



Machine

Learning

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Lecture

PH451, PH551

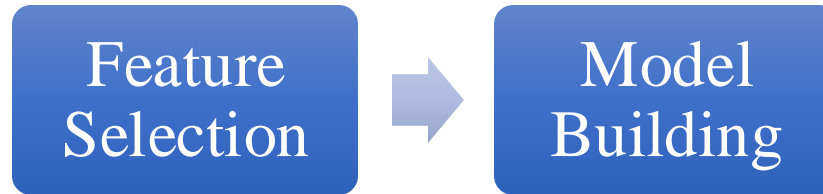
February 18, 2025

Announcements

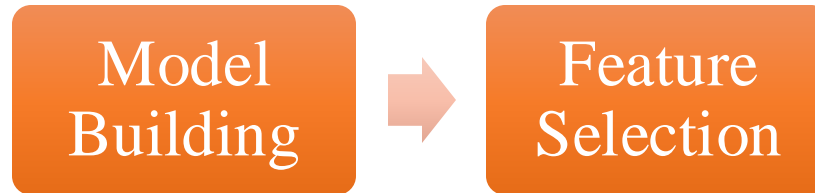
- **Hands-on #4 – due Thu**
- **Hackathon #1 – due Fri**
- **Hands-on #5 – due next Thu**

Recap: Feature Selection

Filters



Wrappers



Embedded-Hybrid



Recap: Wrapper Methods

Selection tied to a model:

- More accurate
- Assess feature interactions
- Search for optimal subset of features

Model
Building



Types:

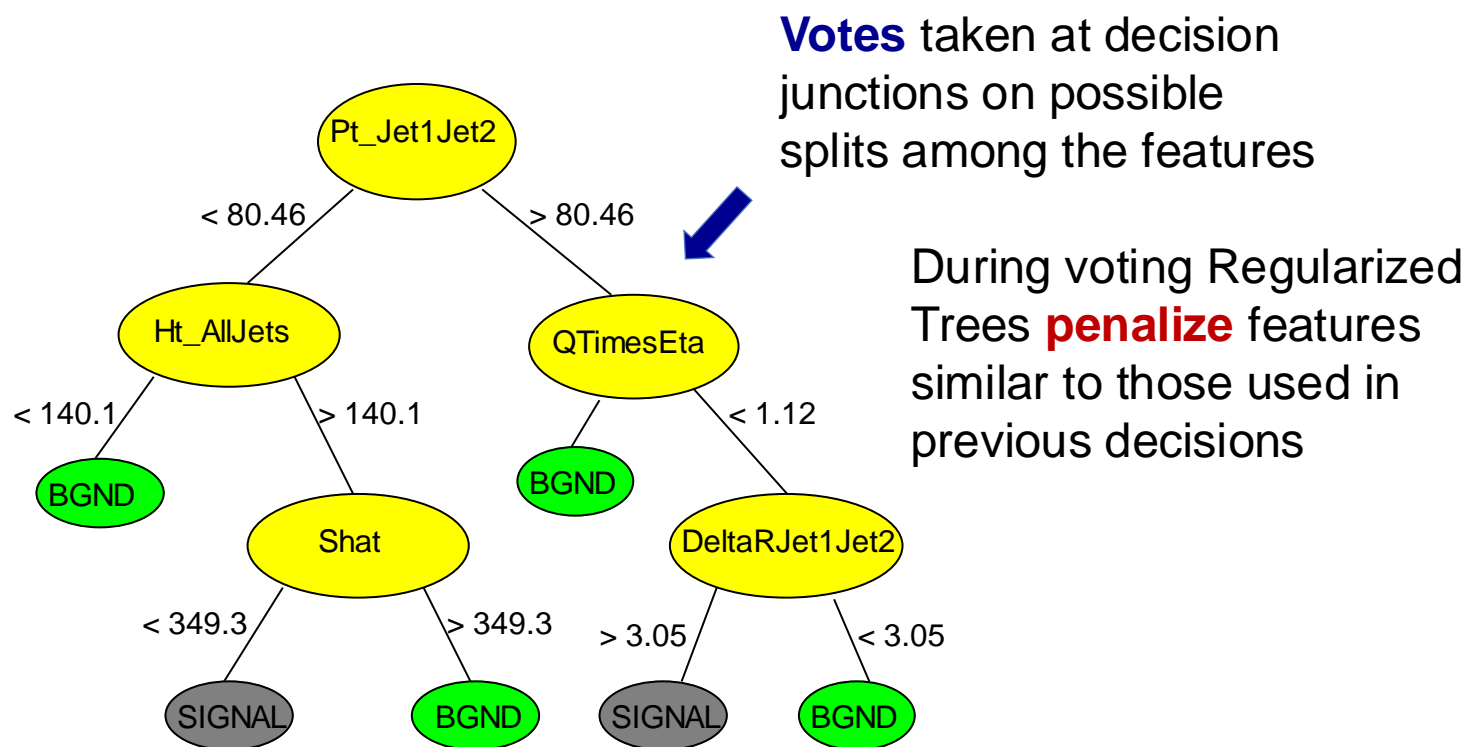
- **Methodical**
- **Probabilistic**
 - random hill-climbing
- **Heuristic**
 - forward backward elimination

