



# Machine Learning

BY SON德拉 IVERSON

# Machine Learning

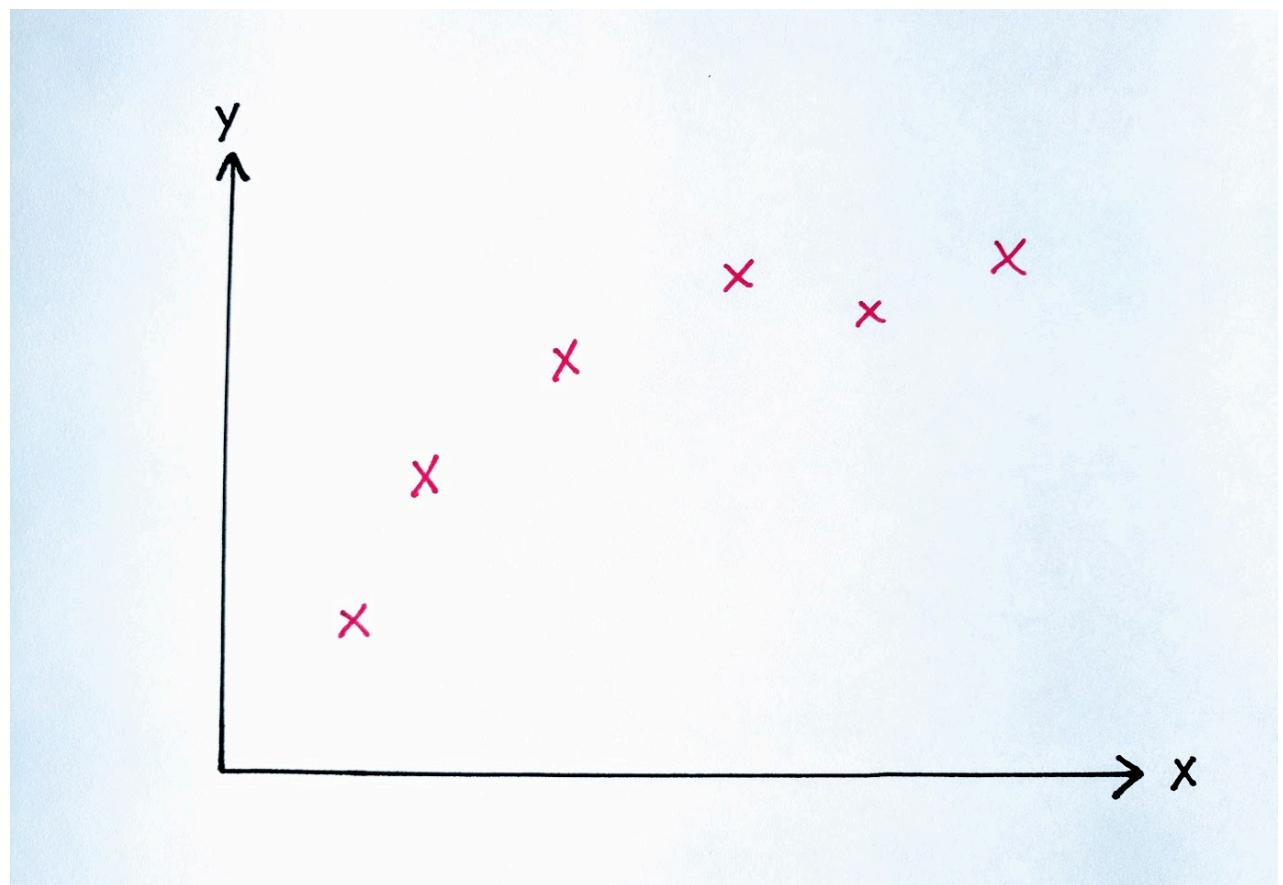
- ▶ The field of study that gives computers the ability to learn without being explicitly programmed. – Arthur Samuel
- ▶ Two Main Types of Algorithms
  - ▶ Supervised Learning – Given inputs and desired output, learn general rule
  - ▶ Unsupervised Learning – No labels given, try to find structure in data

# Machine Learning Algorithms

- ▶ Supervised Learning
  - ▶ **Linear Regression**
  - ▶ Logistic Regression
  - ▶ Neural Networks
- ▶ Unsupervised Learning
  - ▶ Clustering
  - ▶ Anomaly Detection
  - ▶ Recommender Systems

# Linear Regression

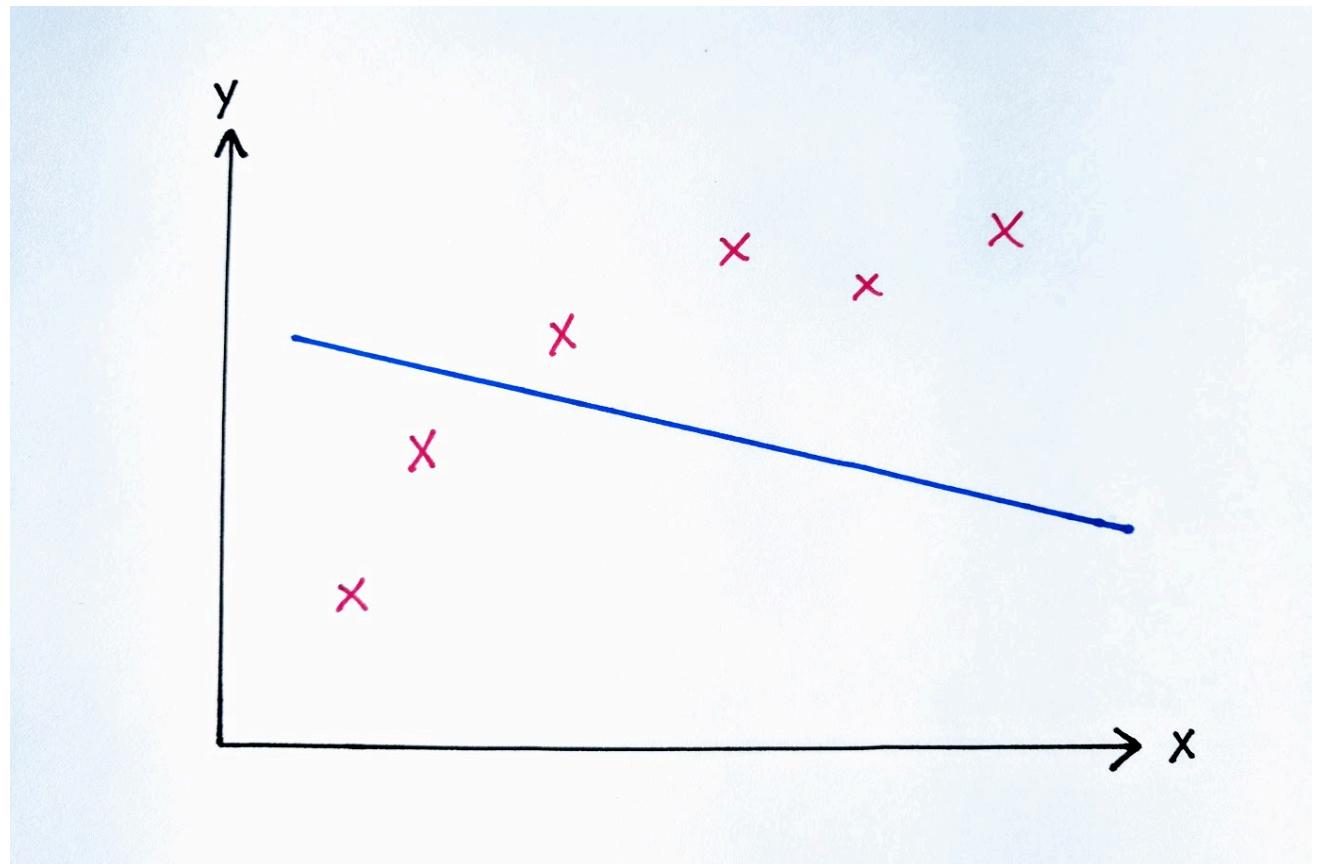
- ▶ For continuous data
- ▶ Example: predicting the selling price of a house based on the size
- ▶  $x$  is called a feature, input, predictor or independent variable
- ▶  $y$  is called a response, output, target or dependent variable



