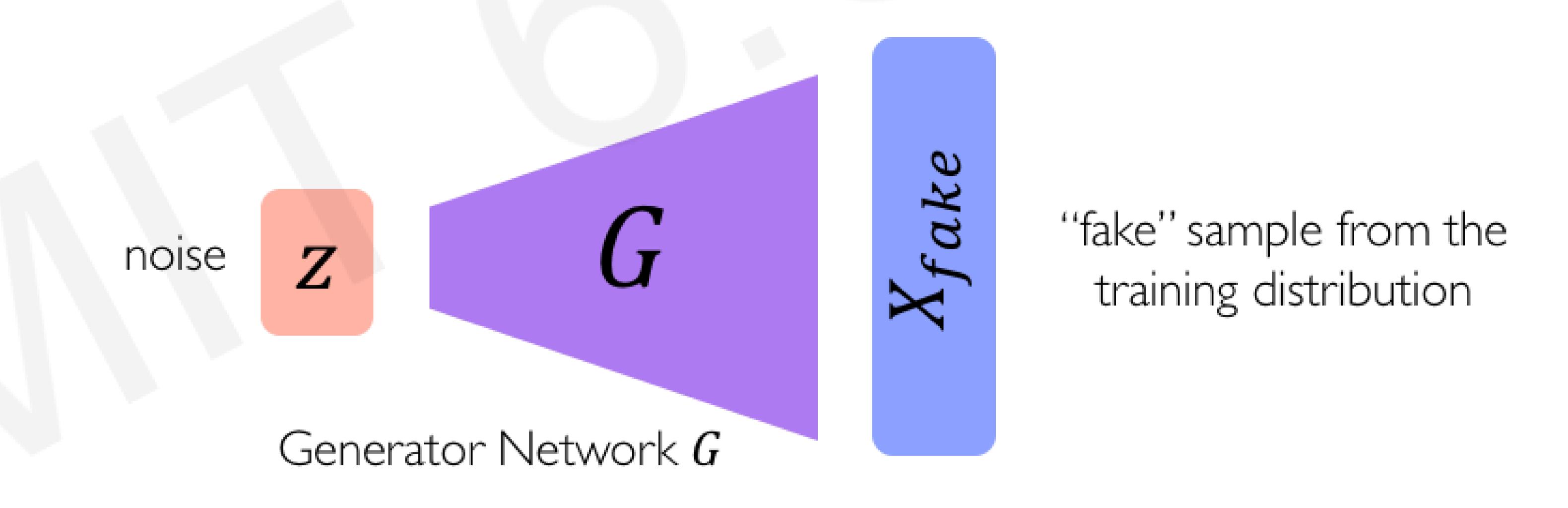
Generative Adversarial Networks (GANs)

What if we just want to sample?

Idea: don't explicitly model density, and instead just sample to generate new instances.

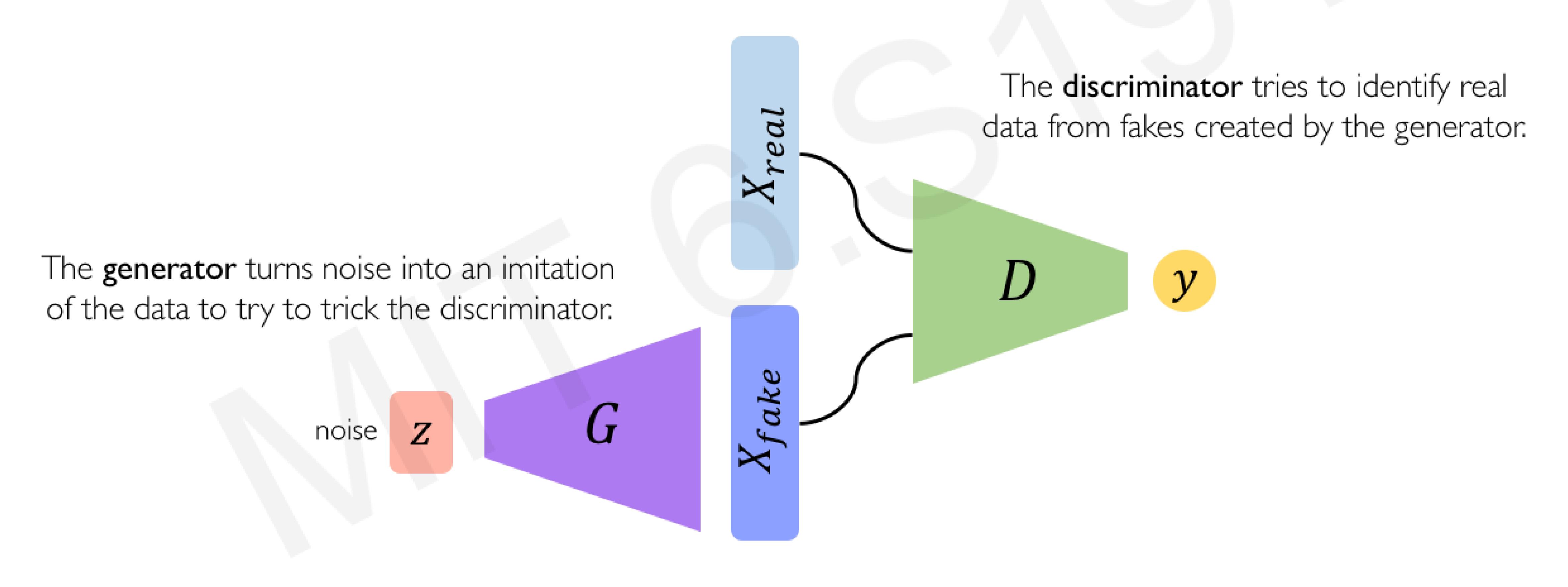
Problem: want to sample from complex distribution – can't do this directly!

Solution: sample from something simple (noise), learn a transformation to the training distribution.



Generative Adversarial Networks (GANs)

Generative Adversarial Networks (GANs) are a way to make a generative model by having two neural networks compete with each other.



Generator starts from noise to try to create an imitation of the data.

