Image Dehazing Transformer with Transmission-Aware 3D Position Embedding

Chunle Guo¹ Qixin Yan² Saeed Anwar³ Runmin Cong⁴ Wenqi Ren⁵ Chongyi Li^{6*}

¹ TMCC, CS, Nankai University ² Tianjin University ³ Australian National University

⁴ Beijing Jiaotong University ⁵ Sun Yat-sen University ⁶ S-Lab, Nanyang Technological University guochunle@nankai.edu.cn qxyan@tju.edu.cn saeedanwarcse@gmail.com

rmcong@bjtu.edu.cn rwq.renwenqi@gmail.com lichongyi25@gmail.com https://li-chongyi.github.io/Proj_DeHamer.html

This supplementary material provides the detailed network structure and parameters of our method and additional experimental results.

- We provide the detailed structures and parameters of our Transformer module, CNN encoder module, and CNN decoder module in Figure 1, Figure 2, and Figure 3, respectively.
- We provide the comparison with winner methods(TND [6] and DW-GAN [3]) of NTIRE 2020/2021 image dehazing challenge in Table 1.
- We provide the comparisons with the other two stateof-the-art methods (CL [10] and DA [9]) in Table 2.
- We provide the comparisons of inference time with state-of-the-art methods (FFANet [8], TND [6], and DW-GAN [3]) for processing an image of size 1600×1200×3 using an NVIDIA 1080Ti GPU in Table 3.
- We provide more visual comparison results in Figure 4, Figure 5, Figure 6, and Figure 7.

Table 1. Quantitative comparisons with the winner methods of NTIRE 2020/2021 on SOTS, Dense-Haze, and NH-Haze datasets. Our method achieves the highest PSNR and SSIM scores on the SOTS dataset and the highest PSNR score on the Dense-Haze dataset when compared with the 2020/2021 champion solutions TDN [6] and DW-GAN [3].

Methods	SOTS		Dense-Haze		NH-Haze	
	PSNR↑	SSIM↑	PSNR↑	SSIM↑	PSNR↑	SSIM↑
TDN ('20)	34.59	0.9754	15.50	0.5081	20.44	0.6683
DW-GAN ('21)	35.94	<u>0.9860</u>	<u>16.49</u>	0.5911	21.51	0.7111
DeHamer (Ours)	36.63	0.9881	16.62	<u>0.5602</u>	20.66	<u>0.6844</u>

^{*}Chongyi Li (lichongyi25@gmail.com) is the corresponding author.

Table 2. Quantitative comparisons with two additional stateof-the-art methods (CL [10] and DA [9]) on SOTS, Dense-Haze, and NH-Haze datasets. Our method achieves the highest PSRN and SSIM scores on the Dense-Haze dataset and highest PSNR score on the NH-Haze dataset, and the second-highest scores across other datasets.

Methods	SOTS		Dense-Haze		NH-Haze	
	PSNR↑	SSIM↑	PSNR↑	SSIM↑	PSNR↑	SSIM↑
CL ('21)	37.17	0.9901	<u>15.80</u>	0.4660	<u>19.88</u>	0.7173
DA ('20)	25.30	0.9423	12.56	0.4994	13.77	0.5349
DeHamer (Ours)	<u>36.63</u>	<u>0.9881</u>	16.62	0.5602	20.66	<u>0.6844</u>

Table 3. Inference time comparisons with state-of-theart methods (FFANet [8], TND [6], and DW-GAN [3]). Compared with these methods, our method achieves the fastest inference speed.

