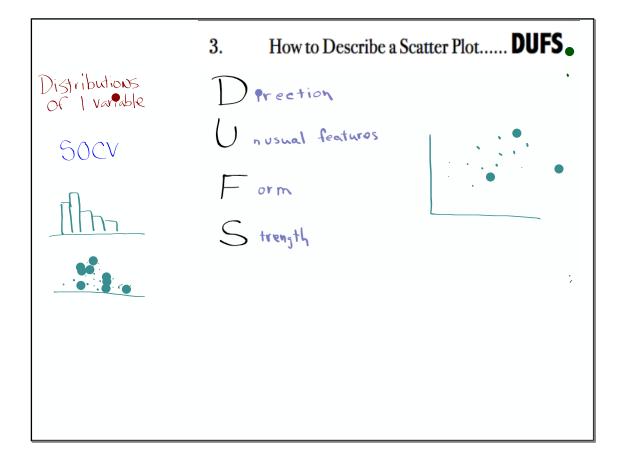
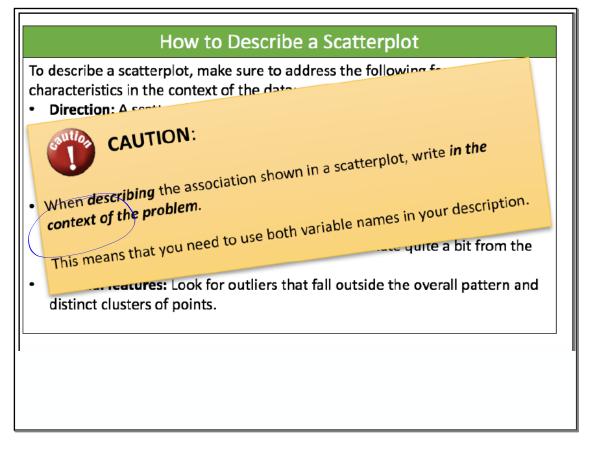
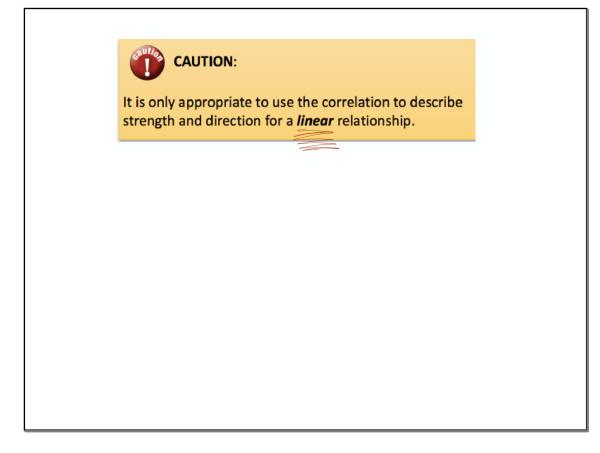
Reminders of some Key ideas and points

