

Probability — deductive reasoning

population

sample

$\Pr(H) = \Pr(T) = 0.5$

ten times

$\Pr(\text{all heads}) = (0.5)^{10}$

Statistical Inference — inductive reasoning

population

sample

$\Pr(H) = \Pr(T) = ?$

ten times

All heads.
Is it a fair coin?

Beginning of Probability

C. de Mére

Pascal

Fermat

Head → A get 1 point
Tail → B get 1 point
4 points → win \$100
Interrupted at 3:2

\$100

A \$75
B \$25

$\Pr(\text{B win}) = \Pr(\text{both tails}) = 0.25$

Classical School

N equally probable possible outcomes

$\Pr(E) = \frac{M}{N}$

$\Pr(\text{heart}) = \frac{13}{52} = 0.25$

$\Pr(\text{odd number from dice toss}) = \frac{3}{6} = 0.5$

$\Pr(\text{different birthdays}) = \frac{365 \times 364 \times \dots \times (365 - n + 1)}{365 \times 365 \times \dots \times 365} = \frac{P_n}{(365)^n}$

Frequency School

Results of coin toss experiment

| | | | | |
|-----------------------------|-----|------|-------|----------|
| No. of Tosses | 5 | 100 | 1000 | 1000000 |
| No. of Heads | 3 | 47 | 508 | 499987 |
| Relative Frequency of Heads | 0.6 | 0.47 | 0.508 | 0.499987 |

relative frequency of Heads $\rightarrow 0.5$ as $n \rightarrow \infty$

$\Pr(E) \stackrel{\text{define}}{=} \lim_{n \rightarrow \infty} \frac{f}{n}$

Subjective School

Probabilities are quantitative expressions of uncertain about a person's knowledge of the occurrence of some event.

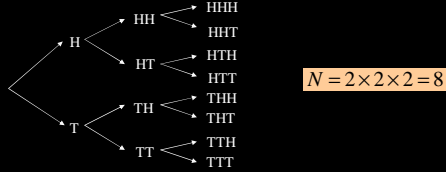
"I thought there is 30% chance that tomorrow will rain."

"Arsenal will have more than 80% chance to win the championship this year."

"I wished I had worked hard in last semester. I think I should have 80% chance to get better grade if I did so."

Counting Procedures

Example: Possible outcomes of tossing three coins



$$N = 2 \times 2 \times 2 = 8$$

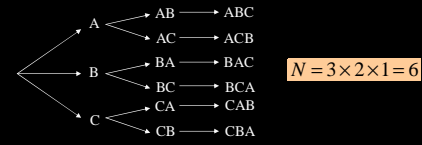
n_1 ways in 1st step n_2 ways in 2nd step n_k ways in kth step

$$n_1 n_2 \cdots n_k \text{ ways in } k \text{ steps}$$

Permutation

A specific arrangement of objects in a definite order.

Example: possible arrangements of three persons in a row

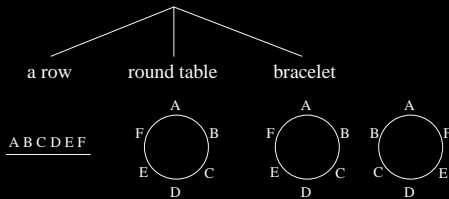


$$N = 3 \times 2 \times 1 = 6$$

$$\text{No. of permutations of } n \text{ distinct objects} \\ = n(n-1)\cdots(2)(1) = n!$$

Permutation

6 objects: A B C D E F

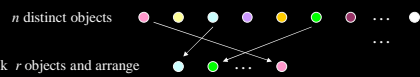


$$6! = 720$$

$$(6-1)! = 120$$

$$(6-1)!/2 = 60$$

Permutation



$$\text{No. of ordered subset:} \\ {}_n P_r = n(n-1)\cdots(n-r+1) = \frac{n!}{(n-r)!}$$

Example: Tierce outcomes in horse race with 14 horses

$$5 \ 6 \ 14$$

$$7 \ 1 \ 2$$

$$1 \ 2 \ 7$$

$$N = {}_{14}P_3 = \frac{14!}{11!} = 2184$$

Combination

A subset of r objects from n distinct objects, without regard to order.

