



Generating images of thermal cats and dogs

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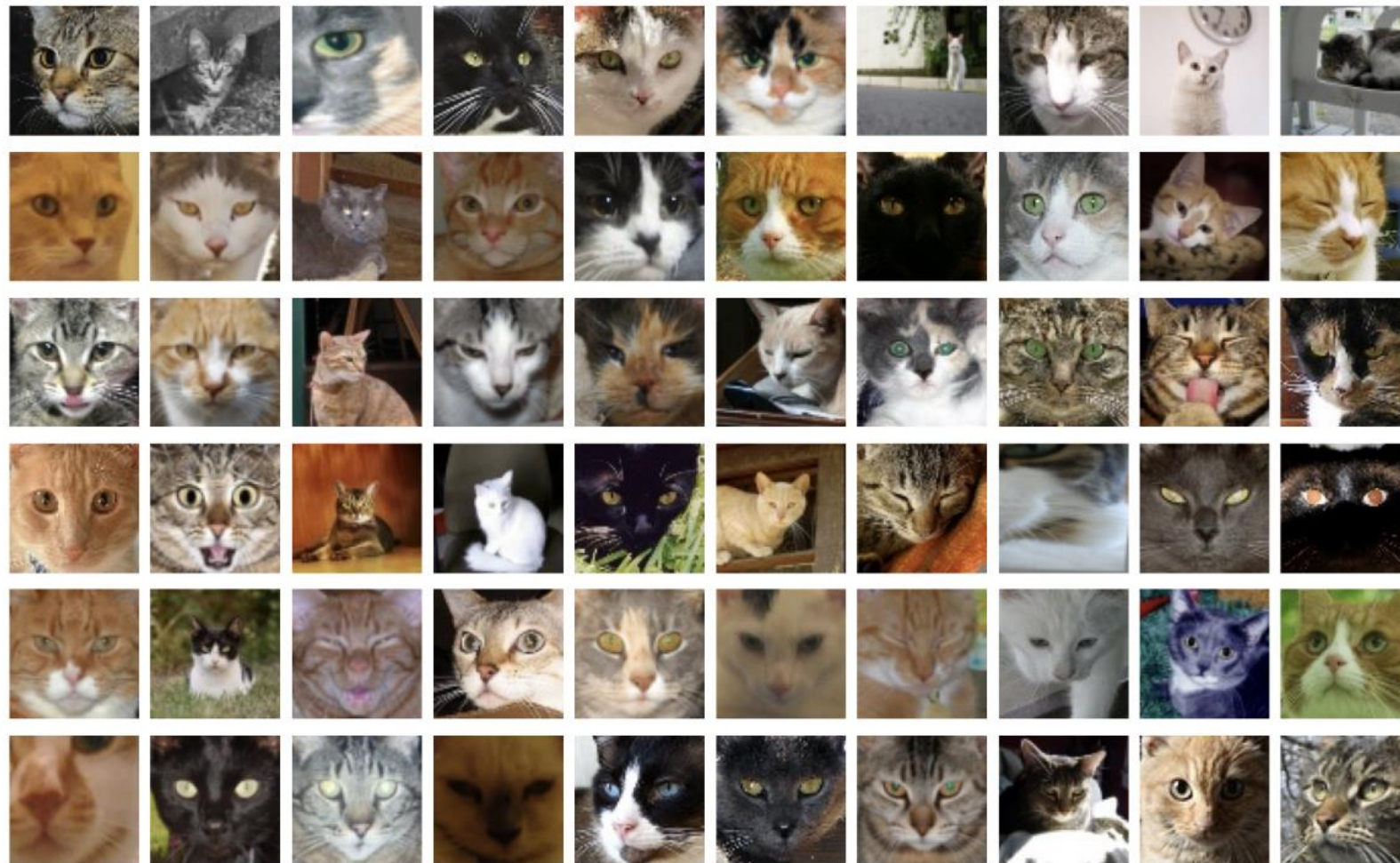
Agenda

