

# Biostatistics Week V

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**ACIBADEM**  
MEHMET ALİ AYDINLAR  
ÜNİVERSİTESİ

# Random Variable

- A random variable (RV) is a variable whose possible values are **numerical outcomes of a random phenomenon**
- There are two types of random variables:
  - ***Discrete*** – flipping a coin, rolling a die, number of pancreatic cancer cases in a year ...
  - ***Continuous*** – systolic blood pressures of hypertensive patients, progression-free survival time of glioblastoma patients, expression level of a certain gene ...

RV

Discrete



Continuous



# Probability Mass Function (PMF)

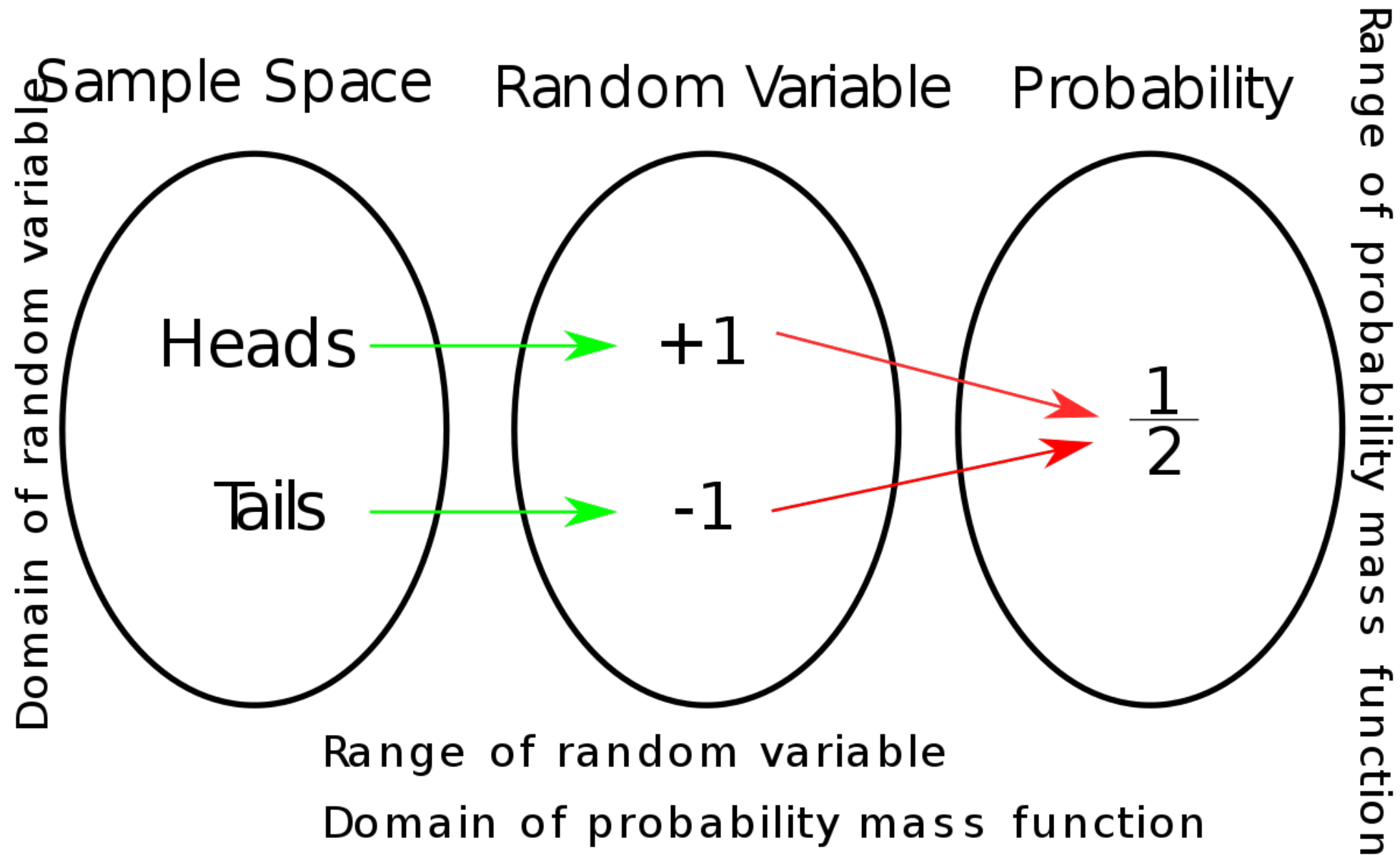
- Probability mass function is the probability distribution of a discrete random variable
- It provides the possible values and their associated probabilities