# Linear Regression

$$h_{\theta}(x) = \theta_0 + \theta_1 x$$

- ▶ Hypothesis Function is the equation of a line

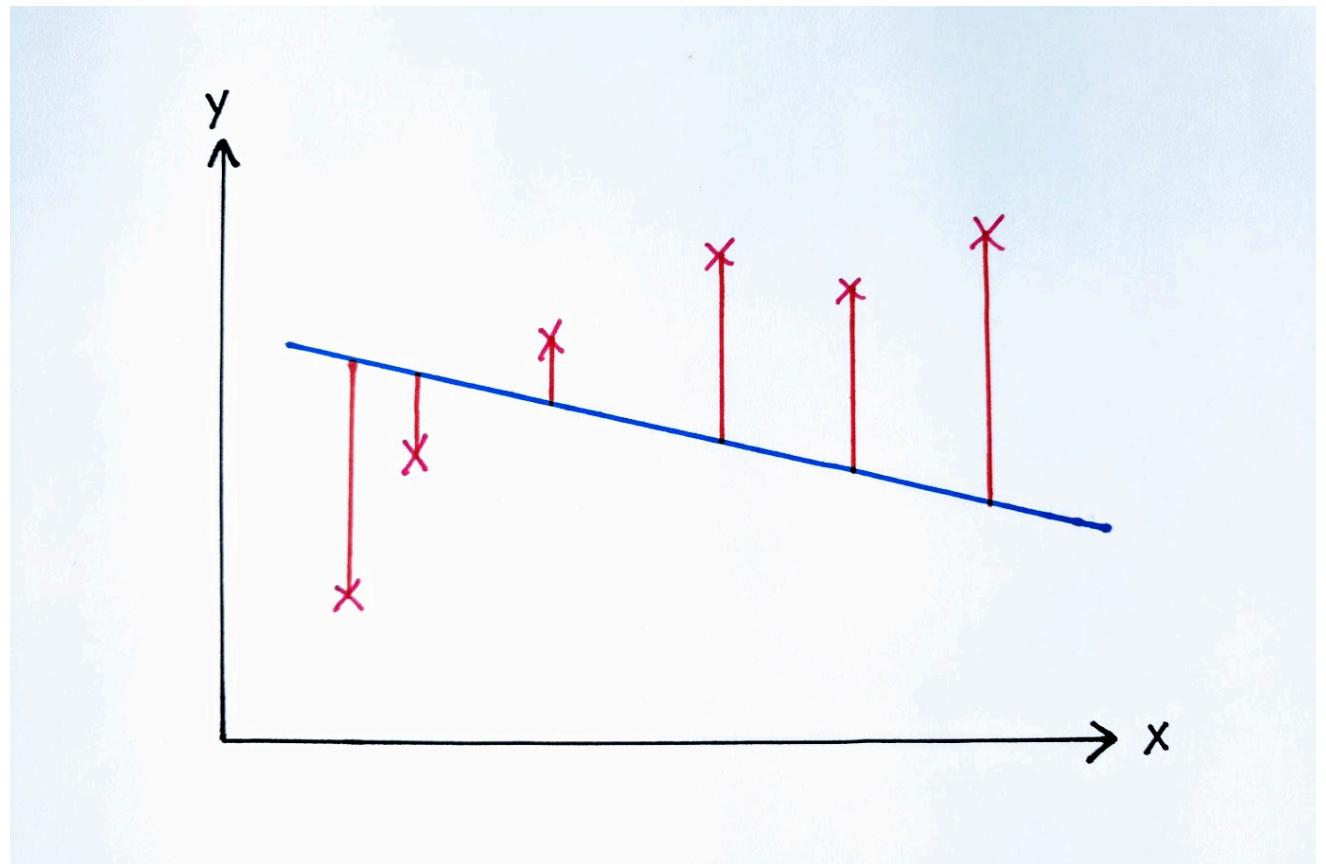


# Linear Regression

- ▶ Calculate cost (i.e. error)

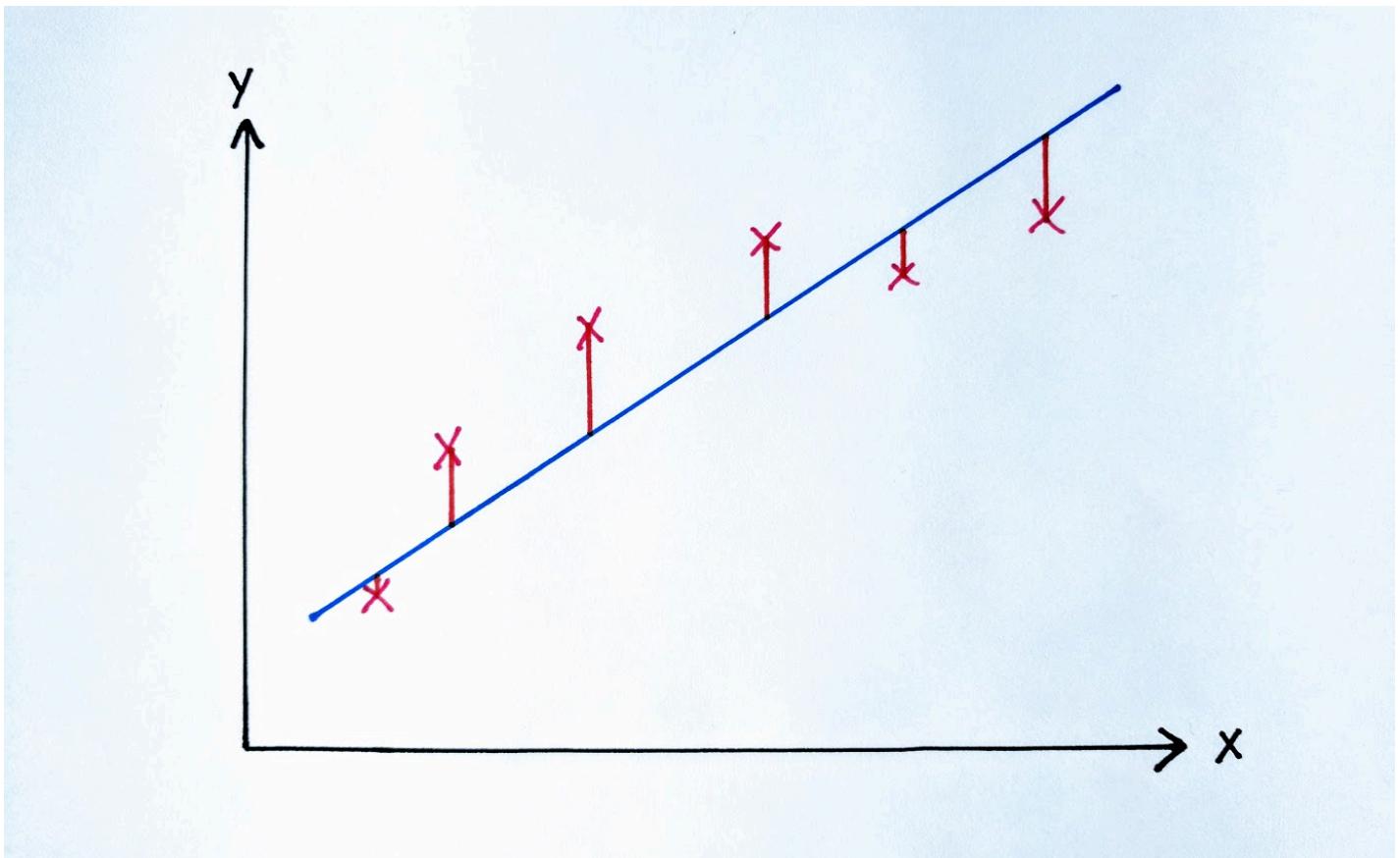
$$\sum_{i=1}^m (h_\theta(x_i) - y_i)^2$$

- ▶ Residual Sum of Squares (RSS)



# Linear Regression

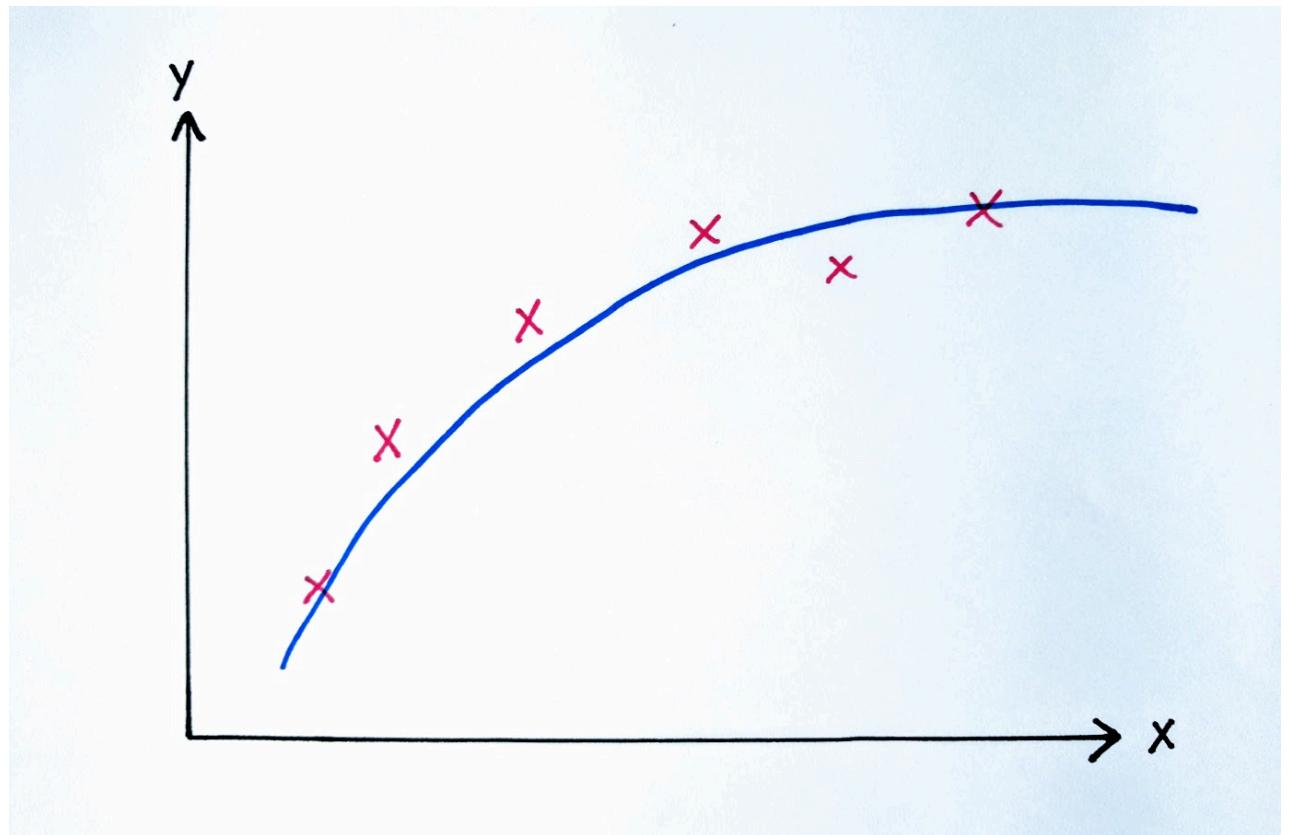
- ▶ Find the equation that minimizes the error for the training data
- ▶ This gives the theta values for the hypothesis function
- ▶ Use that equation to compute the output for new input values