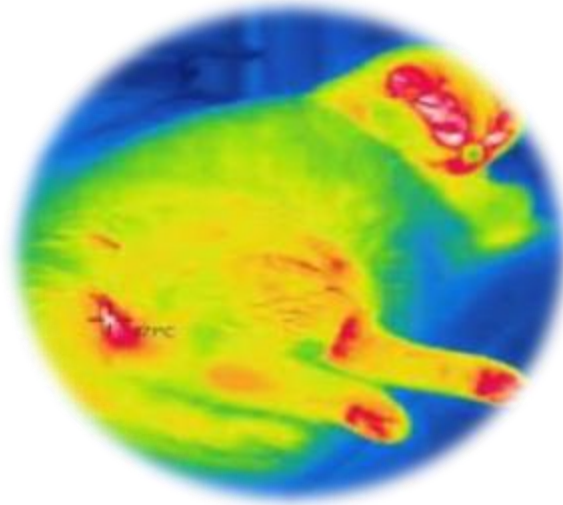
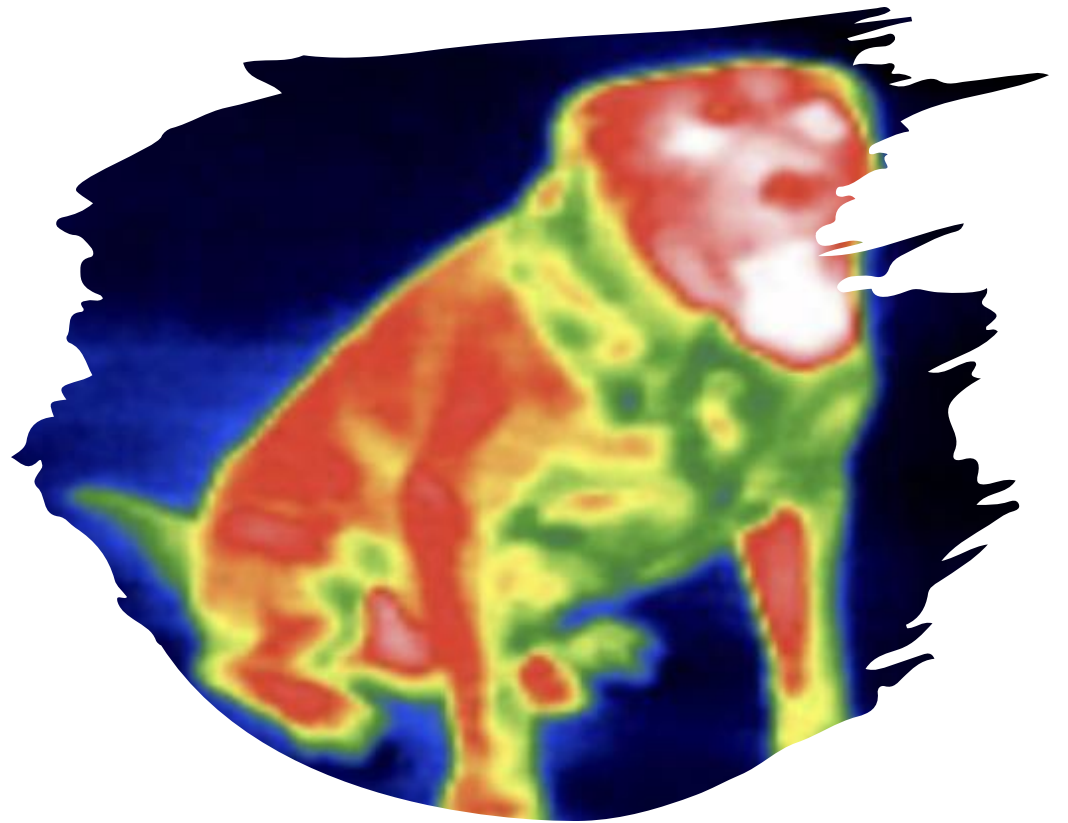
