

## 1.4: Data Exploration

Dr. Bean - Stat 5100

### 1 Why Data Exploration

Data Modeling is a lot like:



In order to avoid disaster, you need to **look** before you **jump**.

Example: Consider four scenarios where we use to create a model that uses values of  $x$  to predict values of  $y$ . We make the assumption in each case that the data can be modeled as

$$Y_i = \beta_0 + \beta_1 X_{i,1} + \epsilon_i \quad (1)$$

This assumption means that we assume that  $X$  and  $Y$  share a linear relationship. That is, as  $X$  increases,  $Y$  will increase proportionally. We will explore this further in Handout 2.1.

Data Explorations BEFORE modeling will help us to detect:

- Skewed distributions
- Outlier points
- Non-linear trends

Often, we can use **variable transformations** to get data that are normal, or at least symmetric, in distribution.

Why symmetric data? Consider the “door hinge” problem.

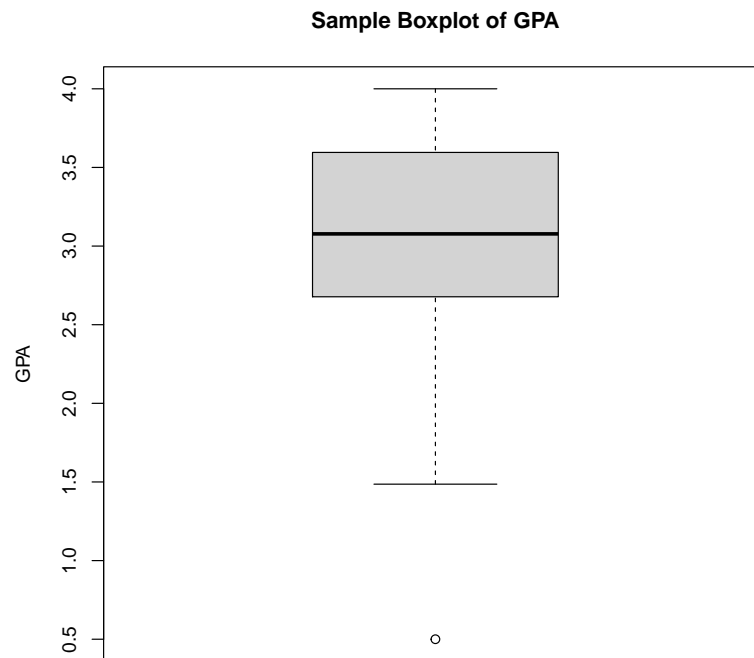
#### Common Exploratory Plots

- **Boxplots::** Show the five quartiles of the data (min, 25th percentile, median, 75th percentile, and maximum).

- Values that are farther than  $1.5 \times \text{IQR}$  (Interquartile Range, which is the 75th percentile minus the 25th percentile) above the 75th percentile or below the 25th percentile are typically plotted as “outlier” points.
- Great way to quickly summarize the range of values.

```
library(stat5100)
data(college)

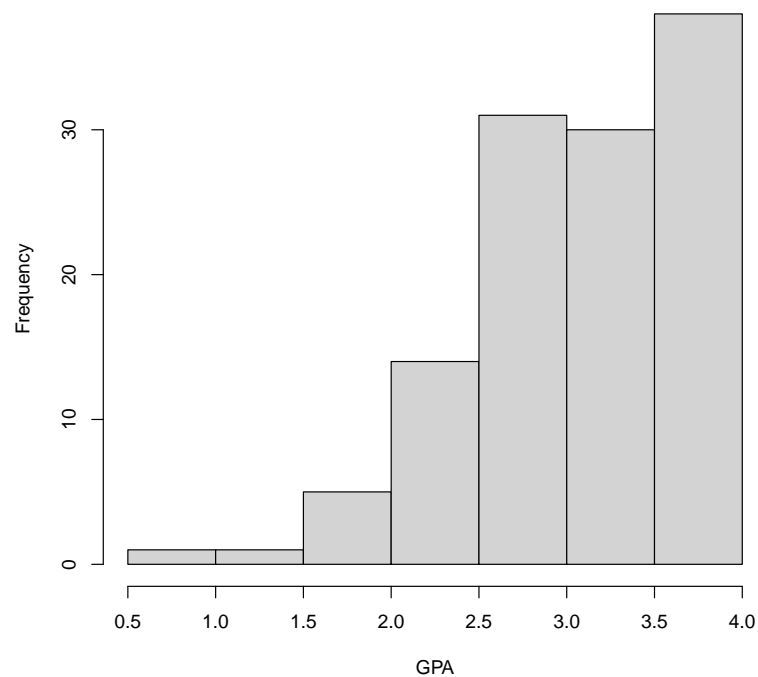
boxplot(college$gpa, main = "Sample Boxplot of GPA", ylab = "GPA")
```



