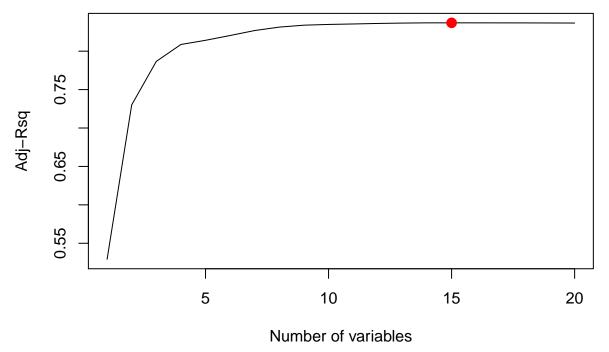
Life_expectancy project code

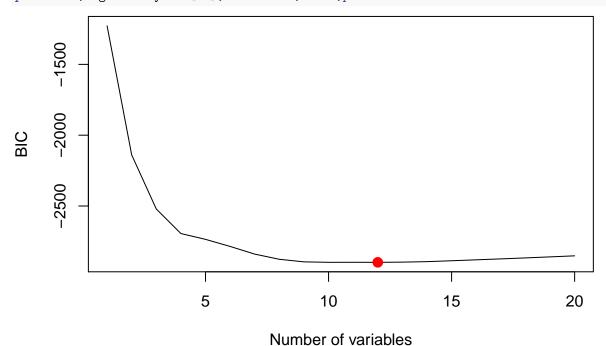
LitianZhou 04/07/2019

```
library(boot)
library(pls)
library(leaps)
library(gam)
library(glmnet)
life exp=read.csv("Life Expectancy Data.csv")
life_exp=na.omit(life_exp)
life_exp=life_exp[2:22] # remove the "country" predictor
attach(life_exp)
set.seed(1)
trainid <- sample(1:nrow(life_exp), nrow(life_exp)/2)</pre>
train <- life_exp[trainid,]</pre>
test <- life_exp[-trainid,]</pre>
require(boot)
glm.fit = glm(Life.expectancy~., data=life_exp)
loocv.err.OLS=cv.glm(life_exp, glm.fit)$delta[1]
loocv.err.OLS
## [1] 12.84627
require(leaps)
set.seed(7)
regfit.full=regsubsets(Life.expectancy~., life_exp,nvmax=20)
reg.summary=summary(regfit.full)
plot(reg.summary$adjr2, xlab = "Number of variables", ylab = "Adj-Rsq",type ="1")
which.max(reg.summary$adjr2) # give 15
## [1] 15
points(15,reg.summary$adjr2[15], col="red",cex=2,pch=20)
```



```
plot(reg.summary$bic, xlab = "Number of variables", ylab = "BIC",type ="l")
which.min(reg.summary$bic) # give 12
```

[1] 12
points(12,reg.summary\$bic[12], col="red",cex=2,pch=20)



```
coef(regfit.full,12)
## (Intercept) Year
```

