

CHAPTER 3: DISTRIBUTIONS OF RANDOM VARIABLES (PART 1)

Day 6 topics:

Section 3.1: Random variables

Section 3.2: Binomial distribution

3.1. Random variables.

Definition 3.1. A **random variable (r.v.)** assigns numerical values to the outcome of a random phenomenon.

Notation:

A random variable is usually denoted with a capital letter such as X , Y , or Z .

Example 3.2. Data points

Suppose you have a dataset of size 3 ($n = 3$) with $x_1 = 5$, $x_2 = 3$, and $x_3 = 6$.

- Each data point is the outcome of a random phenomenon
- Each data point is a numerical value
- The data points are examples of values of a sequence of random variables X_1 , X_2 , and X_3
- For datasets, we almost always assume the data points came from random variables that are independent and have the same **distribution**.
- To calculate the likelihood of data, we need to know the distribution of the random variable that models the data.
- First, let's remind ourselves how to calculate the mean and variance of a dataset:
 - What is the mean of the data points?

$$\bar{x} = 14/3 = 4.66667$$

- What are the variance and standard deviation of the data points?

$$s^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n - 1}$$

$$s^2 = 2.3333$$

$$s = 1.527525 \sim 1.53$$

Example 3.3. Rolling a die

Suppose you roll a fair die. Let the random variable (r.v.) X be the outcome of the roll, i.e. the value of the face showing on the die.

(1) What is the probability distribution of the r.v. X ?

(2) What is the expected outcome of the r.v. X ?

•

$$7/2 = 3.5$$

- *Not a possible outcome!*
- *Do not round.*

(3) Now suppose the 6-sided die is not fair. How would we calculate the expected outcome?

x	$\mathbb{P}(X = x)$	$x\mathbb{P}(X = x)$
1	0.10	0.1
2	0.20	0.4
3	0.05	0.15
4	0.05	0.2
5	0.25	1.25
6	0.35	2.10
sum	1	$\mu = 4.2$

(From Textbook § 2.1.5)

Definition 3.4. A **probability distribution** consists of all disjoint outcomes and their associated probabilities.

Rules for a probability distribution

A probability distribution is a list of all possible outcomes and their associated probabilities that satisfies three rules:

- (1) The outcomes listed must be disjoint.
- (2) Each probability must be between 0 and 1.
- (3) The probabilities must total to 1.

Probability distributions are usually either **discrete** or **continuous**, depending on whether the random variable is discrete or continuous.

(Back to Textbook § 3.1.1)

Definition 3.5. A **discrete** r.v. X takes on a finite number of values or countably infinite number of possible values.

Definition 3.6. A **continuous** r.v. X can take on any real value in an interval of values or unions of intervals.

Class notes:

Sketch histogram of discrete values 1, 2, ..., 6 with continuous approximation
 X could be values from a die or a waiting time in minutes

§ 3.1.2 Expectation

- We call the mean of a random variable its **expected value**
- The expected value is calculated as a weighted average

Definition 3.7. Expected value of a discrete random variable

If X takes on outcomes x_1, \dots, x_k with probabilities $P(X = x_1), \dots, P(X = x_k)$, the expected value of X is the sum of each outcome multiplied by its corresponding probability:

$$\begin{aligned}\mu = E(X) &= x_1P(X = x_1) + \dots + x_kP(X = x_k) \\ &= \sum_{i=1}^k x_iP(X = x_i).\end{aligned}$$

§ 3.1.3 Variability of random variables

- Just like with data, the variability of a r.v. is described with its variance or standard deviation.
- Squared deviations from the mean are weighted by their respective probabilities

Definition 3.8. Variance of a discrete random variable

If X takes on outcomes x_1, \dots, x_k with probabilities $P(X = x_1), \dots, P(X = x_k)$ and expected value $\mu = E(X)$, then the variance of X , denoted by $\text{Var}(X)$ or σ^2 , is

$$\begin{aligned}\text{Var}(X) &= (x_1 - \mu)^2P(X = x_1) + \dots + (x_k - \mu)^2P(X = x_k) \\ &= \sum_{i=1}^k (x_i - \mu)^2P(X = x_i).\end{aligned}$$

Definition 3.9. Standard deviation of a discrete random variable

The standard deviation of X , labeled $SD(X)$ or σ , is

$$SD(X) = \sigma = \sqrt{\sigma^2}$$

Example 3.10. Rolling a fair die: variance

Suppose you roll a fair 6-sided die. Let the random variable (r.v.) X be the outcome of the roll, i.e. the value of the face showing on the die.

