

# **Statistical Sciences**

# DoSS Summer Bootcamp Probability Module 2

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# Recap

#### Learnt in last module:

- Measurable spaces
  - ▶ Sample Space
  - $\triangleright \ \sigma$ -algebra
- Probability measures
  - $\triangleright$  Measures on  $\sigma$ -field
  - Basic results
- Conditional probability
  - ▶ Bayes' rule
  - ▷ Law of total probability



### **Outline**

- Independence of events
  - ▶ Pairwise independence, mutual independence
  - ▷ Conditional independence
- Random variables
- Distribution functions
- Density functions and mass functions
- Independence of random variables



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Recall the Bayes rule:

$$P(A \mid B) = \frac{P(A \cap B)}{P(B)}, \quad P(B) > 0$$

• What if B does not change our belief about A?

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- Equivalently,  $P(A \cap B) = P(A)P(B)$ .

### Independence of two events

Two events A and B are independent if  $P(A \cap B) = P(A)P(B)$ .

#### Remark:



#### Consider more than 2 events:

### Pairwise independence

We say that events  $A_1, A_2, \cdots, A_n$  are pairwise independent if

$$P(A_i \cap A_j) = P(A_i) \cdot P(A_j), \quad \forall i \neq j$$

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# Mutual independence

We say that events  $A_1, A_2, \dots, A_n$  are mutually independent or independent if for all subsets  $I \in \{1, 2, \dots, n\}$ 

$$\frac{P(\cap_{i\in I}A_i)=\prod_{i\in I}P(A_i)}{}$$



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Remark:

mutual indepune = pairwise independen is true



but does not hold.



#### **Example:**

- Toss two fair coins.;
- $A = \{ \text{ First toss is head} \}$ ,  $B = \{ \text{ Second toss is head } \}$ ,  $C = \{ \text{ Outcomes are the same } \}$ ;
- $A = \{HH, HT\}, B = \{HH, TH\}, C = \{HH, TT\};$

#### **Example:**

- Toss two fair coins.;
- A = { First toss is head}, B = { Second toss is head }, C = { Outcomes are the same };
- $A = \{HH, HT\}, B = \{HH, TH\}, C = \{HH, TT\};$
- $P(A \cap B) = P(A)P(B), P(A \cap C) = P(A)P(C), P(B \cap C) = P(B)P(C);$

A, B, C are purruise independit

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#### **Example:**

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- $P(A \cap B) = P(A)P(B), P(A \cap C) = P(A)P(C), P(B \cap C) = P(B)P(C);$



### Conditional independence

Two events A and B are conditionally independent given an event C if

$$P(A \cap B \mid C) = P(A \mid C)P(B \mid C).$$

definition is repland by conditional probability



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### **Example:**

Previous example continued:

• 
$$A = \{HH, HT\}, B = \{HH, TH\}, C = \{HH, TT\};$$

• 
$$P(A \cap B \mid C) = ?, P(A \mid C)P(B \mid C) = ?$$

$$(P(APB|C) = \frac{(P(APPAC))}{(PCC)} = \frac{4}{3}$$



### Conditional independence

Two events A and B are conditionally independent given an event C if

$$P(A \cap B \mid C) = P(A \mid C)P(B \mid C).$$

#### **Example:**

Previous example continued:

- $A = \{HH, HT\}, B = \{HH, TH\}, C = \{HH, TT\};$
- $P(A \cap B \mid C) = ?, P(A \mid C)P(B \mid C) = ?$

#### Remark:

Equivalent definition:

$$P(A \mid B, C) = P(A \mid C).$$



### Random variables

#### Idea:

Instead of focusing on each events themselves, sometimes we care more about functions of the outcomes.



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#### Example:

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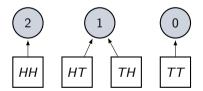


Figure: Mapping from the sample space to the numbers of heads



### **Random Variables**

#### **Example:**

- Select twice from red and black ball with replacement: {RR, RB, BR, BB}
- Care about the number of red balls:  $\{2, 1, 0\}$

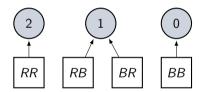


Figure: Mapping from the sample space to the numbers of red balls



### **Random Variables**

#### Merits:

- Mapping the complicated events on  $\sigma$ -field to some numbers on real line.
- Simplify different events into the same structure



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### Random Variables

Consider sample space  $\Omega$  and the corresponding  $\sigma$ -field  $\mathcal{F}$ , for  $X:\Omega\to\mathbb{R}$ , if

$$A \in \mathcal{R}$$
 (Borel sets on  $\mathbb{R}$ )  $\Rightarrow X^{-1}(A) \in \mathcal{F}$ ,

then we call X as a random variable.

