Package 'PatientLevelPrediction'

April 3, 2017

```
Type Package
Title Package for patient level prediction using data in the OMOP Common Data
Version 1.2.0
Date 2016-11-17
Author Jenna Reps [aut],
      Martijn J. Schuemie [aut, cre],
      Marc A. Suchard [aut],
      Patrick B. Ryan [aut],
      Peter R. Rijnbeek [aut]
Maintainer Jenna Reps < reps@ohdsi.org>
Description A package for creating patient level prediction models. Given a
      cohort of interest and an outcome of interest, the package can use data in the
      OMOP Common Data Model to build a large set of features. These features can then
      be assessed to fit a predictive model using a number of machine learning algorithms.
      Several performance measures are implemented for model evaluation.
License Apache License 2.0
Depends R (>= 3.2.2),
      DatabaseConnector (>= 1.3.0),
      Cyclops (>= 1.2.1-2)
Imports ggplot2,
      gridExtra,
      bit,
      ff,
      ffbase (>= 0.12.1),
      plyr,
      survAUC,
      Rcpp (>= 0.11.2),
      RJDBC,
      SqlRender (\geq 1.1.3),
      survival.
      FeatureExtraction,
      xgboost,
      Matrix,
      AUC,
```

PythonInR, futile.options,

2 R topics documented:

futile.logger,	
utils,	
methods,	
BigKnn,	
reshape2	
Suggests testthat,	
pROC,	
gnm,	
knitr,	
rmarkdown,	
scoring,	
Metrics,	
SparseM,	
ResourceSelect	ior
LinkingTo Repp	
NeedsCompilation y	es
RoxygenNote 6.0.1	

R topics documented:

accuracy
applyModel
averagePrecision
brierScore
bySumFf
calibrationLine
computeAuc
computeAucFromDataFrames
createStudyPopulation
diagnosticOddsRatio
evaluatePlp
$exportPlpDataToCsv \\ \ . \\ \ . \\ \ 1$
f1Score
$false Discovery Rate \ \dots \ \dots \ \ 1$
$false Negative Rate \ . \ . \ . \ . \ . \ . \ . \ . \ . \ $
falseOmissionRate
falsePositiveRate
$fitGLMModel \ . \ . \ . \ . \ . \ . \ . \ . \ . \ $
$fit Plp \dots \dots$
getAttritionTable
getCalibration
getModelDetails
getPlpData
getPredictionDistribution
getThresholdSummary
grepCovariateNames
insertDbPopulation
loadPlpData
$loadPlpModel \dots \dots$
loadPlnResult 2

R	topics	documented:
---	--------	-------------

loadPrediction 24 negativeLikelihoodRatio 24 negativePredictiveValue 25 PatientLevelPrediction 25 personSplitter 26 plotFleediction 26 plotFl Measure 27 plotGeneralizability 27 plotPplp 28 plotPrecisionRecall 28 plotPredictionDistribution 29 plotPredictionDistribution 29 plotPredictionDistribution 29 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration 31 plotSparseRoc 32 plotVariableScatterplot 33 ppositiveLikelihoodRatio 34 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfd 35 predictPlobabilities 36 runPlpAnalyses 37 savePlpAnal 39 savePlpAnalyses 37 savePlpAndel 39 <	
negativePredictiveValue 25 PatientLevelPrediction 25 personSplitter 26 plotDemographicSummary 26 plotFl Measure 27 plotGeneralizability 27 plotPlp 28 plotPredictionSidecall 28 plotPredictionDistribution 29 plotPredictionDistribution 29 plotPredictionDistribution 30 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration 31 plotSparseCalibration2 32 plotVariableScatterplot 33 plotVariableScatterplot 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfdf 35 predictPlpAnalyses 37 savePlpAnalyses 37 savePlpModel 39 savePlpModel 39 savePlpModel 39 savePlpCation 40 sensitivity 41	loadPrediction
PatientLevelPrediction 25 personSplitter 26 plotDemographicSummary 26 plotFl Measure 27 plotPlp 28 plotPrecisionRecall 28 plotPrecisionRecall 28 plotPredictedPDF 29 plotPredictionDistribution 29 plotPredictionDistribution 29 plotPredictionDistribution 31 plotSparseColibration 31 plotSparseCalibration 31 plotSparseCalibration 32 plotVariableScatterplot 33 plotVariableScatterplot 33 ppositiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpModel 39 savePlpModel 39 savePlpModel 39 savePlpModel 39 savePlpModel 39 sa	negativeLikelihoodRatio
personSplitter 26 plotDemographicSummary 26 plotFI Measure 27 plotGeneralizability 27 plotPlp 28 plotPredictionDistribution 28 plotPredictionDistribution 29 plotPredictionDistribution 29 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration 31 plotSparseRoc 32 plotVariableScatterplot 33 plotVariableScatterplot 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfd 35 predictPfd 35 predictPfoabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpModel 39 savePrediction 40 sexifivity 41 setKNN 42 setKNN 42 setKNN 42	negativePredictiveValue
plotDemographicSummary 26 plotF1Measure 27 plotGeneralizability 27 plotPlp 28 plotPrecisionRecall 28 plotPredictedPDF 29 plotPredictionDistribution 29 plotPreferencePDF 30 plotRoc 31 plotSparseCalibration 31 plotSparseRoc 32 plotVariableScatterplot 33 plotSparseRoc 32 plotVariableScatterplot 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfd 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setKNN 42 setLassoLogisticRegression 42 setNaiveBayes 43 setRandomForest 44 <td>PatientLevelPrediction</td>	PatientLevelPrediction
plotF1Measure 27 plotGeneralizability 27 plotPlp 28 plotPrecisionRecall 28 plotPredictedPDF 29 plotPredictionDistribution 29 plotPreferencePDF 30 plotSparseCalibration 31 plotSparseCalibration2 32 plotVariableScatterplot 33 plotVariableScatterplot 33 plotAsaSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpModel 39 savePlpBosult 40 sextInsoLogisticRegression 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setNaiveBayes 43 setNaiveBayes 43 setNaiveBayes <td>personSplitter</td>	personSplitter
plotGeneralizability 27 plotPlp 28 plotPrecisionRecall 28 plotPredictionDistribution 29 plotPreferencePDF 30 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfdf 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpResult 40 savePrediction 40 sensitivity 41 setCardientBoostingMachine 41 setLxNN 42 setMLP 43 setNaiveBayes 43 setNaiveBayes 43 setNaiveBayes 43 setNaiveBayes 43 setNaiveBayes 43 <	plotDemographicSummary
plotPlp 28 plotPrecisionRecall 28 plotPredictedPDF 29 plotPredictionDistribution 29 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpModel 39 savePlpModel 39 savePlpResult 40 sevePrediction 40 sensitivity 41 setCardientBoostingMachine 41 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setNaiveBayes 43 setNaiveBayes 43 setNaiveBayes 43 <td>plotF1Measure</td>	plotF1Measure
plotPrecisionRecall 28 plotPredictedPDF 29 plotPredictionDistribution 29 plotPreferencePDF 30 plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfdf 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 sensitivity 41 setKNN 42 setKasoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45 simulatePlpData 46	plotGeneralizability
plotPredictedPDF 29 plotPredictionDistribution 29 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpResult 40 savePrediction 40 sensitivity 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 46	plotPlp
plotPredictionDistribution 29 plotPreferencePDF 30 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 psitiveLikelinoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45	plotPrecisionRecall
plotPreferencePDF 30 plotRoc 31 plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plotAsimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictPfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45	plotPredictedPDF
plotRoc 31 plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plotVariableScatterplot 33 pplotAaSimulationProfile 33 positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45 simulatePlpData 46	plotPredictionDistribution
plotSparseCalibration 31 plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45 simulatePlpData 46	plotPreferencePDF
plotSparseCalibration2 32 plotSparseRoc 32 plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 sevePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setRandomForest 44 setTimeAtRisk 44 similatePlpData 45 simulatePlpData 46	plotRoc
plotSparseRoc 32 plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	plotSparseCalibration
plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45 simulatePlpData 46	plotSparseCalibration2
plotVariableScatterplot 33 plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45 simulatePlpData 46	plotSparseRoc
plpDataSimulationProfile 33 positiveLikelihoodRatio 34 positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setNaiveBayes 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	
positivePredictiveValue 34 predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	
predictFfdf 35 predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	positiveLikelihoodRatio
predictPlp 35 predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	positivePredictiveValue
predictProbabilities 36 runPlpAnalyses 37 savePlpData 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setAssoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	predictFfdf
runPlpAnalyses 37 savePlpData 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 simulatePlpData 45 simulatePlpData 46	predictPlp
savePlpData 39 savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	predictProbabilities
savePlpModel 39 savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	runPlpAnalyses
savePlpResult 40 savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	savePlpData
savePrediction 40 sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	
sensitivity 41 setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	•
setGradientBoostingMachine 41 setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	savePrediction
setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	sensitivity
setKNN 42 setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	setGradientBoostingMachine
setLassoLogisticRegression 42 setMLP 43 setNaiveBayes 43 setRandomForest 44 setTimeAtRisk 44 similarPlpData 45 simulatePlpData 46	
setNaiveBayes43setRandomForest44setTimeAtRisk44similarPlpData45simulatePlpData46	
setRandomForest44setTimeAtRisk44similarPlpData45simulatePlpData46	setMLP
setRandomForest44setTimeAtRisk44similarPlpData45simulatePlpData46	setNaiveBayes
setTimeAtRisk44similarPlpData45simulatePlpData46	•
simulatePlpData	
simulatePlpData	similarPlpData
	•
	specificity
timeSplitter	
•	toSparseM
toSparseM	
toSparseM	-

