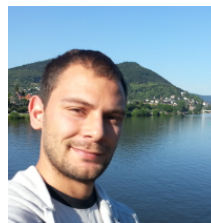


# X-GAN: Latent Optimization by Convex Combinations for Generative Adversarial Network

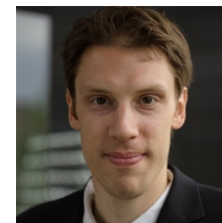
Oliver Blum



Biagio Brattoli



Björn Ommer



Computer Vision Group, IWR & HCI  
Heidelberg University

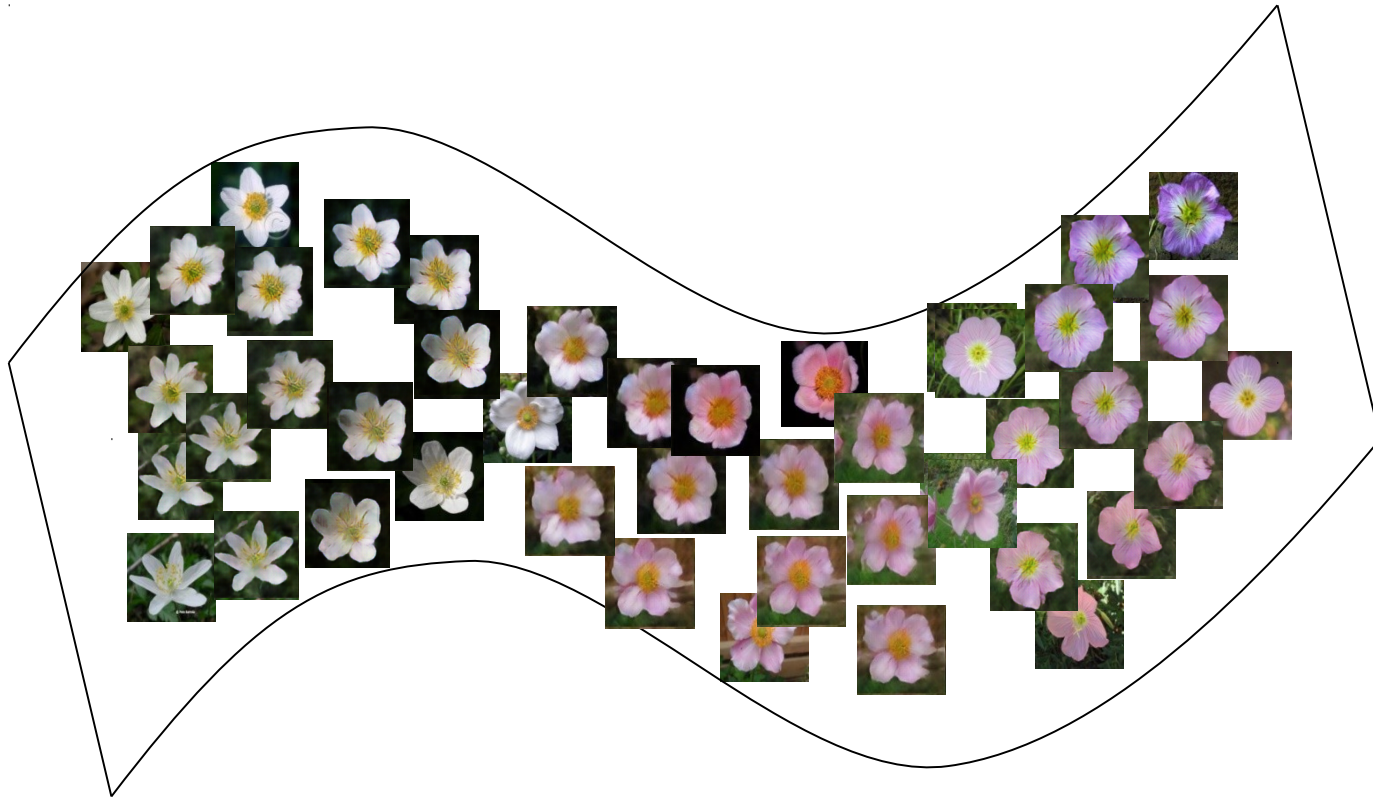


# Motivation



Reconstruction of the continuous Image Manifold

