



# GANimals

## CS 455 Final Project

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# Overview of Challenge

- ❖ There was a need for mass creation of **monstrous animal faces**.
- ❖ Doing this traditionally, either by **hand-drawing** or **photoshop**, proved to be too **time-consuming**.
- ❖ Thus, **GANimals** was born.





# Limitations

## ❖ Hardware

- Most of our computers were **not strong enough** to handle generating the images.

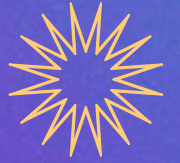
## ❖ Dataset

- Kaggle dataset is limited to **mammal faces only**, but “GANmals” wasn’t as catchy.





# Approach



- ❖ The images were generated using GANs, otherwise known as **Generative Adversarial Networks**.
- ❖ There have been many other applications in art\* that also utilize GANs.


\*Ours is not this good.


**Enter prompt** 14/100


dragon samurai

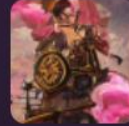
Sunset cliffs Never ending flower Fire and water DNA tornado


**Art Style**


  
Psychedelic


  
Synthwave


  
Ghibli

  
Steampunk

  
Fantasy Art

  
Vibrant

  
HD

  
Psychic

Create



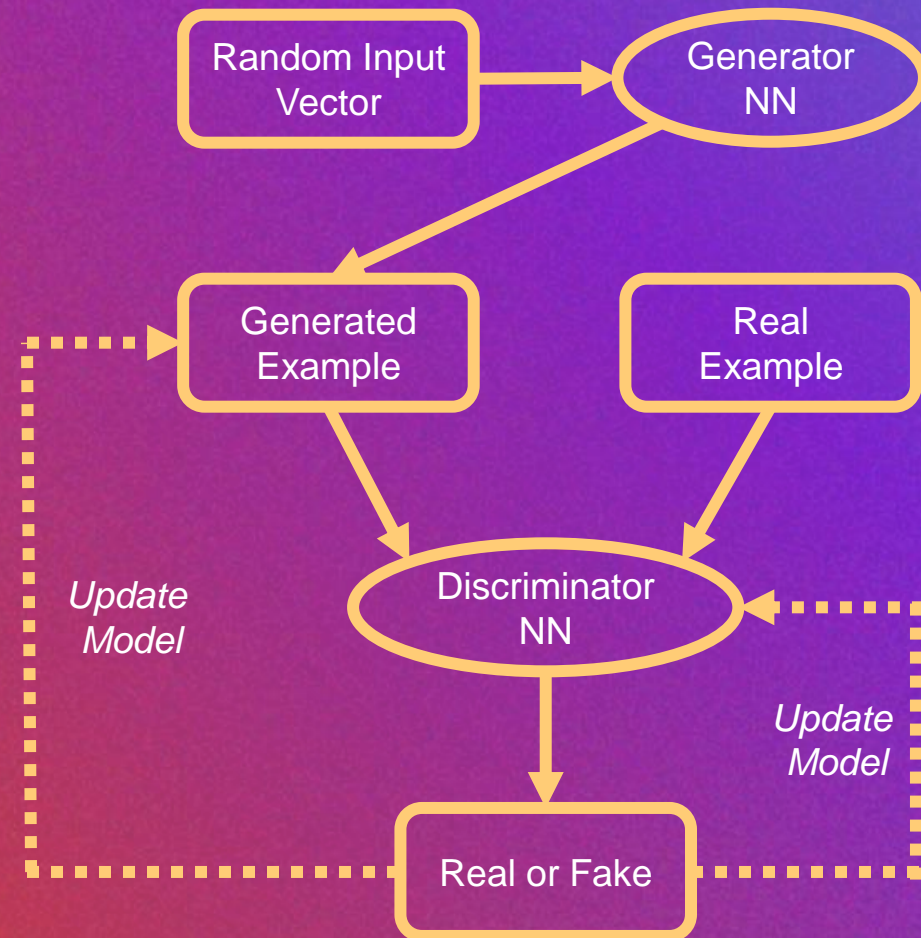


# Algorithm

## ❖ Two neural networks:

- **Generator** – creates fake sample
- **Discriminator** – attempts to discern whether or not it is “real”

