

# GAN Application

Random Number Generator

[https://youtu.be/3N\\_otcqpFIM](https://youtu.be/3N_otcqpFIM)

# GANs

- Generative Adversarial Networks
- Two Models
  - Generator
  - Discriminator

# GANs Example Code

<https://www.tensorflow.org/tutorials/generative/dcgan?>

- Using MNIST Data

```
@tf.function
def train_step(images):
    noise = tf.random.normal([BATCH_SIZE, noise_dim])

    with tf.GradientTape() as gen_tape, tf.GradientTape() as disc_tape:
        generated_images = generator(noise, training=True)

        real_output = discriminator(images, training=True)
        fake_output = discriminator(generated_images, training=True)

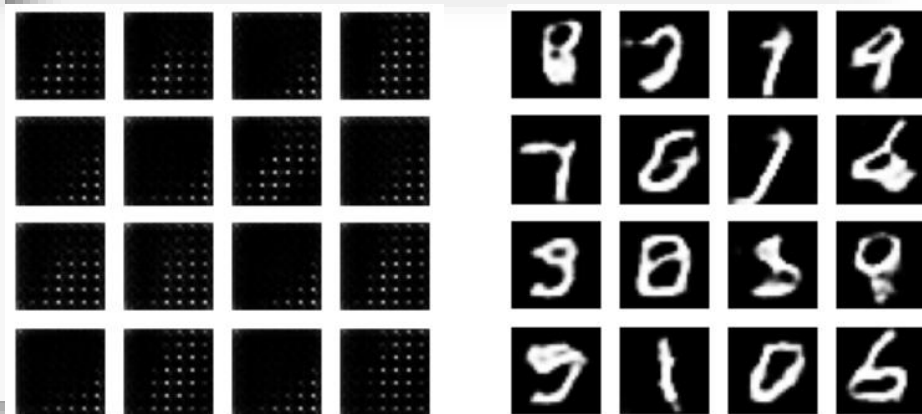
        gen_loss = generator_loss(fake_output)
        disc_loss = discriminator_loss(real_output, fake_output)

    gradients_of_generator = gen_tape.gradient(gen_loss, generator.trainable_variables)
    gradients_of_discriminator = disc_tape.gradient(disc_loss, discriminator.trainable_variables)

    generator_optimizer.apply_gradients(zip(gradients_of_generator, generator.trainable_variables))
    discriminator_optimizer.apply_gradients(zip(gradients_of_discriminator, discriminator.trainable_variables))
```

```
def discriminator_loss(real_output, fake_output):
    real_loss = cross_entropy(tf.ones_like(real_output), real_output)
    fake_loss = cross_entropy(tf.zeros_like(fake_output), fake_output)
    total_loss = real_loss + fake_loss
    return total_loss
```

```
def generator_loss(fake_output):
    return cross_entropy(tf.ones_like(fake_output), fake_output)
```



# RNG

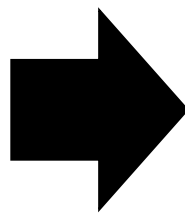
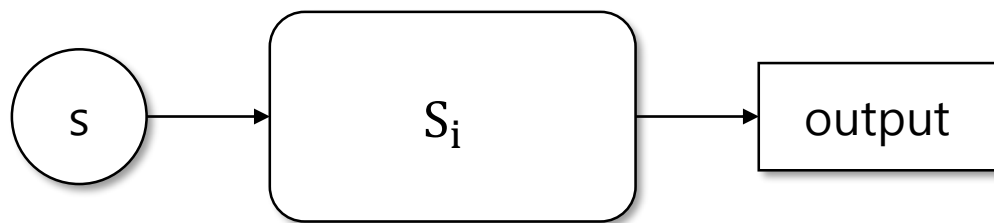
- Random Number Generator
- Two RNGs
  - TRNG(True RNG)
  - PRNG(Pseudo RNG)
    - Seed  $s$

# Design

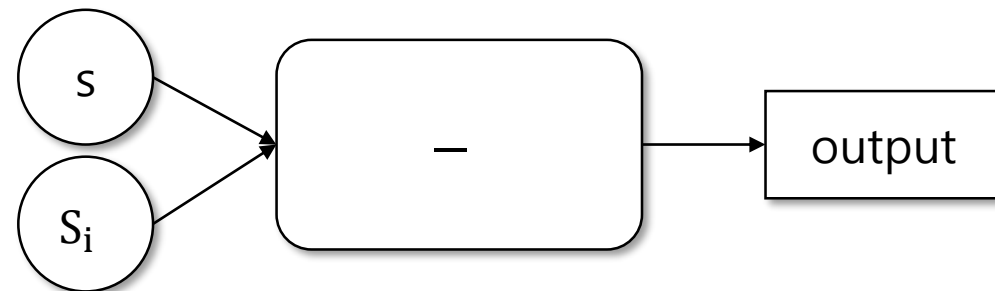
$$\textit{prng}(s) : \mathbb{B} \rightarrow \mathbb{B}^n$$

