

lab 5 Jin Kweon (3032235207)

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For the next two sections of this lab assignment, assume that the data comes from the following model:

$y_i = x_i^T \beta + \epsilon_i$ where $i = 1, \dots, n$ and noises are iid normal with mean 0 and variance σ^2 and x is fixed.

```
lm1 <- lm(mpg ~ disp + hp, data = mtcars)
sum <- summary(lm1)
sum

##
## Call:
## lm(formula = mpg ~ disp + hp, data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.7945 -2.3036 -0.8246  1.8582  6.9363
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  30.735904   1.331566   23.083 < 2e-16 ***
## disp        -0.030346   0.007405   -4.098 0.000306 ***
## hp          -0.024840   0.013385   -1.856 0.073679 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.127 on 29 degrees of freedom
## Multiple R-squared:  0.7482, Adjusted R-squared:  0.7309
## F-statistic: 43.09 on 2 and 29 DF,  p-value: 2.062e-09

lowbound <- sum$coefficients[2,1] - qt(0.975, 29) * sum$coefficients[2,2]
upbound <- sum$coefficients[2,1] + qt(0.975, 29) * sum$coefficients[2,2]

list(lowbound, upbound)

## [[1]]
## [1] -0.04549091
##
## [[2]]
## [1] -0.01520165

confint(lm1, "disp", level = 0.95)

##              2.5 %       97.5 %
## disp -0.04549091 -0.01520165
```

1. Our null is beta being equal to zero. (whether we drop the variable or not)
2. Alternative hypothesis will be when beta is not being equal to zero. It is two-sided as t-test contains negative and positive numbers. "***" stands for a number between 0 and 0.001. "." stands for a number between 0.05 and 0.1.
3. No we do not reject the null. P-value is not in the critical region.

$MSE = bias^2 + variance$.

Q. what do you mean by “you can never prove the null”? ==> cuz we never know the null is really true. We just either reject or not reject the null based on our testing. Q. what is orthogonal polynomials in “raw” in `poly()` function? ==> if we use orthogonal polynomials, then we do gram schmidt to get orthognoal vectors (for inner product $\int_0^1 fg = 0$) Q. So, training set is the entire data set and test set is also the entired data set, as well??? ==> yes

```
#4.
tstat <- (sum$coefficients[2,1] + 0.05) / sum$coefficients[2,2]
pt(tstat, 29, lower.tail = F)

## [1] 0.00638547

print("We would reject the one-way test.")

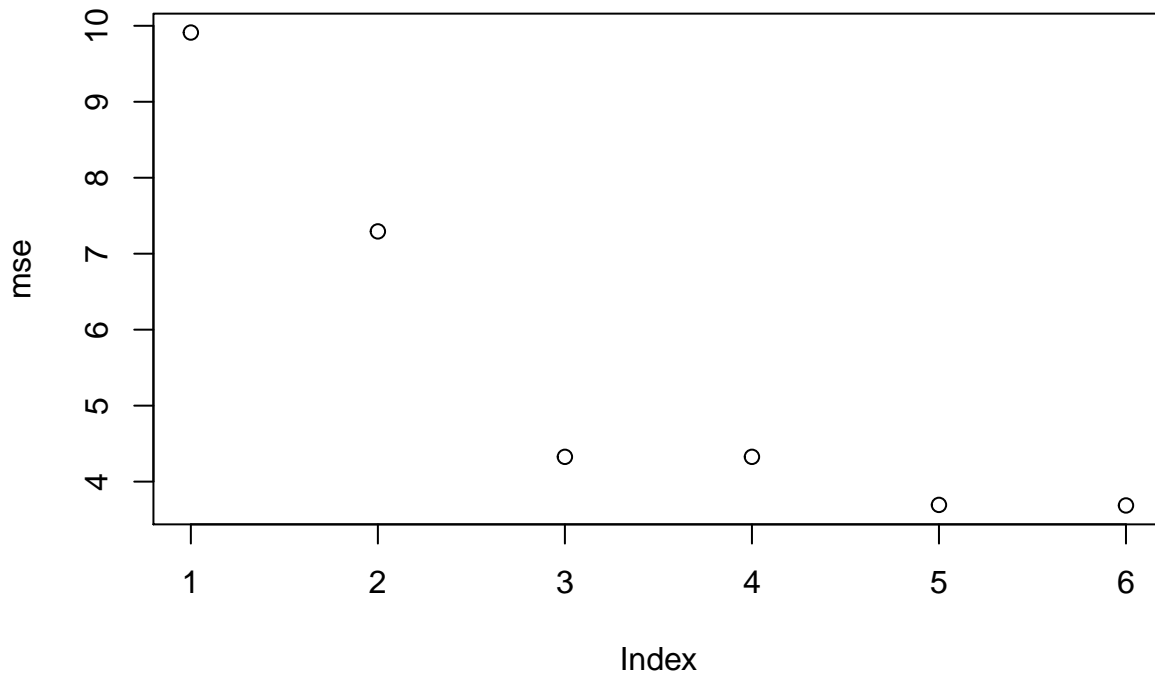
## [1] "We would reject the one-way test."

y <- mtcars$mpg
x <- mtcars$displ
mse <- c()

mse_fn <- function(emp_vec){
  for (i in 1:6){
    emp_vec[i] <- (sum((y - lm(y ~ poly(x, i, raw = T))$fitted.values)^2)) / nrow(mtcars)
  }
  return(emp_vec)
}

mse <- mse_fn(mse)

plot(mse)
```



The more power/predictors, the less mse in general.

Q. And, holdout random sampling should be without replacement. Right? ==> yes

Q. So for hold out, we get beta hats from training and apply these beta hats into the test set X and get y hat and calculate MSE there. Right? ==> Yes.

