

Patrick Smith

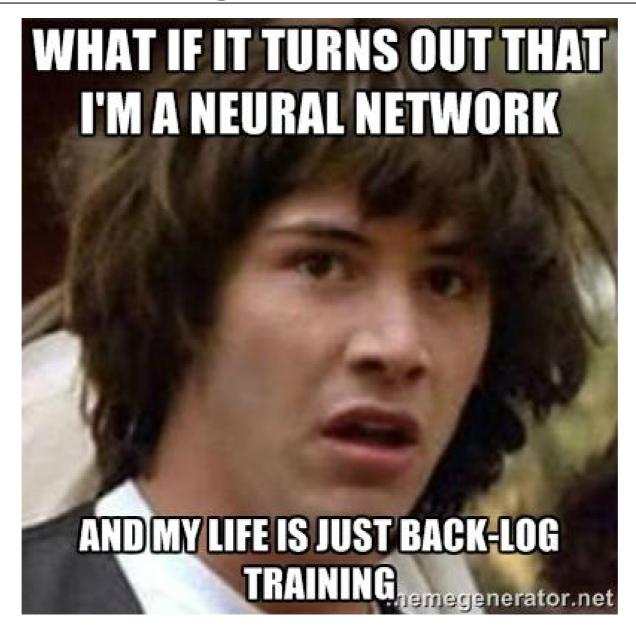
LEARNING OBJECTIVES

- Identify + Learn Additional Unsupervised Learning Methods
- State when/why each unsupervised learning method is utilized

PRE-WORK

- Have a working understand of scikit learn and numpy
- Be able to create functions from scratch in python
- Have a basic understanding of linear algebra concepts such as matrices, eigenvalues, and eigenvectors

OPENING



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Alan Turing, 1947

So far this week, we've learned about different forms of clustering - including hierarchical, k-means, and DBSCAN, as well as PCA.

Today, we're going to walk through some other useful unsupervised learning methods in scikit to add to your data science toolbox.

Overview: Other Unsupervised Learning Methods

"Machine learning is the field of study that gives computers the ability to learn without being explicitly programmed"

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Manifold Learning

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Manifold learning is another form of dimensionality reduction.

This week we learned about dimensionality reduction in a general sense; however, what happens when we want to perform dimensionality reduction on a non-linear dataset? That's where Manifold Learning comes in.

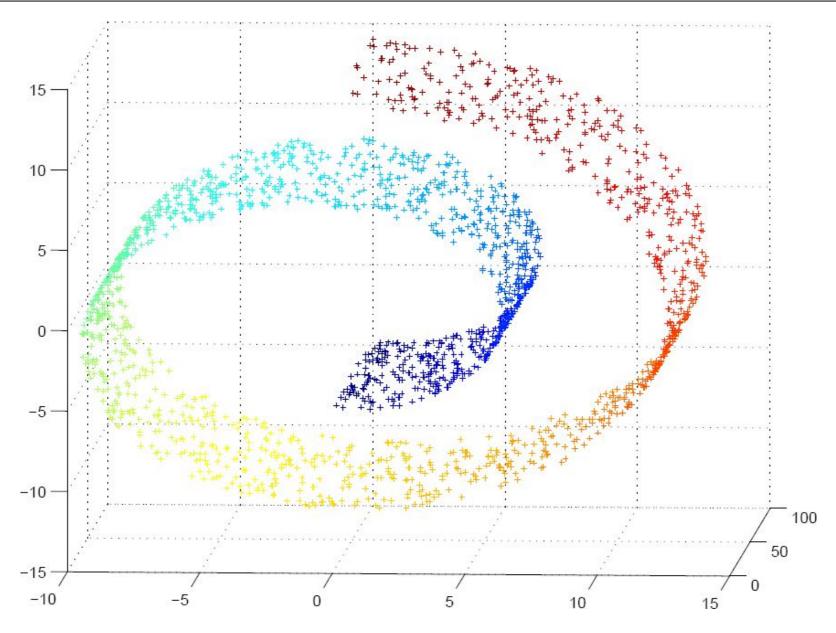
Manifold Learning: Recall PCA

There are 3 assumptions that PCA makes:

- *Linearity:* Our data does not hold nonlinear relationships.
- Large variances define importance: our dimensions are constructed to maximize remaining variance.
- **Principal components are orthogonal:** each component (columns of Z) is completely un-correlated with the other

Manifold Learning:

- Manifold learning is a way to generalize a dimensionality reduction technique like PCA to take into account non-linear structures in the data.
- When we set up Manifold Learning, it learns the structure of a highly dimensional data from the data itself, and then performs the reduction.