4 applyModel

acci	ıra	CV

Calculate the accuracy

Description

Calculate the accuracy

Usage

```
accuracy(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the accuracy

Value

accuracy value

apply	yModel
-------	--------

Apply train model on new data Apply a Patient Level Prediction model on Patient Level Prediction Data and get the predicted risk in [0,1] for each person in the population. If the user inputs a population with an outcome Count column then the function also returns the evaluation of the prediction (AUC, brier score, calibration)

Description

Apply train model on new data Apply a Patient Level Prediction model on Patient Level Prediction Data and get the predicted risk in [0,1] for each person in the population. If the user inputs a population with an outcomeCount column then the function also returns the evaluation of the prediction (AUC, brier score, calibration)

Usage

```
applyModel(population, plpData, plpModel, logConnection = NULL,
  databaseOutput = NULL, silent = F)
```

averagePrecision 5

Arguments

population The population of people who you want to predict the risk for

plpData The plpData for the population

plpModel The trained PatientLevelPrediction model

logConnection A connection to output any logging during the process databaseOutput Whether to save the details into the prediction database

silent Whether to turn off progress reporting

Examples

```
## Not run:
# load the model and data
plpData <- loadPlpData("C:/plpdata")
plpModel <- loadPlpModel("C:/plpmodel")

# use the same population settings as the model:
populationSettings <- plpModel$populationSettings
populationSettings$plpData <- plpData
population <- do.call(createStudyPopulation, populationSettings)

# get the prediction:
prediction <- applyModel(population, plpData, plpModel)$prediction

## End(Not run)</pre>
```

 $average \hbox{Precision}$

Calculate the average precision

Description

Calculate the average precision

Usage

```
averagePrecision(prediction)
```

Arguments

prediction A prediction object as generated using the predictProbabilities function.

Details

Calculates the average precision from a predition object

Value

The average precision

6 bySumFf

brierScore

brierScore

Description

brierScore

Usage

```
brierScore(prediction)
```

Arguments

prediction

A prediction object as generated using the predictProbabilities function.

Details

Calculates the brierScore from prediction object

Value

A list containing the brier score and the scaled brier score

bySumFf

Compute sum of values binned by a second variable

Description

Compute sum of values binned by a second variable

Usage

```
bySumFf(values, bins)
```

Arguments

values An ff object containing the numeric values to be summed bins An ff object containing the numeric values to bin by

Examples

```
values <- ff::as.ff(c(1, 1, 2, 2, 1))
bins <- ff::as.ff(c(1, 1, 1, 2, 2))
bySumFf(values, bins)
```

calibrationLine 7

calibrationLine calibrationLine

Description

calibrationLine

Usage

```
calibrationLine(prediction, numberOfStrata = 10)
```

Arguments

prediction A prediction object as generated using the predictProbabilities function.

numberOfStrata The number of groups to split the prediction into

Details

Calculates the calibration from prediction object

computeAuc

Compute the area under the ROC curve

Description

Compute the area under the ROC curve

Usage

```
computeAuc(prediction, confidenceInterval = FALSE)
```

Arguments

```
\label{eq:prediction} \mbox{A prediction object as generated using the $\operatorname{\textit{predict}}$ functions.} \\ \mbox{confidenceInterval}
```

Should 95 percebt confidence intervals be computed?

Details

Computes the area under the ROC curve for the predicted probabilities, given the true observed outcomes.

computeAucFromDataFrames

Compute the area under the ROC curve

Description

Compute the area under the ROC curve

Usage

```
computeAucFromDataFrames(prediction, status, time = NULL,
  confidenceInterval = FALSE, timePoint, modelType = "logistic")
```

Arguments

prediction A vector with the predicted hazard rate.

status A vector with the status of 1 (event) or 0 (no event).

time Only for survival models: a vector with the time to event or censor (which ever

comes first).

confidenceInterval

Should 95 percebt confidence intervals be computed?

timePoint Only for survival models: time point when the AUC should be evaluated

modelType Type of model. Currently supported are "logistic" and "survival".

Details

Computes the area under the ROC curve for the predicted probabilities, given the true observed outcomes.

createStudyPopulation Create a study population

Description

Create a study population

Usage