3.048802e+02 -1.251579e-01 ## StatusDeveloping Adult.Mortality

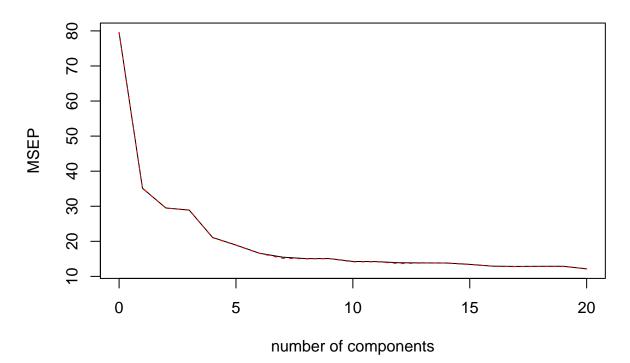
```
-9.240120e-01
##
                                                        -1.646155e-02
##
                      infant.deaths
                                                              Alcohol
                                                        -1.172219e-01
                       8.361896e-02
##
##
            percentage.expenditure
                                                                  BMT
                                                         3.697864e-02
##
                       4.638867e-04
##
                 under.five.deaths
                                                           Diphtheria
##
                      -6.395703e-02
                                                         1.541662e-02
##
                           HIV.AIDS Income.composition.of.resources
##
                      -4.457164e-01
                                                         1.056436e+01
##
                          Schooling
##
                       9.154314e-01
regfit.fwd=regsubsets(Life.expectancy~., life_exp, nvmax=20, method = "forward")
coef(regfit.fwd,12)
##
                                                                  Year
                        (Intercept)
##
                       3.048802e+02
                                                        -1.251579e-01
##
                   StatusDeveloping
                                                      Adult.Mortality
##
                      -9.240120e-01
                                                        -1.646155e-02
##
                      infant.deaths
                                                              Alcohol
                       8.361896e-02
                                                        -1.172219e-01
##
                                                                   BMI
##
            percentage.expenditure
##
                       4.638867e-04
                                                         3.697864e-02
##
                  under.five.deaths
                                                           Diphtheria
##
                      -6.395703e-02
                                                         1.541662e-02
##
                           HIV.AIDS Income.composition.of.resources
                                                         1.056436e+01
##
                      -4.457164e-01
##
                          Schooling
##
                       9.154314e-01
regfit.bwd=regsubsets(Life.expectancy~., life_exp, nvmax=20, method = "backward")
coef(regfit.bwd,12)
##
                        (Intercept)
                                                                  Year
##
                       3.048802e+02
                                                        -1.251579e-01
##
                   StatusDeveloping
                                                      Adult.Mortality
                      -9.240120e-01
                                                        -1.646155e-02
##
##
                      infant.deaths
                                                              Alcohol
                       8.361896e-02
                                                        -1.172219e-01
##
                                                                   BMI
##
            percentage.expenditure
                                                         3.697864e-02
##
                       4.638867e-04
                  under.five.deaths
##
                                                           Diphtheria
##
                      -6.395703e-02
                                                         1.541662e-02
##
                           HIV.AIDS Income.composition.of.resources
##
                      -4.457164e-01
                                                         1.056436e+01
##
                          Schooling
##
                       9.154314e-01
# Using 12 predictors, BSS, forward selection and backward selection all give the identical model
require(leaps)
# predict function from chapter 6 labs
predict.regsubsets <- function(object, newdata, id, ...){</pre>
  form <- as.formula(object$call[[2]])</pre>
  mat <- model.matrix(form, newdata)</pre>
```

```
coefi <- coef(object, id=id)</pre>
  xvars <- names(coefi)</pre>
  mat[,xvars]%*%coefi
}
k=10
set.seed(5)
folds=sample(1:k, nrow(life_exp),replace = TRUE)
cv.errors=matrix(NA,k,19,dimnames=list(NULL,paste(1:19)))
for(j in 1:k){
  best.fit=regsubsets(Life.expectancy~., data=life_exp[folds!=j,], nvmax = 20)
  for(i in 1:19){
    pred=predict(best.fit, life_exp[folds==j,],id=i)
    cv.errors[j,i]=mean((life_exp$Life.expectancy[folds==j]-pred)^2)
  }
}
mean.cv.errors=apply(cv.errors,2,mean)
plot(mean.cv.errors,type="b")
which.min(mean.cv.errors) # give 14, depend on the random seed
## 14
points(14,mean.cv.errors[14], col="red",cex=2,pch=20)
            0
     35
mean.cv.errors
     30
     25
     20
                   0-0-0-0-0-0-0-0-0-
     15
                                              10
                                                                 15
                           5
                                            Index
mean.cv.errors[14]
##
         14
## 12.78913
#use 14 predictors to fit the model
regfit.cv.select=regsubsets(Life.expectancy~.,data=life_exp, nvmax = 14)
