Elements of Probability Theory

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Probability in Dimension 1

Random experiment

An experiment which repeated under same fixed conditions produce different results is called random experiment.

Examples:

1. Rolling a cube-shaped fare die



2. Choosing an integer even number randomly



3. Measuring temperature



Probability space

The mathematical framework in which a random experiment is described consists of a triplet $(\Omega, \Sigma, \mathcal{P})$, called probability space.

The elements of a probability space are:

- Ω: sample space (set with all possible events)
- Σ : σ -algebra on Ω (set with relevant events only)
- P: probability measure (measure of expectation of an event occurrence)

Sample space

A non-empty set which contains all possible events for a certain random experiment is called sample space, being represented by Ω .

Examples:

1. Rolling a cube-shaped fare die (finite Ω)

$$\Omega = \{1, 2, 3, 4, 5, 6\}$$

2. Choosing an integer even number randomly (denumerable Ω)

$$\Omega = \{\cdots, -8, -6, -4, -2, 0, 2, 4, 6, 8, \cdots\}$$

3. Measuring temperature in Kelvin (non-denumerable Ω)

$$\Omega = [a, b] \subset [0, +\infty)$$

σ -algebra of events

In general, not all of the events in Ω are of interest.

Intuitively, a σ -algebra on Ω is the set of relevant outcomes for a random experiment. Formally, Σ is a σ -algebra on Ω if

- $\phi \in \Sigma$ (contains the empty set)
- $\mathcal{A}^c \in \Sigma$ for any $\mathcal{A} \in \Sigma$ (closed under complementation)
- $\bigcup_{i=1}^\infty \mathcal{A}_i \in \Sigma$ for any $\mathcal{A}_i \in \Sigma$ (closed under denumerable unions)

Probability measure

A probability measure is a function $\mathcal{P}:\Sigma \to [0,1]\subset \mathbb{R}$ such that

- $\mathcal{P} \{ \mathcal{A} \} \ge 0$ for any $\mathcal{A} \in \Sigma$ (probability is nonnegative)
- $\mathcal{P}\left\{\Omega\right\} = 1$ (entire space has probability one)
- $\mathcal{P}\left\{\bigcup_{i=1}^{\infty} \mathcal{A}_i\right\} = \sum_{i=1}^{\infty} \mathcal{P}\left\{\mathcal{A}_i\right\}$ for any \mathcal{A}_i mutually disjoint $(\sigma$ -additivity)

Remark:

 $\mathcal{P}\left\{\phi\right\} = 0$ (empty set has probability zero)



A fair coin is thrown twice.

The number of faces is of interest.



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Probability space 1:

$$\Omega_1 = \big\{ (\mathit{H}, \mathit{H}), (\mathit{H}, \mathit{T}), (\mathit{T}, \mathit{H}), (\mathit{T}, \mathit{T}) \big\}$$





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Probability space 1:

$$\begin{split} &\Omega_1 = \big\{ (\textit{H},\textit{H}), (\textit{H},\textit{T}), (\textit{T},\textit{H}), (\textit{T},\textit{T}) \big\} \\ &\mathcal{P}_1 \left\{ (\textit{H},\textit{H}) \right\} = 1/4, \quad \mathcal{P}_1 \left\{ (\textit{H},\textit{T}) \right\} = 1/4, \\ &\mathcal{P}_1 \left\{ (\textit{T},\textit{H}) \right\} = 1/4, \quad \mathcal{P}_1 \left\{ (\textit{T},\textit{T}) \right\} = 1/4 \end{split}$$





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Probability space 2:

$$\Omega_2=\{{\scriptstyle 0,1,2}\}$$



A. C. Morgado, J. B. Pitombeira, P. C. P. Carvalho, P. J. Fernandez, **Análise Combinatória e Probabilidade**. SBM, 2016



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Probability space 2:

$$\Omega_2=\{{\scriptstyle 0,1,2}\}$$

$$\mathcal{P}_{2} \{0\} = 1/4, \quad \mathcal{P}_{2} \{1\} = 1/2, \quad \mathcal{P}_{2} \{2\} = 1/4$$



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A point is randomly choosen in a square (side b).

Event A: point lie above main diagonal

Event B: point lie in main diagonal



A point is randomly choosen in a square (side b).