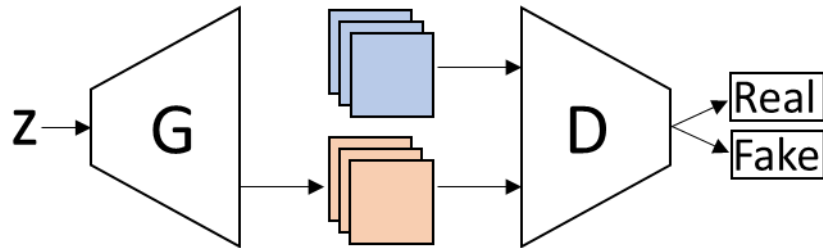
# Motivation



Reconstruction of the continuous Image Manifold

# Issues with Current Models

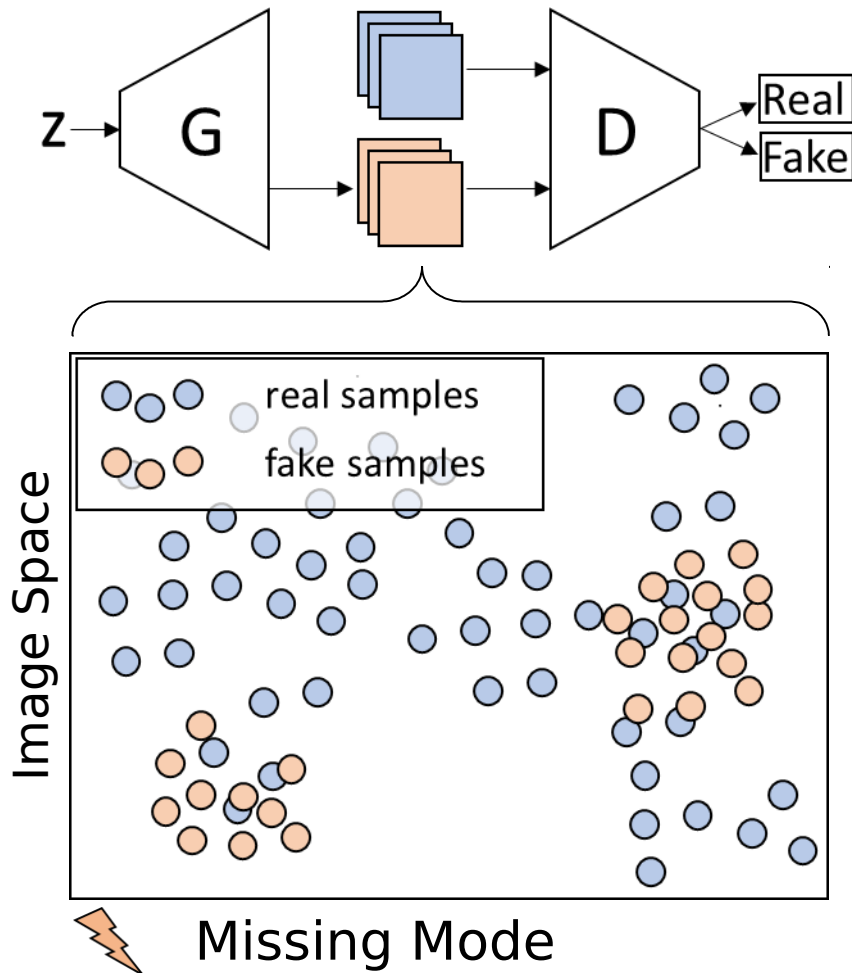
## Generative Adversarial Net





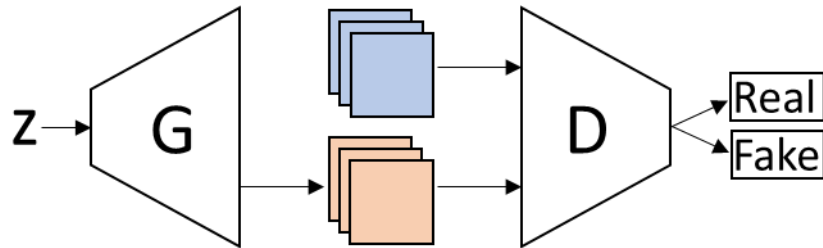
# Issues with Current Models

## Generative Adversarial Net

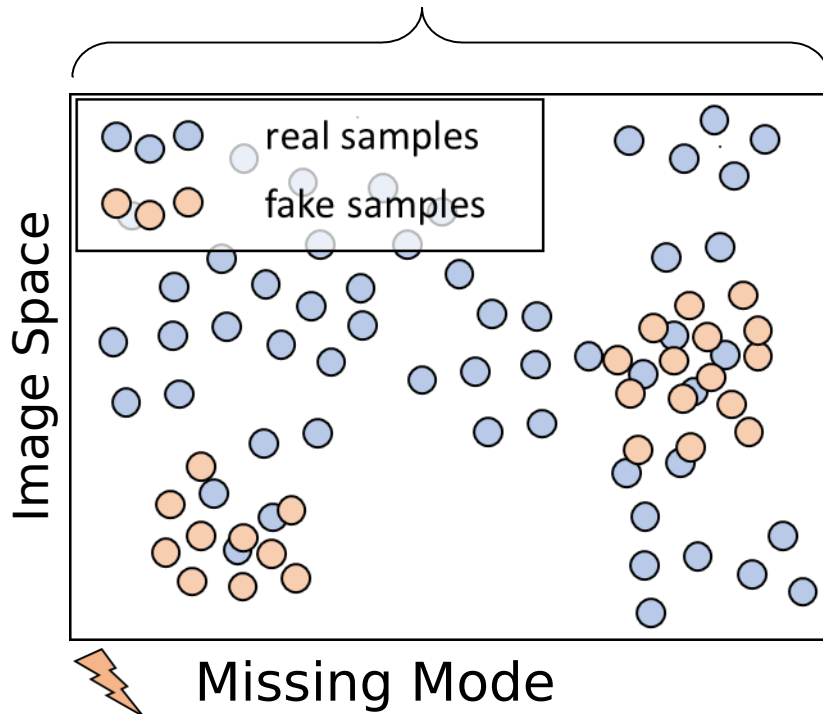
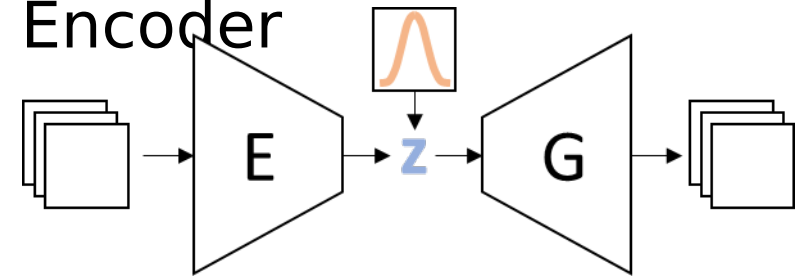


