UVG Dataset: 50/120fps 4K Sequences for Video Codec Analysis and Development

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ABSTRACT

This paper provides an overview of our open Ultra Video Group (UVG) dataset that is composed of 16 versatile 4K (3840×2160) test video sequences. These natural sequences were captured either at 50 or 120 frames per second (fps) and stored online in raw 8-bit and 10-bit 4:2:0 YUV formats. The dataset is published on our website (ultravideo.cs.tut.fi) under a non-commercial Creative Commons BY-NC license. In this paper, all UVG sequences are described in detail and characterized by their spatial and temporal perceptual information, rate-distortion behavior, and coding complexity with the latest HEVC/H.265 and VVC/H.266 reference video codecs. The proposed dataset is the first to provide complementary 4K sequences up to 120 fps and is therefore particularly valuable for cutting-edge multimedia applications. Our evaluations also show that it comprehensively complements the existing 4K test set in VVC standardization, so we recommend including it in subjective and objective quality assessments of nextgeneration VVC codecs.

CCS CONCEPTS

 \bullet Computing methodologies \sim Computer graphics \sim Image compression \bullet Information systems \sim Data management systems \sim Database design and models

KEYWORDS

Open dataset, Raw video, Ultra High Definition (UHD), High Efficiency Video Coding (HEVC), Versatile Video Coding (VVC)

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1 Introduction

Digital video has become ubiquitous in our everyday life thanks to a myriad of multimedia devices and services. Cisco [1] reports that 75% of total IP traffic was dedicated to video in 2017 and estimates it to grow to 82% by 2022. This explosive growth is mainly driven by a proliferation of numerous video applications and their everincreasing quality expectations for immersive user experience.

In past decades, MPEG and ITU-T have addressed the needs of multimedia applications by announcing a series of video coding standards of which *High Efficiency Video Coding* (*HEVC/H.265*) [2] represents state-of-the-art. Currently, MPEG and ITU-T are also seeking coding gain beyond HEVC by developing a new standard called *Versatile Video Coding* (*VVC/H.266*) [3].

The main objective of digital video coding is to reduce the amount of bits needed to represent a video signal without deteriorating its quality. VVC and all prior international video coding standards are based on a classical hybrid video coding scheme that combines spatial/temporal (intra/inter) prediction, transform coding, quantization, and entropy coding.

Typically, video codecs are benchmarked with objective and subjective metrics to find out their *rate-distortion* (*RD*) performance, i.e., a trade-off between coding efficiency and loss of information. Conducting these quality assessments comprehensively calls for representative video datasets that address a broad range of video content (motion, texture, and illumination) and video parameters (resolution, frame rate, and bit depth).

In this paper, we present an open dataset made up of 16 raw 2160p (3840×2160) test video sequences. Eight of them were captured at 120 *frames per second* (*fps*) and the rest at 50 fps. The dataset is available online on our website at

http://ultravideo.cs.tut.fi/#testsequences

in 8-bit and 10-bit 4:2:0 YUV formats under a non-commercial Creative Commons BY-NC license. This work also characterizes all these sequences content-wise by their spatial and temporal perceptual information and performance-wise by their RD performance and coding complexity with the latest HEVC/H.265 and VVC/H.266 reference video codecs.

The rest of this paper is organized as follows. Section 2 gives an overview of the existing public video datasets. The proposed UVG dataset is presented in Section 3 and each UVG sequence is characterized in Section 4. Finally, Section 5 concludes the paper.

2 Existing Open Datasets

The existing open video datasets can be classified into 1) datasets made up of encoded video sequences; and 2) datasets containing original unprocessed video sequences in raw RGB or YUV format.

2.1 Encoded Datasets

These datasets are most commonly used in the *Quality of Experience (QoE)* measurements which are tolerant of quality degradations and visual artefacts introduced by dataset encoding. Therefore, typical applications of interest deal also with already encoded video, such as transmission and transcoding.

VideoSet [4] is composed of 220 5-second sequences, all in four resolutions: 1920×1080 (1080p), 1280×720 (720p), 960×540 (540p), and 640×360 (360p). Each of these 880 video clips were encoded using the AVC/H.264 codec [5] with *Quantization Parameter (QP)* values from 1 to 51. Authors provide the first three *just noticeable difference (JND)* points with 30+ subjects [6].