## ❖ They are trained to compete against each other.





# Key Resources

- ❖ Kaggle Animal Faces Dataset which includes 16,130 high-quality images at 512×512 resolution.
- ❖ The images were split into three categories:
  - Cats
  - Dogs
  - Wildlife

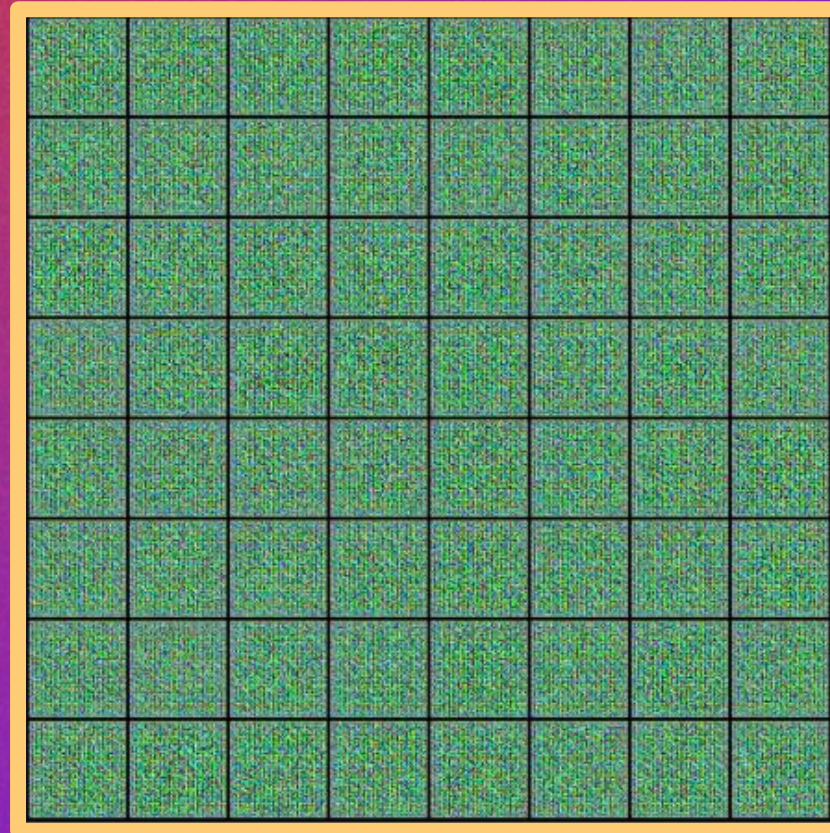




# Solution

## ❖ GENERATOR MODEL

- Input is a **matrix of random numbers** (Latent Tensor)
- Used as a seed for generating “fake” images
- Convert Latent Tensor into an image using **Transposed Convolution** (a.k.a. Deconvolution) which is shown to the right

