



DSFA

Spring 2018

Lecture 35

Nearest Neighbor

Announcements

Nearest Neighbor

How to classify a new individual:

- Find their **nearest neighbor**: the individual closest to them in the data set
- Assign the new individual the **same** label as that nearest neighbor

(Demo)

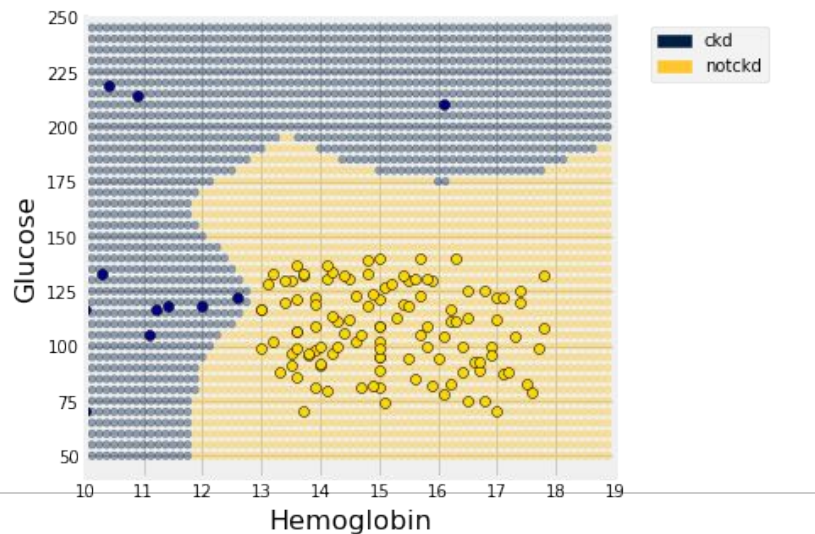
Nearest Neighbor recap

How to classify a new individual:

- Find their **nearest neighbor**: the individual closest to them in the data set
 - (We put data in standard units because scale of one attribute was so different than the other attribute--you will **not** need to do that on your proj3)
 - Compute table of distances from that individual to all other individuals
 - Sort by distance, so that closest is in the first row
 - Assign the new individual the **same** label as that nearest neighbor
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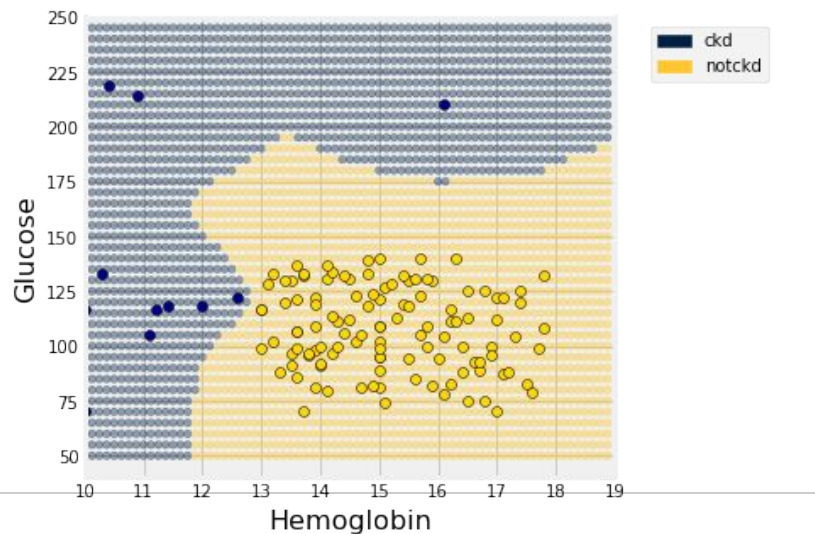
Decision Boundary

- Partition between the two classes
- Computer figured out that boundary, instead of humans having to “hard code” it: **machine learning**

