

Multiple Degradation and Reconstruction Network for Single Image Denoising via Knowledge Distillation

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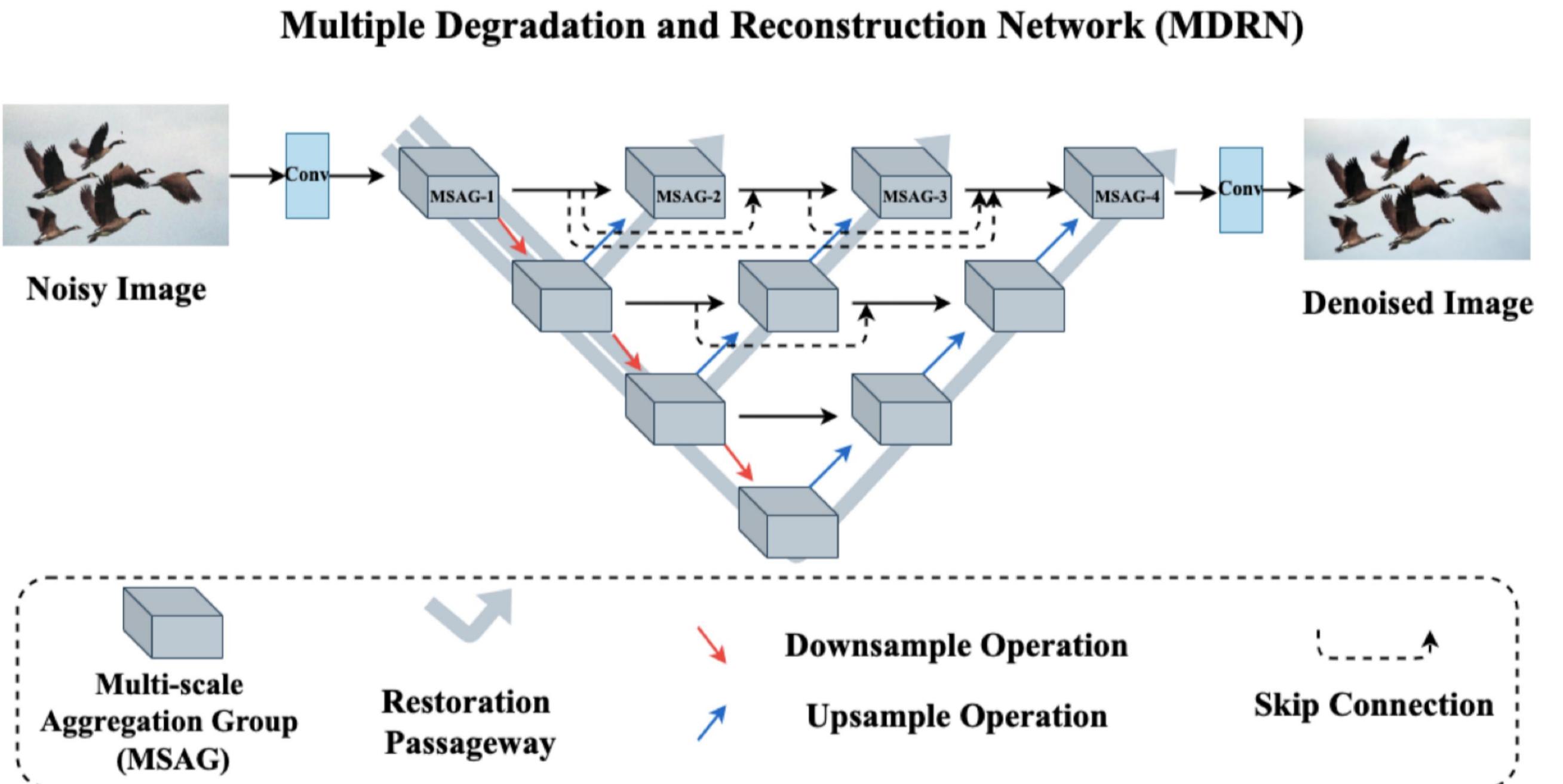


