

## 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)

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

### § 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 1. 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 

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

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

Hence  $E, F$  are independent.

Prop 1. If  $E$  and  $F$  are independent, then

(1)  $E$  and  $F^c$  are independent

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

Pf. (1)

$$P(E \cap F^c) = P(E) - P(EF)$$

$$= P(E) - P(E) \cdot P(F)$$

$$= P(E)(1 - P(F))$$

$$= P(E)P(F^c),$$

(since  $E = (E \cap F) \cup (E \cap F^c)$   
disjoint)

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

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

$$(2) P(EF) = P(E)P(F)$$

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

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

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.