

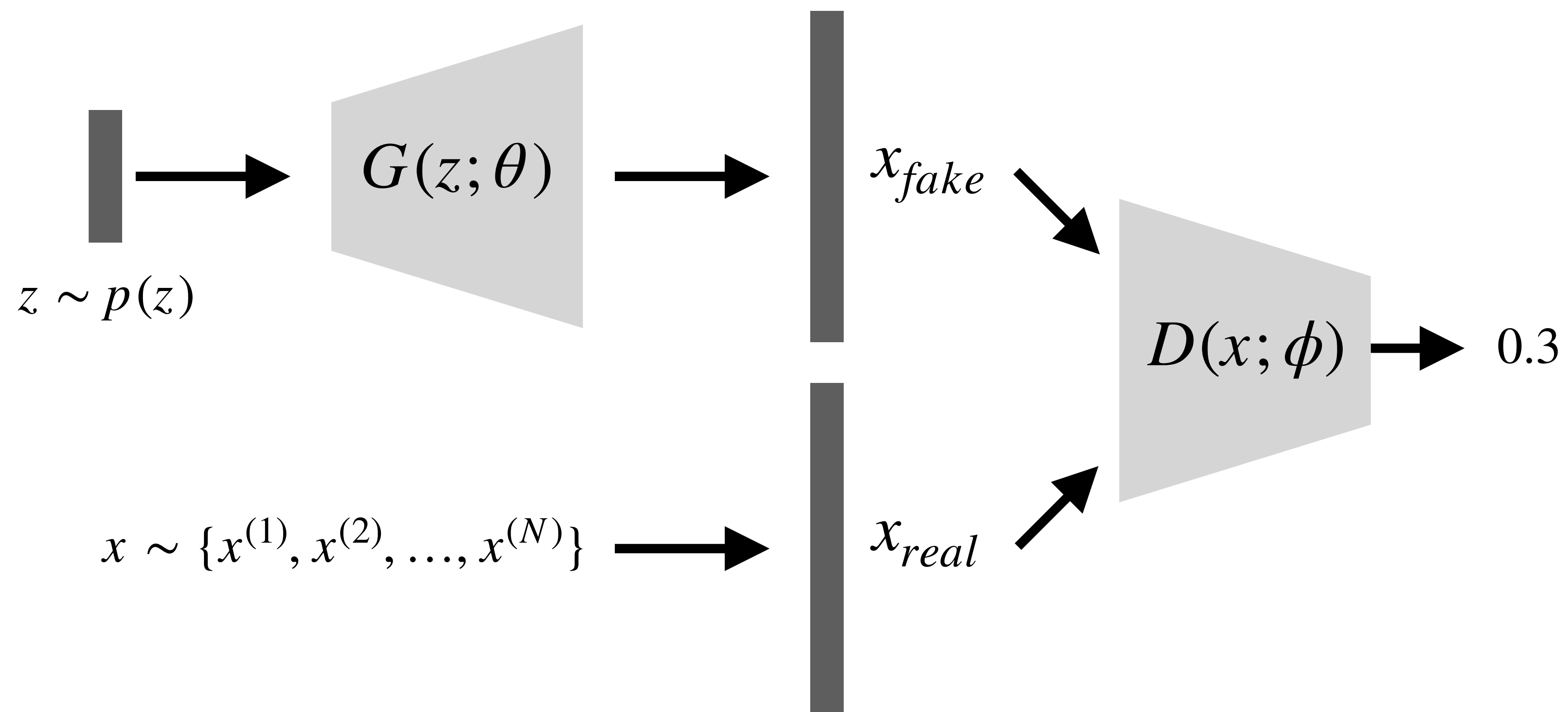
Hierarchical Mixtures of Generators for Adversarial Learning

Alper Ahmetoğlu¹ and Ethem Alpaydın²

¹Department of Computer Engineering
Boğaziçi University
Istanbul, Turkey

²Department of Computer Science
Özyeğin University
Istanbul, Turkey

Generative Adversarial Networks



Problems with GANs

- Vanishing or exploding gradients: $\log(1 - D(G(z)))$ or $-\log D(G(z))$
- Mode Collapse

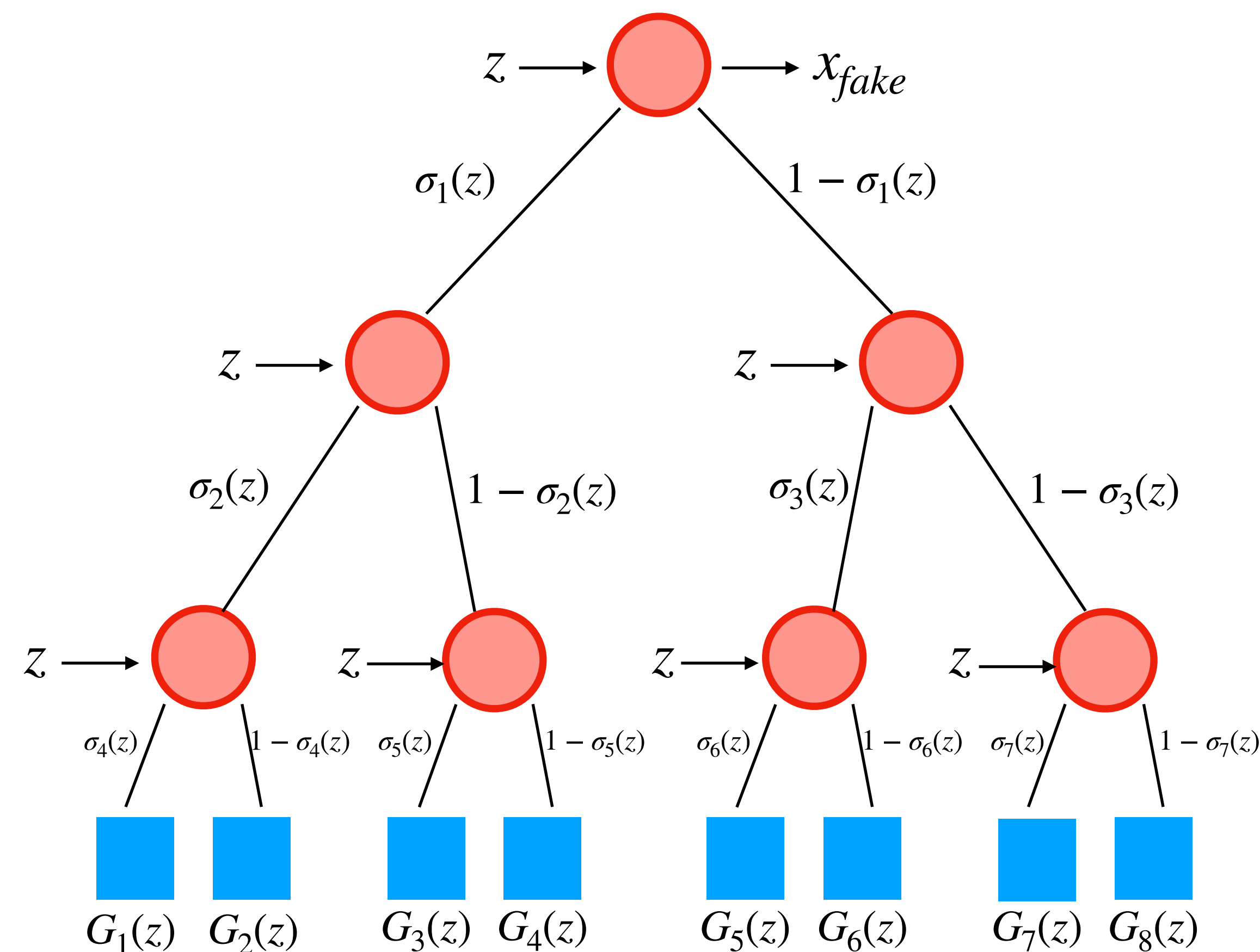
Solution Approaches

- Architecture change: DCGAN, ProgGAN
- Objective change: LSGAN, WGAN
- Regularization methods: Spectral normalization

Combining Multiple Generators in GANs

- Multi-agent diverse GAN (MADGAN, Ghosh et al. 2018)
- Mixture GAN (MGAN, Hoang et al. 2018)
- Mixture of experts GAN (MEGAN, Park et al. 2018)

