

Lab 6.5

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6.5.1

```
library(ISLR2)
View(Hitters)
names(Hitters)
```

```
## [1] "AtBat"      "Hits"       "HmRun"      "Runs"       "RBI"        "Walks"
## [7] "Years"      "CAtBat"     "CHits"      "CHmRun"     "CRuns"      "CRBI"
## [13] "CWalks"     "League"     "Division"   "PutOuts"    "Assists"    "Errors"
## [19] "Salary"     "NewLeague"
```

```
dim(Hitters)
```

```
## [1] 322 20
```

```
sum(is.na(Hitters$Salary))
```

```
## [1] 59
```

```
Hitters <- na.omit(Hitters)
dim(Hitters)
```

```
## [1] 263 20
```

```
sum(is.na(Hitters))
```

```
## [1] 0
```

```
library(leaps)
regfit.full <- regsubsets(Salary ~ ., Hitters)
summary(regfit.full)
```

```
## Subset selection object
## Call: regsubsets.formula(Salary ~ ., Hitters)
## 19 Variables (and intercept)
##           Forced in Forced out
```

```
## AtBat      FALSE      FALSE
## Hits       FALSE      FALSE
## HmRun      FALSE      FALSE
## Runs       FALSE      FALSE
## RBI        FALSE      FALSE
## Walks      FALSE      FALSE
## Years      FALSE      FALSE
## CatBat     FALSE      FALSE
## CHits      FALSE      FALSE
## CHmRun     FALSE      FALSE
## CRuns      FALSE      FALSE
## CRBI       FALSE      FALSE
## CWalks     FALSE      FALSE
## LeagueN    FALSE      FALSE
## DivisionW  FALSE      FALSE
## PutOuts    FALSE      FALSE
## Assists    FALSE      FALSE
## Errors     FALSE      FALSE
## NewLeagueN FALSE      FALSE
## 1 subsets of each size up to 8
## Selection Algorithm: exhaustive
##           AtBat Hits HmRun Runs RBI Walks Years CatBat CHits CHmRun CRuns CRBI
## 1 ( 1 ) " " " " " " " " " " " " " " " " " " " " " "
## 2 ( 1 ) " " "*" " " " " " " " " " " " " " " " "
## 3 ( 1 ) " " "*" " " " " " " " " " " " " " " " "
## 4 ( 1 ) " " "*" " " " " " " " " " " " " " " " "
## 5 ( 1 ) "*" "*" " " " " " " " " " " " " " " " "
## 6 ( 1 ) "*" "*" " " " " " " "*" " " " " " " " "
## 7 ( 1 ) " " "*" " " " " " " "*" " " "*" "*" " " "
## 8 ( 1 ) "*" "*" " " " " " " "*" " " " " "*" "*" "
##           CWalks LeagueN DivisionW PutOuts Assists Errors NewLeagueN
## 1 ( 1 ) " " " " " " " " " " " " " "
## 2 ( 1 ) " " " " " " " " " " " "
## 3 ( 1 ) " " " " " " "*" " " " " "
## 4 ( 1 ) " " " " "*" "*" " " " " " "
## 5 ( 1 ) " " " " "*" "*" " " " " " "
## 6 ( 1 ) " " " " "*" "*" " " " " " "
## 7 ( 1 ) " " " " "*" "*" " " " " " "
## 8 ( 1 ) "*" " " "*" "*" " " " " " "
```

```
regfit.full <- regsubsets(Salary ~ ., data = Hitters , nvmax = 19)
reg.summary <- summary(regfit.full)

names(reg.summary)
```

```
## [1] "which" "rsq" "rss" "adjr2" "cp" "bic" "outmat" "obj"
```

```
reg.summary$rsq
```

```
## [1] 0.3214501 0.4252237 0.4514294 0.4754067 0.4908036 0.5087146 0.5141227
## [8] 0.5285569 0.5346124 0.5404950 0.5426153 0.5436302 0.5444570 0.5452164
## [15] 0.5454692 0.5457656 0.5459518 0.5460945 0.5461159
```

```
par(mfrow = c(2, 2))
plot(reg.summary$rss , xlab = "Number of Variables", ylab = "RSS", type = "l")
plot(reg.summary$adjr2 , xlab = "Number of Variables", ylab = "Adjusted RSq", type = "l")

which.max(reg.summary$adjr2)
```

```
## [1] 11
```

```
points(11, reg.summary$adjr2[11], col = "red", cex = 2, pch = 20)

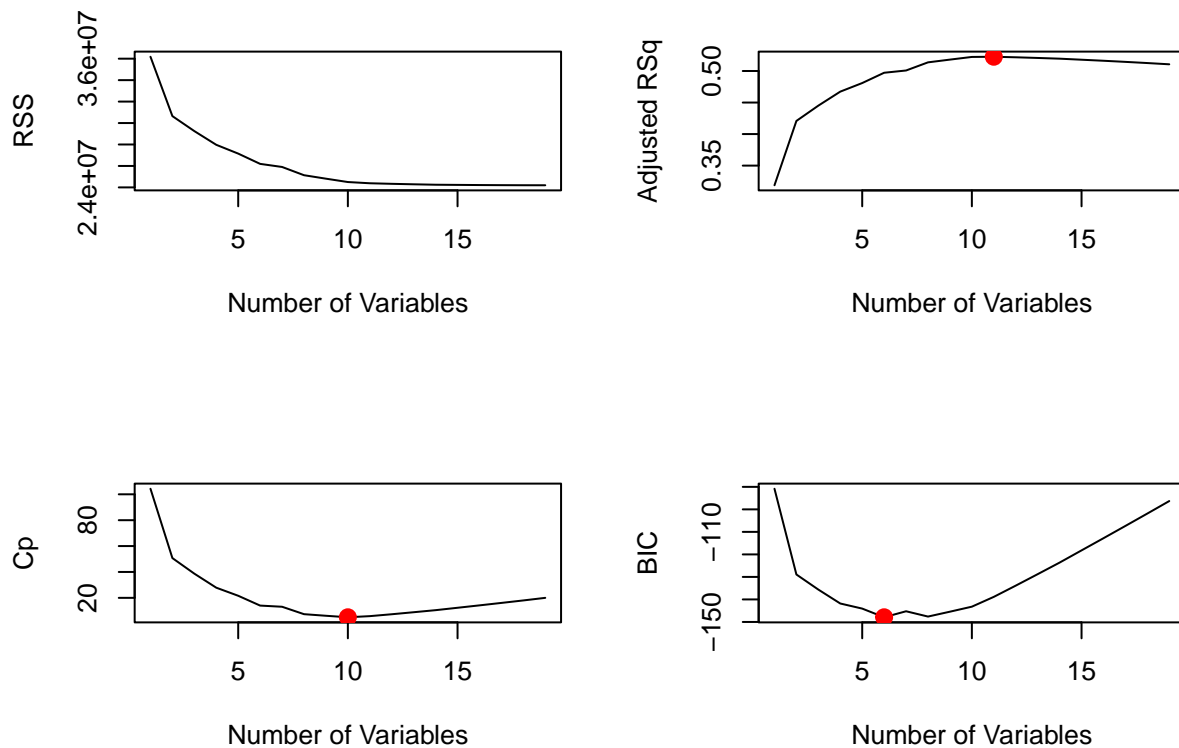
plot(reg.summary$cp , xlab = "Number of Variables", ylab = "Cp", type = "l")
which.min(reg.summary$cp)
```

```
## [1] 10
```

