

## 2.2 Probability Mass Functions

### Definition

The **probability mass function (pmf)** of a discrete random variable  $X$  is a function  $p_X : \mathbb{R} \rightarrow [0, 1]$  defined by:

$$p_X(x) = P(X = x), \quad x \in R_X,$$

where  $R_X = \text{range}(X)$  is the set of values the random variable can take.

The pmf of a discrete random variable defines the **probability distribution** of a random variable.

- We frequently assign names to common probability distributions.

## Range of a random variable

The range of a discrete random variable  $X$  is defined as

$$R_X = \{x \in \mathbb{R} : P(X = x) > 0\}.$$

When  $X$  is discrete, then  $R_X$  contains a countable number of elements.

## Range versus support

The **support** of a discrete random variable is sometimes used synonymously with the range of  $X$  ([Casella and Berger 2002](#); [Blitzstein and Hwang 2019](#); [Reich and Ghosh 2019](#)).

Other resources define the support of  $X$ ,  $S_X$ , as the smallest closed set such that

$$\sum_{x \in S_X} p_X(x) = 1.$$

Under this second definition, the range and support will be the same for discrete random variables but can differ for continuous random variables.

## Notation

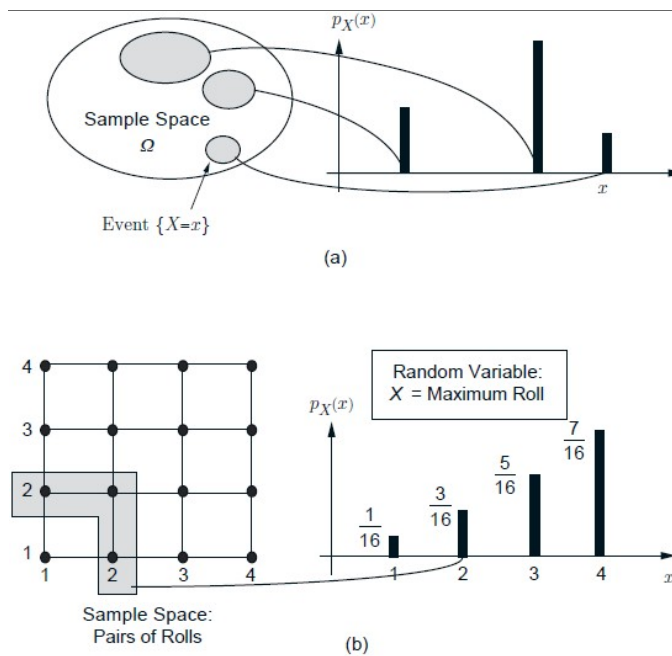
- Uppercase  $X$  will denote a random variable.
- Lowercase  $x$  will denote a specific numeric value.

## Calculation of a discrete pmf

For each possible value  $x$  of  $X$ :

1. Collect all the possible outcomes that give rise to the event  $\{X = x\}$ .
2. Add their probabilities to obtain  $p_X(x)$ .

## Visualization of pmf calculation



## Calculating a pmf

### Simple example

Suppose we toss a fair coin twice in a row and count the number of heads.

$$p_X(x) = \begin{cases} & \text{if } x = 0 \\ & \text{if } x = 1 \\ & \text{if } x = 2. \end{cases}$$

## Properties of pmfs

The pmf of a random variable  $X$ ,  $p_X(x)$ , satisfies the following properties:

- $p_X(x) \geq 0$  for all  $x$
- $\sum_{x \in R_X} p_X(x) = 1$ , where  $R_X$  is the range of  $X$ .
- If  $A \subset R_X$ , then  $P(A) = \sum_{x \in A} p_X(x)$ .

E.g., continuing our simple coin flipping example,

$$P(X > 0) = p_X(1) + p_X(2) = \frac{1}{2} + \frac{1}{4} = \frac{3}{4}.$$

## The Bernoulli random variable

A Bernoulli random variable models a single trial that has only two possible outcomes: success or failure.

- A success occurs with probability  $p \in [0, 1]$  and is assigned the value 1.
- A failure occurs with probability  $1 - p$  and is assigned the value 0.

## The Bernoulli random variable defined

Let  $X$  be a Bernoulli random variable such that the probability of a success is  $p \in [0, 1]$ .

- Then  $X \sim \text{Bernoulli}(p)$  (read as  $X$  is distributed as Bernoulli  $p$ ).

