

# Deep Correlated Image Set Compression Based on Distributed Source Coding and Multi-Scale Fusion

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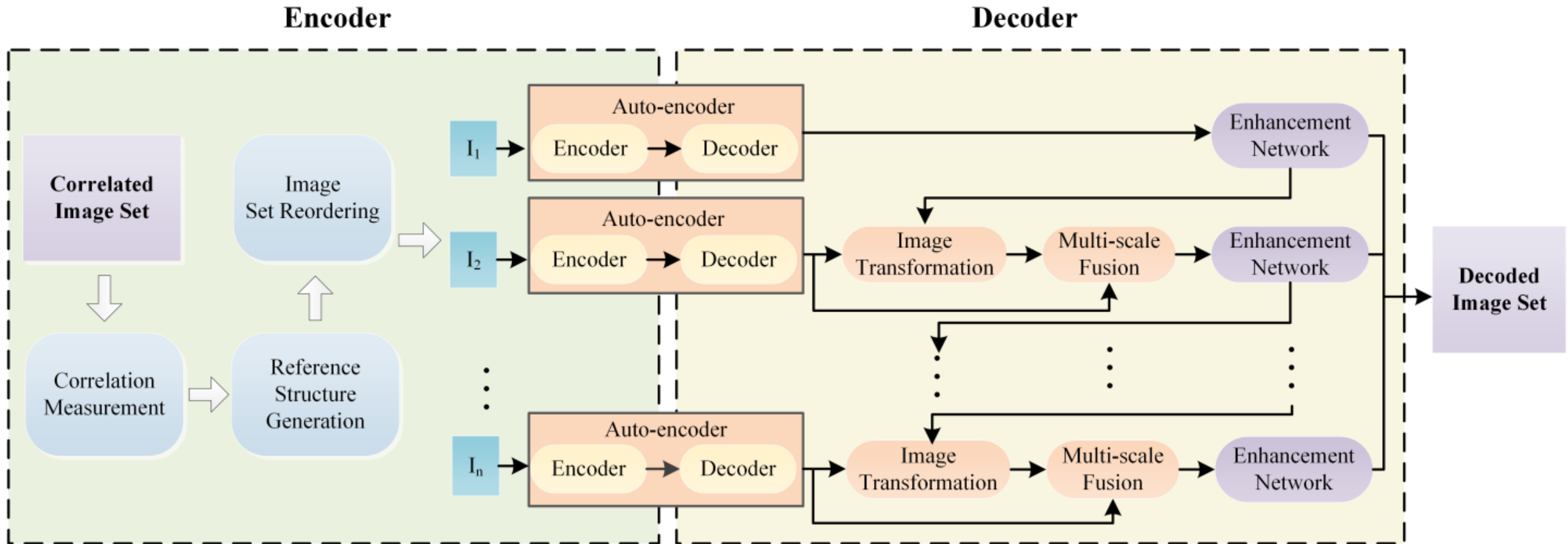
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Santa Clara University

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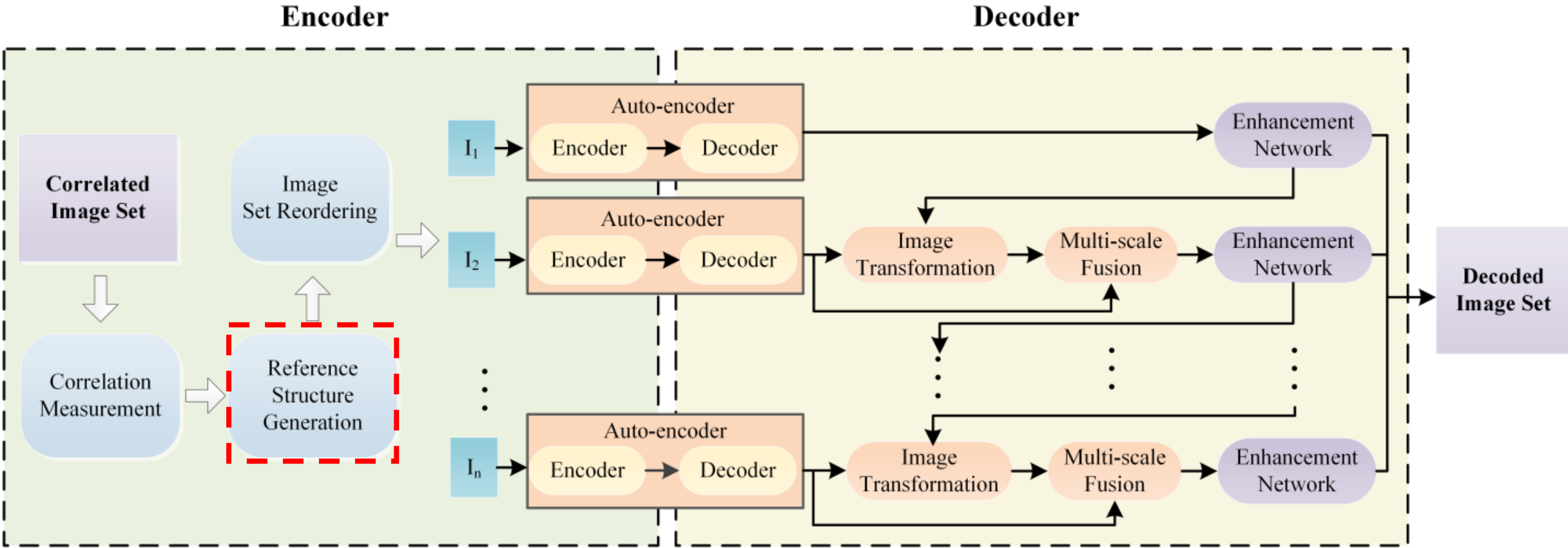
# Contributions

