Weekly Summary Template

Brady Miller

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Include a $very\ brief$ summary of what you learnt in this class here. Today, I learnt the following concepts in class:

- 1. How to make a correlation plot and its interpretation
- 2. Variance inflation
- 3. Step-wise Regression

```
# loading in necessary libraries and datasets
library(ISLR2)
library(dplyr)
```

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':

filter, lag

```
The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union
  library(tidyr)
  library(purrr)
  library(readr)
  library(glmnet)
Loading required package: Matrix
Attaching package: 'Matrix'
The following objects are masked from 'package:tidyr':
    expand, pack, unpack
Loaded glmnet 4.1-6
  library(caret)
Loading required package: ggplot2
Loading required package: lattice
Attaching package: 'caret'
The following object is masked from 'package:purrr':
    lift
  library(car)
Loading required package: carData
Attaching package: 'car'
```

```
The following object is masked from 'package:purrr':
    some

The following object is masked from 'package:dplyr':
    recode

df <- Boston
    attach(Boston)
```

Correlation Plot

```
# creates new data frame that outputs correlation table with numeric values
R <- df %>%
  keep(is.numeric) %>%
  cor()
R
```

```
crim
                                   indus
                                                 chas
crim
        0.42097171
       -0.20046922 1.00000000 -0.53382819 -0.042696719 -0.51660371
zn
indus
        0.40658341 -0.53382819 1.00000000 0.062938027
                                                      0.76365145
       -0.05589158 -0.04269672 0.06293803 1.000000000 0.09120281
chas
        0.42097171 - 0.51660371 \ 0.76365145 \ 0.091202807 \ 1.00000000
nox
       -0.21924670 0.31199059 -0.39167585 0.091251225 -0.30218819
rm
        0.35273425 - 0.56953734 \ 0.64477851 \ 0.086517774 \ 0.73147010
age
       -0.37967009 0.66440822 -0.70802699 -0.099175780 -0.76923011
dis
rad
        0.62550515 - 0.31194783 \quad 0.59512927 - 0.007368241 \quad 0.61144056
        0.58276431 -0.31456332 0.72076018 -0.035586518 0.66802320
tax
ptratio 0.28994558 -0.39167855 0.38324756 -0.121515174 0.18893268
        0.45562148 - 0.41299457 \quad 0.60379972 - 0.053929298 \quad 0.59087892
lstat
       -0.38830461 0.36044534 -0.48372516 0.175260177 -0.42732077
medv
                          age
                                     dis
                                                  rad
                                                                   ptratio
       -0.21924670 0.35273425 -0.37967009 0.625505145 0.58276431
crim
                                                                 0.2899456
zn
        0.31199059 -0.56953734 0.66440822 -0.311947826 -0.31456332 -0.3916785
       -0.39167585 0.64477851 -0.70802699 0.595129275 0.72076018
indus
                                                                 0.3832476
chas
        -0.30218819 0.73147010 -0.76923011 0.611440563 0.66802320
                                                                 0.1889327
nox
        1.00000000 - 0.24026493 \quad 0.20524621 \ - 0.209846668 \ - 0.29204783 \ - 0.3555015
rm
```

