# Intro to Classification Algorithms

GHW February 2023

First, let's check in!

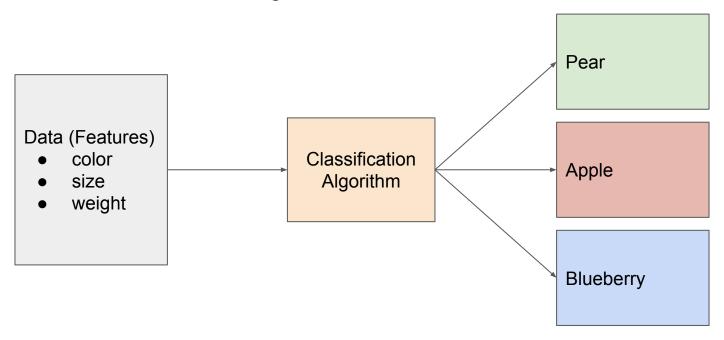
## Most Common Classification Algorithms

- 1. Logistic Regression
- 2. Naive Bayes
- 3. K-Nearest Neighbors (KNN)
- 4. Markov Decision Process (MDP)
- 5. Decision Trees / Random Forest
- 6. K-Means Clustering
- 7. Support Vector Machines

#### What is a Classification Algorithm?

A classification algorithms takes raw data and predicts its category.

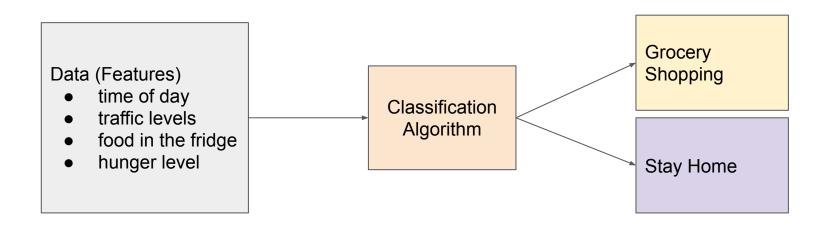
ex: fruit classification algorithm...



#### What is a Classification Algorithm?

A classification algorithms takes raw data and predicts its category.

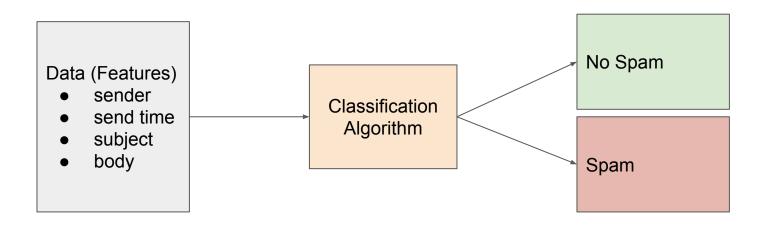
ex: email spam classification algorithm...



#### What is a Classification Algorithm?

A classification algorithms takes raw data and predicts its category.

ex: email spam classification algorithm...



How do we construct an algorithm specific to

our needs?

# Type of Learning

#### Supervised

- Model "learns" from training examples, provided by humans.
- Mostly used for "discrimination" (recognition) tasks.
- Train / test data
  - Typically 80% / 20% split
  - ex: Random Forest (Lecture 3)



## Type of Learning

#### Unsupervised

- Model "learns" from unlabeled data.
- Mostly used for "generative" (imagination) tasks.
- No split between train / test data.

 At the end, we will look at the K-means clustering algorithm in Python for an example! (Note: This is technically a "clustering" algorithm, but it does classify data into categories.)

Age

Eyes

Perspective

Mood

Generative Adversarial Network (GAN)

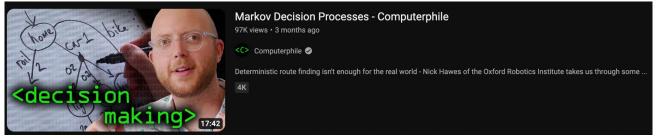


#### Type of Learning

#### Reinforcement Learning

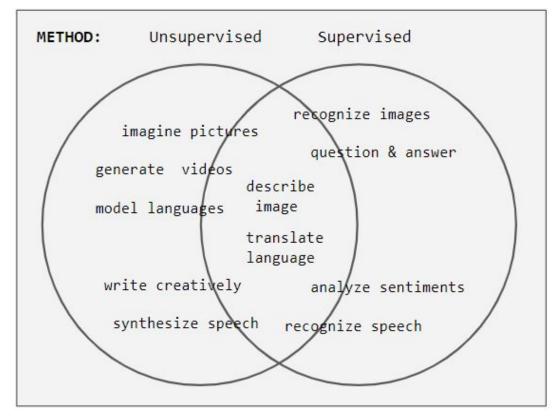
- Model "learns" the best strategy, using a scenario given by humans.
- Scenario: actions + environment
- Mostly used for decision making tasks (ex: robotics).
- Example: Markov Decision Process (MDP) (Lecture 2)





Type of Learning: Select Learning Mode based on

Application!



