Week 3 Homework

Ben Arancibia June 23, 2015

KJ 7.2 Friedman (1991) introduced several benchmark datasets create by simulation. One of these simulations used the following non–linear equation to create data: $y = 10\sin(\pi x_1 x_2) + 20(x_3 - 0.5)^2 + 10x_4 + 5x_5 + N(0, \sigma^2)$

```
library(mlbench)
library(caret)

## Warning: package 'caret' was built under R version 3.1.3

## Loading required package: lattice

## Warning: package 'lattice' was built under R version 3.1.3

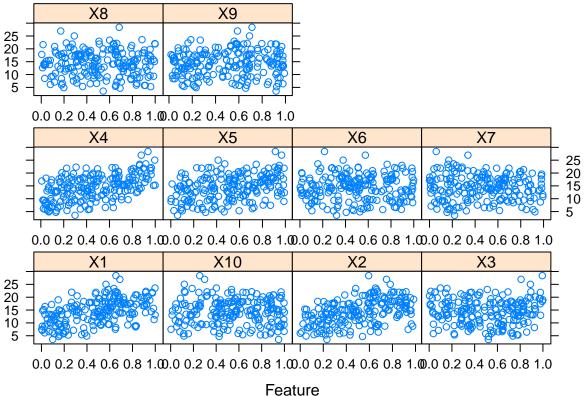
## Loading required package: ggplot2

## Warning: package 'ggplot2' was built under R version 3.1.3

set.seed(200)

trainingData <- mlbench.friedman1(200, sd = 1)

trainingData$x <- data.frame(trainingData$x)
featurePlot(trainingData$x, trainingData$y)</pre>
```



```
testData <- mlbench.friedman1(5000, sd = 1)
testData$x <- data.frame(testData$x)</pre>
```

Tune several models on these data. For example:

```
set.seed(921)
knnModel <- train(x = trainingData$x, y = trainingData$y, method = "knn",
                 preProc = c("center", "scale"),
                 tuneLength = 10)
knnModel
## k-Nearest Neighbors
##
## 200 samples
   10 predictor
##
##
## Pre-processing: centered, scaled
## Resampling: Bootstrapped (25 reps)
##
## Summary of sample sizes: 200, 200, 200, 200, 200, 200, ...
##
## Resampling results across tuning parameters:
##
##
    k
        RMSE
                  Rsquared
                             RMSE SD
                                        Rsquared SD
##
     5 3.488933 0.5019753 0.2658769
                                        0.07412999
##
     7 3.324957 0.5484600 0.2275136 0.06432697
##
     9 3.224541
                  0.5853589 0.2425946 0.06903863
##
    11 3.178450 0.6091595 0.2593909 0.07304742
##
    13 3.183655 0.6183553
                             0.2523970 0.06995701
##
    15 3.188007 0.6250729
                             0.2408867 0.06604822
##
    17 3.214343 0.6281943
                             0.1890541
                                       0.05315353
##
    19 3.208743 0.6403733
                             0.1921773 0.05056336
##
    21 3.215199 0.6487431
                             0.1830866 0.04961129
##
    23 3.235167 0.6528070 0.1870124 0.04751686
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was k = 11.
knnPred <- predict(knnModel, newdata = testData$x)</pre>
postResample(pred = knnPred, obs = testData$y)
```

```
## RMSE Rsquared
## 3.1222641 0.6690472
```

Which models appear to give the best performance? Does MARS select the informative predictors (those named X1–X5)?

K-nearest neighbors models are better when predictors and the response relies on the samples' proximity in the predictor space.