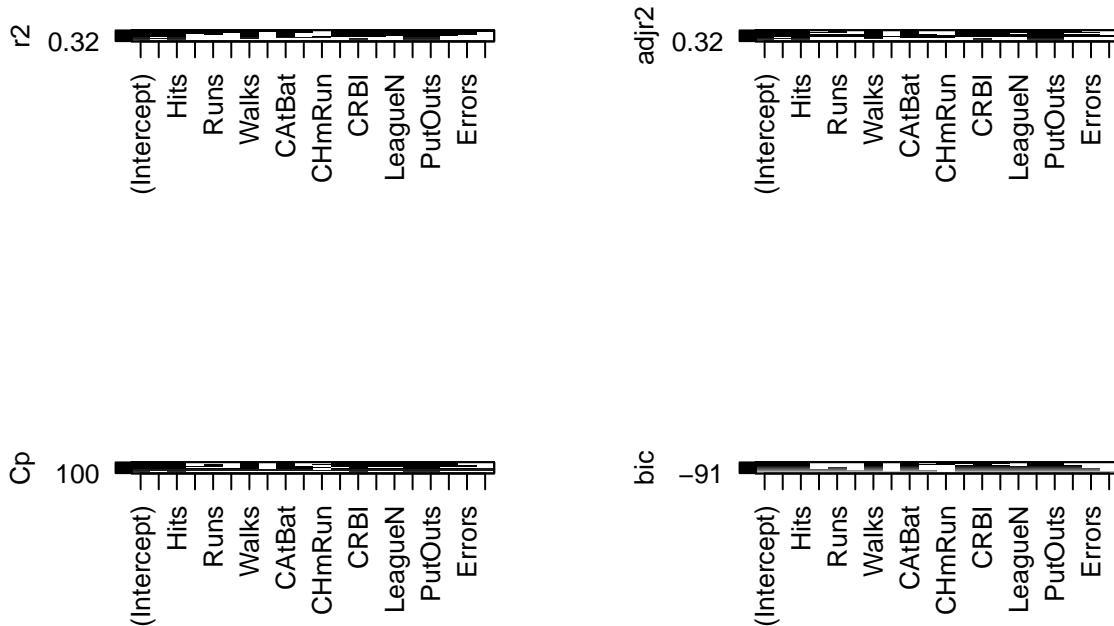
```
points(10, reg.summary$cp[10], col = "red", cex = 2, pch = 20)
which.min(reg.summary$bic)
```

```
## [1] 6
```

```
plot(reg.summary$bic , xlab = "Number of Variables", ylab = "BIC", type = "l")
points(6, reg.summary$bic[6], col = "red", cex = 2, pch = 20)
```



```
plot(regfit.full , scale = "r2")
plot(regfit.full , scale = "adjr2")
plot(regfit.full , scale = "Cp")
plot(regfit.full , scale = "bic")
```



```
coef(regfit.full , 6)
```

```
## (Intercept)      AtBat      Hits      Walks      CRBI      DivisionW
## 91.5117981    -1.8685892    7.6043976    3.6976468    0.6430169   -122.9515338
##      PutOuts
##      0.2643076
```

```
regfit.fwd <- regsubsets(Salary ~ ., data = Hitters , nvmax = 19, method = "forward")
summary(regfit.fwd)
```

```
## Subset selection object
## Call: regsubsets.formula(Salary ~ ., data = Hitters, nvmax = 19, method = "forward")
## 19 Variables (and intercept)
##      Forced in Forced out
## AtBat      FALSE      FALSE
## Hits      FALSE      FALSE
## HmRun      FALSE      FALSE
## Runs      FALSE      FALSE
## RBI      FALSE      FALSE
```

```

## Walks          FALSE      FALSE
## Years          FALSE      FALSE
## CAtBat         FALSE      FALSE
## CHits          FALSE      FALSE
## CHmRun         FALSE      FALSE
## CRuns          FALSE      FALSE
## CRBI           FALSE      FALSE
## CWalks         FALSE      FALSE
## LeagueN        FALSE      FALSE
## DivisionW      FALSE      FALSE
## PutOuts        FALSE      FALSE
## Assists        FALSE      FALSE
## Errors         FALSE      FALSE
## NewLeagueN     FALSE      FALSE
## 1 subsets of each size up to 19
## Selection Algorithm: forward
##           AtBat Hits HmRun Runs RBI Walks Years CAtBat CHits CHmRun CRuns CRBI
## 1 ( 1 ) " " " " " " " " " " " " " " " " " " " "
## 2 ( 1 ) " " "*" " " " " " " " " " " " " " " " "
## 3 ( 1 ) " " "*" " " " " " " " " " " " " " " " "
## 4 ( 1 ) " " "*" " " " " " " " " " " " " " " " "
## 5 ( 1 ) "*" "*" " " " " " " " " " " " " " " " "
## 6 ( 1 ) "*" "*" " " " " " " "*" " " " " " " " "
## 7 ( 1 ) "*" "*" " " " " " " "*" " " " " " " " "
## 8 ( 1 ) "*" "*" " " " " " " "*" " " " " " " "*" "
## 9 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " "*"
## 10 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " "*"
## 11 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " "*"
## 12 ( 1 ) "*" "*" " " " " "*" " " "*" " " " " "*"
## 13 ( 1 ) "*" "*" " " " " "*" " " "*" " " " " "*"
## 14 ( 1 ) "*" "*" "*" "*" " " "*" " " "*" " " " " "*"
## 15 ( 1 ) "*" "*" "*" "*" " " "*" "*" " " " " " " "*"
## 16 ( 1 ) "*" "*" "*" "*" "*" "*" " " "*" "*" " " " " "*"
## 17 ( 1 ) "*" "*" "*" "*" "*" "*" " " "*" "*" " " " " "*"
## 18 ( 1 ) "*" "*" "*" "*" "*" "*" "*" "*" "*" " " " " "*"
## 19 ( 1 ) "*" "*" "*" "*" "*" "*" "*" "*" "*" "*" " " " " "*"
##           CWalks LeagueN DivisionW PutOuts Assists Errors NewLeagueN
## 1 ( 1 ) " " " " " " " " " " " "
## 2 ( 1 ) " " " " " " " " " " " "
## 3 ( 1 ) " " " " " " "*" " " " " "
## 4 ( 1 ) " " " " "*" "*" " " " " "
## 5 ( 1 ) " " " " "*" "*" " " " " "
## 6 ( 1 ) " " " " "*" "*" " " " " "
## 7 ( 1 ) "*" " " "*" "*" " " " " "
## 8 ( 1 ) "*" " " "*" "*" " " " " "
## 9 ( 1 ) "*" " " "*" "*" " " " " "
## 10 ( 1 ) "*" " " "*" "*" "*" " " " "
## 11 ( 1 ) "*" "*" "*" "*" "*" " " " "
## 12 ( 1 ) "*" "*" "*" "*" "*" " " " "
## 13 ( 1 ) "*" "*" "*" "*" "*" "*" " " "
## 14 ( 1 ) "*" "*" "*" "*" "*" "*" " " "
## 15 ( 1 ) "*" "*" "*" "*" "*" "*" " " "
## 16 ( 1 ) "*" "*" "*" "*" "*" "*" "*" " "
## 17 ( 1 ) "*" "*" "*" "*" "*" "*" "*" "*"