Generator

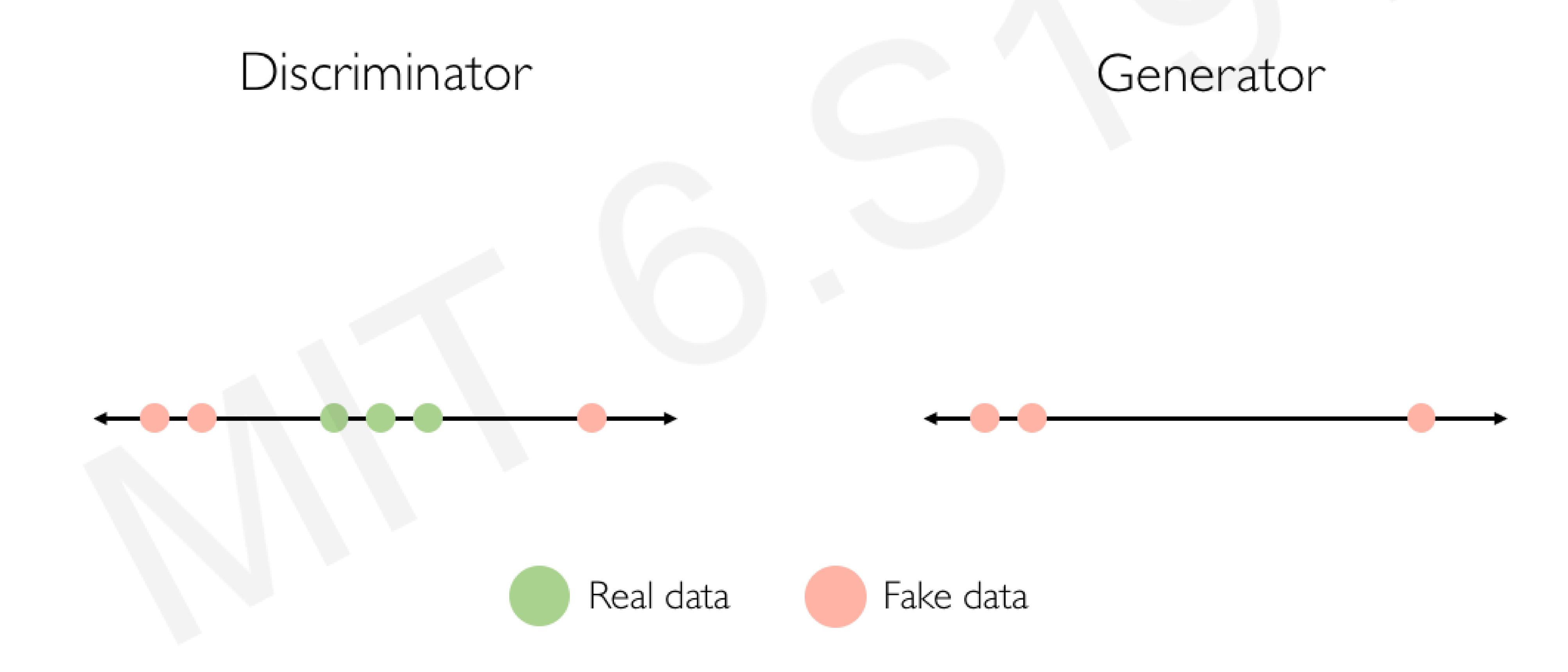


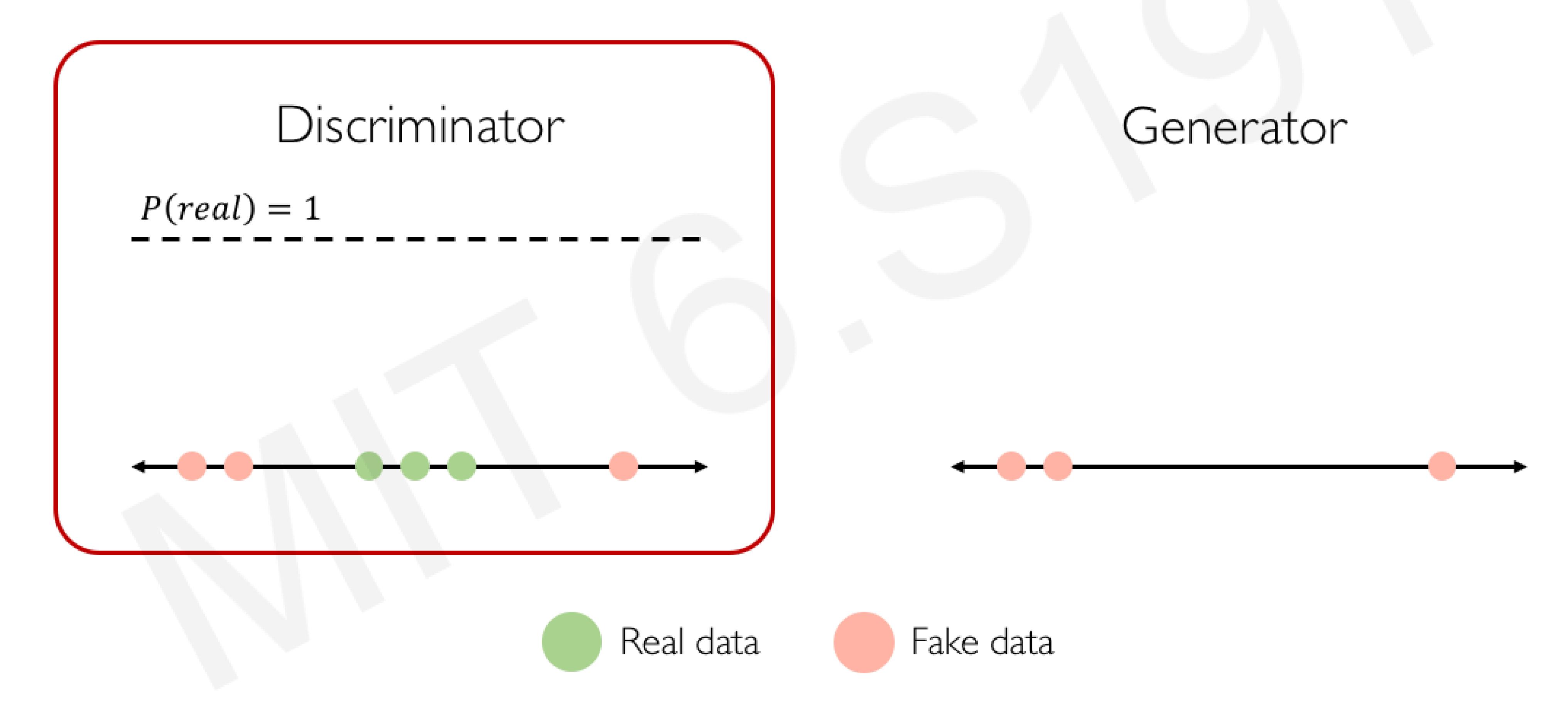


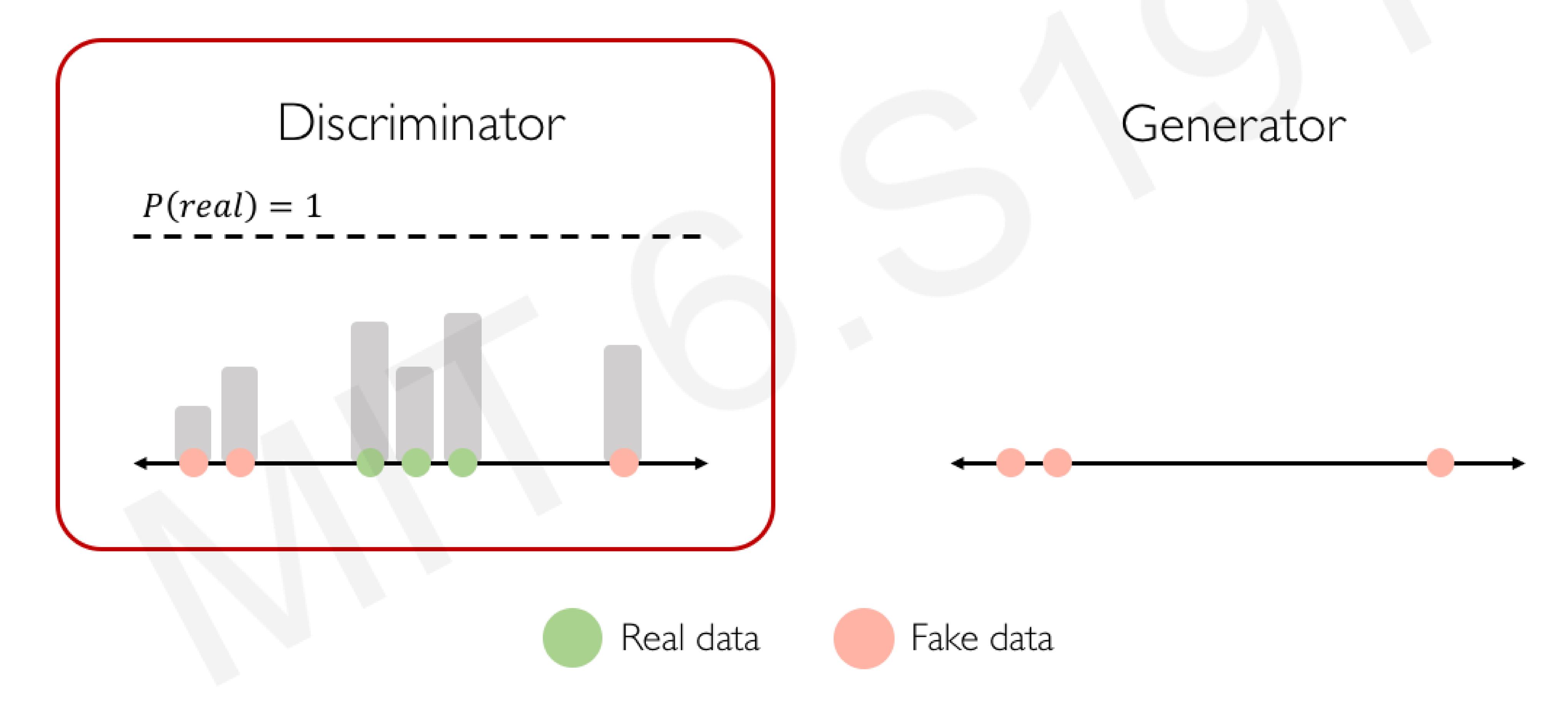
Discriminator looks at both real data and fake data created by the generator.