$$p: \mathbb{Z} \rightarrow [0,1]$$

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

# Properties of a Proper PMF ( $p_X$ )

1.  $p_X(x)$  is defined for all  $x$  over the given domain
2.  $0 \leq p_X(x) \leq 1$
3.  $\sum_x p_X(x) = 1$



# Probability Density Function (PDF)

- Probability density function is the probability distribution of a continuous random variable
- It provides the possible values and their associated probabilities

$$f: \mathbb{R} \rightarrow [0,1]$$

$$f_X(x) = P(X = x)$$

# Properties of a Proper PDF ( $f_X$ )

1.  $f_X$  is continuous over the given range
2.  $0 \leq f_X(x) \leq 1$
3.  $\int_{-\infty}^{\infty} f_X(x) dx = 1$



Cumulative Density Function (CDF)

$$F_X(x) = P(X \leq x)$$

Survival Function

$$S(x) = P(X > x)$$

# Expected Value

- The weighted average of all the possible values of a RV by the associated probabilities
- For discrete RVs:

$$E[X] = \sum_{i=1}^n P(X = x_i) x_i$$

- For continuous RVs:

$$E[X] = \int_{-\infty}^{\infty} f(x) x \, dx$$

# Expected Value

- Expectation can be interpreted as the average outcome value over a large number of repetitions
- Properties:
  - $E[X + c] = E[X] + c$
  - $E[X * c] = E[X] * c$
  - $E[X + Y] = E[X] + E[Y]$
  - $E[X * Y] = E[X] E[Y]$  if X and Y are independent

# Variance

- Expected squared distance of the RV values from the expected value

$$\text{Var}(X) = E[(X - E[X])^2] = E[X^2] - E[X]^2$$

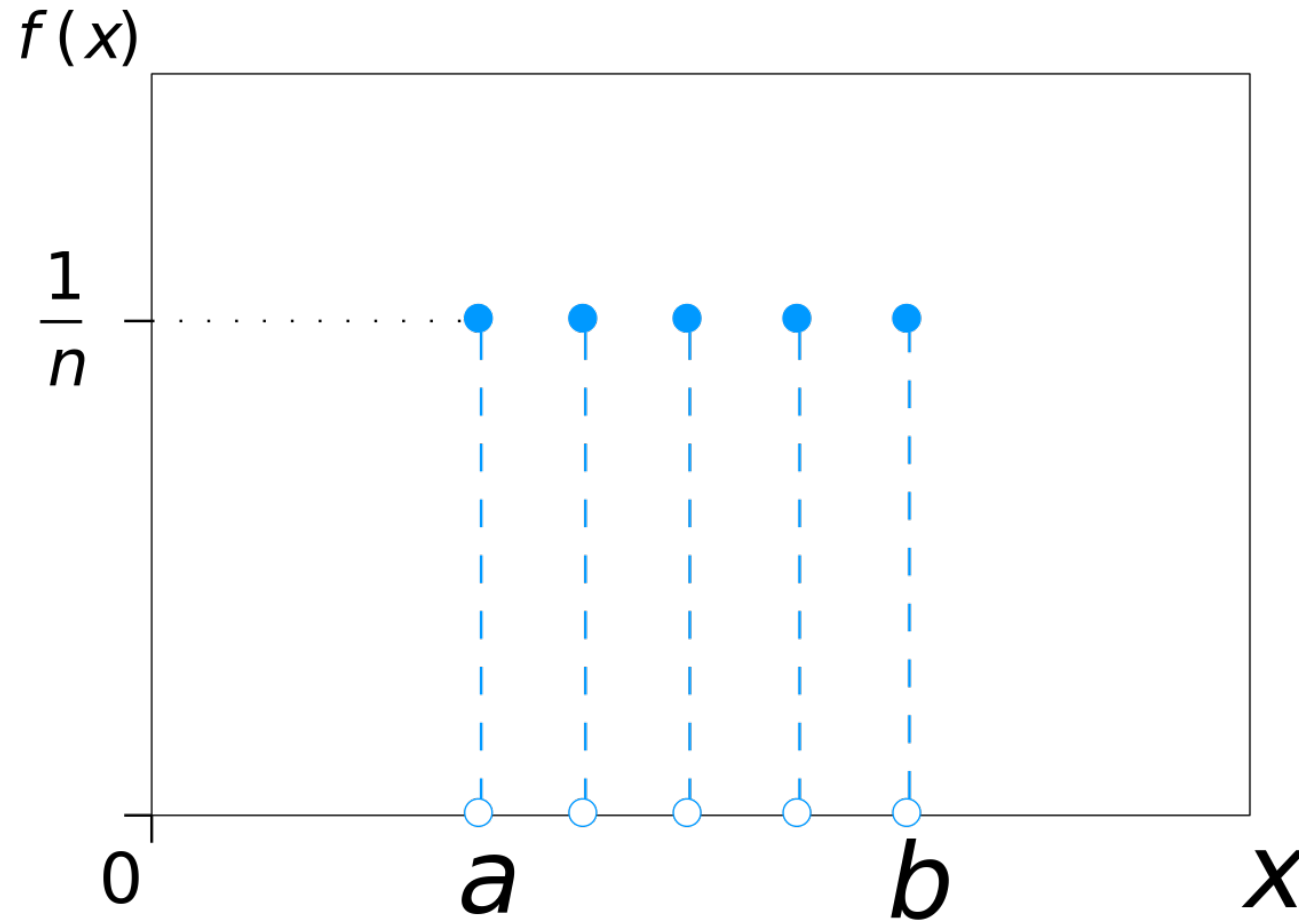
Properties:

- $\text{Var}(X + c) = \text{Var}(X)$
- $\text{Var}(Xc) = \text{Var}(X)c^2$
- $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$  if X and Y are independent.

# Commonly Used Discrete Distributions

- Discrete Uniform Distribution
- Bernoulli Distribution
- Binomial Distribution
- Geometric Distribution
- Hypergeometric Distribution
- Poisson Distribution

# Discrete Uniform Distribution



$$a \leq k \leq b, \quad n = b - a + 1$$

$$\text{PMF} \quad P(X = k) = \frac{1}{n}$$

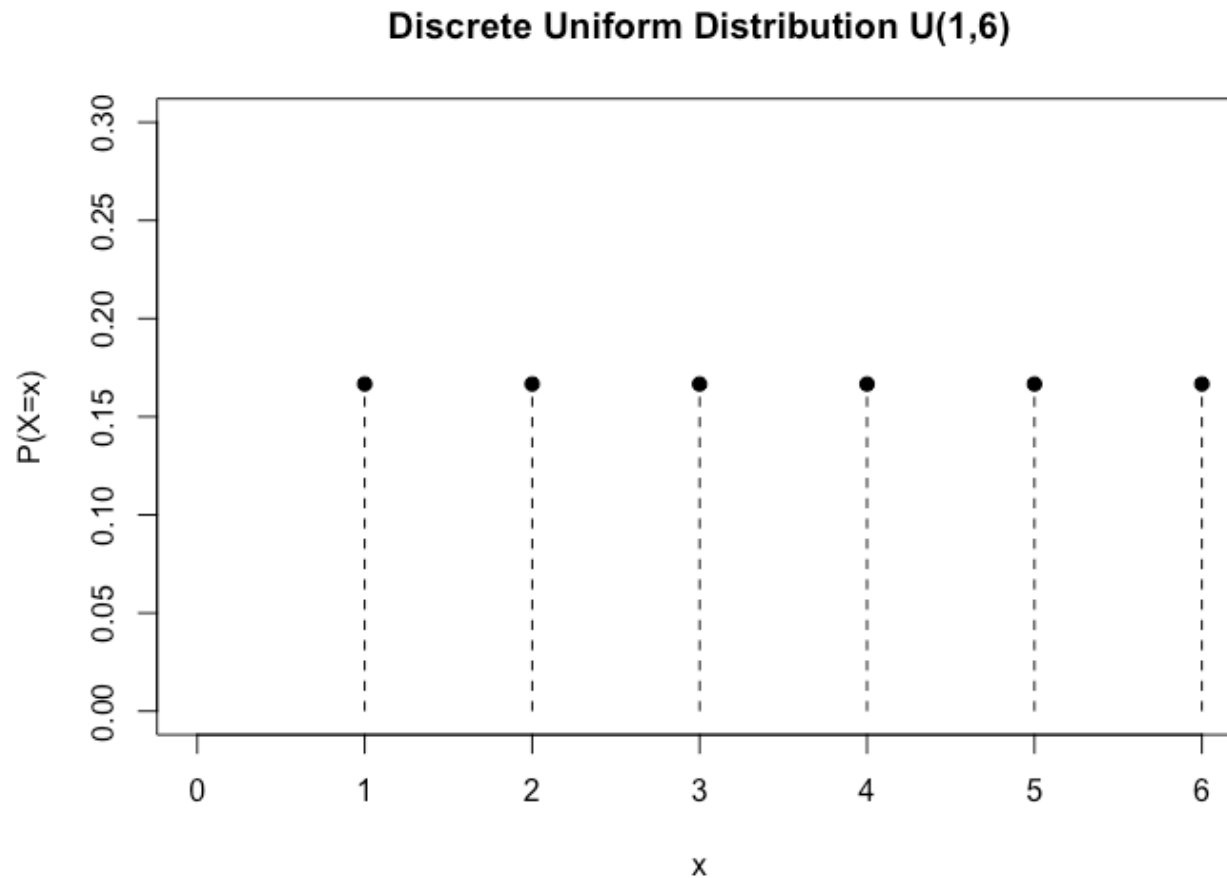
$$\text{E}[X] \quad \frac{a + b}{2}$$

$$\text{Var}(X) \quad \frac{n^2 - 1}{12}$$

$$\text{CDF} \quad P(X \leq k) = \frac{k - a + 1}{n}$$

# Discrete Uniform Distribution

- Rolling a die



$$1 \leq k \leq 6, \quad n = 6$$

**PMF**  $P(X = k) = \frac{1}{6}$

**E[X]**  $\frac{1 + 6}{2} = 3.5$

**Var(X)**  $\frac{6^2 - 1}{12} = \frac{35}{12} \approx 2.92$

# Bernoulli Distribution

Let  $X$  be a random variable with possible values 0 and 1, and let  $P(X = 1) = p$ .

$$pmf = P(X = x) = \begin{cases} p^x(1-p)^{1-x} & x \in 0, 1 \\ 0 & otherwise \end{cases}$$

$$cdf = F_x(x) = P(X \leq x) = p^x(1-p)^{1-x}$$

$$E[X] = p \text{ and } Var(X) = p(1-p)$$

Example: Flipping a fair ( $p = 0.5$ ) coin



# Binomial Distribution

- Used to describe the number of successes in  $n$  binary trials
- $n$ : number of trials
- $p$ : probability of success in one trial

$$X \sim B(n, p)$$

$$\text{PMF } P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}, \quad k = 0, 1, 2, \dots, n$$

