

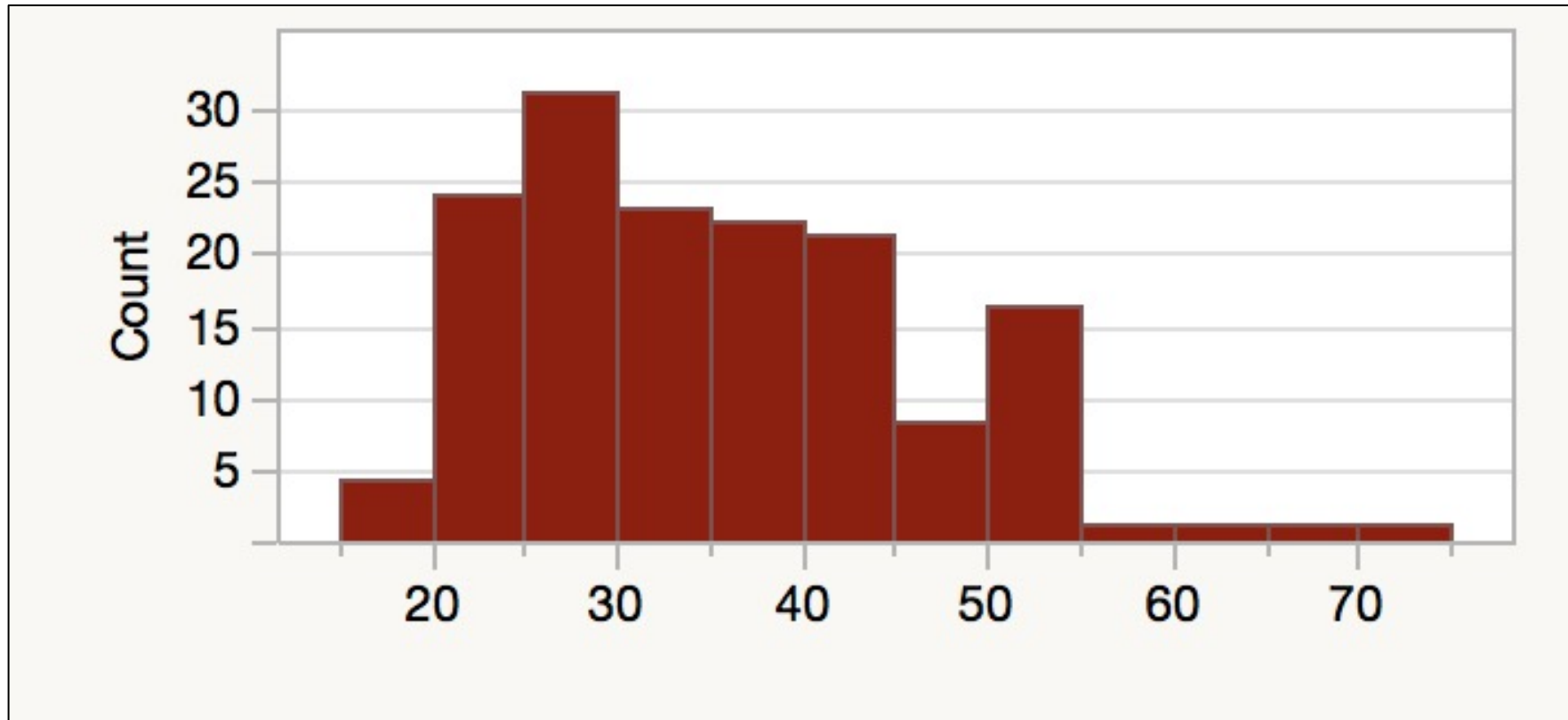
Data Driven Decision Making: Descriptive Statistics

GSBA 545, Fall 2021

Professor Dawn Porter

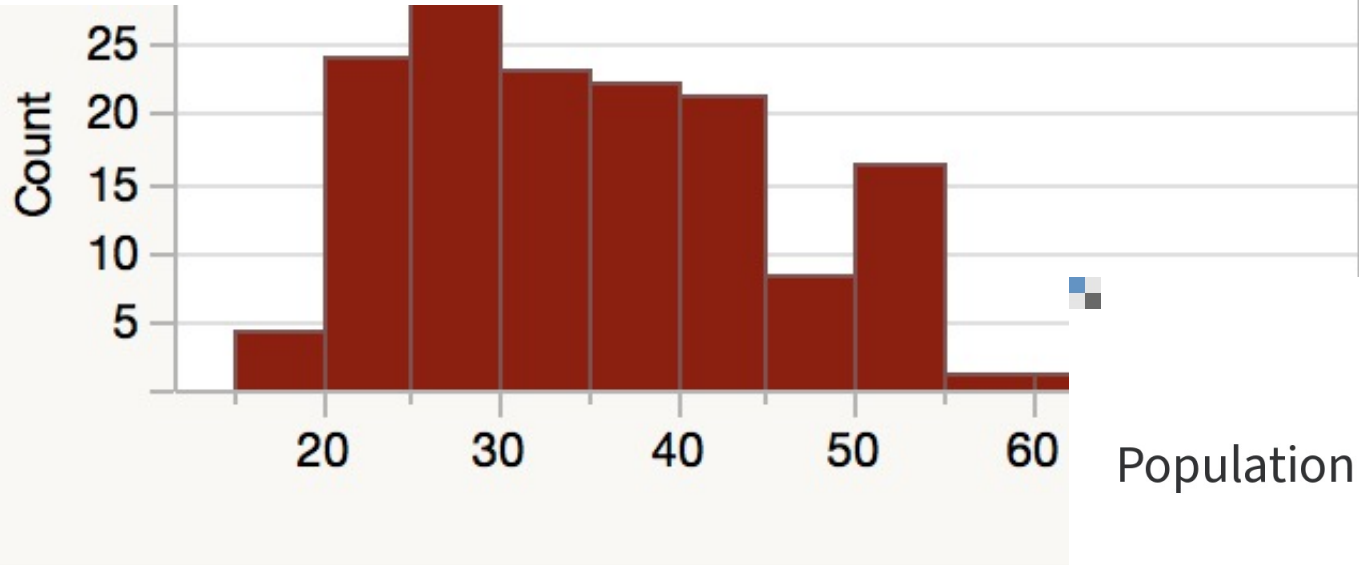
- Basic Terminology & Scales of Measurement
- Numerical Measures
 - Central Tendency
 - Dispersion
- Graphical Methods
 - Histograms
 - Box-and-Whisker plots
 - Bar Charts & Pie Charts
 - Scatterplots

MPG for 153 Hybrid Cars



Hybrid Car data: Population or Sample?

