

Advanced Statistical Modelling: Ridge Regression

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Choosing the penalization parameter λ

The objective of this exercise is to implement Ridge Regression with two different approaches: $MSPE_{val}(\lambda)$ and $MSPE_{k-CV}(\lambda)$. In both cases we are going to take as input data the following:

- Matrix x and vector y corresponding to the training sample.
- Matrix x_{val} and vector y_{val} corresponding to the validation set.
- Vector $lambda.v$ of candidate values for λ .

We are going to output for each element λ in $lambda.v$ and the value of the $MSPE_{val}(\lambda) / MSPE_{k-CV}(\lambda)$. Furthermore, we are going to plot these values against $\log(1 + \lambda) - 1$.

Once we have build these two functions, we are going to use the prostate data used in class. We are going to choose a λ according to:

- Behaviour in the validation set (30 validations not included in the training sample)
- 5-fold, 10-fold cross-validation.
- Compare our results with those obtained with leave-one-out and generalized cross-validation.

Ridge regression based on $MSPE_{val}(\lambda)$

In order to choose the penalization parameter λ , we are going to write a function implementing the ridge regression penalization parameter λ choice based on the minimization of the $MSPE_{val}(\lambda)$.

```
ridge_lambda_search <- function(x, y, x.val, y.val, lambda.v) {
  result <- data.frame(lambda=lambda.v, mspe=rep(0,length(lambda.v)), df=rep(0,length(lambda.v)))
  x.svd <- svd(x)
  for(i in 1:length(lambda.v)) {
    lambda <- lambda.v[i]
    d_inv <- diag(1/(x.svd$d*x.svd$d - lambda))
    xx_inv <- t( solve( t(x) %*% x + lambda*diag(1,ncol(x)) ))
    beta <- xx_inv %*% t(x) %*% y
    y.hat <- x.val %*% beta
    mspe <- sum((y.val - y.hat)^2) / length(y.val)
    df <- sum(x.svd$d^2 / (x.svd$d^2 +lambda))
    result$mspe[i] <- mspe
    result$df[i] <- df
  }
  plot(mspe~log(1+lambda), result, col=2)
  lambda.min <- result$lambda[which.min(result$mspe)]
  abline(v=log(1+lambda.min),col=2,lty=2)
  plot(mspe~df, result, col=3)
  df.min <- result$df[which.min(result$mspe)]
  abline(v=df.min,col=3,lty=2)
  return(result)
}
```

Ridge regression based on $MSP E_{k-CV}(\lambda)$

Now, we will write an R function implementing the ridge regression penalization parameter λ choice based on k-fold cross-validation $MSP E_{kCV}(\lambda)$

```
ridge_lambda_search_cv <- function(x, y, lambda.v, cv=10) {
  result <- data.frame(lambda=lambda.v, mspe=rep(0,length(lambda.v)), df=rep(0,length(lambda.v)))
  for(i in 1:length(lambda.v)) {
    lambda.v[i] <- lambda.v[i]
    folds <- createFolds(1:nrow(x), k = cv)
    result.cv <- data.frame(mspe=rep(0,cv), df=rep(0,cv))
    for(j in 1:cv) {
      fold <- folds[[j]]
      x.train <- x[-fold,]
      y.train <- y[-fold]
      x.val <- x[fold,]
      y.val <- y[fold]
      x.svd <- svd(x.train)
      d <- x.svd$d
      v <- x.svd$v
      d_inv <- diag(1/(d*d - lambda.v[i]))
      xx_inv <- t( solve( t(x.train) %*% x.train + lambda.v[i]*diag(1,ncol(x)) ))
      beta <- xx_inv %*% t(x.train) %*% y.train
      y.hat <- x.val %*% beta
      mspe <- sum((y.val - y.hat)^2) / length(y.val)
      df <- sum(d^2 / (d^2 + lambda.v[i]))
      result.cv$mspe[j] <- mspe
      result.cv$df[j] <- df
    }
    result$mspe[i] <- mean(result.cv$mspe)
    result$df[i] <- mean(result.cv$df)
  }
  plot(mspe~log(1+lambda), result, col=2)
  lambda.min <- result$lambda[which.min(result$mspe)]
  abline(v=log(1+lambda.min),col=2,lty=2)
  plot(mspe~df, result, col=3)
  df.min <- result$df[which.min(result$mspe)]
  abline(v=df.min,col=3,lty=2)

  return(result)
}
```

Comparison between penalization parameters of $MSP E_{val}(\lambda)$ and $MSP E_{k-CV}(\lambda)$

```
ridge_lambda_search_loocv_gcv <- function(x, y, lambda.v) {
  l <- length(lambda.v)
  result <- data.frame(lambda=lambda.v, loocv=rep(0,l), gcv=rep(0,l),df=rep(0,l))
  n <- nrow(x)
  x.svd <- svd(x)
  d <- x.svd$d
  v <- x.svd$v
  u <- x.svd$u
```

```

for(i in 1:l) {
  lambda <- lambda.v[i]
  d_inv <- diag(1/(d^2 - lambda))
  xx_inv <- t( solve( t(x) %*% x + lambda*diag(1,ncol(x)) ))
  beta <- (xx_inv %*% t(x)) %*% y
  y.hat <- x %*% beta
  df <- sum(d^2 / (d^2 +lambda))
  h <- x %*% xx_inv %*% t(x)
  result$loocv[i] <- sum( ( (y - y.hat)/(1 - diag(h)) )^2 ) / n
  result$gcv[i] <- sum( ( (y - y.hat)/(1 - df/n) )^2 ) / n
  result$df[i] <- df
}
plot(loocv~log(1+lambda), result, col=2)
lambda.min <- result$lambda[which.min(result$loocv)]
abline(v=log(1+lambda.min),col=2,lty=2)
plot(loocv~df, result, col=3)
df.min <- result$df[which.min(result$loocv)]
abline(v=df.min,col=3,lty=2)
plot(gcv~log(1+lambda), result, col=2)
lambda.min <- result$lambda[which.min(result$gcv)]
abline(v=log(1+lambda.min),col=2,lty=2)
plot(gcv~df, result, col=3)
df.min <- result$df[which.min(result$gcv)]
abline(v=df.min,col=3,lty=2)
return(result)
}

```

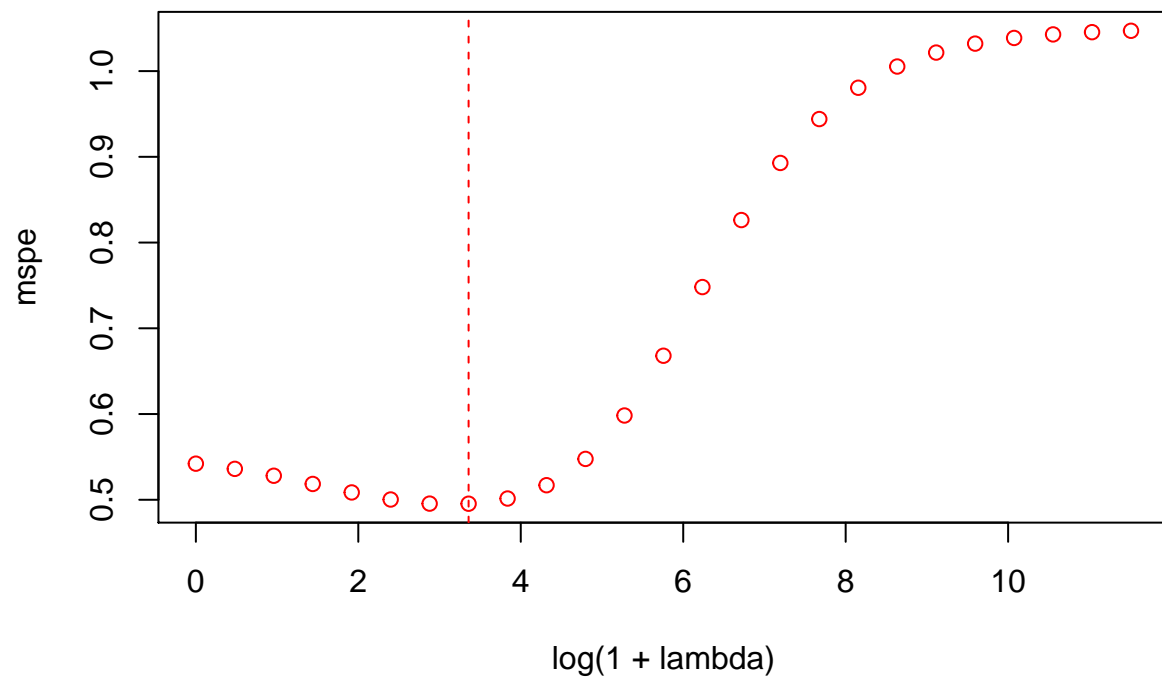
```

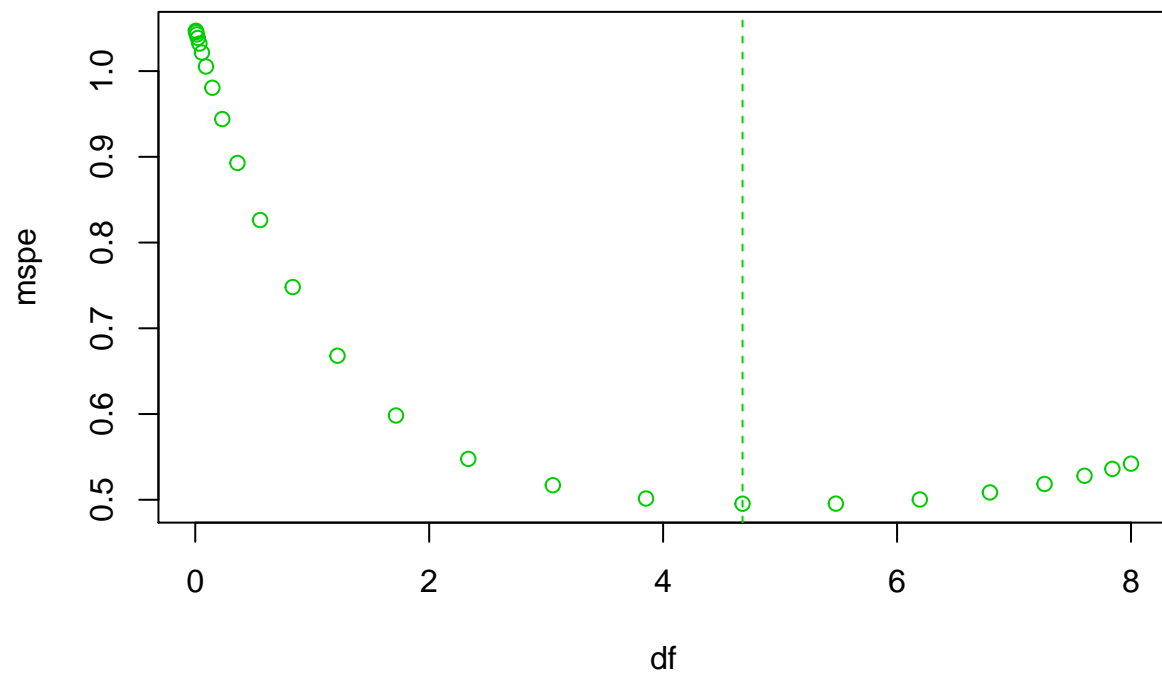
prostate <- fread(file="prostate_data.txt", header=TRUE)
prostate$train <- as.logical(prostate$train)
prostate$No <- NULL
prostate$train <- as.logical(prostate$train)
prostate <- as.data.frame(prostate)
data <- prostate[,1:9]
lambda.max <- 1e5
n.lambdas <- 25
lambda.v <- exp(seq(0,log(lambda.max+1),length=n.lambdas))-1
validation.ind <- prostate$train
validation <- data[!validation.ind,]
training <- data[validation.ind,]

x <- scale(training[,1:8], center = T, scale = T)
y <- scale(training[,9], center = T, scale = F)

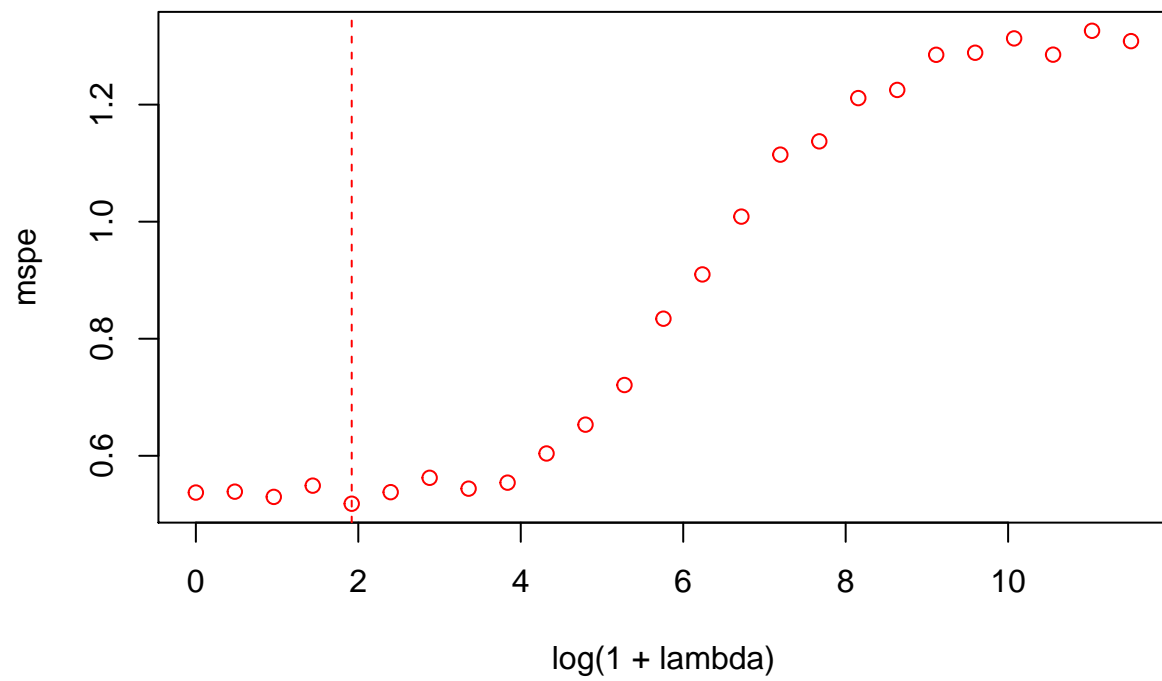
x.val <- scale(validation[,1:8], center = T, scale = T)
y.val <- scale(validation[,9], center = T, scale = F)
result.valid <- ridge_lambda_search(x, y, x.val, y.val, lambda.v)

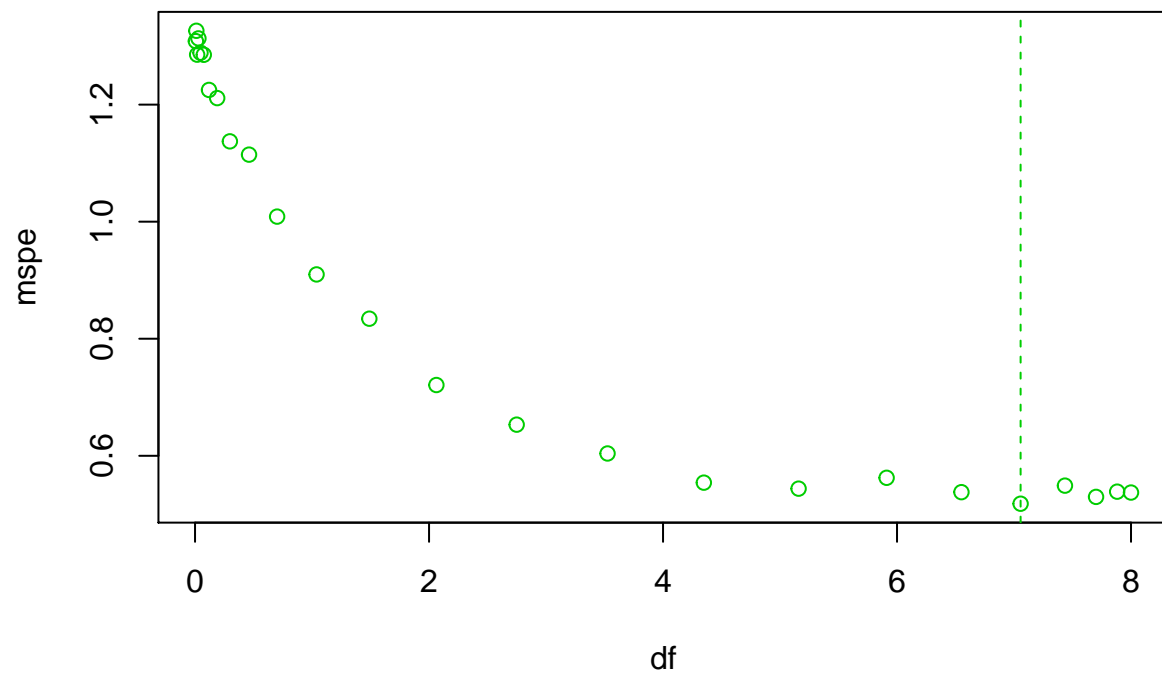
```



