

Section 4 Conditional probability, total probability formula and Bayes' rule

- 一、Conditional probability
- 二、Total probability and Bayes' rule
- 三、Conclusion

—、 Conditional probability

1.e.g.

Given three families, one has **only one boy and no girl**, one **has only one girl and no boy**, and the last family has **one boy and one girl**:

Question? Determine the probability of the following events.

- 1, A: select a family randomly from the three families, what is the probability **that it has a boy**?
- 2, B: select a family randomly from the three families, what is the probability **that it has a girl**?
- 3, C: given a family selected has a girl, what is the probability that it **only** has a boy? $P(A|B) = ?$

Solution: (1) $P(A) = \frac{2}{3}$

(2) $P(B) = \frac{2}{3}$

(3) $P(AB) = \frac{1}{3}$

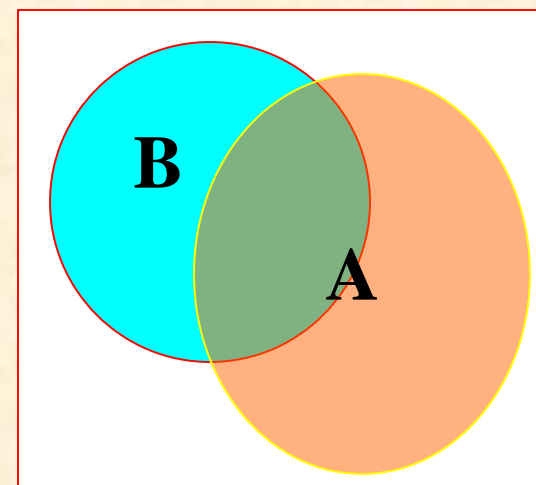
(4) $P(A|B) = \frac{P(AB)}{P(B)} = \frac{1}{2}$

note. 1° $P(A) = \frac{2}{3} \neq P(A|B)$

2° $P(AB) = \frac{1}{3} \neq P(A|B)$

$P(AB) : \Omega =$

Ω



{only one boy family, only one girl family, one boy and one girl family }

$P(A|B) : \Omega_B = B$

3° $P(A|B) = \frac{1}{2} = \frac{1/3}{2/3} = \frac{P(AB)}{P(B)}$

Can this merely be coincidence?

No.

2. Definition 1.8 (conditional probability)

Given two events A , B , if $P(B) > 0$, then

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

is the conditional probability of A , while given B .

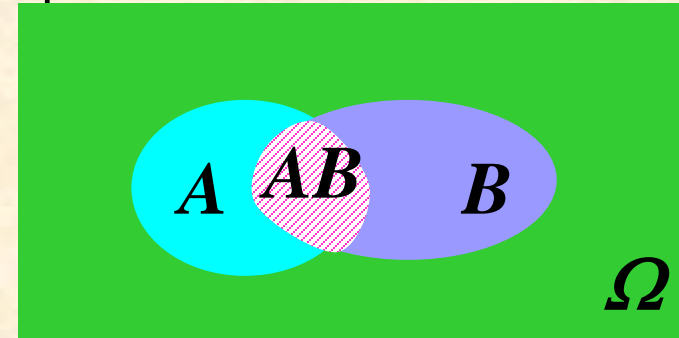
Note. The way to calculate the conditional probability:

- ① reduce the sample space;
- ② definition.

Difference between $P(AB)$ and $P(A|B)$:

$$P(AB) : \Omega$$

$$P(A|B) : \Omega_B = B$$



e.g. , classical probability

$$P(AB) = \frac{|AB|}{|\Omega|}$$

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

e.g.1 (1) Given the families having **3 children**, what is the probability for the event that one family selected randomly has **at least** one girl (suppose that the boy and girl are born equally).

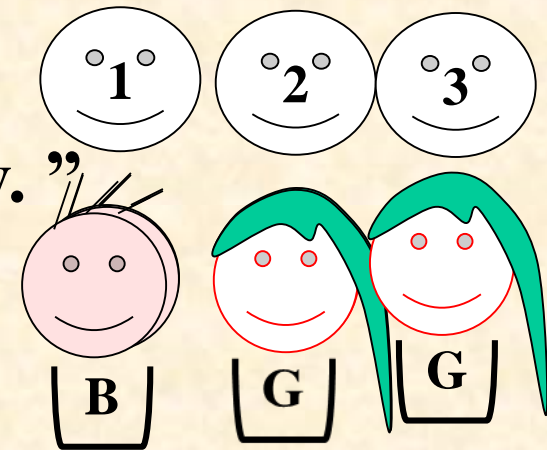
Solution: $|\Omega| : 2^3$.

