

# Probability Calculus

**Anna Janicka**

lecture IV, 28.10.2021

**RANDOM VARIABLES – INTRO:**

# Plan for today

---

- Poisson theorem
- Definition of the distribution of a random variable
- Description of the distribution of a random variable – examples
- Cumulative Distribution Function – intro.



# Poisson Theorem

---

## 1. Poisson Theorem

*If  $p_n \in [0, 1]$ ,  $\lim_{n \rightarrow \infty} np_n = \lambda > 0$ ,  
then for  $k = 0, 1, 2, \dots$ ,*

*we have that*

$$\lim_{n \rightarrow \infty} \binom{n}{k} p_n^k (1 - p_n)^{n-k} = \frac{\lambda^k}{k!} e^{-\lambda}.$$

## 2. Assessment of approximation error

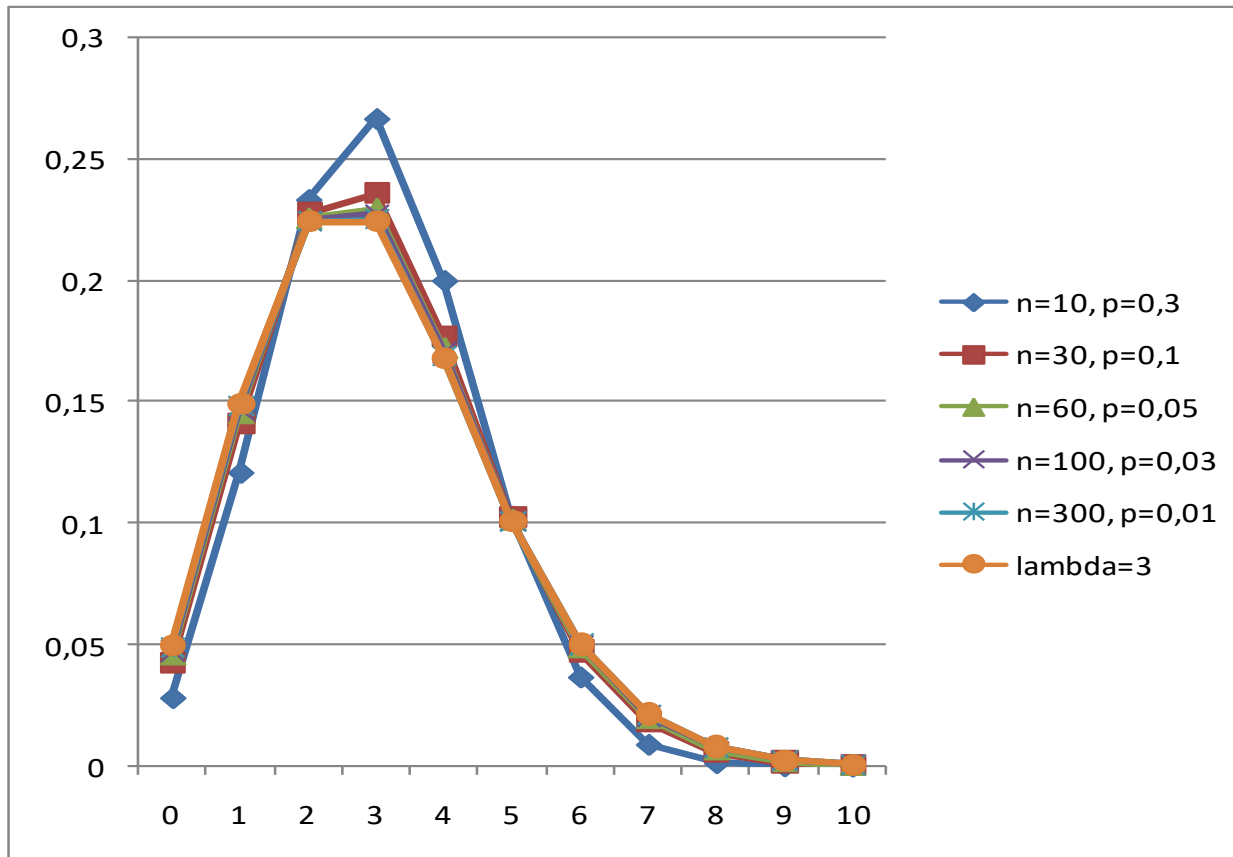
*Let  $S_n$  denote the number of successes in a  
Bernoulli process with  $n$  trials and a  
probability of success in a single trial equal to  $p$ ,  
and let  $\lambda = np$ . For any  $A \subset \{0, 1, 2, \dots\}$ , we have*

