Model Selection - readmission 180d

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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 (968 bytes)
## . outcome_column = character 1= readmission_180d
## . 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 = readRDS(sprintf(
  "./auxiliar/significant_columns/categorical_%s.rds",
  outcome_column
))
num_features_list = readRDS(sprintf(
  "./auxiliar/significant_columns/numerical_%s.rds",
  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. sex
## 02. age
## 03. education_level
## 04. patient_state
## 05. underlying_heart_disease
## 06. heart_disease
## 07. nyha_basal
## 08. prior_mi
## 09. heart_failure
## 10. af
## 11. cardiac_arrest
## 12. transplant
## 13. valvopathy
```

- ## 14. endocardites
- ## 15. diabetes
- ## 16. renal_failure
- ## 17. hemodialysis
- ## 18. copd
- ## 19. comorbidities_count
- ## 20. procedure_type_1
- ## 21. reop_type_1
- ## 22. procedure_type_new
- ## 23. cied_final_1
- ## 24. cied_final_group_1
- ## 25. admission_pre_t0_count
- ## 26. admission_pre_t0_180d
- ## 27. icu_t0
- ## 28. dialysis_t0
- ## 29. n_procedure_t0
- ## 30. admission_t0_emergency
- ## 31. aco
- ## 32. antiarritmico
- ## 33. betabloqueador
- ## 34. ieca_bra
- ## 35. dva
- ## 36. digoxina
- ## 37. estatina
- ## 38. diuretico
- ## 39. vasodilatador
- ## 40. insuf_cardiaca
- ## 41. espironolactona
- ## 42. bloq_calcio
- ## 43. antiplaquetario_ev
- ## 44. insulina
- ## 45. anticonvulsivante
- ## 46. psicofarmacos
- ## 47. antifungico
- ## 48. antiviral
- ## 49. antiretroviral
- ## 50. classe_meds_qtde
- ## 51. meds_cardiovasc_qtde
- ## 52. meds_antimicrobianos
- ## 53. vni
- ## 54. ventilacao_mecanica
- ## 55. cec
- ## 56. transplante_cardiaco
- ## 57. cir_toracica
- ## 58. outros_proced_cirurgicos
- ## 59. icp
- ## 60. intervencao_cv
- ## 61. angioplastia
- ## 62. cateterismo
- ## 63. eletrofisiologia
- ## 64. cateter_venoso_central
- ## 65. proced_invasivos_qtde
- ## 66. cve desf
- ## 67. transfusao
- ## 68. interconsulta
- ## 69. equipe_multiprof
- ## 70. holter
- ## 71. teste_esforco
- ## 72. espiro_ergoespiro
- ## 73. tilt_teste
- ## 74. metodos_graficos_qtde

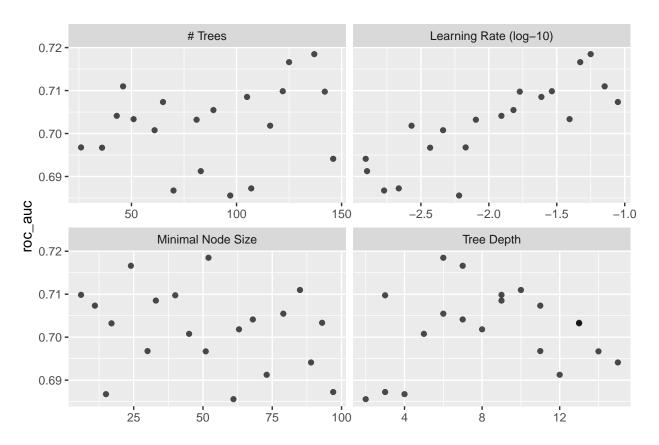
```
## 75. laboratorio
## 76. cultura
## 77. analises_clinicas_qtde
## 78. citologia
## 79. biopsia
## 80. histopatologia_qtde
## 81. angio_rm
## 82. angio_tc
## 83. arteriografia
## 84. cintilografia
## 85. ecocardiograma
## 86. endoscopia
## 87. flebografia
## 88. pet_ct
## 89. ultrassom
## 90. tomografia
## 91. ressonancia
## 92. exames_imagem_qtde
## 93. bic
## 94. mpp
## 95. hospital_stay
Minutes to run: 0
```

Train test split (70%/30%)

Minutes to run: 0.001

Boosted Tree (XGBoost)

