2. KAGGLE Problem- Otto Group Product Classification Challenge

Classify products into the correct category

The Otto Group is one of the world's biggest e-commerce companies, with subsidiaries in more than 20 countries, including Crate & Barrel (USA), Otto.de (Germany) and 3 Suisses (France).

They are selling millions of products worldwide every day, with several thousand products being added to our product line.

A consistent analysis of the performance of their products is crucial. However, due to our diverse global infrastructure, many identical products get classified differently. Therefore, the quality of our product analysis depends heavily on the ability to accurately cluster similar products. The better the classification, the more insights we can generate about our product range.

For this competition, a dataset is provided with 93 features for more than 200,000 products. The objective is to build a predictive model which is able to distinguish between the main product categories.

File descriptions

trainData.csv - the training set
testData.csv - the test set
sampleSubmission.csv - a sample submission file in the correct format

Data fields

- id an anonymous id unique to a product.
- feat 1, feat 2, ..., feat 93 the various features of a product.
- target the class of a product.

Data Set Description

Each row corresponds to a single product. There are a total of 93 numerical features, which represent counts of different events. All features have been obfuscated and will not be defined any further.

There are nine categories for all products. Each target category represents one of our most important product categories (like fashion, electronics, etc.). The products for the training and testing sets are selected randomly.

IMPLEMENTATION:

