A TEMPORAL QUALITY SMOOTH METHOD FOR WAVELET VIDEO CODERS

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ABSTRACT
The wavelet-based video codec is capable of providing scalable bitstreams such that quality of service (QoS) among heterogeneous networks can be manipulated in a unified framework. The three-dimensional wavelet video coder 3D WVC had been developed to provide this fully scalable bitstream. However, the quality control or rate allocation was performed on the scale of one GOP such that the reconstructed pictures suffer temporal quality fluctuations. The theoretical coding model provided ratio parameters among subbands to smooth temporal picture quality. However, it needs further adaptation for dealing with practical WVC. We propose to investigate the rate-distortion relation, between the decomposed subbands ($\Psi$) and the reconstructed pictures in one GOP ($\tilde{P}$), under which the rate allocation for constant quality (RACQ) can be operated. Experiments show that the temporal quality fluctuations can be reduced to 10–30 times smaller as compared to 3D-SPIHT and 5–10 times smaller to the ratio parameter approach.

1. INTRODUCTION
The wavelet-based scalable video coder has been developed by many [1] [2] as an alternative to predictive coding approaches, e.g., MPEG-4 or H.264. By investigating the temporal correlation in addition to 2D wavelet coding, the 2D+t (or 3D) wavelet video coders demonstrate fine granularity properties in temporal, spatial or quality displaying dimensions. Previous works on 3D wavelet video coders provide frameworks [1] [2] to assemble the embedded scalable bitstream. These frameworks aim to achieve high compression efficiency for scalable videos in which a group of pictures (GOP) is the basic processing unit to demonstrate scalable capabilities. Since the embedded bitstream can be truncated at any point to fit the rate constraints, the QoS problem can be simplified to how to allocate bits among subband images. In JPEG2000, EBCOT [3] uses the Lagrangian technique to construct the embedded bitstream such that any truncation points are optimal. The EBCOT coding approach can be extended to 3D wavelet video coder [4] to provide temporal and spatial scalable bitstreams. Note that global optimal bit allocation for one GOP, i.e., 3D WVC, may not present good subjective performances due to temporal quality fluctuations [5]. If the temporal correlation was not exploited, such as that in the motion JPEG2000, the temporal quality fluctuation can be reduced by controlling the picture bitstream under the buffer-channel constraints [6]. This rate control algorithm is similar to EBCOT, operated under independent R-D function, but is proceeded with larger video units. Another [7] proposed to find the rate-distortion ratio among temporal subbands to smooth the temporal video quality. A set of ratio parameters was derived based on the filtering process of Daubechies 9-7 wavelets. Notwithstanding the assumed i.i.d. property among subbands and high bit rate coding model may not be satisfactory in practical coding, the ratio parameters provide a good initial bit allocation for RACQ.

We propose to investigate the relation from the rate allocation in $\Psi$ to the error distribution in reconstructed pictures $\tilde{P}$ for performing the RACQ algorithm. The analysis/synthesis framework of the scalable 3D WVC [2] is utilized to formulate this rate-distortion mapping relation, denoted as $M_s$. A rate allocation algorithm which targets constant picture quality can then be developed based on this $M_s$ function.

This paper is organized as follows. Section 2 discuss the theoretical ratio parameters for constant quality video. The rate-distortion relation among decomposed subbands $\Psi$ and reconstructed pictures $\tilde{P}$ of the scalable 3D WVC and the RACQ algorithm are presented in Section 3. Section 4 is the experimental study. Section 5 concludes this paper.

2. THEORETICAL CODING CONTROL MODEL
For 3D wavelet video coders, potential temporal PSNR fluctuations in the reconstructed picture can be eliminated by adjusting the slopes of rate-distortion operation points among subbands. In high bit rate coding, the distortion relationship between wavelet filter coefficients and reconstructed data
is [3]:

\[
\hat{d}(2n) = d(L) \sum_j f^2_{2j} + d(H) \sum_j g^2_{2j+1} \tag{1}
\]

\[
\hat{d}(2n + 1) = d(L) \sum_i f^2_{2i+1} + d(H) \sum_j g^2_{2j},
\]

where \( g' \) and \( f' \) are synthesis high-pass and low-pass filter coefficients respectively and \( d(H) \) and \( d(L) \) are the coding errors, in mean square sense, of high and low subband respectively. To make reconstructed data with constant error, i.e., \( d_{2n} = d_{2n+1} = \hat{d} \), it can be derived that \([7]\)

\[
\frac{d(L)}{d(H)} = \sum_i g_{2i+1}^2 - \sum_j g_{2j}^2.
\]

For \( k \) level 1D wavelet decomposition, there will be at most \( M = 2^k \) subbands. Let \( S = \{s^0, s^1, \ldots s^{M-1}\} \) denote the corresponding synthesis basis matrix that \( s^i \) is the basis for coefficients in position \( Mk+i \). Based on the i.i.d. assumption, the relation between distortion of the reconstructed data and the wavelet coefficients is

\[
\tilde{D} = S \cdot D,
\]

where \( \tilde{D} = \{d_0d_1 \cdots d_{M-1}\}, D = \{d_0d_1 \cdots d_{M-1}\} \) and \( S = \{\sum_k s^j_{k-i}\} \) \([7]\) respectively. In high rate coding, the rate distortion operation slopes can be assumed to be proportional to subband distortions, i.e.,

\[
\lambda^0 : \lambda^1 : \cdots : \lambda^{M-1} = d(0) : d(1) : \cdots : d(m-1). \tag{4}
\]