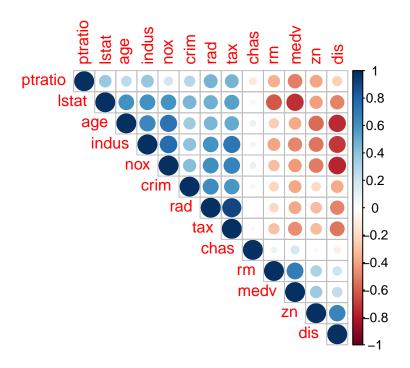
```
1.00000000 -0.74788054
        -0.24026493
                                             0.456022452
                                                          0.50645559 0.2615150
age
dis
         0.20524621 -0.74788054
                                 1.00000000 -0.494587930 -0.53443158 -0.2324705
                     0.45602245 -0.49458793
                                             1.000000000
rad
        -0.20984667
                                                          0.91022819
                                                                       0.4647412
        -0.29204783
                     0.50645559 -0.53443158
                                             0.910228189
                                                           1.00000000
                                                                       0.4608530
tax
ptratio -0.35550149
                     0.26151501 -0.23247054
                                             0.464741179
                                                           0.46085304
                                                                       1.0000000
lstat
        -0.61380827
                     0.60233853 -0.49699583
                                             0.488676335
                                                           0.54399341
                                                                       0.3740443
medv
         0.69535995 -0.37695457
                                 0.24992873 -0.381626231 -0.46853593 -0.5077867
             lstat
                         medv
         0.4556215 -0.3883046
crim
zn
        -0.4129946 0.3604453
         0.6037997 -0.4837252
indus
chas
        -0.0539293 0.1752602
nox
         0.5908789 -0.4273208
        -0.6138083 0.6953599
rm
         0.6023385 -0.3769546
age
dis
        -0.4969958 0.2499287
         0.4886763 -0.3816262
rad
tax
         0.5439934 -0.4685359
        0.3740443 -0.5077867
ptratio
         1.0000000 -0.7376627
lstat
medv
        -0.7376627 1.0000000
```

- Correlation can range between [-1,1]
 - 1. Close to 1 = strong positive correlation (as one variable increases, the other variable increases)
 - 2. Close to -1 = strong negative correlation(as one variable increases, the other variable decreases)
 - 3. Close to 0 = very weak/no correlation between variables
- Can come up with a *p*-value for a correlation value to see if there is a relationship between 2 variables (see if one is a good predictor of another)

```
# creates correlation plot
library(corrplot)

corrplot 0.92 loaded

corrplot(R, type = 'upper', order = 'hclust')
```



- Red = negative correlation
- Blue = positive correlation
- Can't isolate effect of indus and nox as they are related so won't be able to hold one constant while increasing the other (have very high correlation)
- Tax is negatively correlated with indus (total nuberm of business), so if the number of businesses in neighborhood goes up, then the distance from that given house to a business district decreases/goes down

Variance Inflation

```
# creating new model that we can use to investigate variance inflation
new_cols <- colnames(df)[-c(5,13)]
model <- lm(medv ~ ., df %>% select(-c(indus, nox, dis)))
summary(model)

Call:
lm(formula = medv ~ ., data = df %>% select(-c(indus, nox, dis)))

Residuals:
    Min    1Q    Median    3Q    Max
```

```
-16.9388 -3.0974 -0.7082 1.8472 28.3443
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 21.655695
                        4.215323
                                   5.137 4.01e-07 ***
            -0.091908
crim
                        0.034722 -2.647 0.008380 **
zn
             0.008794
                        0.012670
                                   0.694 0.487957
chas
             2.952830
                        0.913519
                                   3.232 0.001309 **
             4.100202
                        0.439135
                                   9.337 < 2e-16 ***
rm
age
             0.020892
                        0.012195
                                   1.713 0.087315 .
             0.251852
                        0.067890
                                   3.710 0.000231 ***
rad
                        0.003469 -3.584 0.000371 ***
tax
            -0.012434
                        0.129206 -6.862 2.03e-11 ***
ptratio
            -0.886594
lstat
            -0.573951
                        0.053177 -10.793 < 2e-16 ***
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
```

Residual standard error: 5.084 on 496 degrees of freedom Multiple R-squared: 0.6999, Adjusted R-squared: 0.6944 F-statistic: 128.5 on 9 and 496 DF, p-value: < 2.2e-16

- Variance inflation = if you have 2 variables that are highly correlated with one another, if you change one variable, then you would expect change in the other variable in the same direction (can't hold that variable constant while increasing the other).
- 1. variance is used to compute standard errors
- 2. if standard error goes up, then p-value goes up, so it becomes more significant & less likely to reject the null hypothesis
- 3. inflating it with sum of other variables
- 4. high variance inflation = gets very high, so it becomes very insignificant

```
# inflation variance factor table
library(car)
library(knitr)
library(rmarkdown)
vif_model <- lm(medv ~ ., df)
vif(vif_model) %>% knitr::kable()
```

	X
crim	1.767486
zn	2.298459
indus	3.987181

	х
chas	1.071168
nox	4.369093
rm	1.912532
age	3.088232
dis	3.954037
rad	7.445301
tax	9.002158
ptratio	1.797060
lstat	2.870776

