

Probability Review

CSEN266

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Random Variable

- Basic Idea

- Don't know how to completely determine what value will occur.
- Can only specify probabilities of values occurring.

- Examples

- Number on rolled dice
- Temperature at specified time of day
- Friday night attendance at a cinema

Two Types of Random Variable

- Discrete RVs

- The number of rolled dice
- Friday night attendance at a cinema

- Continuous RVs

- Temperature
- Blood pressure

Notation

- **RV:** upper case letter
 - e.g. X
- **The value of an RV:** lower case letter
 - e.g. $X = x$
- **Upper case P :** the probability, or probability mass function (PMF) of a discrete RV
 - e.g. $P(X = x)$
- **Lower case p , or f :** the probability density function (PDF) of a continuous RV
 - e.g. $p(x), f(x)$

Discrete Random Variable

- Probability Mass Function (PMF)

- $P(X = x_i) \geq 0, i = 1, 2, \dots, N$

- $\sum_{i=1}^N P(X = x_i) = 1$

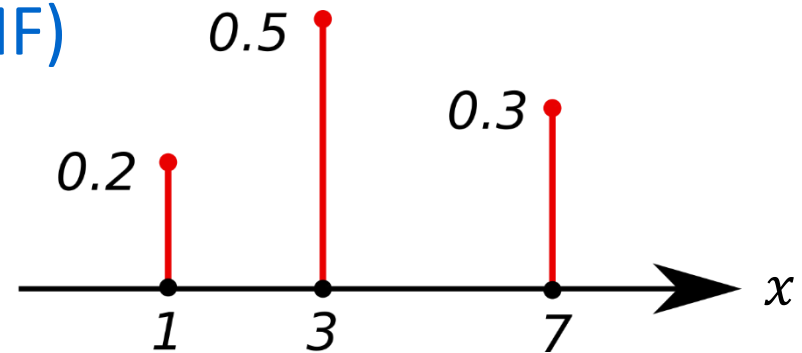


Figure 1

- Cumulative Distribution Function (CDF)

- $F(x) = P(X \leq x)$

- Example in **Figure 1**

- $F(1) = P(X \leq 1) = 0.2$

- $F(3) = P(X \leq 3) = 0.2 + 0.5 = 0.7$

- $F(4) = P(X \leq 4) = ?$
 $= 0.2 + 0.5 = 0.7$

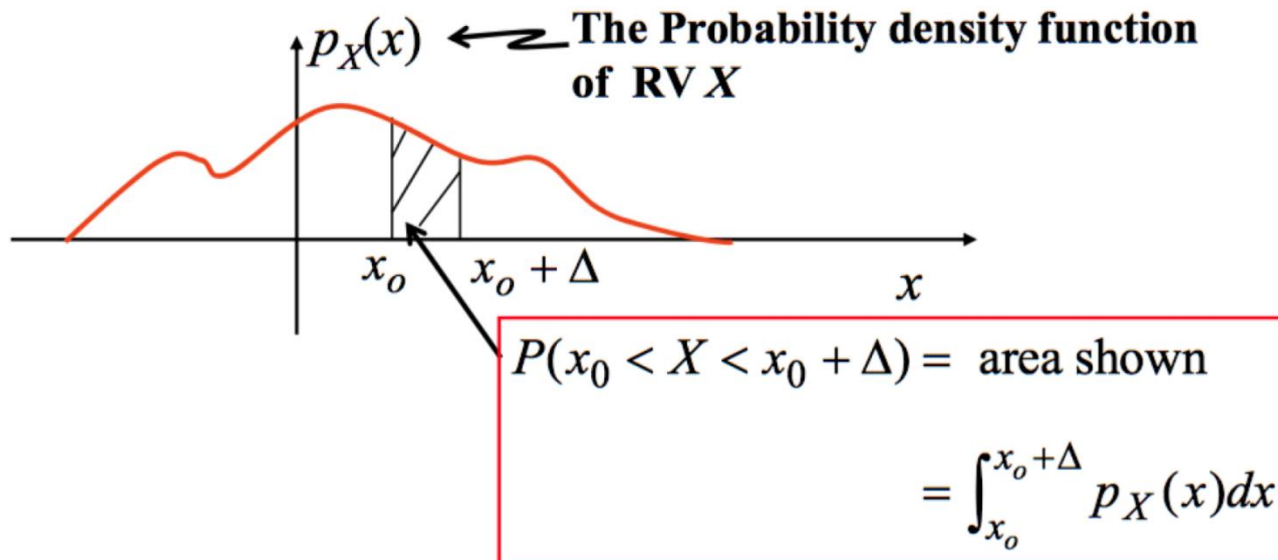
Continuous Random Variable

- Given continuous RV X

Want to know $P(X = x_0)$?

