

Submission 9

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A single generalized boosted regression model.

Loading

```
source("../helpers/predictions.R")
```

```
library(caret)
```

```
## Loading required package: lattice  
## Loading required package: ggplot2
```

```
library(doMC)
```

```
## Loading required package: foreach  
## Loading required package: iterators  
## Loading required package: parallel
```

```
train <- read.csv("../data/processed/processed_train.csv")  
test <- read.csv("../data/original/test.csv")
```

Data

```
set.seed(123)  
training.indices <- createDataPartition(train$Label, p=0.6, list=F)  
training <- train[training.indices,]  
validation <- train[-training.indices,]
```

Models

Fit a GBM with training data

```
predictors <- training[,setdiff(names(training),c("Label","EventId"))]  
registerDoMC(cores=8)  
gbm.fit <- train(x=predictors, y=training$Label, method="gbm")
```

```
## Loading required package: gbm  
## Loading required package: survival  
## Loading required package: splines  
##  
## Attaching package: 'survival'
```

```
##
## The following object is masked from 'package:caret':
##
##   cluster
##
## Loaded gbm 2.1
```

## Iter	TrainDeviance	ValidDeviance	StepSize	Improve
## 1	1.2303	nan	0.1000	0.0277
## 2	1.1848	nan	0.1000	0.0228
## 3	1.1489	nan	0.1000	0.0178
## 4	1.1170	nan	0.1000	0.0160
## 5	1.0889	nan	0.1000	0.0139
## 6	1.0654	nan	0.1000	0.0116
## 7	1.0424	nan	0.1000	0.0114
## 8	1.0245	nan	0.1000	0.0090
## 9	1.0072	nan	0.1000	0.0087
## 10	0.9936	nan	0.1000	0.0066
## 20	0.9022	nan	0.1000	0.0032
## 40	0.8392	nan	0.1000	0.0009
## 60	0.8109	nan	0.1000	0.0004
## 80	0.7928	nan	0.1000	0.0005
## 100	0.7805	nan	0.1000	0.0003
## 120	0.7714	nan	0.1000	0.0002
## 140	0.7634	nan	0.1000	0.0002
## 150	0.7602	nan	0.1000	0.0000

```
gbm.fit
```

```
## Stochastic Gradient Boosting
##
## 150001 samples
##   30 predictors
##   2 classes: 'b', 's'
##
## No pre-processing
## Resampling: Bootstrapped (25 reps)
##
## Summary of sample sizes: 150001, 150001, 150001, 150001, 150001, 150001, ...
##
## Resampling results across tuning parameters:
##
##   interaction.depth  n.trees  Accuracy  Kappa  Accuracy SD  Kappa SD
## 1                    50      0.8        0.6    0.001    0.003
## 1                    100      0.8        0.6    0.001    0.003
## 1                    200      0.8        0.6    0.001    0.003
## 2                     50      0.8        0.6    0.001    0.002
## 2                     100      0.8        0.6    0.001    0.003
## 2                     200      0.8        0.6    0.001    0.003
## 3                     50      0.8        0.6    0.002    0.004
## 3                     100      0.8        0.6    0.001    0.003
## 3                     200      0.8        0.6    0.001    0.003
##
```

```
## Tuning parameter 'shrinkage' was held constant at a value of 0.1
## Accuracy was used to select the optimal model using the largest value.
## The final values used for the model were n.trees = 150,
## interaction.depth = 3 and shrinkage = 0.1.
```

```
val <- validation[,setdiff(names(validation),c("EventId","Label"))]
pred.val <- predict(gbm.fit, val)
confusionMatrix(pred.val, validation$Label)
```

```
## Warning: NAs produced by integer overflow
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction      b      s
##           b 58897 10421
##           s  6836 23845
##
##           Accuracy : 0.827
##           95% CI : (0.825, 0.83)
##           No Information Rate : 0.657
##           P-Value [Acc > NIR] : <2e-16
##
##           Kappa : NA
##           McNemar's Test P-Value : <2e-16
##
##           Sensitivity : 0.896
##           Specificity : 0.696
##           Pos Pred Value : 0.850
##           Neg Pred Value : 0.777
##           Prevalence : 0.657
##           Detection Rate : 0.589
##           Detection Prevalence : 0.693
##           Balanced Accuracy : 0.796
##
##           'Positive' Class : b
##
```

```
pred.test <- predict(gbm.fit, test, type="prob")
pred.df <- PrepPrediction(pred.test, test)
WriteSubmission(pred.df, 9)
```