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Probability and Random Processes

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Random variables: definition

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Definition

Let (Ω, \mathbb{P}) be a probability space. A function

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

is called a real valued *random variable*. Similarly, a function $X : \Omega \to \mathbb{R}^n$ is called a vector-valued random variable.

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Discrete random variables

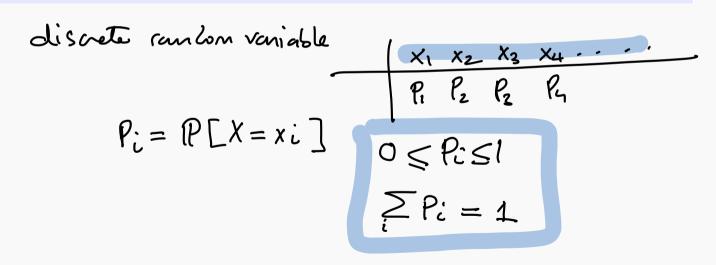
Discrete random variables

Definition

The probability mass function of a random variable X with sets of values x_1, \ldots is defined by

$$p(x) = \mathbb{P}[X = x].$$

Note that for every i, we have $p(x_i) = p_i > 0$.



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Example

A die is rolled. Let X be the random variable that tells us whether the outcome is larger than 4 or not. $\mathcal{P}(X>4) = \mathcal{P}(\{5,6\}) = \frac{1}{3}$.

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- The coin shows H with probability p and T with probability 1 p.
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Suppose n = 2: Then

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$$\mathbb{P}[HT] = \mathbb{P}[\text{ first} H] \mathbb{P}[\text{ second} T] = p(1-p).$$

$$\mathbb{P}[TH] = \mathbb{P}[\text{ first} T] \mathbb{P}[\text{ second} H] = (1-p)p.$$

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$$\mathbb{X} = \text{Hods}$$

$$\text{HH} \longrightarrow 2 \qquad p^{2}$$

$$\text{HT TH} \longrightarrow 1 \quad 2p(1-p)$$

$$\text{TT} \longrightarrow 0 \quad (1-p)^{2}$$

$$(1)$$

Suppose n = 3. Then the number of heads could be 0, 1, 2, 3

Bernoulli tail

HHH
$$\longrightarrow$$
 3 p^3

HHT HTH THH \longrightarrow 2 $3p^2(1-p)$

HTT THT TTH \longrightarrow 1 $3p(1-p)^2$

TTT \longrightarrow 0 $(1-p)^3$

Definition

A random variable X has the Binomial distribution with parameters (n, p) if,

$$\mathbb{P}[X = k] = \begin{cases} \binom{n}{k} p^k (1-p)^{n-k} & \text{if } 0 \le k \le n \\ 0 & \text{otherwise} \end{cases}$$

$$P(HH...HT...T) = p^{k}(1-p)^{n-k}$$

$$(n) = \# warp of picking k items out of n items.$$

M=1,2,3,--

0<9<1

Examples

Suppose X is a Binomial random variable with parameter (n, p) = (5, 1/3). Find $\mathbb{P}[X \ge 4]$ and $\mathbb{P}[X \le 1]$.

$$P(X \ge 4) = P(X = 4 \text{ or } X = 5) =$$

$$P(X = 5) = P' = (1/3)^{5} \qquad \frac{X | 0 | 2 | 3 | 4 | 5}{I(\sqrt{3})}$$

$$P(X = 4) = (5) | P'(1 - P)^{1}$$

$$= 5P'(1 - P) = 5(1/3)^{4} \cdot \frac{7}{3}$$

$$P(X \ge 1) = P(X = 0) + P(X = 1)$$

$$(\frac{7}{2})^{5} + (\frac{7}{3})^{4}(\frac{7}{3})^{4}(\frac{7}{3})^{1}$$

$$P(X=k) = \binom{n}{k} P(1-p),$$
Special case
$$P = \frac{1}{2}$$

$$P(X=k) = {n \choose k} (1/2)^k \cdot (1/2)^{n-k}$$

$$= {n \choose k} \cdot (1/2)^n = \frac{{n \choose k}}{2!}$$

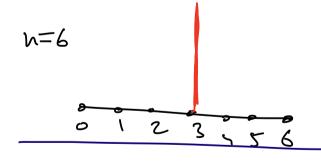
n=5

$$\binom{1}{2} = 2 \qquad \binom{2}{2} = 2$$

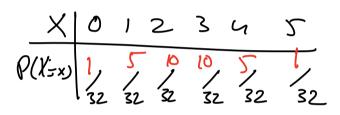
$$\binom{9}{2} = 1 \qquad \binom{2}{2} = 2$$