Q. And, when we add up test and training, it should be entire data. Right? Meaning I should get 80% of entire data as training, and the left over will be test?? ==> Yes cuz we dont model to cheat. Test set cannot be partial of trainig set.

```
for (j in 1:5){
  rand <- c(sample(nrow(mtcars) , size = as.integer(nrow(mtcars) * 0.2), replace = F))

  testsize <- as.integer(nrow(mtcars) * 0.2)
  ytest <- y[rand]
  xtest <- x[rand]

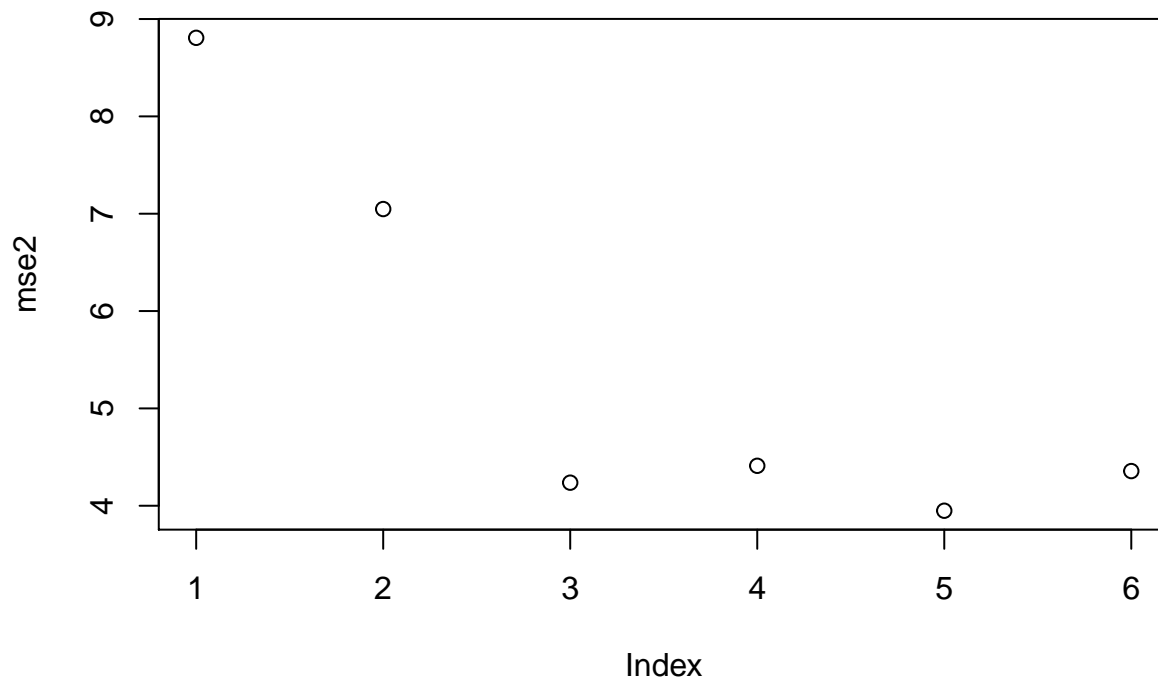
  ytrain <- y[-(rand)]
  xtrain <- x[-(rand)]

  mse2 <- c()

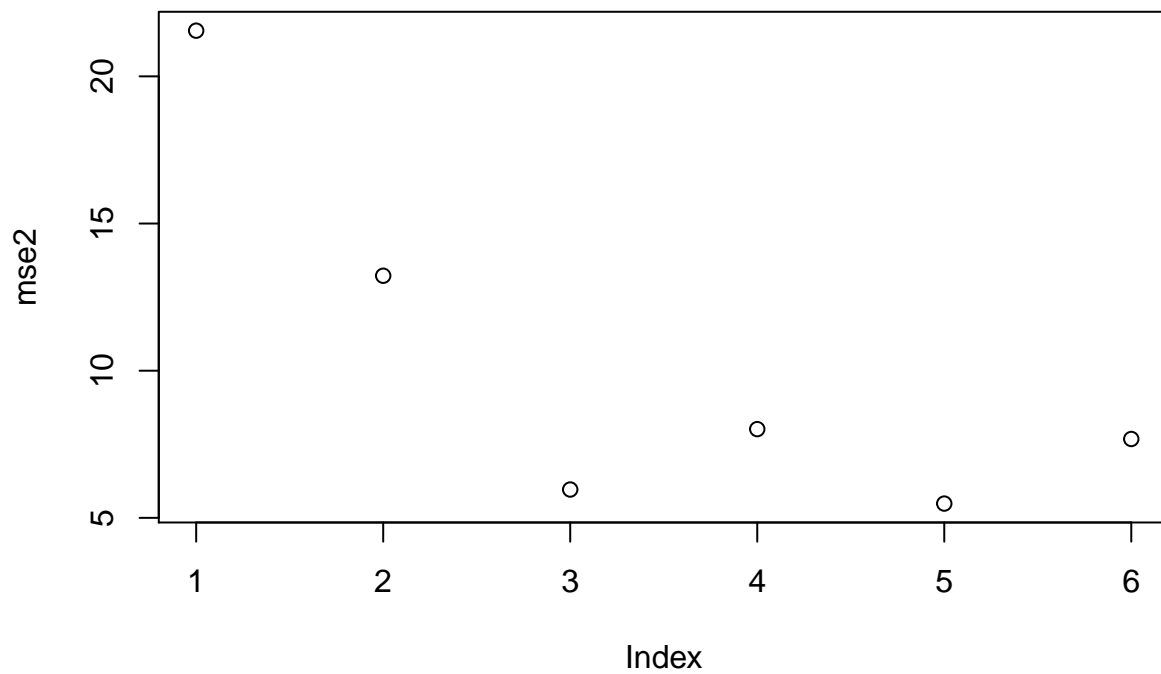
  for (i in 1:6){
    coef <- lm(ytrain ~ poly(xtrain, i, raw = T))$coefficients #get beta hats from training sets.
    yhat_test <- cbind(1, poly(xtest, i, raw = T)) %*% coef #get yhat with beta hats from training and
    mse2[i] <- sum((ytest - yhat_test)^2) / testsize
  }

  print(mse2)
  plot(mse2)
}
```

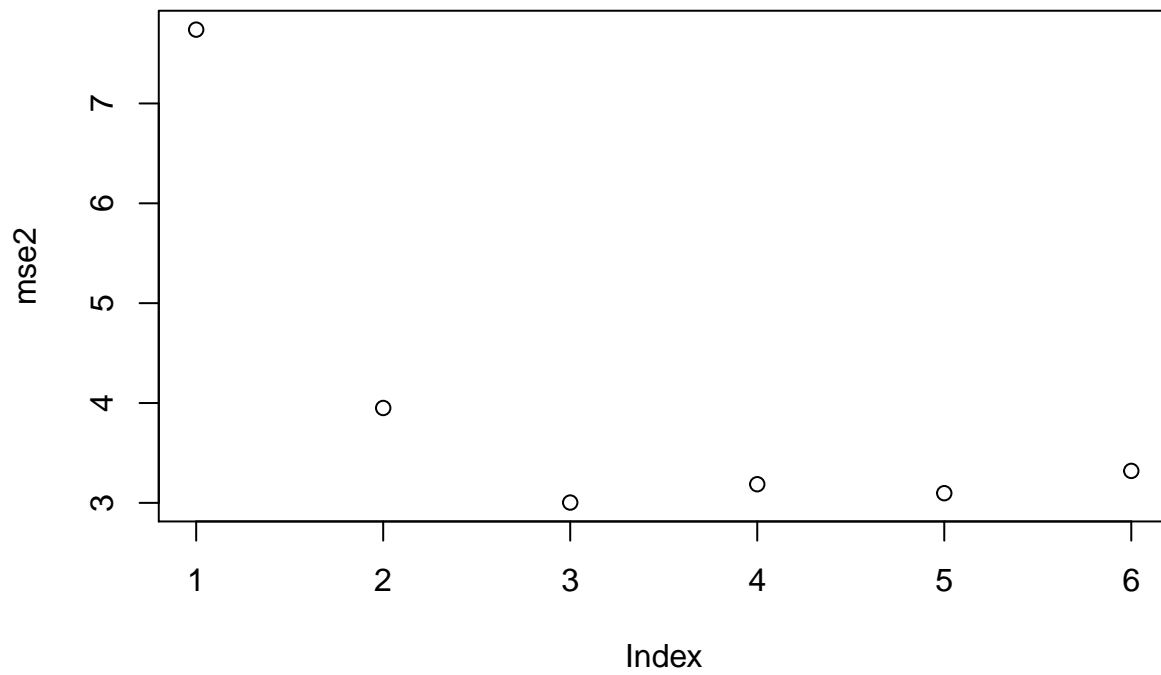
```
## [1] 8.806870 7.047812 4.236967 4.410343 3.948822 4.356148
```



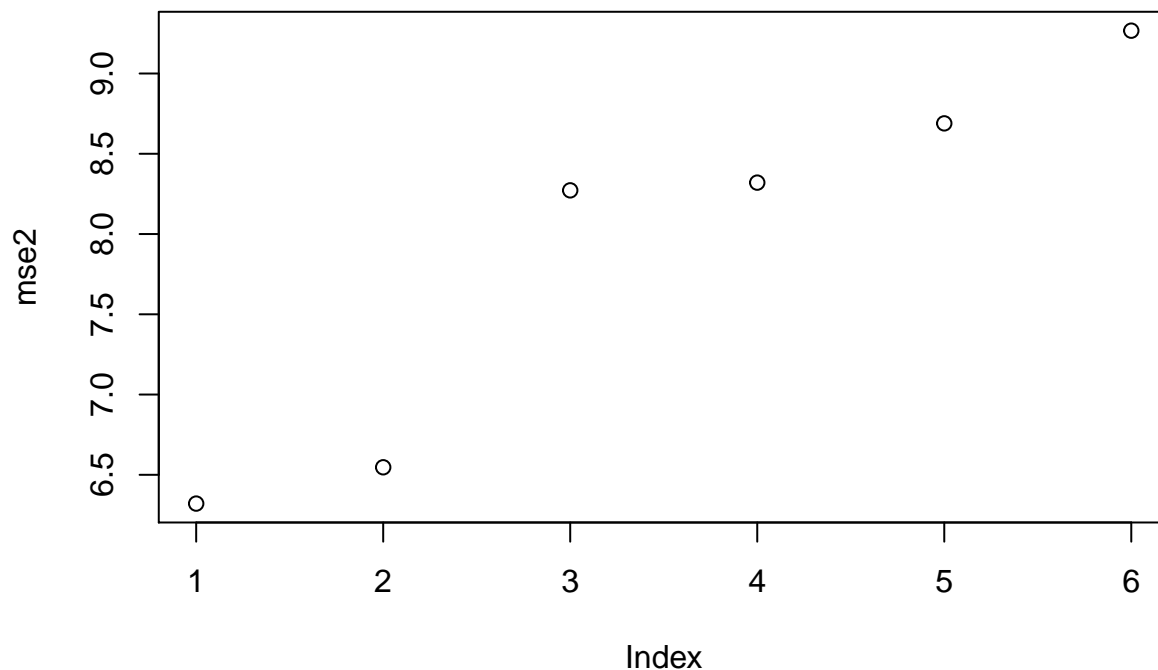
```
## [1] 21.552001 13.225544 5.962679 8.014849 5.485204 7.679725
```