# Issues with Current Models

Generative Adversarial Net

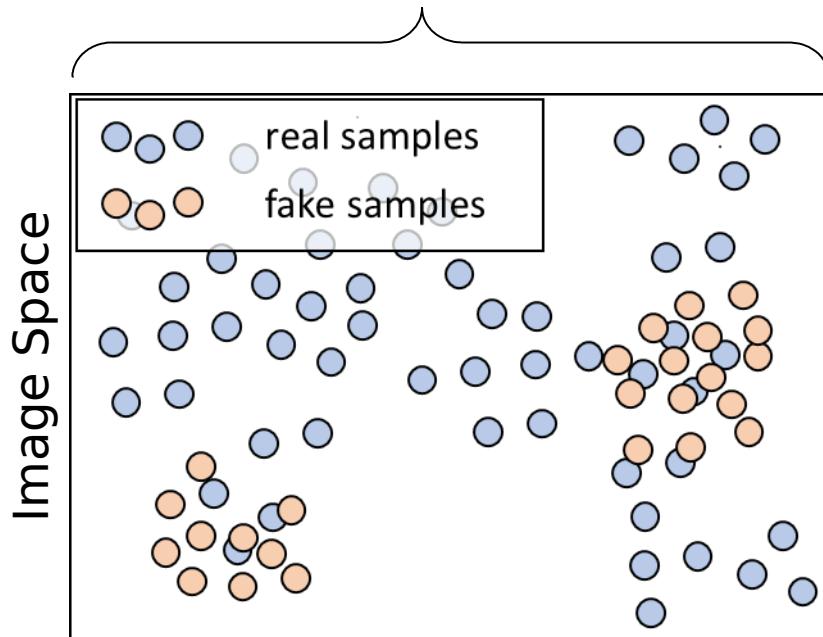
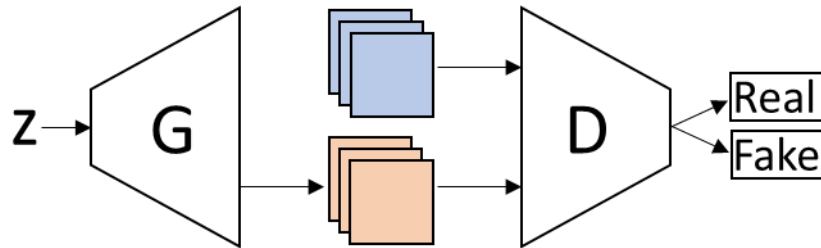


Variational Auto Encoder



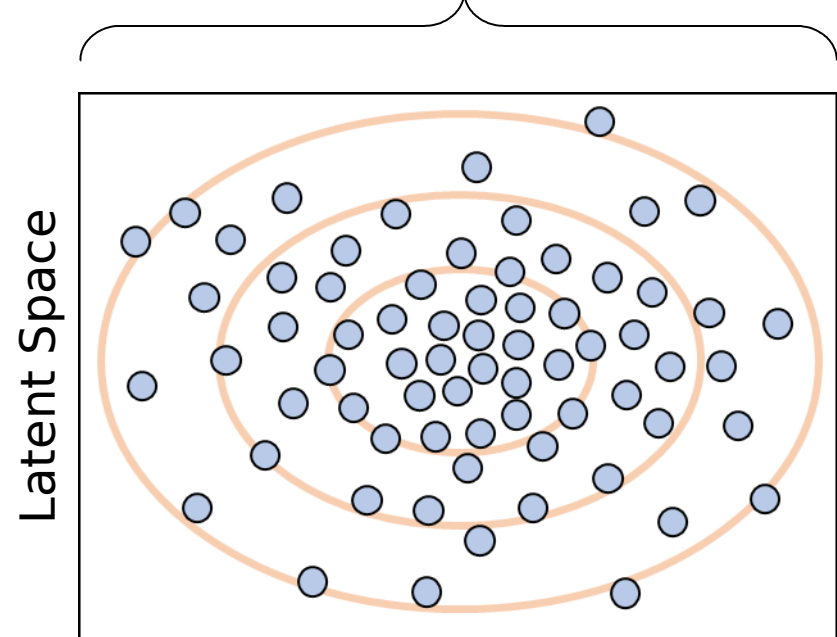
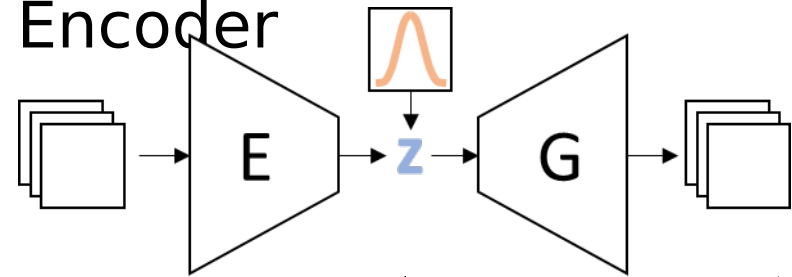
# Issues with Current Models