```
createStudyPopulation(plpData, population = NULL, outcomeId, binary = T,
  includeAllOutcomes = T, firstExposureOnly = FALSE, washoutPeriod = 0,
  removeSubjectsWithPriorOutcome = TRUE, priorOutcomeLookback = 99999,
  requireTimeAtRisk = T, minTimeAtRisk = 365, riskWindowStart = 0,
  addExposureDaysToStart = FALSE, riskWindowEnd = 365,
  addExposureDaysToEnd = F, verbosity = futile.logger::INFO, ...)
```

createStudyPopulation 9

Arguments

plpData An object of type plpData as generated using getDbplpData.

population If specified, this population will be used as the starting point instead of the co-

horts in the plpData object.

outcomeId The ID of the outcome. If not specified, no outcome-specific transformations

will be performed.

binary Forces the outcomeCount to be 0 or 1 (use for binary prediction problems)

includeAllOutcomes

(binary) indicating whether to include people with outcomes who are not ob-

served for the whole at risk period

firstExposureOnly

Should only the first exposure per subject be included? Note that this is typically

done in the createStudyPopulation function,

washoutPeriod The mininum required continuous observation time prior to index date for a

person to be included in the cohort.

removeSubjectsWithPriorOutcome

Remove subjects that have the outcome prior to the risk window start?

priorOutcomeLookback

How many days should we look back when identifying prior outcomes?

requireTimeAtRisk

Should subject without time at risk be removed?

minTimeAtRisk The minimum number of days at risk required to be included

riskWindowStart

The start of the risk window (in days) relative to the index date (+ days of expo-

 $sure\ if\ the\ add {\tt ExposureDaysToStart}\ parameter\ is\ specified).$

addExposureDaysToStart

Add the length of exposure the start of the risk window?

riskWindowEnd The end of the risk window (in days) relative to the index data (+ days of expo-

sure if the addExposureDaysToEnd parameter is specified).

addExposureDaysToEnd

Add the length of exposure the risk window?

verbosity Sets the level of the verbosity. If the log level is at or higher in priority than the

logger threshold, a message will print. The levels are:

• DEBUGHighest verbosity showing all debug statements

- TRACEShowing information about start and end of steps
- INFOShow informative information (Default)
- · WARNShow warning messages
- ERRORShow error messages
- FATALBe silent except for fatal errors

... Other inputs

Details

Create a study population by enforcing certain inclusion and exclusion criteria, defining a risk window, and determining which outcomes fall inside the risk window.

10 diagnosticOddsRatio

Value

A data frame specifying the study population. This data frame will have the following columns:

rowId A unique identifier for an exposure

subjectId The person ID of the subject

cohortStartdate The index date

outcomeCount The number of outcomes observed during the risk window

timeAtRisk The number of days in the risk window

survivalTime The number of days until either the outcome or the end of the risk window

diagnosticOddsRatio

Calculate the diagnostic odds ratio

Description

Calculate the diagnostic odds ratio

Usage

```
diagnosticOddsRatio(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the diagnostic odds ratio

Value

diagnosticOddsRatio value

evaluatePlp 11

Description

Evaluates the performance of the patient level prediction model

Usage

```
evaluatePlp(prediction, plpData)
```

Arguments

prediction The patient level prediction model's prediction

plpData The patient level prediction data

Details

The function calculates various metrics to measure the performance of the model

Value

A list containing the performance values

Description

Export all data in a plpData object to CSV files

Usage

```
exportPlpDataToCsv(plpData, outputFolder)
```

Arguments

plpData An object of type plpData.

outputFolder The folder on the file system where the CSV files will be created. If the folder

does not yet exist it will be created.

12 f1Score

Details

Created a set of CSV files in the output folder with all the data in the plplData object. This function is intended to be used for research into prediction methods. The following files will be created:

cohort.csv Listing all persons and their prediction periods. This file will have these fields: row_id (a unique ID per period), person_id, cohort_start_date, cohort_id, time (number of days in the window).

outcomes.csv Listing all outcomes per period. This file will have these fields: row_id, outcome_id, outcome_count, time_to_event.

exclude.csv Either not exported or a file listing per outcome ID which windows had the outcome prior to the window and should therefore be removed prior to fitting the model. This object will have these fields: rowId, outcomeId.

covariates.csv Listing the baseline covariates per person in the cohorts. This is done using a sparse representation: covariates with a value of 0 are omitted to save space. The covariates file will have three columns: rowId, covariateId, and covariateValue.

covariateRef.csv A file describing the covariates that have been extracted. **metaData** Some information on how the plpData object was constructed.

Examples

```
## Not run:
exportPlpDataToCsv(plpData, "s:/temp/exportTest")
## End(Not run)
```

f1Score

Calculate the f1Score

Description

Calculate the f1Score

Usage

```
f1Score(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the f1Score

Value

f1Score value

falseDiscoveryRate 13

falseDiscoveryRate	Calculate the	falseDiscoveryRate

Description

Calculate the falseDiscoveryRate

Usage

```
falseDiscoveryRate(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the falseDiscoveryRate

Value

falseDiscoveryRate value

falseNegativeRate Calculate the fals	seNegativeRate
--------------------------------------	----------------

Description

Calculate the falseNegativeRate

Usage

```
falseNegativeRate(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the falseNegativeRate

Value

falseNegativeRate value

14 falsePositiveRate

falseOmissionRate

Calculate the falseOmissionRate

Description

Calculate the falseOmissionRate

Usage

```
falseOmissionRate(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the falseOmissionRate

Value

falseOmissionRate value

 ${\tt falsePositiveRate}$

 ${\it Calculate the false Positive Rate}$

Description

Calculate the falsePositiveRate

Usage

```
falsePositiveRate(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the falsePositiveRate

Value

falsePositiveRate value

fitGLMModel 15

Description

Fit a predictive model

Usage

```
fitGLMModel(population, plpData, modelType = "logistic",
   excludeCovariateIds = c(), includeCovariateIds = c(),
   prior = createPrior("laplace", useCrossValidation = TRUE),
   control = createControl(cvType = "auto", fold = 3, startingVariance = 0.01,
   tolerance = 2e-06, cvRepetitions = 1, selectorType = "byPid", noiseLevel =
   "silent", threads = -1, maxIterations = 3000))
```

Arguments

population	A population object generated by createStudyPopulation, potentially filtered
population.	11 population object generated by c. od cooleday. opaza cro., potentiany mitered

by other functions.

plpData An object of type plpData as generated using getDbPlpData.

modelType The type of outcome model that will be used. Possible values are "logistic",

"poisson", or "cox".

excludeCovariateIds

Exclude these covariates from the outcome model.

includeCovariateIds

Include only these covariates in the outcome model.

prior The prior used to fit the model. See createPrior for details.

control The control object used to control the cross-validation used to determine the

hyperparameters of the prior (if applicable). See createControl for details.

Description

Train various models using a default parameter gird search or user specified parameters

Usage

```
fitPlp(population, data, modelSettings, cohortId, outcomeId)
```

16 getAttritionTable

Arguments

population The population created using createStudyPopulation() who will have their risks

predicted

data An object of type plpData - the patient level prediction data extracted from the

CDM.

modelSettings An object of class modelSettings created using one of the function:

• logisticRegressionModel() A lasso logistic regression model

• GBMclassifier() A gradient boosting machine

• RFclassifier() A random forest model

• GLMclassifier () A generalised linear model

• KNNclassifier() A KNN model

cohortId Id of study cohort
outcomeId Id of outcome cohort

Details

The user can define the machine learning model to train (regularised logistic regression, random forest, gradient boosting machine, neural network and)

Value

An object of class plpModel containing:

model The trained prediction model

modelLoc The path to where the model is saved (if saved)

trainAuc The AUC obtained on the training set

trainCalibration

The calibration obtained on the training set

modelSettings A list specifiying the model, preprocessing, outcomeId and cohortId

trainingTime The time taken to train the classifier

getAttritionTable Get the attrition table for a population

Description

Get the attrition table for a population

Usage

getAttritionTable(object)

Arguments

object Either an object of type plpData, a population object generated by functions

like createStudyPopulation, or an object of type outcomeModel.

getCalibration 17

Value

A data frame specifying the number of people and exposures in the population after specific steps of filtering.

getCalibration

Get a sparse summary of the calibration

Description

Get a sparse summary of the calibration

Usage

```
getCalibration(prediction, numberOfStrata = 10, truncateFraction = 0.01)
```

Arguments

prediction A prediction object as generated using the predict functions. numberOfStrata The number of strata in the plot. truncateFraction

This fraction of probability values will be ignored when plotting, to avoid the x-axis scale being dominated by a few outliers.

Details

Generates a sparse summary showing the predicted probabilities and the observed fractions. Predictions are stratefied into equally sized bins of predicted probabilities.

Value

A dataframe with the calibration summary

 ${\tt getModelDetails}$

Get the predictive model details

Description

getModelDetails shows the full model, so showing the betas of all variables included in the model, along with the variable names

Usage

