Probability

Mathematical background for statistics

Download the section 7 .Rmd handout to STAT240/lecture/sect07-probability.

Material in this section is covered by Chapters 9 and 10 on the notes website.

In statistics, we work with a sample of data.

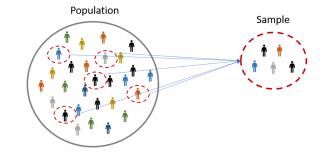
Freeze duration:

118, 151, 121, 96, 110, 117, 132, 104, 125, ...

Heights:

65, 68, 70, 72, 64, 63, ...

We assume random samples are drawn from a **population**.



Goal: learn about population

- What is the true average freeze duration?
- What is the true mean height of US males?

Random sampling introduces uncertainty, which we study with **probability**.

Probability is used all the time:

- I might miss my bus if I don't hurry up.
- I will need to have a good day on the final to get an A.
- On another day, we would've won that game.
- I didn't even leave late or anything, I just hit every red light.

Represent sampling as a random process.

- Any "experiment" with a random outcome
- Flip a coin twice
- Measuring a random item from a population

Repeating this process generates data.

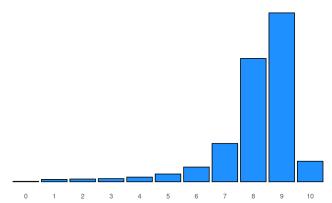
Wil will focus on **numeric** data.

- Height
- Number of siblings
- GPA

A random, numeric process is represented by a random variable (RV).

The Apgar score for newborns is on a scale from 0-10 based on condition immediately after birth.

Distribution of Apgar Scores



A random variable X has possible values given by x.

• X = process, x = specific outcome

But how common is each value?

X	0	1	2	3	4	5
	0.001	0.006	0.007	0.008	0.012	0.020
X	6	7	8	9	10	
	0.038	0.099	0.319	0.437	0.053	

- Probability of a random draw
- Breakdown of population values

Formally, a **probability function** assigns a number between 0 and 1 to each possible value of an RV.

If we ran the random process infinite times, it is the percentage of time we would get the given outcome.

Some probabilities are intuitive.

- P(Heads) from the flip of a fair coin
- P(3) from rolling a fair die

For RVs, write P(X = x).

Remember X and x are different!

The full table is a probability distribution.

X	0	1	2	3	4	5
P(X = x)	0.001	0.006	0.007	0.008	0.012	0.020
X	6	7	8	9	10	
P(X = x)	0.038	0.099	0.319	0.437	0.053	

Properties of probability distributions:

- Probabilities must be between 0 and 1.
- Probabilities must all sum to 1.

If there are a few finite outcomes, we can list the probabilities in a table.

To recap:

- A random process generates data
- A random variable is the numeric representation of the process
- A probability distribution assigns a probability to each value of the RV

These tools represent a population.

Flip a coin three times.

- What are the possible outcomes?
- What is the probability of each outcome?

Let RV *X* count the number of heads out of three coin flips.

- What are the possible values of X?
- What is the probability distribution of X?

A **discrete** RV can only take on specific values.

- Number of people in a family
- Number of goals scored in a match

A **continuous** RV can take on any value in a range.

- Time to walk to class
- Inches of rainfall

The Apgar score is discrete.

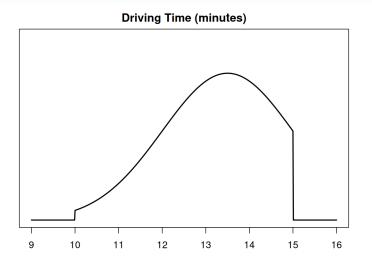
- Values can be enumerated in a table
- Probabilities sum to 1
- Values inbetween are impossible

Just because 0 and 1 are possible values, it doesn't mean (0.1, 0.535, 0.9999) also have probability.

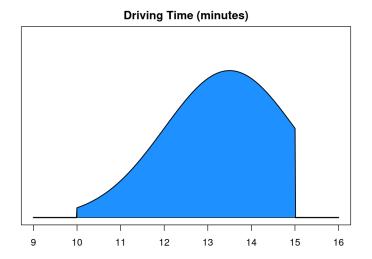
A **continuous** random variable can take on any value in a specific range.

We draw the probability as a smooth curve, similar to geom_density().

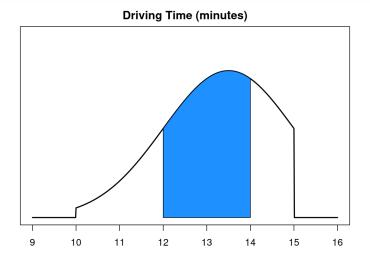
For example, it takes 10 to 15 minutes to drive to campus, depending on traffic.



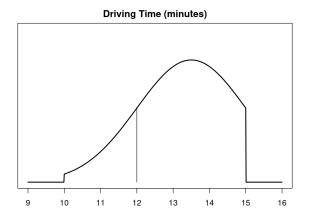
Probabilities are given by the area under the curve.



The total area is 1.

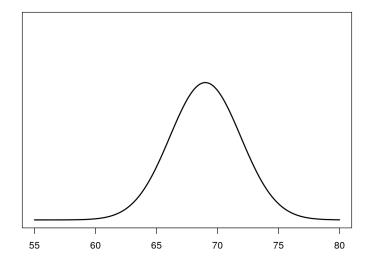


 $P(12 \le \text{driving time } \le 14).$

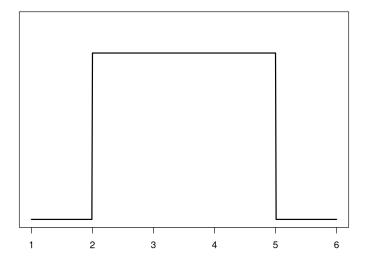


The probability of a single value like P(driving time = 12) is 0.

Height of US males:



Random number between 2 and 5:



We can use RVs to calculate summary measures.

- Expected value: population mean (center)
- Variance: population variance (spread)

We have a population where 90% of the values are 0, and 10% of the values are 1.

$$P(X = 0) = 0.9, \qquad P(X = 1) = 0.1$$

The mean is a weighted average:

$$(0)0.9 + (1)0.1 = 0.1 = E(X)$$

This RV mean is called **expectation**.

In general, the expectation of a discrete RV is:

$$E(X) = \mu = \sum_{x} x \cdot P(X = x)$$

which is a weighted average of the values of X.

If we were able to take a sample of infinite x_i 's and take their mean, \bar{x} would be equal to μ .

The **variance** of a RV is the average squared distance from the RV to its mean.

We start by taking $[x - E(X)]^2$.

Just like with expectation, we take a *weighted* average based on the probabilities.

The variance of a discrete RV is

$$V(X) = \sigma^2 = \sum_{X} [x - E(X)]^2 \cdot P(X = x)$$

The **standard deviation** of X is the square root of the variance.

$$sd(X) = \sqrt{V(X)}$$
 or $\sigma = \sqrt{\sigma^2}$

Expectation and variance have the same interpretation for continuous RVs.

- E(X): population mean
- V(X): population variance

To calculate the mean and variance of a continuous RV, we use calculus, which is not a part of 240.

Let's work with a small discrete example. Recall:

Probabilities must sum to 1

$$E(X) = \sum_{x} x \cdot P(X = x)$$

$$V(X) = \sum_{x} [x - E(X)]^2 \cdot P(X = x)$$