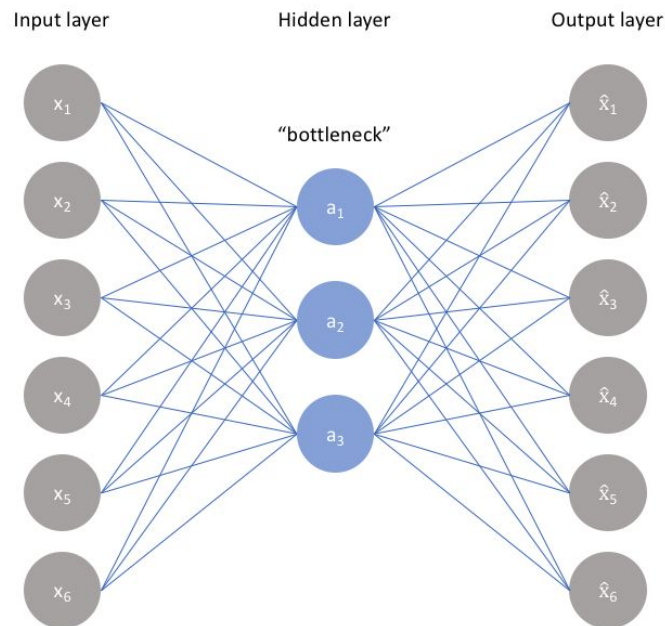


# GANs

<https://github.com/timestocome/DeepLearning-Talks>

# AutoEncoders

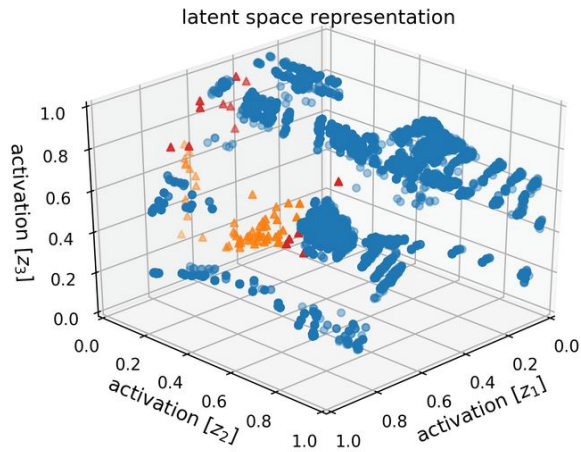
- Compression, noise removal
- Z space, Latent Space, Bottleneck
- data need not be labeled
- still used internally in complex GANS
- ... and other neural networks



# Latent Space, Z space

Not continuous

Can be mapped onto data but  
poorly grouped, can't create similar  
data



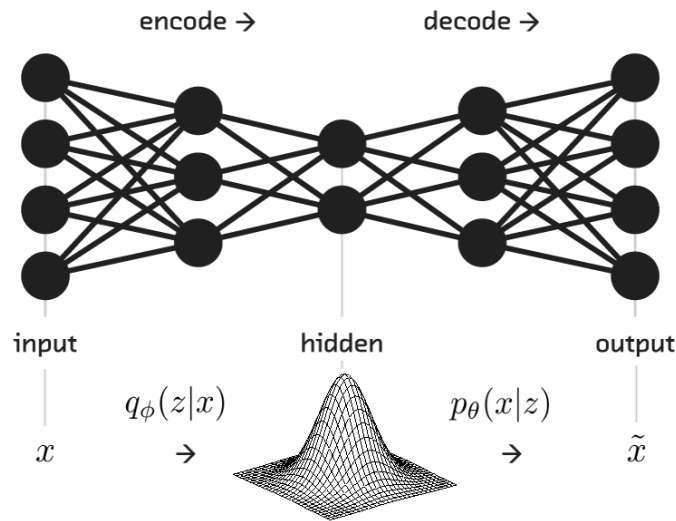
# Variational Auto Encoders

- Continuous Z Space, learns distribution
- Binary Cross Entropy + KL Divergence
- Similar data grouped together
- Can create similar data

How VanGogh is it?

