## **Imperial College** London



# Neural Distributed Image Compression with **Cross-Attention Feature Alignment**



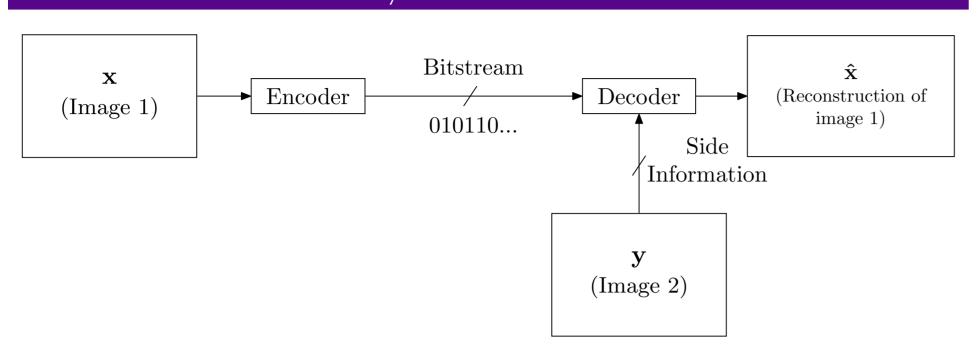


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#### System Model



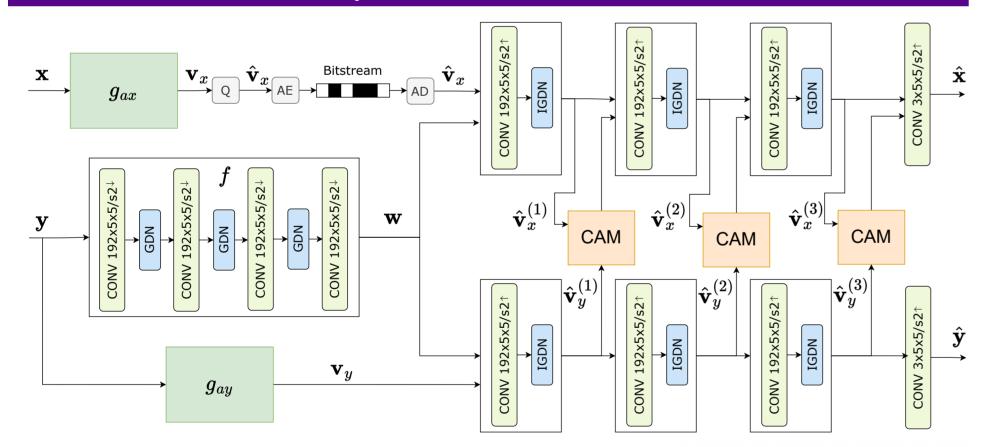
Motivation: Decoder-only side information can provide immense reductions in the transmission rate in lossy compression schemes [1]!

Real-life applications include distributed sensor networks (e.g., autonomous vehicles, multiple static cameras, unmanned aerial vehicles).

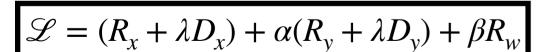
#### Distributed Image Compression

- DSIN [2]: Finds corresponding patches to refine the reconstructed image.
- NDIC [3]: Extracts ``common information" between correlated images.
- Our work, ATN: Employs cross-attention modules (CAMs) to align intermediate latent representations.
  - Computes the attention globally, between patches of the latent representations over all channels, similarly to [4].
  - This is similar to *patch-matching* idea in [2], but our method provides a *differentiable* alternative to search-based algorithm used in [2].

#### Proposed Architecture



- w common features of two images
- $\mathbf{v}_{\chi}$ ,  $\mathbf{v}_{\chi}$  local/private features
- Extract w from y, send only  $\mathbf{v}_{x}$
- Align intermediate latents  $\mathbf{v}_{\chi}^{(i)}$  and  $\mathbf{v}_{v}^{(i)}$  (in  $i^{th}$  layer) using cross-attention module (CAM)
- Generate <u>query</u>  $\mathbf{Q}_{x}$  from  $\mathbf{v}_{x}^{(i)}$ ,  $\underline{key}~\mathbf{K}_{\mathrm{v}}$  and  $\underline{value}~\mathbf{V}_{\mathrm{v}}$  from  $\mathbf{v}_{\mathrm{v}}^{(i)}$  (all learnable weight matrices!)