- This table measures the of the amount of multicolinearity in a set of multiple regression variables.
- high inflation factor = anything greater than 2
- low inflation factor = anything less than 2

Stepwise Regression

- can drop a subset of the variables that don't explain enough the variability in the R^2
- by adding more variables, R^2 goes up -> now you can ask what variables leads to the highest increase in R^2 /more significant/good predictors and have low variance inflation

Coefficients:

(Intercept)	crim	zn	indus	chas	nox
41.617270	-0.121389	0.046963	0.013468	2.839993	-18.758022
rm	age	dis	rad	tax	ptratio
3.658119	0.003611	-1.490754	0.289405	-0.012682	-0.937533
lstat					
-0.552019					

Forward Selection

- Starts with no covariates
- Add each covariate 1 by 1 in order of variables importance, increasing R^2

```
Start: AIC=2246.51
```

 $medv \sim 1$

```
Df Sum of Sq
                         RSS
+ lstat
               23243.9 19472 1851.0
+ rm
           1
              20654.4 22062 1914.2
              11014.3 31702 2097.6
+ ptratio 1
               9995.2 32721 2113.6
+ indus
           1
               9377.3 33339 2123.1
+ tax
           1
               7800.1 34916 2146.5
+ nox
+ crim
           1
               6440.8 36276 2165.8
               6221.1 36495 2168.9
+ rad
           1
+ age
           1
               6069.8 36647 2171.0
               5549.7 37167 2178.1
+ zn
           1
+ dis
           1
                2668.2 40048 2215.9
                1312.1 41404 2232.7
+ chas
                       42716 2246.5
<none>
```

Step: AIC=1851.01

 $medv \sim lstat$

Df Sum of Sq RSS AIC + rm 1 4033.1 15439 1735.6 + ptratio 1 2670.1 16802 1778.4

```
+ chas 1 786.3 18686 1832.2

+ dis 1 772.4 18700 1832.5

+ age 1 304.3 19168 1845.0

+ tax 1 274.4 19198 1845.8

+ zn 1 160.3 19312 1848.8

+ crim 1 146.9 19325 1849.2

+ indus 1 98.7 19374 1850.4

<none> 19472 1851.0

+ rad 1 25.1 19447 1852.4

+ nox 1 4.8 19468 1852.9
```

Step: AIC=1735.58
medv ~ lstat + rm

		Df	Sum of	Sq	RSS	AIC
+	ptratio	1	1711	.32	13728	1678.1
+	chas	1	548	.53	14891	1719.3
+	tax	1	425	.16	15014	1723.5
+	dis	1	351	. 15	15088	1725.9
+	crim	1	311	.42	15128	1727.3
+	rad	1	180	.45	15259	1731.6
+	indus	1	61	.09	15378	1735.6
<1	none>				15439	1735.6
+	zn	1	56	.56	15383	1735.7
+	age	1	20	.18	15419	1736.9
+	nox	1	14	.90	15424	1737.1

Step: AIC=1678.13
medv ~ lstat + rm + ptratio

		Df	Sum of Sq	RSS	AIC
+	dis	1	499.08	13229	1661.4
+	chas	1	377.96	13350	1666.0
+	crim	1	122.52	13606	1675.6
+	age	1	66.24	13662	1677.7
<r< td=""><td>ione></td><td></td><td></td><td>13728</td><td>1678.1</td></r<>	ione>			13728	1678.1
+	tax	1	44.36	13684	1678.5
+	nox	1	24.81	13703	1679.2
+	zn	1	14.96	13713	1679.6
+	rad	1	6.07	13722	1679.9
+	indus	1	0.83	13727	1680.1