Motivation & Contributions

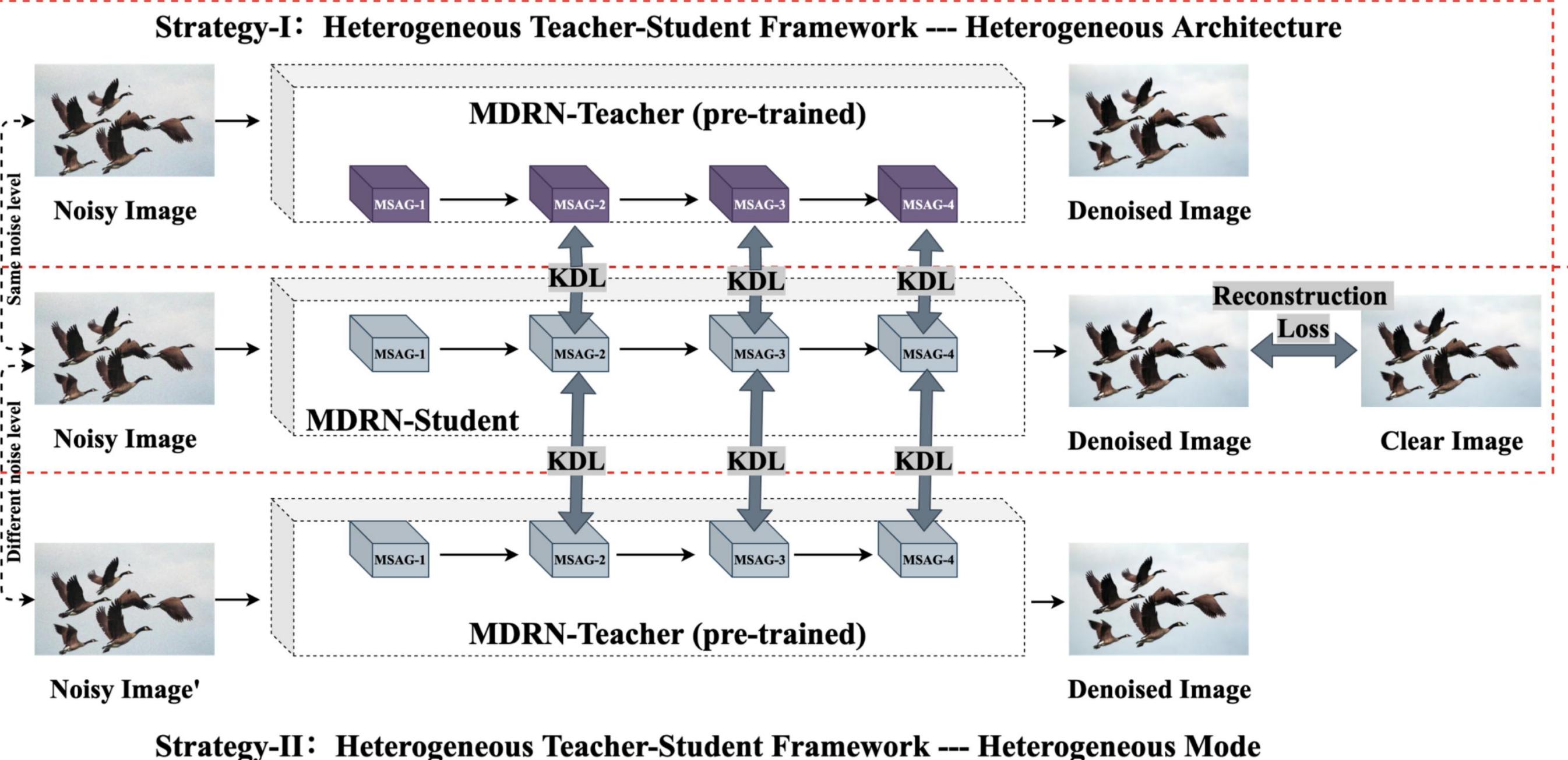
Different from previous works that blindly increase the depth of the network, we explore the degradation mechanism of the noisy image and aim to introduce the knowledge distillation strategy to the model.