Event A: point lie above main diagonal

Event B: point lie in main diagonal

$$\mathcal{P}\left\{A\right\} = \frac{\text{area of upper triangle}}{\text{area of square}} = \frac{0.5 \, b^2}{b^2} = \frac{1}{2}$$



A point is randomly choosen in a square (side b).

Event A: point lie above main diagonal

Event B: point lie in main diagonal

$$\mathcal{P}\left\{A\right\} = \frac{\text{area of upper triangle}}{\text{area of square}} = \frac{0.5 \, b^2}{b^2} = \frac{1}{2}$$

$$\mathcal{P}\left\{B\right\} = \frac{\text{area of main diagonal}}{\text{area of square}} = \frac{0}{b^2} = 0$$



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$$\mathcal{P}\left\{A
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$$\mathcal{P}\left\{B\right\} = \frac{\text{area of main diagonal}}{\text{area of square}} = \frac{0}{b^2} = 0$$

$$\mathcal{P}\left\{ C
ight\} =rac{ ext{area outside main diagonal}}{ ext{area of square}}=rac{b^{2}-0}{b^{2}}=1$$

Remarks on probability

Probability zero

- An impossible event has probability zero
 (e.g. roll a six faces dice, numbered from 1 to 6, and get 7)
- Not every event with probability zero is impossible
 (e.g. randomly pick a point on the main diagonal of a square)

Probability one

- An event which occurrence is certain has probability one (e.g. throw a coin and obtain head or tail)
- Not every event with probability one occurs
 (e.g. randomly pick a point outside square's main diagonal)

Conditional probability

Consider a pair of random events A and B such that $\mathcal{P}\{B\} > 0$.

The <u>conditional probability</u> of A, given the occurrence of B, denoted as $\mathcal{P}\left\{A|B\right\}$, is defined as

$$\mathcal{P}\left\{A\mid B\right\} = \frac{\mathcal{P}\left\{A\cap B\right\}}{\mathcal{P}\left\{B\right\}}.$$

It follows that

$$\mathcal{P}\left\{A\cap B\right\} = \mathcal{P}\left\{A\mid B\right\} \times \mathcal{P}\left\{B\right\}.$$



Somebody rolls a pair of six-sided dice.

A =value rolled on die 1

B =value rolled on die 2

What is the probability that A = 2 given that $A + B \le 5$?



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What is the probability that A = 2 given that $A + B \le 5$?

A=2

| | | | | | В | | |
|---|---|---|---|---|----|----|----|
| | | 1 | 2 | 3 | 4 | 5 | 6 |
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| | 2 | 3 | 4 | 5 | 6 | 7 | 8 |
| A | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| ^ | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| | 5 | 6 | 7 | 8 | 9 | 10 | 11 |
| | 6 | 7 | 8 | 9 | 10 | 11 | 12 |



Conditional probability — Wikipedia, The Free Encyclopedia, 2017.



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$$\mathcal{P}\left\{ A\right\} =6/36$$



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| 71—2 | | | | | | | | | | | | |
|------|---|---|---|---|----|----|----|--|--|--|--|--|
| | | | В | | | | | | | | | |
| | | 1 | 2 | 3 | | | | | | | | |
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | | | | | |
| | 2 | 3 | 4 | 5 | 6 | 7 | 8 | | | | | |
| A | 3 | 4 | 5 | 6 | 7 | 8 | 9 | | | | | |
| Α. | 4 | 5 | 6 | 7 | 8 | 9 | 10 | | | | | |
| | 5 | 6 | 7 | 8 | 9 | 10 | 11 | | | | | |
| | 6 | 7 | 8 | 9 | 10 | 11 | 12 | | | | | |

$$\mathcal{P}\left\{ A\right\} =6/36$$

 $A+B \le 5$

| | | | В | | | | | | | |
|----|---|---|---|---|----|----|----|--|--|--|
| | | 1 | 2 | 3 | 4 | 5 | 6 | | | |
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | | | |
| | 2 | 3 | 4 | 5 | 6 | 7 | 8 | | | |
| Α | 3 | 4 | 5 | 6 | 7 | 8 | 9 | | | |
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| | 5 | 6 | 7 | 8 | 9 | 10 | 11 | | | |
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| | | | В | | | | | | |
|---|---|---|---|---|----|----|----|--|--|
| | | 1 | 2 | 3 | 4 | 5 | 6 | | |
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$$P\{A+B \le 5\} = 10/36$$