MPG for 153 Hybrid Cars

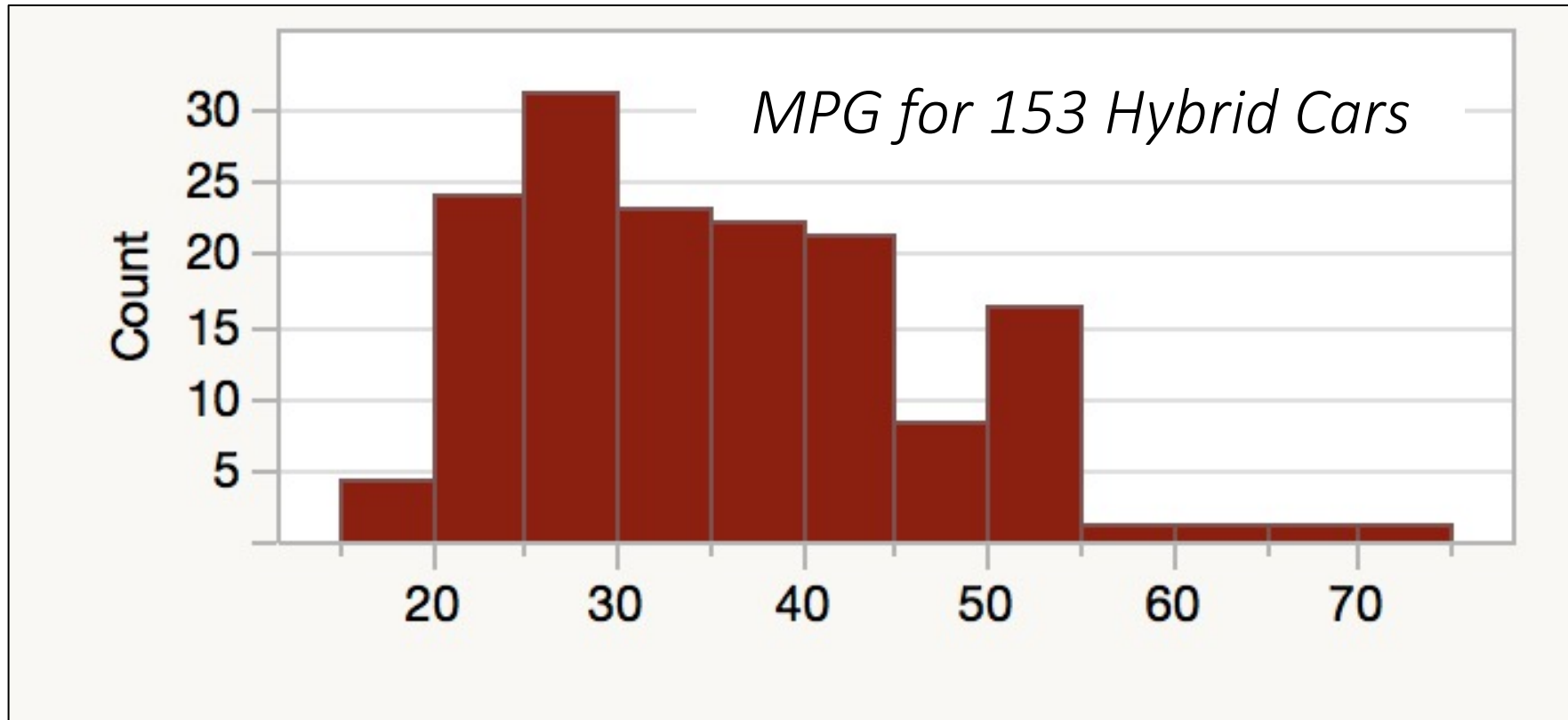


Population

Sample

Total Results: 0

Population



MPG for 153 Hybrid Cars

Population: Set of all items of interest in a statistical problem.

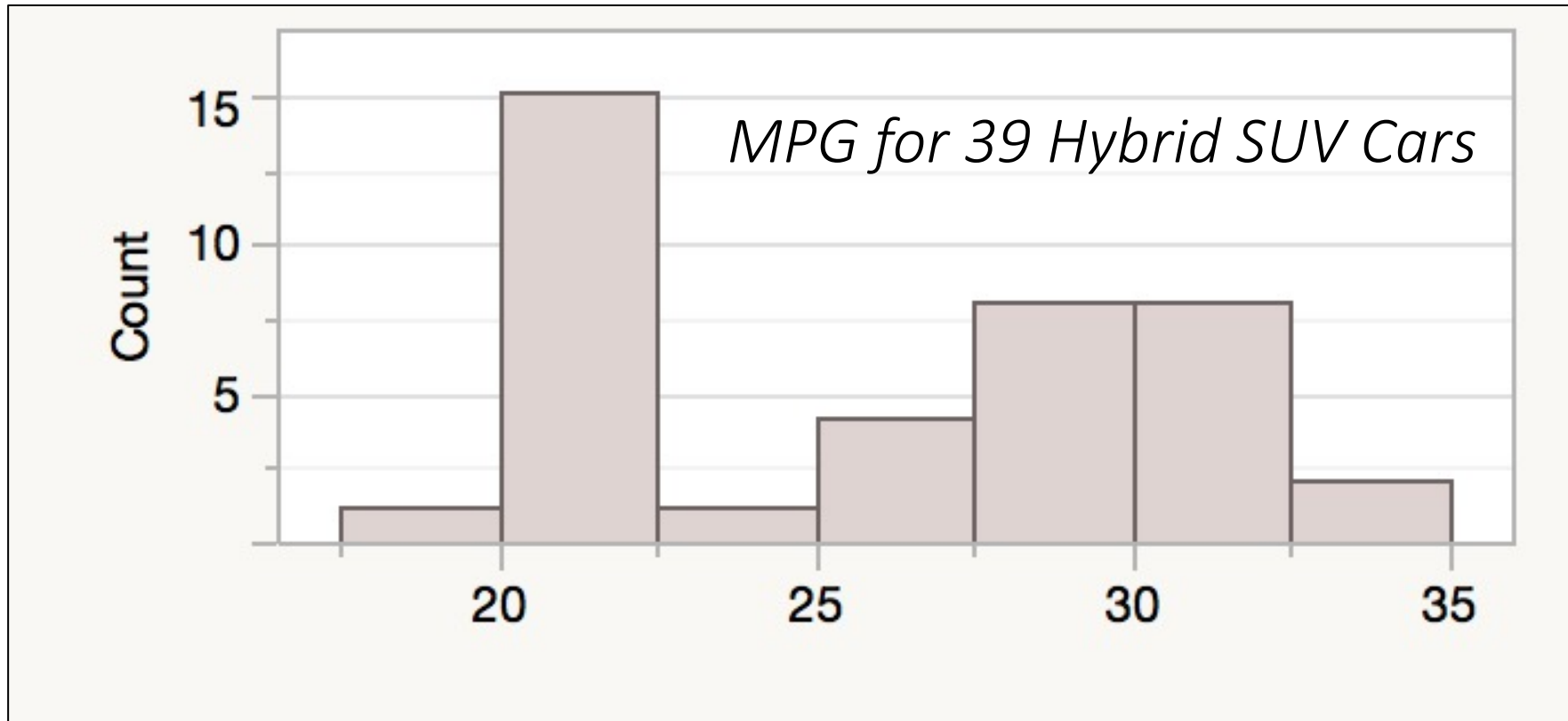
Parameter: Descriptive measure of population

- N = population size
- μ = population average
- σ = population standard deviation

Population: 153 Hybrid Cars

- $N = 153$
- μ = mean = average = 34.80 mpg
- σ = standard deviation = typical fluctuation = 10.97 mpg

Sample



MPG for 39 Hybrid SUV Cars

Sample: Set of data drawn from the population

Statistic: Descriptive measure of sample

- n = sample size
- \bar{x} = sample average
- s = sample standard deviation

Sample: 39 Hybrid SUV Cars

- $n = 39$
- \bar{x} = mean = average = 26.01 mpg
- s = standard deviation = typical fluctuation = 4.60 mpg

Numerical (quantitative)

- Natural measurement system
- Ratios and comparisons make sense



Histograms
Boxplots
Scatterplots

Categorical (qualitative)

- Nominal: no inherent ordering
- Ordinal: ordered, but distance between classes may vary



Bar Charts
Pie Charts
Side-by-side Boxplots

Scales of Measurement

Discrete: Possible number of values is countable

- Number of Hybrid SUV Cars
- Number of Comedy films released in 2017
- Number of games in any given World Series

Continuous: Possible number of values is relatively infinite

- MPG of Hybrid Cars
- Height, weight, distance

Cross-sectional: Snapshot of data at a specific point in time

- Economic indicators for several countries in 2019

Time Series: Result of tracking one or more variables over time

- Economic indicators for only the US from 1900-2019

Histograms and boxplots help uncover distribution shape:

Symmetrical (roughly equal tails)

- Bell-Shaped Distribution.

