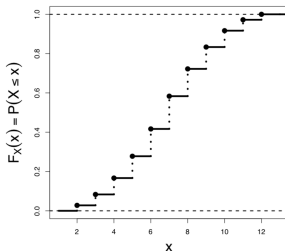


Random variables and distributions



Recall

- Let (Ω, \mathcal{F}, P) be a probability space.
- **Discrete**: Sample space Ω is countable/listable (finite or infinite)

- $\Omega = \{\omega_1, \omega_2, \dots\}, \quad P(\omega_i) = p_i$
- Probability table

Elementary events ω_i	ω_1	ω_2	\dots	ω_n	\dots	
Probability p_i	p_1	p_2	\dots	p_n	\dots	$\sum_i p_i = 1$

$0 \leq p_i \leq 1$

- Probability of an event A

$$P(A) = \sum_{\omega_i \in A} P(\omega_i)$$

- **Continuous**: Ω is a subset of the set of real numbers (an interval)
- We now introduce **random variables**.

Example: A game with two dice

- **Random experiment:** We roll **two fair** dice and write down the outcomes as (i, j) , where i is the result of the first and j is the result of the second die.
- **Probability space** $(\Omega, \mathcal{P}(\Omega), P)$: $\Omega = \{(i, j) \mid i, j = 1, \dots, 6\}$, $P(\{(i, j)\}) = \frac{1}{36}$.
- **Game:** We win \$500 if the sum is 7, otherwise we lose \$100.
 - We denote this **payoff function** by

$$D(i, j) = \begin{cases} 500 & \text{if } i + j = 7 \\ -100 & \text{if } i + j \neq 7. \end{cases}$$

...This function is an example of a **random variable**.

A **random variable** assigns a number to each outcome in the sample space.

Random variables

- A (measurable) function

$$X: \Omega \rightarrow \mathbb{R}$$

which assigns a real number $X(\omega) = x$ to every $\omega \in \Omega$

$$\omega \in \Omega \mapsto X(\omega) = x \in \mathbb{R}$$

is called a **random variable**.

- The number $x \in \mathbb{R}$ is called the **realization** of X .
- The set of values of X is the **image space** (or feature space)

$$\{x \mid X(\omega) = x, \omega \in \Omega\}$$

Random variables

- Two different types of random variables:
 - Discrete
 - If the image space is **finite** or **countably infinite**, the random variable X is **discrete**.
 - Continuous
 - A random variable X that can take values in one or more intervals, i.e. values that are infinite and uncountable are called **continuous**.
- Example:
 - The **number** of cash registers opened in a grocery store is a discrete random variable.
The **time** spent waiting in the queue is a continuous random variable.

Discrete random variables

- The **probability mass function** (pmf) of a discrete random variable X is the function

$$p(a) = P(X = a).$$

... auf Deutsch: **Wahrscheinlichkeitsgewichtungsfunktion** (Gewichte)

- It always holds $0 \leq p(a) \leq 1$
- We allow a to be any real number, i.e. $a \in \mathbb{R}$.
If a is a value that X never takes, then $p(a) = 0$.

- Example**

Let Ω be our earlier sample space for rolling **two fair** dice.

Let M to be the **maximum** value of the two dice:

$$M(i, j) = \max(i, j).$$

- M is a **discrete** random variable. We describe it by listing its possible values and the probabilities associated with those values.

value	a	1	2	3	4	5	6
pmf	$p(a)$	$\frac{1}{36}$	$\frac{3}{36}$	$\frac{5}{36}$	$\frac{7}{36}$	$\frac{9}{36}$	$\frac{11}{36}$

$$\sum_{a=1}^6 p(a) = 1$$

Continuous random variables

- (Ω, \mathcal{F}, P) is a probability space.
 - **Continuous** range of values $[c, d], [0, 1], [0, +\infty), (-\infty, +\infty)$
- A random variable X is **continuous** if there exists a function f such that for any $c \leq d$ it holds

$$P(c \leq X \leq d) = \int_c^d f(x) dx.$$

- $f(x)$ is called the **probability density function**
(**density function** or **density**)