Feature
Selection

Recap: Regularized Trees

Inspired by J. Friedman and Popescu, 2008 work on rules regularization

Decision Tree:



Recap: PCA

Principal Components

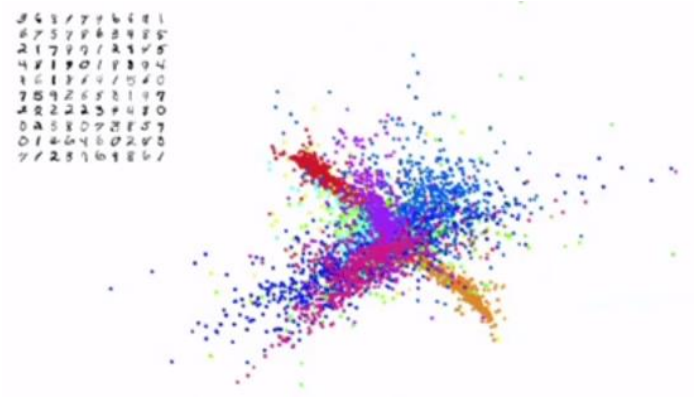
- Linear Method, Pearson/Hotelling 1901/1933
- Find the **hyperplane closest** to data and project into it
 - Minimize the **squared distance** between original data and projection
 - Orthogonal axes that **maximize remaining variance** (principal components)
 - Find with Singular Value Decomposition (SVD)
 - Ignore components of lesser significance

MNIST

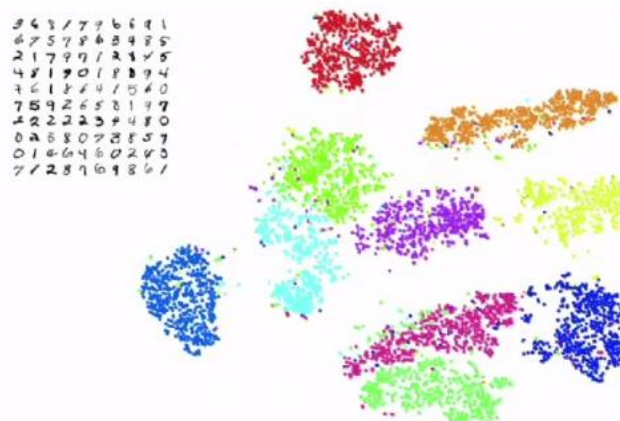
PCA



LLE



t-SNE



Unsupervised Learning



Why?