KoNViD-1k Video Database [7] contains 1200 panoramic 8-second 540p clips encoded with AVC/H.264 codec. The dataset was evaluated across blur, colorfulness, contrast, spatial information, temporal information, and video quality [8].

LIVE Netflix Video Quality of Experience database [9] consists of 112 distorted videos that were generated from fourteen 1080p videos at 24, 25, and 30 fps. The distortions are based on a set of eight different playout patterns including dynamic compression ratio of AVC/H.264 codec and rebuffering. This dataset was made for subjective quality assessment. Each video clip was assessed by 35 to 55 human subjects on a mobile device [10].

UGC dataset [11] includes 1500 20-second video clips sampled from millions of YouTube videos. They are split into 15 categories: animation, cover song, gaming, *high-dynamic-range* (*HDR*), how to, lecture, live music, lyric video, music video, news clip, sports, television clip, vertical video, vlog, and *virtual reality* (*VR*). Each video is available in YUV 4:2:0 format and in 360p, 480p, 720p, 1080p, and 2160p resolutions [12]. However, we classify this dataset into the first category, because it practically deals with encoded YouTube videos that were decoded and saved into YUV format. This approach introduces similar visual artifacts as found in the other encoded datasets.

2.2 Raw Datasets

This category is particularly essential in video codec evaluation and it is also better suited for many signal processing applications such as denoising. However, the acquisition of raw data requires professional camera equipment and extensive storage capacity because data integrity can only be preserved with lossless compression or without any compression.

Xiph.org Video Test Media [13] is a popular dataset for video compression. It contains 120 individual video clips of multiple resolutions (from 240p to 2160p), frame rates (from 25 fps to 60 fps), and content categories (e.g., movie and gaming) including nineteen 2160p sequences.

SJTU Media Lab proposed datasets for 4K [14] and HDR [15] video sequences. The 4K dataset includes fifteen 2160p sequences shot with Sony F65 camera at 30 fps. They are available in two formats: 10-bit YUV 4:4:4 and 8-bit YUV 4:2:0 [16]. The HDR dataset is made up of sixteen 2160p HDR sequences. The raw video data were recorded in Sony RAW 16-bit MAF format at 60 fps [17].

Joint Collaborative Team on Video Coding (JCT-VC) defined a dataset of 18 video sequences in the Common Test Conditions (CTC) [18] for HEVC standardization. The sequences were split into six classes (from A to F) according to spatial resolution and content type. The CTC dataset was updated by the Joint Video Experts Team (JVET) for VVC standardization [19]. Four sequences were removed, and nine new sequences were added including six 2160p sequences.

3 UVG Dataset Description and Characterization

The proposed UVG dataset contains 16 test video sequences. They were captured with Sony F65 video camera at either 50 or 120 fps in 16-bit F65RAW-HFR format and converted to 10-bit and 8-bit 4:2:0 YUV videos with FFmpeg [20]. Tables 1 and 2 tabulate these sequences in alphabetical order and depict their snapshots, essential parameters, content descriptions, specific features, and average RGB color histograms over all frames in each video.

Next, our dataset is first characterized with well-known *spatial* perceptual information (SI) and temporal perceptual information (TI) scores recommended by ITU [21]. Secondly, the RD characteristics of these sequences are evaluated with the latest releases of the HEVC/H.265 and VVC/H.266 reference encoders, i.e., the default 10-bit configurations of HM16.20 [22] and VTM8.0 [23], respectively. Finally, the sequences are assessed with the proposed complexity metric derived from [24]. In all these cases, the relevance of our UVG dataset is validated against the 2160p sequences of the VVC CTC dataset [19], which can be considered the most pertinent reference in current video coding research.

3.1 Spatial and Temporal Perceptual Information

SI indicates the maximum amount of spatial detail in a video. For a given N-frame sequence, SI is calculated by first filtering the luma samples (i, j) of each frame $F_n = F_0, F_1, ..., F_{N-1}$ with the Sobel filter, then computing the standard deviation over the Sobel filtered frame pixels, and finally taking the maximum of them [21] as

$$SI = \max_{0 \le n \le N-1} \{ \underset{1 \le i \le M-2}{\text{std}} \left[\sqrt{[Gh_n(i,j)]^2 + [Gv_n(i,j)]^2} \right] \}, \tag{1}$$

where W is the frame width and H is the frame height. $Gh_n(i,j)$ and $Gv_n(i,j)$ are the convolution results of the horizontal and vertical 3×3 Sobel edge detection kernel filters, respectively.