# Linear Regression

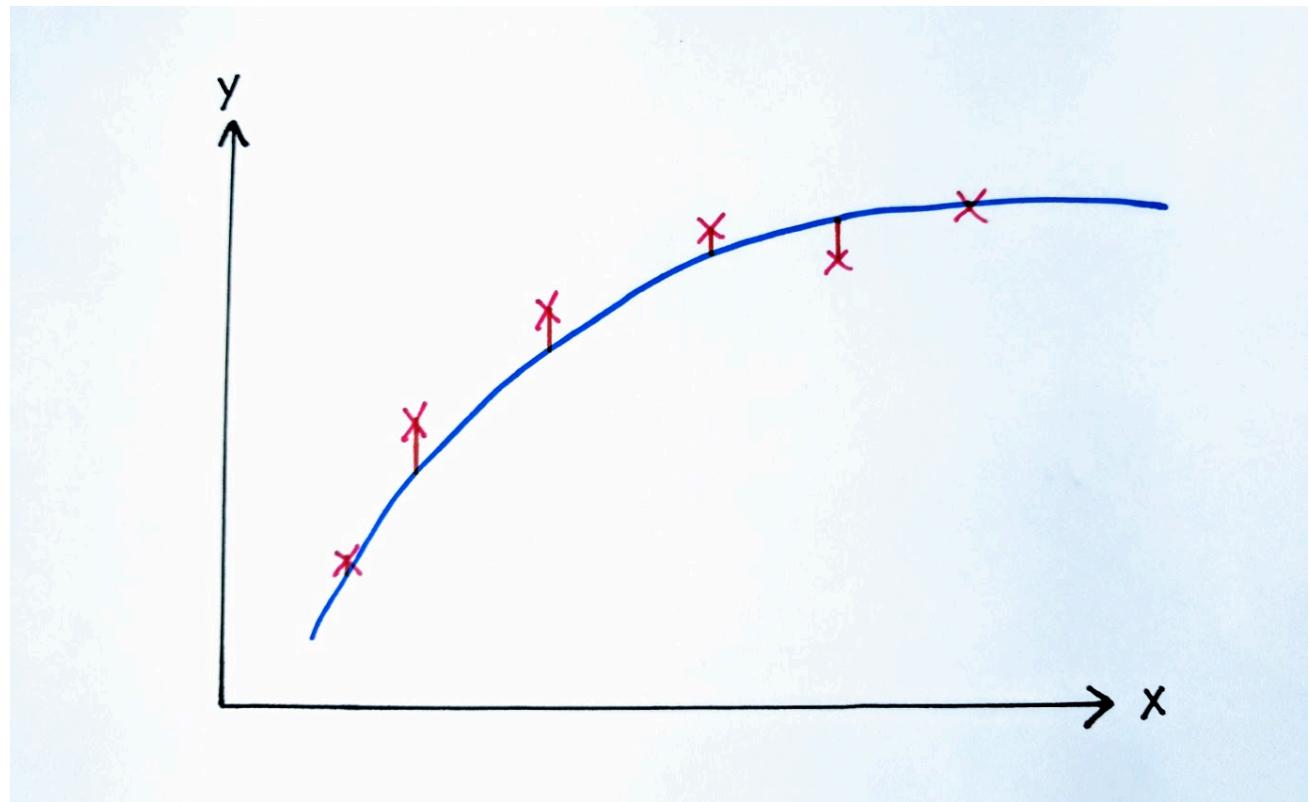
- ▶ Can use higher order terms for quadratic, cubic, etc. equations

$$h_{\theta}(x) = \theta_0 + \theta_1 x + \theta_2 x^2$$



# Linear Regression

- ▶ Calculate and minimize cost with same algorithm
- ▶ Also works in multiple dimensions using multiple features
- ▶ Determines weights (positive or negative) for all features

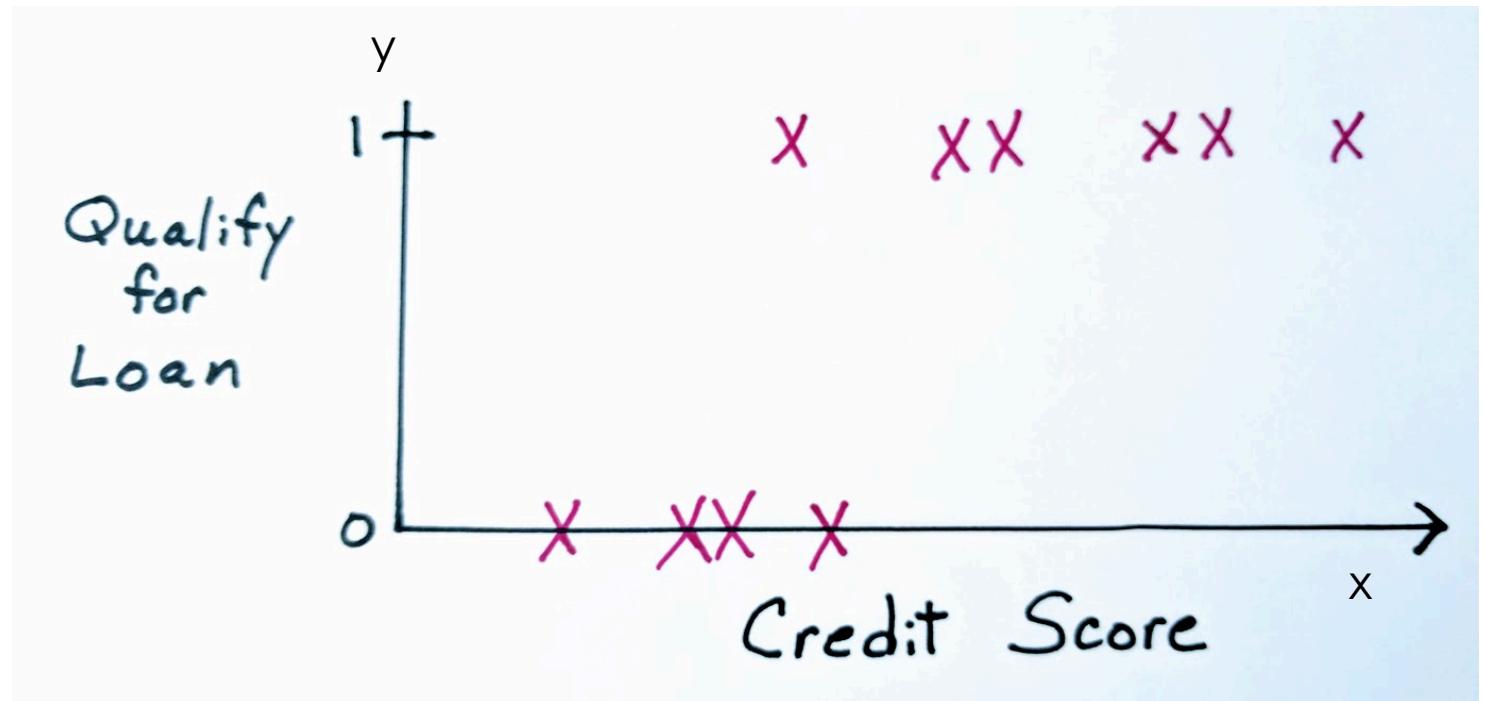


# Machine Learning Algorithms

- ▶ Supervised Learning
  - ▶ Linear Regression
  - ▶ **Logistic Regression**
  - ▶ Neural Networks
- ▶ Unsupervised Learning
  - ▶ Clustering
  - ▶ Anomaly Detection
  - ▶ Recommender Systems

# Logistic Regression

- ▶ Binary Classification Problem
- ▶ Model probability of being in particular class
- ▶  $P(y=1 | x)$



