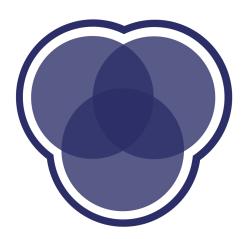
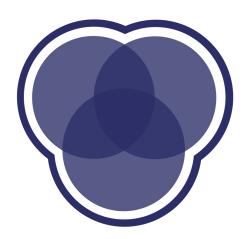


Intro to Machine Learning

Demo



Machine Learning



Machine Learning

According to Wikipedia...

"Machine Learning is a subfield of computer science that gives computers the ability to learn without being explicitly





ML Setup

Hypothesis: Some speculative relationship between the input

space and output space

Input Space: Variable or set of variables(data)

Output Space: Target variable to estimate



Supervised vs Unsupervised

Supervised learning problems...

- Known target variable info
- Positive / Negative examples

Unsupervised learning problems...

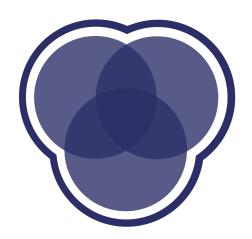
- Unknown target variables
- Difficult to validate





- Supervised learning:
- given $(x_1, y_1), \ldots, (x_n, y_n)$, learn f(x) = y
- Unsupervised learning:
- given x_1, \ldots, x_n , learn patterns or structure
- ▶ Online learning: for i = 1, ..., n, given x_i , predict and observe y_i , learn f(x) = y
- predict and observe y_i , learn T(x) = yActive learning: for i = 1, ..., n,
 - choose x_i , predict and observe y_i , learn f(x) = y
- ▶ **Reinforcement learning:** for i = 1, ..., n, choose x_i , predict y_i , observe reward r_i , learn f(x) = y

Supervised Learning



Setup

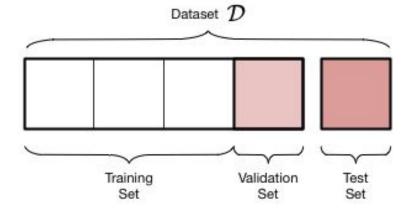
- Training / Validation split
- Feature variable(s)
- Target variable
- Train and Test





Validation Set

- Split data into two sets
- Train model on one and validate on another
- Advantages / Disadvantages?



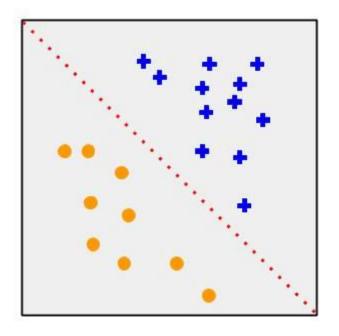


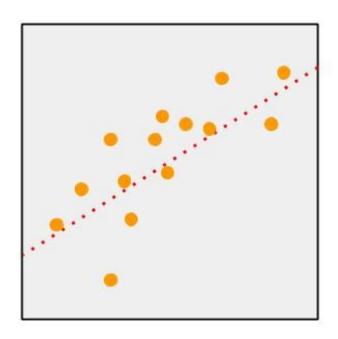
Output Space Properties

- **Continuous -** e.g. temperature, height, probability
- **Discrete -** e.g. car brands, race, Pokémon type, diagnosis