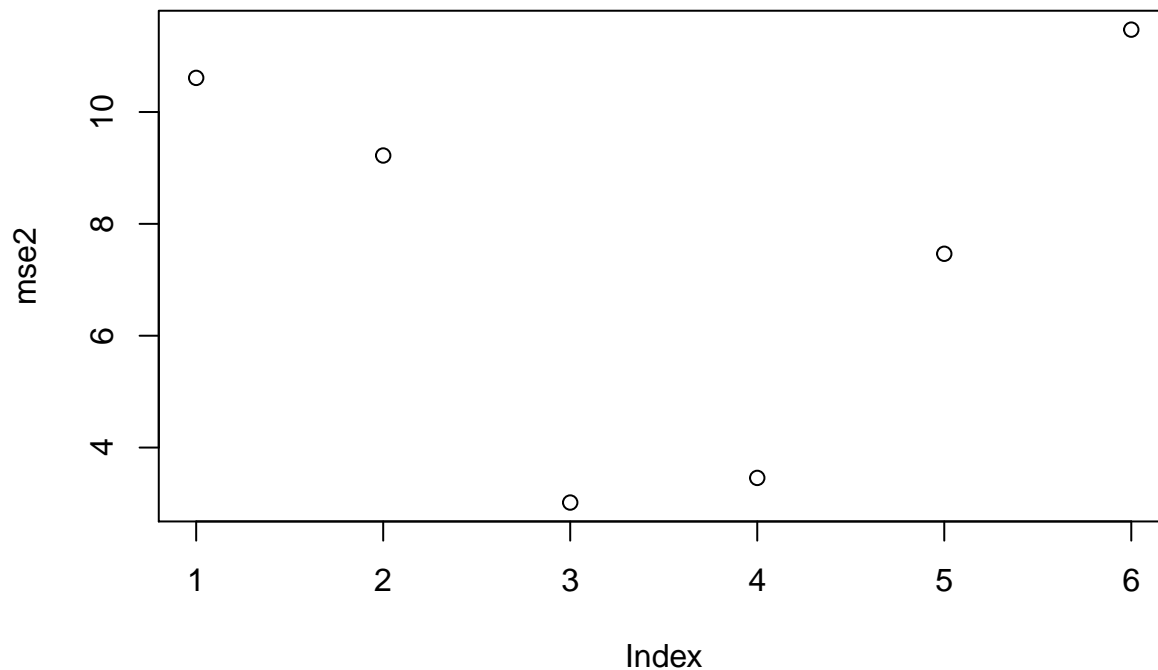
```
## [1] 7.739102 3.950187 3.003422 3.186763 3.097149 3.320194
```



```
## [1] 6.320833 6.546851 8.272179 8.320337 8.689484 9.266805
```



```
## [1] 10.610230  9.222065  3.017085  3.457284  7.465131 11.472516
```



Q. So, fold can have different numbers as they are randomly picked?? ==> Right.

Q. So, can I say cross validation MSE is reasonable, as they are randomly picked and get the average of each fold of MSEs? ==> Yes. (holdout and cross validation are good. Bootstrap is not really common, meaning not really good.)

Q. How to assign different names for each looping cv several times?.... ==> assign them as lists.

Q. For number 5, so, by the definition from [https://en.wikipedia.org/wiki/Cross-validation_\(statistics\)](https://en.wikipedia.org/wiki/Cross-validation_(statistics)), I don't think n-fold cross validation is the same as leave one out cross validation. n-fold cv has n-folds, and thus, there is only 1 element for each fold and we compute n times. However, leave one out cv has 1 observation as

a testing set, and leave n-1 as a training set but those n-1 does not have to be divided into n-1 groups. So, n-1 can be divided into 2 groups, 3 groups, etc. ??? ==> No. they are the same. Wikipedia did not say it but when leave one out cross validation, other rest should have 1 for each fold.

```
fold <- createFolds(mtcars$disp)
fold

## $Fold01
## [1] 3 5 10 24
##
## $Fold02
## [1] 17 22 27
##
## $Fold03
## [1] 7 14 20
##
## $Fold04
## [1] 6 18 26 29
##
## $Fold05
## [1] 9 15
##
## $Fold06
## [1] 11 16
##
## $Fold07
## [1] 13 19 21
##
## $Fold08
## [1] 2 12 25 30
##
## $Fold09
## [1] 8 28 31
##
## $Fold10
## [1] 1 4 23 32

for (k in 1:3){
  cv <- matrix(0, 6, 10)
  fold <- createFolds(mtcars$disp)
  for (i in 1:10){
    ytest <- y[fold[[i]]] #Assign test on the fold
    xtest <- x[fold[[i]]]

    ytrain <- y[-(fold[[i]])] #Assign train on all observations except the ones in the fold
    xtrain <- x[-(fold[[i]])]

    for (j in 1:6){
      coef2 <- lm(ytrain ~ poly(xtrain, j, raw = T))$coefficients #get beta hats from training sets.
      yhat_test2 <- cbind(1, poly(xtest, j, raw = T)) %*% coef2 #get yhat with beta hats from training
      cv[j,i] <- sum((ytest - yhat_test2)^2) / length(ytest)
    }
  }
  print(cv)
}
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
## [1,] 21.502298 13.55437 10.023155 2.084125 7.342543 6.649979 2.366911
## [2,] 11.328696 15.88113 12.609010 1.116814 4.345578 9.650645 1.089579
## [3,]  3.924389 10.09592  5.443970 5.372020 4.857411 3.522418 4.959331
## [4,]  3.947696 10.40392  5.458696 5.625817 4.885671 3.541559 5.135622
## [5,]  3.150453 10.02968 16.216497 5.361999 2.632756 3.295990 4.471313
## [6,]  3.587025 10.10576 17.462760 5.458907 3.726852 3.873860 4.683216
##           [,8]      [,9]      [,10]
## [1,] 7.670977 50.804224 2.1172852
## [2,] 6.892349 34.049111 0.4335074
## [3,] 3.881686 14.103979 3.5310115
## [4,] 4.077362 15.422091 3.5510476
## [5,] 1.902578  8.497730 4.5936129
## [6,] 2.217097  9.061066 4.7348782
##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
## [1,] 19.884546 18.106845 6.042862 21.241582 16.711482 5.945914 2.1209551
## [2,]  8.606172  7.926383 4.324356 19.354133 12.702997 9.907980 0.9645657
## [3,]  3.618587  3.469097 2.551448  6.053712  7.629742 5.266013 5.8681455
## [4,]  3.768515  3.469427 2.633993  6.327842 11.349675 5.348232 6.0764542
## [5,]  2.578725  6.580390 1.267396 10.559475  8.081594 4.991494 5.9368162
## [6,]  3.659892  7.279847 1.373708 10.604071  8.074771 4.997383 6.3313302
##           [,8]      [,9]      [,10]
## [1,] 0.5599166 5.895169 19.656879
## [2,] 11.8034551 4.981495 14.728957
## [3,]  4.8372709 4.276961  9.289342
## [4,]  5.0968753 4.693121 12.138803
## [5,] 12.0819645 2.831920  8.932148
## [6,] 27.6431618 2.808900  9.068566
##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
## [1,] 12.180128 4.4267132  5.449726 0.7246283 12.109235 24.114481 10.310837
## [2,]  4.638807 0.5702108 13.472327 0.7539829  5.691774 12.913576 12.448732
## [3,]  1.241309 0.9880024  6.530877 6.8884509  1.623071  4.338218  7.688937
## [4,]  1.358940 1.0698812  7.098599 6.8885207  1.690765  4.338325  7.826365
## [5,]  2.449048 0.2684121  4.784445 5.5251688  1.983069  5.621179  5.891564
## [6,]  2.446965 0.3748168  6.758127 5.4891847  1.938780  5.742943  6.199265
##           [,8]      [,9]      [,10]
## [1,] 21.237987 18.718264  3.186830
## [2,] 18.537649 14.266714  6.147953
## [3,]  2.744865  8.379066 11.431485
## [4,]  2.835219 12.003791 11.631290
## [5,]  6.468778 12.270887 10.507837
## [6,] 25.074642 13.900627 10.556143
```

```
cvmse <- matrix(0, 6, 10)
for (i in 1:10){
  ytest <- y[fold[[i]]] #Assign test on the fold
  xtest <- x[fold[[i]]]

  ytrain <- y[-(fold[[i]])] #Assign train on all observations except the ones in the fold
  xtrain <- x[-(fold[[i]])]

  for (j in 1:6){
    coef2 <- lm(ytrain ~ poly(xtrain, j, raw = T))$coefficients #get beta hats from training sets.
    yhat_test2 <- cbind(1, poly(xtest, j, raw = T)) %*% coef2 #get yhat with beta hats from training an
```

```

    cvmse[j,i] <- sum((ytest - yhat_test2)^2) / length(ytest)
  }
}