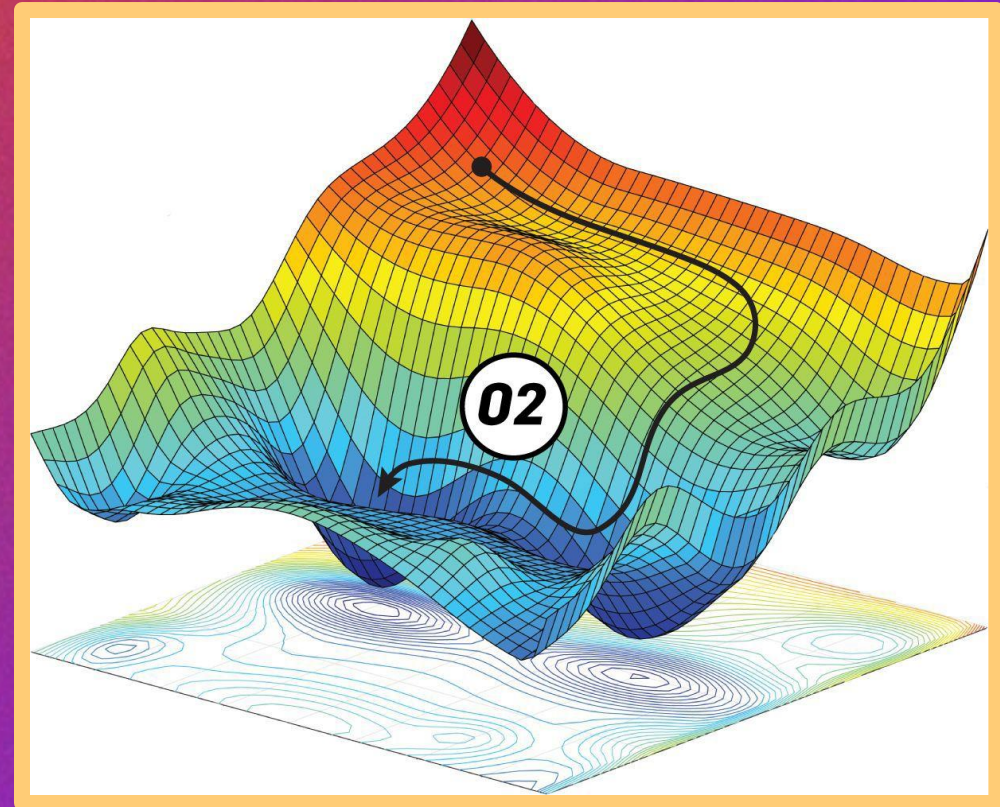




# Solution

## ❖ GENERATOR TRAINING

- 1) Generate a batch of images and then pass it to the **discriminator**.
- 2) Calculate the **loss** by setting the discriminator's **target labels to 1**. (We're trying to trick the discriminator.)
- 3) Use resulting loss to perform **Gradient Descent** to update the weights of the generator.