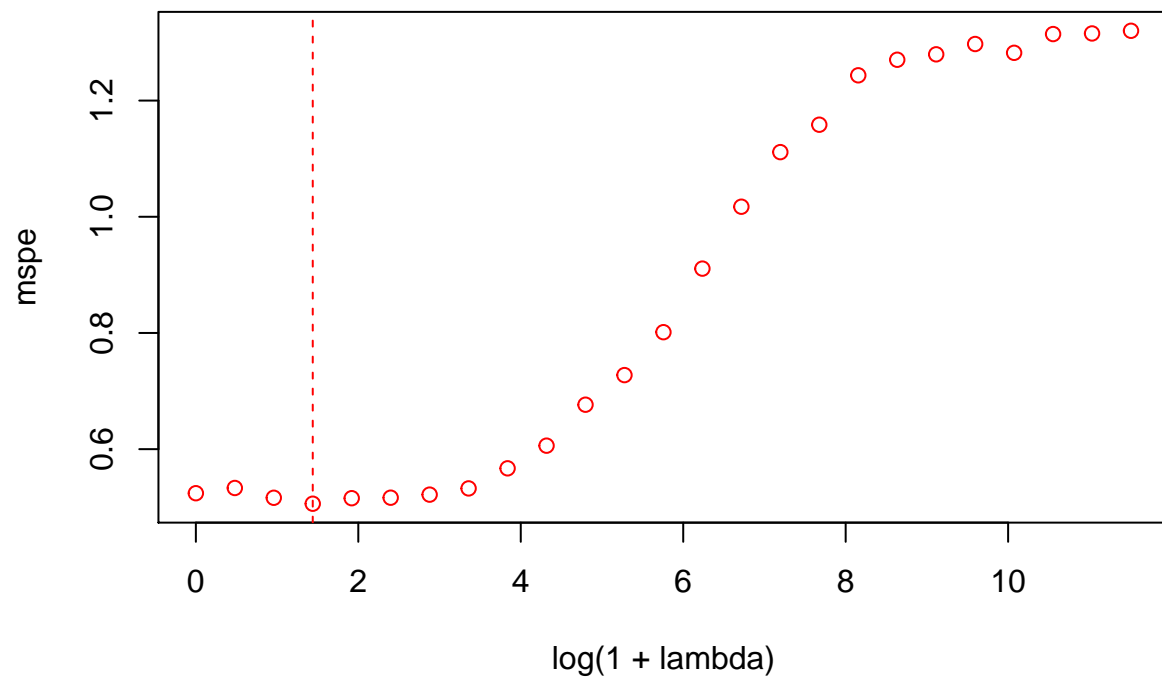


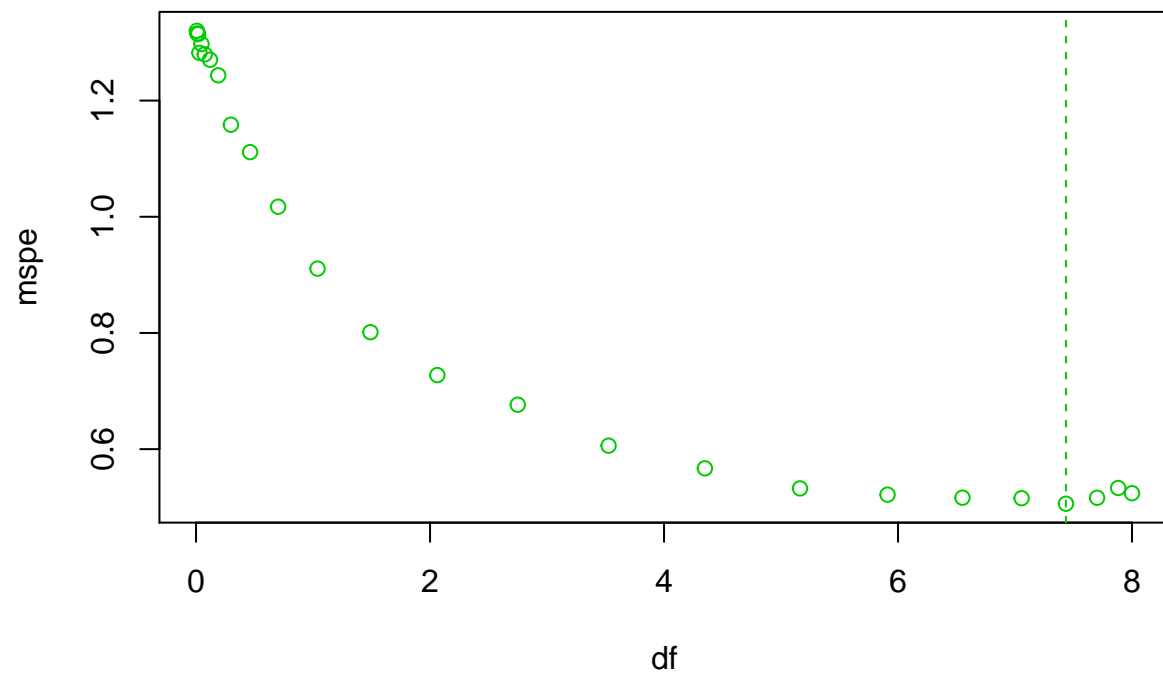
```
# 5-fold and 10-fold CV
# With CV
x <- scale(data[,1:8], center = T, scale = T)
y <- scale(data[,9], center = T, scale = F)
result.5.cv <- ridge_lambda_search_cv(x, y, cv = 10, lambda.v)
```



