

CIS 2033 based on

Dekking et al. A Modern Introduction to Probability and Statistics. 2007

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C4: DISCRETE_RANDOM_VARIABLES

DEFINITION OF A DISCRETE RANDOM VARIABLE

Let Ω be an arbitrary sample space. We can have 2 types of sample spaces:

0) FINITE: $\Omega = \{a_1, a_2, \dots, a_n\}$

1) INFINITE: $\Omega = a_1, a_2, \dots$

As Ω is arbitrary, it can assume any arbitrary collection of arbitrary values.

A discrete random variable X is a function $X: \Omega \rightarrow \mathbb{R}$.

X receives an arbitrary sample space Ω as input, and maps Ω to a set of Real Numbers \mathbb{R} .

DEFINITION OF A PROBABILITY MASS FUNCTION

The Probability Mass Function p of a discrete random variable X is the function $p: \mathbb{R} \rightarrow [0, 1]$. p receives a set of Real Numbers \mathbb{R} as input, and maps the set to the inclusive interval $[0, 1]$. We define p as:

$$p(a) = P(\{X = a\}) \mid -\infty < a < \infty.$$

Explanation of the formula:

- 0) Since Ω is arbitrary, value a assumes an arbitrary range.
- 1) Set $\{X = a\} \subseteq \Omega$ describes an event or events that occur in Ω , for a particular a .
- 2) $P(\{X = a\})$ is the probability that a particular a will occur in Ω .

If X assumes a finite number of values $\{a_1, a_2, \dots, a_n\}$, then:

- 0) $p(a_i) > 0$.
- 1) $p(a_1) + p(a_2) + \dots + p(a_n) = 1$.
- 2) $p(a) = 0$ for all other values of a .

If X assumes an infinite number of values a_1, a_2, \dots , then:

- 0) $p(a_i) > 0$.
- 1) $p(a_1) + p(a_2) + \dots = 1$.
- 2) $p(a) = 0$ for all other values of a .

DEFINITION OF A CUMULATIVE/DISTRIBUTION FUNCTION

The Cumulative/Distribution Function F of a random variable X is the function $F: \mathbb{R} \rightarrow [0, 1]$.

F receives a set of Real Numbers \mathbb{R} as input, and maps the set to the inclusive interval $[0, 1]$.

We define F as:

$$F(a_w) = P(\{X \leq a_w\}) \mid -\infty < a_w < \infty.$$

Another valid definition of F is:

$$F(a_w) = \sum_{k=a_1}^{a_w} [p(k)].$$

Explanation of the formulas:

0) Since Ω is arbitrary, a_w assumes an arbitrary range.

1) Set $[\{X \leq a_w\} \subseteq \Omega] \mid [\forall k \in X, [k \leq a_w]]$.

2) $P(\{X \leq a_w\})$ is the cumulative probability of all such events $k \in X$.

DEFINITION OF A BERNOULLI DISTRIBUTION

A discrete random variable X has a Bernoulli Distribution with these parameters:

Success Probability $p \mid 0 \leq p \leq 1$.

Failure Probability $1-p \mid 0 \leq 1-p \leq 1$.

We define the Probability Mass Function p_x of a Bernoulli Distribution as:

$$p_x(1) = P(\{X = 1\}) = p.$$

$$p_x(0) = P(\{X = 0\}) = 1 - p.$$

Explanation of the formulas:

0) A Bernoulli Distribution is suitable to model experiments with only 2 possible outcomes:

$$1 = \text{"success"}; 0 = \text{"failure"}.$$

1) $p_x(1) = P(\{X = 1\})$ is the probability that the event $1 = \text{"success"}$ shall occur.

2) $p_x(0) = P(\{X = 0\})$ is the probability that the event $0 = \text{"failure"}$ shall occur.

We denote this Bernoulli Distribution as $Ber(p)$.

DEFINITION OF A BINOMIAL DISTRIBUTION

Intuition: The probability of k successful outcomes in n trials.

A discrete random variable X has a Binomial Distribution with these parameters:

$n \mid n = 1, 2, \dots$

Success Probability $p \mid 0 \leq p \leq 1$.

Failure Probability $1 - p \mid 0 \leq 1 - p \leq 1$.

We define the Probability Mass Function p_x of a Binomial Distribution as:

$$p_x(k) = P(\{X = k\}) = \binom{n}{k} [p^k] [1 - p]^{[n-k]} \mid k = 0, 1, \dots, n.$$

Explanation of the formula:

0) $\binom{n}{k} = \frac{n!}{k![n-k]}$: Combination: from n elements, the number of order-insensitive ways to choose k elements.

1) $[p^k]$: The probability of a "success," repeated k times.

2) $[1 - p]^{[n-k]}$: The probability of a "failure," repeated $[n - k]$ times.

We denote this Binomial Distribution by $Bin(n, p)$.

DEFINITION OF A GEOMETRIC DISTRIBUTION

Intuition: The probability of being successful in k th trial. This means that all previous $k-1$ failed.

A discrete random variable X has a Geometric Distribution with these parameters:

Success Probability $p \mid 0 \leq p \leq 1$. Failure Probability $1-p \mid 0 \leq 1-p \leq 1$.

We define the Probability Mass Function p_x of a Geometric Distribution as:

$$p_x(k) = P(\{X = k\}) = [1-p]^{[k-1]} p \mid [k = 1, 2, \dots].$$

Explanation of the formula:

0) $[k]$: The number of "repetitions" of the experiment, until a "success" occurs.

1) $[1-p]^{[k-1]}$: The probability of a "failure," repeated $[k-1]$ times.

2) $[p]$: The probability of a "success," repeated 1 time, after the execution of $[k-1]$ "failures."