Developing models to predict cytotoxicity of metal nanoparticles

TXT

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Setup R packages for model development

Loading required package: sandwich

After installing R and RStudio, open RStudio and run the following codes to install and load necessary packages for developing models and making web applications::

```
packages <- c("svDialogs", "data.table", "data.table", "openxlsx", "mlbench", "caret",</pre>
              "tools", "DT", "magrittr", "ggplot2", "nortest", "tseries", "stringr",
              "RcmdrMisc", "lmtest", "dplyr", "randomForest", "shiny", "shinydashboard")
install.packages(setdiff(packages, rownames(installed.packages())))
library(svDialogs)
library(data.table)
library(openxlsx)
library(mlbench)
library(caret)
## Loading required package: ggplot2
## Loading required package: lattice
library(tools)
library(DT)
library(ggplot2)
library(car)
## Loading required package: carData
library(nortest)
library(tseries)
## Registered S3 method overwritten by 'quantmod':
##
    method
                      from
     as.zoo.data.frame zoo
library(RcmdrMisc)
```

```
library(lmtest)
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
       as.Date, as.Date.numeric
library(dplyr)
##
## Attaching package: 'dplyr'
## The following object is masked from 'package:car':
##
##
       recode
## The following objects are masked from 'package:data.table':
##
       between, first, last
##
## The following objects are masked from 'package:stats':
##
       filter, lag
##
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
library(randomForest)
## randomForest 4.6-14
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:dplyr':
##
##
       combine
## The following object is masked from 'package:ggplot2':
##
##
       margin
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```

Choose excel file containing dataset and read the dataset:

```
DataMetal <- read.xlsx("MetalESN.xlsx", sheet = 1, startRow = 1, colNames = TRUE,
                        rowNames = FALSE, detectDates = FALSE, skipEmptyRows = TRUE,
                        skipEmptyCols = TRUE, rows = NULL, cols = NULL,
                        check.names = FALSE, namedRegion = NULL,
                        na.strings = "NA", fillMergedCells = FALSE)
DataMetal <- select(DataMetal, c("Toxicity",</pre>
                                  "Dose",
                                  "Assay",
                                  "Time",
                                  "Species",
                                  "Cancer",
                                  "Cell_Tissue",
                                  "Cell_line",
                                  "SSA",
                                  "Zeta",
                                  "HSize",
                                  "CoreSize",
                                  "Coating",
                                  "Shape",
                                  "Metal"))
```

Split data into training and test set (70/30)

```
set.seed(1991)
split_size <- floor(0.70 * nrow(DataMetal))
in_rows <- sample(c(1:nrow(DataMetal)), size = split_size, replace = FALSE)
train <- DataMetal[in_rows, ]
test <- DataMetal[-in_rows, ]</pre>
```

Train Random Forest model:

```
train.control <- trainControl(method = "repeatedcv", number = 10, repeats = 5)
RFmodel <- train(Toxicity ~ ., data = train, method = "rf", ntree = 100, trControl = train.control)
print(RFmodel)</pre>
```

```
## Random Forest
##
## 1403 samples
##
    14 predictor
      2 classes: 'NON_TOXIC', 'TOXIC'
##
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 5 times)
## Summary of sample sizes: 1262, 1263, 1263, 1263, 1263, 1262, ...
## Resampling results across tuning parameters:
##
##
    mtry Accuracy
                     Kappa
##
      2
          0.8238213 0.1845892
          0.9154577 0.7138967
##
     68
          0.9158772 0.7215603
##
     134
##
```

```
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was mtry = 134.
Use RFmodel to predict test set:
predictions <- RFmodel %>% predict(test); predictions_train <- RFmodel %>% predict(train)
Get confusion matrix and performance of model:
CMatrix <- confusionMatrix(predictions, as.factor(test$Toxicity))</pre>
Performance <- data.frame(Parameter = row.names(as.data.frame(CMatrix$byClass)),
                          Value = as.data.frame(CMatrix$byClass))
colnames(Performance) <- c("Parameters", "Values")</pre>
## Confusion Matrix and Statistics
##
##
              Reference
## Prediction NON_TOXIC TOXIC
     NON TOXIC
                     480
                             25
##
     TOXIC
##
                      14
                             83
##
##
                  Accuracy: 0.9352
##
                    95% CI: (0.9125, 0.9535)
##
       No Information Rate: 0.8206
##
       P-Value [Acc > NIR] : <2e-16
##
##
                      Kappa: 0.7709
##
##
    Mcnemar's Test P-Value: 0.1093
##
               Sensitivity: 0.9717
##
##
               Specificity: 0.7685
            Pos Pred Value: 0.9505
##
            Neg Pred Value: 0.8557
##
                Prevalence: 0.8206
##
##
            Detection Rate: 0.7973
      Detection Prevalence: 0.8389
##
##
         Balanced Accuracy: 0.8701
##
##
          'Positive' Class : NON_TOXIC
##
Performance
##
                                   Parameters
                                                 Values
## Sensitivity
                                  Sensitivity 0.9716599
```

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Recall 0.9716599

Specificity 0.7685185

Precision 0.9504950

Pos Pred Value 0.9504950

Neg Pred Value 0.8556701

Specificity

Precision

Recall

Pos Pred Value

Neg Pred Value

```
## F1 F1 0.9609610
## Prevalence Prevalence 0.8205980
## Detection Rate Detection Rate 0.7973422
## Detection Prevalence Detection Prevalence 0.8388704
## Balanced Accuracy Balanced Accuracy 0.8700892
```

Save data and models for later use in web application:

```
save(RFmodel, file = "RFmodel.RData")
```