

# Drone Image generation using GAN

(Generative Adversarial Networks)

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# What is GAN?

- Generative Adversarial Network
- Generates images using NN
- Ian Goodfellow, et al, (2014)



## Variety of GANs: CycleGAN (2017)



zebra → horse



horse → zebra

## Variety of GANs: DiscoGAN (2017)



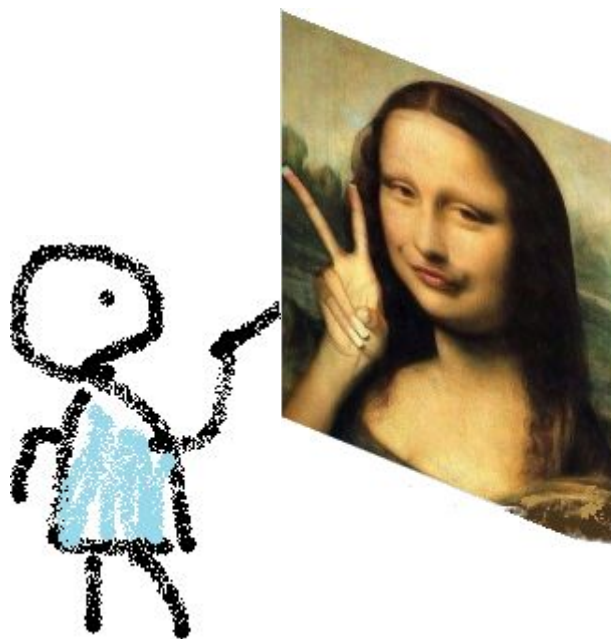
(b) Handbag images (input) & **Generated** shoe images (output)

## Variety of GANs: PGGAN (2018), StyleGAN (2019)





# How does GAN work?



Forger  
(Generator)

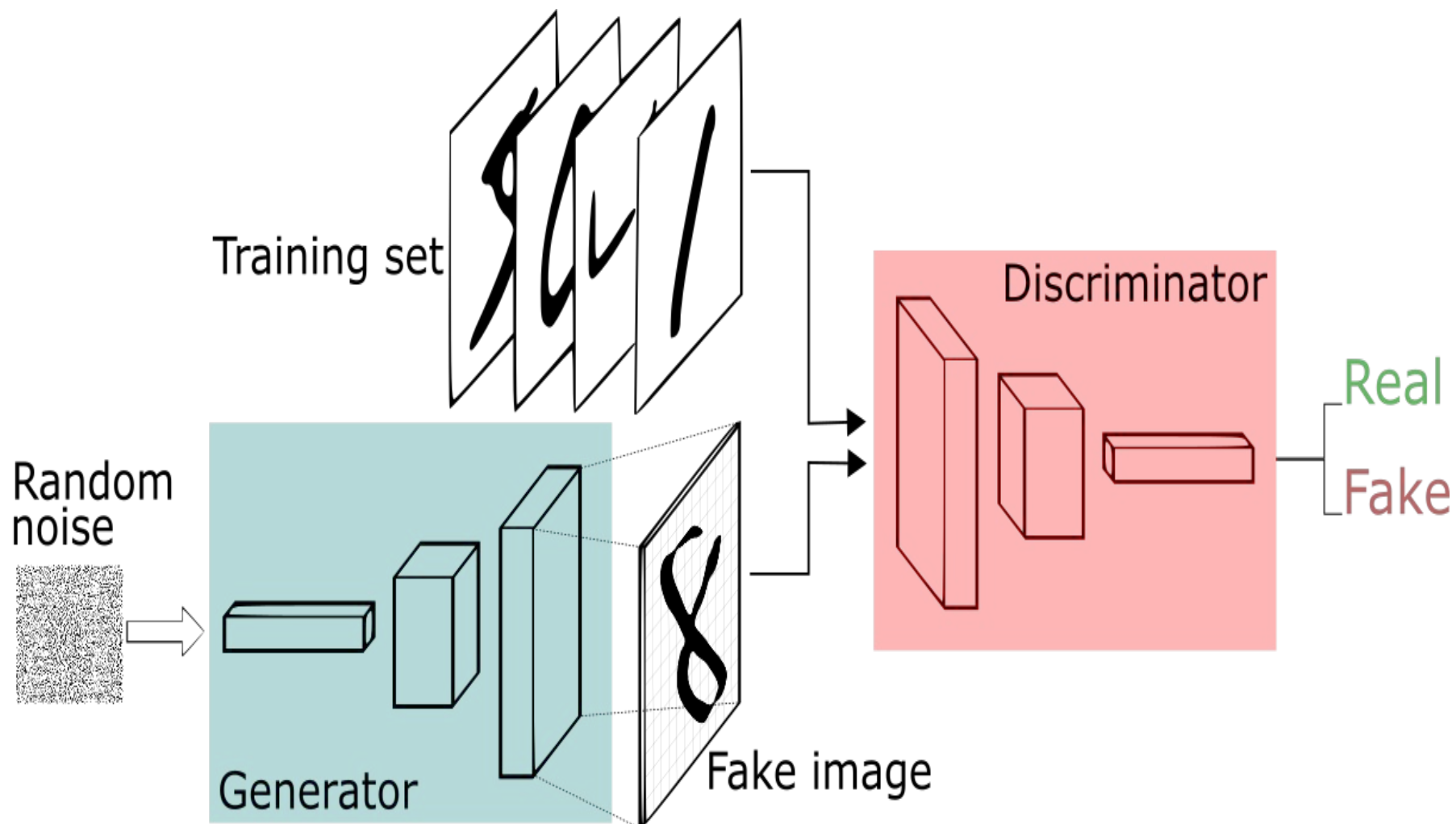


Real? Fake?