coef(regfit.cv.select, 14)
```

```
##
                        (Intercept)
                                                                  Year
                      310.086873211
                                                         -0.127717305
##
##
                   StatusDeveloping
                                                      Adult.Mortality
                       -0.897508949
                                                         -0.016262157
##
##
                      infant.deaths
                                                               Alcohol
                        0.086080629
                                                         -0.129897924
##
            percentage.expenditure
##
                        0.000452335
                                                          0.031993887
##
##
                  under.five.deaths
                                                    Total.expenditure
                       -0.065159011
                                                          0.092009080
##
##
                         Diphtheria
                                                              HIV.AIDS
                        0.015089120
                                                          -0.447801053
##
##
                 thinness.5.9.years Income.composition.of.resources
                       -0.052118455
                                                          10.523194825
##
##
                          Schooling
##
                        0.904719271
fit.lm <- lm(Life.expectancy~., data=train)</pre>
pred.lm <- predict(fit.lm, test)</pre>
(err.lm <- mean((test$Life.expectancy - pred.lm)^2))</pre>
## [1] 13.77854
#err.lm may inflate the error since it uses validation set approach, which does not use whole data to b
require(glmnet)
xmat.train <- model.matrix(Life.expectancy~., data=train)[,-1]</pre>
xmat.test <- model.matrix(Life.expectancy~., data=test)[,-1]</pre>
fit.ridge <- cv.glmnet(xmat.train, train$Life.expectancy, alpha=0)
(lambda <- fit.ridge$lambda.min) # optimal lambda
## [1] 0.7133576
pred.ridge <- predict(fit.ridge, s=lambda, newx=xmat.test)</pre>
(err.ridge <- mean((test$Life.expectancy - pred.ridge)^2)) # test error</pre>
## [1] 14.39299
lifetest.avg <- mean(life_exp$Life.expectancy)</pre>
#ridge.r2 <- 1 - mean((pred.ridge - xmat.test$Life.expectancy)^2) / mean((lifetest.avg - xmat.test$Life</pre>
xmat.train <- model.matrix(Life.expectancy~., data=train)[,-1]</pre>
xmat.test <- model.matrix(Life.expectancy~., data=test)[,-1]</pre>
fit.lasso <- cv.glmnet(xmat.train, train$Life.expectancy, alpha=1)</pre>
(lambda <- fit.lasso$lambda.min) # optimal lambda
## [1] 0.002879837
pred.lasso <- predict(fit.lasso, s=lambda, newx=xmat.test)</pre>
(err.lasso <- mean((test$Life.expectancy - pred.lasso)^2)) # test error</pre>
## [1] 13.78092
coef.lasso <- predict(fit.lasso, type="coefficients", s=lambda)[1:ncol(life_exp),]</pre>
coef.lasso[coef.lasso != 0]
##
                        (Intercept)
                                                                  Year
                       3.744972e+02
                                                        -1.592786e-01
##
##
                   StatusDeveloping
                                                      Adult.Mortality
```

```
-1.382550e+00
                                                        -1.667109e-02
##
                      infant.deaths
##
                                                              Alcohol
                       7.489124e-02
                                                        -1.564712e-01
##
            percentage.expenditure
                                                          Hepatitis.B
##
##
                       3.947122e-04
                                                        -4.639169e-03
                            Measles
                                                                  BMI
##
##
                      -9.974977e-06
                                                         3.011031e-02
                 under.five.deaths
                                                                Polio
##
##
                      -5.695558e-02
                                                         8.483315e-04
                  Total.expenditure
                                                           Diphtheria
##
##
                      -1.158607e-02
                                                         1.553190e-02
                           HIV.AIDS
                                                                  GDP
##
                      -4.171617e-01
                                                         7.686975e-06
##
                                                thinness..1.19.years
##
                         Population
##
                      -9.191059e-10
                                                        -2.193032e-02
##
                 thinness.5.9.years Income.composition.of.resources
##
                      -2.439081e-02
                                                         1.055896e+01
##
                          Schooling
                       9.344881e-01
##
length(coef.lasso[coef.lasso != 0])
## [1] 21
require(pls)
set.seed(1)
fit.pcr <- pcr(Life.expectancy~., data=train, scale=TRUE, validation="CV")</pre>
validationplot(fit.pcr, val.type="MSEP") # I choose M = 7
```