Correspondingly, TI stands for the maximum amount of temporal variation between successive frames F_{n-1} and F_n [21] as

$$TI = \max_{1 \le n \le N-1} \{ \underset{0 \le i \le W-1}{\text{std}} [F_n(i,j) - F_{n-1}(i,j)] \}.$$
 (2)

Table 1. Test sequence characteristics in the proposed UVG data set (sequences 1 - 8).

Snapshots	Chara	RGB color histogram		
Control of the Contro	Name: Beauty Frames: 600 Motion: Slow / plenty of Description: A close-up o Specific features: High co face; Noisy background.	f female face with hair l	•	# 10 ⁵ 2 1 0 0 255
	Name: Bosphorus Frames: 600 Motion: Slow / plenty of Description: A black boat Specific features: Rippli hilly landscape, a seaside	cruising on the waterwang water; foggy backg	round composed of a	4 # 10 ⁵ 2 0 0 255
To the second se	Name: CityAlley Frames: 600 Motion: Slow / little Description: A slow upwa Specific features: Detailer alley; high contrast between	d people on top of a tall		5#10 ⁵ 0 255
Ngitoric	Name: FlowerFocus Frames: 600 Motion: Slow / little Description: A close-up o Specific features: Blurred changes to flowers in the b	green background; after	-	# 10 ⁴ 10 0 255
	Name: FlowerKids Frames: 600 Motion: Slow / plenty of Description: Two kids mo Specific features: Childre the sequence; a lens flare of	oving in a yellow flower on close to the camera;	camera shaking along	# 10 ⁵ 10 5 0 255
	Name: FlowerPan Frames: 600 Motion: Slow / little Description: A pan across Specific features: Three b insects on the flowers.	ands of flowers: red, yel	low, and indigo; flying	# 10 ⁵ 2 1 0 0 255
	Name: HoneyBee Frames: 600 Motion: Fast / little Description: A focus on a Specific features: Fixed of in the foreground and the little	amera focused on the mackground.	niddle; blurred flowers	# 10 ⁴ 10 0 255
Nghuk	Name: Jockey Frames: 600 Motion: Fast / plenty of Description: A galloping specific features: Came background; the focus cha	era tracks the horse	causing fast moving	# 10 ⁴ 10 5 0 0 255

Table 2. Test sequence characteristics in the proposed UVG data set (sequences 9 - 16).

Snapshots Snapshots	characteristics Characteristics	RGB color histogram		
Page 1980 and the same of the	Characteristic and description Name: Lips Frame rate: 120fps Resolution: 2160p			
**CORCA National	Frames: 600 Duration: 5s Bit depth: 8 and 10 Motion: Slow / little Texture: Smooth Contrast: High Description: A close-up of a female putting red lipstick with a brush. Specific features: Part of the face in the foreground is out of focus; high contrast between black background and female face.			2 # 10 ⁵ 1 0 0 255
	Name: RaceNight Frames: 600 Motion: Fast / plenty of Description: A group of h Specific features: Came sequence making the back	ra tracks the horses in ground move; static can	n the first half of the nera in the second half.	2 # 10 ⁵ 1 0 255
Turke Joke Amuse 1 S 7 O 11 12	Name: ReadySetGo Frames: 600 Motion: Fast / plenty of Description: Horses go of Specific features: Camer sideways movement.			2 # 10 ⁵ 1 0 255
	Name: RiverBank Frames: 600 Motion: Slow / little Description: A slow pan a Specific features: Ripplic composed of detailed trees	ng water sparkling wit	Resolution: 2160p Bit depth: 8 and 10 Contrast: High th the sun; background	# 10 ⁴ 10 5 0 0 255
STAN AMARITA	Name: ShakeNDry Frames: 300 Motion: Slow / plenty of Description: A dog shakir Specific features: Water scene at the end of the seq	ng itself dry in a forest. drops flying around; th	Resolution: 2160p Bit depth: 8 and 10 Contrast: Low ne dog leaves the static	2 # 10 ⁵ 1 0 0 255
	Name: SunBath Frames: 300 Motion: Slow / plenty of Description: Tree branche Specific features: Part of background overexposed f	es swinging in front of the branches coming in from direct sunlight.	n and out of the scene;	2#10 ⁶ 1 0 0 255
Nighturk	Name: Twilight Frames: 600 Motion: Slow / plenty of Description: Arrival of a t Specific features: Dark so the sequence.	tram at twilight.	Resolution: 2160p Bit depth: 8 and 10 Contrast: High the scene at the end of	# 10 ⁵ 5 0 0 255
	Name: YachtRide Frames: 600 Motion: Fast / plenty of Description: A close-up o Specific features: Camera sun; the boat leaves the sco	a focused on foam and v	water sparkling with the	# 10 ⁵ 2 1 0 0 255

