# **INFO 1998: Introduction to Machine Learning**



#### **Mid-Semester Feedback**

- How are we doing?
   Improvements/suggestions are welcome!
- Google form will be posted on Ed



## Mid-Semester Check-in for final project

- between lecture 6 and 8
- Need Dataset + Jupyter Notebook (basic data cleaning/exploration)
- 5 points of the final grade
- (Full announcement on Ed)



# **Lecture 6: Intro to Classifiers**

**INFO 1998: Introduction to Machine Learning** 



# **Agenda**

- 1. What is a Classifier?
- 2. K-Nearest Neighbors Classifier
- 3. Fit/Overfitting
- 4. Confusion Matrices





#### Classifiers are able to help answer questions like...

- "What species is this?"
- "What major is a student in based on their classes?"
- "Which Hogwarts House do I belong to?"
- "Am I going to pass this class?"





- Classifiers predict the class/category of a set of data points.
   This class/category is based off of the target variable we are looking at.
- Difference between linear regression and classifiers
  - Linear regression is used to predict the value of a continuous variable
  - Classifiers are used to predict categorical or binary variables





Two categories of classifiers: lazy learners and eager learners

#### Lazy Learners

- Store the training data and wait until a testing data appear
- Classification is conducted based on the most related data in the stored training data
- Less training time, more time in predicting

#### Eager Learners

- Construct a classification model based on the given training data before receiving data for classification
- More training time, less time in predicting





# K-Nearest Neighbors Classifier





#### What is the KNN Classifier?

- Lazy learner classifier
- Easy to interpret
- Fast to calculate
- Good for coarse analysis







#### **How Does It Work?**

Uses the k (a user specified value) nearest data points to predict the unknown one

- A simple assumption: the values **nearest** to a data point are **similar** to it
- k is a **hyperparameter** of the KNN model (a parameter that affects the learning process)!



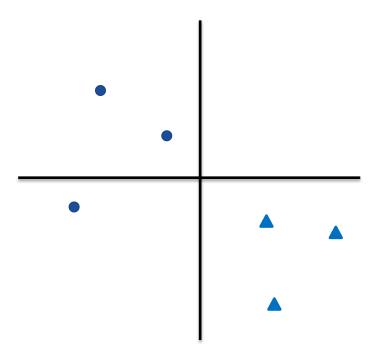


# **How Does It Work?** got an A, maybe I В got an A as well Α Α Α В





**Define** a k value (in this case k = 3)

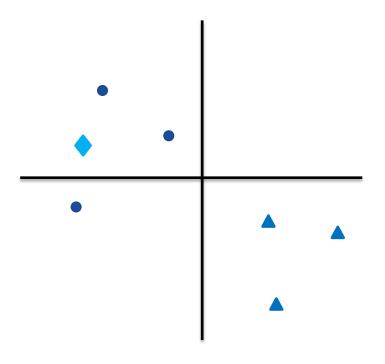






**Define** a k value (in this case k = 3)

**Pick** a point to predict (blue diamond)



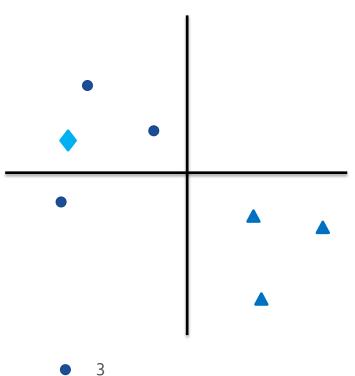




**Define** a k value (in this case k = 3)

**Pick** a point to predict (blue diamond)

**Count** the number of closest points









**Define** a k value (in this case k = 3)

Pick a point to predict (blue diamond)

**Count** the number of closest points

**Increase** the radius until the number of points within the radius adds up to 3







**Define** a k value (in this case k = 3)

Pick a point to predict (blue diamond)

