Model Selection - readmission 60d

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Global parameters

```
k <- params$k # Number of folds for cross validation
grid_size <- params$grid_size # Number of parameter combination to tune on each model
repeats <- params$repeats
RUN_ALL_MODELS <- params$RUN_ALL_MODELS
Hmisc::list.tree(params)

## params = list 5 (952 bytes)
## . outcome_column = character 1= readmission_60d
## . k = double 1= 10
## . grid_size = double 1= 20
## . repeats = double 1= 2
## . RUN_ALL_MODELS = logical 1= TRUE</pre>
Minutes to run: 0
```

Imports

```
library(tidyverse)
library(yaml)
library(tidymodels)
library(usemodels)
library(vip)
library(bonsai)
library(lightgbm)
library(caret)
library(pROC)

source("aux_functions.R")
predict <- stats::predict</pre>
```

Minutes to run: 0

Loading data

```
load('dataset/processed_data.RData')
load('dataset/processed_dictionary.RData')

columns_list <- yaml.load_file("./auxiliar/columns_list.yaml")

outcome_column <- params$outcome_column
features_list <- params$features_list

df <- mutate(df, across(where(is.character), as.factor))

Minutes to run: 0.005</pre>
```

dir.create(file.path("./auxiliar/model_selection/hyperparameters/"),

showWarnings = FALSE,

Minutes to run: 0

```
Eligible features
cat_features_list = read_yaml(sprintf(
  "./auxiliar/significant_columns/categorical_%s.yaml",
  outcome_column
))
num_features_list = read_yaml(sprintf(
  "./auxiliar/significant_columns/numerical_%s.yaml",
  outcome_column
))
features_list = c(cat_features_list, num_features_list)
Minutes to run: 0
eligible_columns = df_names %>%
  filter(momento.aquisicao == 'Admissão t0') %>%
  .$variable.name
exception_columns = c('death_intraop', 'death_intraop_1', 'disch_outcomes_t0')
correlated_columns = c('year_procedure_1', # com year_adm_t0
                       'age_surgery_1', # com age
                       'admission_t0', # com admission_pre_t0_count
                       'atb', # com meds_antimicrobianos
                       'classe_meds_cardio_qtde', # com classe_meds_qtde
                       'suporte_hemod', # com proced_invasivos_qtde,
                       'radiografia', # com exames_imagem_qtde
                       'ecg' # com metodos_graficos_qtde
                       )
eligible_features = eligible_columns %>%
  base::intersect(c(columns_list$categorical_columns, columns_list$numerical_columns)) %>%
  setdiff(c(exception_columns, correlated_columns))
features = base::intersect(eligible_features, features_list)
gluedown::md_order(features, seq = TRUE, pad = TRUE)
## 01. age
## 02. education_level
## 03. underlying_heart_disease
## 04. heart_disease
## 05. nyha_basal
## 06. prior_mi
## 07. heart_failure
## 08. af
## 09. cardiac_arrest
## 10. transplant
## 11. valvopathy
## 12. diabetes
## 13. hemodialysis
```

- ## 14. comorbidities_count
- ## 15. procedure_type_1
- ## 16. reop_type_1
- ## 17. procedure_type_new
- ## 18. cied_final_1
- ## 19. cied_final_group_1
- ## 20. admission_pre_t0_count
- ## 21. admission_pre_t0_180d
- ## 22. icu_t0
- ## 23. dialysis_t0
- ## 24. admission_t0_emergency
- ## 25. aco
- ## 26. antiarritmico
- ## 27. betabloqueador
- ## 28. ieca_bra
- ## 29. dva
- ## 30. digoxina
- ## 31. estatina
- ## 32. diuretico
- ## 33. vasodilatador
- ## 34. insuf_cardiaca
- ## 35. espironolactona
- ## 36. bloq_calcio
- ## 37. antiplaquetario_ev
- ## 38. insulina
- ## 39. anticonvulsivante
- ## 40. psicofarmacos
- ## 41. antifungico
- ## 42. antiviral
- ## 43. classe_meds_qtde
- ## 44. meds_cardiovasc_qtde
- ## 45. meds_antimicrobianos
- ## 46. ventilacao_mecanica
- ## 47. cec
- ## 48. transplante_cardiaco
- ## 49. cir_toracica
- ## 50. outros_proced_cirurgicos
- ## 51. icp
- ## 52. angioplastia
- ## 53. cateterismo
- ## 54. eletrofisiologia
- ## 55. cateter_venoso_central
- ## 56. proced_invasivos_qtde
- ## 57. cve_desf
- ## 58. transfusao
- ## 59. interconsulta
- ## 60. equipe_multiprof
- ## 61. holter
- ## 62. teste_esforco
- ## 63. espiro_ergoespiro
- ## 64. tilt_teste
- ## 65. metodos_graficos_qtde
- ## 66. laboratorio
- ## 67. cultura
- ## 68. analises_clinicas_qtde
- ## 69. citologia
- ## 70. biopsia
- ## 71. histopatologia_qtde
- ## 72. angio_rm
- ## 73. angio_tc
- ## 74. arteriografia

```
## 75. cintilografia
## 76. ecocardiograma
## 77. endoscopia
## 78. pet_ct
## 79. ultrassom
## 80. tomografia
## 81. ressonancia
## 82. exames_imagem_qtde
## 83. bic
## 84. hospital_stay
Minutes to run: 0
```