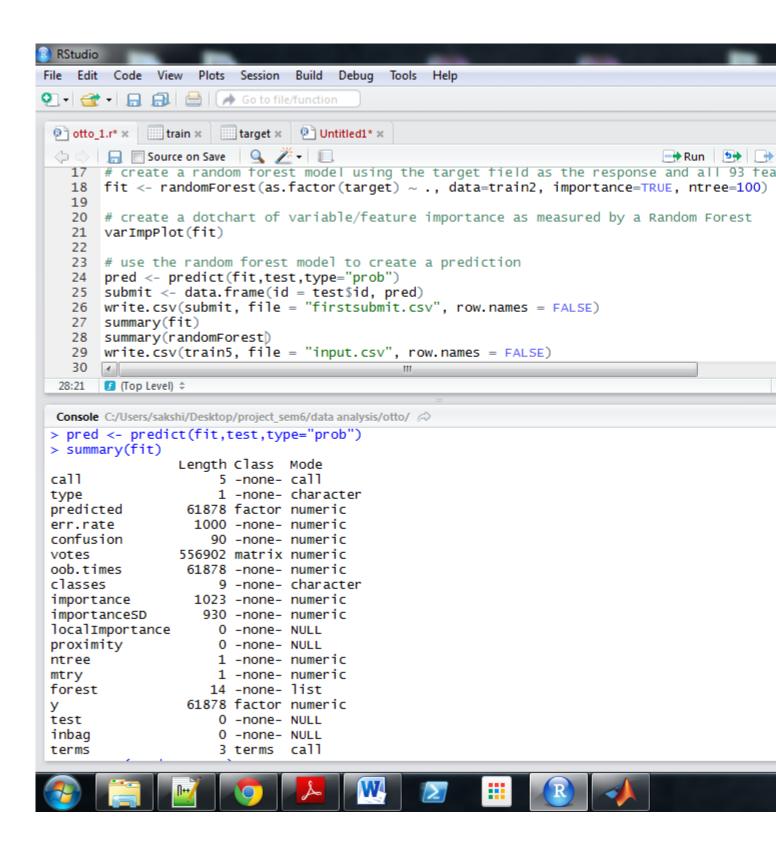
1. RANDOM FOREST GENERATION

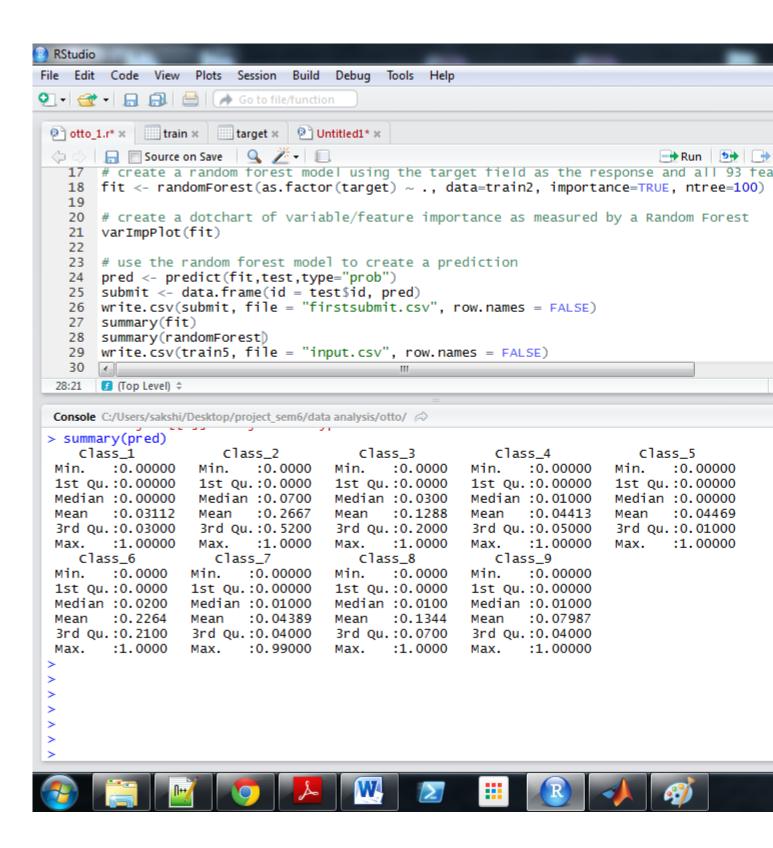
```
install.packages('randomForest')
                                             # install randomForest package.
library(randomForest)
                                             # include random forest library.
 set.seed(12)
                             # set a unique seed number so as to get the same results everytime we run
                                                                                    the below model.
fit <- randomForest(as.factor(target) ~ ., data=train2, importance=TRUE, ntree=100)
# create a random forest model using the target field as the response and all 93 features as inputs
varImpPlot(fit)
                 # create a dotchart of variable/feature importance as measured by a Random Forest
pred <- predict(fit,test,type="prob")</pre>
                                                  # use the random forest model to create a prediction
submit <- data.frame(id = test$id, pred)</pre>
write.csv(submit, file = "random forest submit.csv", row.names = FALSE)
summary(fit)
         Length Class Mode
call
             5 -none- call
              1 -none- character
type
predicted
              61878 factor numeric
err.rate
             1000 -none- numeric
confusion
                90 -none- numeric
           556902 matrix numeric
votes
oob.times
              61878 -none- numeric
               9 -none- character
classes
                1023 -none- numeric
importance
importanceSD
                  930 -none- numeric
localImportance
                   0 -none- NULL
proximity
                 0 -none- NULL
ntree
              1 -none- numeric
               1 -none- numeric
mtry
              14 -none- list
forest
           61878 factor numeric
y
             0 -none- NULL
test
               0 -none- NULL
inbag
terms
               3 terms call
Hence, we have generated random forest using length class mode of 1 none character type.
summary(pred)
  Class 1
                 Class 2
                                                           Class 5
                              Class 3
                                            Class 4
```

Min. :0.00000 Min. :0.0000 Min. :0.0000 Min. :0.00000 Min. :0.00000 1st Qu.:0.00000 1st Qu.:0.00000 1st Qu.:0.00000 Median :0.0700 Median :0.0300 Median :0.01000 Median :0.00000

Mean :0.03112 Mean :0.2667 Mean :0.1288 Mean :0.04413 Mean :0.04469 3rd Qu.:0.03000 3rd Qu.:0.5200 3rd Qu.:0.2000 3rd Qu.:0.05000 3rd Qu.:0.01000 Max. :1.00000 Max. :1.0000 Max. :1.00000 Max. :0.00000 Max. :0.00000 Max. :0.00000 Max. :0.00000 Median :0.01000 Median :0.01000 Median :0.01000 Median :0.0264 Mean :0.04389 Mean :0.1344 Mean :0.07987 3rd Qu.:0.2100 3rd Qu.:0.04000 3rd Qu.:0.0700 3rd Qu.:0.04000 Max. :1.00000 Max. :1.00000 Max. :1.00000 Max. :1.00000 Max. :1.00000