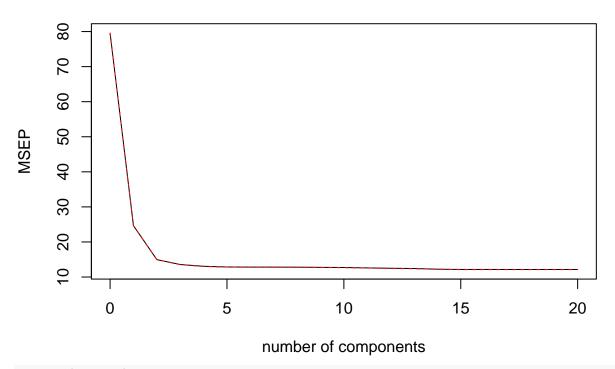
Life.expectancy



```
## Data:
            X dimension: 824 20
## Y dimension: 824 1
## Fit method: svdpc
## Number of components considered: 20
##
## VALIDATION: RMSEP
## Cross-validated using 10 random segments.
##
          (Intercept) 1 comps 2 comps 3 comps 4 comps 5 comps
## CV
                8.919
                         5.928
                                   5.432
                                            5.376
                                                     4.594
                                                               4.352
                                                                        4.075
## adjCV
                8.919
                         5.925
                                   5.429
                                            5.381
                                                     4.586
                                                               4.345
                                                                        4.070
##
          7 comps 8 comps 9 comps 10 comps 11 comps 12 comps 13 comps
## CV
            3.934
                     3.882
                               3.883
                                         3.776
                                                   3.770
                                                              3.726
                                                                        3.721
            3.892
                     3.874
                               3.878
                                         3.761
                                                   3.764
                                                              3.693
                                                                        3.714
## adjCV
##
          14 comps
                   15 comps
                              16 comps 17 comps
                                                  18 comps 19 comps
## CV
             3.718
                       3.664
                                  3.593
                                            3.583
                                                      3.588
                                                                 3.589
## adiCV
             3.712
                       3.661
                                  3.587
                                            3.577
                                                      3.582
                                                                 3.582
##
          20 comps
             3.487
## CV
             3.481
## adjCV
## TRAINING: % variance explained
##
                    1 comps 2 comps 3 comps 4 comps 5 comps 6 comps
## X
                      29.45
                                43.08
                                         51.87
                                                  59.66
                                                           66.19
                                                                     71.25
## Life.expectancy
                      56.59
                                63.56
                                         64.48
                                                  74.17
                                                           76.93
                                                                     79.94
##
                    7 comps 8 comps 9 comps 10 comps 11 comps 12 comps
## X
                      75.65
                               79.78
                                         83.46
                                                   86.40
                                                              89.13
                                                                        91.29
## Life.expectancy
                      81.91
                                81.91
                                         81.96
                                                   83.01
                                                              83.01
                                                                        83.61
##
                                                   16 comps
                    13 comps 14 comps 15 comps
                                                            17 comps
                                                                 99.43
## X
                       93.43
                                  95.37
                                            96.99
                                                      98.49
                       83.62
                                  83.74
                                            84.14
                                                      84.68
                                                                 84.79
## Life.expectancy
##
                    18 comps
                              19 comps
                                        20 comps
## X
                                  99.98
                                           100.00
                       99.79
## Life.expectancy
                       84.80
                                  84.81
                                            85.52
pred.pcr <- predict(fit.pcr, test, ncomp=10)</pre>
(err.pcr <- mean((test$Life.expectancy - pred.pcr)^2)) # test error</pre>
## [1] 15.37257
set.seed(1)
fit.pls <- plsr(Life.expectancy~., data=train, scale=TRUE, validation="CV")</pre>
validationplot(fit.pls, val.type="MSEP")
```

