



Lecture 37

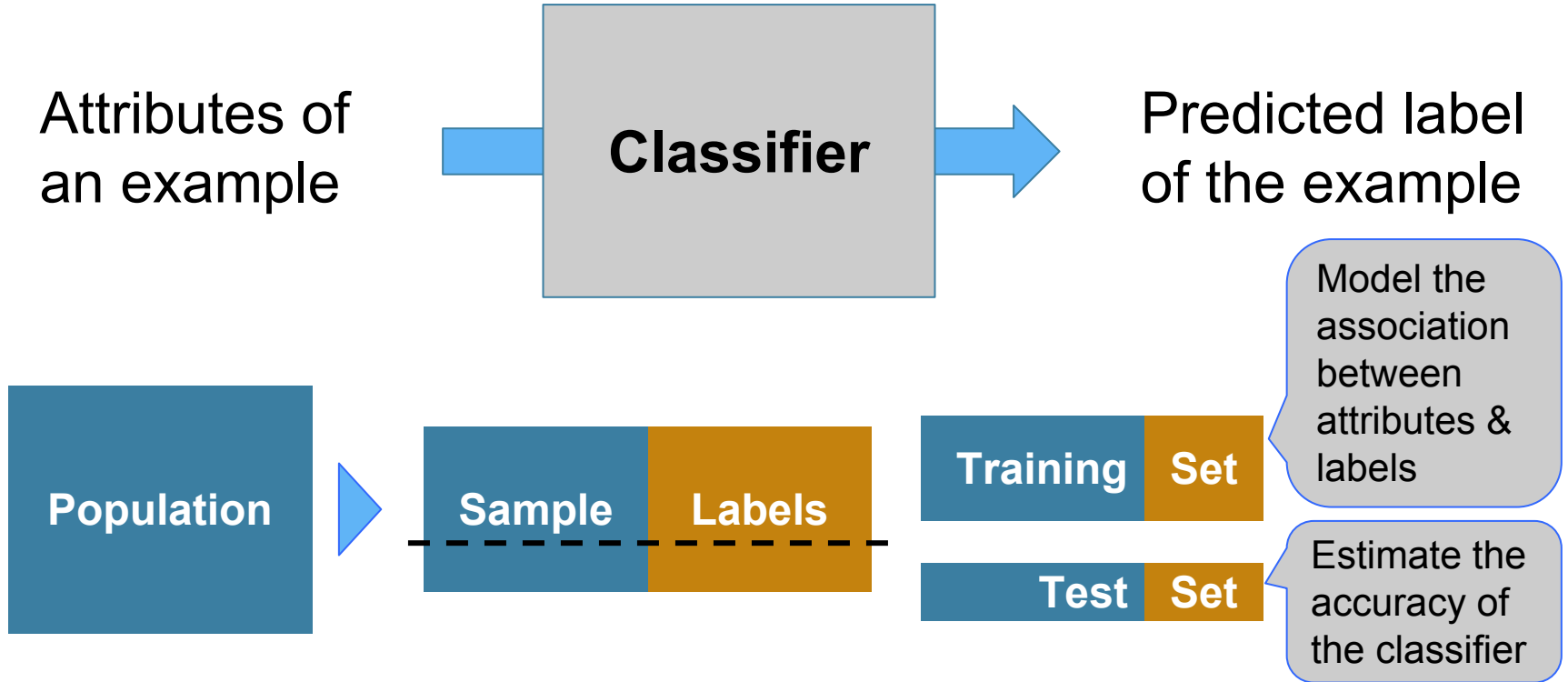
Decisions

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Announcements

Classifiers

Training a Classifier



Nearest Neighbors

Finding the k Nearest Neighbors

To find the k nearest neighbors of an example:

- Find the distance between the example and each example 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
-

The Classifier

To classify a point:

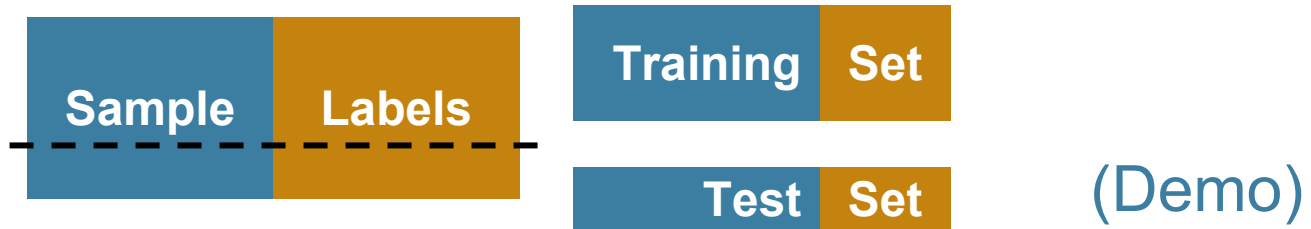
- Find its k nearest neighbors
- Take a majority vote of the k nearest neighbors to see which of the two classes appears more often
- Assign the point the class that wins the majority vote

(Demo)

Evaluation

Accuracy of a Classifier

- The accuracy of a classifier on a labeled data set is the proportion of examples that are labeled correctly
- Need to compare classifier predictions to true labels
- If the labeled data set is sampled at random from a population, then we can infer accuracy on that population



Decisions

Decisions Under Uncertainty

Interpretation by Physicians of Clinical Laboratory Results (1978)

"We asked 20 house officers, 20 fourth-year medical students and 20 attending physicians, selected in 67 consecutive hallway encounters at four Harvard Medical School teaching hospitals, the following question:

"If a test to detect a disease whose prevalence is $1/1000$ has a false positive rate of 5%, what is the chance that a person found to have a positive result actually has the disease, assuming that you know nothing about the person's symptoms or signs?"

Decisions Under Uncertainty

Interpretation by Physicians of Clinical Laboratory Results (1978)

"Eleven of 60 participants, or 18%, gave the correct answer. These participants included four of 20 fourth-year students, three of 20 residents in internal medicine and four of 20 attending physicians. The most common answer, given by 27, was that [the chance that a person found to have a positive result actually has the disease] was 95%.

Conditional Probability

Round One

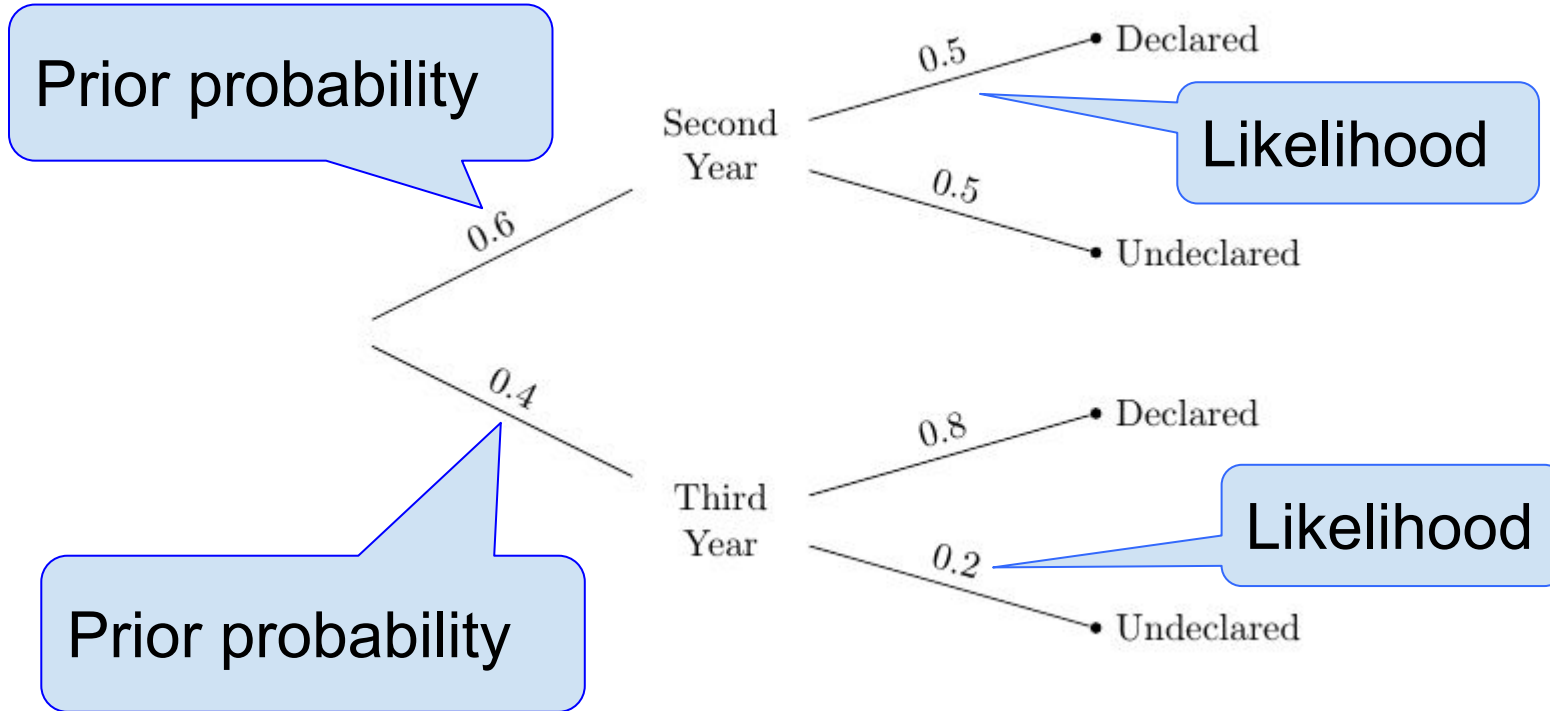
- Scenario:
 - Class consists of second years (60%) and third years (40%)
 - 50% of the second years have declared their major
 - 80% of the third years have declared their major
 - I pick one student at random.
 - Which is more likely: Second year or third year?
 - Second year, because they are 60% of the class
-