- **Supervised learning:**
 - **Class labels**
- **Unsupervised**
 - **No labels**
 - **Part of human learning experience**

Applications

- **Dimensionality Reduction and Visualization**
 - All methods we met so far were unsupervised (PCA/LLE/t-SNE)!
- **Anomaly Detection**
- **Recommendations**
- **Clustering**

Clustering

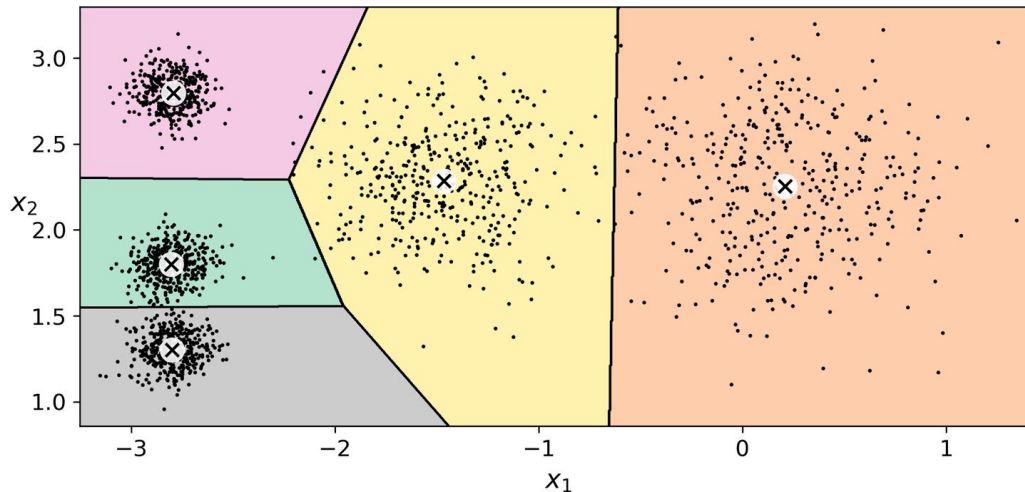
- Find groups of instances similar to each other and different from instances in other groups
- Key questions:
 - How many groups?
 - How to partition the data?

K-Means

Lloyd, 1957

- Popular algorithm
- Relatively quick
- Find cluster centers (centroids)
- Assign instances to closest cluster
- Note: have to specify # of clusters (k)

K-Means Algorithm



- **Randomly** pick k instances as centroids
- **Label, update** centroids, repeat...

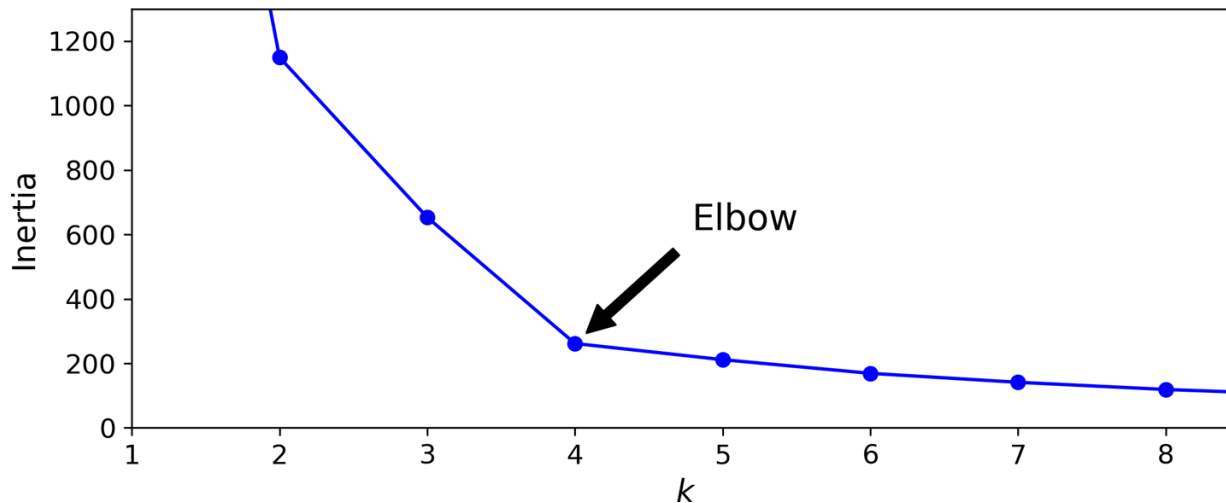
- **Issues:** may converge to local minimum instead of global, strong dependence on initialization
- **Improvements:** K-Means++ (2006) start with centroids that are distant from each other, mini-batches

