Report

Objectives

The main objective is to build a model that can correctly predict in-campus locations using Wi-Fi signal strength from various routers around the buildings. This indoors positioning proves to be tricky as the multiple devices interfere with each other and prevents geo-location using the usual GPS signals.

Data Cleaning and optimization

Initially I tried to run the raw data through the entire modelling algorithm of Random Forest, with the computing power I have, it was able to run over a period of 6hrs. However, this had to be improved for the other algorithms, therefore, I took advantage of R parallelization capabilities and used the *doParallel* library to utilize 7 cores out of the 8 available on my machine. Furthermore, I reduced the number of instances from 19,937 on each of the 529 features to only 5249 instances over the 529 features. This was achieved by only looking at one building (building 0). This combination allowed my computer to runt the random forest modeling algorithm in 5 minutes. With these new settings in place I ran 3 other algorithms: K-Nearest Neighbors (KNN), SVM Radial, and C50.

Results

Random Forest – All data results

```
No pre-processing
Resampling: Cross-Validated (10 fold, repeated 3 times)
Summary of sample sizes: 17948, 17954, 17932, 17936, 17944, 17938, ...
Resampling results:

Accuracy Kappa
0.7744364 0.7741045

Tuning parameter 'mtry' was held constant at a value of 22
```

Random Forest - Building 0

```
No pre-processing
Resampling: Cross-Validated (10 fold, repeated 3 times)
Summary of sample sizes: 4724, 4727, 4718, 4720, 4723, 4730, ...
Resampling results:

Accuracy Kappa
0.7347186 0.7336129

Tuning parameter 'mtry' was held constant at a value of 22
```

```
No pre-processing
Resampling: Cross-Validated (10 fold, repeated 3 times)
Summary of sample sizes: 4724, 4721, 4728, 4722, 4723, 4718, ...
Resampling results across tuning parameters:
  model winnow trials Accuracy
                                       Kappa
  rules FALSE
                           0.6033380 0.6017167
  rules FALSE
rules FALSE
                           0.7238785 0.7227538
                           0.7425067
                                      0.7414588
                           0.6072169 0.6056116
          TRUE
  rules
          TRUE 10
                           0.7242557 0.7231324
  rules
                           0.7409067 0.7398507
  rules
          TRUE 20
         FALSE
  tree
                           0.6122203 0.6106394
         FALSE
                           0.7194127 0.7182689
  tree
         FALSE
                           0.7339312 0.7328461
0.6135000 0.6119255
  tree
           TRUE
  tree
                           0.7202678 0.7191261
  tree
           TRUE
                           0.7336153 0.7325286
  tree
           TRUE
Accuracy was used to select the optimal model using the largest value.
The final values used for the model were trials = \overline{20}, model = rules and winnow = FALSE.
```

SVM Radial Fit - Building 0

```
Call:
svm(formula = MasterID ~ ., data = df3, kernel = "radial", cost = 5, scale = FALSE)

Parameters:
SVM-Type: C-classification
SVM-Kernel: radial
cost: 5

Number of Support Vectors: 5051
```

K-Nearest Neighbors (KNN) – Building 0

```
Pre-processing: centered (520), scaled (520)
Resampling: Cross-Validated (10 fold, repeated 3 times)
Summary of sample sizes: 4721, 4723, 4724, 4719, 4728, 4728, ...
Resampling results across tuning parameters:
  k
      Accuracy
                 Kappa
   5 0.5331114 0.5312129
     0.5137396 0.5117584
  9 0.4979818 0.4959351
  11 0.4886605 0.4865756
  13 0.4728360 0.4706896
  15
     0.4611507 0.4589529
     0.4443897 0.4421246
  17
  19 0.4239463 0.4215928
  21 0.4065500 0.4041257
  23 0.3902922
                 0.3877968
  25 0.3755103 0.3729502
```

Summary of Results

```
Call:
summary.resamples(object = ModelData)
Models: RF, KNN, C50
Number of resamples: 30
Accuracy
                           Median
         Min.
                1st Qu.
                                       Mean
                                               3rd Qu.
                                                            Max. NA's
RF 0.7005650 0.7234963 0.7326427 0.7347186 0.7458378 0.7740113
                                                                    0
KNN 0.4971209 0.5208729 0.5301417 0.5331114 0.5472822 0.5727969
                                                                    0
C50 0.7142857 0.7436695 0.7557172 0.7529214 0.7624909 0.7817837
                                                                    0
Kappa
                                               3rd Qu.
         Min.
                1st Qu.
                           Median
                                       Mean
                                                            Max. NA's
RF 0.6993248 0.7223434 0.7315311 0.7336129 0.7447809 0.7730688
                                                                    0
KNN 0.4950449 0.5189198 0.5282289 0.5312129 0.5454369 0.5710685
                                                                    0
C50 0.7131304 0.7426229 0.7547239 0.7519144 0.7615170 0.7808938
                                                                    0
```

Based on these results, the best algorithm to use is the C50 algorithm, however if run time is not an issue, the Random Forest prediction using the full range of data is the best choice with the highest Accuracy and Kappa scores.

Recommendations

Attempting other algorithms with using cloud computing services from amazon (aws) is an option that is worth mentioning. For this project I attempted to use these but proved to be more problematic and complicated than just cleaning and refining the data.

Taking principal components analysis (PCA) and applying it to this large amount of data can be a useful tool to further increase the time output for each model.