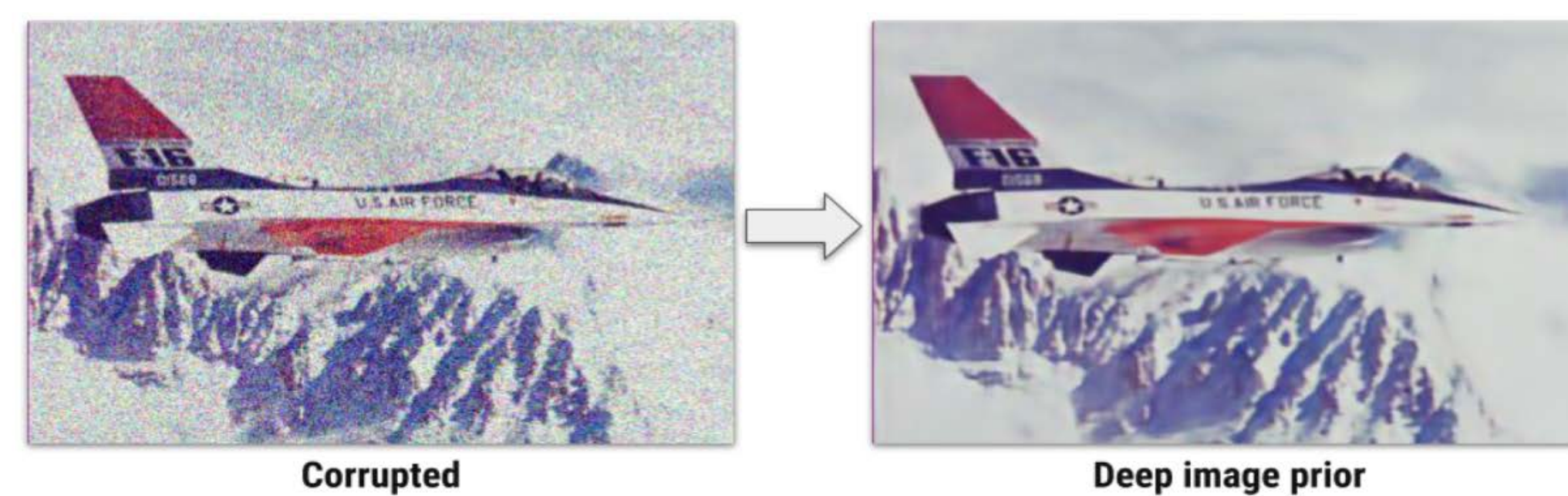


Basic Ideas



denoising with autoencoders

Question : Can we use same idea to restore damaged photos ?

