

Multiple regression:
fitting the best fitting plane

$$\hat{Y} = a + b_1 X_1 + b_2 X_2$$

Where, a = intercept (when $X_1=0$ and $X_2=0$)

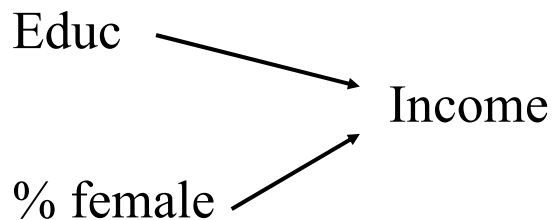
Y = income

X_1 = education

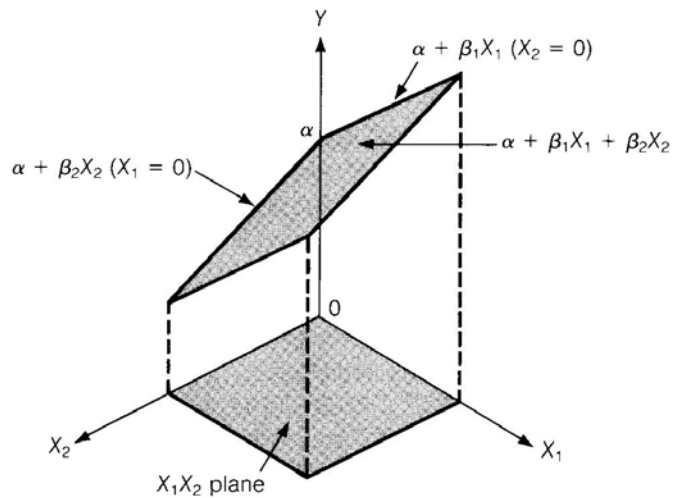
X_2 = % female in occupation

[unit of analysis = occupation]

Multiple regression: model

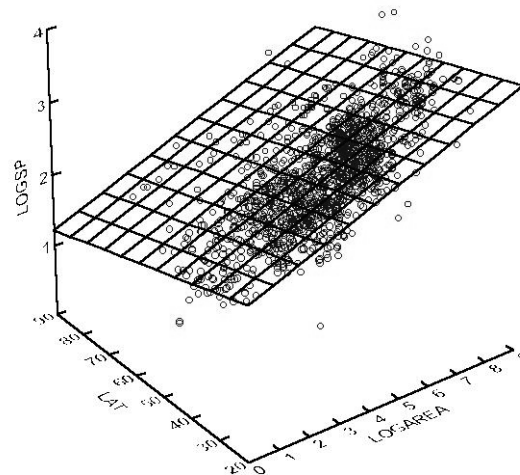


Multiple regression: 3D model



Source: Agresti and Finlay, 1986, p. 317

Multiple regression: 3D model



Multiple regression: interpretation of b's

- b_1 = slope for education (net effect of education on income controlling for percent female; how much income in dollars for each year of education)
- b_2 = slope for % female (net effect of percent female on income controlling for education; how much income in dollars for each percent female)

Multiple regression: interpretation of b's

Coefficients called:

- Metric coefficient
- Net regression coefficient
- Unstandardized regression coefficient

Interpretation:

- Net effect
- Independent effect
- Controlling for

Multiple regression:
standardized regression coefficients

$$\hat{Y} = a + b_1X_1 + b_2X_2 \quad \Rightarrow$$

$$\hat{y} = B_1x_1 + B_2x_2$$

where, $B_{yx} = b_{yx} (s_x/s_y)$

Creating standard
scores:

$$x_1 = (X_1 - \bar{X}_1)/s_{X_1}$$

Multiple regression:
standardized regression coefficients

- B's range from -1 to +1
- Interpretation: a one s.d. change in the independent variable, produces a predicted change of "Beta" s.d.'s in the dependent variable, net of other variables
- More common interpretation: if $B_1 > B_2$ then education is much more important in predicting income than is % female
- "considerably larger, more than twice as important"

Murdock data:

Y = stratification, X₁=political integration,
X₂=money exchange

<u>Society</u>	<u>X₁</u>	<u>X₂</u>	<u>Y</u>	<u>Society</u>	<u>X₁</u>	<u>X₂</u>	<u>Y</u>
001	2	0	1	011	1	1	1
003	3	2	2	013	0	0	0
005	3	3	3	015	2	0	1
007	4	0	2	017	2	3	1
009	0	0	0	019	4	3	2

Calculating standardized and
unstandardized coefficients

Computational formula:

$$r = \frac{n \sum X_i Y_i - (\sum X_i)(\sum Y_i)}{\sqrt{[n \sum X_i^2 - (\sum X_i)^2] [n \sum Y_i^2 - (\sum Y_i)^2]}}$$

Calculating standardized and unstandardized coefficients

- $r_{yx1} = .865$
- $r_{yx2} = .620$
- $r_{x1x2} = .482$

Calculating standardized coefficients

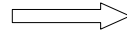
$$B_{yx.z} = \frac{r_{yx} - r_{yz}r_{xz}}{1 - r_{xz}^2}$$

[General formula]

Calculating standardized coefficients

- $B_{y_{x_1}.x_2} = .737$

- $B_{y_{x_2}.x_1} = .265$



$$\hat{y} = .737x_1 + .265x_2$$

Interpretation: political integration is substantially more important than money in determining level of stratification

Calculating unstandardized coefficients

$b_{yx} = B_{yx} (s_y/s_x)$, using s.d.
computational formula:

$$s_y = \sqrt{\frac{\sum Y^2}{n} - \left(\frac{\sum Y}{n}\right)^2}$$

$$\bar{Y}$$

Calculating unstandardized coefficients

Calculate s.d.'s:

- $s_y = .9$
- $s_{x1} = 1.37$
- $s_{x2} = 1.33$

Calculate b's:

$$\begin{aligned}b_{yx1.x2} &= B_{yx1.x2} (s_y/s_{x1}) \\ &= (.737) (.9/1.37) = .484\end{aligned}$$

$$\begin{aligned}b_{yx2.x1} &= B_{yx2.x1} (s_y/s_{x2}) \\ &= (.265) (.9/1.33) = .179\end{aligned}$$

Calculating intercept

$$\bar{Y} = a + b_1 \bar{X}_1 + b_2 \bar{X}_2$$

$$a = \bar{Y} - b_1 \bar{X}_1 - b_2 \bar{X}_2$$

$$a = (1.3) - (.484)(2.1) - (.179)(1.2) = .0688$$

$$a = .069$$

Prediction equation

$$\hat{Y} = .069 + .484X_1 + .179X_2$$

Interpret!

Calculating R^2

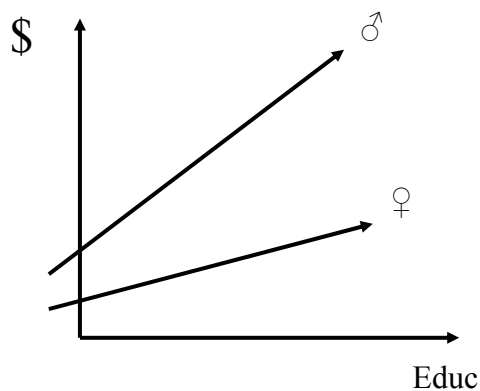
$$\begin{aligned} R^2_{y.x1x2} &= B_{yx1}r_{yx1} + B_{yx2}r_{yx2} \\ &= (.737)(.865) + (.265)(.620) \\ &= .802 \end{aligned}$$

Interpretation: 80 percent of the variation in stratification is explained by political integration and money

Standardized vs. unstandardized coefficients

- Use standardized to compare variables *within* equations
- Use unstandardized to compare same variable *across* equations

Unstandardized coefficients



Caveats

- 1) Don't interpret regression lines beyond where you have data
- 2) Report to three significant nonzero digits (retain larger # of digits in intermediate calculations)
- 3) Multicollinearity: problem with high correlations