```
getModelDetails(predictiveModel, plpData)
```

Arguments

predictiveModel

An object of type predictiveModel as generated using he fitPlp function.

plpData

An object of type plpData as generated using getPlpData.

18 getPlpData

Details

Shows the coefficients and names of the covariates with non-zero coefficients.

getPlpData

Get the patient level prediction data from the server

Description

This function executes a large set of SQL statements against the database in OMOP CDM format to extract the data needed to perform the analysis.

Usage

```
getPlpData(connectionDetails, cdmDatabaseSchema,
 oracleTempSchema = cdmDatabaseSchema, cohortId, outcomeIds,
 studyStartDate = "", studyEndDate = "",
 cohortDatabaseSchema = cdmDatabaseSchema, cohortTable = "cohort",
 outcomeDatabaseSchema = cdmDatabaseSchema, outcomeTable = "cohort",
 cdmVersion = "5", excludeDrugsFromCovariates = F,
 firstExposureOnly = FALSE, washoutPeriod = 0, covariateSettings)
```

Arguments

connectionDetails

An R object of type

connectionDetails created using the function createConnectionDetails in the DatabaseConnector package.

cdmDatabaseSchema

The name of the database schema that contains the OMOP CDM instance. Requires read permissions to this database. On SQL Server, this should specifiy both the database and the schema, so for example 'cdm_instance.dbo'.

oracleTempSchema

For Oracle only: the name of the database schema where you want all temporary tables to be managed. Requires create/insert permissions to this database.

cohortId

A unique identifier to define the at risk cohort. If cohortTable = DRUG_ERA, cohortId is a CONCEPT_ID and all descendant concepts within that CON-CEPT_ID will be used to define the cohort. If cohortTable <> DRUG_ERA, cohortId is used to select the cohort_concept_id in the cohort-like table.

outcomeIds

A list of cohort_definition_ids used to define outcomes.

studyStartDate A calendar date specifying the minimum date that a cohort index date can appear. Date format is 'yyyymmdd'.

studyEndDate

A calendar date specifying the maximum date that a cohort index date can appear. Date format is 'yyyymmdd'. Important: the study end data is also used to truncate risk windows, meaning no outcomes beyond the study end date will be considered

cohortDatabaseSchema

The name of the database schema that is the location where the cohort data used to define the at risk cohort is available. If cohortTable = DRUG_ERA, cohortDatabaseSchema is not used by assumed to be cdmSchema. Requires read permissions to this database.

getPlpData 19

cohortTable The tablename that contains the at risk cohort. If cohortTable <> DRUG ERA,

then expectation is cohort Table has format of COHORT table: cohort concept id,

SUBJECT_ID, COHORT_START_DATE, COHORT_END_DATE.

outcomeDatabaseSchema

The name of the database schema that is the location where the data used to define the outcome cohorts is available. If cohortTable = CONDITION ERA, exposureDatabaseSchema is not used by assumed to be cdmSchema. Requires

read permissions to this database.

outcomeTable The tablename that contains the outcome cohorts. If outcome Table <> CONDI-

> TION OCCURRENCE, then expectation is outcome Table has format of CO-HORT table: COHORT_DEFINITION_ID, SUBJECT_ID, COHORT_START_DATE,

COHORT_END_DATE.

Define the OMOP CDM version used: currently support "4" and "5". cdmVersion

excludeDrugsFromCovariates

Should the target and comparator drugs (and their descendant concepts) be excluded from the covariates? Note that this will work if the drugs are actualy drug

concept IDs (and not cohort IDs).

firstExposureOnly

Should only the first exposure per subject be included? Note that this is typically done in the createStudyPopulation function, but can already be done here for

efficiency reasons.

The mininum required continuous observation time prior to index date for a washoutPeriod person to be included in the at risk cohort. Note that this is typically done in the

createStudyPopulation function, but can already be done here for efficiency

reasons.

covariateSettings

An object of type covariateSettings as created using the createCovariateSettings function in the FeatureExtraction package.

Details

Based on the arguments, the at risk cohort data is retrieved, as well as outcomes occurring in these subjects. The at risk cohort can be identified using the drug_era table, or through user-defined cohorts in a cohort table either inside the CDM instance or in a separate schema. Similarly, outcomes are identified using the condition_era table or through user-defined cohorts in a cohort table either inside the CDM instance or in a separate schema. Covariates are automatically extracted from the appropriate tables within the CDM. Important: The concepts used to define the at risk cohort must not be included in the covariates, including any descendant concepts. If the cohortId arguments represent real concept IDs, you can set the excludeDrugsFromCovariates argument to TRUE and automatically the drugs and their descendants will be excluded from the covariates. However, if the cohortId argument does not represent concept IDs, you will need to manually add the concept_ids and descendants to the excludedCovariateConceptIds of the covariateSettings argument.

Value

Returns an object of type plpData, containing information on the cohorts, their outcomes, and baseline covariates. Information about multiple outcomes can be captured at once for efficiency reasons. This object is a list with the following components:

outcomes A data frame listing the outcomes per person, including the time to event, and the outcome id. Outcomes are not yet filtered based on risk window, since this is done at a later stage.

cohorts A data frame listing the persons in each cohort, listing their exposure status as well as the time to the end of the observation period and time to the end of the cohort (usually the end of the exposure era).

covariates An ffdf object listing the baseline covariates per person in the two cohorts. This is done using a sparse representation: covariates with a value of 0 are omitted to save space.

covariateRef An ffdf object describing the covariates that have been extracted.

metaData A list of objects with information on how the cohortMethodData object was constructed.

The generic () and summary() functions have been implemented for this object.

getPredictionDistribution

Calculates the prediction distribution

Description

Calculates the prediction distribution

Usage

getPredictionDistribution(prediction)

Arguments

prediction

A prediction object as generated using the predictProbabilities function.

Details

Calculates the quantiles from a predition object

Value

The 0.00, 0.1, 0.25, 0.5, 0.75, 0.9, 1.00 quantile pf the prediction, the mean and standard deviation per class

getThresholdSummary

Calculate all measures for sparse ROC

Description

Calculate all measures for sparse ROC

Usage

getThresholdSummary(prediction)

Arguments

prediction

A prediction object as generated using the predictProbabilities function.

grepCovariateNames 21

Details

Calculates the TP, FP, TN, FN, TPR, FPR, accuracy, PPF, FOR and Fmeasure from a predition object

Value

A data frame with all the measures

grepCovariateNames

Extract covariate names

Description

Extracts covariate names using a regular-expression.

Usage

grepCovariateNames(pattern, object)

Arguments

pattern A regular expression with which to name covariate names

object An R object of type plpData or covariateData.

Details

This function extracts covariate names that match a regular-expression for a plpData or covariateData object.

Value

Returns a data.frame containing information about covariates that match a regular expression. This data.frame has the following columns:

covariateId Numerical identifier for use in model fitting using these covariates

covariateName Text identifier

analysisId Analysis identifier

conceptId OMOP common data model concept identifier, or 0

22 loadPlpData

insertDbPopulation	Insert a population into a database
Inoci cooi oparación	ment a population time a database

Description

Insert a population into a database

Usage