... auf Deutsch: **Wahrscheinlichkeitsdichtefunktion**
(**Dichtefunktion** oder **Dichte**)

- Properties:

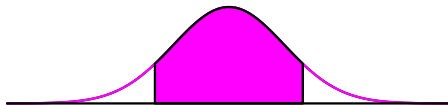
- f is **nonnegative**: $f(x) \geq 0$
- The **area** under f is one: $\int_{-\infty}^{+\infty} f(x) dx = 1$

Continuous random variables

- Visualization

$$P(c \leq X \leq d) = \int_c^d f(x) dx$$

= Area under f between c and d



Probability density function and probability

Continuous vs. discrete random variables

- The probability **density** function $f(x)$ of a continuous random variables (**pdf**) is the analogue to the probability **mass** function $p(x)$ of a discrete random variable (**pmf**).
- Important differences:
 - In contrast to $p(x)$, the probability density function $f(x)$ is **not a probability**. We have to **integrate** it to get probabilities.
 - Since $f(x)$ is not a probability, there is **no restriction** that $f(x)$ is less than or equal to 1.

Cumulative distribution function

- The **cumulative distribution function** (cdf) of a random variable X is the function F_X defined by

$$\underline{F_X(x) = P(X \leq x)}, \quad \text{for all } x \in \mathbb{R}.$$

Auf Deutsch: Verteilungsfunktion

- $F_X(x) = F(x)$ is defined for **all** real values x .
- A **cumulative distribution function** satisfies the following **properties**:
 - $0 \leq F(x) \leq 1$ for all $x \in \mathbb{R}$
 - F is **monotonically increasing**, i.e. from $x \leq y$ it follows $F(x) \leq F(y)$
 - $\lim_{x \rightarrow -\infty} F(x) = 0$ and $\lim_{x \rightarrow +\infty} F(x) = 1$
 - F is **right continuous**, i.e. $\lim_{h \searrow 0} F(x+h) = F(x)$ for all $x \in \mathbb{R}$.
- Important properties**:
 - $P(X > x) = 1 - P(X \leq x) = 1 - F(x)$ and
 - $P(a \leq X \leq b) = F(b) - F(a)$.

Example: Maximum of two dice

- Discrete random variables

(a) Let Ω be our sample space from the example with two fair dice.

Let M be the **maximum** of the two dice: $M(i, j) = \max(i, j)$.

We want to obtain $F_M(a) = F(a) = P(M \leq a)$.

From the distribution table

value	a	1	2	3	4	5	6	
pmf	$p(a)$	$\frac{1}{36}$	$\frac{3}{36}$	$\frac{5}{36}$	$\frac{7}{36}$	$\frac{9}{36}$	$\frac{11}{36}$	1
cdf	$F(a)$	$\frac{1}{36}$	$\frac{4}{36}$	$\frac{9}{36}$	$\frac{16}{36}$	$\frac{25}{36}$	1	

we get, for example

$$F(9) = 1, \quad F(-2) = 0, \quad F(2.5) = \frac{4}{36} \quad \text{and} \quad F(\pi) = \frac{9}{36}.$$

Example: Maximum of two dice

- Discrete random variables

(a) Let Ω be our sample space from the example with two dice.

Let M be the **maximum** of the two dice: $M(i, j) = \max(i, j)$.

We want to obtain $F_M(a) = F(a) = P(M \leq a)$

- We summarize, the **cdf of M** is of the following form

$$F_M(a) = F(a) = \begin{cases} 0, & a < 1 \\ \frac{1}{36}, & 1 \leq a < 2 \\ \frac{4}{36}, & 2 \leq a < 3 \\ \frac{9}{36}, & 3 \leq a < 4 \\ \frac{16}{36}, & 4 \leq a < 5 \\ \frac{25}{36}, & 5 \leq a < 6 \\ 1, & a \geq 6 \end{cases}.$$

F_M is a **step** (piecewise continuous) function

Example: Sum of two dice

(b) Let Ω be our sample space for the same example of rolling two fair dice.