## Generative Adversarial Net



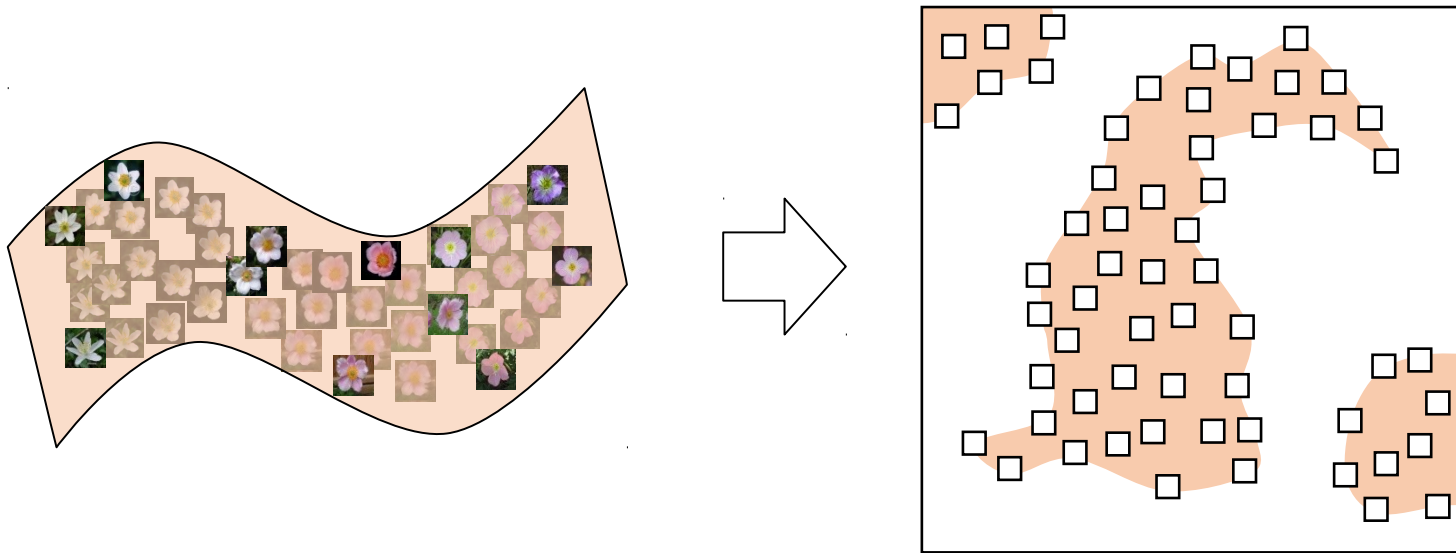
⚡ Missing Mode

## Variational Auto Encoder



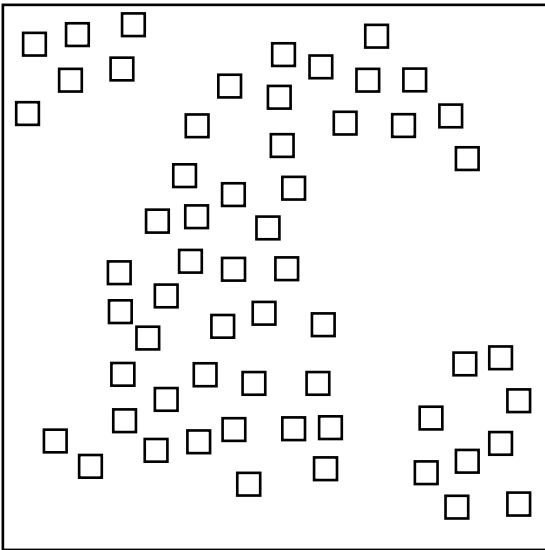
⚡ Overpowering

# Convex Combination Strategy



- □ Latent representations of training images (discrete)
- Latent representation of all realistic images (continuous)

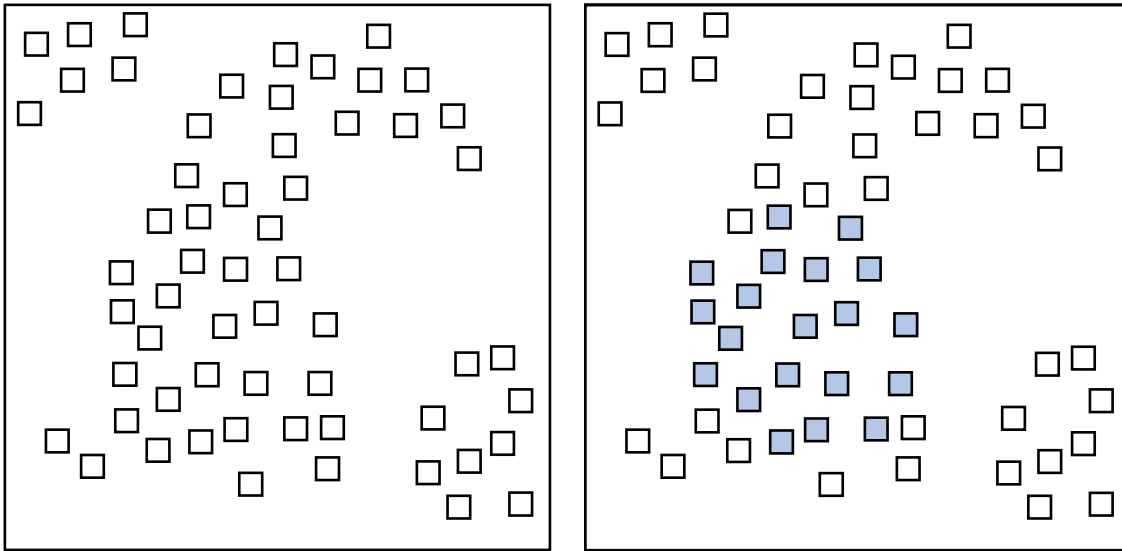
# Convex Combination Strategy



- □ Latent representations of training images (discrete)
- Latent representation of all realistic images (continuous)