Detective  
(Discriminator)

# GAN Implementation



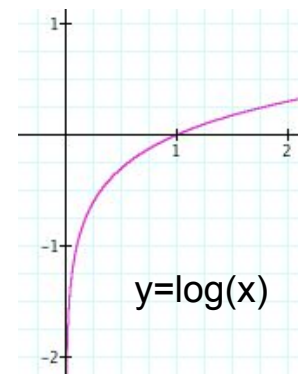
# GAN Loss function

$$\min_G \max_D V(D, G)$$

$$V(D, G) = \mathbb{E}_{x \sim p_{data}(x)} [\log D(x)] + \mathbb{E}_{z \sim p_z(z)} [\log(1 - D(G(z)))]$$

$D(x)$  = probability that Discriminator judges  $x$  is real (0..1)

$G(z)$  = image generated by Generator.



D gets better:  $\rightarrow D(x) \nearrow, D(G(z)) \searrow \rightarrow V \nearrow$



G gets better:  $\rightarrow D(G(z)) \nearrow \rightarrow V \searrow$



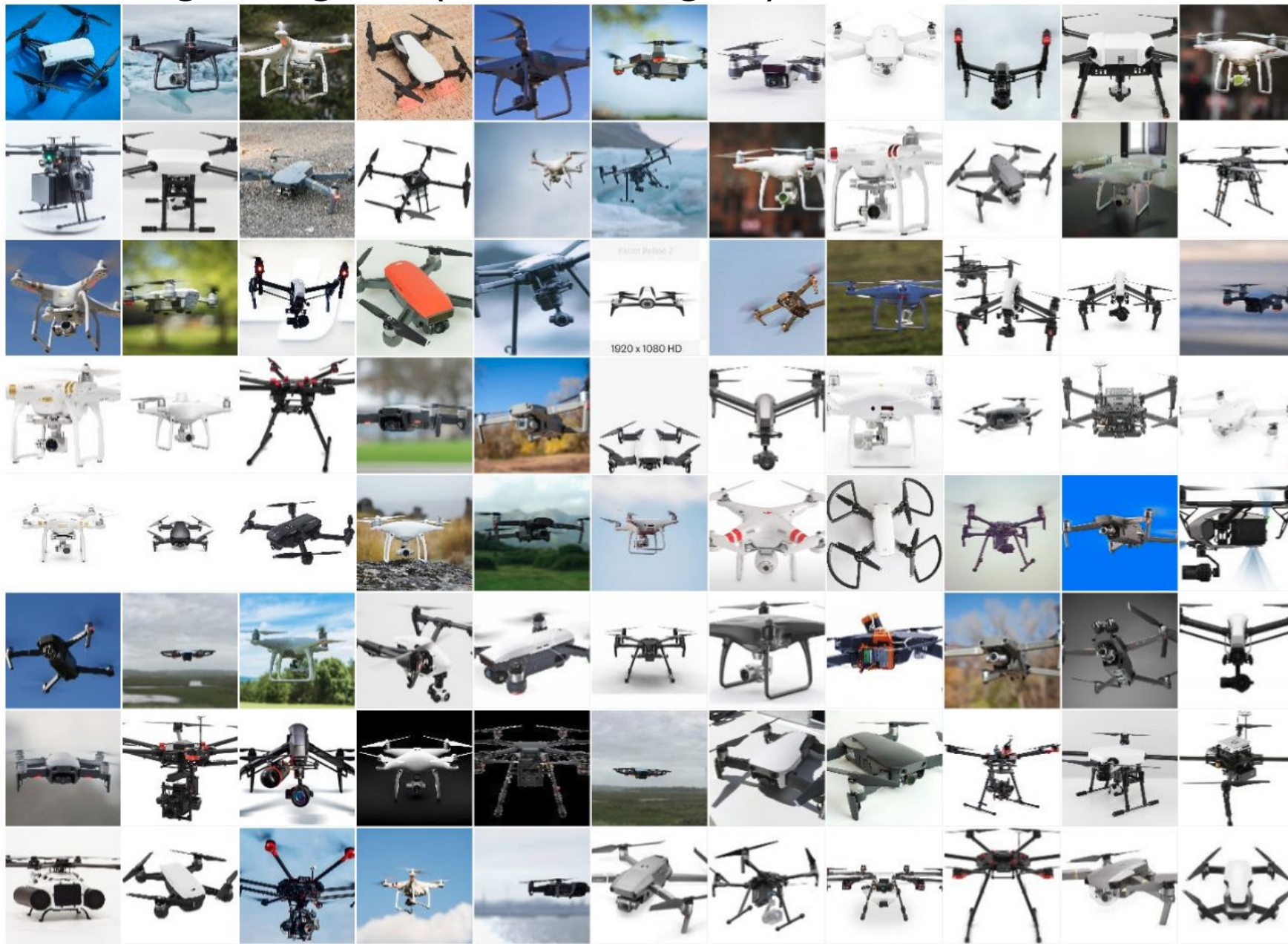
# My Project -- generates drone images with GAN

- Our company have a product to classify if an object in Video is a drone or not.
- It is based on a simple NN classifier (using Caffe)
- It works relatively well considering the limited number of training images.
- Having more images will improve the accuracy.

## Implementation

- Python Keras + TensorFlow
- DCGAN : Deep Convolutional GAN -- Using Convolution layer

# Training images (~800 images)



Training...

(another slide)