

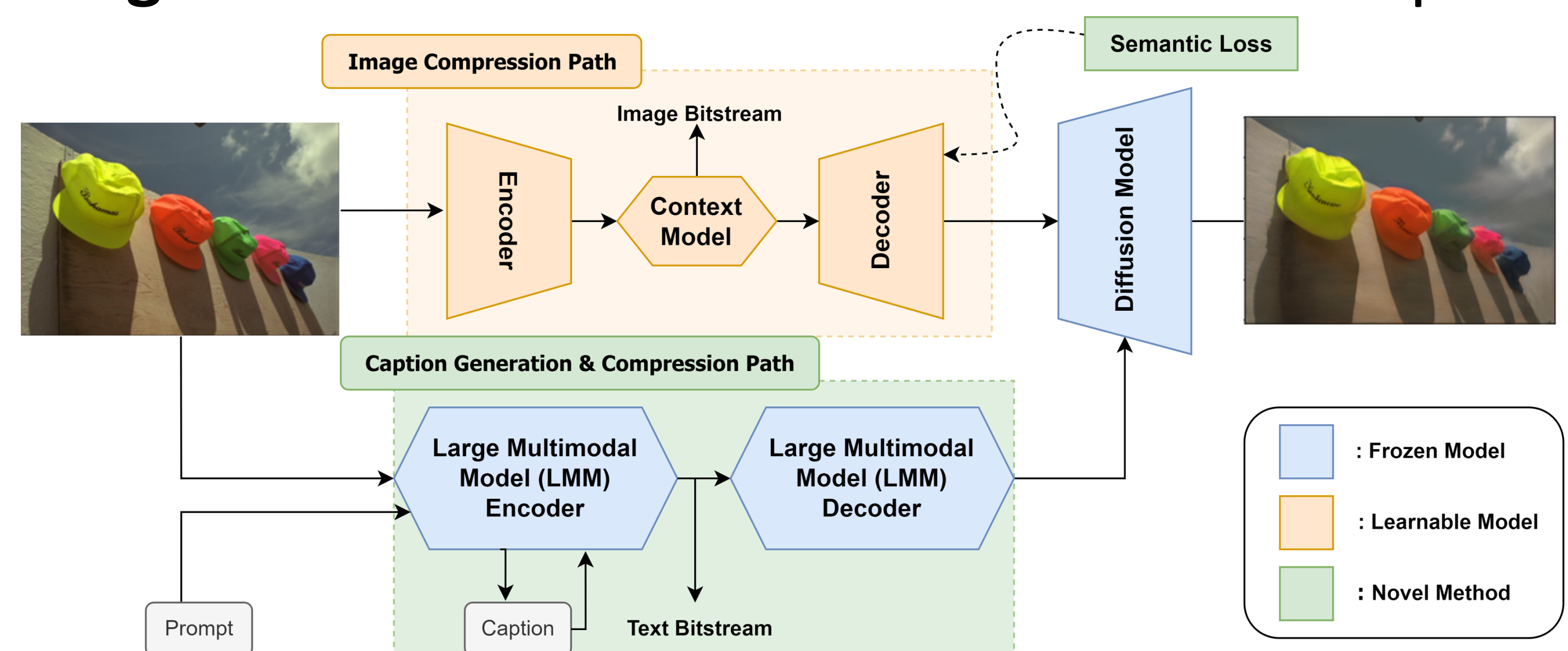
Background

- **Learned Image Compression** is a technique to design neural network-based non-linear transform and entropy model to compress images
 - Its sub-domain, **ultra low-bitrate compression** model, targets to less than **0.1 bpp** while keeping good perceptual quality
 - Some models utilize **text caption** as a sub-information that captures semantic information (MISC[3], Text & Sketch[4])
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- Recently, **Large Language Model** is shown to be effective in text compression and data compression (LLM-ZIP[1], Language Modeling is Compression[2])
- | | T ₁ | T ₂ | T ₃ | T ₄ |
|----------------|----------------|----------------|----------------|----------------|
| Input token | A lighthouse | with | yellow | brick |
| Encoded Number | 7 | 3 | NaN | NaN |

$P(T_i | T_1) = \text{'is ' : 0.5, 'of ' : 0.2, 'with ' : 0.15, ...}$
'with ' is encoded to number 3
- How can we integrate **LLM based compression** in the workflow of **Learned Image Compression**?