```

```
## 18 ( 1 ) "*" "*" "*" "*" "*" "*" "*"
## 19 ( 1 ) "*" "*" "*" "*" "*" "*" "*"

```

```
regfit.bwd <- regsubsets(Salary ~ ., data = Hitters, nvmax = 19, method = "backward")
summary(regfit.bwd)

```

```
## Subset selection object
## Call: regsubsets.formula(Salary ~ ., data = Hitters, nvmax = 19, method = "backward")
## 19 Variables (and intercept)
##              Forced in Forced out
## AtBat          FALSE          FALSE
## Hits           FALSE          FALSE
## HmRun          FALSE          FALSE
## Runs           FALSE          FALSE
## RBI            FALSE          FALSE
## Walks          FALSE          FALSE
## Years          FALSE          FALSE
## CAtBat         FALSE          FALSE
## CHits          FALSE          FALSE
## CHmRun         FALSE          FALSE
## CRuns          FALSE          FALSE
## CRBI          FALSE          FALSE
## CWalks         FALSE          FALSE
## LeagueN        FALSE          FALSE
## DivisionW      FALSE          FALSE
## PutOuts        FALSE          FALSE
## Assists        FALSE          FALSE
## Errors         FALSE          FALSE
## NewLeagueN     FALSE          FALSE
## 1 subsets of each size up to 19
## Selection Algorithm: backward
##              AtBat Hits HmRun Runs RBI Walks Years CAtBat CHits CHmRun CRuns CRBI
## 1 ( 1 ) " " " " " " " " " " " " " " " " "*" " "
## 2 ( 1 ) " " "*" " " " " " " " " " " " " " " "*" " "
## 3 ( 1 ) " " "*" " " " " " " " " " " " " " " "*" " "
## 4 ( 1 ) "*" "*" " " " " " " " " " " " " " " "*" " "
## 5 ( 1 ) "*" "*" " " " " " " "*" " " " " " " " " "*" " "
## 6 ( 1 ) "*" "*" " " " " " " "*" " " " " " " " " "*" " "
## 7 ( 1 ) "*" "*" " " " " " " "*" " " " " " " " " "*" " "
## 8 ( 1 ) "*" "*" " " " " " " "*" " " " " " " " " "*" "*"
## 9 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " " " "*" "*"
## 10 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " " " "*" "*"
## 11 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " " " "*" "*"
## 12 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " " " "*" "*"
## 13 ( 1 ) "*" "*" " " " " " " "*" " " "*" " " " " " " "*" "*"
## 14 ( 1 ) "*" "*" "*" "*" " " " "*" " " " " " " " " "*" "*"
## 15 ( 1 ) "*" "*" "*" "*" " " " "*" " " " " " " " " "*" "*"
## 16 ( 1 ) "*" "*" "*" "*" "*" "*" " " " " " " " " " " "*" "*"
## 17 ( 1 ) "*" "*" "*" "*" "*" "*" " " " " " " " " " " "*" "*"
## 18 ( 1 ) "*" "*" "*" "*" "*" "*" "*" " " " " " " " " " " "*" "*"
## 19 ( 1 ) "*" "*" "*" "*" "*" "*" "*" " " " " " " " " " " "*" "*"
##              CWalks LeagueN DivisionW PutOuts Assists Errors NewLeagueN
## 1 ( 1 ) " " " " " " " " " " " "
## 2 ( 1 ) " " " " " " " " " " " "
```

```
## 3 ( 1 ) " " " " " " "*" " " " " " "
## 4 ( 1 ) " " " " " " "*" " " " " " "
## 5 ( 1 ) " " " " " " "*" " " " " " "
## 6 ( 1 ) " " " " "*" "*" " " " " " "
## 7 ( 1 ) "*" " " "*" "*" " " " " " "
## 8 ( 1 ) "*" " " "*" "*" " " " " " "
## 9 ( 1 ) "*" " " "*" "*" " " " " " "
## 10 ( 1 ) "*" " " "*" "*" "*" " " " " "
## 11 ( 1 ) "*" "*" "*" "*" "*" " " " " "
## 12 ( 1 ) "*" "*" "*" "*" "*" " " " " "
## 13 ( 1 ) "*" "*" "*" "*" "*" "*" " " " "
## 14 ( 1 ) "*" "*" "*" "*" "*" "*" " " " "
## 15 ( 1 ) "*" "*" "*" "*" "*" "*" " " " "
## 16 ( 1 ) "*" "*" "*" "*" "*" "*" " " " "
## 17 ( 1 ) "*" "*" "*" "*" "*" "*" "*" " " "
## 18 ( 1 ) "*" "*" "*" "*" "*" "*" "*" " " "
## 19 ( 1 ) "*" "*" "*" "*" "*" "*" "*" " " "
```

```
coef(regfit.full , 7)
```

```
## (Intercept)      Hits      Walks      CAtBat      CHits      CHmRun
## 79.4509472    1.2833513    3.2274264   -0.3752350    1.4957073    1.4420538
## DivisionW      PutOuts
## -129.9866432    0.2366813
```

```
coef(regfit.fwd , 7)
```

```
## (Intercept)      AtBat      Hits      Walks      CRBI      CWalks
## 109.7873062   -1.9588851    7.4498772    4.9131401    0.8537622   -0.3053070
## DivisionW      PutOuts
## -127.1223928    0.2533404
```

```
coef(regfit.bwd , 7)
```

```
## (Intercept)      AtBat      Hits      Walks      CRuns      CWalks
## 105.6487488   -1.9762838    6.7574914    6.0558691    1.1293095   -0.7163346
## DivisionW      PutOuts
## -116.1692169    0.3028847
```

```
set.seed(1)
train <- sample(c(TRUE , FALSE), nrow(Hitters), replace = TRUE)
test <- (!train)

regfit.best <- regsubsets(Salary ~ .,
  data = Hitters[train , ], nvmax = 19)

test.mat <- model.matrix(Salary ~ ., data = Hitters[test , ])

val.errors <- rep(NA, 19)
for (i in 1:19) {
  coefi <- coef(regfit.best , id = i)
```

```

pred <- test.mat[, names(coefi)] %*% coefi
val.errors[i] <- mean((Hitters$Salary[test] - pred)^2)
}

