

#### Lecture 34, November 16

#### **Implementing a Classifier**

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## Announcements

- Project 3 will be released today. Checkpoint Tuesday 11/22, final deadline Tuesday 11/29
- Homework will be assigned on Friday:
  - Early submission: Wed 11/23 (usual schedule)
  - "Regular" submission: Monday 11/28 after the break
- GSI/Tutor office hours locations from now on:
  - Mondays Etcheverry 3106
  - Wednesdays Etcheverry 3108
  - Other days: no change

## Classification

- Response variable is categorical; values are **classes**
- Binary response: Only two classes, 0 and 1
- Try to classify the response into one of the categories, based on:
  - Values of predictor variables, called **attributes**
  - Training set of data in which the classes of the individuals are known

## **k-Nearest Neighbor Classifier**

- New individual, unknown class
- Find the k closest individuals in the training set
  They are the new individual's "k nearest neighbors"
- Assign the new individual the same class as the majority of the k nearest neighbors (k is usually taken to be an odd number)

## **The Test Set**

- Split original training set at random into two sets
- Use one of the sets for training:
  - Explore as much as you want
  - Develop classifier
- Use the other set (test set) to compare the classifier's predictions and the true classes

## **Rows of Tables**

- Each row contains all the data for one individual
- tbl.row(i) evaluates to ith row of tbl
- tbl.row(i).item(j) is item indexed j of row i
- Type: "row object"; not all elements are of same type
- If all elements are of the same type (e.g. all numerical), then np.array (my\_row) converts my\_row to an array
- tbl.apply(function\_name) applies the function to each row of tbl; each entire row is passed to function\_name (Demo)

#### **Distance Between Two Points**

• Two attributes *x* and *y*:

$$D = \sqrt{(x_0 - x_1)^2 + (y_0 - y_1)^2}.$$

• Three attributes *x*, *y*, and *z*:

$$D = \sqrt{(x_0 - x_1)^2 + (y_0 - y_1)^2 + (z_0 - z_1)^2}$$

• and so on ...

# Finding the k Nearest Neighbors

To find the *k* nearest neighbors of a point:

- Find the distance between the point and each point in the training set
- Augment the training data table with a column containing all the distances
- Sort the augmented table in increasing order of the distances
- Take the top *k* rows of the sorted table

(Demo)

#### **The Classifier**

To classify a point:

- Find its *k* nearest neighbors
- Take a majority vote of the *k* nearest neighbors to see which of the two classes appears more often
- Assign the point the class that wins the majority vote

## **Assessing Accuracy**

Separate the data at random into a training set and a test set

- Use the training set to classify each point in the test set
- Find the fraction of points for which the classification is correct