```
insertDbPopulation(population, cohortIds = 1, connectionDetails,
  cohortDatabaseSchema, cohortTable = "cohort", createTable = FALSE,
  dropTableIfExists = TRUE, cdmVersion = "5")
```

Arguments

population Either an object of type plpData or a population object generated by functions

like createStudyPopulation.

cohortIds The IDs to be used for the treated and comparator cohort, respectively.

connectionDetails

An R object of type

connectionDetails created using the function createConnectionDetails in

the DatabaseConnector package.

cohortDatabaseSchema

The name of the database schema where the data will be written. Requires write permissions to this database. On SQL Server, this should specify both the

database and the schema, so for example 'cdm_instance.dbo'.

cohortTable The name of the table in the database schema where the data will be written.

createTable Should a new table be created? If not, the data will be inserted into an existing

table.

 ${\tt dropTableIfExists}$

If createTable = TRUE and the table already exists it will be overwritten.

cdmVersion Define the OMOP CDM version used: currently support "4" and "5".

Details

Inserts a population table into a database. The table in the database will have the same structure as the 'cohort' table in the Common Data Model.

loadPlpData Load the cohort data from a folder

Description

loadPlpData loads an object of type plpData from a folder in the file system.

Usage

```
loadPlpData(file, readOnly = TRUE)
```

loadPlpModel 23

Arguments

file The name of the folder containing the data.

readOnly If true, the data is opened read only.

Details

The data will be written to a set of files in the folder specified by the user.

Value

An object of class plpData.

Examples

todo

loadPlpModel

loads the plp model

Description

loads the plp model

Usage

loadPlpModel(dirPath)

Arguments

dirPath

The location of the model

Details

Loads a plp model that was saved using savePlpModel()

 ${\tt loadPlpResult}$

Loads the evaluation dataframe

Description

Loads the evaluation dataframe

Usage

loadPlpResult(dirPath)

Arguments

dirPath

The directory where the evaluation was saved

Details

Loads the evaluation

loadPrediction Load

Loads the prediciton dataframe to csv

Description

Loads the prediciton dataframe to csv

Usage

loadPrediction(dirPath)

Arguments

dirPath

The directory to saved the csv

Details

Loads the prediciton csv file

 ${\tt negativeLikelihoodRatio}$

Calculate the negativeLikelihoodRatio

Description

Calculate the negativeLikelihoodRatio

Usage

```
negativeLikelihoodRatio(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the negativeLikelihoodRatio

Value

negativeLikelihoodRatio value

${\tt negativePredictiveValue}$

 $Calculate\ the\ negative Predictive Value$

Description

Calculate the negativePredictiveValue

Usage

```
negativePredictiveValue(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the negativePredictiveValue

Value

negativePredictiveValue value

PatientLevelPrediction

 ${\it Patient Level Prediction}$

Description

PatientLevelPrediction

personSplitter	Split data into random subsets stratified by class
----------------	--

Description

Split data into random subsets stratified by class

Usage

```
personSplitter(population, test = 0.3, nfold = 3, seed = NULL)
```

Arguments

population An object created using createStudyPopulation().

A real number between 0 and 1 indicating the test set fraction of the data

An integer >= 1 specifying the number of folds used in cross validation

seed If set a fixed seed is used, otherwise a random split is performed

Details

Returns a dataframe of rowIds and indexes with a -1 index indicating the rowId belongs to the test set and a positive integer index value indicating the rowId's cross valiation fold within the train set.

Value

A dataframe containing the columns: rowId and index

```
plotDemographicSummary
```

Plot the Observed vs. expected incidence, by age and gender

Description

Plot the Observed vs. expected incidence, by age and gender

Usage

```
plotDemographicSummary(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the Observed vs. expected incidence, by age and gender #'

plotF1Measure 27

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotF1Measure Plot the F1 measure efficiency frontier using the sparse thresholdSummary data frame

Description

Plot the F1 measure efficiency frontier using the sparse thresholdSummary data frame

Usage

```
plotF1Measure(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the F1 measure efficiency frontier using the sparse thresholdSummary data frame

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotGeneralizability Plot the train/test generalizability diagnostic

Description

Plot the train/test generalizability diagnostic

Usage

```
plotGeneralizability(covariateSummary, fileName = NULL)
```

Arguments

covariateSummary

A prediction object as generated using the runPlp function.

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

28 plotPrecisionRecall

Details

Create a plot showing the train/test generalizability diagnostic #'

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotPlp

Plot all the PatientLevelPrediction plots

Description

Plot all the PatientLevelPrediction plots

Usage

```
plotPlp(result, filename)
```

Arguments

result Object returned by the runPlp() function

filename Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a directory with all the plots

Value

TRUE if it ran

plotPrecisionRecall

Plot the precision-recall curve using the sparse thresholdSummary data frame

Description

Plot the precision-recall curve using the sparse thresholdSummary data frame

Usage

```
plotPrecisionRecall(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

plotPredictedPDF 29

Details

Create a plot showing the precision-recall curve using the sparse thresholdSummary data frame

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotPredictedPDF	Plot the Predicted probability density function, showing prediction
	overlap between true and false cases

Description

Plot the Predicted probability density function, showing prediction overlap between true and false cases

Usage

```
plotPredictedPDF(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot,png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the predicted probability density function, showing prediction overlap between true and false cases

Value

A ggplot object. Use the ggsave function to save to file in a different format.

```
plotPredictionDistribution
```

Plot the side-by-side boxplots of prediction distribution, by class#'

Description

Plot the side-by-side boxplots of prediction distribution, by class#'

Usage

```
plotPredictionDistribution(evaluation, type = "train", fileName = NULL)
```

30 plotPreferencePDF

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the side-by-side boxplots of prediction distribution, by class #'

Value

A ggplot object. Use the ggsave function to save to file in a different format.

 ${\it plot} \ {\it Plot} \ the \ preference \ score \ probability \ density \ function, \ showing \ prediction \ probability \ density \ function \ probability \ function \ probability \ density \ function \ probability \ fu$

tion overlap between true and false cases #'

Description

Plot the preference score probability density function, showing prediction overlap between true and false cases #'

Usage

```
plotPreferencePDF(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the preference score probability density function, showing prediction overlap between true and false cases #'

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotRoc 31

plotRoc Plot the ROC curve

Description

Plot the ROC curve

Usage

```
plotRoc(prediction, fileName = NULL)
```

Arguments

prediction A prediction object as generated using the predictProbabilities function.

FileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the Receiver Operator Characteristics (ROC) curve.

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotSparseCalibration Plot the calibration

Description

Plot the calibration

Usage

```
plotSparseCalibration(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the calibration #'

Value

A ggplot object. Use the ggsave function to save to file in a different format.

32 plotSparseRoc

```
plotSparseCalibration2
```

Plot the conventional calibration

Description

Plot the conventional calibration

Usage

```
plotSparseCalibration2(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the calibration #'

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotSparseRoc

Plot the ROC curve using the sparse thresholdSummary data frame

Description

Plot the ROC curve using the sparse thresholdSummary data frame

Usage

```
plotSparseRoc(evaluation, type = "train", fileName = NULL)
```

Arguments

evaluation A prediction object as generated using the runPlp function.

type options: 'train' or test'

fileName Name of the file where the plot should be saved, for example 'plot.png'. See the

function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the Receiver Operator Characteristics (ROC) curve.

Value

A ggplot object. Use the ggsave function to save to file in a different format.

plotVariableScatterplot

Plot the variable importance scatterplot

Description

Plot the variable importance scatterplot

Usage

```
plotVariableScatterplot(covariateSummary, fileName = NULL)
```

Arguments

covariateSummary

A prediction object as generated using the runPlp function.

fileName

Name of the file where the plot should be saved, for example 'plot.png'. See the function ggsave in the ggplot2 package for supported file formats.