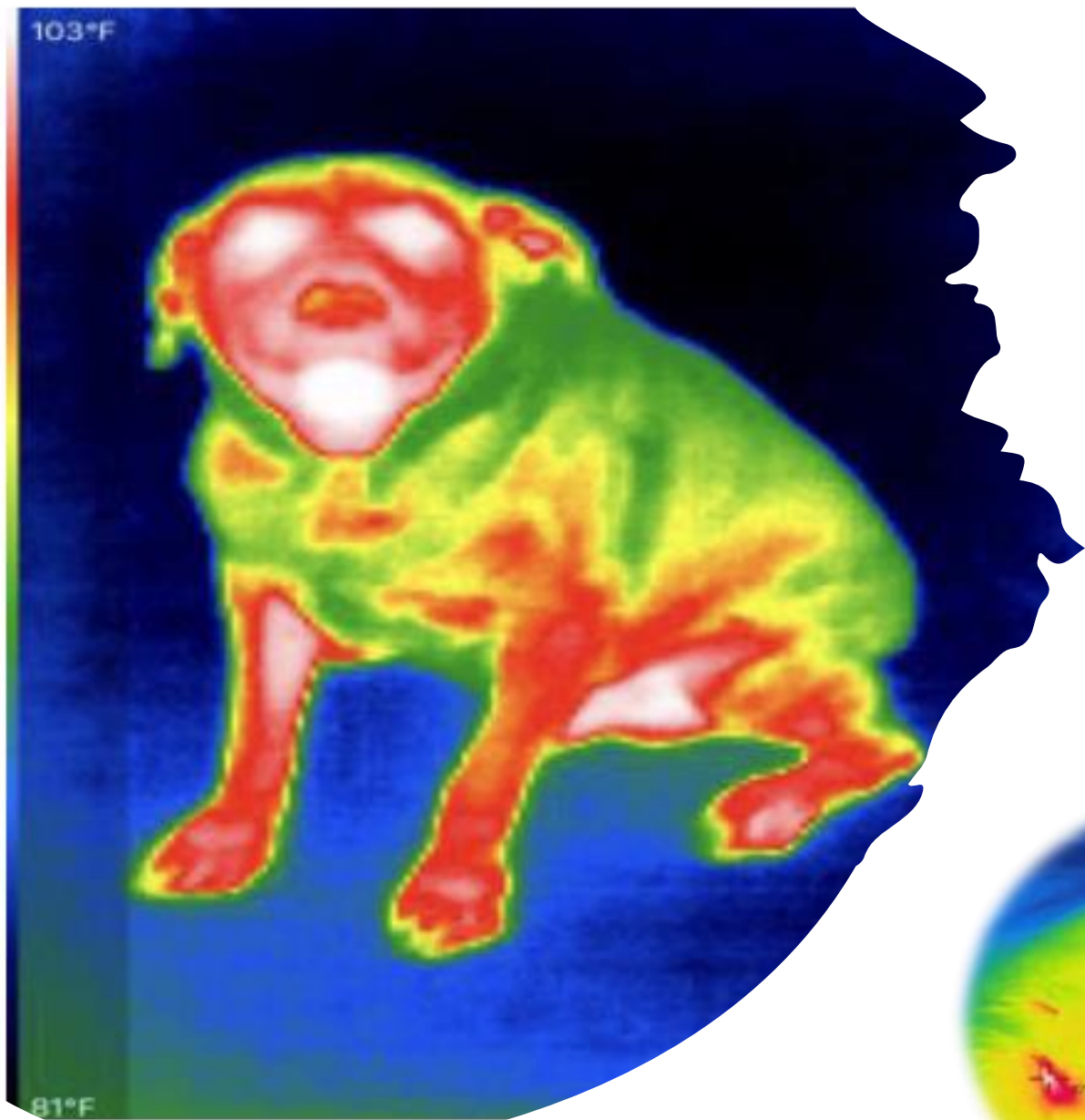
- Dataset
- Augmentation techniques
- Implementations
- Results