```

```
print(cvmse)
```

```

##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
## [1,] 12.180128 4.4267132 5.449726 0.7246283 12.109235 24.114481 10.310837
## [2,]  4.638807 0.5702108 13.472327 0.7539829  5.691774 12.913576 12.448732
## [3,]  1.241309 0.9880024  6.530877 6.8884509  1.623071  4.338218  7.688937
## [4,]  1.358940 1.0698812  7.098599 6.8885207  1.690765  4.338325  7.826365
## [5,]  2.449048 0.2684121  4.784445 5.5251688  1.983069  5.621179  5.891564
## [6,]  2.446965 0.3748168  6.758127 5.4891847  1.938780  5.742943  6.199265
##           [,8]      [,9]      [,10]
## [1,] 21.237987 18.718264  3.186830
## [2,] 18.537649 14.266714  6.147953
## [3,]  2.744865  8.379066 11.431485
## [4,]  2.835219 12.003791 11.631290
## [5,]  6.468778 12.270887 10.507837
## [6,] 25.074642 13.900627 10.556143

```

#Or, use predict function to do it for you.....

```

cvmse2 <- matrix(0, 6, 10)
newmtcar <- as.data.frame(cbind(mpg = mtcars$mpg, disp = mtcars$disp))

```

```

for (i in 1:10){
  ytest2 <- y[fold[[i]]] #Assign test on the fold
  xtest2 <- x[fold[[i]]]

  ytrain2 <- y[-(fold[[i])]] #Assign train on all observations except the ones in the fold
  xtrain2 <- x[-(fold[[i])]]

  for (j in 1:6){
    lms2 <- lm(mpg ~ poly(disp, j, raw = T), data = newmtcar[-(fold[[i])], ]) #get beta hats from train
    yhat_test3 <- predict(lms2, newdata = newmtcar[fold[[i]], ]) #get yhat with beta hats from trainin
    cvmse2[j,i] <- sum((ytest2 - yhat_test3)^2) / length(ytest2)
  }
}
print(cvmse2)

```

```

##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
## [1,] 12.180128 4.4267132 5.449726 0.7246283 12.109235 24.114481 10.310837
## [2,]  4.638807 0.5702108 13.472327 0.7539829  5.691774 12.913576 12.448732
## [3,]  1.241309 0.9880024  6.530877 6.8884509  1.623071  4.338218  7.688937
## [4,]  1.358940 1.0698812  7.098599 6.8885207  1.690765  4.338325  7.826365
## [5,]  2.449048 0.2684121  4.784445 5.5251688  1.983069  5.621179  5.891564
## [6,]  2.446965 0.3748168  6.758127 5.4891847  1.938780  5.742943  6.199265
##           [,8]      [,9]      [,10]
## [1,] 21.237987 18.718264  3.186830
## [2,] 18.537649 14.266714  6.147953
## [3,]  2.744865  8.379066 11.431485
## [4,]  2.835219 12.003791 11.631290
## [5,]  6.468778 12.270887 10.507837

```



```
## [6,] 25.074642 13.900627 10.556143
```

```
identical(cvmse, cvmse2)
```

```
## [1] TRUE
```

#These codes below will not work because I do not use predict function in one data frame. If I do not use

#<https://stackoverflow.com/questions/31879271/how-can-i-add-a-hashtag-sign-to-many-lines-in-r-command>

```
#for (i in 1:10){
```

```
# ytest2 <- y[fold[[i]]] #Assign test on the fold
```

```
# xtest2 <- x[fold[[i]]]
```

```
# ytrain2 <- y[-(fold[[i]])] #Assign train on all observations except the ones in the fold
```

```
# xtrain2 <- x[-(fold[[i]])]
```

```
#
```

```
# for (j in 1:6){
```

```
# fit <- lm(ytrain2 ~ poly(xtrain2, j, raw = T), data = as.data.frame(cbind(ytrain2, xtrain2)))
```

```
# predict(fit , newdata = cbind(1, poly(xtest2, j, raw = T)))
```

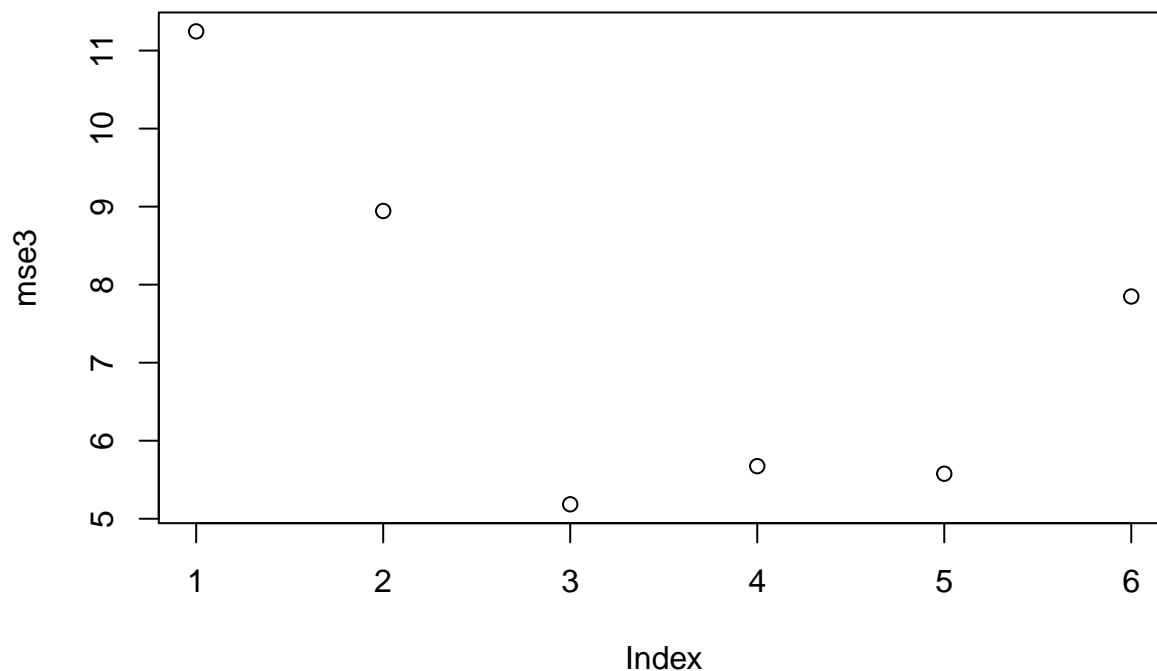
```
# cvmse2[j,i] <- sum((ytest2 - yhat_test3)^2) / length(ytest2)
```

```
# }
```

```
# }
```

```
mse3 <- rowMeans(cvmse)
```

```
plot(mse3)
```



CVMSE does not stay the same, as they are randomly picked. . . .

So, we said leave one out cv because we leave only 1 observation as the testing/validation set and all others as the training set.

In general, Leave p out cross-validation requires training and validating the model C_p^n times, where n is the number of observations in the original sample. There is really no fold concept in leave p out, but leave n out can be equal to n fold cv as we have n groups and have 1 observation for each fold.

