

Lecture 4

Shape and Measures of Central Tendency

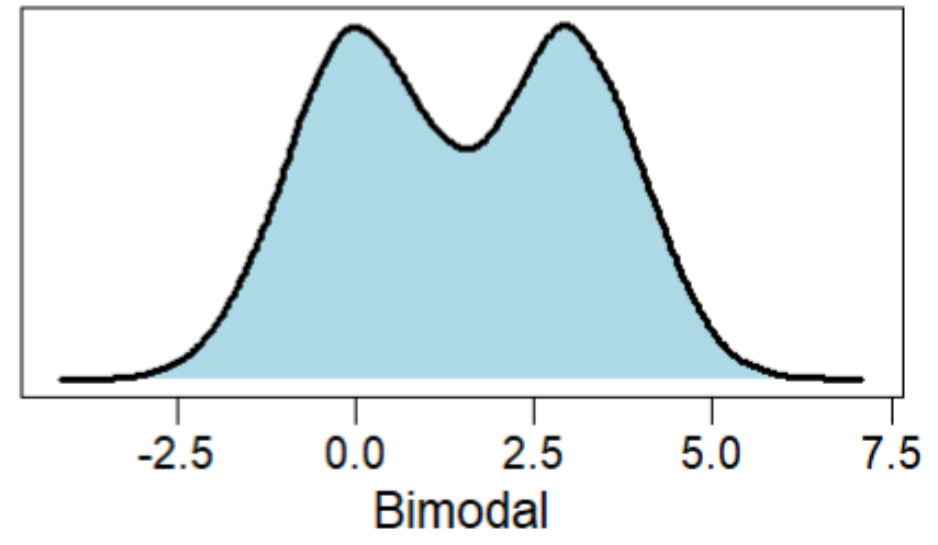
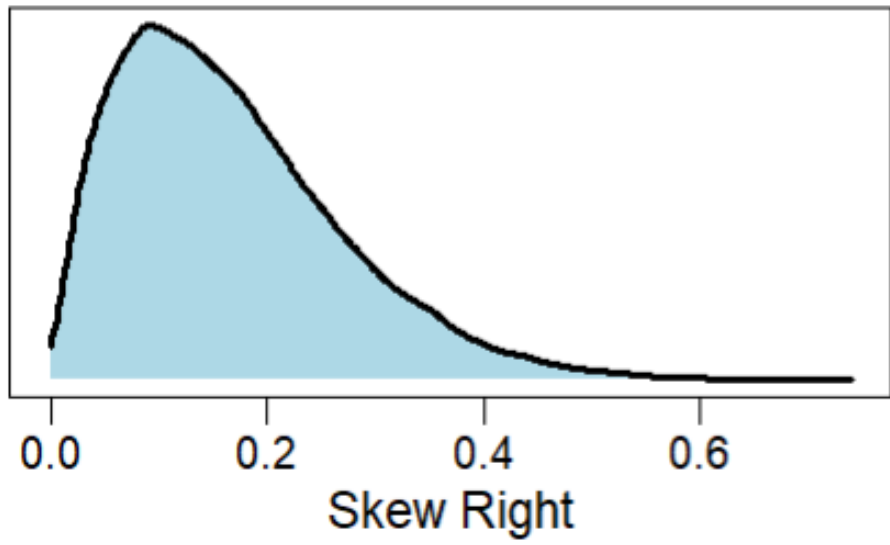
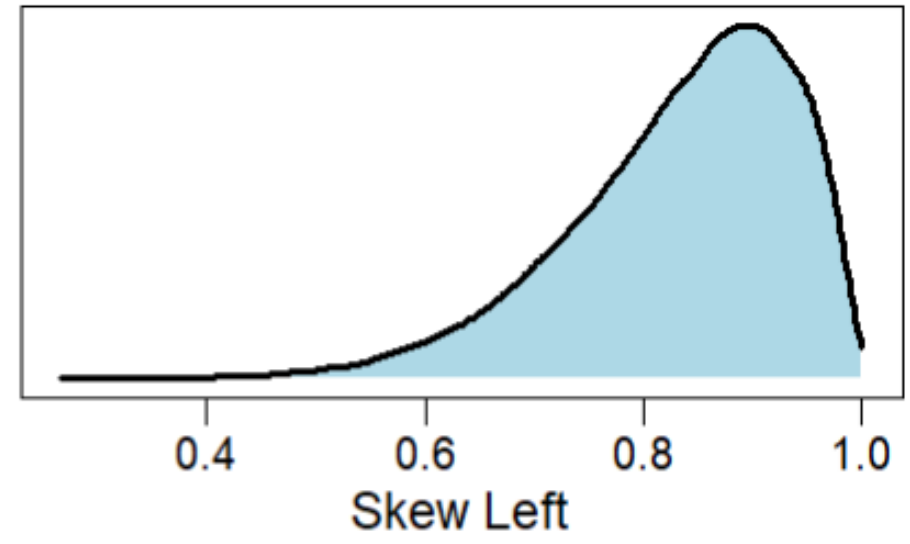
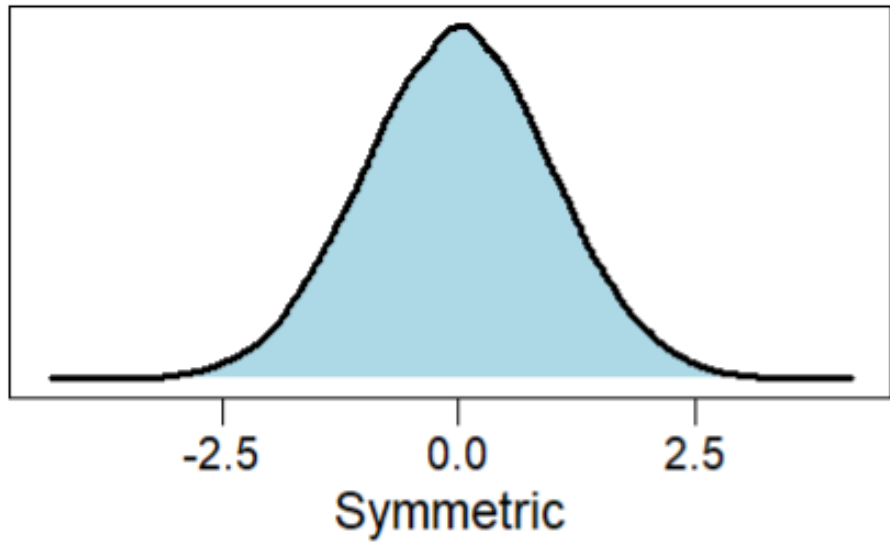
Review from Friday 1/19

