



Fig. 2 Comparison of decoded frames (part magnified). Left: former VVC test model (JEM7.1) [8] 559kbps, 36.4 dB; Middle: H.265/HEVC test model (HM16.19) 552kbps, 35.4dB; H.264/AVC test model (JM19.0) 1106kbps, 35.2dB

The target compression performance is a 30–50% bit-rate reduction compared to H.265/HEVC at the same subjective video quality. However, with latest VTM version 6, the compression performance is 24-35%. It is forecast that this will be almost close to final objective performance of VVC. For subjective and objective comparison, decoded frames of VVC and former standards with rates and PSNR values are shown in Fig. 2.

1.3 Real-entity-oriented coding

In conventional video coding schemes including all above standard technologies, reproducing a signal as

close to the original video signal as possible has been mostly aimed at. With this approach, decoded video quality can never exceed original video signal. Generally, video capturing system inevitably offers noises, quantization, and subsampling distortions. As for fixed-camera videos, it has been reported that extraction of noise-reduced background image information by applying simple temporal filtering drastically gains coding efficiency up to 50% (measured in PSNR with original noisy signal) [10]. That work encourages us to depart from noisy pixel-oriented coding to noiseless real-entity-oriented coding (Fig. 3).

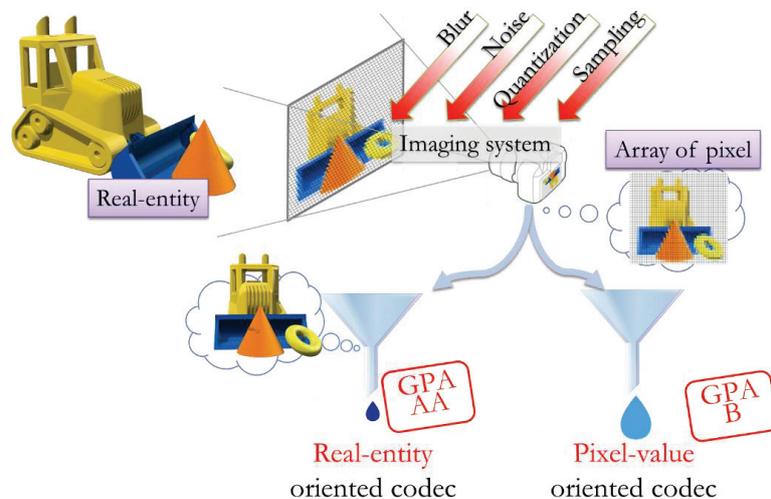


Fig. 3 Real-entity-oriented video coding vs. pixel-value-oriented (conventional) video coding