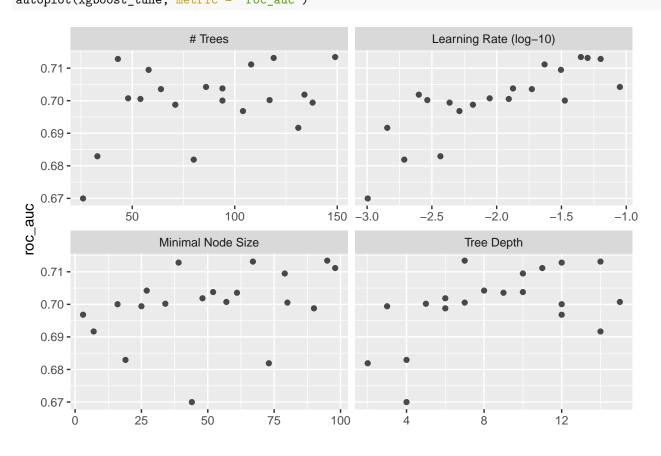
Train test split (70%/30%)

Minutes to run: 0.001

Boosted Tree (XGBoost)

```
xgboost_recipe <-</pre>
 recipe(formula = sprintf("%s ~ .", outcome_column) %>% as.formula, data = df_train) %>%
  step_novel(all_nominal_predictors()) %>%
  step_unknown(all_nominal_predictors()) %>%
  step_other(all_nominal_predictors(), threshold = 0.05, other = ".merged") %>%
  step_dummy(all_nominal_predictors())
xgboost_spec <- boost_tree(</pre>
 trees = tune(),
 min_n = tune(),
 tree_depth = tune(),
  learn_rate = tune(),
) %>%
  set_engine("xgboost",
            nthread = 8) %>%
  set_mode("classification")
xgboost_grid <- grid_latin_hypercube(</pre>
 trees(range = c(25L, 150L)),
 min_n(range = c(2L, 100L)),
 tree_depth(range = c(2L, 15L)),
 learn_rate(range = c(-3, -1), trans = log10_trans()),
  size = grid_size
xgboost_workflow <-</pre>
 workflow() %>%
```

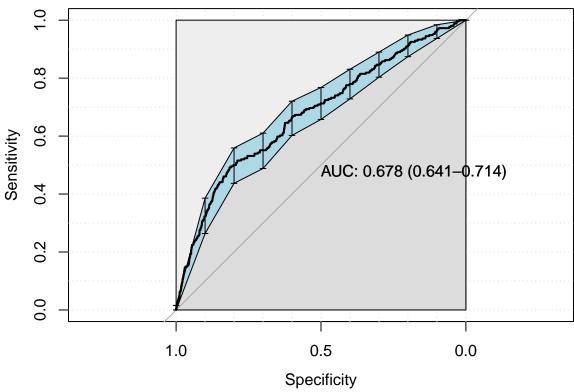
```
add_recipe(xgboost_recipe) %>%
  add_model(xgboost_spec)
xgboost_tune <-
  xgboost_workflow %>%
  tune_grid(resamples = df_folds,
            grid = xgboost_grid)
xgboost_tune %>%
  show_best("roc_auc")
## # A tibble: 5 x 10
     trees min_n tree_depth learn_rate .metric .estimator
                                                            mean
                                                                      n std_err .config
##
                      <int>
                                  <dbl> <chr>
                                                <chr>
                                                                          <dbl> <chr>
     <int> <int>
                                                            <dbl> <int>
## 1
       149
              95
                          7
                                 0.0447 roc_auc binary
                                                           0.713
                                                                     10 0.0122 Preprocessor1_Model19
##
  2
       119
              67
                         14
                                 0.0503 roc_auc binary
                                                           0.713
                                                                     10 0.0118 Preprocessor1_Model14
##
  3
        43
              39
                         12
                                 0.0635 roc_auc binary
                                                           0.713
                                                                     10 0.0115 Preprocessor1_Model08
##
  4
       108
              98
                         11
                                                           0.711
                                                                     10 0.0120 Preprocessor1_Model20
                                 0.0234 roc_auc binary
## 5
        58
              79
                         10
                                 0.0313 roc_auc binary
                                                           0.709
                                                                     10 0.0123 Preprocessor1_Model16
best_xgboost <- xgboost_tune %>%
  select_best("roc_auc")
autoplot(xgboost_tune, metric = "roc_auc")
```