Round Two

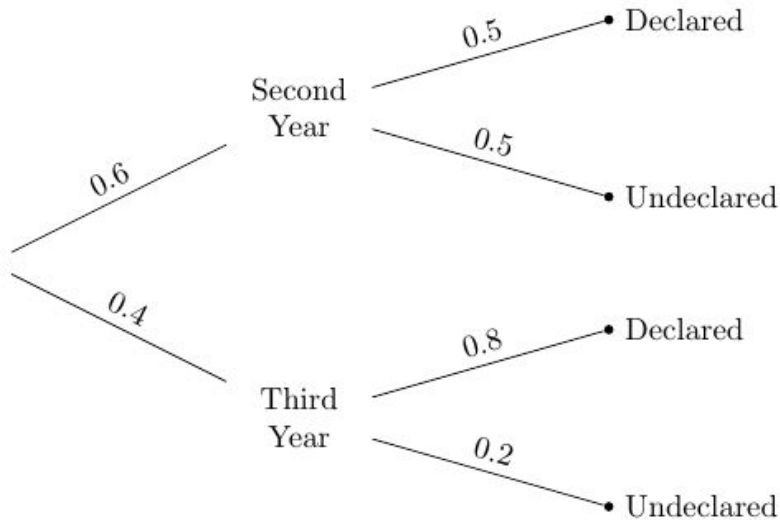
- Slightly different scenario:
 - Class consists of second years (60%) and third years (40%)
 - 50% of the second years have declared their major
 - 80% of the third years have declared their major
 - I pick one student at random...
That student has declared a major!
 - Second Year or Third Year? (Demo)
-

Bayes' Rule

Diagram and Terminology



Bayes' Rule



Pick a student at random.

Posterior probability:

$P(\text{Third Year} \mid \text{Declared})$

$$0.4 \times 0.8$$

= -----

$$(0.6 \times 0.5) + (0.4 \times 0.8)$$

$$= 0.5161\dots$$

Purpose of Bayes' Rule

- Update your prediction based on new information
 - In a multi-stage experiment, find the chance of an event at an earlier stage, given the result of a later stage
-

