

Department of Linguistics and Translation

香港城市大學 City University of Hong Kong

Fundamentals of Statistics for Language Sciences LT2206



Jixing Li Lecture 8: Simple Linear Regression

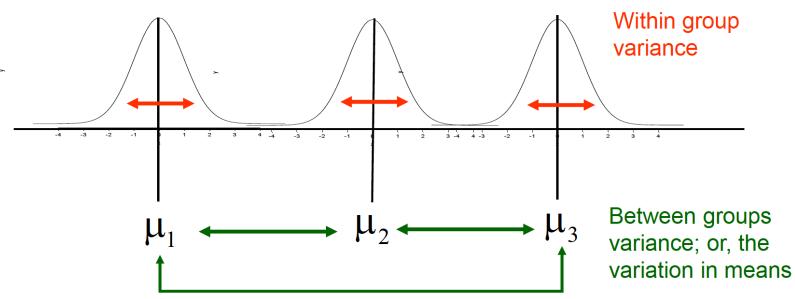
One-way ANOVA

1. Compute sum of squares between group (SSB):

2. Compute sum of squares within group / error (SSE): sum of the squared differences between each individual observation and the group mean of that observation.

 $\sum n_k (\overline{X_k} - \overline{X})^2$

3. Compute sum of squares total (SST): SSB+SSE



The ANOVA table

Source of variation	Degrees of freedom	Sum of squares		F statis	tic
Between	k-1	SSB	MSB =SSB/(k-1)	MSB/MS	E
Within (Error)	n-k	SSE	MSE =SSE/(n-k)		
Total	n-1	SST			

k: number of groups*n*: total number of samples

Under H_0 , F should tend to be close to 1. Under H_a , F should exceed 1, by an amount depending on both n and k.

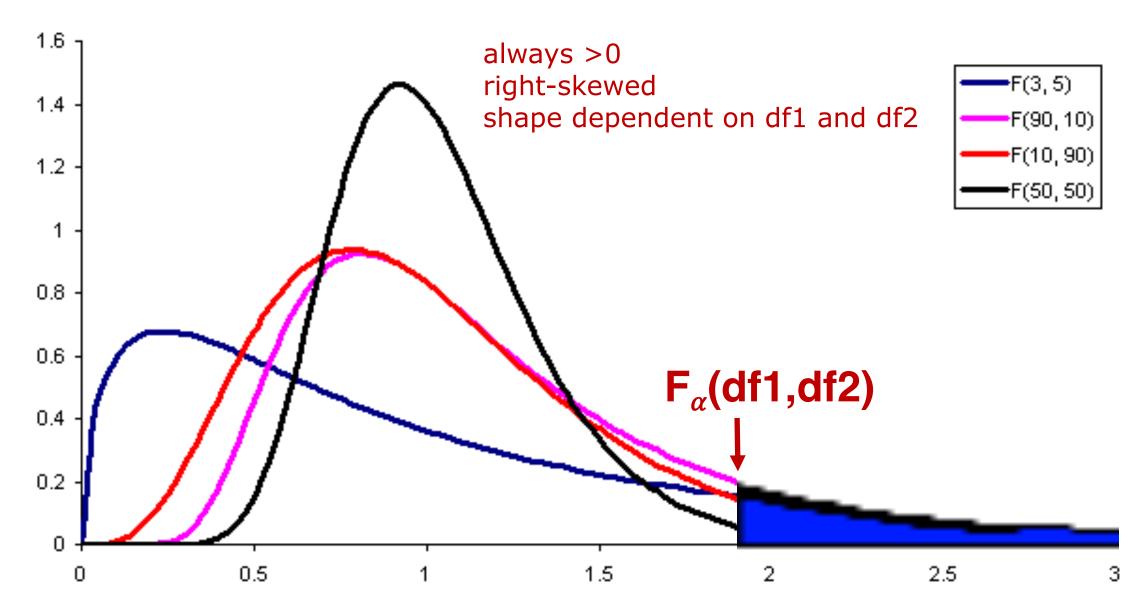
Example: The cocktail party experiment

group	mixed	mean	grand mean
hearing-impaired	2,2,3	2.33	
normal	2,4,4	3.33	2.78
children	2,3,3	2.67	

