





**Deep Learning From Scratch** 

#### Unsupervised learning

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## Layout

- Autoencoders
- Learning unsupervised representations
- Sparse coding
- A manifold learning view
- Deep patient

Unsupervised learning tries to understand the properties of a particular set of data. There are different ways of doing this

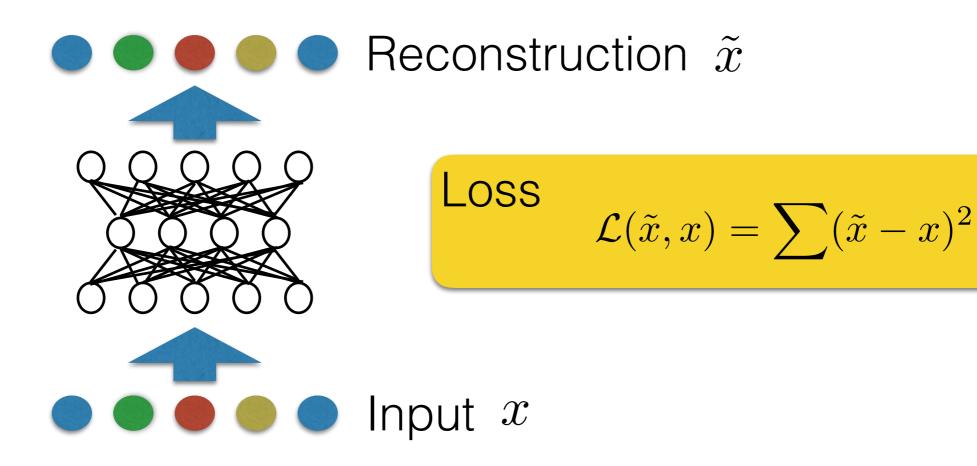
- Clustering Divide data in groups according to some notion of similarity.
- Manifold learning Understanding how data is distributed in the space, parameterising a manifold.

#### Autoencoder

Build a network with the aim of reconstruction.

D.E. Rumelhart, G.E. Hinton, and R.J. Williams. Learning internal representations by error propagation. In Parallel Distributed Processing. Vol 1: Foundations. MIT Press, Cambridge, MA, 1986.

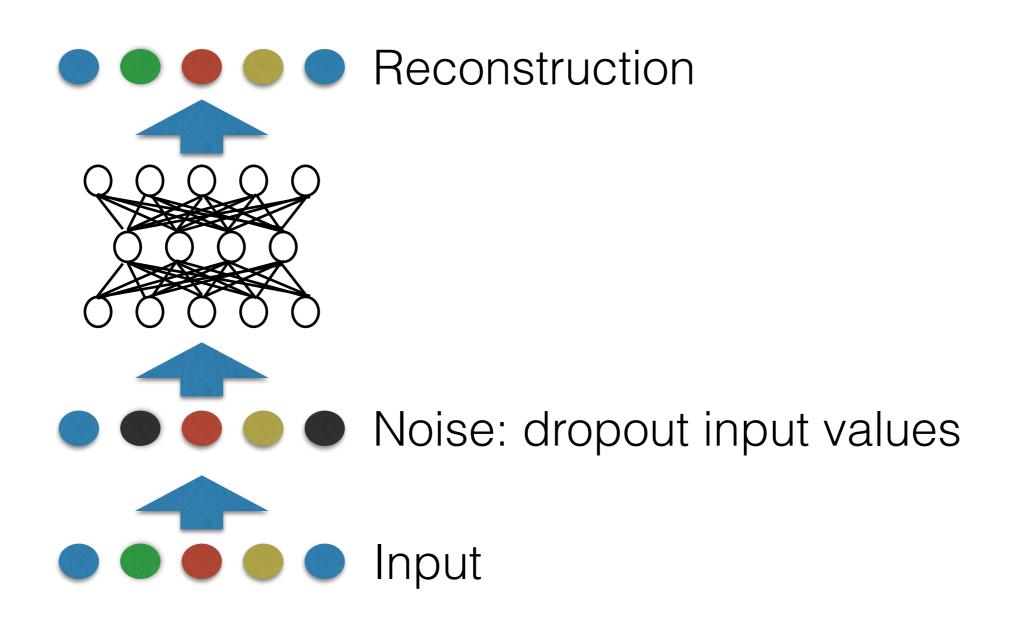
#### Autoencoders



#### Problems

- In large networks it may learn the identity mapping rendering the auto encoder representation useless.
- In order to correct this issue and furthermore give robustness to the auto encoder, demonising auto encoders are proposed.