Performance Metric

Inertia

- Mean squared (L2) distance between each instance and closest centroid

“Elbow” method



Predictive Clustering

Example of a **multi-function regression** model based on trees or rules

- **Decision trees** are equated to clustering trees by P. Langley in 1996, first noted by Fisher in 1993
- **Cluster “hierarchy”** each tree node corresponds to a cluster. Root node contains full dataset partitioned recursively into sub-clusters

Predictive Clustering Trees

Predictive clustering implementation

- **Decision tree and rule induction** system
- Designed for **multi-task learning** and multi-label classification
- Well-suited for both **classification** and **regression** problems

Clustering Concept

Use **decision tree induction** to obtain clusters with:

- **minimal intra-cluster distance**
 - between examples from the same cluster
- **maximal inter-cluster distance**
 - between examples from different clusters
 - In classification trees distance metric is class entropy

Illustrative Example

14 input variables $\{a, b, c, d \dots\}$

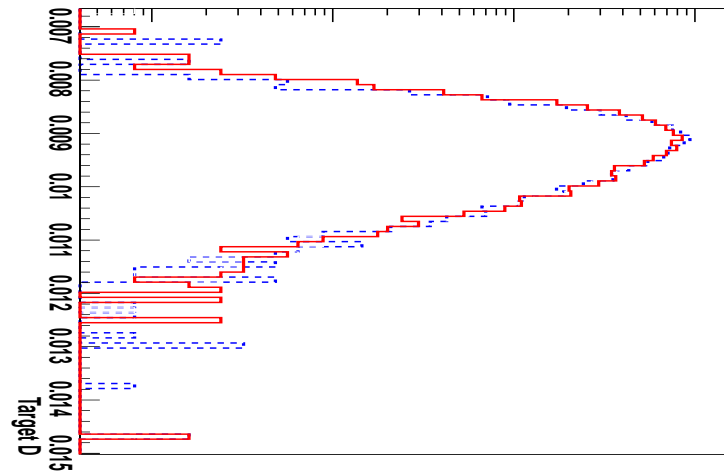
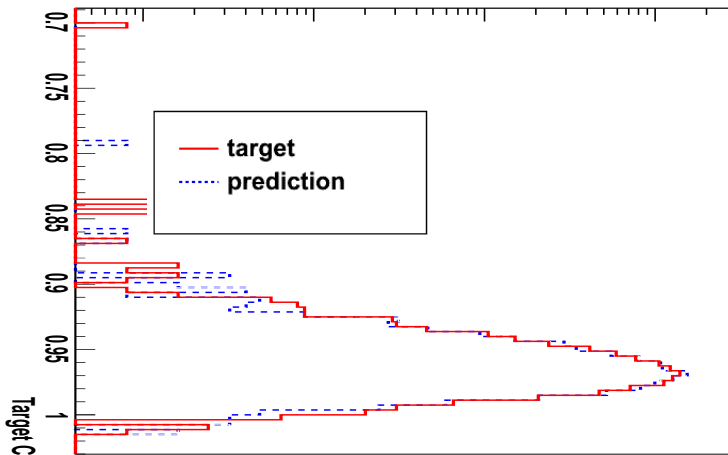
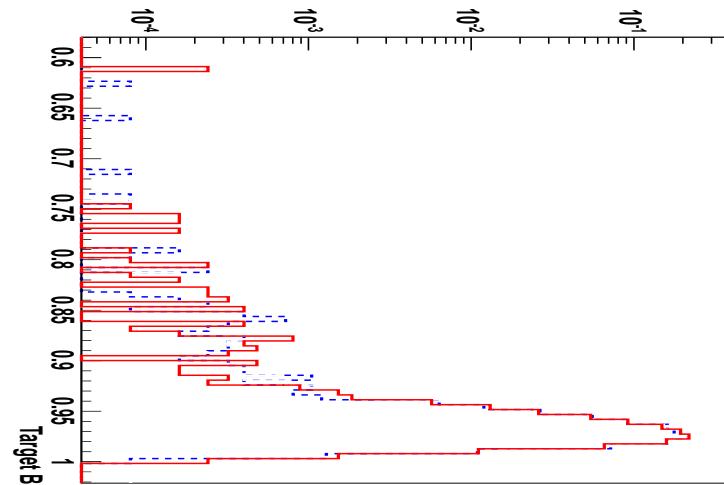
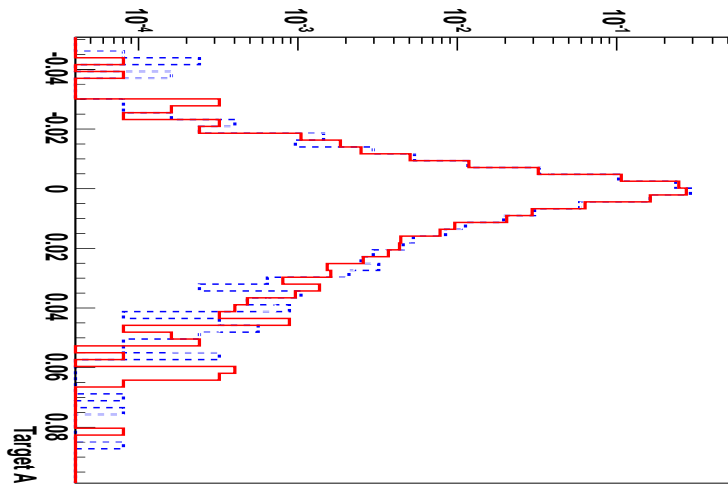
- 4 of them strongly correlated