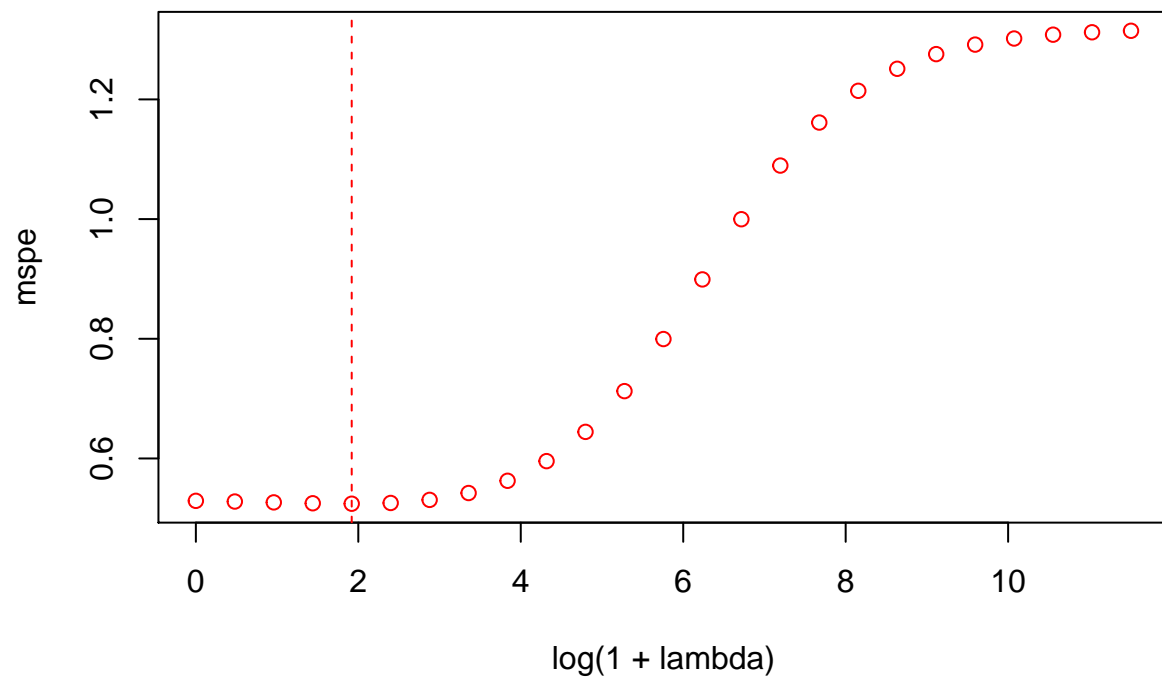


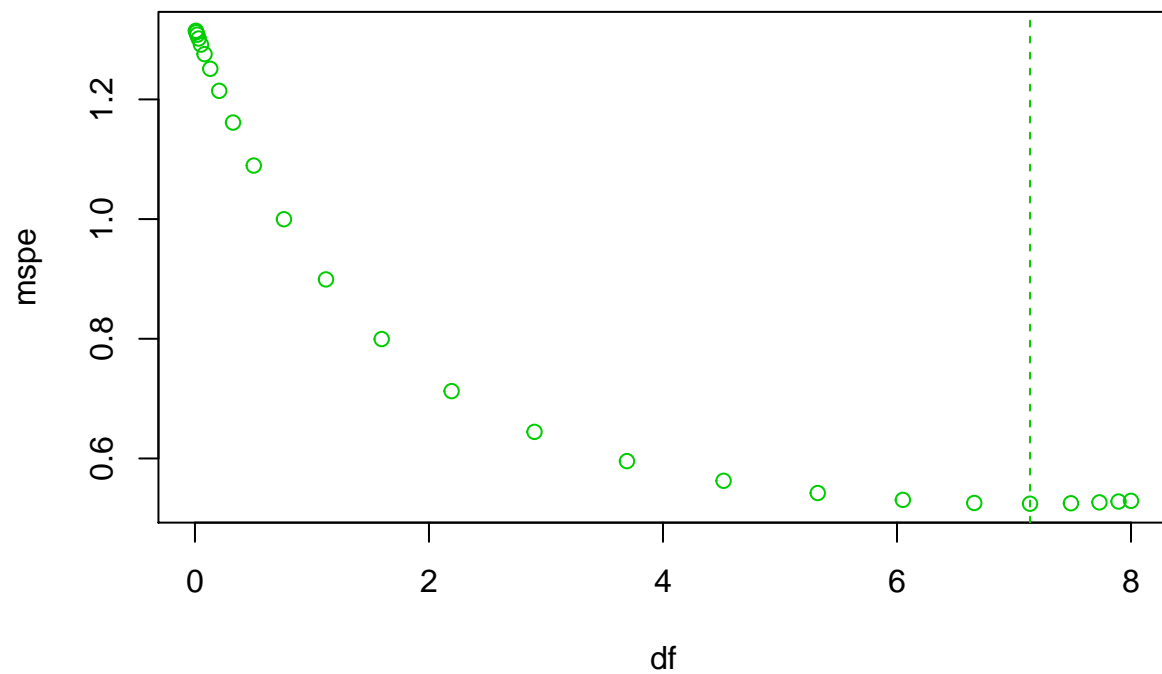
```
result.10.cv <- ridge_lambda_search_cv(x, y, cv = 10, lambda.v)
```



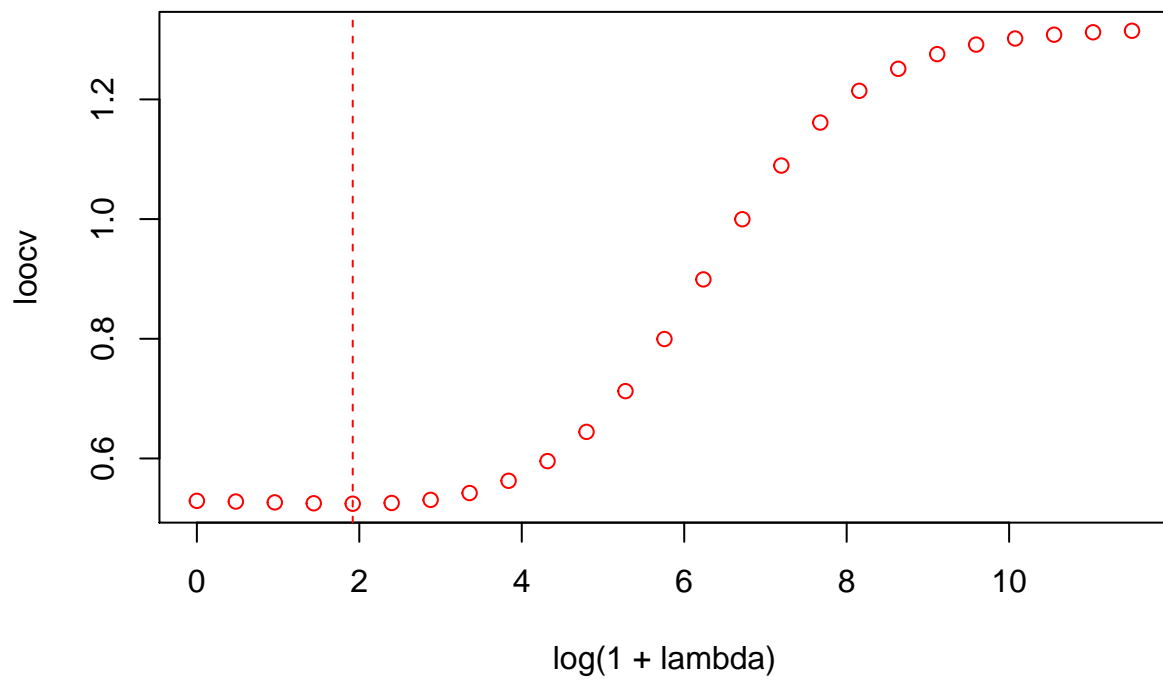


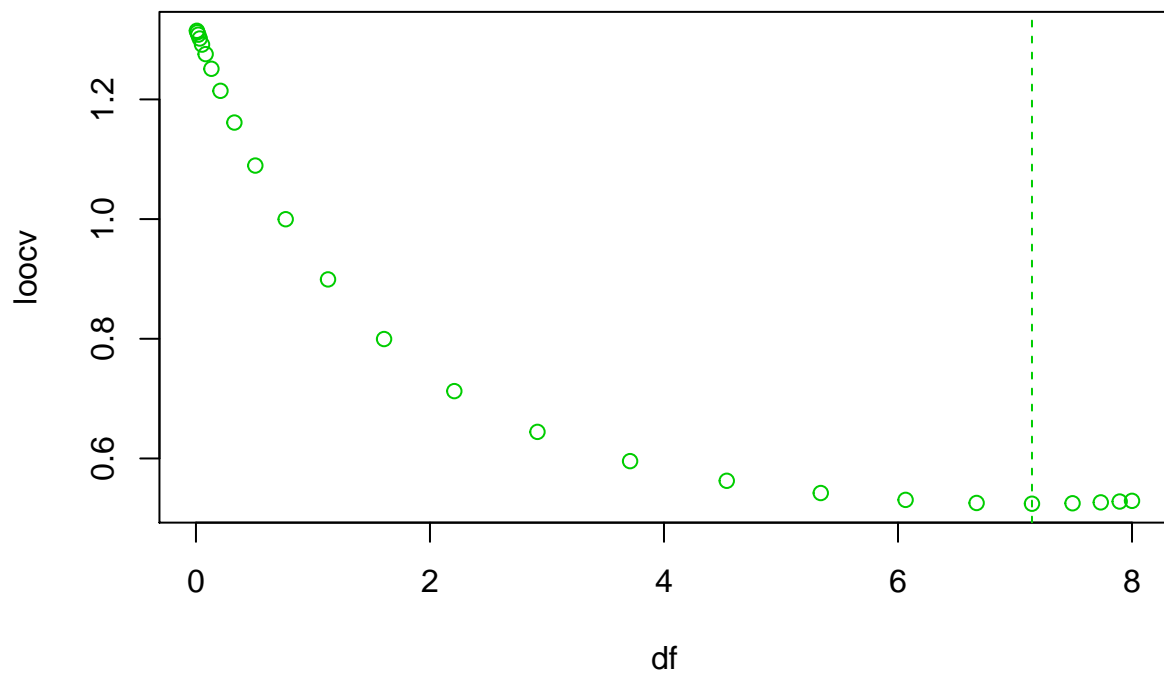
```
# LOOCV and GCV estimates
result.loocv <- ridge_lambda_search_cv(x, y, cv = nrow(x), lambda.v)
```

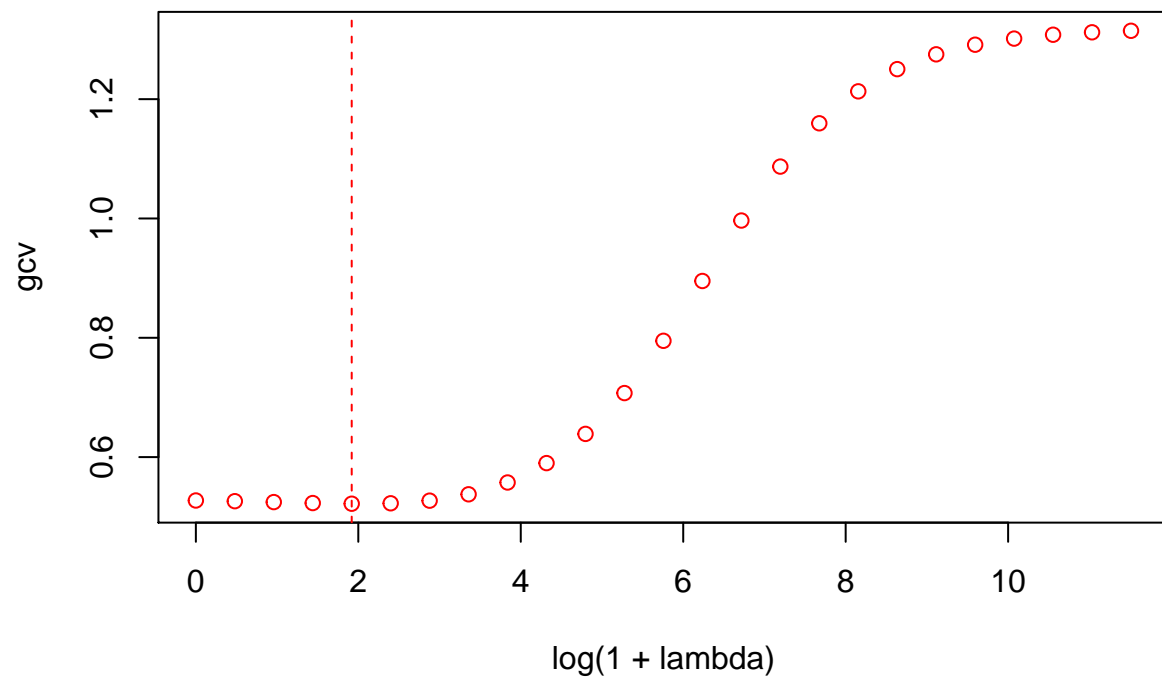


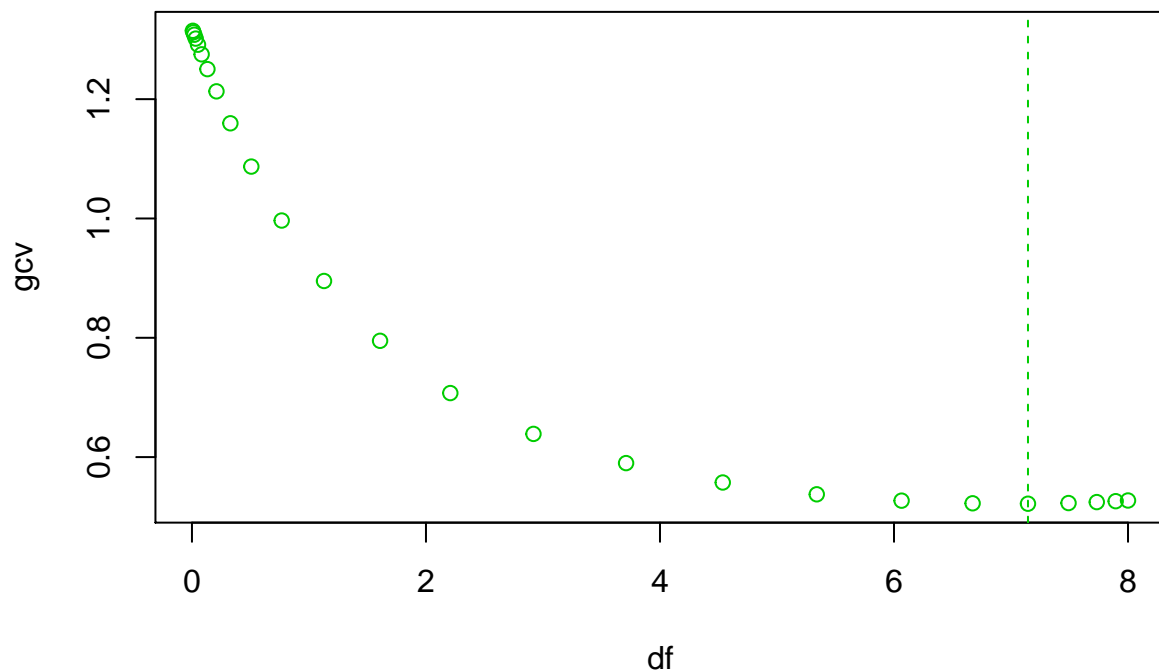


```
result.loocv.gcv <- ridge_lambda_search_loocv_gcv(x, y, lambda.v)
```









We observe that the results of MSPE seem more stable in the LOOCV and GCV compared to 5-fold and 10-fold CV, as one would expect given the small size of the dataset. Furthermore, the validation set results are also relatively stable, although with a different curve than for the LOOCV/GCV.

