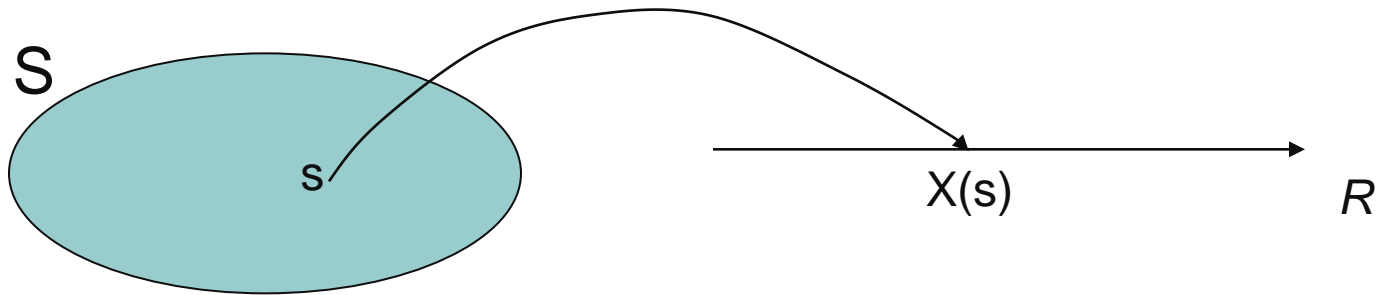


Probability theory

Random variables

In an experiment a number is often attached to each outcome.



Definition:

A **random variable** X is a function defined on S , which takes values on the real axis

$$X: S \rightarrow R$$

Sample space \nearrow S \rightarrow R \nwarrow Real numbers



Probability theory

Random variables

Example:

Random variable	Type
Number of eyes when rolling a die	discrete
The sum of eyes when rolling two dice	discrete
Number of children in a family	discrete
Age of first-time mother	discrete
Time of running 5 km	continuous
Amount of sugar in a coke	continuous
Height of males	continuous

counting

measure

Discrete: can take a finite number of values or an infinite but countable number of values.

Continuous: takes values from the set of real numbers.



Discrete random variable

Probability function

Definition:

Let $X : S \rightarrow R$ be a discrete random variable.

The function $f(x)$ is a **probability function** for X , if

1. $f(x) \geq 0$ for all x

2. $\sum_x f(x) = 1$

3. $P(X = x) = f(x),$

where $P(X=x)$ is the probability for the outcomes $s \in S : X(s) = x$.



Discrete random variable

Probability function

Example: Flip three coins $X : \# \text{ heads}$ $X : S \rightarrow \{0,1,2,3\}$

Outcome	Value of X	Probability function
TTT	$X=0$	$f(0) = P(X=0) = 1/8$
HTT, TTH, THT	$X=1$	$f(1) = P(X=1) = 3/8$
HHT, HTH, THH	$X=2$	$f(2) = P(X=2) = 3/8$
HHH	$X=3$	$f(3) = P(X=3) = 1/8$

Notice! The definition of a probability function is fulfilled:

1. $f(x) \geq 0$
2. $\sum f(x) = 1$
3. $P(X=x) = f(x)$



Discrete random variable

Cumulative distribution function

Definition:

Let $X : S \rightarrow R$ be a discrete random variable with probability function $f(x)$.

The cumulative distribution function for X , $F(x)$, is defined by

$$F(x) = P(X \leq x) = \sum_{t \leq x} f(t) \quad \text{for } -\infty < x < \infty$$

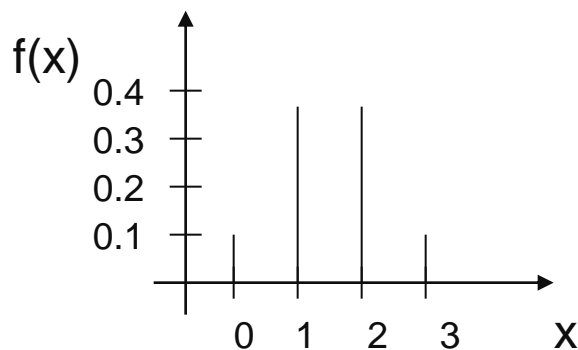
Discrete random variable

Cumulative distribution function

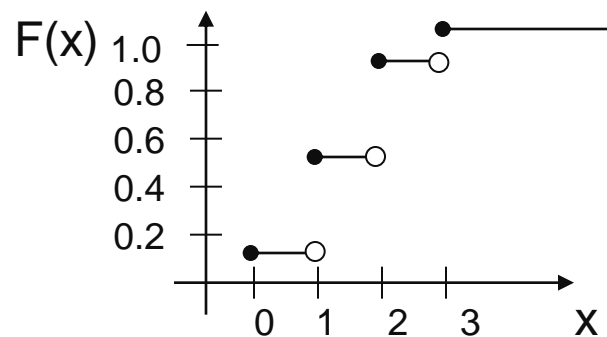
Example: Flip three coins $X : \# \text{ heads} \quad X : S \rightarrow \{0,1,2,3\}$