14 target outputs to estimate $\{A, B, C, D \dots\}$

- 4 of them strongly correlated

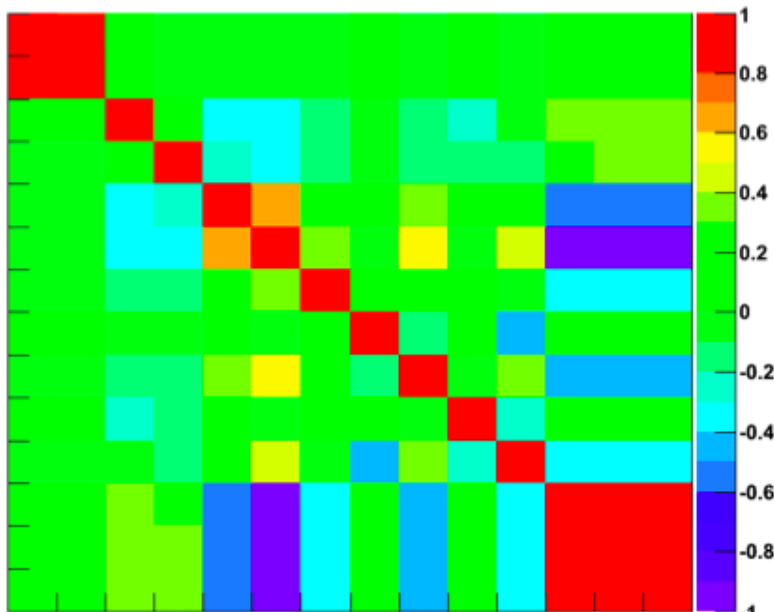
Challenge: build a predictive model to describe simultaneously all the outputs $\{A, B, C, D \dots\}$, provided a corresponding set of inputs.