Ridge regression for the Boston Housing data

The Boston Housing dataset is a classical dataset which contains the values of 506 suburbs of Boston corresponding to 1978. This dataset can be found in many places but we are going to use a version with some corrections that was provided to us, which additionally includes the UTM coordinates of the geographical centers of each neighborhood. Therefore, the variables are the following:

Variable	Description	Type
CRIM	per capita crime rate by town	Numeric
ZN	proportion of residential land zoned for lots over 25,000 sq.ft.	Numeric
INDUS	proportion of non-retail business acres per town	Numeric
CHAS	Charles River dummy variable (= 1 if tract bounds river; 0 otherwise)	Factor
NOX	nitric oxides concentration (parts per 10 million)	Numeric
RM	average number of rooms per dwelling	Numeric
AGE	proportion of owner-occupied units built prior to 1940	Numeric
DIS	weighted distances to five Boston employment centres	Numeric
RAD	index of accessibility to radial highways	Numeric
TAX	full-value property-tax rate per \$10,000	Numeric
PTRATIO	pupil-teacher ratio by town	Numeric
B	$1000(\text{Bk} - 0.63)^2$ where Bk is the proportion of blacks by town	Numeric
LSTAT	% lower status of the population	Numeric

Variable	Description	Type
MEDV	Median value of owner-occupied homes in \$1000's	Numeric

In this exercise we are going to use ridge regression on the Boston Housing dataset to fit the regression model where the response variable is *MEDV* and the explanatory variables are the remaining 13 variables shown in the list. As we can see when loading the data, there are more variables than the ones listed (*TOWN*, *TOWNNO*, *LON*, *LAT*, *CMEDV*). We decided not to use them since the statement explicitly defines which to use.

Beside from eliminating variables, we divided the dataset between train and test. To do so, we have decided to do 3/4 training, 1/4 test.

train

##	MEDV	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX
## 1	24.0	0.00632	18.0	2.31	0	0.5380	6.575	65.2	4.0900	1	296
## 2	21.6	0.02731	0.0	7.07	0	0.4690	6.421	78.9	4.9671	2	242
## 3	34.7	0.02729	0.0	7.07	0	0.4690	7.185	61.1	4.9671	2	242
## 4	33.4	0.03237	0.0	2.18	0	0.4580	6.998	45.8	6.0622	3	222
## 5	36.2	0.06905	0.0	2.18	0	0.4580	7.147	54.2	6.0622	3	222
## 6	28.7	0.02985	0.0	2.18	0	0.4580	6.430	58.7	6.0622	3	222
## 8	27.1	0.14455	12.5	7.87	0	0.5240	6.172	96.1	5.9505	5	311
## 9	16.5	0.21124	12.5	7.87	0	0.5240	5.631	100.0	6.0821	5	311
## 10	18.9	0.17004	12.5	7.87	0	0.5240	6.004	85.9	6.5921	5	311
## 12	18.9	0.11747	12.5	7.87	0	0.5240	6.009	82.9	6.2267	5	311
## 16	19.9	0.62739	0.0	8.14	0	0.5380	5.834	56.5	4.4986	4	307
## 17	23.1	1.05393	0.0	8.14	0	0.5380	5.935	29.3	4.4986	4	307
## 19	20.2	0.80271	0.0	8.14	0	0.5380	5.456	36.6	3.7965	4	307
## 21	13.6	1.25179	0.0	8.14	0	0.5380	5.570	98.1	3.7979	4	307
## 23	15.2	1.23247	0.0	8.14	0	0.5380	6.142	91.7	3.9769	4	307
## 27	16.6	0.67191	0.0	8.14	0	0.5380	5.813	90.3	4.6820	4	307
## 28	14.8	0.95577	0.0	8.14	0	0.5380	6.047	88.8	4.4534	4	307
## 29	18.4	0.77299	0.0	8.14	0	0.5380	6.495	94.4	4.4547	4	307
## 30	21.0	1.00245	0.0	8.14	0	0.5380	6.674	87.3	4.2390	4	307
## 31	12.7	1.13081	0.0	8.14	0	0.5380	5.713	94.1	4.2330	4	307
## 32	14.5	1.35472	0.0	8.14	0	0.5380	6.072	100.0	4.1750	4	307
## 33	13.2	1.38799	0.0	8.14	0	0.5380	5.950	82.0	3.9900	4	307
## 34	13.1	1.15172	0.0	8.14	0	0.5380	5.701	95.0	3.7872	4	307
## 35	13.5	1.61282	0.0	8.14	0	0.5380	6.096	96.9	3.7598	4	307
## 36	18.9	0.06417	0.0	5.96	0	0.4990	5.933	68.2	3.3603	5	279
## 37	20.0	0.09744	0.0	5.96	0	0.4990	5.841	61.4	3.3779	5	279
## 38	21.0	0.08014	0.0	5.96	0	0.4990	5.850	41.5	3.9342	5	279
## 39	24.7	0.17505	0.0	5.96	0	0.4990	5.966	30.2	3.8473	5	279
## 41	34.9	0.03359	75.0	2.95	0	0.4280	7.024	15.8	5.4011	3	252
## 42	26.6	0.12744	0.0	6.91	0	0.4480	6.770	2.9	5.7209	3	233
## 44	24.7	0.15936	0.0	6.91	0	0.4480	6.211	6.5	5.7209	3	233
## 48	16.6	0.22927	0.0	6.91	0	0.4480	6.030	85.5	5.6894	3	233
## 51	19.7	0.08873	21.0	5.64	0	0.4390	5.963	45.7	6.8147	4	243
## 53	25.0	0.05360	21.0	5.64	0	0.4390	6.511	21.1	6.8147	4	243
## 56	35.4	0.01311	90.0	1.22	0	0.4030	7.249	21.9	8.6966	5	226
## 57	24.7	0.02055	85.0	0.74	0	0.4100	6.383	35.7	9.1876	2	313
## 58	31.6	0.01432	100.0	1.32	0	0.4110	6.816	40.5	8.3248	5	256
## 59	23.3	0.15445	25.0	5.13	0	0.4530	6.145	29.2	7.8148	8	284
## 60	19.6	0.10328	25.0	5.13	0	0.4530	5.927	47.2	6.9320	8	284

## 61	18.7	0.14932	25.0	5.13	0	0.4530	5.741	66.2	7.2254	8	284
## 62	16.0	0.17171	25.0	5.13	0	0.4530	5.966	93.4	6.8185	8	284
## 63	22.2	0.11027	25.0	5.13	0	0.4530	6.456	67.8	7.2255	8	284
## 64	25.0	0.12650	25.0	5.13	0	0.4530	6.762	43.4	7.9809	8	284
## 66	23.5	0.03584	80.0	3.37	0	0.3980	6.290	17.8	6.6115	4	337
## 67	19.4	0.04379	80.0	3.37	0	0.3980	5.787	31.1	6.6115	4	337
## 68	22.0	0.05789	12.5	6.07	0	0.4090	5.878	21.4	6.4980	4	345
## 69	17.4	0.13554	12.5	6.07	0	0.4090	5.594	36.8	6.4980	4	345
## 72	21.7	0.15876	0.0	10.81	0	0.4130	5.961	17.5	5.2873	4	305
## 73	22.8	0.09164	0.0	10.81	0	0.4130	6.065	7.8	5.2873	4	305
## 74	23.4	0.19539	0.0	10.81	0	0.4130	6.245	6.2	5.2873	4	305
## 75	24.1	0.07896	0.0	12.83	0	0.4370	6.273	6.0	4.2515	5	398
## 77	20.0	0.10153	0.0	12.83	0	0.4370	6.279	74.5	4.0522	5	398
## 79	21.2	0.05646	0.0	12.83	0	0.4370	6.232	53.7	5.0141	5	398
## 80	20.3	0.08387	0.0	12.83	0	0.4370	5.874	36.6	4.5026	5	398
## 81	28.0	0.04113	25.0	4.86	0	0.4260	6.727	33.5	5.4007	4	281
## 82	23.9	0.04462	25.0	4.86	0	0.4260	6.619	70.4	5.4007	4	281
## 83	24.8	0.03659	25.0	4.86	0	0.4260	6.302	32.2	5.4007	4	281
## 84	22.9	0.03551	25.0	4.86	0	0.4260	6.167	46.7	5.4007	4	281
## 86	26.6	0.05735	0.0	4.49	0	0.4490	6.630	56.1	4.4377	3	247
## 88	22.2	0.07151	0.0	4.49	0	0.4490	6.121	56.8	3.7476	3	247
## 89	23.6	0.05660	0.0	3.41	0	0.4890	7.007	86.3	3.4217	2	270
## 91	22.6	0.04684	0.0	3.41	0	0.4890	6.417	66.1	3.0923	2	270
## 92	22.0	0.03932	0.0	3.41	0	0.4890	6.405	73.9	3.0921	2	270
## 94	25.0	0.02875	28.0	15.04	0	0.4640	6.211	28.9	3.6659	4	270
## 95	20.6	0.04294	28.0	15.04	0	0.4640	6.249	77.3	3.6150	4	270
## 98	38.7	0.12083	0.0	2.89	0	0.4450	8.069	76.0	3.4952	2	276
## 99	43.8	0.08187	0.0	2.89	0	0.4450	7.820	36.9	3.4952	2	276
## 100	33.2	0.06860	0.0	2.89	0	0.4450	7.416	62.5	3.4952	2	276
## 102	26.5	0.11432	0.0	8.56	0	0.5200	6.781	71.3	2.8561	5	384
## 104	19.3	0.21161	0.0	8.56	0	0.5200	6.137	87.4	2.7147	5	384
## 107	19.5	0.17120	0.0	8.56	0	0.5200	5.836	91.9	2.2110	5	384
## 108	20.4	0.13117	0.0	8.56	0	0.5200	6.127	85.2	2.1224	5	384
## 109	19.8	0.12802	0.0	8.56	0	0.5200	6.474	97.1	2.4329	5	384
## 110	19.4	0.26363	0.0	8.56	0	0.5200	6.229	91.2	2.5451	5	384
## 111	21.7	0.10793	0.0	8.56	0	0.5200	6.195	54.4	2.7778	5	384
## 112	22.8	0.10084	0.0	10.01	0	0.5470	6.715	81.6	2.6775	6	432
## 113	18.8	0.12329	0.0	10.01	0	0.5470	5.913	92.9	2.3534	6	432
## 114	18.7	0.22212	0.0	10.01	0	0.5470	6.092	95.4	2.5480	6	432
## 115	18.5	0.14231	0.0	10.01	0	0.5470	6.254	84.2	2.2565	6	432
## 116	18.3	0.17134	0.0	10.01	0	0.5470	5.928	88.2	2.4631	6	432
## 118	19.2	0.15098	0.0	10.01	0	0.5470	6.021	82.6	2.7474	6	432
## 119	20.4	0.13058	0.0	10.01	0	0.5470	5.872	73.1	2.4775	6	432
## 120	19.3	0.14476	0.0	10.01	0	0.5470	5.731	65.2	2.7592	6	432
## 121	22.0	0.06899	0.0	25.65	0	0.5810	5.870	69.7	2.2577	2	188
## 122	20.3	0.07165	0.0	25.65	0	0.5810	6.004	84.1	2.1974	2	188
## 123	20.5	0.09299	0.0	25.65	0	0.5810	5.961	92.9	2.0869	2	188
## 124	17.3	0.15038	0.0	25.65	0	0.5810	5.856	97.0	1.9444	2	188
## 125	18.8	0.09849	0.0	25.65	0	0.5810	5.879	95.8	2.0063	2	188
## 126	21.4	0.16902	0.0	25.65	0	0.5810	5.986	88.4	1.9929	2	188
## 128	16.2	0.25915	0.0	21.89	0	0.6240	5.693	96.0	1.7883	4	437
## 131	19.2	0.34006	0.0	21.89	0	0.6240	6.458	98.9	2.1185	4	437
## 132	19.6	1.19294	0.0	21.89	0	0.6240	6.326	97.7	2.2710	4	437
## 133	23.0	0.59005	0.0	21.89	0	0.6240	6.372	97.9	2.3274	4	437

