

# vignette PlasmodeSim

2022-10-19

Welcome to the vignette about the R package PlasmodeSim. This package is still under development.

## installing plasmodeSim using remotes

To install using `remotes` run:

```
#install.packages("remotes")
remotes::install_github("GidiusVanDeKamp/PlasmodeSim")
```

```
## Skipping install of 'PlasmodeSim' from a github remote, the SHA1 (61ac8e05) has not changed since last
## Use 'force = TRUE' to force installation
```

## Setting up

This documents skips some parts, we have skipped the steps to obtain the `plpResults` and the `plpData`.

```
library(dplyr)

modelSettings <- PatientLevelPrediction::setLassoLogisticRegression()

plpResultLogistic <- PatientLevelPrediction::loadPlpResult("~/R/internshipErasmusMC/simulate-new-patients-outcomes/plpResultLogistic")

plpData<- PatientLevelPrediction::loadPlpData("~/R/internshipErasmusMC/simulate-new-patients-outcomes/plpData")
```

## Example 1

In this example we obtain new outcomes of a fitted logistic model.

```
plpModelLog <- plpResultLogistic$model
probabilites <- PlasmodeSim::newPropsParametersPlpModel(plpModelLog, plpData, plpData$cohorts)

## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.21 secs
## Prediction took 0.182 secs
```

The function `predictPlp` returned this information.

```
newOut <- PlasmodeSim::newOutcomes(200, probabilites)
head(newOut)
```

```
##   rowId newOutcomes
## 1  2052           0
## 2  1714           1
## 3   379           1
## 4  2263           0
## 5   901           0
## 6  1316           0
```

In the output of `newOut` patients are drawn randomly with the same chance, the patients could be drawn multiple times. If this happens they can have a different outcome. The function `newOutcomes` needs a data set where the column that contains the probabilities is called `value`.

## Example 2

We here we show how to simulate new outcomes from an unfitted logistic model.

```
Parameters <- plpModelLog$model$coefficients
UnfittedParameters <- Parameters
UnfittedParameters[1,1] <- -0.4
UnfittedParameters[2:4,1] <- 0.4
head(UnfittedParameters)
```

```
##      betas covariateIds
## 1 -0.4000 (Intercept)
## 2  0.4000      6003
## 3  0.4000      8003
## 4  0.4000      9003
## 5  0.0092    8507001
## 6  0.0000    28060210
```

For the logistic model it is necessary that the parameters are stored in a dataset with a column called `betas` and a column called `covariateIds`.

```
plpModelunfitted <- PlasmodeSim::makeLogisticModel(UnfittedParameters)
newprobs <- PatientLevelPrediction::predictPlp(plpModelunfitted, plpData, plpData$cohorts)
```

```
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.196 secs
## Prediction took 0.168 secs
```

```
newOut <- PlasmodeSim::newOutcomes(200, newprobs)
head(newOut)
```

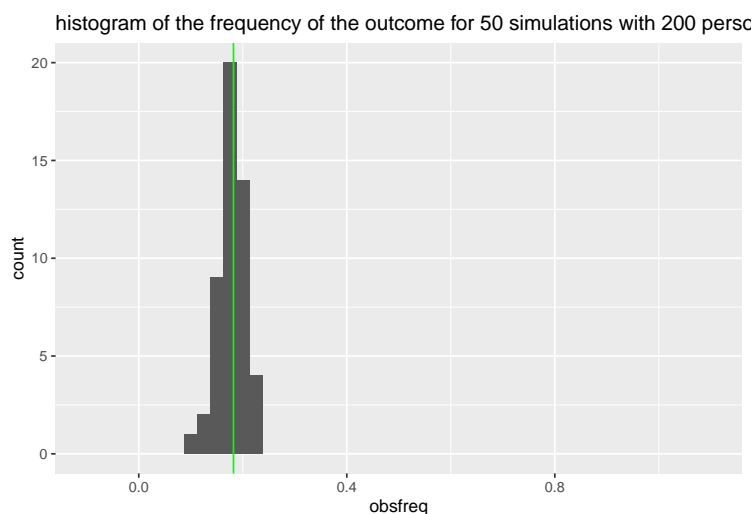
```
##   rowId newOutcomes
## 1   836           0
## 2  1803           1
## 3  2547           1
## 4  1977           1
## 5  1668           0
## 6   910           1
```

## Visual simulations

The function `visualOutcome` simulated new data and then plots the frequency of the outcome. Right now the function `visualOutcome` only works for a logistic model. The green line in the plots is the average outcome in the original dataset.