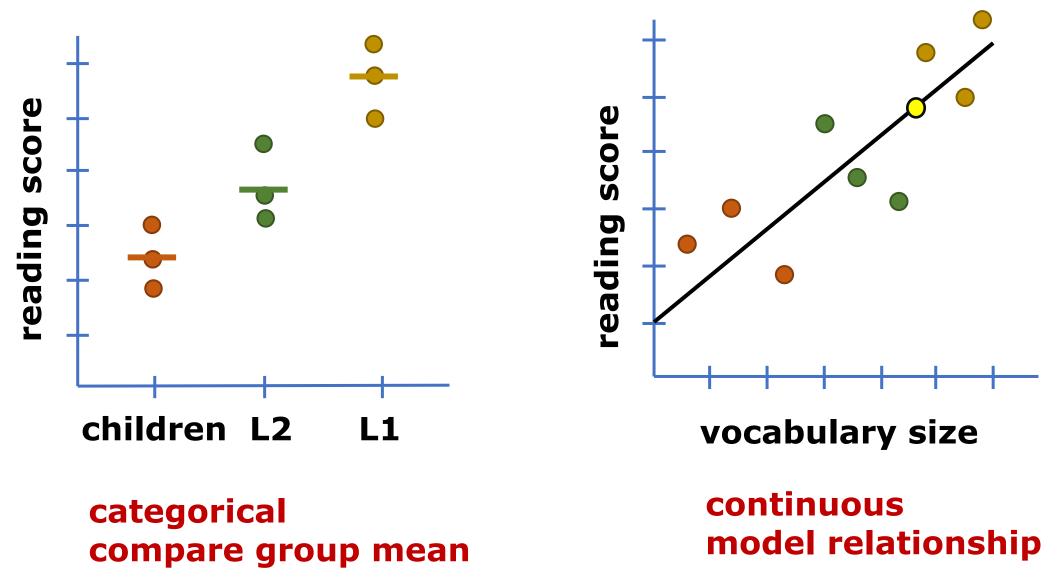
SSB =
$$3*(2.33-2.78)^2 + 3*(3.33-2.78)^2 + 3*(2.67-2.78)^2 = 2.77$$

SSE = $(2-2.33)^2 + (2-2.33)^2 + (3-2.33)^2 + (2-3.33)^2 + (4-3.33)^2 + (4-3.33)^2 + (2-2.67)^2 + (3-2.67)^2 = 4$
k = 3, **n** = 9
MSB= **SSB** / (k-1) = $2.77/2 = 1.385$
MSE = **SSE** / (n-k) = 0.67
F = **MSB/MSE** = 2.07