```

```
val.errors
```

```

## [1] 164377.3 144405.5 152175.7 145198.4 137902.1 139175.7 126849.0 136191.4
## [9] 132889.6 135434.9 136963.3 140694.9 140690.9 141951.2 141508.2 142164.4
## [17] 141767.4 142339.6 142238.2

```

```
which.min(val.errors)
```

```
## [1] 7
```

```
coef(regfit.best , 7)
```

```

## (Intercept)      AtBat      Hits      Walks      CRuns      CWalks
##  67.1085369  -2.1462987   7.0149547   8.0716640   1.2425113  -0.8337844
##  DivisionW      PutOuts
## -118.4364998    0.2526925

```

```

predict.regsubsets <- function(object , newdata , id, ...) {
  form <- as.formula(object$call [[2]])
  mat <- model.matrix(form , newdata)
  coefi <- coef(object , id = id)
  xvars <- names(coefi)
  mat[, xvars] %*% coefi
}

```

```

regfit.best <- regsubsets(Salary ~ ., data = Hitters , nvmax = 19)
coef(regfit.best , 7)

```

```

## (Intercept)      Hits      Walks      CAtBat      CHits      CHmRun
##  79.4509472   1.2833513   3.2274264  -0.3752350   1.4957073   1.4420538
##  DivisionW      PutOuts
## -129.9866432    0.2366813

```

```

k <- 10
n <- nrow(Hitters)
set.seed(1)
folds <- sample(rep(1:k, length = n))
cv.errors <- matrix(NA, k, 19,
  dimnames = list(NULL , paste (1:19)))

for (j in 1:k) {
  best.fit <- regsubsets(Salary ~ ., data = Hitters[folds != j, ], nvmax = 19)
  for (i in 1:19) {
    pred <- predict(best.fit , Hitters[folds == j, ], id = i)
    cv.errors[j, i] <- mean((Hitters$Salary[folds == j] - pred)^2)
  }
}

```

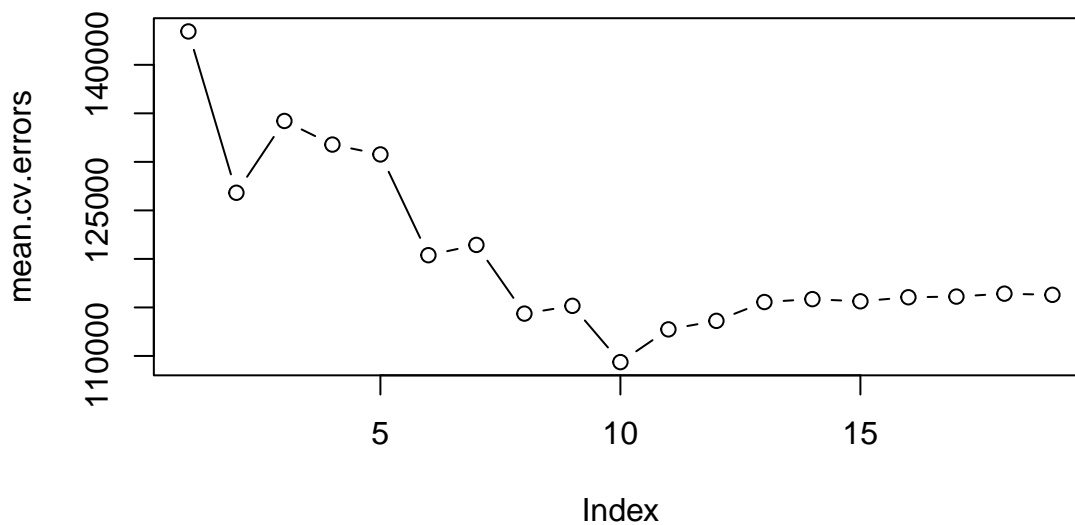


```
}
```

```
mean.cv.errors <- apply(cv.errors , 2, mean)
mean.cv.errors
```

```
##      1      2      3      4      5      6      7      8
## 143439.8 126817.0 134214.2 131782.9 130765.6 120382.9 121443.1 114363.7
##      9     10     11     12     13     14     15     16
## 115163.1 109366.0 112738.5 113616.5 115557.6 115853.3 115630.6 116050.0
##     17     18     19
## 116117.0 116419.3 116299.1
```

```
par(mfrow = c(1, 1))
plot(mean.cv.errors , type = "b")
```



```
reg.best <- regsubsets(Salary ~ ., data = Hitters , nvmax = 19)
coef(reg.best , 10)
```

```
## (Intercept)      AtBat      Hits      Walks      CAtBat      CRuns
## 162.5354420 -2.1686501  6.9180175  5.7732246 -0.1300798  1.4082490
##      CRBI      CWalks  DivisionW      PutOuts      Assists
##  0.7743122 -0.8308264 -112.3800575  0.2973726  0.2831680
```

6.5.2

```
x <- model.matrix(Salary ~ ., Hitters)[, -1]
y <- Hitters$Salary
```

```
library(glmnet)
```

```
## Loading required package: Matrix
```

```
## Loaded glmnet 4.1-8
```

```
grid <- 10^seq(10, -2, length = 100)
ridge.mod <- glmnet(x, y, alpha = 0, lambda = grid)

dim(coef(ridge.mod))
```

```
## [1] 20 100
```

```
ridge.mod$lambda[50]
```

```
## [1] 11497.57
```

```
coef(ridge.mod)[, 50]
```

```
##      (Intercept)      AtBat      Hits      HmRun      Runs
## 407.356050200    0.036957182    0.138180344    0.524629976    0.230701523
##           RBI           Walks           Years      CAtBat      CHits
## 0.239841459    0.289618741    1.107702929    0.003131815    0.011653637
##      CHmRun      CRuns      CRBI      CWalks      LeagueN
## 0.087545670    0.023379882    0.024138320    0.025015421    0.085028114
## DivisionW      PutOuts      Assists      Errors      NewLeagueN
## -6.215440973    0.016482577    0.002612988   -0.020502690    0.301433531
```

```
sqrt(sum(coef(ridge.mod)[-1, 50]^2))
```

```
## [1] 6.360612
```

```
ridge.mod$lambda[60]
```

```
## [1] 705.4802
```

```
coef(ridge.mod)[, 60]
```

```
##      (Intercept)      AtBat      Hits      HmRun      Runs      RBI
## 54.32519950    0.11211115    0.65622409    1.17980910    0.93769713    0.84718546
##           Walks           Years      CAtBat      CHits      CHmRun      CRuns
## 1.31987948    2.59640425    0.01083413    0.04674557    0.33777318    0.09355528
##      CRBI      CWalks      LeagueN      DivisionW      PutOuts      Assists
## 0.09780402    0.07189612    13.68370191   -54.65877750    0.11852289    0.01606037
##      Errors      NewLeagueN
## -0.70358655    8.61181213
```

```
sqrt(sum(coef(ridge.mod)[-1, 60]^2))
```

```
## [1] 57.11001
```

```
predict(ridge.mod , s = 50, type = "coefficients")[1:20, ]
```

```
##      (Intercept)      AtBat      Hits      HmRun      Runs
## 4.876610e+01 -3.580999e-01 1.969359e+00 -1.278248e+00 1.145892e+00
##      RBI      Walks      Years      CAtBat      CHits
## 8.038292e-01 2.716186e+00 -6.218319e+00 5.447837e-03 1.064895e-01
##      CHmRun      CRuns      CRBI      CWalks      LeagueN
## 6.244860e-01 2.214985e-01 2.186914e-01 -1.500245e-01 4.592589e+01
##      DivisionW      PutOuts      Assists      Errors      NewLeagueN
## -1.182011e+02 2.502322e-01 1.215665e-01 -3.278600e+00 -9.496680e+00
```

```
set.seed(1)
train <- sample(1:nrow(x), nrow(x) / 2)
test <- (-train)
y.test <- y[test]

ridge.mod <- glmnet(x[train , ], y[train], alpha = 0, lambda = grid, thresh = 1e-12)
ridge.pred <- predict(ridge.mod , s = 4, newx = x[test , ])
mean((ridge.pred - y.test)^2)
```

```
## [1] 142199.2
```

```
mean((mean(y[train]) - y.test)^2)
```

```
## [1] 224669.9
```

```
ridge.pred <- predict(ridge.mod , s = 1e10 , newx = x[test , ])
mean((ridge.pred - y.test)^2)
```

```
## [1] 224669.8
```

```
ridge.pred <- predict(ridge.mod , s = 0, newx = x[test , ], exact = T, x = x[train , ], y = y[train])
mean((ridge.pred - y.test)^2)
```

```
## [1] 168588.6
```

```
lm(y ~ x, subset = train)
```