- We propose a lightweight and efficient **Multiple Degradation and Reconstruction Network (MDRN)** for image denoising, which can progressively remove image noise.
- We propose a **Multi-Scale Aggregation Group (MSAG)** for feature extraction. MSAB is the basic components of MDRN, which can extract rich multi-scale features with few parameters.
- We propose two **Heterogeneous Knowledge Distillation Strategies (HKDS)** for SID. With the help of HKDS, MDRN can learn richer and more accurate features from the teacher model.

Method



Heterogeneous Knowledge Distillation



➤ Reconstruction Loss

$$\mathcal{L}_{RL} = \|F_{SID}(I_{noisy}) - I_{clear}\|_1$$

➤ Knowledge Distillation Loss

$$\mathcal{L}_{KDL} = \sum_{i=2,3,4} \|S_G^i(I_{noisy}) - T_G^i(I'_{noisy})\|_1$$

Ablation Studies

Method	Para.	Set12 σ=15	BSD68 σ=15	Urban100 σ=15
Noise Level	-	σ=50	σ=50	σ=50
FFDNet	851K	32.75	27.32	31.63
MDRN (2B)	770K	32.87	27.35	31.72
MDRN (2B) + HADS	770K	32.96	27.43	31.78

Method	Para.	Kodak24 σ=30	CBSD68 σ=30	Urban100 σ=30
Noise Level	-	σ=70	σ=70	σ=70
FFDNet	851K	31.39	27.68	30.31
MDRN (2B)	770K	31.40	27.69	30.38
MDRN (2B) + HADS	770K	31.47	27.73	30.42



Results

Method	Set12			BSD68			Urban100		
	σ=15	σ=25	σ=50	σ=15	σ=25	σ=50	σ=15	σ=25	σ=50
BM3D	32.37	29.97	26.72	31.08	28.57	25.62	32.34	29.70	25.94
RED30	32.83	30.48	27.34	31.72	29.26	26.35	32.75	30.21	26.64
TNRD	32.50	30.06	26.81	31.42	28.92	25.97	31.98	29.29	25.71
IRCNN	32.77	30.38	27.14	31.63	29.15	26.19	32.49	29.82	26.14
DnCNN	32.86	30.43	27.18	31.73	29.23	26.23	32.68	29.97	26.28
FFDNet	32.75	30.43	27.32	31.63	29.19	26.29	32.42	29.92	26.52
MLEFGN	33.04	30.66	27.54	31.81	29.34	26.39	33.21	30.64	27.22
MFENANN	32.95	30.63	27.55	31.73	29.29	26.38	-	-	-
DRNet	33.01	30.64	27.46	31.81	29.35	26.39	-	-	-
MDRN (Ours)	33.06	30.67	27.56	31.83	29.36	26.41	33.22	30.67	27.24
MDRN+ (Ours)	33.10	30.71	27.60	31.86	29.39	26.44	33.31	30.78	27.31

Method	Kodak24			CBSD68			Urban100		
	σ=30	σ=50	σ=70	σ=30	σ=50	σ=70	σ=30	σ=50	σ=70
CBM3D	30.89	28.63	27.27	29.73	27.38	26.00	30.36	27.94	26.31
RED30	29.71	27.62	26.36	28.46	26.35	25.08	29.02	26.40	24.74
TNRD	28.83	27.17	24.94	27.64	25.96	23.83	27.40	25.52	22.63
IRCNN	31.24	28.93	N/A	30.22	27.86	N/A	30.28	27.69	N/A
DnCNN	31.39	29.16	27.64	30.40	28.01	26.56	30.28	28.16	26.17
MemNet	29.67	27.65	26.40	28.39	26.33	25.08	28.93	26.53	24.93
FFDNet	31.39	29.10	27.68	30.31	27.96	26.53	30.53	28.05	26.39
MLEFGN	31.67	29.38	27.94	30.56	28.21	26.75	31.32	28.92	27.28
MDRN (Ours)	31.68	29.40	27.96	30.57	28.23	26.77	31.35	28.96	27.32
MDRN+ (Ours)	31.73	29.44	28.01	30.61	28.27	26.82	31.41	29.00	27.37

Visual Comparison