Let X be the sum of two dice: $X(i, j) = i + j$.

- e.g. $\{X = 8\} = \{(2, 6), (3, 5), (4, 4), (5, 3), (6, 2)\}$ and $P(X = 8) = \frac{5}{36}$
- In general, the pmf of X is of the form

$$P(X = x) = \frac{6 - |7 - x|}{36}, \quad \text{for } x = 2, 3, \dots, 12$$

... HW Check!

the distribution table

value	x	2	3	4	5	6	7	8	9	10	11	12	
pmf	$p(x)$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$	1

Example: Sum of two dice

(b) Let Ω be our sample space for the same example of rolling two dice.

Let X be the **sum** of two dice: $X(i, j) = i + j$.

- e.g. $\{X = 8\} = \{(2, 6), (3, 5), (4, 4), (5, 3), (6, 2)\}$ and $P(X = 8) = \frac{5}{36}$
- In general, the **pmf of X** is of the form

$$P(X = x) = \frac{6 - |7 - x|}{36}, \quad \text{for } x = 2, 3, \dots, 12$$

... HW Check!

the distribution table

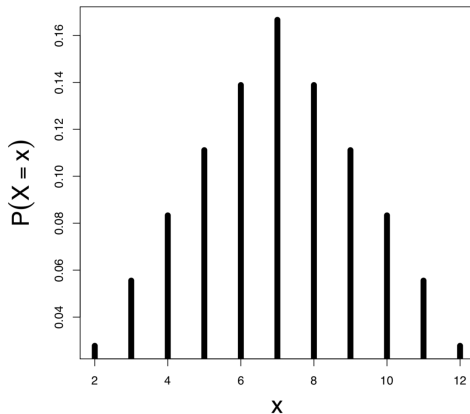
value	x	2	3	4	5	6	7	8	9	10	11	12	
pmf	$p(x)$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$	1
cdf	$F(x)$	$\frac{1}{36}$	$\frac{3}{36}$	$\frac{6}{36}$	$\frac{10}{36}$	$\frac{15}{36}$	$\frac{21}{36}$	$\frac{26}{36}$	$\frac{30}{36}$	$\frac{33}{36}$	$\frac{35}{36}$	1	

- e.g. $F(5.3) = P(X \leq 5.3)$
 $= P(X = 2) + P(X = 3) + P(X = 4) + P(X = 5) = \frac{10}{36}$

Example: Sum of two dice

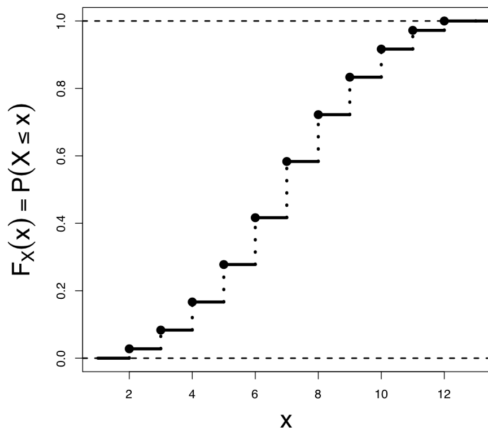
- The probability mass function

$$P(X = x) = \frac{6 - |7 - x|}{36}, \quad \text{for } x = 2, 3, \dots, 12$$



Example: Sum of two dice

- The graph of the cumulative distribution function is a **step function**.



Example: Payoff function

HW We roll two dice and write down the outcomes as (i, j) , where i is the result of the first and j is the result of the second die. We play a game in which we win \$500 if the sum is 7, otherwise we lose \$100.

- Find the pmf and the cdf of the payoff function D .
- What would be the expected win in this game?

