

Applied Regression Analysis and Other Multivariable Methods

Chapter 5.7 - 6

Yi Zhou

2015-11-11

Outline

Chapter 5 Straight-line Regression Analysis

5-7 Inferences concerning the slope and intercept

5-8 Interpretations of test for slope and intercept

5-9 Inferences concerning the regression line

Multivariate situation

Chapter 6 The correlation coefficient and straight-line regression analysis

6-1.2 Definition of r

6-3 Bivariate Normal Distribution

6-4.5 r and the strength of the straight-line relationship

6-6.7 Test of hypotheses and CIs for the correlation coefficient

5-7 Inferences concerning the slope and intercept

- ▶ The origin of *inference*:
 - ▶ To assess the fitted line
 - ▶ To consider the uncertainties from the sample
- ▶ CIs and test statistical hypotheses for the unknown parameters
- ▶ **Assumption:** the random variable Y has a normal distribution at each fixed value of X .
- ▶ Given $\mu_{Y|X} = \beta_0 + \beta_1 X$, $\hat{\beta}_0$ and $\hat{\beta}_1$ are normally distributed with respective means β_0 and β_1 .
- ▶ *Theorem:* linear functions of independent normal distributed observations are themselves normally distributed.
- ▶ $H_0 : \beta_1 = \beta_1^{(0)}$, $H_0 : \beta_0 = \beta_0^{(0)}$, using t distribution with $(n - 2)$ d.f. (see eq.(5.12), eq.(5.13)).

5-8 Interpretations of test for slope and intercept

- ▶ Assumption: normality, independence, variance homogeneity hold.
- ▶ Zero slope: $H_0 : \beta_1 = 0$
 - ▶ Accept: X provides little or no help;
The relationship is not linear.
 - ▶ Reject: X provides significant information;
There is a definite linear component and might be a curvilinear term.
- ▶ Zero intercept: $H_1 : \beta_0 = 0$

5-9 Inferences concerning the regression line

- ▶ $\mu_{Y|X_0} = \mu_{Y|X_0}^{(0)}$
- ▶ Confidence intervals for $\mu_{Y|X_0}$, confidence band.
- ▶ A better estimation of $\mu_{(Y|X)}$ will be done near \bar{X}
- ▶ Prediction interval(PI) for \hat{Y}_{X_0} , prediction band.

Multivariate situation

$$\mu_{Y|X} = \beta_0 + \beta_1 X_1 + \cdots + \beta_p X_p$$

- ▶ Ordinary Least Square methods: $\frac{\partial Q}{\partial \beta_i}$
- ▶ $H_0 : \beta_i = \beta_i^{(0)}$, using t distribution with $(n - p - 1)$ d.f.
- ▶ $H_0 : \beta_i = 0$
- ▶ $H_0 : \beta_1 = \beta_2 = \cdots = \beta_p = 0$, F test, $F_{p, n-p-1}$.
- ▶ $H_0 : \beta_1 = \beta_2 = \cdots = \beta_r = 0$, part of X_i , F test, $F_{r, n-p-1}$

6-1.2 Definition of r

- ▶ The sample correlation coefficient: r
- ▶ The population correlation coefficient: ρ
- ▶ Mathematical properties of r is on pp.81
- ▶ Interpretation of r
- ▶ positive / negative quadrants

6-3 Bivariate Normal Distribution

- ▶ A bell-shaped surface
- ▶ The conditional distribution of Y at X , that is the distribution of Y for fixed X , is univariate normal.
- ▶ Joint p.d.f

$$P(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp\left\{\frac{-z}{2(1-\rho^2)}\right\}$$

$$\text{with } z = \frac{(x_1 - \mu_1)^2}{\sigma_1^2} - \frac{2\rho(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1\sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2}$$

$$\text{and } \rho = \text{cor}(x_1, x_2) = \frac{V_{12}}{\sigma_1\sigma_2}$$

- ▶ Marginal probability:

$$P(x_1) = \int_{-\infty}^{\infty} P(x_1, x_2) dx_2 = \frac{1}{\sigma_1\sqrt{2\pi}} \exp\left\{-\frac{(x_1 - \mu_1)^2}{2\sigma_1^2}\right\}$$

$$P(x_2) = \int_{-\infty}^{\infty} P(x_1, x_2) dx_1 = \frac{1}{\sigma_2\sqrt{2\pi}} \exp\left\{-\frac{(x_2 - \mu_2)^2}{2\sigma_2^2}\right\}$$

6-4 r and the strength of the straight-line relationship

- ▶ SSY and SSE

$$r^2 = \frac{SSY - SSE}{SSY}$$

- ▶ If $SSE = 0$, then $r^2 = 1$
- ▶ If $r^2 = 0$, then it implies the absence of linear relationship.
- ▶ **Misconceptions** of r^2
 - ▶ Not a measure of magnitude of the slope
 - ▶ Not a measure of the appropriateness of the straight-line model

6-6.7 Test of hypotheses and CIs for the correlation coefficient

- ▶ $H_0 : \rho = 0$
- ▶ $H_0 : \rho = \rho_0, \rho_0 \neq 0$
- ▶ CI
- ▶ $H_0 : \rho_1 = \rho_2$, independent random samples
- ▶ $H_0 : \rho_{12} = \rho_{13}$, not independent