# Biostatistics Week IV

Ege Ülgen, M.D.

28 October 2021

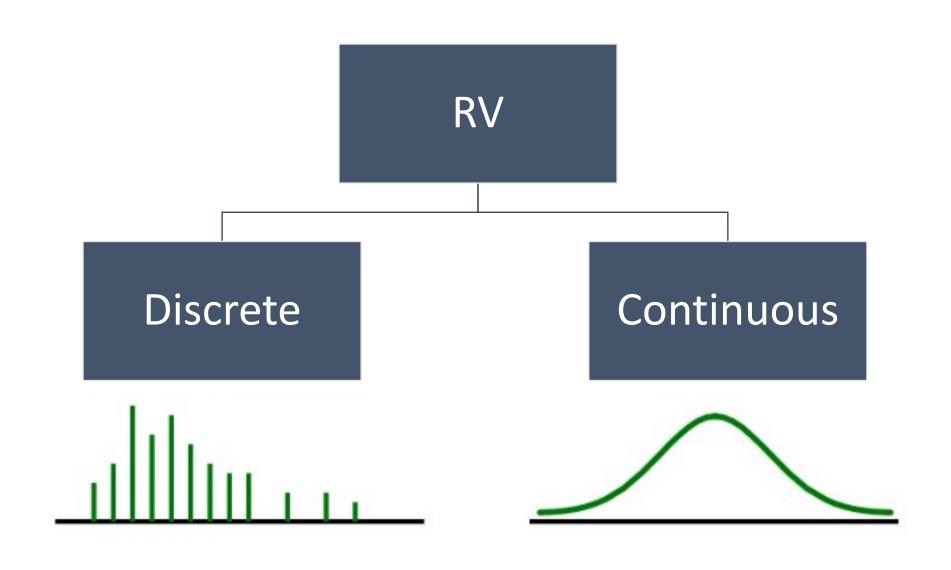


### Random Variable

 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

...



## Probability Mass Function (PMF)

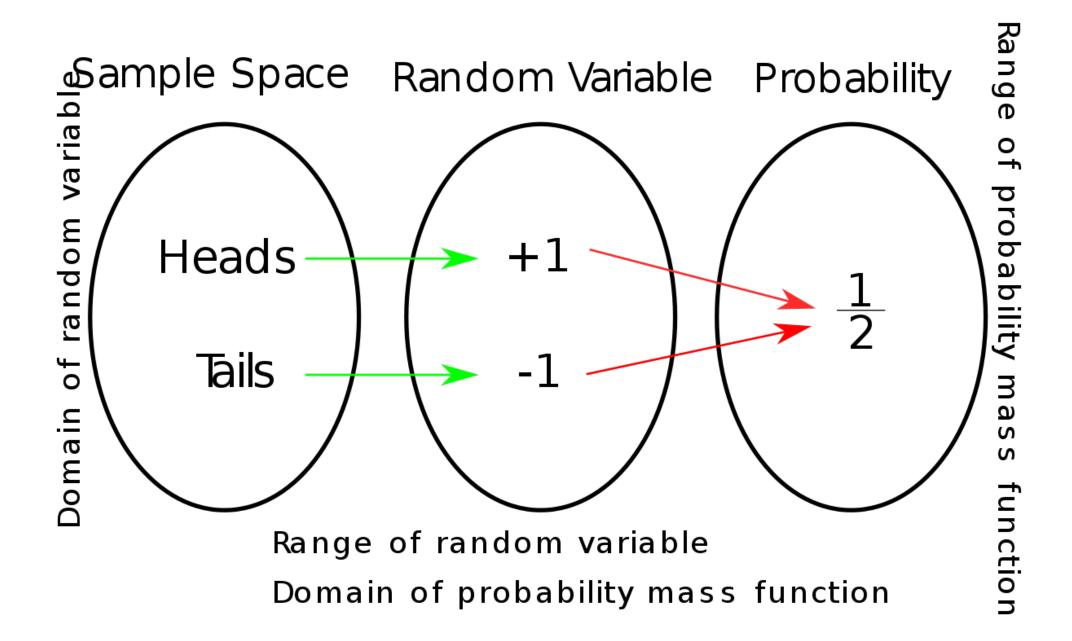
- Probability mass function is the probability distribution of a discrete random variable
- It provides the possible values and their associated probabilities

$$p: \mathbb{R} \to [0,1]$$

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

## Properties of a Proper PMF $(p_X)$

- 1.  $p_X(x)$  is defined for all x over the given domain
- 2.  $0 \le p_X(x) \le 1$
- 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