- A deep similar image set compression scheme based on DSC and multi-scale fusion is proposed
- A side information generation scheme is proposed based on the decoded reference image
- An image enhancement network is proposed to further improve the quality of reconstructed image
- Experimental results validate the superior performance of both subjective and objective quality

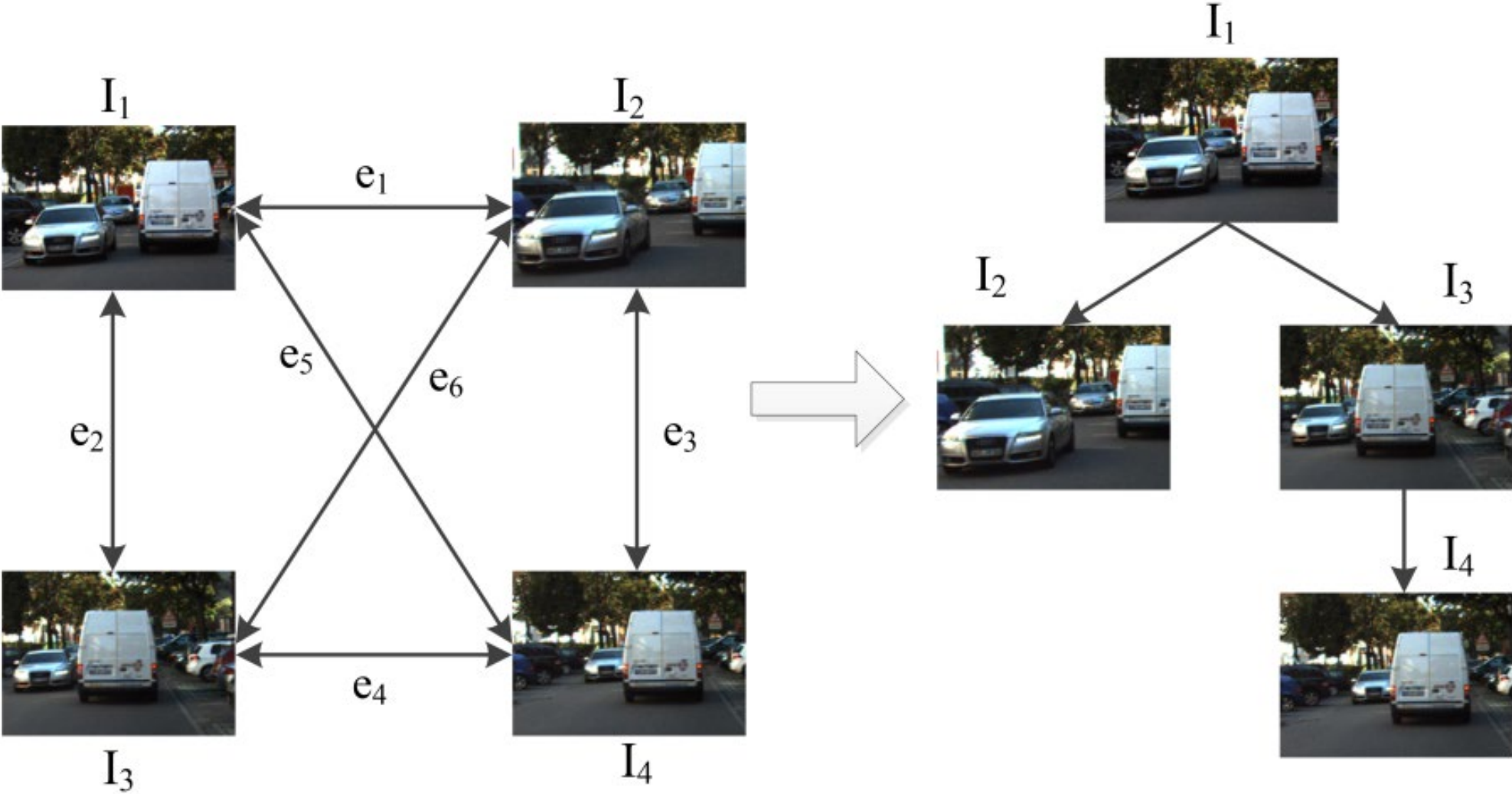
# Network Architecture



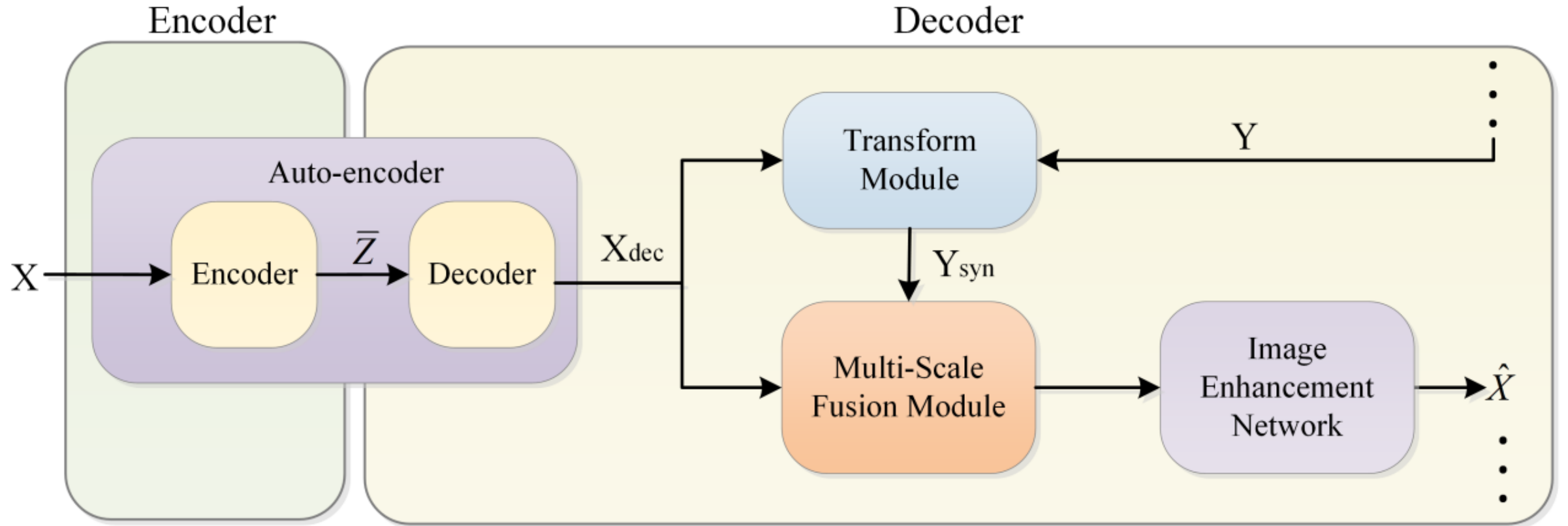
# Generation of Reference Structure



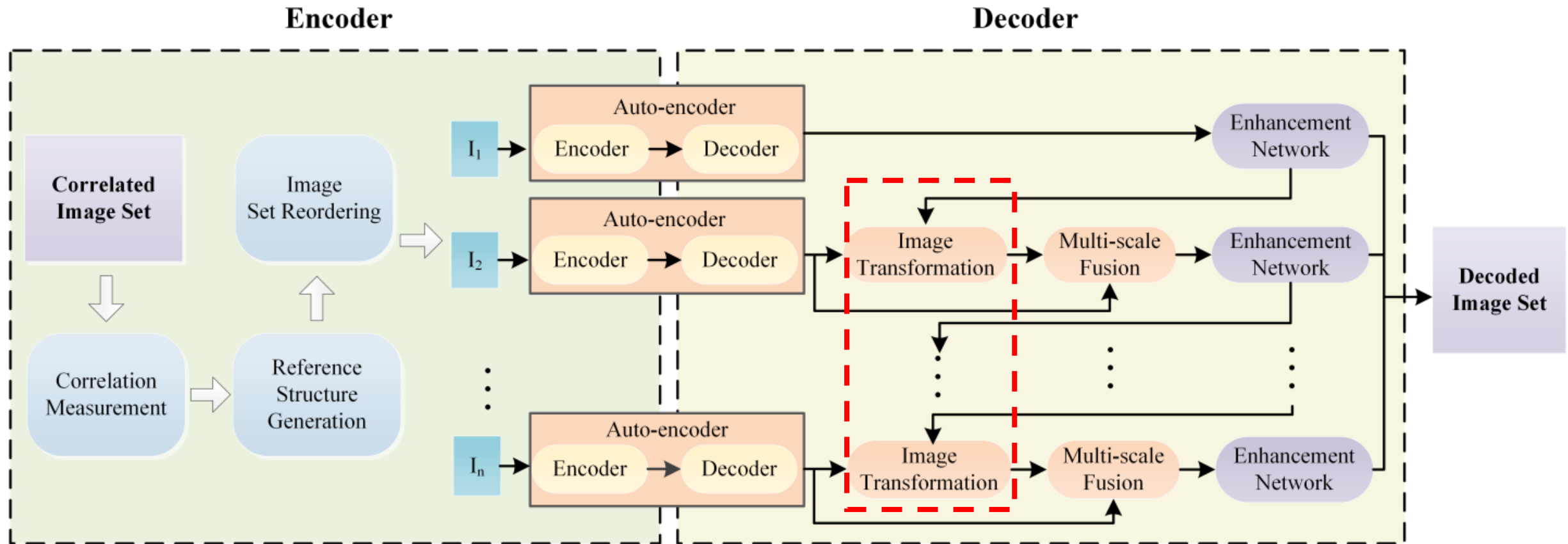