Step: AIC=1661.39

```
medv ~ lstat + rm + ptratio + dis
       Df Sum of Sq
                      RSS
                             AIC
             759.56 12469 1633.5
+ nox
+ chas
             267.43 12962 1653.1
+ indus 1
             242.65 12986 1654.0
+ tax
        1
             240.34 12989 1654.1
+ crim
        1
            233.54 12995 1654.4
            144.81 13084 1657.8
+ zn
        1
        1
            61.36 13168 1661.0
+ age
                    13229 1661.4
<none>
              22.40 13206 1662.5
+ rad
Step: AIC=1633.47
medv ~ lstat + rm + ptratio + dis + nox
       Df Sum of Sq RSS
                             AIC
             328.27 12141 1622.0
+ chas
        1
+ zn
        1
             151.71 12318 1629.3
+ crim
        1
            141.43 12328 1629.7
            53.48 12416 1633.3
+ rad
<none>
                    12469 1633.5
+ indus 1
            17.10 12452 1634.8
+ tax
        1
              10.50 12459 1635.0
+ age
        1
              0.25 12469 1635.5
Step: AIC=1621.97
medv ~ lstat + rm + ptratio + dis + nox + chas
       Df Sum of Sq RSS
           164.406 11977 1617.1
+ zn
       1 116.330 12025 1619.1
+ crim
        1 58.556 12082 1621.5
+ rad
<none>
                    12141 1622.0
+ indus 1
            26.274 12115 1622.9
             4.187 12137 1623.8
+ tax
        1
              2.331 12139 1623.9
+ age
        1
Step: AIC=1617.07
medv ~ lstat + rm + ptratio + dis + nox + chas + zn
       Df Sum of Sq
                      RSS
                             AIC
       1 170.902 11806 1611.8
+ crim
```

```
<none>
                    11977 1617.1
             31.773 11945 1617.7
+ tax
        1
             28.311 11948 1617.9
+ rad
        1
+ indus 1 27.377 11949 1617.9
            0.071 11977 1619.1
+ age
        1
Step: AIC=1611.8
medv ~ lstat + rm + ptratio + dis + nox + chas + zn + crim
       Df Sum of Sq RSS
                             AIC
        1 155.006 11651 1607.1
+ rad
<none>
                    11806 1611.8
+ indus 1 24.957 11781 1612.7
             1.418 11804 1613.7
+ tax
        1
            0.178 11806 1613.8
+ age
        1
Step: AIC=1607.11
medv ~ lstat + rm + ptratio + dis + nox + chas + zn + crim +
   rad
       Df Sum of Sq RSS
                          AIC
            298.573 11352 1596.0
+ tax
<none>
                    11651 1607.1
+ indus 1
            44.346 11606 1607.2
+ age
        1
            0.581 11650 1609.1
Step: AIC=1595.98
medv ~ lstat + rm + ptratio + dis + nox + chas + zn + crim +
   rad + tax
       Df Sum of Sq RSS
                             AIC
<none>
                    11352 1596.0
             1.6865 11350 1597.9
+ age
        1
+ indus 1
             1.0784 11351 1597.9
  summary(forward_model)
Call:
lm(formula = medv ~ lstat + rm + ptratio + dis + nox + chas +
    zn + crim + rad + tax, data = df)
```

Residuals:

```
Min 1Q Median 3Q Max -15.1814 -2.7625 -0.6243 1.8448 26.3920
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept)
             41.451747
                          4.903283
                                     8.454 3.18e-16 ***
lstat
             -0.546509
                          0.047442 -11.519
                                            < 2e-16 ***
              3.672957
                          0.409127
                                     8.978 < 2e-16 ***
rm
ptratio
             -0.930961
                          0.130423
                                   -7.138 3.39e-12 ***
                          0.187675
                                    -8.078 5.08e-15 ***
dis
             -1.515951
nox
            -18.262427
                          3.565247
                                    -5.122 4.33e-07 ***
              2.871873
chas
                          0.862591
                                     3.329 0.000935 ***
              0.046191
                          0.013673
                                     3.378 0.000787 ***
zn
             -0.121665
                          0.032919
                                    -3.696 0.000244 ***
crim
              0.283932
                          0.063945
                                     4.440 1.11e-05 ***
rad
             -0.012292
                          0.003407 -3.608 0.000340 ***
tax
___
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
```

Residual standard error: 4.789 on 495 degrees of freedom Multiple R-squared: 0.7342, Adjusted R-squared: 0.7289 F-statistic: 136.8 on 10 and 495 DF, p-value: < 2.2e-16