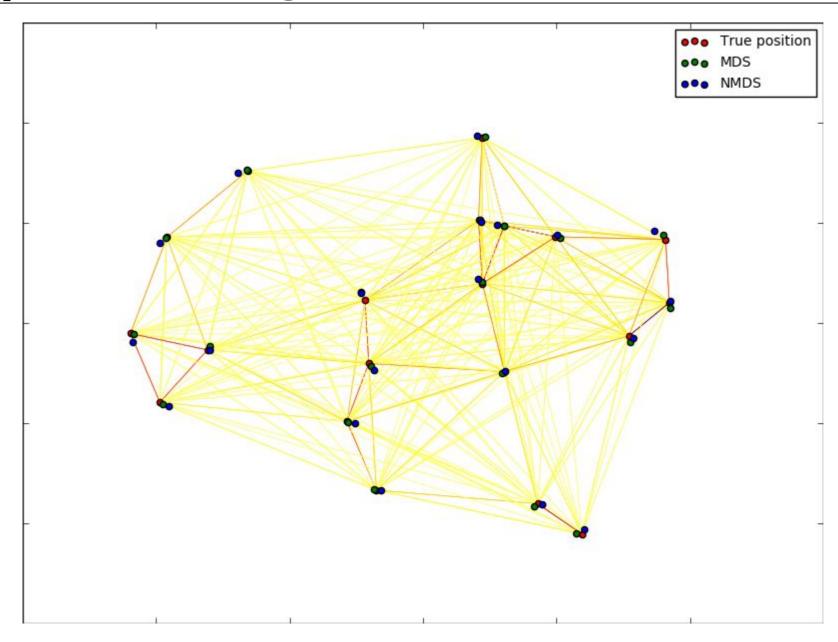


Manifold Learning: Isomapping

- Isomapping is one of the oldest and simplest implementations of manifold learning it seeks to reduce the dimensionality while mainting the distance between points.
- When we set up Manifold Learning, it learns the structure of a highly dimensional data from the data itself, and then performs the reduction.

Manifold Learning: Multidimensional Scaling

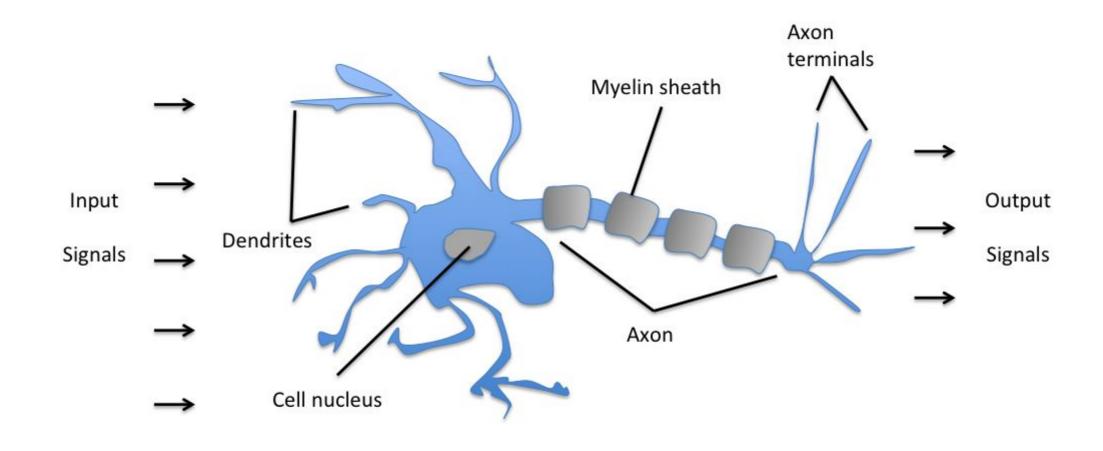
- In a basic sense, multidimensional scaling is a method to provide a visual representation of similarities among a group of data.
- With multidimensional scaling, we're taking a highly dimensional data set and representing it in as little as two dimensions.