before tomorrow's Ch. 3 Test

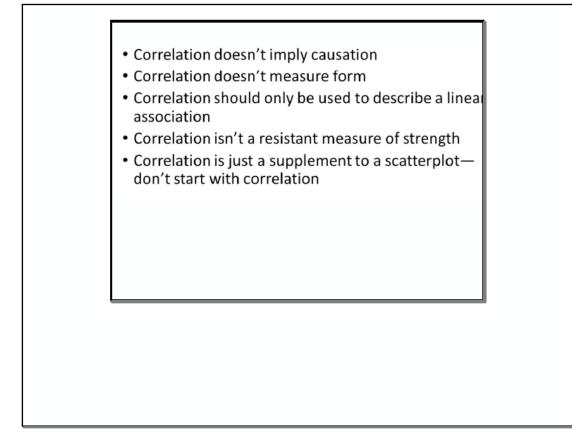


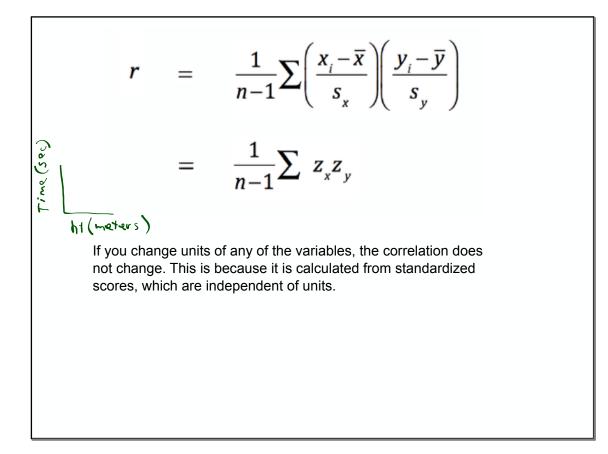


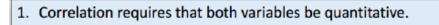
For a linear association between two quantitative variables, the correlation r measures the direction and strength of the association. direction Strength



The correlation of r=-.927 confirms that the linear association between years since 1900 and 100-meter record time is strong and negative T Strength direction