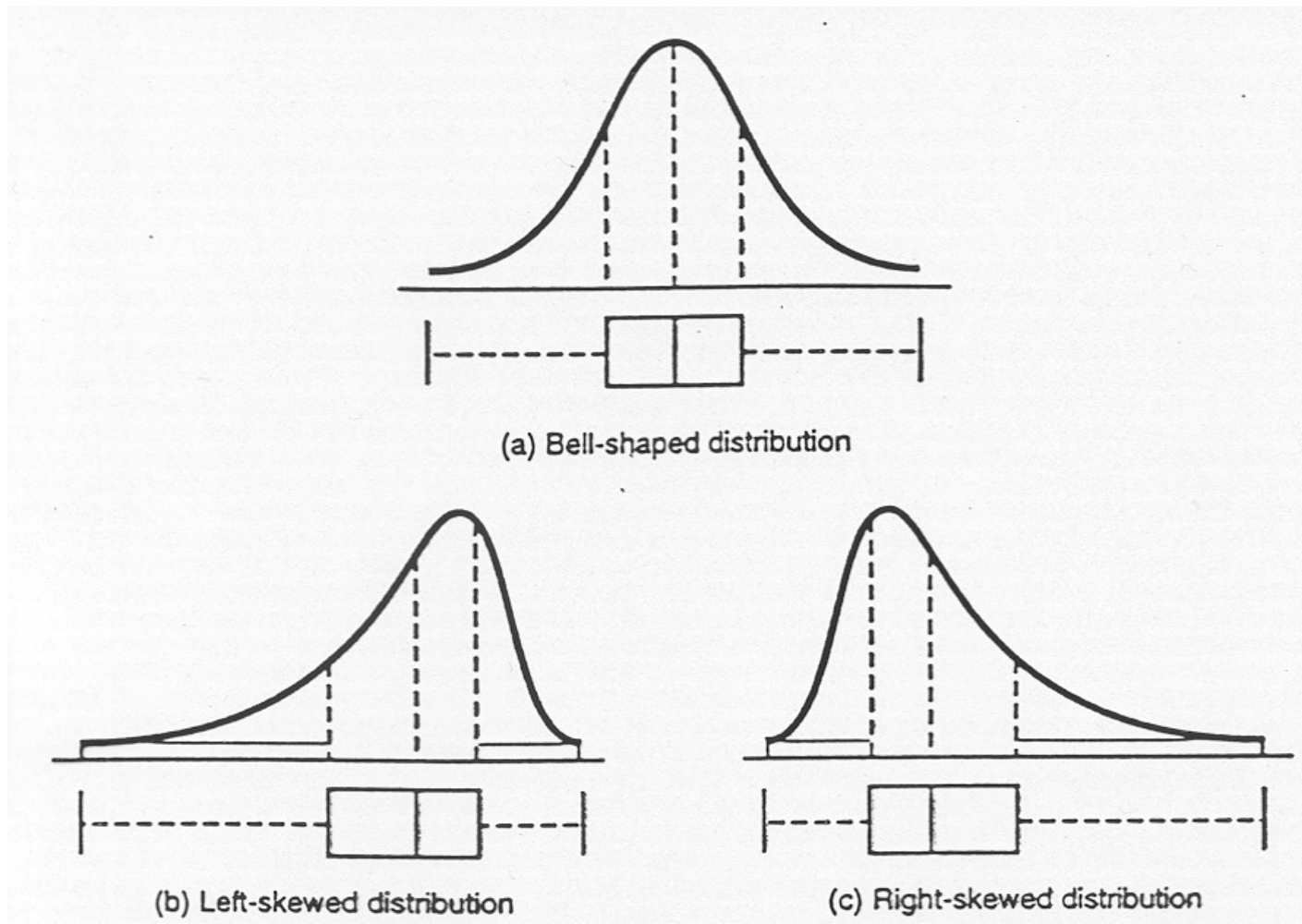
Positively Skewed – skewed right (long tail on right)

- Income Distributions.

Negatively Skewed – skewed left (long tail on left)

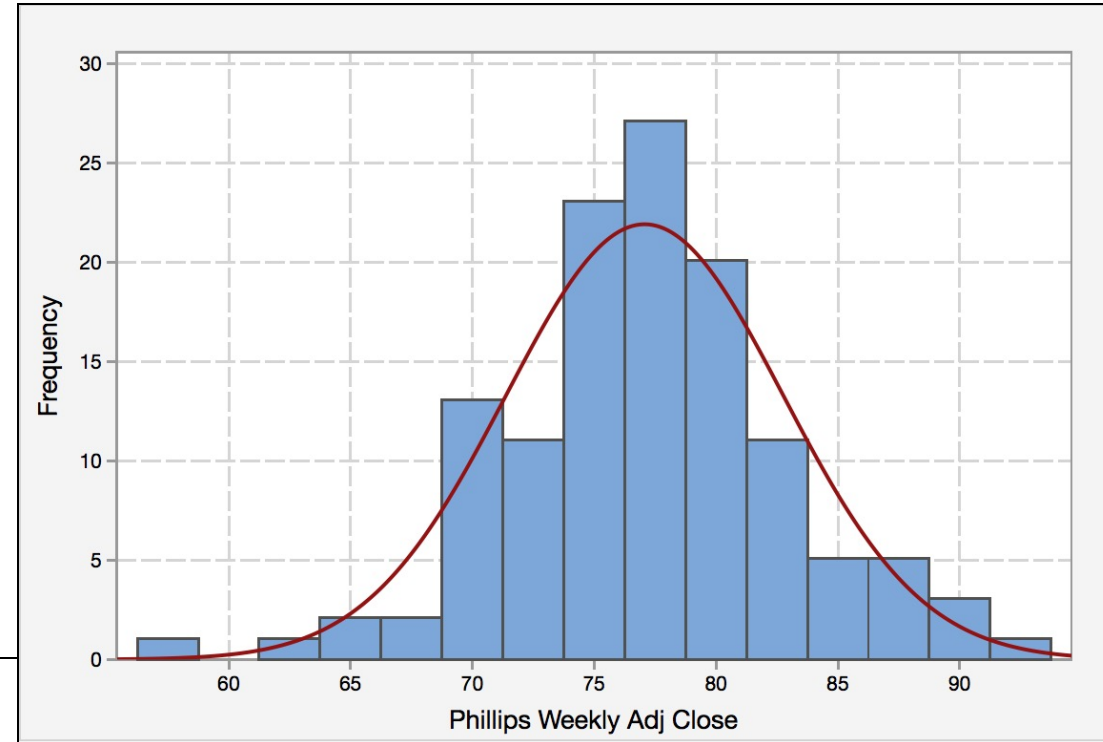
- Scores on an easy exam.

Distribution Shapes



Phillips Stock Prices*:

Mean \approx Median and Median is *somewhat* close to being about halfway between 25th and 75th percentiles.



