Relationships Between Two Variables: Scatterplots and Correlation

- Example: Consider the population of cars manufactured in the U.S.
 What is the relationship (1) between engine size and horsepower?
 (2) Between engine size and gas mileage?
- Do cars with large engines tend to have high horsepower?
- Do cars with large engines tend to have high gas mileage?
- We will see both graphical and numerical ways to summarize this information about the *relationship between two variables*.

Scatterplots

- A *scatterplot* is a graph that shows the relationship between two *quantitative* variables.
- Each individual in the data set has *two variables* measured on it.
- For each individual, the values of one variable are plotted on the horizontal axis, with the values of the other variable on the vertical axis.
- On the plot, there is a dot for each observation in the data set.
- See example:

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Scatterplot of Horsepower and Vehicle Weight for 32 cars.





Explanatory and Response Variables

- Sometimes one variable (called the *explanatory variable*) may naturally *explain* or *predict* the value of the other variable (called the *response variable*).
- If so, the explanatory variable is denoted *X* and plotted on the X-axis. The response variable is denoted *Y* and plotted on the Y-axis.
- In the engine size / gas mileage example, which is naturally the response variable?
- Other times, there is no natural explanatory-response relationship between the two variables – either one can go on the horizontal axis. (Example: Height/Weight)

Interpreting Scatterplots

- Same process as interpreting other graphs
- Look for overall pattern in the plot and look for deviations from that pattern.
- Describe general pattern (not counting outliers)
- Then look closely at individual outlying values to determine their cause.

Positive and Negative Associations

- A scatterplot can show us the *direction* of the relationship between two variables.
- Two variables have a *positive association* if observations having large values for one variable also tend to have large values for the other variable.
- Also, when variables are positively associated, observations having small values for one variable also tend to have small values for the other variable.
- The scatterplot for such positively associated variables has a pattern that slopes upward from left to right. (Example: Horsepower and Vehicle Weight)

Positive and Negative Associations (Continued)

- Two variables have a *negative association* if observations having large values for one variable tend to have *small* values for the other variable.
- The scatterplot for such negatively associated variables has a pattern that slopes downward from left to right.

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Scatterplot of Gas Mileage and Vehicle Weight for 32 cars.





Form and Strength of Association

- The *form* of the relationship between two variables may be approximately *linear* or *curved*.
- Once we identify the *form* of the relationship, we can characterize the *strength* of the relationship between the two variables.
- The relationship is *strong* if most of the data values *closely* follow the major trend in the plot.
- The relationship is *weak* if the data values show a great deal of *random scatter* around the major trend in the plot.

Form and Strength of Association (Continued)

- In the previous two scatterplots shown, what are the *forms* of the relationships?
- Which of the two scatterplots shows a *stronger* relationship between the two variables plotted?
- Are there any notable outliers in the scatterplots?

What is the best description of the relationship between vehicle weight and gas mileage, based on the scatterplot?

- A. There is a fairly strong, roughly linear, positive association between vehicle weight and gas mileage.
- B. There is a very weak, roughly linear, negative association between vehicle weight and gas mileage.
- C. There is a very weak, curved, positive association between vehicle weight and gas mileage.
- D. There is a fairly strong, roughly linear, negative association between vehicle weight and gas mileage.

Correlation

- Straight-line relationships between two variables are often of interest in data analyses.
- Correlation is a numerical measure of the strength and direction of the linear relationship between two quantitative variables.
- This could give us a more precise measure of the association than a scatterplot.
- Correlation coefficient (denoted *r*) is a number between -1 and 1.
- A positive value of *r* indicates the two variables are *positively* linearly associated.
- A negative value of *r* indicates the two variables are *negatively* linearly associated.

More on Correlation

- The correlation also tells us how *strong* the linear relationship is.
- A value of *r* near -1 or near 1 indicates the two variables are *strongly* linearly associated.
- A value of *r* near 0 indicates the two variables are *weakly* linearly associated.
- See example pictures:

If two variables are strongly negatively linearly associated, what might be a value of *r* for a sample of data on these two variables?

A. *r* = 0.78 B. *r* = 0.13 C. *r* = -0.95

D. *r* = -0.04

Cautions about Correlation

- The correlation does not change if we change the units of measurement for the variables (example: measuring vehicle weight in tons).
- Correlation ignores any distinction between explanatory and response variables.
- Correlation only describes the *linear* association between two variables, not any curved relationship!
- Correlation may be strongly affected (either increased or decreased) by a few outlying observations.

If two variables are very strongly associated, what might be a value of *r* for a sample of data on these two variables? Choose the best answer:

A. *r* = 0.98

B. *r* = 0.03

- C. Impossible to say for sure, but definitely near 1 or near -1.
- D. Impossible to say for sure.

The Effect of Outliers on Correlation

- Recall the scatterplot of Horsepower and Vehicle Weight.
- The Maserati Bora is somewhat outlying with respect to the overall pattern.
- The correlation for the entire data set is 0.659.
- If we delete the Maserati Bora observation, will the correlation be larger or smaller?





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The Effect of Outliers on Correlation

- The correlation for the entire data set is 0.659.
- If we delete the Maserati Bora observation, will the correlation be stronger or weaker?
- If we delete the Maserati Bora observation, the correlation becomes 0.725.
- The Maserati Bora observation had dampened the strength of the linear relationship.
- In other cases, an outlier could *increase* the correlation between the variables (see "Thought Question 3" picture)

Consider the *left graph* in the "Thought Question 3" slide. Suppose the correlation between the two variables is 0.57. What might be the correlation *if we deleted* the solitary outlier?

- A. 0.47
- **B. 0.02**
- C. -0.83

D. 0.74

Consider the *right graph* in the "Thought Question 3" slide. Suppose the correlation between the two variables is 0.83. What might be the correlation *if we deleted* the outlying value?

- A. 0.72
- **B. -0.12**
- **C. -0.64**

D. 0.95

More about Correlation

- The ordinary correlation coefficient can only be calculated when both variables are quantitative.
- It doesn't make sense to talk about a "high correlation between gender and preferred TV network."
- There may be a *relationship* between two categorical variables, but it's not measured by correlation.
- There are other statistics that measure the association between categorical variables, or between a categorical variable and a quantitative variable (we won't cover these).
- *Also note:* To be complete, we should report means and standard deviations for each variable, along with the correlation.