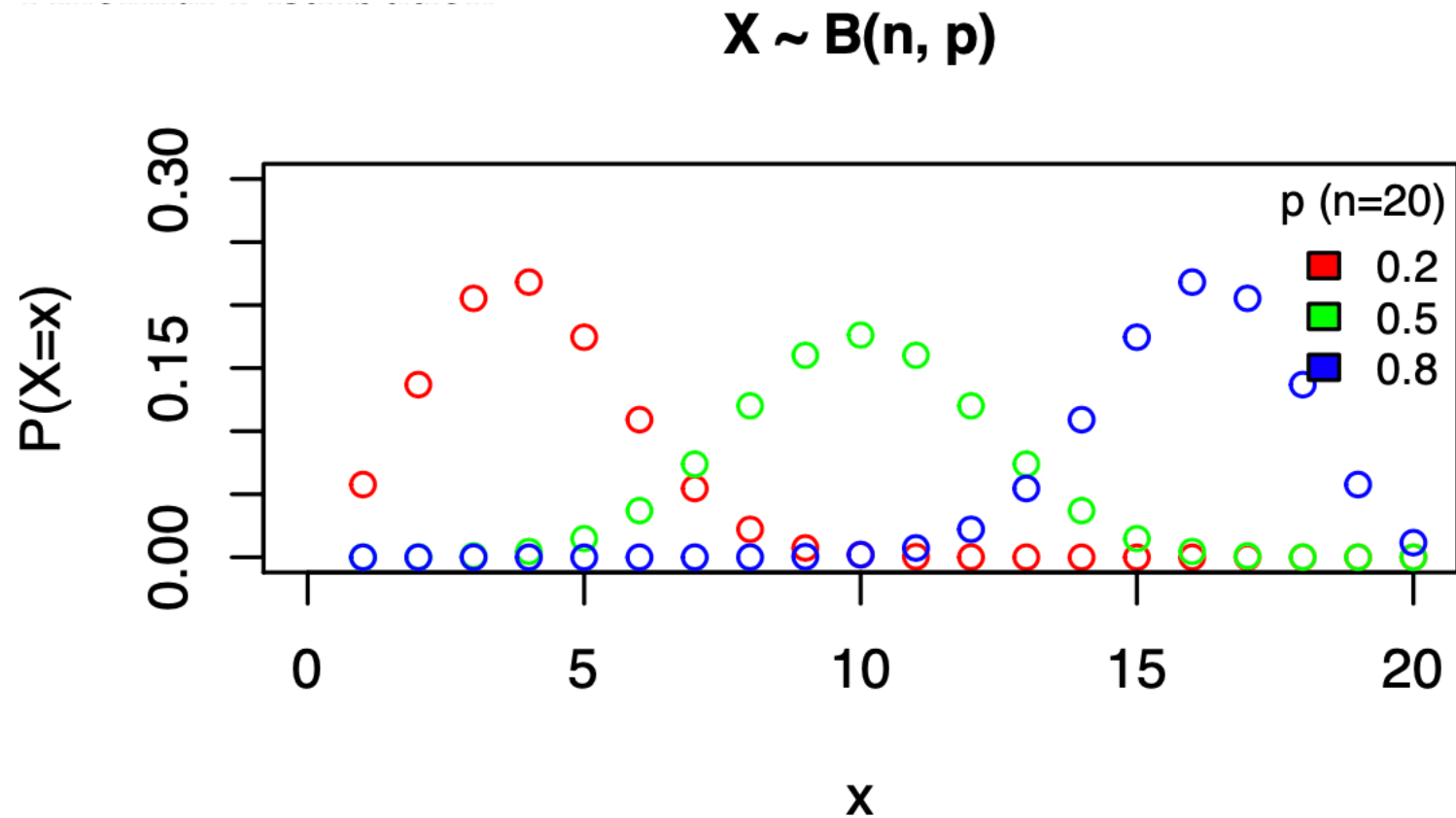
$$E[X] = np$$

$$\text{Var}(X) = np(1-p)$$

$$\text{As } X = \sum_{i=1}^n Y_i \text{ where } Y_i \sim \text{Bernoulli}(p) (\text{iid})$$

# Binomial Distribution

- e.g., flipping a coin 20 times



$$P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}$$

# Binomial Distribution Example

- A novel treatment has a success rate of 80%. Out of 10 patients who underwent the novel treatment:

- a) What is the probability that exactly 6 recovers?
- b) What is the probability that at least 9 recovers?
- c) What is the expected value and variance?

a)  $P(X = 6) = \binom{10}{6} 0.8^6 (1 - 0.8)^{10-6} = 0.88$

b)  $P(X \geq 9) = P(X = 9) + P(X = 10)$   
 $= \binom{10}{9} 0.8^9 (1 - 0.8)^{10-9} + \binom{10}{10} 0.8^{10} (1 - 0.8)^{10-10}$   
 $= 0.2684 + 0.1073 = 0.3758$

c)  $E[X] = np = 10 \times 0.8 = 8$   
 $np(1 - p) = 10 \times 0.8 \times 0.2 = 1.6$