- **Histograms:** Use bins to show the number of observations in a range.
  - Help us to visualize the distribution of the data by imagining a smooth curve running along the top of the bins.
  - Word of caution: the choice of bin width can drastically change the shape of a histogram.

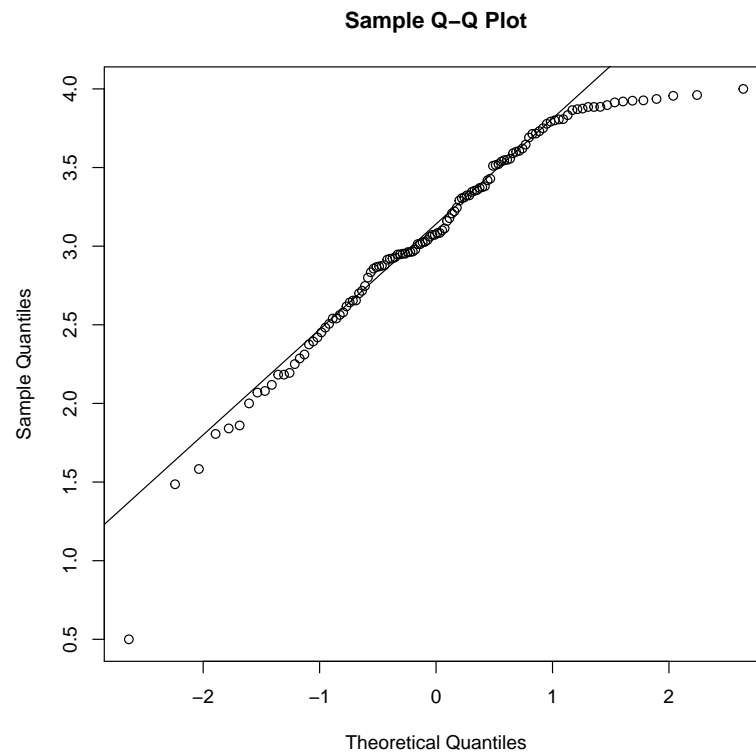
```
hist(college$gpa, main = "Sample Histogram of GPA", xlab = "GPA")
```

Sample Histogram of GPA



- **QQ Plot:** “Quantile Comparison” plots help to easily compare the observed distribution of points to a theoretical (typically normal) distribution.
  - Plots the data quantiles against the theoretical quantiles of similar observations that are normal in distribution.
  - Points that closely follow the diagonal line indicate that the observed data follow the theoretical distribution.
  - While they don’t help to visualize shape, qqplots are superior to histograms as a visual check for normality.

```
qqnorm(college$gpa, main = "Sample Q-Q Plot")  
qqline(college$gpa)
```



*(The above shows that GPA is clearly not normally distributed. If they were, then the data points would lie on the line very nicely.)*

- **Scatterplots:** Plots paired observations from two variables as points on a two-dimensional plot.
  - Excellent way to determine if two variables share a relationship.
  - Can combine in a **scatterplot matrix** when looking at relationships between more than two variables.
  - Subject to **overplotting** when you have thousands of observations that you are trying to plot at the same time.

```
plot(college$act, college$gpa, main = "Sample Scatterplot of GPA vs ACT",  
     xlab = "ACT Score", ylab = "GPA")
```



See **Handout 1.4.2** for an extended example in R of data explorations.