## **C3** Indoor Wifi Locationing

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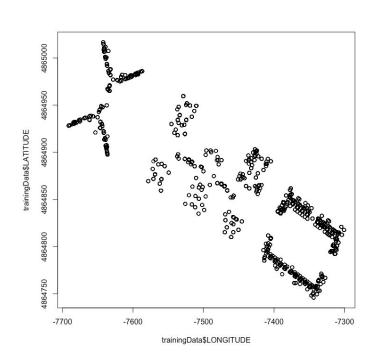
Aug 3rd, 2018





#### Indoor Location Methods Based on WiFi Fingerprinting

http://archive.ics.uci.edu/ml/datasets/UJIIndoorLoc



#### **UJIIndoorLoc Data**

- 21,048 total observations split among 3 buildings w/ 4-5 floors.
- 2 529 total features. 520 Wireless Access Points (WAPs) as predictors.
- 2 Datasets from original study: trainingData & validationData.

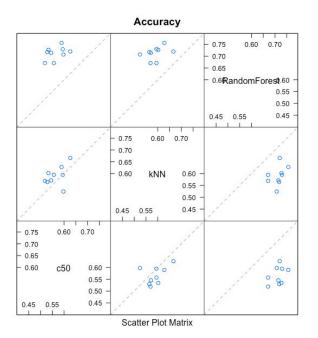


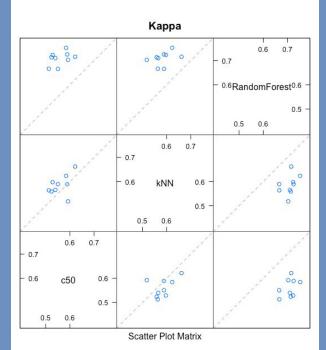
### Classification Methodology

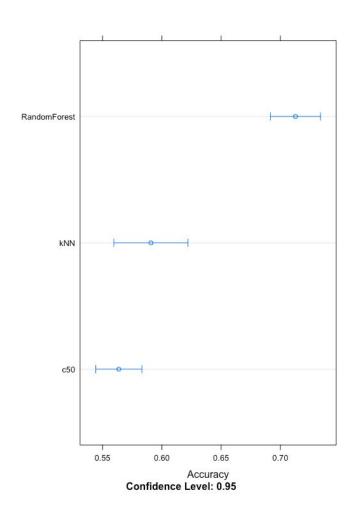
Evaluate multiple machine learning models for best accuracy.

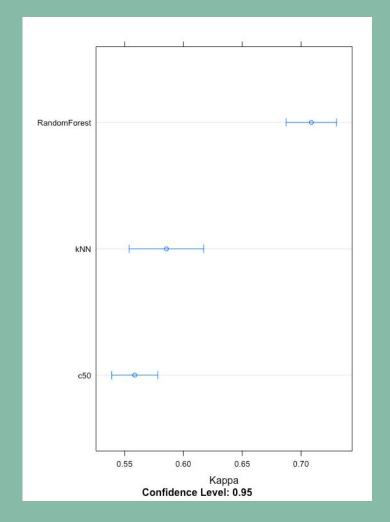
- ☐ Used only "trainingData.csv" from the source data since ~20k observations of 529 variables.
  - Only 312 WAPs in common with validationData.csv
  - ValidationData had a number of missing attributes.
- ☐ Combined Floor & SpaceID features into composite variable as target to be predicted.
  - Experimented with other composite variables.
- ☐ Subsetted observations by 3 buildings and sampled 20% of building 1 due to lower density.
- Removed features (longitude, latitude, building ID, relative position, timestamp, phone ID, and user ID).
- ☐ Trained and tested using C50, RandomForest, kNN & SVM algorithms to determine best classification model.

#### C50, RandomForest & kNN











## RandomForest Was the Best Performer

30 resamples

ACCURACY	Min	Mean	Max
c50	0.440	0.564	0.683
kNN	0.524	0.591	0.667
RandomForest	0.671	0.713	0.756

KAPPA	Min	Mean	Max
c50	0.434	0.559	0.678
kNN	0.518	0.586	0.662
RandomForest	0.666	0.709	0.753



# Trained Model Accurate with New Data

#### RandomForest

	Accuracy	Kappa
Training	0.713	0.709
Prediction	0.777	0.774



### Conclusions

#### A lot of iterations.

- Best model for predicting Floor\_SpaceID composite variable was RandomForest with mtry = 64.
- Size of data set and dimensionality resulted in preliminary attempts that had very high computational costs and poor accuracy.
- □ Subsetting by building and sampling 20% of data offered good accuracy with reasonable computation times as a trade-off.
- The SpaceID identifying an office or classroom in the data is unconventional and not unique (multiple spaces on multiple floors in multiple buildings with same SpaceID).
- WAPs are unevenly dispersed geographically and in some areas likely too dense to allow for accurate triangulation/positioning. Poor communications topology.

