

Why Probability?

We want to combine uncertain statements/observations to end up with logical inferences.

- frequentist
- Bayesian approach (subjective)

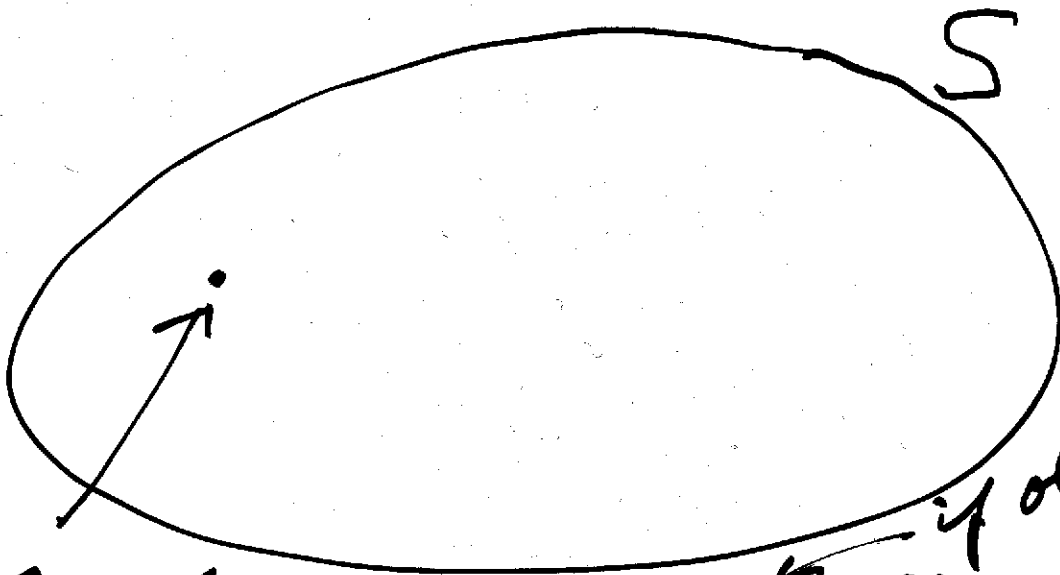
We talk about probabilities of events & denote the probability of an event A by $P(A)$.

Impossible Event - $\emptyset \rightarrow P(\emptyset) = 0$

Sure event $- S \rightarrow P(S) = 1$
(Ω)

Events either occur or they do not.

Suppose we are interested in the analysis of some situation. We decide what we want to observe. The set of ^(ideal) possible observations is called the sample space. It is usually denoted by S . Note that S is a set of outcomes or sample points.



sample point $s, \omega, \underline{x}, \underline{z}, g$ if observing #'s vectors

eg depth of a lake at a certain point over a certain time



You might be interested in

$$X = \max_{a \leq t \leq b} g(t)$$

Notice $X: S \rightarrow \mathbb{R}$. This is a random variable (rv)

Another rv is

$$Y = \min_{a \leq t \leq b} g(t)$$

The vector

$$\begin{pmatrix} X \\ Y \end{pmatrix} : S \rightarrow \mathbb{R}^2$$

is a random vector (rvec).

eg Toss a coin & observe
H or T.

$$S = \{ \underset{\uparrow}{H}, \underset{\uparrow}{T} \}$$

sample points

Set $X(H) = 1$ & $X(T) = 0$.

This is the rv "# of H's
on one toss".

eg Toss a coin 3 times & observe the sequence of faces.

$$S = \{HHH, TTT, THT, \dots\}$$

8 sample points

Let $X = \#$ of H's tossed.

