Two-Dimensional Audio Compression Method Using Video Coding Schemes

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Abstract: As video compression is one of the core technologies that enables seamless media streaming within the available network bandwidth, it is crucial to employ media codecs to support powerful coding performance and higher visual quality. Versatile Video Coding (VVC) is the latest video coding standard developed by the Joint Video Experts Team (JVET) that can compress original data hundreds of times in the image or video; the latest audio coding standard, Unified Speech and Audio Coding (USAC), achieves a compression rate of about 20 times for audio or speech data. In this paper, we propose a pre-processing method to generate a two-dimensional (2D) audio signal as an input of a VVC encoder, and investigate the applicability to 2D audio compression using the video coding scheme. To evaluate the coding performance, we measure both signal-to-noise ratio (SNR) and bits per sample (bps). The experimental result shows the possibility of researching 2D audio encoding using video coding schemes.

Keywords: audio compression; video compression; next generation audio coding; versatile video coding (VVC); mu-law encoding; linear mapping

1. Introduction

As consumer demands for realistic and rich media services on low-end devices are rapidly increasing in the field of multimedia delivery and storage applications, the need for audio or video codecs with powerful coding performance is emphasized, which can achieve minimum bitrates and maintain higher perceptual quality compared with the original data. As the state-of-the-art video coding standard, Versatile Video Coding (VVC) [1] was developed by the Joint Video Experts Team (JVET) of the ITU-T Video Coding Experts Group (VCEG) and ISO/IEC Moving Picture Experts Group (MPEG). VVC can achieve a bitrate reduction of 50% with similar visual quality compared with the previous method, named High Efficiency Video Coding (HEVC) [2]. In the field of audio coding technology, Unified Speech and Audio Coding (USAC) [3] was developed by the MPEG audio group as the latest audio coding standard, which integrates speech and audio coding schemes. Whereas VVC can accomplish significant coding performance in the original image or video data, USAC can provide about 20 times compression performance of the original audio or speech data.

In this study, we conducted various experiments to verify the possibility of compressing two-dimensional (2D) audio signals using video coding schemes. In detail, we converted 1D audio signal into 2D audio signal with the proposed 2D audio conversion process as an input of a VVC encoder. We used both signal-to-noise ratio (SNR) and bits per sample (bps) for performance evaluations.

The remainder of this paper is organized as follows: In Section 2, we overview the video coding tools, which are newly adopted in VVC. In Section 3, we describe the proposed
methods for converting a 1D audio signal into a 2D audio signal. Finally, the experimental results and conclusions are provided in Sections 4 and 5, respectively.

2. Overview of VVC

VVC can provide powerful coding performance compared with HEVC. One of the main differences between HEVC and VVC is the block structure. Both HEVC and VVC commonly specify coding tree unit (CTU) as the largest coding unit, which has a changeable size depending on the encoder configuration. In addition, a CTU can be split into four coding units (CUs) by a quad tree (QT) structure to adapt to a variety of block properties. In HEVC, a CU can be further partitioned into one, two, or four prediction units (PUs) according to the PU splitting type. After obtaining the residual block derived from the PU-level intra- or inter-prediction, a CU can be partitioned into multiple transform units (TUs) according to a residual quad-tree (RQT) structure similar to that of CU split.

VVC substitutes the concepts of multiple partition unit types (CU, PU, and TU) with a QT-based multi-type tree (QTMTT) block structure, where the MTT is classified into binary tree (BT) split and ternary tree (TT) split to support more flexibility for CU partition shapes. This means that a QT can be further split by the MTT structure after a CTU is first partitioned by a QT structure. As depicted in Figure 1, VVC specifies four MTT split types: vertical binary split (SPLIT_BT_VER), horizontal binary split (SPLIT_BT_HOR), vertical ternary split (SPLIT_TT_VER), and horizontal ternary split (SPLIT_TT_HOR), as well as QT split. In VVC, a QT or MTT node is considered a CU for prediction and transform processes without any further partitioning schemes. Note that CU, PU, and TU have the same block size in the VVC block structure. In other words, a CU in VVC can have either a square or rectangular shape, whereas a CU in HEVC always has a square shape.