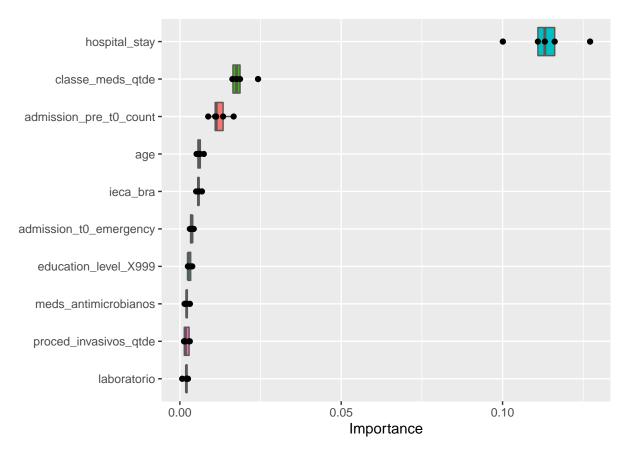
```
final_xgboost_workflow <-
    xgboost_workflow %>%
    finalize_workflow(best_xgboost)

last_xgboost_fit <-
    final_xgboost_workflow %>%
    last_fit(df_split)

final_xgboost_fit <- extract_workflow(last_xgboost_fit)</pre>
```



```
## [1] "Optimal Threshold: 0.09"
  Confusion Matrix and Statistics
##
##
       reference
##
  data
      0 3650 129
##
##
      1 826 125
##
##
                  Accuracy : 0.7981
                    95% CI : (0.7864, 0.8095)
##
##
      No Information Rate: 0.9463
       P-Value [Acc > NIR] : 1
##
##
##
                     Kappa : 0.1341
##
   Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity: 0.8155
##
##
               Specificity: 0.4921
            Pos Pred Value: 0.9659
##
            Neg Pred Value : 0.1314
##
                Prevalence: 0.9463
##
            Detection Rate: 0.7717
##
      Detection Prevalence: 0.7989
##
         Balanced Accuracy: 0.6538
##
##
          'Positive' Class : 0
##
extract_vip(final_xgboost_fit, pred_wrapper = predict,
            reference_class = "0")
```



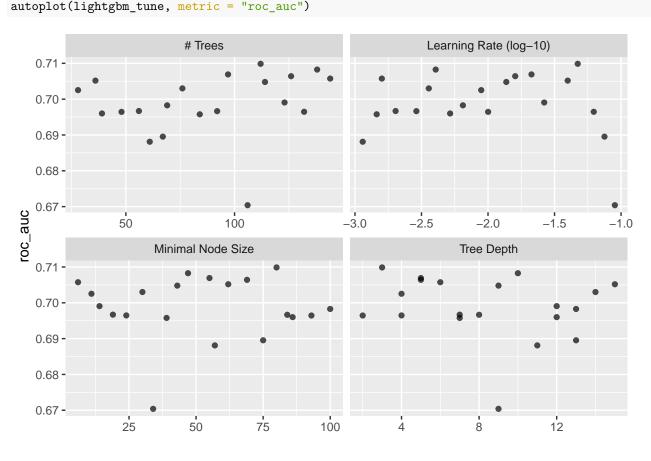
```
xgboost_parameters <- xgboost_tune %>%
show_best("roc_auc", n = 1) %>%
select(-.metric, -.estimator, -.config, -mean, -n, -std_err) %>%
as.list
```

Minutes to run: 4.706

Boosted Tree (LightGBM)

```
lightgbm_recipe <-</pre>
  recipe(formula = sprintf("%s ~ .", outcome_column) %% as.formula, data = df_train) %>%
  step_novel(all_nominal_predictors()) %>%
  step_unknown(all_nominal_predictors()) %>%
  step_other(all_nominal_predictors(), threshold = 0.05, other = ".merged") %>%
  step_dummy(all_nominal_predictors())
lightgbm_spec <- boost_tree(</pre>
  trees = tune(),
 min_n = tune(),
 tree_depth = tune(),
 learn_rate = tune(),
  sample_size = 1
) %>%
  set_engine("lightgbm",
             nthread = 8) %>%
  set_mode("classification")
lightgbm_grid <- grid_latin_hypercube(</pre>
  trees(range = c(25L, 150L)),
  min_n(range = c(2L, 100L)),
  tree_depth(range = c(2L, 15L)),
  learn_rate(range = c(-3, -1), trans = log10_trans()),
  size = grid_size
```

```
)
lightgbm_workflow <-
  workflow() %>%
  add_recipe(lightgbm_recipe) %>%
  add_model(lightgbm_spec)
lightgbm_tune <-
  lightgbm_workflow %>%
  tune_grid(resamples = df_folds,
            grid = lightgbm_grid)
lightgbm_tune %>%
  show_best("roc_auc")
## # A tibble: 5 x 10
##
     trees min_n tree_depth learn_rate .metric .estimator
                                                            mean
                                                                      n std_err .config
##
     <int> <int>
                      <int>
                                  <dbl> <chr>
                                                <chr>>
                                                            <dbl> <int>
                                                                          <dbl> <chr>
##
  1
       112
              80
                                0.0473 roc_auc binary
                                                           0.710
                                                                     10 0.0134 Preprocessor1_Model16
##
  2
       138
              47
                         10
                                                                     10 0.0112 Preprocessor1_Model10
                                0.00404 roc_auc binary
                                                           0.708
##
  3
        97
              55
                          5
                                0.0212 roc_auc binary
                                                           0.707
                                                                     10 0.0124 Preprocessor1_Model11
       126
                                                           0.706
## 4
              69
                          5
                                0.0160 roc_auc binary
                                                                     10 0.0117 Preprocessor1_Model14
## 5
       144
               6
                          6
                               0.00159 roc_auc binary
                                                           0.706
                                                                     10 0.0130 Preprocessor1_Model01
best_lightgbm <- lightgbm_tune %>%
  select_best("roc_auc")
```



