

Review

- Sample spaces having equally likely outcomes.

$$P(E) = \frac{\# E}{\# S}$$

Exer 1. A deck of 52 cards is dealt out. What is the probability that the first ace occurs in the 14th card.

Solution: Let E denote the event that the first ace occurs in the 14th card. Let S denote the sample space.

$$\text{Then } \# S = 52!$$

$$\# E = 48 \times 47 \times \cdots \times 36 \times 4 \times (38!)$$

$$\text{Hence } P(E) = \frac{\# E}{\# S} = \frac{48 \times 47 \times \cdots \times 36 \times 4}{52 \times 51 \times \cdots \times 39}.$$

Chap 3. Conditional probability and independence

§ 3.1 Conditional probability.

Example 1: Let us roll two dices. Suppose the first die is a 3. Given this information, what is the prob. that the sum of 2 dices equals 8?

Sol: F — the event that the first die is 3

E — the event that the sum of 2 dices equals 8.

$$F = \{(3, 1), (3, 2), (3, 3), (3, 4), (3, 5), (3, 6)\}$$

$$E = \{(i, j) \in \{1, 2, 3, 4, 5, 6\}^2 : i + j = 8\}.$$

$$E \cap F = \{(3, 5)\}.$$

The prob. of each outcome in F is $\frac{1}{6}$, so is the outcome $\{(3, 5)\}$.

Hence the (conditional) prob of E given F is $\frac{1}{6}$.

Def. (conditional prob.).

Let E, F be two events for a random experiment.
Suppose $P(F) > 0$. Then the conditional prob. of E
given F is

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

Remark: If $P(F) = 0$,

then $P(E|F)$ is not well-defined.

But in practice, you may assign
any value from $[0, 1]$ to $P(E|F)$.

Prop. (Multiplicative rule)

$$\bullet P(E_1, E_2) = P(E_1) P(E_2 | E_1)$$

$$\bullet P(E_1, E_2, \dots, E_n)$$

$$= P(E_1) \cdot P(E_2 | E_1) \cdot P(E_3 | E_1, E_2) \cdots$$

$$\cdot P(E_n | E_1, E_2, \dots, E_{n-1})$$

pf. Since $P(E_2 | E_1) = \frac{P(E_1, E_2)}{P(E_1)}$, so

$$P(E_1, E_2) = P(E_1) \cdot P(E_2 | E_1).$$

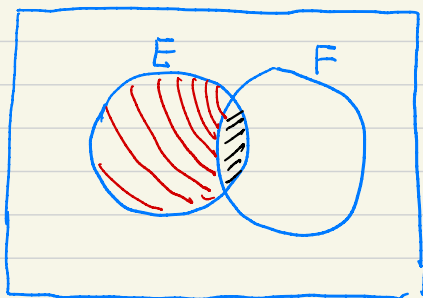
To see the second identity,

$$\text{RHS} = P(E_1) \cdot \frac{P(E_1, E_2)}{P(E_1)} \cdot \frac{P(E_1, E_2, E_3)}{P(E_1, E_2)} \cdots \frac{P(E_1, \dots, E_n)}{P(E_1, \dots, E_{n-1})}$$

$$= P(E_1, \dots, E_n). \quad \square$$

§ 3.2 Bayes' formula.

Let E, F be two events.



$$E = \underbrace{(E \cap F)}_{\text{(black)}} \cup \underbrace{(E \cap F^c)}_{\text{(red)}}$$

Hence

$$P(E) = P(E \cap F) + P(E \cap F^c)$$

$$\text{But } P(E \cap F) = P(F) \cdot P(E|F),$$

$$P(E \cap F^c) = P(F^c) \cdot P(E|F^c).$$

We obtain

$$P(E) = P(F) \cdot P(E|F) + P(F^c) \cdot P(E|F^c).$$

(Total probability formula).

Hence to determine the prob. of E ,
we may first conduct the "conditioning"
upon whether or not the event F has
occured .

Next we give a generalization of this formula.

Let F_1, F_2, \dots, F_n be a sequence of events

such that they are mutually exclusive,

and $\bigcup_{k=1}^n F_k = S$ (we say F_1, \dots, F_n
are exhaustive)

Then we have

$$P(E) = \sum_{k=1}^n P(F_k) \cdot P(E|F_k).$$

pf: Notice that $E = \bigcup_{k=1}^n (E \cap F_k)$
(with disjoint union)

Hence

$$\begin{aligned} P(E) &= \sum_{k=1}^n P(E \cap F_k) \\ &= \sum_{k=1}^n P(F_k) P(E|F_k). \quad \square \end{aligned}$$

Prop. (Bayes' formula).

Assume F_1, \dots, F_n are mutually exclusive and exhaustive.

Then for any $(1 \leq i \leq n)$,

$$P(F_i|E) = \frac{P(F_i) \cdot P(E|F_i)}{\sum_{k=1}^n P(F_k) P(E|F_k)}$$

$$\text{Pf: } \sum_{k=1}^n P(F_k) P(E|F_k) = P(E)$$

$$P(F_i) \cdot P(E|F_i) = P(E|F_i)$$

