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# Comparing Students

- Student A takes the SAT and scores 600 on the math component, 100 points higher than the average student.
- Student B takes the ACT and scores 27 on the math component, 9 points higher than the average student.
- Which student performed better?

# How many standard deviations?

	SAT	ACT
Mean	500	18
SD	100	6
Student	600	27

The standard deviation helps compare two different metrics.

- SAT:  $600 - 500 = 100$  (or 1 standard deviation)
- ACT:  $27 - 18 = 9$  (or 1.5 standard deviations)

# The z-Score (or standard score)

$$z = \frac{y - \bar{y}}{s}$$

- The z-score measures the distance of the value from the mean in standard deviations.
- A positive z-score indicates the value is above the mean.
- A negative z-score indicates the value is below the mean.
- A small z-score indicates the value is close to the mean when compared to the rest of the data values.
- A large z-score indicates the value is far from the mean when compared to the rest of the data values.

# z-Scores for our Students

	SAT	ACT
Mean	500	18
SD	100	6
Student	600	27

SAT

$$z = \frac{600 - 500}{100} = 1$$

ACT

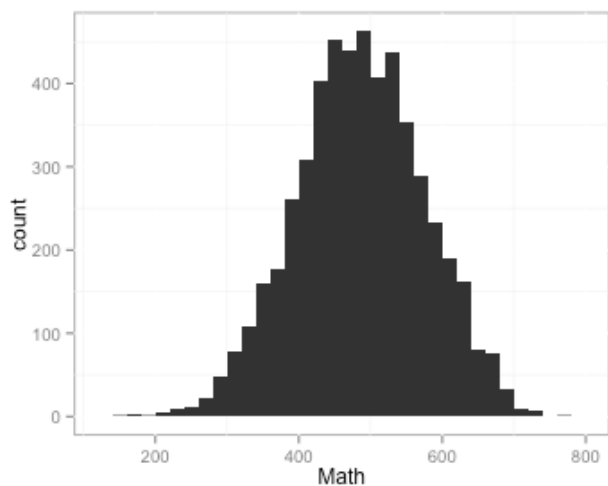
$$z = \frac{27 - 18}{6} = 1.5$$

# Shifting

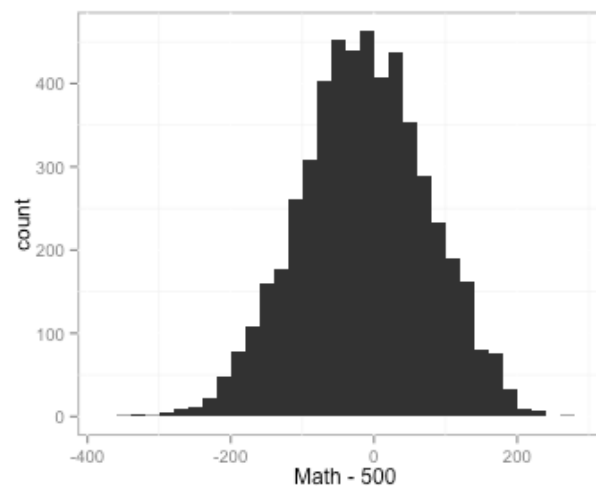
If the same number is subtracted or added to all data values, then:

- The measures of the spread - standard deviation, range, and IQR - are all unaffected.
- The measures of position - mean, median, and mode - are all changed by that number.

```
ggplot(pisausa, aes(x=Math)) +  
geom_histogram(binwidth=20)
```



```
ggplot(pisausa, aes(x=Math - 500)) +  
geom_histogram(binwidth=20)
```