Dataset

- 29,843 RGB images
- 64x64 pixels

Cat Dataset

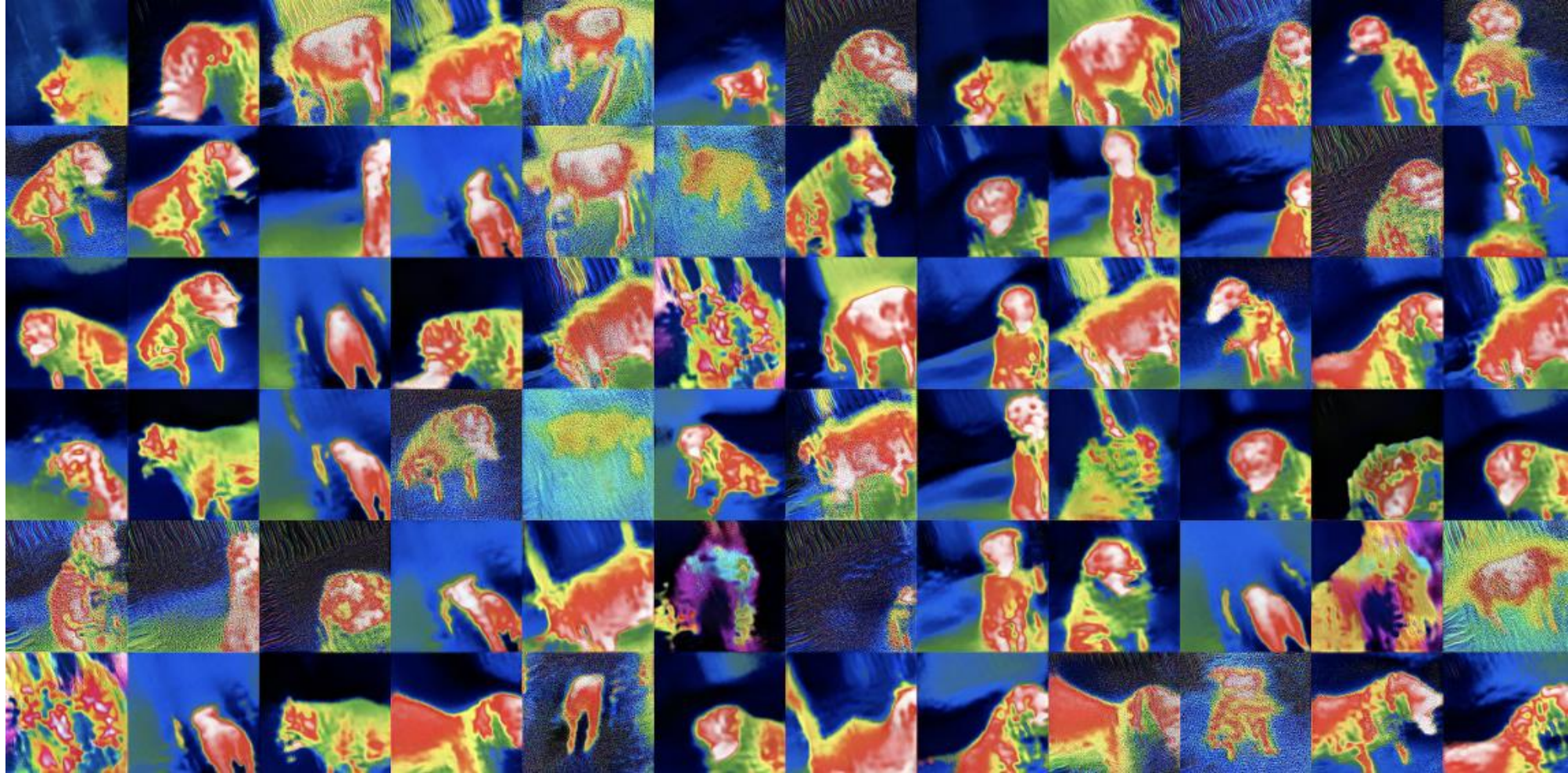




Dataset

- About 200 images
- Mostly 512x512 pixels

Augmentation

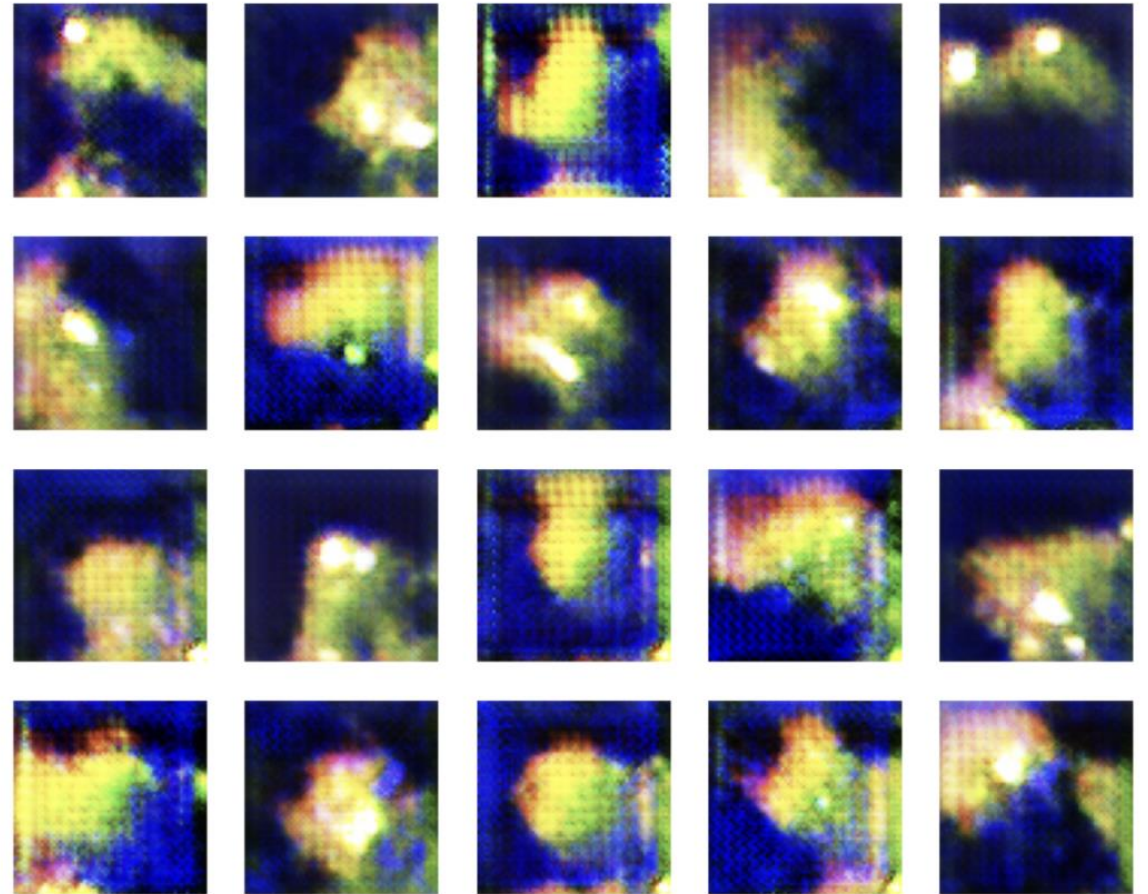


Implementations

- Cycle-Gan
- Diffusion-Based model
- Style-Gan
- Style-Gan2-ADA
- For all models we changed photos to size to maintain the same size (128x128 or 256x256 depends on model)