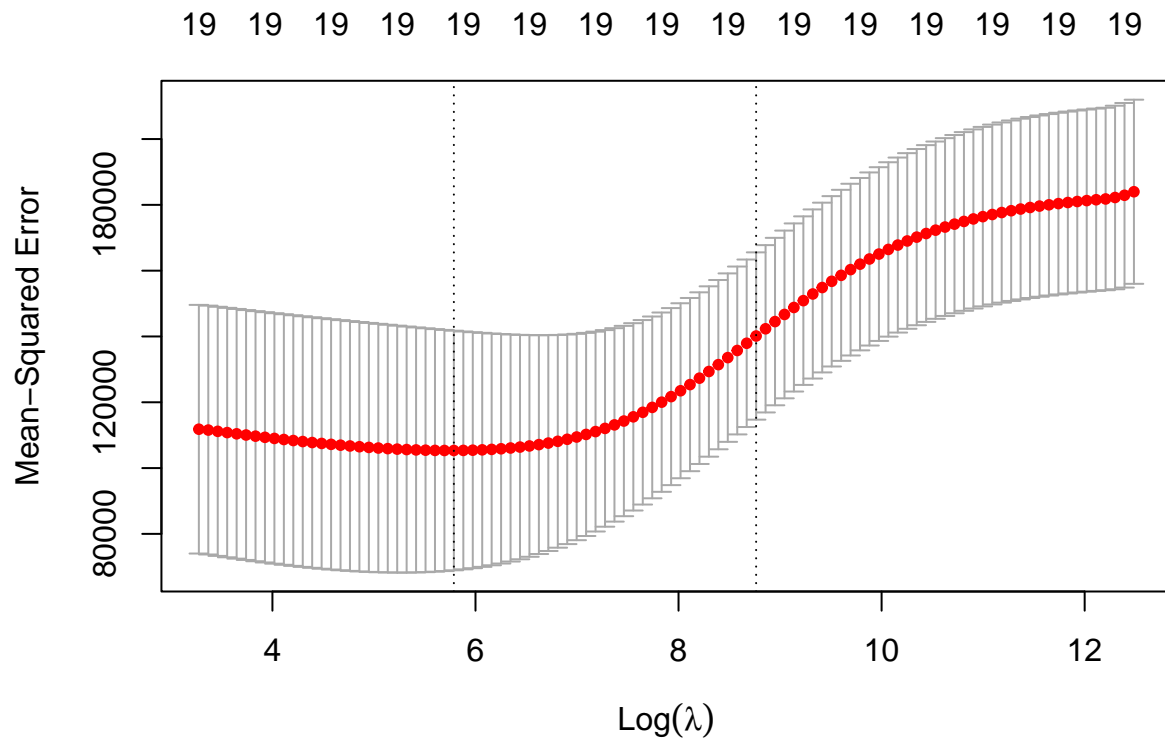
```
##
## Call:
## lm(formula = y ~ x, subset = train)
##
## Coefficients:
## (Intercept)      xAtBat      xHits      xHmRun      xRuns      xRBI
```

```
##      274.0145      -0.3521      -1.6377      5.8145      1.5424      1.1243
##      xWalks      xYears      xCatBat      xCHits      xCHmRun      xCRuns
##      3.7287      -16.3773      -0.6412      3.1632      3.4008      -0.9739
##      xCRBI      xCWalks      xLeagueN      xDivisionW      xPutOuts      xAssists
##      -0.6005      0.3379      119.1486      -144.0831      0.1976      0.6804
##      xErrors      xNewLeagueN
##      -4.7128      -71.0951
```

```
predict(ridge.mod , s = 0, exact = T, type = "coefficients",
  x = x[train, ], y = y[train])[1:20, ]
```

```
## (Intercept)      AtBat      Hits      HmRun      Runs      RBI
## 274.0200994 -0.3521900 -1.6371383 5.8146692 1.5423361 1.1241837
##      Walks      Years      CAtBat      CHits      CHmRun      CRuns
## 3.7288406 -16.3795195 -0.6411235 3.1629444 3.4005281 -0.9739405
##      CRBI      CWalks      LeagueN      DivisionW      PutOuts      Assists
## -0.6003976 0.3378422 119.1434637 -144.0853061 0.1976300 0.6804200
##      Errors      NewLeagueN
## -4.7127879 -71.0898914
```

```
set.seed(1)
cv.out <- cv.glmnet(x[train , ], y[train], alpha = 0)
plot(cv.out)
```



```
bestlam <- cv.out$lambda.min
bestlam
```

```
## [1] 326.0828
```