```
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 <-</pre>
 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> <int> <dbl> <chr> <chr>
                                                <dbl> <int> <dbl> <chr>
## 1 137 52
                  6 0.0562 roc_auc binary
                                                125 24
                    7 0.0471 roc_auc binary
                                                ## 2
                                                ## 3
      46
          85
                    10
                          0.0712 roc_auc binary
## 4
     122
           6
                    9
                          0.0291 roc_auc binary
                                                       10 0.0138 Preprocessor1_Model01
                                                0.710
## 5
     142
           40
                     3
                          0.0169 roc_auc binary
                                                0.710
                                                       10 0.0104 Preprocessor1_Model08
best_xgboost <- xgboost_tune %>%
 select_best("roc_auc")
autoplot(xgboost_tune, metric = "roc_auc")
```

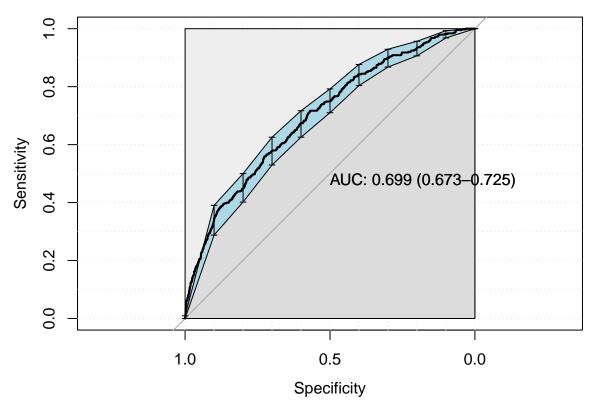


```
final_xgboost_workflow <-
    xgboost_workflow %>%
    finalize_workflow(best_xgboost)

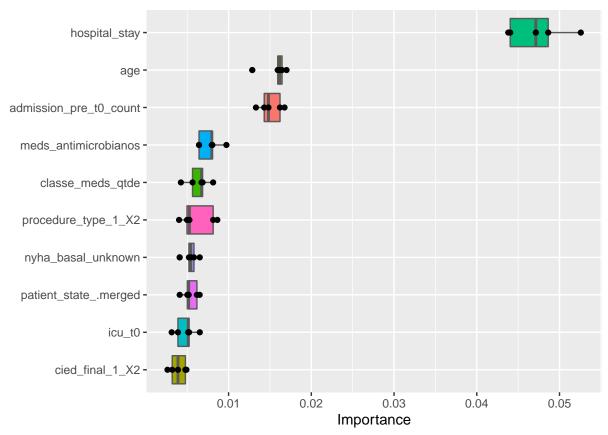
last_xgboost_fit <-
    final_xgboost_workflow %>%
    last_fit(df_split)

final_xgboost_fit <- extract_workflow(last_xgboost_fit)

xgboost_auc <- validation(final_xgboost_fit, df_test)</pre>
```



```
## [1] "Optimal Threshold: 0.10"
##
  Confusion Matrix and Statistics
##
       reference
##
  data
           0
##
      0 3118
              193
##
      1 1174 245
##
##
                  Accuracy: 0.711
                    95% CI : (0.6978, 0.7239)
##
##
      No Information Rate: 0.9074
      P-Value [Acc > NIR] : 1
##
##
##
                     Kappa : 0.1425
##
##
   Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity: 0.7265
##
               Specificity: 0.5594
##
            Pos Pred Value: 0.9417
            Neg Pred Value: 0.1727
##
##
                Prevalence: 0.9074
##
            Detection Rate: 0.6592
##
      Detection Prevalence : 0.7000
##
         Balanced Accuracy : 0.6429
##
##
          'Positive' Class: 0
##
extract_vip(final_xgboost_fit, pred_wrapper = predict,
            reference_class = "0")
```

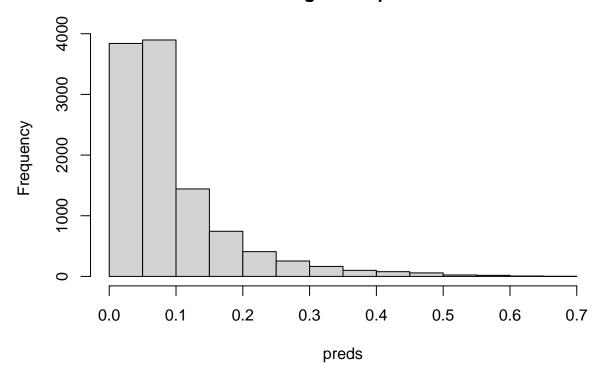


```
xgboost_parameters <- xgboost_tune %>%
show_best("roc_auc", n = 1) %>%
select(-.metric, -.estimator, -.config, -mean, -n, -std_err) %>%
as.list

# saveRDS(
# xgboost_parameters,
file = sprintf(
# "./auxiliar/model_selection/hyperparameters/xgboost_%s.rds",
# outcome_column
# )
# )