Indov	Saguanaa	VTM8.0 – QP 32			HM16.20 vs. VTM8.0			
Index	Sequence	Kbps	PSNR _Y	$PSNR_U$	PSNR _v	BD-BR _Y	$BD-BR_U$	BD - BR_V
U1	Beauty	7337	34.85	37.08	39.46	-22.99%	-34.39%	-34.13%
U2	Bosphorus	5323	41.02	47.62	46.55	-25.70%	-65.02%	-55.12%
U3	CityAlley	1050	40.49	48.10	47.18	-30.91%	-56.13%	-58.07%
U4	FlowerFocus	635	39.86	44.26	44.33	-50.15%	-55.42%	-61.45%
U5	FlowerKids	3729	39.98	43.77	45.27	-32.34%	-39.08%	-45.56%
U6	FlowerPan	2778	38.21	40.89	41.91	-38.92%	-38.44%	-40.05%
U7	HoneyBee	2756	38.90	42.70	42.51	-45.49%	-37.39%	-38.43%
U8	Jockey	4432	39.24	43.74	43.15	-46.58%	-51.95%	-50.70%
U9	Lips	3662	35.06	37.62	40.37	-46.60%	-74.29%	-53.50%
U10	RaceNight	4621	36.27	43.85	43.49	-30.81%	-46.22%	-38.20%
U11	ReadySetGo	11449	40.14	44.49	44.14	-32.12%	-41.82%	-40.90%
U12	RiverBank	7860	36.14	45.41	45.09	-23.55%	-69.34%	-71.26%
U13	ShakeNDry	13852	37.71	41.97	42.21	-24.12%	-34.61%	-41.98%
U14	SunBath	4065	42.83	46.85	48.87	-32.35%	-40.34%	-44.76%
U15	Twilight	1208	40.52	48.28	47.60	-31.94%	-62.63%	-58.28%
U16	YachtRide	19105	38.54	45.60	44.44	-27.18%	-63.71%	-60.65%
C1	Tango2	3662	38.86	47.58	45.07	-37.37 %	-54.83 %	-51.06 %
C2	FoodMarket4	3968	41.11	45.78	46.12	-36.28 %	-36.66 %	-38.50 %
C3	Campfire	7028	36.51	36.10	39.70	-38.34 %	-29.29 %	-48.19 %
C4	CatRobot	4186	38.34	40.8	41.25	-44.09 %	-52.79 %	-46.38 %
C5	DaylightRoad2	4151	36.41	43.97	41.38	-43.13 %	-49.11 %	-39.88 %
C6	ParkRunning3	19177	36.52	32.85	34.37	-37.29 %	-15.87 %	-14.33 %

Table 3. Rate-Distortion characterization of the UVG and CTC datasets.

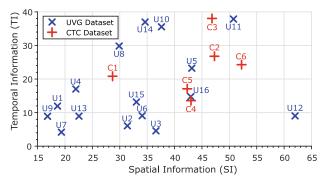


Figure 1: SI and TI for the UVG and CTC video sequences.

Figure 1 plots the SI and TI scores for the proposed UVG video sequences (U1-U16) in blue and the 2160p CTC sequences (C1-C6) in red. The indexes are linked to associated sequences in Table 3. The results illustrate that the proposed sequences are distributed across a wide range of SI and TI scores. This diversity makes the UVG dataset highly complementary to that of the CTC.