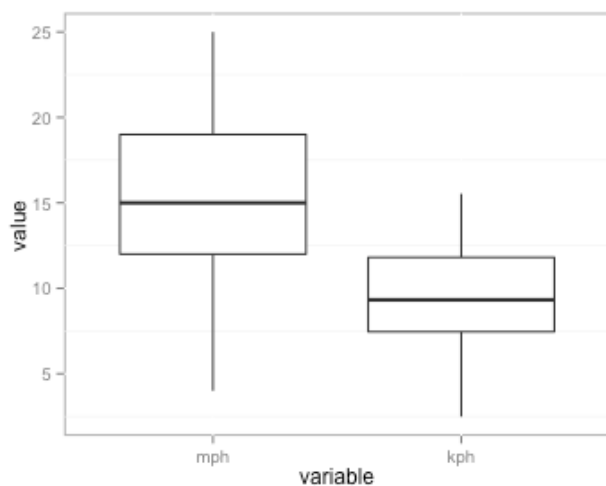


# Scaling

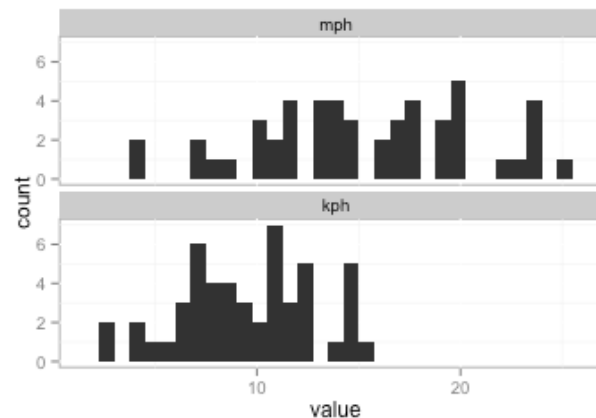
If we multiply all data values by the same number, what happens to the position and spread?

When we multiply (or divide) all the data values by a constant, all measures of position and all measures of spread are multiplied (or divided) by that same constant.

```
ggplot(speeds, aes(x=variable, y=value)) +  
geom_boxplot()
```



```
ggplot(speeds, aes(x=value)) +  
geom_histogram() +  
facet_wrap(~ variable, ncol=1)
```



# Models

“All models are wrong, but some are useful.”

- George Box

- $-1 < z < 1$ : Not uncommon
- $z = \pm 3$ : Rare
- $z = 6$ : Shouts out for attention!



# The Normal Model

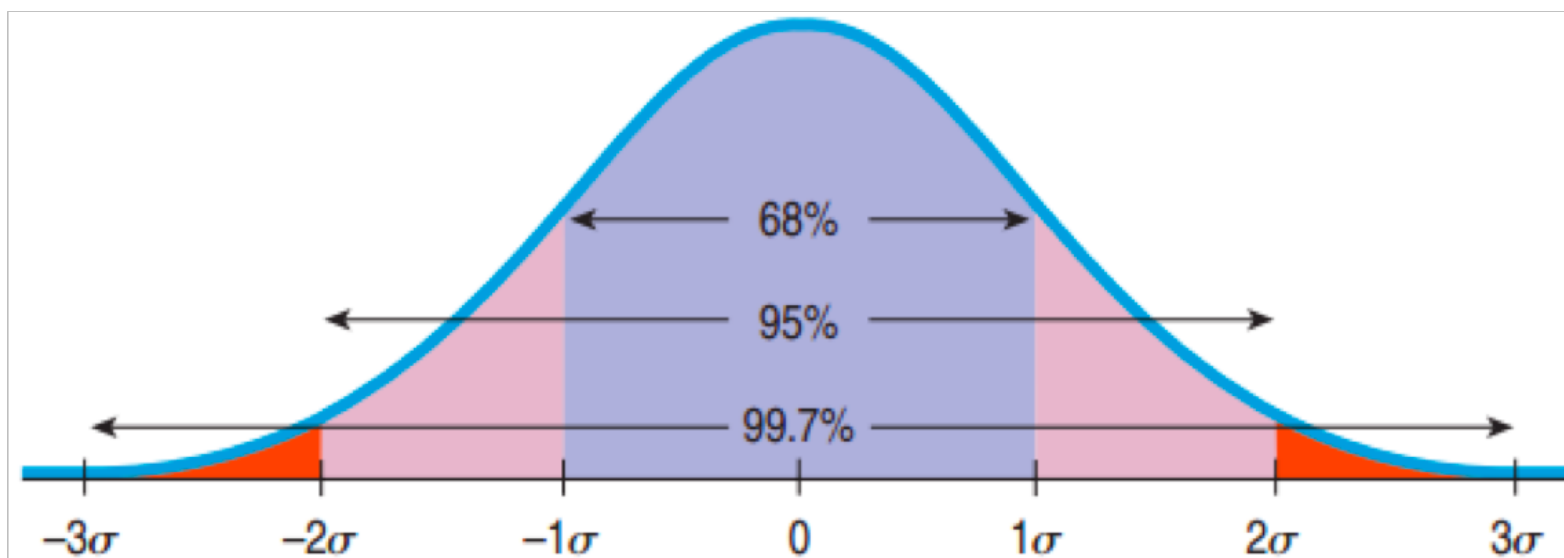
- Bell Shaped: unimodal, symmetric
- A Normal model for every mean and standard deviation.
  - $\mu$  (read “mew”) represents the population mean.
  - $\sigma$  (read “sigma”) represents the population standard deviation.
  - $N(\mu, \sigma)$  represents a Normal model with mean  $m$  and standard deviation  $s$ .