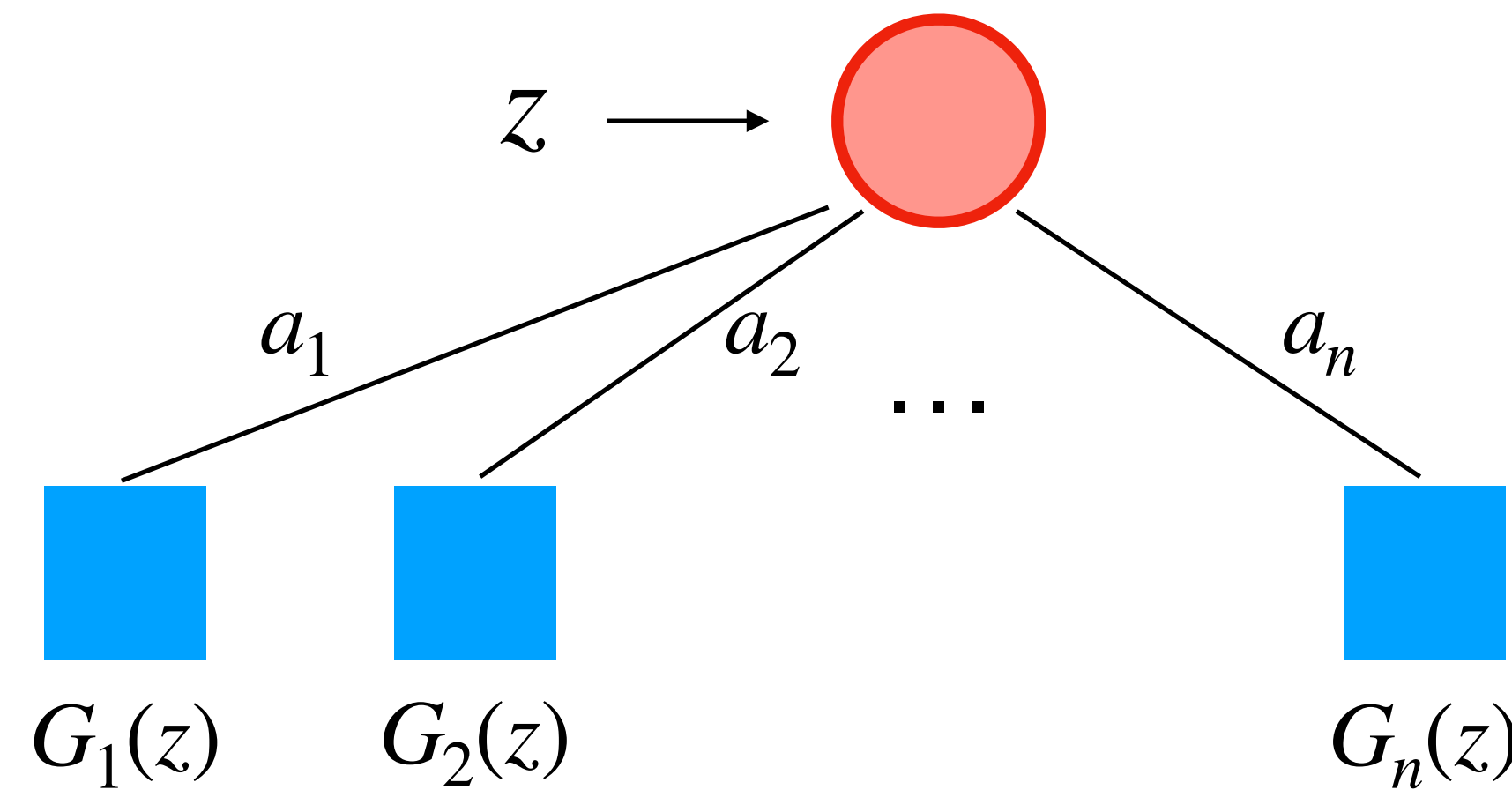
Hierarchical Mixture of Generators (HMoG)