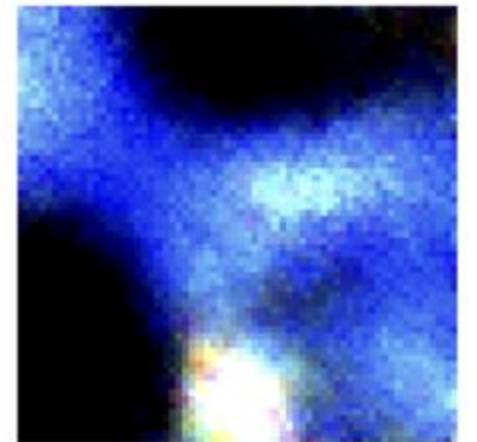
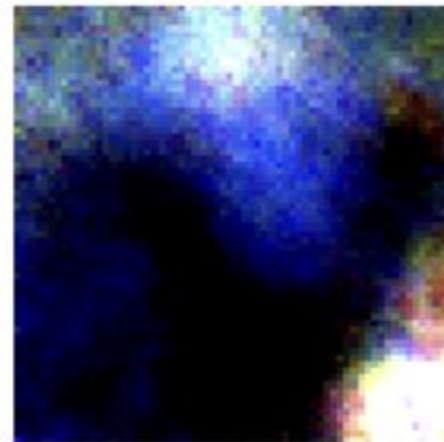
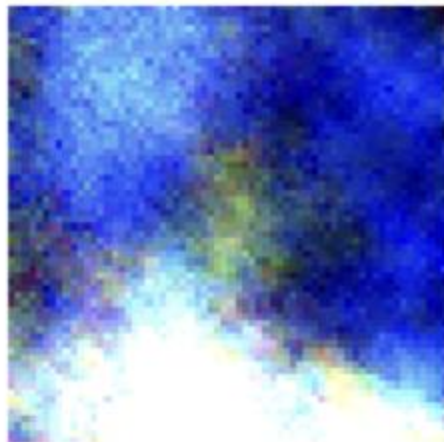
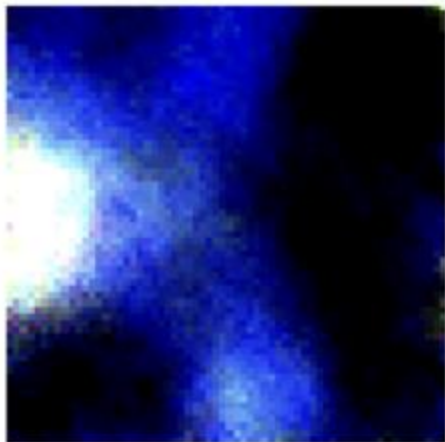
Implementations – Cycle-Gan

- Parameters:
 - Adam optimizers,
 - Learning rate of 2×10^{-4}
 - Loss function : MSE



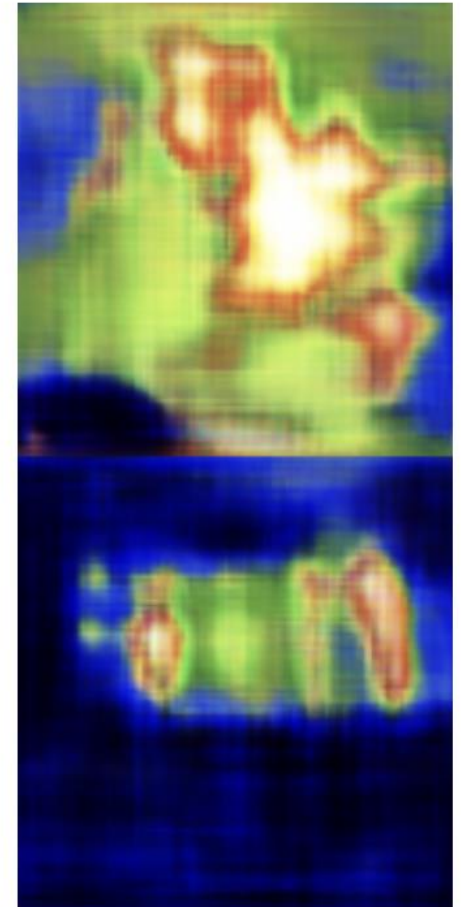
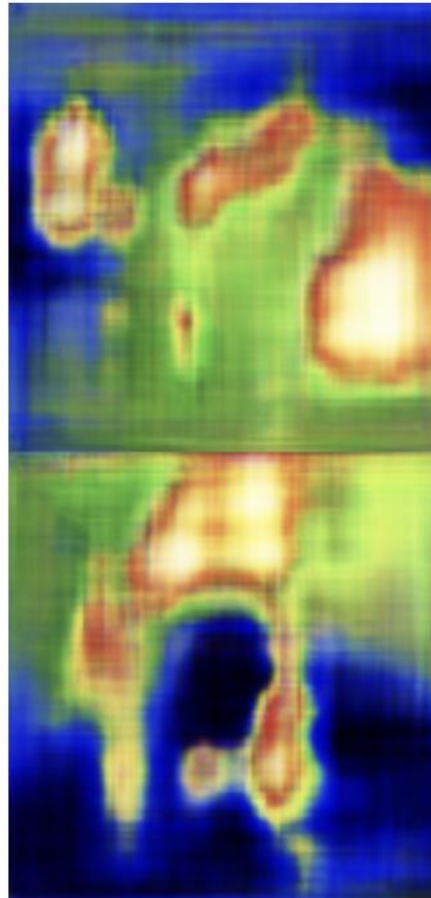
Implementations – Diffusion-based