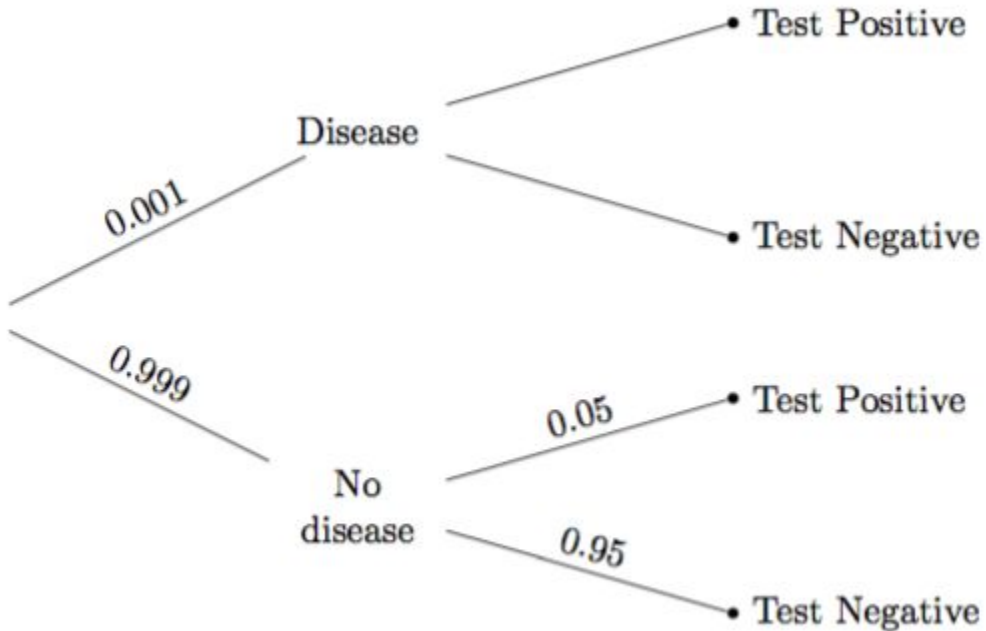
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Example: Doctors & Clinical Tests

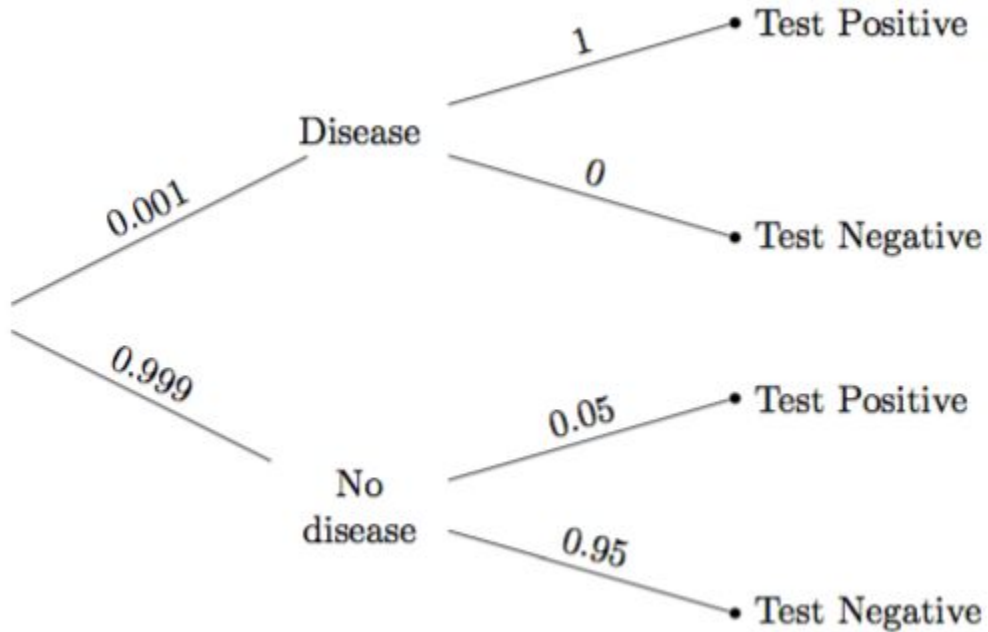


Problem did not give the *true positive* rate.

That's the chance the test says "positive" if the person has the disease.

It was assumed to be 100%.

Data and Calculation



$P(\text{Disease given Test } +)$

=

$$0.001 * 1$$

$$(0.001 * 1) + (0.999 * 0.05)$$

$$= 0.0196270\dots$$

(Demo)

Decisions

Subjective Probabilities

A probability of an outcome is...

