## Image and Video Compression using Generative Sparse Representation with Fidelity Controls

Lebin Zhou, Wei Wang, Wei Jiang

Futurewei Technologies Inc., San Jose, CA

LQ Features  $F_t^{LQ}$ 



## Introduction

**ACCV HANOI F** 2024 VIETNAM **DEC 8-12** 

Dual-stream framework for learned image compression (LIC) and video compression (LVC) by Sparse Visual Representation (SVR)

- Main stream: high-quality (HQ) codebook-based SVR for HQ baseline reconstruction with high perceptual quality
- Control stream: fidelity-preserving controls from low-quality (LQ) input to guide conditional generation in main stream
  Advantages over conventional pipeline:
  Balanced perceptual quality and reconstruction fidelity with flexible bitrate control



- Fully recovered SVR-based reconstruction of all frames that mitigates error propagation
- Unified pipeline for both LIC and LVC
- General framework to accommodate different conventional fidelity-preserving image/video compression tools



## Experiments

LIC: JPEG-AI Dataset

## LQ fidelity-preserving LQ by MLIC [1] SVR codebook from AdaCode [2]

LVC: Mixed set from AOM, JVET, MPEG, AVS LQ fidelity-preserving by VVC (qp=42) SVR codebook from FeMaSR [3]

[1] Jiang, W., et al. Mlic: Multi-reference entropy model for learned image compression. ACM Multimedia (2023)
[2] Liu, K., Jiang, Y., Choi, I., Gu, J.: Learning image-adaptive codebooks for class- agnostic image restoration. ICCV (2023)
[3] Chen, C., et al., Real-rorld blind super-resolution via feature matching with implicit high-resolution priors. ACM Multimedia (2022)

![](_page_0_Picture_18.jpeg)

LPIPS | PSNR | SSIM 0.383 | 30.02 | 0.827 0.133 | 30.57 | 0.825 0.133 | 30.48 | 0.826

![](_page_0_Picture_20.jpeg)

LPIPS | PSNR | SSIM 0.534 | 30.51 | 0.825 0.149 | 32.48 | 0.858 0.152 | 32.02 | 0.857

![](_page_0_Picture_22.jpeg)

LPIPS | PSNR | SSIM 0.304 | 28.26 | 0.812 0.100 | 28.34 | 0.834 0.100 | 27.80 | 0.840