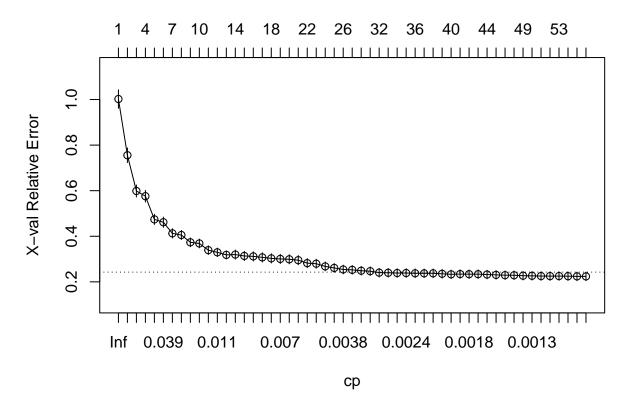
Regression-tree.R

rocka

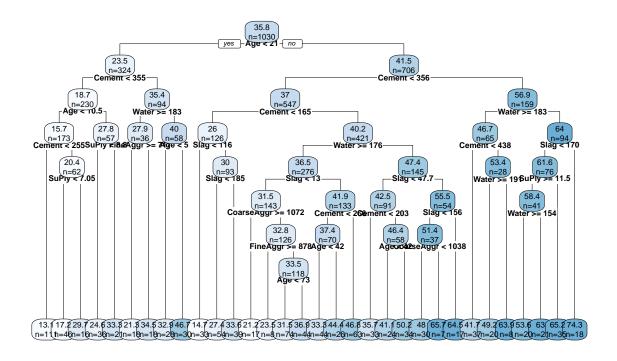
2023-12-10

```
#Regression Tree
library(rpart)
library(rpart.plot)
## Warning: package 'rpart.plot' was built under R version 4.3.2
library(DMwR2)
## Warning: package 'DMwR2' was built under R version 4.3.2
## Registered S3 method overwritten by 'quantmod':
    method
##
    as.zoo.data.frame zoo
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
      filter, lag
## The following objects are masked from 'package:base':
##
      intersect, setdiff, setequal, union
library(tidyverse)
## Warning: package 'ggplot2' was built under R version 4.3.2
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v forcats 1.0.0 v readr
                                  2.1.4
## v ggplot2 3.4.4 v stringr 1.5.0
## v lubridate 1.9.2 v tibble 3.2.1
             1.0.2 v tidyr
## v purrr
                                 1.3.0
```

size of tree



1se regression tree for concrete compressive strength (in MPa)



var(df\$response)