##	134	18.4	0.32982	0.0	21.89	0	0.6240	5.822	95.4	2.4699	4	437
##	136	18.1	0.55778	0.0	21.89	0	0.6240	6.335	98.2	2.1107	4	437
##	137	17.4	0.32264	0.0	21.89	0	0.6240	5.942	93.5	1.9669	4	437
##	140	17.8	0.54452	0.0	21.89	0	0.6240	6.151	97.9	1.6687	4	437
##	142	14.4	1.62864	0.0	21.89	0	0.6240	5.019	100.0	1.4394	4	437
##	143	13.4	3.32105	0.0	19.58	1	0.8710	5.403	100.0	1.3216	5	403
##	144	15.6	4.09740	0.0	19.58	0	0.8710	5.468	100.0	1.4118	5	403
##	145	11.8	2.77974	0.0	19.58	0	0.8710	4.903	97.8	1.3459	5	403
##	146	13.8	2.37934	0.0	19.58	0	0.8710	6.130	100.0	1.4191	5	403
##	147	15.6	2.15505	0.0	19.58	0	0.8710	5.628	100.0	1.5166	5	403
##	148	14.6	2.36862	0.0	19.58	0	0.8710	4.926	95.7	1.4608	5	403
##	149	17.8	2.33099	0.0	19.58	0	0.8710	5.186	93.8	1.5296	5	403
##	150	15.4	2.73397	0.0	19.58	0	0.8710	5.597	94.9	1.5257	5	403
##	152	19.6	1.49632	0.0	19.58	0	0.8710	5.404	100.0	1.5916	5	403
##	153	15.3	1.12658	0.0	19.58	1	0.8710	5.012	88.0	1.6102	5	403
##	156	15.6	3.53501	0.0	19.58	1	0.8710	6.152	82.6	1.7455	5	403
##	157	13.1	2.44668	0.0	19.58	0	0.8710	5.272	94.0	1.7364	5	403
##	158	41.3	1.22358	0.0	19.58	0	0.6050	6.943	97.4	1.8773	5	403
##	159	24.3	1.34284	0.0	19.58	0	0.6050	6.066	100.0	1.7573	5	403
##	162	50.0	1.46336	0.0	19.58	0	0.6050	7.489	90.8	1.9709	5	403
##	163	50.0	1.83377	0.0	19.58	1	0.6050	7.802	98.2	2.0407	5	403
##	165	22.7	2.24236	0.0	19.58	0	0.6050	5.854	91.8	2.4220	5	403
##	166	25.0	2.92400	0.0	19.58	0	0.6050	6.101	93.0	2.2834	5	403
##	167	50.0	2.01019	0.0	19.58	0	0.6050	7.929	96.2	2.0459	5	403
##	168	23.8	1.80028	0.0	19.58	0	0.6050	5.877	79.2	2.4259	5	403
##	169	23.8	2.30040	0.0	19.58	0	0.6050	6.319	96.1	2.1000	5	403
##	170	22.3	2.44953	0.0	19.58	0	0.6050	6.402	95.2	2.2625	5	403
##	171	17.4	1.20742	0.0	19.58	0	0.6050	5.875	94.6	2.4259	5	403
##	172	19.1	2.31390	0.0	19.58	0	0.6050	5.880	97.3	2.3887	5	403
##	173	23.1	0.13914	0.0	4.05	0	0.5100	5.572	88.5	2.5961	5	296
##	174	23.6	0.09178	0.0	4.05	0	0.5100	6.416	84.1	2.6463	5	296
##	175	22.6	0.08447	0.0	4.05	0	0.5100	5.859	68.7	2.7019	5	296
##	176	29.4	0.06664	0.0	4.05	0	0.5100	6.546	33.1	3.1323	5	296
##	178	24.6	0.05425	0.0	4.05	0	0.5100	6.315	73.4	3.3175	5	296
##	179	29.9	0.06642	0.0	4.05	0	0.5100	6.860	74.4	2.9153	5	296
##	180	37.2	0.05780	0.0	2.46	0	0.4880	6.980	58.4	2.8290	3	193
##	181	39.8	0.06588	0.0	2.46	0	0.4880	7.765	83.3	2.7410	3	193
##	182	36.2	0.06888	0.0	2.46	0	0.4880	6.144	62.2	2.5979	3	193
##	184	32.5	0.10008	0.0	2.46	0	0.4880	6.563	95.6	2.8470	3	193
##	186	29.6	0.06047	0.0	2.46	0	0.4880	6.153	68.8	3.2797	3	193
##	187	50.0	0.05602	0.0	2.46	0	0.4880	7.831	53.6	3.1992	3	193
##	188	32.0	0.07875	45.0	3.44	0	0.4370	6.782	41.1	3.7886	5	398
##	189	29.8	0.12579	45.0	3.44	0	0.4370	6.556	29.1	4.5667	5	398
##	190	34.9	0.08370	45.0	3.44	0	0.4370	7.185	38.9	4.5667	5	398
##	191	37.0	0.09068	45.0	3.44	0	0.4370	6.951	21.5	6.4798	5	398
##	192	30.5	0.06911	45.0	3.44	0	0.4370	6.739	30.8	6.4798	5	398
##	193	36.4	0.08664	45.0	3.44	0	0.4370	7.178	26.3	6.4798	5	398
##	195	29.1	0.01439	60.0	2.93	0	0.4010	6.604	18.8	6.2196	1	265
##	196	50.0	0.01381	80.0	0.46	0	0.4220	7.875	32.0	5.6484	4	255
##	197	33.3	0.04011	80.0	1.52	0	0.4040	7.287	34.1	7.3090	2	329
##	198	30.3	0.04666	80.0	1.52	0	0.4040	7.107	36.6	7.3090	2	329
##	199	34.6	0.03768	80.0	1.52	0	0.4040	7.274	38.3	7.3090	2	329
##	200	34.9	0.03150	95.0	1.47	0	0.4030	6.975	15.3	7.6534	3	402
##	202	24.1	0.03445	82.5	2.03	0	0.4150	6.162	38.4	6.2700	2	348

##	203	42.3	0.02177	82.5	2.03	0	0.4150	7.610	15.7	6.2700	2	348
##	204	48.5	0.03510	95.0	2.68	0	0.4161	7.853	33.2	5.1180	4	224
##	205	50.0	0.02009	95.0	2.68	0	0.4161	8.034	31.9	5.1180	4	224
##	206	22.6	0.13642	0.0	10.59	0	0.4890	5.891	22.3	3.9454	4	277
##	208	22.5	0.25199	0.0	10.59	0	0.4890	5.783	72.7	4.3549	4	277
##	209	24.4	0.13587	0.0	10.59	1	0.4890	6.064	59.1	4.2392	4	277
##	212	19.3	0.37578	0.0	10.59	1	0.4890	5.404	88.6	3.6650	4	277
##	213	22.4	0.21719	0.0	10.59	1	0.4890	5.807	53.8	3.6526	4	277
##	214	28.1	0.14052	0.0	10.59	0	0.4890	6.375	32.3	3.9454	4	277
##	216	25.0	0.19802	0.0	10.59	0	0.4890	6.182	42.4	3.9454	4	277
##	218	28.7	0.07013	0.0	13.89	0	0.5500	6.642	85.1	3.4211	5	276
##	219	21.5	0.11069	0.0	13.89	1	0.5500	5.951	93.8	2.8893	5	276
##	220	23.0	0.11425	0.0	13.89	1	0.5500	6.373	92.4	3.3633	5	276
##	221	26.7	0.35809	0.0	6.20	1	0.5070	6.951	88.5	2.8617	8	307
##	223	27.5	0.62356	0.0	6.20	1	0.5070	6.879	77.7	3.2721	8	307
##	225	44.8	0.31533	0.0	6.20	0	0.5040	8.266	78.3	2.8944	8	307
##	226	50.0	0.52693	0.0	6.20	0	0.5040	8.725	83.0	2.8944	8	307
##	227	37.6	0.38214	0.0	6.20	0	0.5040	8.040	86.5	3.2157	8	307
##	228	31.6	0.41238	0.0	6.20	0	0.5040	7.163	79.9	3.2157	8	307
##	230	31.5	0.44178	0.0	6.20	0	0.5040	6.552	21.4	3.3751	8	307
##	231	24.3	0.53700	0.0	6.20	0	0.5040	5.981	68.1	3.6715	8	307
##	234	48.3	0.33147	0.0	6.20	0	0.5070	8.247	70.4	3.6519	8	307
##	235	29.0	0.44791	0.0	6.20	1	0.5070	6.726	66.5	3.6519	8	307
##	236	24.0	0.33045	0.0	6.20	0	0.5070	6.086	61.5	3.6519	8	307
##	237	25.1	0.52058	0.0	6.20	1	0.5070	6.631	76.5	4.1480	8	307
##	239	23.7	0.08244	30.0	4.93	0	0.4280	6.481	18.5	6.1899	6	300
##	240	23.3	0.09252	30.0	4.93	0	0.4280	6.606	42.2	6.1899	6	300
##	242	20.1	0.10612	30.0	4.93	0	0.4280	6.095	65.1	6.3361	6	300
##	243	22.2	0.10290	30.0	4.93	0	0.4280	6.358	52.9	7.0355	6	300
##	244	23.7	0.12757	30.0	4.93	0	0.4280	6.393	7.8	7.0355	6	300
##	246	18.5	0.19133	22.0	5.86	0	0.4310	5.605	70.2	7.9549	7	330
##	247	24.3	0.33983	22.0	5.86	0	0.4310	6.108	34.9	8.0555	7	330
##	248	20.5	0.19657	22.0	5.86	0	0.4310	6.226	79.2	8.0555	7	330
##	249	24.5	0.16439	22.0	5.86	0	0.4310	6.433	49.1	7.8265	7	330
##	250	26.2	0.19073	22.0	5.86	0	0.4310	6.718	17.5	7.8265	7	330
##	251	24.4	0.14030	22.0	5.86	0	0.4310	6.487	13.0	7.3967	7	330
##	252	24.8	0.21409	22.0	5.86	0	0.4310	6.438	8.9	7.3967	7	330
##	254	42.8	0.36894	22.0	5.86	0	0.4310	8.259	8.4	8.9067	7	330
##	255	21.9	0.04819	80.0	3.64	0	0.3920	6.108	32.0	9.2203	1	315
##	256	20.9	0.03548	80.0	3.64	0	0.3920	5.876	19.1	9.2203	1	315
##	258	50.0	0.61154	20.0	3.97	0	0.6470	8.704	86.9	1.8010	5	264
##	260	30.1	0.65665	20.0	3.97	0	0.6470	6.842	100.0	2.0107	5	264
##	262	43.1	0.53412	20.0	3.97	0	0.6470	7.520	89.4	2.1398	5	264
##	264	31.0	0.82526	20.0	3.97	0	0.6470	7.327	94.5	2.0788	5	264
##	265	36.5	0.55007	20.0	3.97	0	0.6470	7.206	91.6	1.9301	5	264
##	266	22.8	0.76162	20.0	3.97	0	0.6470	5.560	62.8	1.9865	5	264
##	267	30.7	0.78570	20.0	3.97	0	0.6470	7.014	84.6	2.1329	5	264
##	268	50.0	0.57834	20.0	3.97	0	0.5750	8.297	67.0	2.4216	5	264
##	269	43.5	0.54050	20.0	3.97	0	0.5750	7.470	52.6	2.8720	5	264
##	270	20.7	0.09065	20.0	6.96	1	0.4640	5.920	61.5	3.9175	3	223
##	271	21.1	0.29916	20.0	6.96	0	0.4640	5.856	42.1	4.4290	3	223
##	273	24.4	0.11460	20.0	6.96	0	0.4640	6.538	58.7	3.9175	3	223
##	274	35.2	0.22188	20.0	6.96	1	0.4640	7.691	51.8	4.3665	3	223
##	275	32.4	0.05644	40.0	6.41	1	0.4470	6.758	32.9	4.0776	4	254

