DCC2023 Poster Session

Temporal Down-sampling based Video Coding with Frame-Recurrent Enhancement

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Proposals

Experimental results

Coclusion



Traditional video coding VS Downsampling-based coding method

Traditional video codig method

Downsampling-based coding method







Related works—Downsampling-based coding method



Super-resolution (SR)

Unlike these works, we propose a temporal down sampling method to select down-sampled frames and use inter-frame information for enhancement.

- Ho, M. M., He, G., Wang, Z., & Zhou, J. (2020, January). Down-sampling based video coding with degradation-aware restorationreconstruction deep neural network. In International Conference on Multimedia Modeling (pp. 99-110). Springer, Cham.
- Fatemeh Nasiri, Wassim Hamidouche, Luce Morin, Gildas Cocherel, and Nicolas Dhollande, "A study on the impact of training data in cnn based super-resolution for low bitrate end-to-end video coding," in 2020 Tenth International Conference on Image Processing Theory, Tools and Applications (IPTA). IEEE, 2020, pp. 1–5.





Related work—EDVR: Video Restoration with Enhanced Deformable Convolutional Networks



Xintao Wang, Kelvin CK Chan, Ke Yu, Chao Dong, and Chen Change Loy, "Edvr: Video restoration with enhanced deformable convolutional networks," in Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition Workshops, 2019, pp. 0–0.



Proposals



II The Proposal 1—TDS Framework

Temporal Down-sampling based Video Coding



(a) Encoder structure



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II The Proposal 1—TDS Framework

Temporal Down-sampling based Video Coding





The processing steps of EDVR for deblurring

(b) Decoder structure

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Original All intra(AI) configuration **VS** Our proposal modules based on the AI configuration







The Proposal 2—SRFR Framework



Up-sampling with frame-Recurrent enhancement



Enhancing the middle frame by referring to the first and last frames

The second frame is enhanced by referring to the first and third frames, and the fourth frame is enhanced by referring to the third and fifth frames.

The essence of enhancement is to refine the current frame by referring to nearby frames.



Experimental Results

Experimental Results







Figure 2. Rate-distortion curves for comparing VVC and our method. Results for sequences compressed by All Intra: (a) Fourpeople; (b) Partyscene; (c) KristenAndSara; (d) BQSquare.



Experimental Results





The second frame





Bitrate:1122.384 PSNR:30.1601 SSIM: 0.8928





Bitrate: 1122.384 PSNR: 32.3399 SSIM: 0.9004



The second frame



PSNR:23.7439

Bitrate:2603.088 PSNR:23.9060 SSIM: 0.7190 Bitrate:2603.088 Bi



PSNR:32.1974

SSIM: 0.8912

Bitrate:3122.208 PSNR:33.3524 SSIM: 0.9076



Bitrate:2603.088 PSNR:34.5032 SSIM: 0.9279



The second frame

PartyScene.



Bitrate:4934.4 PSNR:19.0278 SSIM: 0.3392



Bitrate:4934.4 PSNR:27.2456 SSIM: 0.8839

PSNR:31.9639

SSIM: 0.8920



Bitrate:6136.848

PSNR:26.8176

SSIM: 0.8259

Bitrate:4934.4

PSNR:28.6274 SSIM: 0.8784

Original Image (a) GT (b) Bicubic (c) EDVRDeblur (d) VVC (e) Ours **Figure 3.** Subjective results comparisons between the proposed method and the other three methods in PSNR (dB)/SSIM. Results for sequences: (1) KristenAndSara; (2) Fourpeople; (c)





Resolution	Sequence	Low Delay P		All Intra	
		BD-rate	BD-psnr	BD-rate	BD-psnr
1920x1080	Cactus	-60.47	0.468	-59.803	1.933
	BQTerrance	-39.261	0.543	-77.574	3.776
832x480	BQMall	-61.248	2.115	-75.61	5.102
	partyScene	-82.764	4.705	-85.455	7.238
416x240	BQSquare	-58.396	2.332	-80.975	7.677
1280x720	KristenAndSara	-65.751	4.517	-68.7	5.296
	Fourpeople	-58.582	0.679	-73.889	6.673
	Johnny	-61.863	0.679	-63.591	3.956
Average		-61.041875	2.00475	-73.199625	5.206375

Table 1. Objective comparison between our proposed method using AI configuration and the standard VVC in QPs = {22,27,32,37,42,47}. Ours outperforms VVC in the average BD-rate, BD-psnr. The table shows the results of the calculations for middle down-sampled frames.



Conclusion

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IV Conclusion



We design a temporal down-sampling based video coding framework (TDS).It can be combined with all the existing coding standards.

A method of super-resolution with frame recurrent image enhancement (SRFR) is applied to improve the downsampled frames. The temporal information from high resolution frames can be fully used to improve the video quality through frame recurrent.

When compared to VVC, the BD-rate of the down-sampled frames can be reduced from 39.261% ~ 85.455 % in AL and LDP configuration.





For your listening

