Creating a machine learning analysis plan

Instructions

If you haven't already, download forestfires.tsv data and documentation. Use the caret package to prepare the forest fire data for a machine learning analysis. When you finish preparing the data, save your script and upload it to Canvas.

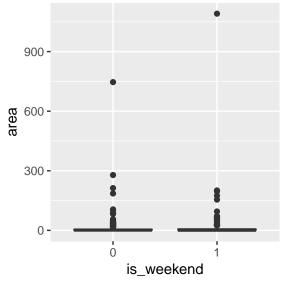
```
### Load the forest fires datasets on your machine by setting the working directory
### or specify the directory of your data
### This code is commented out and will NOT run
### ff = read.delim("forestfires.tsv", sep = '\t', header = TRUE)
```

1. Make at least one new feature and plot it against the burn area. From visual inspection does there appear to be a relationship between the new feature and the burn area?

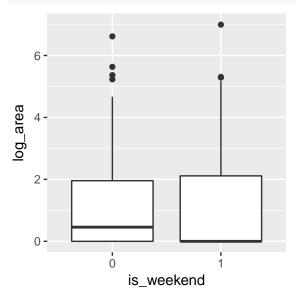
```
ff$is_weekend = ifelse(ff$day %in% c("sat", "sun"), 1, 0)
ff$is_weekend = factor(ff$is_weekend)

ff$log_area = log(ff$area + 1)

ggplot(ff, aes(x = is_weekend, y = area)) +
    geom_boxplot()
```



```
ggplot(ff, aes(x = is_weekend, y = log_area)) +
  geom_boxplot()
```



Since the area variable is so skewed, it is easier to see relationships on the log scale (second plot). The relationship between is_weekend and burn area is not overwhelming but the median is lower on weekend. There is also a gapairs plot of the is_weekend variable in the tutorial.

2. Use createDataPartition to split 80% of the forest fire data into a training set.

```
in_train = createDataPartition(y = ff$log_area, p = 0.8, list = FALSE)
ff_train = ff[in_train, ]
ff_test = ff[-in_train, ]
```

3. Use preProcess to prepare your data for analysis. What, if any, variables were removed for near zero variance?

One variable (rain) is removed. Following the tutorial, using the model.matrix method helps find other near zero variance categories in the factor variables.

```
##
               freqRatio percentUnique zeroVar
## X
                1.085714
                              2.1634615
                                          FALSE FALSE
## Y
                1.687500
                              1.4423077
                                          FALSE FALSE
## month
                1.145985
                              2.8846154
                                          FALSE FALSE
## day
                1.212121
                              1.6826923
                                          FALSE FALSE
## FFMC
                1.090909
                             23.5576923
                                          FALSE FALSE
## DMC
                             45.1923077
                                          FALSE FALSE
                1.142857
## DC
                1.142857
                             45.9134615
                                          FALSE FALSE
## ISI
                1.055556
                             26.9230769
                                          FALSE FALSE
                1.333333
                             43.2692308
                                          FALSE FALSE
## temp
## RH
                1.350000
                             17.5480769
                                          FALSE FALSE
                                          FALSE FALSE
## wind
                1.045455
                              5.0480769
## rain
              204.000000
                              1.6826923
                                          FALSE TRUE
               66.000000
                                          FALSE FALSE
## area
                             48.7980769
## is weekend
                1.849315
                              0.4807692
                                          FALSE FALSE
## log_area
               66.000000
                             48.7980769
                                          FALSE FALSE
```