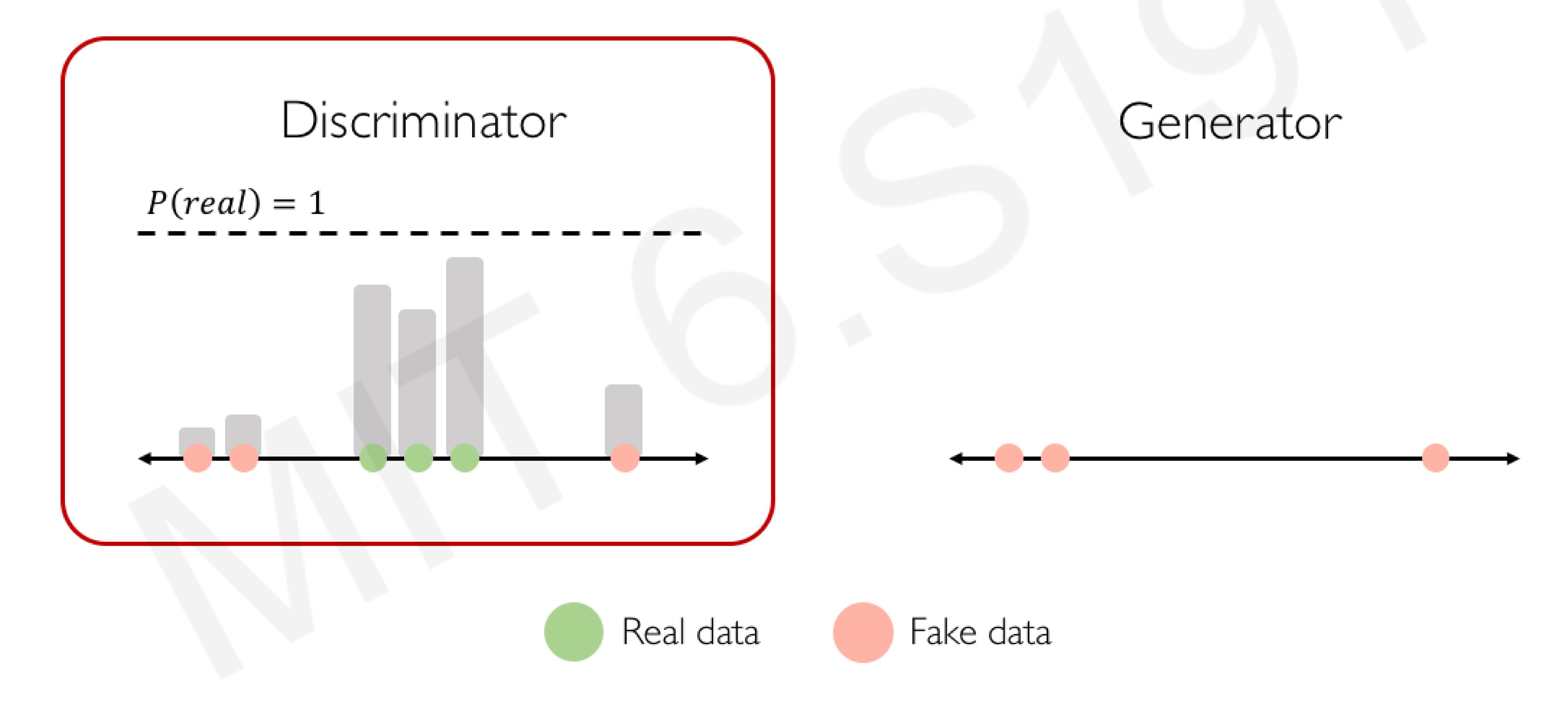
Discriminator Generator Fake data

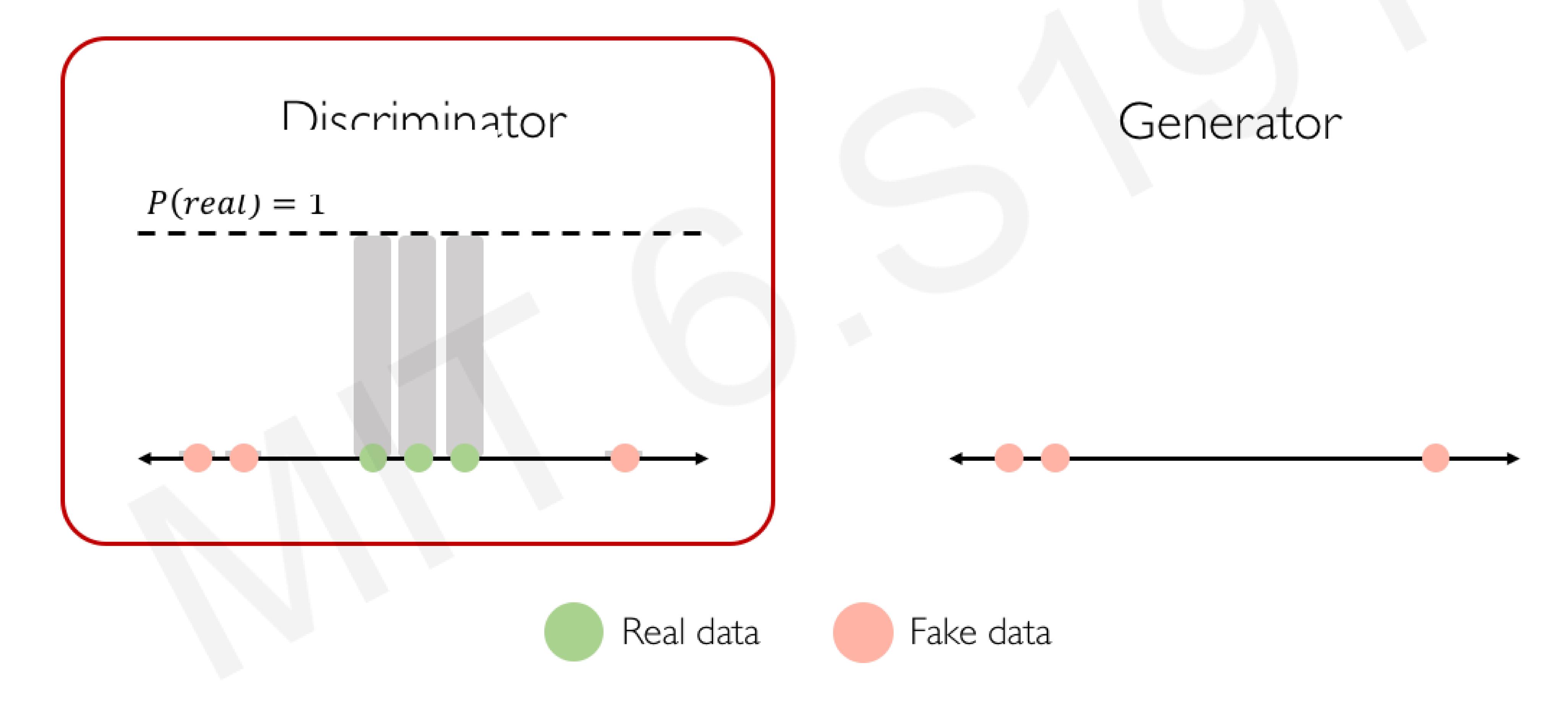
Discriminator looks at both real data and fake data created by the generator.