Illustrative Example

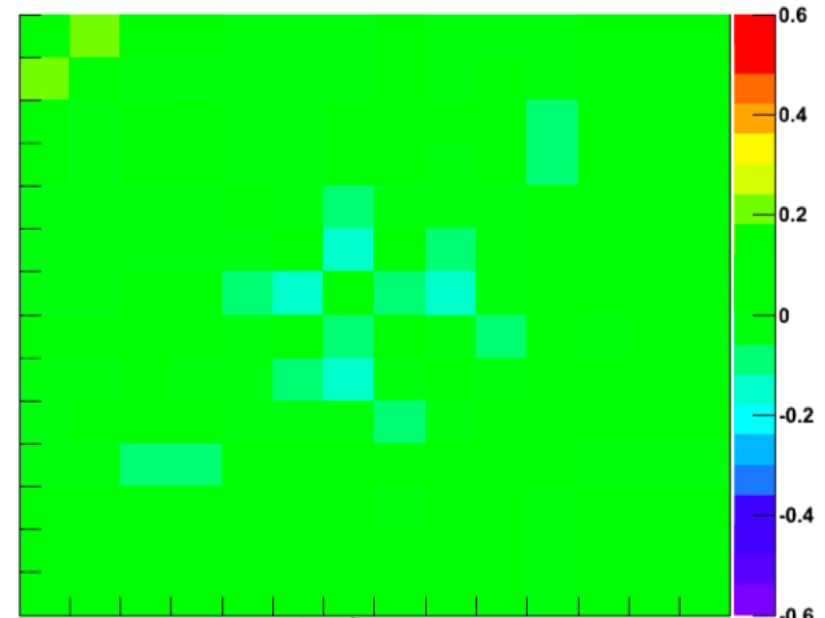


Illustrative Example

Target Correlations



Prediction-Target Difference

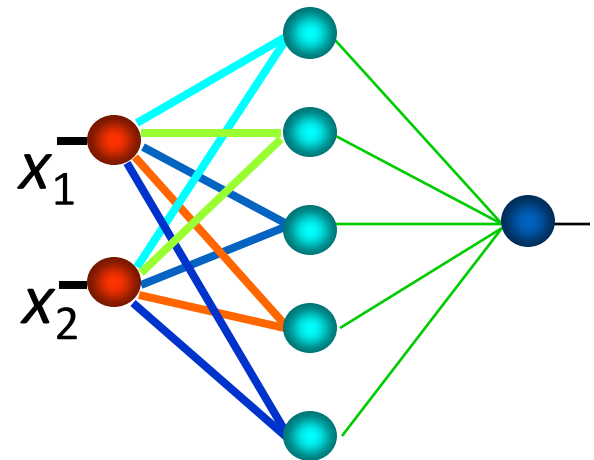
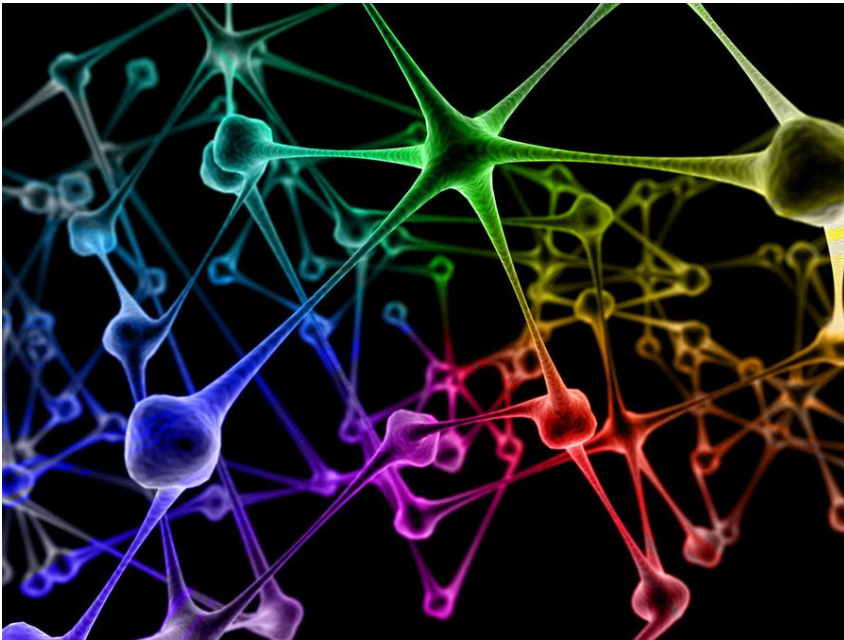