A = “there is **at least one girl** in the family.”

\bar{A} = “all the children are boy.”

$$P(\bar{A}) = \frac{1}{2^3} = \frac{1}{8}$$

$$\therefore P(A) = 1 - P(\bar{A}) = 1 - \frac{1}{8} = \frac{7}{8}$$



(2) For the families having 3 children, given the family selected has at least one girl, what is the probability for the event that **this family selected **has at least one boy**?**

Solution: denote A=“there is at least one girl in the family.”

B=“there is at least one boy in the family.”

$$\text{then } P(B) = 1 - P(\bar{B}) = 1 - \frac{1}{2^3} = \frac{7}{8},$$

Denote C=“the family has one boy and two girls.”

D=“the family has two boys and one girl.”

$$\text{then } AB = C + D \quad (C \cap D = \emptyset)$$

$$\begin{aligned}\therefore P(AB) &= P(C) + P(D) \\ &= 2 \times \frac{3}{2^3} = \frac{6}{8}\end{aligned}$$

$$\begin{aligned}\text{Thus, } P(B|A) &= \frac{P(AB)}{P(A)} \\ &= \frac{6/8}{7/8} = \frac{6}{7}.\end{aligned}$$

e.g.2 The probability for the dog that can live more than 8 years **is 0.8**, and that more than 10 years **is 0.4**. Given one dog who is already **8** years, what is the probability for it can live more than **10** years?

Solution: Suppose $A = \text{“live more than 8 years”}$;
 $B = \text{“live more than 10 years”}$;

$$\text{then } P(B|A) = \frac{P(AB)}{P(A)} \cdot \quad (\because B \subset A, \therefore AB = B)$$

Because $P(A) = 0.8$, $P(B) = 0.4$, $P(AB) = P(B)$,

$$\text{Thus, } P(B|A) = \frac{P(AB)}{P(A)} = \frac{0.4}{0.8} = \frac{1}{2}.$$

3. Properties of conditional probability

(1) non-negativity: $0 \leq P(A|B) \leq 1$;

Proof: $\because AB \subset B \therefore 0 \leq P(AB) \leq P(B)$

$$\because P(B) > 0 \quad \therefore 0 \leq \frac{P(AB)}{P(B)} \leq 1$$

i.e., $0 \leq P(A|B) \leq 1$.

(2) normalization : $P(\Omega|B) = 1$;

Proof: $\because \Omega B = B$

$$\therefore P(\Omega|B) = \frac{P(\Omega B)}{P(B)} = \frac{P(B)}{P(B)} = 1$$

(3) additivity :

Given $A_i, A_j, i, j=1,2,\dots$, are mutually exclusive, i.e., $A_i \cap A_j = \phi$

$$\text{Then, } P\left(\left(\sum_{k=1}^{\infty} A_k\right) \middle| B\right) = \sum_{k=1}^{\infty} P(A_k | B)$$

$$\text{Proof: } P\left(\left(\sum_{k=1}^{\infty} A_k\right) \middle| B\right) = \frac{P\left(\left(\sum_{k=1}^{\infty} A_k\right) B\right)}{P(B)}$$

$$= \frac{\sum_{k=1}^{\infty} P(A_k B)}{P(B)} = \sum_{k=1}^{\infty} P(A_k | B)$$

$$(4) : P((A_1 \cup A_2)|B) \\ = P(A_1|B) + P(A_2|B) - P(A_1A_2|B)$$

Proof: Since the denominators are the same, it is equivalent to prove the numerators are the same

$$P((A_1 \cup A_2)B) = P(A_1B \cup A_2B) \\ = P(A_1B) + P(A_2B) - P(A_1A_2B)$$

(5) The conditional probability of the opposite event:

$$P(A|B) = 1 - P(\bar{A}|B).$$

4. Multiplication

if $P(B) > 0$, then $P(AB) = P(B)P(A|B)$

if $P(A) > 0$, then $P(AB) = P(A)P(B|A)$

Meaning:

The probability of the event AB **equals the probability of one event(i.e.,A) times the conditional probability of the other event(i.e., B) given the event(i.e., A) happens** .