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

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

## Properties of a Proper PDF $(f_X)$

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

## Cumulative Density Function (CDF)

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

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

$$Var(X) = E[(X - E[X])^2] = E[X^2] - E[X]^2$$

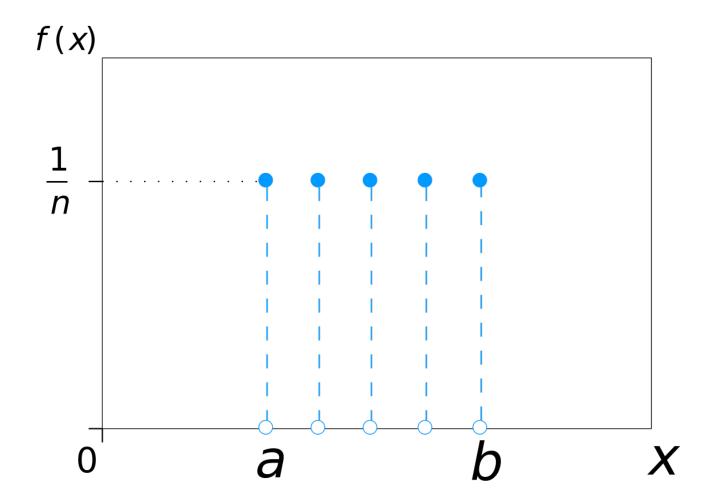
Properties:

- Var(X+c) = Var(X)
- $Var(Xc) = Var(X)c^2$
- Var(X + Y) = Var(X) + 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



$$a \le k \le b$$
,  $n = b - a + 1$ 

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

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

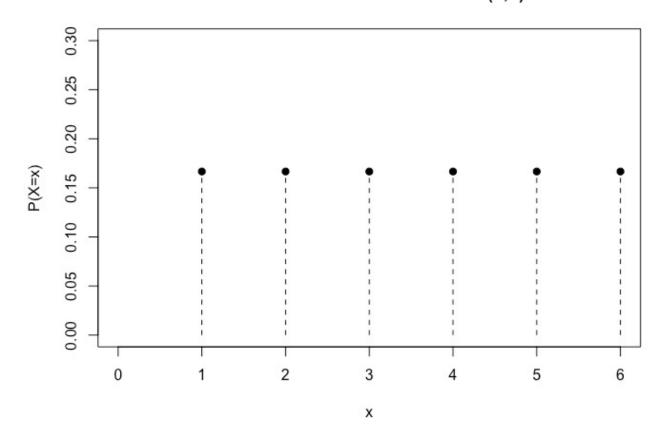
Var(X)  $\frac{n^2-1}{12}$ 

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

### Discrete Uniform Distribution

### Rolling a die

#### Discrete Uniform Distribution U(1,6)



$$1 \le k \le 6$$
,  $n = 6$ 

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.

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

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

$$E[X] = p$$
 and  $Var(X) = p(1 - p)$ 

Example: Flipping a fair (p = 0.5) coin

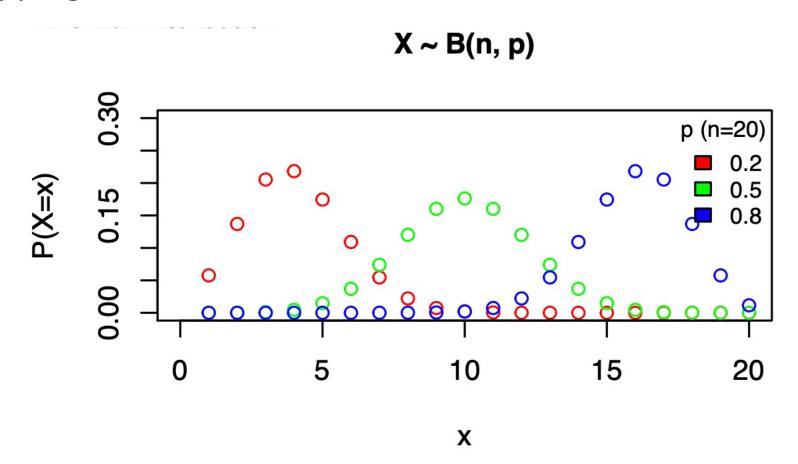
### Binomial Distribution

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

$$X^{\sim}B(n,p)$$
 PMF  $P(X=k)=\binom{n}{k}p^k(1-p)^{n-k},\ k=0,1,2,...,n$  E[X]  $np$  As  $X=\sum_{i=1}^n Y_i\ where\ Y_i\sim Bernoulli(p)(iid)$  Var(X)  $np(1-p)$ 