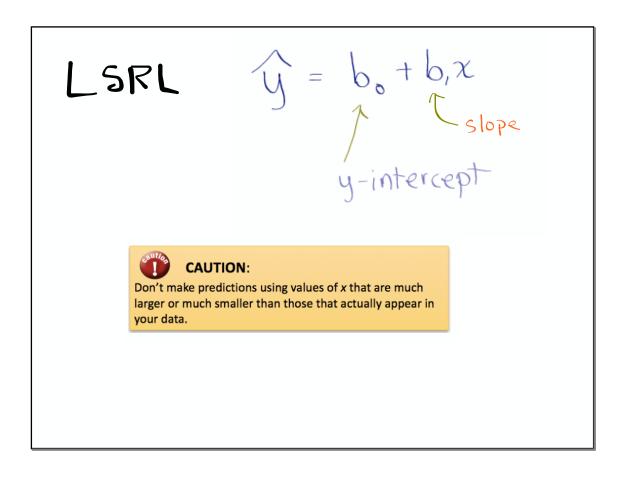


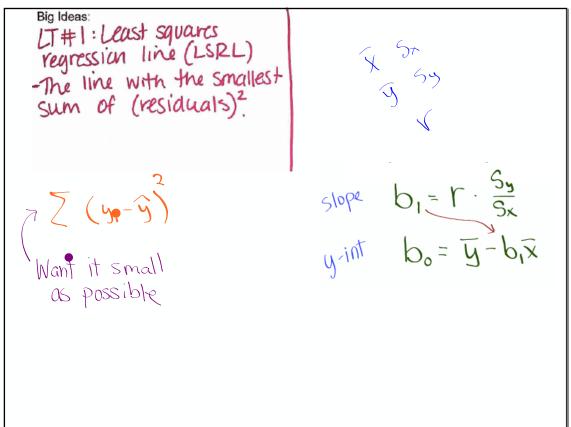


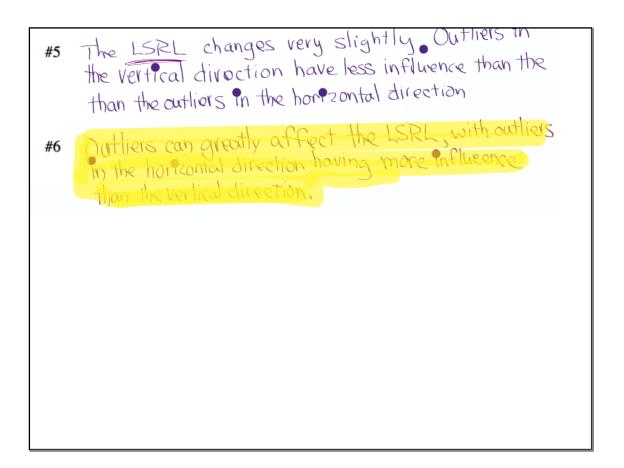


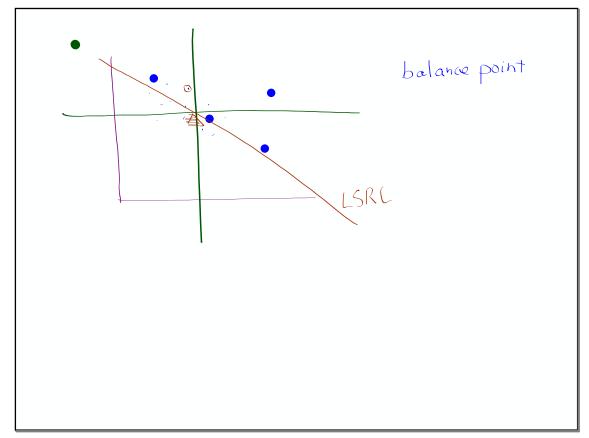
- 2. Correlation makes no distinction between explanatory and response variables.
- 3. *r* does not change when we change the units of measurement of *x*, *y*, or both.

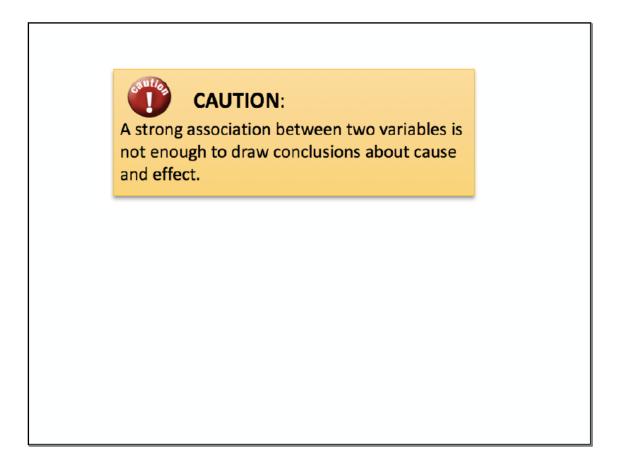
4. The correlation r has no unit of measurement. It's just a number.