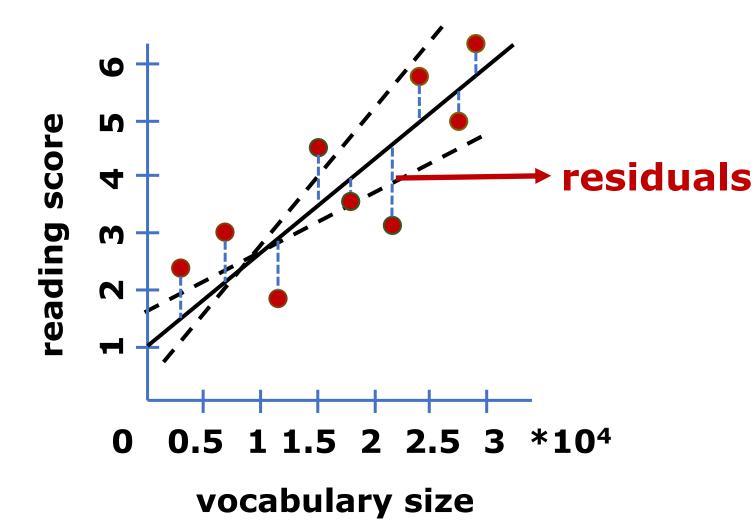
F-distribution



ANOVA vs. Regression



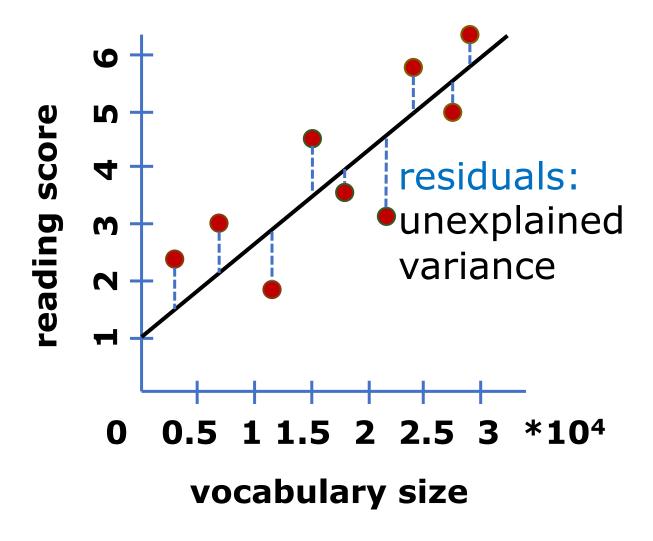
Fit the regression line



The best fit regression line minimizes the sum of squared residuals

→ least-squares regression line

Interpreting regression model

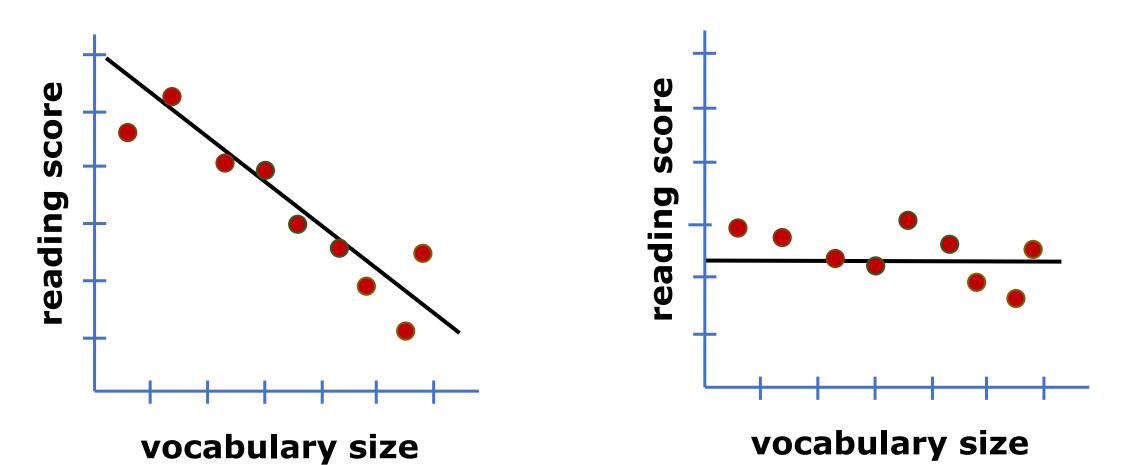


$$y = 2x + 1 \qquad y = b_1 x + b_0$$

slope: vocabulary size
increase by 10,000,
reading score increases
by 2 points on average.

