### **Generative Adversarial Networks**

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## What is a Generative Model?

#### Generative

- Produces samples from a distribution.
- Input:
  - source of randomness,
  - class label,
  - latent variable, etc.
- Output: a sample from the target distribution.

#### **Discriminative**

- Given a sample, determines if it comes from a distribution
- Input: a sample
- Output:
  - True/False,
  - a class label, etc.

## Training a Generative Model

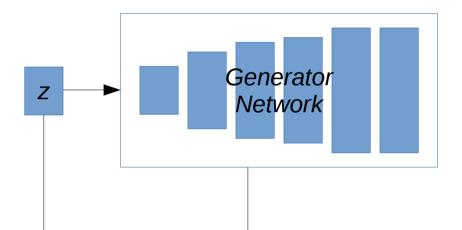
#### Generative

- Data
  - (Image, Label)
  - (Image, Label)
  - **–** ...
- Label -> Image
  - Not a function!
  - It is a distribution.
  - One-to-many relationship.

#### **Discriminative**

- Data
  - (Image, Label)
  - (Image, Label)
  - ...
- Image -> Label
  - Non-bijective function.
  - SGD to learn function.

## Make the Generator a Function



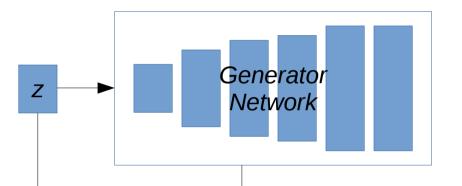
## Source of Randomness

Needed to produce different samples from the distribution.

#### **Generator Network**

Function from random vector z to an image x. We are learning weights so that x appears to be from target distribution Y.

## Score the output



Score the output.

## Source of Randomness

Needed to produce different samples from the distribution.

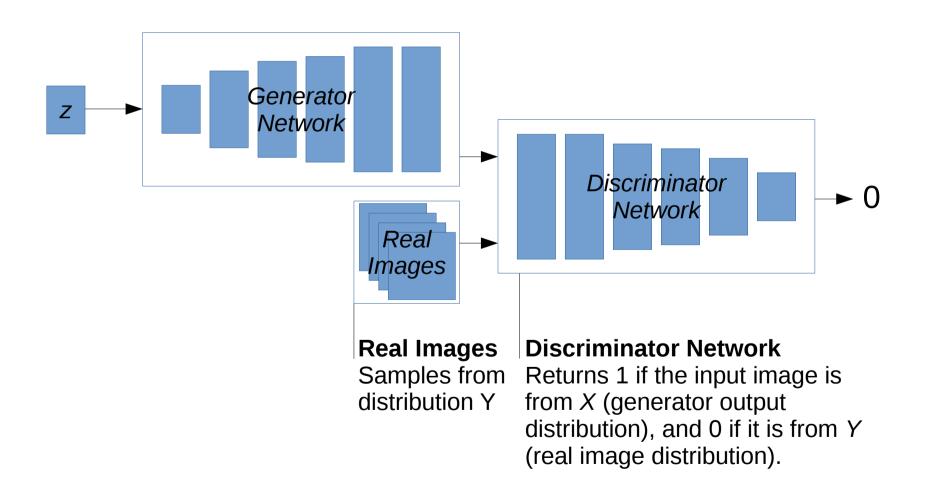
#### **Generator Network**

Function from random vector *z* to an image *x*. We are learning weights so that *x* appears to be from target distribution *Y*.

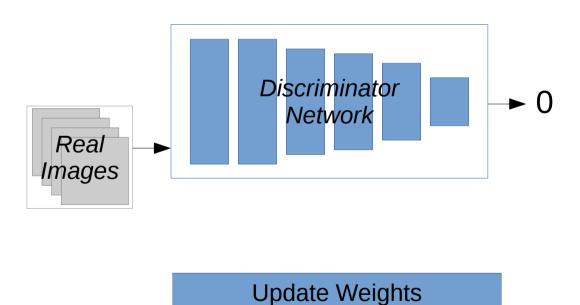
#### **Score the Output**

Given a sample *x* from *X*, we need a differentiable function that can determine how likely *x* is to be from target distribution *Y*. How can we construct such a function?