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$$\mathcal{P}\left\{A\right\} = 6/36$$

 $A+B \le 5$

| | | В | | | | | | | | | |
|---|------------------|-------------------|---|---|---|----------------------|--|--|--|--|--|
| | 1 | 2 | 3 | 4 | 5 | 6 | | | | | |
| 1 | 2 | 3 | 4 | 5 | 6 | 7 | | | | | |
| 2 | 3 | 4 | 5 | 6 | 7 | 8 | | | | | |
| 3 | 4 | 5 | 6 | 7 | 8 | 9 | | | | | |
| 4 | 5 | 6 | 7 | 8 | 9 | 10 | | | | | |
| 5 | 6 | 7 | 8 | 9 | 10 | 11 | | | | | |
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| | 2 3 4 5 | 1 2 3 3 4 4 5 5 6 | 1 2 3 2 3 4 3 4 5 4 5 6 5 6 7 | 1 2 3 4 2 3 4 5 3 4 5 6 4 5 6 7 5 6 7 8 | 1 2 3 4 1 2 3 4 5 2 3 4 5 6 3 4 5 6 7 8 5 6 7 8 9 | 1 2 3 4 5 6 7 8 9 10 | | | | | |

$$P\{A+B \le 5\} = 10/36$$

| 4 | A+B<5 |
|---|-------|
| | |

| | | | | | • | | |
|---|---|---|---|---|----|----|----|
| | | 1 | 2 | 3 | 4 | 5 | 6 |
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
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 $A+B \le 5$

| | | В | | | | | | | | | |
|---|---|---|---|---|----|----|-----|--|--|--|--|
| 1 | | 1 | 2 | 3 | 4 | 5 | 6 | | | | |
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | | | | |
| | 2 | 3 | 4 | 5 | 6 | 7 | 8 | | | | |
| | 3 | 4 | 5 | 6 | 7 | 8 | 9 | | | | |
| A | 4 | 5 | 6 | 7 | 8 | 9 | 10 | | | | |
| | 5 | 6 | 7 | 8 | 9 | 10 | -11 | | | | |
| | 6 | 7 | 8 | 9 | 10 | 11 | 12 | | | | |

$$\mathcal{P}\left\{A+B\leq 5\right\}=10/36$$

 $A \mid A+B \leq 5$

| | | В | | | | | |
|---|---|---|---|---|----|----|----|
| • | | 1 | 2 | 3 | 4 | 5 | 6 |
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| | | | | | | | |

$$\mathcal{P}\left\{A\,|\,A+B\leq 5\right\}=3/10$$



Conditional probability - Wikipedia, The Free Encyclopedia, 2017.

Independence of events

If the occurrence of an event B does not affect the occurrence of an event A one has

$$\mathcal{P}\left\{A\mid B\right\} = \mathcal{P}\left\{A\right\}.$$

In this way, once $\mathcal{P}\left\{A\cap B\right\}=\mathcal{P}\left\{A\mid B\right\}\times\mathcal{P}\left\{B\right\}$ it is true that

$$\mathcal{P}\left\{A\cap B\right\} = \mathcal{P}\left\{A\right\} \times \mathcal{P}\left\{B\right\}.$$

Events A and B in which the latter holds are said to be independent.

Remark:

This notion generalizes itself naturally to n events.



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen" Event 2: ♠ "spade"

Are these events independent?



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen"

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Are these events independent?