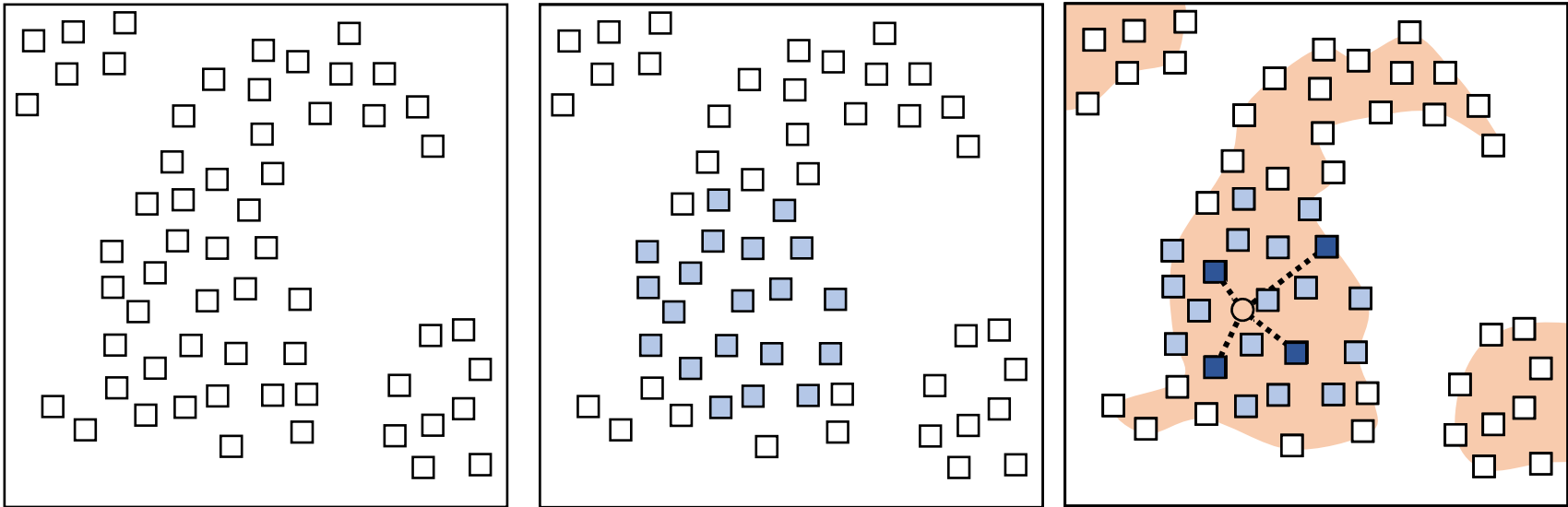


# Convex Combination Strategy



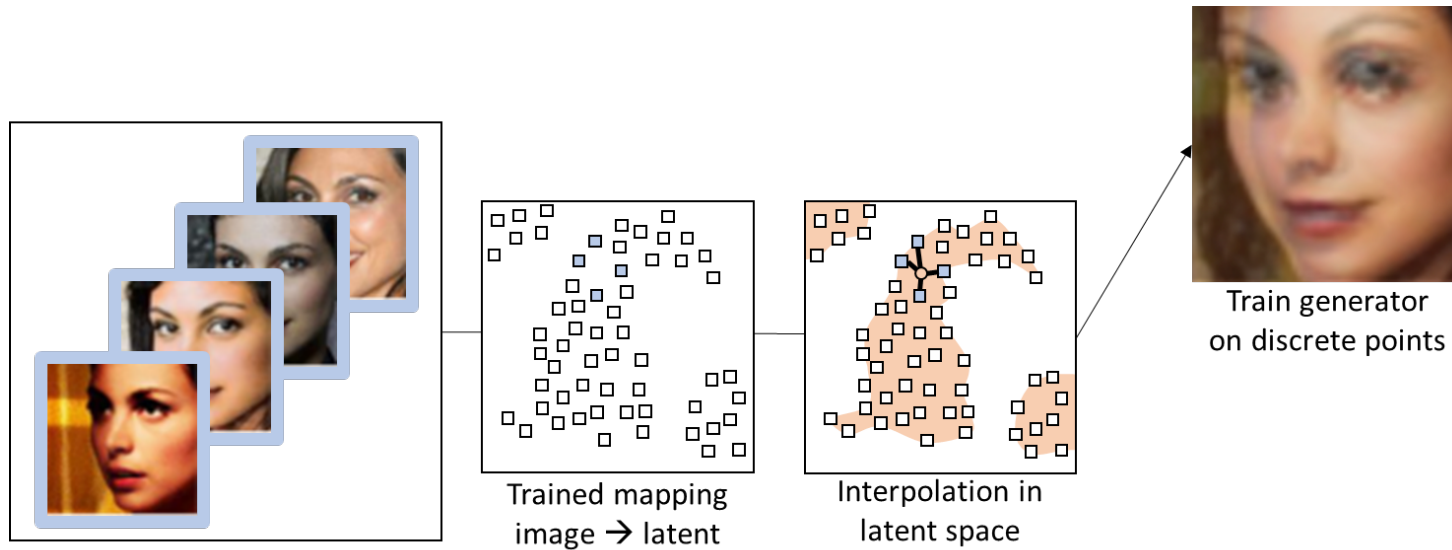
- □ Latent representations of training images (discrete)
- Latent representation of all realistic images (continuous)

