

# Probability Distribution Functions - Discrete variable

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## Discrete random variable: functions and properties

- Variable: counts of events.
- Example: number of women in 2nd year of the Biomedical Engineering degree.

### Probability mass function

The probability mass function assigns probability to each possible value of the random variable.

$$\begin{aligned} P : \mathbb{R} &\longrightarrow [0, 1] \\ k &\longrightarrow P(X = k) = p \end{aligned}$$

$$P(X = k) \in [0, 1]$$

$$\sum_{i=-\infty}^{+\infty} P(X = k) = 1$$

Example: number of boys in sets of 3 siblings. Assuming equiprobability and independence among births:

$X$  = number of boys.

$$P(\varphi) = P(\sigma) = 0.5$$

### Probability Distribution Function

The probability distribution function assigns **cumulative probability** to each possible value of the random variable.

$\Omega$	$X$	$p$
♀♀♀	0	0.125
♀♀♂	1	0.125
♀♂♀	1	0.125
♂♀♀	1	0.125
♀♂♂	2	0.125
♂♀♀	2	0.125
♂♂♀	2	0.125
♂♂♂	3	0.125

$X$	$P(X = k)$
0	0.125
1	0.375
2	0.375
3	0.125

$$\begin{aligned}
 F : \mathbb{R} &\longrightarrow [0, 1] \\
 k &\longrightarrow F(k) = P(X \leq k) = p
 \end{aligned}$$

$$P(k_1 < X \leq k_2) = F(k_2) - F(k_1)$$

$k$	$P(X = k)$	$P(X \leq k)$
0	0.125	0.125
1	0.375	0.5
2	0.375	0.875
3	0.125	1

## Expectation of a Random Variable $E(X)$

Expectation: the expected value of a random variable.

Prediction: a guess about the value of the random variable.

Prediction error: the difference between the prediction and the actual value of the random variable.

The expectation **minimizes the prediction error** when the number of predictions tend to infinite.

Discrete variable: expectation appears as a weighted mean.

$$E(X) = \sum_{k=-\infty}^{+\infty} k \cdot P(X = k)$$

- If we randomly selected a set of 3 siblings, how many boys do we expect to find?

$$E(X) = \sum_{k=0}^3 k \cdot P(X = k) =$$

$$0 \cdot P(X = 0) + 1 \cdot P(X = 1) + 2 \cdot P(X = 2) + 3 \cdot P(X = 3) =$$

$$0 \cdot 0.125 + 1 \cdot 0.375 + 2 \cdot 0.375 + 3 \cdot 0.125 = 1.5$$

The expectation is 1.5 boys. Notice that the expectation may be a value that the random variable cannot take.

### Properties of the Expectation

Let  $X$  and  $Y$  be two random variables and  $C$  a constant.

- Linearity.

$$E(X + C) = E(X) + C$$

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

- Multiplicity.

$$E(X \cdot C) = E(X) \cdot C$$

- If  $X$  and  $Y$  are independent:

$$E(X \cdot Y) = E(X) \cdot E(Y)$$

### Variance of a random variable

- Variance: indicates the dispersion of the variable around its expectation.

It is a statistical distance:

- How far are the values from the expectation?
- What is the mean distance of the random variable to its expectation?
- How much representative of the random variable is the expectation?

If the **variance is large** the data will be very **spread** and the expectation can not be taken as representative of the random variable.

- General expression

$$V(X) = E[(X - E(X))^2] = E(X^2) - E(X)^2$$

- Variance in discrete variables

$$V(X) = \sum_{k=-\infty}^{+\infty} k^2 \cdot P(X = k) - E(X)^2$$

- Example: number of boys in 3 siblings

$k^2$	$P(x = k)$	$k^2 P(x = k)$
0	0.125	0
1	0.375	0.375
4	0.375	1.5
9	0.125	1.125
		$\Sigma = 3$

$$V(X) = \sum_{k=0}^3 k^2 P(X = k) - E(X)^2 = 3 - 1.5^2 = 0.75$$

The random variable “number of boys in 3 siblings” has a expectation of 1.5 boys and a variance of 0.75 boys<sup>2</sup>.

### Properties of the Variance

Let  $X$  and  $Y$  be two random variables and  $C$  a constant.

- Invariant to changes in location parameter

$$V(X + C) = V(X)$$

- Scale

$$V(X \cdot C) = C^2 \cdot V(X)$$

- Variance of the sum of two random variables

$$V(X + Y) = V(X) + V(Y) + 2Cov(X, Y)$$

where  $Cov(X, Y)$  is the covariance which is a measure of *joint variability*.

$$Cov(X, Y) = E(XY) - E(X)E(Y)$$

If two variables, X and Y, are *independent*,  $Cov(X, Y) = 0$ .