Artificial Neural Networks

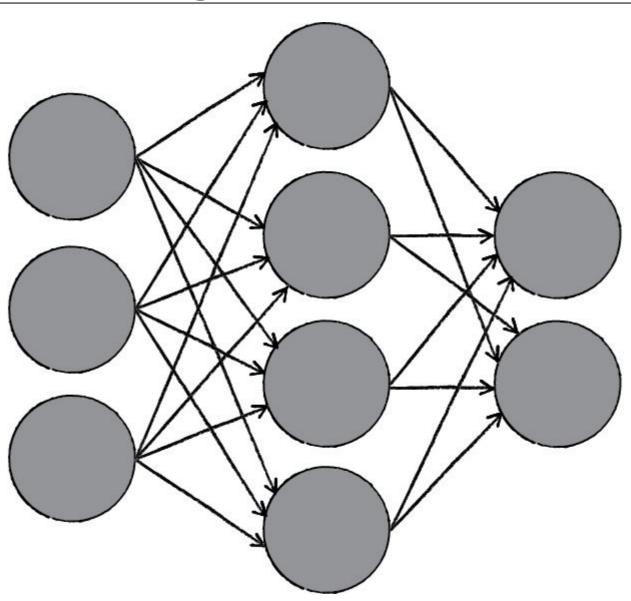
Artificial Neural Networks are simulated models of neural networks in the brain.

In a practical sense, neural networks have become most useful to tackle problems that are simple for the human brain but difficult for machines, IE: pattern recognition.



Schematic of a biological neuron.

- When a neural network receives information, it processes it through a system of nodes
- While this may seem simplistic, a great, intertwined network of these nodes are utilizing parallel computations can be exceptionally powerful.
- Neural networks learn and adapt just the way a human brain would to changing information.



We can implement artificial neural networks as a supervised learning method or an unsupervised learning method.

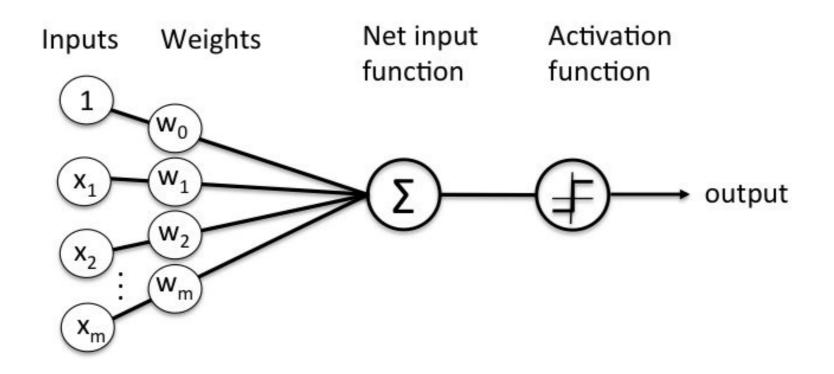
In the former, we train the network to learn.

Let's take the example of a segment of farm animals - a cow, horse, pig, and goat. We can tell train the network to recognize each animal and then have it provide us it's best guesses, and in return we give it the correct answers, and the network then compares its answers to the correct answers and learns the correction classifications.

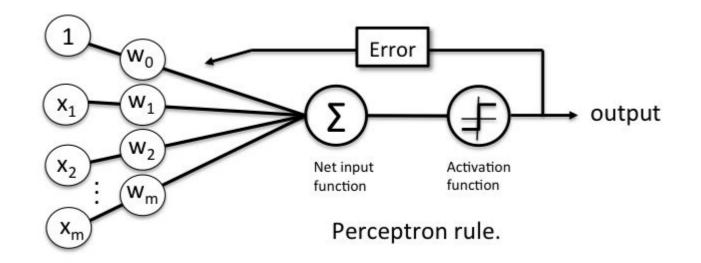
With unsupervised neural networks, we train the neural network in a situation where there is not a set of known answers.

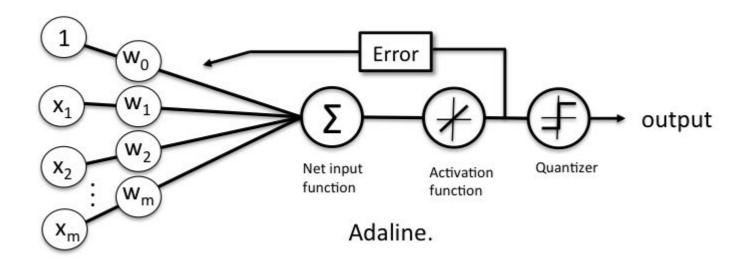
In scikit, we can implement a form of unsupervised neural networks by utilizing Restricted Boltzmann Machines, which are models that learn a probability distribution

