing of frames in the sequence is independently on each other. 2-D filter (working both in horizontal and vertical directions) can be applied on each pixel. The decision map determines the process of filtering for the specific pixel. All the block-based video codecs (and the related blocking-artifacts), are covered by this method such that it filters 4×4 boundary of the frame. The algorithm can be applied on MPEG and on H.264 coded sequences. At first 8×8 block edges are scanned and the 4×4 block edges are processed subsequently.

\(F\) is the activity factor for the six-pixel vector \(P = \{p_0, p_1, p_2, p_3, p_4, p_5\}\). \(G(\text{QP})\) is a threshold and it is function of \(\text{QP}\): stronger is the quantization and higher should be the value of \(G\). \(F(p)\) represents the number of detected edges inside the vector \(P\). \(T_2\) represents a fixed threshold max and min are the maximum and minimum values of \(P_0\) and According to \(F(p)\) the vector \(P\) can divide the processing of the algorithm in three filtering mode types:

- Filtering decision step
- First filtering pass
- Second filtering pass

**Decision modes:** The two values \(XY\) is called filtering mode which assigns each pixel of the frame. \(X\) represents horizontal filtering mode and \(Y\) represents vertical filtering mode. \(X\) and \(Y\) take the values from the set \{N, D, S\}, where N means no filtering, D means default filtering and finally S means strong filtering.

![Fig. 2: Vector filtering classification](image)

The outputs of this first step are two filtering mode decision matrices, one for horizontal and other for the vertical direction (for each pixel the couple of value \(XY\) is defined). Filtering modes are calculated based on the variation of vertical and horizontal six-pixel vectors in its each 4×4 block boundary. First the activity of the six-pixel vector must be checked and if it is high, it means that there are variations in the set of pixels and they are to be filtered using strong filtering mode, otherwise by default filtering mode.

In second step, the final decision is made by estimating the activity of the pixel set is caused due to blocking artifacts or natural sharpness of the image itself. Statistics of the vector (based on neighbour pixel values) are computed if the pixel in the vector are candidates for strong or default filtering mode. If the value of the pixel has difference among them it cannot be explained due to the blocking artifacts effects. The decision map does not has filtering mode and the pixel value is altered. Otherwise, the decision will be strong or default filtering mode depending upon the first decision step in Fig. 2. Each set is characterized by their filtering modes estimated to its pixels. Pixels located around the horizontal block boundary of \(G_2\), have the filtering modes: \{NN, ND, NS, SN, SD, SS\}, as the horizontal default filter is not possible for \(G_2\) pixels.

Similarly the possible filtering modes of other sets are:
G1: \{NN, NS, ND, DN, DD, DS, SN, SD, SS\}
G3: \{NN, NS, DN, DS, SN, SS\}
G4: \{NN, NS, SN, SS\}

Filtering: For the quality we introduce the motion of Filtering Window (FW) in Fig. 3. To designate a 6×6 pixel box centred at the intersection of four 4×4 pixel blocks. In Fig. 4a FW is first placed at the upper left corner of MB and shifted based on scanning order. In Fig. 4b 8×8 block edges are filtered by the remaining 4×4 block edges. This process considers the blocking artifacts in video coded with 8×8 block DCT, like MPEG-2.

To avoid multi-filtering, 16 pixels in FW are filtered by 2-D filter. To reduce the complexity of nonseparable filters, MB is processed in two passes. In Fig. 5b, the dark yellow pixels represent formerly filtered pixel from the upper and left MBs, while the light yellow pixels are the filtered pixels after running the first pass on current MB. The white regions represent the remaining unfiltered pixels. These pixels are filtered later in a second pass, with a simplified set of filters as defined below.
Fig. 5: (a) Pixel filtered in step 1 inside the filtering window (b) inside the MB

Fig. 6: Weighting coefficients for SS Mode

Fig. 7: Weighting coefficients for DS/SD Modes

**First filtering pass:** In first pass eight pixels FW(p12, p13, p21, p24, p31, p34, p42, p43) are filtered. in Fig. 5a. Filtering modes for G2 and G3 pixels are:

\{NN, ND, NS, SN, SD, SS\} + \{NN, NS, DN, DS, SN, SS\}

For filtering modes with an N (no filter) in any direction (ND, NS, DN, SN), only one dimensional filters are required. For instance, ND and DN modes apply a 1-D default filter on the target pixel in vertical and horizontal direction respectively. The ND mode can be assigned to the pixels belonging to G2, (p12, p13, p42, p43). In this case, the filter is applied vertically on the target pixel.

As in the DN filtering mode of the pixels belonging to G3 (p21, p24, p31, p34), the filtered pixel values are computed symmetrically to the ND filtering.

In the cases like where the filtering mode belongs to \{DS, SD, SS\}, a 2-D filtering is applied on the desired pixel. The introduced 2-D filters are the simplified versions from the combination of the horizontal and vertical 1-D filters. In order to preserve a small amount of computations, the weighted matrix of the 2-D filter is simplified. Some coefficients which are having a small weight are cut and others are rounded, while preserving similar filter characteristics. Figure 6 and 7 show the simplified 2-D filters used to process p21 and p12 in SS, DS and SD modes. Other G2 and G3 pixels are filtered in the same way.