Generalization: Given A, B, C , and $P(AB) > 0$, then

$$P(ABC) = P(A)P(B|A)P(C|AB).$$

In general, suppose A_1, A_2, \dots, A_n are events, if

$$P(A_1 A_2 \cdots A_{n-1}) > 0,$$

then

$$P(A_1 A_2 \cdots A_n) = P(A_1)P(A_2|A_1)P(A_3|A_1 A_2) \cdots \\ \cdots P(A_n|A_1 A_2 \cdots A_{n-1}).$$

二、 Theory of the total probability and the Bayes' Rule

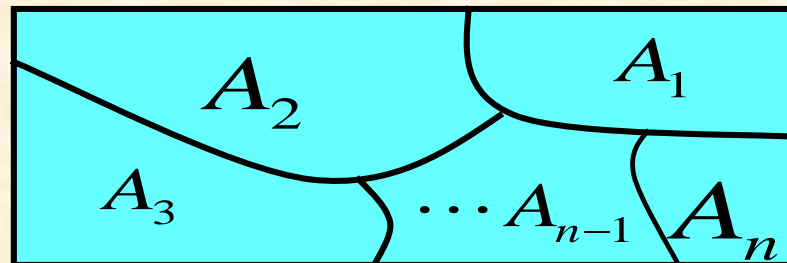
1. The partition of the sample space

Given A_1, A_2, \dots, A_n are n events of E , and Ω : sample space.
If

$$(1) \quad A_i A_j = \emptyset, \quad i \neq j, \quad i, j = 1, 2, \dots, n;$$

$$(2) \quad A_1 \cup A_2 \cup \dots \cup A_n = \Omega,$$

then, A_1, A_2, \dots, A_n constitute a partition of the sample space.



2. Theory of the total probability

If the events A_1, \dots, A_n constitute a partition of the sample space Ω such that $P(A_i) \neq 0$, then for any event B of Ω ,

$$\begin{aligned} P(B) &= P(B | A_1)P(A_1) + P(B | A_2)P(A_2) \\ &\quad + \dots + P(B | A_n)P(A_n) \\ &= \sum_{i=1}^n P(A_i)P(B | A_i) \end{aligned}$$



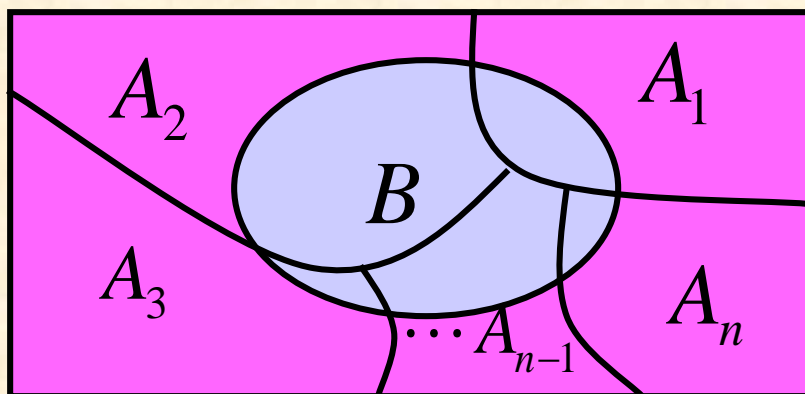
Total probability

Proof:
$$\begin{aligned} B &= B\Omega = B \cap (A_1 \cup A_2 \cup \cdots A_n) \\ &= BA_1 \cup BA_2 \cup \cdots \cup BA_n. \end{aligned}$$

Since $A_i A_j = \emptyset \Rightarrow (BA_i)(BA_j) = \emptyset$

$$\Rightarrow P(B) = P(BA_1) + P(BA_2) + \cdots + P(BA_n)$$

$$\Rightarrow P(B) = P(A_1)P(B | A_1) + P(A_2)P(B | A_2) + \cdots + P(A_n)P(B | A_n)$$