$$P(X = x_0) = 0$$

- Because the probability cannot sum up to infinity
- Use probability density function (PDF) to describe the distribution of the continuous RV X



The Mean (or Expectation) of a RV

- For discrete RVs

- $\mu_X = E\{X\} = \sum_{i=1}^N x_i P(X = x_i)$

- For continuous RVs

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

Example

- Consider two independent coin tosses, each with a $\frac{3}{4}$ probability of a head, and let X be the number of heads obtained
- Its PMF is

$$P(X = k) = \begin{cases} (1/4)^2 & \text{if } k = 0, \\ 2 \cdot (1/4) \cdot (3/4) & \text{if } k = 1, \\ (3/4)^2 & \text{if } k = 2, \end{cases}$$

- Its expectation is

$$E(X) = \sum_k k \times P(X = k) = 0 \cdot \left(\frac{1}{4}\right)^2 + 1 \cdot \left(2 \cdot \frac{1}{4} \cdot \frac{3}{4}\right) + 2 \cdot \left(\frac{3}{4}\right)^2 = \frac{24}{16} = \frac{3}{2}.$$

The Variance of a RV

- **Variance:** characterizes how much the RV deviates from the mean
- **Discrete RV**

$$\sigma^2 = E\{(X - \mu_X)^2\} = \sum_{i=1}^N (x_i - \mu_X)^2 \times P(X = x_i)$$

- **Continuous RV**

$$\sigma^2 = E\{(X - \mu_X)^2\} = \int_{-\infty}^{+\infty} (x - \mu_X)^2 \cdot p_X(x) dx$$

Joint Probability

- Two RVs

- $$P(X = x, Y = y) = P(X = x) \times P(Y = y|X = x)$$
$$= P(Y = y) \times P(X = x|Y = y)$$

- If RVs X and Y are **independent**, then

$$P(X = x, Y = y) = P(X = x) \times P(Y = y)$$

Example

- We toss a fair coin two successive times. A and B are the events:

$A = \{\text{the 1}^{\text{st}} \text{ toss is a head}\}$

$B = \{\text{the 2}^{\text{nd}} \text{ toss is a tail}\}$

- Find $P(A, B)$

Answer:

Events A and B are independent

$$P(A, B) = P(A) \times P(B) = \frac{1}{2} \times \frac{1}{2} = \frac{1}{4}$$

Conditional Probability

- $P(X = x|Y = y) = \frac{P(X=x,Y=y)}{P(Y=y)}$

- $P(Y = y|X = x) = \frac{P(X=x,Y=y)}{P(X=x)}$

Example

- We toss a fair coin three successive times. A and B are the events:

$A = \{\text{more heads than tails come up}\}$

$B = \{\text{1}^{\text{st}} \text{ toss is a head}\}$

- Find the probability $P(A)$

- Answer:

- Total number of combinations: $2^3 = 8$

- Event A : HHT, HTH, THH, HHH

- $P(A) = \frac{4}{8} = \frac{1}{2}$

Example

- We toss a fair coin three successive times. A and B are the events:

$A = \{\text{more heads than tails come up}\}$

$B = \{1^{\text{st}} \text{ toss is a head}\}$

- Find the probability $P(B)$

- Answer:

- Total number of combinations: $2^3 = 8$

- Event B : HHH, HHT, HTH, HTT

- $P(B) = \frac{4}{8} = \frac{1}{2}$

Example

- We toss a fair coin three successive times. A and B are the events:

$A = \{\text{more heads than tails come up}\}$

$B = \{1^{\text{st}} \text{ toss is a head}\}$

- Find the probability $P(A, B)$

- **Answer:**

- Event A : HHT, HTH, THH, HHH

- Event B : HHH, HHT, HTH, HTT

- $P(A, B) = P(A) \times P(B|A) = \frac{1}{2} \times \frac{3}{4} = \frac{3}{8}$

- Events A and B are not independent

Joint Probability of N RV's

- Chain Rule

- $P(X_1 = x_1, X_2 = x_2, \dots, X_N = x_N)$

$$= P(X_1 = x_1) \times P(X_2 = x_2 | X_1 = x_1)$$

$$\times P(X_3 = x_3 | X_1 = x_1, X_2 = x_2) \times \dots$$

$$\times P(X_N = x_N | X_1 = x_1, X_2 = x_2, \dots, X_{N-1} = x_{N-1})$$

Joint Probability of N RV's

- If RVs X_1, X_2, \dots, X_N are **independent**, then

$$\begin{aligned} &P(X_1 = x_1, X_2 = x_2, \dots, X_N = x_N) \\ &= P(X_1 = x_1) \times P(X_2 = x_2) \times \dots \times P(X_N = x_N) \\ &= \prod_{i=1}^N P(X_i = x_i) \end{aligned}$$

Total Probability

- Assume the observed data point is: $X = x$
- The observation may come from one of N hypotheses (classes): $H_i, i = 1, \dots, N$
- $P(X = x)$
 $= P(H_1)P(X = x|H_1) + P(H_2)P(X = x|H_2) + \dots$
 $+ P(H_N)P(X = x|H_N)$
- $P(X = x) = \sum_{i=1}^N P(H_i)P(X = x|H_i)$

Example

- You enter a chess tournament where your probability of winning a game is 0.3 against half the players (call them type 1), 0.4 against a quarter of the players (call them type 2), and 0.5 against the remaining quarter of the players (call them type 3). You play a game against a randomly chosen opponent. What is the probability of winning?

Let A_i be the event of playing with an opponent of type i . We have

$$P(A_1) = 0.5, P(A_2) = 0.25, P(A_3) = 0.25$$

Let also B be the event of winning. We have

$$\mathbf{P}(B | A_1) = 0.3, \quad \mathbf{P}(B | A_2) = 0.4, \quad \mathbf{P}(B | A_3) = 0.5.$$

Thus, by the total probability theorem, the probability of winning is

$$\begin{aligned} \mathbf{P}(B) &= \mathbf{P}(A_1)\mathbf{P}(B | A_1) + \mathbf{P}(A_2)\mathbf{P}(B | A_2) + \mathbf{P}(A_3)\mathbf{P}(B | A_3) \\ &= 0.5 \cdot 0.3 + 0.25 \cdot 0.4 + 0.25 \cdot 0.5 \\ &= 0.375. \end{aligned}$$

Bayes' Rule

- Prior Probability
- $P(H_i)$: the probability of hypothesis H_i
- Posteriori probability
- $P(H_i|X = x)$: the probability of hypothesis H_i given the observation that $X = x$

- Bayes Rule

$$P(H_i|X = x) = \frac{P(H_i)P(X = x|H_i)}{P(X = x)}$$

$$\text{where } P(X = x) = \sum_{i=1}^N P(H_i)P(X = x|H_i)$$

Example

- Let us return to the chess problem. Here A_i is the event of getting an opponent of type i , and

$$\mathbf{P}(A_1) = 0.5, \quad \mathbf{P}(A_2) = 0.25, \quad \mathbf{P}(A_3) = 0.25.$$

Also, B is the event of winning, and

$$\mathbf{P}(B | A_1) = 0.3, \quad \mathbf{P}(B | A_2) = 0.4, \quad \mathbf{P}(B | A_3) = 0.5.$$

Suppose that you win. What is the probability $\mathbf{P}(A_1 | B)$ that you had an opponent of type 1?

Using Bayes' rule, we have

$$\begin{aligned} \mathbf{P}(A_1 | B) &= \frac{\mathbf{P}(A_1)\mathbf{P}(B | A_1)}{\mathbf{P}(A_1)\mathbf{P}(B | A_1) + \mathbf{P}(A_2)\mathbf{P}(B | A_2) + \mathbf{P}(A_3)\mathbf{P}(B | A_3)} \\ &= \frac{0.5 \cdot 0.3}{0.5 \cdot 0.3 + 0.25 \cdot 0.4 + 0.25 \cdot 0.5} \\ &= 0.4. \end{aligned}$$