9 Statistical Inference in Multiple Regression Applied regression analysis and other multivariate methods

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Applied Rgression Analysis And Other Multivariable Methods Kleinbaum/ Kupper/ Nizam/ Rosenberg

9.1 Preview

Q: The contributions of X_i to Y

- Overall test: F test
- ► Test for additional of a *single* variable: **partial F test**
- ► Test for additional of a group variable: multiple partial F test

F test in regression analysis context:

$$F = \frac{\hat{\sigma}_0^2 (H_0 \text{ is true})}{\hat{\sigma}^2 (MS \text{ residual})}$$

9.1 Preview

- ► Full model
- Reduced model

$$F = \frac{\hat{\sigma}_0^2 \text{ (reduced model)}}{\hat{\sigma}^2 \text{ (full model)}} \text{ (right??)}$$

9.2 Test for significant overall regression

- ► H₀: all k independent variables considered together do not explain a significant amount of the variation in Y.
- ► *H*₀: there is no significant overall regression using all *k* independent variables in the model
- H_0 : $\beta_1 = \beta_2 = \cdots = \beta_k = 0$

$$F = \frac{MS \ Regression}{MS \ Residual} \sim F_{k,n-k-1} \ (9.1)$$

9.2 Test for significant overall regression

$$F = \frac{R^2/k}{(1 - R^2)/(n - k - 1)} \sim F_{k,n-k-1}$$
 (9.2)
$$R^2 = \frac{SSY - SSE}{SSY}$$

- ► $(SSY SSE)/k = MS \ Regression : \hat{\sigma}_0^2 = \hat{\sigma}^2 \ under \ H_0$ (reduced model).
- ► $SSE/(n-k-1) = MS \ Residual : \hat{\sigma}^2$ under the assumed model (full model).

9.3 Partial F test

Partial F test: whether the *addition* of any specific independent variable (only one) is significantly contributes to the prediction of Y.

- ▶ H_0 : X^* does not significantly add to the prediction of Y given that $X_1, X_2, ..., X_p$ are already in the model.
- ► H_0 : $\beta^* = 0$ in the model $Y = \beta_0 + \beta_1 X_1 + \dots + \beta_p X_p + \beta^* X^* + E$

$$F(X^*|X_1,\ldots,X_p) = \frac{SS(X^*|X_1,\ldots,X_p)}{MS(X_1,\ldots,X_p,X^*)} \sim F_{1,n-p-2} \quad (9.4)$$

$$SS(X^*|X_1,...,X_p)$$

= SS Regression (full model) - SS Regression (reduced model)

9.3 Partial F test

- ► Full model: $Y = \beta_0 + \beta_1 X_1 + \cdots + \beta_p X_p + \beta^* X^* + E$
- ▶ Reduced model: $Y = \beta_0 + \beta_1 X_1 + \cdots + \beta_p X_p + E$

$$F(X^*|X_1,\ldots,X_p) = \frac{MS(X^*|X_1,\ldots,X_p)}{MS(X_1,\ldots,X_p,X^*)} \sim F_{1,n-p-1}$$

- $ightharpoonup F(X_2|X_1), F(X_3|X_1, X_2), \text{ and } MS \ Residual(X_1, X_2)$
- T test alternative
- Application: the control of extraneous variable.
 - ▶ To work backward by deleting X^* variables, at one time, until a best model is obtained. (Chapter 16, P_{447})

9.4 Multiple partial F test

Multiple partial F test: whether the simultaneously addition of any specific independent variables (more than one) are significantly contributes to the prediction of Y.

- ▶ H_0 : $X_1^*, X_2^*, \dots, X_s^*$ do not significantly add to the prediction of Y given that X_1, X_2, \dots, X_p are already in the model.
- ▶ H_0 : $\beta_1^* = \beta_2^* = \cdots = \beta_s^* = 0$ in the model $Y = \beta_0 + \beta_1 X_1 + \cdots + \beta_p X_p + \beta_1^* X_1^* + \cdots + \beta_s^* X_s^* + E$

$$F(X_1^*, \dots, X_s^* | X_1, \dots, X_p) = \frac{SS(X_1^*, \dots, X_s^* | X_1, \dots, X_p)/s}{MS(X_1, \dots, X_p, X_1^*, \dots, X_s^*)}$$
$$\sim F_{s, n-p-s-1} \quad (9.6)$$

9.4 Multiple partial F test

$$F(X_1^*, \dots, X_s^* | X_1, \dots, X_p) = \frac{MS(X_1^*, \dots, X_s^* | X_1, \dots, X_p)}{MS(X_1, \dots, X_p, X_1^*, \dots, X_s^*)}$$

$$SS(X_1^*,\ldots,X_s^*|X_1,\ldots,X_p)$$

- = SS Reg(full model) SS Reg(reduced model)
- = SSE(reduced model) SSE(full model)
 - Application: the simultaneous importance of particular subsets of a set of predictor variables.
 - Whether a chunk of variables (having some trait in common) is important when considered together.

9.5 Strategies

- ► Variables-add-in-odder tests
- Variables-add-last tests

9.6 Additional inference methods for multiple regression

- ▶ Test involving the intercept: H_0 : $\beta_0 = 0$
- ► Confidence interval about regression coefficients:

$$\hat{\beta^*} \pm t_{n-k-1}(S_{\hat{\beta^*}})$$

- ▶ Whether the mean value of Y equals to a hypothesized value at a given set of values for the independent variables.
 - ► T test (9.8)
 - ► CI
- Prediction of Y at a specific set of predictor values: PI
- ► Inference method for linear functions of regression coefficients: the joint effect of the predictors