**Count** the number of closest points

**Increase** the radius until the number of points within the radius adds up to 3

**Predict** the blue diamond to be a blue circle!

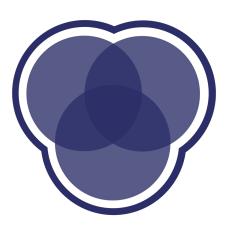


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# **Demo**







# Fit/Overfitting



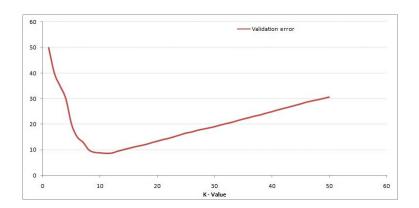


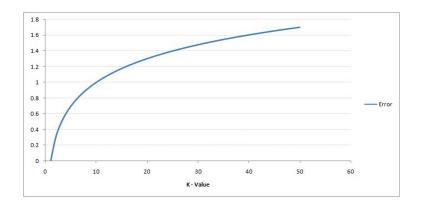
## **Overfitting**

When the model corresponds too closely to training data and then isn't transferable to other data.

#### Can fix by:

- Splitting data into training and validation sets
- Increasing k





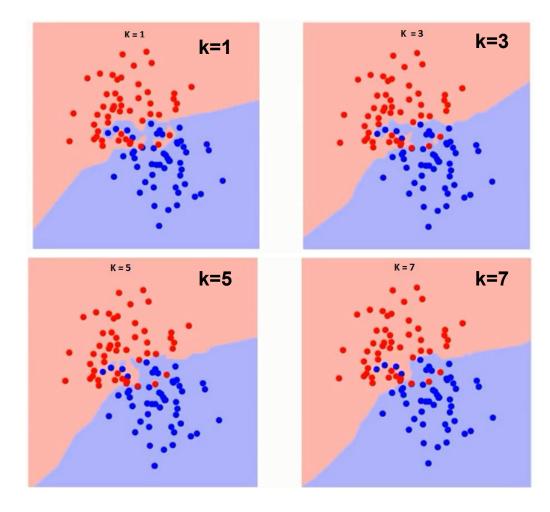




# Relationship Between k and Fit

The **k** value you use has a relationship to the fit of the model

A higher k gives a smoother line, but too large of a k and it is the average of all the data (or the label that is most common/likely)





# **Confusion Matrix**





#### What is a Confusion Matrix?

Table used to describe the performance of a classifier on a set of binary test data for which the true values are known

	p' (Predicted)	n' (Predicted)
P (Actual)	True Positive	False Negative
n (Actual)	False Positive	True Negative





# Sensitivity

Called the true positive rate

Tells us how many positives are correctly identified as positives

Optimize for: Initial diagnosis of fatal disease

	p' (Predicted)	n' (Predicted)
P (Actual)	True Positive	False Negative
n (Actual)	False Positive	True Negative

**Sensitivity** = True Positive/ (True Positive + False Negative)





# **Specificity**

Called the true negative rate

Tells us how many negatives are correctly identified as negatives

Optimize for: testing for a disease with a risky treatment

	P' (Predicted)	n' (Predicted)
P (Actual)	True Positive	False Negative
n (Actual)	False Positive	True Negative

Specificity = True Negative/ (True Negative + False Positive)





#### Question [pollev.com/vchen100]

Which is an example of when you would want higher specificity?

- A. DNA tests for a death penalty case
- B. Deciding which iPhone to buy
- C. Airport security





## **Overall Accuracy**

#### Proportion of correct predictions

	p' (Predicted)	n' (Predicted)
P (Actual)	True Positive	False Negative
n (Actual)	False Positive	True Negative

Accuracy =
(True Positive + True Negative) / Total





#### **Overall Error Rate**

#### Proportion of incorrect predictions

	p' (Predicted)	n' (Predicted)
P (Actual)	True Positive	False Negative
n (Actual)	False Positive	True Negative

Error = (False Positive + False Negative) / Total





#### **Precision**

Proportion of correct positive predictions among all positive predictions

	p' (Predicted)	n' (Predicted)
P (Actual)	True Positive	False Negative
n (Actual)	False Positive	True Negative

Precision = True Positive /
(True Positive + False Positive)





## **Coming Up**

- Assignment 6: Due on Oct 26th, 2022
- Mid-Semester Check-In: More details on ED Discussion!
- Feedback Survey: Please fill it out!
- Next Lecture: Applications of Supervised Learning