summary(fit.pcr)

Life.expectancy



summary(fit.pls)

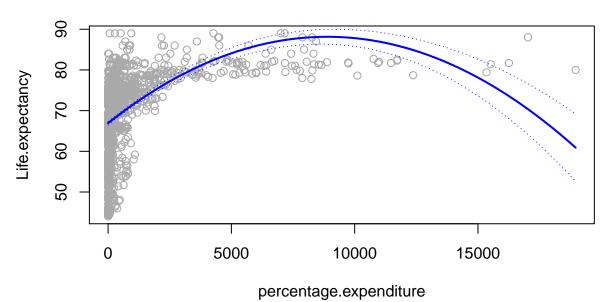
```
## Data:
            X dimension: 824 20
## Y dimension: 824 1
## Fit method: kernelpls
## Number of components considered: 20
##
## VALIDATION: RMSEP
  Cross-validated using 10 random segments.
##
          (Intercept)
                       1 comps 2 comps
                                         3 comps
                                                    4 comps
                                                            5 comps
                                                                       6 comps
## CV
                8.919
                          4.965
                                   3.869
                                             3.683
                                                      3.614
                                                               3.587
                                                                         3.582
                8.919
                          4.960
## adjCV
                                   3.866
                                             3.677
                                                      3.607
                                                               3.581
                                                                         3.576
##
          7 comps 8 comps
                            9 comps 10 comps 11 comps 12 comps
                                                                      13 comps
## CV
            3.582
                     3.579
                               3.572
                                         3.566
                                                    3.549
                                                              3.535
                                                                         3.525
            3.576
                     3.573
                               3.566
                                         3.560
                                                    3.542
                                                              3.526
## adjCV
                                                                         3.517
##
          14 comps
                    15 comps
                               16 comps
                                         17 comps
                                                    18 comps
                                                              19 comps
             3.498
                        3.487
                                  3.488
                                             3.487
                                                       3.486
                                                                  3.487
## CV
             3.490
                                  3.482
                                             3.481
                                                       3.480
## adjCV
                        3.480
                                                                  3.481
##
          20 comps
             3.487
## CV
## adjCV
             3.481
##
## TRAINING: % variance explained
                     1 comps
                              2 comps
                                       3 comps
                                                4 comps 5 comps
                                                                   6 comps
## X
                       28.66
                                39.26
                                         49.10
                                                   54.46
                                                            58.75
                                                                      64.63
                                                   84.58
## Life.expectancy
                       69.91
                                81.89
                                         83.91
                                                            84.77
                                                                      84.81
##
                    7 comps
                              8 comps
                                       9 comps
                                                10 comps
                                                           11 comps
                                                                      12 comps
                       69.49
                                74.42
                                         77.38
                                                    80.49
                                                              82.20
                                                                         83.61
                                         84.90
## Life.expectancy
                       84.83
                                84.85
                                                    84.97
                                                              85.15
                                                                         85.31
                     13 comps 14 comps 15 comps 16 comps 17 comps
```

```
89.26
## X
                        86.71
                                   87.60
                                                        92.80
                                                                   94.02
## Life.expectancy
                        85.35
                                   85.47
                                             85.51
                                                        85.51
                                                                   85.52
##
                     18 comps
                               19 comps
                                          20 comps
## X
                        95.14
                                   97.84
                                            100.00
## Life.expectancy
                        85.52
                                   85.52
                                             85.52
pred.pls <- predict(fit.pls, test, ncomp=5) # min Cv at M=5</pre>
(err.pls <- mean((test$Life.expectancy - pred.pls)^2)) # test error</pre>
## [1] 14.24667
```

fit the CV chosen model into poly, regression spline and GAM:

```
poly.fit=lm(Life.expectancy~poly(percentage.expenditure,2))
summary(poly.fit)
##
## Call:
## lm(formula = Life.expectancy ~ poly(percentage.expenditure, 2))
## Residuals:
##
       Min
                10 Median
                                3Q
                                       Max
           -4.079
                    1.315
                            5.288
                                   21.765
  -22.946
##
## Coefficients:
##
                                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                     69.3023
                                                0.1889 366.83
                                                                 <2e-16 ***
## poly(percentage.expenditure, 2)1 146.2843
                                                7.6718
                                                         19.07
                                                                  <2e-16 ***
## poly(percentage.expenditure, 2)2 -96.1937
                                                7.6718 -12.54
                                                                 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.672 on 1646 degrees of freedom
## Multiple R-squared: 0.2404, Adjusted R-squared: 0.2394
## F-statistic: 260.4 on 2 and 1646 DF, p-value: < 2.2e-16
percentage.expenditurelims = range(percentage.expenditure)
percentage.expenditure.grid=seq(from=percentage.expenditurelims[1], to=percentage.expenditurelims[2])
preds=predict(poly.fit,newdata = list(percentage.expenditure=percentage.expenditure.grid), se=TRUE)
se.bands=cbind(preds\fit+2*preds\fit, preds\fit-2*preds\fit)
par(oma=c(0,0,3,0))
plot(percentage.expenditure,Life.expectancy, xlim=percentage.expenditurelims, col="darkgrey")
title("Degree-2 Polynomial",outer=T)
lines(percentage.expenditure.grid, preds$fit, lwd="2",col="blue")
matlines(percentage.expenditure.grid, se.bands, lwd=1, col="blue",lty=3)
```