intercept: the expected y
when x=0, may or may
not make sense

Interpreting regression model



negative b1: vocabulary size increases, reading score decreases

b1 close to 0: vocabulary size does not influence reading score

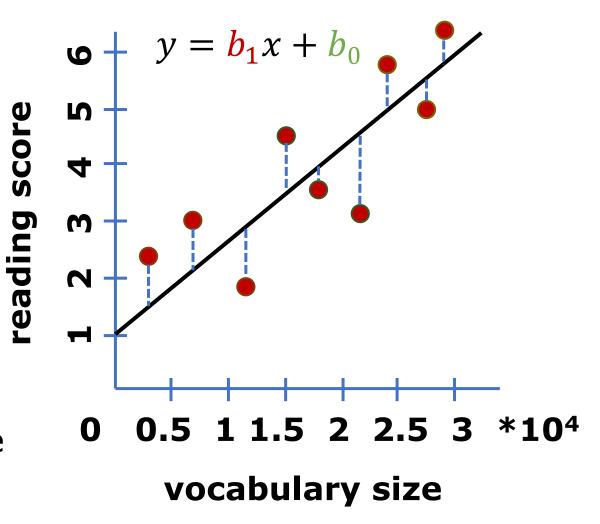
Estimating regression coefficients

Solution (calculus):

$$\mathbf{b_1} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$

$$\mathbf{b_0} = \bar{y} - b_1 \bar{x}$$

 x_i = value of independent variable y_i = value of dependent variable \bar{x} = mean of the independent variable \bar{y} = mean of the dependent variable

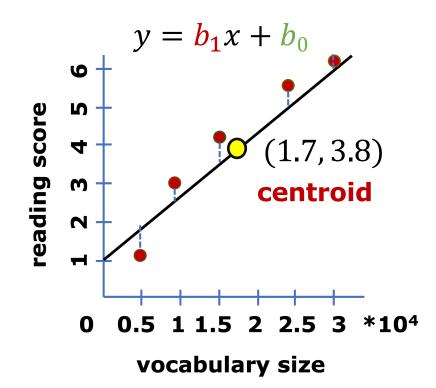


Estimating regression coefficients

$$\mathbf{b_1} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$

$$\mathbf{b_0} = \bar{y} - b_1 \bar{x}$$

reading score	1,3,4,5,6 (M=3.8)		
vocabulary size	0.5,1,1.5,2.5,3 (M=1.7)		