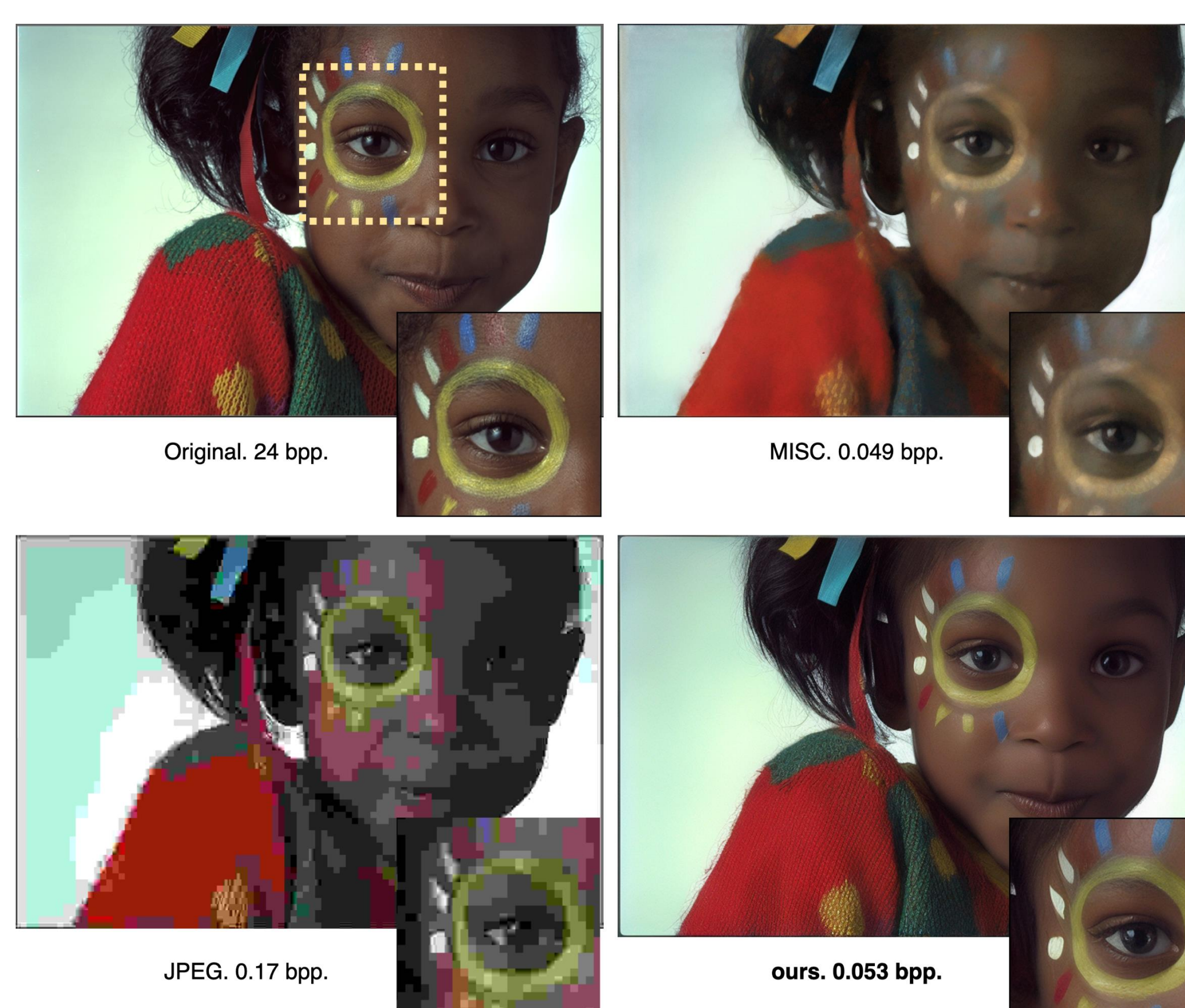