Q, When you said “test on the model that is not in the sample,” are you talking about the number itself or the iteration? So, let's say original data is 1 1 2 3 4 5 and when we get sample 1 (first one) 2 2 2 3 5, then do our test set includes the second one but does not include the first one. Right? ==> Right. Order matters, not the number itself. Also, when we have 2 2 2 for training, we use all of them although it is repetition since we sample with replacement. Right? ==> yes!

```
bootmse <- matrix(0, 6, 400)

sd <- c()
for (i in 1:400){
  trainrow <- sample(nrow(mtcars), nrow(mtcars), replace = T) #sample the row number for training.
  ytrain3 <- y[trainrow]
  xtrain3 <- x[trainrow]

  ytest3 <- y[-(trainrow)]
  xtest3 <- x[-(trainrow)]
  for (j in 1:6){
    coef3 <- lm(ytrain3 ~ poly(xtrain3, j, raw = T))$coefficients #get beta hats from training sets.
    yhat_test3 <- cbind(1, poly(xtest3, j, raw = T)) %*% coef3 #get yhat with beta hats from training sets.
    bootmse[j,i] <- sum((ytest3 - yhat_test3)^2) / length(xtest3)
  }
}
print(bootmse)
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]
## [1,] 22.061052   9.675936 11.348389 13.971967 12.204276  6.507804
## [2,] 15.323618  30.074190  9.720899 10.209569  9.268578  7.201565
## [3,]  8.885566  18.833766  5.342195  5.849273  5.101093  3.641401
## [4,]  8.982156 155.488191  5.322446  5.867028  5.061890  3.943989
## [5,]  7.277555 341.923319  6.687805  5.383381  5.612236  7.915487
## [6,]  7.398866  91.794788 20.770506  5.045753 13.636480 11.588735
##           [,7]      [,8]      [,9]     [,10]     [,11]     [,12]
## [1,]  2.454001 10.191379 11.358867 15.447946 10.808976 15.115690
## [2,] 29.676763  7.329827  5.739296 10.191321  9.231082  8.833404
## [3,]  7.849467  4.388817  4.337770  5.978359  4.222998  4.305203
## [4,] 18.050804  5.906793  4.359050  6.104798  4.321362 15.270928
## [5,]  4.982361  5.152581  3.296925  8.014713 13.005971 15.537225
## [6,] 109.030707  9.447026  4.001078 11.605186 12.985683 35.089194
##           [,13]     [,14]     [,15]     [,16]     [,17]     [,18]
## [1,] 10.684626 19.529821 11.300665 18.039699 12.875749 14.337423
## [2,] 12.766664 14.018362  8.482566 13.932012  7.617037 12.388737
## [3,]  5.159947  7.379558  4.015818  7.550341  4.647218  6.082730
## [4,]  4.935652 14.240991  4.165758  7.679880  5.345001 13.303783
## [5,]  4.194980 12.043695  3.137561  5.354689  4.377889  4.564543
## [6,] 129.294045 23.563379 13.717248  5.365954  4.501377 221.233080
##           [,19]     [,20]     [,21]     [,22]     [,23]     [,24]
## [1,] 17.401995 11.828188 13.006104 10.853381  8.688040  8.793158
## [2,] 14.935623  9.776974 19.468908 10.214770  9.522490 44.125412
## [3,]  9.409739  5.349423  6.517791  8.466495  7.382548 30.228018
## [4,] 13.476752  6.895574 21.432296  8.729652  7.404956 186.501530
## [5,] 14.630792  6.560532 15.388338  8.432194  8.614807 116.116696
## [6,] 13.513108  6.607089 93.953932  8.489613 10.909780 600.328162
##           [,25]     [,26]     [,27]     [,28]     [,29]     [,30]     [,31]
## [1,] 13.144907 22.821452 14.576224 16.085422  8.297724 10.410327  5.772798
## [2,]  9.406218 14.722293  8.888291 13.622665  8.767289  7.175251  4.921687
```

```

## [3,] 5.900342 6.307970 3.970212 8.181632 5.329385 7.246015 6.681623
## [4,] 5.902593 6.120140 3.986550 9.707979 8.639913 9.064673 6.587509
## [5,] 5.998782 8.510205 7.369389 9.645525 5.047696 7.504969 6.457262
## [6,] 10.575941 8.649776 7.370740 9.319717 5.911432 10.423298 6.526221
##      [,32]      [,33]      [,34]      [,35]      [,36]      [,37]      [,38]
## [1,] 4.801875 22.233946 8.569925 11.575210 9.542118 11.408036 7.224547
## [2,] 26.675755 16.147871 7.796876 7.067143 7.243052 7.681736 7.063104
## [3,] 14.331731 6.145081 3.884755 7.476884 3.032069 3.624076 4.879164
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## [6,] 7.658195 6.386936 7.592840 9.116316 19.531208 10.598455 9.842479
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## [1,] 10.991913 7.387589 17.311178 10.85415 12.816748 14.775230 10.034097
## [2,] 8.098101 3.753363 20.139577 21.73955 12.952156 11.051902 8.058506
## [3,] 7.225568 6.768444 8.443964 11.97438 8.378885 7.384873 5.634659
## [4,] 7.468197 6.353975 11.660650 189.42449 9.371159 8.281980 5.897203
## [5,] 4.869639 5.558471 54.461955 236.42741 6.397053 6.684599 5.945422
## [6,] 9.378025 5.582139 63.252255 336.07652 9.670347 6.680005 9.136890
##      [,254]  [,255]  [,256]  [,257]  [,258]  [,259]  [,260]
## [1,] 9.550531 15.700214 13.675054 6.784281 13.202587 5.717368 11.248808
## [2,] 8.538005 10.679419 10.420994 9.836177 17.872577 7.470367 11.220036
## [3,] 5.375315 7.133857 6.283022 6.422829 5.976511 6.493448 8.268887
## [4,] 5.825879 7.162541 7.368162 6.295778 28.142397 6.532440 8.215251
## [5,] 4.693904 8.112349 6.583517 6.656861 9.732927 6.936342 7.164600
## [6,] 57.772323 28.363423 7.635638 14.208351 9.834617 14.094327 7.843295
##      [,261]  [,262]  [,263]  [,264]  [,265]  [,266]  [,267]
## [1,] 8.618345 8.708279 12.494380 15.120162 11.762819 6.721778 24.258087
## [2,] 4.661435 8.292578 20.331743 10.499696 7.816188 3.641034 17.952483
## [3,] 3.499116 4.063089 7.348715 4.856321 4.529725 3.055878 6.691019
## [4,] 3.531243 4.489605 6.308158 5.126473 4.438795 5.884852 11.280167
## [5,] 4.731627 6.951029 4.714251 5.850093 5.030206 6.065092 6.933810
## [6,] 5.435820 8.426058 51.679942 8.503906 5.101385 6.069117 8.925998
##      [,268]  [,269]  [,270]  [,271]  [,272]  [,273]
## [1,] 8.329825 12.29964 21.70193 6.664988 1.304611e+01 11.706474
## [2,] 7.271534 44.64130 18.15580 7.587181 4.852247e+01 9.199103
## [3,] 7.813481 14.49891 13.45056 6.232582 3.983131e+00 4.350988
## [4,] 7.693100 157.42019 10.11643 7.990088 7.856203e+02 4.436825
## [5,] 6.501228 62.61464 15.43167 6.784930 5.927783e+03 6.820101
## [6,] 6.652034 4412.68421 765.99531 6.376702 3.307395e+05 10.934784
##      [,274]  [,275]  [,276]  [,277]  [,278]  [,279]  [,280]
## [1,] 15.950398 10.660487 10.943134 12.197619 20.129739 14.032798 13.737020
## [2,] 12.023522 7.517881 12.372681 8.459273 14.408401 12.134072 11.346733
## [3,] 8.653414 6.109746 8.373326 6.724531 8.749184 8.601721 8.427007
## [4,] 9.883952 6.081543 10.150489 7.368039 11.272195 8.562462 8.721303
## [5,] 7.494846 4.408356 8.937218 7.143616 9.097138 9.134924 8.149434
## [6,] 8.067923 4.481921 9.091342 8.007685 10.176934 9.533463 11.184276
##      [,281]  [,282]  [,283]  [,284]  [,285]  [,286]  [,287]
## [1,] 12.875246 14.118784 7.701721 6.909087 20.302109 13.590789 12.853320
## [2,] 9.447111 18.385773 10.356595 8.677709 17.379668 9.887297 9.891466
## [3,] 5.212696 8.268924 5.528690 6.457428 10.081632 5.851260 4.403852
## [4,] 6.187011 12.778495 5.719527 6.695590 10.296840 9.438946 4.502928
## [5,] 7.118340 5.847900 7.064413 6.158669 7.651490 8.272529 5.753368
## [6,] 7.436443 122.151449 13.163852 7.459971 7.776064 8.957063 6.338131
##      [,288]  [,289]  [,290]  [,291]  [,292]  [,293]  [,294]
## [1,] 16.024871 15.307394 10.107252 10.349553 11.081630 10.470163 9.935529
## [2,] 10.929883 9.949402 9.011008 7.887450 9.818130 8.456000 8.256523
## [3,] 5.866074 5.972936 9.234263 4.794259 6.675226 7.377225 5.261869
## [4,] 6.206539 9.882331 9.146539 4.902405 7.010059 7.411178 5.442795
## [5,] 4.600601 6.928499 8.396430 4.454133 6.226294 6.048097 4.394903
## [6,] 5.447123 6.884568 11.309062 7.045291 6.216782 6.165439 20.557009