Example: Choose three letters from {A, B, C, D}

| Combination | Permutations |
|-------------|--|
| {A, B, C} | (A, B, C), (A, C, B), (B, A, C), (B, C, A), (C, A, B), (C, B, A) |
| {A, B, D} | (A, B, D), (A, D, B), (D, A, B), (D, B, A), (B, A, D), (B, D, A) |
| {A, C, D} | (A, C, D), (A, D, C), (C, A, D), (C, D, A), (D, A, C), (D, C, A) |
| {B, C, D} | (B, C, D), (B, D, C), (C, B, D), (C, D, B), (D, B, C), (D, C, B) |

$${}_4 C_3 = \frac{{}_4 P_3}{3!} = \frac{24}{6} = 4$$

$${}_4 P_3 = \frac{4!}{(4-3)!} = \frac{24}{1} = 24$$

$${}_n C_r = \frac{{}_n P_r}{r!} = \frac{n!}{r!(n-r)!}$$

Combination

Example: 6 numbers chosen from {1, 2, ..., 49}



$$N = {}_{49} C_6 = \frac{49!}{6!43!} = 13983816$$

Example: draw five poker cards, Pr(full house) = ?



Full house:

$$\text{no. of possible outcomes: } N = {}_{52} C_3 = 2598960$$

$$\text{no. of possible full houses: } M = {}_{13} C_2 \times 2 \times {}_4 C_3 \times {}_4 C_2 = 3744$$

$$\text{Pr(full house)} = \frac{3744}{2598960} = 0.00144$$

Set

Set — collections of objects $E = \{ \text{dice}, \text{dice}, \text{dice}, \text{dice}, \text{dice} \}$

Element — object in the set $\text{dice} \in E$

Venn Diagram

$B \subseteq A$
B is a subset of A

Set operations

Union
 $A \cup B$

$x \in A \cup B$
equivalent to
 $x \in A$ or $x \in B$

Intersection
 $A \cap B$

$x \in A \cap B$
equivalent to
 $x \in A$ and $x \in B$

Complement
 \bar{A}

$x \in \bar{A}$
equivalent to
 $x \notin A$

Set operations

$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$

Set operations

$A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$

Event

Sample Space — all possible outcomes

Event — subset of sample space

Ω

$\Omega = \{ \text{outcomes of a die toss} \}$
 $E = \{ \text{even number} \}$

Event

Disjoint — two events have no common element

$A \cap B = \emptyset$

Ω

$A \cap C \neq \emptyset$

$A = \{ \text{even number} \}$
 $B = \{ \text{odd number} \}$
 $C = \{ \text{multiple of 3} \}$

Event

Mutually exclusive Exhaustive

Partition

Occurrence — E occurs if the outcome belongs to E.

outcome

occur not occur

$A \cap C$ $A \cap B \cap C$

$A \cap C$

$A = \{ \text{even number} \}$
 $B = \{ \text{odd number} \}$
 $C = \{ \text{multiple of 3} \}$

Komogorov's Axiom System

$\Pr(E) \geq 0$ for any event E $\Pr(\Omega) = 1$

$\Pr(A \cup B) = \Pr(A) + \Pr(B)$ if $A \cap B = \emptyset$

$\Pr(\emptyset) = 0$ If $A \subseteq B$, then $\Pr(A) \leq \Pr(B)$

$\Pr(\bar{A}) = 1 - \Pr(A)$ $\Pr(E) \leq 1$ for any event E

$\Pr(A \cup B) = \Pr(A) + \Pr(B) - \Pr(A \cap B)$

Example

$\Pr(M) = 0.1$ $\Pr(M \cap C) = 0.05$

$\Pr(A) = 0.2$ $\Pr(M \cap A) = 0.03$ $\Pr(M \cap C \cap A) = 0.01$

$\Pr(C) = 0.25$ $\Pr(C \cap A) = 0.1$

$\Pr(\text{none of these behaviors}) = 0.62$

$\Pr(M \cup C \cup A) = 0.38$

Conditional Probability

randomly pick one

$\Pr(\text{red circle}) = 5/10 = 0.5$
 $\Pr(\text{blue triangle}) = 3/10 = 0.3$
 $\Pr(\text{red}) = 6/10 = 0.6$
 $\Pr(\text{triangle}) = 4/10 = 0.4$

You are told that it is blue. You are told that it is red.

$\Pr(\text{triangle}) = 3/4 = 0.75$ $\Pr(\text{triangle}) = 1/6 = 0.17$

$\Pr(\text{triangle} | \text{blue}) = \frac{3/10}{4/10} = \frac{\Pr(\text{blue triangle})}{\Pr(\text{blue})}$ $\Pr(\text{triangle} | \text{red}) = \frac{1/10}{6/10} = \frac{\Pr(\text{red triangle})}{\Pr(\text{red})}$