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

- Parameters: Numbers that help specify the model (i.e.  $\mu, \sigma$ )
- Statistics: Numbers that summarize the data (e.g.  $\bar{y}, s$ , median, mode)
- $N(0, 1)$  is called the standard Normal model, or the standard Normal distribution.
- The Normal model should only be used if the data is approximately symmetric and unimodal.

# The 68-95-99.7 Rule

- 68% of the values fall within 1 standard deviation of the mean.
- 95% of the values fall within 2 standard deviations of the mean.
- 99.7% of the values fall within 3 standard deviations of the mean.

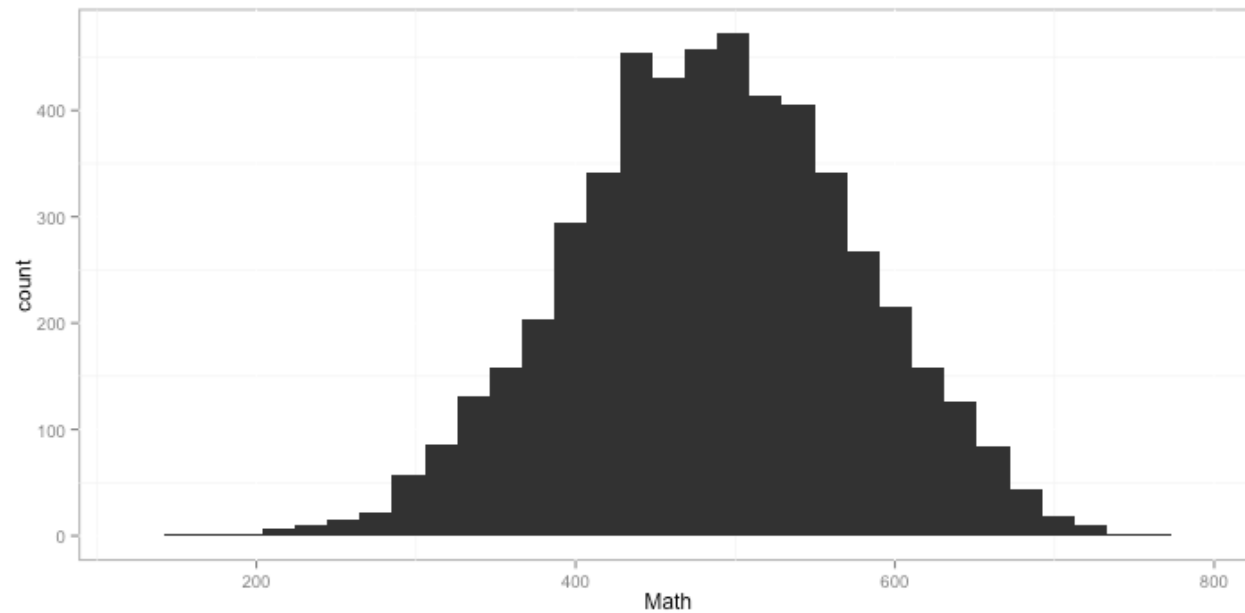


Finding percentiles: <http://shiny.albany.edu/stat/stdnormal/>

# Checking Normality

At a minimum, plot a histogram.