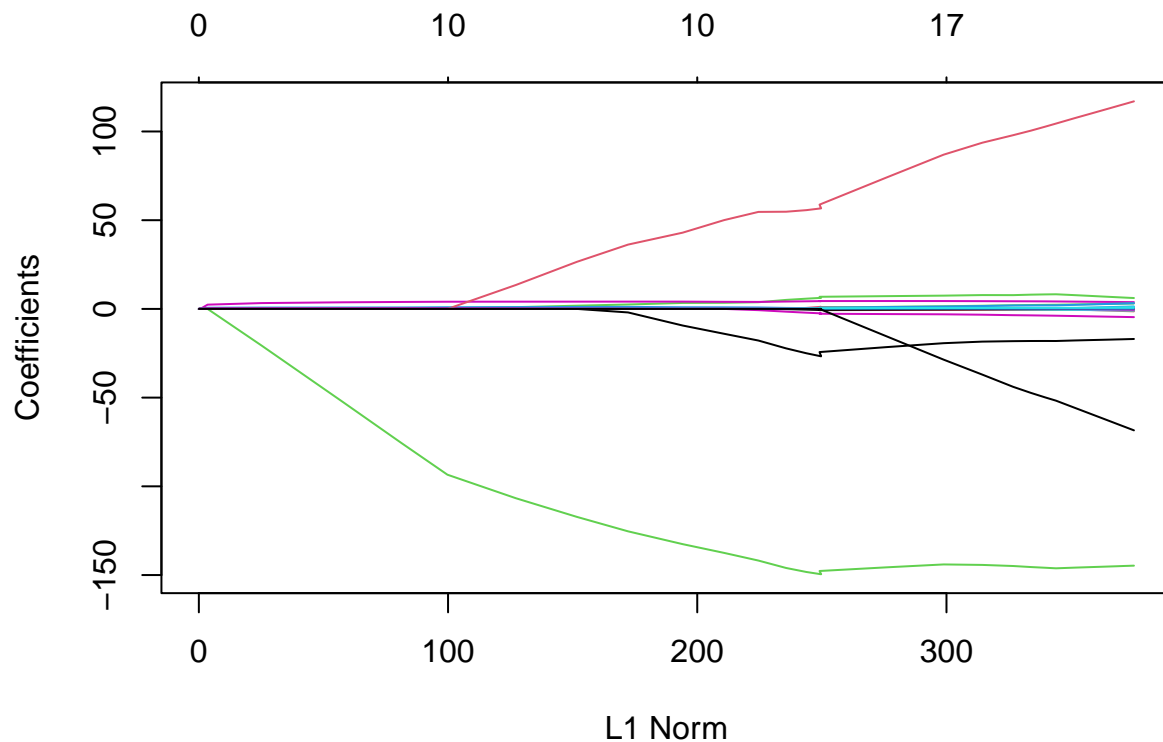
```
ridge.pred <- predict(ridge.mod , s = bestlam , newx = x[test , ])
mean((ridge.pred - y.test)^2)
```

```
## [1] 139856.6
```

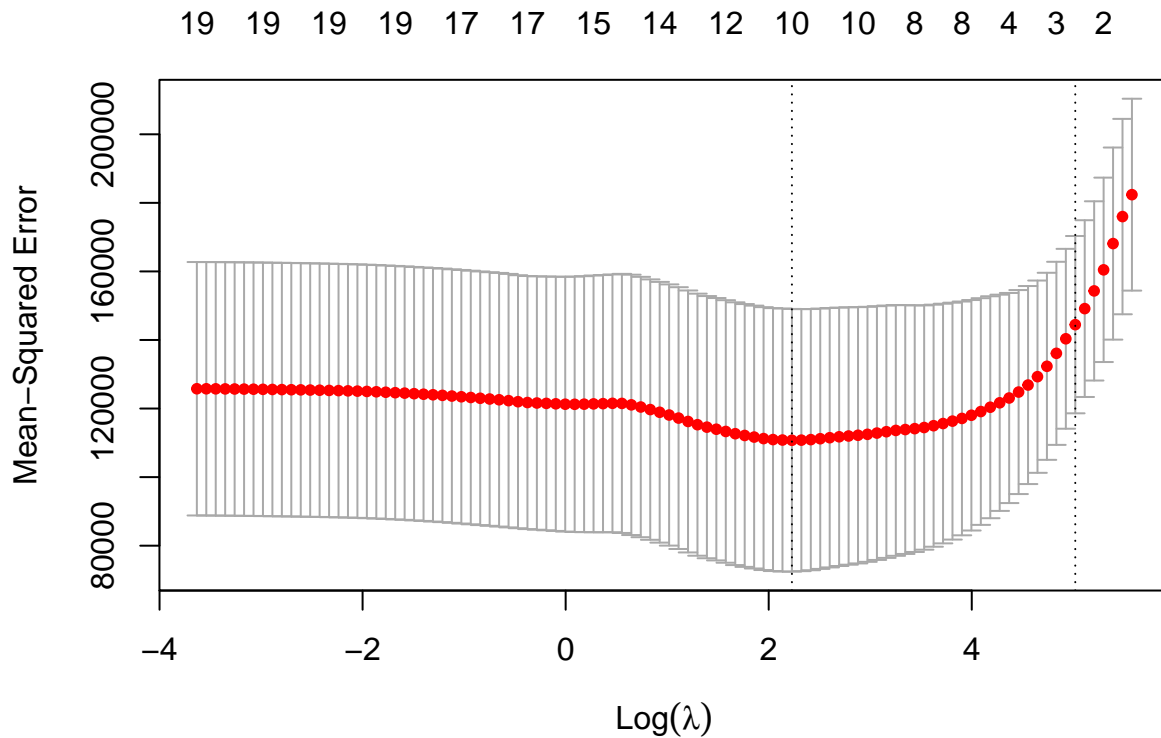
```
out <- glmnet(x, y, alpha = 0)
predict(out , type = "coefficients", s = bestlam)[1:20, ]
```

```
## (Intercept)      AtBat      Hits      HmRun      Runs      RBI
## 15.44383120  0.07715547  0.85911582  0.60103106  1.06369007  0.87936105
##      Walks      Years    CAtBat    CHits    CHmRun    CRuns
##  1.62444617  1.35254778  0.01134999  0.05746654  0.40680157  0.11456224
##      CRBI      CWalks    LeagueN  DivisionW    PutOuts    Assists
##  0.12116504  0.05299202  22.09143197 -79.04032656  0.16619903  0.02941950
##      Errors  NewLeagueN
## -1.36092945  9.12487765
```

```
lasso.mod <- glmnet(x[train , ], y[train], alpha = 1, lambda = grid)
plot(lasso.mod)
```



```
set.seed(1)
cv.out <- cv.glmnet(x[train , ], y[train], alpha = 1)
plot(cv.out)
```



```
bestlam <- cv.out$lambda.min
lasso.pred <- predict(lasso.mod , s = bestlam , newx = x[test , ])
mean((lasso.pred - y.test)^2)
```

```
## [1] 143673.6
```

```
out <- glmnet(x, y, alpha = 1, lambda = grid)
lasso.coef <- predict(out , type = "coefficients", s = bestlam)[1:20, ]
lasso.coef
```

```
##      (Intercept)      AtBat      Hits      HmRun      Runs
##      1.27479059 -0.05497143  2.18034583  0.00000000  0.00000000
##           RBI      Walks      Years      CAtBat      CHits
##      0.00000000  2.29192406 -0.33806109  0.00000000  0.00000000
##      CHmRun      CRuns      CRBI      CWalks      LeagueN
##      0.02825013  0.21628385  0.41712537  0.00000000  20.28615023
##      DivisionW      PutOuts      Assists      Errors      NewLeagueN
## -116.16755870  0.23752385  0.00000000 -0.85629148  0.00000000
```

```
lasso.coef[lasso.coef != 0]
```

```
##      (Intercept)      AtBat      Hits      Walks      Years
##      1.27479059   -0.05497143   2.18034583   2.29192406   -0.33806109
##      CHmRun      CRuns      CRBI      LeagueN      DivisionW
##      0.02825013   0.21628385   0.41712537   20.28615023  -116.16755870
##      PutOuts      Errors
##      0.23752385   -0.85629148
```

