

# *Probability and Inferential Statistics*

## Vorlesung SS 16

### Distribution

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## Distribution

The definition of a random variable  $X$  on a probability space  $(\Omega, \mathcal{A}, P)$  with values in  $(\Omega', \mathcal{A}')$  implies that all inverse images

$$X^{-1}(A') := \{\omega \in \Omega: X(\omega) \in A'\}, \quad A' \in \mathcal{A}',$$

are elements of the  $\sigma$ -algebra  $\mathcal{A}$  on  $\Omega$ . Because the measure  $P: \mathcal{A} \rightarrow [0, 1]$  assigns a probability to *all* elements of  $\mathcal{A}$ , the probabilities  $P[X^{-1}(A')]$  of these inverse images are determined by  $P$ .

**Definition 1.** Let  $X: (\Omega, \mathcal{A}, P) \rightarrow (\Omega', \mathcal{A}')$  be a random variable. Then the function  $P_X: \mathcal{A}' \rightarrow [0, 1]$  defined by

$$P_X(A') = P[X^{-1}(A')], \quad \forall A' \in \mathcal{A}', \quad (1)$$

is called the *distribution* of  $X$  (with respect to  $P$ ).

## Example: Indicator Variable

Let  $(\Omega, \mathcal{A}, P)$  be a probability space and  $A \in \mathcal{A}$ . Then  $1_A$  is called the *indicator (variable) of (the event) A*, if

$$1_A(\omega) := \begin{cases} 1, & \text{if } \omega \in A \\ 0, & \text{otherwise.} \end{cases} \quad (2)$$

The inverse image of the sets  $\{1\}$  and  $\{0\}$  under  $1_A$  are

$$1_A^{-1}(\{1\}) = \{1_A=1\} = A \quad \text{and} \quad 1_A^{-1}(\{0\}) = \{1_A=0\} = A^c.$$

Hence, the values of the distribution  $P_{1_A}$  are

$$P_{1_A}(\{1\}) = P(A), \quad P_{1_A}(\{0\}) = P(A^c), \quad P_{1_A}(\Omega') = P(\Omega) = 1, \quad P_{1_A}(\emptyset) = P(\emptyset) = 0.$$

## Notation

### Notation

If  $A' \in \mathcal{A}'$ , we use the notation

$$P(X \in A') := P[X^{-1}(A')] \quad (3)$$

for the probability of the event  $\{X \in A'\} = X^{-1}(A')$ , i. e., the event that  $X$  takes on a value in the subset  $A'$  of  $\Omega'$ . If  $\{x\} \in \mathcal{A}'$ , then we use

$$P(X=x) := P[X^{-1}(\{x\})] \quad (4)$$

for the probability of the event  $\{X=x\} = X^{-1}(\{x\}) = \{\omega \in \Omega: X(\omega) = x\}$ . If we write  $P(X=x)$ , then we always assume  $\{x\} \in \mathcal{A}'$ , even if not mentioned explicitly.

## New Probability Space

### New Probability Space

The definition of a random variable implies that *every* random variable  $X: (\Omega, \mathcal{A}, P) \rightarrow (\Omega', \mathcal{A}')$  has a distribution  $P_X: \mathcal{A}' \rightarrow [0, 1]$ .

Furthermore,  $P_X$  is also a probability measure and  $(\Omega', \mathcal{A}', P_X)$  is also a probability space.

## Example

Consider again tossing a coin two times and the random variable  $X$  “number of heads”. The values

$$P_X(\{0\}) = P[X^{-1}(\{0\})] = P[\{(t, t)\}] = 1/4,$$

$$P_X(\{1\}) = P[X^{-1}(\{1\})] = P[\{(h, t), (t, h)\}] = 1/2,$$

$$P_X(\{2\}) = P[X^{-1}(\{2\})] = P[\{(h, h)\}] = 1/4$$

of the distribution  $P_X$  are the probabilities of the events that  $X$  takes on the values 0, 1, and 2, respectively. Furthermore, the values

$$P_X(\{0, 1\}) = P[X^{-1}(\{0, 1\})] = P[\{(t, t), (h, t), (t, h)\}] = 3/4,$$

$$P_X(\{0, 2\}) = P[X^{-1}(\{0, 2\})] = P[\{(t, t), (h, h)\}] = 1/2,$$

$$P_X(\{1, 2\}) = P[X^{-1}(\{1, 2\})] = P[\{(h, t), (t, h), (h, h)\}] = 3/4$$

of  $P_X$  are the probabilities of the events that  $X$  takes on the values in the set  $\{0, 1\}$ ,  $\{0, 2\}$ , and  $\{1, 2\}$ , respectively. The value  $P_X(\Omega') = P[X^{-1}(\Omega')] = P(\Omega) = 1$  of the distribution  $P_X$  is the probability of the sure event and  $P_X(\emptyset) = P[X^{-1}(\emptyset)] = P(\emptyset) = 0$  is the value of  $P_X$  for the impossible event.