By controlling R-D operation slopes (and hence \( D \)) according to the specified filters \( S_i \), it can lead to smooth temporal video quality, i.e., \( \tilde{D} = \mathbf{I} = \{11 \cdots 1\} \). The ratio parameter in eq. 4 is derived under high-bit rate assumptions and is analyzed in statistical approaches. Further adjustments are required when performing bit allocation in GOP-based coding framework. Experiments show that the ratio parameter approach (STD-EWVC) did provide good initial bit allocations for constant quality video. However, the adaptation capability should be exploited to accommodate the non-ideal properties of video signals.

3. SCALABLE WAVELET VIDEO CODER

3.1. Rate-Distortion Relation in 3D WVC

The temporal analysis, for example, of an eight picture GOP in 3D WVC is shown in Fig. 1. Solid rectangular boxes without shading denote the source pictures of the \( k \)-th GOP, \( P_k = \{p_k(i) = p(i + (k-1) \cdot N)\}_{i=1,\ldots,N} \). Let \( l_{i,j} \) and \( h_{i,j} \) be the \( j \)-th pair of the approximate and detail pictures on the \( i \)-th decomposition level. Final subband pictures, \( \Psi_k = \{l_{\log_2N,1}, h_{\log_2N,1}\} \cup \)

\[
\{h_{i,j} | i = \log_2N - 1, \cdots , 2, 1, j = 1, 2, \cdots, \left[ \frac{N}{2^i} \right] \} \tag{5}
\]

under which the RACQ would be operated, are marked as shaded boxes. Intermediate subband pictures are marked as dashed boxes. Solid and dotted lines between subband pictures denote the low- and high-pass analysis filtering in the temporal decomposition process respectively. Note that motion compensated temporal filtering (MCTF \([1]\) could be utilized before performing the decomposition. It needs to transmit the motion vector information but expects a higher compression ratio.

In this 3D WVC, the control target to smooth quality among reconstructed pictures in \( \tilde{P}_k \) is achieved by performing proper rate allocation in \( \Psi_k \). The mapping between \( \tilde{P}_k \) and \( \Psi_k \) in the synthesis (reconstruction) process of the 3D WVC (Fig. 1) is found to be

\[
\mathbf{M}_S(\psi_{i,j}) = \{\tilde{p}_k(m) | m = \tilde{m}_b + 1, \tilde{m}_b + 2, \cdots, \tilde{m}_b + \tilde{N} \}, \tag{6}
\]

where \( \tilde{m}_b = 2^i \cdot (j-1) \), \( \psi_{i,j} \in \Psi_k \) and \( \mathbf{M}_S(\psi_{i,j}) \subset \tilde{P}_k \) would return the set of pictures in which the rate adjustment in \( \psi_{i,j} \) would distribute. For example, if rates allocated to \( h_{2,1} \) were adjusted, then the resultant error distribution would include pictures specified by \( \mathbf{M}_S(h_{2,1}) = \{\tilde{p}_k(1), \tilde{p}_k(2), \tilde{p}_k(3), \tilde{p}_k(4)\} \).

3.2. Rate Allocation for Constant Quality

In JPEG2000 EBCOT \([3]\), the Lagrange multiplier optimization process is used to solve the following rate-distortion optimization problem:

\[
\min_{R(i), n \leq i \leq n+N-1} \sum_{i=n}^{n+N-1} D(i),
\]

subject to \( \sum_{i=n}^{n+N-1} R(i) \leq R_0, \tag{7} \)
where $R(i)$ and $D(i)$ are the rate-distortion (in mean square error) pair for the $i$-th codeblock of subband images. Since the rate-distortion functions $R(D)$s among codeblocks are independent, the Lagrange multiplier optimization procedure can be operated on these $R(D)$s directly. The total distortion in subband codeblocks, $\sum_i D(i)$ equals to that in the spatial domain, i.e., $\sum_{x,y}(p(x,y) - \hat{p}(x,y))^2$. However, the optimal bit allocation in one GOP of WVC may lead to unpleasant quality fluctuations [5] among reconstructed pictures in $\hat{P}$. In addition, the quality in the spatial domain $\hat{P}$ does not correspond to the quality modifications in subband images $\Psi$. The STD-EWVC [7] provides ratio parameters $\alpha_{ij}$s, i.e., $\lambda_i = \alpha_{ij} \cdot \lambda_j$ for $i \neq j$, to utilize the unconstrained Lagrangian optimization process. However, the theoretical approach to yield constant temporal quality may need to be modified due to non-ideal signal properties. We propose to perform rate adjustment, based on the initial bit allocation, among subband images $\Psi_k$ to provide practical smooth video quality in $\hat{P}$. This RACQ algorithm is described below.