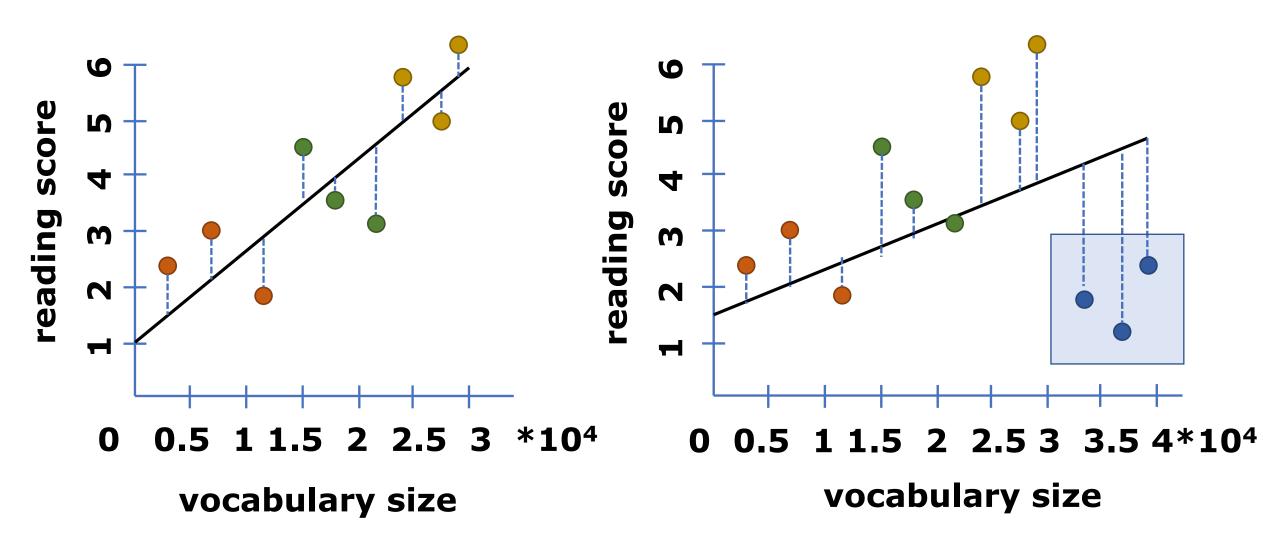


$$\mathbf{b_1} = \frac{(0.5 - 1.7)(1 - 3.8) + (1 - 1.7)(3 - 3.8) + (1.5 - 1.7)(4 - 3.8) + (2.5 - 1.7)(5 - 3.8) + (3 - 1.7)(6 - 3.8)}{(0.5 - 1.7)^2 + (1 - 1.7)^2 + (1.5 - 1.7)^2 + (2.5 - 1.7)^2 + (3 - 1.7)^2}$$

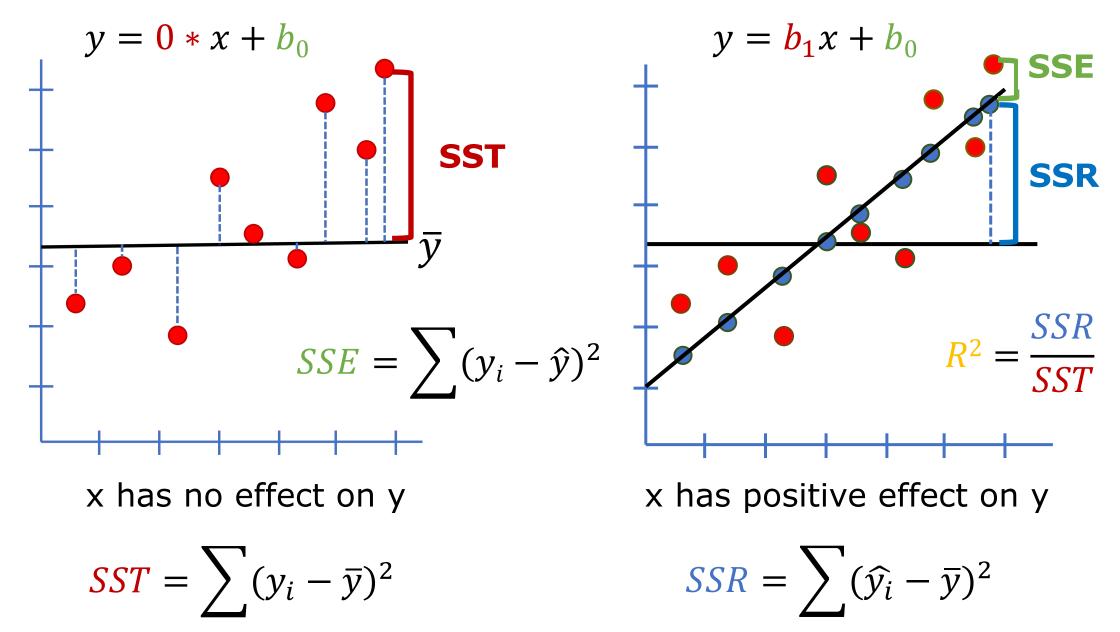
= 1.53

 $b_0 = 3.8 - 1.53 * 1.7 = 1.199$

Goodness of fit



A tale of two models



R² and F ratio

$$SSR = \sum (\hat{y_i} - \bar{y})^2 \qquad MSR = \frac{SSR}{k - 1} \qquad F = \frac{MSR}{MSE}$$
$$SSE = \sum (y_i - \hat{y})^2 \qquad MSE = \frac{SSE}{n - 2}$$

$$SST = \sum (y_i - \bar{y})^2$$

 $\frac{R^2}{SST} = \frac{SSR}{SST}$

k: number of model parameters (slope, intercept) *n*: number of data points

Example: Alice dataset

Participants listened to the first chapter of *Alice in Wonderland* in the fMRI scanner.

Question: How is the frequency of each word in the story affect the brain activity from 4 brain regions?