# Feature-based MST Generation



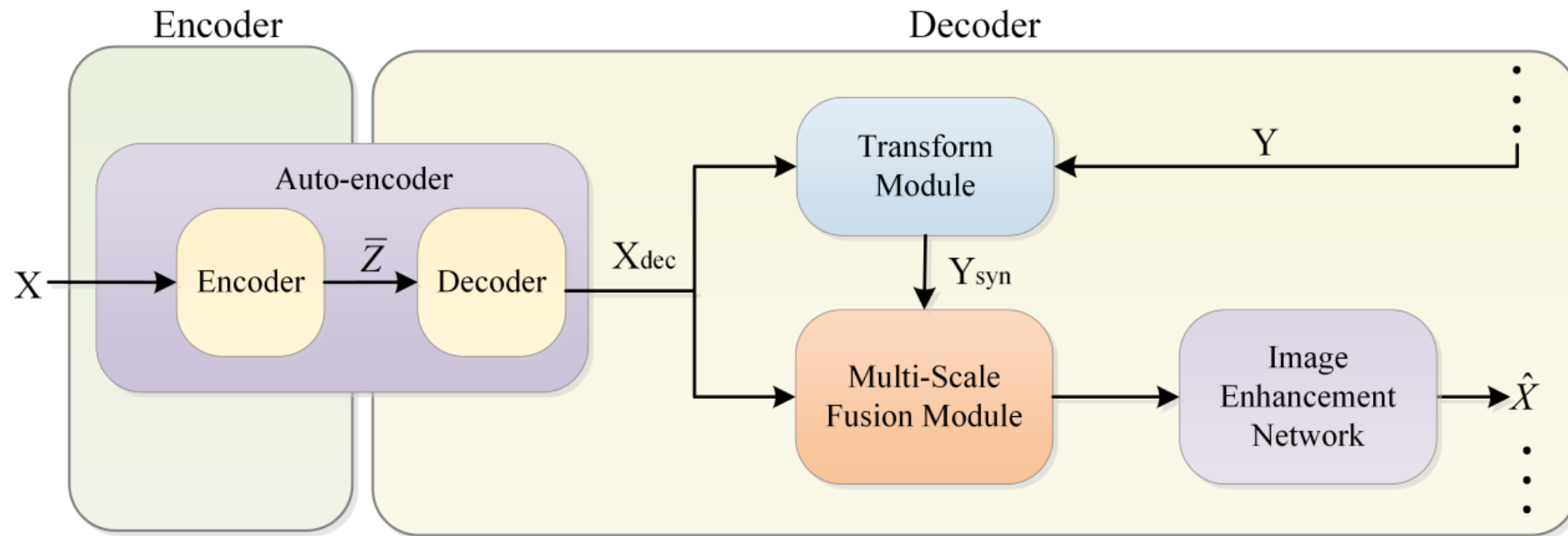
# Detailed Image Compression Module



# Reference Image Transform Module



# Reference Image Transform Module

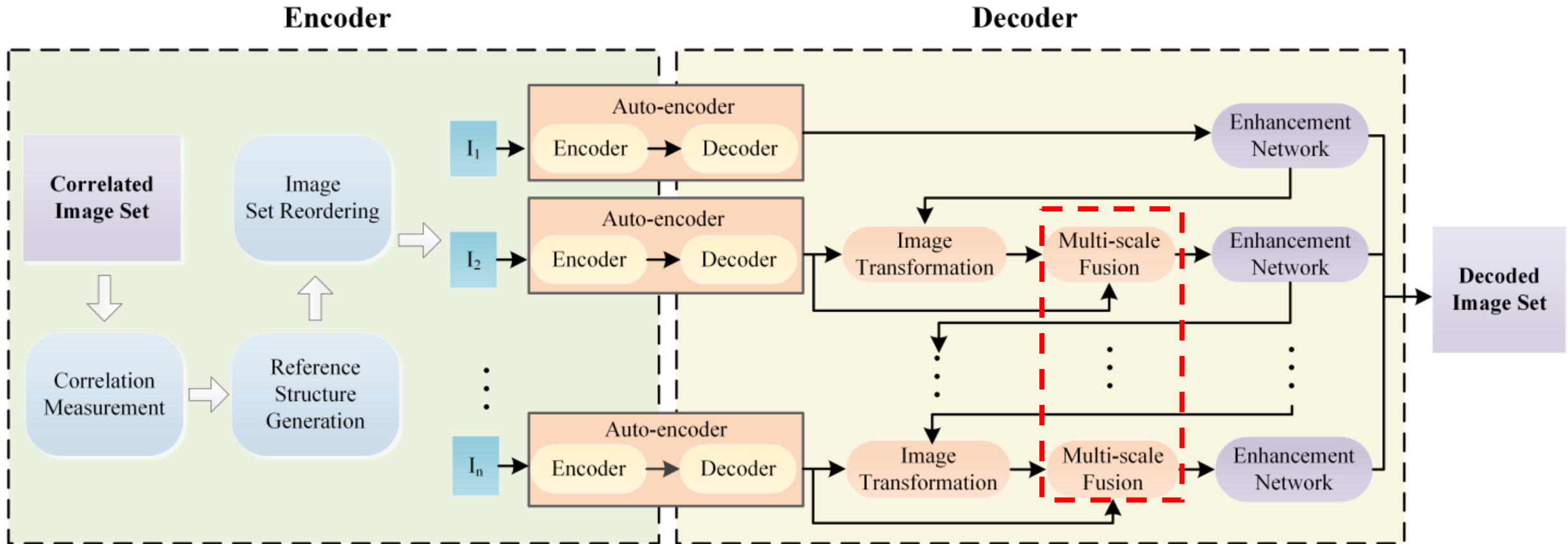


$$f(i) = \arg \max_j \text{corr} (\pi (X_{dec}(i)), \pi (Y_{dec}(j))) \quad (1)$$