Degree-2 Polynomial

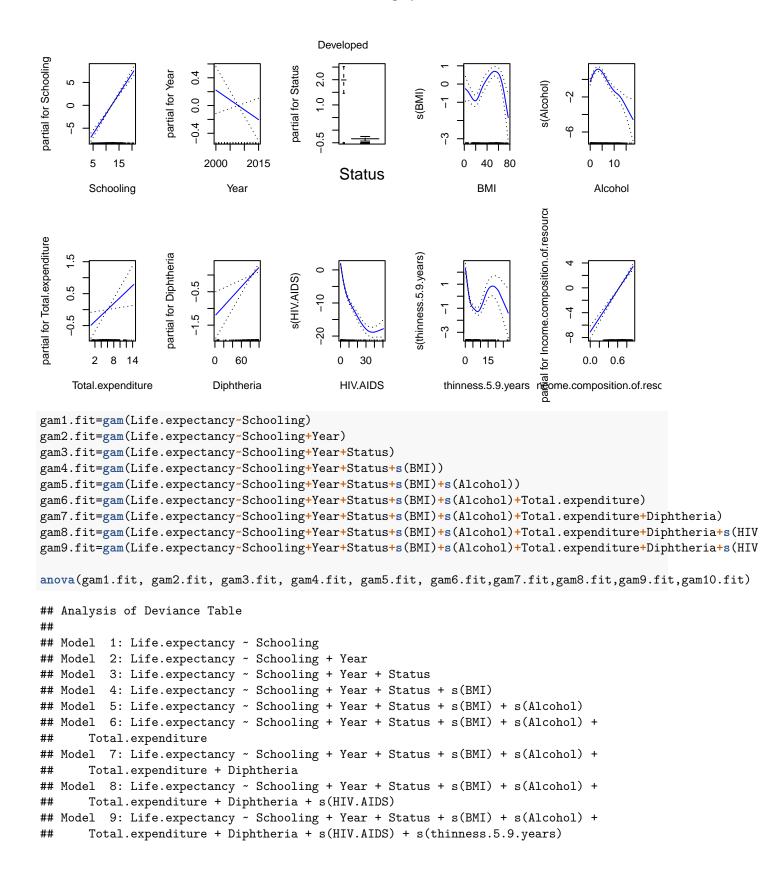


graph shows that as people increase there expenditure on health, the life expenctancy first increase, then decrease (or unchange).

```
require(gam)
# Choose from best subset model of 14 predictors. Then remove "Adult.Mortality", "infant.deaths" , "unde
gam10.fit=gam(Life.expectancy~Schooling+Year+Status+s(BMI)+ s(Alcohol)+Total.expenditure+Diphtheria+s(H
par(mfrow=c(2,5))
par(oma=c(0,0,3,0))
plot.Gam(gam10.fit, se=TRUE, col="blue")
title("GAM: linear + smoothing spline fit", outer=TRUE)
```