$$\mathsf{cards} = \left\{ \begin{array}{c} 2 \lozenge, 3 \lozenge, 4 \lozenge, 5 \lozenge, 6 \lozenge, 7 \lozenge, 8 \lozenge, 9 \lozenge, 10 \lozenge, J \lozenge, Q \lozenge, K \lozenge, A \lozenge, \\ 2 \clubsuit, 3 \clubsuit, 4 \clubsuit, 5 \clubsuit, 6 \clubsuit, 7 \clubsuit, 8 \clubsuit, 9 \clubsuit, 10 \clubsuit, J \clubsuit, Q \clubsuit, K \clubsuit, A \clubsuit, \\ 2 \heartsuit, 3 \heartsuit, 4 \heartsuit, 5 \heartsuit, 6 \heartsuit, 7 \heartsuit, 8 \heartsuit, 9 \heartsuit, 10 \heartsuit, J \heartsuit, Q \heartsuit, K \heartsuit, A \heartsuit, \\ 2 \spadesuit, 3 \spadesuit, 4 \spadesuit, 5 \spadesuit, 6 \spadesuit, 7 \spadesuit, 8 \spadesuit, 9 \spadesuit, 10 \spadesuit, J \spadesuit, Q \spadesuit, K \spadesuit, A \spadesuit \end{array} \right\}$$



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$$\mathcal{P}_1\left\{Q\right\} = 4/52$$



A card is drawn from a deck with 52 unknown cards.

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$$\mathcal{P}_1\left\{ Q
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$$\mathcal{P}_1 \{Q\} = 4/52$$

 $\mathcal{P}_1 \{ \spadesuit \} = 13/52$
 $\mathcal{P}_1 \{ Q \spadesuit \} = 1/52$



A card is drawn from a deck with 52 unknown cards.

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Are these events independent?

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$$\begin{array}{l} \mathcal{P}_1 \left\{ Q \right\} = 4/52 \\ \mathcal{P}_1 \left\{ \spadesuit \right\} = 13/52 \\ \mathcal{P}_1 \left\{ Q \spadesuit \right\} = 1/52 = \underbrace{4/52}_{\mathcal{P}_1 \left\{ Q \right\}} \times \underbrace{13/52}_{\mathcal{P}_1 \left\{ \spadesuit \right\}} \\ \end{array}$$



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen"

Event 2: ♠ "spade"

Are these events independent?

Probability space 1 (fair deck):

$$\mathsf{cards} = \left\{ \begin{array}{c} 2\lozenge, 3\lozenge, 4\lozenge, 5\lozenge, 6\lozenge, 7\lozenge, 8\lozenge, 9\lozenge, 10\lozenge, J\lozenge, Q\lozenge, K\lozenge, A\lozenge, \\ 2\$, 3\$, 4\$, 5\$, 6\$, 7\$, 8\$, 9\$, 10\$, J\$, Q\$, K\$, A\$, \\ 2\heartsuit, 3\heartsuit, 4\heartsuit, 5\heartsuit, 6\heartsuit, 7\heartsuit, 8\heartsuit, 9\heartsuit, 10\heartsuit, J\heartsuit, Q\heartsuit, K\heartsuit, A\heartsuit, \\ 2\$, 3\$, 4\$, 5\$, 6\$, 7\$, 8\$, 9\$, 10\$, J\$, Q\$, K\$, A\$ \end{array} \right\}$$

$$\mathcal{P}_1 \{Q\} = 4/52$$
 $\mathcal{P}_1 \{ \} = 13/52$
 $\mathcal{P}_1 \{Q\} = 1/52 = 4/52 \times 13/52$
 $\mathcal{P}_1 \{Q\} = 1/52 = 4/52 \times 13/52$

Events are independent!



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen" Event 2: ♠ "spade"

Are these events independent?



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen"

Event 2: 🏚 "spade"

Are these events independent?



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen"

Event 2: ♠ "spade"

Are these events independent?

$$\mathcal{P}_2\left\{Q\right\} = 28/52$$



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen"

Event 2: ♠ "spade"

Are these events independent?

Probability space 2 (unfair deck):

$$P_2\{Q\} = 28/52$$

 $P_2\{A\} = 1/2$



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen"

Event 2: ♠ "spade"

Are these events independent?

Probability space 2 (unfair deck):

$$\mathcal{P}_2 \{Q\} = 28/52$$

$$\mathcal{P}_2 \{\spadesuit\} = 1/2$$

$$\mathcal{P}_2 \{Q\spadesuit\} = 1/2$$



A card is drawn from a deck with 52 unknown cards.

Event 1: Q "queen"

Event 2: ♠ "spade"

Are these events independent?