```

##	[,295]	[,296]	[,297]	[,298]	[,299]	[,300]	[,301]
## [1,]	5.870481	10.833617	11.214380	26.838140	11.362318	10.590375	10.896324
## [2,]	2.267006	10.085223	10.678098	23.616368	10.573109	8.024063	24.575355
## [3,]	2.618094	7.640165	6.785832	6.609471	6.934673	5.472579	8.268958
## [4,]	3.102777	9.026305	6.855697	6.569259	6.893617	8.057728	11.368956
## [5,]	3.576913	7.593470	6.906833	27.210301	6.085483	4.980388	6.364792
## [6,]	3.536167	9.319001	6.752331	120.145992	6.293469	7.131954	52.927854
##	[,302]	[,303]	[,304]	[,305]	[,306]	[,307]	[,308]
## [1,]	22.910797	18.456168	9.577370	6.009893	7.217176	27.38810	12.743185
## [2,]	18.336153	14.134199	35.850235	18.879956	8.423409	15.54423	7.514032
## [3,]	5.383874	6.762612	5.148437	8.253716	4.685714	5.90784	5.293841
## [4,]	24.797268	6.944723	117.796142	22.741784	5.050727	71.98669	8.631611
## [5,]	8.896270	5.936715	753.421641	5.083239	6.779959	22.86464	8.074329
## [6,]	87.533122	8.353264	245.712526	82.184246	12.343815	19.88627	9.705777
##	[,309]	[,310]	[,311]	[,312]	[,313]	[,314]	[,315]
## [1,]	16.846533	13.112033	12.31484	11.651403	12.273583	15.365576	15.405068
## [2,]	11.469297	8.603912	23.39308	5.567253	9.799580	10.866880	12.522402
## [3,]	6.493822	3.586231	10.23640	3.562301	6.108872	6.506803	9.236670
## [4,]	6.945395	4.856093	23.98594	5.422422	7.337576	6.588133	10.664151
## [5,]	6.085384	4.260150	12.00750	8.610542	6.189288	4.936640	8.529447
## [6,]	6.981074	5.208297	15.35022	7.821223	6.138451	7.029763	10.504981
##	[,316]	[,317]	[,318]	[,319]	[,320]	[,321]	[,322]
## [1,]	11.553905	15.224978	13.261643	11.214206	12.716349	10.938404	8.068320
## [2,]	11.894930	11.630520	9.813226	9.887571	8.880782	8.977323	4.548533
## [3,]	5.592204	5.533237	5.418861	4.664543	5.090697	6.835267	3.859564
## [4,]	11.966905	6.123639	5.465343	7.421203	5.424896	6.998996	3.851217
## [5,]	3.909930	4.912063	5.711565	6.376829	4.117358	5.906269	5.956427
## [6,]	4.792733	6.349009	7.077550	8.220724	4.239085	5.810517	5.950045
##	[,323]	[,324]	[,325]	[,326]	[,327]	[,328]	[,329]
## [1,]	8.805191	9.756635	10.265868	15.274848	6.277155	13.942521	16.163324
## [2,]	7.693190	7.697984	6.322640	13.089154	3.438916	10.659524	13.167190
## [3,]	7.936772	5.451131	6.034554	8.086126	6.725776	5.130321	6.000593
## [4,]	8.218747	5.125485	6.037600	12.819045	6.758496	5.935018	8.810055
## [5,]	8.644834	10.454898	4.716686	11.100950	7.892234	5.789532	5.262800
## [6,]	16.175023	13.913075	5.039167	11.858942	7.914459	7.171331	4.644955
##	[,330]	[,331]	[,332]	[,333]	[,334]	[,335]	
## [1,]	12.024519	10.485701	18.63805	20.116405	10.65966	9.744398	
## [2,]	12.169450	8.346681	22.63152	24.128375	36.92948	7.933238	
## [3,]	8.134328	5.433535	23.36649	15.176875	10.26973	7.081669	
## [4,]	11.720561	6.263097	570.62824	18.736380	272.72519	7.233219	
## [5,]	9.964003	8.214581	43.48214	9.718741	522.34068	6.447522	
## [6,]	12.636641	10.002827	233288.29360	49.548627	55.69484	12.191354	
##	[,336]	[,337]	[,338]	[,339]	[,340]	[,341]	
## [1,]	11.379833	11.440108	11.240765	13.886713	14.753011	10.136626	
## [2,]	25.357820	6.901492	11.954009	9.446994	11.658710	8.324708	
## [3,]	7.075439	5.046857	6.437542	4.197125	6.253560	5.672774	
## [4,]	168.115974	6.554589	8.573186	4.803645	8.373727	6.407874	
## [5,]	660.574227	5.440074	8.056105	6.199658	6.758282	5.771901	
## [6,]	1027.753935	7.872000	8.779095	12.060299	7.251532	5.932084	
##	[,342]	[,343]	[,344]	[,345]	[,346]	[,347]	[,348]
## [1,]	10.970347	11.814599	7.623767	11.093897	8.601760	10.089547	15.613013
## [2,]	11.361809	9.984047	8.343379	7.240152	8.583638	19.807765	13.382973
## [3,]	5.004052	6.688559	6.339115	3.886577	4.967890	9.841728	7.104401
## [4,]	5.319049	6.629360	6.320744	3.725008	5.579866	12.900865	9.336579