Continuous random variable: Cumulative distribution function

- The **cumulative distribution function** of a **continuous** random variable X is given by

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$$

- $$\begin{aligned} P(a \leq X \leq b) &= F(b) - F(a) \\ &= \int_{-\infty}^b f(x) dx - \int_{-\infty}^a f(x) dx = \int_a^b f(x) dx \end{aligned}$$
- It also holds $F'(x) = f(x)$ for all $x \in \mathbb{R}$

Example

- Let X has the probability density function

$$f(x) = \begin{cases} 3, & x \in [0, \frac{1}{3}] \\ 0, & \text{else} \end{cases}.$$

(a) Compute $P(0.1 \leq X \leq 0.2)$ and $P(0.1 \leq X \leq 1)$.

(b) Find the cumulative distribution function of X .

- Answer:

$$(a) P(0.1 \leq X \leq 0.2) = \int_{0.1}^{0.2} f(x) dx = \int_{0.1}^{0.2} 3 dx = 0.3$$

(or area of rectangle = $3 \cdot 0.1 = 0.3$)

$$P(0.1 \leq X \leq 1) = \int_{0.1}^1 f(x) dx = \int_{0.1}^{\frac{1}{3}} f(x) dx + \int_{\frac{1}{3}}^1 f(x) dx = 3 \int_{0.1}^{\frac{1}{3}} dx + 0 = 0.7$$

(b) The cumulative distribution function $F(x) = P(X \leq x)$ is given by

$$F(x) = \begin{cases} 0, & x < 0 \\ 3x, & 0 \leq x < \frac{1}{3} \\ 1, & x \geq \frac{1}{3} \end{cases}.$$

More examples

HW Let X has range $[0, 1]$ and probability density function $f(x) = ax^2$.

- (a) What is the value of a ?
- (b) Compute the cumulative distribution function (cdf) $F_X(x)$.
- (c) Compute $P(1 \leq X \leq 2)$.

HW Let Y has range $[0, b]$ and its cdf is given by

$$F_Y(y) = \begin{cases} 0, & y < 0 \\ \frac{y^2}{9}, & 0 \leq y < b \\ 1, & y \geq b \end{cases}$$

- (a) What is the value of b ?
- (b) Find the probability density function $f(y)$ of Y .

Generalized inverse and quantile function

- One often needs the **inverse function** of a cdf F .
- As a cdf is not necessary **strictly monotonically** increasing, one needs the notion of the **generalized inverse** of F

$$F^{-1}(p) := \inf \{x \mid F(x) \geq p\} \quad \text{for } p \in (0, 1).$$

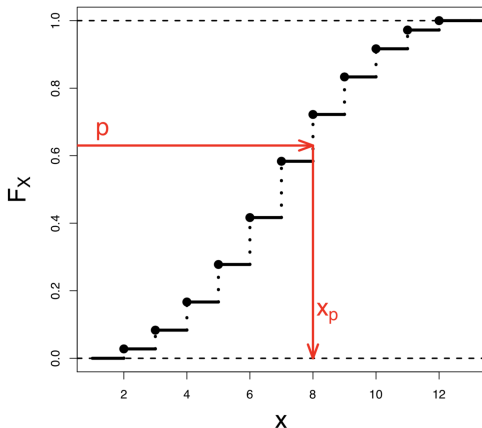
- If F is **strictly monotonically increasing**, then F^{-1} is the (usual) inverse function of F .
- A function that assigns the value $F^{-1}(p)$ to every $p \in (0, 1)$ is called **quantile function**

$$x_p = F^{-1}(p) \quad \text{for } p \in (0, 1) \iff F(x_p) = p \quad \text{for } p \in (0, 1)$$

- x_p is a **p -quantil** of F

Example: Sum of two dice

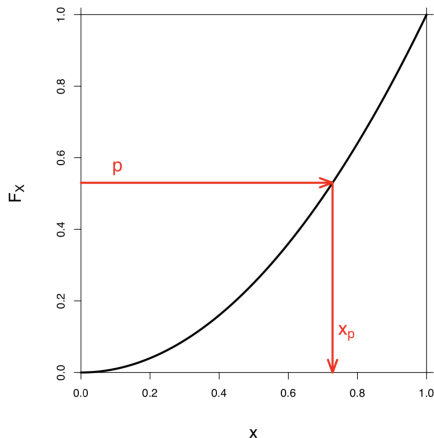
- The graph of the cumulative distribution function is a **step function**.