Details

Create a plot showing the variable importance scatterplot #'

Value

A ggplot object. Use the ggsave function to save to file in a different format.

```
plpDataSimulationProfile
```

A simulation profile

Description

A simulation profile

Usage

data(plpDataSimulationProfile)

positivePredictiveValue

positive Likelihood Ratio

 $Calculate\ the\ positive Likelihood Ratio$

Description

Calculate the positiveLikelihoodRatio

Usage

```
positiveLikelihoodRatio(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the positiveLikelihoodRatio

Value

positiveLikelihoodRatio value

```
positivePredictiveValue
```

Calculate the positivePredictiveValue

Description

Calculate the positivePredictiveValue

Usage

```
positivePredictiveValue(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the positivePredictiveValue

predictPlp 35

Value

positivePredictiveValue value

predictFfdf	Generated predictions from a regression model		
-------------	---	--	--

Description

Generated predictions from a regression model

Usage

```
predictFfdf(coefficients, population, covariates, modelType = "logistic")
```

Arguments

coefficients	A names numeric vector where the names are the covariateIds, except for the first value which is expected to be the intercept.
population	A data frame containing the population to do the prediction for
covariates	A data frame or ffdf object containing the covariates with predefined columns (see below).
modelType	Current supported types are "logistic", "poisson", or "survival".

Details

These columns are expected in the outcome object:

rowId	(integer)	Row ID is used to link multiple covariates (x) to a single outcome (y)
time	(real)	For models that use time (e.g. Poisson or Cox regression) this contains time
		(e.g. number of days)

These columns are expected in the covariates object:

rowId	(integer)	Row ID is used to link multiple covariates (x) to a single outcome (y)
covariateId	(integer)	A numeric identifier of a covariate
covariateValue	(real)	The value of the specified covariate

Description

Predict the risk of the outcome using the input plpModel for the input plpData

36 predictProbabilities

Usage

```
predictPlp(plpModel, population, plpData, index = NULL)
```

Arguments

plpModel An object of type plpModel - a patient level prediction model

population The population created using createStudyPopulation() who will have their risks

predicted

plpData An object of type plpData - the patient level prediction data extracted from the

CDM.

index A data frame containing rowId: a vector of rowids and index: a vector of doubles

the same length as the rowlds. If used, only the rowlds with a negative index

value are used to calculate the prediction.

Details

The function applied the trained model on the plpData to make predictions

Value

A dataframe containing the prediction for each person in the population with an attribute metaData containing prediction details.

predictProbabilities Create predictive probabilities

Description

Create predictive probabilities

Usage

```
predictProbabilities(predictiveModel, population, covariates)
```

Arguments

predictiveModel

An object of type predictiveModel as generated using fitPlp.

population The population to calculate the prediction for

covariates The covariate part of PlpData containing the covariates for the population

Details

Generates predictions for the population specified in plpData given the model.

Value

The value column in the result data.frame is: logistic: probabilities of the outcome, poisson: Poisson rate (per day) of the outcome, survival: hazard rate (per day) of the outcome.

runPlpAnalyses 37

runPlpAnalyses	Develop patient-level predcition models for multiple outcomes, target popuations and settings

Description

Develop patient-level predcition models for multiple outcomes, target popuations and settings

Usage

```
runPlpAnalyses(outputFolder = getwd(), connectionDetails = NULL,
 cdmDatabaseSchema = NULL, oracleTempSchema = cdmDatabaseSchema,
 cohortDatabaseSchema = cdmDatabaseSchema, cohortTable = "cohort"
 outcomeDatabaseSchema = cdmDatabaseSchema, outcomeTable = "cohort",
 cdmVersion = "5", studyStartDate = "", studyEndDate = "",
 atRiskCohortIds = 1, outcomeIds = 2,
 covariateSettings = list(FeatureExtraction::createCovariateSettings(useCovariateDemographics
 = T, useCovariateDemographicsGender = T, useCovariateDemographicsRace = T,
 useCovariateDemographicsAge = T, useCovariateDemographicsYear = F,
 useCovariateDemographicsMonth = T, useCovariateConditionOccurrence = T,
 useCovariateConditionOccurrence365d = T),
 FeatureExtraction::createCovariateSettings(useCovariateDemographics = T,
 useCovariateDemographicsGender = T, useCovariateDemographicsRace = T,
 useCovariateDemographicsAge = T,
                                        useCovariateDemographicsYear = F,
 useCovariateDemographicsMonth = T, useCovariateDrugExposure = T,
 useCovariateDrugExposure365d = T)),
  timeAtRisks = list(setTimeAtRisks(riskWindowEnd = 365),
  setTimeAtRisks(riskWindowEnd = 365 * 2)), modelSettings = NULL,
  internalValidation = "time", testFraction = 0.25, nfold = 3,
  splitSeed = NULL, indexes = NULL, verbosity = futile.logger::INFO)
```

Arguments

outputFolder The directory to save the results and data to - needs read/write privileges connectionDetails

An R object of type connectionDetails created using the function createConnectionDetails in the DatabaseConnector package.

cdmDatabaseSchema

The name of the database schema that contains the OMOP CDM instance. Requires read permissions to this database. On SQL Server, this should specify both the database and the schema, so for example 'cdm_instance.dbo'.

oracleTempSchema

For Oracle only: the name of the database schema where you want all temporary tables to be managed. Requires create/insert permissions to this database.

cohortDatabaseSchema

The name of the database schema that is the location where the cohort data used to define the at risk cohort is available. If cohortTable = DRUG_ERA, cohortDatabaseSchema is not used by assumed to be cdmSchema. Requires read permissions to this database.

38 runPlpAnalyses

cohortTable The tablename that contains the at risk cohort. If cohortTable <> DRUG_ERA, then expectation is cohortTable has format of COHORT table: cohort_concept_id,

SUBJECT_ID, COHORT_START_DATE, COHORT_END_DATE.

outcomeDatabaseSchema

The name of the database schema that is the location where the data used to define the outcome cohorts is available. If cohortTable = CONDITION_ERA, exposureDatabaseSchema is not used by assumed to be cdmSchema. Requires

read permissions to this database.

outcomeTable The tablename that contains the outcome cohorts. If outcomeTable <> CONDI-

TION_OCCURRENCE, then expectation is outcomeTable has format of CO-HORT table: COHORT_DEFINITION_ID, SUBJECT_ID, COHORT_START_DATE,

COHORT_END_DATE.

cdmVersion Define the OMOP CDM version used: currently support "4" and "5".

studyStartDate A calendar date specifying the minimum date that a cohort index date can ap-

pear. Date format is 'yyyymmdd'.

studyEndDate A calendar date specifying the maximum date that a cohort index date can ap-

pear. Date format is 'yyyymmdd'. Important: the study end data is also used to truncate risk windows, meaning no outcomes beyond the study end date will be

considered.

atRiskCohortIds

A vector containing the unique identifiers to define the at risk cohorts. Each at risk cohortId is used to select the cohort_concept_id in the cohort-like table.

outcomeIds A list of cohort_definition_ids used to define outcomes.

covariateSettings

An object of type covariate Settings as created using the create Covariate Settings function in the Feature Extraction package. This can be a list of multiple set-

tings.