## Distribution Function

If we consider a univariate real-valued random variable  $X$ , then the *distribution function*  $F_X$  assigns to each  $x \in \mathbb{R}$  the probability  $P(X \leq x)$  of the event  $\{X \leq x\} = \{\omega \in \Omega: X(\omega) \leq x\}$  that  $X$  takes on a value *smaller or equal* than  $x$ . The distribution function uniquely determines the distribution  $P_X$ .

**Definition 2.** Let  $X: (\Omega, \mathcal{A}, P) \rightarrow (\mathbb{R}, \mathcal{B}, P_X)$  denote a real-valued random variable with distribution  $P_X$ . Then the (*cumulative*) *distribution function*  $F_X: \mathbb{R} \rightarrow [0, 1]$  of  $X$  is defined by:

$$F_X(x) := P_X([-\infty, x]) = P(X \leq x), \quad \forall x \in \mathbb{R}. \quad (5)$$

$F_X$  assigns to each  $x \in \mathbb{R}$  the probability to take on a value *less than or equal to*  $x$ .

## Probabilities and Intervals

The definition of a distribution function implies that we can compute the probability  $P(a < X \leq b)$  of  $X$  taking a value in the interval  $]a, b]$  by

$$P(a < X \leq b) = F_X(b) - F_X(a), \quad \text{if } a < b, \quad (6)$$

because

$$P(a < X \leq b) = P_X(]-\infty, b] \setminus ]-\infty, a]) = P_X(]-\infty, b]) - P_X(]-\infty, a]).$$

## Discrete Random Variable and Probability Function

The distribution of a discrete random variable can be described by its *probability function* that is now introduced. Remember, if  $X: (\Omega, \mathcal{A}, P) \rightarrow (\Omega', \mathcal{A}')$  is a random variable, then the distribution  $P_X$  of  $X$  is a probability measure on  $(\Omega', \mathcal{A}')$ .

**Definition 3** (Discrete Random Variable and its Probability Function). Let  $X: (\Omega, \mathcal{A}, P) \rightarrow (\Omega', \mathcal{A}')$  be a random variable and assume that  $\Omega'_0 \subset \Omega'$  is finite or countable with  $P_X(\Omega'_0) = 1$  and  $\{x\} \in \mathcal{A}'$  for all  $x \in \Omega'_0$ . Then  $X$  and its distribution  $P_X$  are called *discrete*, and the function  $p_X: \Omega' \rightarrow [0, 1]$  defined by

$$p_X(x) = \begin{cases} P_X(\{x\}), & \text{if } x \in \Omega'_0, \\ 0, & \text{if } x \in \Omega' \setminus \Omega'_0, \end{cases} \quad (7)$$

is called the *probability function of  $X$* .

We also use the notation  $P(X=x) = p_X(x)$ .

The distribution  $P_X$  assigns its values to each element of  $\mathcal{A}'$ , which are subsets of  $\Omega'$ . In contrast, the probability function  $p_X$  assigns its values to each  $x \in \Omega'_0$ .

## Example 1 - Flipping 2 Coins

### Example

Consider again tossing a coin two times and the random variable  $X$  “number of heads”. Then

$$p_X(0) = P_X(\{0\}) = 1/4,$$

$$p_X(1) = P_X(\{1\}) = 1/2,$$

$$p_X(2) = P_X(\{2\}) = 1/4$$

are the values of  $p_X$ . They are the probabilities of the events that  $X$  takes on the value 0, 1, and 2, respectively. These probabilities are also denoted  $P(X=0)$ ,  $P(X=1)$ ,  $P(X=2)$ , respectively.

## Example 2 - Binomial Distribution

Let  $(\Omega, \mathcal{A}, P)$  be a probability space and  $X := \sum_{i=1}^n 1_{A_i}$ , where  $1_{A_i}$  is the indicator variable of  $A_i \in \mathcal{A}$ ,  $i = 1, \dots, n$ . Assume

- (a)  $P(A_i) = p$  for all  $i = 1, \dots, n$ , and
- (b) the events  $A_1, \dots, A_n$  are independent.