Statistics

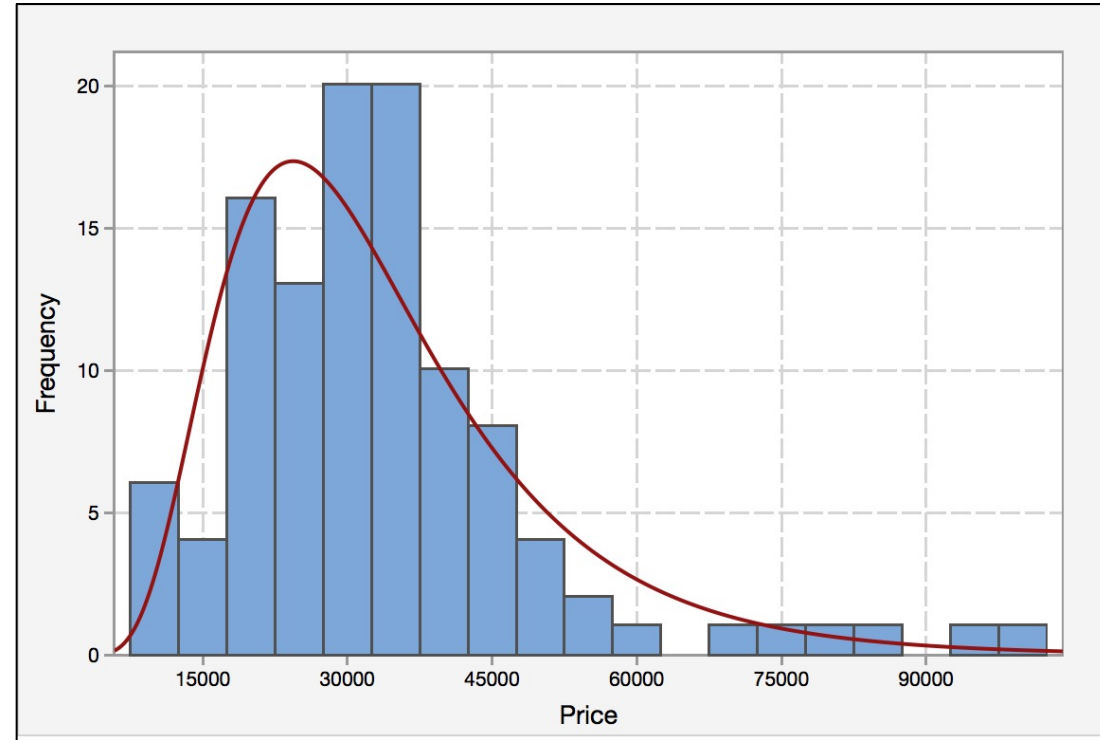
Variable	N	Mean	StDev	Minimum	Q1	Median	Q3	Maximum
Adj Close	125	77.0677	5.6917	58.3264	73.9521	77.2600	79.8241	91.3296

*Weekly closing prices, 1/6/14 – 5/23/16

Right-Skewed Distribution

LA Used Car Prices:

Mean > Median and Mean is closer to the 75th percentile than to the 25th percentile.



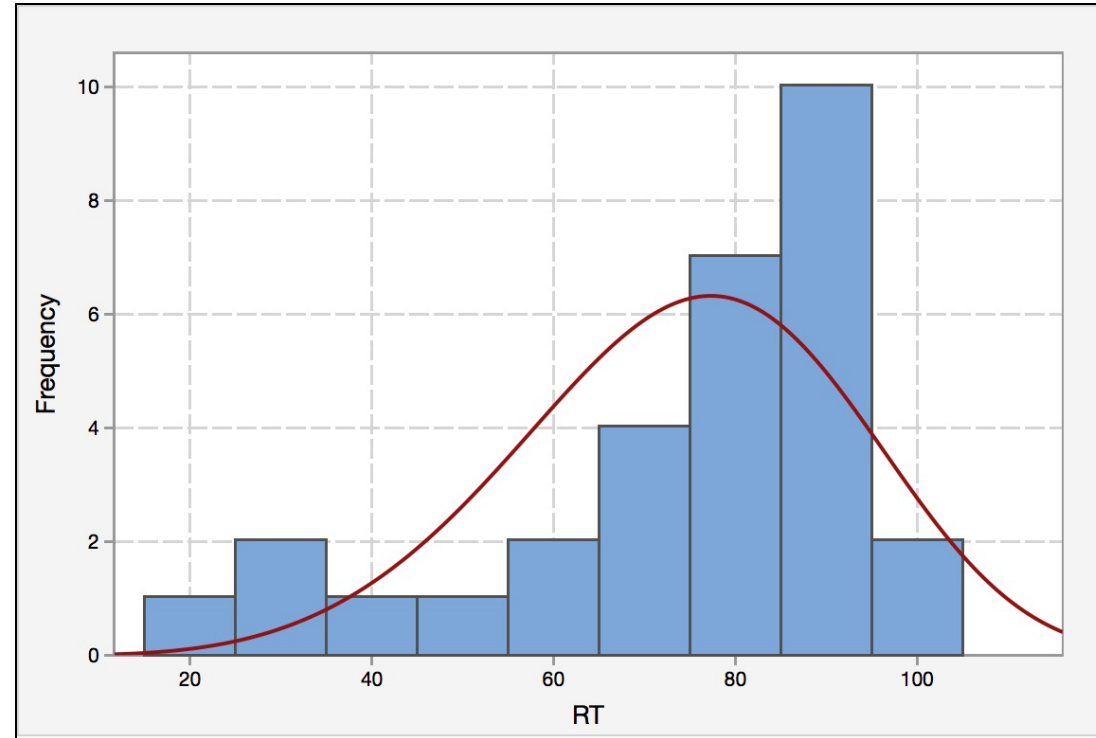
Statistics

Variable	N	Mean	StDev	Minimum	Q1	Median	Q3	Maximum
Price	110	33598	16314	9950	22991	31210	39220	99999

Left-Skewed Distribution

Top Movies in China, Rotten Tomatoes Score:

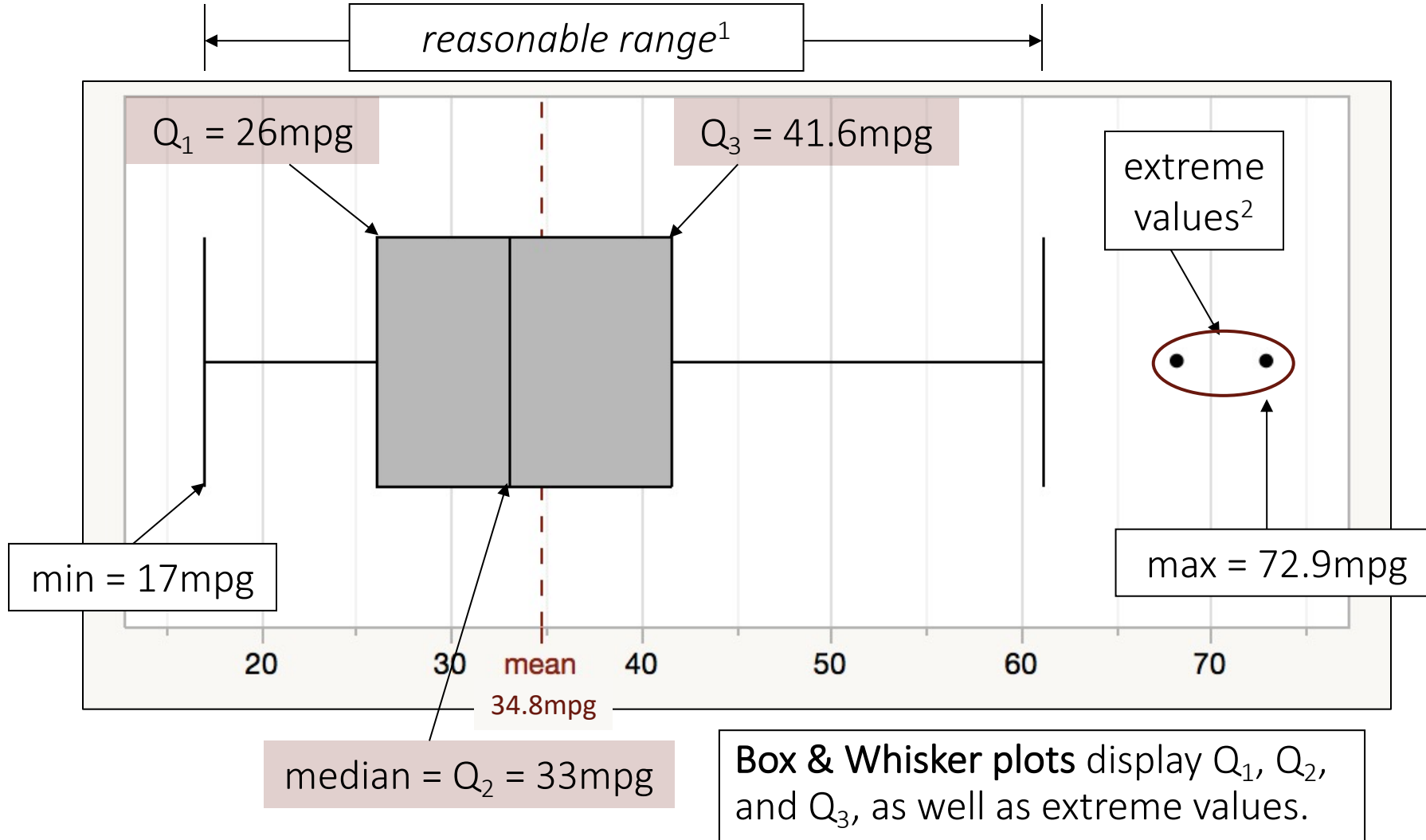
Mean < Median and Median is closer to 75th than to 25th percentile.



