

# Probability

## Sample spaces and events

Definition: The sample space of an experiment denoted by  $S$ , is the set of all possible outcomes of that experiment.

Example: observing the gender of the next child born at the local hospital, where  $S = \{M, F\}$

## Events

Definition: An event is any collection (subset) of outcomes contained in the sample space  $S$ .

## Properties of Probability

Given an experiment and a sample space  $S$ , the objective of probability is to assign to each event  $A$  a number  $P(A)$ , called the probability of the event  $A$ , which will give a precise measure of the chance that  $A$  will occur.

## Properties of Probability

- ① For any event  $A$ ,  $P(A) \geq 0$ .
- ②  $P(S) = 1$
- ③ If  $A_1, A_2, \dots, A_i, \dots$  is an infinite collection of disjoint events, then  $P(A_1 \cup A_2 \cup \dots) = \sum_{i=1}^{+\infty} P(A_i)$ .

$$\textcircled{4} P(\emptyset) = 0. \quad \textcircled{5} P(A) \leq 1, \forall A \in \mathcal{S}.$$

Consider an event  $A$ , with  $n(A)$  denoting the number of outcomes contained in  $A$ . Then

$$P(A) = \frac{n(A)}{N}, \text{ where } N \text{ is the out comes number of the experiments.}$$

## Discrete Random Variables and probability Distributions

Random variables: In any experiment, there are numerous characteristics that can be observed or measured, but in most cases an experimenter will focus on some specific aspect or aspects of a sample.

Definition: For a given sample space  $S'$  of some experiment, a random variable (r.v) is a function whose domain is the sample space and whose range is the set of real numbers.

Example

## Bernoulli random variable

Definition: Any random variable whose only possible values are 0 and 1 is called a Bernoulli random variable.

## Types of Random variables

We distinguished between two different types of random variables.

Definition: A discrete random variable is a random variable whose possible values either constitute a finite set or else be listed in an infinite sequence in which there is a first element, a second element, and so on.

A random variable is continuous if both of the following apply:

① Its set of possible values consists either of all numbers in a single interval on the number line (possible from  $-\infty$  to  $+\infty$ ) or all numbers in a disjoint union of such intervals (e.g.,  $[0, 10] \cup [20, 30]$ ).

② no possible value of the variable has positive probability, that is,  $P(X=c) = 0$  for any possible value  $c$ .

③

# Probability Distribution for discrete

## Random Variables

Example: Suppose that the probability distribution of  $X$  is as given in the following table; the first row of the table lists the possible  $X$  values and the second row gives the probability of each such value.

$x$	0	1	2	3	4	5	6
$P(x)$	0,05	0,10	0,15	0,25	0,20	0,15	0,10

$$\begin{aligned} \text{We have } \therefore P(X \leq 2) &= P(X=0 \text{ or } 1 \text{ or } 2) \\ &= P(0) + P(1) + P(2) \\ &= 0,05 + 0,10 + 0,15 = 0,30 \end{aligned}$$

$$\begin{aligned} P(X \geq 3) &= P(X=3 \text{ or } 4 \text{ or } 5 \text{ or } 6) \\ &= 1 - P(X \leq 2) = 1 - 0,3 = 0,7. \end{aligned}$$

Definition: The probability distribution or probability mass function is defined for every number  $x$  by  $P(x) = P(X=x) = P(\{\omega \in \Omega : X(\omega) = x\})$

Example: Six lots of components are ready to be shipped by a certain supplier. The number of defective components in each lot is as

Lot	1	2	3	4	5	6
Number of defectives	0	2	0	1	2	0

One of these lots is to be randomly selected for shipment to a particular customer.

Let  $X$  be the number of defectives in the selected lot. The three possible  $X$  values are: 0, 1, and 2

$$P(0) = P(X=0) = P(\text{lot 1 or 3 or 6 is sent}) \\ = \frac{3}{6} = 0,5$$

$$P(1) = P(X=1) = P(\text{lot 4 is sent}) = \frac{1}{6} = 0,167$$

$$P(X=2) = P(2) = P(\text{lot 2 or 5 is sent}) = \frac{2}{6} = 0,333$$

$X$	0	1	2
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$P(X)$	0,5	0,167	0,333
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The values of  $X$  along with their probabilities collectively specify the probability distribution.

## The Binomial Probability Distribution

There are many experiments that conform either exactly or approximately to the following list of requirements:

- ① The experiment consists of a sequence of  $n$  smaller experiments called trials.
- ② - Each trial can result in one of the same possible outcomes (dichotomous trials), which we generically denote by success (S) and failure (F).
- ③ The trials are independent so that the outcome on any particular trial does not influence the outcome on any other trial.
- ④ - The probability of success  $p(S)$  is constant from trial; we denote this probability by  $P$ .

