

S2:

PROBABILITY:

Statistical theory and practice rest largely on the concept of probability.

Probability is a measure of estimation of likelihood of occurrence of an event.

Ex 1: If a fair coin is tossed, the probability of getting heads is 0.5.

Remark: Probabilities are between 0 and 1. (both included).

Ex 2: If a die is ~~thrown~~<sup>casted</sup>, the probability of getting a 3 is  $\frac{1}{6} = 0.1\hat{6}$  ( $\approx 0.166666\dots$ ).

Remark: Die (singular)  $\rightarrow$  Dice (Plural).

Any probability pertains to the results of a situation which we call an experiment. An experiment is any process by which data are obtained through the observation of uncontrolled events in nature or through controlled procedures in a laboratory.

Any experiment can result in various outcomes. The sample space ( $\Omega$ ) is the set of all possible outcomes that may occur as a result of a particular experiment.

Exs: Coin  $\rightarrow \Omega = \{ \text{heads, tail} \} = \{ H, T \}$ .

Die  $\rightarrow \Omega = \{ 1, 2, 3, 4, 5, 6 \}$ .

~~A subset of the sample space~~

An event is a subset of the sample space.

Ex: Die: ~~Event~~ •  $\{ \text{Even} \}$ ,  $\{ \text{Odd} \}$ ,  $\{ \geq 3 \}$ ,  $\{ 1 \}$ .

- Events can be combined. Given two events  $A$  and  $B$ , then:
  - their union  $A \cup B$
  - their intersection  $A \cap B$
  - the complementary  $A^c$ .

## THE PROBABILITY FUNCTION:

The probability function (denoted by  $P$ ) is the function that quantifies the likelihood of an event. Given an event  $A$ ,  ~~$P(A)$~~

Properties:

1.  $0 \leq P(A) \leq 1$ .

2.  $P(\Omega) = 1$ .

3. If  $A \cap B = \emptyset$  (2 exclusive events), then

$$P(A \cup B) = P(A) + P(B).$$

3'. If  $A_1, \dots, A_n$  are pair wise exclusive events ( $A_i \cap A_j = \emptyset$ ), then

$$P(A_1 \cup \dots \cup A_n) = P(A_1) + P(A_2) + \dots + P(A_n).$$

Notice that  $\forall A, A^c \cap A = \emptyset, A^c \cup A = \Omega$ . Hence,

$$P(A^c) + P(A) = 1 \Rightarrow P(A^c) = 1 - P(A).$$

Ex: In the die problem,

$$P(\{1,4\}) = \frac{1}{6}$$

$$P(\{1,2,4\}) = \frac{2}{6} = \frac{1}{3}$$

$$P(\{3,4,5,6\}) = \frac{4}{6} = \frac{2}{3}$$

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Notice that the coin and the die satisfy the property of equiprobability: The probability of all results are equal and are  $\frac{1}{n}$ , where  $n$  is the number of possible results. This phenomenon is quite common, but not always true: a biased coin (or die) will not have this property.

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• Another general property of the probability function is that, if two events are independent, then the probability of that both happen is the product of their probabilities.

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

Ex: ~~A~~ A fair die, we <sup>cast</sup> ~~throw~~ it twice. What is the probability that the first result is 1 and the second is 3?

$$P(F_1 \cap F_3) = P(F_1) P(F_3) = \frac{1}{6} \left( = \left( \frac{1}{6} \cdot \frac{1}{6} \right) \right).$$

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\*\*ADVANCED Conditional probability: Given two events A and B, the probability that A occurs given that the event B is certain to occur is denoted by  $P(A|B)$ , and is calculated by

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

Ex: If we know that the result of a die came up even, what is the probability of that 2 came up?

$$P(\{2\} | \{2, 4, 6\}) = \frac{P(\{2\} \cap \{2, 4, 6\})}{P(\{2, 4, 6\})} = \frac{\frac{1}{6}}{\frac{3}{6}} = \frac{1}{3}.$$

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## RANDOM VARIABLES:

We have seen in the previous examples that the results of experiments might be non-numerical (heads, tails) and numerical (1, 2, 3, 4, 5, 6).

A random variable is ~~the~~<sup>a</sup> quantification of the results of an experiment. A random variable is a numerical quantity the value of which is determined by an experiment.

Often we will denote random variables by capital letters  $X, Y, Z$ .

Ex: In the ~~die~~<sup>coin</sup> example, we can build a random variable  $X = \begin{cases} 0 & \text{heads} \\ 1 & \text{tails.} \end{cases}$

Ex: In the die example,  $X =$  result of casting the die.

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With the random variables we are still in the game of computing probabilities.