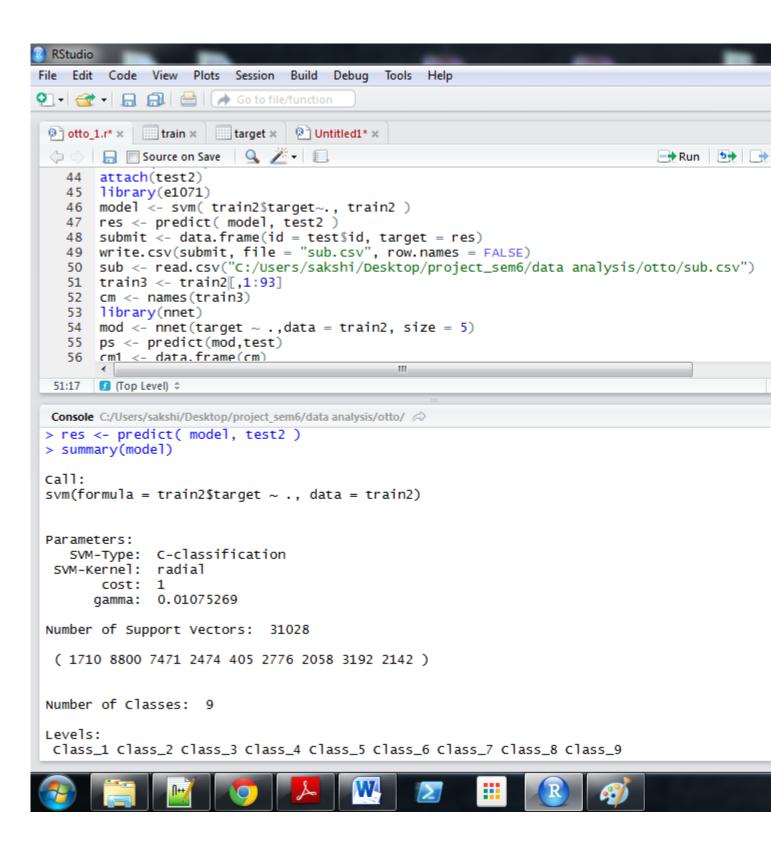
Mean and Median of all quartiles in each class is derived for the prediction.