timeAtRisks A list detailing the time at risk intervals that willbe used to create the prediciton

models (the period of time we wish to predict the outcome occurence within)

created using the function $\mathtt{setTimeAtRisks}.$

modelSettings A list of model settings created using the setGradientBoostingMachine, setRandomForest,

 ${\tt setLassoLogisticRegression}, {\tt setNaiveBayes} \ or \ {\tt setKNN}.$

internal Validation

verbosity

The type of internal validation for the model. Either 'person' which stratifies by outcome to partion into test/train sets or 'time' which picks a set date and all people with an at risk cohort start date prior to this join the train set and people

after join the test set.

testFraction The fracion of the target population to include into the test set

nfold The number of cross validation folds to apply when finding the optimal hyper-

parameters

splitSeed (default NULL) The seed used to do the random split for internal Validation='person'

indexes The nfold validation indexes

Sets the level of the verbosity. If the log level is at or higher in priority than the logger threshold, a message will print. The levels are:

• DEBUGHighest verbosity showing all debug statements

- TRACEShowing information about start and end of steps
- INFOShow informative information (Default)
- · WARNShow warning messages

savePlpData 39

- ERRORShow error messages
- FATALBe silent except for fatal errors

savePlpData

Save the cohort data to folder

Description

savePlpData saves an object of type plpData to folder.

Usage

```
savePlpData(plpData, file, envir = NULL)
```

Arguments

plpData An object of type plpData as generated using getDbPlpData.

file The name of the folder where the data will be written. The folder should not yet

exist.

envir The environment for to evaluate variables when saving

Details

The data will be written to a set of files in the folder specified by the user.

Examples

todo

savePlpModel

Saves the plp model

Description

Saves the plp model

Usage

```
savePlpModel(plpModel, dirPath)
```

Arguments

plpModel A trained classifier returned by running runPlp()\$model

dirPath A location to save the model to

Details

Saves the plp model to a user specificed folder

40 savePrediction

savePlpResult

Saves the result from runPlp into the location directory

Description

Saves the result from runPlp into the location directory

Usage

```
savePlpResult(result, dirPath)
```

Arguments

result The result of running runPlp()
dirPath The directory to save the csv

Details

Saves the result from runPlp into the location directory

savePrediction

Saves the prediction dataframe to csv

Description

Saves the prediction dataframe to csv

Usage

```
savePrediction(prediction, dirPath)
```

Arguments

prediction The prediciton data.frame
dirPath The directory to save the csv

Details

Saves the prediction data frame returned by predict.R to a csv file

sensitivity 41

sensitivity	Calculate the sensitivity	

Description

Calculate the sensitivity

Usage

```
sensitivity(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the sensitivity

Value

sensitivity value

```
setGradientBoostingMachine
```

Create setting for gradient boosting machine model using gbm_xgboost implementation

Description

Create setting for gradient boosting machine model using gbm_xgboost implementation

Usage

```
setGradientBoostingMachine(ntrees = c(10, 100), nthread = 20,
  max_depth = 6, min_rows = 20, learn_rate = 0.1, seed = NULL)
```

Arguments

ntrees	The number of trees to build
nthread	The number of computer threads to (how many cores do you have?)
max_depth	Maximum number of interactions - a large value will lead to slow model training
min_rows	The minimum number of rows required at each end node of the tree
learn_rate	The boosting learn rate
seed	An option to add a seed when training the final model

Examples

```
\label{loss_model_gbm} $$\ensuremath{^{-}}$ setGradientBoostingMachine(ntrees=c(10,100), nthread=20, \\ max\_depth=c(4,6), learn\_rate=c(0.1,0.3)) $$
```

setKNN

Create setting for knn model

Description

Create setting for knn model

Usage

```
setKNN(k = 1000, indexFolder = file.path(getwd(), "knn"))
```

Arguments

k The number of neighbors to consider

indexFolder The directory where the results and intermediate steps are output

Examples

```
model.knn \leftarrow setKNN(k=10000)
```

 ${\tt setLassoLogisticRegression}$

Create setting for lasso logistic regression

Description

Create setting for lasso logistic regression

Usage

```
setLassoLogisticRegression(variance = 0.01, seed = NULL)
```

Arguments

variance a single value used as the starting value for the automatic lambda search

seed An option to add a seed when training the model

Examples

```
model.lr <- setLassoLogisticRegression()</pre>
```

setMLP 43

setMLP

Create setting for neural network model with python

Description

Create setting for neural network model with python

Usage

```
setMLP(size = 4, alpha = 1e-05, seed = NULL)
```

Arguments

size The number of hidden nodes

alpha The 12 regularisation seed A seed for the model

Examples

```
## Not run:
model.mlp <- setMLP(size=4, alpha=0.00001, seed=NULL)
## End(Not run)</pre>
```

setNaiveBayes

Create setting for naive bayes model with python

Description

Create setting for naive bayes model with python

Usage

```
setNaiveBayes()
```

Examples

```
## Not run:
model.nb <- setNaiveBayes()
## End(Not run)</pre>
```

44 setTimeAtRisk

setRandomForest Create setting for random forest model with python (very fast)	
--	--

Description

Create setting for random forest model with python (very fast)

Usage

```
setRandomForest(mtries = -1, ntrees = c(10, 500), max_depth = 17, varImp = T, seed = NULL)
```

Arguments

mtries	The number of features to include in each tree (-1 defaults to square root of total features)
ntrees	The number of trees to build
max_depth	Maximum number of interactions - a large value will lead to slow model training
varImp	Perform an initial variable selection prior to fitting the model to select the useful variables
seed	An option to add a seed when training the final model

Examples

 ${\tt setTimeAtRisk}$

setTimeAtRisk

Description

create the timeAtRisks for the multiple analysis studies

Usage

```
setTimeAtRisk(includeAllOutcomes = T, firstExposureOnly = F,
washoutPeriod = 0, removeSubjectsWithPriorOutcome = T,
priorOutcomeLookback = 99999, riskWindowStart = 1,
addExposureDaysToStart = F, riskWindowEnd = 365,
addExposureDaysToEnd = F, requireTimeAtRisk = T,
minTimeAtRisk = riskWindowEnd - riskWindowStart)
```

similarPlpData 45

Arguments

includeAllOutcomes

Do you want to include people who have the outcome but are not observed for the whole at risk period?

firstExposureOnly

Only consider the first time occurence of the outcome?

washoutPeriod The minimum prior observation in days a person required to be included removeSubjectsWithPriorOutcome

Remove people who have the outcome some period before the time at risk?

priorOutcomeLookback

The number of days prior to investigate for the variable removeSubjectsWith-PriorOutcome

riskWindowStart

The number of days after the at risk population subject's index date to start the time at risk period

 ${\it add} {\it Exposure Days To Start}$

Should the risk window start be relative to the index end date instead?

riskWindowEnd The number of days after the at risk population subject's index date to end the time at risk period

addExposureDaysToEnd

Should the risk window end be relative to the index end date instead?

requireTimeAtRisk

Should you only include people with a minimum time at risk period?

minTimeAtRisk If requireTimeAtRisk is TRUE, then this is the minimum number of days a person must be at risk

similarPlpData

Extract new plpData using plpModel settings use metadata in plp-Model to extract similar data and population for new databases:

Description

Extract new plpData using plpModel settings use metadata in plpModel to extract similar data and population for new databases:

Usage