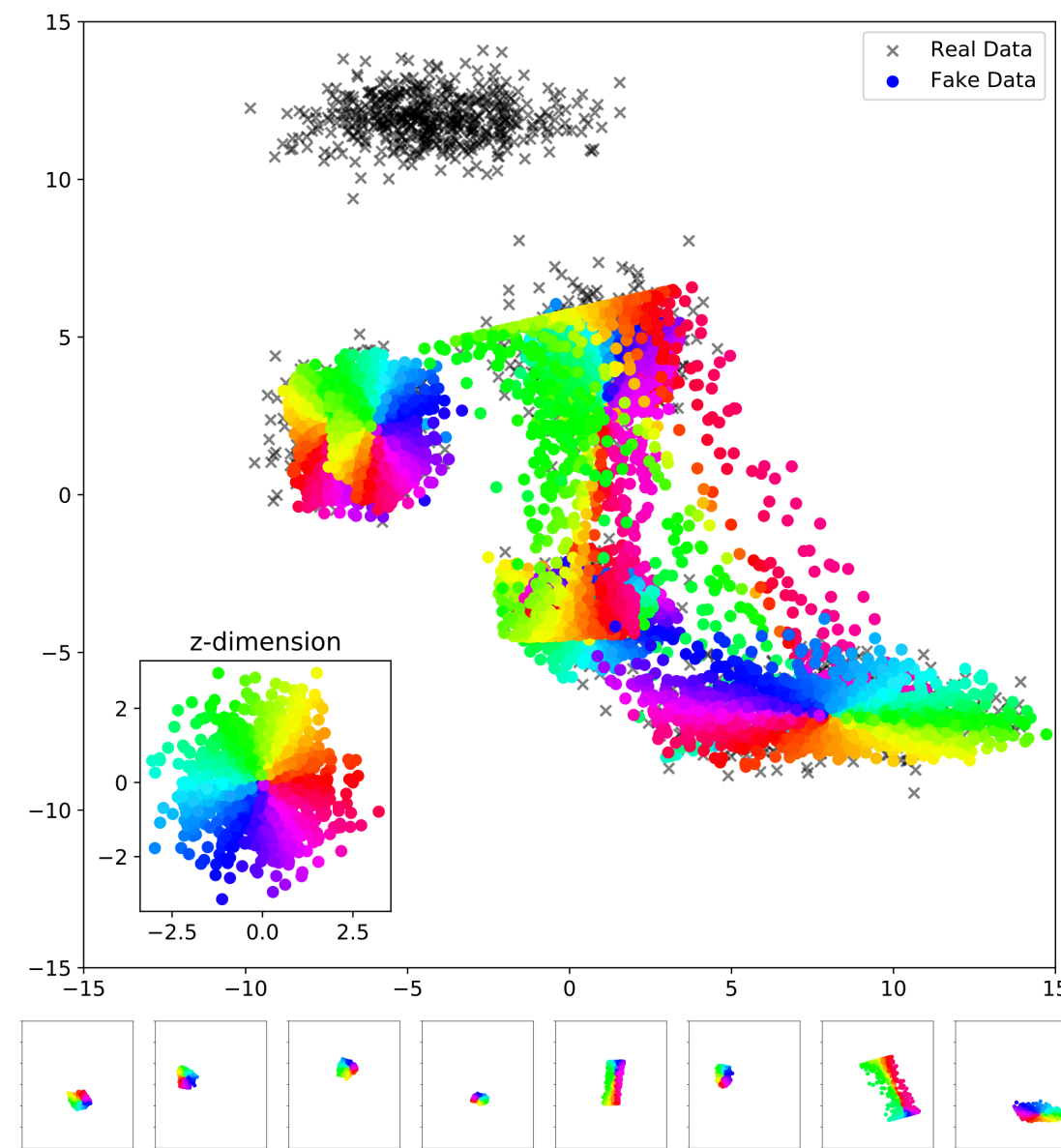
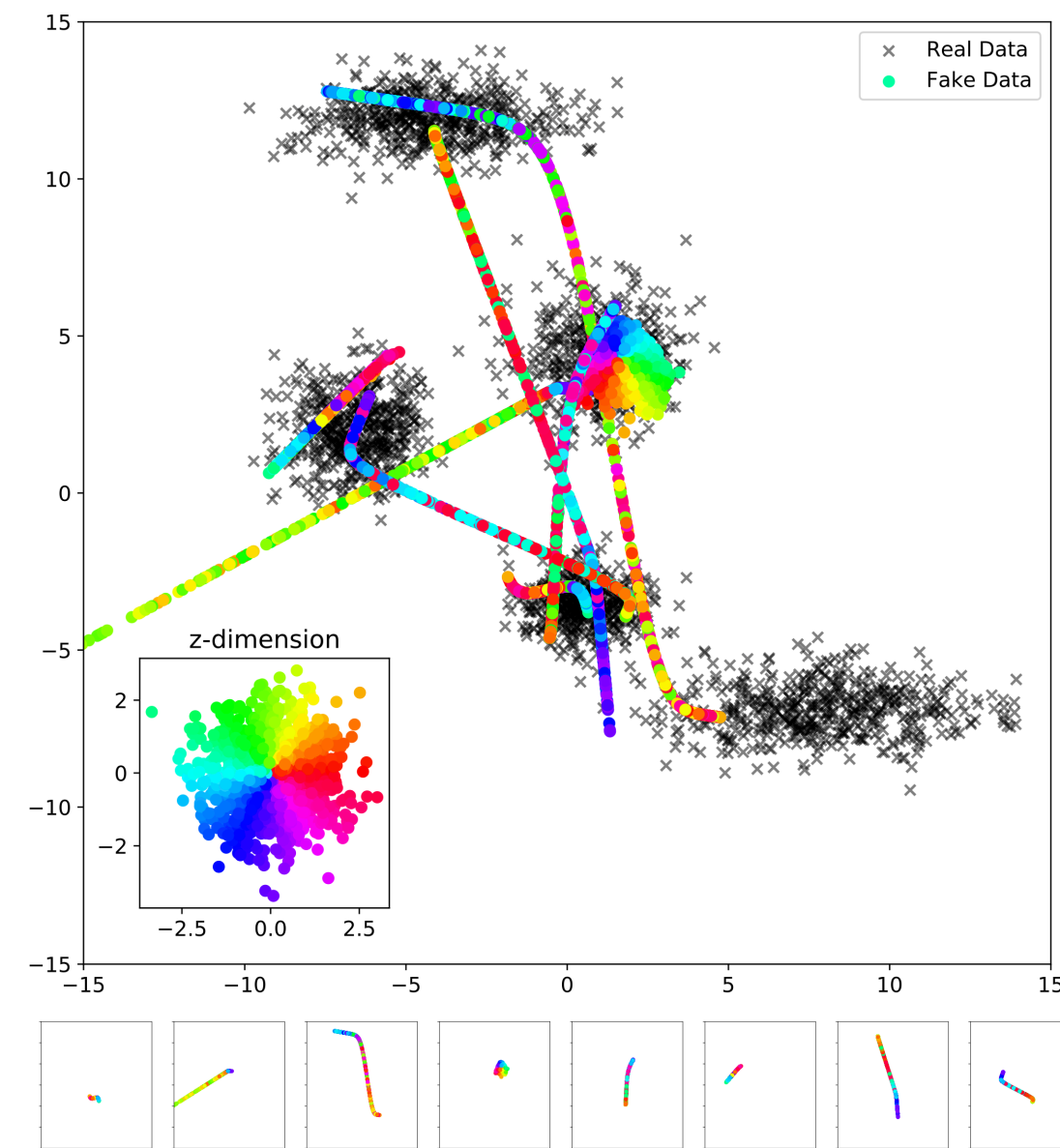
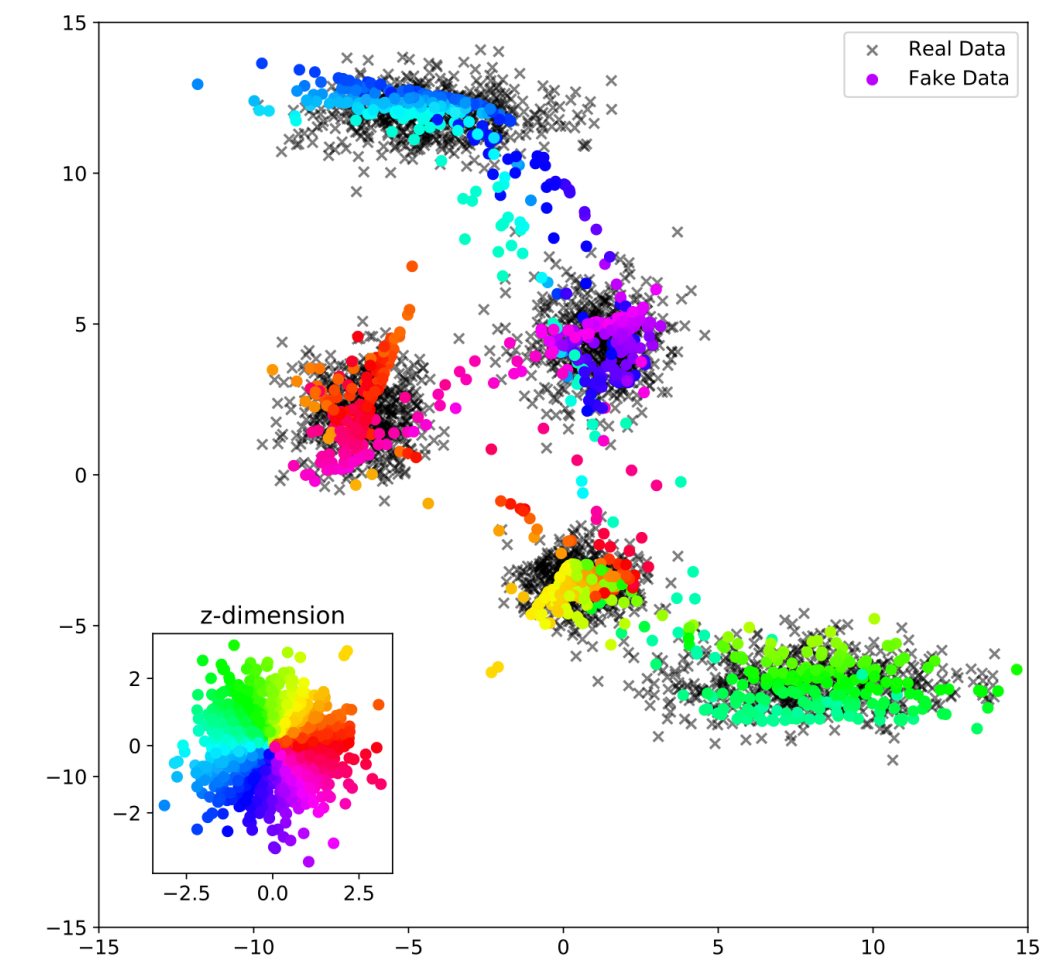
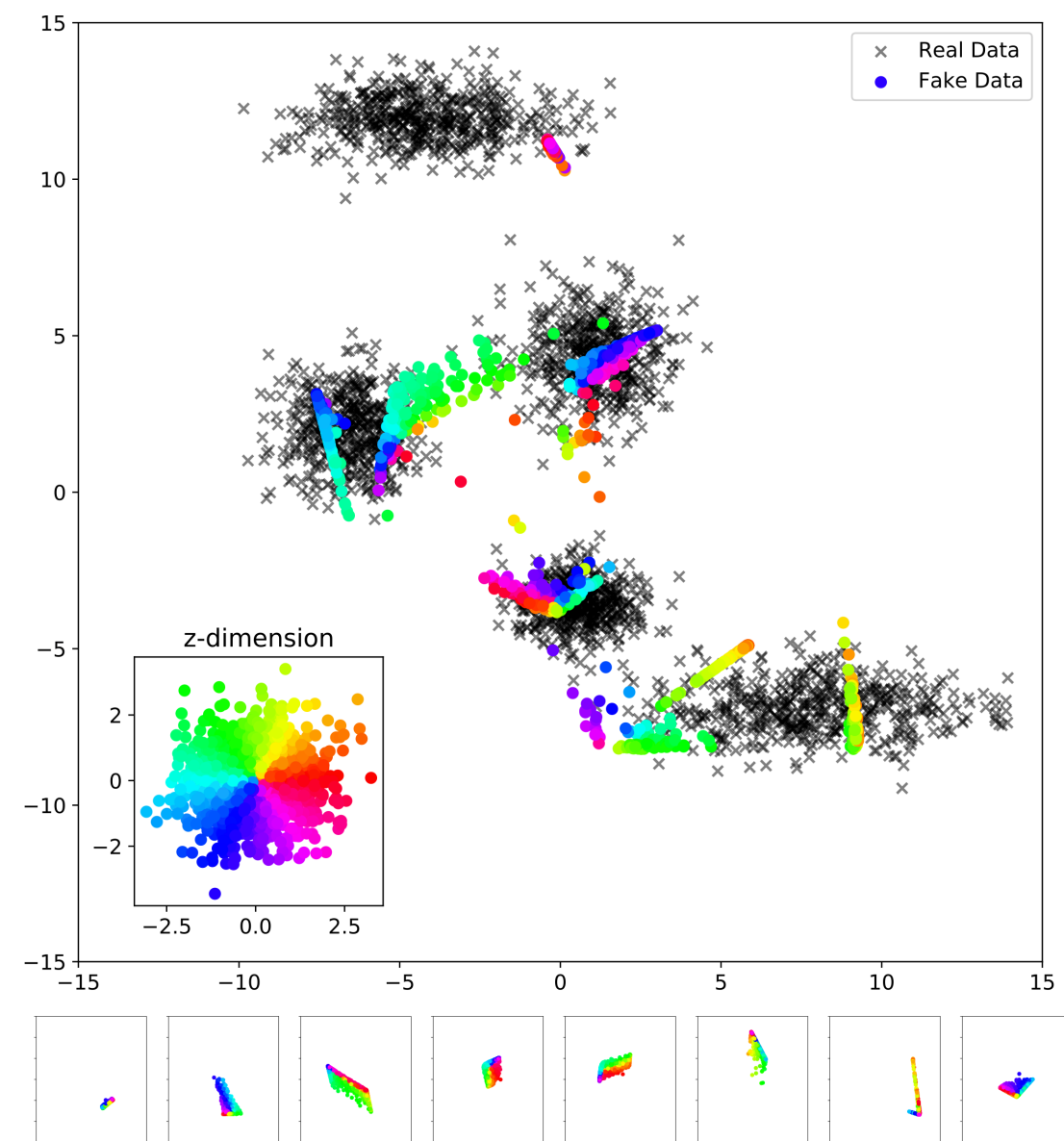