#### Denoising Autoencoders

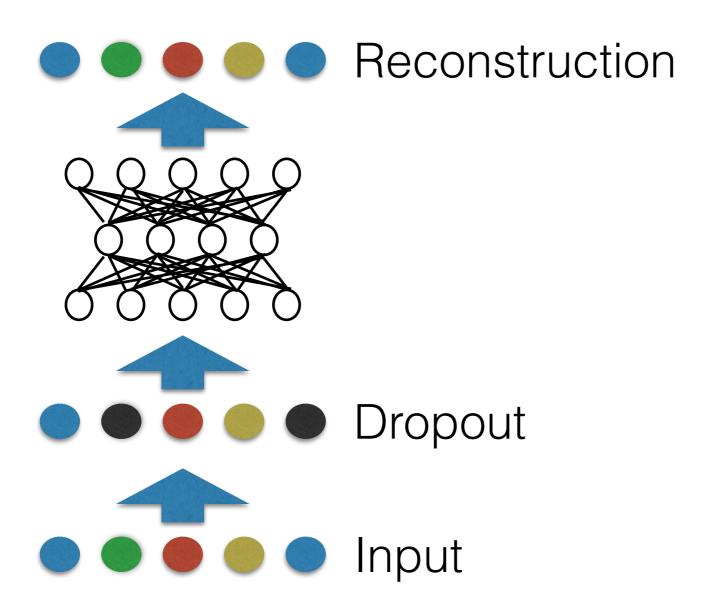


### Hands on

## Learning representations

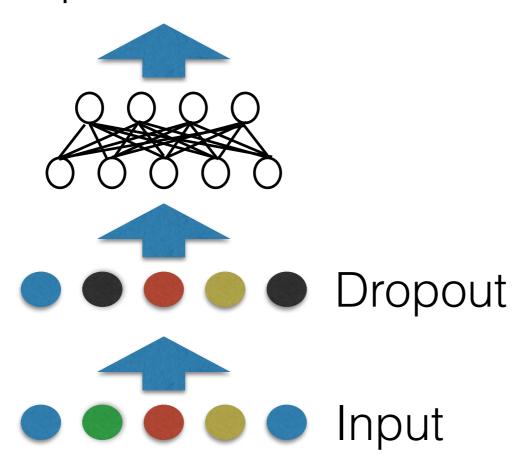
- What can we use these representations for?
- Transfer learning
  - Pure transfer
  - Pretraining
- Compression