- Parameters:
 - 50 epochs using an Adam optimizer
 - learning rate of 1×10^{-4}
 - Mean Squared Error (MSE) as the loss function.



Implementations – Style-Gan

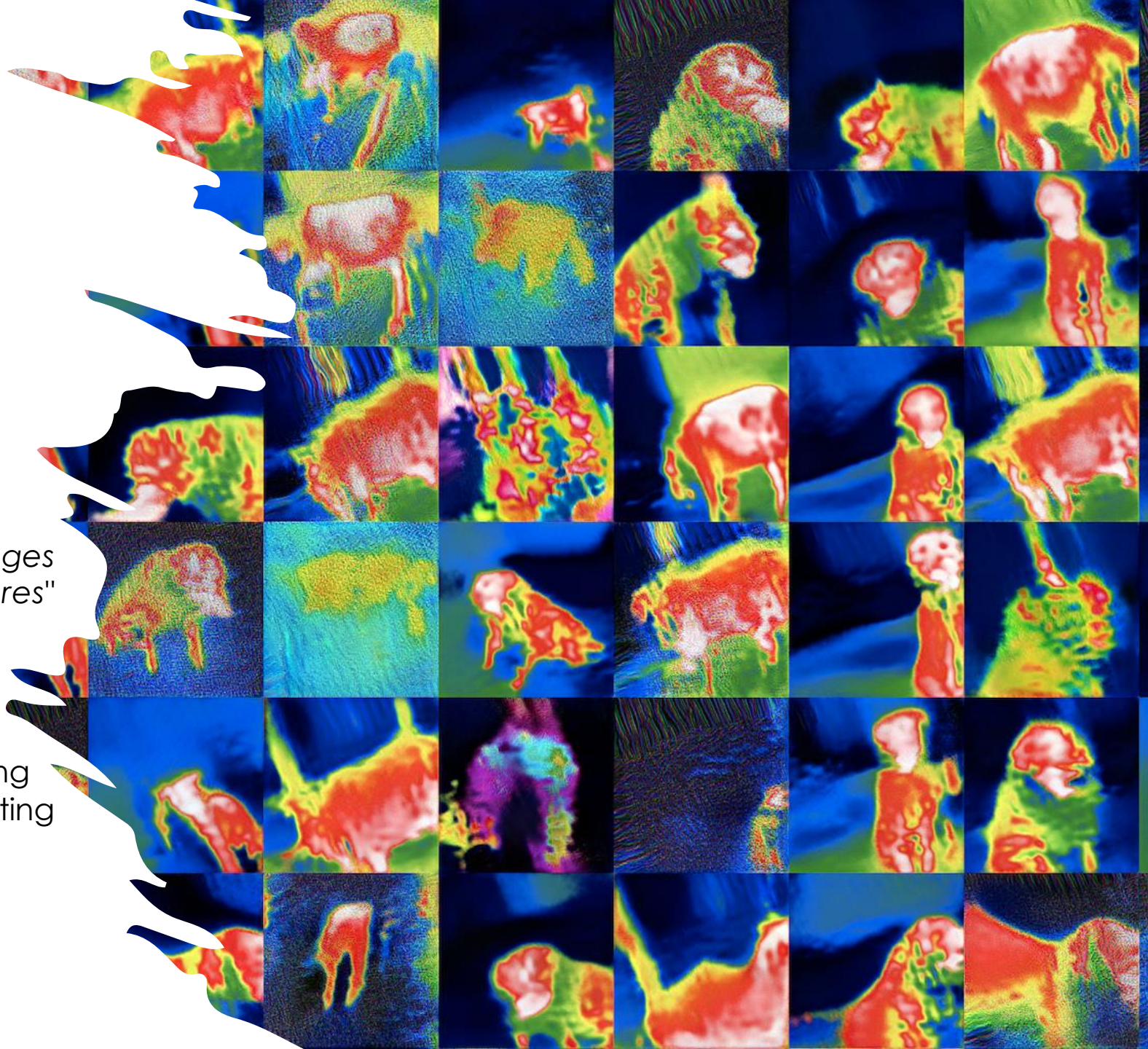
- Parameters:
 - 100 epochs,
 - batch size of 8,
 - Adam optimizers,
 - learning rate 2×10^{-4}
 - Loss fun: binary cross-entropy loss