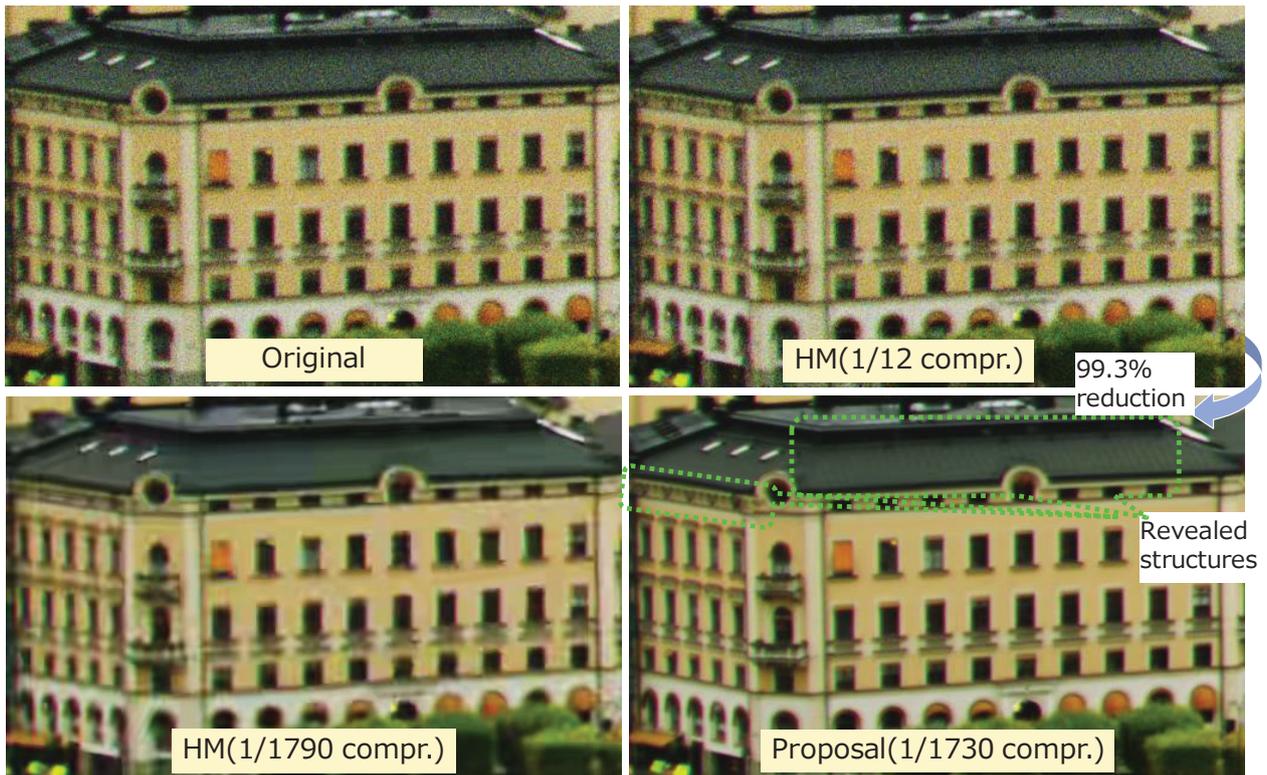


Fig. 4 Rigid object tracking-based real-entity-oriented video coding vs. conventional pixel-value-oriented video coding

2 Real-Entity-Oriented Coding Results

2.1 With rigid object tracking

With use of real-entity-oriented coding, video frames of rigid objects are registered using precise deformation and matching to successfully obtain denoised “real-entity image” and encoded and reproduced [11]. The results are shown in Fig. 4. Top-left picture is from original (uncompressed) video, bottom-left-bottom picture is from 1:1790 compressed video using HM. Bottom-right picture is from similar bit-rate, 1:1730 compressed video using proposed method. The visual quality is even better than the original, not to mention HM. And surprisingly, hidden textures (in dotted green boxes) are revealed in the proposal. Top-right picture is from high bit-rate (1:12 compression) video using HM. It is clearly observed that bits for coding noise wastes too much bits.

2.2 With water deskewing

Another target of the application of real-entity-oriented coding is shown in Fig. 5. The moving water heavily distorts the shapes of the objects under the water. By estimating the objects shape as if the water is not moving and assigning that image as additional reference frame (ARF), the coding efficiency greatly improves [12]. Fig. 6 shows the visual results of temporal median filter and proposed ARF, as well as original and still-water images. Median filter can usually eliminate unnecessary foreground objects but is quite blurred (coding efficiency is

Table 1 Water-bottom video coding gains vs. VTM4.0

ARF	Sequence	Y BD-rate
still-water (reference)	aquafish	-14.08%
	dish1	-48.19%
	kiban1	-53.36%
	pebble	-53.78%
	average	-42.35%
median	aquafish	5.76%
	dish1	13.77%
	kiban1	20.20%
	pebble	12.17%
	average	12.98%
Proposal	aquafish	-13.20%
	dish1	-45.91%
	kiban1	-47.51%
	pebble	-47.55%
	average	-38.54%

so low as expected). Our proposal looks crisp and similar to still-water image, and has undistorted shapes (cf. circular ring area and white lines in kiban1). Table 1 shows the comparison of coding efficiencies in BD-Rate [13] compared to VTM4.0 according to the ARF types, i.e., still-water, median filter, and proposal. Still-water gives the best performance, and our proposal the second with 39% bit-rate savings in average of four sequences. (Please be noted that still-water image cannot be used in real situations.)

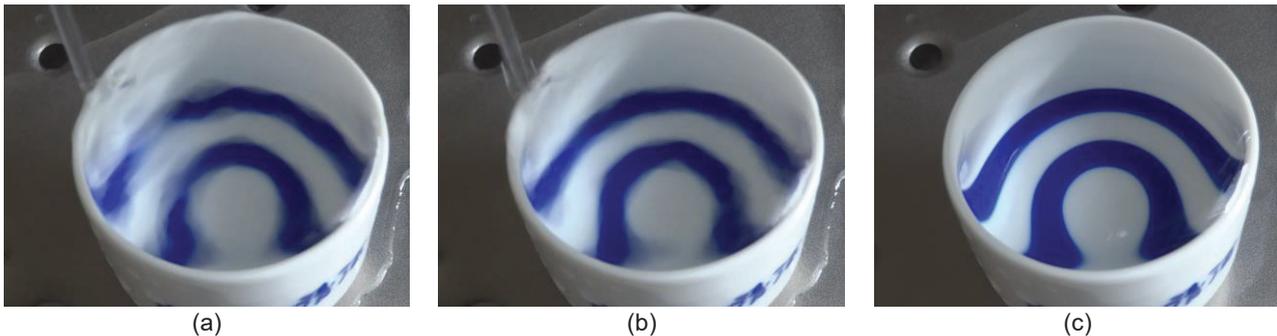


Fig. 5 Example of water-bottom video. (a) and (b): consecutive frames with moving water, (c): with still-water



Fig. 6 Magnified image comparison for sequence 'kiban1'. From left to right, original frame (skewed by water), median filter (non-skewed but blurry), proposed Additional Reference Frame (ARF), and still-water (for reference)

3 CONCLUSIONS

Real-entity-oriented approach presented in this paper is still in early-stage but a steady departure from conventional pixel-oriented technology, and will change the video quality metric because the reproduced video has higher quality than the original. The impact of the new approach is promising. It will help reduce the amount of huge video traffic with several digits of order, which supports the network services as a whole and also help incubating totally new video services over the network.

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