The

GAM: linear + smoothing spline fit



```
## Model 10: Life.expectancy ~ Schooling + Year + Status + s(BMI) + s(Alcohol) +
##
       Total.expenditure + Diphtheria + s(HIV.AIDS) + s(thinness.5.9.years) +
       Income.composition.of.resources
##
##
      Resid. Df Resid. Dev
                               Df Deviance Pr(>Chi)
## 1
           1647
                     60009
## 2
           1646
                     59985 1.0000
                                      24.5 0.149053
## 3
                     59160 1.0000
                                     824.7 < 2.2e-16 ***
           1645
                                    7765.8 < 2.2e-16 ***
## 4
           1641
                     51395 4.0001
## 5
           1637
                     48722 4.0001
                                    2672.9 < 2.2e-16 ***
## 6
           1636
                     48597 1.0000
                                    125.1 0.001098 **
                     47673 1.0000
## 7
           1635
                                     923.1 < 2.2e-16 ***
## 8
           1631
                     22994 4.0002 24679.9 < 2.2e-16 ***
## 9
           1627
                     21220 3.9998
                                   1773.8 < 2.2e-16 ***
## 10
                     19099 1.0000
                                    2120.8 < 2.2e-16 ***
           1626
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
trainid <- sample(1:nrow(life_exp), nrow(life_exp)/2)</pre>
train <- life_exp[trainid,]</pre>
test <- life_exp[-trainid,]</pre>
#according to the ANOVA test, we find the gam6 has a low p-value, to keep the model easier to grasp, we
# get the test error by VS approach
gam.fit.train=gam(Life.expectancy~Schooling+Year+Status+s(BMI)+ s(Alcohol)+Diphtheria+s(HIV.AIDS)+ s(th
gam.final.pred = predict(gam.fit.train, test)
(gam.error = mean((test$Life.expectancy - gam.final.pred)^2))
## [1] 12.7334
# fit the model with 9 predictors:
gam.final.fit=gam(Life.expectancy~Schooling+Year+Status+s(BMI)+ s(Alcohol)+Diphtheria+s(HIV.AIDS)+ s(th
summary(gam.final.fit)
##
## Call: gam(formula = Life.expectancy ~ Schooling + Year + Status + s(BMI) +
       s(Alcohol) + Diphtheria + s(HIV.AIDS) + s(thinness.5.9.years) +
##
       Income.composition.of.resources)
##
## Deviance Residuals:
                1Q Median
       Min
                                3Q
## -14.707 -1.973
                     0.102
                             1.948
                                    12.569
## (Dispersion Parameter for gaussian family taken to be 11.7727)
##
       Null Deviance: 127529.3 on 1648 degrees of freedom
##
## Residual Deviance: 19154.26 on 1627 degrees of freedom
## AIC: 8769.594
## Number of Local Scoring Iterations: 2
## Anova for Parametric Effects
##
                                     Df Sum Sq Mean Sq F value
                                                                    Pr(>F)
## Schooling
                                         50254
                                                  50254 4268.698 < 2.2e-16 ***
                                      1
                                                          22.028 2.912e-06 ***
## Year
                                            259
                                                    259
                                       1
```

```
## Status
                                      1
                                          1186
                                                  1186 100.731 < 2.2e-16 ***
## s(BMI)
                                      1
                                          2424
                                                  2424 205.879 < 2.2e-16 ***
## s(Alcohol)
                                      1
                                           903
                                                   903
                                                        76.690 < 2.2e-16 ***
## Diphtheria
                                           715
                                                   715
                                                         60.697 1.178e-14 ***
                                      1
## s(HIV.AIDS)
                                      1
                                         23362
                                                 23362 1984.455 < 2.2e-16 ***
## s(thinness.5.9.years)
                                           364
                                                   364
                                                         30.961 3.073e-08 ***
                                      1
## Income.composition.of.resources
                                          2399
                                                  2399 203.785 < 2.2e-16 ***
                                      1
## Residuals
                                   1627 19154
                                                    12
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Anova for Nonparametric Effects
                                   Npar Df Npar F
                                                       Pr(F)
## (Intercept)
## Schooling
## Year
## Status
## s(BMI)
                                         3 15.117 1.048e-09 ***
## s(Alcohol)
                                         3 26.929 < 2.2e-16 ***
## Diphtheria
## s(HIV.AIDS)
                                         3 107.825 < 2.2e-16 ***
## s(thinness.5.9.years)
                                         3 55.714 < 2.2e-16 ***
## Income.composition.of.resources
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
The MSE of those methods are pretty close, while GAM has the best performace.
err.all <- c(err.lm, loocv.err.OLS, err.ridge, err.lasso, err.pcr, err.pls, gam.error)
names(err.all) <- c("OLS-VS", "OLS-LOOCV", "ridge", "lasso", "pcr", "pls", "gam-VS")</pre>
barplot(err.all )
```

