

Chapter 3 Part I Linear Regression (Problem 12)

kartheek raj mulasa

12/21/2019

Problem 12: This problem involves simple linear regression without an intercept.

(a) Recall that the coefficient estimate $\hat{\beta}$ for the linear regression of Y onto X without an intercept is given by (3.38). Under what circumstance is the coefficient estimate for the regression of X onto Y the same as the coefficient estimate for the regression of Y onto X?

Answer

In generic linear equation $y=mx+c$, $C=0$ as intercept is zero then equation $y=mx$ viceversa x onto y gives equation $x=my$ both linear equations have the same slope occurs when $y=x$.

Given the linear regression equation in (3.38) we can say denominator term must be equal as numerator same for both events $\beta = \frac{\sum \{i\}x\{i\}y\{i\}}{\sum \{i\}x\{i\}x\{i\}}$

$$\sum_{i=1}^n x_i^2 = \sum_{i=1}^n y_i^2$$

(b) Generate an example in R with $n = 100$ observations in which the coefficient estimate for the regression of X onto Y is different from the coefficient estimate for the regression of Y onto X.

Answer

creating the observations

```
set.seed(1)
x<-sample(1:1000,100,replace = T)
y<-x*2+sample(1:1000,100,replace = T)
length(x)
```

```
## [1] 100
```

```
length(y)
```

```
## [1] 100
```

Building linear models x onto y and y onto x.

```
slm<-lm(y~x+0)
slm1<-lm(x~y+0)
summary(slm)
```

```
##
## Call:
## lm(formula = y ~ x + 0)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -641.5  -122.1   138.6   299.6   707.6
##
## Coefficients:
```

```
## Estimate Std. Error t value Pr(>|t|)
## x 2.74064 0.05449 50.3 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 330.8 on 99 degrees of freedom
## Multiple R-squared: 0.9623, Adjusted R-squared: 0.962
## F-statistic: 2530 on 1 and 99 DF, p-value: < 2.2e-16
summary(slm1)
```

```
##
## Call:
## lm(formula = x ~ y + 0)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -243.14  -92.29  -23.92   73.22  260.33
##
## Coefficients:
## Estimate Std. Error t value Pr(>|t|)
## y 0.351137 0.006981 50.3 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 118.4 on 99 degrees of freedom
## Multiple R-squared: 0.9623, Adjusted R-squared: 0.962
## F-statistic: 2530 on 1 and 99 DF, p-value: < 2.2e-16
```

Generated linear models yields different coefficients

(c) Generate an example in R with $n = 100$ observations in which the coefficient estimate for the regression of X onto Y is the same as the coefficient estimate for the regression of Y onto X .

Answer

Generating the 100 observations

```
x<-100:200
x<-x^2
y<-200:100
y<-y^2
```

Building linear models x onto y and y onto x .

```
slm2<-lm(x~y+0)
slm3<-lm(y~x+0)
summary(slm2)
```

```
##
## Call:
## lm(formula = x ~ y + 0)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -20166  -7470   5532  18842  32459
##
## Coefficients:
```

```
## Estimate Std. Error t value Pr(>|t|)
## y 0.75414 0.06567 11.48 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 16460 on 100 degrees of freedom
## Multiple R-squared: 0.5687, Adjusted R-squared: 0.5644
## F-statistic: 131.9 on 1 and 100 DF, p-value: < 2.2e-16
summary(slm3)
```

```
##
## Call:
## lm(formula = y ~ x + 0)
##
## Residuals:
## Min 1Q Median 3Q Max
## -20166 -7470 5532 18842 32459
##
## Coefficients:
## Estimate Std. Error t value Pr(>|t|)
## x 0.75414 0.06567 11.48 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 16460 on 100 degrees of freedom
## Multiple R-squared: 0.5687, Adjusted R-squared: 0.5644
## F-statistic: 131.9 on 1 and 100 DF, p-value: < 2.2e-16
```

Both models yield same coefficient.