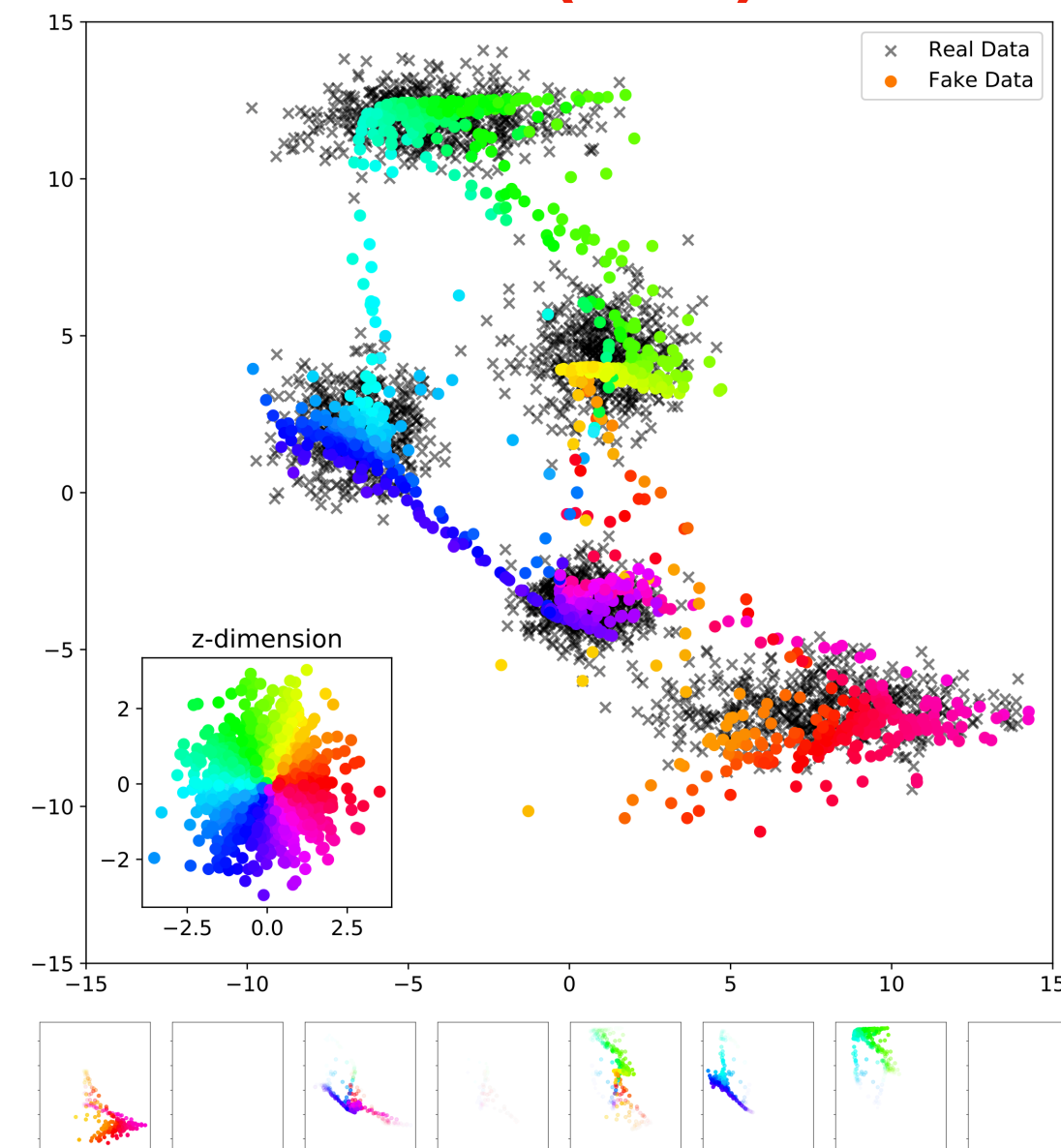
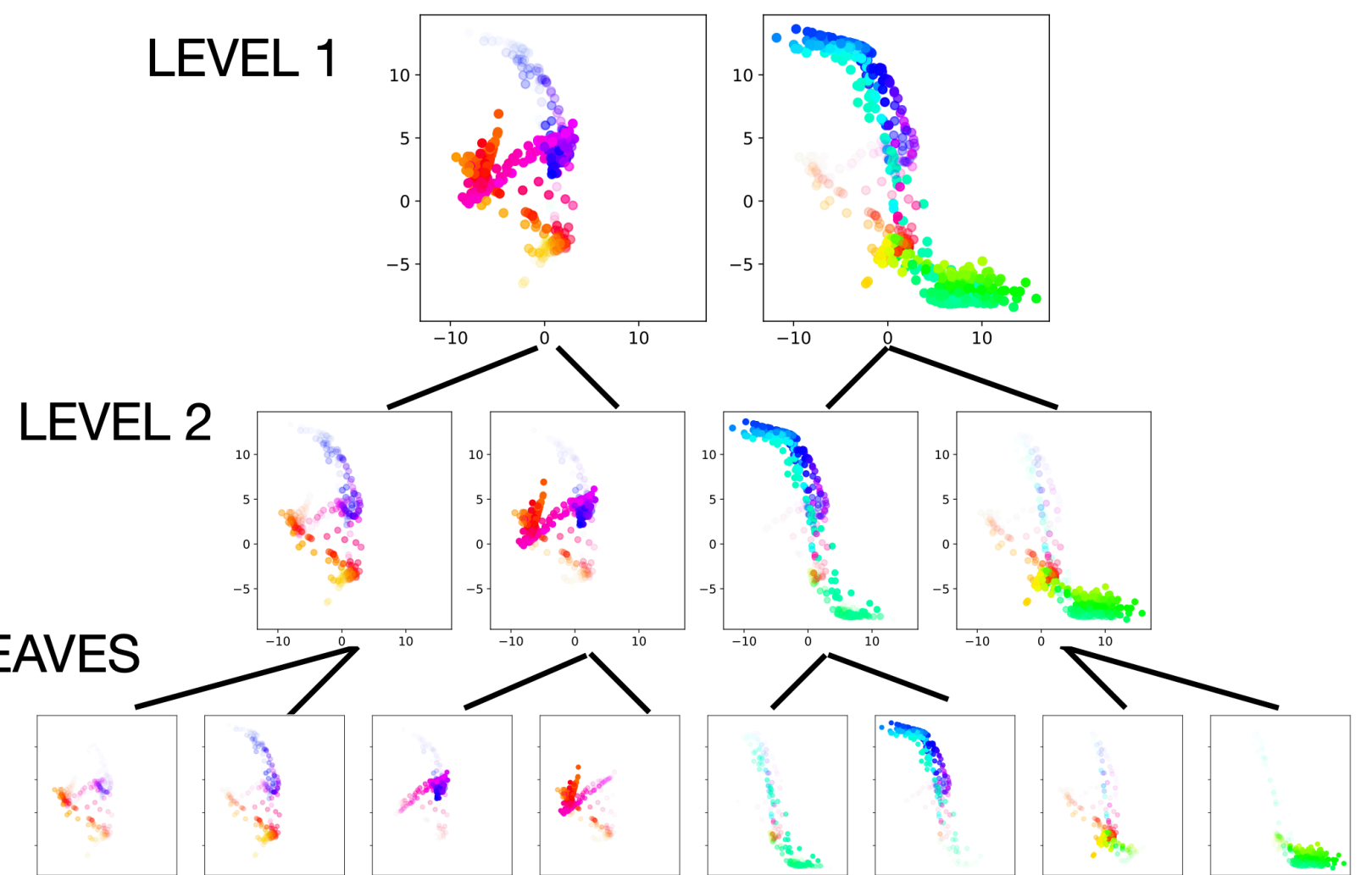
$$\sigma_1(z) = \frac{1}{1 + \exp[-(v_1 z + v_{10})]}$$