![Figure 1. VVC block structure (QTMTT).](image)

Table 1 shows newly adopted coding tools between HEVC and VVC. In general, intra-prediction generates a predicted block from the reconstructed neighboring pixels of the current block. As shown in Figure 2, VVC can provide 67 intra-prediction modes, where modes 0 and 1 are planar and DC mode, respectively, and the others are in angular prediction mode to represent edge direction. According to [4], the VVC test model (VTM) [5] achieves 25% higher compression performance than the HEVC test model (HM) [6] under the all-intra (AI) configuration recommended by JVET Common Test Conditions (CTC) [7]. This improvement was mainly realized by the newly adopted coding tools, such as position dependent intra-prediction combination (PDPC) [8], cross component linear model intra-prediction (CCLM) [9], wide angle intra-prediction (WAIP) [10], and matrix-based intra-prediction (MIP) [11]; the computational complexity of encoder substantially increased approximately 26 times [4].
Inter-prediction fetches a predicted block from previously decoded reference frames using motion estimation (ME) and motion compensation (MC) processes. The motion parameter can be transmitted to the decoder in either an explicit or implicit manner. When a CU is coded as SKIP or MERGE mode, the encoder does not transmit any coding parameters, such as significant residual coefficients, motion vector differences, or a reference picture index. In cases of SKIP and MERGE modes, the motion information for the current CU is derived from the neighboring CUs including spatial and temporal candidates. VVC adopted several new coding tools for SKIP/MERGE and inter-prediction, such as affine inter-prediction [12], geometric partitioning mode (GPM) [13], merge with motion vector difference (MMVD) [14], decoder-side motion vector refinement (DMVR) [15], combined inter- and intra-prediction (CIIP) [16], and bi-prediction with CU-level weight (BCW) [17]. VVC also adopted adaptive motion vector resolution (AMVR) [18], symmetric motion vector difference (SMVD) [19], and advanced motion vector prediction (AMVP) to save the bitrate of the motion parameters. Although these tools can significantly improve the coding performance, the computational complexity of a VVC encoder is up to nine times higher under the random-access (RA) configuration compared with that of HEVC [4].

**Table 1.** The list of video coding tools. Comparison between HEVC and VVC.

| Category          | HEVC                                      | VVC                                           |
|-------------------|-------------------------------------------|-----------------------------------------------|
| Intra prediction  | 35 Intra-prediction modes                 | 67 intra-prediction modes                     |
|                   |                                           | Wide angle intra-prediction (WAIP)            |
|                   |                                           | Position-dependent intra-prediction combination (PDPC) |
|                   |                                           | Matrix-based intra-prediction (MIP)           |
|                   |                                           | Multi reference line prediction (MRL)         |
|                   |                                           | Intra sub-block partitioning (ISP)            |
|                   |                                           | Cross-component linear model (CCLM)           |
| Skip/Merge        | Regular skip/merge                        | Regular skip/merge                            |
|                   |                                           | Sub-block-based temporal motion vector predictors (SbTMVP) |
|                   |                                           | Merge with motion vector difference (MMVD)    |
|                   |                                           | Decoder-side motion vector refinement (DMVR)  |
|                   |                                           | Bi-directional optical flow (BDOF)            |
|                   |                                           | Combined inter- and intra-prediction (CIIP)   |
| Inter prediction  | Advanced motion vector prediction (AMVP)  | Advanced motion vector prediction (AMVP)       |
|                   |                                           | Symmetric motion vector difference (SMVD)     |
|                   |                                           | Affine inter-prediction                       |
|                   |                                           | Adaptive motion vector resolution (AMVR)       |
|                   |                                           | Bi-prediction with CU-level weight (BCW)      |