**breaking up
the whole
into parts**

Note. The condition of the total probability:

$$\sum_{i=1}^n A_i = \Omega$$

can be replaced by

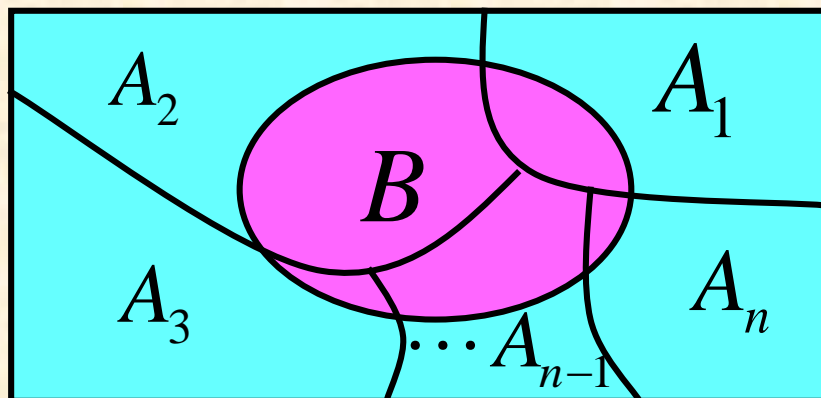
$$B \subset \sum_{i=1}^n A_i.$$

3. Meaning of the total probability

If the event B happening is due to n events A_i ($i=1,2,\dots,n$), and A_i, A_j ($i \neq j$) are **mutually exclusive**, then

$P(B)$ has relationship with $P(BA_i)$ ($i=1,2,\dots,n$).

$$P(B) = \sum_{i=1}^n P(BA_i).$$



e.g.4

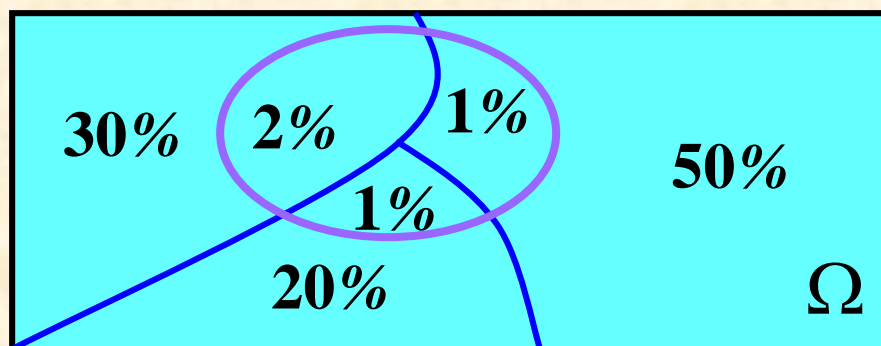
Given a batch of products coming from **three firms**, 30%, 50% and 20% of the products are from firm 1, firm 2 and firm 3, respectively. The defect rates of the three firms are 2%, 1%, 1% respectively. **What is the probability that one product selected randomly is defected?**

Solution: A = “one product selected randomly is defected”.

B_i = “the product selected randomly is from firm i , $i=1,2,3$ ”.

then, $B_1 \cup B_2 \cup B_3 = \Omega$,

$B_i B_j = \emptyset, \quad i, j = 1, 2, 3.$



By the **theory of total probability** .

$$P(A) = P(B_1)P(A|B_1) + P(B_2)P(A|B_2) + P(B_3)P(A|B_3).$$

$$P(B_1) = 0.3, \quad P(B_2) = 0.5, \quad P(B_3) = 0.2,$$

$$P(A|B_1) = 0.02, \quad P(A|B_2) = 0.01, \quad P(A|B_3) = 0.01,$$

$$\text{Thus, } P(A) = P(B_1)P(A|B_1) + P(B_2)P(A|B_2) + P(B_3)P(A|B_3)$$

$$= 0.02 \times 0.3 + 0.01 \times 0.5 + 0.01 \times 0.2 = 0.013.$$

4. Bayes' Rule

Theorem If the events A_1, A_2, \dots, A_n constitute a partition of Ω , such that $P(A_j) \neq 0, j = 1, \dots, n$, then for any event B of Ω , with $P(B) \neq 0$,

$$P(A_i | B) = \frac{P(B | A_i)P(A_i)}{\sum_{j=1}^n P(B | A_j)P(A_j)}, \quad i = 1, 2, \dots, n.$$

Such a formula is called **Bayes' Rule**.

Proof: $P(A_i|B) = \frac{P(A_i B)}{P(B)} \quad i = 1, 2, \dots, n.$

$$= \frac{P(B | A_i)P(A_i)}{P(B)}$$

$$= \frac{P(A_i)P(B | A_i)}{\sum_{i=1}^n P(A_i)P(B | A_i)}$$

e.g. 5 One box contains **12 balls** , **9** are new, **3** old.

In the first set **3 balls are used** . After the first set, the balls are **put back**. Then the second set begins, still 3 balls are chosen from the 12 balls.

Question? Determine the probability of the following events.