Here 
$$X^{-1}(A) = \{\omega : X(\omega) \in A\}.$$

We can also say X is  $\mathcal{F}$ -measurable.



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$$(\Omega, \mathcal{F}, \mathbb{P}) \xrightarrow{\times} (\mathbb{R}, \mathbb{R}, \mathbb{M}_{\times})$$

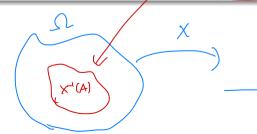
Probability measure  $P(\cdot)$  on  $\mathcal{F}$  can induce a measure  $\mu(\cdot)$  on  $\mathcal{R}$ :

### Probability measure on ${\cal R}$

We can define a probability  $\mu$  on  $(R, \mathcal{R})$  as follows:

$$\forall A \in \mathcal{R}, \quad \mu(A) := P(X^{-1}(A)) = P(X \in A).$$

Then  $\mu$  is a probability measure and it is called the distribution of X.





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### Probability measure on ${\cal R}$

We can define a probability  $\mu$  on  $(R, \mathcal{R})$  as follows:

$$\forall A \in \mathcal{R}, \quad \mu(A) := P(X^{-1}(A)) = P(\underline{X \in A}).$$

Then  $\mu$  is a probability measure and it is called the distribution of X.

#### Remark:

Verify that  $\mu$  is a probability measure.

• 
$$\mu(\mathbb{R}) = 1$$
.

• If 
$$A_1, A_2, \dots \in \mathcal{R}$$
 are disjoint, then  $\mu(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} \mu(A_i)$ .



Consider the special set that belongs to  $\mathcal{R}$ ,  $(-\infty, x]$ :

#### Cumulative Distribution Function

The cumulative distribution function of random variable X is defined as follows:

$$F(x) := P(X \le x) = P(X^{-1}((-\infty, x])), \quad \forall x \in \mathbb{R}.$$

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#### Cumulative Distribution Function

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#### **Properties of CDF:**

- $\lim_{x \to \infty} F(x) = 1$ ,  $\lim_{x \to -\infty} F(x) = 0$   $\leftarrow$  can show by continuity of probability
- $F(\cdot)$  is non-decreasing
- $F(\cdot)$  is right-continuous  $\leftarrow$
- Let  $F(x^-) = \lim_{y \nearrow x} F(y)$ , then  $F(x^-) = P(X < x)$
- $P(X = x) = F(x) F(x^{-}) \leftarrow 0$  if



there is a jump at X



# Proofs of properties of CDF (first 2 properties):

When 
$$X / \infty$$
  $\{X \leq X\}$   $A$ .  $A$ .

By the continuety from helow,

 $\lim_{X \neq \infty} P(X \leq X) = P(Q) = 1$ .

When  $X \geq -\infty$ ,  $\{X \leq X\} \geq \emptyset$ .

By the continuity from whome,

 $\lim_{X \neq \infty} P(X \leq X) = P(\emptyset) = 0$ .

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Let 
$$\chi_1 \neq \chi_2$$
.

Then  $\{\chi \neq \chi_1\} \subset \{\chi \neq \chi_2\}$ 

So,  $\{\chi \neq \chi_2\} = \{\chi \neq \chi_1\} \cup \{\chi_1 < \chi \neq \chi_2\}$ 

disjoint.

By countable additivity

$$F(x_2) = P(x_2 x_2) = P(x_2 x_1) + P(x_1 < x_2 x_2)$$

$$\geq P(x_2 x_1) = F(x_1)$$

#### Classification of the random variables:

- Discrete random variable: X takes either a finite or countable number of possible numbers.
- Continuous random variable: The CDF is continuous everywhere.



