Bayes' Rule

STAT 330 - Iowa State University

#### **Outline**

In this lecture students will be introduced to Bayes' rule for "flipping" around conditional probabilities. Bayes' rule is very useful in many applications.

# Bayes' Rule

## Bayes' Rule

### Theorem (Bayes' Rule)

If  $B_1, \ldots, B_k$  is a cover or partition of  $\Omega$ , and A is an event, then

$$\mathbb{P}(B_j|A) = \frac{\mathbb{P}(A|B_j)\mathbb{P}(B_j)}{\sum_{j=1}^k \mathbb{P}(A|B_j)\mathbb{P}(B_j)}$$

Why?

$$\mathbb{P}(B_j|A) = \frac{\mathbb{P}(A \cap B_j)}{\mathbb{P}(A)} = \frac{\mathbb{P}(A|B_j)\mathbb{P}(B_j)}{\sum_{j=1}^k \mathbb{P}(A|B_j)\mathbb{P}(B_j)}$$

(Definition of conditional probability in numerator, LoTP in denominator)

- ullet Bayes rule o way to "flip" conditional probabilities.
- If we know  $\mathbb{P}(A|B_i)$  and  $\mathbb{P}(B_i)$ , then we can obtain  $\mathbb{P}(B_i|A)$
- Extremely useful for real world applications!

# **Applying Bayes Rule**

### Example:

My email is divided into 3 folders: Normal, Important, Spam. From past experience, the probability of emails belonging to these folders is 0.2, 0.1, and 0.7 respectively.

- Out of normal emails, the word "free" occurs with probability 0.01.
- Out of important emails, "free" occurs with probability 0.01.
- Out of spam emails, "free" occurs with probability 0.9.

My spam filter reads an email that contains the word "free". What is the probability that this email is spam?

# Applying Bayes Rule Cont.

#### **Define events:**

N= email is normal, I= email is important, S= email is spam F= email contains "free",  $\overline{F}=$  email doesn't contain "free"

#### Given:

$$\mathbb{P}(N) = 0.2, \ \mathbb{P}(I) = 0.1, \ \mathbb{P}(S) = 0.7$$
  
 $\mathbb{P}(F|N) = 0.01$   
 $\mathbb{P}(F|S) = 0.9$ 

$$\mathbb{P}(S|F) = ?$$
 (This is what we want to know)

# Applying Bayes Rule Cont.

What is the probability that my email is spam given that it contains the word "free"?

$$\mathbb{P}(S|F) = \frac{\mathbb{P}(S \cap F)}{\mathbb{P}(F)}$$

$$= \frac{\mathbb{P}(S)\mathbb{P}(F|S)}{\mathbb{P}(S)\mathbb{P}(F|S) + \mathbb{P}(I)\mathbb{P}(F|I) + \mathbb{P}(N)P(F|N)}$$

$$= \frac{\mathbb{P}(S)\mathbb{P}(F|S) + \mathbb{P}(I)\mathbb{P}(F|I) + \mathbb{P}(N)P(F|N)}{\mathbb{P}(S)\mathbb{P}(F|S) + \mathbb{P}(I)\mathbb{P}(F|I) + \mathbb{P}(N)P(F|N)}$$

# Applying Bayes Rule Cont.

### Conceptual understanding

- Before knowing anything
  - ightarrow probability that email is spam was  $\mathbb{P}(S)=0.7$ .
- After knowing that the email contains the word "free"
  - ightarrow update probability based on this knowledge.
- After knowing the email contains "free"
  - $\rightarrow$  probability of the email being spam is  $\mathbb{P}(S|F) = 0.995$ .
- We could calculate  $\mathbb{P}(N|F)$  and  $\mathbb{P}(I|F)$  also and *classify* the email to the category with the highest probability.
- In machine learning/statistics, this procedure is called a naive Bayes classifier.

### Recap

Students should be able to recognize when Bayes' rule is appropriate. They should be able to apply it to practical problems.