The pmf of  $X$  is:

$$p_X(x) = \begin{cases} p & \text{if } x = 1 \\ 1-p & \text{if } x = 0 \\ 0 & \text{otherwise.} \end{cases}$$

$R_X :$

## Bernoulli examples

- Website clicks: A user either clicks on an ad (success) or doesn't (failure).
- Medical diagnosis: A test either detects a disease (positive result) or not (negative result).
- Email spam filter: An income email is marked as spam (success) or not (failure).
- Sensor activation: A motion sensor either detects movement (success) or doesn't (failure) in a given time frame.

## Independent and identically distributed

We often assume that random variables are **independent and identically distributed** (i.i.d.).

i.i.d. random variables are defined by two characteristics:

1. All of the random variables are independent.
  - Conceptually, the value of one random variable has no impact on the others (and so on).
2. Each of the random variables has the exact same probability distribution.

## The Binomial random variable

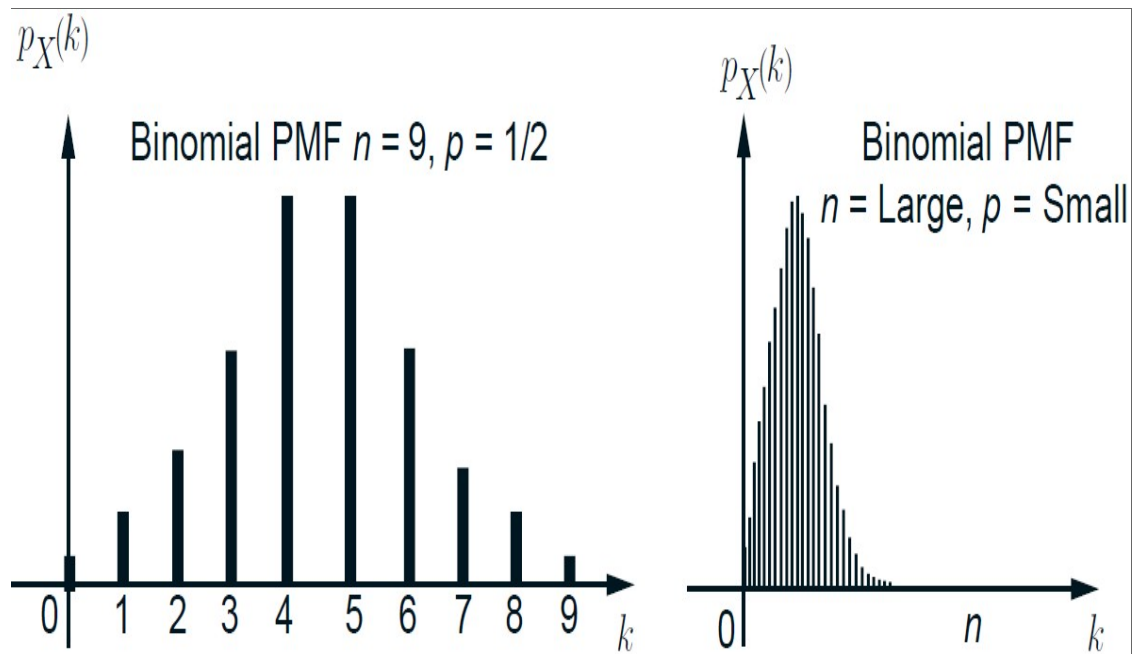
A Binomial random variable models the number of successes in a sequence of  $n$  i.i.d. Bernoulli( $p$ ) trials.

If  $X \sim \text{Binomial}(n, p)$ , then the pmf is

$$p_X(k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad \text{for } k = 0, 1, 2, \dots, n.$$

$R_X$  :

## Visualizing a Binomial pmf



The pmf of a Binomial random variable

### Binomial examples

Product defects: counting the number of defective items in a batch of manufactured goods.

Email campaigns: counting how many recipients open an email out of a fixed number sent.

Vaccination effectiveness: out of 100 vaccinated individuals, how many avoid infection?

Memory tests: how many words a subject recalls correctly out of a list.



## Computing a binomial probability

Only 5% of a lot of 20 electrical fuses are supposed to be defective. What is the probability that at least four defective fuses are observed?