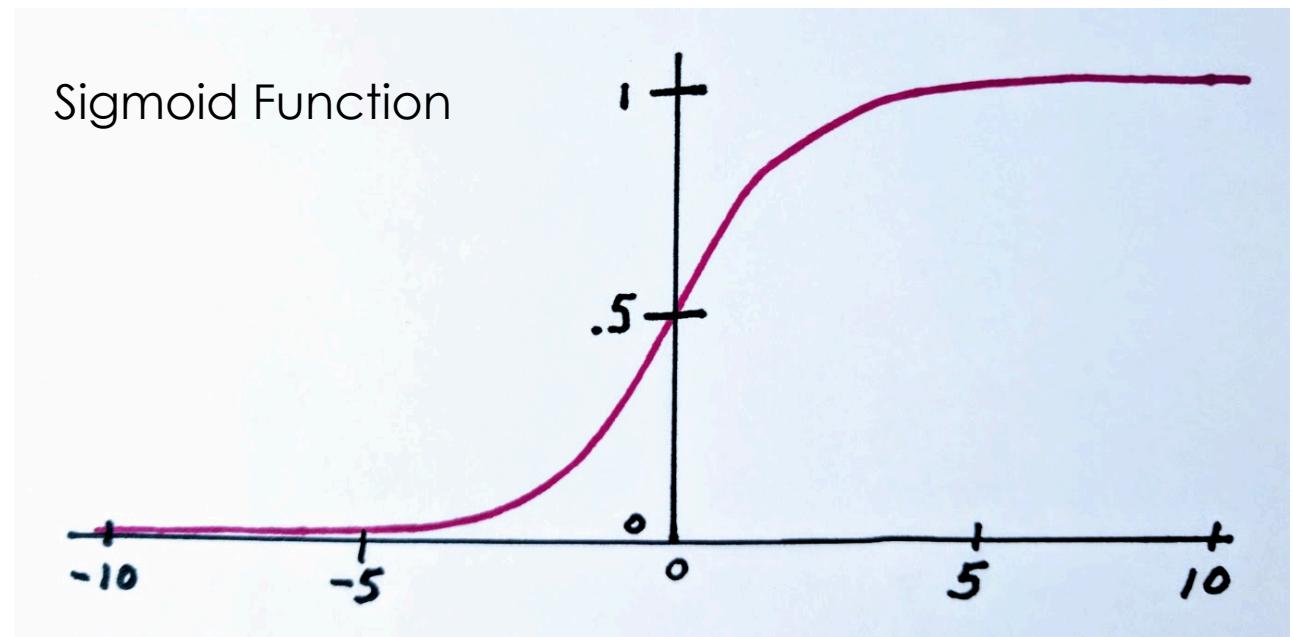
# Logistic Regression

$$h_{\theta}(x) = g(\theta^T x)$$

$$z = \theta^T x$$

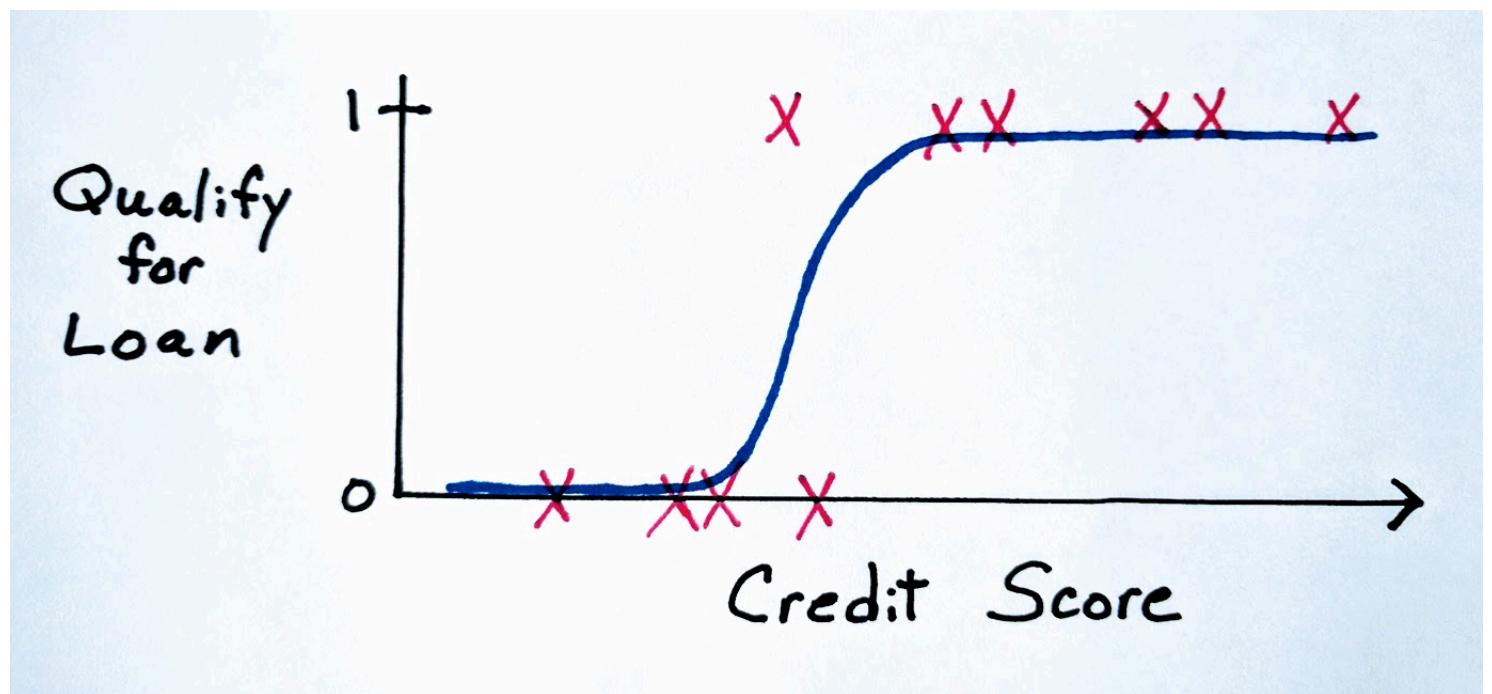
$$g(z) = \frac{1}{1 + e^{-z}}$$

$$\theta^T x = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_n x_n$$



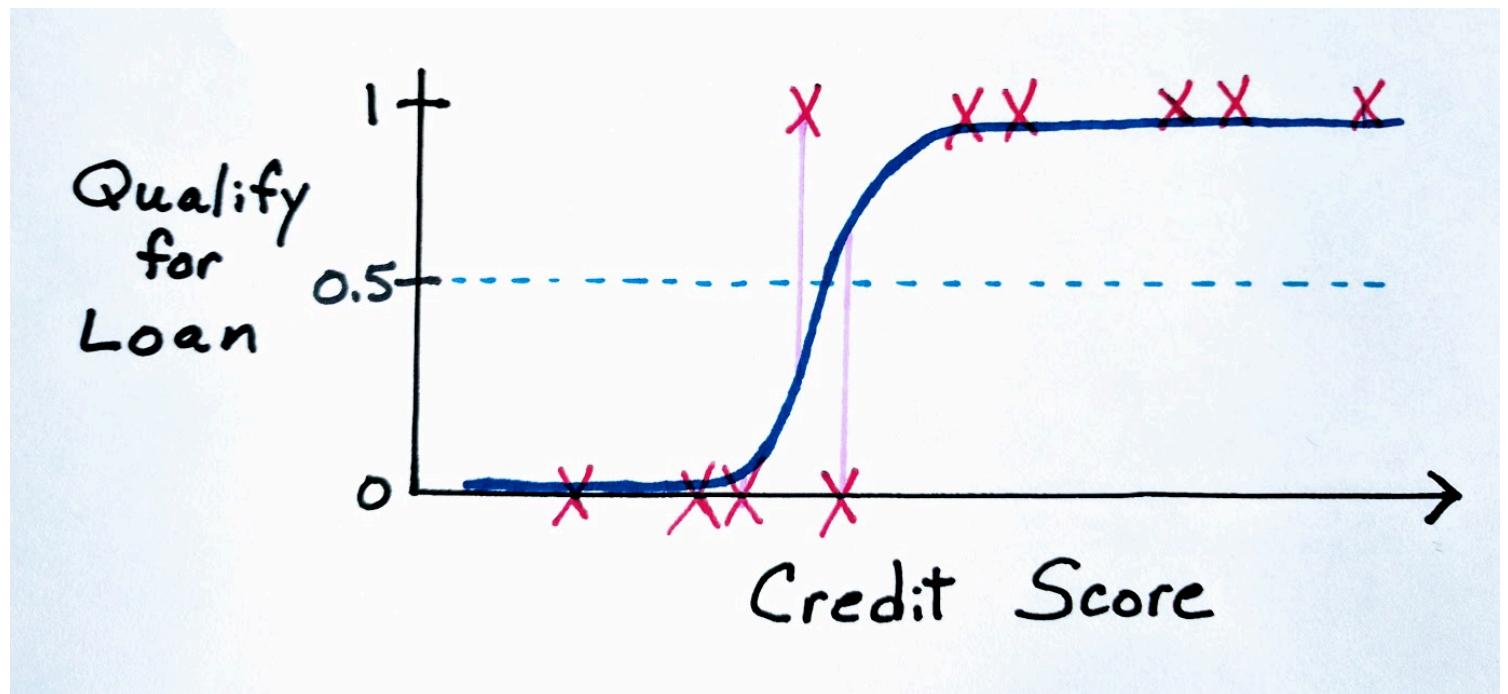
# Logistic Regression

- ▶ Find best fit of sigmoid function to test data
- ▶ Estimate coefficients using Maximum Likelihood Estimate



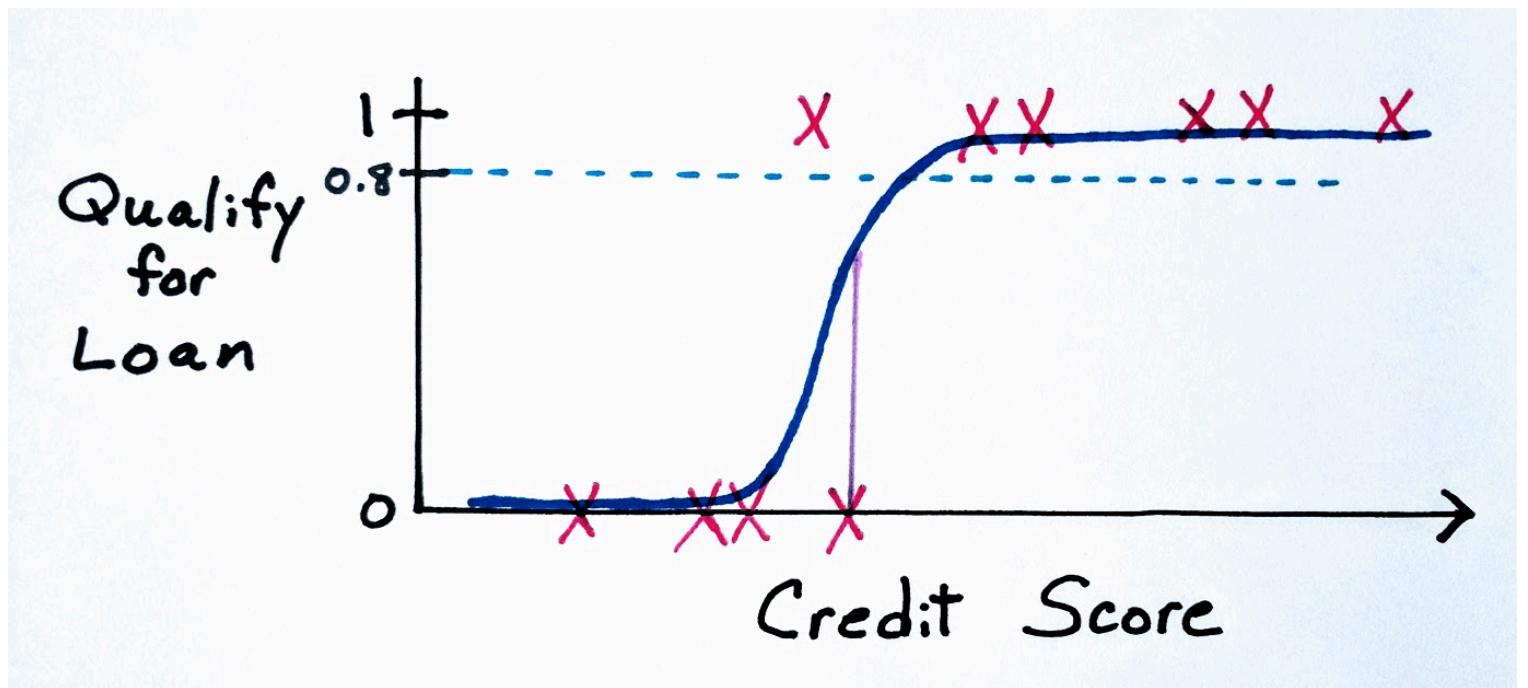
# Logistic Regression

- ▶ If probability  $\geq 0.5$ , output = 1 (true)
- ▶ If probability  $< 0.5$ , output = 0 (false)