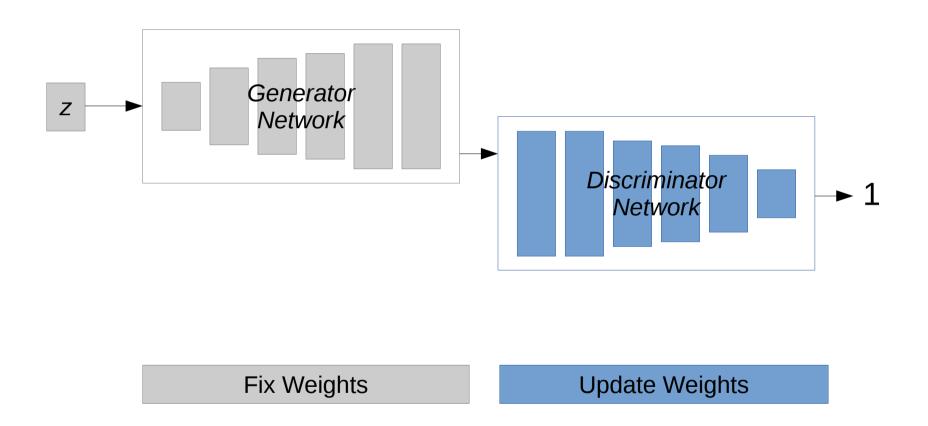
### The Discriminator Network



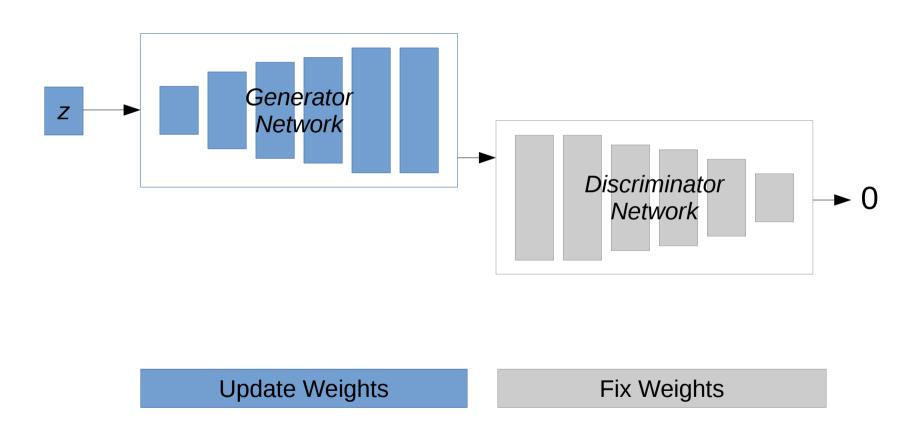
# Training the Network (1/3)



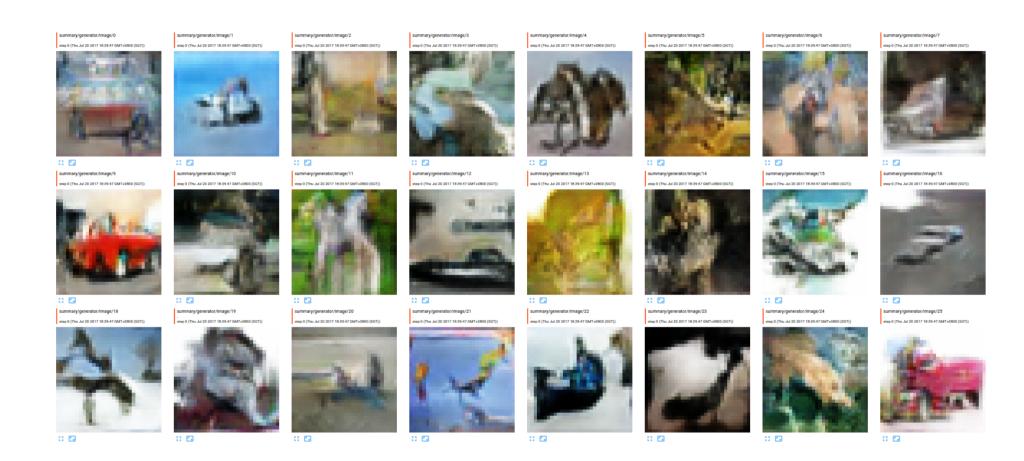
# Training the Network (2/3)



# Training the Network (3/3)



# Output!



# **Comparing Output**