Conditional Probability

For any two events A and B, the conditional probability of A given that B has occurred is written as $\Pr(A|B)$ and defined as

$$\Pr(A|B) = \frac{\Pr(A \cap B)}{\Pr(B)}$$

Multiplication theorem

$$\Pr(A \cap B) = \Pr(B) \Pr(A|B)$$

$$\Pr(A \cap B \cap C) = \Pr(C) \Pr(B|C) \Pr(A|B \cap C)$$

Independence

Two events A and B are called *independent* if and only if

$$\Pr(A \cap B) = \Pr(A)\Pr(B)$$

$$\Rightarrow \Pr(B) = \Pr(B|A) \text{ if } \Pr(A) > 0$$

Example : opinion poll on abortion

| | Favor (F) | Opposed (O) |
|---------------|-----------|-------------|
| White (W) | 0.459 | 0.441 |
| Non-white (N) | 0.051 | 0.049 |

$$\Pr(W) = 0.459 + 0.441 = 0.9$$

$$\Pr(F) = 0.459 + 0.051 = 0.51$$

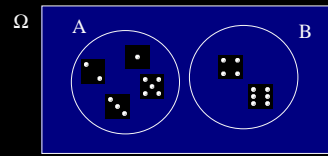
$$\Pr(W \cap F) = 0.459 = \Pr(W)\Pr(F) = 0.459$$

W and F are independent.

Independence vs Mutually Exclusive

Disjoint – two events have no common element

$$A \cap B = \emptyset$$



A and B are not independent !

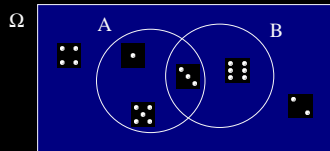
$$\Pr(A \cap B) = 0$$

$$\neq \Pr(A)\Pr(B) \text{ if } \Pr(A) > 0, \Pr(B) > 0$$

Independence vs Mutually Exclusive

Independence – events have no addition information to each other

$$\Pr(A \cap B) = \Pr(A)\Pr(B)$$



A = { odd }

B = { multi of 3 }

$$A \cap B = \{3\} \neq \emptyset$$

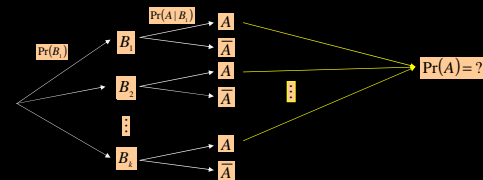
A and B are independent !

$$\Pr(A \cap B) = 1/6 = 1/2 \times 1/3 = \Pr(A)\Pr(B)$$

Law of Total Probability

B_1, B_2, \dots, B_k — partition of the sample space

mutually exclusion — cannot occur at the same time
exhaustive — at least one of them will occur



$$\Pr(A) = \Pr(B_1)\Pr(A|B_1) + \dots + \Pr(B_k)\Pr(A|B_k)$$

Law of Total Probability

Example: Morse code transmission

Bayes

dot : dash = 3 : 4

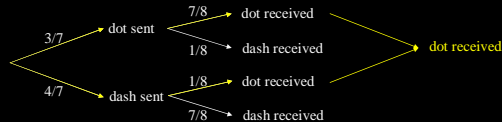
$$\Pr(\text{dot}) = 3/7$$

$$\Pr(\text{dash}) = 4/7$$

error rate = 1/8

$$\Pr(\text{receive dot} | \text{dash}) = 1/8$$

$$\Pr(\text{receive dash} | \text{dot}) = 1/8$$



$$\Pr(\text{receive dot}) = \frac{3}{7} \times \frac{7}{8} + \frac{4}{7} \times \frac{1}{8} = \frac{25}{56}$$

$$\Pr(\text{dot sent} | \text{dot received}) = ?$$

Bayes Rule

B_1, B_2, \dots, B_k — partition of the sample space

$$\Pr(B_i | A) = \frac{\Pr(B_i \cap A)}{\Pr(A)} = \frac{\Pr(B_i)\Pr(A|B_i)}{\Pr(A)}$$

$$= \frac{\Pr(B_i)\Pr(A|B_i)}{\Pr(B_1)\Pr(A|B_1) + \Pr(B_2)\Pr(A|B_2) + \dots + \Pr(B_k)\Pr(A|B_k)}$$

Example: Morse code transmission

$$\Pr(\text{dot sent} | \text{dot received}) = \frac{\Pr(\text{dot sent})\Pr(\text{dot received} | \text{dot sent})}{\Pr(\text{dot received})}$$

$$= \frac{3/7 \times 7/8}{25/56} = 0.84$$

Bayes Rule

Example: virus test

A — Test shows positive result as carrier

B — Infected by the virus

0.3% infected → $\Pr(B) = 0.003$

2% false positive → $\Pr(A | \bar{B}) = 0.02$

1% false negative → $\Pr(\bar{A} | B) = 0.01$

prior

$$\Pr(B | A) = \frac{\Pr(B) \Pr(A | B)}{\Pr(B) \Pr(A | B) + \Pr(\bar{B}) \Pr(A | \bar{B})}$$
$$= \frac{0.003 \times 0.99}{0.003 \times 0.99 + 0.997 \times 0.02} = 0.1296$$

posterior