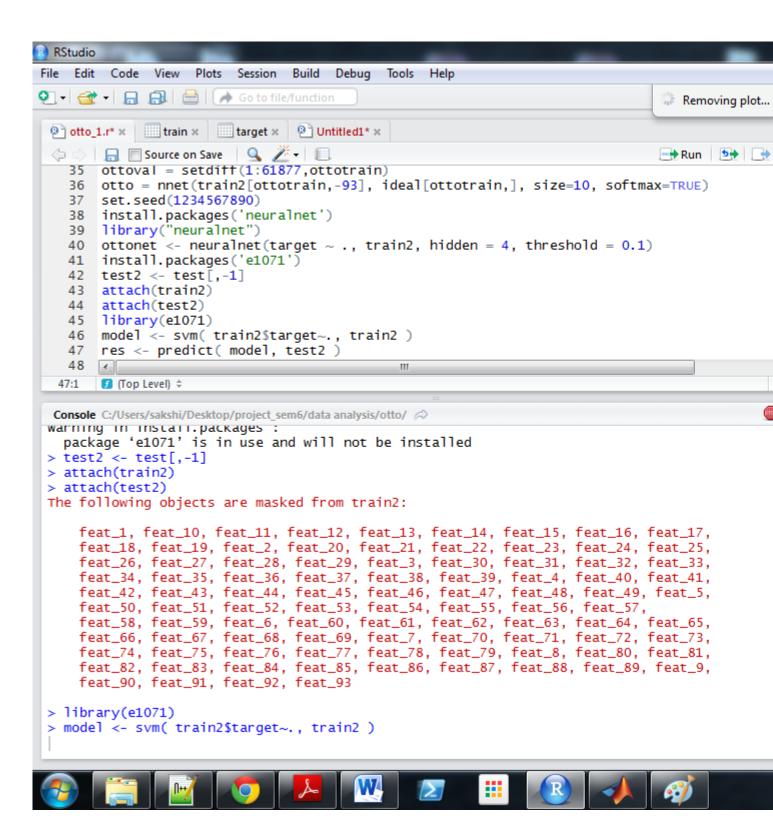


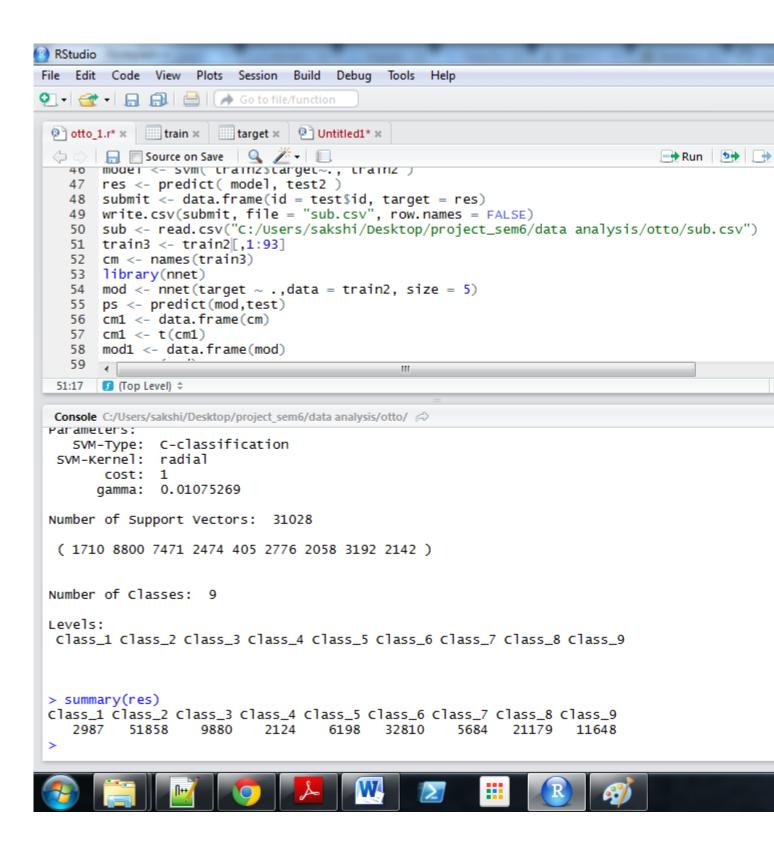


2. SUPPORT VECTOR MACHINE (SVM)

```
install.packages('e1071')
                                       # install SVM package
test2 < -test[,-1]
                                       # removing the id column
attach(train2)
attach(test2)
library(e1071)
                                         #include SVM library
model <- svm( train2$target~., train2 )
                                         # train the data model with class named 'target'
res <- predict( model, test2 )
                                                  # predict the classification for test data
  submit <- data.frame(id = test$id, target = res)</pre>
                                                      # create a data frame by re combining the result
                                                                                      with id column
write.csv(submit, file = "sub.csv", row.names = FALSE)
sub <- read.csv("C:/Users/sakshi/Desktop/project_sem6/data analysis/otto/sub.csv")
 summary(model)
Call:
svm(formula = train2$target ~ ., data = train2)
Parameters:
 SVM-Type: C-classification
SVM-Kernel: radial
    cost: 1
   gamma: 0.01075269
Number of Support Vectors: 31028
( 1710 8800 7471 2474 405 2776 2058 3192 2142 )
Number of Classes: 9
Levels:
Class_1 Class_2 Class_3 Class_4 Class_5 Class_6 Class_7 Class_8 Class_9
Hence, we have used C-classification type of sym with radial kernel for the classification and 31028
support vectors are derived.
summary(res)
Class_1 Class_2 Class_3 Class_4 Class_5 Class_6 Class_7 Class_8 Class_9
 2987 51858 9880 2124 6198 32810 5684 21179 11648
```