[1] 279.0818

```
printcp(rpart.modela.1se)
```

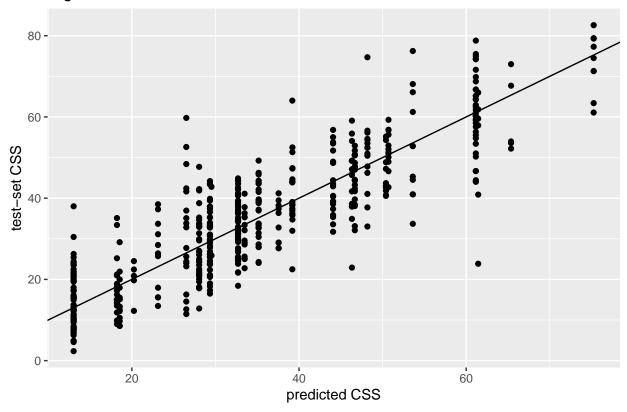
```
##
## Regression tree:
## rpart(formula = response ~ ., data = df, method = "anova", cp = 0.001)
##
## Variables actually used in tree construction:
                            CoarseAggr FineAggr
## [1] Age
                 Cement
                                                  Slag
                                                             SuPly
                                                                        Water
##
## Root node error: 287175/1030 = 278.81
##
## n= 1030
##
##
            CP nsplit rel error xerror
## 1 0.2480823
                        1.00000 1.00217 0.040518
## 2
     0.1714067
                    1
                        0.75192 0.75551 0.032159
## 3
     0.0683911
                    2
                       0.58051 0.59872 0.026329
                    3 0.51212 0.57584 0.025762
## 4 0.0645348
## 5 0.0398308
                    4 0.44759 0.47410 0.022379
## 6 0.0391064
                    5 0.40775 0.46161 0.022415
```

```
0.36865 0.41282 0.020684
## 7 0.0260850
## 8 0.0217650
                     7 0.34256 0.40551 0.020089
## 9 0.0198351
                     8 0.32080 0.37354 0.019389
                    9 0.30096 0.36847 0.019389
## 10 0.0198292
## 11 0.0112837
                    10
                        0.28113 0.33909 0.018947
## 12 0.0101590
                  11 0.26985 0.32956 0.018673
## 13 0.0096024
                  12
                       0.25969 0.31831 0.017869
## 14 0.0085087
                  13 0.25009 0.31966 0.020159
## 15 0.0082445
                   14
                        0.24158 0.31362 0.020397
                  15 0.23334 0.31180 0.020444
## 16 0.0076775
## 17 0.0075075
                   16
                        0.22566 0.30733 0.020398
## 18 0.0071104
                    17
                        0.21815 0.30307 0.020384
                        0.21104 0.30057 0.020275
## 19 0.0069893
                    18
                   19
                       0.20405 0.29900 0.020276
## 20 0.0069440
## 21 0.0063950
                    20 0.19711 0.29509 0.020173
## 22 0.0061580
                    21 0.19071 0.28211 0.020024
## 23 0.0054207
                    22 0.18455 0.27966 0.019892
                    23
## 24 0.0043104
                       0.17913 0.26777 0.019487
## 25 0.0040058
                    24
                        0.17482 0.26109 0.019494
## 26 0.0035609
                    25
                        0.17082 0.25427 0.019208
## 27 0.0030721
                    26
                        0.16726 0.25271 0.019170
## 28 0.0030072
                    28
                        0.16111 0.24880 0.018947
## 29 0.0026963
                    29
                         0.15810 0.24685 0.018898
## 30 0.0026865
                    31
                       0.15271 0.24033 0.018937
(MSE<-278.81*0.24033)
## [1] 67.00641
(rmse<-sqrt(MSE))</pre>
## [1] 8.185744
# Use a validation set to estimate test-set MSE
set.seed(7)
train = sample(1:nrow(df), nrow(df)/2)
set.seed(SEED)
rpart.concrete.train <- rpart(response~</pre>
                                ., data = df[train,],
                              method="anova", cp=0)
rpart.concrete.train <- rt.prune(rpart.concrete.train, se=1)</pre>
Yhat <- predict(rpart.concrete.train, newdata = df[-train,])</pre>
concrete.test <- df[-train, "response"]</pre>
(MSE <- mean((Yhat-concrete.test)^2))</pre>
```

[1] 68.21776

```
(RMSE <- sqrt(MSE))</pre>
## [1] 8.259404
#repeating with different seed for creating training set
set.seed(5)
train <- sample(1:nrow(df), nrow(df)/2)</pre>
set.seed(SEED)
rpart.concrete.train <- rpart(response~</pre>
                                   ., data = df[train,],
                                method="anova", cp=0)
rpart.concrete.train <- rt.prune(rpart.concrete.train, se=1)</pre>
Yhat <- predict(rpart.concrete.train, newdata = df[-train,])</pre>
concrete.test <- df[-train, "response"]</pre>
(MSE <- mean((Yhat-concrete.test)^2))</pre>
## [1] 64.34217
(RMSE <- sqrt(MSE))</pre>
## [1] 8.021357
ggplot(data.frame(Yhat, concrete.test), aes(x=Yhat ,y=concrete.test)) +
  geom_point() +
  geom_abline(slope=1,intercept=0) +
  labs(x="predicted CSS",
       y="test-set CSS",
       title="regression tree")
```

regression tree



```
# Results: MSE RMSE

# Cross-validated MSE 67.00641 8.185744

# Validation set MSE (set.seed(7)) 68.21776 8.259404

# Validation set MSE (set.seed(5)) 64.34217 8.021357
```