##	276	32.0	0.09604	40.0	6.41	0	0.4470	6.854	42.8	4.2673	4	254
##	277	33.2	0.10469	40.0	6.41	1	0.4470	7.267	49.0	4.7872	4	254
##	279	29.1	0.07978	40.0	6.41	0	0.4470	6.482	32.1	4.1403	4	254
##	280	35.1	0.21038	20.0	3.33	0	0.4429	6.812	32.2	4.1007	5	216
##	281	45.4	0.03578	20.0	3.33	0	0.4429	7.820	64.5	4.6947	5	216
##	282	35.4	0.03705	20.0	3.33	0	0.4429	6.968	37.2	5.2447	5	216
##	283	46.0	0.06129	20.0	3.33	1	0.4429	7.645	49.7	5.2119	5	216
##	285	32.2	0.00906	90.0	2.97	0	0.4000	7.088	20.8	7.3073	1	285
##	286	22.0	0.01096	55.0	2.25	0	0.3890	6.453	31.9	7.3073	1	300
##	287	20.1	0.01965	80.0	1.76	0	0.3850	6.230	31.5	9.0892	1	241
##	288	23.2	0.03871	52.5	5.32	0	0.4050	6.209	31.3	7.3172	6	293
##	289	22.3	0.04590	52.5	5.32	0	0.4050	6.315	45.6	7.3172	6	293
##	290	24.8	0.04297	52.5	5.32	0	0.4050	6.565	22.9	7.3172	6	293
##	291	28.5	0.03502	80.0	4.95	0	0.4110	6.861	27.9	5.1167	4	245
##	292	37.3	0.07886	80.0	4.95	0	0.4110	7.148	27.7	5.1167	4	245
##	294	23.9	0.08265	0.0	13.92	0	0.4370	6.127	18.4	5.5027	4	289
##	296	28.6	0.12932	0.0	13.92	0	0.4370	6.678	31.1	5.9604	4	289
##	297	27.1	0.05372	0.0	13.92	0	0.4370	6.549	51.0	5.9604	4	289
##	299	22.5	0.06466	70.0	2.24	0	0.4000	6.345	20.1	7.8278	5	358
##	301	24.8	0.04417	70.0	2.24	0	0.4000	6.871	47.4	7.8278	5	358
##	303	26.4	0.09266	34.0	6.09	0	0.4330	6.495	18.4	5.4917	7	329
##	304	33.1	0.10000	34.0	6.09	0	0.4330	6.982	17.7	5.4917	7	329
##	305	36.1	0.05515	33.0	2.18	0	0.4720	7.236	41.1	4.0220	7	222
##	306	28.4	0.05479	33.0	2.18	0	0.4720	6.616	58.1	3.3700	7	222
##	307	33.4	0.07503	33.0	2.18	0	0.4720	7.420	71.9	3.0992	7	222
##	308	28.2	0.04932	33.0	2.18	0	0.4720	6.849	70.3	3.1827	7	222
##	309	22.8	0.49298	0.0	9.90	0	0.5440	6.635	82.5	3.3175	4	304
##	311	16.1	2.63548	0.0	9.90	0	0.5440	4.973	37.8	2.5194	4	304
##	313	19.4	0.26169	0.0	9.90	0	0.5440	6.023	90.4	2.8340	4	304
##	314	21.6	0.26938	0.0	9.90	0	0.5440	6.266	82.8	3.2628	4	304
##	316	16.2	0.25356	0.0	9.90	0	0.5440	5.705	77.7	3.9450	4	304
##	317	17.8	0.31827	0.0	9.90	0	0.5440	5.914	83.2	3.9986	4	304
##	318	19.8	0.24522	0.0	9.90	0	0.5440	5.782	71.7	4.0317	4	304
##	319	23.1	0.40202	0.0	9.90	0	0.5440	6.382	67.2	3.5325	4	304
##	320	21.0	0.47547	0.0	9.90	0	0.5440	6.113	58.8	4.0019	4	304
##	321	23.8	0.16760	0.0	7.38	0	0.4930	6.426	52.3	4.5404	5	287
##	322	23.1	0.18159	0.0	7.38	0	0.4930	6.376	54.3	4.5404	5	287
##	323	20.4	0.35114	0.0	7.38	0	0.4930	6.041	49.9	4.7211	5	287
##	324	18.5	0.28392	0.0	7.38	0	0.4930	5.708	74.3	4.7211	5	287
##	327	23.0	0.30347	0.0	7.38	0	0.4930	6.312	28.9	5.4159	5	287
##	328	22.2	0.24103	0.0	7.38	0	0.4930	6.083	43.7	5.4159	5	287
##	330	22.6	0.06724	0.0	3.24	0	0.4600	6.333	17.2	5.2146	4	430
##	331	19.8	0.04544	0.0	3.24	0	0.4600	6.144	32.2	5.8736	4	430
##	332	17.1	0.05023	35.0	6.06	0	0.4379	5.706	28.4	6.6407	1	304
##	333	19.4	0.03466	35.0	6.06	0	0.4379	6.031	23.3	6.6407	1	304
##	334	22.2	0.05083	0.0	5.19	0	0.5150	6.316	38.1	6.4584	5	224
##	335	20.7	0.03738	0.0	5.19	0	0.5150	6.310	38.5	6.4584	5	224
##	336	21.1	0.03961	0.0	5.19	0	0.5150	6.037	34.5	5.9853	5	224
##	337	19.5	0.03427	0.0	5.19	0	0.5150	5.869	46.3	5.2311	5	224
##	338	18.5	0.03041	0.0	5.19	0	0.5150	5.895	59.6	5.6150	5	224
##	340	19.0	0.05497	0.0	5.19	0	0.5150	5.985	45.4	4.8122	5	224
##	341	18.7	0.06151	0.0	5.19	0	0.5150	5.968	58.5	4.8122	5	224
##	343	16.5	0.02498	0.0	1.89	0	0.5180	6.540	59.7	6.2669	1	422
##	345	31.2	0.03049	55.0	3.78	0	0.4840	6.874	28.1	6.4654	5	370

##	346	17.5	0.03113	0.0	4.39	0	0.4420	6.014	48.5	8.0136	3	352
##	347	17.2	0.06162	0.0	4.39	0	0.4420	5.898	52.3	8.0136	3	352
##	349	24.5	0.01501	80.0	2.01	0	0.4350	6.635	29.7	8.3440	4	280
##	350	26.6	0.02899	40.0	1.25	0	0.4290	6.939	34.5	8.7921	1	335
##	351	22.9	0.06211	40.0	1.25	0	0.4290	6.490	44.4	8.7921	1	335
##	352	24.1	0.07950	60.0	1.69	0	0.4110	6.579	35.9	10.7103	4	411
##	353	18.6	0.07244	60.0	1.69	0	0.4110	5.884	18.5	10.7103	4	411
##	355	18.2	0.04301	80.0	1.91	0	0.4130	5.663	21.9	10.5857	4	334
##	356	20.6	0.10659	80.0	1.91	0	0.4130	5.936	19.5	10.5857	4	334
##	358	21.7	3.84970	0.0	18.10	1	0.7700	6.395	91.0	2.5052	24	666
##	359	22.7	5.20177	0.0	18.10	1	0.7700	6.127	83.4	2.7227	24	666
##	360	22.6	4.26131	0.0	18.10	0	0.7700	6.112	81.3	2.5091	24	666
##	361	25.0	4.54192	0.0	18.10	0	0.7700	6.398	88.0	2.5182	24	666
##	363	20.8	3.67822	0.0	18.10	0	0.7700	5.362	96.2	2.1036	24	666
##	365	21.9	3.47428	0.0	18.10	1	0.7180	8.780	82.9	1.9047	24	666
##	366	27.5	4.55587	0.0	18.10	0	0.7180	3.561	87.9	1.6132	24	666
##	367	21.9	3.69695	0.0	18.10	0	0.7180	4.963	91.4	1.7523	24	666
##	369	50.0	4.89822	0.0	18.10	0	0.6310	4.970	100.0	1.3325	24	666
##	370	50.0	5.66998	0.0	18.10	1	0.6310	6.683	96.8	1.3567	24	666
##	373	50.0	8.26725	0.0	18.10	1	0.6680	5.875	89.6	1.1296	24	666
##	374	13.8	11.10810	0.0	18.10	0	0.6680	4.906	100.0	1.1742	24	666
##	375	13.8	18.49820	0.0	18.10	0	0.6680	4.138	100.0	1.1370	24	666
##	376	15.0	19.60910	0.0	18.10	0	0.6710	7.313	97.9	1.3163	24	666
##	379	13.1	23.64820	0.0	18.10	0	0.6710	6.380	96.2	1.3861	24	666
##	380	10.2	17.86670	0.0	18.10	0	0.6710	6.223	100.0	1.3861	24	666
##	382	10.9	15.87440	0.0	18.10	0	0.6710	6.545	99.1	1.5192	24	666
##	383	11.3	9.18702	0.0	18.10	0	0.7000	5.536	100.0	1.5804	24	666
##	384	12.3	7.99248	0.0	18.10	0	0.7000	5.520	100.0	1.5331	24	666
##	385	8.8	20.08490	0.0	18.10	0	0.7000	4.368	91.2	1.4395	24	666
##	386	7.2	16.81180	0.0	18.10	0	0.7000	5.277	98.1	1.4261	24	666
##	387	10.5	24.39380	0.0	18.10	0	0.7000	4.652	100.0	1.4672	24	666
##	388	7.4	22.59710	0.0	18.10	0	0.7000	5.000	89.5	1.5184	24	666
##	389	10.2	14.33370	0.0	18.10	0	0.7000	4.880	100.0	1.5895	24	666
##	390	11.5	8.15174	0.0	18.10	0	0.7000	5.390	98.9	1.7281	24	666
##	392	23.2	5.29305	0.0	18.10	0	0.7000	6.051	82.5	2.1678	24	666
##	394	13.8	8.64476	0.0	18.10	0	0.6930	6.193	92.6	1.7912	24	666
##	396	13.1	8.71675	0.0	18.10	0	0.6930	6.471	98.8	1.7257	24	666
##	397	12.5	5.87205	0.0	18.10	0	0.6930	6.405	96.0	1.6768	24	666
##	398	8.5	7.67202	0.0	18.10	0	0.6930	5.747	98.9	1.6334	24	666
##	399	5.0	38.35180	0.0	18.10	0	0.6930	5.453	100.0	1.4896	24	666
##	400	6.3	9.91655	0.0	18.10	0	0.6930	5.852	77.8	1.5004	24	666
##	401	5.6	25.04610	0.0	18.10	0	0.6930	5.987	100.0	1.5888	24	666
##	402	7.2	14.23620	0.0	18.10	0	0.6930	6.343	100.0	1.5741	24	666
##	404	8.3	24.80170	0.0	18.10	0	0.6930	5.349	96.0	1.7028	24	666
##	405	8.5	41.52920	0.0	18.10	0	0.6930	5.531	85.4	1.6074	24	666
##	407	11.9	20.71620	0.0	18.10	0	0.6590	4.138	100.0	1.1781	24	666
##	408	27.9	11.95110	0.0	18.10	0	0.6590	5.608	100.0	1.2852	24	666
##	409	17.2	7.40389	0.0	18.10	0	0.5970	5.617	97.9	1.4547	24	666
##	410	27.5	14.43830	0.0	18.10	0	0.5970	6.852	100.0	1.4655	24	666
##	411	15.0	51.13580	0.0	18.10	0	0.5970	5.757	100.0	1.4130	24	666
##	412	17.2	14.05070	0.0	18.10	0	0.5970	6.657	100.0	1.5275	24	666
##	417	7.5	10.83420	0.0	18.10	0	0.6790	6.782	90.8	1.8195	24	666
##	418	10.4	25.94060	0.0	18.10	0	0.6790	5.304	89.1	1.6475	24	666
##	419	8.8	73.53410	0.0	18.10	0	0.6790	5.957	100.0	1.8026	24	666