3. NEURAL NETWORKS

```
# include all columns except the class label column
train3 <- train2[,1:93]
                                            # get the names of all the columns
cm <- names(train3)
library(nnet)
                                                    # include the neural network library ie. nnet
  mod <- nnet(target ~ .,data = train2, size = 5)
                                                     # generate a neural network model with number
                                                             of hidden layers as 5 and train the data.
ps <- predict(mod,test)
                                                    # predict the test data with the model generated.
cm1 <- data.frame(cm)
cm1 <- t(cm1)
                                                     # take the transpose of the result
mod1 <- data.frame(mod)</pre>
summary(mod)
write.csv(ps1, file = "subneuralfin.csv", row.names = FALSE)
mod <- nnet(target \sim ., data = train2, size = 5)
# weights: 524
initial value 140521.042097
iter 10 value 64189.244074
iter 20 value 51283.502796
iter 30 value 49777.149148
iter 40 value 48315.724506
iter 50 value 46965.225870
iter 60 value 45911.909257
iter 70 value 45208.936264
iter 80 value 44719.998361
iter 90 value 44343.831812
iter 100 value 43675.211675
final value 43675.211675
stopped after 100 iterations
Hence, the training of neural network goes till 100 iterations in which our input matrix consist of
93x61487 dimension and target matrix of 9x61487 with 524 weights using softmax modelling.
a 93-5-9 network with 524 weights
options were - softmax modelling
 b->h1 i1->h1 i2->h1 i3->h1 i4->h1 i5->h1 i6->h1 i7->h1 i8->h1 i9->h1 i10->h1
 -4.42 0.40 -0.25 0.11 0.97 1.19 -0.31 2.82 -0.22 3.56 0.05
i11->h1 i12->h1 i13->h1 i14->h1 i15->h1 i16->h1 i17->h1 i18->h1 i19->h1 i20->h1 i21->h1
 -6.06 -2.82 3.56 1.57 3.64 -0.75 -5.78 1.22 0.87 -0.12 1.31
i22->h1 i23->h1 i24->h1 i25->h1 i26->h1 i27->h1 i28->h1 i29->h1 i30->h1 i31->h1 i32->h1
 0.41 -2.76 -0.73 2.98 -2.74 -2.85 0.87 -1.03 0.53 -1.20 -2.45
i33->h1 i34->h1 i35->h1 i36->h1 i37->h1 i38->h1 i39->h1 i40->h1 i41->h1 i42->h1 i43->h1
```