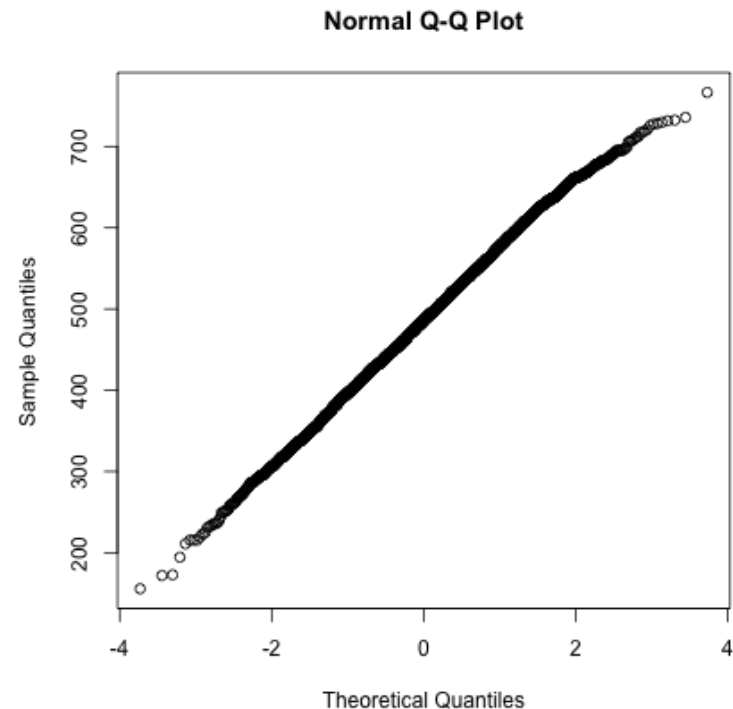
```
ggplot(pisausa, aes(x=Math)) + geom_histogram()
```



# Normal Probability Plots: PISA Math

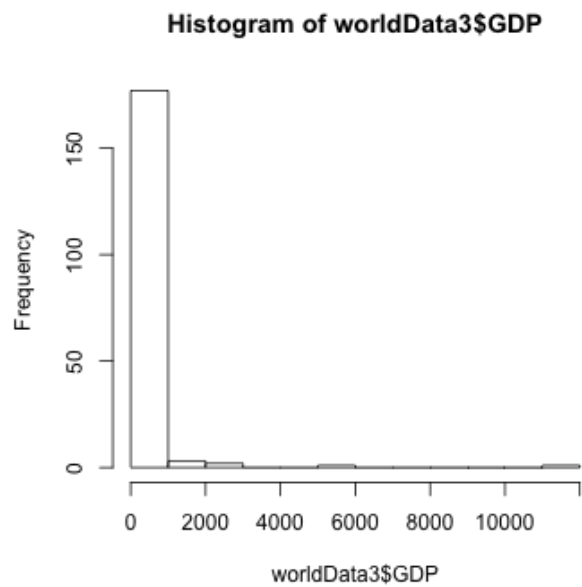
- Plots each value against the z-score that would be expected had the distribution been perfectly normal.
- If the plot shows a line or is nearly straight, then the Normal model works.
- If the plot strays from being a line, then the Normal model is not a good model.

```
qqnorm(pisausa$Math)
```