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$$\textit{prng}^\nabla(s, S_i) : X \rightarrow \mathbb{B}$$

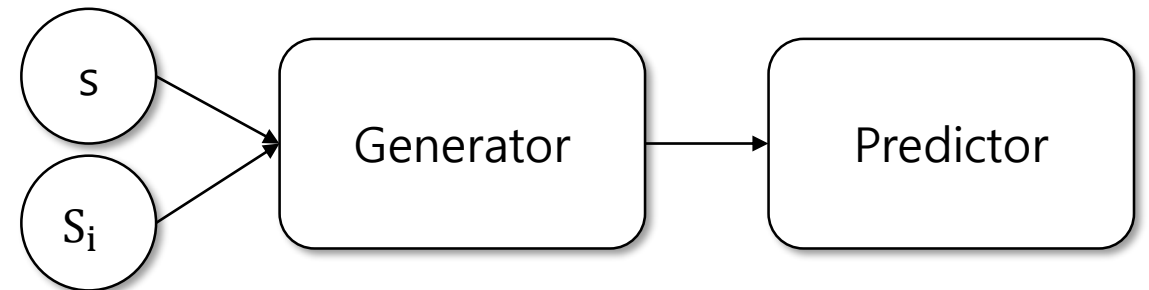
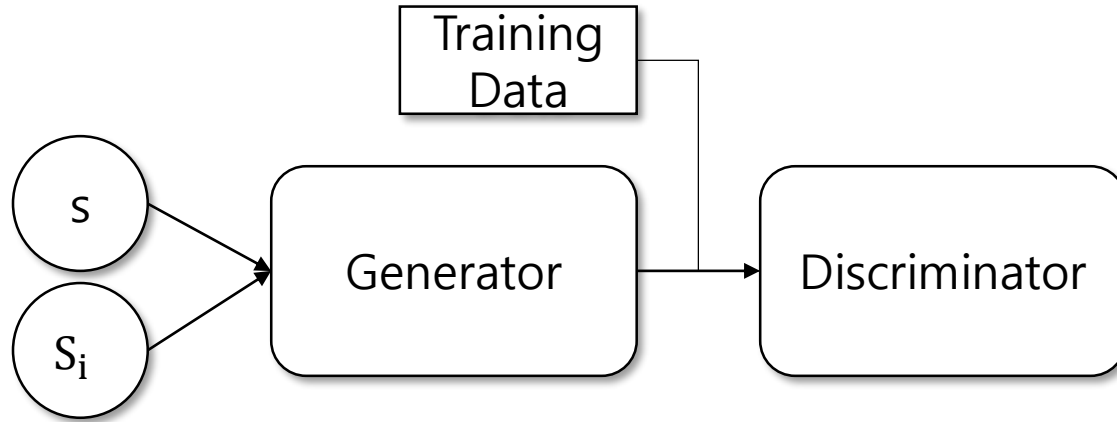


$$G^\nabla(s, \mathbf{o}_t) : \mathbb{B}^{t+1} \rightarrow \mathbb{B}^n$$



# Design

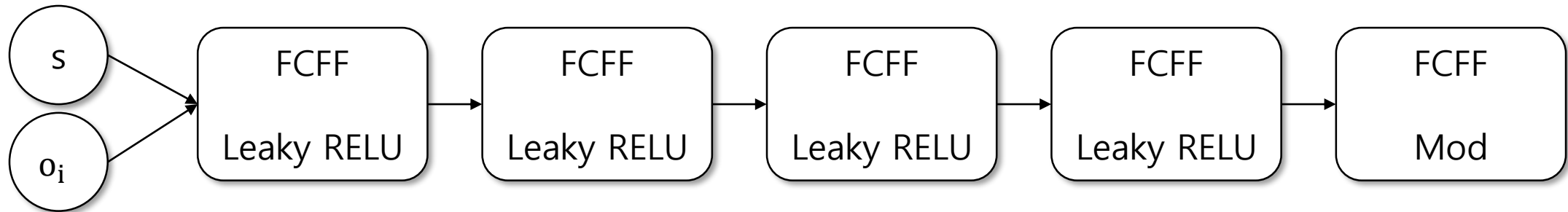
- Two Models
  - Generator
  - Discriminator



# Design G

- Two Models
  - Generator
  - Discriminator

$$G^\nabla(s, o_1) : \mathbb{B}^2 \rightarrow \mathbb{B}^8$$



Loss = absolute difference loss

Optimizer = Adam

Logistic (a.k.a.  
Sigmoid or Soft  
step)