**Figure 2.** Intra-prediction modes in VVC.
Per the JVET CTC [7], VVC supports three encoding configurations [20]: all-intra (AI), low-delay (LD), and random-access (RA). In the AI configuration, each picture is encoded as an intra picture and does not use temporal reference pictures. Conversely, in case of the LD configuration, only the first frame in a video sequence is encoded as an intra frame; subsequent frames are encoded using the inter-prediction tools to exploit temporal reference pictures as well as intra-prediction tools. In the RA coding mode, an intra frame is encoded in an approximately one second interval in accordance with the intra period, and other frames are encoded as hierarchical B frames along the group of pictures (GOP) size. This means that a delay may occur in RA configuration because the display order is different from the encoding order.

Although the aforementioned VVC coding tools provide significant compression performance compared with HEVC, their computational complexity [21] of the encoder is high. Therefore, many researchers have studied various fast encoding algorithms for block structure, intra-prediction, and inter-prediction of VVC while maintaining the visual quality. In terms of block structure, Fan et al. proposed a fast QTMT partition algorithm based on variance and a Sobel operator to choose only one partition from five QTMT partitions in the process of intra-prediction [22]. Jin et al. proposed a fast QTBT partition method through a convolutional neural network (CNN) [23]. To reduce the computational complexity of intra-prediction, Yang et al. proposed a fast intra-mode decision method with gradient descent search [24] and Dong et al. proposed an adaptive mode pruning method by removing the non-promising intra-modes [25]. In terms of inter-prediction, Park and Kang proposed a fast affine motion estimation method using statistical characteristics of parent and current block modes [26].

3. VVC-Based 2D Audio Compression Method

3.1. Overall Frameworks of 2D Audio Coding

In general, the original 1D audio signal is generated from pulse code modulation (PCM), which involves sampling, quantization, and binarization processes. In this study, we used a 48 kHz sampled audio signal, and each sample was already mapped to signed 16 bits binary number by PCM coding. As the inputs of VVC specify a video sequence, which is an unsigned 2D signal with a max 12 bits integer number [7], we converted the original 1D audio signal into an unsigned 2D audio signal as the pre-processing before VVC encoding.

As shown in Figure 3, the overall architectures for 2D audio compression can be divided into the encoder and decoder side, where the encoder side consists of the 2D audio conversion process and the VVC encoder. In the 2D audio conversion process, signed 1D audio signals with a bit-depth of 16 bits are mapped to unsigned 1D audio signals with the bit-depth of 8 or 10 bits and then unsigned 2D audio signals are generated by 1D-to-2D packing as an input of the VVC encoder. After receiving a compressed bitstream, the decoder side conducts VVC decoding and inverse 2D audio conversion process to reconstruct the 1D audio signal with a bit-depth of 16 bits. Note that it may strongly influence the coding performance of the VVC depending on how the 1D signal is converted to a 2D signal.

3.2. Proposed 2D Audio Conversion Process

To generate unsigned 1D audio signals, we exploited three approaches as follows:

1. Non-linear mapping (NLM) to generate unsigned 8 bits per sample;
2. Linear mapping (LM) to generate unsigned 10 bits per sample;
3. Adaptive linear mapping (ALM) to generate unsigned 10 bits per sample.
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Non-linear mapping (NLM) by Mu-law [27] is primarily used in 8 bits PCM digital telecommunication systems as one of two versions of ITU-T G.711. This reduces the dynamic range of an original signed 1D audio signal, as expressed in Equation (1):

\[ NLM(I) = \text{sgn}(I) \cdot \frac{\ln(1 + (2^N - 1)|I|)}{\ln(1 + (2^N - 1))}, \]  

(1)

where \( \text{sgn}(I) \) and \( N \) denote the sign function and the bit-depth of output samples, respectively.