Statistics

Variable	N	Mean	StDev	Minimum	Q1	Median	Q3	Maximum
RT	30	74.433	21.716	18.000	68.500	81.000	91.000	98.000

Box & Whisker Plot

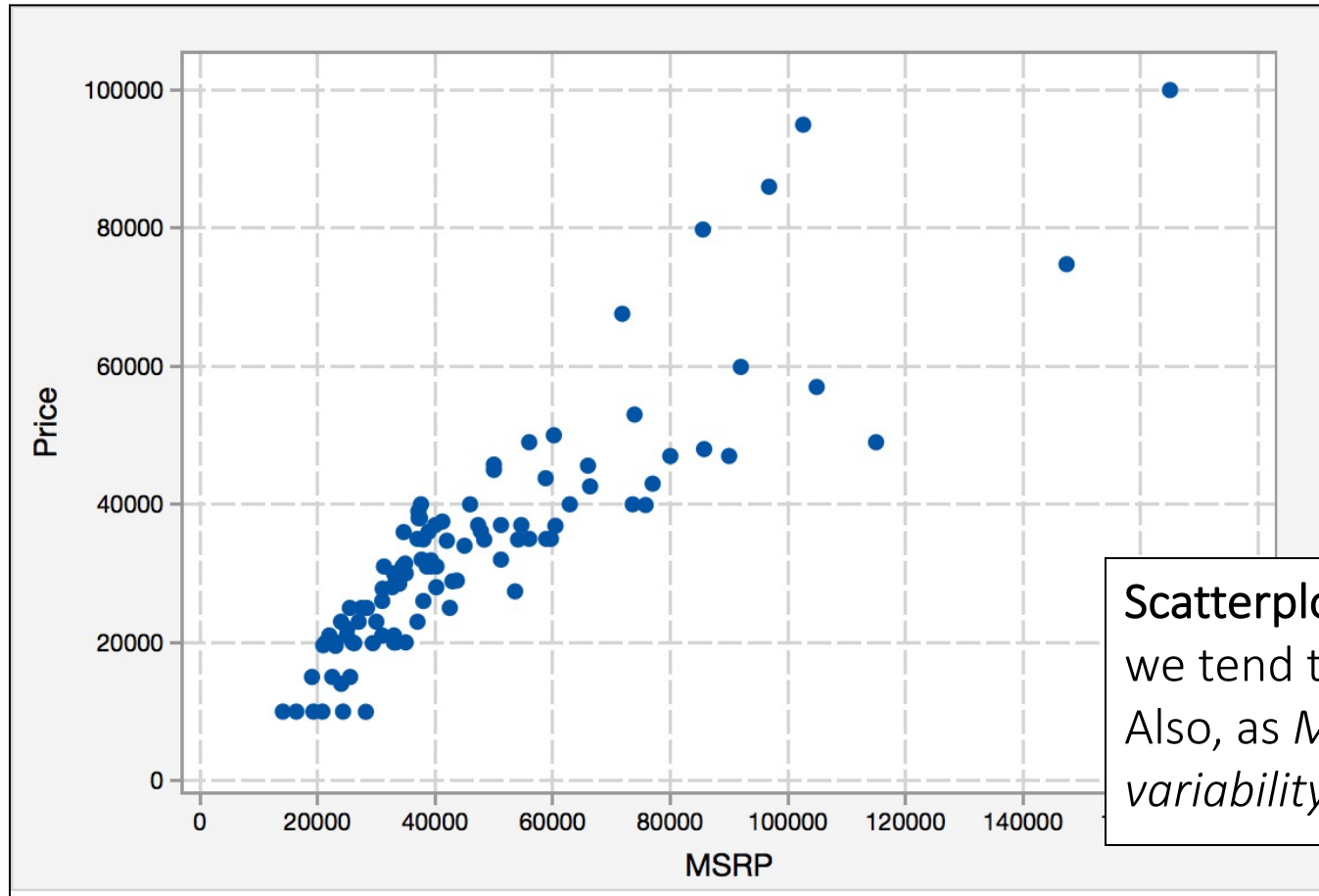


¹ Bounds of the *reasonable range* are:

$$\text{Median} \pm 1.5 \text{ IQR}$$

² *Extreme values* are defined as being at least 3 IQRs from the median.

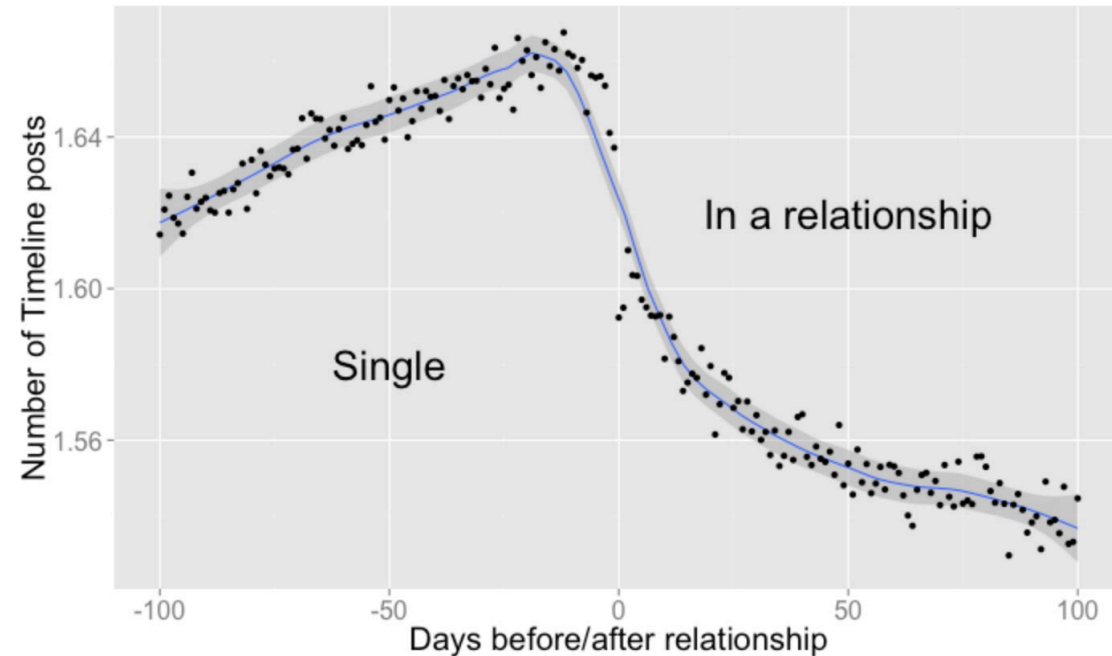
Scatterplot: Example



Scatterplot shows that, as *MSRP* increases, we tend to see higher prices for used cars. Also, as *MSRP* increases, there is more *variability* in the prices of used cars.