Rectified linear  
unit (ReLU)<sup>[15]</sup>



Leaky rectified  
linear unit (Leaky  
ReLU)<sup>[17]</sup>

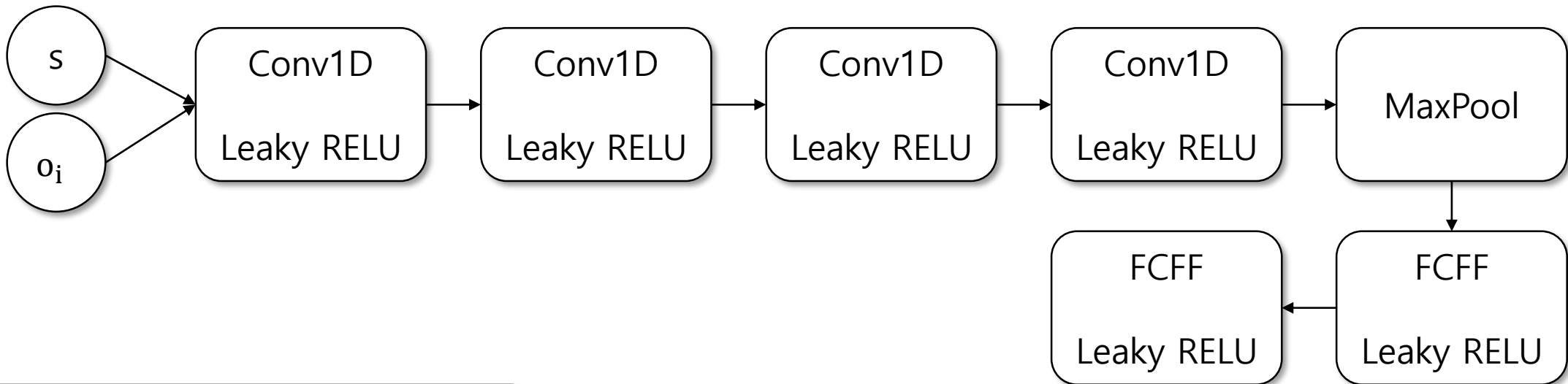


# Design D or P

- Two Models
  - Generator
  - Discriminator

$$D(\mathbf{r}) : \mathbb{B}^8 \rightarrow [0, 1]$$

$$P(\mathbf{r}_{split}) : \mathbb{B}^7 \rightarrow \mathbb{B}$$



Loss = absolute difference loss

Optimizer = Adam

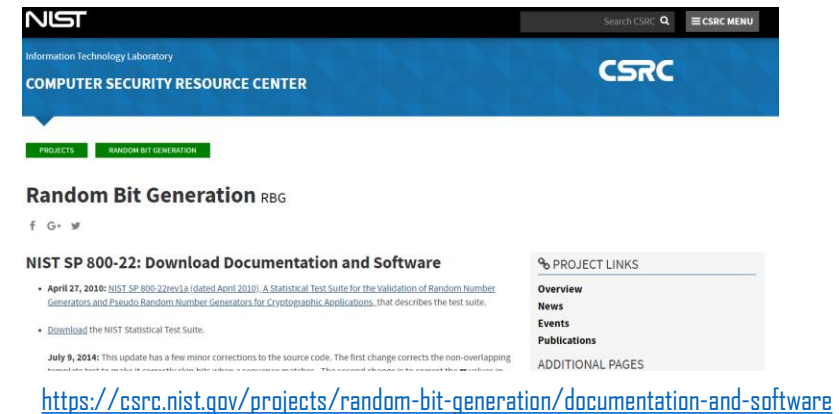


# Experimental Procedure

1. Generate data with untrained Generator → **Data 1**
2. Generate data with trained Generator → **Data 2**
3. Compare **Data 1** and **Data 2** using NIST test suite

## NIST test suite

188 tests,  
Each repeated 10 times  
1,000,000 input bits for each repetition



# Evaluation

$i$	$T$	$\langle T_I \rangle$	$\langle F_I \rangle$	$\langle F_{I\%} \rangle / \%$	$\langle F_p \rangle$	$\langle F_T \rangle$	$\langle F_{\%} \rangle / \%$
$D_{before}$	188	1800	1796	99.8	188	188	100.0
$D_{after}$	188	1800	61	3.5	4.3	6.9	3.9
$P_{before}$	188	1800	1798	99.9	188	188	100.0
$P_{after}$	188	1830	56	3.0	2.7	4.5	2.5

1. Best pass rate was around 98%.
2. Non-cryptographic PRNGs.