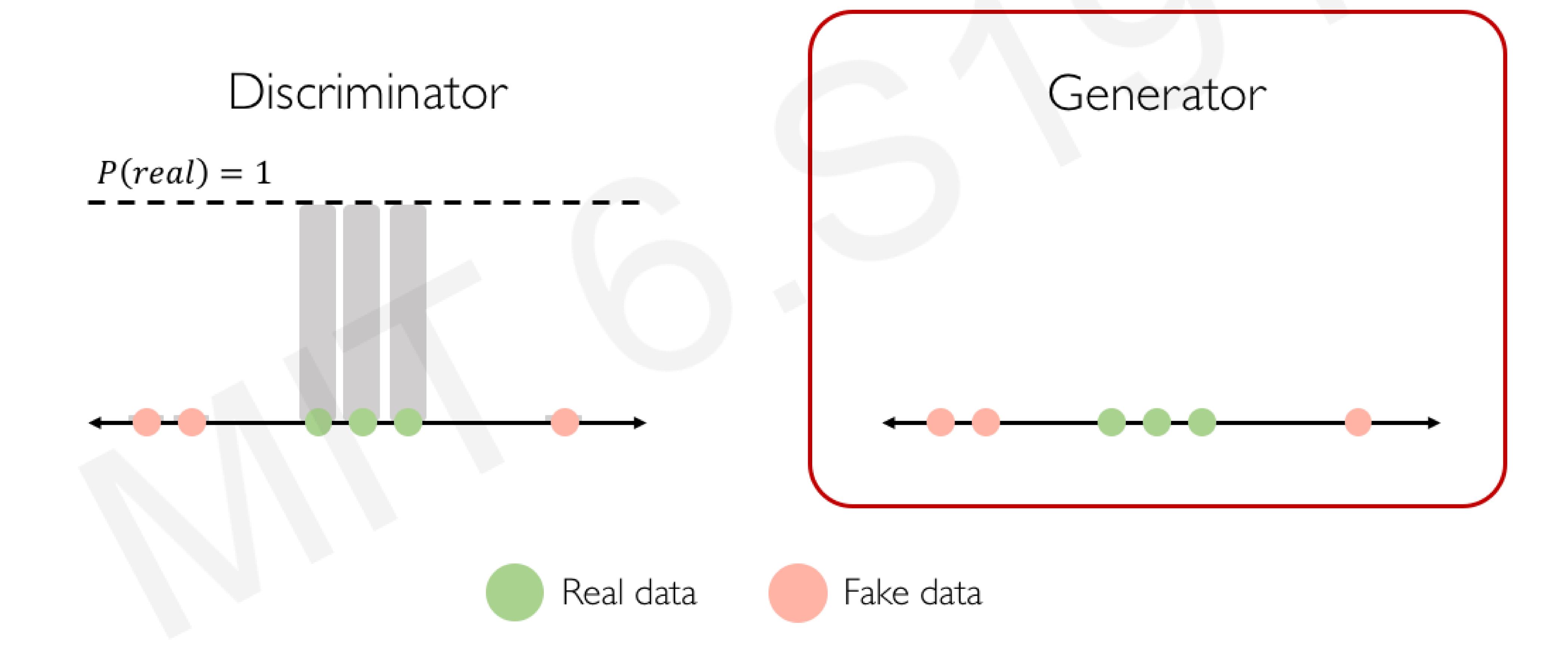


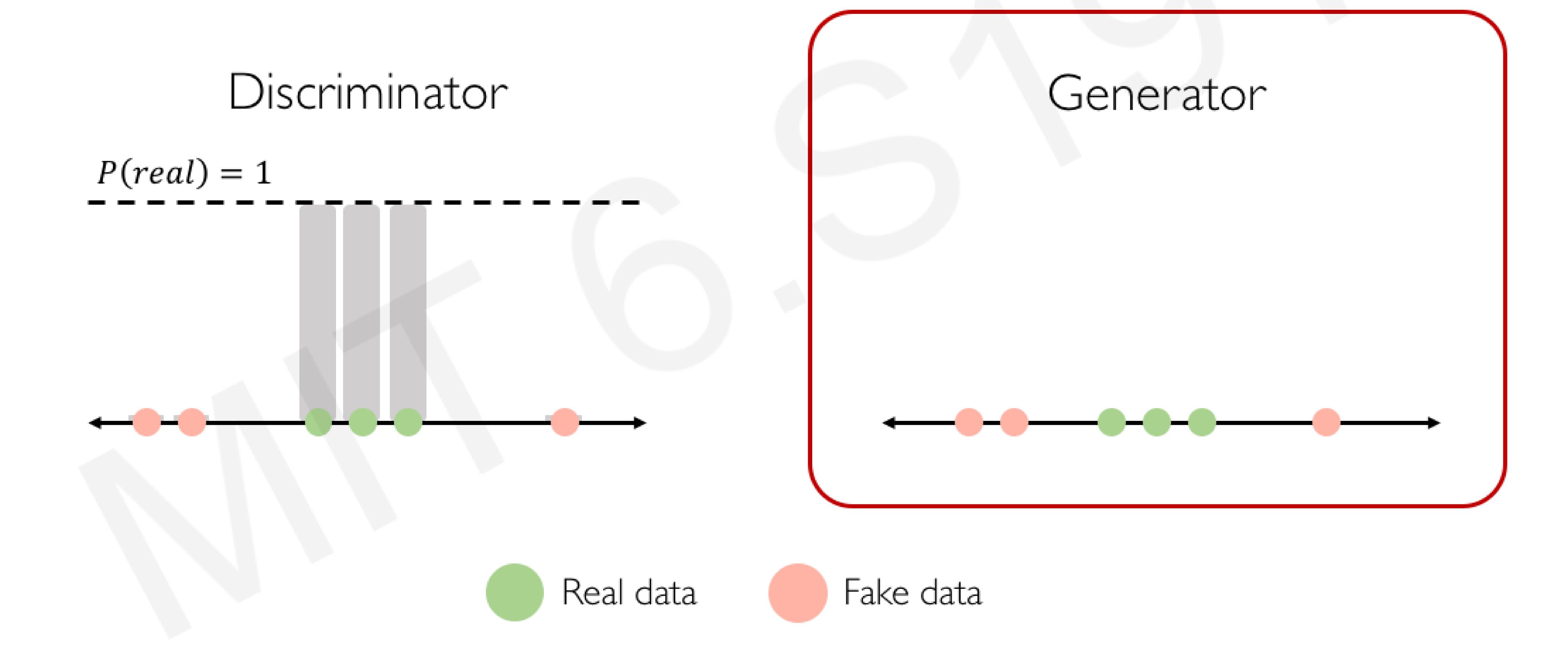


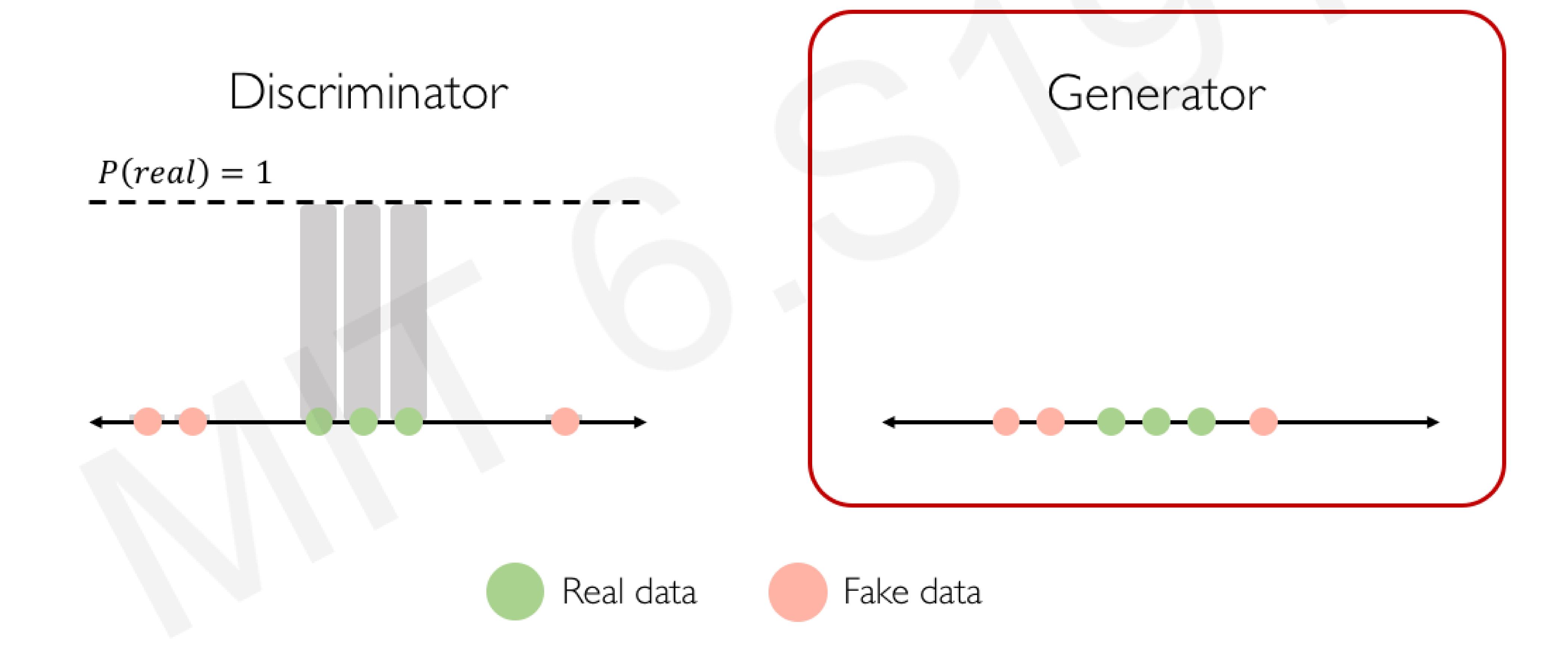


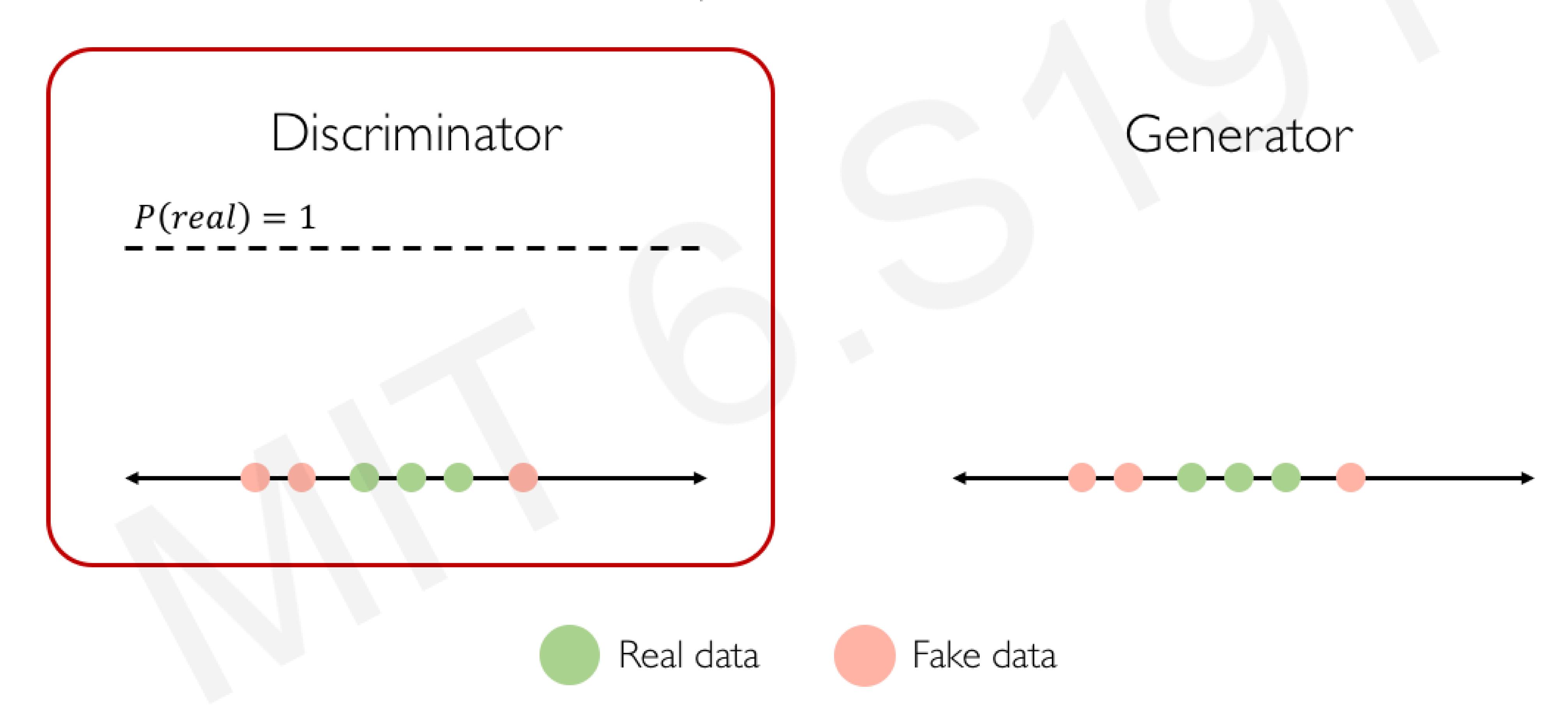


