

vignetteSurvivalTimes

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first we make sure we have the uptodate package

```
remotes::install_github("GidiusVanDeKamp/PlasmodeSim")
```

```
## cli (3.3.0 -> 3.4.1) [CRAN]
```

```
## package 'cli' successfully unpacked and MD5 sums checked
```

```
## Warning: cannot remove prior installation of package 'cli'
```

```
## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying C:
```

```
## \Users\gidiu\AppData\Local\R\win-library\4.2\00LOCK\cli\libs\x64\cli.dll to C:
```

```
## \Users\gidiu\AppData\Local\R\win-library\4.2\cli\libs\x64\cli.dll: Permission
```

```
## denied
```

```
## Warning: restored 'cli'
```

```
##
```

```
## The downloaded binary packages are in
```

```
## C:\Users\gidiu\AppData\Local\Temp\RtmpYPR2fZ\downloaded_packages
```

```
## * checking for file 'C:\Users\gidiu\AppData\Local\Temp\RtmpYPR2fZ\remotes260c17cd3a71\GidiusVanDeKamp
```

```
## * preparing 'PlasmodeSim':
```

```
## * checking DESCRIPTION meta-information ... OK
```

```
## * checking for LF line-endings in source and make files and shell scripts
```

```
## * checking for empty or unneeded directories
```

```
## Omitted 'LazyData' from DESCRIPTION
```

```
## * building 'PlasmodeSim_0.1.0.tar.gz'
```

```
##
```

Loading the plpData

```
connectionDetails <- Eunomia::getEunomiaConnectionDetails()
```

```
# create the database
```

```
Eunomia::createCohorts(  
  connectionDetails = connectionDetails,  
  cdmDatabaseSchema = 'main',  
  cohortDatabaseSchema = 'main',  
  cohortTable = 'cohort'  
)
```

```
## Creating cohort: Celecoxib
```

```
## |
```

```
## Creating cohort: Diclofenac
```

```
## |
```

```
## Creating cohort: GiBleed
```

```
## |
```

```
## Creating cohort: NSAIDs
```

```
## |
```

```
## Cohorts created in table main.cohort
```

```
## cohortId      name
```

```
## 1           1 Celecoxib
```

```
## 2           2 Diclofenac
```

```
## 3           3   GiBleed
```

```
## 4           4   NSAIDs
```

```
##
```

```
## 1      A simplified cohort definition for new users of celecoxib, designed specifically for Eunomia.
```

```
## 2      A simplified cohort definition for new users of diclofenac, designed specifically for Eunomia.
```

```
## 3 A simplified cohort definition for gastrointestinal bleeding, designed specifically for Eunomia.
```

```
## 4      A simplified cohort definition for new users of NSAIDs, designed specifically for Eunomia.
```

```
## count
```

```
## 1  1844
```

```
## 2   850
```

```
## 3   479
```

```
## 4  2694
```

```
# -----  
# Points PatientLevelPredictionPackage to the Eunomia database  
# Tells Eunomia to extract the cohort stored with id = 4 as the target cohort  
# and cohort with id = 3 as the outcome cohort. The other settings (...Schema)  
# tell the database where to look for the target and the outcome cohorts  
# -----
```

```
databaseDetails <- PatientLevelPrediction::createDatabaseDetails(  
  connectionDetails = connectionDetails,  
  cdmDatabaseId = "eunomia",  
  cdmDatabaseSchema = 'main',  
  cdmDatabaseName = 'Eunomia',  
  cohortDatabaseSchema = 'main',  
  cohortTable = 'cohort',  
  target = 4,
```

```

outcomeDatabaseSchema = 'main',
outcomeTable = 'cohort',
outcomeId = 3,
cdmVersion = 5
)

# Use ?FeatureExtraction::createCovariateSettings to see what the options are
# There are a lot...
covariateSettings <- FeatureExtraction::createCovariateSettings(
  useDemographicsGender = TRUE,
  useDemographicsAgeGroup = TRUE,
  useConditionGroupEraLongTerm = TRUE,
  useDrugGroupEraLongTerm = TRUE,
  endDays = -1,
  longTermStartDays = -365
)

restrictPlpDataSettings <- PatientLevelPrediction::createRestrictPlpDataSettings(
  studyStartDate = '20000101',
  studyEndDate = '20200101',
  firstExposureOnly = TRUE,
  washoutPeriod = 30
)

# issue with studyStartDate/studyEndDate
restrictPlpDataSettings <- PatientLevelPrediction::createRestrictPlpDataSettings(
  firstExposureOnly = TRUE,
  washoutPeriod = 30
)

plpData <- PatientLevelPrediction::getPlpData(
  databaseDetails = databaseDetails,
  covariateSettings = covariateSettings,
  restrictPlpDataSettings = restrictPlpDataSettings
)

