

CS1800  
Discrete Structures  
Fall 2017

Lecture 26  
11/6/17

## Last time

- Finish Entropy
- Start Cond. Prob.
  - Bayes Law

## Today

- Finish Cond. Prob.
  - & Bayes Law
- Start Markov chains
  - & Page Rank

## Next time

- Finish MC & PR

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- Start Module 4:
    - Algorithms & Analysis

# Conditional Probability

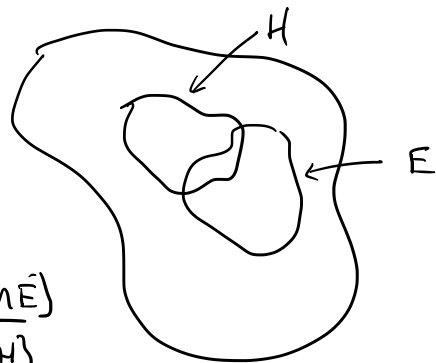
← prob of E given H

$$Pr[H \cap E] = Pr[H] \cdot \underline{Pr[E|H]} \quad \leftarrow$$

$$\Rightarrow Pr[E|H] = \frac{Pr[H \cap E]}{Pr[H]}$$

$$= Pr[E] \cdot Pr[H|E] \quad \leftarrow$$

$$\Rightarrow Pr[H|E] = \frac{Pr[H \cap E]}{Pr[E]}$$



Independence

$$Pr[E|H] = Pr[E]$$

$$Pr[H|E] = Pr[H]$$

$$\Rightarrow Pr[H \cap E] =$$

$$Pr[H] \cdot Pr[E|H]$$

$$= Pr[H] \cdot Pr[E]$$

Bayes Law:

$$\underline{Pr[E]} \cdot \underline{Pr[H|E]} = Pr[H \cap E] = \underline{Pr[H]} \cdot \underline{Pr[E|H]}$$

$$Pr[H|E] = \frac{Pr[E|H] \cdot Pr[H]}{Pr[E]}$$

H = hypothesis: do have Zika?

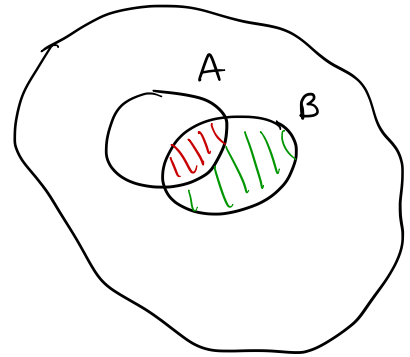
E = evidence: did blood test come back pos?

## Bayes Law

$$Pr(B|A) \cdot Pr(A) = Pr(A \cap B) = Pr(A|B) \cdot Pr(B)$$

$$Pr(A|B) = \frac{Pr(B|A) \cdot Pr(A)}{Pr(B)}$$

$$= \frac{Pr(B|A) \cdot Pr(A)}{Pr(B|A) \cdot Pr(A) + Pr(B|\bar{A}) \cdot Pr(\bar{A})}$$



what is  $Pr(B)$ ?

$$\begin{aligned} Pr(B) &= \text{red} + \text{green} \\ &= Pr(B|A) \cdot Pr(A) + Pr(B|\bar{A}) \cdot Pr(\bar{A}) \end{aligned}$$

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$$Pr(H|E) = \frac{Pr(E|H) \cdot Pr(H)}{Pr(E|H) \cdot Pr(H) + Pr(E|\bar{H}) \cdot Pr(\bar{H})}$$

Example: Zika in FL 2016

• Prevalence of Zika in S. FL.  $P(\text{Zika}) = 10^{-5}$  (1 in 100,000)

• accuracy of blood test is 99%

i.e.  $P(\text{pos. test} | \text{Zika}) = 0.99$  ← test subjects who had Zika

$P(\text{pos. test} | \text{no Zika}) = 0.01$  ← control group who don't have Zika

• You test positive: what is the chance that you have Zika?

⇒ Not  $P(\text{pos. test} | \text{Zika}) = 0.99$

⇒ you want  $P(\text{Zika} | \text{pos. test})!$

$$P(\text{zika} \mid \text{pos test}) = \frac{P(\text{pos. test} \mid \text{zika}) \cdot P(\text{zika})}{P(\text{pos test})}$$
$$= \frac{P(\text{pos. test} \mid \text{zika}) \cdot P(\text{zika})}{P(\text{pos test} \mid \text{zika}) \cdot P(\text{zika}) + P(\text{pos test} \mid \text{not zika}) \cdot P(\text{not zika})}$$

$$= \frac{0.99 \times 10^{-5}}{0.99 \times 10^{-5} + 0.01 \times (1 - 10^{-5})}$$

$$= \frac{0.0000099}{0.0000099 + 0.0099999}$$

$$\approx 0.00099$$

$$\approx 0.1\% \quad \text{i.e., only 1 in 1,000!}$$

Seems wildly counter intuitive, but...

10,000,000 people in FL.  $10^7$

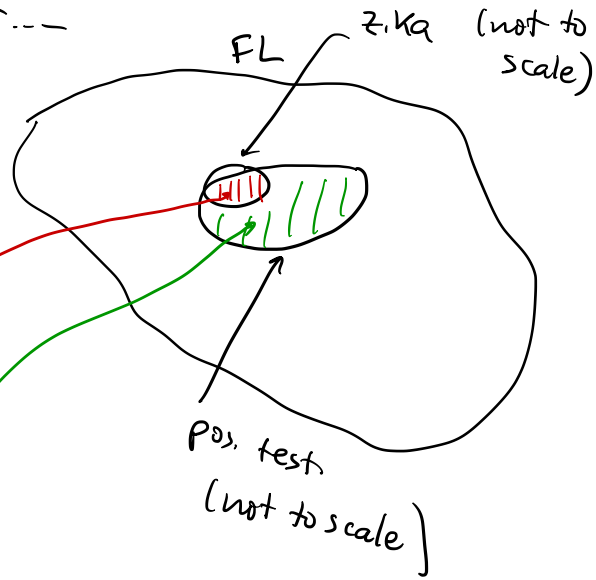
w/ zika:  $10^7 \cdot 10^{-5} = 10^2 = 100$

w/o zika: 9,999,900

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test pos w/ zika:  $100 \cdot 0.99 = 99$

test pos. w/o zika:  $9,999,900 \times 0.01$   
 $= 99,999$

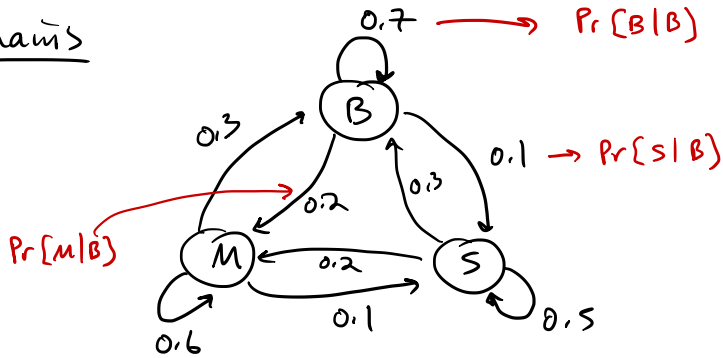


$\therefore$  If you test pos.,

you're about 1,000 times more likely  
to be among 99,999 who don't have zika  
but test pos. than among the 99  
who do have zika and test pos.

# Markov chains

- B: Bertucci's
- M: Margaritas
- S: Sato



3 reds  $\Rightarrow$  must add to 1