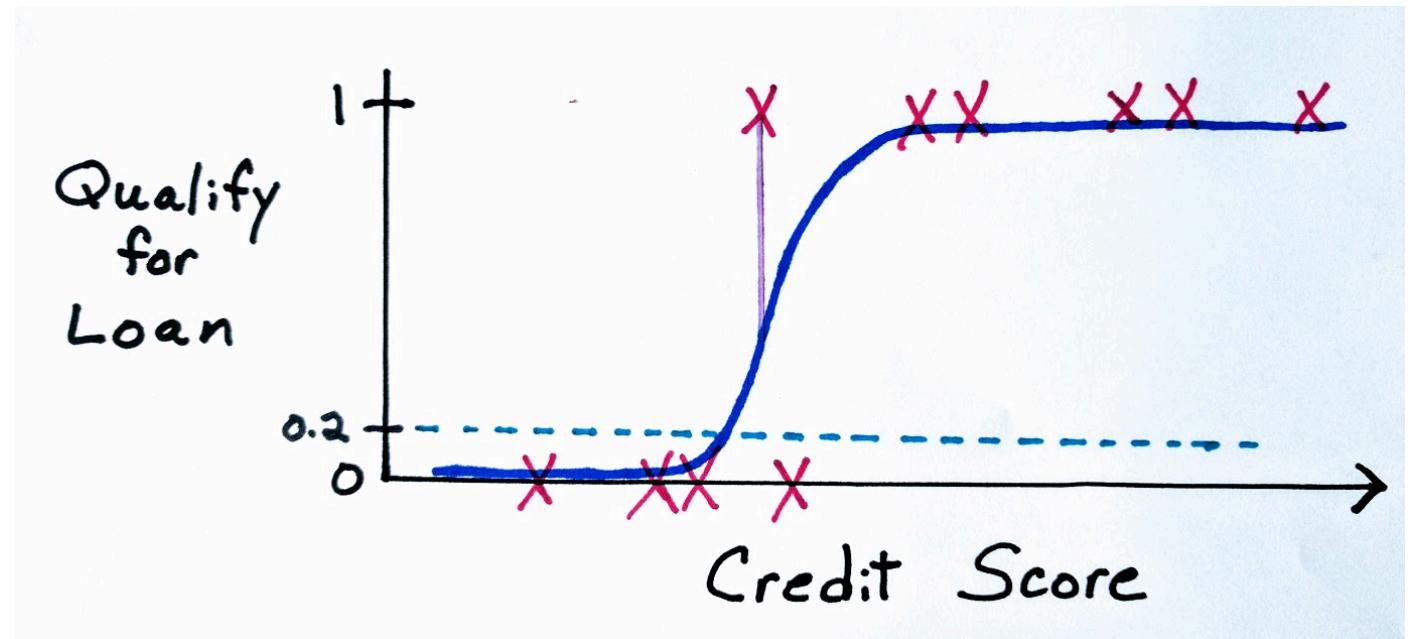
# Logistic Regression

- ▶ Decision Boundary
- ▶ Use higher value ( $> 0.5$ ) for higher probability of being true



# Logistic Regression

- ▶ Decision Boundary
- ▶ Use smaller value (< 0.5) to include more values as true

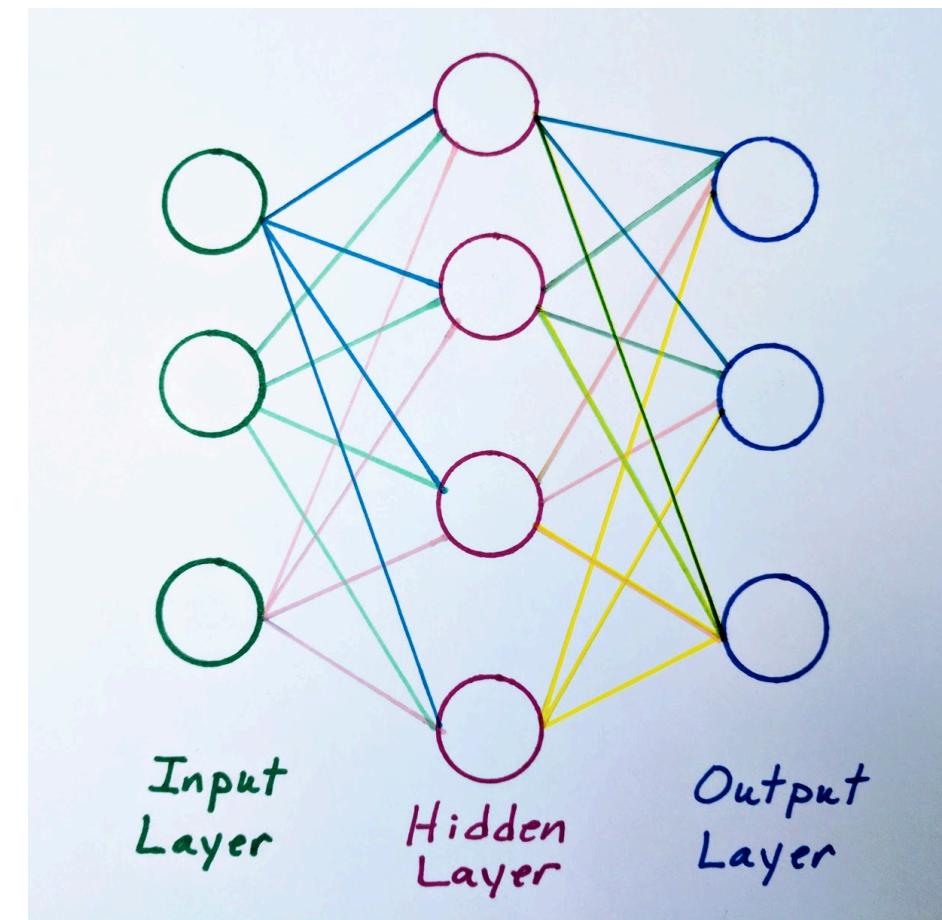


# Machine Learning Algorithms

- ▶ Supervised Learning
  - ▶ Linear Regression
  - ▶ Logistic Regression
  - ▶ **Neural Networks**
- ▶ Unsupervised Learning
  - ▶ Clustering
  - ▶ Anomaly Detection
  - ▶ Recommender Systems

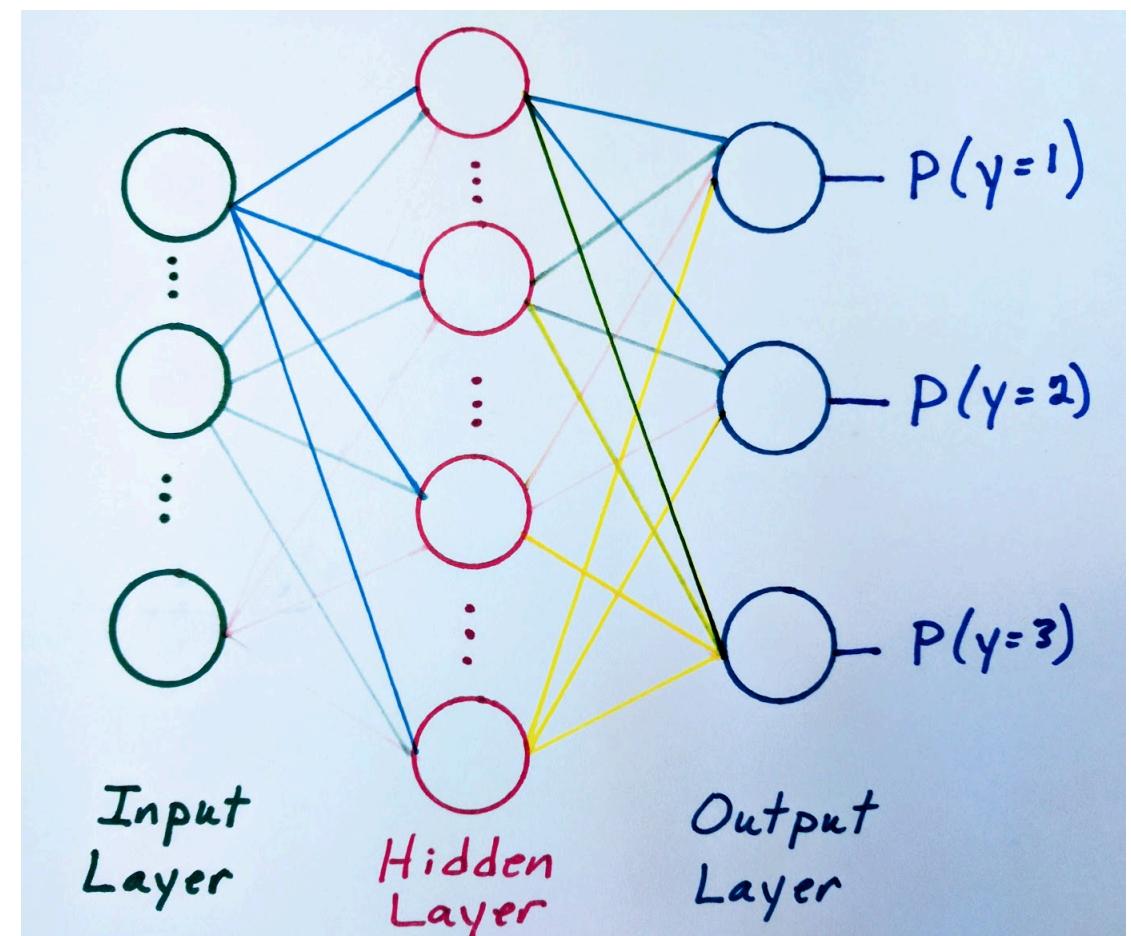
# Neural Networks

- ▶ Can learn complex relationships
- ▶ Modeled on neurons in the brain
- ▶ Each node modeled as a sigmoid function or a Rectified Linear Unit (ReLU), etc.



