

IST687 - Music Classification Project

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4/18/2019

Executive Summary

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Introduction

Related Work

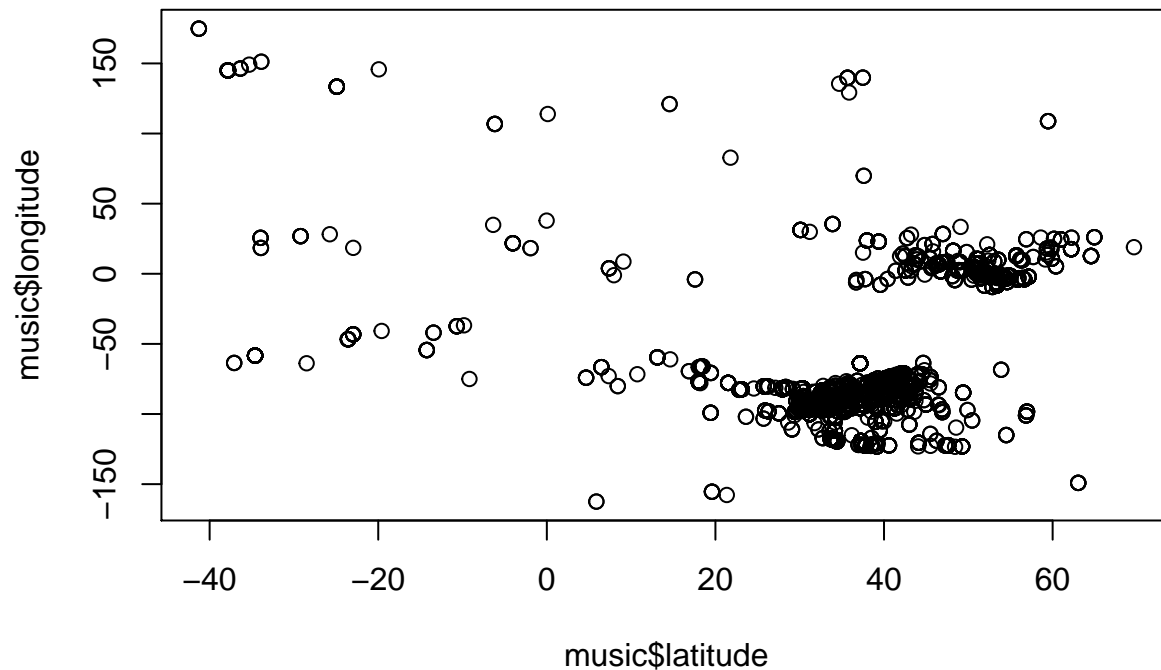
Thierry Bertin-Mahieux, Daniel P.W. Ellis, Brian Whitman, and Paul Lamere. The Million Song Dataset. In Proceedings of the 12th International Society for Music Information Retrieval Conference (ISMIR 2011), 2011.

Dataset

```
#Prior to importing, a new column artist.hotttnesss.label was adding with  
#Hot(>.4590), Warm(<.4590 and >.3357), Cold(<.3357). Four rows with blanks in  
#familiarity were also deleted.  
music <- read.csv(file="/Library/Mobile Documents/com~apple~CloudDocs/Syracuse/IST687/Project/Music Pro  
#Copy original data to a new dataframe music1 and exclude unneeded data  
music1 <- music[-c(1:5,7,16,19,21:25,30,34)]
```