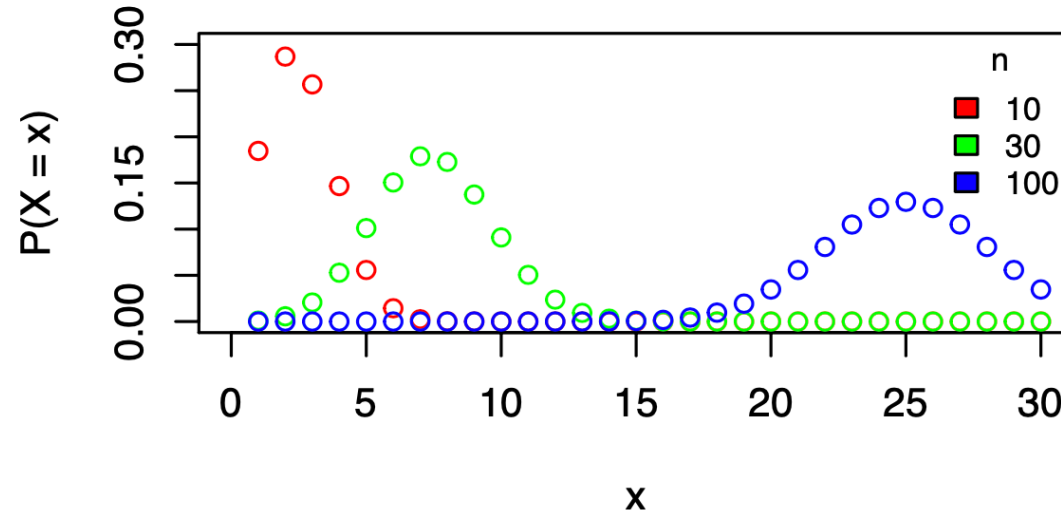
# Hypergeometric Distribution

- Describes the probability of  $k$  successes in  $n$  draws, **without replacement**\*, from a finite population of size  $N$  that contains exactly  $K$  objects with that feature

\*Contrary to the binomial distribution which describes the probability of  $k$  successes in  $n$  draws **with replacement**

# Hypergeometric Distribution

$X \sim \text{Hypergeometric}(200, 50, n)$



$$P(X = x) = \frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}}$$

$E[X] = np$  and  $Var(X) = np(1-p)\frac{N-n}{N-1}$  where  $p = K/N$

Example: Drawing  $n$  balls from an urn that contains 50 white (desired) and 150 red balls (the above plots) and getting  $x$  white balls