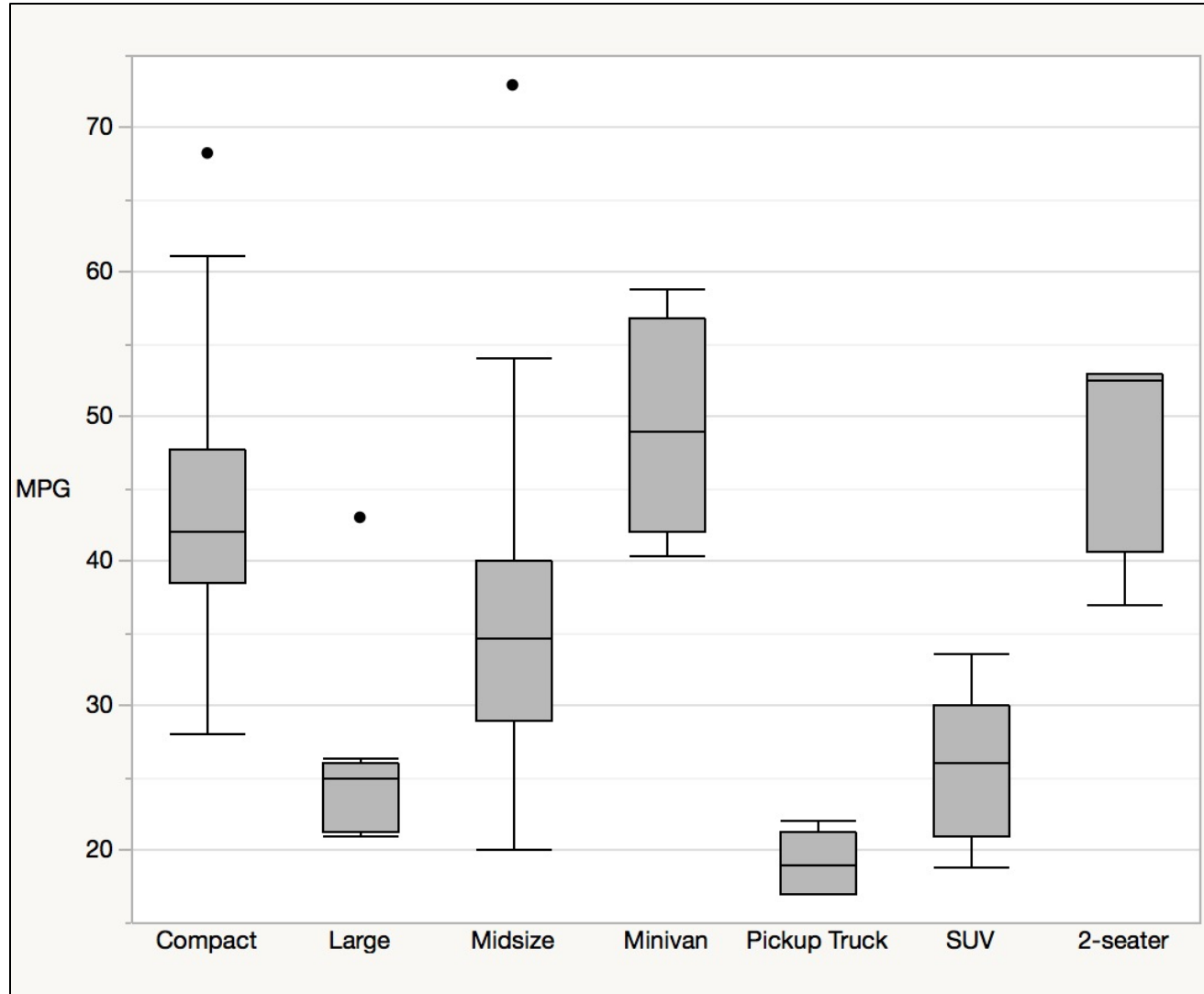
Scatterplot: Example

Facebook: In the 100 or so days before you're likely to start a relationship with someone, the number of interactions between users is expected to rise consistently. Then, right before the relationship begins, there's a free-fall in the number of timeline posts. After the relationship is established, the freefall is followed by a steady decline in the number of wall posts.