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

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

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

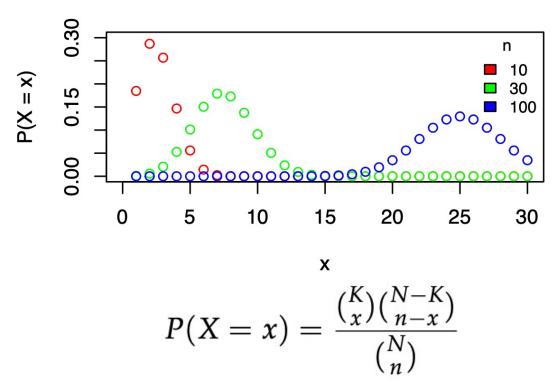
## Hypergeometric Distribution

 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 ~ Hypergeometric(200, 50, n)



E[X] = np and  $Var(X) = np(1-p)\binom{N-n}{N-1}$  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

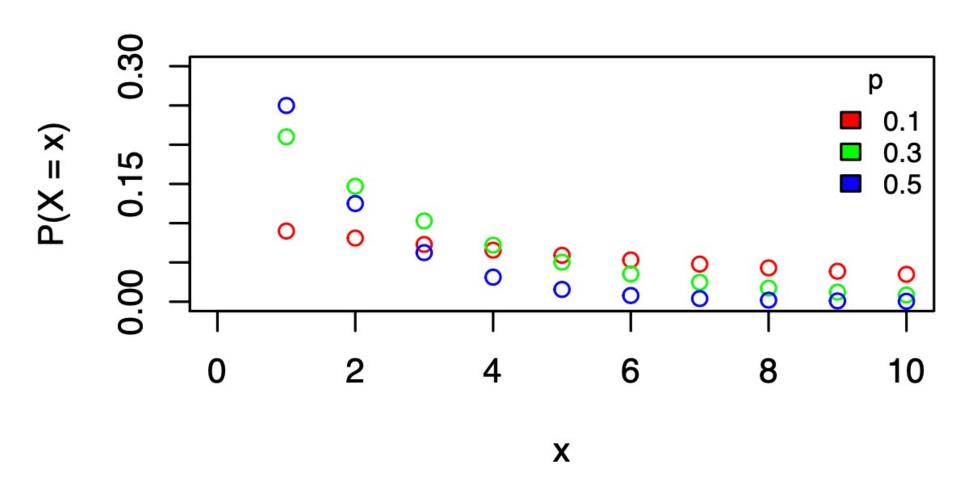
 The probability distribution of the number X of Bernoulli trials needed to get one success

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

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

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

### X ~ Geometric(p)



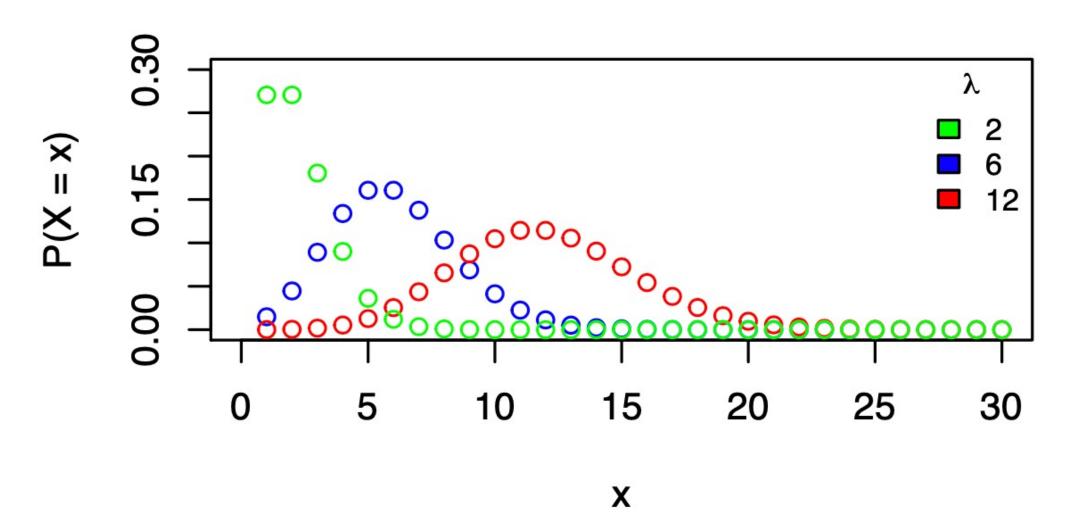
### Poisson Distribution

- 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

PMF 
$$P(X = x) = \frac{e^{-\lambda} \lambda^{x}}{x!}, x = 0, 1, 2, ...$$
E[X]  $\lambda$ 

$$Var(X)$$
  $\lambda$ 

### $X \sim Pois(\lambda)$



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

### 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?

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

b) 
$$P(X \ge 2) = 1 - P(X < 2) = 1 - (P(X = 0) + P(X = 1))$$
  
 $= 1 - (\frac{e^{-4}4^0}{0!} + \frac{e^{-4}4^1}{1!})$   
 $= 1 - (0.0183 + 0.0733) = 0.9084$ 

## 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