Example: continuous random variable

- A p -quantil of F is obtained from

$$F_X(x_p) = P(X \leq x_p) = \int_{-\infty}^{x_p} f(t) dt = p \iff x_p = F_X^{-1}(p)$$



Quartiles

- A 50%-quantil $x_{0.5}$ of F is called **median**.
- A 25%-quantil $x_{0.25}$ of F is called **lower quartile**
- A 75%-quantil $x_{0.75}$ of F is called **upper quartile**

Expected value

- The expected value (expectation)
 - For a **discrete** random variable with the values $x_1, x_2 \dots$ and pmfs $p(x_i)$, the **expected value** is defined by:

$$\mathbb{E}(X) = \sum_{i=1}^{\infty} x_i \cdot p(x_i) = x_1 \cdot p(x_1) + x_2 \cdot p(x_2) + \dots$$

- Let X be a **continuous** random variable with the density $f(x)$.
The **expected value** of X is defined by:

$$\mathbb{E}(X) = \int_{-\infty}^{+\infty} x \cdot f(x) dx$$

It is assumed that the **sum** and the **integral**
are absolute **convergent**!

- It is a **weighted mean** (average) of the possible outcomes of X .
- It is a measure of the **central tendency**.

Moment of order one of X .

Properties of $\mathbb{E}(X)$

(e1) $\mathbb{E}(aX + b) = a\mathbb{E}(X) + b, \quad a, b \in \mathbb{R}$

(e2) $\mathbb{E}(aX + bY) = a\mathbb{E}(X) + b\mathbb{E}(Y), \quad a, b \in \mathbb{R}$

(e3) For a function h it holds

$$\mathbb{E}(h(X)) = \sum_{i=1}^{\infty} h(x_i) \cdot p(x_i) \quad \text{respectively} \quad \mathbb{E}(h(X)) = \int_{-\infty}^{+\infty} h(x) \cdot f(x) dx.$$

Particularly:

- $\mathbb{E}(X^k)$... the **moment** of order k of X
- $\mu_k = \mathbb{E}((X - \mathbb{E}(X))^k)$... the **central moment** of order k of X
- $\mu_2 = \mathbb{E}((X - \mathbb{E}(X))^2) = \text{Var}(X)$... is the **variance** of X
- $\sigma = \sqrt{\text{Var}(X)}$... is the **standard deviation** of X

(e4) Let X and Y be **independent**. Then it holds $\mathbb{E}(X \cdot Y) = \mathbb{E}(X) \cdot \mathbb{E}(Y)$.

Properties of $\mathbb{V}ar(X)$

$$(v1) \quad \mathbb{V}ar(X) = \mathbb{E}((X - \mathbb{E}(X))^2) = \mathbb{E}(X^2) - (\mathbb{E}(X))^2$$

$$(v2) \quad \mathbb{V}ar(aX + b) = a^2 \mathbb{V}ar(X), \quad a, b \in \mathbb{R}$$

$$(v3) \quad \mathbb{V}ar(X + b) = \mathbb{V}ar(X), \quad b \in \mathbb{R}$$

(v4) Let X and Y be **independent**. Then it holds

$$\mathbb{V}ar(X + Y) = \mathbb{V}ar(X) + \mathbb{V}ar(Y).$$

Note: if X and Y are **not** independent, an **additional term** arises in the sum! We will get back to this later!