Practice: Histogram

- $X = \{-1.49, -0.65, -0.6, -0.54, -0.45, 0.01, 0.17, 0.27, 0.51, 1.34\}$

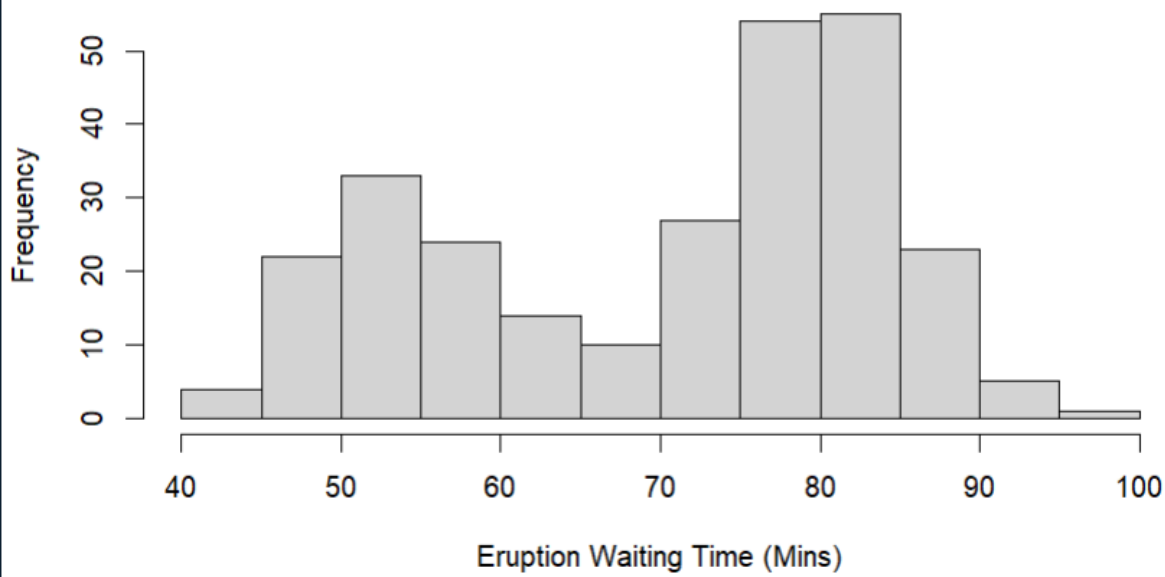
Construct a histogram using $K = 4$ bins/intervals:

Shape of a distribution



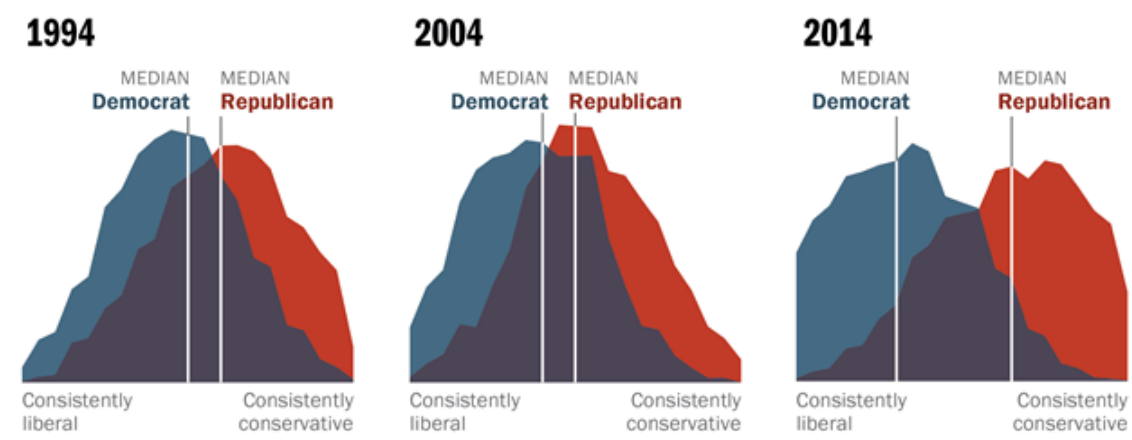
- Bimodal distributions can arise when
 - A population is polarized on a controversial issue
 - When observations come from two different populations

Histogram of Eruption Waiting Times



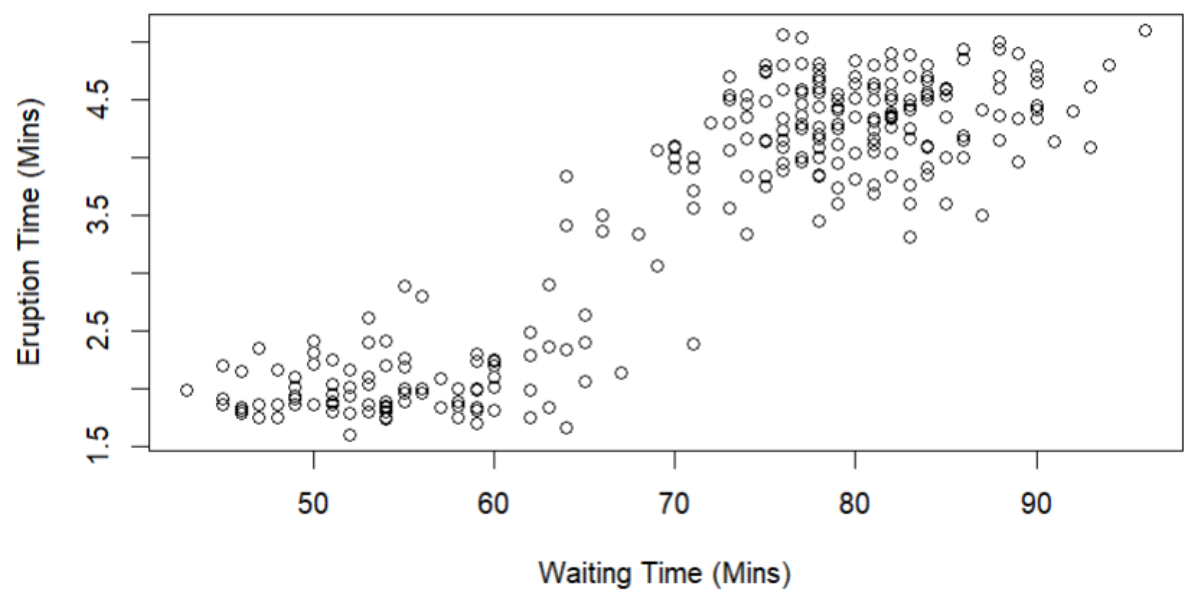
Democrats and Republicans More Ideologically Divided than in the Past

Distribution of Democrats and Republicans on a 10-item scale of political values



Source: 2014 Political Polarization in the American Public
 Notes: Ideological consistency based on a scale of 10 political values questions (see Appendix A). The blue area in this chart represents the ideological distribution of Democrats; the red area of Republicans. The overlap of these two distributions is shaded purple. Republicans include Republican-leaning independents; Democrats include Democratic-leaning independents (see Appendix B).

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Measures of Central Tendency

- The (arithmetic) **mean** is the average of a set of observations
it measures the center of mass of a distribution (the balancing point)

$$\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n} = \sum_{i=1}^n \frac{x_i}{n}$$

- We can also express the mean in terms of the frequency F or the relative frequency RF

$$\bar{x} = \frac{1}{n} \sum_x x F(x) \quad \text{or} \quad \bar{x} = \sum_x x RF(x)$$

Where the sum is over all distinct values of the variable x

- the mean is usually not equal to any of the values observed in the sample
- The mean is highly influenced by **outliers** - observations that take on extreme values relative to the distribution

The mean

- $X = \{$

Measures of Central Tendency

- The **median** is the middle value of a set of observations

Ex.)