Then X is a rv. The possible values of X are 0, 1, 2, 3

Now measure X . This yields a number x .

eg Take a course & your grade is observed.

$$S = \{x \mid 0 \leq x \leq 100\} = [0, 100]$$
$$= \{\omega \mid 0 \leq \omega \leq 100\}$$

Let $X =$ your grade. Notice

$$X(\omega) = \omega \quad X(x) = x$$

eg Roll a die & observe the
of dots.

$$S = \{1, 2, 3, 4, 5, 6\}$$

↑
a sample point

Consider the subset

$$A = \{2, 4, 6\}$$

= the event of rolling
an even # of dots

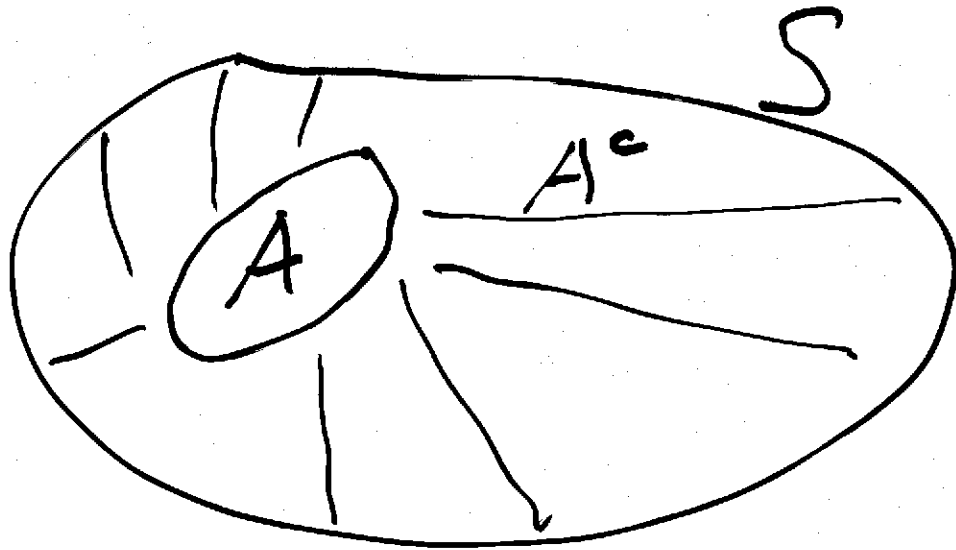
= { rolling an even #
of dots }

Def'n An event is a subset
of S .

Note $\circ S$ is an event (the sure event)

- ② An event occurs if and only if the observation is an element of the event.
- ③ The empty set never occurs & is the impossible event \emptyset .

We need to know more about sets / events.



The event A is made up of "possible outcomes", i.e. sample points.

A^c is made up of points outside A .

Let A & B be events. If they are useful in predicting each other they are called dependent events. Otherwise, they are independent.

Notation AB (or $A \overset{\text{and/intersection}}{\cap} B$)
= the event that both A & B occur.

A_1, A_2, \dots = the event that all the A_i 's occur.

$A \cup B$ = the event that at least one of the two or/union occurs.

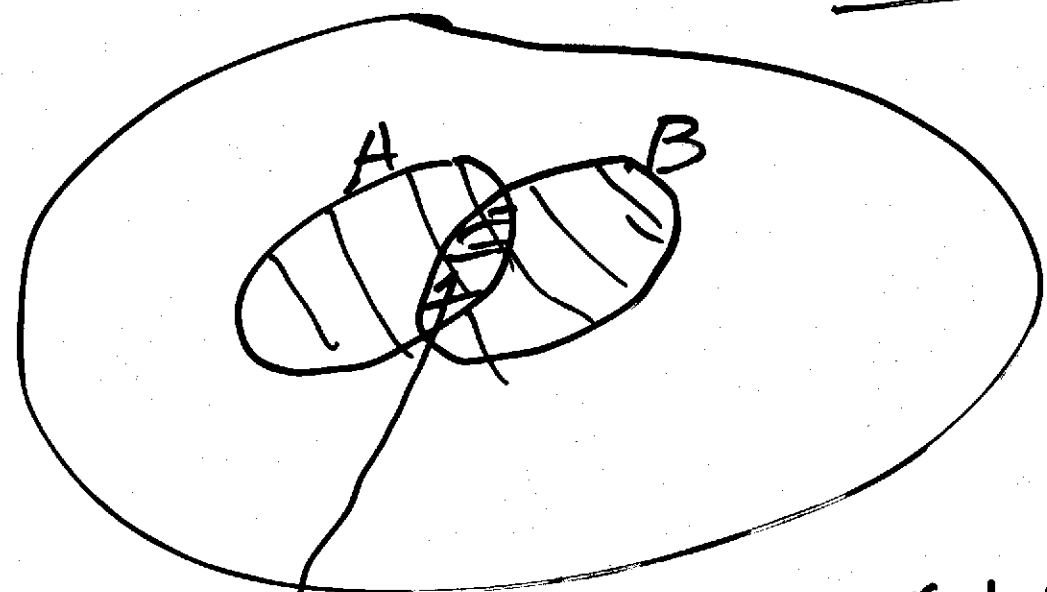
$A_1 \cup A_2 \cup \dots$ = the event that at