Definition: An experiment for which conditions 1-4 are satisfied is called a binomial experiment.

Example: The same coin is tossed successively and independently  $n$  times. We arbitrarily use  $S$  to denote the outcome  $H$  (heads) and  $F$  to denote the outcome  $T$  (tails).

Then this experiment satisfies condition 4.

### The Binomial Random Variable and Distribution

Definition: The binomial random variable  $X$  associated with a binomial experiment consisting of  $n$  trials is defined as

$X =$  the number of  $S$ 's among the  $n$  trials.

Suppose, for example, that  $n = 3$ . Then there are eight possible outcomes for the experiment:

SSS SSF SFS SFF FSS FSF FFS FFF

From the definition of  $X$ ,  $X(SSS) = 3$

$X(SFF) = 1$ , and so on. Possible values for

$X$  in an  $n$ -trial experiment are  $0, 1, 2, \dots, n$ .

We will often write  $X \sim \text{Bin}(n, p)$  to indicate that  $X$  is a binomial r.v. based on  $n$  trials with success probability  $p$ .

Example: Consider first the case  $n=4$  for which each outcome, its probability, and corresponding  $x$  values are listed in the Table -A-

$$\begin{aligned} \text{For example, } P(SSFS) &= P(S) \cdot P(S) \cdot P(F) \cdot P(S) \\ &= p \times p \times (1-p) \times p \\ &= p^3(1-p). \end{aligned} \quad \left[ \begin{array}{l} \text{independent} \\ \text{trials} \end{array} \right]$$

Outcome	$x$	Probability	Outcome	$x$	Probability
SSSS	4	$p^4$	FSSS	3	$p^3(1-p)$
SSSF	3	$p^3(1-p)$	FSSF	2	$p^2(1-p)^2$
SSFS	3	$p^3(1-p)$	FSFS	2	$p^2(1-p)^2$
SSFF	2	$p^2(1-p)^2$	FSFF	1	$p(1-p)^3$
SFSS	3	$p^3(1-p)$	FFSS	2	$p^2(1-p)^2$
SFSF	2	$p^2(1-p)^2$	FFSF	1	$p(1-p)^3$
SFFS	2	$p^2(1-p)^2$	FFFS	1	$p(1-p)^3$
SFFF	1	$p(1-p)^3$	FFFF	0	0

We wish  $b(x, 4, p)$  (probability mass function) for  $x=0, 1, 2, 3$  and 4.

For  $b(3, 4, p)$ , we have: (From the table A-)

$$\begin{aligned} b(3, 4, p) &= P(FSSS) + P(SFSS) + P(SSFS) + P(SSSF) \\ &= 4 p^3(1-p) \end{aligned}$$

$$b(3; 4, p) = \left. \begin{array}{l} \text{number of outcomes} \\ \text{with } x=3 \end{array} \right\} \times \left. \begin{array}{l} \text{Probability of} \\ \text{any particular outcome} \\ \text{with } x=3 \end{array} \right\}$$

(8)



Exercise: Calculate: (From the table - A -)

$$b(2; 4, p), b(4; 4, p), b(1; 4, p).$$

Theorem

$$b(x; n, p) = \begin{cases} C_n^x p^x (1-p)^{n-x}, & x=0, 1, 2, \dots, n \\ 0 & \text{otherwise.} \end{cases}$$

Example: Each of six randomly selected cola drinkers is given a glass containing cola S and one containing cola F. The glasses are identical in appearance except for a code on the bottom to identify the cola. Suppose there is actually no tendency among cola drinkers to prefer one cola to the other, then  $p = 0.5$ , so with

$X =$  the number among the six who prefers S  $\Rightarrow X \sim B(6; 0.5)$

Thus

$$\begin{aligned} P(X=3) &= C_6^3 (0.5)^3 (0.5)^3 = \frac{6!}{3!3!} (0.5)^3 (0.5)^3 \\ &= \frac{6 \times 5 \times 4 \times 3!}{3! (2 \times 2)} (0.5)^3 \times (0.5)^3 \\ &= 20 (0.5)^6 = \boxed{0.313} \end{aligned}$$

The probability that at last three prefer s

$$P(X \leq 3) = \sum_{x=3}^6 b(x; 6, 0.5)$$

$$= \sum_{x=3}^6 C_6^x (0.5)^x (0.5)^{6-x}$$

$$= C_6^3 (0.5)^3 (0.5)^3 + C_6^4 (0.5)^4 (0.5)^2 + C_6^5 (0.5)^5 (0.5) + C_6^6 (0.5)^6$$

where:  $C_n^k = \frac{n!}{k!(n-k)!}$

$$n! = n(n-1)(n-2) \dots \times 2 \times 1.$$