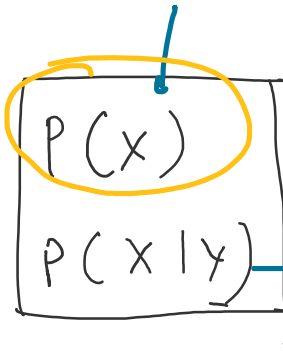


Generative Adversarial Network

01 July 2025 13:19

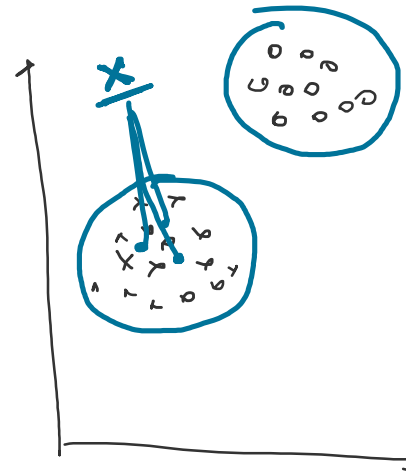
→ 2014

Ian Goodfellow



to classify

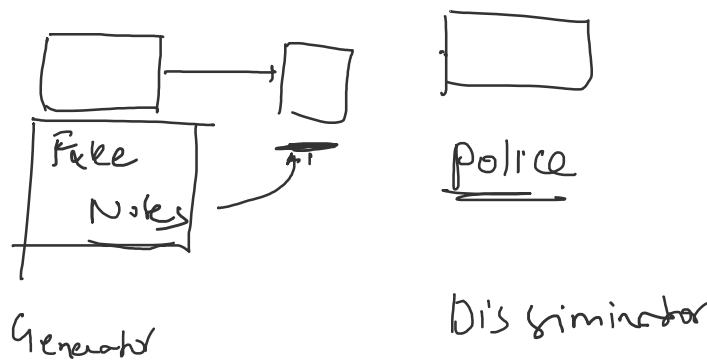
learn distribution of



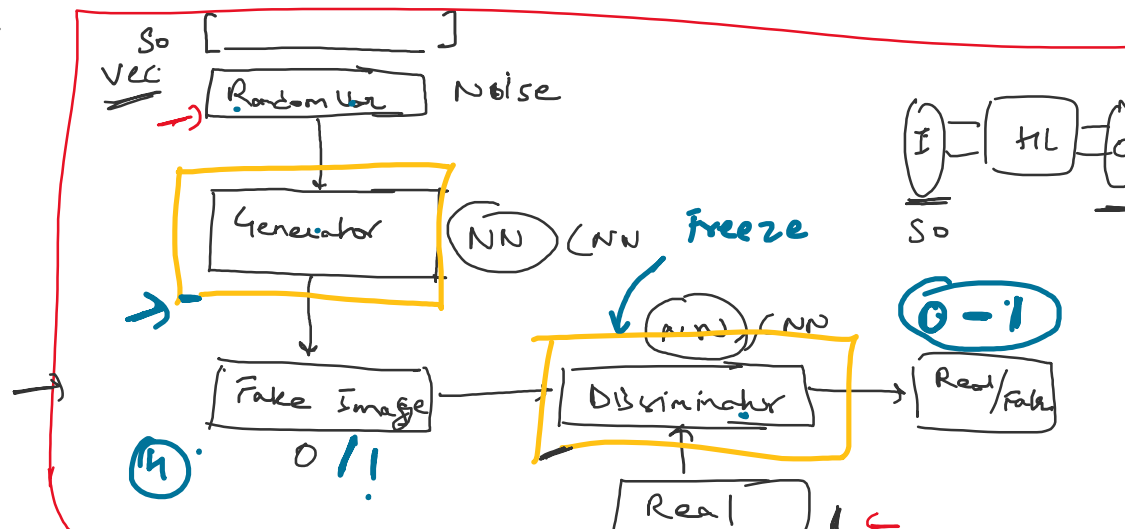
GAN → Generative Adversarial Networks

Chattgpt → GPT

- ① Generator → Generates a Fake
- ② Discriminator → Helps to discriminate b/w Fake & Real



Model



GAN

Training of GANs

- ① Random var : $\mathbb{R}^n \rightarrow I_0 \quad z \sim P(z)$
- ② Generate Fake Image using Generator by giving it R.V.
- ③ Train Your Discriminator by using Fake & Real image

0
1
- ④ Train Your Generator by
 - ① Freeze Your Discriminator
 - ② Label Your Fake Images as 1 and train Your GAN
- ⑤ Repeat all the steps for some epoch

Objfn

$$\min_{\theta_g} \max_{\theta_d} \left[E_{x \sim P_{data}} \log D_{\theta_d}(x) + E_{z \sim P(z)} \log (1 - D_{\theta_d}(G_{\theta_g}(z))) \right]$$

DCGANs → Deep Convolutional GAN

Discriminator → Convolutional Neural Network

Generator → CNN Convolution Transpose

Down Sampling : **Convolution**

Upsampling : Convolution Transpose

Up Sampling

Max Pooling

1	2
3	4

1	1	2	2
1	1	2	2
3	3	4	4
3	3	4	4

Cifar 32x32 → 224x224

Transfer learning

Convolution Transpose

1	2
0	3

Input
Image

*

1	3
2	1

kernel

=

1	3
2	1

2	6
4	2

0	0
0	0

3	9
6	3

	1	0	2
	0	3	1
	2	0	1

img

*

0	2
1	1

kernel

stride=1			
	0	2	0
	1	1	8
	0	7	4
	2	2	1

Stride=2

0	2	0	0	0
1	1	0	0	2
0	0			

6x6