6.5.3

```
library(pls)
```

```
##
## Attaching package: 'pls'

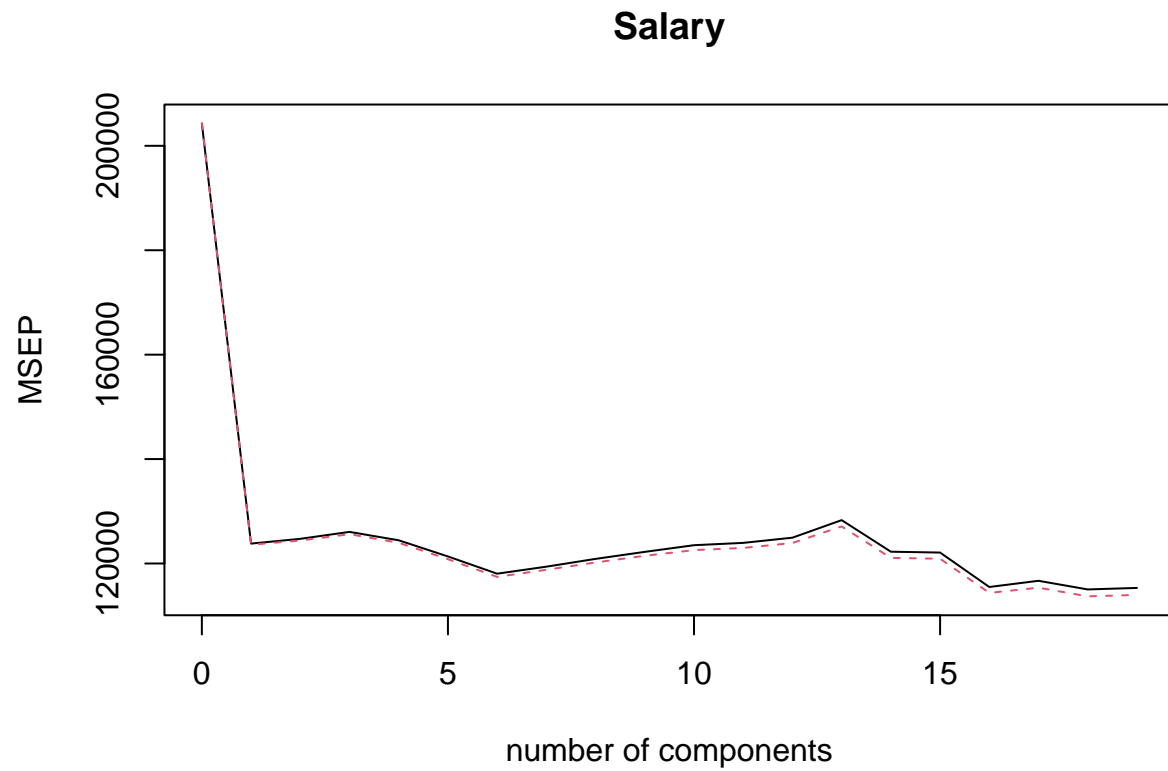
## The following object is masked from 'package:stats':
##
##      loadings
```

```
set.seed(2)
pcr.fit <- pcr(Salary ~ ., data = Hitters, scale = TRUE, validation = "CV")
summary(pcr.fit)
```

```
## Data:      X dimension: 263 19
## Y dimension: 263 1
## Fit method: svdpc
## Number of components considered: 19
##
## VALIDATION: RMSEP
## Cross-validated using 10 random segments.
##      (Intercept)  1 comps  2 comps  3 comps  4 comps  5 comps  6 comps
## CV              452    351.9    353.2    355.0    352.8    348.4    343.6
## adjCV           452    351.6    352.7    354.4    352.1    347.6    342.7
##      7 comps  8 comps  9 comps 10 comps 11 comps 12 comps 13 comps
## CV          345.5    347.7    349.6    351.4    352.1    353.5    358.2
## adjCV        344.7    346.7    348.5    350.1    350.7    352.0    356.5
##      14 comps 15 comps 16 comps 17 comps 18 comps 19 comps
## CV          349.7    349.4    339.9    341.6    339.2    339.6
## adjCV        348.0    347.7    338.2    339.7    337.2    337.6
##
## TRAINING: % variance explained
##      1 comps  2 comps  3 comps  4 comps  5 comps  6 comps  7 comps  8 comps
## X           38.31    60.16    70.84    79.03    84.29    88.63    92.26    94.96
## Salary       40.63    41.58    42.17    43.22    44.90    46.48    46.69    46.75
##      9 comps 10 comps 11 comps 12 comps 13 comps 14 comps 15 comps
## X           96.28    97.26    97.98    98.65    99.15    99.47    99.75
## Salary       46.86    47.76    47.82    47.85    48.10    50.40    50.55
##      16 comps 17 comps 18 comps 19 comps
```

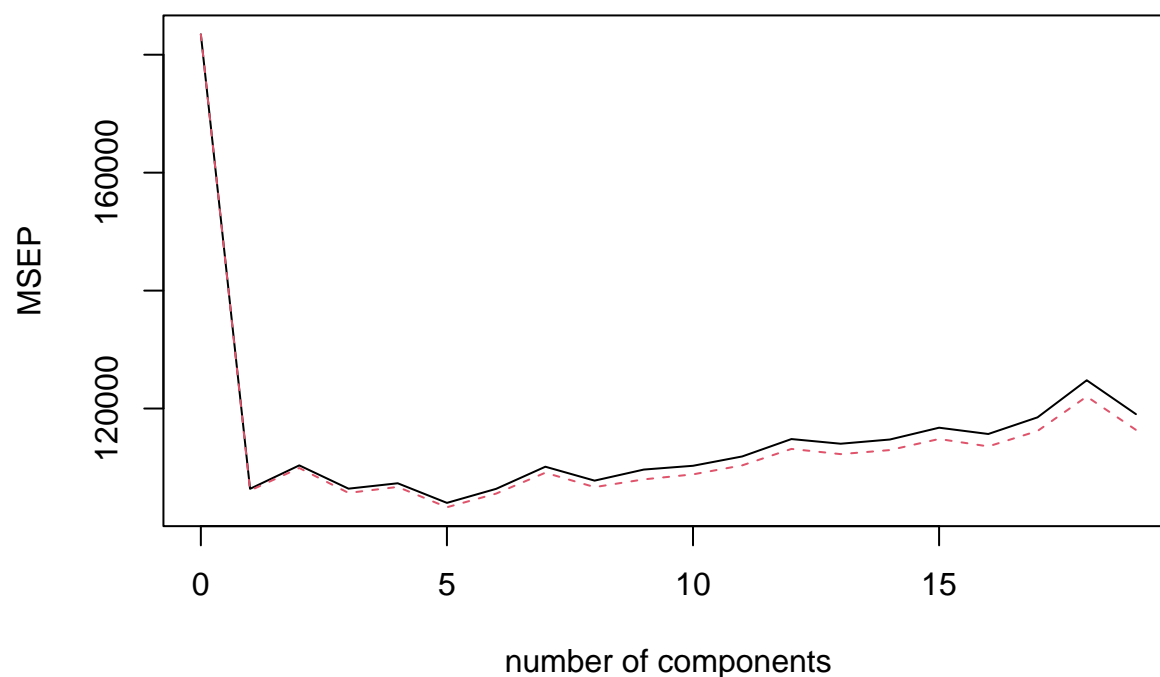
## X	99.89	99.97	99.99	100.00
## Salary	53.01	53.85	54.61	54.61

```
validationplot(pcr.fit , val.type = "MSEP")
```



```
set.seed(1)
pcr.fit <- pcr(Salary ~ ., data = Hitters , subset = train , scale = TRUE , validation = "CV")
validationplot(pcr.fit , val.type = "MSEP")
```