# Geometric Distribution

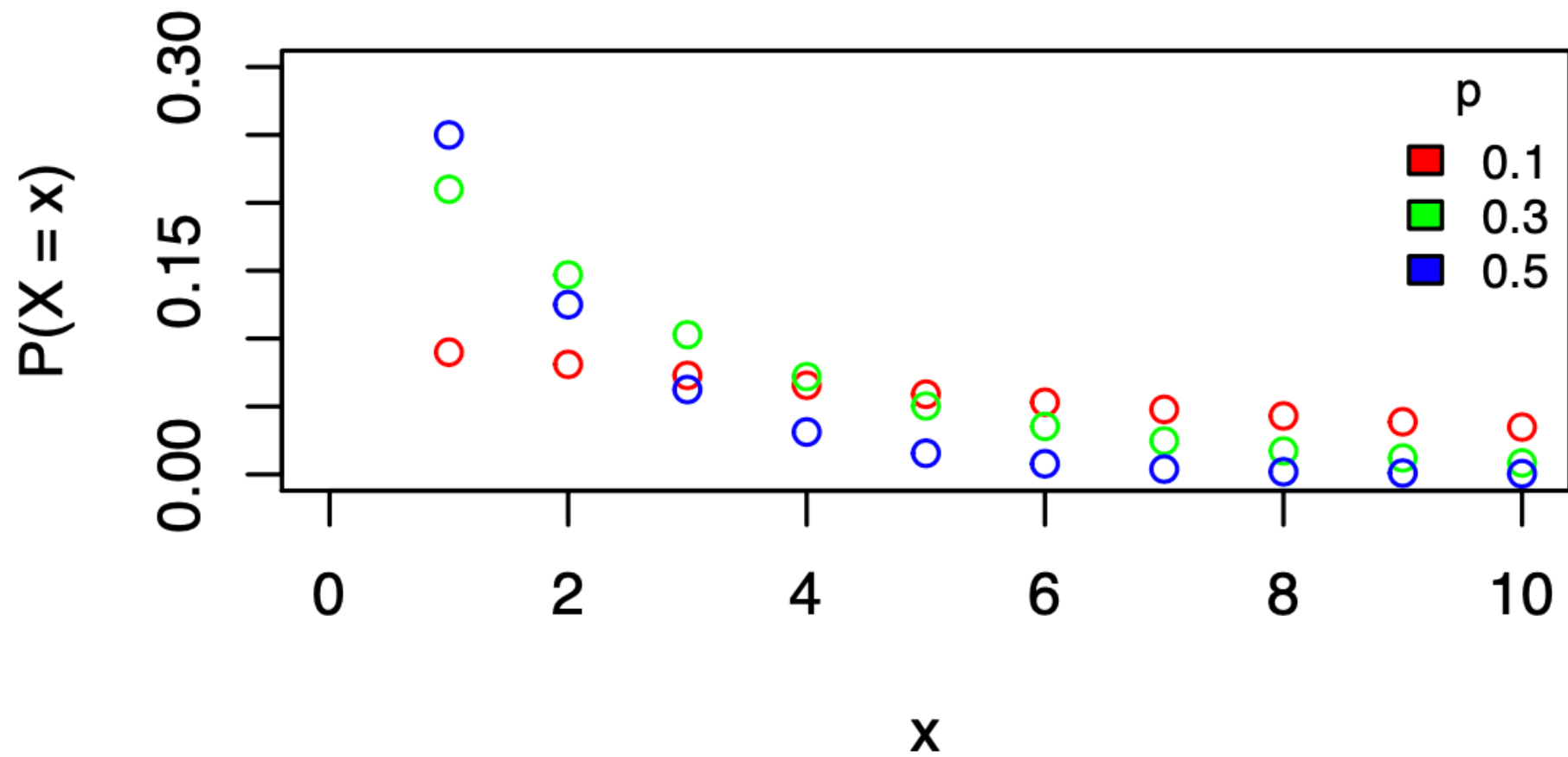
- The probability distribution of the number  $X$  of Bernoulli trials needed to get one success

$$P(X = x) = p(1 - p)^{x-1}$$

$$E[X] = \frac{1}{p}, \text{Var}(X) = \frac{1-p}{p^2}$$

Example: Number of times a coin is flipped before getting heads.

**$X \sim \text{Geometric}(p)$**



# Poisson Distribution

- expresses the probability of a given number of events occurring in a **fixed interval of time or space** if these events occur with a known constant rate and independent of time
- useful to model counts. E.g.,
  - number of rare diseases diagnosed in a certain year
  - number of mutations in a certain region within a chromosome
  - number of births per hour in a certain day