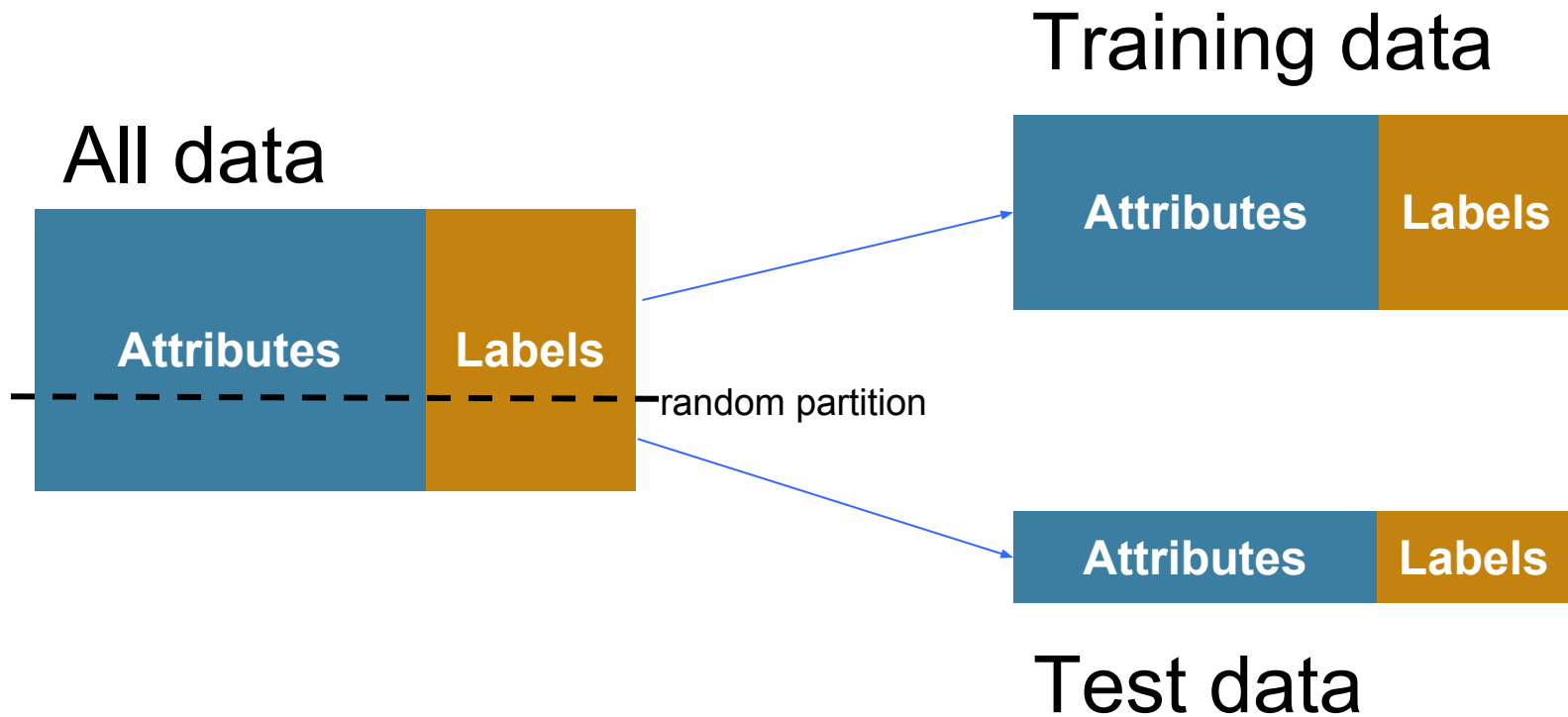


Evaluating a Classifier

How do we evaluate whether classifier is doing a good job on all those points where we have no data?



Train vs. Test



Train vs. Test

- Use **training data to create** the classifier
- Use **test data to evaluate** the finished classifier
- **Never** allow classifier to see test data until the very end:
think of classifier as a cheater who would be happy to
just memorize the answers

(Demo)

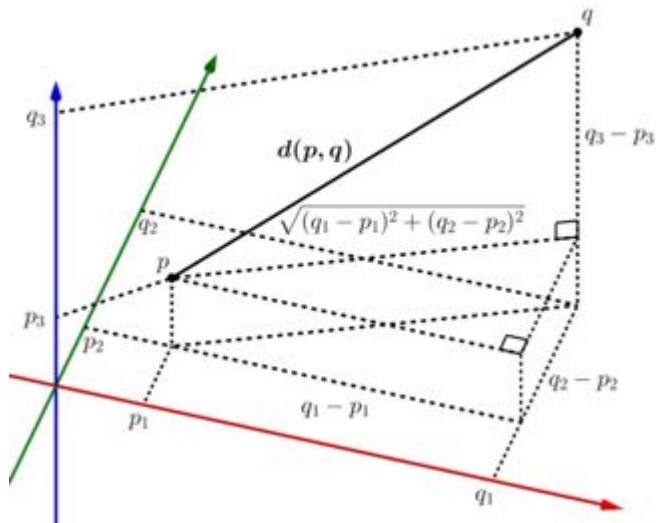
Multiple Neighbors

- If data are noisy, asking just the closest neighbor might not be ideal for accuracy
- Instead, ask the k closest neighbors, and take the majority label

(Demo)

Multiple Attributes

- We've used 2 attributes so far
- But nothing special about 2, just have to compute distances in higher dimensional spaces



(Demo)