Similarly, linear mapping (LM) maps the dynamic range of an original signed 1D audio signal into a specified sample range depending on the maximum value (Max) of samples. As shown in Figure 4, the sample’s dynamic range of LM is from -Max to Max. This range can be divided into uniform intervals and each interval is represented with the type of unsigned N bits by Equation (2).

\[ LM(I) = \left\lfloor \frac{I + \text{Max}(|I|)}{2 \cdot \text{Max}(|I|)} \cdot (2^N - 1) \right\rfloor, \]  

(2)

where \( |x| \) and \( \text{Max}(|I|) \) indicate the rounding operation and the maximum value of all samples, respectively. Note that the encoder side transmits the value of \( \text{Max}(|I|) \) with the bit-depth of 16 bits for signed 1D conversion. Since this method has to search for the maximum value in the overall audio samples, this process is slow compared with the length of the audio signal. In addition, an unused sample range, called the dead zone, may exist in the range of the unsigned 1D signal, as shown in Figure 4. To resolve the aforementioned problems, we considered adaptive linear mapping (ALM). As shown in Figure 5, ALM firstly searches for both the Min (I) and Max (I) values among the signed 1D audio samples and then performs the LM conversion to represent the unsigned N bits 1D audio signal, as expressed in Equation (3). Because a video encoder generally compresses input data in the unit of a 2D frame, ALM can reduce the encoding delay in the unit of a frame and avoid the dead zone problem compared with LM. Therefore, the encoder side
has to transmit both $\text{Min}(I_k)$ and $\text{Max}(I_k)$ values for each frame, where $k$ is the index of the frames.

$$ALM(I_k) = \left\lfloor \frac{I_k - \text{Min}(I_k)}{\text{Max}(I_k) - \text{Min}(I_k)} \left( 2^N - 1 \right) \right\rfloor$$ (3)

Figure 4. Graphical representation of linear mapping where the orange area indicates the unused sample range called the dead zone.

Figure 5. Graphical representation of adaptive linear mapping.

After performing unsigned 1D conversion, the 2D audio signal is generated from the unsigned 1D audio signals used as the inputs of the VVC encoder. For example, if the spatial resolution (width $\times$ height) of a frame is $256 \times 256$, the unsigned 1D audio signals, whose total number is $256 \times 256$, are arranged a frame according to the raster-scan order, as shown in Figure 6. Figure 7 shows the 2D packing results according to the aforementioned 2D conversion approaches. Although 2D audio signals showed different and complex texture patterns between consecutive frames, we confirmed that the 2D signals had a horizontally strong directionality within a frame.
4. Experimental Results

In this study, we used an original audio signal (48 kHz, 16 bits) whose length was approximately 195 s and consisted of five music, five speech, and five mixed items, as described in Table 2. To compress a 2D audio signal using the VVC encoder, we first set the spatial resolution to $256 \times 256$. Because the original audio signal had 9,371,648 samples, we generated 143 frames with a $256 \times 256$ spatial resolution. As demonstrated in the spectrograms in Figure 8, ALM is superior to other conversion methods in terms of SNR.
After generating 2D audio sequences with 143 frames, they were compressed by the VVC encoder under the experimental environments presented in Table 3.

Table 2. Description of audio signal items.

| Category  | Item Name          | Description               |
|-----------|--------------------|---------------------------|
| Music     | salvation          | Classical chorus music    |
|           | te15               | Classical music           |
|           | Music_1            | Rock music                |
|           | Music_3            | Pop music                 |
|           | phi7               | Classical music           |
| Speech    | es01               | English speech            |
|           | louis_raquin_15    | French speech             |
|           | Wedding_speech     | Korean speech             |
|           | te1_mg54_speech    | German speech             |
|           | Arirang_speech     | Korean speech             |
| Mixed     | twinkle_ff51       | Speech with pop music     |
|           | SpeechOverMusic_1  | Speech with chorus music   |
|           | SpeechOverMusic_4  | Speech with pop music     |
|           | HarryPotter        | Speech with background music|
|           | lion               | Speech with background music|

Figure 8. Spectrograms of the music_1 audio signal.