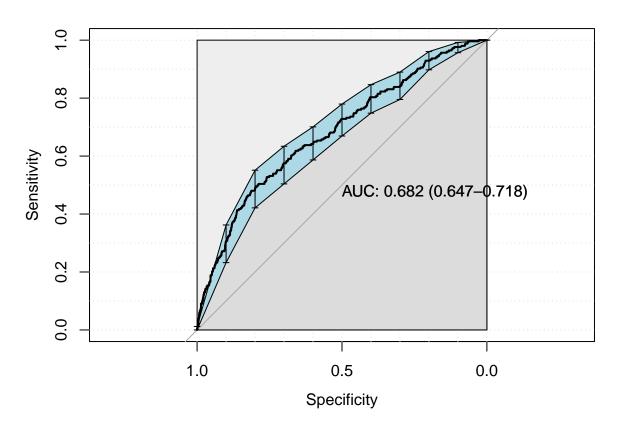
```
final_lightgbm_workflow <-
    lightgbm_workflow %>%
    finalize_workflow(best_lightgbm)

last_lightgbm_fit <-</pre>
```

```
final_lightgbm_workflow %>%
  last_fit(df_split)

final_lightgbm_fit <- extract_workflow(last_lightgbm_fit)

lightgbm_auc <- validation(final_lightgbm_fit, df_test)</pre>
```

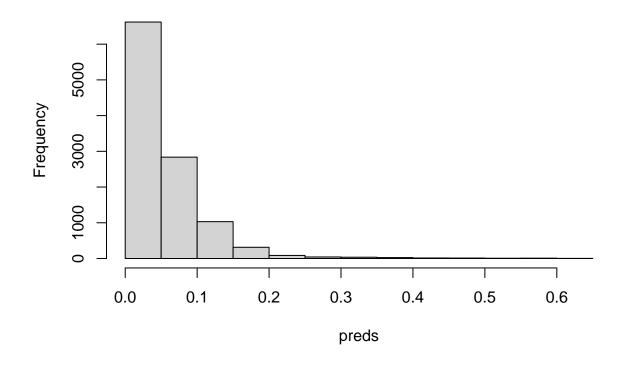


```
## [1] "Optimal Threshold: 0.08"
## Confusion Matrix and Statistics
##
##
       reference
##
  data
           0
                1
##
      0 3640 132
      1 836 122
##
##
##
                  Accuracy : 0.7953
##
                    95% CI: (0.7836, 0.8068)
##
      No Information Rate: 0.9463
##
      P-Value [Acc > NIR] : 1
##
##
                     Kappa : 0.1272
##
   Mcnemar's Test P-Value : <2e-16
##
##
##
               Sensitivity: 0.8132
##
               Specificity: 0.4803
##
            Pos Pred Value: 0.9650
##
            Neg Pred Value : 0.1273
##
                Prevalence: 0.9463
##
            Detection Rate: 0.7696
      Detection Prevalence: 0.7975
##
##
         Balanced Accuracy: 0.6468
##
```

```
##
          'Positive' Class: 0
##
lightgbm_parameters <- lightgbm_tune %>%
  show_best("roc_auc", n = 1) %>%
  select(-.metric, -.estimator, -.config, -mean, -n, -std_err) %>%
  as.list
Hmisc::list.tree(lightgbm_parameters)
   lightgbm_parameters = list 4 (736 bytes)
## . trees = integer 1= 112
## . min_n = integer 1= 80
## . tree_depth = integer 1= 3
## . learn_rate = double 1= 0.047266
con <- file(sprintf('./auxiliar/model_selection/hyperparameters/%s.yaml', outcome_column), "w")</pre>
write_yaml(lightgbm_parameters, con)
close(con)
```

Minutes to run: 3.685

Histogram of preds



Minutes to run:

0.012

GLM

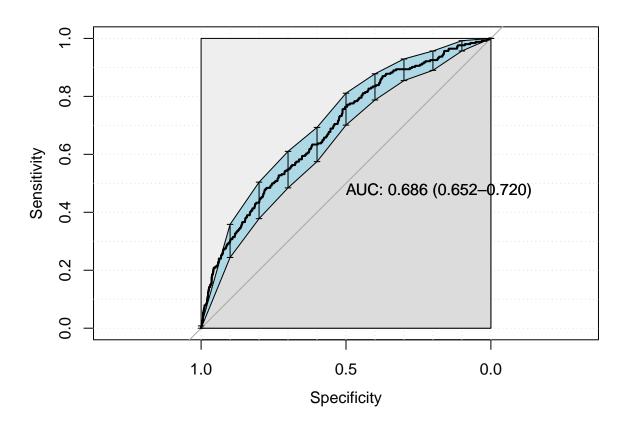
```
glmnet_recipe <-
    recipe(formula = sprintf("%s ~ .", outcome_column) %>% as.formula, data = df_train) %>%
    step_novel(all_nominal_predictors()) %>%
    step_unknown(all_nominal_predictors()) %>%
    step_other(all_nominal_predictors(), threshold = 0.05, other = ".merged") %>%
    step_dummy(all_nominal_predictors()) %>%
    step_zv(all_predictors()) %>%
    step_normalize(all_numeric_predictors())
```

```
glmnet_spec <-
  logistic_reg(penalty = 0) %>%
  set_mode("classification") %>%
  set_engine("glmnet")

glmnet_workflow <-
  workflow() %>%
  add_recipe(glmnet_recipe) %>%
  add_model(glmnet_spec)

glm_fit <- glmnet_workflow %>%
  fit(df_train)

glmnet_auc <- validation(glm_fit, df_test)</pre>
```



```
## [1] "Optimal Threshold: 0.05"
  Confusion Matrix and Statistics
##
##
       reference
##
  data
           0
                1
##
      0 2291
               62
      1 2185 192
##
##
                  Accuracy : 0.5249
##
                    95% CI: (0.5106, 0.5393)
##
      No Information Rate: 0.9463
##
      P-Value [Acc > NIR] : 1
##
##
##
                     Kappa : 0.0542
##
##
    Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity: 0.51184
```