Salary



```
pcr.pred <- predict(pcr.fit , x[test , ], ncomp = 5)
mean((pcr.pred - y.test)^2)
```

```
## [1] 142811.8
```

```
pcr.fit <- pcr(y ~ x, scale = TRUE, ncomp = 5)
summary(pcr.fit)
```

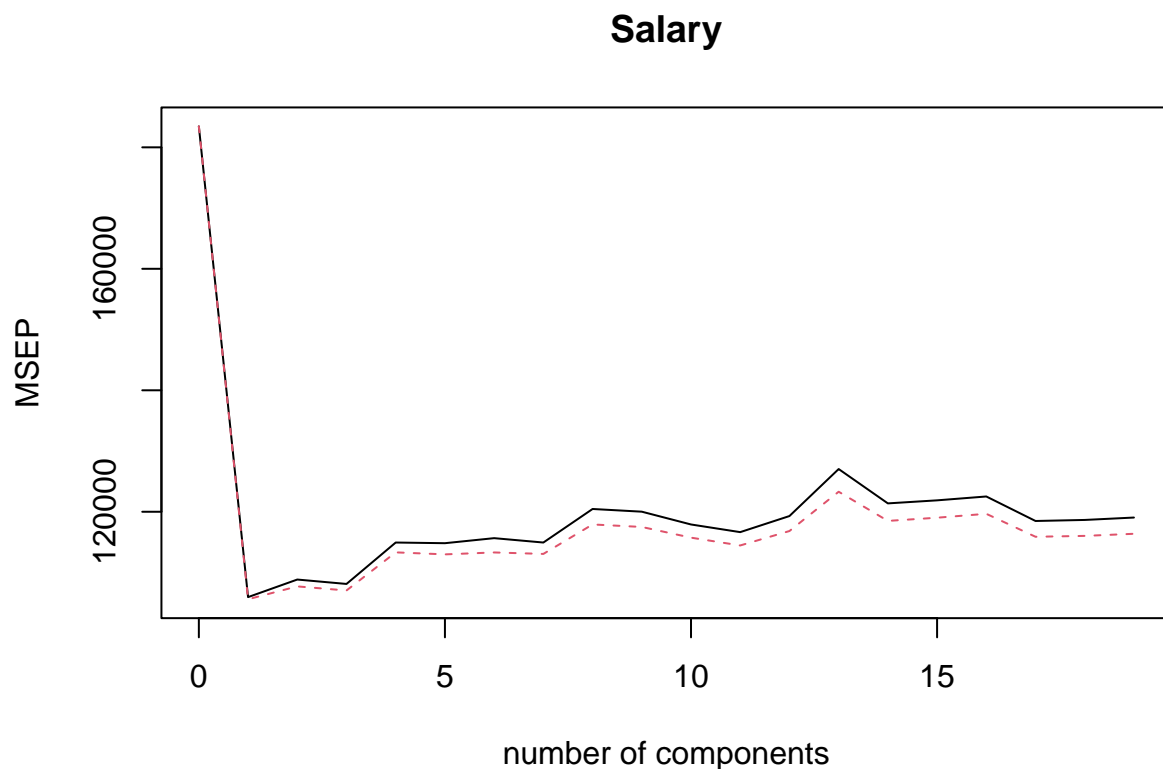
```
## Data:      X dimension: 263 19
## Y dimension: 263 1
## Fit method: svdpc
## Number of components considered: 5
## TRAINING: % variance explained
##      1 comps  2 comps  3 comps  4 comps  5 comps
## X      38.31   60.16   70.84   79.03   84.29
## y      40.63   41.58   42.17   43.22   44.90
```

```
set.seed(1)
pls.fit <- plsr(Salary ~ ., data = Hitters , subset = train , scale = TRUE, validation = "CV")
summary(pls.fit)
```

```
## Data:      X dimension: 131 19
## Y dimension: 131 1
## Fit method: kernelpls
```

```
## Number of components considered: 19
##
## VALIDATION: RMSEP
## Cross-validated using 10 random segments.
##      (Intercept)  1 comps  2 comps  3 comps  4 comps  5 comps  6 comps
## CV           428.3   325.5   329.9   328.8   339.0   338.9   340.1
## adjCV         428.3   325.0   328.2   327.2   336.6   336.1   336.6
##      7 comps  8 comps  9 comps 10 comps 11 comps 12 comps 13 comps
## CV           339.0   347.1   346.4   343.4   341.5   345.4   356.4
## adjCV         336.2   343.4   342.8   340.2   338.3   341.8   351.1
##      14 comps 15 comps 16 comps 17 comps 18 comps 19 comps
## CV           348.4   349.1   350.0   344.2   344.5   345.0
## adjCV         344.2   345.0   345.9   340.4   340.6   341.1
##
## TRAINING: % variance explained
##      1 comps  2 comps  3 comps  4 comps  5 comps  6 comps  7 comps  8 comps
## X           39.13   48.80   60.09   75.07   78.58   81.12   88.21   90.71
## Salary       46.36   50.72   52.23   53.03   54.07   54.77   55.05   55.66
##      9 comps 10 comps 11 comps 12 comps 13 comps 14 comps 15 comps
## X           93.17   96.05   97.08   97.61   97.97   98.70   99.12
## Salary       55.95   56.12   56.47   56.68   57.37   57.76   58.08
##      16 comps 17 comps 18 comps 19 comps
## X           99.61   99.70   99.95  100.00
## Salary       58.17   58.49   58.56   58.62
```

```
validationplot(pls.fit , val.type = "MSEP")
```



```
pls.pred <- predict(pls.fit , x[test , ], ncomp = 1)
mean((pls.pred - y.test)^2)
```

```
## [1] 151995.3
```

```
pls.fit <- plsr(Salary ~ ., data = Hitters , scale = TRUE, ncomp = 1)
summary(pls.fit)
```

```
## Data:      X dimension: 263 19
## Y dimension: 263 1
## Fit method: kernelpls
## Number of components considered: 1
## TRAINING: % variance explained
##           1 comps
## X           38.08
## Salary      43.05
```