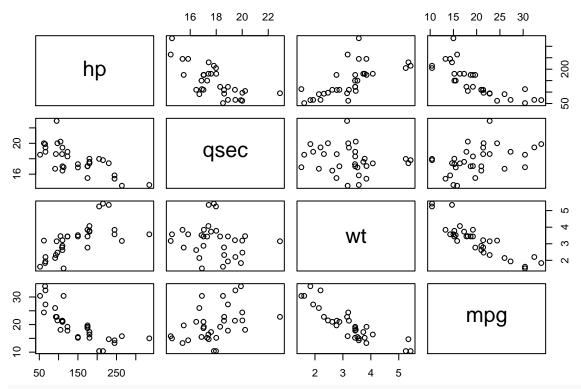
lab04-ols-regression-Katia-Williams

Katia Williams 2/12/2018

Part 1:EDA

```
names(mtcars)
              "cyl" "disp" "hp"
## [1] "mpg"
                                  "drat" "wt"
                                               "qsec" "vs"
                                                                   "gear"
## [11] "carb"
cars \leftarrow mtcars[c(4, 7, 6, 1)]
head(cars)
##
                    hp qsec
                              wt mpg
## Mazda RX4
                   110 16.46 2.620 21.0
## Mazda RX4 Wag
                   110 17.02 2.875 21.0
## Datsun 710
                    93 18.61 2.320 22.8
## Hornet 4 Drive
                   110 19.44 3.215 21.4
## Hornet Sportabout 175 17.02 3.440 18.7
## Valiant
                   105 20.22 3.460 18.1
summary(cars)
                                       wt
         hp
                       qsec
                                                      mpg
  Min. : 52.0
                  Min. :14.50
                                  Min.
                                       :1.513
                                                 Min.
                                                       :10.40
  1st Qu.: 96.5
                  1st Qu.:16.89
                                  1st Qu.:2.581
                                                 1st Qu.:15.43
## Median :123.0
                  Median :17.71
                                  Median :3.325
                                                 Median :19.20
## Mean
         :146.7
                  Mean :17.85
                                  Mean
                                       :3.217
                                                 Mean
                                                       :20.09
## 3rd Qu.:180.0
                  3rd Qu.:18.90
                                  3rd Qu.:3.610
                                                 3rd Qu.:22.80
          :335.0
                         :22.90
## Max.
                  Max.
                                  Max.
                                        :5.424
                                                 Max.
                                                       :33.90
cor(cars)
##
               hp
                       qsec
        ## qsec -0.7082234 1.0000000 -0.1747159 0.4186840
        0.6587479 -0.1747159 1.0000000 -0.8676594
## mpg -0.7761684 0.4186840 -0.8676594 1.0000000
pairs(cars)
```

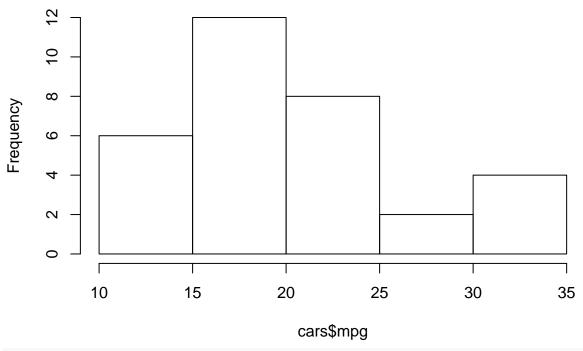


prcomp(cars)

```
## Standard deviations (1, .., p=4):
## [1] 68.7374255 3.8533877 1.2237768 0.4021126
##
## Rotation (n x k) = (4 x 4):
## PC1 PC2 PC3 PC4
## hp 0.997453468 0.06643932 0.02542544 -0.005093549
## qsec -0.018396105 -0.11140955 0.96740653 -0.226658563
## wt 0.009406474 -0.14735595 0.20966894 0.966559195
## mpg -0.068261860 0.98054064 0.13970358 0.119846874
```