## The Geometric random variable

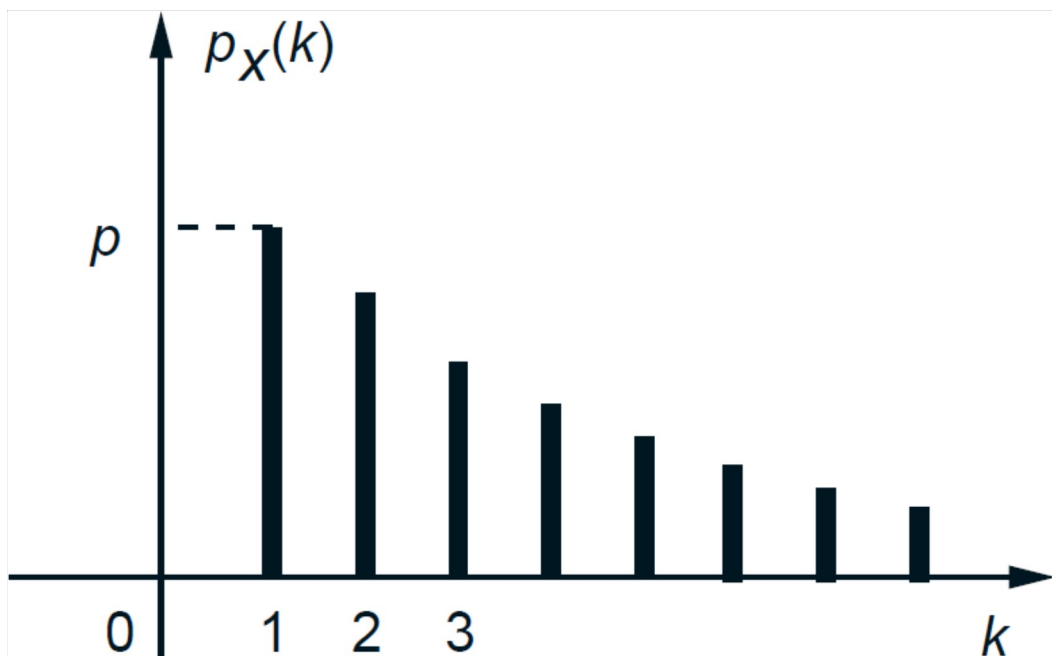
A Geometric random variable counts the number of i.i.d. Bernoulli( $p$ ) trials until the first success.

If  $X \sim \text{Geometric}(p)$ , then the pmf is

$$p_X(x) = (1 - p)^{x-1}p, \quad x = 1, 2, 3, \dots$$

$R_X$  :

## Visualizing a Geometric pmf



The pmf of a Geometric random variable

## Geometric examples

How many calls must a call center agent make until they reach a customer who answers?

How many items must we inspect until we find a defective one?

How many free throws must a person take until they make one?

## Geometric computation

A call center agent must keep making calls until they reach a customer who answers. The probability of a customer answering is 0.10. What is the probability that the agent needs to make at least 20 calls?

## The Poisson distribution

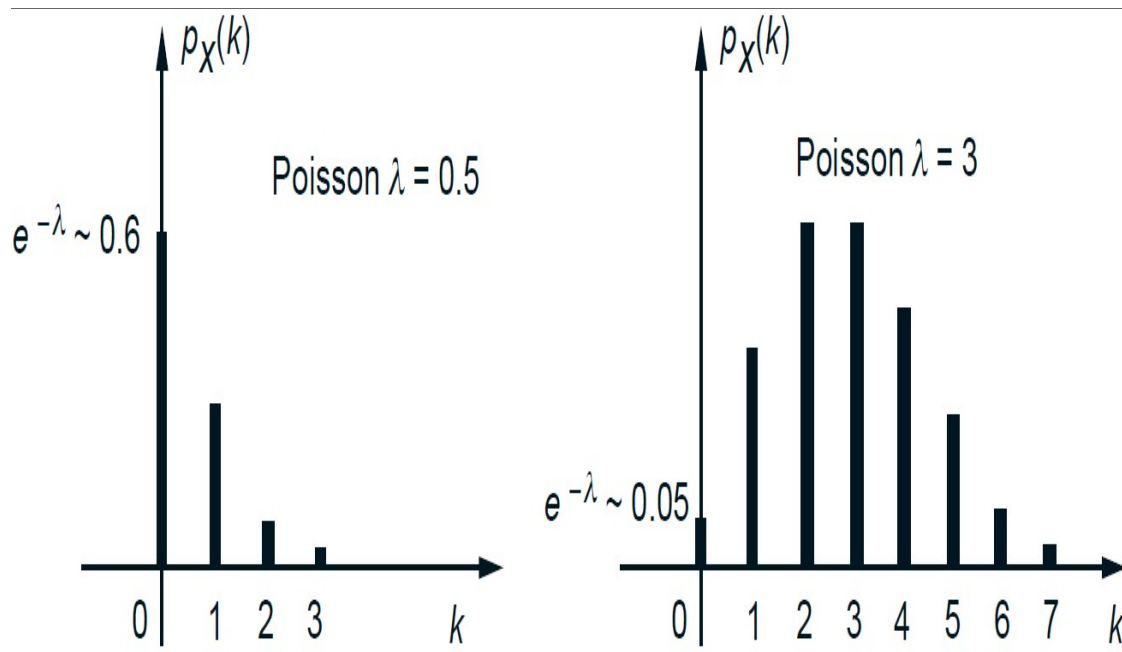
A **Poisson random variable** models the number of events occurring in a fixed interval of time or space, assuming events occur independently and at a constant average rate,  $\lambda > 0$ .