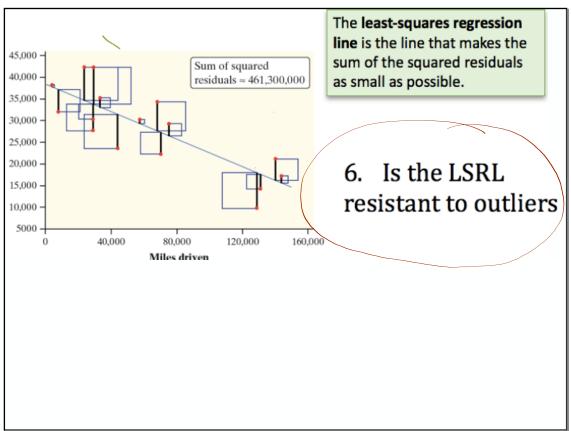


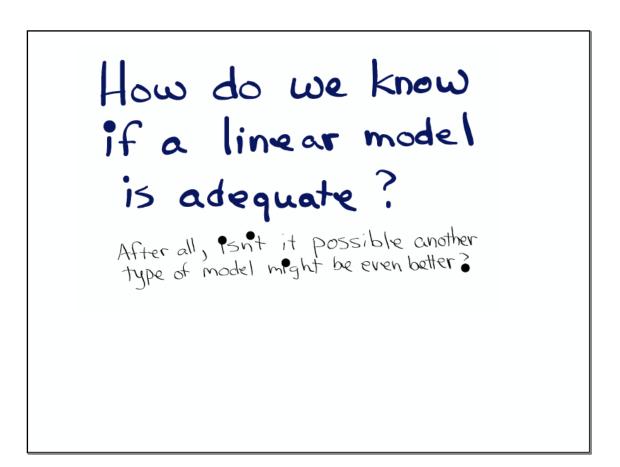








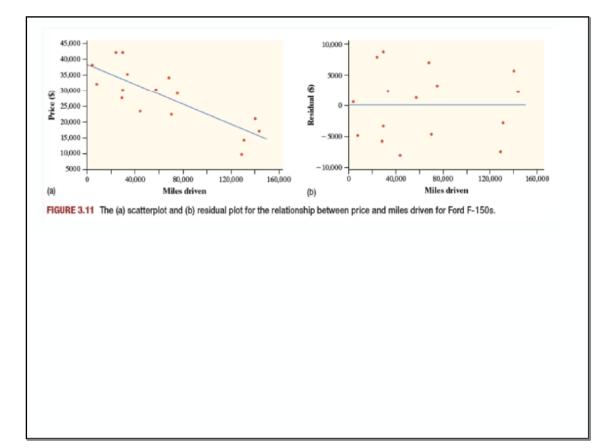




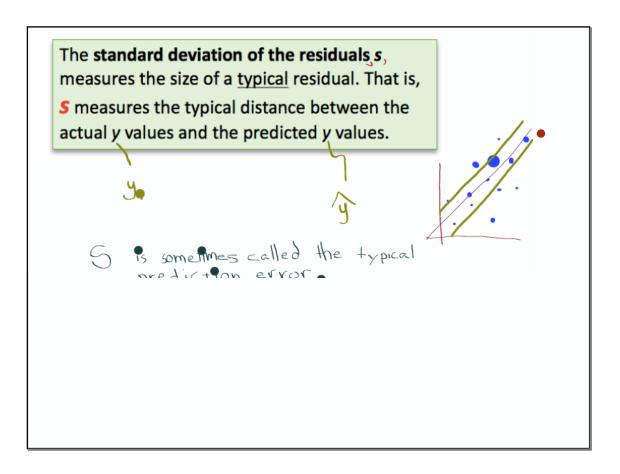
Determining if a Linear Model Is Appropriate: Residual Plots

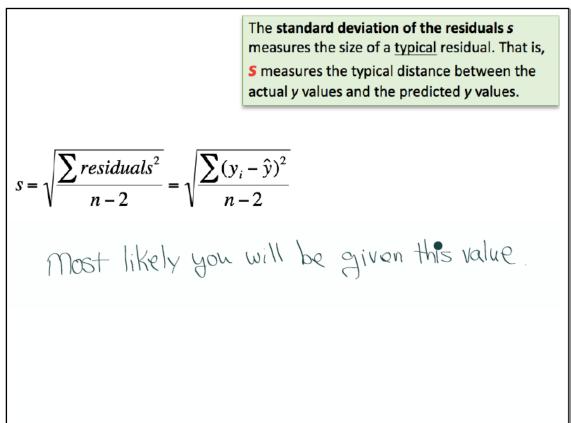
A residual plot magnifies the deviations of the points from the line, making it easier to see unusual observations and patterns. If a regression model is appropriate:

- The residual plot should show no obvious patterns.
- The residuals should be relatively small in size.



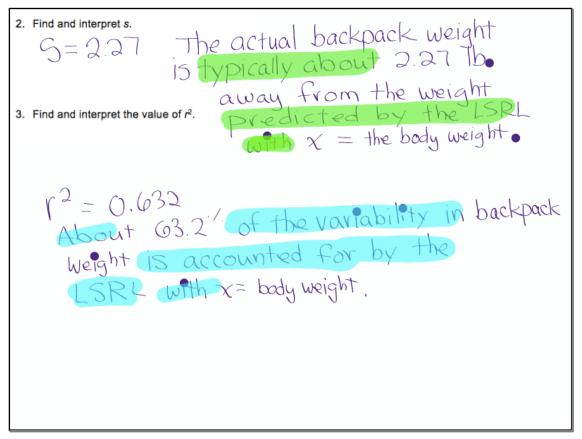
If so, we can use S and r² to determine how good the predictions will be (How well does the line work) Standard Deviations of the residuals r² Coefficient of Determination