Data = 1,1,4,5,6

How to compute the median:

1. Compute the median by first ordering the observations from smallest value to largest value and choose the number in the middle

Median = 4

Mode = 1

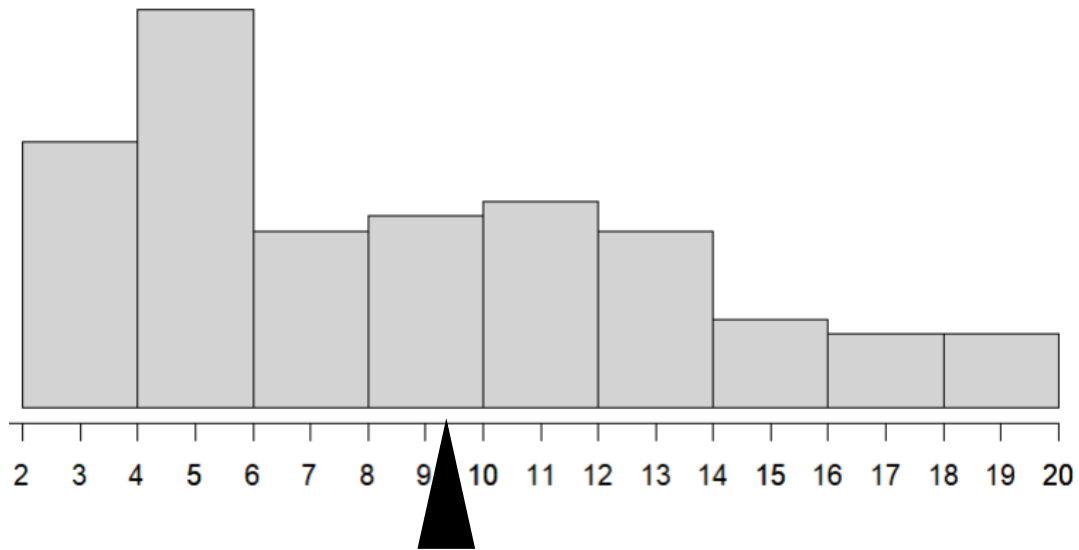
Data = 1,1,4,5,6,6

2. If the n is odd the median is the middle number
 - If n is even the median is the sum of the two middle values divided by 2