# Neural Networks

- ▶ Example: Identifying the digits 1, 2, and 3
- ▶ Input layer: pixel values of image of a digit
- ▶ Output layer: probability of digit being 1, 2, and 3



# Neural Networks

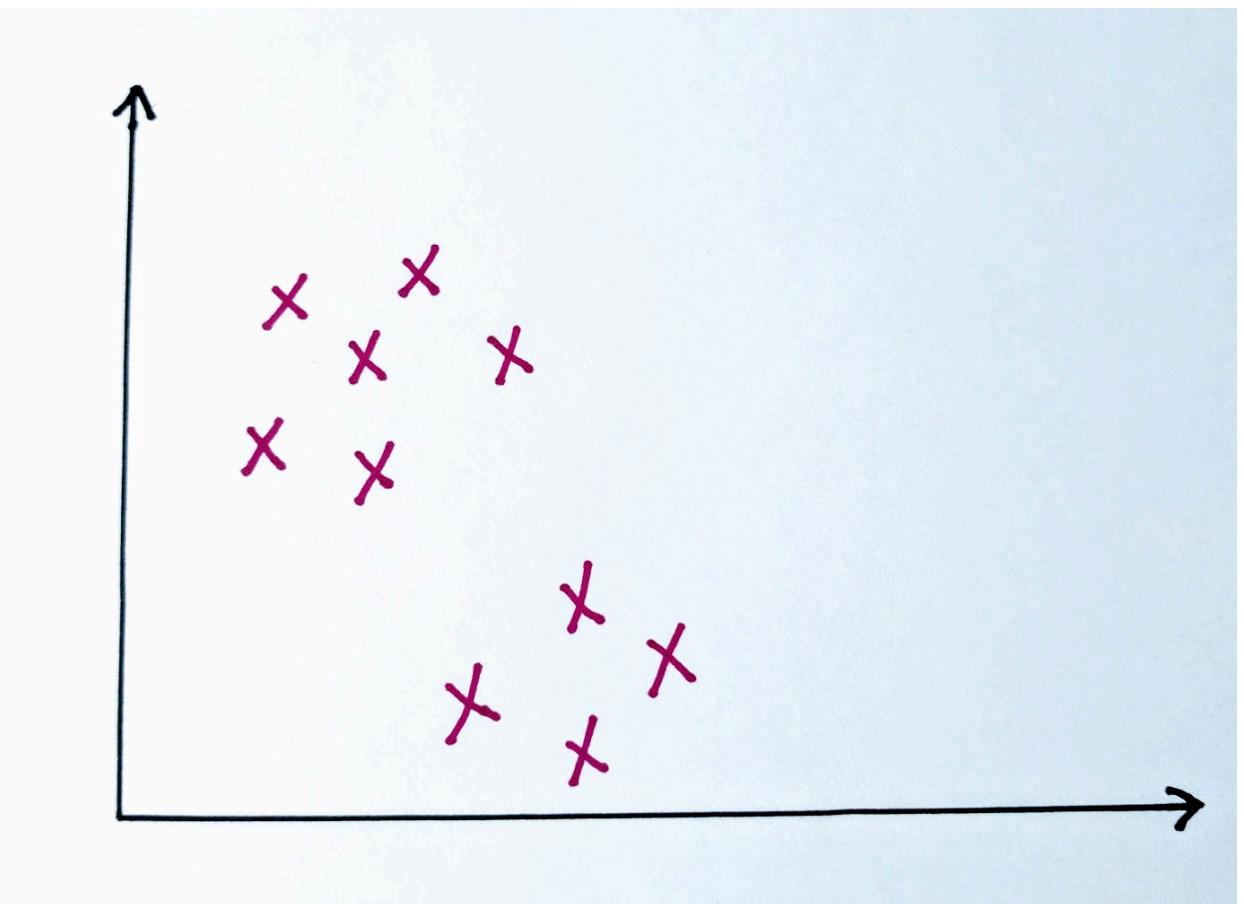
- ▶ Neural network trained similar to logistic regression
- ▶ Includes a back propagation step where the error is distributed to the previous layer

# Machine Learning Algorithms

- ▶ Supervised Learning
  - ▶ Linear Regression
  - ▶ Logistic Regression
  - ▶ Neural Networks
- ▶ Unsupervised Learning
  - ▶ **Clustering**
  - ▶ Anomaly Detection
  - ▶ Recommender Systems

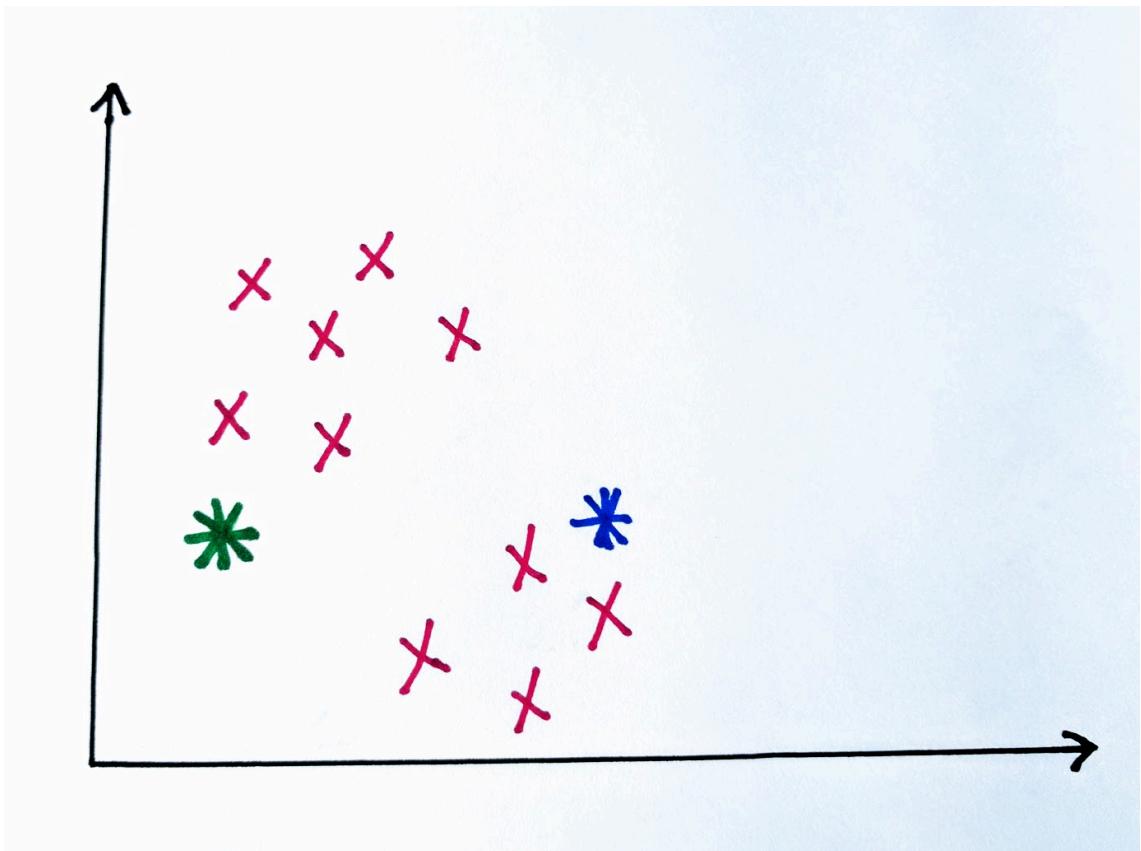
# Clustering

- ▶ Unsupervised – data is unlabeled
- ▶ K-Means Algorithm
  - ▶  $K$  = number of clusters/groups
  - ▶ Mean = average