Example

It is given

X	x	3	4	5	6
pmf	$p(x)$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{8}$	$\frac{1}{8}$

- (a) Compute $\mathbb{E}(X)$ and $\mathbb{E}(3X - 2)$.
- (b) Compute $\mathbb{E}(X^2)$ and $\text{Var}(X)$.

Answer:

(a) $\mathbb{E}(X) = 3 \cdot \frac{1}{4} + 4 \cdot \frac{1}{2} + 5 \cdot \frac{1}{8} + 6 \cdot \frac{1}{8} = \frac{33}{8}$ and

$$\mathbb{E}(3X - 2) = 3 \cdot \mathbb{E}(X) - 2 = 3 \cdot \frac{33}{8} - 2 = \frac{83}{8}$$

... We have used the properties of the expectation.

II way:

$Y = 3X - 2$	y	7	10	13	16
pmf	$p(y)$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{8}$	$\frac{1}{8}$

$$\mathbb{E}(3X - 2) = \mathbb{E}(Y) = 7 \cdot \frac{1}{4} + 10 \cdot \frac{1}{2} + 13 \cdot \frac{1}{8} + 16 \cdot \frac{1}{8} = \frac{83}{8}$$

Example cont.

$$(b) \mathbb{E}(X^2) = \frac{9}{4} + \frac{16}{2} + \frac{25}{8} + \frac{36}{8} = \frac{143}{8}$$

$$\mathbb{V}ar(X) = \mathbb{E}(X^2) - (\mathbb{E}(X))^2 = \frac{143}{8} - \left(\frac{33}{8}\right)^2 = \frac{55}{64}.$$

Transformations

- Let X be a continuous random variable, i.e., its density f_X is known.
- What is the **distribution of a transformation**

$$Y = g(X)$$

with $g : \mathbb{R} \rightarrow \mathbb{R}$?

- **Method:** Use the cumulative distribution function of X
 - Determine F_X .
 - Determine F_Y for $Y = g(X)$.
 - Find $f_Y(y) = F'_Y(y)$.

Another example

Let X has the density function

$$f(x) = \begin{cases} ax^4, & x \in [0, 1] \\ 0, & \text{otherwise} \end{cases}.$$

- (a) Find a .
- (b) Compute $\mathbb{E}(X)$ and $\mathbb{V}ar(X)$.
- (c) Find the median value of X .
- (d) Let X_1, X_2, \dots, X_{25} are independent and identically distributed (i.i.d.) copies of X . Let \bar{X} be their average (mean), i.e.

$$\bar{X} = \frac{1}{25} \sum_{i=1}^{25} X_i.$$

Compute the standard deviation of \bar{X} .

- (e) Let $Y = 2X^3 + 1$. Find the density of Y .

Another example: Solution

(a) From $1 = \int_{\mathbb{R}} f(x) dx = \int_0^1 ax^4 dx$ we obtain $a = 5$.

(b) The expectation and variance are

$$\mu = \mathbb{E}(X) = \int_0^1 5x^5 dx = \frac{5}{6}$$

$$\text{Var}(X) = \int_0^1 \left(x - \frac{5}{6}\right)^2 5x^4 dx = 5 \int_0^1 \left(x^6 - \frac{5}{3}x^5 + \frac{25}{36}x^4\right) dx = \frac{5}{252} \approx 0.02$$

$$\sigma = \sqrt{\text{Var}(X)} \approx 0.14$$

(c) First we calculate the cumulative distribution function

$$F_X(x) = \begin{cases} 0, & x < 0 \\ x^5, & 0 \leq x < 1 \\ 1, & x \geq 1 \end{cases}$$

Median $x_{0.5}$ solves the equation

$$F_X(x_{0.5}) = 0.5$$

Then, $F_X(x_{0.5}) = x_{0.5}^5 = 0.5$ and $x_{0.5} = \sqrt[5]{0.5} \approx 0.87$.