- The lower the AIC value, the better
- Add the variables that have the result in the lowest AIC value at that point
- The AIC next to each variable, says which variable that would be added next would make the AIC value the lowest
- lstat is first one added as it makes the AIC value the smallest out of all the variables
- forward selection -> keep building model by including variables
- at the end, if you add age or indus, the AIC value increases, which means you should stop before adding those variables (don't use those variables in the model)

Backward Selection

- Start with full model
- Remove variables 1 by 1 to see which results in a decrease in the AIC value

```
Start: AIC=1599.85
medv ~ crim + zn + indus + chas + nox + rm + age + dis + rad +
```

tax + ptratio + lstat

- zn

1

261.75 11614 1605.5

```
Df Sum of Sq
                        RSS
                              AIC
- indus
                 1.08 11350 1597.9
                1.69 11351 1597.9
- age
<none>
                      11349 1599.8
- chas
               245.31 11595 1608.7
          1 256.28 11606 1609.2
- tax
          1 263.59 11613 1609.5
- zn
          1 311.49 11661 1611.6
- crim
          1 430.71 11780 1616.7
- rad
- nox
              546.10 11896 1621.6
          1
- ptratio 1
            1157.70 12507 1647.0
          1 1258.52 12608 1651.1
- dis
          1 1744.36 13094 1670.2
- rm
- lstat
          1 2733.54 14083 1707.0
Step: AIC=1597.9
medv ~ crim + zn + chas + nox + rm + age + dis + rad + tax +
   ptratio + lstat
         Df Sum of Sq RSS
                              AIC
- age
                 1.69 11352 1596.0
<none>
                      11350 1597.9
- chas
               251.21 11602 1607.0
          1
          1 262.99 11614 1607.5
- zn
          1 299.68 11650 1609.1
- tax
          1 313.07 11664 1609.7
- crim
          1 453.61 11804 1615.7
- rad
- nox
          1
              574.23 11925 1620.9
- ptratio 1 1168.01 12518 1645.5
- dis
          1 1333.19 12684 1652.1
            1750.50 13101 1668.5
- rm
          1
- lstat
          1 2743.21 14094 1705.4
Step: AIC=1595.98
medv ~ crim + zn + chas + nox + rm + dis + rad + tax + ptratio +
    lstat
         Df Sum of Sq
                       RSS
                              AIC
                      11352 1596.0
<none>
- chas
               254.21 11606 1605.2
          1
```

```
298.57 11651 1607.1
- tax
           1
- crim
           1
               313.27 11666 1607.8
               452.16 11804 1613.7
- rad
           1
               601.74 11954 1620.1
- nox
           1
- ptratio 1
             1168.51 12521 1643.5
              1496.35 12848 1656.6
- dis
- rm
           1
              1848.38 13201 1670.3
- lstat
              3043.23 14395 1714.2
  summary(backward_model)
Call:
lm(formula = medv ~ crim + zn + chas + nox + rm + dis + rad +
    tax + ptratio + lstat, data = df)
Residuals:
     Min
               1Q
                    Median
                                 3Q
                                         Max
-15.1814 -2.7625
                  -0.6243
                             1.8448
                                     26.3920
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                                  8.454 3.18e-16 ***
(Intercept) 41.451747
                         4.903283
crim
                         0.032919 -3.696 0.000244 ***
             -0.121665
              0.046191
                         0.013673 3.378 0.000787 ***
zn
chas
              2.871873
                         0.862591 3.329 0.000935 ***
nox
            -18.262427
                         3.565247 -5.122 4.33e-07 ***
             3.672957
                         0.409127 8.978 < 2e-16 ***
rm
dis
             -1.515951
                         0.187675 -8.078 5.08e-15 ***
             0.283932
                         0.063945
                                   4.440 1.11e-05 ***
rad
                         0.003407 -3.608 0.000340 ***
             -0.012292
tax
                         0.130423 -7.138 3.39e-12 ***
             -0.930961
ptratio
                         0.047442 -11.519 < 2e-16 ***
             -0.546509
lstat
___
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
Residual standard error: 4.789 on 495 degrees of freedom
```

Residual standard error: 4.789 on 495 degrees of freedom Multiple R-squared: 0.7342, Adjusted R-squared: 0.7289 F-statistic: 136.8 on 10 and 495 DF, p-value: < 2.2e-16