Very close to Zero

Outline

- **Neural Networks**
- **Shallow Learning**

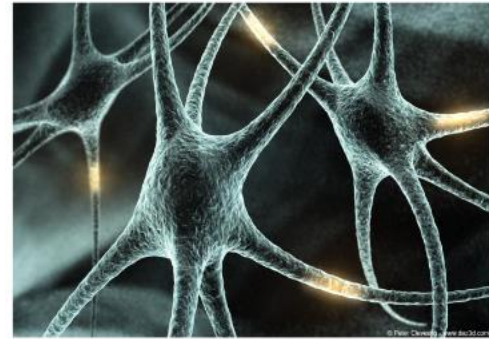
Neural Networks



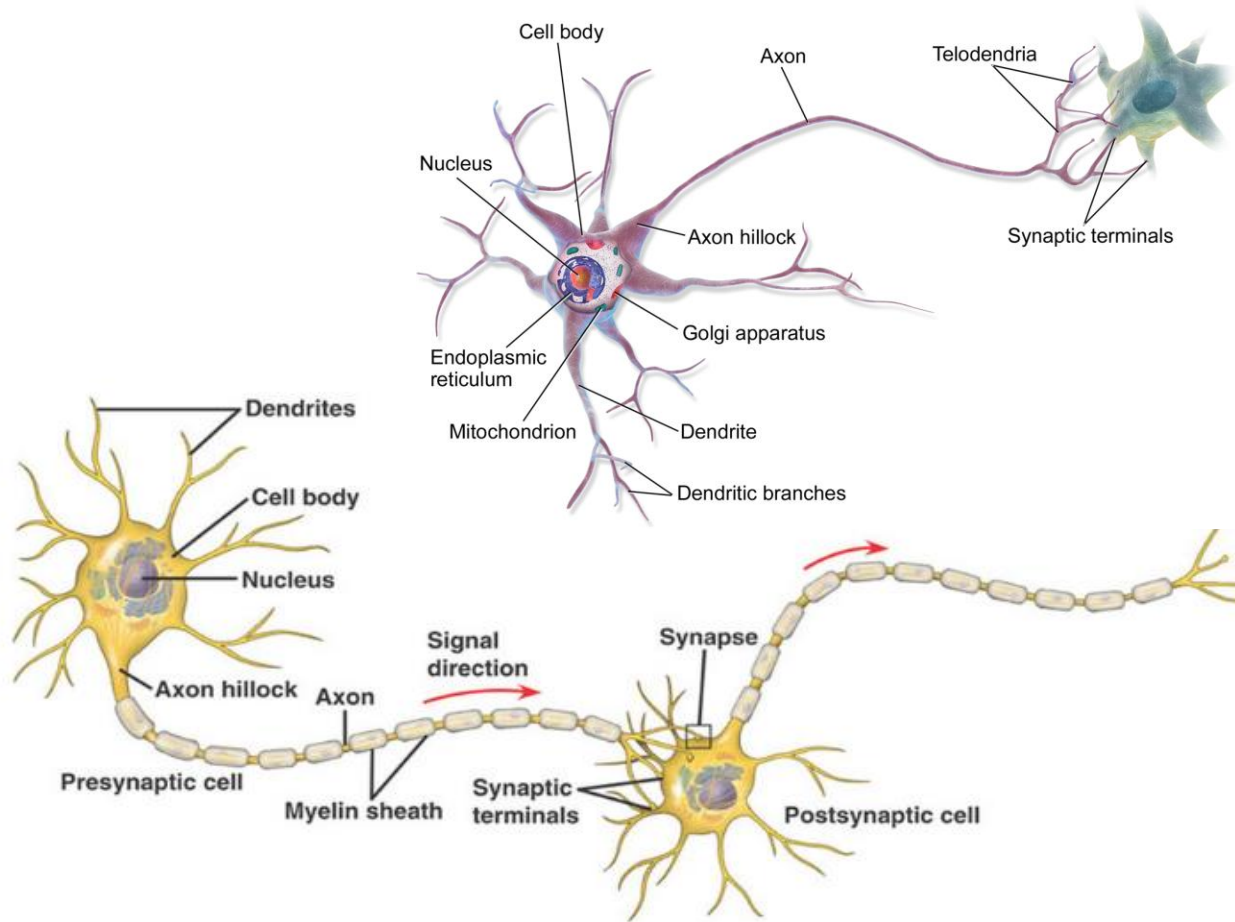
Neural Networks

Biological systems built of very complex webs of interconnected neurons

- Highly connected to other neurons
- Performs computations by combining signals from other neurons
- Outputs of these computations may be transmitted to one or more other neurons