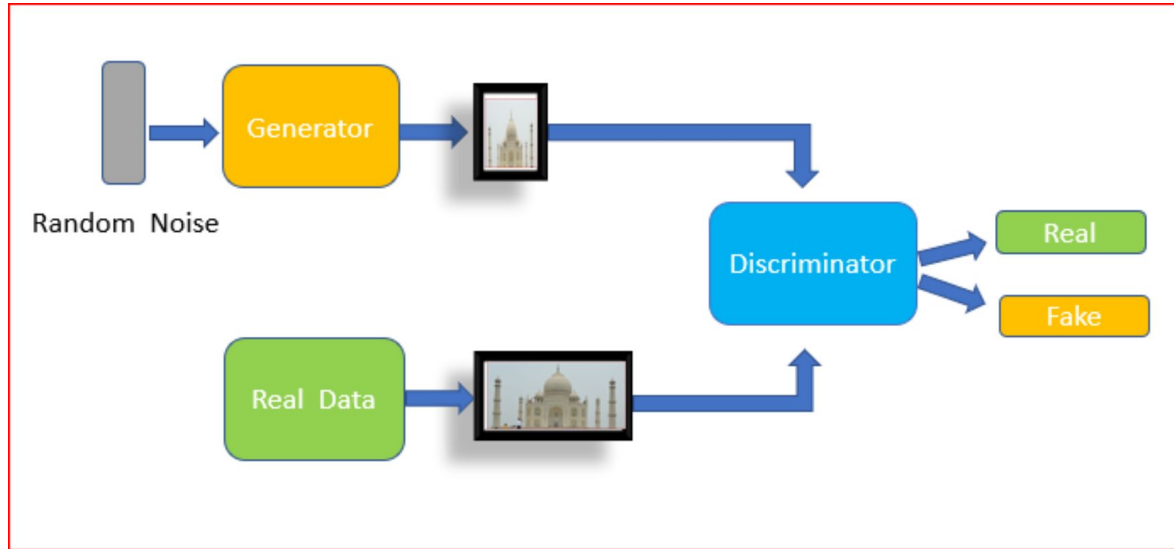
How is this measured?

$$D(p||q) = \sum_{x \in X} p(x) \log \frac{p(x)}{q(x)}.$$



# GANs

- Discriminator ( Critic ) determines error
- Generator can be used to create new data ( medical imaging )



# Training GANs

for each batch:

- take a random real example, and generated fake and classify them

- compute error and update Discriminator

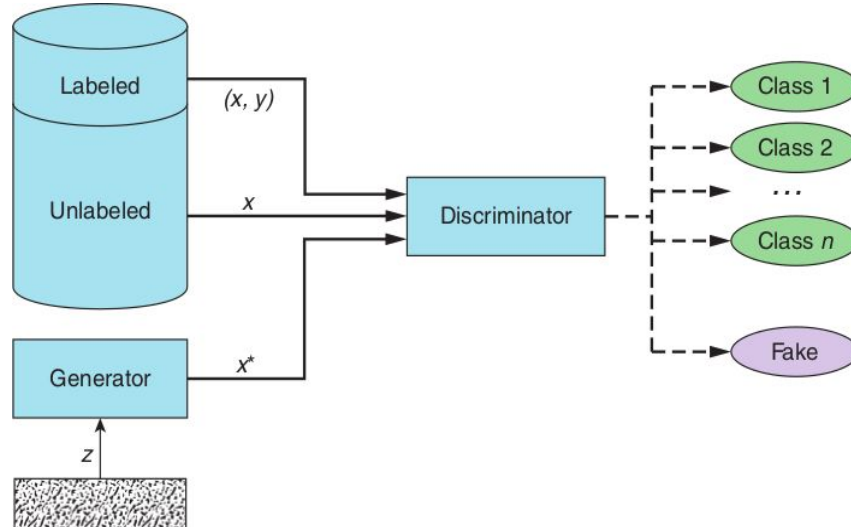
- take random noise and create fake example for Discriminator to classify

- compute error and update Generator

# Semi Supervised GAN

Can be used with partially labeled data ( Kannada/MNIST )

( different distribution btwn train/test/val data, used only 4k labeled images of 40k, 81% accuracy)



# Code

Semi Supervised GAN Kannada code

<https://github.com/timestocome/DeepLearning-Talks/tree/master/GAN%20Talk>

- RL GANs — World Models <https://worldmodels.github.io/>
- Really Awesome GAN resources  
<https://github.com/nightrome/really-awesome-gan>
- Keras GANs <https://github.com/eriklindernoren/Keras-GAN>
- Numpy GAN [https://github.com/shinseung428/gan\\_numpy](https://github.com/shinseung428/gan_numpy)



# Resources

## Paper

- Generative Adversarial Nets

<https://papers.nips.cc/paper/5423-generative-adversarial-nets.pdf>

## Books

- Generative Deep Learning: Teaching Machines to Paint, Write, Compose, and Play (O'Reilly)
- GANs in Action (Manning)

# Misc

Papers with Code: <https://paperswithcode.com>