Median = $\frac{4+5}{2} = 4.5$

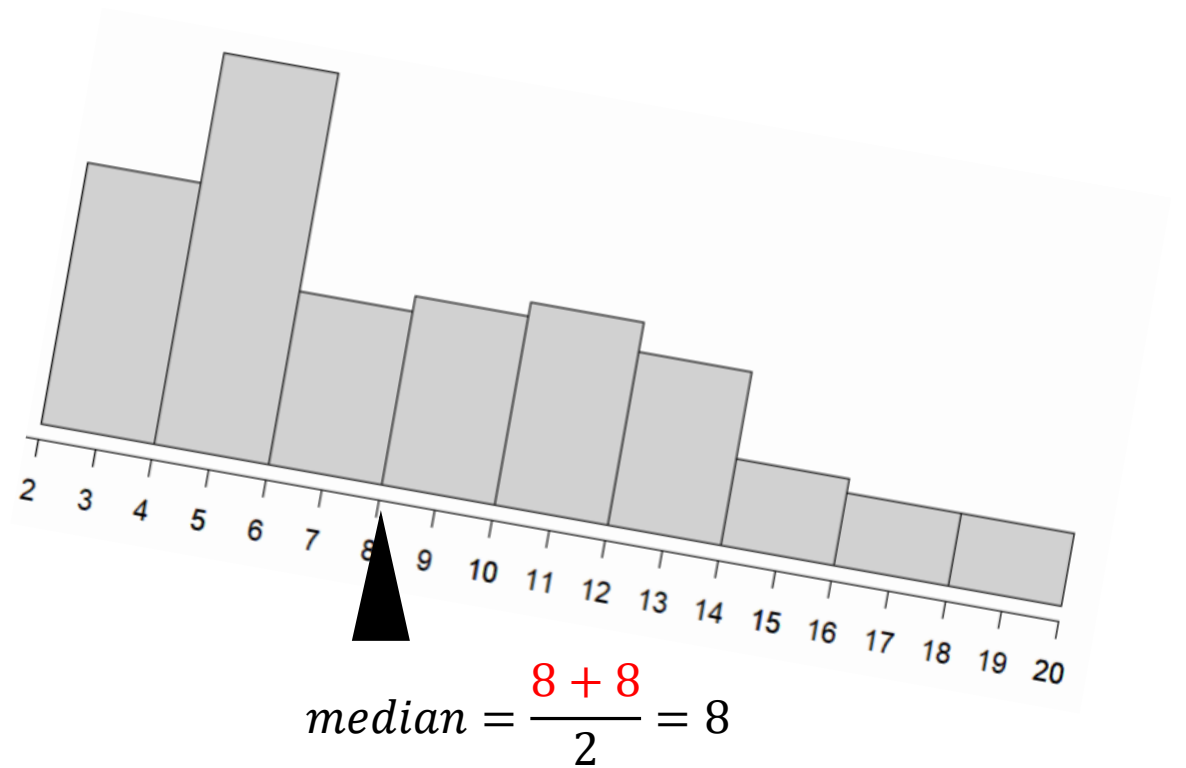
Mode 1, 6

Data: 3 3 3 3 3 3 3 4 4 4 4 4 4 4 4 4 5 5 5 5 5 5 5 5 5 5
 5 5 5 5 5 5 5 5 5 5 6 6 6 6 6 6 6 6 6 6 7 7 7 7 7 7 7
 7 8 8 8 8 8 9 9 9 9 9 9 9 10 10 10 10 10 10 10 11 11 11 11 11 12
 12 12 12 12 12 12 12 13 13 13 14 14 14 14 14 14 15 15 15 16 16 16 16 17 17 17 18
 18 20 20 20 20 20



$$\bar{x} = 9.2$$

The mean is the center of gravity



$$median = \frac{8 + 8}{2} = 8$$

The median is the middle value

Measure of Central Tendency

- The **mode** is the value with the largest relative frequency (i.e the value that occurs most often)
 - Can be used with categorical data (mean and median cannot)
 - e.g the most frequent category
 - It may not be unique if two or more values have the same frequency
 - **Caution** for quantitative data, the mode may not anywhere near the center of the distribution.

Ex.)

Data = 1,1,4,5,6

Mode = 1

Data = 1,1,4,5,6,6

Mode 1, 6

Practice:

- Roll a six-sided die $n = 10$ times and record the number rolled each time
- Data = 1,2,3,3,4,4,4,5,6,6

x	$f(x)$	$rf(x)$
1	1	0.1
2	1	0.1
3	2	0.2
4	3	0.3
5	1	0.1
6	2	0.2

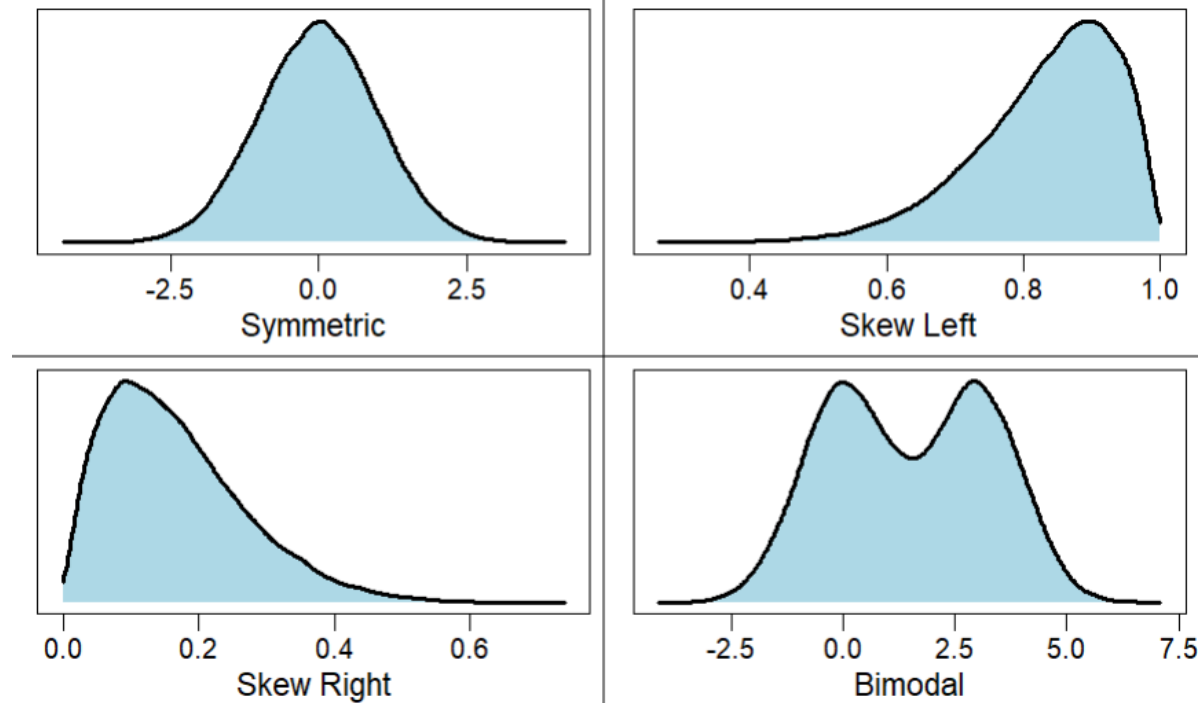
Compute the **mean** using all 3 equations:

Compute the **median**

Compute the **mode**

Comparing the Mean, Median, and Mode

- The shape of a distribution influences whether the mean is larger or smaller.
- Skew left = mean < median
- Skew right = mean > median
- When a distribution is symmetric the mean will equal the median



