

Review.

- Conditional probability

$$P(E|F) = \frac{P(EF)}{P(F)}, \text{ if } P(F) > 0.$$

↓
(conditional prob. of E given F)

- $P(E_1, E_2, \dots, E_n) = P(E_1) \cdot P(E_2|E_1) \cdot P(E_3|E_1, E_2) \cdots P(E_n|E_1, \dots, E_{n-1})$
(Multiplicative rule).

- Suppose F_1, F_2, \dots, F_n are mutually exclusive, and exhaustive (i.e. $\bigcup_{i=1}^n F_i = S$).

Then

- $P(E) = \sum_{i=1}^n P(F_i) \cdot P(E|F_i)$
(law of total probability)

- $P(F_i|E) = \frac{P(F_i) P(E|F_i)}{\sum_{k=1}^n P(F_k) P(E|F_k)}$
(Bayes' formula).

Remark: Suppose $F \subset S$ is an event in a sample space with $P(F) > 0$.

Then $P(\cdot|F)$ is a probability on S .

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

$$(2) 0 \leq P(E|F) \leq 1.$$

$$(3) P\left(\bigcup_{n=1}^{\infty} E_n\right|F) = \sum_{n=1}^{\infty} P(E_n|F),$$

if E_1, E_2, \dots are mutually exclusive.

(1), (2) are obvious. To see (3)

$$\begin{aligned} P\left(\bigcup_{n=1}^{\infty} E_n\right|F) &= \frac{P\left(\left(\bigcup_{n=1}^{\infty} E_n\right) \cap F\right)}{P(F)} \\ &= \frac{P\left(\bigcup_{n=1}^{\infty} (E_n \cap F)\right)}{P(F)} \\ &= \sum_{n=1}^{\infty} \frac{P(E_n \cap F)}{P(F)} \quad (\text{since } E_n \cap F \text{ are disjoint}) \\ &= \sum_{n=1}^{\infty} P(E_n|F). \end{aligned}$$

Example 1.

A bin contains 3 types of disposable flashlights. The probability that a type 1 flashlight will give more than 100 hours of use is .7, with the corresponding probabilities for type 2 and type 3 flashlights being .4 and .3, respectively. Suppose that 20 percent of the flashlights in the bin are type 1, 30 percent are type 2, and 50 percent are type 3.

- (a) What is the probability that a randomly chosen flashlight will give more than 100 hours of use?
(b) Given that a flashlight lasted more than 100 hours, what is the conditional probability that it was a type j flashlight, $j = 1, 2, 3$?

Solution: Let E be the event that a random chosen flashlight will give more than 100 hours.

Let F_i ($i=1,2,3$) be the event that a random chosen flashlight is of type i .

We need to find out (a) $P(E)$; and
(b) $P(F_i|E)$.

From the conditions of the question, we know

$$P(E|F_1) = 0.7, \quad P(E|F_2) = 0.4$$

$$P(E|F_3) = 0.3.$$

$$P(F_1) = 0.2, \quad P(F_2) = 0.3, \quad P(F_3) = 0.5.$$

$$\begin{aligned} \text{Hence } P(E) &= \sum_{i=1}^3 P(F_i) P(E|F_i) \\ &= 0.2 \times 0.7 + 0.3 \times 0.4 + 0.5 \times 0.3 \end{aligned}$$

$$\begin{aligned} P(F_1|E) &= \frac{P(F_1) \cdot P(E|F_1)}{P(E)} \\ &= \frac{0.2 \times 0.7}{0.2 \times 0.7 + 0.3 \times 0.4 + 0.5 \times 0.3} \\ &= \frac{14}{41} \end{aligned}$$

Similarly

$$P(F_2|E) = \frac{12}{41}, \quad P(F_3|E) = \frac{15}{41}. \quad \square$$

§ 3.3. Independent events.

Let E, F be two events. In general, knowing that F has occurred changes the chance of E 's occurrence, that is, possibly $P(E|F) \neq P(E)$.

If $P(E|F) = P(E)$, we say E is independent of F .