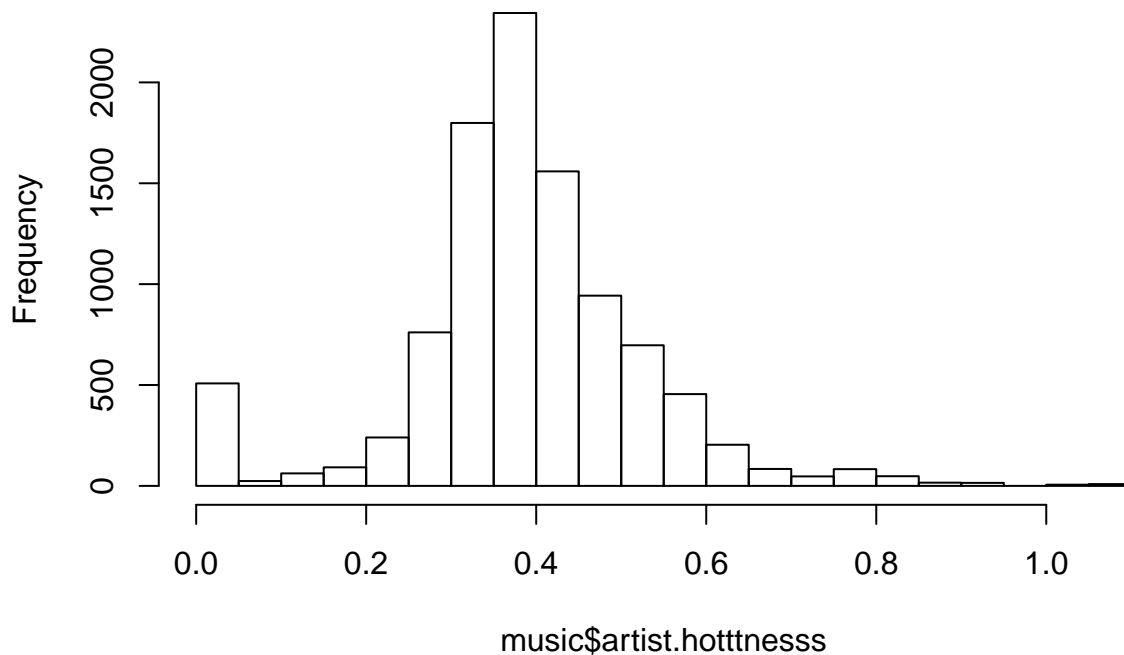
Features

```
#View the number of Cold/Warm/Hot labels  
table(music1$artist.hotttnesss.label)  
  
##  
## Cold Hot Warm  
## 2870 2376 4750  
  
#Plot artists latitude and longitude  
plot(music$latitude,music$longitude)
```



```
#Plot artist hotttnesss
hist(music$artist.hotttnesss,breaks=20)
```

Histogram of music\$artist.hotttnesss



```
music$artist.hotttnesss #THIS IS
INCOMPLETE CODE FOR PLOTTING ADDITIONAL DATA... #Create a map of the world
mapWorld <- borders("world", colour="gray50", fill="white")

#Code from https://rpubs.com/spoonerf/global_map #Need to figure out what to put in locs
locs<-read.csv("my_locations.csv") locs<- sp_dups<-data.frame(ddply(locs,.(Longitude,Latitude),nrow))
sp_dups$loc_id <- 1:length(sp_dups$Longitude) sp_dups_df<-merge(sp_dups, locs, by=c("Longitude","Latitude"))
```

```

loc<-data.frame(sp_dups_dfLongitude,sp_dups_dfLatitude,sp_dups_df$V1) loc<-unique(loc) colnames(loc)<-
c("Longitude", "Latitude", "V1")

coordinates(loc)<-c("Longitude","Latitude") proj4string(loc) <- CRS("+proj=longlat")

loc_df<-data.frame(loc)

theme_opts <- list(theme(panel.grid.minor = element_blank(), panel.grid.major = element_blank(),
panel.background = element_blank(), plot.background = element_rect(fill="white"), panel.border =
element_blank(), axis.line = element_blank(), axis.text.x = element_blank(), axis.text.y = element_blank(),
axis.ticks = element_blank(), axis.title.x = element_blank(), axis.title.y = element_blank(), plot.title =
element_text(size=22)))

library(maps) library(mapdata)

ggplot(data=loc_df, aes(Longitude, Latitude, group=NULL,fill=NULL,size=V1))+#, fill=hole)) + bor-
ders(fill="light grey",colour="light grey")+ geom_point(color="black",alpha=I(7/10))+ scale_size(range=c(1,7),
guide = "legend",labs(size="No. of Populations"))+ coord_equal()+ theme_opts

