

Referensi

- [1] Fu, Y., Guo, G. and Huang, T.S. (2010). Age synthesis and estimation via faces: A survey. *IEEE transactions on pattern analysis and machine intelligence*. (pp. 1955-1976)
- [2] Shu, X., Tang, J., Lai, H., Liu, L., & Yan, S. (2015). Personalized age progression with aging dictionary. In *Proceedings of the IEEE international conference on computer vision* (pp. 3970-3978).
- [3] Kemelmacher-Shlizerman, I., Suwajanakorn, S., & Seitz, S. M. (2014). Illumination-aware age progression. In *Proceedings of the IEEE conference on computer vision and pattern recognition* (pp. 3334-3341).
- [4] Ian Goodfellow, Jean Pouget-Abadie, Mehdi Mirza, Bing Xu, David Warde-Farley, Sherjil Ozair, Aaron Courville, and Yoshua Bengio. Generative Adversarial Nets. *Advances in Neural Information Processing Systems* 27, pages 2672–2680, 2014.
- [5] Liu, S., Sun, Y. and Yan, S. (2017). Face aging with contextual generative adversarial nets. In *Proceedings of the 25th ACM international conference on Multimedia* (pp. 82-90).
- [6] Antipov, G., Baccouche, M., & Dugelay, J. L. (2017, September). Face aging with conditional generative adversarial networks. In *2017 IEEE international conference on image processing (ICIP)* (pp. 2089-2093). IEEE.
- [7] Zhu, J. Y., Park, T., Isola, P., & Efros, A. A. (2017). Unpaired image-to-image translation using cycle-consistent adversarial networks. In *Proceedings of the IEEE international conference on computer vision* (pp. 2223-2232).
- [8] Zhang, Z., Song, Y., & Qi, H. (2017). Age progression/regression by conditional adversarial autoencoder. In *Proceedings of the IEEE conference on computer vision and pattern recognition* (pp. 5810-5818).
- [9] Hamzah, N., & Zaman, F. H. K. (2020, September). Face Aging on Realistic Photo in Cross-Dataset Implementation. In *IOP Conference Series: Materials Science and Engineering* (Vol. 917, No. 1, p. 012080). IOP Publishing.
- [10] Ren, L., Liu, S., Sun, Y., Dong, J., Liu, L., & Yan, S. (2017) . Time Traveler: a real-time face aging system. In *Proceedings of the 25th ACM international conference on Multimedia* (pp. 1245-1246).
- [11] Radford, A., Metz, L., & Chintala, S. (2015). Unsupervised representation learning with deep convolutional generative adversarial networks. *arXiv preprint arXiv:1511.06434*.
- [12] Reed, S., Akata, Z., Yan, X., Logeswaran, L., Schiele, B., & Lee, H. (2016, June). Generative adversarial text to image synthesis. In *International Conference on Machine Learning* (pp. 1060-1069). PMLR
- [13] Zhu, J. Y., Krähenbühl, P., Shechtman, E., & Efros, A. A. (2016, October). Generative visual manipulation on the natural image manifold. In *European conference on computer vision* (pp. 597-613). Springer, Cham.
- [14] Ronneberger, O., Fischer, P., & Brox, T. (2015, October). U-net: Convolutional networks for biomedical image segmentation. In *International Conference on Medical image computing and computer-assisted intervention* (pp. 234-241). Springer, Cham.
- [15] Taigman, Y., Polyak, A., & Wolf, L. (2016). Unsupervised cross-domain image generation. *arXiv preprint arXiv:1611.02200*.
- [16] Heusel, M., Ramsauer, H., Unterthiner, T., Nessler, B., & Hochreiter, S. (2017). Gans trained by a two time-scale update rule converge to a local nash equilibrium. *Advances in neural information processing systems*, 30.
- [17] Kingma, D. P., & Ba, J. (2014). Adam: A method for stochastic optimization. *arXiv preprint arXiv:1412.6980*.