```
similarPlpData(plpModel = NULL, newConnectionDetails = NULL,
  newCdmDatabaseSchema = NULL, newCohortDatabaseSchema = NULL,
  newCohortTable = NULL, newCohortId = NULL,
  newOutcomeDatabaseSchema = NULL, newOutcomeTable = NULL,
  newOutcomeId = NULL)
```

46 simulatePlpData

Arguments

plpModel The trained PatientLevelPrediction model or object returned by runPlp() newConnectionDetails The connectionDetails for the new database newCdmDatabaseSchema The database schema for the new CDM database newCohortDatabaseSchema The database schema where the cohort table is stored newCohortTable The table name of the cohort table newCohortId The cohort_definition_id for the cohort of at risk people ${\tt newOutcomeDatabaseSchema}$ The database schema where the outcome table is stored newOutcomeTable The table name of the outcome table The cohort_definition_id for the outcome newOutcomeId

Examples

```
## Not run:
# set the connection
connectionDetails <- DatabaseConnector::createConnectionDetails()</pre>
# load the model and data
plpModel <- loadPlpModel("C:/plpmodel")</pre>
# extract the new data in the 'newData.dbo' schema using the model settings
newDataList <- similarPlpData(plpModel=plpModel,</pre>
                               newConnectionDetails = connectionDetails,
                               newCdmDatabaseSchema = 'newData.dbo',
                               newCohortDatabaseSchema = 'newData.dbo',
                               newCohortTable = 'cohort',
                               newCohortId = 1,
                               newOutcomeDatabaseSchema = 'newData.dbo',
                               newOutcomeTable = 'outcome',
                               newOutcomeId = 2)
# get the prediction:
prediction <- applyModel(newDataList$population, newDataList$plpData, plpModel)$prediction</pre>
## End(Not run)
```

simulatePlpData

Generate simulated data

Description

 ${\tt simulateplpData}\ creates\ a\ plpData\ object\ with\ simulated\ data.$

Usage

```
simulatePlpData(plpDataSimulationProfile, n = 10000)
```

specificity 47

Arguments

 $\verb|plpDataSimulationProfile| \\$

An object of type plpDataSimulationProfile as generated using the createplpDataSimulationProfile function.

n The size of the population to be generated.

Details

This function generates simulated data that is in many ways similar to the original data on which the simulation profile is based. The contains same outcome, comparator, and outcome concept IDs, and the covariates and their 1st order statistics should be comparable.

Value

An object of type plpData.

specificity

Calculate the specificity

Description

Calculate the specificity

Usage

```
specificity(TP, TN, FN, FP)
```

Arguments

TP	Number of true positives
TN	Number of true negatives
FN	Number of false negatives
FP	Number of false positives

Details

Calculate the specificity

Value

specificity value

48 toSparseM

timeSplitter	Split test/train data by time and then partitions training set into random folds stratified by class

Description

Split test/train data by time and then partitions training set into random folds stratified by class

Usage

```
timeSplitter(population, test = 0.3, nfold = 3, seed = NULL)
```

Arguments

population	An object created t	using createStudyPopula	ation().

A real number between 0 and 1 indicating the test set fraction of the data

An integer >= 1 specifying the number of folds used in cross validation

seed If set a fixed seed is used, otherwise a random split is performed

Details

Returns a dataframe of rowIds and indexes with a -1 index indicating the rowId belongs to the test set and a positive integer index value indicating the rowId's cross valiation fold within the train set.

Value

A dataframe containing the columns: rowId and index

toSparseM	Convert the plpData in COO format into a sparse R matrix	

Description

Converts the standard plpData to a sparse matrix

Usage

```
toSparseM(plpData, population, map = NULL)
```

Arguments

An object of type plpData with covariate in coo format - the patient level pre-

diction data extracted from the CDM.

population The population to include in the matrix

map A covariate map (telling us the column number for covariates)

toSparsePython 49

Details

This function converts the covariate file from ffdf in COO format into a sparse matrix from the package Matrix

Value

Returns a list, containing the data as a sparse matrix, the plpData covariateRef and a data.frame named map that tells us what covariate corresponds to each column This object is a list with the following components:

data A sparse matrix with the rows corresponding to each person in the plpData and the columns corresponding to the covariates.

covariateRef The plpData covariateRef.

map A data.frame containing the data column ids and the corresponding covariateId from covariateRef.

Examples

#TODO

toSparsePython

Convert the plpData in COO format into a sparse python matrix

Description

Converts the standard plpData to a sparse matrix firectly into python

Usage

```
toSparsePython(plpData, population, map = NULL)
```

Arguments

plpData An object of type plpData with covariate in coo format - the patient level pre-

diction data extracted from the CDM.

population The population to include in the matrix

map A covariate map (telling us the column number for covariates)

Details

This function converts the covariate file from ffdf in COO format into a sparse matrix from the package Matrix

50 toSparsePython

Value

Returns a list, containing the python object name of the sparse matrix, the plpData covariateRef and a data.frame named map that tells us what covariate corresponds to each column This object is a list with the following components:

data The python object name containing a sparse matrix with the rows corresponding to each person in the plpData and the columns corresponding to the covariates.

covariateRef The plpData covariateRef.

map A data.frame containing the data column ids and the corresponding covariateId from covariateRef.

Examples

#TODO

Index

*Topic datasets	negativeLikelihoodRatio, 24
plpDataSimulationProfile, 33	negativePredictiveValue, 25
,	,
accuracy, 4	PatientLevelPrediction, 25
applyModel, 4	PatientLevelPrediction-package
averagePrecision, 5	(PatientLevelPrediction), 25
	personSplitter, 26
brierScore, 6	plotDemographicSummary, 26
bySumFf, 6	plotF1Measure, 27
	plotGeneralizability, 27
calibrationLine, 7	plotPlp, 28
computeAuc, 7	plotPrecisionRecall, 28
computeAucFromDataFrames, 8	plotPredictedPDF, 29
createControl, 15	plotPredictionDistribution, 29
createPrior, 15	plotPreferencePDF, 30
createStudyPopulation, 8	plotRoc, 31
	plotSparseCalibration, 31
diagnosticOddsRatio, 10	plotSparseCalibration2, 32
1 (D1 11	plotSparseRoc, 32
evaluatePlp, 11	plotVariableScatterplot, 33
exportPlpDataToCsv, 11	plpDataSimulationProfile, 33
£15 12	positiveLikelihoodRatio, 34
f1Score, 12	positivePredictiveValue, 34
falseDiscoveryRate, 13	predict, 7, 17
falseNegativeRate, 13	predictFfdf, 35
falseOmissionRate, 14	predictPlp, 35
falsePositiveRate, 14	predictProbabilities, 5–7, 20, 31, 36
fitGLMModel, 15	predicti robabilities, 5-7, 20, 51, 50
fitPlp, 15, 17, 36	runPlp, 26–33
matattaitianTahla 16	runPlpAnalyses, 37
getAttritionTable, 16	rum ipmaiyses, 57
getCalibration, 17	savePlpData, 39
getModelDetails, 17	savePlpModel, 39
getPlpData, 17, 18	savePlpResult, 40
getPredictionDistribution, 20	savePrediction, 40
getThresholdSummary, 20	sensitivity, 41
ggsave, 27–33	setGradientBoostingMachine, 41
<pre>grepCovariateNames, 21</pre>	setKNN, 42
incomtDbDomolation 22	setLassoLogisticRegression, 42
insertDbPopulation, 22	setMLP, 43
loadPlpData, 22	setNaiveBayes, 43
loadPlpModel, 23	setRandomForest, 44
loadPlpResult, 23	setTimeAtRisk, 44
·	
loadPrediction, 24	similarPlpData,45

52 INDEX

```
simulatePlpData, 46
specificity, 47
timeSplitter, 48
toSparseM, 48
toSparsePython, 49
```