```

Methods

```

#Do analysis to determine hot/warm/cold artists based on hotttnesss

#The random forest analysis (106-163) is from a training video by Bharatendra Rai
#at https://www.youtube.com/watch?v=dJclNIN-TPo
#Data Partition - ind = independent samples
#The code below runs in console but not R Markdown
set.seed(123)
ind<- sample(2,nrow(music1), replace=TRUE,prob=c(0.7,0.3))
train <- music1[ind==1,]
test <- music1[ind==2,]

#Run randomForest on music1
library(randomForest)

## randomForest 4.6-14

## Type rfNews() to see new features/changes/bug fixes.

set.seed(222)
rf <- randomForest(music1[, -21],music1[, 21])
print(rf)

##
## Call:
## randomForest(x = music1[, -21], y = music1[, 21])
##              Type of random forest: classification
##              Number of trees: 500
## No. of variables tried at each split: 4
##
##              OOB estimate of  error rate: 19.18%
## Confusion matrix:
##           Cold  Hot  Warm class.error
## Cold 2083   14  773   0.2742160
## Hot    8 1968  400   0.1717172
## Warm  455  267 4028   0.1520000

```

```
attributes(rf)
```

```
## $names
## [1] "call"          "type"          "predicted"
## [4] "err.rate"      "confusion"     "votes"
## [7] "oob.times"     "classes"       "importance"
## [10] "importanceSD"  "localImportance" "proximity"
## [13] "ntree"         "mtry"          "forest"
## [16] "y"            "test"          "inbag"
##
## $class
## [1] "randomForest"
```

```
rf$confusion
```

```
##      Cold  Hot Warm class.error
## Cold 2083   14  773   0.2742160
## Hot   8 1968  400   0.1717172
## Warm  455  267 4028   0.1520000
```

```
#Run randomForest again with tune mtry data from below
#Need HELP to fix the next line of code so it works...
#rf <- randomForest(artist.hotttnesss.label ~.,data=music1,ntree=200,mtry=8,
#importance=TRUE,proximity=TRUE)
```

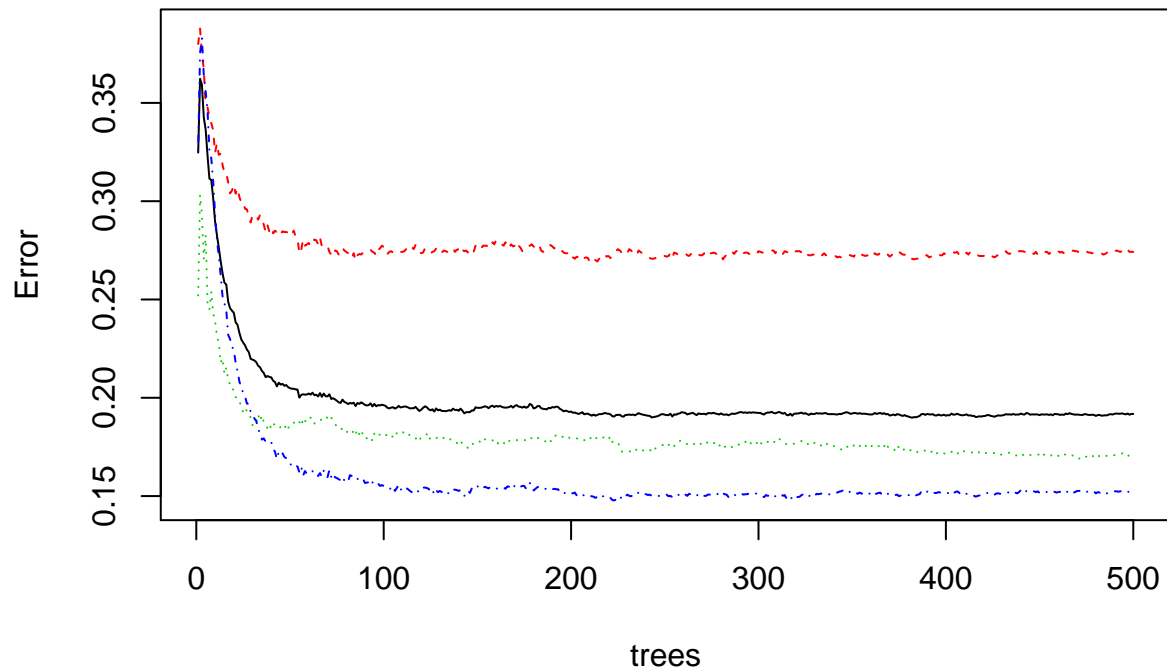
```
#Prediction & Confusion Matrix - train data
#library(caret)
#p1<-predict(rf,train)
# For some reason this is returning an error buit p2 below is working
#confusionMatrix(p1,train)
```

```
#Predition & Confusion Matrix - test data
#p2<-predict(rf,test)
#confusionMatrix(p2,test$artist.hotttnesss.label)
```

```
#Error rate of Random Forest
```

```
plot(rf)
```

