

Bayes's formula

S sample space E and F events

"conditioning on F "

$$E = EF \cup EF^c$$

Proof

$$E = ES = E(F \cup F^c) \\ = EF \cup EF^c$$

And EF and EF^c are mutually exclusive

$$P(E) = P(EF) + P(EF^c) \quad P(E|F) = \frac{P(EF)}{P(F)} \\ = P(E|F)P(F) + P(E|F^c)P(F^c)$$

" $P(E)$ is weighted average of $P(E|F)$ and $P(E|F^c)$ "

Recall trick "reversing the conditional probability"

$$P(A|B) = \frac{P(AB)}{P(B)} = \frac{P(B|A)P(A)}{P(B)}$$

Bayes's formula

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)} = \frac{P(B|A)P(A)}{P(B|A)P(A) + P(B|A^c)P(A^c)}$$

More generally let H_1, \dots, H_n be a collection of mutually exclusive, exhaustive events

Mutually exclusive: $H_i H_j = \emptyset$ ($i \neq j$)

Exhaustive: $H_1 \cup H_2 \cup \dots \cup H_n = S$

S

H_1	H_3	H_5
H_2	H_4	H_6

Consider H_i to be "hypotheses"

Now take any event E which we regard as some "Evidence"

Bayes's
formula

$$P(H_k | E) = \frac{P(E | H_k) P(H_k)}{\sum_{i=1}^n P(E | H_i) P(H_i)}$$

$P(E | H_i)$ - prob of evidence given hypothesis

$P(H_i)$ - "prior probability" of each hypothesis

Insurance company divides people into two
class $\left\{ \begin{array}{l} \text{accident prone} \quad A \\ \text{not accident prone} \quad A^c \end{array} \right.$

$$P(B|A) = P(\text{prone person has accident in 1 year}) = 0.4$$

$$P(B|A^c) = P(\text{not accident prone has accident in 1 year}) = 0.2$$

$P(A) = 30\%$ of population is accident prone

$S = \{ \text{set of people} \}$ $A = \{ \text{accident prone} \}$

$B = \{ \text{people who have an accident in one year} \}$

Q: $P(B)$?

$$\begin{aligned} P(B) &= P(B|A)P(A) + P(B|A^c)P(A^c) \\ &= (.4)(.3) + (.2)(.7) = .26 \end{aligned}$$

Q: Suppose a person has an accident B
What is probability that this person is
accident prone?

$$\begin{aligned} P(A|B) &= \frac{P(B|A)P(A)}{P(B|A)P(A) + P(B|A^c)P(A^c)} \quad \{ .26 \\ &= \frac{(.4)(.3)}{.26} = \frac{6}{13} \end{aligned}$$

We have some prior probabilities

$$C = \{\text{cheating}\} \quad C^c = \{\text{not cheating}\}$$

E = way Am they played

$$P(C|E) = \frac{P(E|C)P(C)}{P(E|C)P(C) + P(E|C^c)P(C^c)}$$

Under what conditions is $P(C|E) > P(C)$

Need
$$\frac{P(E|C)}{P(E|C)P(C) + P(E|C^c)P(C^c)} > 1$$

$$P(E|C) > P(E|C)P(C) + P(E|C^c)P(C^c)$$

$$P(E|C) [1 - P(C)] > P(E|C^c) P(C^c)$$

$$P(E|C) > P(E|C^c) \Rightarrow \text{increases confidence in } C$$

<u>Example</u>	3 cards	RR	red on both sides
		RB	red / black
		BB	black on both sides

Card randomly selected and placed down on table, so we only get to see one side.

Suppose we see red facing up.
What is prob that the other side is black?

Hypothesis = RR or RB or BB

Evidence = {see red} or {see black}

$$P(RB | \{\text{see red}\}) = \frac{P(\text{see red} | RB) P(RB)}{P(\text{see red} | RB) P(RB) + P(\text{see red} | RR) P(RR) + P(\text{see red} | BB) P(BB)}$$

$$= \frac{\left(\frac{1}{2}\right) \left(\frac{1}{3}\right)}{\left(\frac{1}{2}\right) \left(\frac{1}{3}\right) + (1) \left(\frac{1}{3}\right) + 0} = \frac{1}{3}$$