least one of the A_i 's occurs.

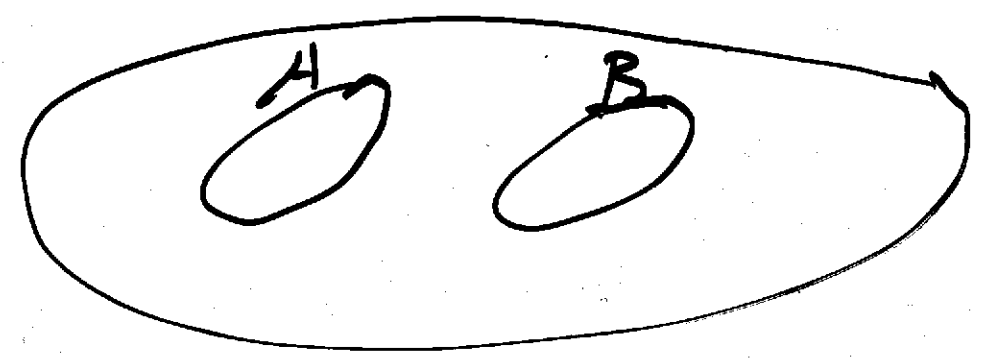
Notation $\bigcup_{i=1}^{\infty} A_i$, $\bigcap_{i=1}^{\infty} A_i$

Note

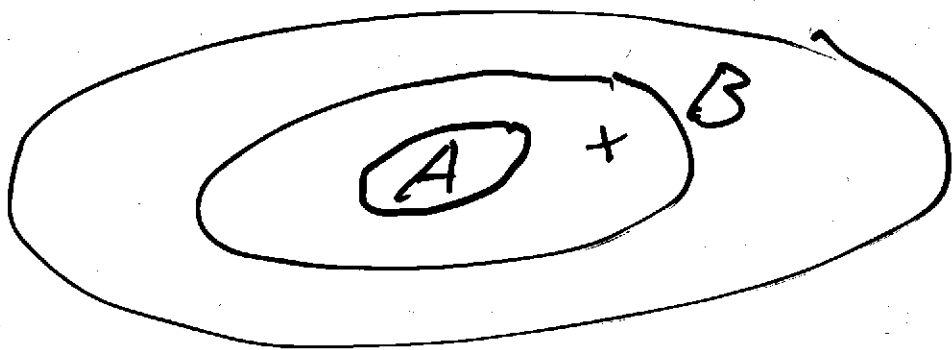
Venn diagram



AB $\parallel \parallel = A \cup B$



no overlap = disjoint



A occurs implies B occurs,

$$A \Rightarrow B$$

$$(A \subset B)$$

Note Suppose $A \Rightarrow B$
and $B \Rightarrow A$. Then
 $A = B$

DeMorgan's Laws

$$(1) \left[\bigcap_i A_i \right]^c = \bigcup_i A_i^c \leftarrow \text{try to show}$$

$$(2) \left(\bigcup_i A_i \right)^c = \bigcap_i A_i^c$$

Def'n A & B are independent
if $P(AB) = P(A)P(B)$

Def'n A_1, A_2, A_3, \dots are
ind if for any $A_{i_1}, A_{i_2}, \dots, A_{i_m}$
 $P(A_{i_1}, A_{i_2}, \dots, A_{i_m})$
 $= P(A_{i_1})P(A_{i_2}) \dots P(A_{i_m})$