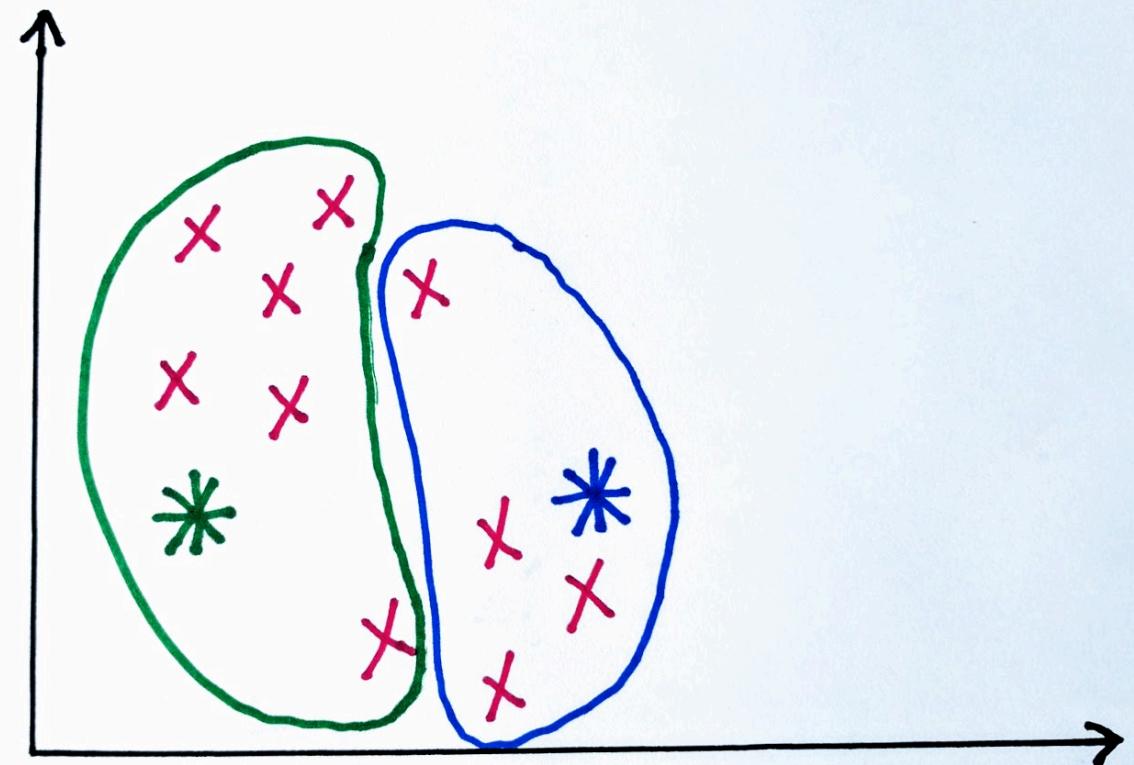
# Clustering, K-Means Algorithm

- ▶ Randomly pick K points
- ▶ Use for the centers of clusters



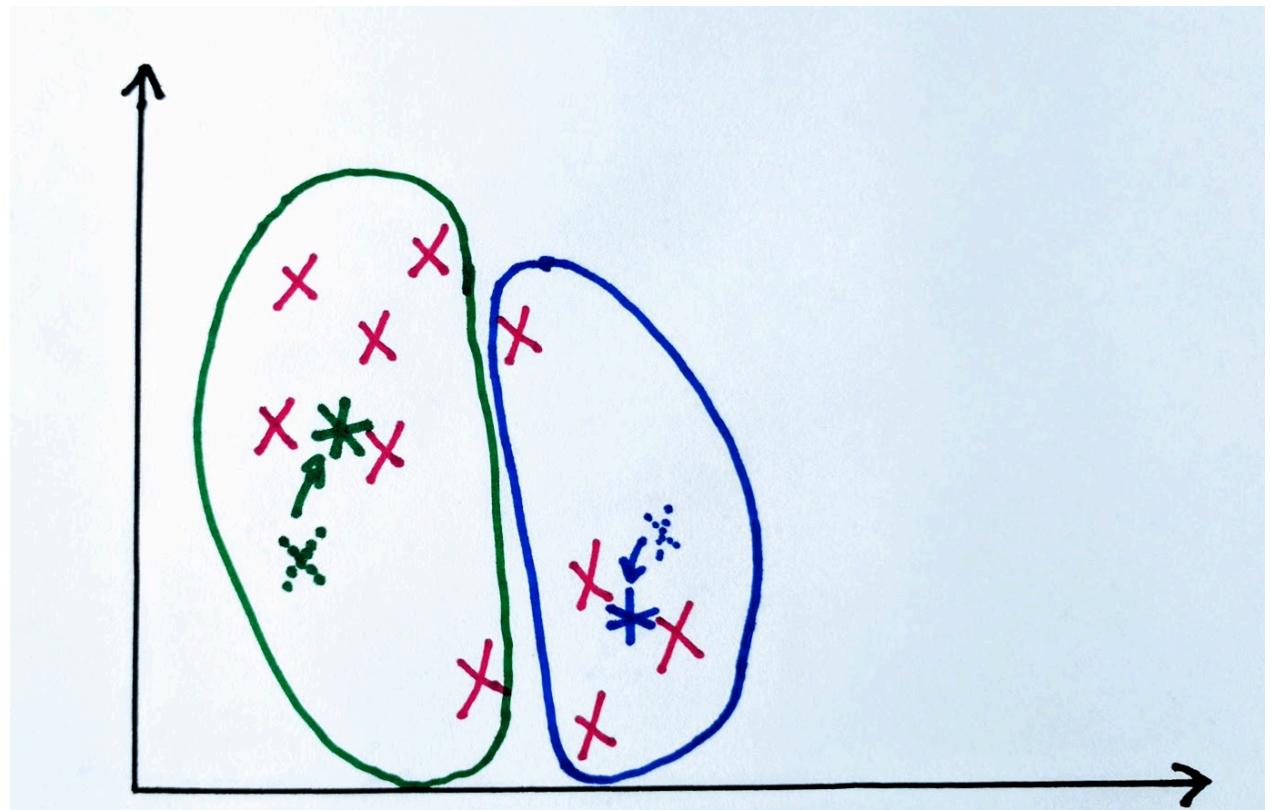
# Clustering, K-Means Algorithm

- ▶ Assign each point to closest cluster center



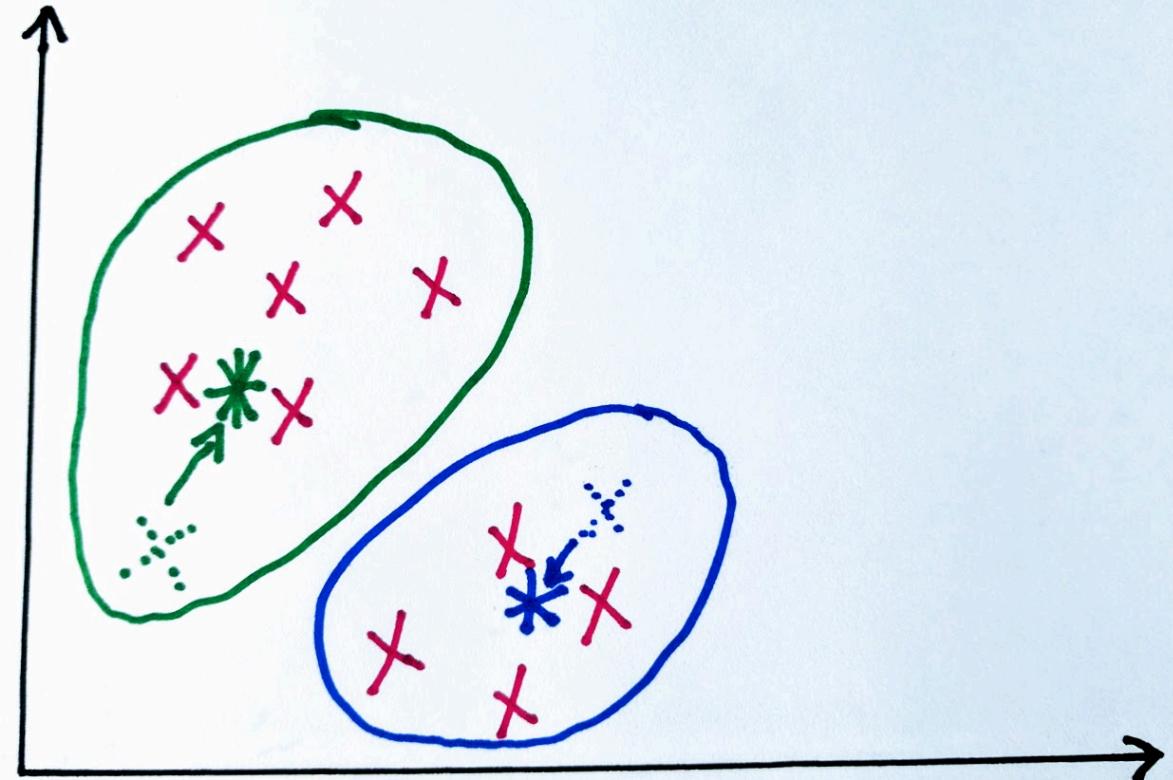
# Clustering, K-Means Algorithm

- ▶ Calculate mean (average) for all points in each cluster
- ▶ Move cluster center to new mean