Probability space 2 (unfair deck):

$$\mathcal{P}_{2} \{Q\} = 28/52$$

$$\mathcal{P}_{2} \{ \mathbf{A} \} = 1/2$$

$$\mathcal{P}_{2} \{Q\mathbf{A} \} = 1/2 \neq \underbrace{28/52}_{\mathcal{P}_{2} \{Q\}} \times \underbrace{1/2}_{\mathcal{P}_{2} \{\mathbf{A} \}}$$



A card is drawn from a deck with <u>52 unknown cards</u>.

Event 1: Q "queen"

Event 2: \spadesuit "spade"

Are these events independent?

Probability space 2 (unfair deck):

$$\mathcal{P}_{2} \{Q\} = 28/52$$

$$\mathcal{P}_{2} \{ \mathbf{A} \} = 1/2$$

$$\mathcal{P}_{2} \{ \mathbf{Q} \mathbf{A} \} = 1/2 \neq \underbrace{28/52}_{\mathcal{P}_{2} \{ \mathbf{Q} \}} \times \underbrace{1/2}_{\mathcal{P}_{2} \{ \mathbf{A} \}}$$

Events are not independent!

Further remarks on probability

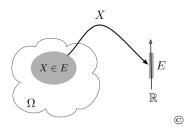
The last example shows that:

- Different probability spaces, for the same random experiment, can produce different predictions
- A probability space that does not accurately describe a random event can produce completely erroneous predictions
- The notion of independence strongly depends on the probability measure employed

Random variable

A mapping $X : \Omega \to \mathbb{R}$ is called a random variable (RV) if the preimage of every real number under X is a relevant event, i.e.,

$$X^{-1}(x) = \left\{ \omega \in \Omega : \ X(\omega) \le x \right\} \in \Sigma, \quad \text{for every } x \in \mathbb{R}.$$



A collection of events in Ω is mapped to an interval E on the real line under such mapping.

RV are numerical characteristics of interesting events.

Remark:

A random variable is a function from Ω to \mathbb{R} , not a real number.

Examples of random variables

1. Rolling die experiment

$$\Omega = \{1, 2, 3, 4, 5, 6\}$$

•
$$X(\omega) = \begin{cases} 1 & \text{if } \omega \text{ is even} \\ 0 & \text{if } \omega \text{ is odd} \end{cases}$$
 (random variable)

$$\Sigma = \{\phi, \{1, 3, 5\}, \{2, 4, 6\}, \Omega\}$$

2. Temperature (in Kelvin) measurement experiment

$$\Omega = [a, b] \subset [0, +\infty)$$

$$\Sigma = \mathcal{B}_{[a,b]}$$
 (Borel σ -algebra)

•
$$X(\omega) = -459.67 + 1.8 \omega$$
 (random variable)





Probability distribution

The <u>probability distribution</u> of X, denoted by F_X , is defined as the probability of the elementary event $\{X \le x\}$, i.e.,

$$F_X(x) = \mathcal{P}\left\{X \leq x\right\}.$$

 F_X is also known as <u>cumulative distribution function (CDF)</u> and has the following properties:

- $0 \le F_X(x) \le 1$
- \bullet F_X is a monotonic, non decreasing, right continuous function

•
$$\mathcal{P}\{x_1 < X \le x_2\} = F_X(x_2) - F_X(x_1) = \int_{x_1}^{x_2} dF_X(x)$$

- $\int_{\mathbb{R}} dF_X(x) = 1$
- $F_X(-\infty) = 0$ and $F_X(+\infty) = 1$

Probability density function

If the function F_X is differentiable, its derivative $p_X(x) = dF_X(x)/dx$ is called probability density function (PDF) of X, and one has

$$F_X(x) = \int_{-\infty}^x p_X(\xi) \, d\xi.$$

Note also that:

- $p_X(x) \ge 0$ for every $x \in \mathbb{R}$

Remark:

Intuitively, $p_X(x) dx$ can be thought of as the probability of X falling within the infinitesimal interval [x, x + dx].

Types of random variables

Discrete random variable
 Distribution is discrete.
 Assumes a denumerable number of values.
 Typically associated with counting processes.

Continuous random variable
 Distribution is continuous.
 Assumes a non-denumarable number of values.
 Typically associated with measuring processes.

- Mixed random variable
 Distribution has points of discontinuity.
 Assumes a non-denumarable number of values.
 A "mixture" of the two previous types.
- Singular random variable
 Distribution is not differentiable at any point.
 It has theoretical interest only.