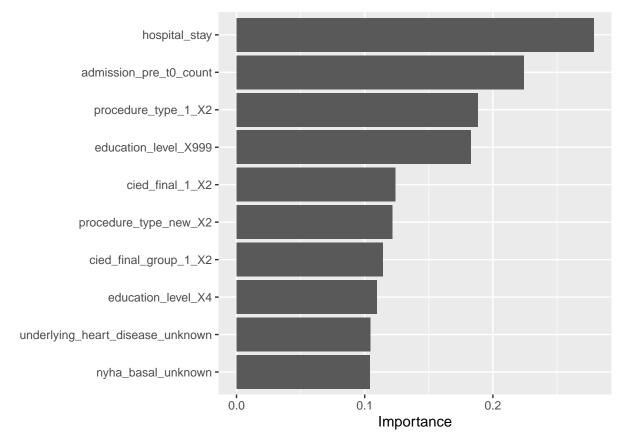
```
## Neg Pred Value : 0.08077
## Prevalence : 0.94630
## Detection Rate : 0.48436
## Detection Prevalence : 0.49746
## Balanced Accuracy : 0.63387
##
## 'Positive' Class : 0
##

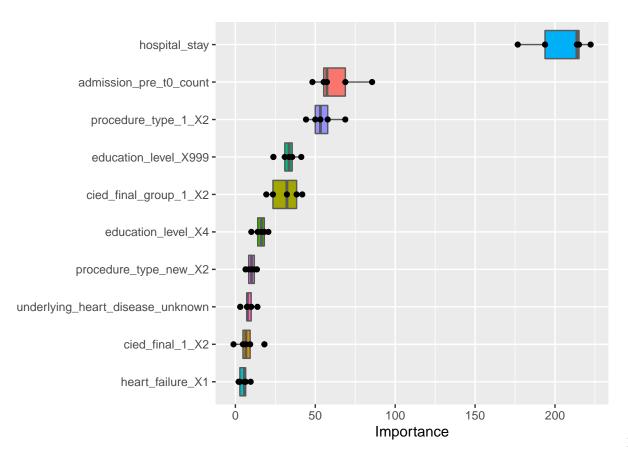
pfun_glmnet <- function(object, newdata) predict(object, newx = newdata)</pre>
```

Specificity: 0.75591 Pos Pred Value: 0.97365

##

##





Minutes to run:

1.636

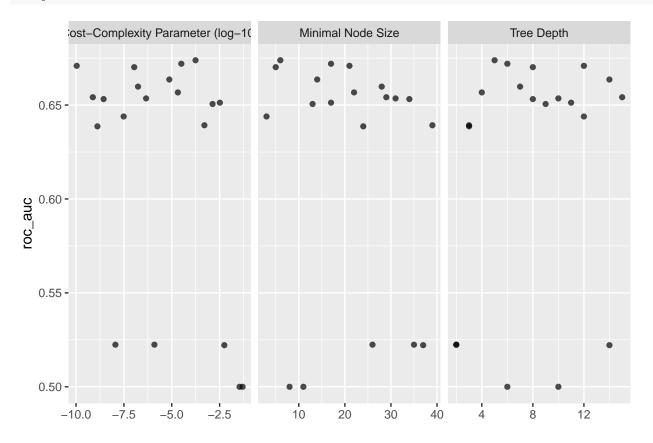
Decision Tree

```
tree_recipe <-
 recipe(formula = sprintf("%s ~ .", outcome_column) %>% as.formula, data = df_train) %>%
  step_novel(all_nominal_predictors()) %>%
  step_unknown(all_nominal_predictors()) %>%
  step_other(all_nominal_predictors(), threshold = 0.05, other = ".merged") %>%
  step_dummy(all_nominal_predictors()) %>%
  step_zv(all_predictors())
tree_spec <-
  decision_tree(cost_complexity = tune(),
                tree_depth = tune(),
                min_n = tune()) %>%
  set mode("classification") %>%
 set_engine("rpart")
tree_grid <- grid_latin_hypercube(cost_complexity(),</pre>
                                  tree_depth(),
                                  min_n(),
                                  size = grid_size)
tree_workflow <-
  workflow() %>%
  add_recipe(tree_recipe) %>%
  add_model(tree_spec)
tree_tune <-
 tree_workflow %>%
 tune_grid(resamples = df_folds,
            grid = tree_grid)
```

```
tree_tune %>%
  collect_metrics()
```

```
# A tibble: 40 x 9
##
      cost_complexity tree_depth min_n .metric
                                                 .estimator mean
                                                                       n std_err .config
                           <int> <int> <chr>
                                                                           <dbl> <chr>
##
                <dbl>
                                                 <chr>>
                                                             <dbl> <int>
                                                                      10 0.00165 Preprocessor1_Model01
             1.08e-10
                                                             0.932
##
    1
                              12
                                     21 accuracy binary
                                     21 roc_auc binary
##
    2
             1.08e-10
                              12
                                                             0.671
                                                                      10 0.0121 Preprocessor1_Model01
             3.07e-8
                                                                      10 0.00227 Preprocessor1_Model02
##
    3
                              12
                                     3 accuracy binary
                                                            0.930
                                     3 roc_auc binary
                                                                      10 0.0160 Preprocessor1_Model02
             3.07e-8
                              12
                                                             0.644
             1.24e- 6
                               2
                                     26 accuracy binary
##
    5
                                                            0.940
                                                                      10 0.00189 Preprocessor1_Model03
##
    6
             1.24e- 6
                               2
                                     26 roc_auc binary
                                                            0.522
                                                                      10 0.0159 Preprocessor1_Model03
                               6
##
    7
             3.16e- 5
                                     17 accuracy binary
                                                            0.939
                                                                      10 0.00180 Preprocessor1_Model04
             3.16e- 5
    8
                               6
                                     17 roc_auc binary
                                                            0.672
                                                                      10 0.0162 Preprocessor1_Model04
             7.47e- 6
                                                            0.929
##
    9
                              14
                                     14 accuracy binary
                                                                      10 0.00144 Preprocessor1_Model05
##
  10
             7.47e- 6
                              14
                                     14 roc_auc binary
                                                            0.664
                                                                      10 0.0129 Preprocessor1_Model05
  # ... with 30 more rows
## # i Use 'print(n = ...)' to see more rows
```

autoplot(tree_tune, metric = "roc_auc")



```
tree_tune %>%
show_best("roc_auc")
```

```
## # A tibble: 5 x 9
##
     cost_complexity tree_depth min_n .metric .estimator mean
                                                                     n std_err .config
##
               <dbl>
                          <int> <int> <chr>
                                               <chr>>
                                                           <dbl> <int>
                                                                         <dbl> <chr>
##
            1.75e- 4
                               5
                                     6 roc_auc binary
                                                           0.674
                                                                    10 0.0140 Preprocessor1_Model07
  1
            3.16e- 5
##
  2
                               6
                                    17 roc_auc binary
                                                           0.672
                                                                        0.0162 Preprocessor1_Model04
##
  3
            1.08e-10
                              12
                                    21 roc_auc binary
                                                           0.671
                                                                       0.0121 Preprocessor1_Model01
##
  4
            1.09e- 7
                               8
                                     5 roc_auc binary
                                                           0.670
                                                                        0.0164 Preprocessor1_Model10
            7.47e- 6
                                                           0.664
## 5
                              14
                                    14 roc_auc binary
                                                                    10 0.0129 Preprocessor1_Model05
```

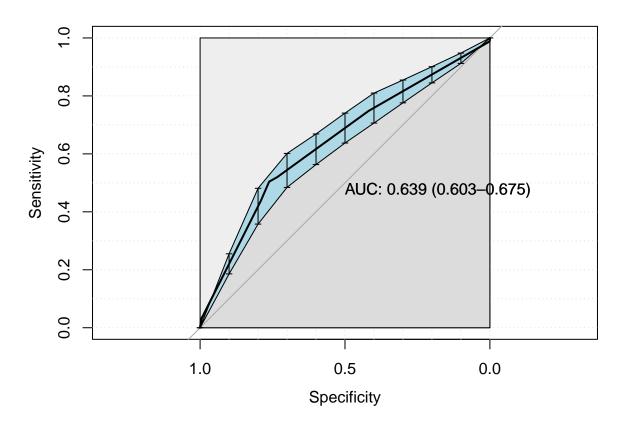
```
best_tree <- tree_tune %>%
    select_best("roc_auc")

final_tree_workflow <-
    tree_workflow %>%
    finalize_workflow(best_tree)

last_tree_fit <-
    final_tree_workflow %>%
    last_fit(df_split)

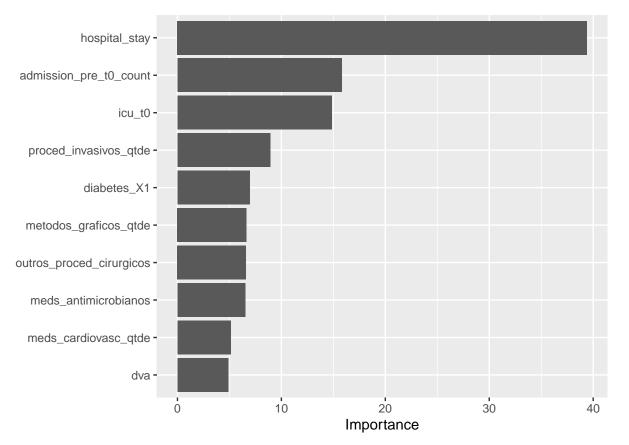
final_tree_fit <- extract_workflow(last_tree_fit)

tree_auc <- validation(final_tree_fit, df_test)</pre>
```



```
## [1] "Optimal Threshold: 0.08"
## Confusion Matrix and Statistics
##
##
      reference
          0 1
##
  data
     0 3410 126
##
##
      1 1066 128
##
##
                  Accuracy: 0.748
##
                    95% CI : (0.7354, 0.7603)
##
      No Information Rate: 0.9463
      P-Value [Acc > NIR] : 1
##
##
##
                     Kappa: 0.0968
##
   Mcnemar's Test P-Value : <2e-16
##
##
##
               Sensitivity: 0.7618
```

```
##
            Neg Pred Value : 0.1072
##
                Prevalence: 0.9463
##
            Detection Rate: 0.7209
##
      Detection Prevalence: 0.7476
##
         Balanced Accuracy: 0.6329
##
##
          'Positive' Class: 0
##
extract_vip(final_tree_fit, pred_wrapper = predict,
            reference_class = "0", use_matrix = FALSE,
            method = 'model')
```



