

$P(\cdot | F)$ is a probability measure

Exam on Friday. Wednesday will be review.

Office hours M 1-2, 3-4, W 9:30-10:30

Let F be some fixed event

define $Q(E) = P(E | F)$

$Q(\cdot) = P(\cdot | F)$ satisfies all properties of a probability measure

Prop $0 \leq P(E | F) \leq 1$ $0 \leq Q(E) \leq 1$

$P(S | F) = 1$ $Q(S) = 1$

if E_1, E_2, \dots are mutually exclusive

$$P\left(\bigcup_{i=1}^{\infty} E_i | F\right) = \sum_{i=1}^{\infty} P(E_i | F)$$

$$Q\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} Q(E_i)$$

Anything that follows from axioms, has a "conditional" version

$$P(E^c | F) = 1 - P(E | F)$$

$$P(E_1 \cup E_2 | F) = P(E_1 | F) + P(E_2 | F) - P(E_1 E_2 | F)$$

Proof of Prop

$$0 \leq P(E|F) \leq 1$$

$$P(E|F) = \frac{P(EF)}{P(F)} \geq 0$$

$$P(EF) \leq P(F) \quad \text{b/c } EF \subset F$$

$$\frac{P(EF)}{P(F)} \leq 1$$

$$P(S|F) = \frac{P(SF)}{P(F)} = \frac{P(F)}{P(F)} = 1$$

$$P\left(\bigcup_{i=1}^{\infty} E_i | F\right) = \frac{P\left(\left(\bigcup_{i=1}^{\infty} E_i\right) F\right)}{P(F)}$$

$$\left(\bigcup_{i=1}^{\infty} E_i\right) F = \bigcup_{i=1}^{\infty} (E_i F) \quad \text{distributive law}$$

$$E_i F E_j F \subset E_i E_j = \emptyset \quad \text{b/c } E_i \text{ mut. exc.}$$

$$\frac{P\left(\bigcup_{i=1}^{\infty} E_i F\right)}{P(F)} = \sum_{i=1}^{\infty} \frac{P(E_i F)}{P(F)} = \sum_{i=1}^{\infty} P(E_i | F)$$

Independent events E_1, E_2 $P(E_1 E_2) = P(E_1) P(E_2)$

Conditionally independent events

$$P(E_1 E_2 | F) = P(E_1 | F) P(E_2 | F)$$

Examples Insurance company

A = accident prone

A_1 = has accident in 1st year

A_2 = has accident in 2nd year

$$P(A) = .3$$

$$P(A_i | A) = .4$$

$$P(A_i | A^c) = .2$$

Assume A_1, A_2
conditionally independent

Previously we saw $P(A_i) = .26$

New Q: $P(A_2 | A_1)$ Accident in 2nd year
given accident in first year.

Condition on A , whether person is accident.

Also, take all probabilities conditional on A_1

$$P(A_2 | A_1) = P(A_2 | A A_1) P(A | A_1)$$

$$+ P(A_2 | A^c A_1) P(A^c | A_1)$$

$$P(A_2 | A A_1) \stackrel{1}{=} P(A_2 | A) = .4$$

because different years independent

$$P(A_2 | A^c A_1) = P(A_2 | A^c) = .2$$

$$P(A | A_1) = \frac{P(A_1 | A) P(A)}{P(A_1)} = \frac{(.4)(.3)}{.26} = \frac{6}{13}$$

$$P(A^c | A_1) = 1 - P(A | A_1) = \frac{7}{13}$$

$$P(A_2 | A_1) = (.4) \frac{6}{13} + .2 \frac{7}{13} \approx .29$$

Laplace's rule of succession:

$k+1$ coins in a box i th coin comes up heads with probability i/k . ($i=0, \dots, k$)

We flip coin n times, and get n heads. What is prob. that $(n+1)$ st flip is heads?

$$C_i = i\text{th coin is selected. } P(C_i) = \frac{1}{k+1}$$

$F_n =$ first n flips are heads.

$H =$ $(n+1)$ st flip is heads.

$$P(F_n | C_i) = \left(\frac{i}{k}\right)^n \quad (\text{Bernoulli trials with } p = i/k)$$

$$P(H | C_i) = i/k$$

The different flips are conditionally independent,
given C_i

$$P(H|F_n) = \sum_{i=0}^k P(H|F_n C_i) P(C_i|F_n)$$

$$P(H|F_n C_i) = P(H|C_i) = i/k$$

because H and F_n are independent given C_i

Bayes Formula

$$P(C_i|F_n) = \frac{P(F_n|C_i) P(C_i)}{P(F_n)}$$

$$= \frac{P(F_n|C_i) P(C_i)}{\sum_{j=0}^k P(F_n|C_j) P(C_j)} = \frac{(i/k)^n \frac{1}{k+1}}{\sum_{j=0}^k (j/k)^n \frac{1}{k+1}}$$

$$= \frac{(i/k)^n}{\sum_{j=0}^k (j/k)^n}$$

$$P(H|F_n) = \sum_{i=0}^k (i/k) \frac{(i/k)^n}{\sum_{j=0}^k (j/k)^n} = \frac{\sum_{i=0}^k (i/k)^{n+1}}{\sum_{j=0}^k (j/k)^n}$$

if k is large

$$P(H|F_n) \approx \frac{n+1}{n+2}$$