# Convex Combination Strategy



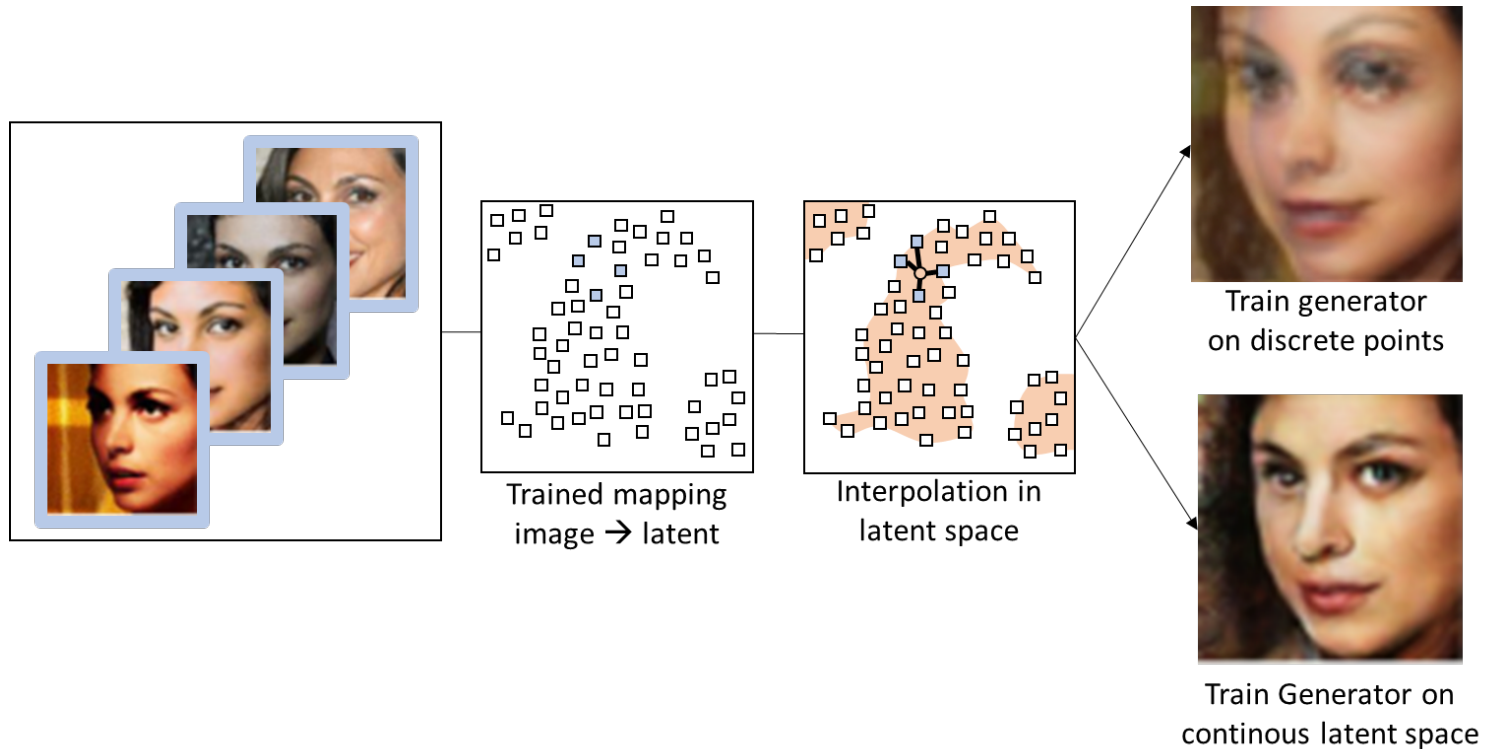
- □ Latent representations of training images (discrete)
- Latent representation of all realistic images (continuous)