```

```

## |

## Warning: The 'oracleTempSchema' argument is deprecated. Use 'tempEmulationSchema' instead.
## This warning is displayed once every 8 hours.

## Constructing features on server
## |
## Fetching data from server
## Fetching data took 0.379 secs

```

Defining a training set.

For simulating new censored survival times we need more than one probability, we use the baselinehazard, stored in the plpModel, we also need a model for the censoring. First we make the training set.

```

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'
)
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
)
TrainingSet <- PlasmodeSim::MakeTraingSet(
  plpData = plpData,
  executeSettings = executeSettings,
  populationSettings = populationSettings,
  splitSettings = splitSettings,
  sampleSettings = sampleSettings,
  preprocessSettings = preprocessSettings,
  featureEngineeringSettings = featureEngineeringSettings,
  outcomeId = 3
)

```

```
## 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 v
## 103 covariates in train data
## Test Set:
## 656 patients with 119 outcomes
## Removing 2 redundant covariates
## Normalizing covariates
## Tidying covariates took 0.536 secs
## Train Set:
## Fold 1 658 patients with 120 outcomes - Fold 2 658 patients with 120 outcomes - Fold 3 658 patients v
## 101 covariates in train data
## Test Set:
## 656 patients with 119 outcomes
```

Fitting the model

To fit the model we make the modelsettings and we run the function `fitModelWithCensoring`.

```
modelSettings <- PatientLevelPrediction::setCoxModel()

fitCensor <- PlasmodeSim::fitModelWithCensoring(
  Trainingset = TrainingSet$Train,
  modelSettings = modelSettings
  # now i have only one model setting
  # should i change this to two seperate settings
)
```

```
## Running Cyclops
## Done.
## GLM fit status: OK
## Creating variable importance data frame
## Prediction took 0.186 secs
## Running Cyclops
## Done.
## GLM fit status: OK
## Creating variable importance data frame
## Prediction took 0.141 secs
```

Generating new outcomes times

```
population <- PatientLevelPrediction::createStudyPopulation(
  plpData = plpData,
  outcomeId = 3,
  populationSettings = populationSettings
)
```

```
## Outcome is 0 or 1
```

```
NewOutcomes <- PlasmodeSim::simulateSurvivaltimesWithCensoring(
  censorModel = fitCensor,
  plpData = plpData,
  population = population,
  populationSettings = populationSettings,
  numberToSimulate = 10
)
```

```
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.175 secs
## Prediction took 0.268 secs
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.181 secs
## Prediction took 0.268 secs
```

```
head(NewOutcomes)
```

```
##   rowId survivalTime outcomeCount
## 1   2367           18             1
## 2    500          7293             0
## 3   2086            20             1
## 4   2279          7300             0
## 5    287          7300             0
## 6    290          5888             0
```

#for simulation the uncensored data

```
newdata <- PlasmodeSim::simulateSurvivaltimes(
  plpModel = fitCensor$outcomesModel,
  plpData = plpData,
  numberToSimulate = 10,
  population = population,
  populationSettings = populationSettings
)
```

```
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.183 secs
## Prediction took 0.273 secs
```

```
head(newdata)
```

```
##   rowId outcome
## 1   1961    7300
## 2   1134    7300
## 3   1720    7300
## 4   2369    7300
## 5    418    7300
## 6   2599    7300
```

Defining an unfitted model

```

# makeCoxModel.

plpModel <- fitCensor$outcomesModel
coeff <- plpModel$model$coefficients
survival <- plpModel$model$baselineSurvival$surv
times <- plpModel$model$baselineSurvival$time

unfittedmodel <- PlasmodeSim::makeCoxModel(
  coefficients = coeff,
  baselinehazard = survival,
  timesofbaselinhazard = times,
  featureEngineering = NULL # = NULL is the standart setting.
)

newdata <- PlasmodeSim::simulateSurvivaltimes(
  plpModel = unfittedmodel,
  plpData = plpData,
  numberToSimulate = 10,
  population = population,
  populationSettings = populationSettings
)