Implementations

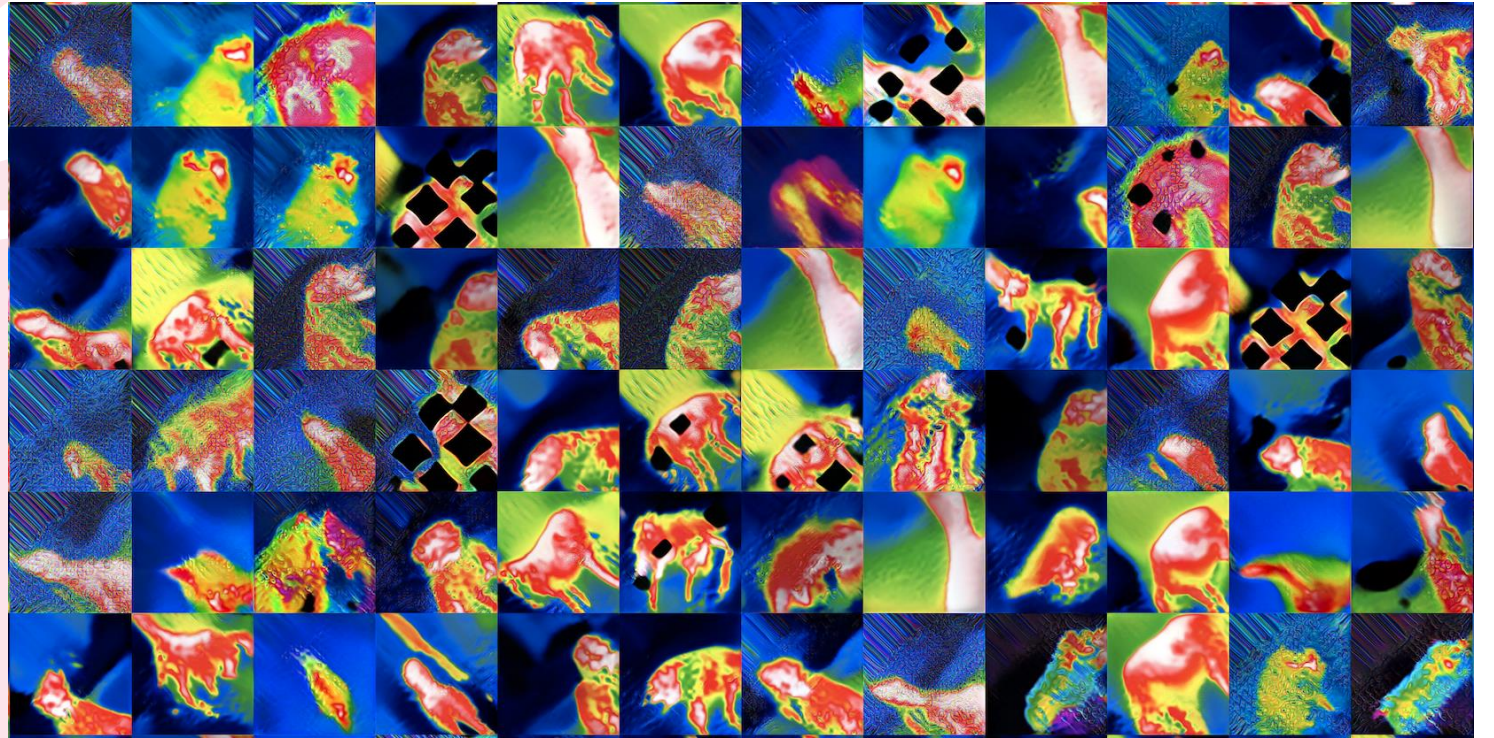
– StyleGan2-Ada

- Parameters:
 - 512-dimensional latent space,
 - 1 000 000 images
 - batch size of 16,
 - Adam optimizer
- *"The approach does not require changes to loss functions or network architectures"*
 - non-saturating logistic loss with R1 regularization
- dynamically adjust augmentation strength on the discriminator, improving training stability and mitigating overfitting on limited dataset.