Outcome	Value of X	Probability function	Cumulative dist. Func.
TTT	$X=0$	$f(0) = P(X=0) = 1/8$	$F(0) = P(X \leq 0) = 1/8$
HTT, TTH, THT	$X=1$	$f(1) = P(X=1) = 3/8$	$F(1) = P(X \leq 1) = 4/8$
HHT, HTH, THH	$X=2$	$f(2) = P(X=2) = 3/8$	$F(2) = P(X \leq 2) = 7/8$
HHH	$X=3$	$f(3) = P(X=3) = 1/8$	$F(3) = P(X \leq 3) = 1$

Probability function:



Cumulative distribution function:





Continuous random variable

A continuous random variable X has **probability 0** for all outcomes!!

Mathematically: $P(X = x) = f(x) = 0$ for all x

Hence, we cannot represent the probability function $f(x)$ by a table or bar chart as in the case of discrete random variables.

Instead we use a continuous function – a **density function**.



Continuous random variable

Density function

Definition:

Let $X: S \rightarrow R$ be a continuous random variables.

A **probability density function** $f(x)$ for X is defined by:

1. $f(x) \geq 0$ for all x

2. $\int_{-\infty}^{\infty} f(x) dx = 1$

3. $P(a < X < b) = \int_a^b f(x) dx$

Note!! Continuity: $P(a < X < b) = P(a \leq X \leq b) = P(a \leq X < b) = P(a < X \leq b)$

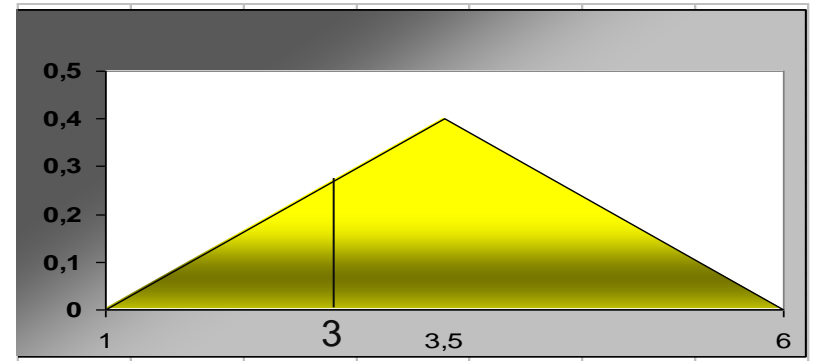
Continuous random variable

Density function

Example: X: service life of car battery in years (continuous)

Density function:

$$f(x) = \begin{cases} -0.16 + 0.16x & \text{for } 1 \leq x \leq 3.5 \\ 0.96 - 0.16x & \text{for } 3.5 < x \leq 6 \\ 0 & \text{otherwise} \end{cases}$$



Probability of a service life longer than 3 years:

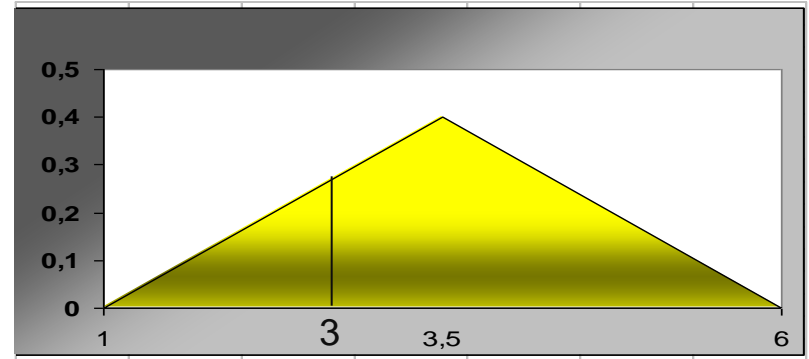
$$\begin{aligned} P(X > 3) &= \int_3^{\infty} f(x) dx \\ &= \int_3^{3.5} (-0.16 + 0.16x) dx + \int_{3.5}^6 (0.96 - 0.16x) dx \\ &= \dots = 0.68 \end{aligned}$$

Continuous random variable

Density function

Alternatively:

$$f(x) = \begin{cases} -0.16 + 0.16x & \text{for } 1 \leq x \leq 3.5 \\ 0.96 - 0.16x & \text{for } 3.5 < x \leq 6 \\ 0 & \text{otherwise} \end{cases}$$



Probability of a service life longer than 3 years:

$$P(X > 3) = 1 - P(X \leq 3)$$

$$= 1 - \int_{-\infty}^3 f(x) dx$$

$$= 1 - \int_1^3 (-0.16 + 0.16x) dx$$

$$= \dots = 1 - 0.32 = 0.68$$

Continuous random variable Density function

Definition:

Let $X : S \rightarrow R$ be a continuous random variable with density function $f(x)$.