# Pretraining and transfer

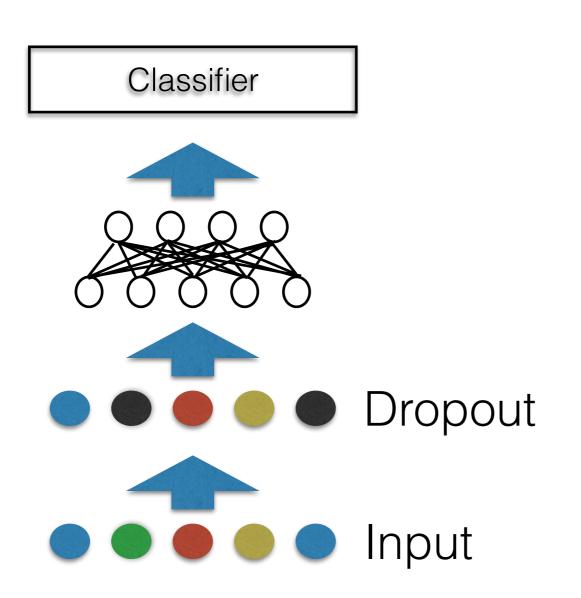


# Pretraining and transfer

Representation



# Pretraining and transfer



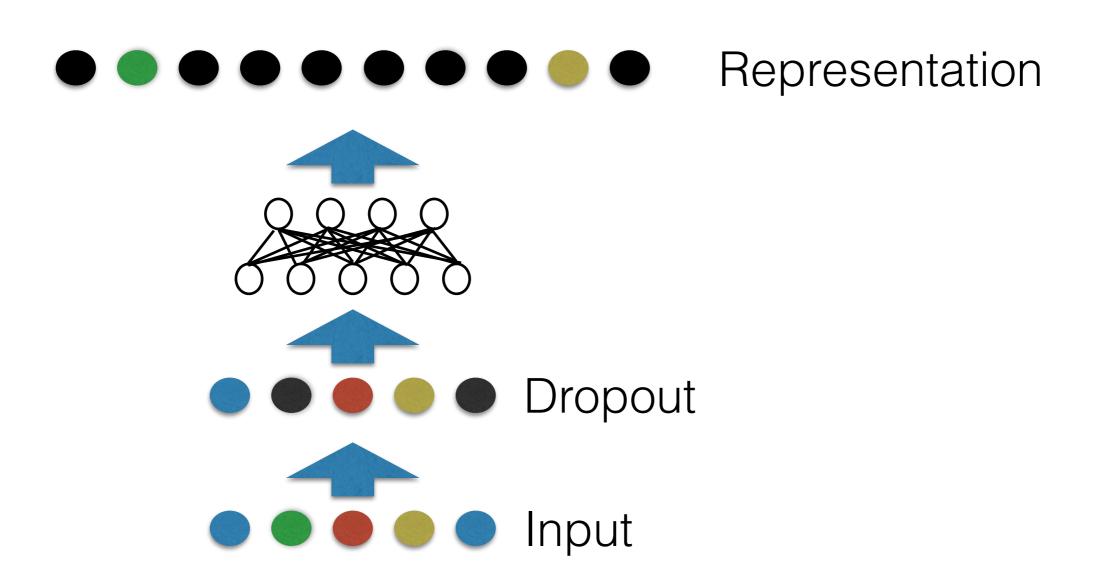
### Hands on

# Compression

 Sparse coding - we might enforce that inner representation is sparse. This is, the learned representation has very few active units.

• How?