Methods	FFANet	TDN	DW-GAN	DeHamer (Ours)
Inference Time (s)	2.62	0.63	0.48	0.41

References

- B. Cai, X. Xu, K. Jia, C. Qing, and D. Tao. An end-to-end system for single image haze removal. *TIP*, 25(11):5187– 5198, 2016. 4, 5, 6, 7
- [2] H. Dong, J. Pan, L. Xiang, Z. Hu, X. Zhang, F. Wang, and M. H. Yang. Multi-scale boosted dehazing network with dense feature fusion. In *CVPR*, pages 2157–2167, 2020. 4, 5, 6, 7
- [3] M. Fu, H. Liu, Y. Yu, J. Chen, and K. Wang. Dw-gan: A discrete wavelet transform gan for nonhomogeneous dehazing. In *CVPRW*, pages 203–212, 2021.
- [4] K. He, J. Sun, and X. Tang. Single image haze removal using dark channel prior. *TPAMI*, 33(12):2341–2353, 2011. 4, 5, 6, 7
- [5] B. Li, X. Peng, Z. Wang, J. Xu, and D. Feng. Aod-net: Allin-one dehazing network. In *ICCV*, pages 4770–4778, 2017. 4, 5, 6, 7



Figure 1. Detailed network structure and parameters of the proposed Transformer module.



Figure 2. Detailed network structure and parameters of the proposed CNN encoder module.



Figure 3. Detailed network structure and parameters of the proposed CNN decoder module.

- [6] J. Liu, H. Wu, Y. Xieand Y. Qu, and L. Ma. Trident dehazing network. In *CVPRW*, pages 430–431, 2020.
- [7] X. Liu, Y. Ma, Z. Shi, and J. Chen. Griddehazenet: Attention-based multi-scale network for image dehazing. In *CVPR*, pages 7314–7323, 2019. 4, 5, 6, 7
- [8] X. Qin, Z. Wang, Y. Bai, X. Xie, and H. Jia. FFA-Net: Feature fusion attention network for single image dehazing. In *AAAI*, pages 11908–11915, 2020. 1, 4, 5, 6, 7
- [9] Y. Shao, L. Li, W. Ren, C. Gao, and N. Sang. Domain adaptation for image dehazing. In *CVPR*, pages 2808–2817, 2020.
- [10] H. Wu, Y. Qu, S. Lin, J. Zhou, R. Qiao, Z. Zhang, Y. Xie, and L. Ma. Contrastive learning for compact single image dehazing. In *CVPR*, pages 10551–10560, 2021. 1
- Z. Zheng, W. Ren, X. Cao, X. Hu, T. Wang, F. Song, and X. Jia. Ultra-high-definition image dehazing via multi-guided bilateral learning. In *CVPR*, pages 16185–16194, 2021. 4, 5, 6, 7



(a) input



(c) DehazeNet [1]



(e) GridDehazeNet [7]



(g) MSBDN [2]



(i) DeHamer (Ours)



(b) DCP [4]



(d) AODNet [5]



(f) FFANet [8]



(h) UHD [11]



(j) GT





(a) input



(c) DehazeNet [1]



(e) GridDehazeNet [7]



(g) MSBDN [2]



(i) DeHamer (Ours)



(b) DCP [4]



(d) AODNet [5]



(f) FFANet [8]



(h) UHD [11]



(j) GT

Figure 5. Visual comparisons on a synthetic hazy image sampled from SOTS-Outdoor testing set. Zoom in for best view.



(a) input



(c) DehazeNet [1]



(e) GridDehazeNet [7]



(g) MSBDN [2]



(i) DeHamer (Ours)



(b) DCP [4]



(d) AODNet [5]



(f) FFANet [8]



(h) UHD [11]



(j) GT

Figure 6. Visual comparisons on a synthetic hazy image sampled from SOTS-indoor testing set. Zoom in for best view.



(b) DCP [4]



(d) AODNet [5]



(f) FFANet [8]



(a) input



(c) DehazeNet [1]



(e) GridDehazeNet [7]



(g) MSBDN [2]



(i) DeHamer (Ours)



(h) UHD [11]



(j) GT

Figure 7. Visual comparisons on a synthetic hazy image sampled from SOTS-indoor testing set. Zoom in for best view.