rf



#The error rate is not improving after ~100 trees

#Tune mtry

```
#t <- tuneRF(train[,-21],train[,21],
```

```
#      stepFactor=.5,
```

```
#      plot=TRUE,
```

```
#      ntreeTry=200,
```

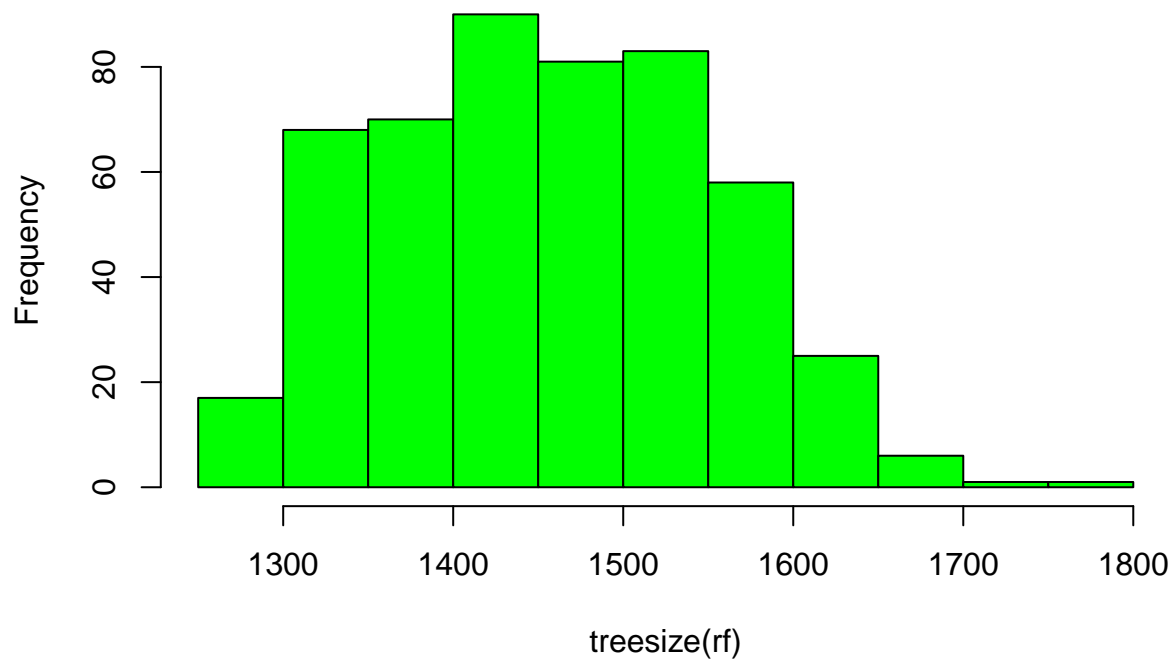
```
#      trace=TRUE,
```

```
#      improve=0.05)
```

