

$$y = \beta_0 + \beta_1 x + \epsilon$$

↓ ↘
 dependent Independent
 Variable Variable

dependent
Variable

β_0 = intercept (Value of y when $x=0$)

β_1 = Slope
(change in y for unit change in x)

Assumptions

1. y/x is normally distributed
↳ y given x is normally distributed
2. Observations need to be independent.
3. $E(\epsilon) = 0$
4. ϵ and x (Independent Variable)
is not correlated.

Goodness of fit

R^2 (R-square) (or) Coefficient
of Determination.

$$R^2 = \frac{SSR}{SST} = \frac{\text{Sum of Squared Residuals}}{\text{Sum of Squared total.}}$$

= total variation explained by the model.

$$0 \leq R^2 \leq 1 \text{ (or) } 0 \leq R^2 \leq 100\%.$$

$$\text{Wage} = \beta_0 + \beta_1(\text{educ}) + \epsilon$$

Wage = annual salary

educ = # of years of education.

β_1 = effect of education on wage
while all other factors are
held constant

ϵ — include all other factors
that affect wage

$$\text{Wage} = \beta_0 + \beta_1(\text{educ}) + \beta_2(\text{experience}) + \beta_3(\text{Skills}) + \epsilon$$

β_0 — intercept (wage when all other
 $x=0$)

β_2 = effect of experience on wage
while all other factors are
fixed/held constant

β_3 = effect of skills on wage
when all others are held
constant

In MLR (Multiple linear regression)

All X 's (i.e. independent variables)
should not be correlated.

$$\text{Wage} = \beta_0 + \beta_1 \underset{\substack{\rightarrow \\ X_1}}{\text{education}} +$$
$$\beta_2 \underset{\substack{\rightarrow \\ X_2}}{\text{experience}} + \beta_3 \underset{\substack{\rightarrow \\ X_3}}{\text{experience + education}}$$
$$X_3 = X_2 + X_1$$

Consumption and Income

$$\text{Consumption} = \beta_0 + \beta_1(\text{income}) + \beta_2(\text{income})^2 + \epsilon$$

$$x_1 = \text{income}$$

$$x_2 = \text{income}^2$$

$$\frac{\Delta \text{Conc}}{\Delta \text{inc}} = \beta_1 + 2\beta_2(\text{income})$$

Generalized form of MLR.

$$Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k + \epsilon$$

β_0 - Intercept (Value of Y when $x_i = 0$)

$\beta_1 \dots \beta_k$ - Slope parameters for
 $x_1 \dots x_k$

ϵ - error term (accounts for all the factors that are not included in the model).

Assumptions

1. y given x_i is normally distributed
2. Observations are independent (random sample)

3. No perfect collinearity

- None of the independent variables are constant
- No linear relationship among any of the independent variables.

Sales(y) Advertising (x)

5

3

10

3

15

3

20

3

25

3



NOT

$$4. E(\varepsilon | x_1, x_2, \dots, x_k) = 0$$

5. Homoskedasticity

$$\text{Var}(\varepsilon | x_1, x_2, \dots, x_k) = \sigma^2$$

$$Y = \beta_0 + \beta_1 x_1 + \beta_2 \underline{x_2} + \dots$$

$\swarrow \beta_k x_k$

$H_0: \beta_1 = 0 \rightarrow$

If x_1 has an effect on Y

If $p < 0.05$ reject H_0

$$Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 \\ + \beta_4 x_4$$

x_1 and x_4 are highly correlated



$$\begin{aligned} \text{College GPA} &= \beta_0 + \beta_1 (\text{hsGPA}) \\ &\quad + \beta_2 (\text{Bclipped class}) \\ &\quad + \beta_3 (\text{business major}) + \\ &\quad \beta_4 (\text{clubs}) + \epsilon, \end{aligned}$$

$$\begin{aligned} \beta_3 &= 0.094 & \text{business major} \\ &\downarrow & = 1 & \text{if} \\ &&& \text{in business School} \\ &&& = 0 & \text{otherwise.} \end{aligned}$$

→ Compared to an engg major
a business major will have
a 0.094 times effect on
Col GPA.