Kolmogorov Axioms for P

1 $P(S) = 1$

2 $P(A) \geq 0$

3 $P(A_1 \cup A_2 \cup \dots) = P(A_1) + P(A_2) + \dots$
↑ ↗ ↘
disjoint

Proposition $P(\emptyset) = 0$
↑
empty event

Proof Let A be an event. Then

$$A = \underset{\uparrow}{A} \cup \underset{\nearrow}{\emptyset} \cup \underset{\nearrow}{\emptyset} \cup \dots$$

disjoint

\Rightarrow implies $P(A) = P(A \cup \emptyset \cup \dots)$
 $= P(A) + P(\emptyset) + \dots$, by Law 3 of P

$\Rightarrow P(\emptyset) = 0$

"The End" "□" "qed"

Proposition If A_1, \dots, A_m are disjoint then

$$P(A_1 \cup \dots \cup A_m) = P(A_1) + \dots + P(A_m)$$

Proof $P(A_1 \cup \dots \cup A_m) = P(A_1 \cup \dots \cup A_m \cup \emptyset \cup \dots)$

$$= P(A_1) + \dots + P(A_n) + \underbrace{P(\emptyset)}_0 + \underbrace{P(\emptyset)}_0 + \dots$$

$$= P(A_1) + \dots + P(A_n)$$

qed

Application

Roll a fair die & observe the # of dots. Then

$$P(\{3\}) = \frac{1}{6}$$

Proof $S = \{1\} \cup \{2\} \cup \{3\} \cup \dots \cup \{6\}$

↑ disjoint

$$\Rightarrow P(S) = \underbrace{P(\{1\}) + \dots + P(\{6\})}_{=1}$$

↙ equal
die is fair

$$= 6 P(\{3\})$$

$$\Rightarrow P(\{3\}) = \frac{1}{6}$$

qed

Important The previous argument shows that a finite sample space with equiprobable sample points yields

$$P(A) = \frac{|A|}{|S|}$$

of outcomes in A

A bit of counting

$$3 \times 2 \times 1 = 3! \quad (3 \text{ factorial})$$

$$n \times (n-1) \cdots \times 2 \times 1 = n!$$

$$0! = 1$$

$$\binom{n}{k} = \frac{n!}{k! (n-k)!}$$

"n choose k"

$$(x_1 + x_2)^m = \sum_{k=0}^m \binom{m}{k} x_1^k x_2^{m-k}$$

binomial theorem

$\binom{m}{k}$ = # of subsets of size k
of set of m elements

So $\binom{m}{0} + \binom{m}{1} + \binom{m}{2} + \dots + \binom{m}{m}$
= total # of subset of a set
of size m
= $(1+1)^m = 2^m$

Set

$$\binom{m}{n_1, n_2, \dots, n_k} = \frac{m!}{n_1! \dots n_k!}$$

where $n_1 + n_2 + \dots + n_k = m$

is the multinomial coefficient. ($k=2$ yields the binomial coefficient).

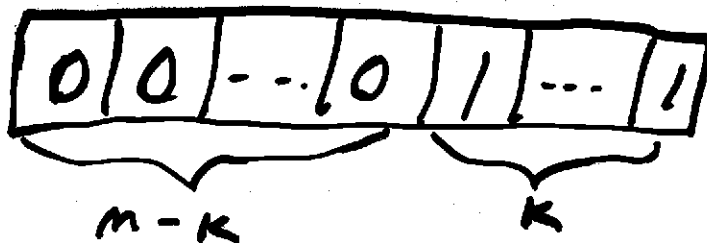
Then

$$(x_1 + \dots + x_k)^m = \sum_{m_1 + \dots + m_k = m} \binom{m}{m_1, \dots, m_k} x_1^{m_1} \dots x_k^{m_k}$$



people in boxes / cells

of arrangements = $m!$



of arrangements $\times (m-k)! \times (k!) = m!$

$$\Rightarrow \# \text{ of arrangements} = \frac{m!}{k! (m-k)!} = \binom{m}{k}$$