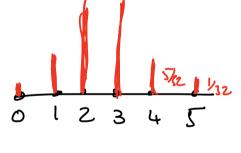
$$\binom{\Gamma}{2} = 10$$

$$\left(\frac{5}{3}\right) = 10$$

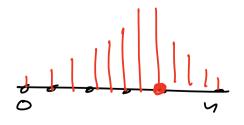


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Geometric distribution

Consider a coin that turns up heads with probability p and tails with probability 1-p. The coin is flipped until the a heads shows ups. Let X be the number of the flips needed. Assuming that the outcome of the flips are independent, we have

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$$X = \text{ first occurance}$$

$$\Rightarrow f \quad H \qquad \mathbb{P}[X = n] = (1 - p)^{n-1}p.$$

$$\mathbb{P}(X = 1) = P$$

$$\mathbb{P}(X = 2) = (1 - p) \cdot P$$

$$\mathbb{P}(X = n) = (1 - p)^{n-1}. \quad P$$

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Geometric distribution

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$$\mathbb{P}\left[X=n\right]=\left(1-p\right)^{n-1}p.$$

Definition

A discrete random variable X has Geometric distribution with parameter p if

$$\mathbb{P}[X=k] = \begin{cases} p^*(1-p) & \text{if } k \geq 1 \\ 0 & \text{otherwise} \end{cases}$$

Example

Let X be a random variable with the geometric distribution with parameter p. Find $\mathbb{P}[X > k]$.

$$P[X>k] = P(X=h+1,h+2,...)$$

$$= \sum_{i=u+1}^{\infty} P(X=i) = \sum_{i=u+1}^{\infty} P(1-p)^{i-1}$$

$$= P[(1-p)^{u} + (1-p)^{u+1} + (1-p)^{u+2} + ...]$$

$$= P(1-p)^{k} [1 + (1-p) + (1-p)^{2} + ...] = P(1-p)^{k} \cdot \frac{1}{1-(1-p)}$$

$$1 + 9 + 9^{2} + 9^{3} + ... = \frac{1}{1-9} = P(1-p)^{k}.$$

$$P(X)h) = P(first h trials were)$$

= $(1-p)^{k}$.

Poisson random variables

There are n=200 people in a seminar. Suppose that the mobile phone of each participant rings during the meeting with probability p=0.01. Assuming the independence of the ringing events, find the probability of the event that k phones ring during the meeting.

N parliagats

P probabil of xuccan for each parliagat

N large
$$p$$
 small $np = \lambda$
 $X = \# \text{success}$

distribution of X $P(X = k)$

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$$P(\chi = k) = \binom{n}{k} \cdot P^{k} (1-P)^{n-k}$$

$$P \xrightarrow{N \to \infty} NP = \lambda \Rightarrow P = \frac{\lambda}{n}$$

$$P(\chi = 0) = \binom{n}{0} \cdot P^{0} (1-P)$$

$$= (1-\frac{\lambda}{n})^{n}$$

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Poisson distribution

Definition

A discrete random variable X has Poisson distribution with parameter λ if

$$P[X = k] = \begin{cases} e^{-\lambda} \frac{\lambda^{k}}{k!} & \text{if } MM \mid k = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

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Distribution function of a random variable

Definition

Let $X : \Omega \to \mathbb{R}$ be a random variable. The *probability distribution function* of X is the function $F_X : \mathbb{R} \to [0,1]$ defined by

$$F_X(t) = \mathbb{P}[X \leq t].$$

Examples: the distribution function of a Bernoulli random variable