#No. of nodes for the trees

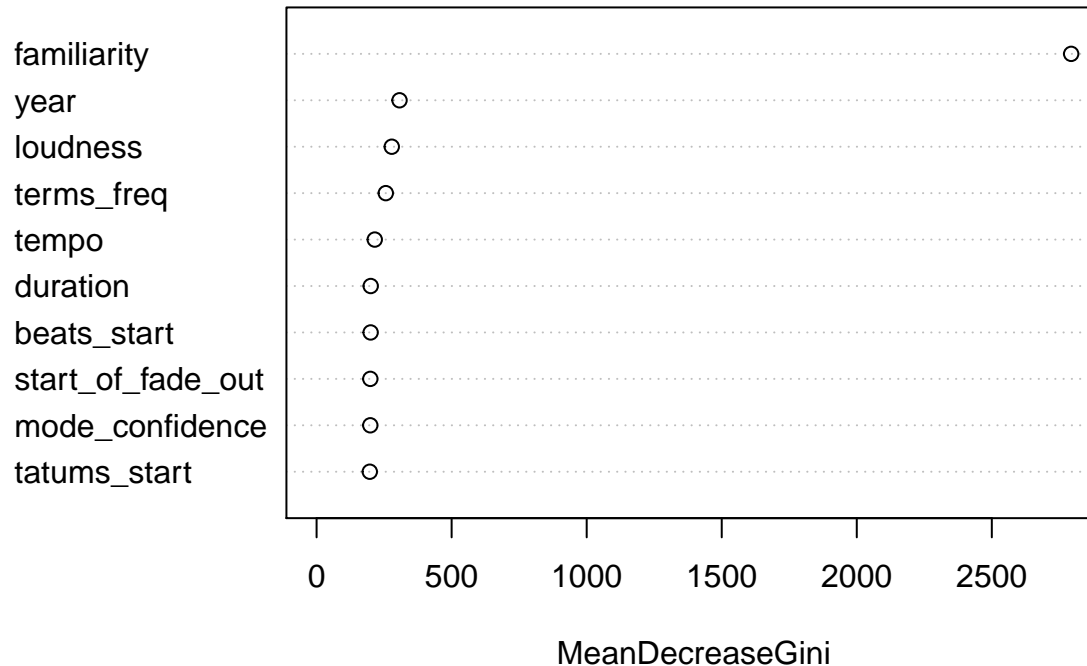
```
hist(treesize(rf),  
     main="Number of Nodes for the Trees",  
     col="green")
```

Number of Nodes for the Trees



```
# Variable Importance
# Familiarity is much more important than the other variables. Should it be removed and run again?
varImpPlot(rf,
            sort=T,
            n.var=10,
            main="Top 10 - Variable Importance")
```

Top 10 – Variable Importance



```
importance(rf)
```

```
##               MeanDecreaseGini
## bars_confidence      190.72247
## beats_confidence     166.50136
## beats_start          200.11836
## duration             200.41104
## end_of_fade_in       158.14247
## familiarity          2793.89271
## key                  123.67454
## key_confidence       186.25796
## latitude             158.41154
## longitude            144.10180
## loudness             278.31880
## mode_confidence      198.81990
## start_of_fade_out    198.91655
## tatums_confidence    175.52538
## tatums_start        196.93868
## tempo               215.29801
## terms_freq          256.20849
## time_signature        57.60156
## time_signature_confidence 144.43782
## year                306.99483
```

```
varUsed(rf)
```

```
## [1] 40656 36125 42149 40892 32952 73071 29208 40149 22266 21818 45831
## [12] 41661 40861 37682 41306 44176 30635 13853 31723 20627
```

```
#Multidimensional Scaling Plot
#The code below causes R to lock up...
```

```
#MDSplot(rf,train$artist.hotttnesss.label)
```

Results

Conclusion

Appendices