4.93 3.25 4.67 -1.97 0.94 1.89 -5.09 3.28 2.47 -2.36 -5.66

-2.30 -1.24 0.22 1.41 1.00 -2.58 -2.81 0.70 -0.96 -6.95 0.84

i44->h1 i45->h1 i46->h1 i47->h1 i48->h1 i49->h1 i50->h1 i51->h1 i52->h1 i53->h1 i54->h1

```
i55->h1 i56->h1 i57->h1 i58->h1 i59->h1 i60->h1 i61->h1 i62->h1 i63->h1 i64->h1 i65->h1
-3.20 -6.63 -3.73 3.68 -5.94 -6.27 -3.37 -0.37 1.35 -0.19 0.85
i66->h1 i67->h1 i68->h1 i69->h1 i70->h1 i71->h1 i72->h1 i73->h1 i74->h1 i75->h1 i76->h1
-2.39 -0.09 2.93 3.24 -0.53 1.83 2.31 0.57 -0.09 4.11 0.44
i77->h1 i78->h1 i79->h1 i80->h1 i81->h1 i82->h1 i83->h1 i84->h1 i85->h1 i86->h1 i87->h1
 -2.63 1.24 -4.13 2.13 -3.29 -3.92 0.61 0.57 -5.16 -1.14 -2.12
i88->h1 i89->h1 i90->h1 i91->h1 i92->h1 i93->h1
-2.26 3.05 6.43 -1.21 0.54 -1.51
 b->h2 i1->h2 i2->h2 i3->h2 i4->h2 i5->h2 i6->h2 i7->h2 i8->h2 i9->h2 i10->h2
-12.77 -5.37 4.00 0.67 0.19 -8.69 -5.41 -4.40 -9.55 2.18 -2.43
i11->h2 i12->h2 i13->h2 i14->h2 i15->h2 i16->h2 i17->h2 i18->h2 i19->h2 i20->h2 i21->h2
 8.45 -1.88 1.48 -6.23 -1.36 0.37 0.93 1.34 5.39 1.94 1.43
i22->h2 i23->h2 i24->h2 i25->h2 i26->h2 i27->h2 i28->h2 i29->h2 i30->h2 i31->h2 i32->h2
 i33->h2 i34->h2 i35->h2 i36->h2 i37->h2 i38->h2 i39->h2 i40->h2 i41->h2 i42->h2 i43->h2
-3.21 -1.51 3.65 -1.28 -4.22 2.10 10.98 -14.01 -0.31 8.54 -2.51
i44->h2 i45->h2 i46->h2 i47->h2 i48->h2 i49->h2 i50->h2 i51->h2 i52->h2 i53->h2 i54->h2
-0.02 3.21 0.32 10.31 -2.55 3.25 6.46 -0.42 -0.89 -3.34 0.68
i55->h2 i56->h2 i57->h2 i58->h2 i59->h2 i60->h2 i61->h2 i62->h2 i63->h2 i64->h2 i65->h2
 1.50 -3.75 6.48 4.49 -0.17 12.85 -0.49 -6.17 2.17 -7.33 0.32
i66->h2 i67->h2 i68->h2 i69->h2 i70->h2 i71->h2 i72->h2 i73->h2 i74->h2 i75->h2 i76->h2
-3.34 -0.86 -10.30 3.70 0.11 1.40 -13.06 4.08 6.08 3.77 -2.37
i77->h2 i78->h2 i79->h2 i80->h2 i81->h2 i82->h2 i83->h2 i84->h2 i85->h2 i86->h2 i87->h2
-8.46 -6.34 -4.50 2.06 2.40 1.20 4.01 -12.70 -7.90 0.84 2.14
i88->h2 i89->h2 i90->h2 i91->h2 i92->h2 i93->h2
 -4.50 -1.89 2.82 -0.73 -0.84 5.02
b->h3 i1->h3 i2->h3 i3->h3 i4->h3 i5->h3 i6->h3 i7->h3 i8->h3 i9->h3 i10->h3
 0.15  0.44  -3.23  2.28  -0.09  -0.73  -3.09  2.68  0.43  -2.66  -1.04
i11->h3 i12->h3 i13->h3 i14->h3 i15->h3 i16->h3 i17->h3 i18->h3 i19->h3 i20->h3 i21->h3
 2.66 -0.03 -0.50 -2.89 -3.34 -0.07 2.84 0.64 0.31 0.78 2.43
i22->h3 i23->h3 i24->h3 i25->h3 i26->h3 i27->h3 i28->h3 i29->h3 i30->h3 i31->h3 i32->h3
 1.07 -1.39 0.49 -2.63 0.83 1.18 4.94 2.83 0.02 4.90 -1.71
i33->h3 i34->h3 i35->h3 i36->h3 i37->h3 i38->h3 i39->h3 i40->h3 i41->h3 i42->h3 i43->h3
-1.20 -4.93 1.47 1.06 0.51 -0.20 -3.98 -4.56 7.18 0.67 -13.15
i44->h3 i45->h3 i46->h3 i47->h3 i48->h3 i49->h3 i50->h3 i51->h3 i52->h3 i53->h3 i54->h3
 i55->h3 i56->h3 i57->h3 i58->h3 i59->h3 i60->h3 i61->h3 i62->h3 i63->h3 i64->h3 i65->h3
 2.14 -9.53 3.24 3.92 5.27 2.71 3.79 3.19 -1.98 -0.88 0.49
i66->h3 i67->h3 i68->h3 i69->h3 i70->h3 i71->h3 i72->h3 i73->h3 i74->h3 i75->h3 i76->h3
 -0.52 1.35 3.31 1.23 -1.17 1.76 -2.07 1.80 1.73 3.09 2.70
i77->h3 i78->h3 i79->h3 i80->h3 i81->h3 i82->h3 i83->h3 i84->h3 i85->h3 i86->h3 i87->h3
 0.02 3.12 -0.33 -1.43 1.89 5.28 -0.36 1.66 0.00 -3.39 -1.17
i88->h3 i89->h3 i90->h3 i91->h3 i92->h3 i93->h3
-2.00 0.28 6.24 1.60 3.53 3.97