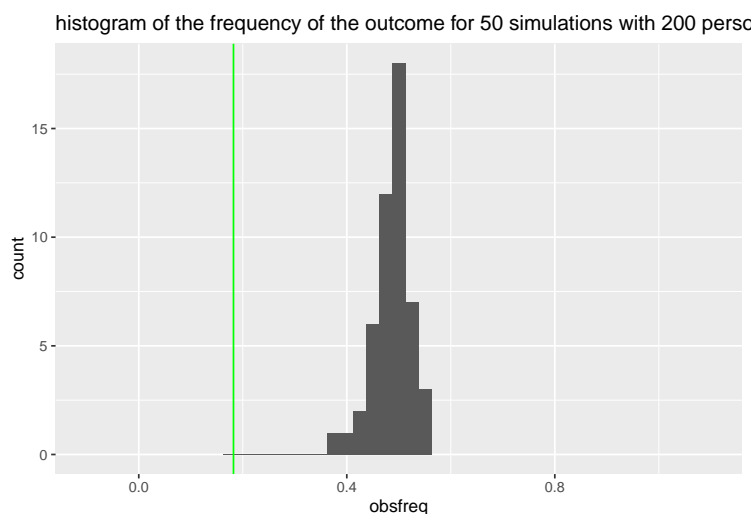
```
PlasmodeSim::visualOutcome(plpData,50,200,Parameters)
```

```
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.17 secs  
## Prediction took 0.177 secs
```



```
PlasmodeSim::visualOutcome(plpData,50,200,UnfittedParameters)
```

```
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.183 secs  
## Prediction took 0.179 secs
```



Here we have plotted 50 times the frequency of the outcome for a simulated dataset with 200 people.

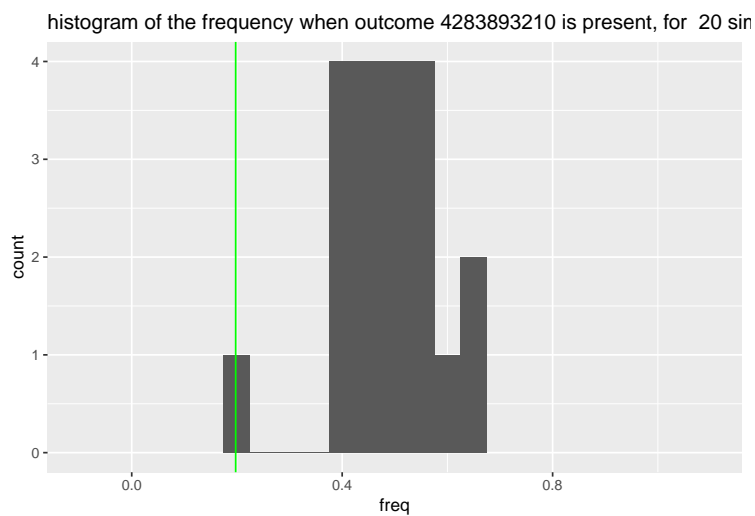
## Visual of a specific covariate

```
covariateIdToStudy<- plpResultLogistic$covariateSummary$covariateId[3]  
UnfittedParameters[3,]
```

```
##      betas covariateIds  
## 3      0.4          8003
```

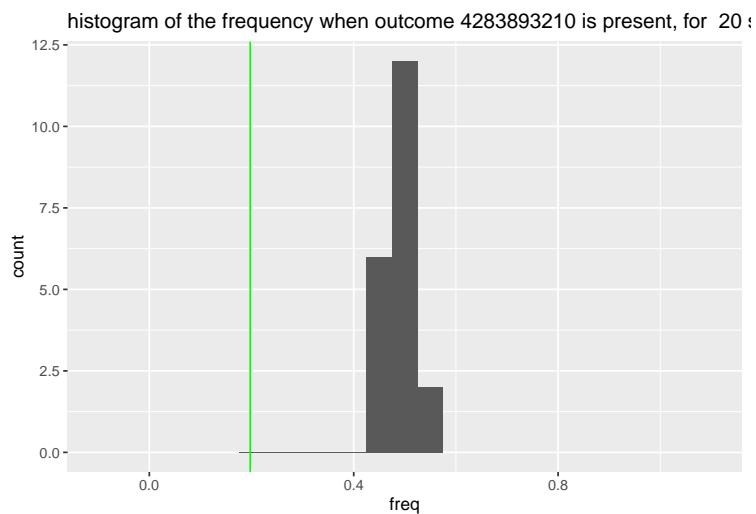
```
PlasmodeSim::visualOutcomeCovariateId(plpData, covariateIdToStudy, 20, 200, UnfittedParameters)
```

```
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.192 secs  
## Prediction took 0.188 secs
```



```
PlasmodeSim::visualOutcomeCovariateId2(plpData, covariateIdToStudy, 20, 200, UnfittedParameters)
```

```
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.18 secs  
## Prediction took 0.179 secs
```



As one can see `visualOutcomeCovariateId` and `visualOutcomeCovariateId2` are very similar, they both calculate and plot the frequency for a group with a specific covariate present. The small difference is that `visualOutcomeCovariateId` filters a newly simulated dataset set to only keep the patients where the covariate is present, and `visualOutcomeCovariateId2` only simulates new outcomes for patients that have the covariate present. We see they are almost the identical only `visualOutcomeCovariateId2` is spread out less because the groups for calculating the frequency with are bigger.