To evaluate the coding performance, we measured the SNR between the original and reconstructed 1D audio signals after decoding the compressed bitstream. Table 4 shows that NLM produced the worst coding performance due to the incorrectness of the signed 1D conversion. In addition, 10 bits LM had a higher SNR than ALM regardless of the encoding mode, as shown in Table 4 and Figure 9.
Table 3. Experimental environments for 2D audio encoding in the VVC.

|                     | 8 bits NLM | 10 bits LM | 10 bits ALM |
|---------------------|------------|------------|-------------|
| Input bit-depth     | 8 bits     | 10 bits    | 10 bits     |
| VTM version         | VTM10.0    |            |             |
| Chroma format       | YUV4:0:0   |            |             |
| Encoding mode       | All-intra, Low-delay-B, Random-access | | |
| Target bps          | 0.5 and 1.0 | 0.5        |             |

Table 4. Experimental results according to the different conversion modes.

| Encoding Mode       | Target bps | 8 bits NLM | 10 bits LM | 10 bits ALM |
|---------------------|------------|------------|------------|
|                     |            | SNR (dB)   | bps        | SNR (dB)   | bps        | SNR (dB)   | bps        |
| All-intra           | 1.0        | 14.09      | 1.01       | 30.55      | 1.03       | 29.30      | 1.02       |
|                     | 0.5        | 6.78       | 0.49       | 23.21      | 0.50       | 22.18      | 0.52       |
| Low-delay-B         | 1.0        | 13.78      | 1.01       | 30.17      | 0.97       | 29.66      | 1.03       |
|                     | 0.5        | 7.15       | 0.49       | 23.55      | 0.49       | 22.64      | 0.52       |
| Random-access       | 1.0        | 13.63      | 0.99       | 30.28      | 0.99       | 29.21      | 1.03       |
|                     | 0.5        | 7.55       | 0.52       | 23.62      | 0.51       | 21.82      | 0.50       |

Figure 9. Rate–distortion (RD) curves according to the different conversion methods.

We further investigated if the objective performance was well-reflected in the subjective quality. In terms of subjective assessment, we conducted a MUSHRA test [28] which is commonly used to evaluate subjective quality in the MPEG audio group. In this experiment, we used 10 audio items with a length of 10 s, and seven listeners participated. To enable comparison with the state-of-the-art audio coding method, the same audio signal was coded by USAC [29]. Because the encoded frames have different quantization parameters (QPs) according to the temporal layer of LD and RA configuration, SNR per frame can fluctuate between consecutive frames. Therefore, we performed MUSHRA tests under the AI configuration without QP variations. In the AI configuration, we set the target to 0.5 bps and conducted MUSHRA listening tests for the original, USAC, ALM, LM, and NLM. As shown in Figure 10, the subjective quality of 10 bits ALM was higher than that of 10 bits LM, which opposes the results in Table 4.
Although the average score of ALM or LM was lower than that of USAC, it was necessary to further investigate the possibility of researching 2D audio encoding with respect to 2D packing and video codec optimization. Firstly, the coding performance is strongly dependent on how the 1D signal is converted to a 2D signal. Because we used simple 2D packing in this study, it was necessary to consider a more elaborate 2D packing method suitable for video coding tools. For example, if we generate the 2D signals to efficiently perform the block based inter prediction, the coding performance of LD or RA configurations can be improved. Secondly, the current VVC has to transmit many unnecessary coding parameters due to the adoption of new tools, even though these tools are not used in the 2D audio encoding process. If a video encoder is optimized with core coding tools and light syntax structure, the coding performance will be improved.

5. Conclusions

In this paper, we proposed a pre-processing method to generate a 2D audio signal as an input of a VVC encoder, and investigated its applicability to 2D audio compression using the video coding scheme. To evaluate the coding performance, we measured both signal-to-noise ratio (SNR) and bits per sample (bps). Additionally, we conducted a MUSHRA test to evaluate subjective quality. The experimental results showed the possibility of researching 2D audio encoding using video coding schemes.

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![Scores of the MUSHRA listening test with a 95% confidence interval for different methods.](image-url)
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