Find the variance and standard deviation of X .

$$\mu = 3.5$$

x	$\mathbb{P}(X = x)$	$x - \mu$	$(x - \mu)^2$	$\mathbb{P}(X = x)(x - \mu)^2$	R
1	1/6	-2.5	6.25	1.04166667	
2	1/6	-1.5	2.25	0.3750	
3	1/6	-0.5	0.25	0.04166667	
4	1/6	0.5	0.25	0.04166667	
5	1/6	1.5	2.25	0.3750	
6	1/6	2.5	6.25	1.04166667	
<i>sum</i>	1			$\sigma^2 = 2.916667$	

$$\sigma = 1.707825$$

R commands

x	$\mathbb{P}(X = x)$	$x - \mu$	$(x - \mu)^2$	$\mathbb{P}(X = x)(x - \mu)^2$	R
$x < -1 : 6$	$px < -1/6$	$x - mu$	$(x - mu)^2$	$px * (x - mu)^2$	
<i>sum</i>	1			$sum(px * (x - mu)^2)$	

Example 3.11. Vaccinated people testing positive for Covid-19

About 25% of people that test positive for Covid-19 are vaccinated for Covid-19.

Define the r.v. X to be 1 if someone that tests positive is vaccinated and 0 if they are not vaccinated.

(1) Make a table for the probability distribution for the r.v. X

(2) What is the expected value of X ?

$$\mathbb{E}[X] = p = 0.25$$

(3) What is the variance of X ?

$$\text{Var}[X] = pq = 0.25 \cdot 0.75 = 0.1875$$

$$SD[X] = 0.4330127$$

§ 3.1.4 Linear combinations of random variables

Seems abstract, but actually fundamental to basic statistical theory - using it all the time!

Definition 3.12. *Linear combinations of random variables.*

If X and Y are random variables and a and b are constants, then

$$aX + bY$$

is a linear combination of the random variables.

Theorem 3.13. *Expected value of a linear combination of random variables.*

If X and Y are random variables and a and b are constants, then

$$\mathbb{E}(aX + bY) = a\mathbb{E}(X) + b\mathbb{E}(Y) = a\mu_X + b\mu_Y$$

Example 3.14. *Expected money for rolling 3 dice*

Let the random variables X_1, X_2, X_3 be the values shown on 3 fair 6-sided dice rolls. Suppose you are given in dollars the amount of the first roll, plus twice the value of the second roll, plus 4 times the value of the third roll. How much money do you expect to get?

$$\mathbb{E}(X_1 + 2X_2 + 4X_3) = \mathbb{E}(X_1) + 2\mathbb{E}(X_2) + 4\mathbb{E}(X_3) = 7 \cdot 3.5 = 24.5$$

Expected value of the sample mean

Theorem 3.15. Variance of a linear combination of random variables.

If X and Y are INDEPENDENT random variables and a and b are constants, then

$$\text{Var}(aX + bY) = a^2 \text{Var}(X) + b^2 \text{Var}(Y)$$

Example 3.16. Variance of money for rolling 3 dice

Let the random variables X_1, X_2, X_3 be the values shown on 3 fair 6-sided dice rolls. Suppose you are given in dollars the amount of the first roll, plus twice the value of the second roll, plus 4 times the value of the third roll.

What are the variance and standard deviation of the amount you get from the 3 rolls?

$$\text{Var}(X_1 + 2X_2 + 4X_3) = \text{Var}(X_1) + 4\text{Var}(X_2) + 16\text{Var}(X_3) = 21 \cdot 17.5 = 367.5$$

$$\text{SD}(X_1 + 2X_2 + 4X_3) = \sqrt{367.5} = 19.17029$$

Variance of the sample mean

Example 3.17. Vaccinated people testing positive for Covid-19 (revisited)

About 25% of people that test positive for Covid-19 are vaccinated for Covid-19.

Define the r.v. X_i to be 1 if someone that tests positive is vaccinated and 0 if they are not vaccinated.

Suppose 3 people have tested positive for Covid-19 (independently of each other).

Let T denote the number of people that are vaccinated amongst the 3 that tested positive.

(1) Using the r.v.'s X_i , write a mathematical equation for calculating T .

$$T = \sum_{i=1}^3 X_i$$

(2) What is the expected value of T ?

$$\mathbb{E}[T] = 3 \cdot p = 3 \cdot 0.25 = 0.75$$

(3) What is the variance of T ?

$$\text{Var}[T] = npq = 3 \cdot 0.25 \cdot 0.75 = 3 \cdot 0.1875 = 0.5625$$

$$SD[T] = 0.5625$$

(4) What is the probability distribution of T ?

x	$\mathbb{P}(X = x)$
0	$0.75^3 = 0.422$
1	$3 \cdot 0.25 \cdot 0.75^2 = 0.422$
2	$3 \cdot 0.25^2 \cdot 0.75 = 0.141$
3	$0.25^3 = 0.016$

3.2. Binomial distribution.

- Many situations involve modeling independent random events that have 2 possible outcomes (binary), such as
 - Repeatedly flipping a coin
 - Whether a person that tested positive with Covid-19 is vaccinated or not
- Repeated events are referred to as **trials**
- The 2 possible outcomes are referred to as **successes** and **failures**.
- We denote the probability of a success as p .
- We denote the probability of a failure as $q = 1 - p$.