The cumulative distribution function for X , $F(x)$, is defined by

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

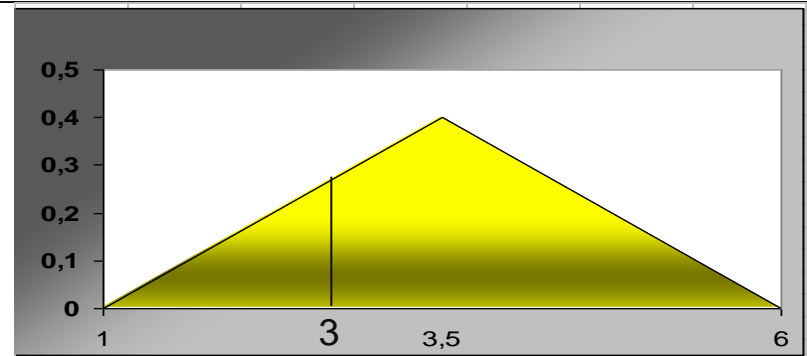
Note: $F'(x) = f(x)$

$$F(3) = P(X \leq 3)$$

$$= \int_{-\infty}^3 f(x) dx$$

$$= \int_1^3 -0.16 + 0.16x dx$$

$$= \dots = 0.32$$

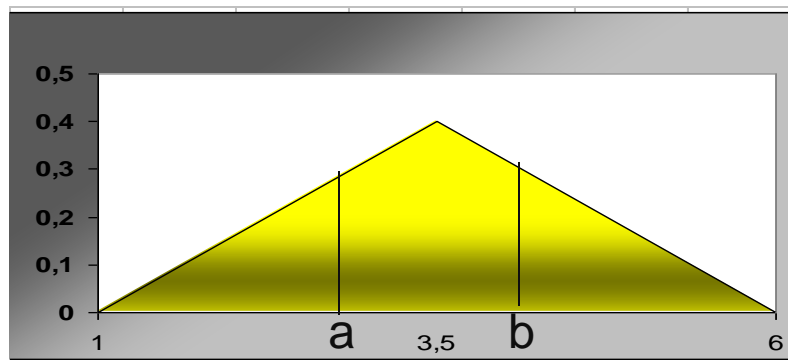


Continuous random variable

Density function

From the definition of the cumulative distribution funct. we get:

$$\begin{aligned} P(a < X < b) &= P(a \leq X \leq b) = P(a \leq X < b) = P(a < X \leq b) \\ &= P(X \leq b) - P(X \leq a) \\ &= F(b) - F(a) \end{aligned}$$





Continuous random variable

Density function

Discrete random variable

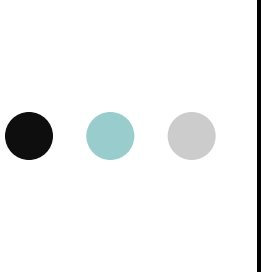
- Sample space is finite or has countable many outcomes
- Probability function $f(x)$ is often given by table
- Calculation of probabilities

$$P(a < X < b) = \sum_{a < t < b} f(t)$$

Continuous random variable

- The sample space contains infinitely many outcomes
- Density function $f(x)$ is a continuous function
- Calculation of probabilities

$$P(a < X < b) = \int_a^b f(t) dt$$



Joint distribution

Joint probability function

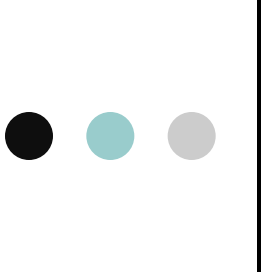
Definition:

Let X and Y be two **discrete** random variables.
The **joint probability function** $f(x,y)$ for X and Y is defined by

1. $f(x,y) \geq 0$ for all x and y
2. $\sum_x \sum_y f(x,y) = 1$
3. $P(X = x, Y = y) = f(x,y)$ (the probability that both $X = x$ and $Y = y$)

For a set A in the xy -plane:

$$P((X, Y) \in A) = \sum_{(x,y) \in A} f(x,y)$$



Joint distribution

Marginal probability function

Definition:

Let X and Y be two **discrete** random variables with joint probability function $f(x,y)$.