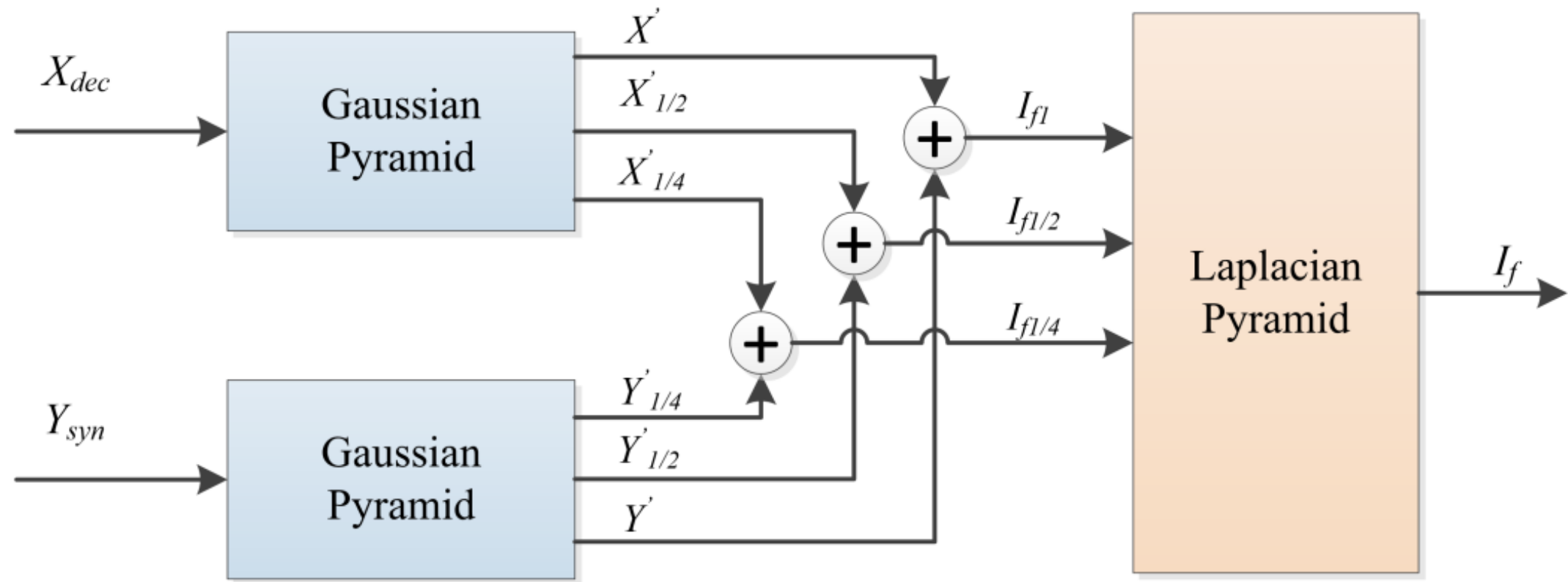
$$Y_{syn}(i) = Y(f(i)) \quad (2)$$



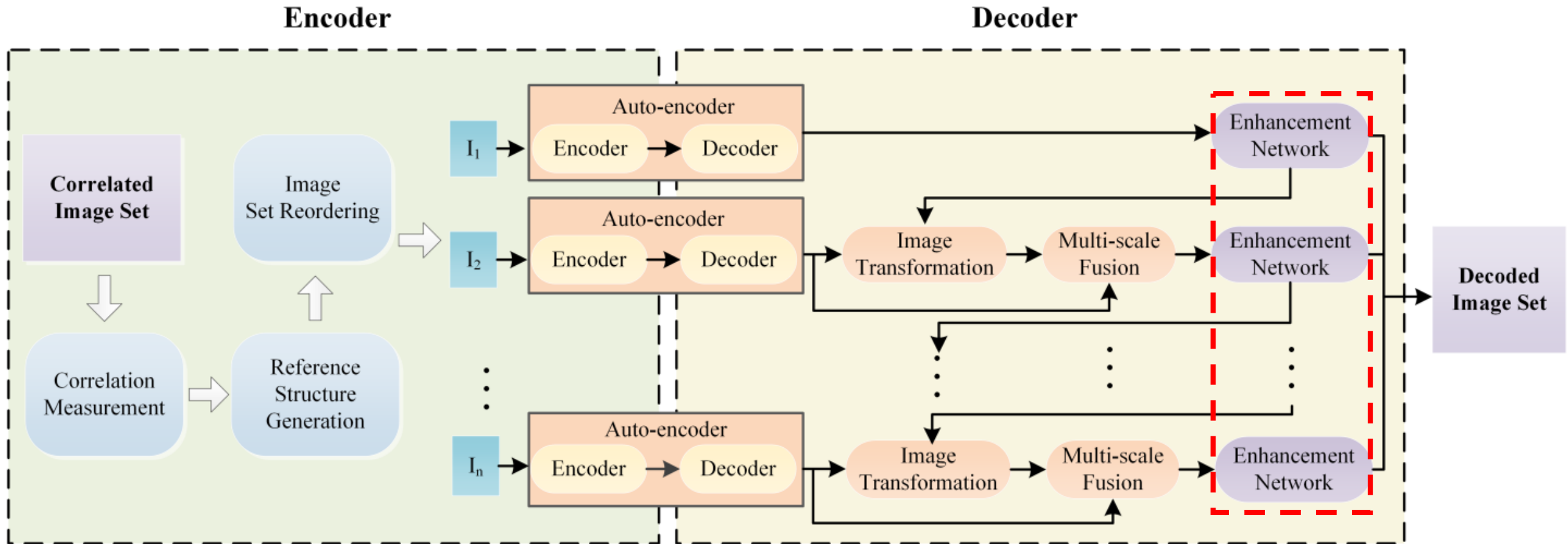
# Multi-Scale Fusion



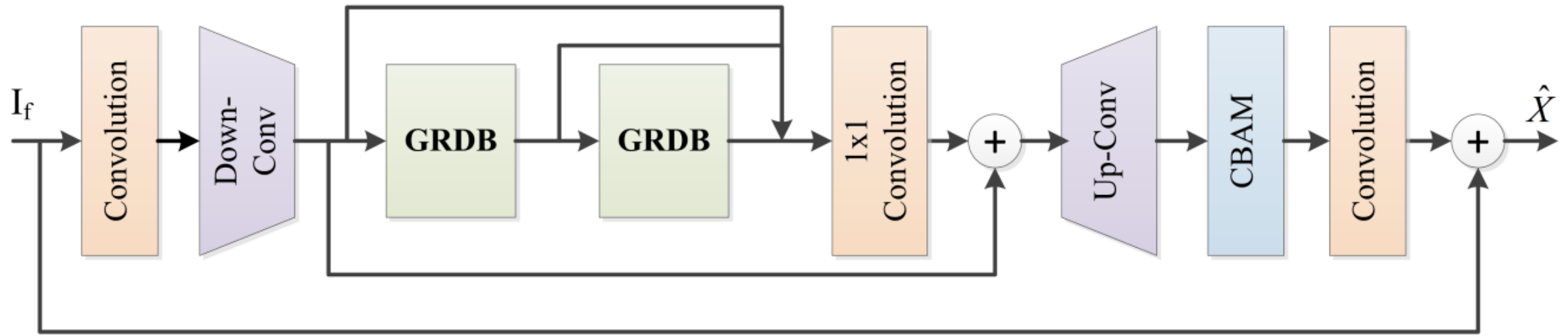
# Multi-Scale Fusion



# Decompressed Image Enhancement



# Decompressed Image Enhancement



# Experimental Settings

- Evaluated on the KITTI datasets, 789 scenes with 21 stereo image pairs per scene taken in sequence
- Implemented in TensorFlow
- Adam solver is adopted
- Initial learning rate of  $10^{-4}$ , batch size of 1
- PSNR and MS-SSIM used as performance criteria