Comparing the Mean, Median, and Mode

- The median is a robust estimate of the mean
- The median is not usually affected by the presence of outliers
- The median is usually preferred for highly skewed distributions
- Ex.) take using the following 9 data points: 0.3, 0.4, 0.8, 1.4, 1.8, 2.1, 5.9, 11.6, 16.9

The **mean** is about 4.58

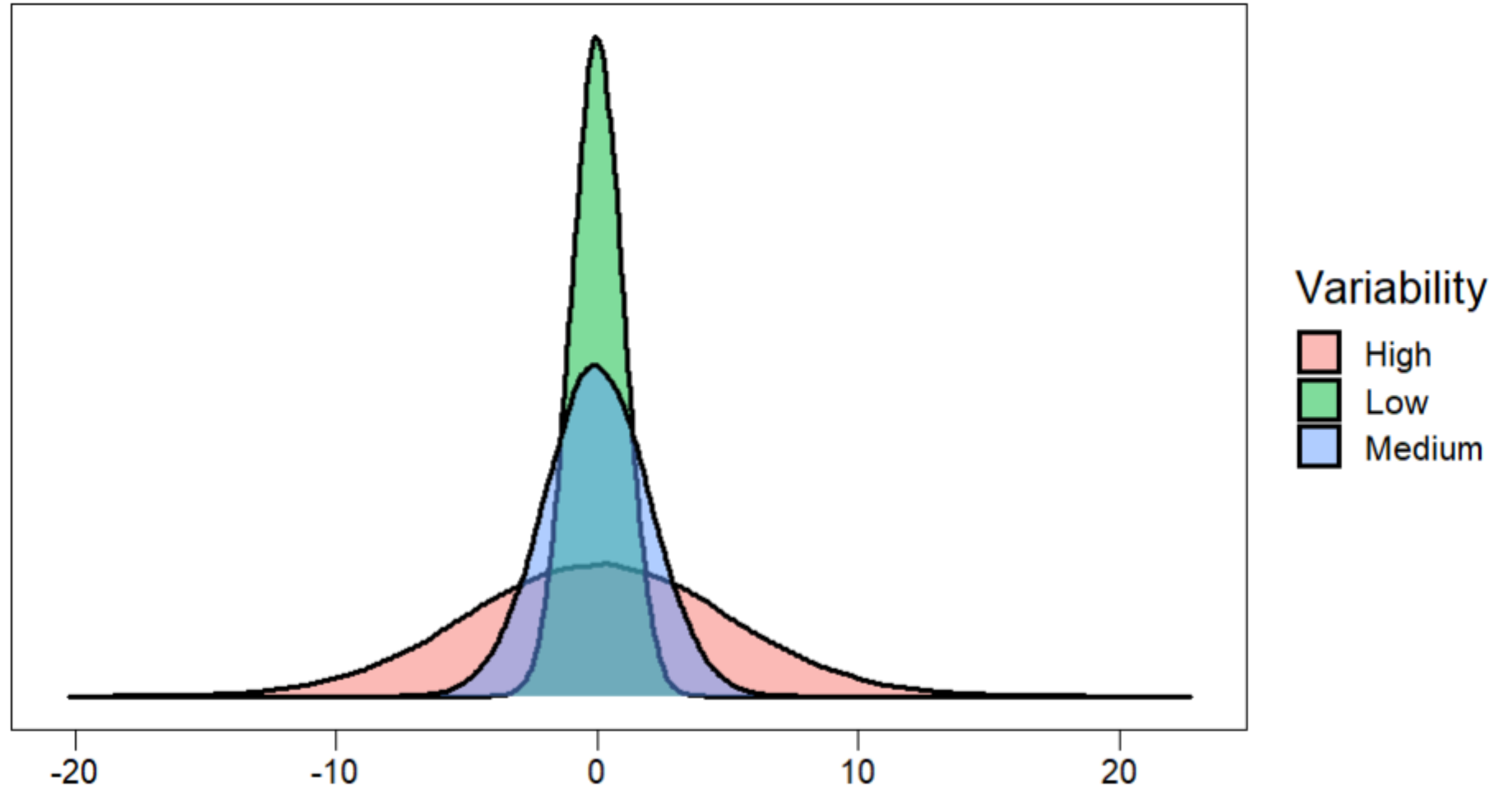
The **median** is 1.8

- Change one of the data points to be an outlier, for example, we change **16.9** to **90**