## 421	16.7	11.08740	0.0	18.10	0	0.7180	6.411	100.0	1.8589	24	666
## 422	14.2	7.02259	0.0	18.10	0	0.7180	6.006	95.3	1.8746	24	666
## 423	20.8	12.04820	0.0	18.10	0	0.6140	5.648	87.6	1.9512	24	666
## 425	11.7	8.79212	0.0	18.10	0	0.5840	5.565	70.6	2.0635	24	666
## 426	8.3	15.86030	0.0	18.10	0	0.6790	5.896	95.4	1.9096	24	666
## 427	10.2	12.24720	0.0	18.10	0	0.5840	5.837	59.7	1.9976	24	666
## 428	10.9	37.66190	0.0	18.10	0	0.6790	6.202	78.7	1.8629	24	666
## 429	11.0	7.36711	0.0	18.10	0	0.6790	6.193	78.1	1.9356	24	666
## 430	9.5	9.33889	0.0	18.10	0	0.6790	6.380	95.6	1.9682	24	666
## 431	14.5	8.49213	0.0	18.10	0	0.5840	6.348	86.1	2.0527	24	666
## 432	14.1	10.06230	0.0	18.10	0	0.5840	6.833	94.3	2.0882	24	666
## 433	16.1	6.44405	0.0	18.10	0	0.5840	6.425	74.8	2.2004	24	666
## 434	14.3	5.58107	0.0	18.10	0	0.7130	6.436	87.9	2.3158	24	666
## 435	11.7	13.91340	0.0	18.10	0	0.7130	6.208	95.0	2.2222	24	666
## 436	13.4	11.16040	0.0	18.10	0	0.7400	6.629	94.6	2.1247	24	666
## 437	9.6	14.42080	0.0	18.10	0	0.7400	6.461	93.3	2.0026	24	666
## 438	8.7	15.17720	0.0	18.10	0	0.7400	6.152	100.0	1.9142	24	666
## 439	8.4	13.67810	0.0	18.10	0	0.7400	5.935	87.9	1.8206	24	666
## 440	12.8	9.39063	0.0	18.10	0	0.7400	5.627	93.9	1.8172	24	666
## 441	10.5	22.05110	0.0	18.10	0	0.7400	5.818	92.4	1.8662	24	666
## 442	17.1	9.72418	0.0	18.10	0	0.7400	6.406	97.2	2.0651	24	666
## 443	18.4	5.66637	0.0	18.10	0	0.7400	6.219	100.0	2.0048	24	666
## 444	15.4	9.96654	0.0	18.10	0	0.7400	6.485	100.0	1.9784	24	666
## 446	11.8	10.67180	0.0	18.10	0	0.7400	6.459	94.8	1.9879	24	666
## 447	14.9	6.28807	0.0	18.10	0	0.7400	6.341	96.4	2.0720	24	666
## 449	14.1	9.32909	0.0	18.10	0	0.7130	6.185	98.7	2.2616	24	666
## 450	13.0	7.52601	0.0	18.10	0	0.7130	6.417	98.3	2.1850	24	666
## 451	13.4	6.71772	0.0	18.10	0	0.7130	6.749	92.6	2.3236	24	666
## 452	15.2	5.44114	0.0	18.10	0	0.7130	6.655	98.2	2.3552	24	666
## 454	17.8	8.24809	0.0	18.10	0	0.7130	7.393	99.3	2.4527	24	666
## 455	14.9	9.51363	0.0	18.10	0	0.7130	6.728	94.1	2.4961	24	666
## 457	12.7	4.66883	0.0	18.10	0	0.7130	5.976	87.9	2.5806	24	666
## 458	13.5	8.20058	0.0	18.10	0	0.7130	5.936	80.3	2.7792	24	666
## 459	14.9	7.75223	0.0	18.10	0	0.7130	6.301	83.7	2.7831	24	666
## 461	16.4	4.81213	0.0	18.10	0	0.7130	6.701	90.0	2.5975	24	666
## 462	17.7	3.69311	0.0	18.10	0	0.7130	6.376	88.4	2.5671	24	666
## 463	19.5	6.65492	0.0	18.10	0	0.7130	6.317	83.0	2.7344	24	666
## 466	19.9	3.16360	0.0	18.10	0	0.6550	5.759	48.2	3.0665	24	666
## 467	19.0	3.77498	0.0	18.10	0	0.6550	5.952	84.7	2.8715	24	666
## 468	19.1	4.42228	0.0	18.10	0	0.5840	6.003	94.5	2.5403	24	666
## 469	19.1	15.57570	0.0	18.10	0	0.5800	5.926	71.0	2.9084	24	666
## 470	20.1	13.07510	0.0	18.10	0	0.5800	5.713	56.7	2.8237	24	666
## 471	19.9	4.34879	0.0	18.10	0	0.5800	6.167	84.0	3.0334	24	666
## 472	19.6	4.03841	0.0	18.10	0	0.5320	6.229	90.7	3.0993	24	666
## 473	23.2	3.56868	0.0	18.10	0	0.5800	6.437	75.0	2.8965	24	666
## 474	29.8	4.64689	0.0	18.10	0	0.6140	6.980	67.6	2.5329	24	666
## 475	13.8	8.05579	0.0	18.10	0	0.5840	5.427	95.4	2.4298	24	666
## 476	13.3	6.39312	0.0	18.10	0	0.5840	6.162	97.4	2.2060	24	666
## 477	16.7	4.87141	0.0	18.10	0	0.6140	6.484	93.6	2.3053	24	666
## 478	12.0	15.02340	0.0	18.10	0	0.6140	5.304	97.3	2.1007	24	666
## 479	14.6	10.23300	0.0	18.10	0	0.6140	6.185	96.7	2.1705	24	666
## 480	21.4	14.33370	0.0	18.10	0	0.6140	6.229	88.0	1.9512	24	666
## 483	25.0	5.73116	0.0	18.10	0	0.5320	7.061	77.0	3.4106	24	666
## 484	21.8	2.81838	0.0	18.10	0	0.5320	5.762	40.3	4.0983	24	666

##	485	20.6	2.37857	0.0	18.10	0	0.5830	5.871	41.9	3.7240	24	666
##	486	21.2	3.67367	0.0	18.10	0	0.5830	6.312	51.9	3.9917	24	666
##	488	20.6	4.83567	0.0	18.10	0	0.5830	5.905	53.2	3.1523	24	666
##	490	7.0	0.18337	0.0	27.74	0	0.6090	5.414	98.3	1.7554	4	711
##	492	13.6	0.10574	0.0	27.74	0	0.6090	5.983	98.8	1.8681	4	711
##	493	20.1	0.11132	0.0	27.74	0	0.6090	5.983	83.5	2.1099	4	711
##	494	21.8	0.17331	0.0	9.69	0	0.5850	5.707	54.0	2.3817	6	391
##	495	24.5	0.27957	0.0	9.69	0	0.5850	5.926	42.6	2.3817	6	391
##	496	23.1	0.17899	0.0	9.69	0	0.5850	5.670	28.8	2.7986	6	391
##	497	19.7	0.28960	0.0	9.69	0	0.5850	5.390	72.9	2.7986	6	391
##	498	18.3	0.26838	0.0	9.69	0	0.5850	5.794	70.6	2.8927	6	391
##	499	21.2	0.23912	0.0	9.69	0	0.5850	6.019	65.3	2.4091	6	391
##	500	17.5	0.17783	0.0	9.69	0	0.5850	5.569	73.5	2.3999	6	391
##	501	16.8	0.22438	0.0	9.69	0	0.5850	6.027	79.7	2.4982	6	391
##	502	22.4	0.06263	0.0	11.93	0	0.5730	6.593	69.1	2.4786	1	273
##	503	20.6	0.04527	0.0	11.93	0	0.5730	6.120	76.7	2.2875	1	273
##	505	22.0	0.10959	0.0	11.93	0	0.5730	6.794	89.3	2.3889	1	273
##	506	11.9	0.04741	0.0	11.93	0	0.5730	6.030	80.8	2.5050	1	273
##		PTRATIO		B	LSTAT							
##	1	15.3	396.90	4.98								
##	2	17.8	396.90	9.14								
##	3	17.8	392.83	4.03								
##	4	18.7	394.63	2.94								
##	5	18.7	396.90	5.33								
##	6	18.7	394.12	5.21								
##	8	15.2	396.90	19.15								
##	9	15.2	386.63	29.93								
##	10	15.2	386.71	17.10								
##	12	15.2	396.90	13.27								
##	16	21.0	395.62	8.47								
##	17	21.0	386.85	6.58								
##	19	21.0	288.99	11.69								
##	21	21.0	376.57	21.02								
##	23	21.0	396.90	18.72								
##	27	21.0	376.88	14.81								
##	28	21.0	306.38	17.28								
##	29	21.0	387.94	12.80								
##	30	21.0	380.23	11.98								
##	31	21.0	360.17	22.60								
##	32	21.0	376.73	13.04								
##	33	21.0	232.60	27.71								
##	34	21.0	358.77	18.35								
##	35	21.0	248.31	20.34								
##	36	19.2	396.90	9.68								
##	37	19.2	377.56	11.41								
##	38	19.2	396.90	8.77								
##	39	19.2	393.43	10.13								
##	41	18.3	395.62	1.98								
##	42	17.9	385.41	4.84								
##	44	17.9	394.46	7.44								
##	48	17.9	392.74	18.80								
##	51	16.8	395.56	13.45								
##	53	16.8	396.90	5.28								
##	56	17.9	395.93	4.81								

## 57	17.3	396.90	5.77
## 58	15.1	392.90	3.95
## 59	19.7	390.68	6.86
## 60	19.7	396.90	9.22
## 61	19.7	395.11	13.15
## 62	19.7	378.08	14.44
## 63	19.7	396.90	6.73
## 64	19.7	395.58	9.50
## 66	16.1	396.90	4.67
## 67	16.1	396.90	10.24
## 68	18.9	396.21	8.10
## 69	18.9	396.90	13.09
## 72	19.2	376.94	9.88
## 73	19.2	390.91	5.52
## 74	19.2	377.17	7.54
## 75	18.7	394.92	6.78
## 77	18.7	373.66	11.97
## 79	18.7	386.40	12.34
## 80	18.7	396.06	9.10
## 81	19.0	396.90	5.29
## 82	19.0	395.63	7.22
## 83	19.0	396.90	6.72
## 84	19.0	390.64	7.51
## 86	18.5	392.30	6.53
## 88	18.5	395.15	8.44
## 89	17.8	396.90	5.50
## 91	17.8	392.18	8.81
## 92	17.8	393.55	8.20
## 94	18.2	396.33	6.21
## 95	18.2	396.90	10.59
## 98	18.0	396.90	4.21
## 99	18.0	393.53	3.57
## 100	18.0	396.90	6.19
## 102	20.9	395.58	7.67
## 104	20.9	394.47	13.44
## 107	20.9	395.67	18.66
## 108	20.9	387.69	14.09
## 109	20.9	395.24	12.27
## 110	20.9	391.23	15.55
## 111	20.9	393.49	13.00
## 112	17.8	395.59	10.16
## 113	17.8	394.95	16.21
## 114	17.8	396.90	17.09
## 115	17.8	388.74	10.45
## 116	17.8	344.91	15.76
## 118	17.8	394.51	10.30
## 119	17.8	338.63	15.37
## 120	17.8	391.50	13.61
## 121	19.1	389.15	14.37
## 122	19.1	377.67	14.27
## 123	19.1	378.09	17.93
## 124	19.1	370.31	25.41
## 125	19.1	379.38	17.58
## 126	19.1	385.02	14.81

## 128	21.2	392.11	17.19
## 131	21.2	395.04	12.60
## 132	21.2	396.90	12.26
## 133	21.2	385.76	11.12
## 134	21.2	388.69	15.03
## 136	21.2	394.67	16.96
## 137	21.2	378.25	16.90
## 140	21.2	396.90	18.46
## 142	21.2	396.90	34.41
## 143	14.7	396.90	26.82
## 144	14.7	396.90	26.42
## 145	14.7	396.90	29.29
## 146	14.7	172.91	27.80
## 147	14.7	169.27	16.65
## 148	14.7	391.71	29.53
## 149	14.7	356.99	28.32
## 150	14.7	351.85	21.45
## 152	14.7	341.60	13.28
## 153	14.7	343.28	12.12
## 156	14.7	88.01	15.02
## 157	14.7	88.63	16.14
## 158	14.7	363.43	4.59
## 159	14.7	353.89	6.43
## 162	14.7	374.43	1.73
## 163	14.7	389.61	1.92
## 165	14.7	395.11	11.64
## 166	14.7	240.16	9.81
## 167	14.7	369.30	3.70
## 168	14.7	227.61	12.14
## 169	14.7	297.09	11.10
## 170	14.7	330.04	11.32
## 171	14.7	292.29	14.43
## 172	14.7	348.13	12.03
## 173	16.6	396.90	14.69
## 174	16.6	395.50	9.04
## 175	16.6	393.23	9.64
## 176	16.6	390.96	5.33
## 178	16.6	395.60	6.29
## 179	16.6	391.27	6.92
## 180	17.8	396.90	5.04
## 181	17.8	395.56	7.56
## 182	17.8	396.90	9.45
## 184	17.8	396.90	5.68
## 186	17.8	387.11	13.15
## 187	17.8	392.63	4.45
## 188	15.2	393.87	6.68
## 189	15.2	382.84	4.56
## 190	15.2	396.90	5.39
## 191	15.2	377.68	5.10
## 192	15.2	389.71	4.69
## 193	15.2	390.49	2.87
## 195	15.6	376.70	4.38
## 196	14.4	394.23	2.97
## 197	12.6	396.90	4.08