3.2.1. Bernoulli distribution.

Definition 3.18. Bernoulli random variable.

If X is a random variable that takes value 1 with probability of success p and 0 with probability $1 - p$, then X is a Bernoulli random variable.

probability table; ONE trial

- We call the probability of success p the **parameter** of the Bernoulli distribution.
- Each value of p identifies a specific Bernoulli distribution out of the **family** of Bernoulli r.v.'s where p is any value between 0 and 1 (inclusive).
- If a r.v. X is modeled by a Bernoulli distribution, then we write in short-hand

$$X \sim \text{Bern}(p)$$

Theorem 3.19. Mean and SD of a Bernoulli r.v.

If X is a Bernoulli r.v. with probability of success p , then

$$\begin{aligned}\mathbb{E}(X) &= p \\ \text{Var}(X) &= p(1 - p) \\ \text{SD}(X) &= \sqrt{p(1 - p)}\end{aligned}$$

Covid example

3.2.2. Binomial distribution.

Recall Example 3.17 from Day 7:

- About 25% of people that test positive for Covid-19 are vaccinated for Covid-19. p
- Define the r.v. X_i to be 1 if someone that tests positive is vaccinated and 0 if they are not vaccinated. S, F
- Suppose 3 people have tested positive for Covid-19 (independently of each other). **fixed n ; independent trials**
- Let T denote the number of people that are vaccinated amongst the 3 that tested positive. **total number S**

The random variable T above is an example of a Binomial random variable.

In general, a random variable X is **Binomial** if the following hold:

- (1) The trials are independent.
- (2) The number of trials, n , is fixed.
- (3) Each trial outcome can be classified as a *success* or *failure*.
- (4) The probability of a success, p , is the same for each trial.
- (5) The r.v. X is the total number of successes in the n trials.

Definition 3.20. *Distribution of a **Binomial** random variable.*

Let X be the total number of successes in n independent trials, each with probability p of a success.

Then probability of observing exactly k successes in n independent trials is

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

- The parameters of a binomial distribution are p and n .
- If a r.v. X is modeled by a binomial distribution, then we write in shorthand

$$X \sim \text{Bin}(n, p)$$

Theorem 3.21. *Mean and SD of a Binomial r.v.*

If X is a binomial r.v. with probability of success p , then

$$\begin{aligned}\mu &= \mathbb{E}(X) = np \\ \sigma^2 &= \text{Var}(X) = np(1-p) \\ \sigma &= \text{SD}(X) = \sqrt{np(1-p)}\end{aligned}$$

Covid example 3.17

Example 3.22. Vaccinated people testing positive for Covid-19 (revisited)

About 25% of people that test positive for Covid-19 are vaccinated for Covid-19.

Suppose 10 people have tested positive for Covid-19 (independently of each other). Let X denote the number of people that are vaccinated amongst the 10 that tested positive.

(1) What is the expected value of X ?

$$\mathbb{E}[X] = 10 \cdot p = 10 \cdot 0.25 = 2.5$$

(2) What is the SD of X ?

$$\text{Var}[X] = npq = 10 \cdot 0.25 \cdot 0.75 = 10 \cdot 0.1875 = 1.875$$

$$SD[X] = 1.369$$

(3) What is the probability that exactly 4 of the 10 people that tested positive are vaccinated?

$$P(X = 4) = \binom{10}{4} 0.25^4 (0.75)^{10-4} = \frac{10!}{4!(10-4)!} p^k (1-p)^{10-k} = 0.145998$$

```
dbinom(x, size, prob)
```

```
dbinom(x= 4, size=10, prob = .25)= 0.145998
```

- (4) *What is the probability that at most 3 of the 10 people that tested positive are vaccinated?*

$$P(X \leq 3) =$$

$$P(X \leq q) =$$

```
pbinom(q, size, prob, lower.tail = TRUE)
pbinom(3, size=10, prob = .25)= 0.2502823
```

- (5) *What is the probability that at least 5 of the 10 people that tested positive are vaccinated?*

$$P(X \geq 5) = P(X > 4) = 0.078$$

$$P(X \geq 5) = 1 - P(X \leq 4) = 1 - 0.922$$

$$P(X > q) =$$

```
pbinom(q, size, prob, lower.tail = FALSE)
pbinom(4, size=10, prob = .25, lower.tail = FALSE)
= 0.07812691
```

$$P(X \leq q) =$$

```
pbinom(4, size=10, prob = .25, lower.tail = TRUE)
= 0.9218731
```