# Visual Quality Comparison

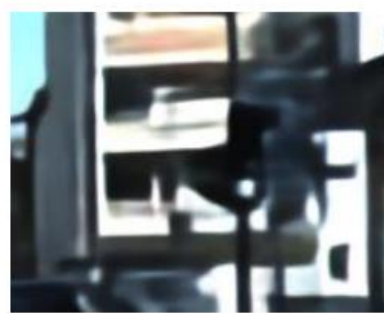
Ground Truth

Proposed

Lee

BPG

JPEG2000



0.0612bpp

0.0593bpp

0.0641bpp

0.0624bpp



0.0550bpp

0.0548bpp

0.0579bpp

0.0533bpp



0.0546bpp

0.0579bpp

0.0533bpp

0.0528bpp

# Objective Quality Comparison

	0.02 bpp	0.04 bpp	0.06 bpp	0.08 bpp	0.10 bpp	0.20 bpp
JPEG2000	25.32	25.96	26.53	26.95	27.25	27.91
BPG	-	-	27.28	27.65	27.91	28.52
Lee[17]	26.24	27.13	27.65	28.02	28.25	28.84
Proposed	<b>27.67</b>	<b>28.29</b>	<b>28.79</b>	<b>29.12</b>	<b>29.33</b>	<b>29.63</b>

PSNR comparison of different methods(dB)

	0.02 bpp	0.04 bpp	0.06 bpp	0.08 bpp	0.10 bpp	0.20 bpp
JPEG2000	0.9007	0.9130	0.9228	0.9289	0.9332	0.9400
BPG	-	-	0.9274	0.9342	0.9380	0.9465
Lee[17]	0.9103	0.9242	0.9324	0.9383	0.9420	0.9480
Proposed	<b>0.9230</b>	<b>0.9338</b>	<b>0.9415</b>	<b>0.9466</b>	<b>0.9494</b>	<b>0.9529</b>