The **marginal probability function** for X is given by

$$g(x) = \sum_y f(x,y) \quad \text{for all } x$$

The **marginal probability function** for Y is given by

$$h(y) = \sum_x f(x,y) \quad \text{for all } y$$

Joint distribution

Marginal probability function

Example 3.16 (modified):

The joint probability function $f(x,y)$ for X and Y is given by

$y \backslash x$	0	1	2
0	$3/28$	$9/28$	$3/28$
1	$3/14$	$3/14$	0
2	$1/28$	0	0

- $h(1) = P(Y = 1) = 3/14 + 3/14 + 0 = 3/7$

- $g(2) = P(X = 2) = 3/28 + 0 + 0 = 3/28$

- $P(X+Y < 2) = 3/28 + 9/28 + 3/14 = 18/28 = 9/14$



Joint distribution

Joint density function

Definition:

Let X and Y be two **continuous** random variables. The **joint density function** $f(x,y)$ for X and Y is defined by

1. $f(x,y) \geq 0$ for all x and y
2. $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = 1$
3. $P(a < X < b, c < Y < d) = \int_c^d \int_a^b f(x,y) dx dy$

For a region A in the xy -plane: $P[(X,Y) \in A] = \int \int_A f(x,y) dx dy$



Joint distribution

Marginal density function

Definition:

Let X and Y be two **continuous** random variables with joint density function $f(x,y)$.

The **marginal density function** for X is given by

$$g(x) = \int_{-\infty}^{\infty} f(x, y) dy \quad \text{for all } x$$

The **marginal density function** for Y is given by

$$h(y) = \int_{-\infty}^{\infty} f(x, y) dx \quad \text{for all } y$$

Joint distribution

Marginal density function

Example 3.15 + 3.17 (modified):

Joint density $f(x,y)$ for X and Y :

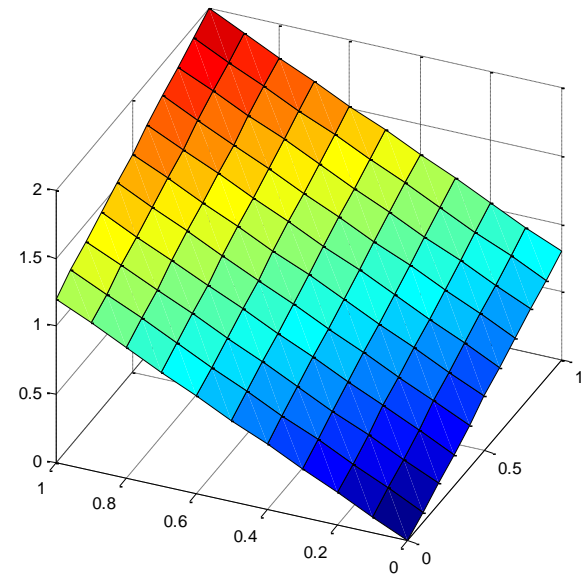
$$f(x, y) = \begin{cases} \frac{2}{5}(2x+3y) & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Marginal density function for X :

$$g(x) = \int_{-\infty}^{\infty} f(x, y) dy$$

$$= \int_0^1 \frac{2}{5}(2x+3y) dy$$

$$= \left[\frac{2}{5}2xy + \frac{1}{5}3y^2 \right]_0^1 = \frac{4}{5}x + \frac{3}{5}$$



Joint distribution

Conditional density and probability functions

Definition:

Let X and Y be random variables (continuous or discrete) with joint density/probability function $f(x,y)$. Then the

Conditional density/probability function for Y given $X=x$ is

$$f(y|x) = f(x,y) / g(x) \quad g(x) \neq 0$$

where $g(x)$ is the marginal density/probability function for X , and the conditional density/probability function for X given $Y=y$ is

$$f(x|y) = f(x,y) / h(y) \quad h(y) \neq 0$$

where $h(y)$ is the marginal density/probability function for Y .

Joint distribution

Conditional probability function

Examples 3.16 + 3.18 (modified):

Joint probability function

$f(x,y)$ for X and Y is given by:

$y \backslash x$	0	1	2
0	$3/28$	$9/28$	$3/28$
1	$3/14$	$3/14$	0
2	$1/28$	0	0

• marginal pf. $g(x) = \begin{cases} \frac{10}{28} & \text{for } x = 0 \\ \frac{15}{28} & \text{for } x = 1 \\ \frac{3}{28} & \text{for } x = 2 \end{cases}$

• $P(Y=1 \mid X=1) = f(1|1)$
 $= f(1,1) / g(1)$
 $= (3/14) / (15/28)$
 $= 6/15$



Joint distribution Independence

Definition:

Two random variables X and Y (**continuous** or **discrete**) with joint density/probability functions $f(x,y)$ and marginal density/probability functions $g(x)$ and $h(y)$, respectively, are said to be **independent** if and only if

$$f(x,y) = g(x) h(y) \quad \text{for all } x,y$$

or if $f(x|y) = g(x)$ (x indep. of y) or $f(y|x)=h(y)$ (y indep. of x)