# Sparse coding



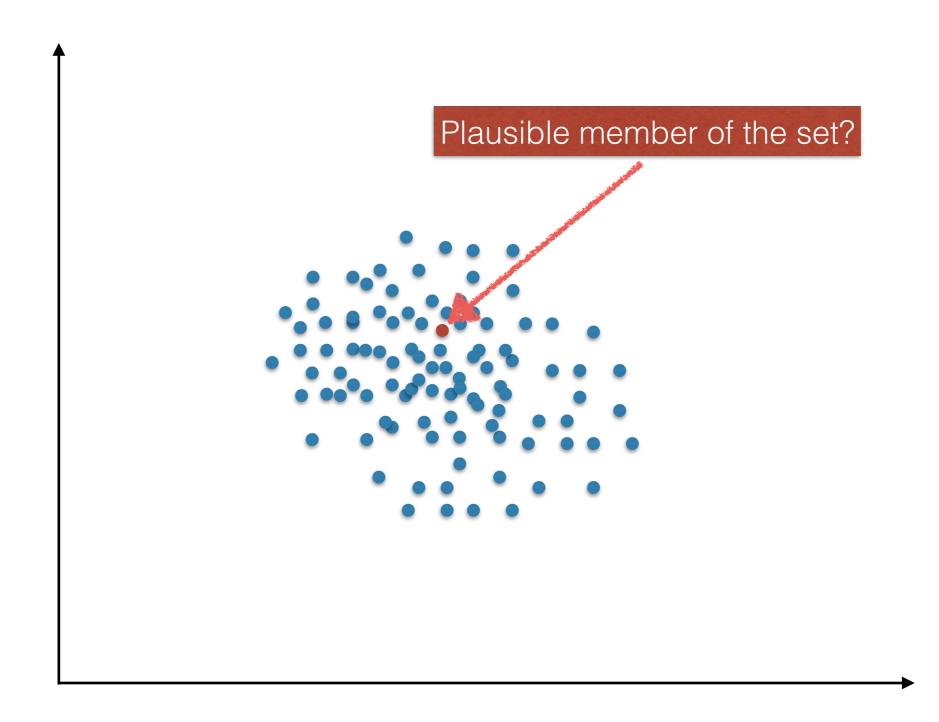
### Hands on

## Manifold learning

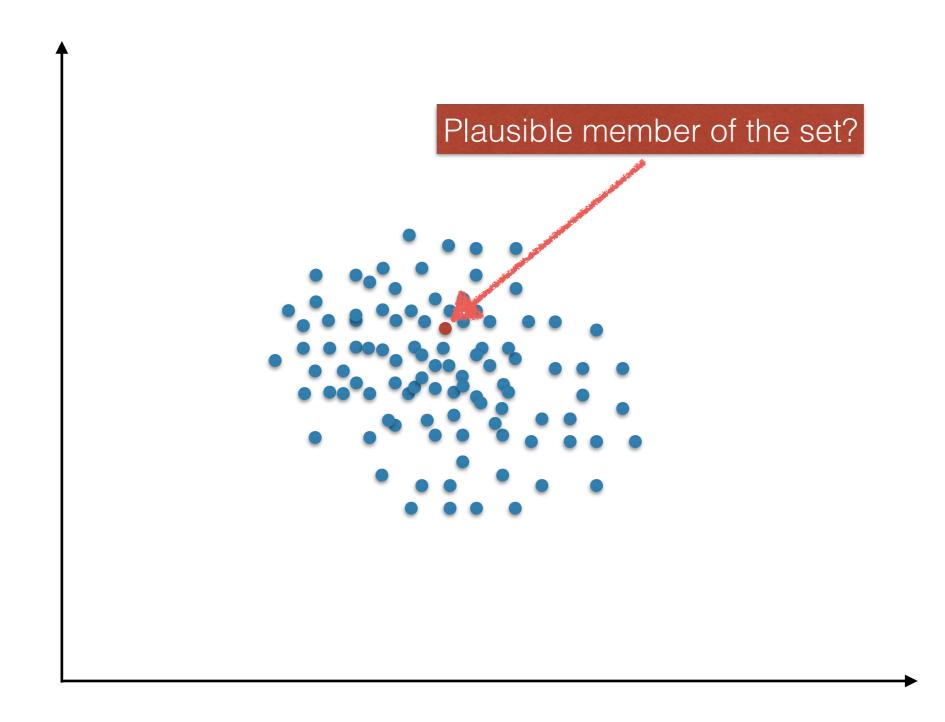
From PCA to Non-linear dimensionality reduction