Minutes to run: 4.487

##

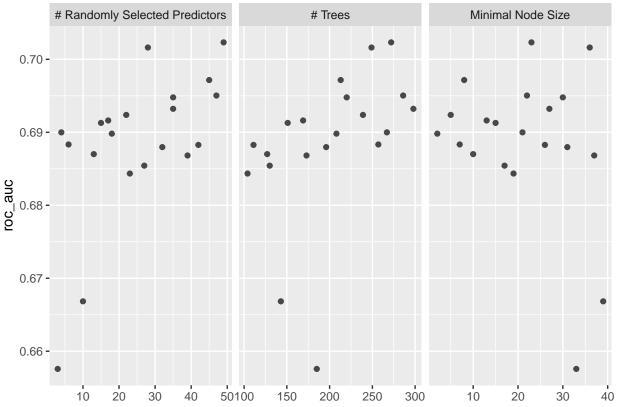
##

Specificity: 0.5039

Pos Pred Value: 0.9644

Random Forest

```
rand_forest(mtry = tune(),
            trees = tune(),
            min_n = tune()) \%
  set_mode("classification") %>%
  set_engine("randomForest",
            probability = TRUE,
            nthread = 8)
rf_grid <- grid_latin_hypercube(mtry(range = c(1L, 50L)),</pre>
                              trees(range = c(100L, 300L)),
                             min_n(),
                             size = grid_size)
rf_workflow <-
  workflow() %>%
  add_recipe(rf_recipe) %>%
  add_model(rf_spec)
rf_tune <-
 rf workflow %>%
 tune_grid(resamples = df_folds,
           grid = rf_grid)
rf_tune %>%
  collect_metrics()
## # A tibble: 40 x 9
##
      mtry trees min_n .metric .estimator mean n std_err .config
##
     <int> <int> <int> <chr>
                            <chr> <dbl> <int> <dbl> <chr>
                 39 accuracy binary 0.941 10 0.00192 Preprocessor1_Model01 39 roc_auc binary 0.667 10 0.0148 Preprocessor1_Model01
## 1
      10 143
##
   2
        10 143
## 3
      35 298 27 accuracy binary
                                         0.940 10 0.00203 Preprocessor1_Model02
## 4
      35 298 27 roc_auc binary
                                         42 111
                                         0.940 10 0.00196 Preprocessor1_Model03
##
   5
                  26 accuracy binary
## 6
      42 111
                   26 roc_auc binary
                                         0.688 10 0.0165 Preprocessor1_Model03
## 7
            272
      49
                   23 accuracy binary
                                         0.939 10 0.00190 Preprocessor1_Model04
## 8
        49
            272
                   23 roc_auc binary
                                         ## 9
        17
             169
                   13 accuracy binary
                                         0.941 10 0.00218 Preprocessor1_Model05
## 10
        17
             169
                   13 roc_auc binary
                                         0.692
                                                 10 0.0119 Preprocessor1_Model05
## # ... with 30 more rows
## # i Use 'print(n = ...)' to see more rows
autoplot(rf_tune, metric = "roc_auc")
```



```
rf_tune %>%
  show_best("roc_auc")
## # A tibble: 5 x 9
##
      mtry trees min_n .metric .estimator mean
                                                      n std_err .config
##
     <int> <int> <int> <chr>
                                <chr>
                                            <dbl> <int>
                                                          <dbl> <chr>
## 1
        49
             272
                     23 roc_auc binary
                                            0.702
                                                     10 0.0123 Preprocessor1_Model04
                     36 roc_auc binary
                                                     10 0.0112 Preprocessor1_Model10
## 2
             249
                                            0.702
        28
## 3
        45
             213
                     8 roc_auc binary
                                            0.697
                                                     10 0.0164 Preprocessor1_Model14
                                            0.695
## 4
        47
             286
                     22 roc_auc binary
                                                     10 0.0142 Preprocessor1_Model19
## 5
        35
             220
                     30 roc_auc binary
                                            0.695
                                                     10 0.0138 Preprocessor1_Model15
best_rf <- rf_tune %>%
  select_best("roc_auc")
final_rf_workflow <-</pre>
  rf_workflow %>%
  finalize_workflow(best_rf)
last_rf_fit <-</pre>
  final_rf_workflow %>%
  last_fit(df_split)
final_rf_fit <- extract_workflow(last_rf_fit)</pre>
rf_auc <- validation(final_rf_fit, df_test)</pre>
```

```
Sensitivity

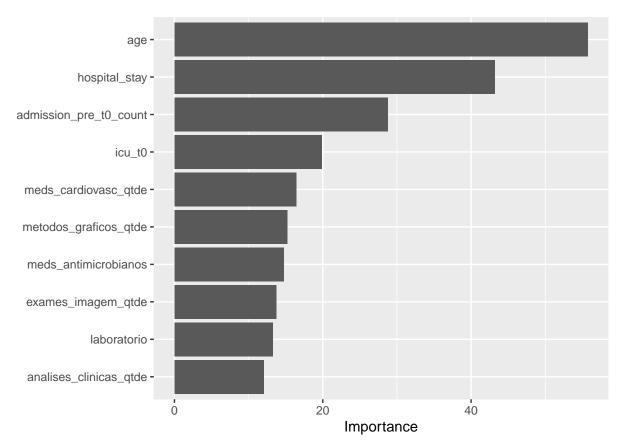
AUC: 0.664 (0.628–0.700)

1.0 0.5 0.0

Specificity
```