3.2 Rate-Distortion Characterization

Table 3 presents the RD characterization results of the UVG and CTC 2160p sequences with VTM8.0 and HM16.20. The four columns in the middle report the bitrate in *kilobits per second* (*Kbps*) and distortion in *peak signal-to-noise ratios* (*PSNRs*) for all sequences when encoding them with VTM8.0 and the QP value of 32. PSNR results (in dBs) are separately given for the luma (*Y*) and two chroma (*U* and *V*) components. The used coding configuration is *Random Access* (*RA*) specified in the CTC [18], [19].

Encoding the UVG dataset with VTM8.0 produced bit rates from 635 to 19105 Kbps and luma PSNRs from 34.85 to 42.83 dB

for a QP value of 32. The respective metrics with the CTC dataset were from 3662 to 19177 Kbps and from 36.41 to 41.11 dB. The increased variances in rate and distortion attest to the content diversity of the UVG dataset in VVC encoding.

Furthermore, the rightmost three columns of Table 3 report the coding gain of VTM8.0 over HM16.20 for the same objective visual quality measured in PSNR. In practice, the bitrate gains were computed with a popular *Bjøntegaard Delta Bit Rate* (*BD-BR*) [25] metric under the RA coding configuration. BD-BR gives the difference in percent between the RD curves of VTM8.0 and HM16.20, which were interpolated through four RD points that correspond to the QP values of 22, 27, 32, and 37.

On average, VTM8.0 saves -39.42% (from -36.28% to -44.09%) of BD-BRy over that of HM16.20 with the CTC dataset and the respective average is -33.86% (from -22.99% to -50.15%) with the UVG dataset. These results show that our dataset also provides edge cases (U1 and U4) beyond those of the CTC (C2 and C4) for coding efficiency comparison.

3.3 Spatial and Temporal Encoding Complexity

Our characterization is complemented by the *spatial encoding* complexity (S_{cplx}) and temporal encoding complexity (T_{cplx}) calculations based on [24]. They were yielded by encoding the sequences with HM16.20 using the QP values of 22, 27, 32, and 37 under the *All Intra* (*AI*) and RA configurations [18], respectively.

For a given sequence, S_{cplx} is calculated by averaging the bit counts over the four QP values in the AI case and normalizing the result (AI_{size}) with the bit count of the uncompressed 10-bit 4:2:0 YUV sequence (U_{size}) as

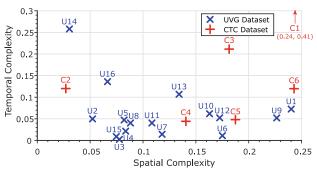


Figure 2: Complexity scores of the UVG and CTC sequences.

$$S_{cplx} = \frac{AI_{size}}{U_{size}}. (3)$$

Correspondingly, T_{cplx} is obtained by first averaging the bit counts over the four QP values in the RA case and then computing the ratio between average bit counts of bi-predicted inter frames (B_{size}) and intra frames (I_{size}) as

$$T_{cplx} = \frac{\operatorname{avg}(B_{size})}{\operatorname{avg}(I_{size})}.$$
 (4)

The intra period was set to 32, so every 32^{th} frame was intra frame and the others were bi-predicted inter frames. The normalization by I_{size} value takes into account the correlation between B_{size} and I_{size} values. High I_{size} values tend to increase B_{size} values because the high spatial complexity increases residuals in inter prediction.

Figure 2 plots the S_{cplx} and T_{cplx} scores for the UVG and CTC 2160p sequences (see Table 3) in blue and red, respectively. Again, the UVG sequences are spread across the entire (S_{cplx}, T_{cplx}) plane of encoding complexities. These results are well in line with our previous findings on the usefulness of the UVG dataset.

4 Conclusion

This paper presented our open UVG dataset that is composed of 16 versatile 4K (2160p) natural 50/120 fps video sequences in 8-bit and 10-bit 4:2:0 YUV formats. This online dataset is available on our website (ultravideo.cs.tut.fi). To the best of our knowledge, this is the first dataset that includes 4K 120fps raw videos, which make it particularly relevant for future video codec development, where the inevitable trend is towards higher resolutions and frame rates. Our dataset is also validated to complement the existing CTC 4K test set in terms of VVC and HEVC coding characteristics and complexity. Therefore, we believe it would diversify objective and subjective quality assessment in VVC standardization and related application development.

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