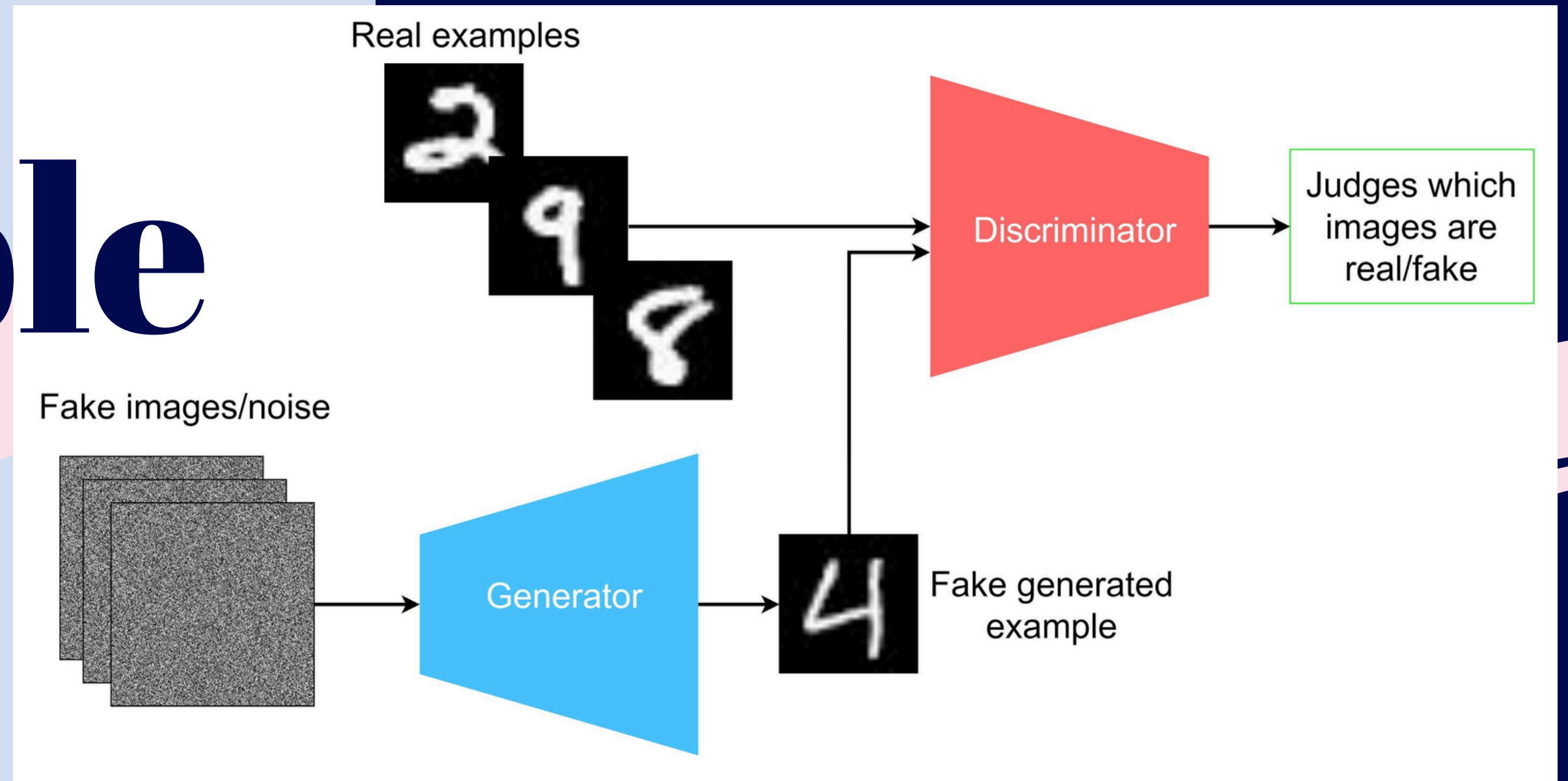


# GAN models

*DL 2025 LAB 10*

# example



generator = noise → image  
discriminator = image → label

# objective

## GENERATOR

Minimize probability of discriminator labeling it correctly

## DISCRIMINATOR

Maximize the probability that it labels the image correctly

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

$D(\mathbf{x})$  is probability that  $\mathbf{x}$  came from the data.



# **min**

$$\log(1 - D(G(z)))$$

generator objective



## **THEORETICALLY**

This will work well because from the generator's perspective we want the discriminator to perform well on our generations

## **REAL**

It will start really saturated, meaning initial steps will be slow, and making learning slower


$$\max$$
$$\log D(G(z))$$

generator objective



## THEORETICALLY

This will work well because from the generator's perspective we want the discriminator to perform well on our generations

## REAL

It will achieve the same as the last equation, but with faster initial steps

**max**

$$\log D(G(z))$$

generator objective

**max**

$$\log D(x)$$

$$\log (1 - D(G(z)))$$

discriminator objective