Notice that

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

$$\Leftrightarrow P(EF) = P(E) \cdot P(F)$$

$$\Leftrightarrow P(F|E) = P(F)$$


Def. We say that E and F are independent if

$$P(EF) = P(E) \cdot P(F).$$

Example 2. A card is randomly chosen from a deck of 52 playing cards.

E — the event that the chosen card is an Ace "A"

F — the event that the chosen card is a spade.

Determine whether or not E and F are independent. 

Solution:

$$P(E) = \frac{4}{52}, \quad P(F) = \frac{13}{52} = \frac{1}{4}.$$

$$P(EF) = \frac{1}{52} = P(E)P(F).$$

Hence E, F are independent.

Prop 3. If E and F are independent, then

(1) E and F^c are independent

(2) E^c and F^c are independent.

Pf. (1)

$$\begin{aligned} P(E \cap F^c) &= P(E) - P(EF) \\ &= P(E) - P(E) \cdot P(F) \\ &= P(E)(1 - P(F)) \\ &= P(E)P(F^c), \end{aligned}$$

Hence E, F^c are independent.

(2) can be obtained from (1).

- Independence of 3 or more events.

Def. We say 3 events E, F, G are independent if

$$(1) \quad P(EFG) = P(E)P(F)P(G).$$

$$(2) \quad P(EF) = P(E) \cdot P(F)$$

$$P(EG) = P(E)P(G)$$

$$P(FG) = P(F)P(G).$$

Def. Let E_1, E_2, \dots, E_n be a finite family of events.

Say E_1, \dots, E_n are independent if for any sub-collection

$E_{i_1}, E_{i_2}, \dots, E_{i_r}$ (with i_1, \dots, i_r being distinct),

$$P(E_{i_1}, E_{i_2}, \dots, E_{i_r}) = P(E_{i_1}) \cdots P(E_{i_r}).$$

Def. We say an infinite family of events are independent if every finite subfamily of them is independent.

Def. (Independence of sub-experiments).

An experiment might consist of some sub-experiments.

For instance, the experiment that rolling a coin continuously consists of a sequence of sub-experiments, where the n -th sub-experiment is the n -th toss of the coin, $n=1, 2, \dots$.

We say these sub-experiments are independent if

E_1, E_2, \dots, E_n are independent whenever E_i is an event whose occurrence depends only on the i -th sub-experiment.

These sub-experiments are said to be trials if the set of possible outcomes of each sub-experiment are the same.

Chap 4. Random Variables.

§4.1 Introduction to random Variables.

Def. For a random experiment, a random Variable (r.v.)

X is a real-valued function defined on the sample space S . That is,

$X: S \rightarrow \mathbb{R}$ is a function.

Example 1. Flip 3 fair coins. Let X be the number of the heads that appear.

$$X = \# \{ \text{heads that appear} \}$$

Eg. if the outcome is (T, H, T) , then $X=1$
if the outcome is (H, T, H) , then $X=2$.

Example 2. Two fair dice are rolled.

Y = the product of the numbers that appear.

If the outcome is $(2, 5)$, then $Y=10$.

Example 3.

Z = the life-time (in hours)
of a randomly chosen flashlight.

§ 4.2. Discrete random Variables.

Def. A r.v. X is said to be discrete if X can take on at most countably many different values.

Def. For a discrete r.v. X , the prob. mass function of X is defined by

$$\begin{aligned}
 p(a) &= P\{X=a\} \\
 &= P\{\omega \in S : X(\omega) = a\}, \\
 &\quad \forall a \in \mathbb{R}.
 \end{aligned}$$

Example 4: $X = \#$ { ^{that} heads appear in rolling
3 fair coins }

$$\{X=0\} = \{(T, T, T)\}$$

$$\{X=1\} = \{(H, T, T), (T, H, T), (T, T, H)\}$$

$$\{X=2\} = \{(H, H, T), (H, T, H), (T, H, H)\}$$

$$\{X=3\} = \{(H, H, H)\}.$$

So $p(0) = \frac{1}{8}$, $p(1) = \frac{3}{8}$, $p(2) = \frac{3}{8}$, $p(3) = \frac{1}{8}$

and $p(a) = 0$ for all $a \notin \{0, 1, 2, 3\}$.