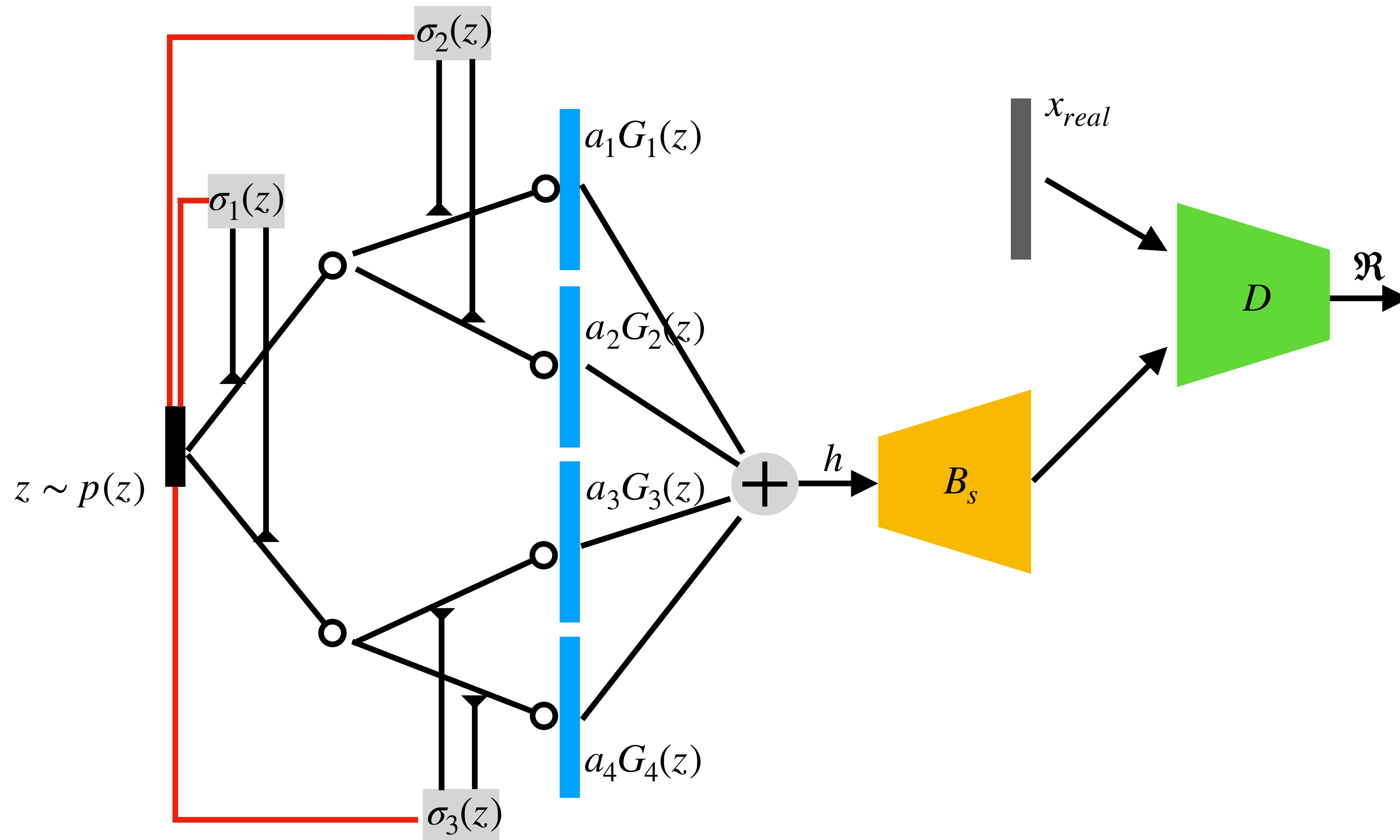
$$G_1(z) = W_1 z + w_{10}$$

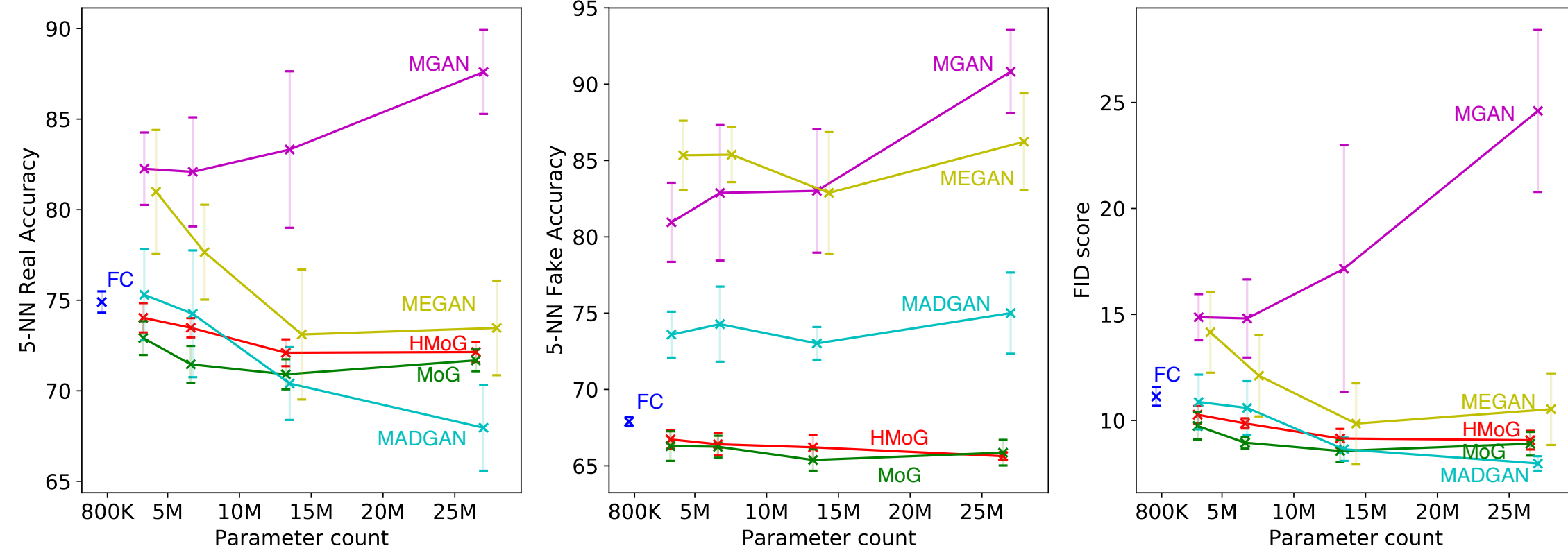
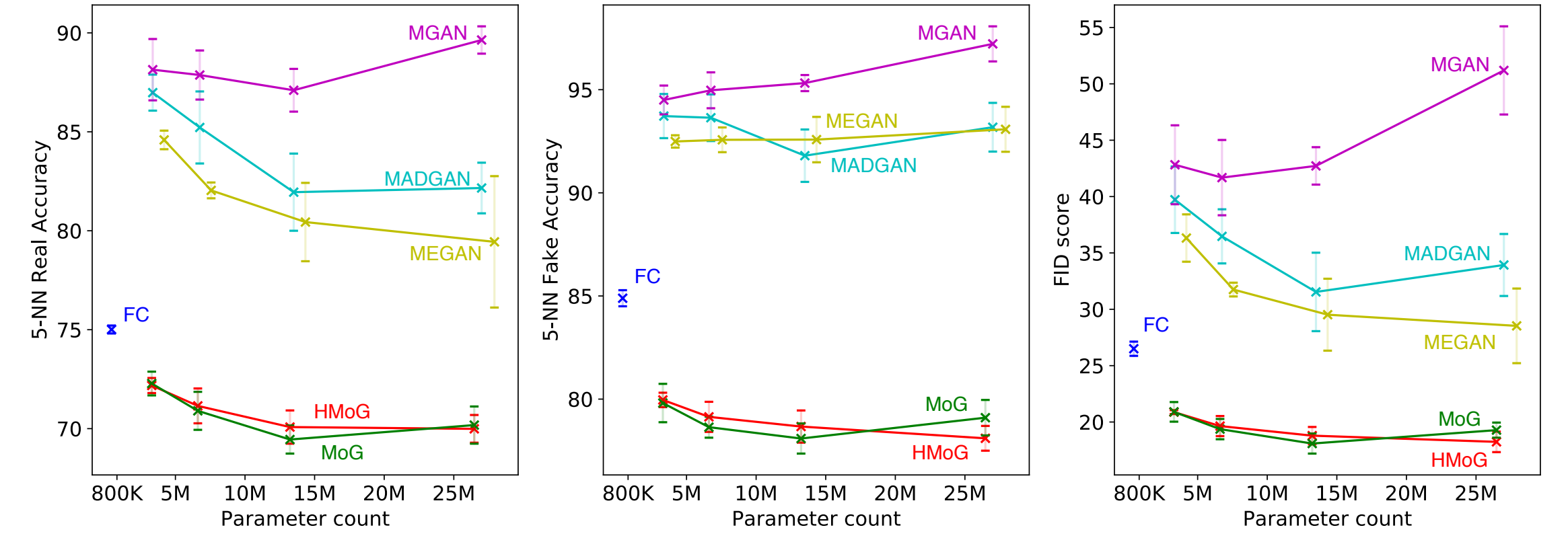