Dataset

Heterogeneous

Animals Data

Training Set 25,000
Test Set 2,000

Homogeneous

Faces Data

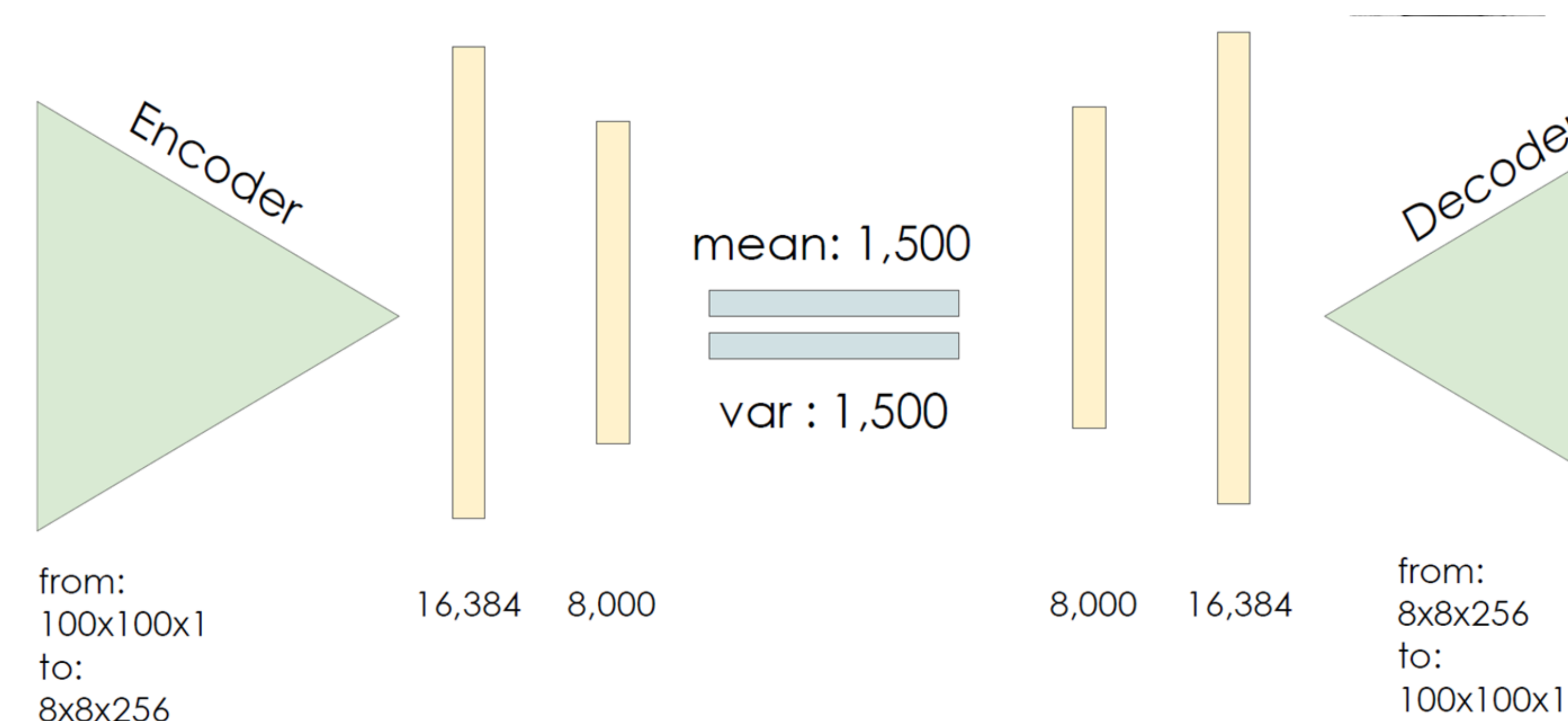
Training Set 39,000
Test Set 2,000

Input Data



Model, Implementation & Results

Variational Autoencoders



Encoder

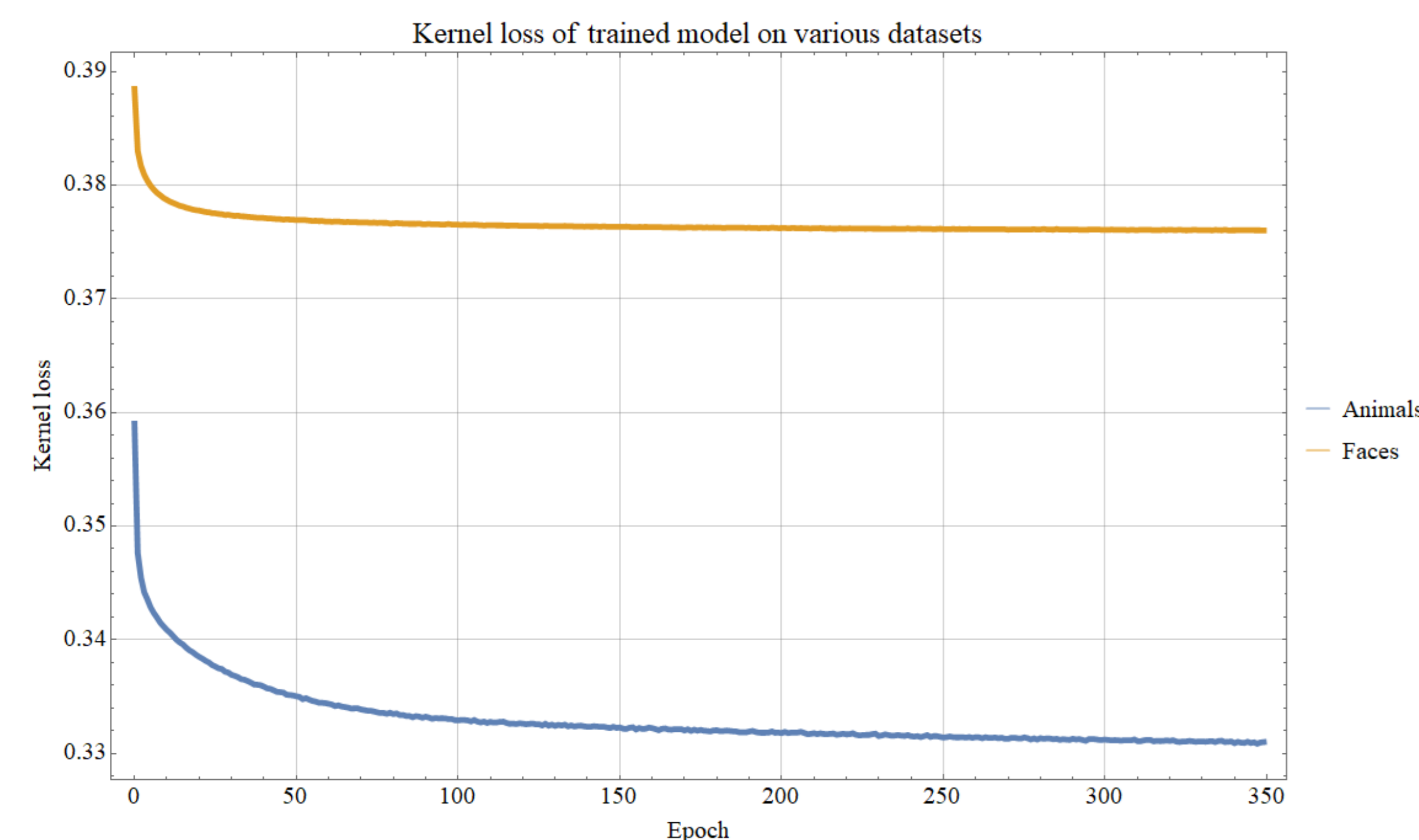
- 7 Convolutional Layers
- 3 Pool Layers
- 3 BatchNorm Layers
- VGG16 architecture

Decoder

- 6 Deconvolutional Layers
- 5 BatchNorm Layers
- DC-GAN architecture

Structural Similarity investigation

$$SSIM(x, y) = \frac{(2\mu_x\mu_y + c_1)(2\sigma_{xy} + c_2)}{(\mu_x^2 + \mu_y^2 + c_1)(\sigma_x^2 + \sigma_y^2 + c_2)}$$



Performances

Experiment 1: Heterogeneity vs Homogeneity



	SSIM	Train Set	Test Set
Animals		0.8874 ± 0.0493	0.6869 ± 0.1606
Faces		0.9168 ± 0.0524	0.9012 ± 0.0537

Experiment 2 : Various interested cases

Dataset	SSIM (Test Set)
40% occultation	0.8692 ± 0.0590
2x20% boxes	0.8553 ± 0.0617



Outlook

- Variation of SSIM subjects to occultation ratio.
- SSIM is arbitrary occultation shapes.

References

- Fawzi et.al. Image Inpainting through Neural Networks hallucinations. 2016
- Burlin et.al. Deep Image Inpainting. 2017
- Ham et.al. Variational Image Inpainting. 2018
- Pinho et.al. Unsupervised Learning for Concept Detection in Medical Images: A Comparative Analysis. doi : 10.3390/app8081213. 2018
- Yu et.al. Generative Image Inpainting with Contextual Attention. 2018
- Nazeri et.al. EdgeConnect: Generative Image Inpainting with Adversarial Edge Learning. 2019
- Erofeev in Towards Data Science Image Inpainting: Humans vs. AI. 2018
- Takahashi et.al. Adaptive Image Inpainting Algorithm based on generalized Principal Component Analysis. 2016