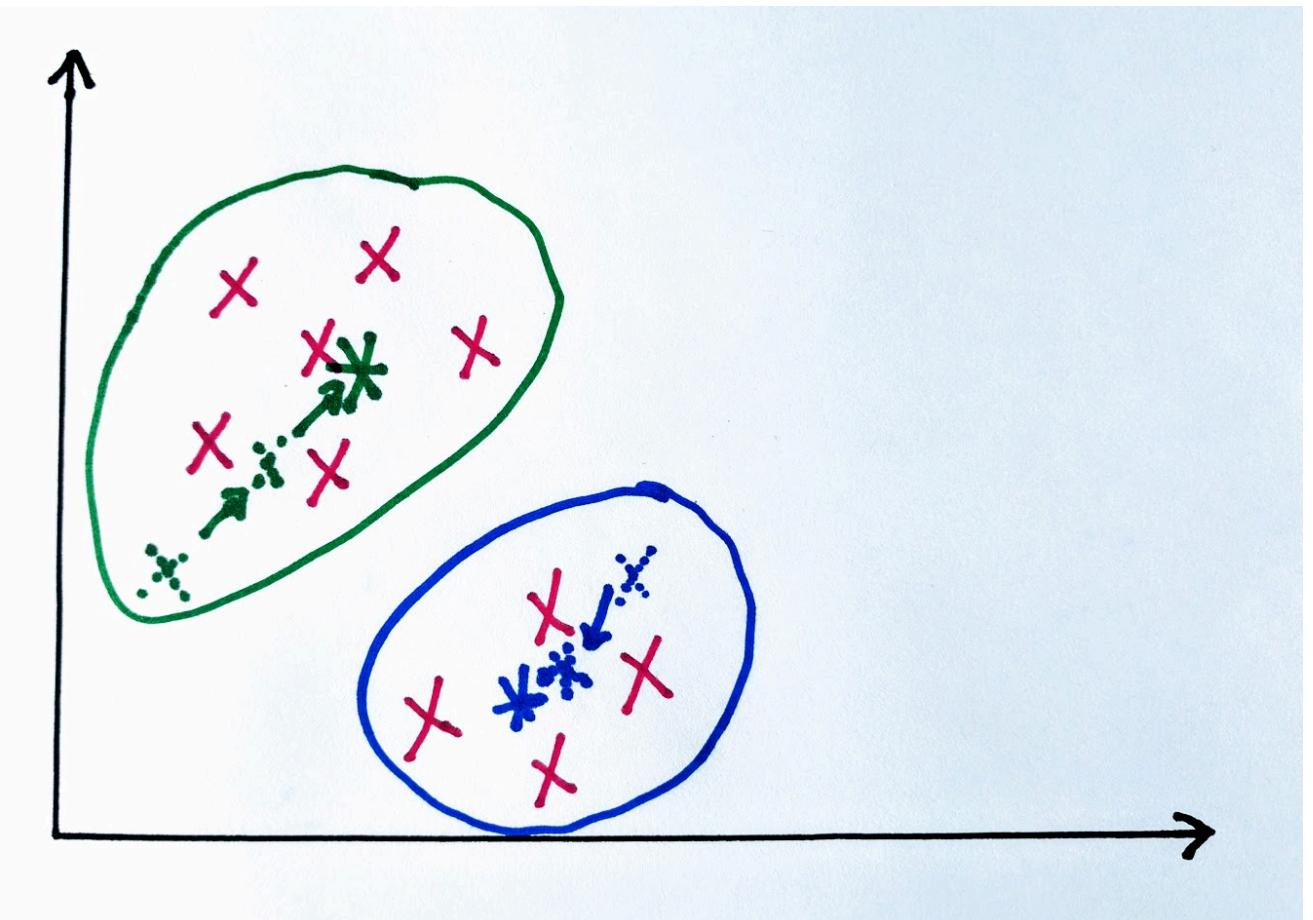
# Clustering, K-Means Algorithm

- ▶ Repeat the process:
- ▶ Assign each point to new closest cluster center



# Clustering, K-Means Algorithm

- ▶ Continue to repeat:
- ▶ Calculate new means
- ▶ Use new means as cluster centers
- ▶ Assign points to closest center



# Clustering, K-Means

- ▶ May not converge to optimal clustering
- ▶ Try multiple random starting points
- ▶ Use actual data points for random starting points

# Machine Learning Algorithms

- ▶ Supervised Learning
  - ▶ Linear Regression
  - ▶ Logistic Regression
  - ▶ Neural Networks
- ▶ Unsupervised Learning
  - ▶ Clustering
  - ▶ **Anomaly Detection**
  - ▶ Recommender Systems

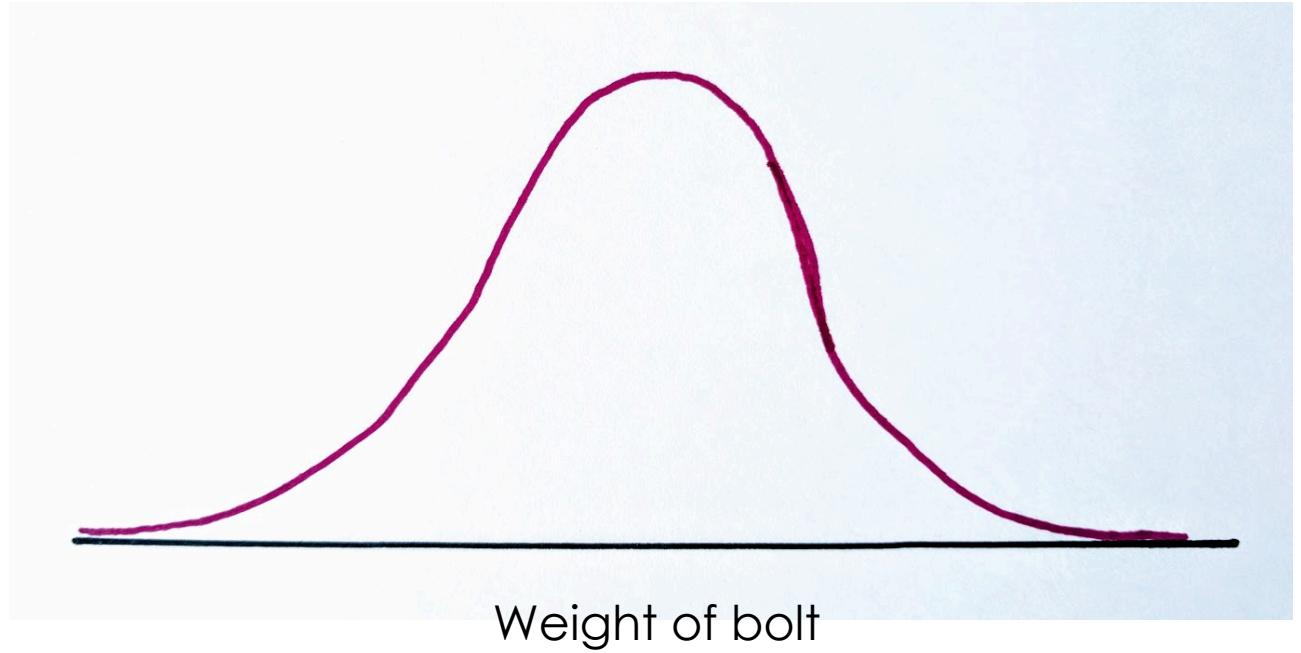
# Anomaly Detection

- ▶ Detect abnormal/anomalous items
  - ▶ Defective products from assembly line
  - ▶ Attacker trying to access computer system
  - ▶ A failing computer within a data center
- ▶ If have many examples of failure, can use logistic regression
- ▶ If failure can take many forms and/or only have a few examples of failure, can use Anomaly Detection

# Anomaly Detection

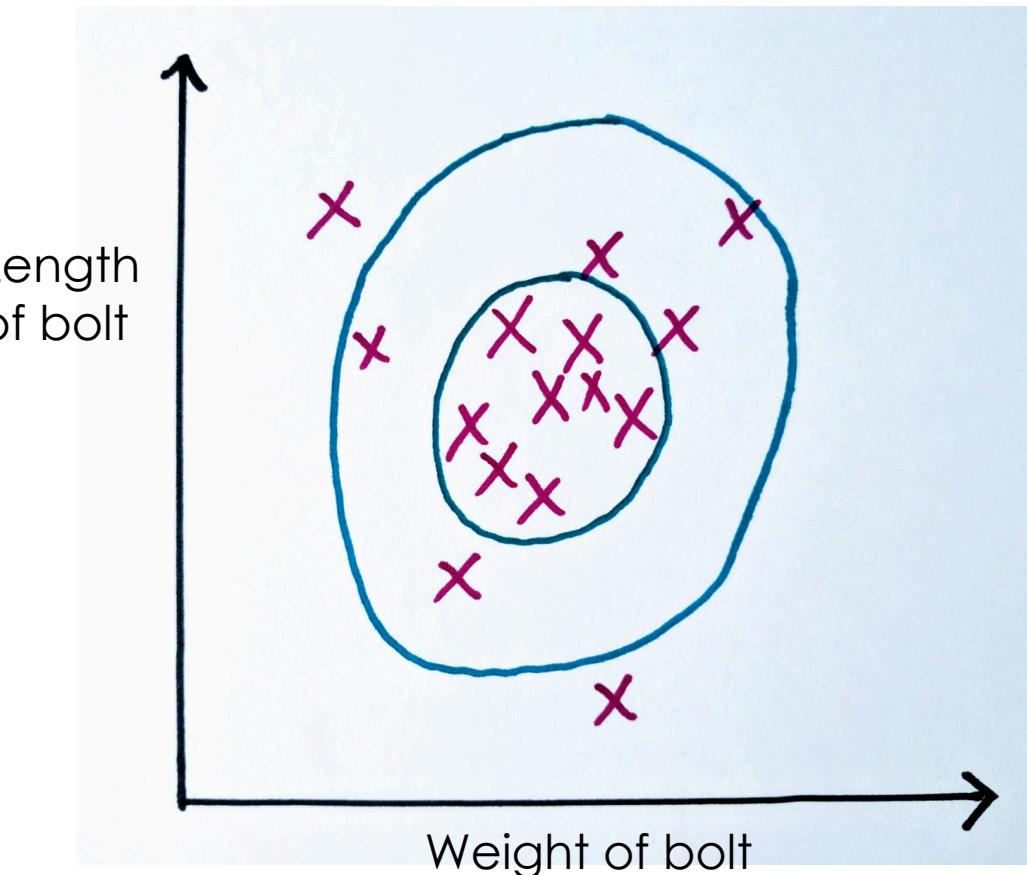
- ▶ Uses Gaussian Distribution,  
i.e. a normal distribution

$$\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$



# Anomaly Detection

- ▶ Calculate mean, mu
- ▶ Calculate standard deviation, sigma
- ▶ Determine epsilon (threshold between normal and anomalous data) with training data



# Machine Learning Algorithms

- ▶ Supervised Learning
  - ▶ Linear Regression
  - ▶ Logistic Regression
  - ▶ Neural Networks
- ▶ Unsupervised Learning
  - ▶ Clustering
  - ▶ Anomaly Detection
- ▶ **Recommender Systems**

# Recommender Systems

- ▶ Content-Based Filtering
  - ▶ Characterize all products by all features
  - ▶ Example with one feature:
    - ▶ Movie A has a high value for the feature “action movie”
    - ▶ User B likes action movies
    - ▶ Recommend Movie A to User B
  - ▶ Uses linear regression to find the movies with features are the closest match to the user's stated preferences

# Recommender Systems

## ► Collaborative Filtering

- ▶ Does not analyze the content of the product (such as a movie)
- ▶ Learns the features of all products by using the rankings from all users along with those users' preferences
- ▶ Recommend a product with highest value of feature weights \* user preferences

# Machine Learning

►Thank you