

Interpreting output of the R function lm

3.6.3 Multiple Linear Regression

(*) If H_0 is true, then F statistic will have (Snedecor's) F distribution with $df = (2, 503)$.

In order to fit a multiple linear regression model using least squares, we again use the `lm()` function. The syntax `lm(y~x1+x2+x3)` is used to fit a model with three predictors, x_1 , x_2 , and x_3 . The `summary()` function now outputs the regression coefficients for all the predictors.

```
> lm.fit=lm(medv~lstat+age,data=Boston)
> summary(lm.fit)
```

Call:

```
lm(formula = medv ~ lstat + age, data = Boston)
```

Residuals:

Min	1Q	Median	3Q	Max
-15.98	-3.98	-1.28	1.97	23.16

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	33.2228	0.7308	45.46	<2e-16 ***
lstat X_1	-1.0321	0.0482	-21.42	<2e-16 ***
age X_2	0.0345	0.0122	2.83	0.0049 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 6.17 on 503 degrees of freedom

Multiple R-squared: 0.551 Adjusted R-squared: 0.549

F-statistic: 309 on 2 and 503 DF, p-value: <2e-16

If $H_0: \beta_2 = 0$ is true, then

$$\frac{\hat{\beta}_2 - 0}{\widehat{SE}(\hat{\beta}_2)} \sim \text{Student's } t \text{ distr. with } df = n - 3 = 503$$

p-value for the test $H_0: \beta_2 = 0$ (2-sided)

$H_A: \beta_2 \neq 0$

R^2 :
coeff. of determination

(*) $\hat{\sigma}$ - an estimator of σ

(*) $H_0: \beta_1 = 0$ and $\beta_2 = 0$ (no associations w. any of the two covariates)
 H_A : at least one of β_1 or $\beta_2 \neq 0$, (*)

CC balance vs ethnicity example:

3 ~~sub~~groups : AA, Asian, Caucasian.

Model 1: The mean is the same for every group, i.e., $E(Y_i) = \mu \forall i$.

Model 2: subgroup-specific means, i.e.,

$E(Y_i) = \mu_j$ if the i th person belongs to group j .

Representation 1:

$$y_i = \mu_1 + \epsilon_i \quad \text{if } i\text{th person is AA}$$

$$y_i = \mu_2 + \epsilon_i \quad \text{if } \text{--- " --- Asian}$$

$$y_i = \mu_3 + \epsilon_i \quad \text{if } \text{--- " --- Caucasian}$$

Representation 2:

$$y_i = \beta_0 + \epsilon_i$$

$$y_i = \beta_0 + \beta_1 + \epsilon_i$$

$$y_i = \beta_0 + \beta_2 + \epsilon_i$$

Example: 2 people from each group. Write out the design matrices for MLR.

$$\begin{pmatrix} Y_1 \\ Y_2 \\ Y_3 \\ Y_4 \\ Y_5 \\ Y_6 \end{pmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \end{bmatrix} + \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \\ \epsilon_4 \\ \epsilon_5 \\ \epsilon_6 \end{bmatrix} \quad \text{equiv.} \quad \begin{pmatrix} Y_1 \\ Y_2 \\ Y_3 \\ Y_4 \\ Y_5 \\ Y_6 \end{pmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \\ \epsilon_4 \\ \epsilon_5 \\ \epsilon_6 \end{bmatrix}$$