(1) B=“in the **second** set, all the **3** balls chosen are **new**.”;

(2) Given the 3 balls chosen in the second set are new, what is the probability of the event that **all the three balls** chosen **in the first set** are **new** ;

Solution: suppose A_i = “in the first set, there are i new balls are chosen ” ($i = 0, 1, 2, 3$)

(1) in the **second** set, all the **3** balls chosen are new;

A_i = “first set, i new balls”

B = “second set, all 3 balls are new”

$$\therefore P(A_i) = \frac{C_9^i \cdot C_3^{3-i}}{C_{12}^3} \quad (i = 0, 1, 2, 3)$$

$$P(B|A_i) = \frac{C_{9-i}^3}{C_{12}^3}$$

$$\begin{aligned} \therefore P(B) &= \sum_{i=0}^3 P(A_i)P(B|A_i) \\ &= \sum_{i=0}^3 \frac{C_9^i \cdot C_3^{3-i}}{C_{12}^3} \cdot \frac{C_{9-i}^3}{C_{12}^3} = 0.146. \end{aligned}$$

First set

new: 9

old: 3

Second set

new: $9-i$

old: $3+i$

**(the new balls
used become
old)**

(2) Given the 3 balls chosen in the second set are new, what is the probability of the event that **all the three balls** chosen **in the first set** are **new** ;

$$P(A_3|B) = \frac{P(A_3B)}{P(B)}$$

$$\therefore P(A_3B) = P(A_3)P(B|A_3)$$

$$= \frac{C_9^3 \cdot C_3^0}{C_{12}^3} \cdot \frac{C_6^3}{C_{12}^3}$$

$$\therefore P(A_3|B) = \frac{P(A_3B)}{P(B)} = \frac{5}{21} = 0.24$$

If in the first set,
all the balls
chosen are new,
then in the
second set

new: 9-3

old: 3+3

(balls are put
back)

三、Conclusion

1. Conditional probability

$$P(B|A) = \frac{P(AB)}{P(A)} \longrightarrow \text{multiplication principle}$$
$$P(AB) = P(A)P(B|A)$$

Theory of total probability

$$P(A) = P(B_1)P(A|B_1) + P(B_2)P(A|B_2) + \cdots + P(B_n)P(A|B_n)$$

Bayes' Rule

$$P(B_i|A) = \frac{P(B_i)P(A|B_i)}{\sum_{j=1}^n P(B_j)P(A|B_j)}, \quad i = 1, 2, \cdots, n.$$

2. The difference between $P(A|B)$ and $P(AB)$.

$P(AB)$ is calculated by considering Ω , while $P(B|A)$ is calculated by considering Ω_A .

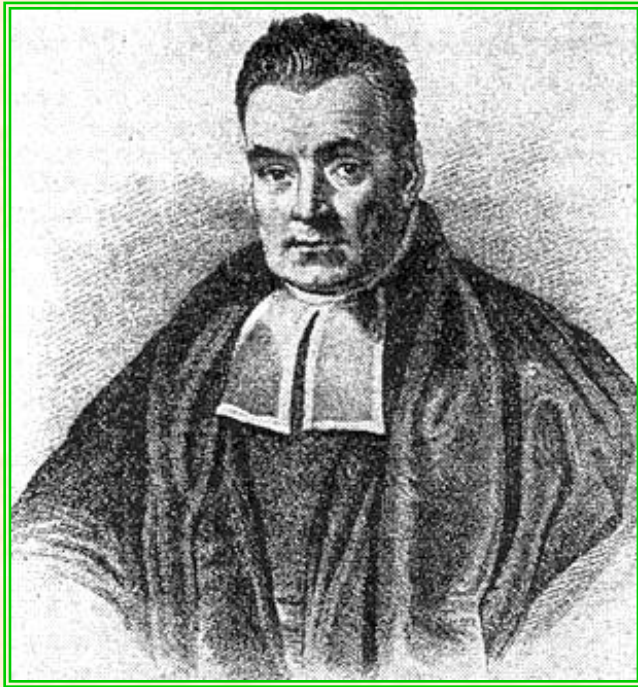
For classical probability,

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

$$P(AB) = \frac{|AB|}{|\Omega|}.$$

Normally, $P(AB)$ is smaller than $P(B|A)$.

Information of Bayes



Thomas Bayes

**Born: 1702 in London,
England**

**Died: 17 April 1761 in
Tunbridge Wells, Kent,
England**