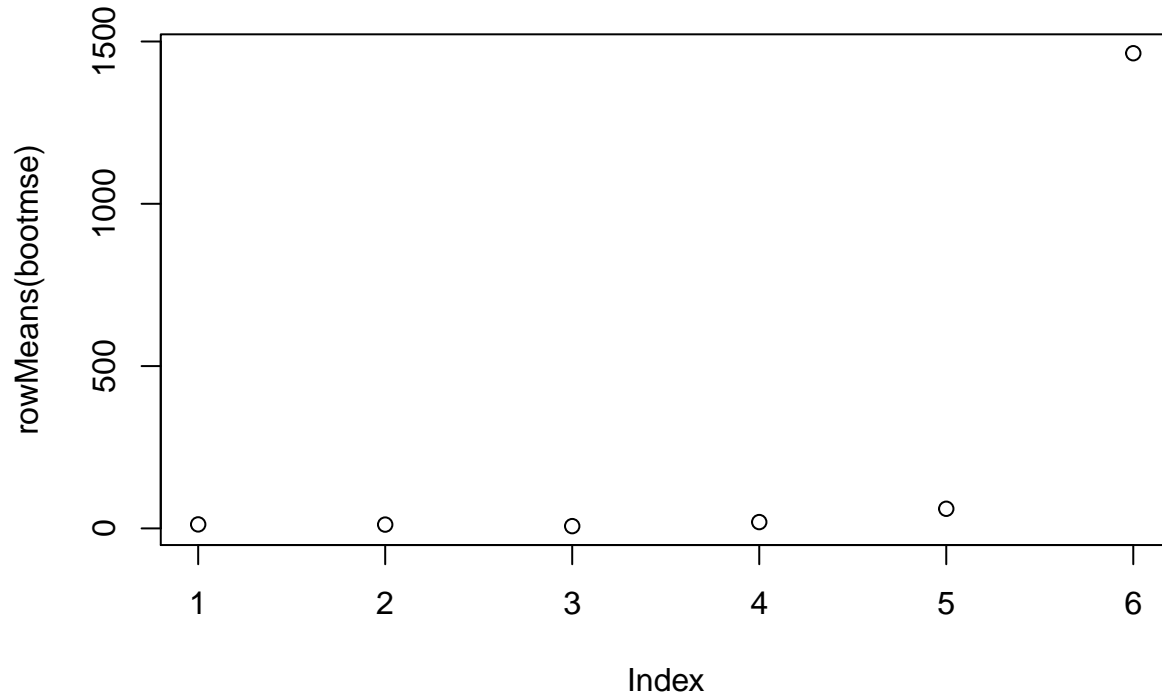

```

## [5,] 6.639587 6.673730 5.595077 9.672382 9.175156 6.469284 9.711606
## [6,] 11.874830 7.161540 5.766550 11.022695 9.810918 40.615458 11.154984
##      [,349]      [,350]      [,351]      [,352]      [,353]      [,354]      [,355]
## [1,] 10.826821 12.223814 17.278020 13.609141 7.834282 18.970465 14.982673
## [2,] 9.106068 9.324562 13.585461 9.831456 7.210286 12.236338 12.176412
## [3,] 4.757201 4.620481 7.927933 3.363326 5.706771 7.973599 5.848932
## [4,] 5.693403 5.151561 10.419715 3.358043 6.864938 11.686433 8.151632
## [5,] 5.937932 6.098298 7.009454 5.473939 6.041762 8.105256 6.170600
## [6,] 6.134669 4.926955 7.425000 12.815518 11.740001 25.952409 7.689667
##      [,356]      [,357]      [,358]      [,359]      [,360]      [,361]      [,362]
## [1,] 11.497322 10.501092 10.569327 9.434930 15.654087 16.106387 8.803101
## [2,] 23.130586 7.007299 10.142193 9.350733 11.111734 12.868147 20.078782
## [3,] 6.620669 5.277879 5.554640 2.348927 5.001391 6.955297 8.639476
## [4,] 12.443527 5.734161 5.494168 2.840513 8.005259 7.459067 7.940646
## [5,] 7.031139 4.754566 7.772649 5.738182 7.746285 6.093197 5.969406
## [6,] 60.775898 5.077683 11.567041 13.734149 10.086201 6.234699 34.123640
##      [,363]      [,364]      [,365]      [,366]      [,367]      [,368]      [,369]
## [1,] 8.315837 14.499944 12.898798 14.391634 5.976088 9.843923 9.910608
## [2,] 45.264016 13.477730 9.923561 8.025571 4.564360 5.793588 10.285510
## [3,] 4.428088 5.719988 4.961538 9.844617 4.103561 3.355424 6.356835
## [4,] 250.129593 14.707354 5.556770 9.188482 4.077083 3.978001 6.850317
## [5,] 340.940477 22.039559 5.836078 8.606525 4.067208 3.166241 6.170631
## [6,] 269.422807 83.535188 6.119679 9.118902 4.522955 3.148069 6.446900
##      [,370]      [,371]      [,372]      [,373]      [,374]      [,375]      [,376]
## [1,] 11.545978 10.153708 6.623473 14.079037 18.167697 11.375473 8.811657
## [2,] 11.282954 7.927041 4.884685 9.449935 11.598363 10.418267 49.366498
## [3,] 7.581365 6.038316 4.015798 5.684504 7.738879 6.600821 33.019592
## [4,] 7.583588 6.009099 4.100665 6.960690 8.131033 7.546629 96.138395
## [5,] 6.396928 4.860988 3.370794 4.611608 6.916604 7.400711 126.183171
## [6,] 6.432343 5.691302 3.354337 5.136565 7.169217 12.751489 233.864810
##      [,377]      [,378]      [,379]      [,380]      [,381]      [,382]      [,383]
## [1,] 13.712188 10.653137 21.59679 13.301624 7.716521 15.978325 12.001883
## [2,] 9.617986 10.116029 18.13920 11.015218 9.692353 12.001408 11.840355
## [3,] 6.739118 5.824591 12.42098 8.236249 6.446687 5.802834 7.322771
## [4,] 6.767123 5.959100 12.94564 9.391878 6.646294 6.769178 7.338425
## [5,] 9.832990 5.311870 10.77204 7.356002 7.199210 4.606543 6.378437
## [6,] 11.540895 17.746889 10.74259 8.223165 14.122109 6.714497 6.332432
##      [,384]      [,385]      [,386]      [,387]      [,388]      [,389]      [,390]
## [1,] 10.018202 10.912672 25.089735 12.932698 7.412306 10.977876 12.917063
## [2,] 10.411932 9.069478 19.401370 9.611202 6.561326 10.726173 11.328249
## [3,] 8.867902 6.155662 4.359674 4.409806 3.628771 5.527416 4.640135
## [4,] 9.337199 6.266118 4.299121 6.164126 4.369509 6.630426 5.335193
## [5,] 7.554817 6.546479 50.053861 7.621379 5.543002 5.045933 5.557230
## [6,] 8.178050 6.370325 54.566964 9.849158 13.755760 5.134218 7.729562
##      [,391]      [,392]      [,393]      [,394]      [,395]      [,396]      [,397]
## [1,] 13.905184 14.918251 9.032093 8.704580 7.423394 11.143023 8.615147
## [2,] 10.587813 13.085620 9.019910 7.162476 5.335974 6.821579 9.344797
## [3,] 7.152971 7.905699 4.940435 4.789088 6.828571 8.274596 4.202222
## [4,] 8.916794 7.932046 7.094410 6.351646 7.791007 9.078131 4.235783
## [5,] 7.020938 7.646224 6.197446 4.441777 6.016400 8.265502 3.113043
## [6,] 7.347223 8.994479 14.231067 4.723176 5.966268 11.294002 19.385735
##      [,398]      [,399]      [,400]
## [1,] 12.478926 15.012322 13.850896
## [2,] 10.302635 13.409601 13.657370

```

```
## [3,] 6.348131 8.744346 9.002654
## [4,] 6.381438 8.802053 9.462265
## [5,] 8.805261 6.777795 8.285385
## [6,] 8.362911 6.777458 8.878981
```

```
plot(rowMeans(bootmse))
```