Real Neurons



Neural Network

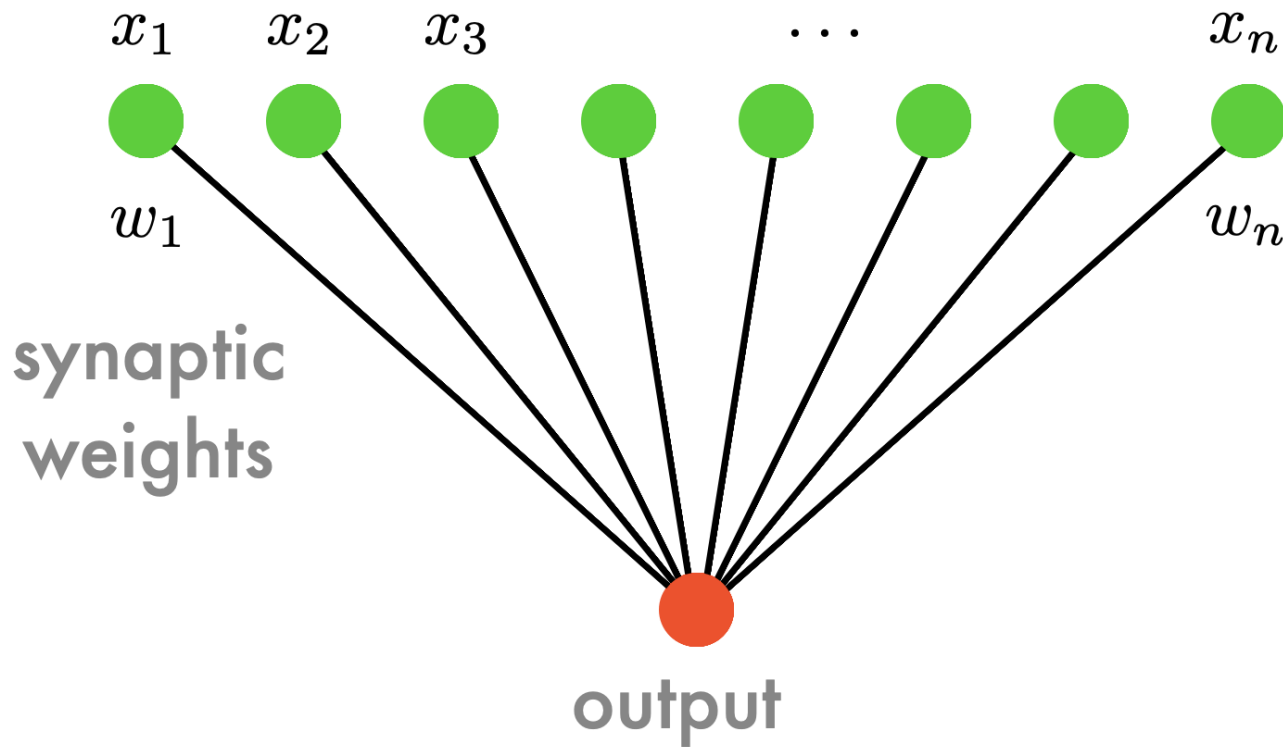
Humans:

- Neuron switching time $\sim 0.001\text{s}$
- Number of neurons $\sim 10^{10}$
- Connections per neuron $\sim 10^4$
- Scene recognition time $\sim 0.1\text{s}$

ANNs:

- Many **threshold switching** units
- Many **weighted interconnections**
- Highly **parallel, distributed** processes

Perceptron



Frank Rosenblatt, 1957

Perceptron Learning

Perceptrons

- **Threshold Logic Unit/Step Function**
 - Linear combination of inputs
 - Classify above threshold
- **Hebbian Learning Rule:**
 - “Fire together, wire together”
- **Linear decision boundary**
 - XoR Classification Problem Minsky and Papert 1969
- **Stack into MultiLayer Perceptrons (MLPs)**