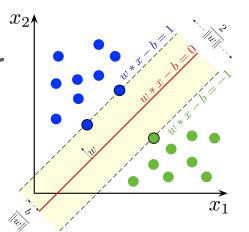
Number of Features

How many variables do you have? (age, height, ....)

Can you reduce the number of features to simplify your model? (Feature Selection)

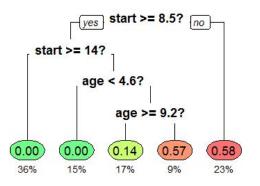
#### More features:

consider a
support vector
machine



#### Less features:

consider a **decision tree** 



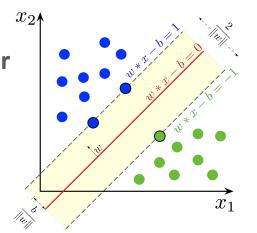
#### Linearity

Is your data linear?

Can you transform your data to make it linear? (We will discuss this later!)

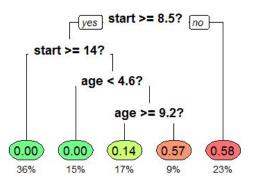
#### Linear:

consider a
support vector
machine



#### Nonlinear:

consider a **decision tree** 



Generally speaking, more training = more accuracy (but not always!)

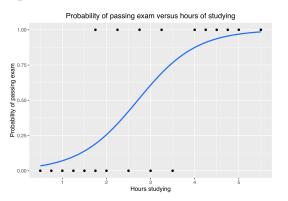
#### **Training Time**

How much data do you have, and how fast is your computer?

#### Faster Training:

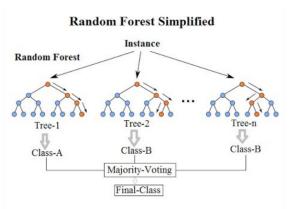
consider a logistic

regression



#### **Slower Training:**

consider a random forest



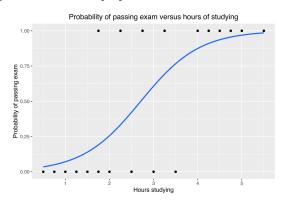
# of Parameters (parameters = model specs. # iterations, error, etc.)

Split into train / validation / test to examine the parameter space.

How much flexibility do you want in your training?

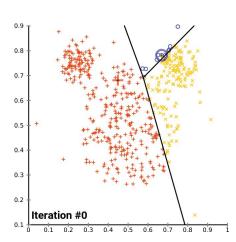
#### Less Parameters:

#### logistic regression (4)



#### More Parameters:

# K-means clustering (8)

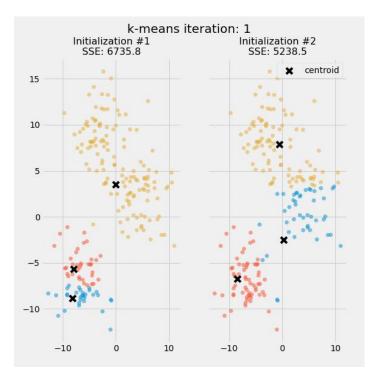


10-minute break

then: K-means Clustering!

## What is K-Means Clustering

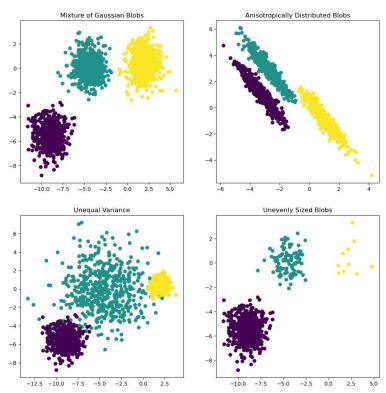
- Clustering algorithm that categories data into "k" groups.
- Random initialization means that the answer will be slightly different each time (nondeterministic).
- Each iteration works to minimize the squared errors from a centroid.



https://realpython.com/k-means-clustering-python/

# K-Means Clustering Assumptions

- There a K clusters, each of roughly the same size
- Features are isotropically distributed.
- Each feature has a spherical variance.
- However, there is flexibility in many of these assumptions (see right panel from sklearn documentation).



Ground truth clusters

https://scikit-learn.org/stable/auto\_examples/cluster/plot\_kmeans\_assumptions.html

Let's do an example!