

Measures of Correlation

In linear regression, there is a dependent variable Y and an independent regressor variable X . We think of Y as being approximately a function of X . We want to know if there is a relation of the form

$$E(Y|X=x) = a + bx$$

$$\sum_y P(Y=y|X=x)$$

For example

$$Y = a + bX + \varepsilon$$

where $\varepsilon \sim N(0, \sigma^2)$

or even $Y = a + bX$

$$\text{In these cases } \text{Cov}(X, Y) = \text{Cov}(X, a + bX) \\ = b \text{Var}(X)$$

$$b = \frac{\text{Cov}(X, Y)}{\text{Var}(X)}$$

We want estimates

$$\text{of } \text{Cov}(X, Y).$$

The covariance of X and Y is

$$\begin{aligned}\text{Cov}(X, Y) &= E[(X - E(X))(Y - E(Y))] \\ &= E[XY] - E[X]E[Y]\end{aligned}$$

The correlation

of X and Y is defined by

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}}$$

we have

$$① -1 \leq \text{Corr}(X, Y) \leq +1$$

$$② \text{ If } X \text{ and } Y \text{ are independent, } \text{Cov}(X, Y) = 0$$

but $\text{Cov}(X, Y) = 0$ does not imply that X and Y are independent.

$$③ \text{ If } Y = aX + b, \text{ then}$$

$$\text{Corr}(X, Y) = \frac{b\text{Var}(X)}{\sqrt{\text{Var}(X) \cdot b^2\text{Var}(X)}}$$

$$= \frac{b}{|b|}$$

$$= \begin{cases} 1 & \text{if } b > 0 \\ 0 & \text{if } b = 0 \\ -1 & \text{if } b < 0 \end{cases}$$

We want measures of correlation

for samples $\vec{x} = (x_1, \dots, x_n)$
and $\vec{y} = (y_1, \dots, y_n)$.

Three such measures are in use.

① Pearson's Correlation Coefficient

$$\text{Let } \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i, \bar{y} = \frac{1}{n} \sum_{i=1}^n y_i \quad (\hat{\sum}_{i=1}^n x_i)$$

$$S_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - \frac{(\sum_{i=1}^n x_i)^2}{n}$$

$$S_{yy} = \sum_{i=1}^n (y_i - \bar{y})^2 = \sum_{i=1}^n y_i^2 - \frac{(\sum_{i=1}^n y_i)^2}{n}$$

$$S_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \sum_{i=1}^n x_i y_i - \frac{(\sum_{i=1}^n x_i)(\sum_{i=1}^n y_i)}{n}$$

Pearson Corr. Coeff. =

$$r = \frac{S_{xy}}{\sqrt{S_{xx} S_{yy}}}$$

The code for calculating r

`Cor (<Data >)`, ~~method = "pearson"~~
`method = "pearson"`

or `Cor (<Data>)`

② Spearman's Correlation coefficient

First we define the ranks of the variables.

The rank of x_i is j if x_i is the j th smallest of x_1, x_2, \dots, x_n .

The rank of y_i is j if y_i is the j th smallest of y_1, \dots, y_n

Spearman's correlation coefficient r_s is Pearson's correlation coefficient applied to

$$(\text{rank}(x_1), \text{rank}(x_2), \dots, \text{rank}(x_n))$$

$$(\text{rank}(y_1), \text{rank}(y_2), \dots, \text{rank}(y_n))$$

The R code for calculating r_s

is

`Cor(<Data>, method = "spearman")`

Fact

If the x_i 's are unique (no repetitions) and the y_i 's are themselves unique

then

$$r_s = 1 - \frac{6 \sum_{i=1}^n d_i^2}{n(n^2-1)}$$

where $d_i = \text{rank}(y_i) - \text{rank}(x_i)$

③ Kendall's correlation coefficient

We consider ~~pair~~ ^{vectors of} observations

$$\begin{pmatrix} x_i \\ y_i \end{pmatrix} \quad i=1, \dots, n$$

A pair of vectors

$$\begin{pmatrix} x_i \\ y_i \end{pmatrix} \quad \begin{pmatrix} x_j \\ y_j \end{pmatrix}$$

is concordant

if ~~(x_i, x_j)~~

if either $x_i < x_j$ and $y_i < y_j$

or $x_i > x_j$ and $y_i > y_j$

A pair of vectors is discordant if

either $x_i < x_j$ and $y_i > y_j$

or

$x_i > x_j$ and $y_i < y_j$

Let n_c be the number of concordant pairs

and n_d be the number of discordant pairs.

Note: $n_c + n_d = \binom{n}{2}$

Kendall's correlation coefficient

τ "tau"

is

$$\tau = \frac{n_c - n_d}{\binom{n}{2}} = \frac{n_c - n_d}{\frac{n(n-1)}{2}}$$

Kendall's τ can be calculated from the ranks.

We can calculate it in R using

`Cor (Data, method = "kendall")`

Inference on Correlation Coefficients

Pearson

We need an assumption about the joint distribution of X and Y .

We assume (X, Y) is bivariate normal distributed with

parameters $\mu_x, \sigma_x, \mu_y, \sigma_y, \rho$

where ρ is the correlation.

We have $E(Y|X=x) = \mu_y + \rho \frac{\sigma_y}{\sigma_x} (x - \mu_x)$

$$E(Y|X=x) = \mu_y + \rho \frac{\sigma_y}{\sigma_x} (x - \mu_x)$$

Theorem

Under $H_0: \rho = 0$ the statistic

$$\frac{r \sqrt{n-2}}{\sqrt{1-r^2}}$$
 is t -distributed with $n-2$ degrees of freedom

where r is Pearson's correlation

n is number of observations

In R cor.test (Data), method = "pearson" 7