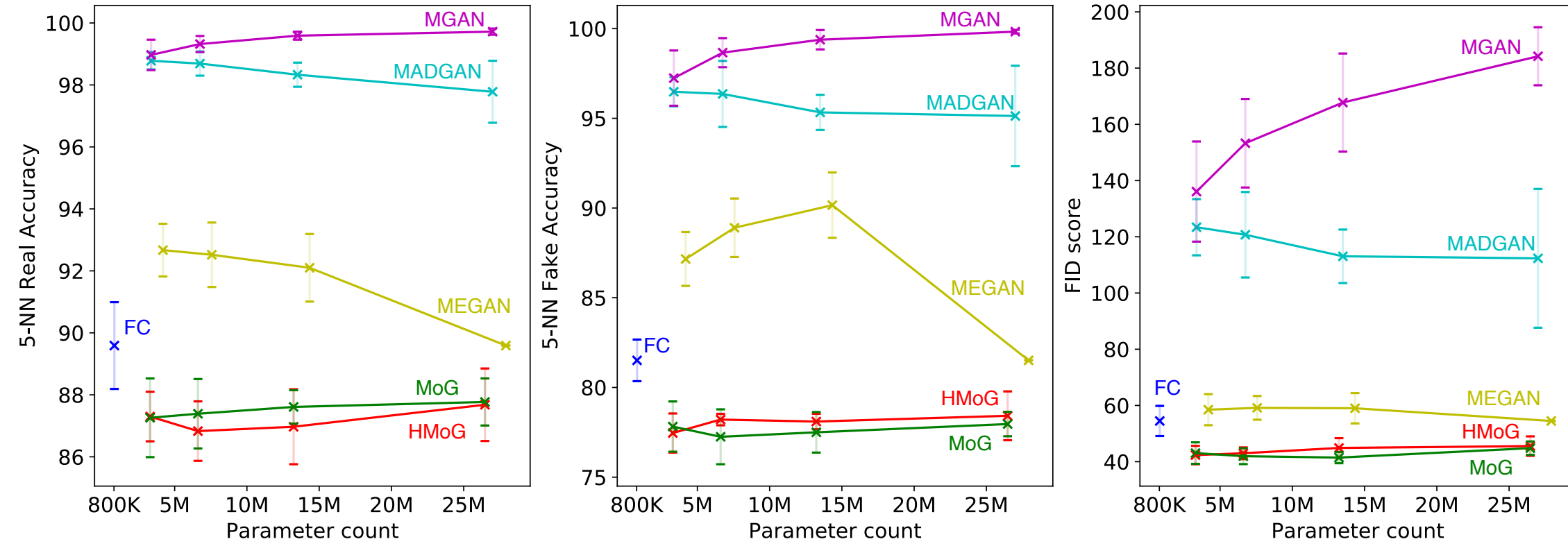
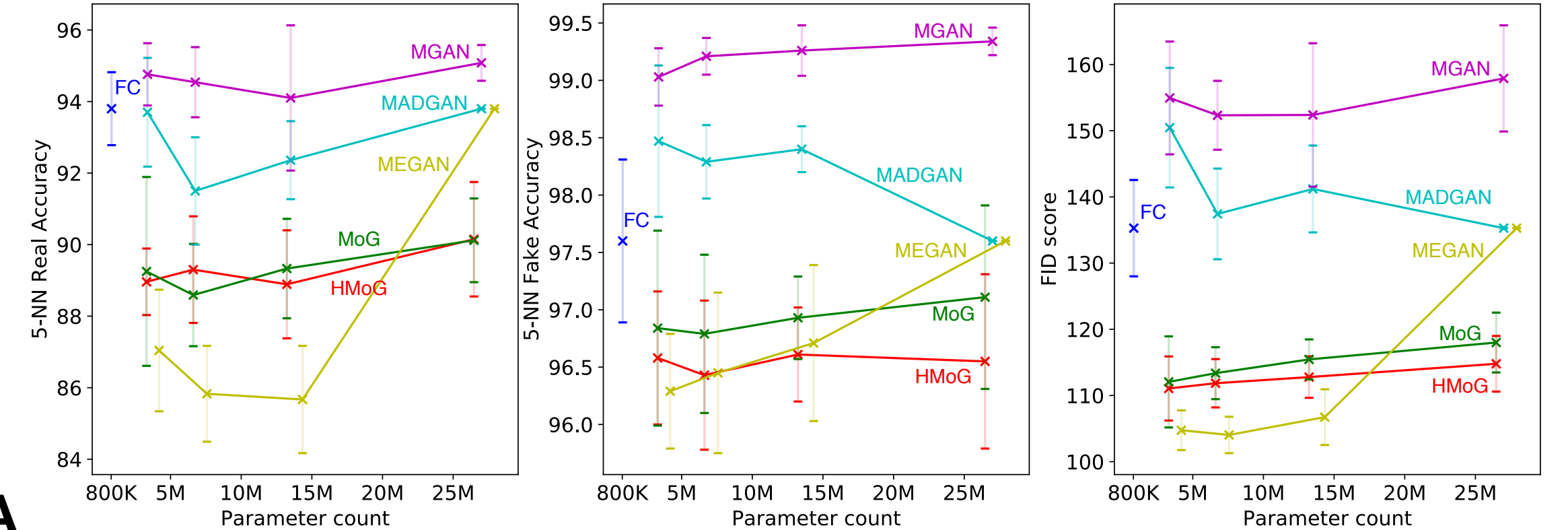
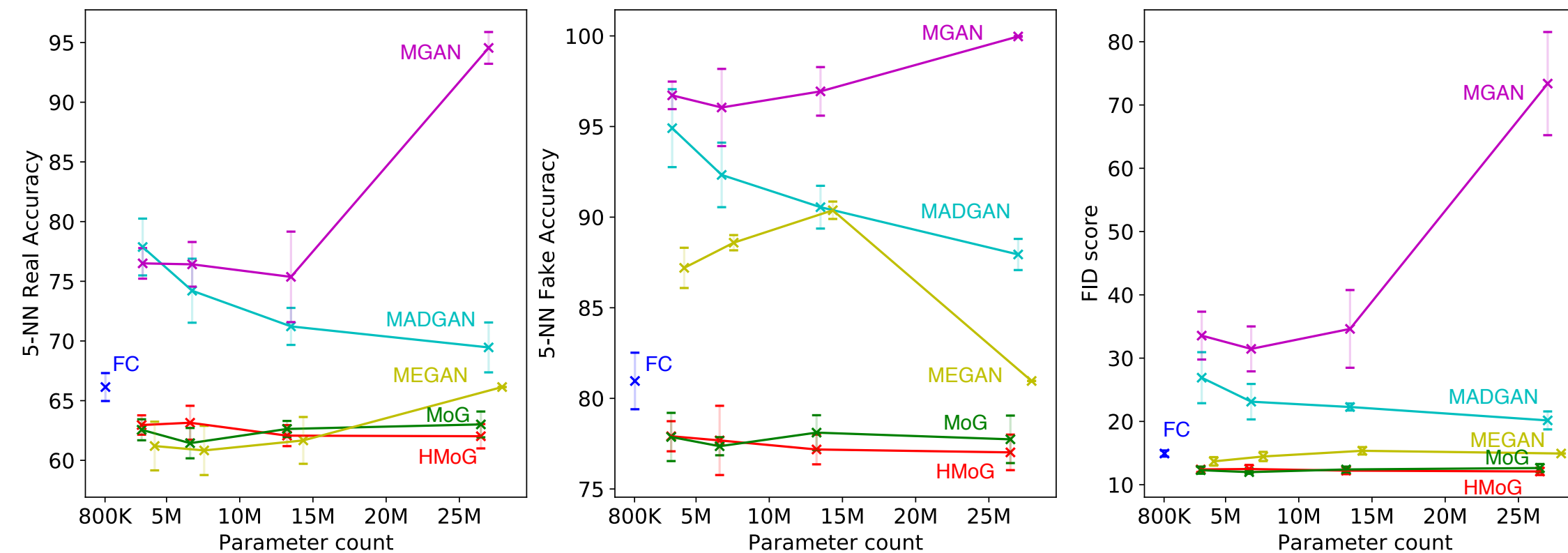
Flat Mixture of Generators (MoG)