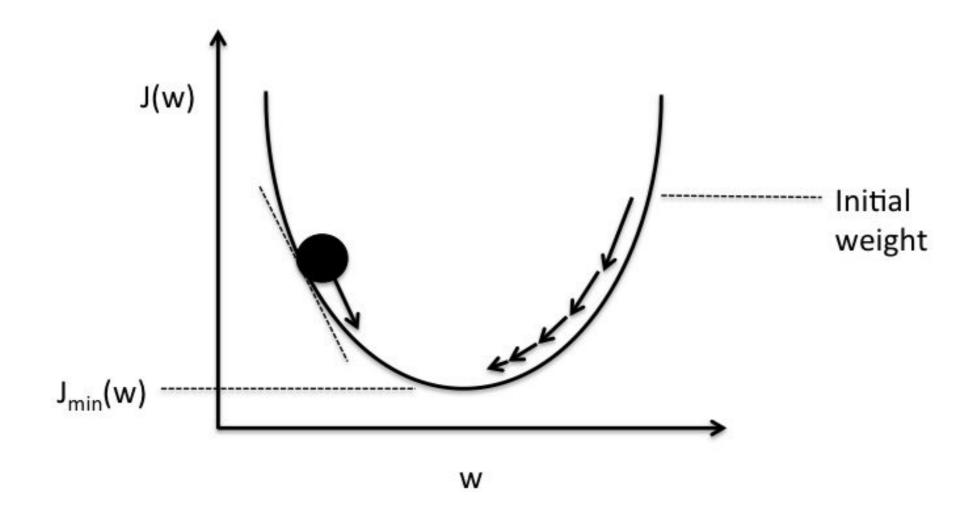
Guided Practice: Constructing a Neural Network



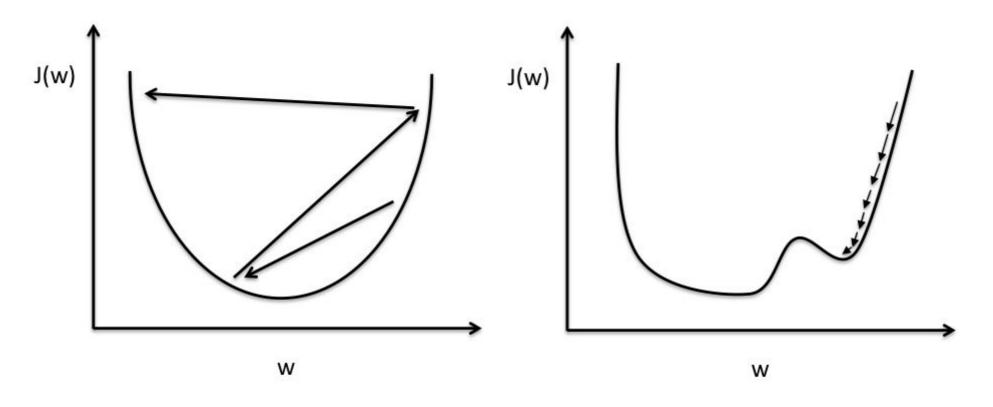
Schematic of Rosenblatt's perceptron.





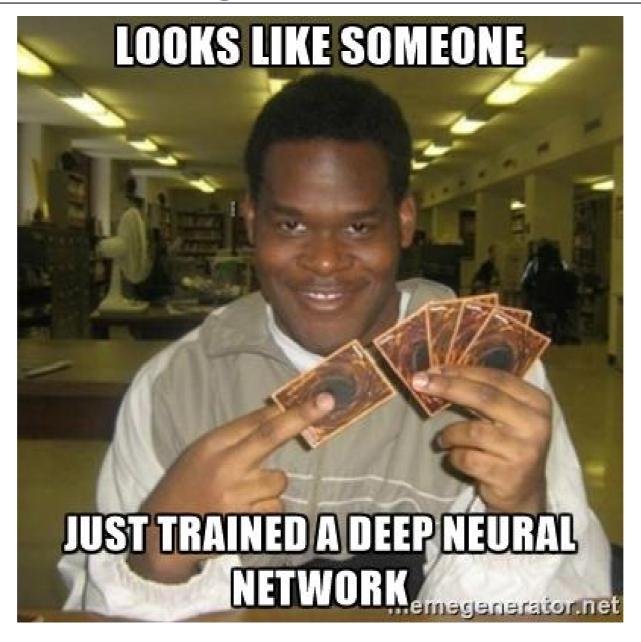


Schematic of gradient descent.



Large learning rate: Overshooting.

Small learning rate: Many iterations until convergence and trapping in local minima.



Conclusion

Q&A