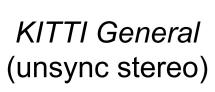


# $\mathbf{v}_{CAM}^{(i)}$ $\mathbf{W}_y^K$ Patch embedding embedding

#### **Experimental Setup**

KITTI Stereo (sync stereo)

Cityscape (sync stereo)

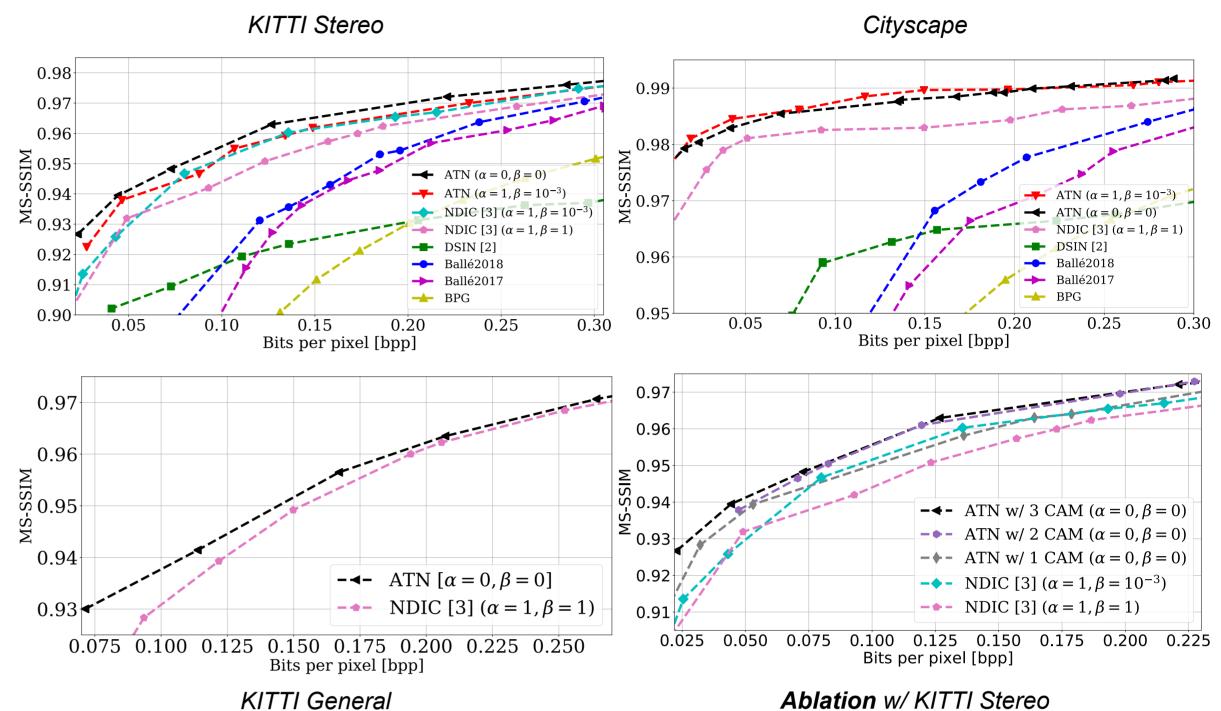






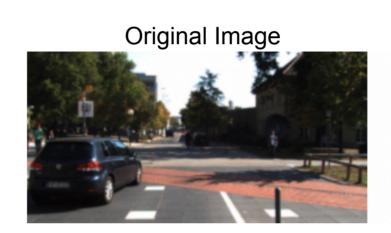


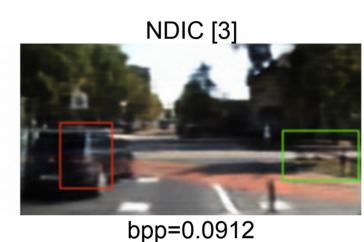
#### Rate-Distortion Performances

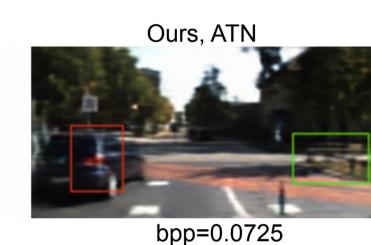


#### Visual Examples

Synchronized stereo cameras







Unsynchronized stereo cameras







#### Common and Local Information

NDIC [3]





bpp=0.0911



Local Information

**Images** 

Common

Information





### References

- [1] A. Wyner and J. Ziv, "The rate-distortion function for source coding with side information at the decoder", IEEE Trans. Inf. Theory, 1976.
- [2] S. Ayzik and S. Avidan, "Deep image compression using decoder side information", ECCV, 2020.
- [3] N. Mital, E. Ozyilkan, A. Garjani, and D. Gunduz, "Neural distributed image compression using common information", DCC, 2022.
- [4] A. Dosovitskiy, L. Beyer, A. Kolesnikov, D. Weissenborn, X. Zhai, T. Unterthiner, M. Dehghani, M. Minderer, G. Heigold, S. Gelly, J. Uszkoreit, and N. Houlsby, ``An image is worth 16x16 words: Transformers for image recognition at scale. In 9th International Conference on Learning Representations", ICLR, 2021.