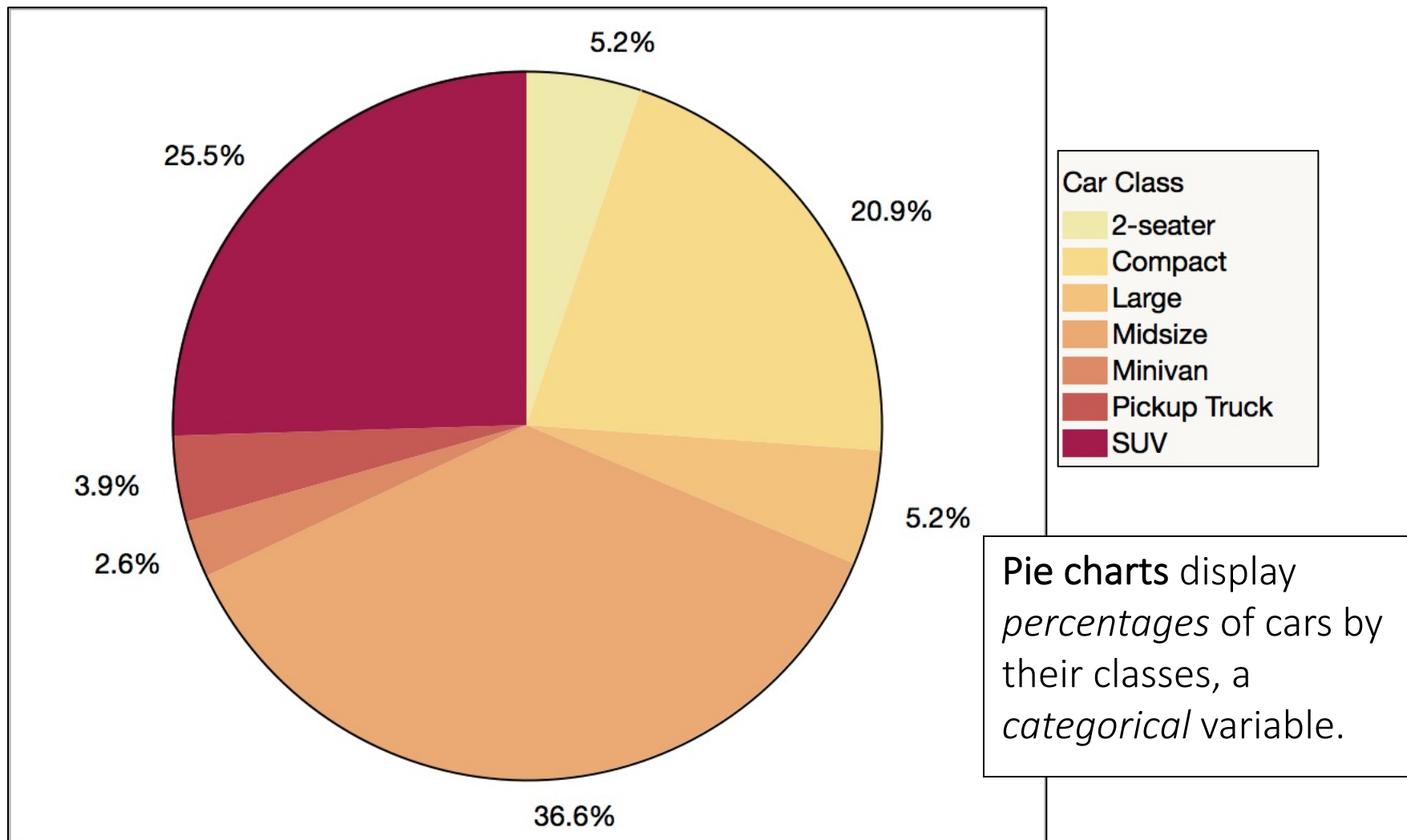
*Chartporn.org

Side-by-Side Boxplots

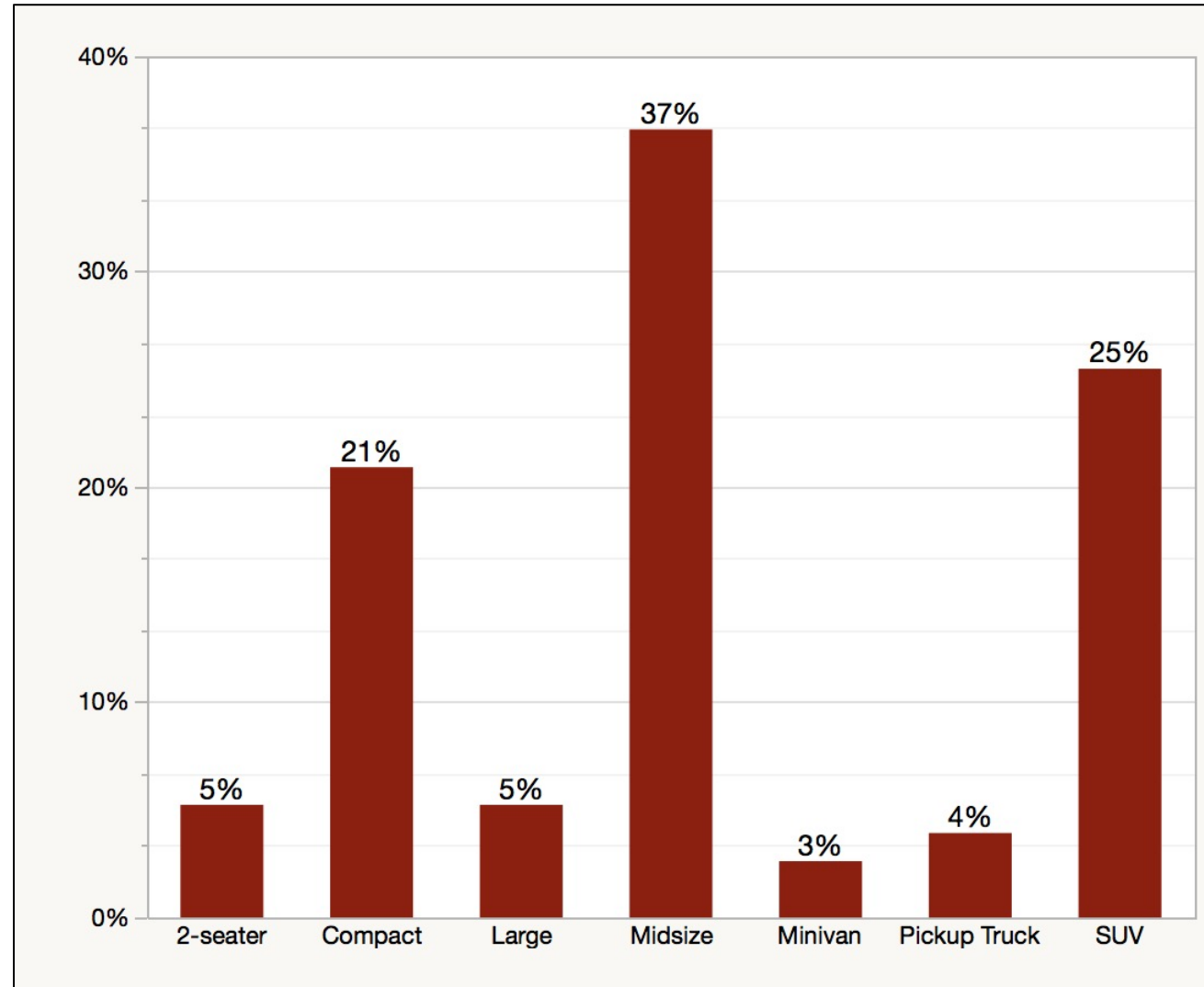


Hybrid Car MPG:

Use side by side boxplots to display relationships between categorical variables (car type) and a quantitative variable (MPG).



Bar charts display *percentages* or *counts* of different cars by their classes, a *categorical* variable.



How do we describe a dataset, especially if it is rather large, without having to present a table of meaningless numbers?

Generally, just two numbers will suffice:

1. Measure of central tendency (i.e. typical value, or location),
2. Measure of dispersion (fluctuation).

Common measures of **central tendency**:

Mean (μ):

Average or expected value

Median (M_d):

Middle point of ordered observations

Mode (M_o):

Most frequent value

The **mean** of a **population** of N measurements x_1, \dots, x_N :

$$\mu = \frac{1}{N} \sum_{i=1}^N x_i = \frac{1}{N} (x_1 + x_2 + \dots + x_N)$$

Eg: Viewing our data set of the *Hybrid Cars' MPG* as a **population**, the population mean is

$$\mu = \frac{1}{153} \sum_{i=1}^{153} x_i = \frac{1}{153} (41.26 + 54.1 + \dots + 37) = 34.7975 \text{ mpg}$$

The **mean** of a **sample** of n measurements x_1, \dots, x_n :

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

Eg: Assessing only the SUV cars from the *Hybrid Car MPG* dataset, of which there are 39 rows, the **sample mean** is

$$\bar{x} = \frac{1}{39} \sum_{i=1}^{39} x_i = \frac{1}{39}(18.82 + 21 \dots + 33.64) = \boxed{26.0077 \text{ mpg}}$$

We can use \bar{x} as an estimate of μ , but we then need to assess the *accuracy* of this and draw conclusions, or *make inferences*, about μ .

Problem: \bar{x} is extremely sensitive to outliers.

- Outliers may be due to errors in recording data
- May be real (but exceptional) observations
- Usually set aside outliers before computing
- Can also use *median*

Whenever a dataset has extreme values, the **median** is the preferred measure of central location.

Given n measurements arranged in order of magnitude,

Median = Middle value if n is odd, or
Average of two middle values if n is even.

Eg: CEO compensation for 5 food processing firms:

Pillsbury	698,000
Borden	1,200,000
Campbell Soup	646,000
Hershey Foods	573,000
Ralston Purina	750,000

Converting to multiples of \$1,000 and arranging in order:

573, 646, 698, 750, 1200

Median compensation is? \$698,000

Mean compensation is? \$773,400

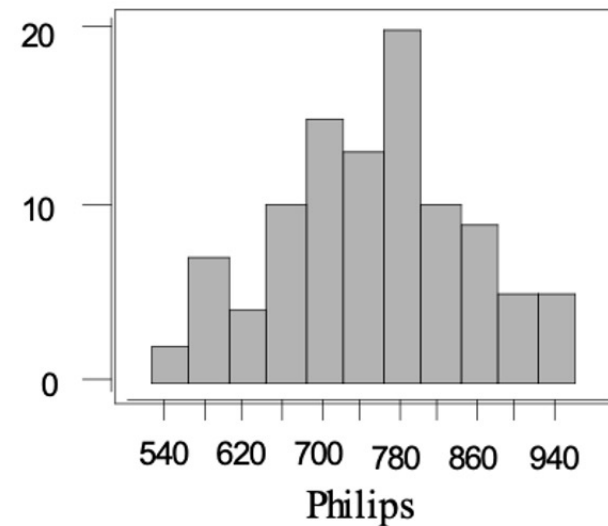
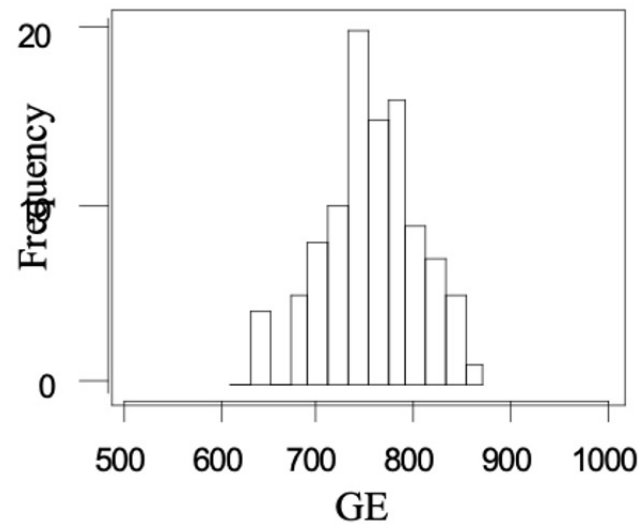
- *Mean > median* because of outlier, Borden.