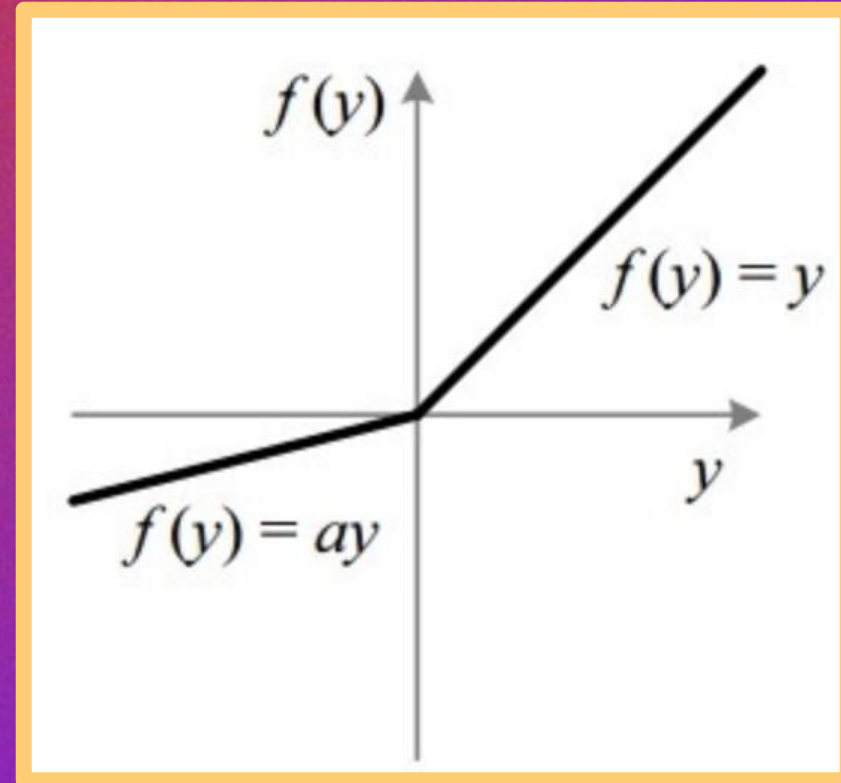




# Solution

## ❖ DISCRIMINATOR MODEL

- Convolutional Neural Network
- Outputs a single number per image between 0 and 1
- Stride = 2
- Input = 64 x 64 (Dataset was scaled down)
- Activation Function: Leaky Rectified Linear Unit (Leaky ReLU) shown to right





# Solution

## ❖ DISCRIMINATOR TRAINING

- Loss Function = Binary Cross-Entropy
- 1) Pass a batch of real images, compute loss, and then set target labels to 1.
  - 2) Pass a batch of generated images, compute loss, and then set target labels to 0.
  - 3) Add the two losses and use overall loss to perform Gradient Descent to update the weights of the discriminator.

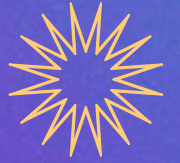
$$H_p(q) = -\frac{1}{N} \sum_{i=1}^N y_i \cdot \log(p(y_i)) + (1 - y_i) \cdot \log(1 - p(y_i))$$

Binary Cross-Entropy / Log Loss





# Demonstration



- ❖ Due to limitations of hardware, a live demo was **not viable**.
- ❖ However, we were able to generate up to **50 epochs** before **danger** was imminent.
- ❖ Please **prepare yourselves** for the following images.