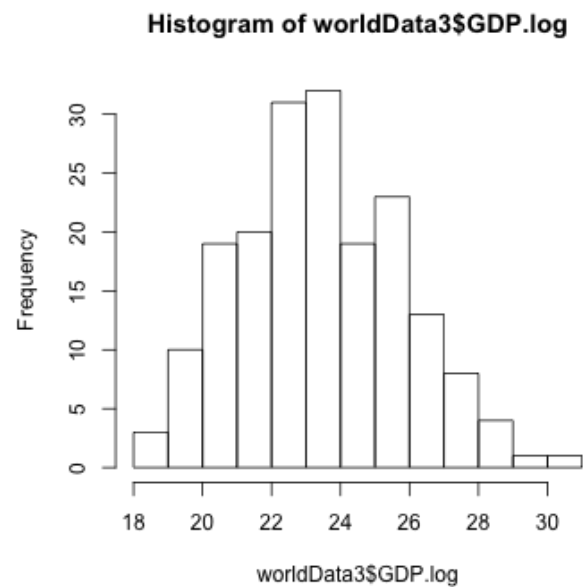


# Histograms: GDP

```
hist(worldData3$GDP)
```

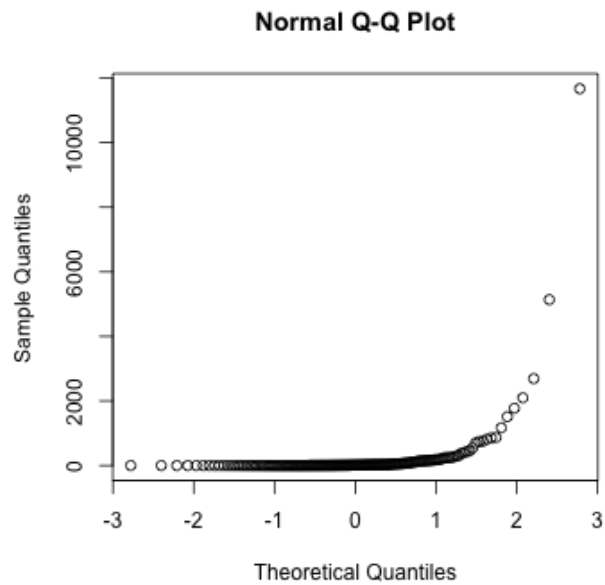


```
hist(worldData3$GDP.log)
```

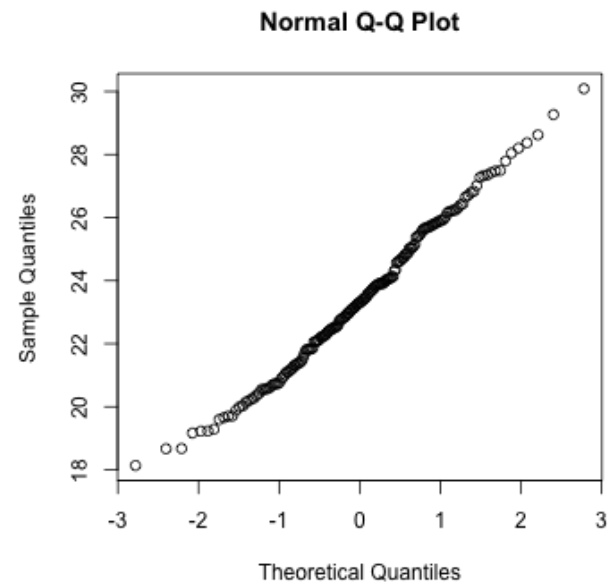


# Normal Probability Plots: GDP

```
qqnorm(worldData3$GDP)
```



```
qqnorm(worldData3$GDP.log)
```



# What can go wrong?

- Don't use the Normal model when the distribution is not unimodal and symmetric.
- Always look at the picture first.
- Don't use the mean and standard deviation when outliers are present.
- Check by making a picture.
- Don't round your results in the middle of the calculation.
- Always wait until the end to round.
- Don't worry about minor differences in results.
- Different rounding can produce slightly different results.