Removing Borden, $mean = \$666,750 < \$672,000 = median$

- Divides data set into two equal parts
- Half of data lies below median, half lies above it
- Resistant to outliers

Mean and **median** do not completely summarize a dataset... we also need to know how spread out the data is.

Lightbulb Lifetimes (hrs): GE vs Philips



GE has less variability.

- GE exhibits better quality control: not much variation
- Philips has more fluctuation although average is same as GE

Range: Largest minus smallest measurement

- Crude measure with little info about dispersion of values
- No resistance to outliers



Eg: Range of Hybrid Car MPG dataset

- Highest value: 72.92 mpg (Prius Alpha V)
- Lowest value: 17 mpg (Silverado 2WD)

$$\text{Range} = 72.92 \text{ mpg} - 17 \text{ mpg} = 55.92 \text{ mpg}$$

Interquartile Range (IQR)

Interquartile range (IQR): $Q_3 - Q_1 = 75^{\text{th}} \text{ \%ile} - 25^{\text{th}} \text{ \%ile}$

- Width of “middle half” of dataset when ordered from smallest to largest
- Resistant to outliers (robust measure)

*better indicator
than range for spread.*

1st quartile: $Q_1 = 25^{\text{th}}$ percentile, 25% of values lie below
Median of the lower half of the data.

2nd quartile: Q_2 or 50th percentile = Median.

3rd quartile: Q_3 or 75th percentile, 75% of values lie below
Median of the upper half of the data.

Eg: IQR of *Hybrid Car MPG* dataset

- $Q_3 = 41.565$ mpg
- $Q_1 = 26$ mpg

$$\text{IQR} = 41.565 \text{ mpg} - 26 \text{ mpg} = 15.565 \text{ mpg}$$

*⇒ more reliable
than previous
slide*

Interquartile Range (IQR)

20 customer satisfaction ratings:

1 3 5 5 7 8 8 8 8 8 8 9 9 9 9 9 10 10 10 10

*range would be
of 9 which is
useless info
but it's still
good to know
what max*

Find the IQR for customer satisfaction ratings:

$$\text{IQR} = 9 - 7.5 = 1.5$$

What is the 50th percentile?

$$\text{Average of } 8 \text{ \& } 8 = 8$$

What is the 25th percentile?

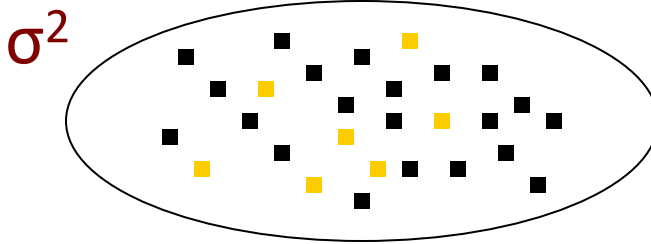
$$\text{Average of } 7 \text{ \& } 8 = 7.5$$

What is the 75th percentile?

$$\text{Average of } 9 \text{ \& } 9 = 9$$

Variance & Standard Deviation

Population X_1, X_2, \dots, X_N



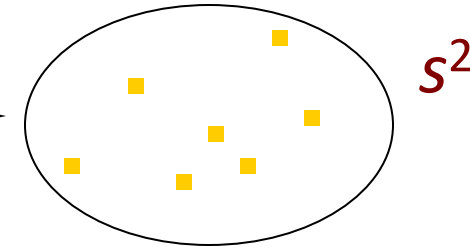
Population Variance:

$$\sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$$

Population Standard Deviation:

$$\sigma = \sqrt{\sigma^2}$$

Sample x_1, x_2, \dots, x_n



Sample Variance:

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

Sample Standard Deviation:

$$s = \sqrt{s^2}$$

*-1 inflates the
variance estimate slightly.
Stdev. P & stdev. S
in excel.*

Variance & Standard Deviation

MPG of 153 Hybrid Cars:

Mean: $\mu = \frac{1}{N} \sum_{i=1}^N x_i = \frac{1}{153}(41.26 + 54.1 + \dots + 37) = 34.8 \text{ mpg}$

Variance: $\sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$
 $= \frac{1}{153}[(41.26 - 34.8)^2 + \dots + (37 - 34.8)^2]$
 $= 120.3958 \text{ mpg}^2$

Standard Deviation: $\sigma = \sqrt{\sigma^2} = 10.9725 \text{ mpg}$

MPG of 39 SUV Hybrid Cars:

Mean: $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i = \frac{1}{39}(18.82 + 21 + \dots + 33.64) = 26 \text{ mpg}$

Variance: $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$
 $= \frac{1}{38}[(18.82 - 26)^2 + \dots + (33.64 - 26)^2]$
 $= 21.149 \text{ mpg}^2$

Standard Deviation: $s = \sqrt{s^2} = 4.599 \text{ mpg}$

Coefficient of Variation

The **coefficient of variation** indicates how large the standard deviation is in relation to the mean and is useful for comparing levels of fluctuation between different variables.

The **coefficient of variation** for a **population** is computed as:

$$c_v = \left[\frac{\sigma}{\mu} \times 100 \right] \%$$

And for a **sample** it is:

$$c_v = \left[\frac{s}{\bar{x}} \times 100 \right] \%$$

way of comparing
2 set of records which
do not have same middle.

MPG of the Hybrid Car dataset:

The **coefficient of variation** for the MPG of the **population** of 153 Hybrid Cars is:

$$c_v = \left[\frac{10.9725}{34.7975} \times 100 \right] \% = \boxed{31.53\%}$$

And the **coefficient of variation** for the MPG of the **sample** of 39 SUV Hybrid Cars is:

$$c_v = \left[\frac{4.5988}{26.0077} \times 100 \right] \% = \boxed{17.68\%}$$

→ The sample of SUV data is relatively less variable than the population.

Comparison of two stocks, Pfizer and Johnson & Johnson:

Monthly adjusted closing PFE and JNJ stock prices (4/1/08 – 3/1/18) had:

	PFE	JNJ
	Adj Close	Adj Close
\bar{x}	22.18	76.67
s	8.36	29.33

The **coefficient of variation** for PFE is: $c_{v,PFE} = \left[\frac{8.36}{22.18} \times 100 \right] \% = 37.68\%$

And the **coefficient of variation** for the JNJ is: $c_{v,JNJ} = \left[\frac{29.33}{76.67} \times 100 \right] \% = 38.25\%$

→ The two stock prices seem to be relatively equally risky!

A normal population with mean μ and standard deviation σ has approximately

68.26% of the population measurements within one standard deviation of the mean:

$$[\mu - \sigma, \quad \mu + \sigma]$$

95.44% of the population measurements within two standard deviations of the mean:

$$[\mu - 2\sigma, \quad \mu + 2\sigma]$$

99.74% of the population measurements within three standard deviations of the mean:

$$[\mu - 3\sigma, \quad \mu + 3\sigma]$$