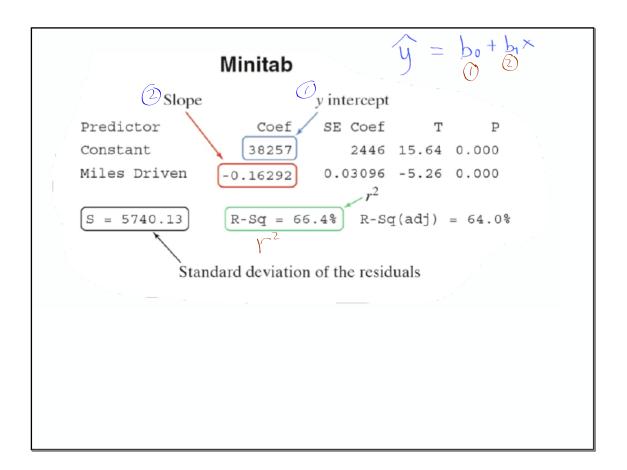


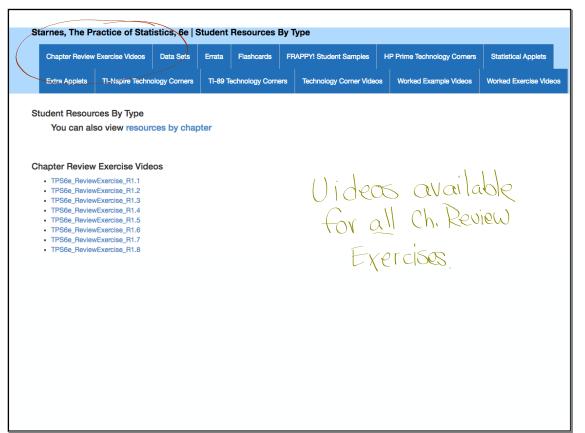


Coefficient of Determination

 r^2 measures the fraction of the variability in the y variable that is accounted for by the LSRL using x.







	Formulas
4	(I) Descriptive Statistics
	$\overline{x} = \frac{\sum x_i}{n}$
	$s_{x} = \sqrt{\frac{1}{n-1} \Sigma \left(x_{i} - \overline{x}\right)^{\Sigma}}^{'}.$
	$s_{P} = \sqrt{\frac{(n_{1}-1)s_{1}^{2} + (n_{2}-1)s_{2}^{2}}{(n_{1}-1) + (n_{2}-1)}}$
	$\hat{y} = b_0 + b_1 x$
	$b_{1} = \frac{\sum (x_{i} - \overline{x})(y_{i} - \overline{y})}{\sum (x_{i} - \overline{x})^{2}}$
	$b_0 = \overline{y} - b_1 \overline{x}$
	$r = \frac{1}{n-1} \sum \left(\frac{x_i - \overline{x}}{s_x} \right) \left(\frac{y_i - \overline{y}}{s_y} \right)$
	$b_1 = r \frac{s_y}{s_x}$
	$s_{b_1} = \frac{\sqrt{\frac{\sum (y_i - \hat{y}_i)^2}{n-2}}}{\sqrt{\sum (x_i - \overline{x})^2}}$

$$\overline{x} = \frac{\sum x_i}{n}$$

$$s_x = \sqrt{\frac{1}{n-1} \sum (x_i - \overline{x})^2}$$

$$s_p = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{(n_1 - 1) + (n_2 - 1)}}$$

$$\hat{y} = b_0 + b_1 x$$

$$b_{0} = \overline{y} - b_{1}\overline{x}$$

$$r = \frac{1}{n-1}\Sigma\left(\frac{x_{i} - \overline{x}}{s_{x}}\right)\left(\frac{y_{i} - \overline{y}}{s_{y}}\right)$$

$$b_{1} = r\frac{s_{y}}{s_{x}}$$