Then the probability that  $X$  takes on the value  $x$  is

$$P(X=x) = b_{n,p}(x) = \binom{n}{x} p^x (1-p)^{n-x}, \quad (8)$$

where  $x = 0, 1, \dots, n$  and

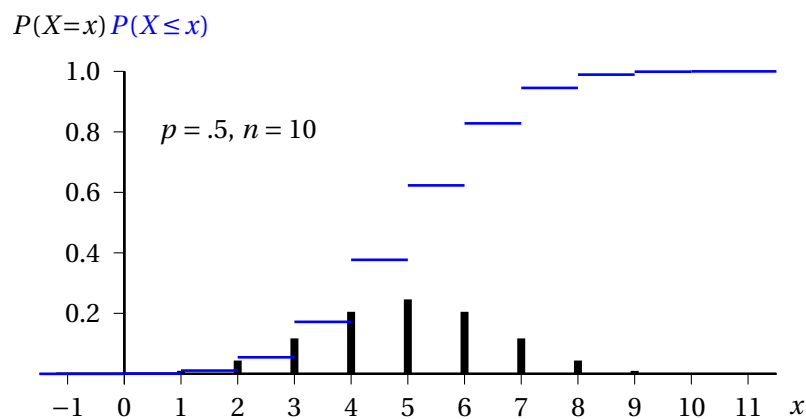
$$\binom{n}{x} := \frac{n!}{x! \cdot (n-x)!} \quad (9)$$

is the *binomial coefficient*. Note that  $0! := 1$  and  $x! := x \cdot (x-1) \cdot \dots \cdot 1$  denotes the *factorial of  $x$* .

If (8) holds, then  $P_X$  is called the *binomial distribution*.

An application is flipping a coin  $n$  times, where  $X$  is the random variable “number of heads”.

## Graph of the Binomial Distribution



Probability function and distribution function of a binomial distribution

## Continuous Random Variable and its Density

**Definition 4.** Let  $X : \Omega \rightarrow \mathbb{R}$  be a real-valued random variable on  $(\Omega, \mathcal{A}, P)$ , and  $F_X$  its distribution function. Then  $X$  is called *continuous*, if there is a Riemann-integrable function  $f_X : \mathbb{R} \rightarrow \mathbb{R}$ , such that

$$F_X(x) := \int_{-\infty}^x f_X(t) dt, \quad x \in \mathbb{R}.$$

A function  $f_X$  satisfying this equation is called a *density* of  $X$ .

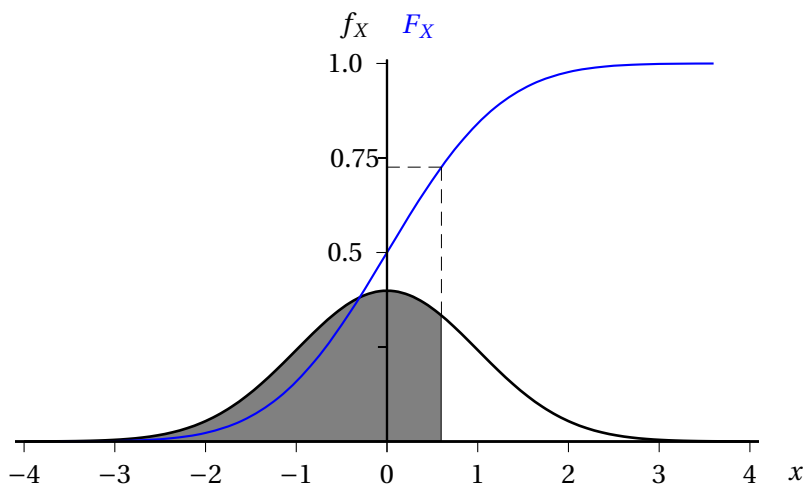
### Example 3: Normal Distribution

**Beispiel 1.** A well-known density of a random variable  $X$  is the density of a *normal distribution*, which is defined by

$$f_X(x) := \frac{1}{\sqrt{2\pi\sigma^2}} \cdot \exp\left(\frac{-(x-\mu)^2}{2\sigma^2}\right),$$

where  $\mu$  and  $\sigma^2$  are real numbers. (Later on we will see that  $\mu$  is the *expectation* of  $X$  and  $\sigma^2$  its *variance*.)

### Density and Distribution Function of the Normal Distribution



Density and distribution function of a

normal distribution