Another example: Solution

- (d) We compute the expectation $\mathbb{E}(\bar{X})$ and the variance $\mathbb{V}\text{ar}(\bar{X})$ of the **sample mean** $\bar{X} = \frac{1}{25}(X_1 + X_2 + \cdots + X_{25})$.

$$\mathbb{E}(\bar{X}) = \frac{1}{25}\mathbb{E}(X_1 + \cdots + X_{25}) = \mathbb{E}(X) = \frac{5}{6}$$

$$\begin{aligned}\mathbb{V}\text{ar}(\bar{X}) &= \frac{1}{25^2}\mathbb{V}\text{ar}(X_1 + X_2 + \cdots + X_{25}) = \frac{1}{25^2}(\mathbb{V}\text{ar}(X_1) + \cdots + \mathbb{V}\text{ar}(X_{25})) \\ &= \frac{1}{25^2} \cdot 25\mathbb{V}\text{ar}(X) = \frac{\mathbb{V}\text{ar}(X)}{25} < \mathbb{V}\text{ar}(X).\end{aligned}$$

Also, $\sigma_{\bar{X}} = \frac{\sigma}{5} \approx 0.028$.

Important: $\mathbb{V}\text{ar}(\bar{X}) \leq \mathbb{V}\text{ar}(X)$!

- (e) In order to find the pdf of $Y = 2X^3 + 1$, we use the cdf of X

$$\begin{aligned}F_Y(y) &= P(Y \leq y) = P(2X^3 + 1 \leq y) = P(X \leq \sqrt[3]{\frac{y-1}{2}}) \\ &= F_X\left(\sqrt[3]{\frac{y-1}{2}}\right) = \begin{cases} 0, & y < 1 \\ \left(\frac{y-1}{2}\right)^{\frac{5}{3}}, & 1 \leq y < 3 \\ 1, & y \geq 3 \end{cases}.\end{aligned}$$

Also,

$$f_Y(y) = F'_Y(y) = \begin{cases} \frac{5}{6} \cdot \left(\frac{y-1}{2}\right)^{\frac{2}{3}}, & 1 < y < 3 \\ 0, & \text{else} \end{cases}.$$

More examples & some multiple-choice questions

Examples

HW Let X be a random variable that takes on values 0, 2 and 3 with probabilities 0.3, 0.1 and 0.6 respectively. Let $Y = 3(X - 1)^2$.

- (a) Compute $\mathbb{E}(X)$ and $\text{Var}(X)$.
- (b) Compute $\mathbb{E}(Y)$.
- (c) Let $F_Y(y)$ be the cdf of Y . What is $F_Y(7)$?

HW Let X be a random variable with the cdf

$$F(x) = \begin{cases} 0 & \text{for } x < 0 \\ x(2-x) & \text{for } 0 \leq x < 1 \\ 1 & \text{for } x > 1 \end{cases}.$$

- (a) Compute $P(X \leq 0.4)$.
- (b) Compute $\mathbb{E}(X)$.
- (c) Find the moment of order four of X .
- (d) Find the median, upper and lower quartiles.

Examples

HW Let X be a random variable with the range $[0, 1]$ and cumulative distribution function

$$F(x) = 2x^2 - x^4, \quad \text{for } 0 \leq x \leq 1.$$

- (a) Verify that F is a cumulative distribution function.
- (b) Compute $P(\frac{1}{4} \leq X \leq \frac{3}{4})$.
- (c) What is the probability density of X ?

HW Let X be a random variable with the cumulative distribution function F_X . Let X_1 and X_2 be independent and identically distributed (i.i.d.) copies from X . Let $Y = \max\{X_1, X_2\}$. Find the distribution function F_Y of Y with respect to F_X .