Mathematical expectation operator

The mathematical expectation of a random variable X is defined as

$$\mathbb{E}\left\{X\right\} = \int_{\mathbb{R}} x \, dF_X(x).$$

The mathematical expectation is a linear operator since

$$\mathbb{E}\left\{\alpha_1 X_1 + \alpha_2 X_2\right\} = \alpha_1 \mathbb{E}\left\{X_1\right\} + \alpha_2 \mathbb{E}\left\{X_2\right\},\,$$

for any pairs of number α_1, α_2 and random variables X_1, X_2 .

<u>Theorem</u> (law of the unconscious statistician):

Given a measurable mapping $h : \mathbb{R} \to \mathbb{R}$ and a random variable X the expected value of h(X) is given by

$$\mathbb{E}\left\{h\left(X\right)\right\} = \int_{\mathbb{R}} h\left(x\right) \, dF_X(x).$$

Mean value

The <u>mean value</u> of the random variable X is defined as

$$\mu_X = \mathbb{E} \{X\}$$

$$= \int_{\mathbb{R}} x \, dF_X(x)$$

$$= \int_{\mathbb{R}} x \, p_X(x) \, dx.$$

(measure of the central tendency)

Remark:

The mean value μ_X is the constant which best approximate the random variable X. The error of this approximation is the standard deviation σ_X .

Variance

The <u>variance</u> of the random variable X is defined as

$$\sigma_X^2 = \mathbb{E}\left\{ (X - \mu_X)^2 \right\}$$

$$= \int_{\mathbb{R}} (x - \mu_X)^2 dF_X(x)$$

$$= \int_{\mathbb{R}} (x - \mu_X)^2 p_X(x) dx.$$

(measure of dispersion about the mean)

Note that variance can also be written as

$$\sigma_X^2 = \mathbb{E}\left\{X^2\right\} - \left(\mathbb{E}\left\{X\right\}\right)^2.$$

Remark:

 $\frac{\sigma_X^2}{\sigma_X^2}$ has the same unit as X^2 .

Standard deviation and variation coefficient

Other second-order statistics of X are the standard deviation

$$\sigma_X = \sqrt{\sigma_X^2},$$

and the variation coefficient

$$\delta_X = \sigma_X/\mu_X, \quad \mu_X \neq 0.$$

(both are measures of dispersion about the mean)

Remark:

 σ_X has the same unit as X and δ_X is dimensionless.

Probability Distributions

Uniform

- Notation: $\mathcal{U}(a,b)$
- Support: [*a*, *b*]
- Parameters:

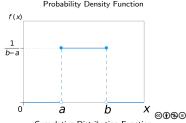
•
$$-\infty < a < b < +\infty$$
 — boundaries

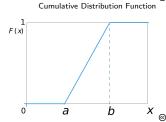
PDF:

$$p_X(x) = \frac{1}{b-a} \mathbb{1}_{[a,b]}(x)$$

- Statistics:

 - $\mu = \frac{1}{2}(a+b)$ $\sigma^2 = \frac{1}{12}(b-a)^2$





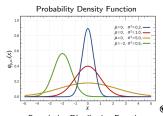


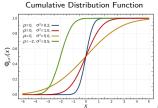
Uniform distribution — Wikipedia, The Free Encyclopedia, 2021.

Gaussian

- Notation: $\mathcal{N}(\mu, \sigma^2)$
- Support: $(-\infty, +\infty)$
- Parameters:
 - \bullet μ mean
 - σ^2 variance
- PDF:

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







Normal distribution — Wikipedia, The Free Encyclopedia, 2021.

Given the mean μ and standard deviation σ of a random variable. Is the distribution well-defined?

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$$p_X(x) = \frac{1}{\sqrt{6\pi}} \exp\left\{-\frac{(x-1)^2}{6}\right\}$$
 and $p_X(x) = \frac{1}{6} \mathbb{1}_{[-2,4]}(x)$ have $\mu = 1$ and $\sigma = \sqrt{3}$.

Given the mean μ and standard deviation σ of a random variable. Is the distribution well-defined?

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What type of information determines a distribution?

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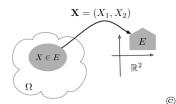
What type of information determines a distribution?