- removes variables that would give it a higher AIC until there are no more of that type
- forward selection and backward selection DON'T always give us the same model

Using both selection methods

• use the full model for both and direction as 'both'

```
selected_model <- step(full_model, direction = 'both',</pre>
                         scope = formula(full_model))
Start: AIC=1599.85
medv ~ crim + zn + indus + chas + nox + rm + age + dis + rad +
    tax + ptratio + lstat
         Df Sum of Sq
                        RSS
                                AIC
                  1.08 11350 1597.9
- indus
- age
                  1.69 11351 1597.9
<none>
                       11349 1599.8
- chas
               245.31 11595 1608.7
           1
               256.28 11606 1609.2
- tax
           1
- zn
           1
               263.59 11613 1609.5
           1 311.49 11661 1611.6
- crim
- rad
           1
             430.71 11780 1616.7
               546.10 11896 1621.6
- nox
- ptratio 1 1157.70 12507 1647.0
- dis
           1
              1258.52 12608 1651.1
             1744.36 13094 1670.2
- rm
           1
- lstat
           1
              2733.54 14083 1707.0
Step: AIC=1597.9
medv ~ crim + zn + chas + nox + rm + age + dis + rad + tax +
   ptratio + lstat
          Df Sum of Sq
                        RSS
                                AIC
           1
                  1.69 11352 1596.0
- age
<none>
                       11350 1597.9
+ indus
                  1.08 11349 1599.8
           1
                251.21 11602 1607.0
- chas
           1
- zn
               262.99 11614 1607.5
               299.68 11650 1609.1
- tax
           1
             313.07 11664 1609.7
- crim
          1
- rad
               453.61 11804 1615.7
           1
           1
               574.23 11925 1620.9
- nox
- ptratio 1
             1168.01 12518 1645.5
- dis
           1
              1333.19 12684 1652.1
```

```
- rm 1 1750.50 13101 1668.5
- lstat 1 2743.21 14094 1705.4
```

Step: AIC=1595.98

medv ~ crim + zn + chas + nox + rm + dis + rad + tax + ptratio +
 lstat

	Df	Sum of Sq	RSS	AIC	
<none></none>			11352	1596.0	
+ age	1	1.69	11350	1597.9	
+ indus	1	1.08	11351	1597.9	
- chas	1	254.21	11606	1605.2	
- zn	1	261.75	11614	1605.5	
- tax	1	298.57	11651	1607.1	
- crim	1	313.27	11666	1607.8	
- rad	1	452.16	11804	1613.7	
- nox	1	601.74	11954	1620.1	
- ptratio	1	1168.51	12521	1643.5	
- dis	1	1496.35	12848	1656.6	
- rm	1	1848.38	13201	1670.3	
- lstat	1	3043.23	14395	1714.2	

summary(selected_model)

Call:

```
lm(formula = medv ~ crim + zn + chas + nox + rm + dis + rad +
tax + ptratio + lstat, data = df)
```

Residuals:

```
Min 1Q Median 3Q Max -15.1814 -2.7625 -0.6243 1.8448 26.3920
```

Coefficients:

Estimate Std. Error t value Pr(>|t|) (Intercept) 41.451747 4.903283 8.454 3.18e-16 *** crim -0.121665 0.032919 -3.696 0.000244 *** zn 0.046191 0.013673 3.378 0.000787 *** chas nox -18.262427 3.565247 -5.122 4.33e-07 *** 8.978 < 2e-16 *** 3.672957 0.409127 rm0.187675 -8.078 5.08e-15 *** dis -1.515951

```
rad 0.283932 0.063945 4.440 1.11e-05 ***
tax -0.012292 0.003407 -3.608 0.000340 ***
ptratio -0.930961 0.130423 -7.138 3.39e-12 ***
lstat -0.546509 0.047442 -11.519 < 2e-16 ***
```

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

Residual standard error: 4.789 on 495 degrees of freedom Multiple R-squared: 0.7342, Adjusted R-squared: 0.7289 F-statistic: 136.8 on 10 and 495 DF, p-value: < 2.2e-16