Ex: Die:  $P(X < 2) = (P(X=1)) = \frac{1}{6}$ .

$$P(X > 3) = (P(X=3 \text{ or } X=4 \text{ or } X=5 \text{ or } X=6)) = \frac{4}{6} = \frac{2}{3}.$$

Ex: Height of students.

$$P(X < 180) = ?$$

$$P(160 < X < 200) = ?$$

## DISCRETE RANDOM VARIABLES:

Notice that the possible results of casting a die or the height of students differ substantially. The height of students is quite special: what is the probability that  $P(X=190)$ ? Discuss it.

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Let  $p(x) = P(X=x)$ . We will say that  $X$  is a discrete random variable if

- $p(x) \geq 0 \forall x$ .
- $p(x) = 0$  for all  $x$  out of a finite or numerable set.
- $(\sum_i p(x) = 1.)$
- $(P(X \in A) = \sum_{x \in A} p(x).)$

Ex: • Die  $\rightarrow$  discrete.

• Height  $\rightarrow$  non-discrete.

• Coin  $\rightarrow$  discrete.

•  $X = \{ \text{the day of my next son's birth} \}$ . (Consider me as immortal! :-).  $\rightarrow$  That's a discrete (but not finite) random variable.

## Distribution function:

Given  $X$ , its distribution function  $F(x)$  is defined by  $F(x) = P(X \leq x)$ . [The probability that  $X \leq x$ ].

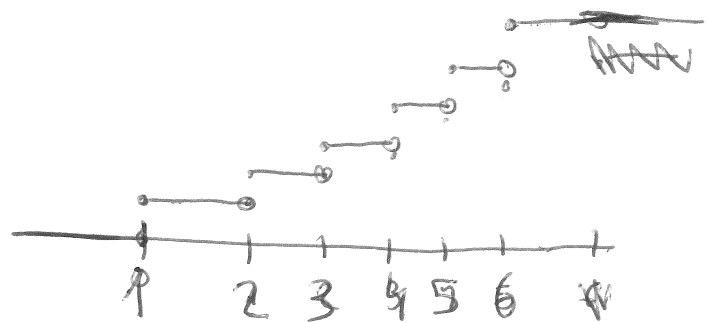
Sometimes distribution functions are called cumulative distribution functions.

### Properties:

- $0 \leq F(x) \leq 1$ .
- $\lim_{x \rightarrow -\infty} F(x) = 0$ ,  $\lim_{x \rightarrow +\infty} F(x) = 1$ .

### Example: Die:

$$F(x) = \begin{cases} 0 & x < 1 \\ 1/6 & x < 2 \\ 2/6 & x < 3 \\ 3/6 & x < 4 \\ 4/6 & x < 5 \\ 5/6 & x < 6 \\ 6/6 & x \geq 6. \end{cases}$$



### Independent random variables:

$X$  and  $Y$  are independent random variables if

$$P(a \leq X \leq b, c \leq Y \leq d) = P(a \leq X \leq b) \cdot P(c \leq Y \leq d).$$



## Expected value:

The expected value of a <sup>(discrete)</sup> random variable  $X$  is the "expected result" of it. It is the weighted mean of all its possible values by their probabilities.

$$E(X) = \sum_i x_i p(x_i) = \sum_i x_i P(X=x_i)$$

Ex. Die:  $E(X) = \frac{1}{6} \cdot 1 + \frac{1}{6} \cdot 2 + \frac{1}{6} \cdot 3 + \dots + \frac{1}{6} \cdot 6 = 3.5$ .

## Properties:

- $E(f(X)) = \sum_i f(x) p(x)$

- $E(aX+b) = aE(X)+b$

- $E(X+Y) = E(X)+E(Y)$

- If  $X_1, \dots, X_n$  form  $\bar{X} = \frac{1}{n}(X_1 + \dots + X_n)$  and  $E(X_i) = \mu$  then  $E(\bar{X}) = \mu$ .

The random variable  $\bar{X}$  will be called the mean random variable.

## Variance and standard deviation

Variance:  $\sigma^2(X) = E((X - E(X))^2) = \sum_i (x_i - \mu)^2 p(x)$

Standard deviation:  $\sigma(X) = \sqrt{\sigma^2(X)} = \sqrt{\sum_i (x_i - \mu)^2 p(x)}$

Properties:

$$\bullet \sigma^2(X) = E(X^2) - E(X)^2$$

$$\bullet \sigma^2(aX+b) = a^2 \sigma^2(X)$$

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Before finishing, let's point out several properties of the expected value and the variance of independent random variables:

$$\bullet E(XY) = E(X)E(Y)$$

$$\bullet \sigma^2(X+Y) = \sigma^2(X) + \sigma^2(Y)$$

$$\bullet X_1, \dots, X_n \sim N, \sigma^2 \Rightarrow E(\bar{X}) = \mu \text{ and } \sigma^2(\bar{X}) = \frac{\sigma^2}{n}$$