If  $X \sim \text{Poisson}(\lambda)$ , then the pmf is

$$p_X(x) = \frac{\lambda^x e^{-\lambda}}{x!}, \quad x = 0, 1, 2, \dots$$

$R_X :$

## Visualizing a Poisson pmf



The pmf of a Poisson random variable

## Poisson examples

Emails received: the number of emails received per hour, with an average rate of  $\lambda = 5$ .

Website hits: the number of website in a minute if the average is 50 hits per minute.

Accidental deaths by horse kick: Ladislaus Bortkiewicz analyzed the number of soldiers in the Prussian army killed by horse kicks. The average was 0.61 deaths per corps per year, and the distribution of deaths followed a Poisson pattern.

## Poisson computation

A small bakery receives an average of 4 online orders per hour. They can fulfill 8 online orders in an hour, if necessary. What is the probability that they receive more than 8 orders in an hour?

## The Negative Binomial distribution

The Negative Binomial distribution models the total number of trials needed to achieve  $r$  successes in a sequence of i.i.d. Bernoulli( $p$ ) trials.

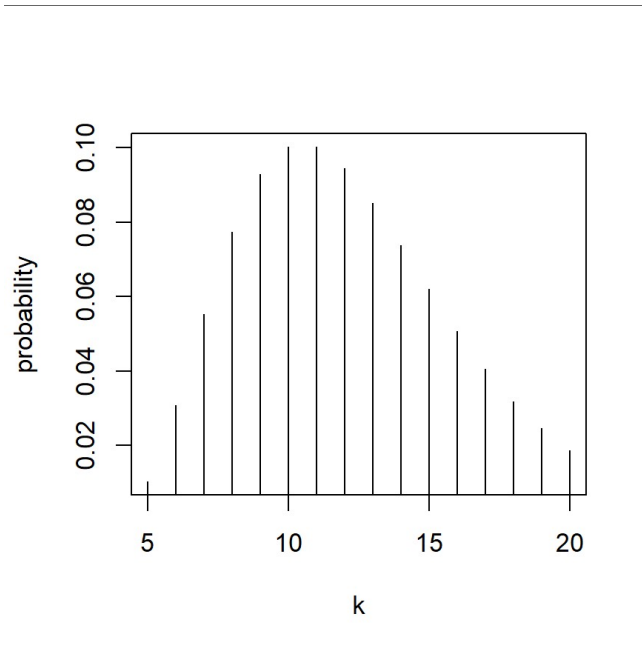
If  $X \sim \text{NB}(r, p)$ , then the pmf is

$$p_X(k) = \binom{k-1}{r-1} p^r (1-p)^{k-r}, \quad x = r, r+1, \dots$$

$R_X$  :

Note: there are multiple characterizations, e.g., the number failures until the  $r$ th success.

## Visualizing a Negative Binomial pmf



The pmf of a Negative Binomial random variable

### Negative Binomial examples

Modeling disease outbreaks: model the number of cases until a certain number of recoveries or deaths occur.

Click-through rates: used in online advertising to model the number of ad impressions until a certain number of clicks occur.

Failure analysis: model the number of trials until a component fails a set number of times.

## Negative Binomial computation

Suppose a basketball player has a 60% chance of making a free throw. What is the probability that it takes at least 7 shots to make 3 successful free throws?

## References

- Blitzstein, Joseph K, and Jessica Hwang. 2019. *Introduction to Probability, 2nd Edition*. Chapman; Hall/CRC.
- Casella, George, and Roger Berger. 2002. *Statistical Inference, 2nd Edition*. Chapman; Hall/CRC.
- Reich, Brian J, and Sujit K Ghosh. 2019. *Bayesian Statistical Methods*. Chapman; Hall/CRC.