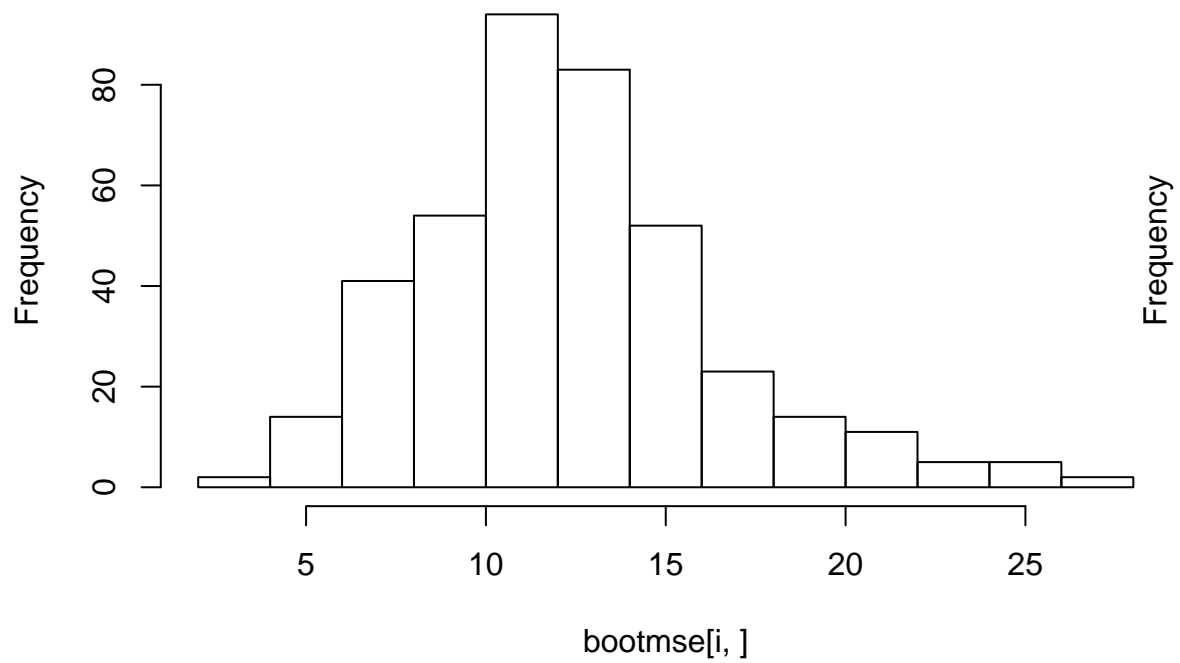


```
apply(bootmse, 1, sd)
```

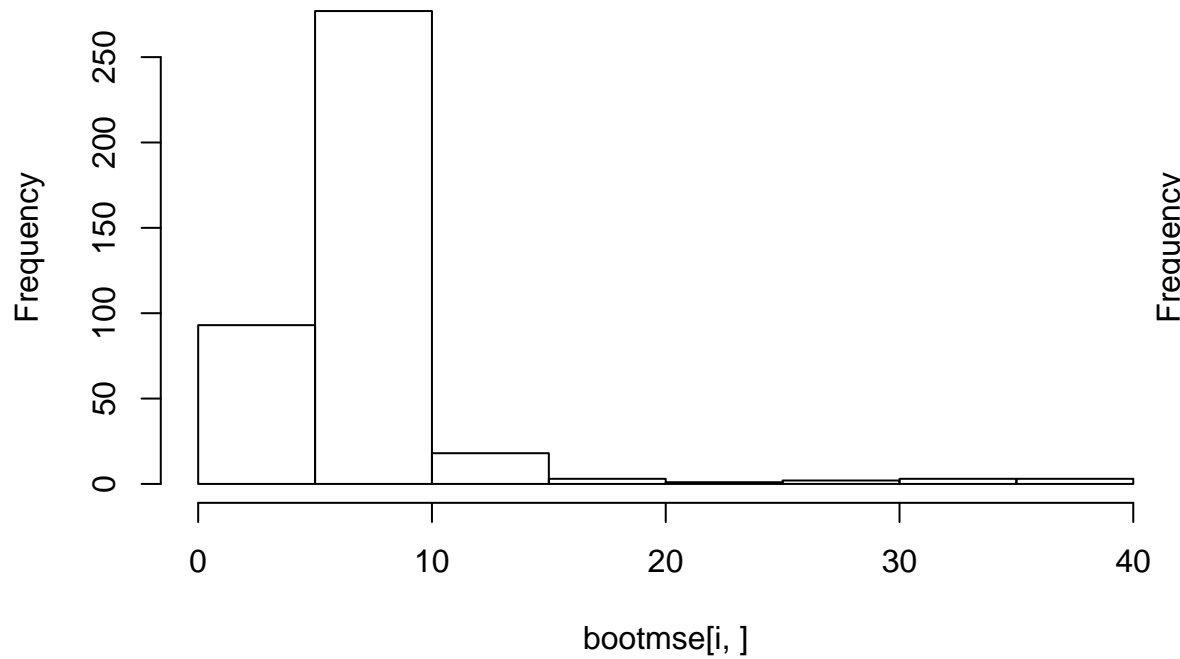
```
## [1] 4.082137 7.424487 4.358966 66.527133 626.811130
## [6] 20211.953385
```

```
for (i in 1:6){
  hist(bootmse[i, ])
}
```

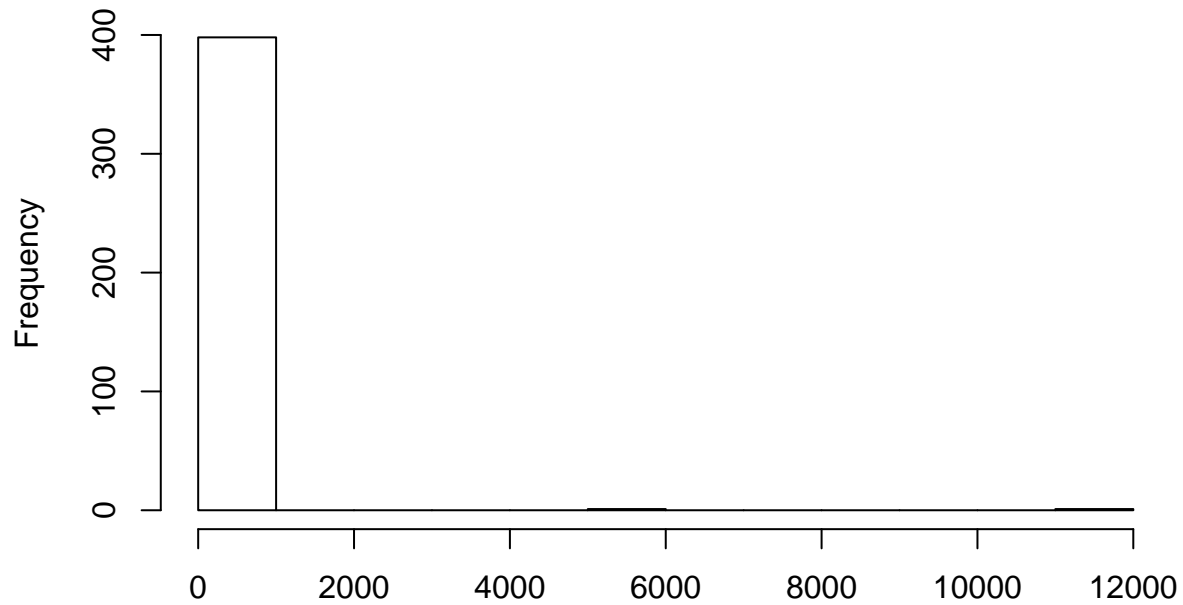
Histogram of bootmse[i,]



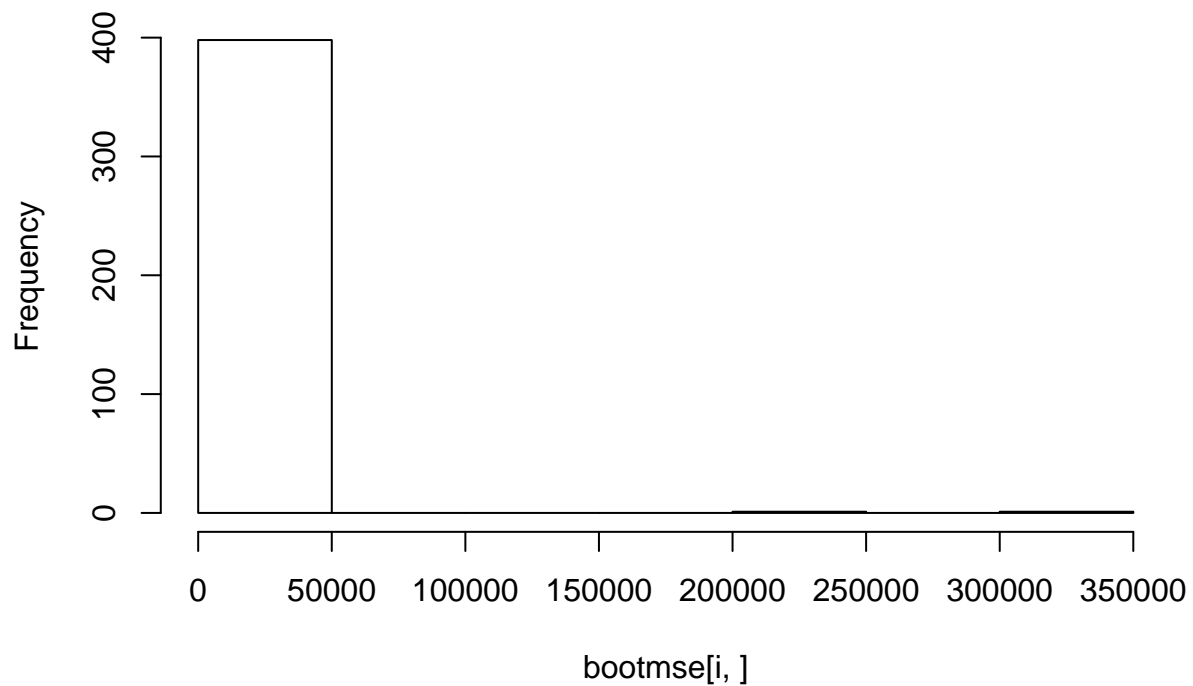
Histogram of bootmse[i,]



Histogram of bootmse[i,]



Histogram of bootmse[i,]



Sample with replacement!!!

first, sample with replacement from the data (this serves as the training set); second, train the model on the sampled data; third, test the model on the data that is not in the sample and compute the performance metric.

SD is pretty big as the polynomial has higher power in general, so it means not stable. There will be big

gaps between your data. (I think it is because bootstrap works well for big data, but we have less than 40 datas for ours)