## 198	12.6	354.31	8.61
## 199	12.6	392.20	6.62
## 200	17.0	396.90	4.56
## 202	14.7	393.77	7.43
## 203	14.7	395.38	3.11
## 204	14.7	392.78	3.81
## 205	14.7	390.55	2.88
## 206	18.6	396.90	10.87
## 208	18.6	389.43	18.06
## 209	18.6	381.32	14.66
## 212	18.6	395.24	23.98
## 213	18.6	390.94	16.03
## 214	18.6	385.81	9.38
## 216	18.6	393.63	9.47
## 218	16.4	392.78	9.69
## 219	16.4	396.90	17.92
## 220	16.4	393.74	10.50
## 221	17.4	391.70	9.71
## 223	17.4	390.39	9.93
## 225	17.4	385.05	4.14
## 226	17.4	382.00	4.63
## 227	17.4	387.38	3.13
## 228	17.4	372.08	6.36
## 230	17.4	380.34	3.76
## 231	17.4	378.35	11.65
## 234	17.4	378.95	3.95
## 235	17.4	360.20	8.05
## 236	17.4	376.75	10.88
## 237	17.4	388.45	9.54
## 239	16.6	379.41	6.36
## 240	16.6	383.78	7.37
## 242	16.6	394.62	12.40
## 243	16.6	372.75	11.22
## 244	16.6	374.71	5.19
## 246	19.1	389.13	18.46
## 247	19.1	390.18	9.16
## 248	19.1	376.14	10.15
## 249	19.1	374.71	9.52
## 250	19.1	393.74	6.56
## 251	19.1	396.28	5.90
## 252	19.1	377.07	3.59
## 254	19.1	396.90	3.54
## 255	16.4	392.89	6.57
## 256	16.4	395.18	9.25
## 258	13.0	389.70	5.12
## 260	13.0	391.93	6.90
## 262	13.0	388.37	7.26
## 264	13.0	393.42	11.25
## 265	13.0	387.89	8.10
## 266	13.0	392.40	10.45
## 267	13.0	384.07	14.79
## 268	13.0	384.54	7.44
## 269	13.0	390.30	3.16
## 270	18.6	391.34	13.65

## 271	18.6	388.65	13.00
## 273	18.6	394.96	7.73
## 274	18.6	390.77	6.58
## 275	17.6	396.90	3.53
## 276	17.6	396.90	2.98
## 277	17.6	389.25	6.05
## 279	17.6	396.90	7.19
## 280	14.9	396.90	4.85
## 281	14.9	387.31	3.76
## 282	14.9	392.23	4.59
## 283	14.9	377.07	3.01
## 285	15.3	394.72	7.85
## 286	15.3	394.72	8.23
## 287	18.2	341.60	12.93
## 288	16.6	396.90	7.14
## 289	16.6	396.90	7.60
## 290	16.6	371.72	9.51
## 291	19.2	396.90	3.33
## 292	19.2	396.90	3.56
## 294	16.0	396.90	8.58
## 296	16.0	396.90	6.27
## 297	16.0	392.85	7.39
## 299	14.8	368.24	4.97
## 301	14.8	390.86	6.07
## 303	16.1	383.61	8.67
## 304	16.1	390.43	4.86
## 305	18.4	393.68	6.93
## 306	18.4	393.36	8.93
## 307	18.4	396.90	6.47
## 308	18.4	396.90	7.53
## 309	18.4	396.90	4.54
## 311	18.4	350.45	12.64
## 313	18.4	396.30	11.72
## 314	18.4	393.39	7.90
## 316	18.4	396.42	11.50
## 317	18.4	390.70	18.33
## 318	18.4	396.90	15.94
## 319	18.4	395.21	10.36
## 320	18.4	396.23	12.73
## 321	19.6	396.90	7.20
## 322	19.6	396.90	6.87
## 323	19.6	396.90	7.70
## 324	19.6	391.13	11.74
## 327	19.6	396.90	6.15
## 328	19.6	396.90	12.79
## 330	16.9	375.21	7.34
## 331	16.9	368.57	9.09
## 332	16.9	394.02	12.43
## 333	16.9	362.25	7.83
## 334	20.2	389.71	5.68
## 335	20.2	389.40	6.75
## 336	20.2	396.90	8.01
## 337	20.2	396.90	9.80
## 338	20.2	394.81	10.56

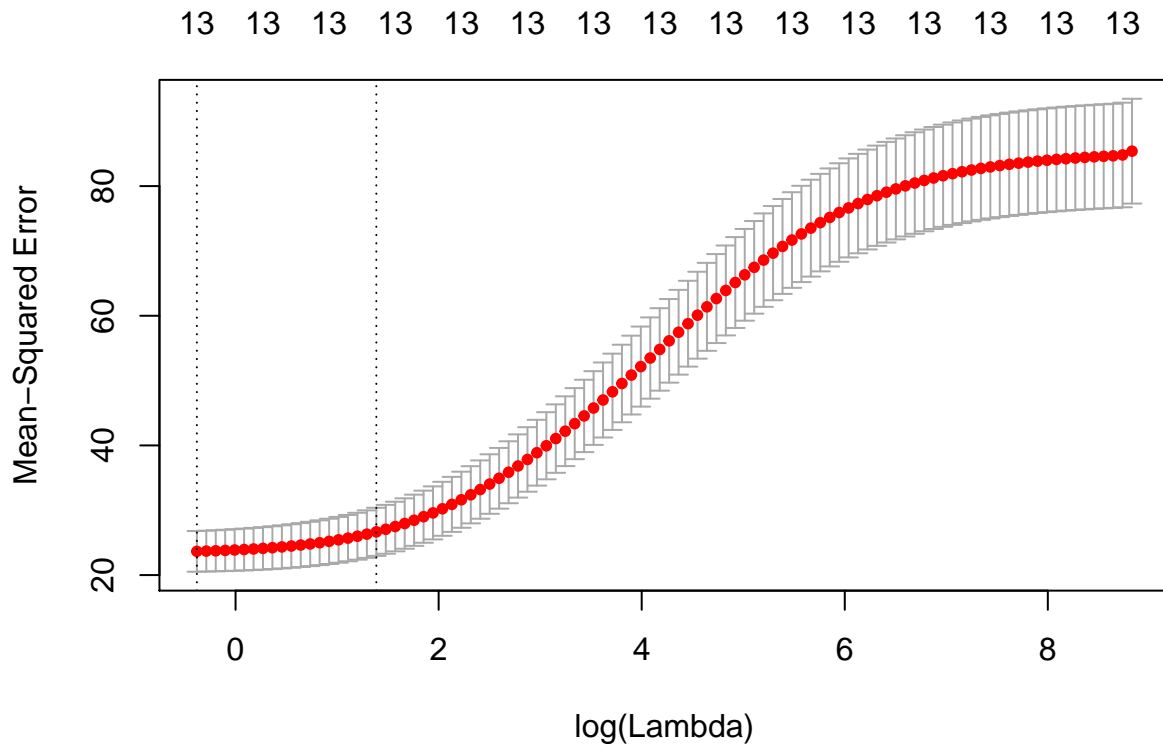
## 340	20.2	396.90	9.74
## 341	20.2	396.90	9.29
## 343	15.9	389.96	8.65
## 345	17.6	387.97	4.61
## 346	18.8	385.64	10.53
## 347	18.8	364.61	12.67
## 349	17.0	390.94	5.99
## 350	19.7	389.85	5.89
## 351	19.7	396.90	5.98
## 352	18.3	370.78	5.49
## 353	18.3	392.33	7.79
## 355	22.0	382.80	8.05
## 356	22.0	376.04	5.57
## 358	20.2	391.34	13.27
## 359	20.2	395.43	11.48
## 360	20.2	390.74	12.67
## 361	20.2	374.56	7.79
## 363	20.2	380.79	10.19
## 365	20.2	354.55	5.29
## 366	20.2	354.70	7.12
## 367	20.2	316.03	14.00
## 369	20.2	375.52	3.26
## 370	20.2	375.33	3.73
## 373	20.2	347.88	8.88
## 374	20.2	396.90	34.77
## 375	20.2	396.90	37.97
## 376	20.2	396.90	13.44
## 379	20.2	396.90	23.69
## 380	20.2	393.74	21.78
## 382	20.2	396.90	21.08
## 383	20.2	396.90	23.60
## 384	20.2	396.90	24.56
## 385	20.2	285.83	30.63
## 386	20.2	396.90	30.81
## 387	20.2	396.90	28.28
## 388	20.2	396.90	31.99
## 389	20.2	372.92	30.62
## 390	20.2	396.90	20.85
## 392	20.2	378.38	18.76
## 394	20.2	396.90	15.17
## 396	20.2	391.98	17.12
## 397	20.2	396.90	19.37
## 398	20.2	393.10	19.92
## 399	20.2	396.90	30.59
## 400	20.2	338.16	29.97
## 401	20.2	396.90	26.77
## 402	20.2	396.90	20.32
## 404	20.2	396.90	19.77
## 405	20.2	329.46	27.38
## 407	20.2	370.22	23.34
## 408	20.2	332.09	12.13
## 409	20.2	314.64	26.40
## 410	20.2	179.36	19.78
## 411	20.2	2.60	10.11

## 412	20.2	35.05	21.22
## 417	20.2	21.57	25.79
## 418	20.2	127.36	26.64
## 419	20.2	16.45	20.62
## 421	20.2	318.75	15.02
## 422	20.2	319.98	15.70
## 423	20.2	291.55	14.10
## 425	20.2	3.65	17.16
## 426	20.2	7.68	24.39
## 427	20.2	24.65	15.69
## 428	20.2	18.82	14.52
## 429	20.2	96.73	21.52
## 430	20.2	60.72	24.08
## 431	20.2	83.45	17.64
## 432	20.2	81.33	19.69
## 433	20.2	97.95	12.03
## 434	20.2	100.19	16.22
## 435	20.2	100.63	15.17
## 436	20.2	109.85	23.27
## 437	20.2	27.49	18.05
## 438	20.2	9.32	26.45
## 439	20.2	68.95	34.02
## 440	20.2	396.90	22.88
## 441	20.2	391.45	22.11
## 442	20.2	385.96	19.52
## 443	20.2	395.69	16.59
## 444	20.2	386.73	18.85
## 446	20.2	43.06	23.98
## 447	20.2	318.01	17.79
## 449	20.2	396.90	18.13
## 450	20.2	304.21	19.31
## 451	20.2	0.32	17.44
## 452	20.2	355.29	17.73
## 454	20.2	375.87	16.74
## 455	20.2	6.68	18.71
## 457	20.2	10.48	19.01
## 458	20.2	3.50	16.94
## 459	20.2	272.21	16.23
## 461	20.2	255.23	16.42
## 462	20.2	391.43	14.65
## 463	20.2	396.90	13.99
## 466	20.2	334.40	14.13
## 467	20.2	22.01	17.15
## 468	20.2	331.29	21.32
## 469	20.2	368.74	18.13
## 470	20.2	396.90	14.76
## 471	20.2	396.90	16.29
## 472	20.2	395.33	12.87
## 473	20.2	393.37	14.36
## 474	20.2	374.68	11.66
## 475	20.2	352.58	18.14
## 476	20.2	302.76	24.10
## 477	20.2	396.21	18.68
## 478	20.2	349.48	24.91

```
## 479      20.2 379.70 18.03
## 480      20.2 383.32 13.11
## 483      20.2 395.28  7.01
## 484      20.2 392.92 10.42
## 485      20.2 370.73 13.34
## 486      20.2 388.62 10.58
## 488      20.2 388.22 11.45
## 490      20.1 344.05 23.97
## 492      20.1 390.11 18.07
## 493      20.1 396.90 13.35
## 494      19.2 396.90 12.01
## 495      19.2 396.90 13.59
## 496      19.2 393.29 17.60
## 497      19.2 396.90 21.14
## 498      19.2 396.90 14.10
## 499      19.2 396.90 12.92
## 500      19.2 395.77 15.10
## 501      19.2 396.90 14.33
## 502      21.0 391.99  9.67
## 503      21.0 396.90  9.08
## 505      21.0 393.45  6.48
## 506      21.0 396.90  7.88
```

```
x <- model.matrix(MEDV~., train)[,-1]
y = train$MEDV
y.test = y[-trainIndex]
```

```
#perform cross-validation to choose tuning parameter lambda
cv.out <- cv.glmnet(x, y, alpha = 0)
plot(cv.out)
```



```
bestlambda <- cv.out$lambda.min #lambda that results in lowest cross validation error
grid <- 10^seq(10,-2,length = 100) #from 10^10 to 10^-2
ridge.mod <- glmnet(x,y,alpha= 0, lambda = grid, thresh = 1e-12)
#make predictions for lambda = bestlambda
ridge.pred <- predict(ridge.mod, s=bestlambda, newx = x[-trainIndex,])
ridge.MSE <- mean((ridge.pred - y.test)^2)
```

```
library(plotmo)
```

```
glmcoef<-coef(ridge.mod,bestlambda )
coef.increase<-dimnames(glmcoef[glmcoef[,1]>0,0])[[1]]
coef.decrease<-dimnames(glmcoef[glmcoef[,1]<0,0])[[1]]

#get ordered list of variables as they appear at smallest lambda
allnames<-names(coef(ridge.mod)[,
                    ncol(coef(ridge.mod))[order(coef(ridge.mod)[,
                    ncol(coef(ridge.mod))],decreasing=TRUE)])

#remove intercept
allnames<-setdiff(allnames,allnames[grepl("Intercept",allnames)])

#assign colors
cols<-rep("gray",length(allnames))
cols[allnames %in% coef.increase]<-"green" # higher medv is good
cols[allnames %in% coef.decrease]<-"red" # lower medv is not
```

```
plot_glmnet(ridge.mod, label=TRUE, s=bestlambda, col=cols)
```