The **mean** becomes 12.7

While the **median** is still 1.8

Variability of A Distribution: Measures of Spread

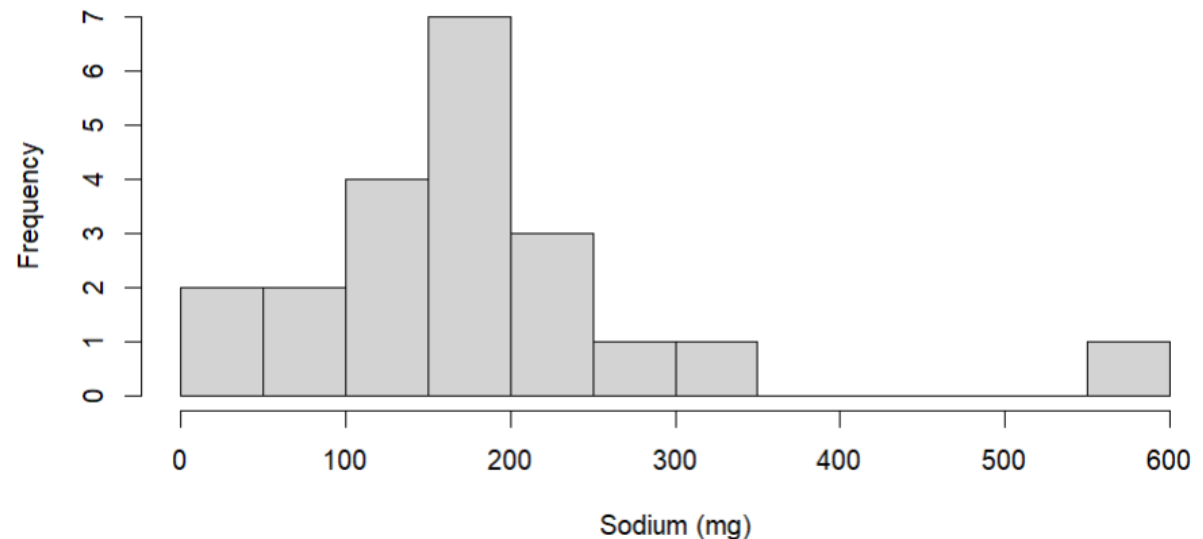
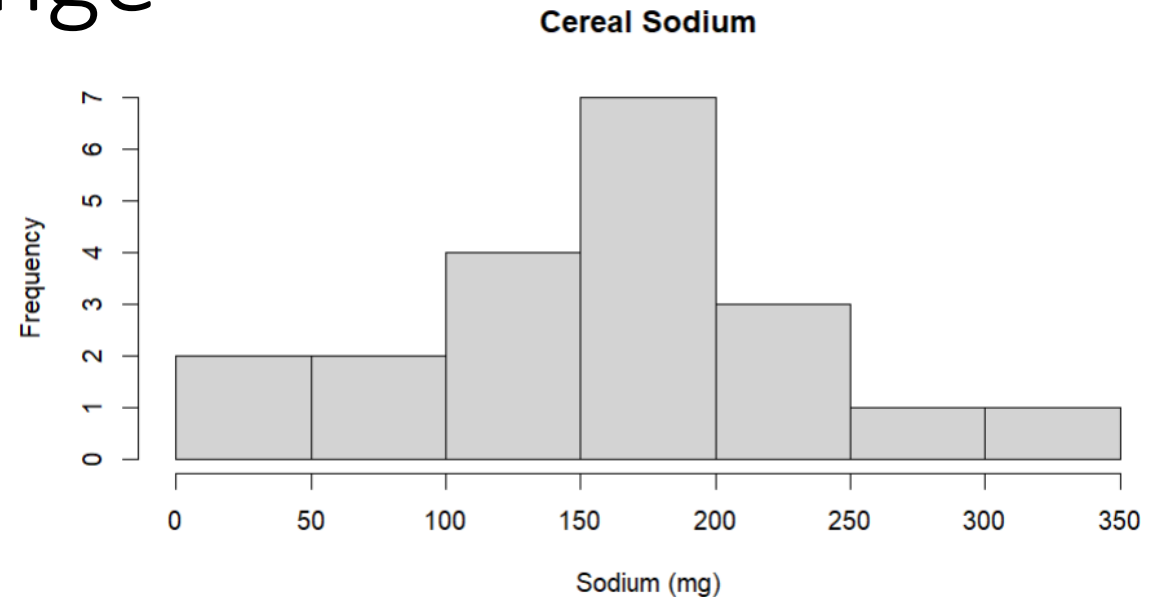


Measures of Spread: Range

- The **range** is a measure of the distance between the smallest and largest values in the data

The range can be computed with only two data points the minimum value and maximum value

- If the range of a set of data is large, then the data vary more
- The range is severely affected by the presence of outliers
- We typically do not use the range to measure variability



Measures of Spread: Deviation

- A better measure of variability that uses *all* the data is based on **deviations**
- **deviations** are the distances of each value from the mean of the data:

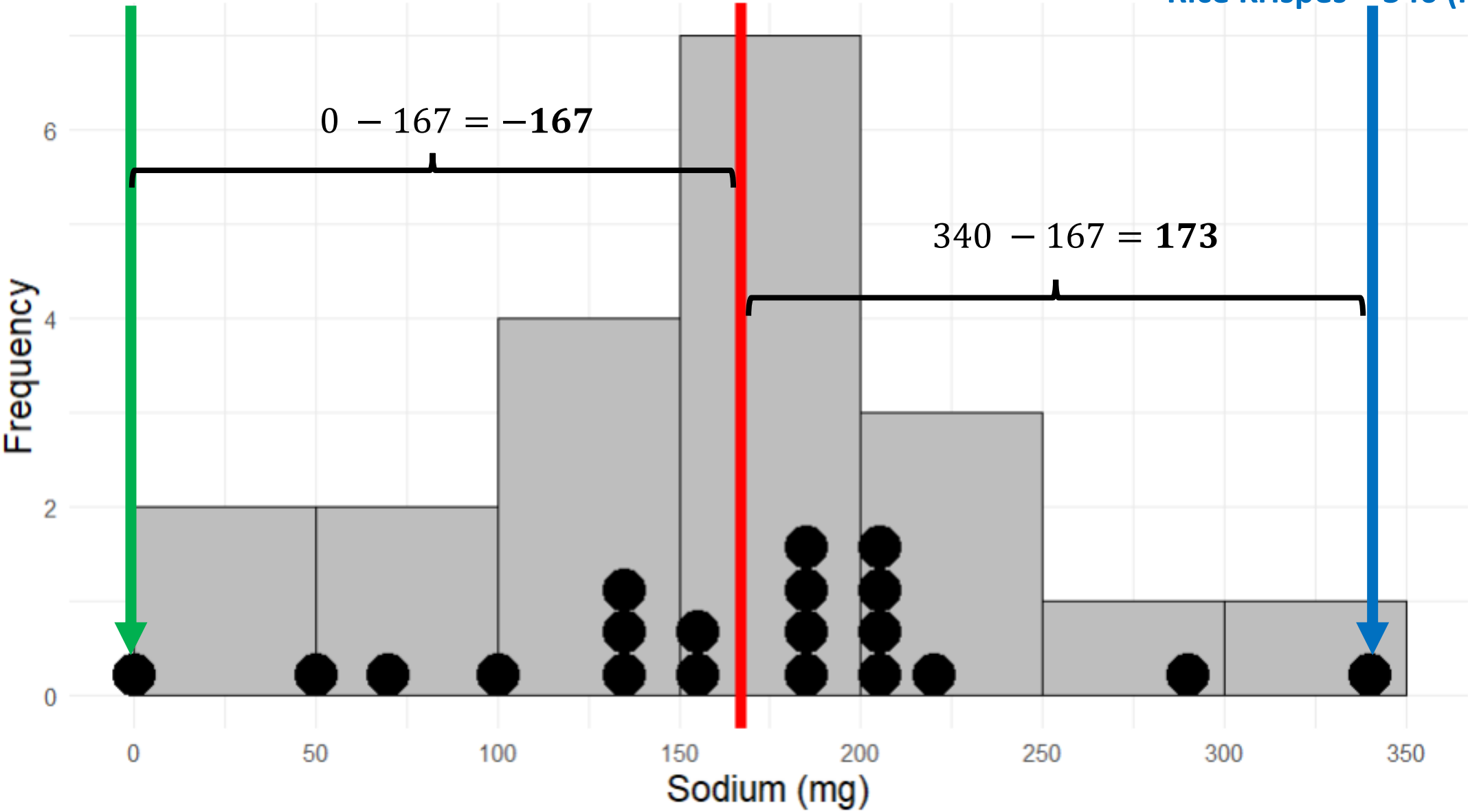
Deviation of an observation $x_i = (x_i - \bar{x})$

- Every observation will have a deviation from the mean

Frosted Mini Wheats = 0 (mg)

Mean = 167 (mg)

Rice Krispes = 340 (mg)



Measures of Spread: Variance

- The sum of all deviations is zero. $\sum_{i=1}^n (x_i - \bar{x}) = 0$
- We typically use either the **squared deviations** or their **absolute value**
Squared deviation of an observation $x_i = (x_i - \bar{x})^2$
- The **Variance** of a distribution is the average squared deviation from the mean

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

- The sum $\sum_{i=1}^n (x_i - \bar{x})^2$ is called the sum of squares

Measures of Spread: Standard Deviation

- Since the variance uses the squared deviation, we usually take its square root called the **standard deviation**

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

- The standard deviation represents (roughly) the average distance of an observation from the mean
- The greater s is the greater the variability in the data is
- We denote the population parameter for the variance and standard deviation using σ for s and σ^2 for s^2

Why divide by $n - 1$?

- We divide by $n - 1$ because we have only $n - 1$ pieces of independent information for s^2
- Since the sum of the deviations must add to zero, then if we know the first $n - 1$ deviations we can always figure out the last one
- Ex.) suppose we have two data points and the deviation of the first data point is $x - \bar{x} = -5$
 - Then the deviation of the second data point has to be 5 for the sum of deviations to be zero.

Try it out: Computing s and s^2

- Roll a six-sided die $n = 10$ times and record the number rolled each time
- Data = 1,2,3,3,4,4,4,5,6,6
- Mean = 3.8