b->h4 i1->h4 i2->h4 i3->h4 i4->h4 i5->h4 i6->h4 i7->h4 i8->h4 i9->h4 i10->h4
-2.22 -0.12 1.11 -0.30 0.15 -2.73 -6.02 -1.49 0.21 -0.79 -0.66
i11->h4 i12->h4 i13->h4 i14->h4 i15->h4 i16->h4 i17->h4 i18->h4 i19->h4 i20->h4 i21->h4
 2.55 0.21 0.09 0.35 1.83 0.27 -1.71 0.67 -1.40 -1.42 1.38
i22->h4 i23->h4 i24->h4 i25->h4 i26->h4 i27->h4 i28->h4 i29->h4 i30->h4 i31->h4 i32->h4
```

```
0.04 -2.47 -0.53 1.89 5.04 0.64 -1.63 1.00 1.18 -0.01 -1.12
i33->h4 i34->h4 i35->h4 i36->h4 i37->h4 i38->h4 i39->h4 i40->h4 i41->h4 i42->h4 i43->h4
 0.17 \ -1.96 \ -2.06 \ 0.71 \ -2.72 \ -0.48 \ -1.79 \ 0.23 \ 0.96 \ 2.63 \ 0.98
i44->h4 i45->h4 i46->h4 i47->h4 i48->h4 i49->h4 i50->h4 i51->h4 i52->h4 i53->h4 i54->h4
 0.13 -1.80 0.49 -0.84 0.29 0.75 -1.79 0.29 -1.60 -0.11 0.53
i55->h4 i56->h4 i57->h4 i58->h4 i59->h4 i60->h4 i61->h4 i62->h4 i63->h4 i64->h4 i65->h4
-1.40 -2.48 1.62 -0.35 0.48 4.25 2.20 0.16 0.68 0.09 -1.18
i66->h4 i67->h4 i68->h4 i69->h4 i70->h4 i71->h4 i72->h4 i73->h4 i74->h4 i75->h4 i76->h4
 i77->h4 i78->h4 i79->h4 i80->h4 i81->h4 i82->h4 i83->h4 i84->h4 i85->h4 i86->h4 i87->h4
-0.91 -2.98 0.17 0.39 -0.55 0.13 -0.55 -2.57 0.37 -0.12 0.73
i88->h4 i89->h4 i90->h4 i91->h4 i92->h4 i93->h4
 0.71 -0.79 -1.71 1.70 1.43 2.14
 b->h5 i1->h5 i2->h5 i3->h5 i4->h5 i5->h5 i6->h5 i7->h5 i8->h5 i9->h5 i10->h5
 -0.13 -1.09 1.65 -1.01 -0.47 -1.18 -0.60 -1.69 0.34 3.72 0.30
i11->h5 i12->h5 i13->h5 i14->h5 i15->h5 i16->h5 i17->h5 i18->h5 i19->h5 i20->h5 i21->h5
-1.74 0.55 0.00 0.34 -0.01 0.70 -0.03 0.04 -2.03 -1.79 0.43
i22->h5 i23->h5 i24->h5 i25->h5 i26->h5 i27->h5 i28->h5 i29->h5 i30->h5 i31->h5 i32->h5
 -0.49 -3.95 0.05 -0.12 -9.46 -0.46 -0.34 1.79 2.52 -1.37 -1.33
i33->h5 i34->h5 i35->h5 i36->h5 i37->h5 i38->h5 i39->h5 i40->h5 i41->h5 i42->h5 i43->h5
 0.33 -1.47 1.44 1.77 0.14 -0.58 0.25 0.61 1.16 -5.20 2.96
i44->h5 i45->h5 i46->h5 i47->h5 i48->h5 i49->h5 i50->h5 i51->h5 i52->h5 i53->h5 i54->h5
-1.46 -0.57 0.81 -2.54 1.11 -1.23 -0.77 -1.71 2.25 2.58 -0.10
i55->h5 i56->h5 i57->h5 i58->h5 i59->h5 i60->h5 i61->h5 i62->h5 i63->h5 i64->h5 i65->h5
 0.66 -0.30 -5.14 -1.85 3.33 -4.60 -2.43 1.79 0.40 0.14 0.97
i66->h5 i67->h5 i68->h5 i69->h5 i70->h5 i71->h5 i72->h5 i73->h5 i74->h5 i75->h5 i76->h5
-0.57 0.11 -1.20 -2.29 0.02 -1.20 0.45 0.01 -0.81 -0.44 -0.73
i77->h5 i78->h5 i79->h5 i80->h5 i81->h5 i82->h5 i83->h5 i84->h5 i85->h5 i86->h5 i87->h5
 6.41 -4.03 -0.72 -0.55 -1.97 0.37 3.81 -3.97 0.52 0.48 -1.51
i88->h5 i89->h5 i90->h5 i91->h5 i92->h5 i93->h5
 0.23 1.81 -1.79 1.30 -0.99 0.42
b->01 h1->01 h2->01 h3->01 h4->01 h5->01
-0.41 -1.85 -0.95 3.69 -0.76 0.68
b->o2 h1->o2 h2->o2 h3->o2 h4->o2 h5->o2
0.62 0.96 -0.56 -2.43 1.50 1.88
b->o3 h1->o3 h2->o3 h3->o3 h4->o3 h5->o3
0.90 0.03 -0.68 -3.33 0.61 2.38
b->o4 h1->o4 h2->o4 h3->o4 h4->o4 h5->o4
0.69 -0.38 -1.03 -3.11 2.93 -0.75
b->o5 h1->o5 h2->o5 h3->o5 h4->o5 h5->o5
-0.26 5.15 -1.46 -4.66 -4.25 -4.47
b->o6 h1->o6 h2->o6 h3->o6 h4->o6 h5->o6
0.27 -2.13 2.58 1.67 2.39 -1.58
b->07 h1->07 h2->07 h3->07 h4->07 h5->07
2.93 -2.86 1.72 -1.08 -1.52 -0.07
b->08 h1->08 h2->08 h3->08 h4->08 h5->08
-0.54 1.59 1.31 3.66 -2.21 -0.64
b->09 h1->09 h2->09 h3->09 h4->09 h5->09
-3.92 -0.80 -1.66 5.59 1.56 2.67
```