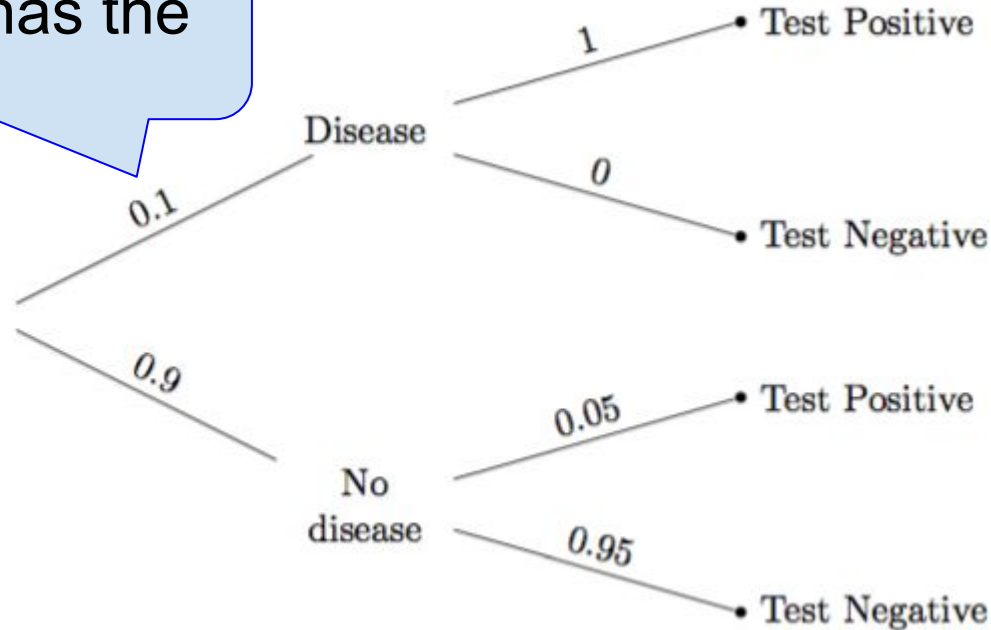
- The frequency with which it will occur in repeated trials, *or*
- The subjective degree of belief that it will (or has) occurred

Why use subjective priors?

- In order to quantify a belief that is relevant to a decision
 - When the subject of your prediction was not selected randomly from the population
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A Subjective Opinion

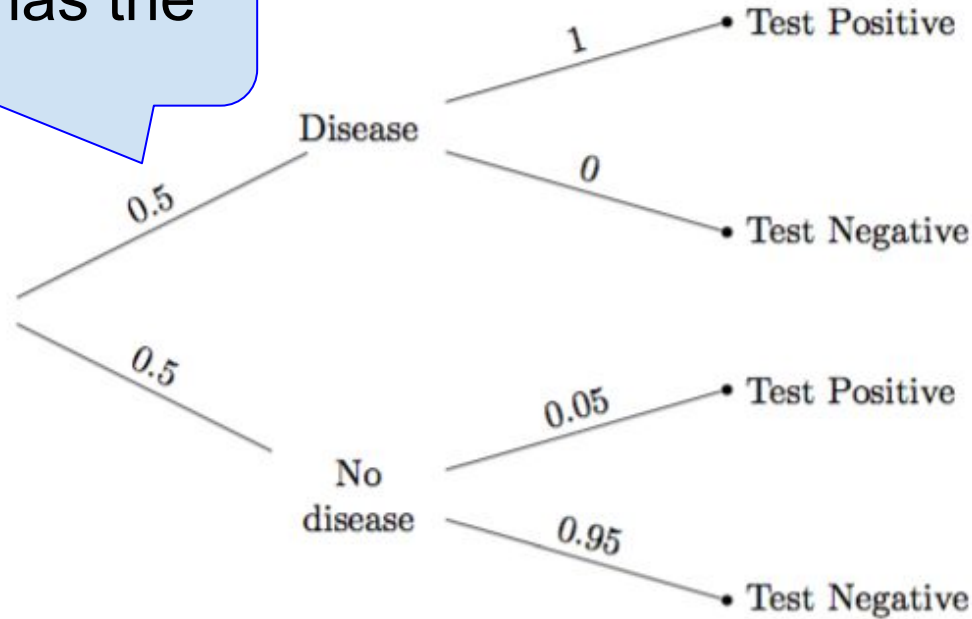
prior probability that the person has the disease



(Demo)

A Different Subjective Opinion

prior probability that the person has the disease



(Demo)