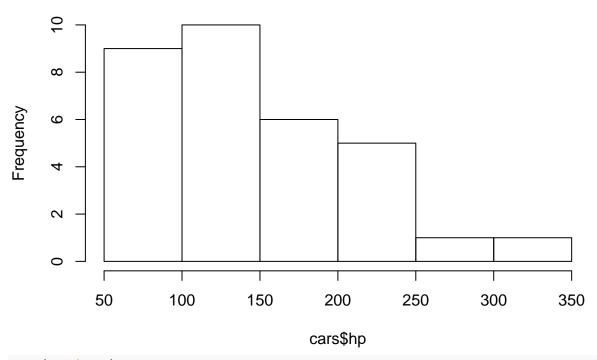
hist(cars\$mpg)

Histogram of cars\$mpg



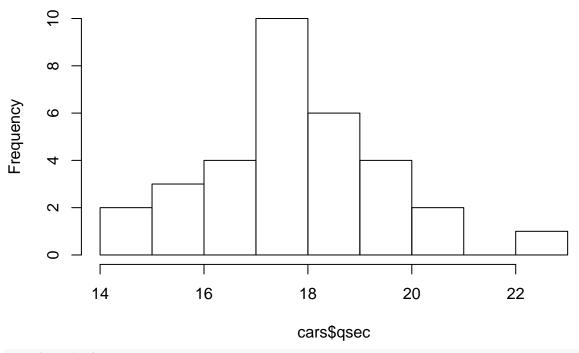
hist(cars\$hp)

Histogram of cars\$hp



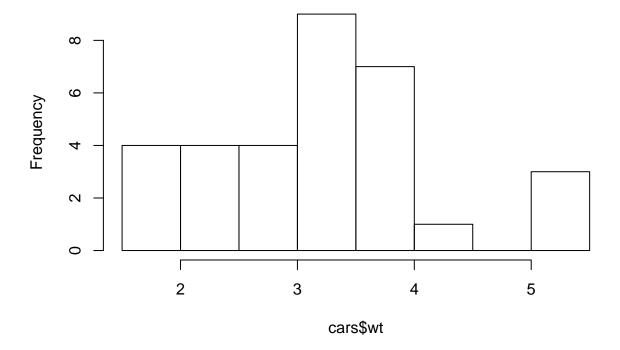
hist(cars\$qsec)

Histogram of cars\$qsec



hist(cars\$wt)

Histogram of cars\$wt



Part 2: OLS Outputs

wt -4.35879720

 \mathbf{X}

```
get_X <- function(df, n, p) {</pre>
 rawX <- df[c(1:p)]
 rawX['ones'] = rep(1, n)
 return (as.matrix(rawX[c(p+1, c(1:p))]))
X <- get_X(cars,length(cars$hp), 3)</pre>
head(X)
##
                     ones hp qsec wt
## Mazda RX4
                     1 110 16.46 2.620
                     1 110 17.02 2.875
## Mazda RX4 Wag
## Datsun 710
                      1 93 18.61 2.320
## Hornet 4 Drive 1 110 19.44 3.215
## Hornet Sportabout 1 175 17.02 3.440
## Valiant
              1 105 20.22 3.460
\mathbf{Y}
get_y <- function(df) {</pre>
return(as.matrix(df[,ncol(df)]))
y <- get_y(cars)</pre>
head(y)
##
        [,1]
## [1,] 21.0
## [2,] 21.0
## [3,] 22.8
## [4,] 21.4
## [5,] 18.7
## [6,] 18.1
Coefficients
Assumption: X has full rank
get_betahats <- function(X, y) {</pre>
 return (solve( t(X) %*% X ) %*% t(X) %*% y)
Coefficients <- get_betahats(X, y)</pre>
Coefficients
               [,1]
## ones 27.61052686
## hp -0.01782227
## qsec 0.51083369
```