Regression vs Classification







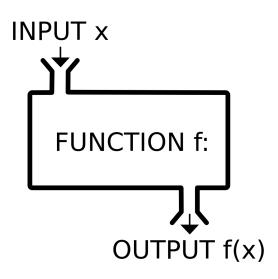
Classification

Regression

What is Learned

Function

Weighted Sum

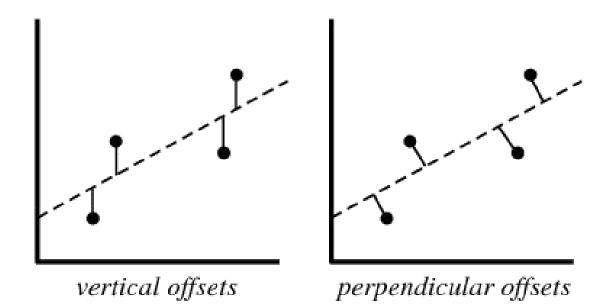


$$y = B_0 + B_I x_I + \dots + B_p x_p$$

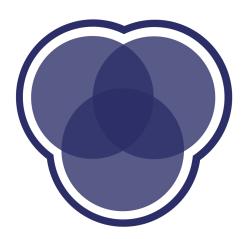


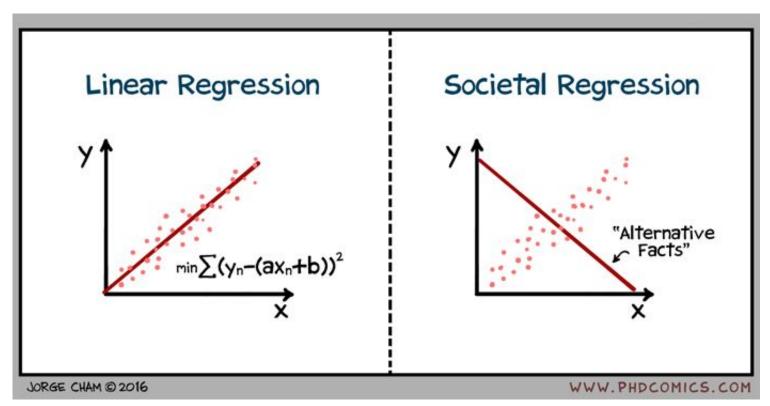
Objective function

- All ML problems are optimization problems
 - Format: Minimize/Maximize Obj in terms of x.
 - Subject to set of constraints
- Objective functions represent assumptions
- Value of objective is an estimation of error

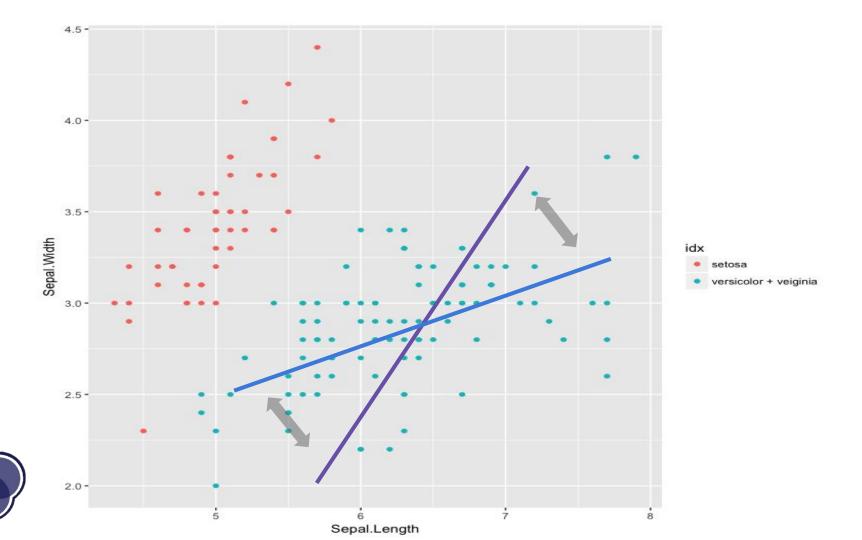


Linear Regression









Linear Regression

$$y = B_0 + B_I x_I + \dots + B_p x_p + \varepsilon$$

What are the assumptions?

- Linear relationship
 - B, the coefficient vector, does not depend on x
- There is an unremovable noise
- This noise is normally distributed about the line

Objective: Least Squares Error (L2)

theoretical
$$\sum_{i=0}^n (y_i - (B_0 + B_1 x_1 + ... + B_p x_p))^2$$
 observed

What does this minimize?

Why this form?



Coming Up

Your problem set: Continue working on Project Part A

Next week: Introduction to classification

See you then!