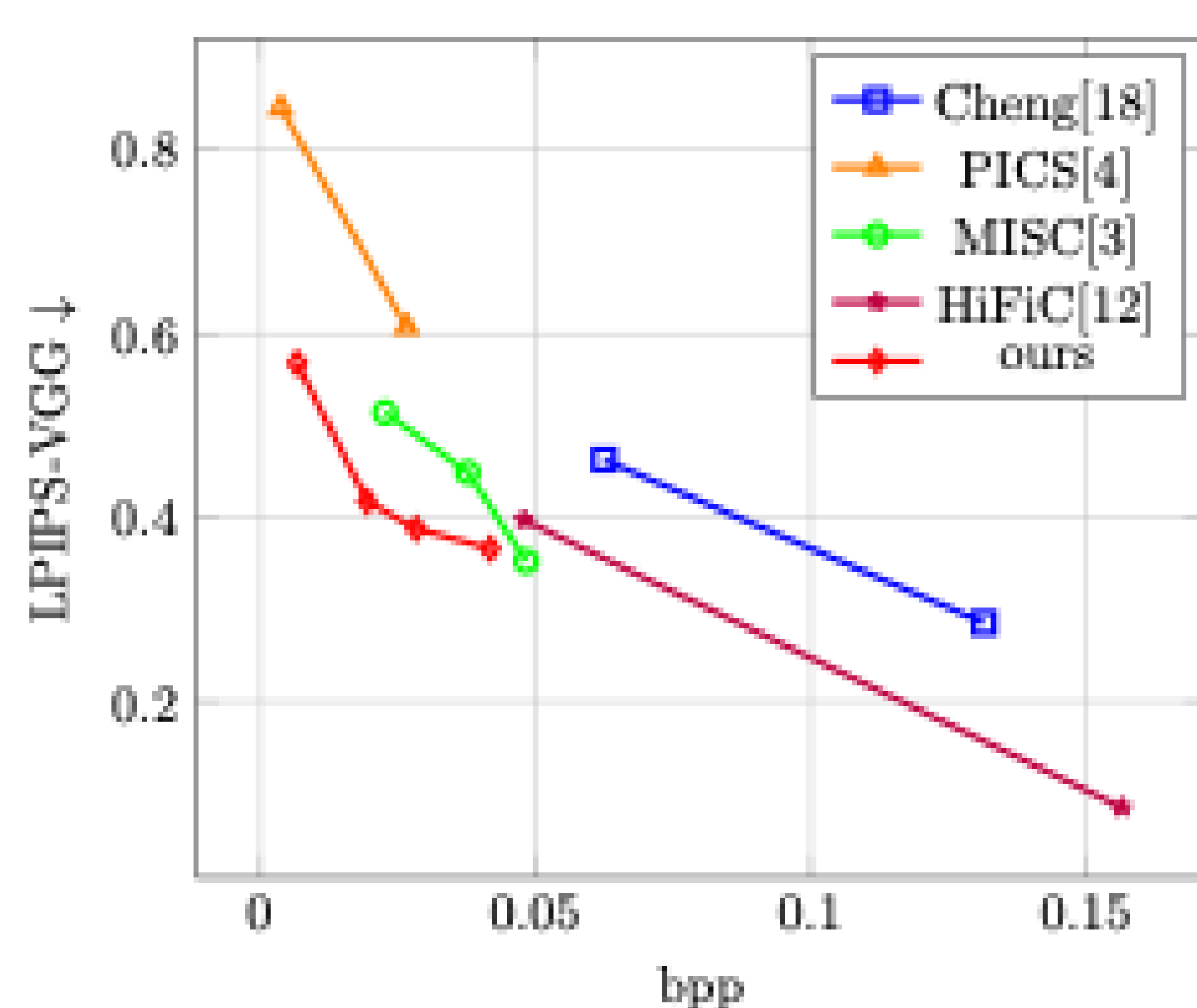
Methodology

