

Poster: Device-Cloud Collaborative Video Storage for **Real-time Ultra HD Video Playback**



(a) Storage Phase

SR-int

► E Reference-based Online QoE-based ral Enhance Key Frame Planne Decoder

(b) Playback Phase

Figure 1: DuoSR System Architecture

ABSTRACT

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With the rapid advancement of mobile hardware, smartphones are now capable of capturing and playing videos in 4K and even 8K resolution, improving immersive and enjoyable viewing experience for users. However, storing these ultra-high-resolution videos poses a significant burden on the local storage of mobile devices. An alternative method is storing video on cloud with the benefits of scalable storage space, cross-device data access, data sharing and backup. Recently, the policy has been integrated into mainstream mobile operation systems. Nevertheless, downloading video from cloud may face high latency under poor network conditions, significantly diminishing the user experience.

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Recent advances in client-side computation create a new opportunity for utilizing network-based super-resolution techniques to improve the quality of the displayed lowresolution videos during playback[1]. However, super-resolution is an ill-posed problem, indicating that even the same lowresolution video may correspond to multiple high-resolution videos. This implies that relying solely on low-resolution information to recover high-definition details is a challenging task.

Studies have revealed that the utilization of reference high-definition information can significantly enhance the effectiveness of super-resolution algorithms[2]. We observe that a device-cloud collaborative paradigm offers new opportunities to address the challenges of storing and playing high-resolution videos. By integrating cloud storage with ondevice video enhancement and using high-resolution patches

from the cloud as reference information, the quality of displayed local video can be significantly improved, thereby enhancing the playback experience.

We propose the **DuoSR** system, which encompasses two phases: storage phase and playback phase.

In the storage phase (Figure 1a), We have designed a downsampling neural network and joint training it with the superresolution network to optimize video compression. Concurrently, we design a neural-enhanced quality prediction module to predict the video quality under different reference patch and different key frame. The module generates a quality prediction table, facilitating cloud quality planner during the playback phase.

In the playback phase (Figure 1b), initially, when the user accesses a video, real-time device and network information are uploaded to the cloud. Based on these data, the cloud, aiming to maximize the Quality of Experience (QoE), determines the key frames for super-resolution and the reference information to be transmitted. Upon receiving the information about the key frames and the reference information on the device, RefVSR is performed on the key frames to reconstruct high resolution key frames. For non-key frames, we reuse the results from key frames, employing the video codec's motion vectors and residual information to upsample them.

REFERENCES

- [1] Y. Huang, Y. Chen, and et al. 2021. Video rescaling networks with joint optimization strategies for downscaling and upscaling. In CVPR. 3527-3536
- [2] Y. Jiang, K. Chan, and et al. 2022. Reference-based Image and Video Super-Resolution via C^2 -Matching. *TPAMI* (2022).

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