# Convex Combination Strategy



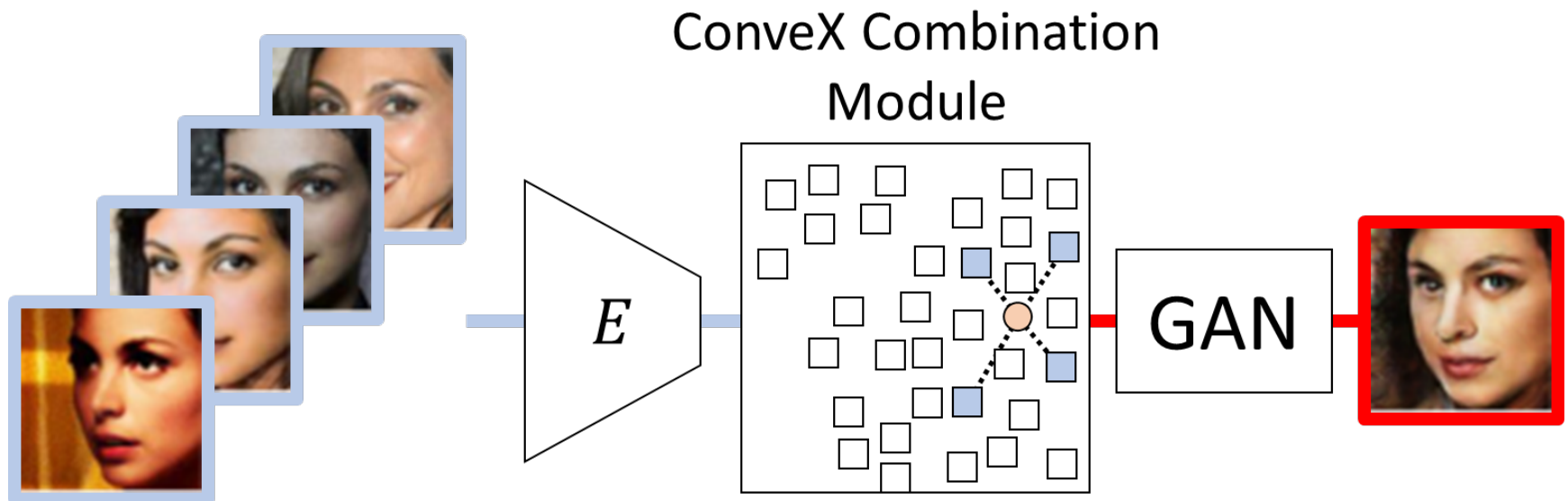
Training with discrete latent vectors

# Convex Combination Strategy



Training with continuous latent vectors

# Convex Combination Strategy

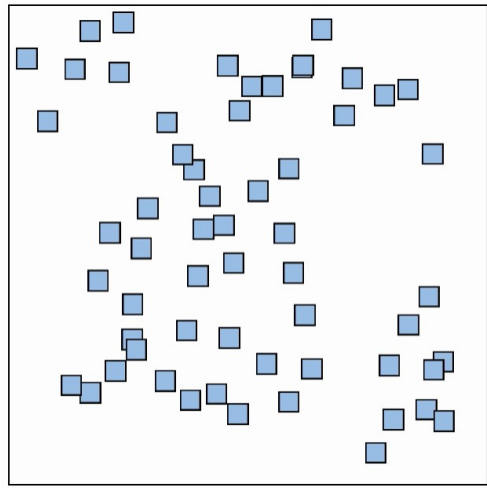
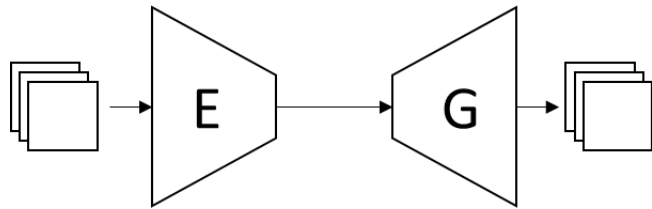


Convex Combinations to sample synthetic feature vectors in latent space

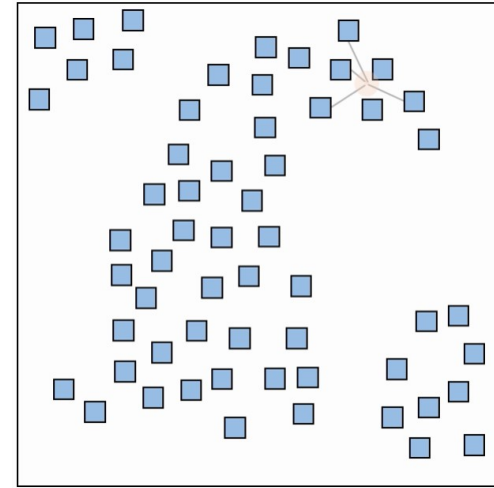
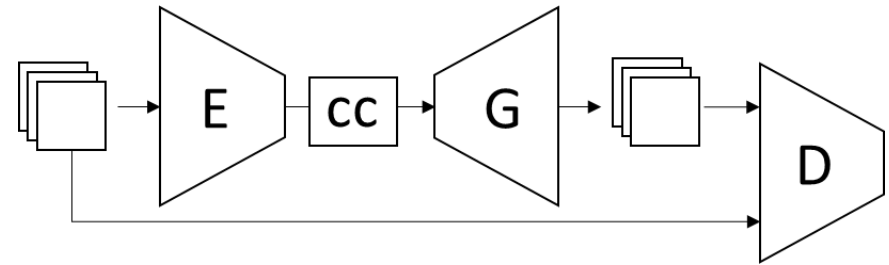


