## Bivariate Data



VISUALIZATION
LINEAR CORRELATION
SIMPLE LINEAR REGRESSION RESIDUAL ANALYSIS

# Visualizing Bivariate Data 

ASSESSING ASSOCIATIONS BETWEEN BIVARIATE (i.e. Paired) QUANTITATIVE DATA WITH SCATTER PLOTS


## Defining Linear Correlation

(refer to p. 85-88 in text)

- Linear Correlation

Positive or Negative
Strong or Weak

- No Linear Correlation

No Correlation
Non-linear Relationship

- Notice: Correlation does not imply Causation (p. 89)



## Further Explorations of Bivariate Data

Possible Non-Linear
Relationship

Possible Outliers and/or Influential Points


## Determining Linear Correlation

## (b)

ASSESSING ASSOCIATIONS BETWEEN BIVARIATE (i.e. Paired) QUANTITATIVE DATA WITH PEARSON'S LINEAR CORRELATION COEFFICIENT ( $r$ )

The Linear
Correlation
Coefficient
(refer to p. 90-95 in text)

Properties

- Formula for Calculating $r$ (p. 95):
$r=\frac{S_{x y}}{\sqrt{S_{x x} S_{y y}}}=\frac{n\left(\sum x y\right)-\left(\sum x\right) *\left(\sum y\right)}{\sqrt{n\left(\sum x^{2}\right)-\left(\sum x\right)^{2}} * \sqrt{n\left(\sum y^{2}\right)-\left(\sum y\right)^{2}}}$
- Properties of $r$ (p. 94):
- $-1 \leq r \leq 1$
- If all values of either variable are converted to a different scale, the value of $r$ does not change
- The value of $r$ is not affected by the choice of $x$ or $y$ (Can you see why this is the case?)
- $r$ measures the strength of a linear relationship only!!
- $r$ is very sensitive to outliers in the sense that a single outlier can dramatically affect its value


## Examples (continued)



City MPG Data

|  |  |  |  | Weight | City | 4095 | 18 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | 2839 | 4035 | 18 | 3860 | 18 |
| 5501 | 1923 | 14991 | 2898 | 3315 | 22 |  |  |
| 5945 | 1961 | 14836 | 3123 | 3115 | 22 | 4020 | 17 |
| 6629 | 1979 | 14478 | 3195 | 4115 3650 | 21 | 2875 | 25 22 |
| 7556 | 2030 | 14539 | 3239 | 3650 3565 | 21 | 3915 4205 | 18 |
| 8716 | 2112 | 14395 | 3129 | 4030 | 18 | 4415 | 17 |
| 9369 | 2192 | 14599 | 3100 | 3710 | 19 | 3060 | 26 |
| 9920 | 2235 | 14969 | 3008 | 3710 | 19 | 3060 | 26 |
| 10167 | 2351 | 15107 | 2983 | 3135 | 24 | 3745 | 27 |
| 11084 | 2411 | 14831 | 3069 | 4105 | 17 | 4180 | 17 |
| 12504 | 2475 | 15081 | 3151 | 4170 | 17 | 3235 | 23 |
| 13746 | 2524 | 15127 | 3127 | 3190 | 22 | 3475 | 22 |
| 13656 | 2674 | 15856 | 3179 | 4180 | 17 | 2865 | 24 |
| 13850 | 2833 | 15938 | 3207 | 2760 | 26 | 3600 | 22 |
| 14145 | 2863 | 16081 | 3345 | 3195 | 24 | 2595 | 30 |
|  |  |  |  | 2980 | 24 | 3465 | 22 |
|  |  |  |  |  |  | 3630 | 20 |
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## Simple Linear Regression

## (0)

ASSESSING ASSOCIATIONS BETWEEN BIVARIATE (i.e. Paired) QUANTITATIVE DATA WITH SIMPLE LINEAR REGRESSION

## A Line of "Best Fit"

(refer to p. 100-105)
(11)

- Given a collection of paired sample data, simple linear regression attempts to algebraically describe the relationship between the two variables $x$ and $y$.
- This algebraic description is often denoted as $\hat{y}=b_{0}+b_{1} x$

Does this equation look familiar?
What do you think $b_{1}$ represents? What about $b_{0}$ ?
How would you describe this representation in terms of the relationship that exists between $x$ and $y$ ?

- The graph of the above equation is the Least Squares line

Also known as the "line of best fit" or the "regression line"
The least squares line fits the sample points best
The slope and intercept can be determined using the formulas on p. 103


# Residual Analysis 



ASSESSING ASSOCIATIONS BETWEEN BIVARIATE (i.e. Paired) QUANTITATIVE DATA WITH SIMPLE LINEAR REGRESSION



## Further Investigations

- For the enrollment dataset, we may attempt to fit a quadratic model to account for the slight curvature
- For the city mpg dataset, there appears to be an outlier that may need evaluated more thoroughly
- Both models could be used for prediction purposes, however, when using regression equations for prediction, we must always consider the strength of the linear relationship that exists between the variables and, also, common errors such as extrapolation

Example:
A simple random sample of 30 winning 5 k times for competitive male runners aged $15-24$ years resulted in a mean 5 k time of 16.79 min . The sample linear correlation coefficient between the age of the runner and the 5 k time for this sample was 0.903 . The simple linear regression equation that fits this sample data was found to be $\hat{y}=21.506-0.276 x$ where $x$ represents the age of the runner in years and $y$ represents the 5 k time for the runner in minutes. Can we use the regression equation above to predict the 5 k time for a 65 year old competitive male runner? Why or why not?

## A Complete Simple Linear Regression Analysis


$\checkmark$ General Steps for a Complete Regression Analysis:
$\checkmark$ Construct a scatter plot and verify that the pattern of the points is approximately a straight line pattern without outliers
$\checkmark$ Assess the linear correlation between two variables of interest and create a regression equation and least squares line
$\checkmark$ Plot the least squares line and verify that the fitting is appropriate
$\checkmark$ Construct a residual plot and verify that there is no pattern (other than a straight line pattern) and also verify that the residual plot does not become thicker or thinner
$\checkmark$ Use a histogram to confirm that the values of the residuals have a distribution that is approximately normal
$\checkmark$ Consider any effects of a pattern over time