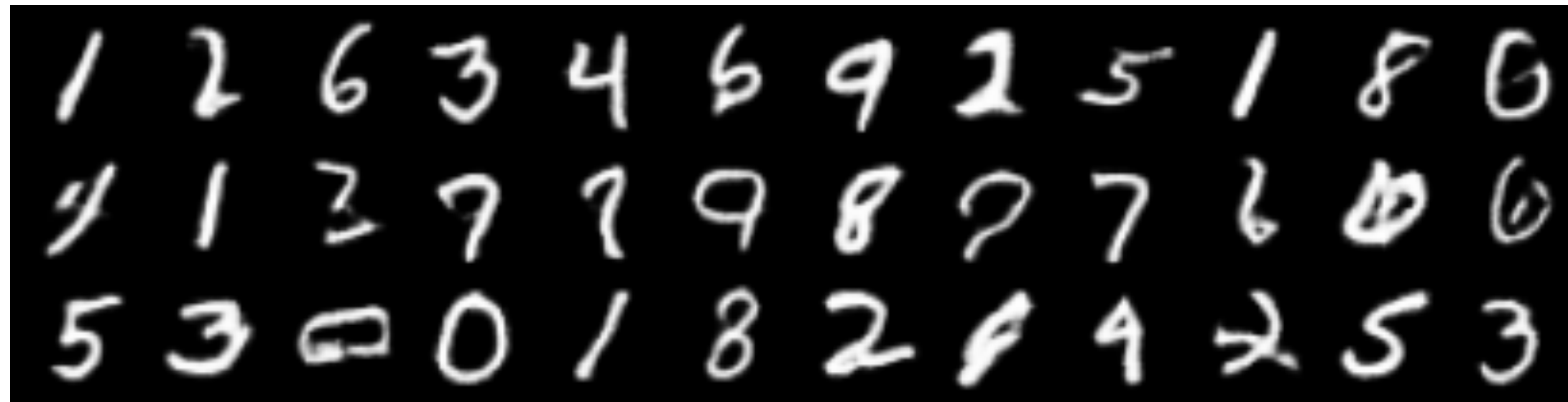
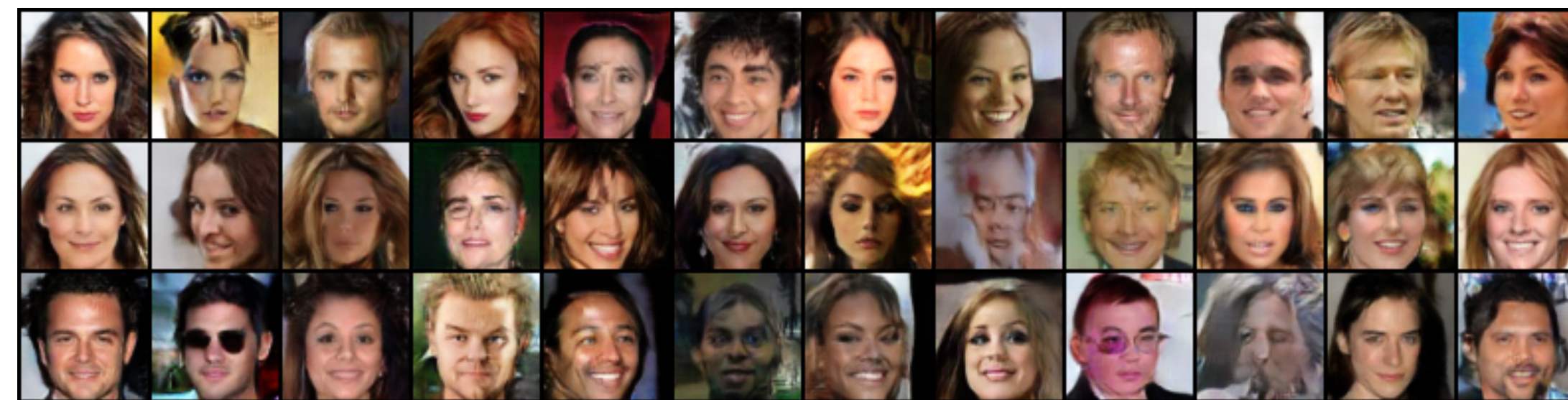


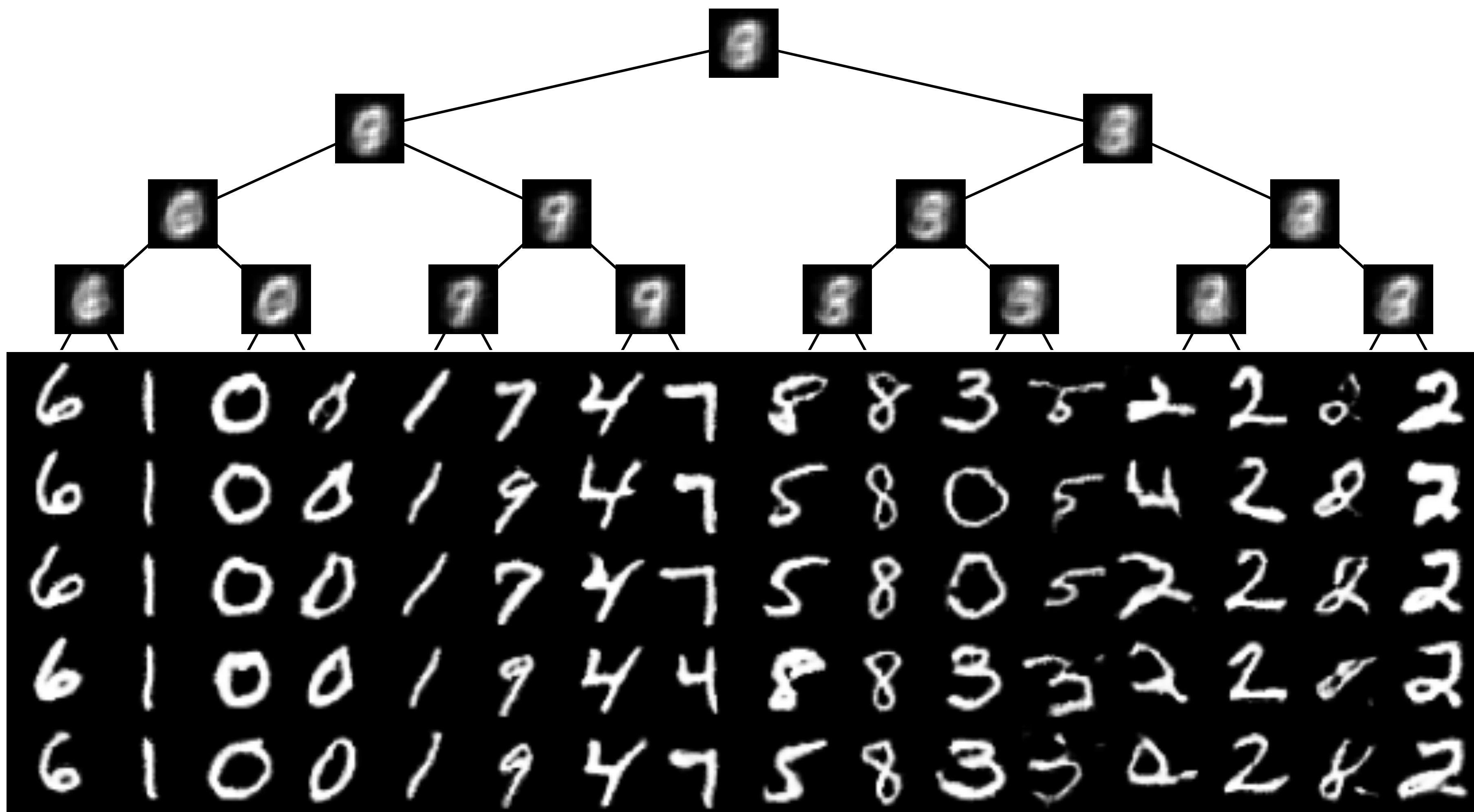
$$a(z) = \mathbf{softmax}(vz + v_0)$$

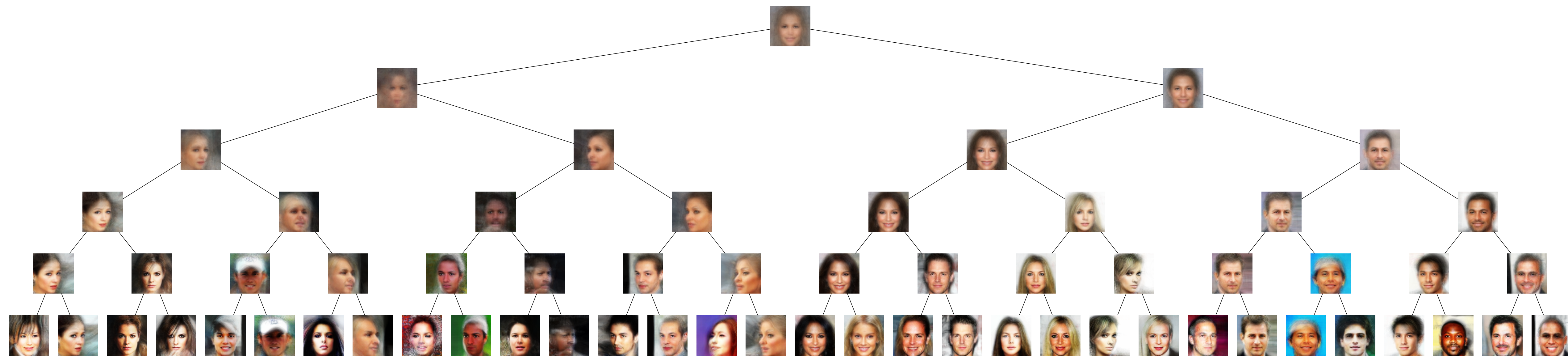
MADGAN**MGAN****HMoG (ours)****MEGAN****MoG (ours)****LEVEL 1****LEVEL 2****LEAVES**