**Second filtering pass:** At the end of the first pass, pixels belonging to G2 and G3 are filtered throughout the MB. During this second pass, we filter the remaining pixels which are belonging to G1 and G4, by applying the appropriate filter.
This filter is done according to the pre assigned filtering mode and also by using the updated pixels from the first pass. In Fig. 8, $p(i-1), p(i+1), p(i+2), p(j-1), p(j+1), p(j+2)$ represent pixels filtered during the first pass, while white pixels are updated according to their assigned filtering mode as follows:

**DN:**
$$P_{ij} = (p(i-1) + 3p(i) + p(i+2)) \gg 3$$

**NS:**
$$P_{ij} = 2p(i-1) + p(i+2) + p(j) \gg 2$$

**SD:**
$$P_{ij} = 6p(i-j) + 4p(i+2) + 4p(j) + 2p(j+1) + p(i-j) - p(j+2) \gg 2$$

**DD:**
$$P_{ij} = 8p(i-j) + 4p(i+2) + 4p(j+1) + p(i-j) + p(j+2) - p(i+2) \gg 4$$

**SS:**
$$P_{ij} = 2p(i-j) + 2p(i) + 2p(j+1) + p(j+2) + p(i+2) \gg 3$$

For symmetric filtering modes, the filtered values of $p_{ij}$ are simply computed in a symmetric manner. Once the decision map is ready, the filtering can be started. A 6x6 filtering window is considered and centered at the intersection of four 4x4 pixel blocks. The filtering window is first filtered 8x8 block edges and then the 4x4 block. For each filtering pixels shown in Fig. 8 are filtered, once the first filtering pass is completed, the second filtering pass is applied to remaining pixels.

**Experimental results:** The sequences have been encoded starting from a raw YUV file. A frame of the sequence "Foreman" from the uncompressed YUV sequence in Fig. 9. A frame of the sequence "Foreman" from the MPEG-4 compressed sequence in Fig. 10.
Table 1: Performance of deblocking filter

| Video sequence | Bitrate | PSNR compressed | PSNR deblocked | PSNR gain | Encoder                  |
|----------------|---------|-----------------|----------------|-----------|--------------------------|
| Foreman        | 168 kbps| 27.68           | 27.83          | 0.15      | MPEG-4                   |
| News           | 179 kbps| 31.84           | 31.91          | 0.07      | MPEG-4                   |
| Coastguard     | 211 kbps| 25.62           | 25.69          | 0.07      | MPEG-4                   |
| Foreman        | 63 kbps | 26.90           | 27.06          | 0.16      | H.264 (without in-loop deblocking filter) |
| News           | 47 kbps | 29.73           | 29.87          | 0.14      | H.264 (without in-loop deblocking filter) |
| Coastguard     | 144 kbps| 27.25           | 27.35          | 0.10      | H.264 (without in-loop deblocking filter) |
| Foreman        | 63 kbps | 27.57           | 27.60          | 0.03      | H.264 (with in-loop deblocking filter) |
| News           | 38 kbps | 28.83           | 28.86          | 0.03      | H.264 (with in-loop deblocking filter) |

The same frame, after applying the deblocking filter in Fig. 11 Working of the deblocking filter has been tested in three different scenarios:

- Videos are encoded using MPEG-4 codec
- Videos have been encoded with H.264 codec, with the in-loop deblocking filter turned off
- Videos have been encoded with the H.264 codec, with the in-loop deblocking filter turned on

The encoding has been tuned in order to generate clear blocking artifacts. In H.264 encoding block is evident and in remaining it is visible.

The bit rate is directly connected to amount of compression used (the lower the bit rate, the worse the video and, generally speaking, the more visible the encoding artifacts). The PSNR (Peak Signal-to-Noise Ratio) is an objective measure of video quality. The PSNR for two images can be computed as follows:

\[ \text{MSE} = \frac{1}{mn} \sum_{i,j} [i(i,j) - k(i,j)]^2 \]

MSE is called Mean Square Error. The PSNR is defined as:

\[ \text{PSNR} = 10 \log_{10} \left( \frac{\text{MAX}}{\text{MSE}} \right) \]

where, MAX is the maximum value of the image, which is for example 255 for 8-bit images.

**RESULTS**

The PSNR value reported here is the average PSNR value calculated for each frame. To compute this value, the compressed sequences (after encoding) and the deblocked sequences (after our filter) have been tested against the uncompressed YUV sequence.

**DISCUSSION**

PSNR does not take into account of the HVS (Human Visual System model) and is not reliable measure of the objective quality of processed image to original image not only the small improvement in PSNR value, the deblocking filter has proved to be reduced of annoying blocking artifacts by video compression. It can be clearly seen that the blocks are smoothed out human eye perceives a better quality of the deblocked frame.

**CONCLUSION**

From the results reported in the Table 1, it is proved that deblocking filter improves the PSNR to 0.1 dB for video encoder using MPEG-4 and H.264 without using its own in-loop deblocking filter. Video sequences encoded by H.264 with its own deblocking filter is enabled and the improvement is lower (0.02-0.03 dB). This is because most of the blocking artifacts is already removed by H.264 in-loop deblocking filter and the video sequences cannot be further improved.

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