[1] "Optimal Threshold: 0.01"

```
Confusion Matrix and Statistics
##
##
       reference
##
  data
           0
##
      0 2450
               78
##
      1 2026 176
##
                  Accuracy : 0.5552
##
                    95% CI: (0.5409, 0.5694)
##
##
       No Information Rate: 0.9463
       P-Value [Acc > NIR] : 1
##
##
##
                     Kappa : 0.052
##
##
    Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity : 0.54736
##
               Specificity: 0.69291
##
            Pos Pred Value: 0.96915
##
            Neg Pred Value: 0.07993
##
                Prevalence: 0.94630
##
            Detection Rate: 0.51797
##
      Detection Prevalence: 0.53446
##
         Balanced Accuracy: 0.62014
##
##
          'Positive' Class: 0
pfun_rf <- function(object, newdata) predict(object, data = newdata)</pre>
extract_vip(final_rf_fit, pred_wrapper = predict,
            reference_class = "1", use_matrix = FALSE,
            method = 'model')
```

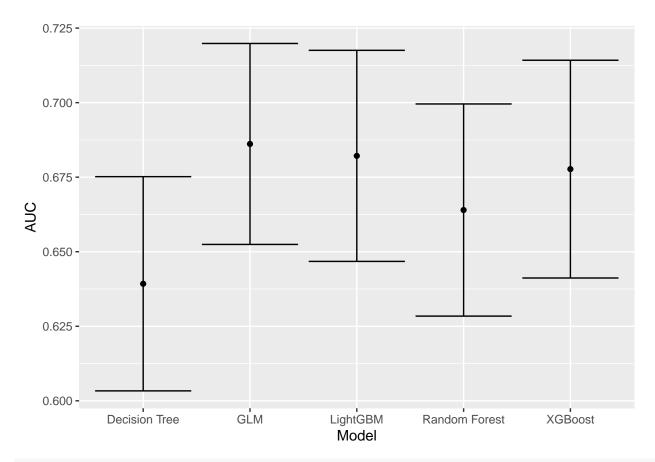


```
# extract_vip(final_rf_fit, pred_wrapper = predict,
# reference_class = "1", use_matrix = FALSE,
# method = 'permute')
```

Minutes to run: 62.692

Models Comparison

```
if (RUN_ALL_MODELS) {
  df_auc <- tibble::tribble(</pre>
    ~Model, ~`AUC`, ~`Lower Limit`, ~`Upper Limit`,
    'XGBoost', as.numeric(xgboost_auc$auc), xgboost_auc$ci[1], xgboost_auc$ci[3],
    'LightGBM', as.numeric(lightgbm_auc$auc), lightgbm_auc$ci[1], lightgbm_auc$ci[3],
    'GLM', as.numeric(glmnet_auc$auc), glmnet_auc$ci[1], glmnet_auc$ci[3],
    'Decision Tree', as.numeric(tree_auc$auc), tree_auc$ci[1], tree_auc$ci[3],
    'Random Forest', as.numeric(rf_auc$auc), rf_auc$ci[1], rf_auc$ci[3]
  ) %>%
    mutate(Target = outcome_column)
} else {
  df_auc <- tibble::tribble(</pre>
    ~Model, ~`AUC`, ~`Lower Limit`, ~`Upper Limit`,
    'LightGBM', as.numeric(lightgbm_auc$auc), lightgbm_auc$ci[1], lightgbm_auc$ci[3]
  ) %>%
    mutate(Target = outcome_column)
}
df_auc %>%
  ggplot(aes(x = Model, y = AUC, ymin = `Lower Limit`, ymax = `Upper Limit`)) +
    geom_point() +
    geom_errorbar()
```



write_csv(df_auc, sprintf("./auxiliar/model_selection/performance/%s.csv", outcome_column))

Minutes to run: 0.002