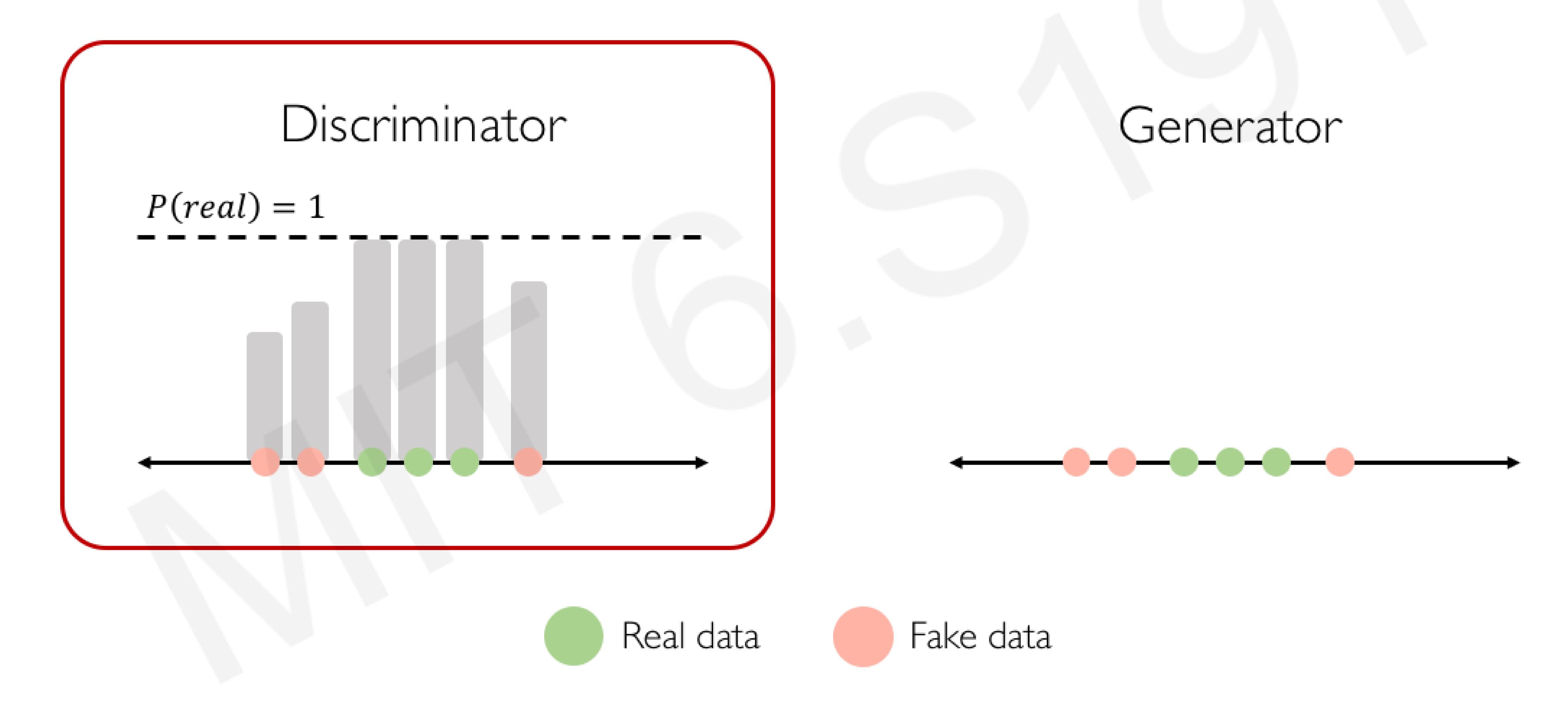


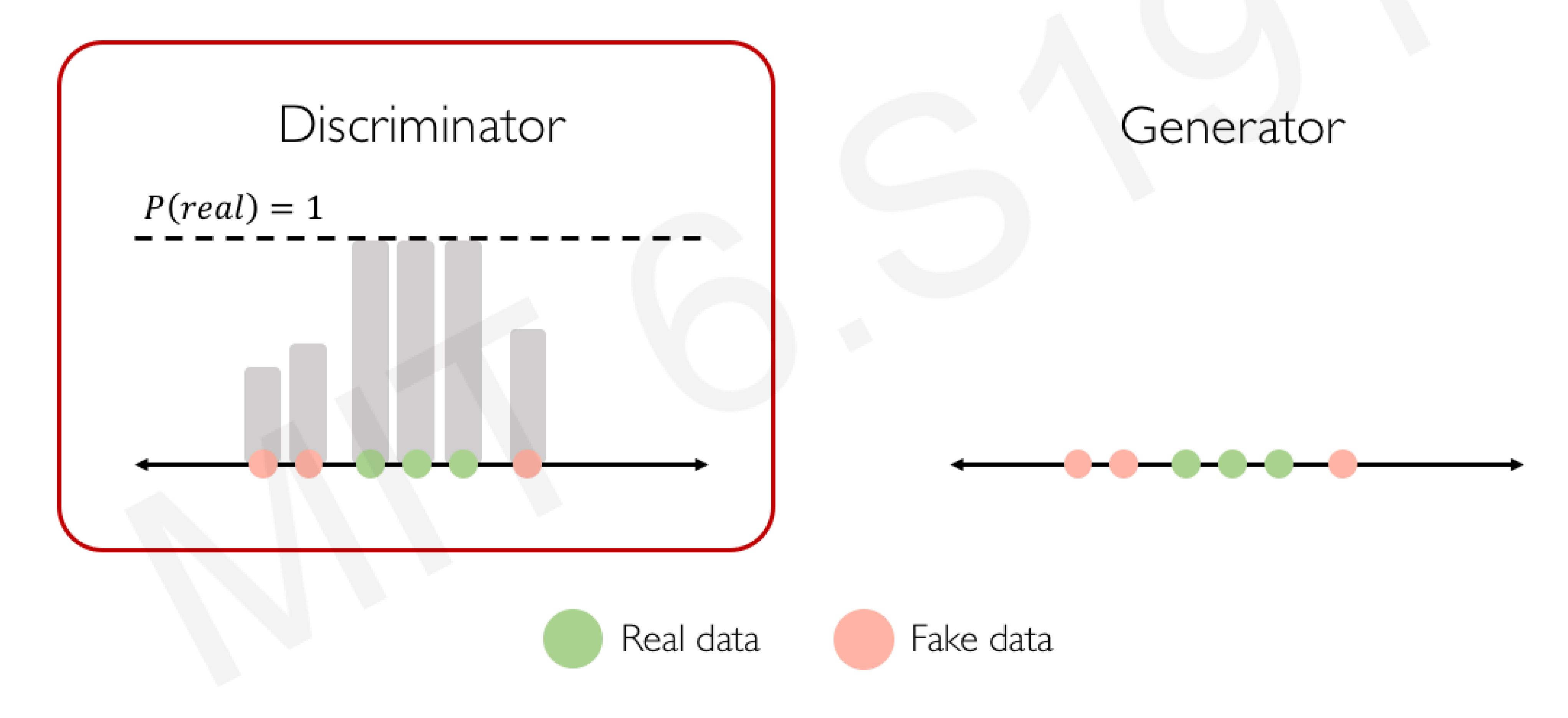


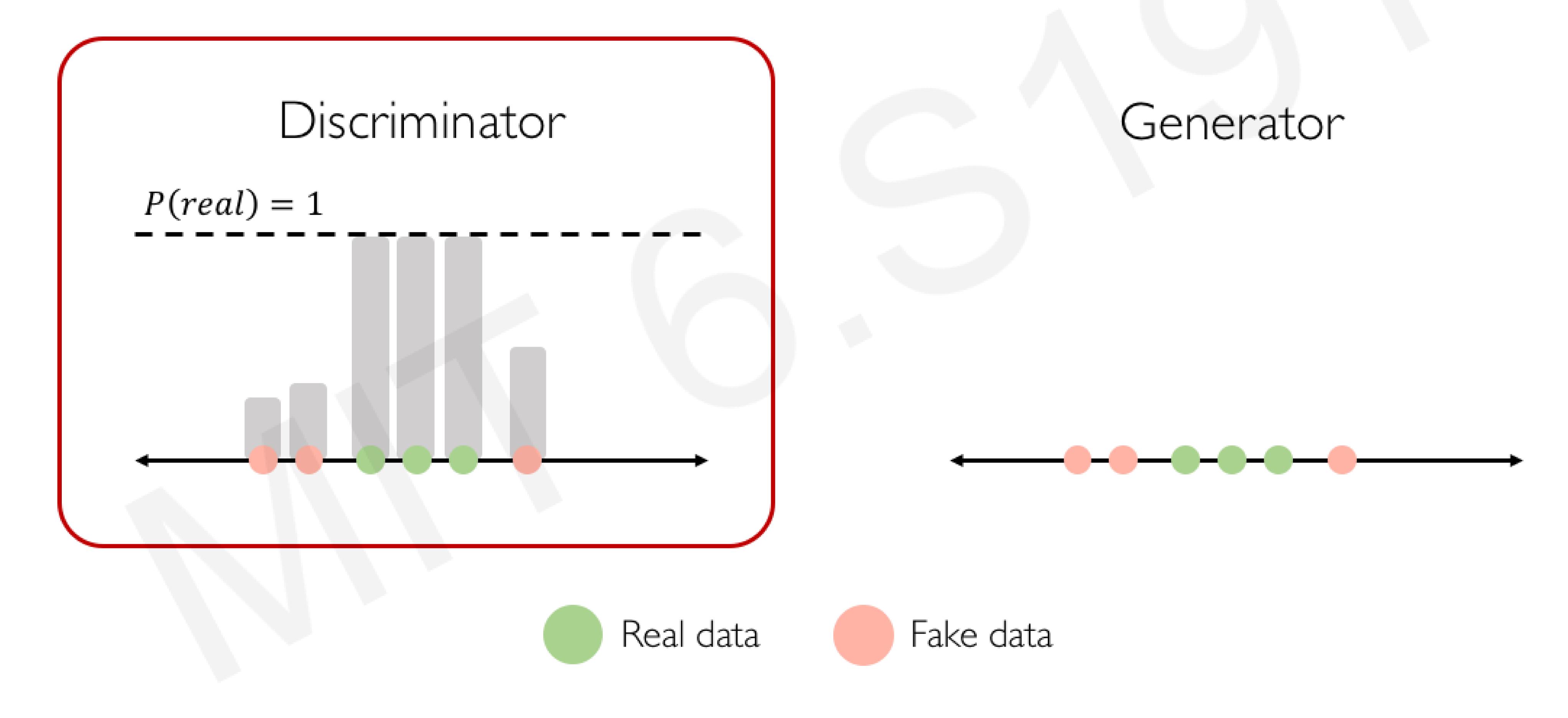