$$\left| \mathbb{P}(S_n \in A) - \sum_{k \in A} \frac{\lambda^k}{k!} e^{-\lambda} \right| \leq \frac{\lambda^2}{n}.$$



# Poisson Theorem – cont.

## The Poisson and Bernoulli processes



# Random variables – basics

---

1. Motivation – functions of the results of an experiment

## 2. Definition of a **random variable**

*A real-valued random variable is any function  $X : \Omega \rightarrow \mathbb{R}$ , such that for all  $a \in \mathbb{R}$  the set  $X^{-1}((-\infty, a])$  is an event, i.e.  $X^{-1}((-\infty, a]) \in \mathcal{F}$ .*

$$X^{-1}((-\infty, a]) = \{\omega \in \Omega : X(\omega) \leq a\}$$

## 3. Examples

- number of heads
- sum of points on dice

■ the distance to a given point



# Random variables – distribution

---

4. Functions of random variables
5. Examples of descriptions of random variables.
6. Definition of a random v. **distribution**

*The probability distribution of a random variable  $X$  (real-valued) is the probability  $\mu_X$  on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ , such that  $\mu_X(A) = \mathbb{P}(X \in A)$ .*

7. Different r.v. have the same distributions



notation:  $X \sim \mu$

*we forget about  $\Omega$*

# Random variables – examples

---

## 8. Examples of random variables

- die roll
- **discrete distributions**
- Binomial distribution
- Geometric distribution
- Poisson distribution
- uniform distribution over an interval: a **continuous distribution**
- another continuous distribution



# Continuous random variables

---

## 9. Definition of a **continuous random variable** and a **density function**

*A random variable  $X$  has a continuous distribution, if there exists a function  $g : \mathbb{R} \rightarrow \mathbb{R}_+$ , such that for any set  $A \in \mathcal{B}(\mathbb{R})$ ,*

$$\mu_X(A) = \mathbb{P}(X \in A) = \int_A g(x) dx.$$

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

## 10. The properties of density functions

- nonnegative
- normalized

---

■ determines the distribution unequivocally





# Random variable examples – cont.

---

## 11. More examples of continuous random variables

- uniform distribution
- exponential distribution
- standard normal distribution
- general normal distribution
- (Dirac delta)



# Random variables – the CDF

---

## 1. The definition of a CDF

The Cumulative distribution function of a random variable  $X : \Omega \rightarrow \mathbb{R}$  is a function  $F_X : \mathbb{R} \rightarrow [0, 1]$ , such that  $F_X(t) = \mathbb{P}(X \leq t)$ .

depends on the distribution only!  
→ CDF of distribution



# Random variables – the CDF

---

## 2. Examples of CDFs

- Dirac delta
- Two-point distribution – discrete distribution
- Exponential distribution
- Normal distribution – no simple form...



# CDFs

---

## 3. Properties of the CDF

*The cumulative distribution function  $F_X$  of a random variable  $X$  has the following properties:*

- (i)  $F_X$  is nondecreasing,*
- (ii)  $\lim_{t \rightarrow \infty} F_X(t) = 1$  and  $\lim_{t \rightarrow -\infty} F_X(t) = 0$ ,*
- (iii)  $F_X$  is right-continuous.*

## 4. CDF $\rightarrow$ distribution

*For any function  $F : \mathbb{R} \rightarrow \mathbb{R}$  satisfying the conditions (i)-(iii) above, there exists a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  and a random variable  $X : \Omega \rightarrow \mathbb{R}$  such that  $F$  is the CDF of  $X$ . Furthermore, the distribution of  $X$*

is determined unequivocally.