StyleGan2-Ada on more aggressive augmentation

- larger and more frequent square occlusions,
- increased rotation angles,
- greater scaling variations



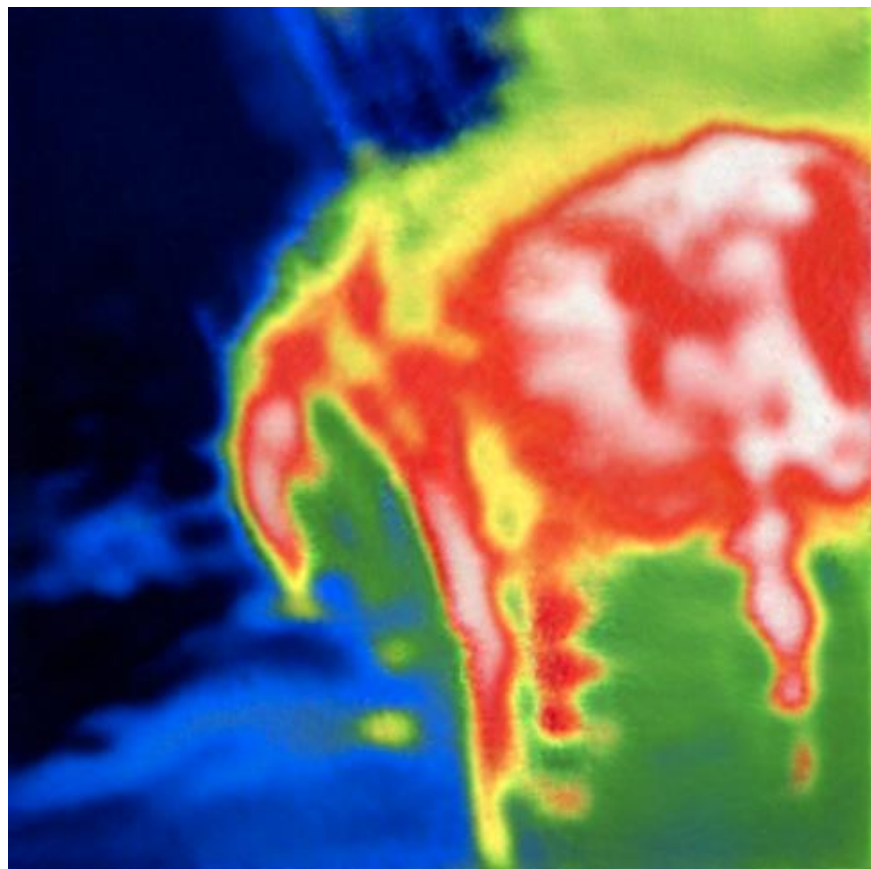
StyleGan2-Ada on RGB



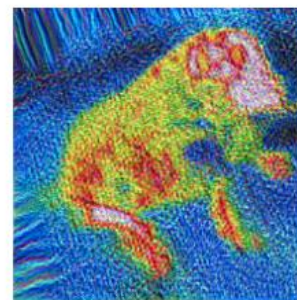
Style- mixing using StyleGan2- Ada models

Variant	Description	Mixing ratio control	Synthesizer	Outputs
Single Seed	One shared latent vector (z)	Blending done by proportional layer mixing (e.g., 50% RGB, 50% Thermal)	Thermal Generator	1 Mixed Image
Independent Seeds	Two random latent seeds	Early layers from RGB, late from Thermal using hard cutoff (after layer 6).	Thermal Generator	RGB, Thermal, Mixed
Multi-Seed Shared	Same seed used in both models	Early structure layers from RGB, later ones from Thermal (after layer 4)	Thermal Generator	RGB, Thermal, Mixed

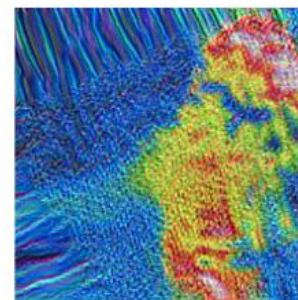
Results of style-mixing



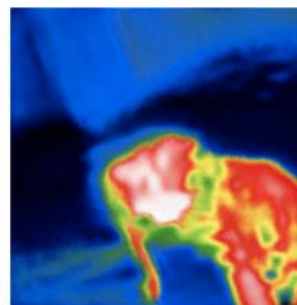
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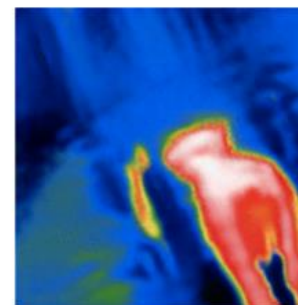
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Comparison of the best models – FID score

Model	FID Score
CycleGAN	350.37
StyleGAN	300
StyleGAN2-ADA	101.37
StyleGAN2-ADA (RGB photos only)	8.12



*Thank you very
much!*

WE WILL BE WORKING FOR
BETTER THERMLA CATS...