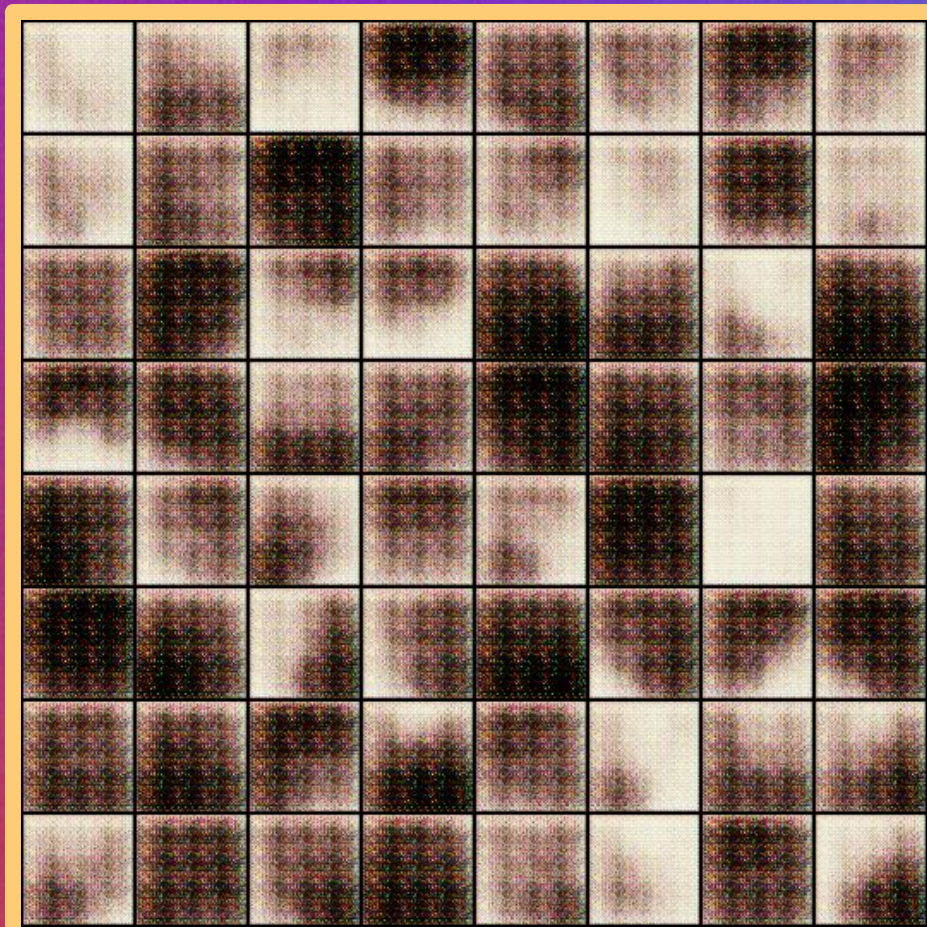




# Demonstration



- ❖ This is the first epoch that we generated.
- ❖ As you can see, there's not much to discern yet.





# Demonstration



- ❖ At 25 epochs, we're really seeing some movement.





# Demonstration

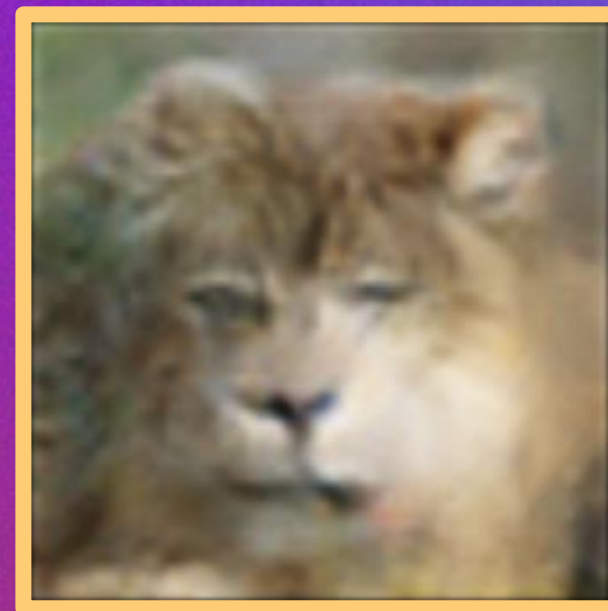


- ❖ 50 epochs in, and there isn't really a noticeable difference between 25 epochs and this.



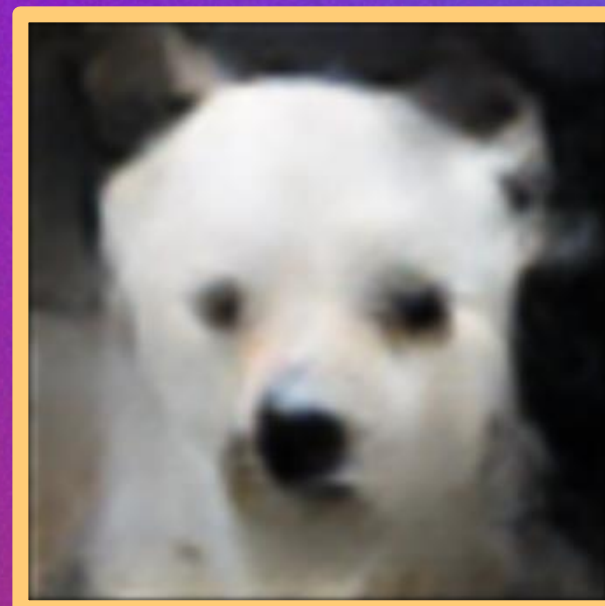


# Demonstration



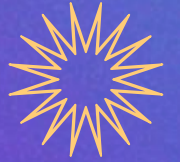


# Demonstration





# Conclusion



- ❖ Overall, it seems like our call for the creation of monstrous animal faces was answered.
- ❖ Operation GANimals was a success.





# Thank You!

Any Questions?