# Implementation: Training

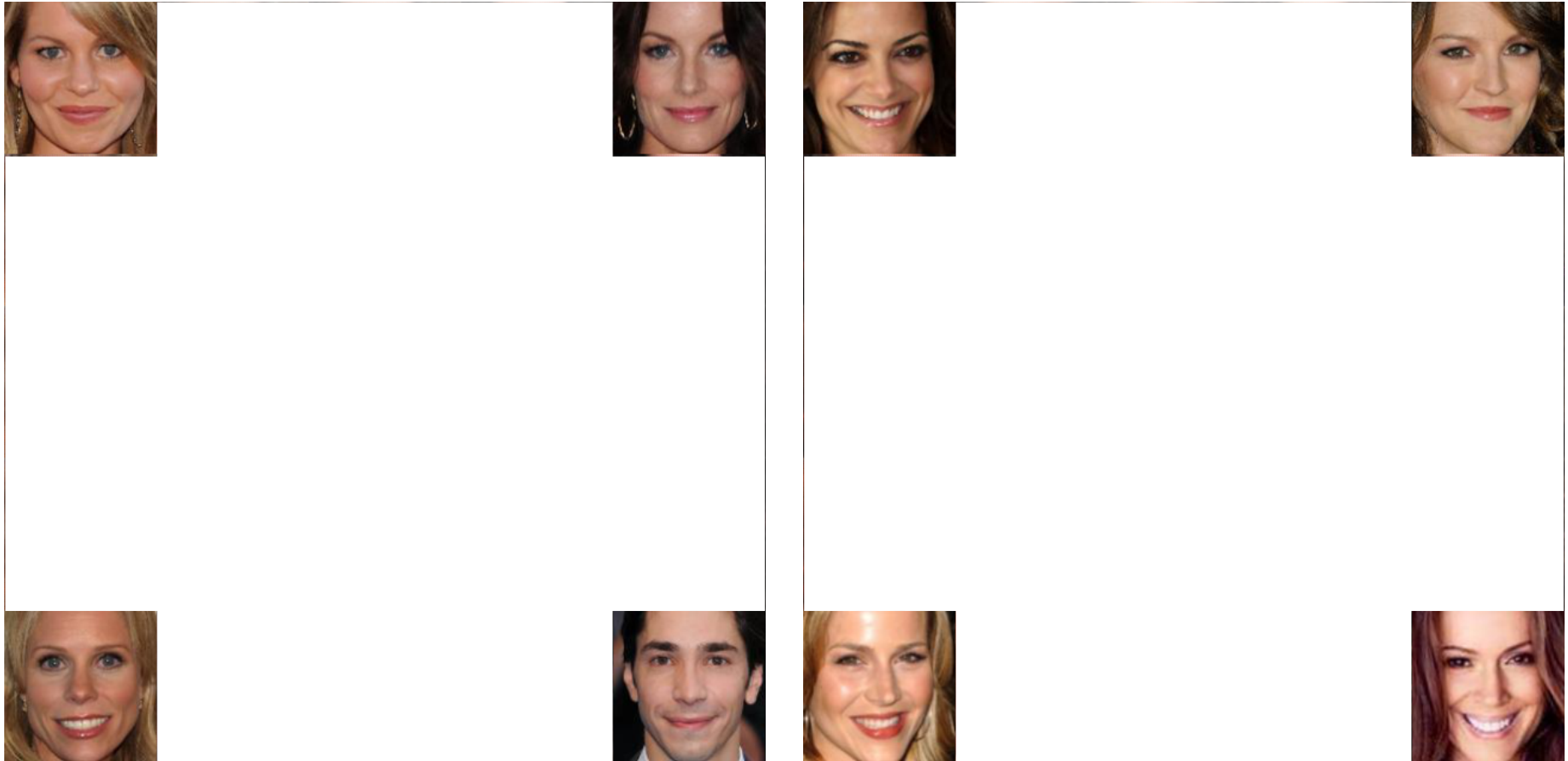
Auto-Encoder: Real data



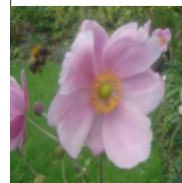
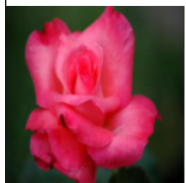
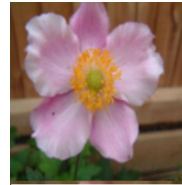
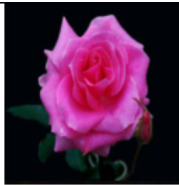
GAN: Sampled data



# Results: Image morphing



# Results: Image morphing



# Results: Visual Comparison

**MD-GAN****CVAE-GAN****X-GAN****real**

# Results: Inception Score

Score quantifying image realism

Model	Inception Score
CVAE-GAN	1.757
<b>X-GAN</b>	<b>1.831</b>
Real Data (upper bound)	2.372

*Tim Salimans, Ian Goodfellow, Wojciech Zaremba, Vicki Cheung, Alec Radford, and Xi Chen. Improved techniques for training gans. In Advances in Neural Information Processing Systems, pages 2234-2242, 2016.*



# Results: D\* - Score

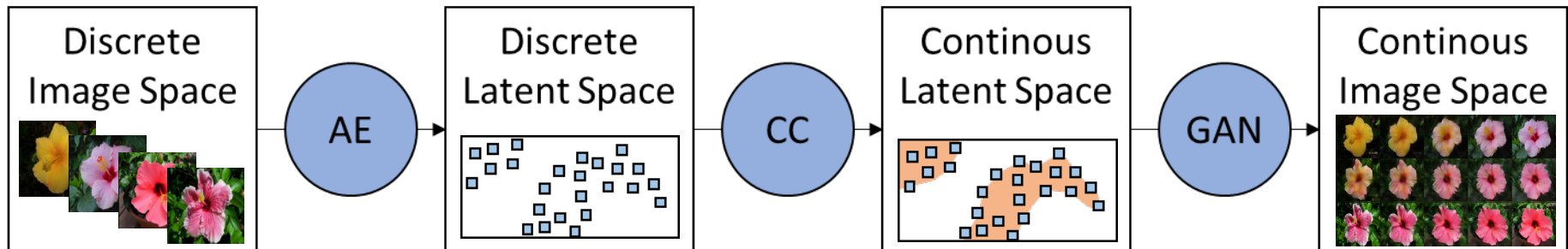
Estimate number of test images that clearly belong to a missing mode

Model	D* - Score
MD-GAN	2299 / 4412
CVAE-GAN	87 / 4412
<b>X-GAN</b>	<b>15 / 4412</b>

*Tong Che, Yanran Li, Athul Paul Jacob, Yoshua Bengio, and Wenjie Li. Mode regularized generative adversarial networks. arXiv preprint arXiv:1612.02136, 2016.*

# Conclusion

- Novel convex combination strategy:



- X-GAN overcomes current issues
  - No latent space prior => No overpowering
  - Sample from entire manifold => mitigate missing mode
- X-GAN outperforms state of the art approaches