summary(ps)

Class 1 Class 2 Class 3 Class 4 Min. :0.0003175 Min. :0.0004027 Min. :0.0000802 Min. :0.0000436 1st Qu.:0.0007202 1st Qu.:0.0005650 1st Qu.:0.0001154 1st Qu.:0.0005764 Median: 0.0046876 Median: 0.0331889 Median: 0.0174847 Median: 0.0092618 Mean :0.0275962 Mean :0.2562176 Mean :0.1332563 Mean :0.0458221 3rd Qu.:0.0109750 3rd Qu.:0.6104054 3rd Qu.:0.2462515 3rd Qu.:0.0583670 Max. :0.4460607 Max. :0.7104585 Max. :0.5125156 Max. :0.5673775 Class 5 Class 6 Class 8 Class 7 Min. :0.0000000 Min. :0.0004977 Min. :0.0003894 Min. :0.0002611 1st Qu.:0.0000001 1st Qu.:0.0016959 1st Qu.:0.0021075 1st Qu.:0.0008596 Median: 0.0001054 Median: 0.0217462 Median: 0.0073123 Median: 0.0090453 Mean :0.0430220 Mean :0.2294358 Mean :0.0462082 Mean :0.1365114 3rd Qu.:0.0006822 3rd Qu.:0.1708350 3rd Qu.:0.0154313 3rd Qu.:0.0233542 Max. :0.9113590 Max. :0.9739285 Max. :0.8554974 Max. :0.9658130 Class 9

Min. :0.0000296 1st Qu.:0.0022342 Median :0.0033974 Mean :0.0819304 3rd Qu.:0.0097583 Max. :0.8958361

For each class mean and median of each quartile is derived after the prediction.