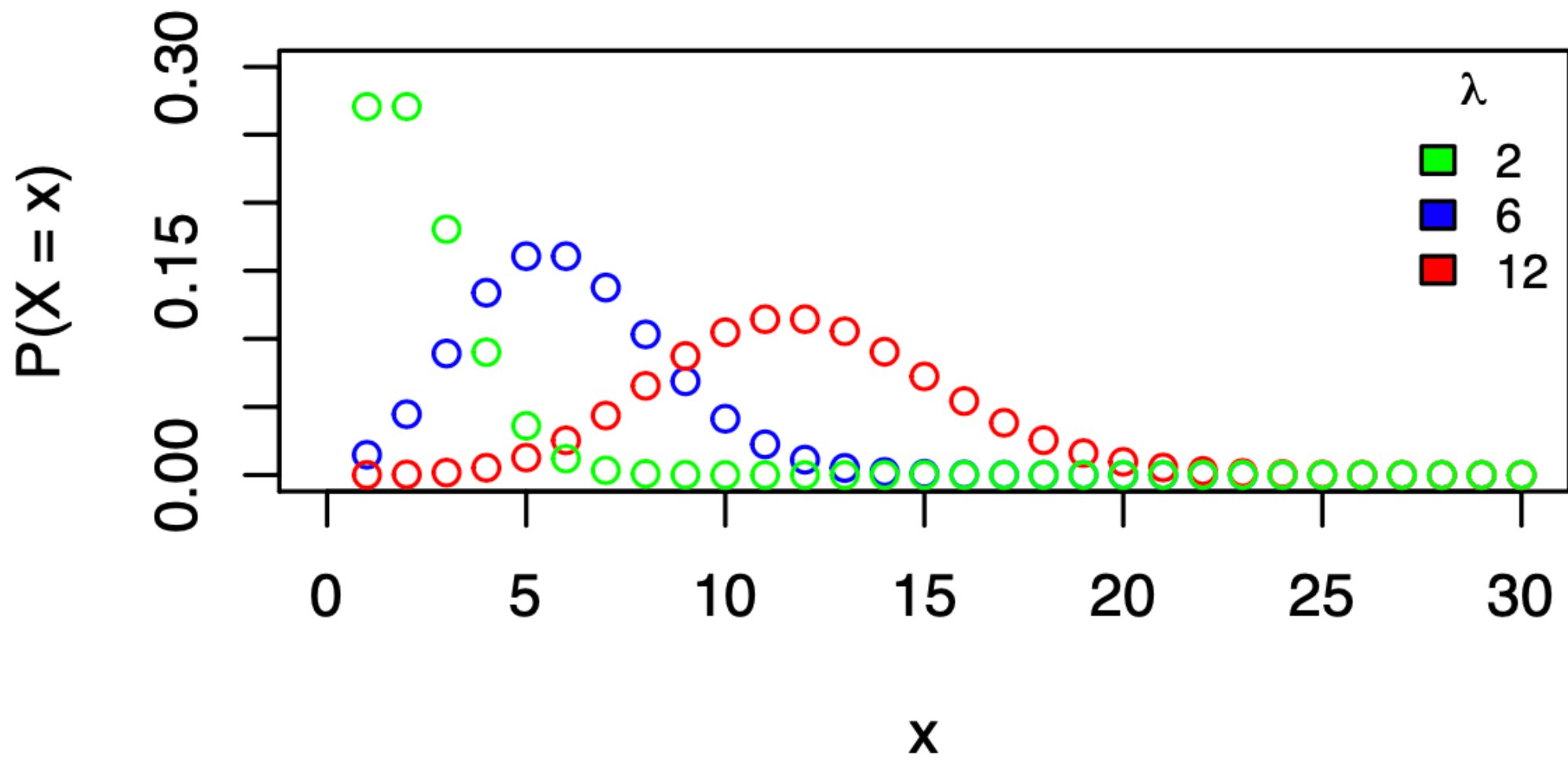
<b>PMF</b>	$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}, x = 0, 1, 2, \dots$
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<b>E[X]</b>	$\lambda$
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<b>Var(X)</b>	$\lambda$
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$$X \sim \text{Pois}(\lambda)$$



$$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}$$

# Poisson Distribution

In a city, the mean number of people dying from a rare disease is 4 in a week.  
In a certain week,

- a) What is the probability that no one dies from the disease?
- b) What is the probability that at least 2 people die from the disease?

$$\text{a) } P(X = 0) = \frac{e^{-4} 4^0}{0!} \approx 0.0183$$

$$\begin{aligned} \text{b) } P(X \geq 2) &= 1 - P(X < 2) = 1 - (P(X = 0) + P(X = 1)) \\ &= 1 - \left( \frac{e^{-4} 4^0}{0!} + \frac{e^{-4} 4^1}{1!} \right) \\ &= 1 - (0.0183 + 0.0733) = 0.9084 \end{aligned}$$

# Poisson Distribution

- As  $n$  gets larger, and  $p$  gets smaller, binomial distribution approximates to Poisson distribution

# Brief Summary

- A RV is a variable whose possible values are numerical outcomes of a random phenomenon
- RV can either be discrete or continuous
- Commonly used discrete distributions include:
  - Discrete Uniform Distribution
  - Bernoulli Distribution
  - Binomial Distribution
  - Hypergeometric Distribution
  - Geometric Distribution
  - Poisson Distribution