hw3 draft

Yujia

2022-10-04

Typed Problem 1

Build a linearly dependant matrix m1:

```
x1 \leftarrow c(6,2,9,1)
x2 \leftarrow c(2,4,1,2)
x3 \leftarrow c(4,-2,8,-1)
m1 \leftarrow cbind(x1,x2,x3)
lmod \leftarrow lm(x1~x3+x2)
summary(lmod)
## Warning in summary.lm(lmod): essentially perfect fit: summary may be unreliable
##
## Call:
## lm(formula = x1 ~ x3 + x2)
##
## Residuals:
##
                        2
                                    3
            1
## -1.846e-16 5.430e-17 1.086e-16 2.172e-17
##
## Coefficients:
                 Estimate Std. Error
##
                                        t value Pr(>|t|)
## (Intercept) 4.441e-16 4.964e-16 8.950e-01
                                                    0.535
               1.000e+00 4.734e-17 2.112e+16
## x3
                                                   <2e-16 ***
                1.000e+00 1.748e-16 5.721e+15
                                                   <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.22e-16 on 1 degrees of freedom
## Multiple R-squared:
                              1, Adjusted R-squared:
## F-statistic: 4.158e+32 on 2 and 1 DF, p-value: < 2.2e-16
x1, x2, x3 are linearly dependent. Regress x1 on x2 and x3, we will get that the R-squared equals to 1 and
```

x1, x2, x3 are linearly dependent. Regress x1 on x2 and x3, we will get that the R-squared equals to 1 and RSS=0. Also, since x1 is depend on both x1 and x2, the regression p-value of x2 and x3 are both less than 0.05. In general, A set of vectors $\{x_1, x_2, \cdots, x_p\}$ is linearly dependent if there exist numbers $\beta_1, \beta_2, \cdots, \beta_p$, not all equal to zero, such that

```
\beta_1 x_1 + \beta_2 x_2 + \dots + \beta_p x_P = 0
```

It can be re-written as

$$x_1 = -(\beta_2 x_2 + \dots + \beta_p x_P)/\beta_1$$

Therefore, we can get a perfect fit and the rss will be 0.

Typed Problem 2

Build a linearly independent matrix m2:

```
## [1] -0.02293641
```

The correlation of this sample is very closed to 0. We can conclude that in general, if numerical predictors are linearly independent, then they are uncorrelated.

Typed Problem 3

 \mathbf{a}

```
myOLS <- function(Y, X, is1 = TRUE) {
  if (is1 == TRUE)
n <- nrow(X)
p <- ncol(X)
if (is1 == FALSE){x1 <- cbind(rep(1, n),X)}
else {x1 <- X}
beta <- solve(crossprod(x1,x1),crossprod(x1,Y))</pre>
xtxi <- solve(t(x1) %*% x1)</pre>
H <- x1 %*% xtxi %*% t(x1)
Rss <- t(Y) %*% (diag(1,n)-H) %*% Y
sigma <- sqrt(Rss/(n-p))</pre>
se <- sqrt(diag(xtxi))*sigma</pre>
list(beta,se)
}
n = 30
set.seed(0)
p = 3
X = matrix(runif(n*p), nrow=n)*2-1
b = seq(1,p,by=1)
Y = X%*%b + rnorm(n)
fit1 = lm(Y ~ X); summary(fit1); # regression with an intercept
##
## Call:
## lm(formula = Y ~ X)
##
## Residuals:
##
       Min
                 1Q Median
                                 ЗQ
                                         Max
## -2.3701 -0.3304 0.1082 0.4938 2.3930
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                             0.1708 -0.756 0.45648
## (Intercept)
                -0.1291
## X1
                  0.9214
                             0.2987
                                       3.085 0.00478 **
## X2
                  2.4021
                             0.3468
                                       6.926 2.36e-07 ***
                  2.7482
                                      7.960 1.94e-08 ***
## X3
                             0.3452
## ---
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9235 on 26 degrees of freedom
## Multiple R-squared: 0.802, Adjusted R-squared: 0.7792
## F-statistic: 35.11 on 3 and 26 DF, p-value: 2.711e-09
myOLS(Y,X,FALSE)
## Warning in sqrt(diag(xtxi)) * sigma: Recycling array of length 1 in vector-array arithmetic is depre
   Use c() or as.vector() instead.
## [[1]]
##
              [,1]
## [1,] -0.1291359
## [2,] 0.9214173
## [3,]
        2.4020865
## [4,] 2.7482181
##
## [[2]]
## [1] 0.1676341 0.2930771 0.3403307 0.3387858
fit0 = lm(Y ~ -1 + X); summary(fit0) # regression without an intercept
## Call:
## lm(formula = Y \sim -1 + X)
##
## Residuals:
##
       Min
                 1Q
                    Median
## -2.48956 -0.47760 -0.00264 0.37870 2.29687
##
## Coefficients:
##
     Estimate Std. Error t value Pr(>|t|)
## X1
                  0.2939 3.038 0.00523 **
       0.8929
       2.3868
                  0.3435
                           6.949 1.81e-07 ***
       2.7167
                  0.3400
                           7.991 1.38e-08 ***
## X3
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9161 on 27 degrees of freedom
## Multiple R-squared: 0.7982, Adjusted R-squared: 0.7758
## F-statistic: 35.61 on 3 and 27 DF, p-value: 1.583e-09
myOLS(Y,X,TRUE)
## Warning in sqrt(diag(xtxi)) * sigma: Recycling array of length 1 in vector-array arithmetic is depre
    Use c() or as.vector() instead.
## [[1]]
##
            [,1]
## [1,] 0.892872
## [2,] 2.386783
## [3,] 2.716730
##
## [[2]]
## [1] 0.2939026 0.3434637 0.3399867
```