HW An enthusiastic football fan gives away Tototips every week, using the digits 0 (draw), 1 (home win), 2 (victory) with the help of the probability function

$$P(X = k) = \begin{cases} \frac{1}{4} + ak + bk^2 & k = 0, 1, 2 \\ 0 & \text{else} \end{cases}$$

with unknown values a and b . However, it is known that for his tips it holds $P(X = 1) = 1/4$. Determine a and b and the corresponding distribution function.

A few multiple-choice questions

- (1) Let X be a random variable with probability density function of the form

$$f(x) = \begin{cases} -2x, & -1 \leq x \leq 0 \\ 0, & \text{else} \end{cases}.$$

Compute $P(-\frac{3}{4} \leq X < -\frac{1}{2})$.

- a. 5/16
- b. 1/2
- c. 7/8
- d. 19/64

- (2) Let

$$F(x) = \begin{cases} 0 & x < 0 \\ 2x & 0 \leq x < 0.5 \\ 1 & x \geq 0.5 \end{cases}$$

be the cumulative distribution function of a random variable X and let $Y = 2X + 1$. Then, the expectation $\mathbb{E}(Y)$ equals

- (a) 0.25
- (b) 0.5
- (c) 1.5
- (d) 0.75

A few multiple-choice questions

- (3) A random variable X has a probability distribution as follows:

X	0	1	2	3
$P(X)$	$2k$	$3k$	$13k$	$2k$

where k is a positive constant. The probability $P(X < 2.0)$ is equal to

- a. 0.90
- b. 0.25
- c. 0.65
- d. 0.15

- (4) Let X be a random variable that takes values $-2, -1, 0, 1$ and 2 , each with probability $1/5$. Let $Y = X^2$. Then,

- a. $\text{Cov}(X, Y) > 0$
- b. $\text{Cov}(X, Y) < 0$
- c. $\text{Var}Y < 2\text{Var}X$
- d. $\text{Var}Y = 2\text{Var}X$

... Note: We will learn soon what is the **covariance** $\text{Cov}(X, Y)$!

A few multiple-choice questions

- (5) Let us consider ballgames which cannot end in a tie. Suppose a bookie will give you \$6 for every \$1 you risk if you pick the winners in three ballgames. Thus, for every \$1 you bet you will either lose \$1 or gain \$5. What is the bookie's expected earnings per dollar wagered?
- a. $-\$2/8$
 - b. $\$34/8$
 - c. $\$2/8$
 - d. $\$21/27$
- (6) Let X and Y be two independent random variables. Suppose that we know $\text{Var}(2X - Y) = 6$ and $\text{Var}(X + 2Y) = 9$. Then, the variance $\text{Var}(Y)$ equals
- a. 1
 - b. 2
 - c. 3
 - d. 5

A few multiple-choice questions

(7) Let

$$F(x) = \begin{cases} 0, & x < 1 \\ 1 - \frac{1}{x^3}, & x \geq 1 \end{cases}$$

be the cumulative distribution function of a random variable X .
Then, its second central moment is

- a. not finite
- b. 0.75
- c. 6.75
- d. 3

(8) Forty people are invited to a party. Each person accepts the invitation, independently of all others, with probability $1/4$. Let X be the number of accepted invitations. Then, the expectation of $X^2 - 8X + 5$ equals

- a. -167.5
- b. 1.25
- c. 27.5
- d. 32.5

A few multiple-choice questions

(9) Anna rolls two fair dice and observe two numbers X and Y .

If $Z = X - Y$, what is the probability $P(Z < 4)$?

- a. $\frac{1}{18}$
- b. $\frac{17}{18}$
- c. $\frac{11}{12}$
- d. $\frac{1}{3}$

(10) Let

$$f(x) = \frac{1}{\pi(1+x^2)}, \quad x \in \mathbb{R}.$$

be the probability density function of a random variable X . Then,

- a. the expectation $\mathbb{E}X = 0$
- b. the variance $\mathbb{V}arX = 1$
- c. the variance $\mathbb{V}arX = \sqrt{\pi}$
- d. the expectation $\mathbb{E}X$ does not exist

Thank you for your attention!