- cumulative distribution function
- probability density function (if exists)
- quantile function
- characteristic function
- moment-generating function (if exists)

Probability in Dimension n

Random vector

Let $\mathbf{x}=(x_1,\cdots,x_n)\in\mathbb{R}^n$. A <u>random vector</u> $\mathbf{X}=(X_1,\cdots,X_n)$ is a collection of n random variables $X_i:\Omega\to\mathbb{R}$ that together may be considered a (measurable) mapping $\mathbf{X}:\Omega\to\mathbb{R}^n$.



A collection of event in Ω is mapped into a region on the Euclidean space under such mapping.

Joint probability distribution

The joint probability distribution of random vector $\mathbf{X} = (X_1, \dots, X_n)$, denoted by $F_{\mathbf{X}}$, is defined as

$$F_{\mathbf{x}}(x_1,\cdots,x_n)=\mathcal{P}\left\{\left\{X_1\leq x_1\right\}\cap\cdots\cap\left\{X_n\leq x_n\right\}\right\}.$$

Thus,

$$\mathcal{P}\left\{\mathbf{a}<\mathbf{X}\leq\mathbf{b}\right\}=\int_{a_1}^{b_1}\cdots\int_{a_n}^{b_n}dF_{\mathbf{X}}(x_1,\cdots,x_n),$$

in which $\{ \mathbf{a} < \mathbf{X} \le \mathbf{b} \} = \{ a_1 < X_1 \le b_1 \} \cap \cdots \cap \{ a_n < X_n \le b_n \}.$

 $F_{\rm X}$ is also known as joint cumulative distribution function.

Joint probability density function

If $p_{\mathbf{x}}(x_1, \dots, x_n) = \partial^n F_{\mathbf{x}}/\partial x_1 \dots \partial x_n$ exists, for any x_1, \dots, x_n , then it is called joint probability density function of \mathbf{X} , and one has

$$F_{\mathbf{X}}(x_1,\cdots,x_n)=\int_{-\infty}^{x_1}\cdots\int_{-\infty}^{x_n}p_{\mathbf{X}}(\xi_1,\cdots,\xi_n)\,d\xi_1\cdots d\xi_n.$$

Note also that:

- $p_{\mathbf{X}}(x_1, \dots, x_n) \geq 0$ for every $(x_1, \dots, x_n) \in \mathbb{R}^n$
- $\int_{-\infty}^{+\infty} \cdots \int_{-\infty}^{+\infty} p_{\mathbf{X}}(x_1, \cdots, x_n) dx_1 \cdots dx_n = 1$

Marginal probability density function

The marginal probability density function of X_i is defined as

$$p_{X_i}(x_i) = \underbrace{\int_{-\infty}^{+\infty} \cdots \int_{-\infty}^{+\infty}}_{n-1 \text{ times}} p_{\mathbf{x}}(x_1, \cdots, x_n) dx_1 \cdots dx_{i-1} dx_{i+1} \cdots dx_n,$$

for $i = 1, \dots, n$.

Conditional distribution

Consider a pair of jointly distributed random variables X and Y. The <u>conditional distribution</u> of X, given the occurrence of the value y of Y, is defined as

$$F_{X|Y}(x \mid y) = \frac{F_{X,Y}(x,y)}{F_Y(y)}.$$

Thus

$$F_{X,Y}(x,y) = F_{X|Y}(x|y) \times F_Y(y),$$

and

$$p_{X,Y}(x,y) = p_{X|Y}(x|y) \times p_Y(y).$$

Remark:

This definition extends naturally to the n-dimensional case.

Independence of distributions

The random variables X and Y are said to be <u>independent</u> if the realization of X does not affect the probability distribution of Y, i.e.,

$$F_{X|Y}(x|y) = F_X(x).$$

Therefore, for independent random variable one has

$$F_{XY}(x,y) = F_X(x) \times F_Y(y),$$

and

$$p_{XY}(x,y) = p_X(x) \times p_Y(y).$$

Remark:

This definition extends naturally to the n-dimensional case.