- We generate text caption, and compress it into bitstream **with one large multimodal model**
 - Input images are fed to fine-tuned **low-bitrate image compressor**
 - Output (distorted) images are then refined with **generative model** conditioned with the caption
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- Our contributions are:
 - We show that **captioning and its compression** can be done in **one LMM**
 - Developed efficient fine-tune methods with **perceptual and semantic loss**



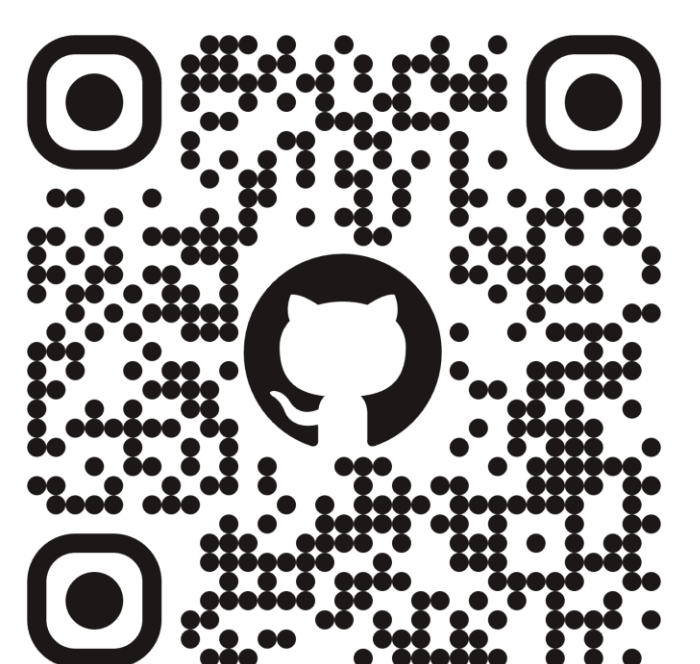
Results & Takeaways

- Our model **achieves more than 65% text compression ratio** and more than **40% bitrate saving** than existing methods (in LPIPS BD rate)
- Example images show our methods eliminate color distortion under ultra low-bitrate (0.053 bpp!)

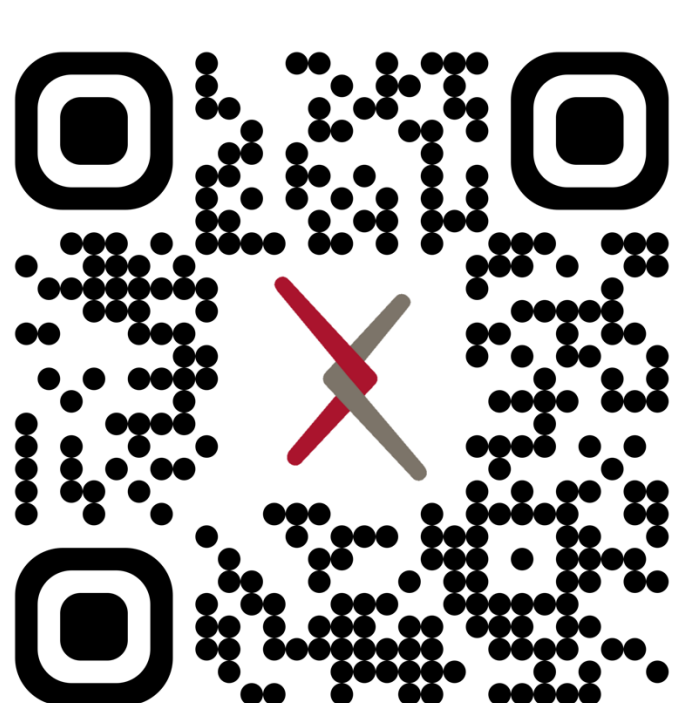


Code & Demo
Available on

GitHub



arXiv



[1] C. S. K. Valmeekam, K. Narayanan, D. Kalathil, J.-F. Chamberland, and S. Shakkottai, "LLMZip: Lossless Text Compression using Large Language Models," arXiv preprint arXiv:2306.04050, June 2023.

[2] G. Deletang, A. Ruoss, P.-A. Duquenne, E. Catt, T. Genewein, C. Mat-tern, J. Grau-Moya, L. K. Wenliang, M. Aitchison, L. Orseau, M. Hutter, and J. Veness, "Language modeling is compression," in ICLR, 2024.

[3] C. Li, G. Lu, D. Feng, H. Wu, Z. Zhang, X. Liu, G. Zhai, W. Lin, and W. Zhang, "MISC: Ultra-low Bitrate Image Semantic Compression Driven by Large Multimodal Model," arXiv preprint arXiv:2402.16749, no. arXiv:2402.16749, 2024.

[4] E. Lei, Y. B. Uslu, H. Hassani, and S. S. Bidokhti, "Text + sketch: Image compression at ultra low rates," ICML 2023 Workshop Neural Compression, 2023.

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