Fitted values

```
fitted_values <- X %*% Coefficients fitted_values
```

```
##
                            [,1]
## Mazda RX4
                       22.638351
## Mazda RX4 Wag
                       21.812925
## Datsun 710
                       25.347261
## Hornet 4 Drive
                       21.567151
## Hornet Sportabout
                       18.191756
## Valiant
                       20.986807
## Duster 360
                       15.774770
## Merc 240D
                       22.817657
## Merc 230
                       23.885291
## Merc 280
                       19.772382
## Merc 280C
                       20.078882
## Merc 450SE
                       15.550720
## Merc 450SL
                       17.134877
## Merc 450SLC
                       17.121271
## Cadillac Fleetwood 10.258066
## Lincoln Continental 9.239679
## Chrysler Imperial
                        9.112356
## Fiat 128
                       26.790835
## Honda Civic
                       29.104951
## Toyota Corolla
                       28.619277
## Toyota Corona
                       25.359114
## Dodge Challenger
                       18.211984
## AMC Javelin
                       18.802141
## Camaro Z28
                       14.378236
## Pontiac Firebird 16.441769
## Fiat X1-9
                       27.654741
## Porsche 914-2
                       25.191797
## Lotus Europa
                       27.634839
## Ford Pantera L
                       16.495149
## Ferrari Dino
                       20.335683
## Maserati Bora
                       13.537332
## Volvo 142E
                       23.051950
```

Residuals

```
residuals <- fitted_values - y
head(residuals)</pre>
```

```
## [,1]
## Mazda RX4 1.6383509
## Mazda RX4 Wag 0.8129245
## Datsun 710 2.5472611
## Hornet 4 Drive 0.1671510
## Hornet Sportabout -0.5082436
## Valiant 2.8868073
```

RSS

Sigma2

```
n <- length(X[,1])
p <- 3
sigma2 <- RSS/(n - p - 1)
sigma2
## [,1]
## [1,] 6.644975</pre>
```

TSS

```
ybar <- mean(y)
TSS <- (1/n)*sum((y - ybar)**2)
TSS
## [1] 35.18897</pre>
```

.... [1] 00.1000.

They are the same!

ESS

```
ESS <- (1/n)*sum((fitted_values - ybar)**2)
ESS
## [1] 29.37462
```

R2

```
R2 <- ESS/TSS
R2

## [1] 0.8347678

(cor(y, fitted_values))**2

## [,1]

## [1,] 0.8347678
```

Checking with lm()

```
reg <- lm(mpg ~ hp + qsec + wt, data = mtcars)
reg
##
## Call:
## lm(formula = mpg ~ hp + qsec + wt, data = mtcars)
## Coefficients:
## (Intercept)
                                 qsec
                                                wt
     27.61053 -0.01782 0.51083
##
                                       -4.35880
Coefficients
##
              [,1]
## ones 27.61052686
## hp -0.01782227
## qsec 0.51083369
## wt -4.35879720
```

Part 3: QR Decomposition

```
qr_ols <- function(X, y) {</pre>
  QRD <- qr(X)
  Q <- qr.Q(QRD)
  R \leftarrow qr.R(QRD)
  QTy <- t(Q) %*% y
  return (backsolve(r=R, QTy, upper.tri=TRUE))
b \leftarrow qr_ols(X, y)
## [1,] 27.61052686
## [2,] -0.01782227
## [3,] 0.51083369
## [4,] -4.35879720
qr.solve(X, y)
##
## ones 27.61052686
## hp -0.01782227
## qsec 0.51083369
## wt -4.35879720
```

Part 4

```
centcars <- data.frame(scale(cars))
centX <- get_X(centcars, n, p)</pre>
```

```
centy <- get_y(centcars)
centbetas <- qr_ols(centX, centy)
centbetas

## [,1]
## [1,] 2.245792e-18</pre>
```

[2,] -2.027471e-01 ## [3,] 1.514582e-01 ## [4,] -7.076380e-01