So that, if two variables, X and Y, are independent:

$$V(X + Y) = V(X) + V(Y)$$

- Variance of the difference of two random variables

$$V(X - Y) = V(X) + V(Y) - 2Cov(X, Y)$$

# Probability distribution models

## Bernoulli

- Binary results. The variable usually takes values 0 or 1.
- Probability distribution defined by one single parameter  $p = P(X = 1)$ .
- Probability mass function

$$\begin{aligned}P(X = 1) &= p \\P(X = 0) &= 1 - p\end{aligned}$$

- Expectation and Variance

$$\begin{aligned}E(X) &= p \\V(X) &= p(1 - p)\end{aligned}$$

**Example. Flipping a coin.  $X = \text{“Head side”}$**

$$p = 0.5, E(X) = 0.5, V(X) = 0.5 \cdot 0.5 = 0.25$$

## Binomial

The addition of  $n$  independent and identically distributed Bernoulli variables.

$$X = \sum_{i=1}^n Y_i, \quad Y_i \sim \text{Bernoulli}(p)$$

- Range of values:  $\{0, 1, \dots, n\}$
- Probability mass function

$$P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}$$

- Expectation and Variance

$$\begin{aligned}E(X) &= np \\V(X) &= np(1 - p)\end{aligned}$$

**Example. Flipping 3 times a coin.**

X="number of heads" follows a  $Bin(3, 0.5)$

$$E(X) = n \cdot p = 3 \cdot 0.5 = 1.5$$

- Probability of 2 heads

$$P(X = 2) = \binom{3}{2} 0.5^2 0.5 = 0.375$$

- Probability of less than 2 heads

$$P(X < 2) = P(X \leq 1) = F(1) = P(X = 0) + P(X = 1) = \binom{3}{0} 0.5^0 0.5^3 + \binom{3}{1} 0.5^1 0.5^2 = 0.125 + 0.375 = 0.5$$

## Geometric

X: "number of failures before a success".

Every trial follows an independent bernoulli distribution with parameter  $p$ .

- Range of values:  $\{0, 1, 2, \dots\}$
- Probability mass function

$$P(X = k) = (1 - p)^k p$$

- Probability distribution function

$$F(k) = 1 - (1 - p)^{k+1}$$

- Expectation and Variance

$$E(X) = \frac{1-p}{p} \quad V(X) = \frac{1-p}{p^2}$$

Alternatively, one could define  $W$ ="Number of trials until one success is observed"

$$W = X + 1$$

- Range of values:  $\{1, 2, \dots\}$
- Probability mass function

$$P(W = k) = (1 - p)^{k-1} p$$

- Probability distribution function

$$F(k) = 1 - (1 - p)^k$$

- Expectation and Variance

$$E(X) = \frac{1}{p} \quad V(X) = \frac{1-p}{p^2}$$

**Example.**  $W$  = “Number of rolls of a die until one 5 is observed”

$$p = \frac{1}{6}$$

- Probability of (exactly) 6 rolls

$$P(W = 6) = \left(\frac{5}{6}\right)^5 \frac{1}{6} = 0.067$$

- Probability of “number 5 is observed in 3 rolls as much”

$$P(W \leq 3) = F(3) = 1 - \left(1 - \frac{1}{6}\right)^3 = 0.421$$

- Expected number of rolls

$$E(W) = \left(\frac{1}{\frac{1}{6}}\right)^{-1} = 6$$

## Negative Binomial

Number of failures until  $r$  successes are observed.

Generalization of the geometric distribution: sum of  $r$  independent and identical Geometric distributions.

- Range of values:  $\{0, 1, 2, \dots\}$
- Probability mass function

$$P(X = k) = \binom{k+r-1}{k} (1-p)^k p^r$$

$$E(X) = \frac{r(1-p)}{p}$$

$$V(X) = \frac{r(1-p)}{p^2}$$



**Example. X = “Number of non-5 results until three 5 are observed when rolling a die”**

$$r = 3, p = 1/6$$

$$X \sim NB(3, 1/6)$$

- Probability of 10 rolls (7 failures, 3 successes).

$$P(X = 7) = \binom{9}{7} \left(\frac{5}{6}\right)^7 \left(\frac{1}{6}\right)^3 = 0.017$$

- Expected number of “non-5” results until three “5” are obtained.

$$E(X) = \frac{3 \left(\frac{5}{6}\right)}{\frac{1}{6}} = 15$$

## Poisson

Let X follow a Binomial distribution  $Bin(n, p)$

$$E(X) = np$$

$$P(X = k) = \binom{n}{k} p^k (1-p)^{n-k} = \binom{n}{k} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k}$$

If the number of trials ( $n$ ) tends to infinite and the probability of success ( $p$ ) tends to zero in such a way that the expectation ( $\lambda$ ) remains the same:

$$\lim_{n \rightarrow \infty} P(X = k) = e^{-\lambda} \frac{\lambda^k}{k!}$$

In this case X follows a Poisson distribution of parameter  $\lambda$

- Commonly applied to rare events.
- Usually X is expressed with relation to a differential (time, space,..).
- Range of values:  $\{0, 1, 2, \dots, \infty\}$
- Expectation and Variance

$$E(X) = V(X) = \lambda$$

### Example

An emergency service from a certain hospital usually receives 10 visits per hour on average. The service collapses if it receives more than 20 visits in an hour.

- Probability that the service collapses.

$$P(X > 20) = 1 - P(X \leq 20) = 1 - \sum_{i=1}^{20} e^{-10} \frac{10^i}{i!} = 0.0016$$

- Probability of no visits in an hour.

$$P(X = 0) = e^{-10} \frac{10^0}{0!} = 4.54 \cdot 10^{-5}$$