MS-SSIM comparison of different methods

# Ablation Studies: Multi-Scale Fusion



(a) 0.0435 bpp, w/o MSF



(b) 0.0413 bpp, with MSF



(c) 0.0487 bpp, w/o MSF



(d) 0.0512 bpp, with MSF

**Visual quality comparison with and without multi-scale fusion module**



# Ablation Studies: Image Reordering and Image Enhancement



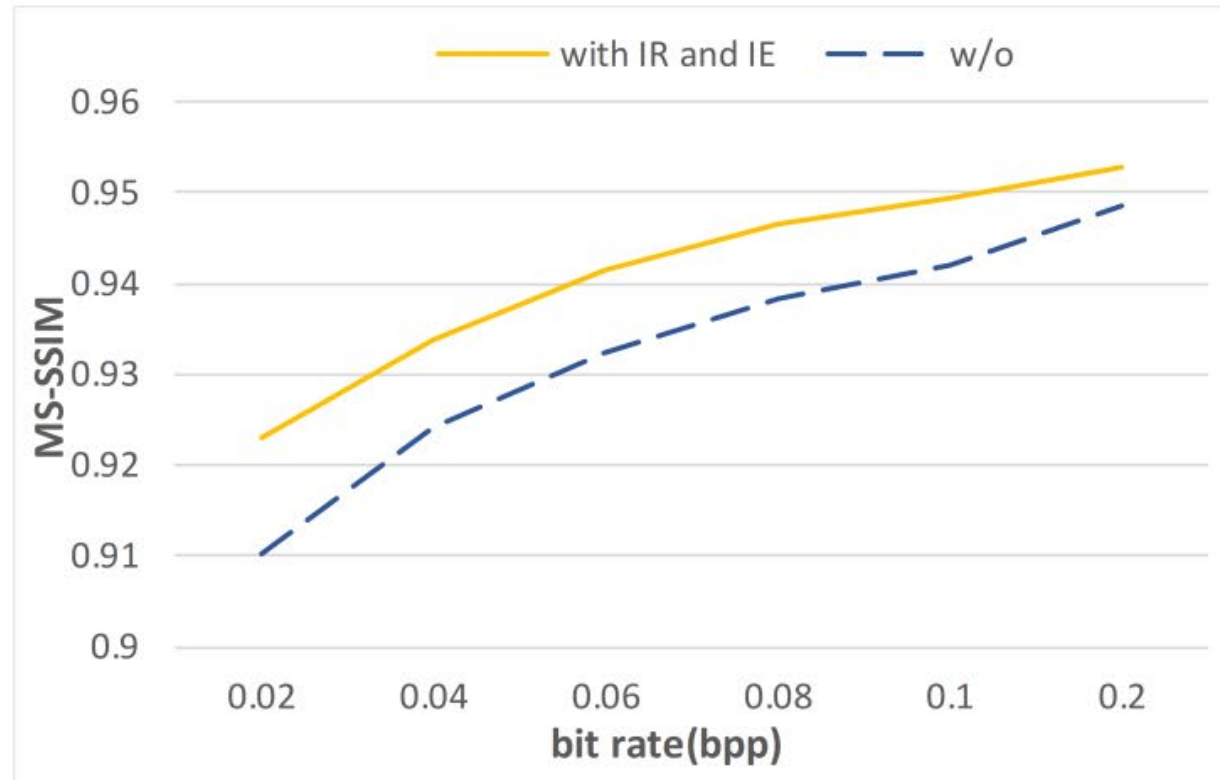
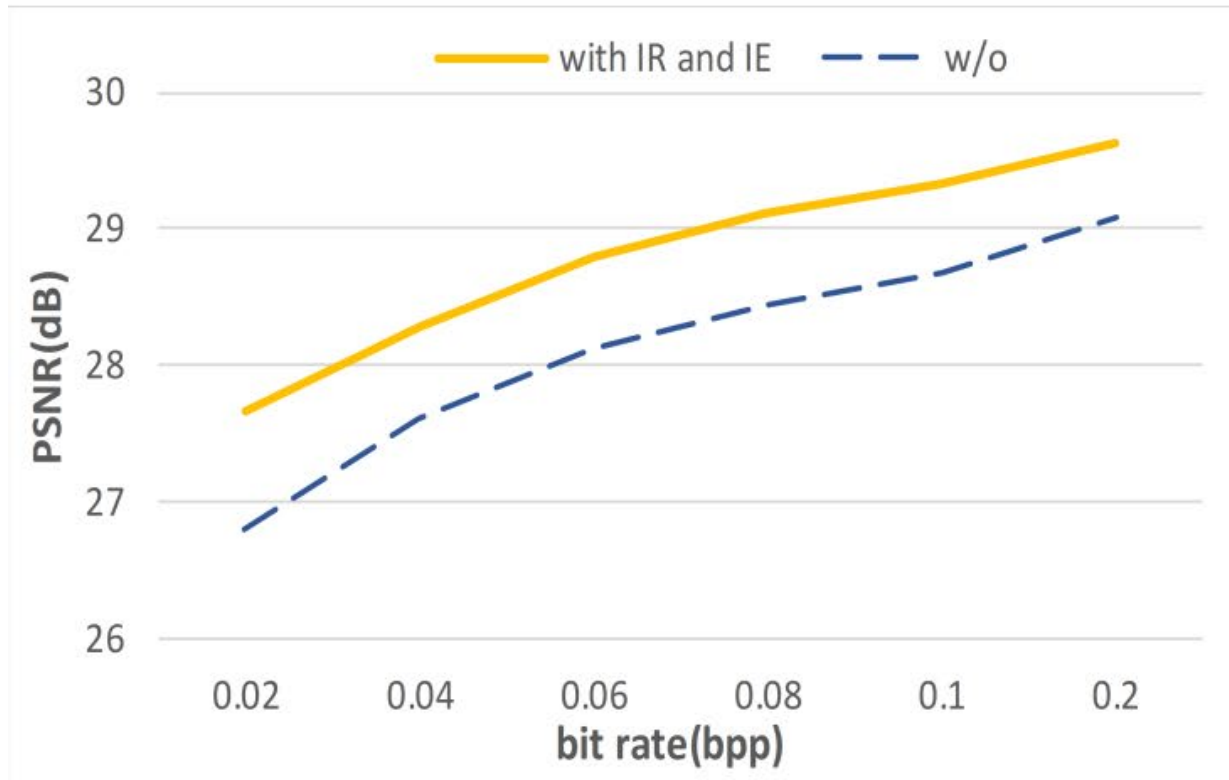
(a) 0.0601 bpp, w/o IR and IEN



(b) 0.0573 bpp, with IR and IEN

**Visual quality comparison with and without IR and IE modules**

# Ablation Studies: Image Reordering and Image Enhancement



Objective quality comparison with and without IR and IEN modules

# Ablation Studies: Image Reordering and Image Enhancement

	0.02 bpp	0.04 bpp	0.06 bpp	0.08 bpp	0.10 bpp	0.20 bpp
with GRDN	26.95	27.75	28.31	28.65	28.87	29.27
with IEN	<b>27.67</b>	<b>28.29</b>	<b>28.79</b>	<b>29.12</b>	<b>29.33</b>	<b>29.63</b>

PSNR comparison with different image enhancement modules(dB)



# Conclusions

- A deep similar image set compression scheme based on Distributed Source Coding and multi-scale fusion is proposed
- Our method can efficiently learn the correlation between an image and its side information and fuse additional side information at different scales
- Extensive experimental results compared with other mainstream methods validate the superior performance of our scheme in both subjective and objective quality



**Thank you!**

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