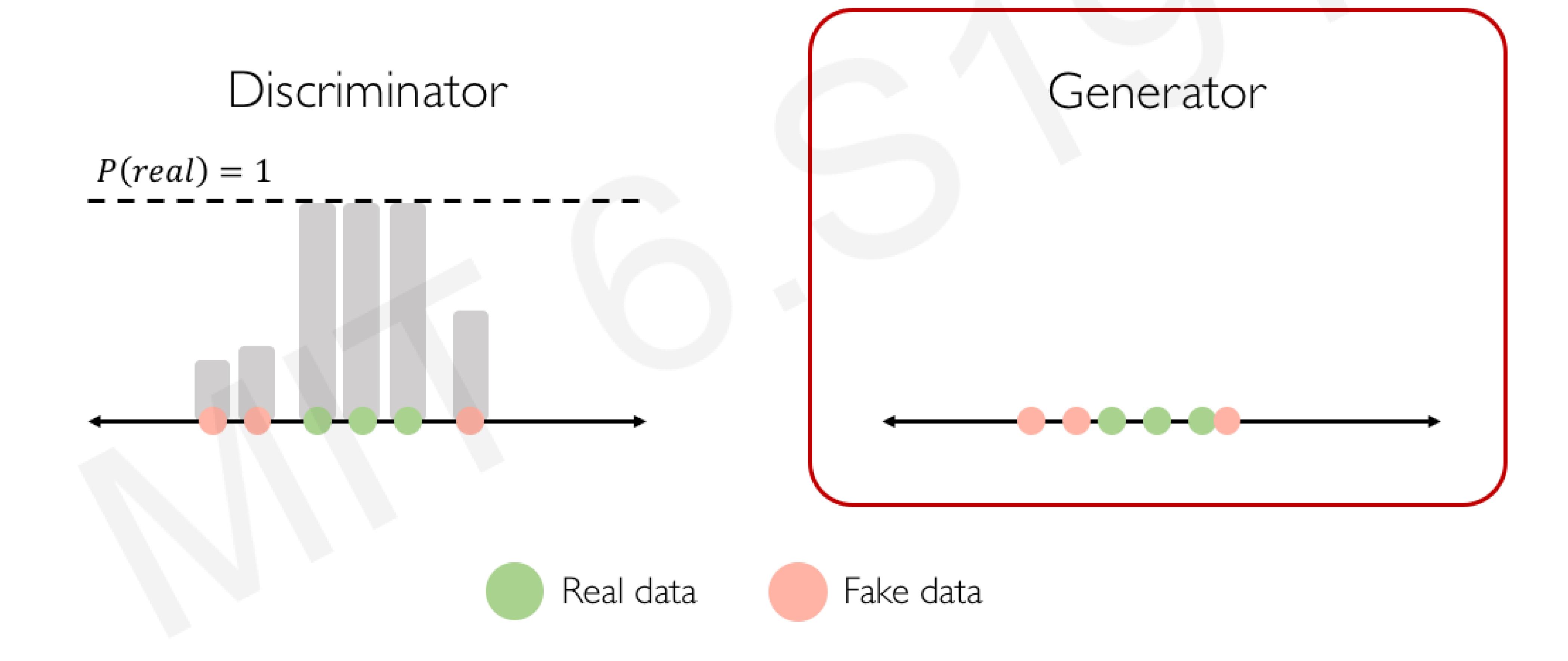


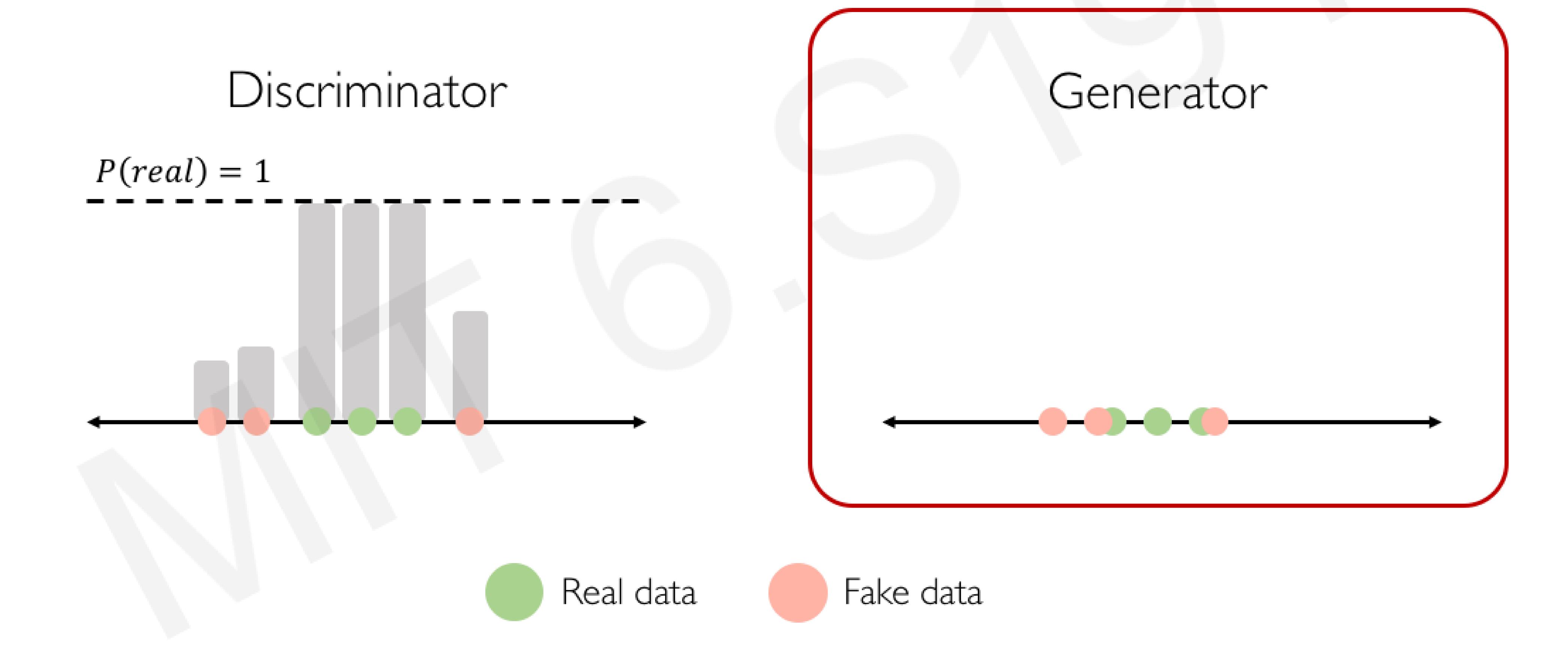


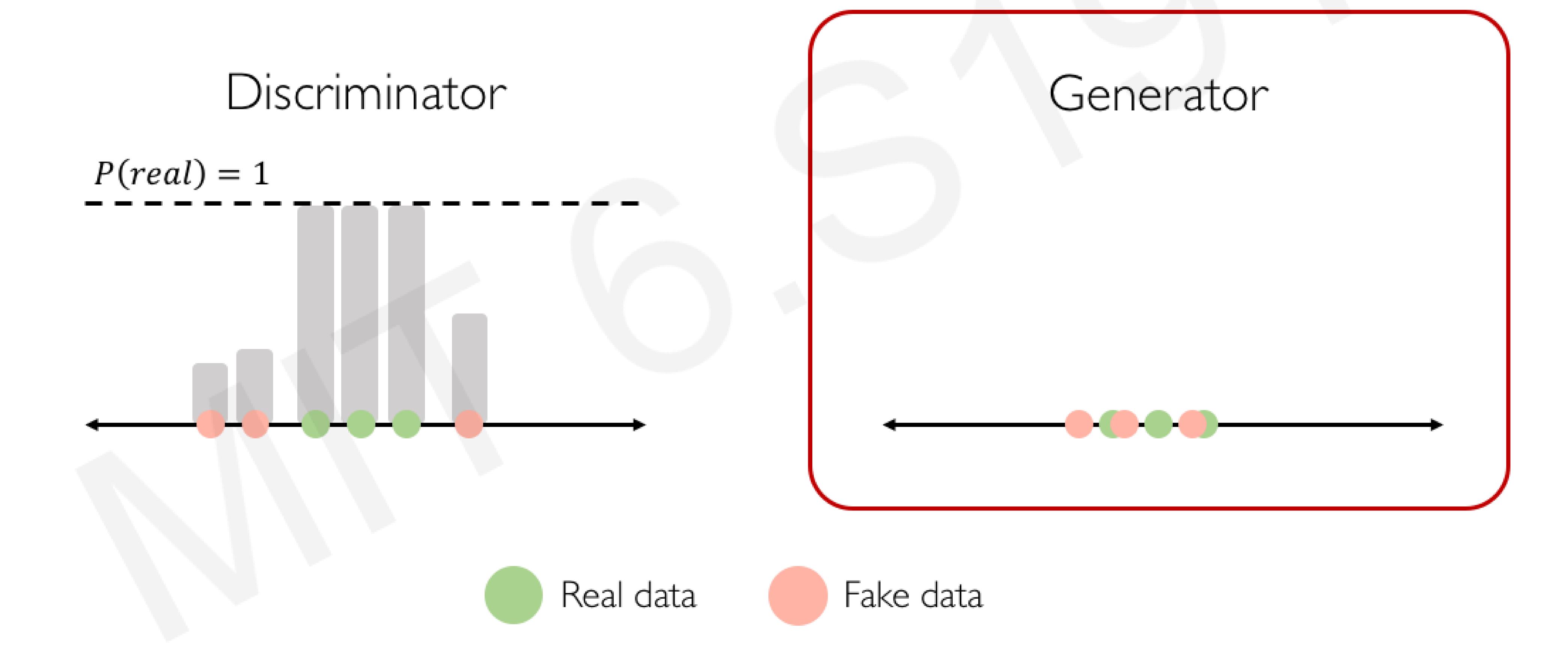




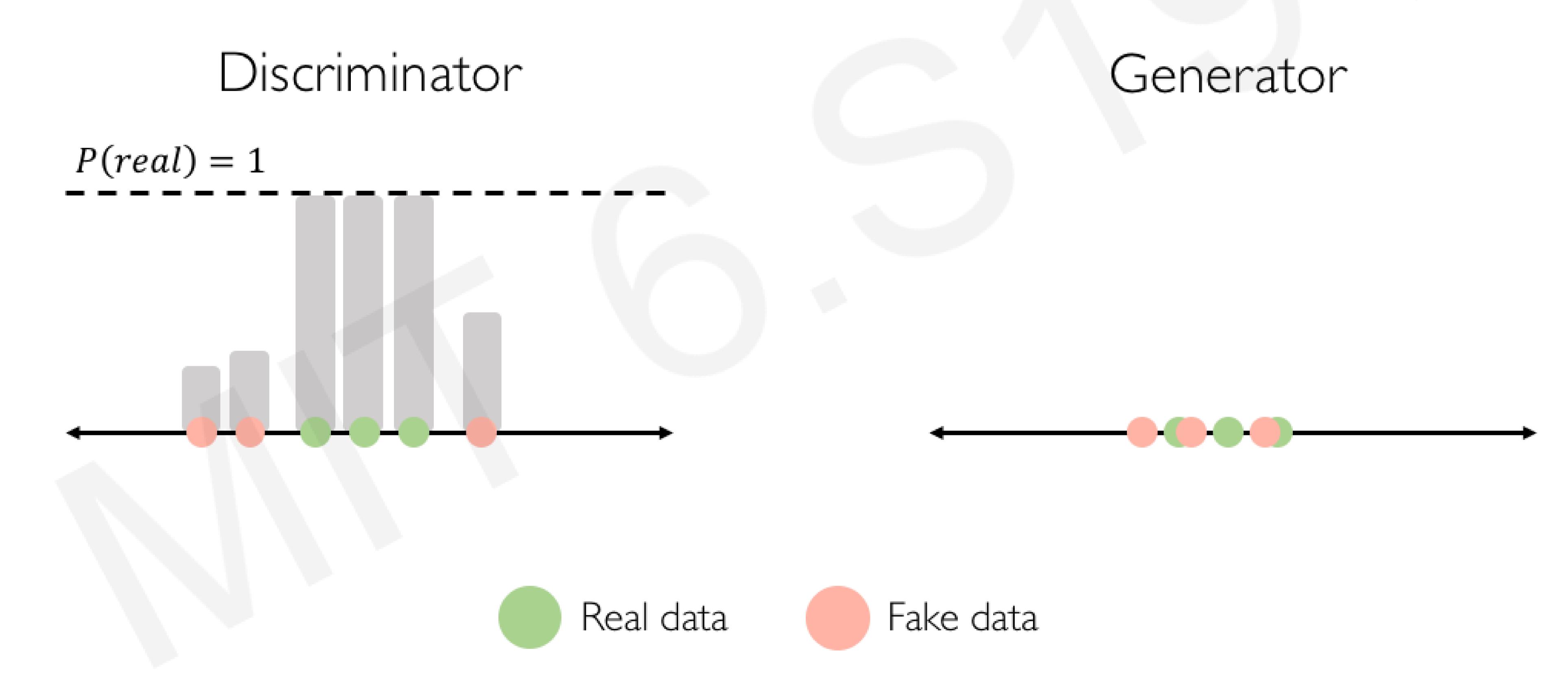








Discriminator tries to identify real data from fakes created by the generator. **Generator** tries to create imitations of data to trick the discriminator.



Generating new data with GANs

After training, use generator network to create **new data** that's never been seen before.