```

```
## Prediction took 0.207 secs
```

```
head(newdata)
```

```
##   rowId outcome
## 1    54      67
## 2  1229     7300
## 3   552     7300
## 4  2376     7300
## 5  1447     7300
## 6  1709       30
```

Defining an unfitted model with censoring

```

#we can swap outcomes with censoring.
unfittedcensor<- list(censorModel = unfittedmodel,
  outcomesModel = fitCensor$outcomesModel)

NewOutcomes <- PlasmodeSim::simulateSurvivaltimesWithCensoring(
  censorModel = unfittedcensor,
  plpData = plpData,
  population = population,
  populationSettings = populationSettings,
  numberToSimulate = 200
)

```

```
## Removing infrequent and redundant covariates and normalizing
```

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

```
## Prediction took 0.266 secs
## Prediction took 0.202 secs
```

```
head(NewOutcomes)
```

```
##   rowId survivalTime outcomeCount
## 1   658           51             0
## 2  1864          7300             0
## 3  1317          7300             0
## 4  1193           86             1
## 5  1518          7300             0
## 6  1883          7300             0
```

Adjusting the BaselineSurvival

```
adjustedModel <- PlasmodeSim::AdjustBaselineSurvival(
  plpModel = plpModel,
  TrainingSet = TrainingSet$Train,
  plpData = plpData,
  populationSettings = populationSettings,
  timeToFixat = 3592,
  proptoFixwith = 0.87,
  intervalSolution = c(-100,100)
)
```

```
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.18 secs
## Prediction took 0.246 secs
```

```
NewOutcomes <- PlasmodeSim::simulateSurvivaltimesWithCensoring(
  censorModel = list(censorModel = fitCensor$outcomesModel,
                    outcomesModel = adjustedModel),
  plpData = plpData,
  population = population,
  populationSettings = populationSettings,
  numberToSimulate = 2000
)
```

```
## Prediction took 0.197 secs
## Removing infrequent and redundant covariates and normalizing
## Removing infrequent and redundant covariates covariates and normalizing took 0.184 secs
## Prediction took 0.275 secs
```

```
head(NewOutcomes)
```

```
##   rowId survivalTime outcomeCount
## 1   705           56             1
## 2   306           14             1
## 3  1066          7300             0
## 4   678          7300             0
## 5   532           64             0
## 6  1942           58             1
```


Plotting Kaplan Meier estimates

the function `kaplanMeierPlot` visualised the kamplanmeier estimate of a given dataset. It works with `ggplot`. we can easily compare the simulated data sets with the real dataset by putting them in one plot. For the true data set we set the colour to red.

```
NewOutcomes <- PlasmodeSim::simulateSurvivaltimesWithCensoring(  
  censorModel = fitCensor,  
  plpData = plpData,  
  population = population,  
  populationSettings = populationSettings,  
  numberToSimulate = 656  
)
```

```
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.19 secs  
## Prediction took 0.266 secs  
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.191 secs  
## Prediction took 0.27 secs
```

```
NewOutcomes2 <- PlasmodeSim::simulateSurvivaltimesWithCensoring(  
  censorModel = fitCensor,  
  plpData = plpData,  
  population = population,  
  populationSettings = populationSettings,  
  numberToSimulate = 656  
)
```

```
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.179 secs  
## Prediction took 0.269 secs  
## Removing infrequent and redundant covariates and normalizing  
## Removing infrequent and redundant covariates covariates and normalizing took 0.178 secs  
## Prediction took 0.27 secs
```

```
ggplot2::ggplot()+  
  PlasmodeSim::KaplanMeierPlot( NewOutcomes )+  
  PlasmodeSim::KaplanMeierPlot( NewOutcomes2 )+  
  PlasmodeSim::KaplanMeierPlot( TrainingSet$Test$labels, colour = 'red' )+  
  ggplot2::xlim(c(0,120))
```

```
## Warning: Removed 223 rows containing missing values ('geom_step()').
```

```
## Warning: Removed 236 rows containing missing values ('geom_step()').
```

```
## Warning: Removed 283 rows containing missing values ('geom_step()').
```