Statistics of random vectors

• second-order random vector

$$\mathbb{E}\left\{\parallel \mathbf{X}\parallel^2\right\} = \int_{\mathbb{R}^n} \parallel \mathbf{x}\parallel^2 dF_{\mathbf{x}}(\mathbf{x}) < +\infty$$

• mean vector

$$\mathbf{m}_{\mathbf{X}} = \mathbb{E}\left\{\mathbf{X}\right\} = \int_{\mathbb{R}^n} \mathbf{x} \, dF_{\mathbf{x}}(\mathbf{x}) \in \mathbb{R}^n$$

• correlation matrix

$$[R_{\mathbf{XY}}] = \mathbb{E}\left\{\mathbf{XY}^T\right\} = \int_{\mathbb{R}^n} \mathbf{x} \mathbf{x}^T dF_{\mathbf{XY}}(\mathbf{x}, \mathbf{y}) \in \mathbb{R}^{n \times n}$$

• covariance matrix

$$[K_{\mathbf{XY}}] = \mathbb{E}\left\{ \left(\mathbf{X} - \mathbf{m_{\mathbf{X}}}\right) \left(\mathbf{Y} - \mathbf{m_{\mathbf{Y}}}\right)^{T} \right\} \in \mathbb{R}^{n \times n}$$

$$= [R_{\mathbf{XY}}] - \mathbf{m_{\mathbf{X}}} \mathbf{m_{\mathbf{Y}}}^{T}$$

Remark:

Matrices $[R_{XY}]$ and $[K_{XY}]$ are symetric positive semi-definite when X = Y.

Correlation of random variables

The random vectors $\mathbf{X}=(X_1,\cdots,X_n)$ and $\mathbf{Y}=(Y_1,\cdots,Y_n)$ are said to be <u>uncorrelated</u> if covariance matrix $[K_{\mathbf{XY}}]$ is null, i.e.,

$$[R_{\boldsymbol{X}\boldsymbol{Y}}] = \boldsymbol{m}_{\boldsymbol{X}} \boldsymbol{m}_{\boldsymbol{Y}}^{T}.$$

If two random vectors are independent, then they are uncorrelated.

 $independence \Longrightarrow uncorrelation$

But uncorrelated random vectors are not independent in general.

uncorrelation ⇒ independence

Remark:

Uncorrelated random vectors which the joint distribution is Gaussian are independent.

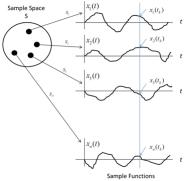
Notions of Random Processes

Random process

A real-valued random process (also called stochastic process) defined on probability space $(\Omega, \Sigma, \mathcal{P})$, indexed by $t \in \mathcal{T}$, is a mapping

$$(t,\omega)\in\mathcal{T}\times\Omega\to X(t,\omega)\in\mathbb{R},$$

such that, for fixed t, the output is a random variable $X(t,\cdot)$, while for fixed ω , $X(\cdot,\omega)$ is a function of t (sample function).



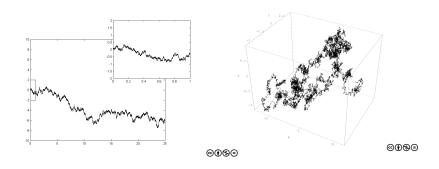
Interpretation and classification

Note that, by definition, a real-valued random process $X(t,\omega)$ is a collection of real-valued random variables indexed by a parameter, and can be thought of as a time-dependent random variable.

According to the nature of the set of indices \mathcal{T} , a stochastic process is classified as:

- discrete if \mathcal{T} is countable
- continuous if \mathcal{T} is uncountable

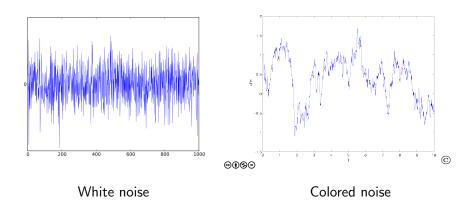
Wiener process (Brownian motion process)





Wiener process — Wikipedia, The Free Encyclopedia, 2021.

White noise and colored noise





White noise — Wikipedia, The Free Encyclopedia, 2021.

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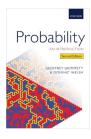
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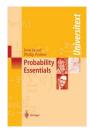


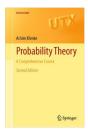
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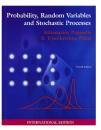


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