1. For each N-picture GOP, say $\hat{P}_k$, it performs the temporal decomposition with the framework in Fig. 1 and returns $N$ subband pictures specified by $\Psi_k$ in eq. (5).

2. With the bit budget for one GOP, it performs the unconstrained Lagrangian multiplier optimization based on R-D relation of pictures in $\Psi_k$. Outputs of this step are denoted as:

$$\{r(\log_2 N,1),r(h_{i,j})|i=\log_2 N,\cdots,2,1;j=1,2,\cdots,[\frac{N}{2}]\},$$

where $r(\psi_{i,j})$ denotes the rate allocated for subband picture $\psi_{i,j}$.

3. For $i = \log_2 N - 1, \cdots, 2, 1,$ for $j = 1, 3, \cdots, [\frac{N}{2}] - 1,$

$$\Delta_Q = \text{Q} \text{ave}(\mathbf{M}_S(h_{i,j})) - \text{Q} \text{ave}(\mathbf{M}_S(h_{i,j+1}));$$

where $(\Delta_Q > T_Q) \{\text{while} \ (\Delta_Q > 0) \ \text{move} \ \delta_b \ \text{bits from} \ r(h_{i,j}) \ \text{to} \ r(h_{i,j+1}); \ \text{else} \ \text{move} \ \delta_b \ \text{bits from} \ r(h_{i,j+1}) \ \text{to} \ r(h_{i,j}); \}

$$\text{Q} \text{ave}(\mathbf{M}_S(\psi_{i,j}))$$

denotes the average PSNR of the reconstructed pictures in $\mathbf{M}_S(\psi_{i,j})$.

Note that the initial bit allocation may dominate the $\text{Q} \text{ave}$ but it would not affect much in temporal quality smoothness performed by RACQ. In addition to this initial bit allocation, the scale of other control factors, $\delta_b$ and $T_Q$, would also have impacts on the PSNR variations in $\hat{P}_k$. From our experiments, the ratios $\lambda_{LLL} : \lambda_{LHL} : \lambda_{LH} : \lambda_{HH} = 2.827 : 1.999 : 1.414 : 1$ provides satisfactory performances either in $\text{Q} \text{ave}$ or PSNR variances.

4. EXPERIMENTAL STUDY

The Haar wavelet basis is used to temporally decompose pictures in $P$. To provide scalable video bitstreams, the 2D SPIHT [8] wavelet coder is used to encode subband pictures in $\Psi$ (eq. (5)). In evaluating the performance of the proposed RACQ algorithm, the PSNR performance of the high bit rate modeling approach STD-EWVC is provided for comparisons.

The test sequences are QCIF (176 × 120) “football” (high activity) and “table tennis” (medium activity) at rate 30 picture/sec. The compression ratios are selected to be 0.6 bpp and 0.3 bpp (bits per pixel) respectively. As shown in Fig. 2(a), the PSNR variances are largely reduced by the RACQ algorithm (thick gray line) as compared to the 3D-SPIHT (solid line). For one GOP, the dynamic ranges are reduced from 3-4 dB to be within 0.5 dB. As compared to the high bit rate approach STD-EWVC (dotted line), the RACQ also demonstrates much smoother temporal picture quality. Fig. 2(b) demonstrates the PSNR performance at 0.3 bpp. As shown, the proposed RACQ algorithm still provides stable and smooth temporal quality as compared to the other two algorithms. Numerical analyses show that the temporal picture quality variations can be reduced to 10-30 times smaller as compared to 3D-SPIHTS and 5-10 times smaller to STD-EWVC. To further inspect coding control performances, the MCTF process [1] was performed prior to utilizing RACQ algorithm. As shown in Fig. 3, the MCTF-RACQ did improve the average PSNR and the temporal quality smoothness. Further simulations demonstrate that the MCTF does not improve much in the overall PSNR or temporal quality smoothness since the operational functions of MCTF and temporal video decomposition are almost the same.

Note that in performing the RACQ algorithm under the 3D WVC system, any rate adjustment among subband pictures would distribute errors with units of pairwise pictures so that quality variations between the pairwise pictures in $\hat{P}$ can not be eliminated further. In addition, the RACQ may sometimes be operated in the rate-distortion boundary such that no further rate adjustment toward constant quality is possible. These two factors, hindering further rate adjustment toward constant quality, are unavoidable by nature in the 3D WVC. As shown in the 3rd, 4-th and 5-th GOPs (16 pictures) of Fig. 3, the temporal picture quality could not be further smoothed even large dynamic quality ranges remain.

5. CONCLUSIONS

We had proposed a quality control method for the fully scalable 3D WVC. For one GOP, the rate-distortion relation $\Psi$ between temporally decomposed subbands and reconstructed pictures had been exploited. By utilizing this re-
Fig. 2. Temporal PSNR performance among RACQ, STD-EWV & 3D-SPIHT on the “football” seq.

Fig. 3. Temporal PSNR performances among RACQ, MCTF-RACQ and 3D-SPIHT on the “mother and daughter” seq.

6. REFERENCES