## Survival times outcomes.

For simulating new survival times we need more than one probability, we use the baselinehazard, stored in the `plpModel`.

```
plpResult <- PatientLevelPrediction::loadPlpResult("~/R/internshipErasmusMC/simulate-new-patients-outcomes/plpResult")
plpData <- PatientLevelPrediction::loadPlpData("~/R/internshipErasmusMC/simulate-new-patients-outcomes/plpData")
plpModel <- plpResult$model # a fitted Cox model

populationSettings <- PatientLevelPrediction::createStudyPopulationSettings(
  binary = TRUE,
  includeAllOutcomes = FALSE,
  firstExposureOnly = FALSE,
  washoutPeriod = 180,
  removeSubjectsWithPriorOutcome = FALSE,
  priorOutcomeLookback = 99999,
  requireTimeAtRisk = TRUE,
  minTimeAtRisk = 1,
  riskWindowStart = 1,
  startAnchor = 'cohort start',
  riskWindowEnd = 7300,
  endAnchor = 'cohort start'
)

newdata <- PlasmodeSim::simulateSurvivaltimes(plpModel, plpData, 3, populationSettings)
head(newdata)
# function that simulates uncensored data, simulateSurvivaltimes(plpModel, plpData, 20)
```

## Simulation censored survival data

first we have to set some settings.

```
executeSettings <- PatientLevelPrediction::createExecuteSettings(
  runSplitData = TRUE,
  runSampleData = FALSE,
  runfeatureEngineering = FALSE,
  runPreprocessData = TRUE,
  runModelDevelopment = TRUE,
  runCovariateSummary = TRUE
)

splitSettings <- PatientLevelPrediction::createDefaultSplitSetting(
  testFraction = 0.25,
  trainFraction = 0.75,
  splitSeed = 123,
  nfold = 3,
)
```

```

    type = 'stratified'
  )
sampleSettings <- PatientLevelPrediction::createSampleSettings(
  type = 'none'
)
featureEngineeringSettings <- PatientLevelPrediction::createFeatureEngineeringSettings(
  type = 'none'
)
preprocessSettings <- PatientLevelPrediction::createPreprocessSettings(
  minFraction = 0,
  normalize = TRUE,
  removeRedundancy = TRUE
)
modelSettings <- PatientLevelPrediction::setCoxModel()

```

now that we have our settings. We define a training set and fit the model. The model actually consists of two `plpModels` stored in a list.

```

TrainingSet <- PlasmodeSim::MakeTrainingSet(plpData, executeSettings, populationSettings, splitSettings,

```

```

## Outcome is 0 or 1
## seed: 123
## Creating a 25% test and 75% train (into 3 folds) random stratified split by class
## Data split into 656 test cases and 1974 train cases (658, 658, 658)
## Train Set:
## Fold 1 658 patients with 120 outcomes - Fold 2 658 patients with 120 outcomes - Fold 3 658 patients with
## 103 covariates in train data
## Test Set:
## 656 patients with 119 outcomes
## Removing 2 redundant covariates
## Normalizing covariates
## Tidying covariates took 0.577 secs
## Train Set:
## Fold 1 658 patients with 120 outcomes - Fold 2 658 patients with 120 outcomes - Fold 3 658 patients with
## 101 covariates in train data
## Test Set:
## 656 patients with 119 outcomes

```

```

fitcensor <- PlasmodeSim::fitCensoring(TrainingSet$Train, modelSettings)

```

```

## Running Cyclops
## Done.
## GLM fit status: OK
## Creating variable importance data frame
## Prediction took 0.139 secs
## Running Cyclops
## Done.
## GLM fit status: OK
## Creating variable importance data frame
## Prediction took 0.138 secs

```

```
NewOutcomes <- PlasmodeSim::simulateSurvivaltimesWithCensoring(fitcensor, plpData, populationSettings, 1000000)
```

```
## Outcome is 0 or 1
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.193 secs
## Prediction took 0.271 secs
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.182 secs
## Prediction took 0.276 secs
```

```
head(NewOutcomes)
```

##	rowId	outcomeCount	outcomes
## 1	1290	1	57
## 2	500	1	8
## 3	1864	1	27
## 4	428	0	226
## 5	315	0	7300
## 6	392	0	5629

## Make an unfitted model

```
# makeCoxModel.
coeff <- plpModel$model$coefficients
survival <- plpModel$model$baselineSurvival$surv
times <- plpModel$model$baselineSurvival$time

unfittedmodel <- PlasmodeSim::makeCoxModel(coeff, survival, times)

unfittedmodel$preprocessing <- plpModel$preprocessing

newdata <- PlasmodeSim::simulateSurvivaltimes(plpModel , plpData, 3, populationSettings)
```

[illegible]

```
## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para
## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para
## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para

## Outcome is 0 or 1

## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para
## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para

## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.167 secs

## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para
## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para
## Currently in a tryCatch or withCallingHandlers block, so unable to add global calling handlers. Para

## Prediction took 0.186 secs

#head(newdata)
```