#### About manifolds

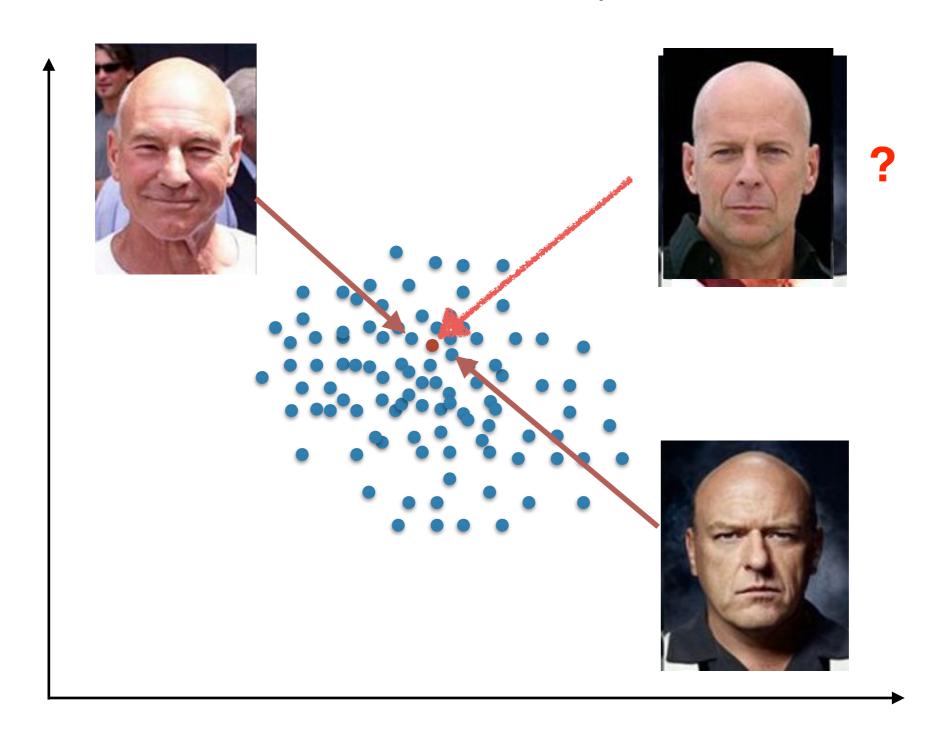
 Consider a general data space. Generally we accept that objects in this space that are close to the samples we have are together

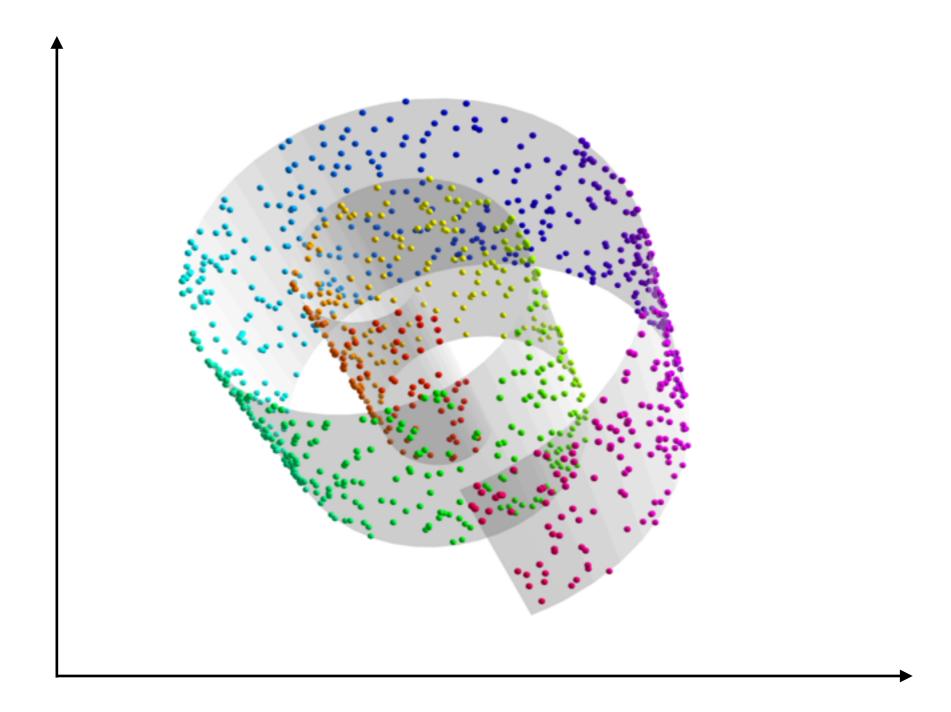


• Consider the space of images.



The Face Space





### Hands on

## Application

 Deep Patient: An Unsupervised Representation to Predict the Future of Patients from the Electronic Health Records