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- Discrete random variable: X takes either a finite or countable number of possible numbers.
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### Another perspective (function):

- Discrete random variable: focus on the probability assigned on each possible values
- Continuous random variable: consider the derivative of the CDF (The continuous monotone CDF is differentiable almost everywhere)



# Density functions and mass functions Piscote case

### Probability mass function

The probability mass function of X at some possible value x is defined by

$$p_X(x) = P(X = x).$$

#### Relationship between PMF and CDF:

$$F(x) = P(X \le x) = \sum_{y \le x} p_X(y)$$
this sum is valid

suce this is countable sum

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#### Relationship between PMF and CDF:

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#### **Example:**

Toss a coin



continuous

### Probability density function

The probability density function of X at some possible value x is defined by

$$f_X(x) = \frac{d}{dx}F(x).$$

#### Relationship between PDF and CDF:

$$F(x) = P(X \le x) = \int_{y \le x} f_X(y) \ dy = \int_{-\infty}^x f_X(y) \ dy$$

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#### **Example:**



### Define independence of random variables based on independence of events:

### Independence of random variables

Suppose  $X_1, X_2, \dots, X_n$  are random variables on  $(\Omega, \mathcal{F}, P)$ , then





 $\begin{array}{ll} X_1, X_2, \cdots, X_n \text{ are independent} \\ \Leftrightarrow & \{X_1 \in A_1\}, \{X_2 \in A_2\}, \cdots, \{X_n \in A_n\} \text{ are independent}, \\ \Leftrightarrow & P(\cap_{i=1}^n \{X_i \in A_i\}) = \prod_{i=1}^n P(\{X_i \in A_i\}) \end{array}$ 



$$\Leftrightarrow P(\cap_{i=1}^n \{X_i \in A_i\}) = \prod_{i=1}^n P(\{X_i \in A_i\})$$



#### Example:

Toss a fair coin twice, denote the number of heads of the i-th toss as  $X_i$ , then  $X_1$  and  $X_2$  are independent.

- A; can be {0} or {1}
- $\{(0,0),(0,1),(1,0),(1,1)\}$

• 
$$P(\{X_1 \in A_1\} \cap \{X_2 \in A_2\}) = \frac{1}{4}$$
  
•  $P(\{X_1 \in A_1\}) = P(\{X_2 \in A_2\}) = \frac{1}{2}$ 



### **Example:**

Toss a fair coin twice, denote the number of heads of the *i*-th toss as  $X_i$ , then  $X_1$  and  $X_2$  are independent.

- $A_i$  can be  $\{0\}$  or  $\{1\}$
- $\{(0,0),(0,1),(1,0),(1,1)\}$
- $P({X_1 \in A_1} \cap {X_2 \in A_2}) = \frac{1}{4}$
- $P({X_1 \in A_1}) = P({X_2 \in A_2}) = \frac{1}{2}$

#### Remark:

How to check independence in practice?



### Corollary of independence

If  $X_1, \dots, X_n$  are random variables, then  $X_1, X_2, \dots, X_n$  are independent if

$$P(X_1 \leq x_1, \cdots, X_n \leq x_n) = \prod_{i=1}^n P(X_i \leq x_i)$$



### Corollary of independence

If  $X_1, \dots, X_n$  are random variables, then  $X_1, X_2, \dots, X_n$  are independent if

$$P(X_1 \leq x_1, \cdots, X_n \leq x_n) = \prod_{i=1}^n P(X_i \leq x_i)$$

#### Remark:

### Independence of discrete random variables

Suppose  $X_1, \dots, X_n$  can only take values from  $\{a_1, \dots\}$ , then  $X_i$ 's are independent if

$$P(\cap\{X_i=a_i\})=\prod_{i=1}^n P(X_i=a_i)$$



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### **Problem Set**

**Problem 1:** Give an example where the events are pairwise independent but not mutually independent.

**Problem 2:** Verify that the measure  $\mu(\cdot)$  induced by  $P(\cdot)$  is a probability measure on  $\mathcal{R}$ .

**Problem 3:** Prove properties 3 - 5 of CDF  $F(\cdot)$ .

**Problem 4:** Bob and Alice are playing a game. They alternatively keep tossing a fair coin and the first one to get a H wins. Does the person who plays first have a better chance at winning?