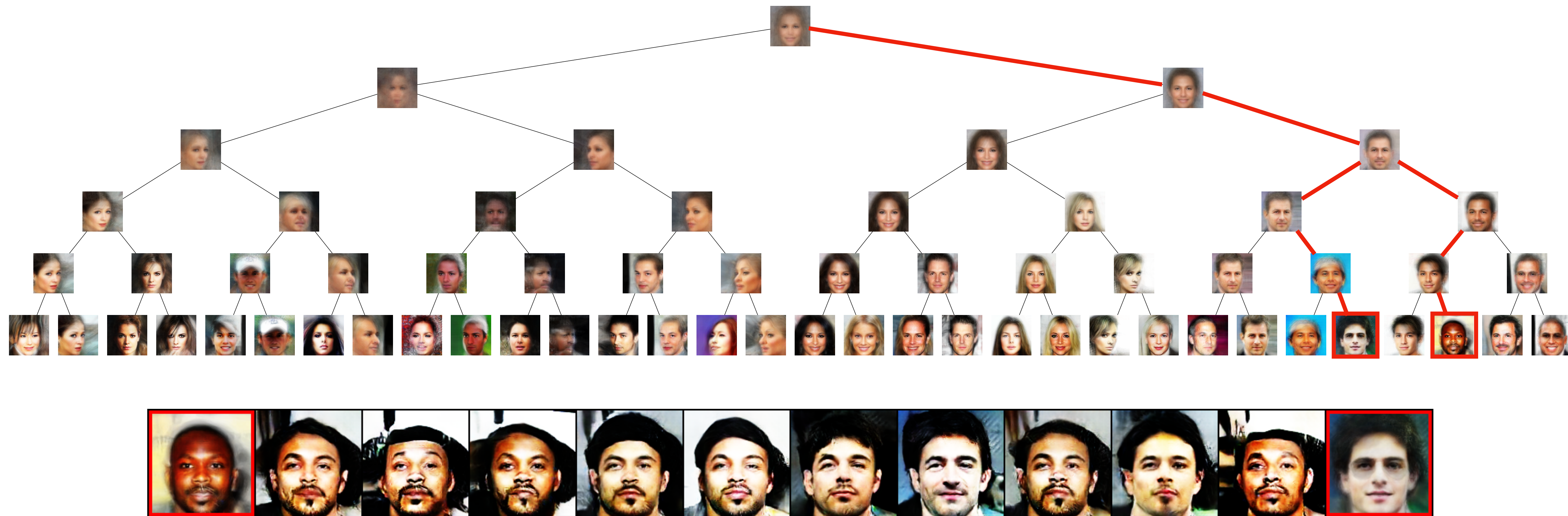


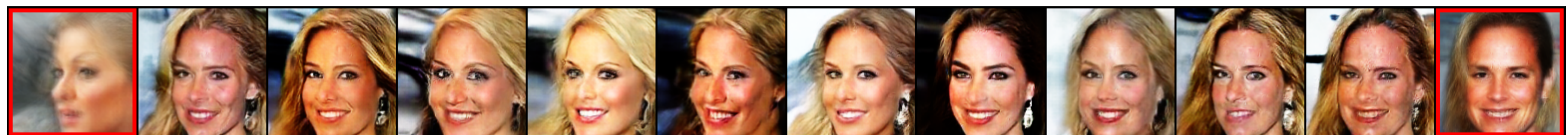
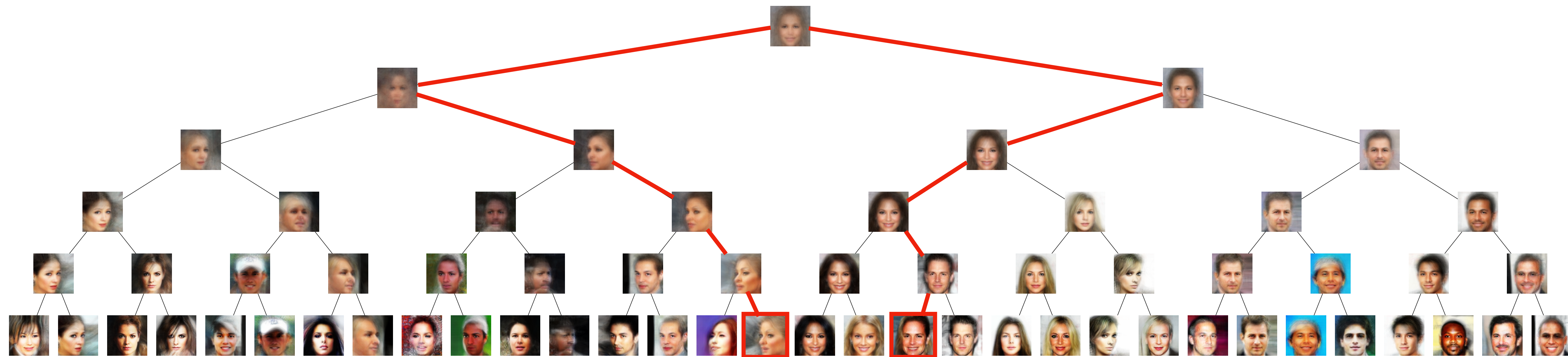
MNIST**FashionMNIST****UTZap50K****Oxford Flowers****CelebA**

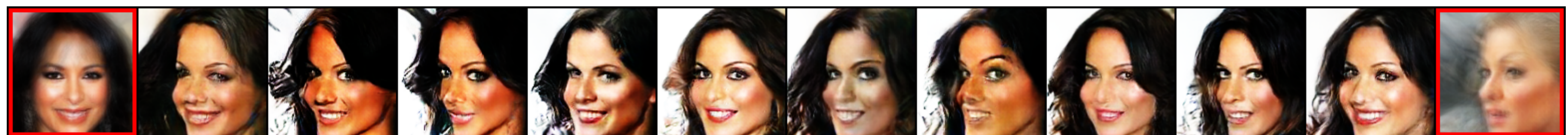
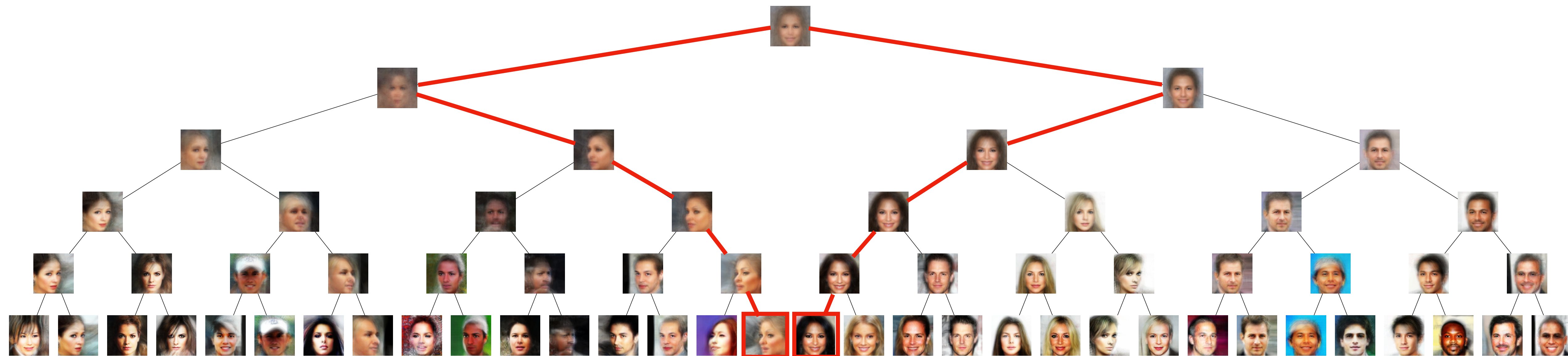
MNIST**FashionMNIST****UTZap50K****Oxford Flowers****CelebA**











Conclusion

- We propose hierarchical mixtures of generators (HMoG).
- HMoG performs better than other multiple generator frameworks based on FID and 5-NN classifier tests.
- HMoG can be easily integrated with other GAN architectures.
- HMoG is interpretable to some extent.

`alpera.xyz/hmog`

Thank you for your time.

Questions and comments are welcome!