preds <- predict(final_xgboost_fit, new_data = df_train, type = "prob") %>%
rename_at(vars(starts_with(".pred_")), ~ str_remove(., ".pred_")) %>%
.$`1`
hist(preds)
```

Histogram of preds



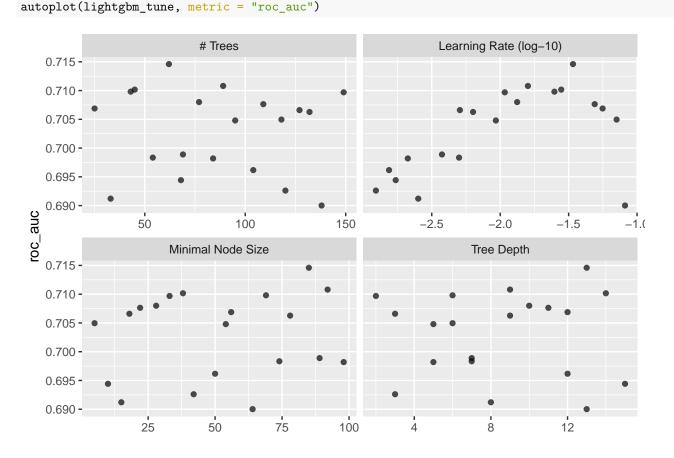
Minutes to run:

5.351

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 <-</pre>
  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
                                                                      n std_err .config
                                                           mean
##
                      <int>
     <int> <int>
                                 <dbl> <chr>
                                                <chr>
                                                            <dbl> <int>
                                                                          <dbl> <chr>
## 1
        62
              85
                         13
                                 0.0340 roc_auc binary
                                                           0.715
                                                                    10 0.0116 Preprocessor1_Model17
##
  2
        89
              92
                          9
                                 0.0159 roc_auc binary
                                                           0.711
                                                                     10 0.0119 Preprocessor1_Model19
                         14
## 3
        45
              38
                                0.0279 roc_auc binary
                                                           0.710
                                                                     10 0.0117 Preprocessor1_Model08
                                                                     10 0.0114 Preprocessor1_Model14
##
  4
        43
              69
                          6
                                 0.0249 roc_auc binary
                                                           0.710
## 5
       149
              33
                          2
                                 0.0108 roc_auc binary
                                                           0.710
                                                                     10 0.0102 Preprocessor1_Model07
best_lightgbm <- lightgbm_tune %>%
  select_best("roc_auc")
```

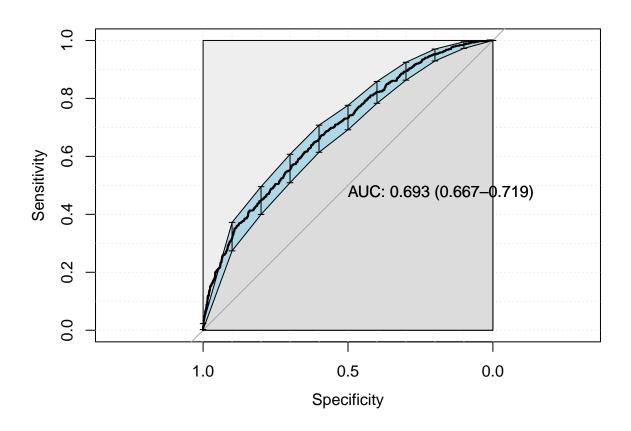


```
final_lightgbm_workflow <-
    lightgbm_workflow %>%
    finalize_workflow(best_lightgbm)

last_lightgbm_fit <-
    final_lightgbm_workflow %>%
    last_fit(df_split)

final_lightgbm_fit <- extract_workflow(last_lightgbm_fit)</pre>
```

[1] "Optimal Threshold: 0.08"



```
Confusion Matrix and Statistics
##
##
       reference
##
  data
           0
##
      0 2651 154
##
      1 1641 284
##
##
                  Accuracy : 0.6205
                    95% CI: (0.6065, 0.6344)
##
##
       No Information Rate: 0.9074
       P-Value [Acc > NIR] : 1
##
##
##
                     Kappa: 0.1054
##
    Mcnemar's Test P-Value : <2e-16
##
##
##
               Sensitivity: 0.6177
##
               Specificity: 0.6484
##
            Pos Pred Value: 0.9451
            Neg Pred Value: 0.1475
##
##
                Prevalence: 0.9074
##
            Detection Rate: 0.5605
##
      Detection Prevalence: 0.5930
##
         Balanced Accuracy: 0.6330
##
##
          'Positive' Class: 0
lightgbm_parameters <- lightgbm_tune %>%
  show_best("roc_auc", n = 1) \%
  select(-.metric, -.estimator, -.config, -mean, -n, -std_err) %>%
```

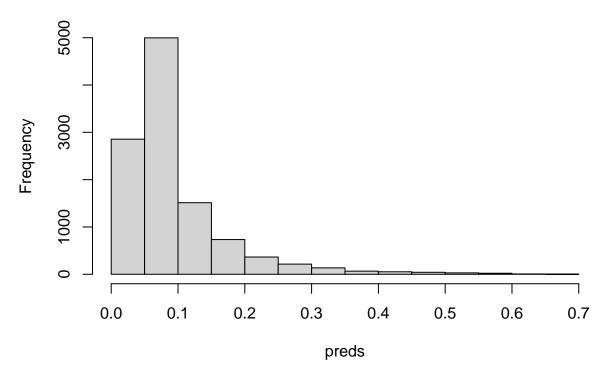
```
as.list
Hmisc::list.tree(lightgbm_parameters)

## lightgbm_parameters = list 4 (736 bytes)
## . trees = integer 1= 62
## . min_n = integer 1= 85
## . tree_depth = integer 1= 13
## . learn_rate = double 1= 0.03403

saveRDS(
    lightgbm_parameters,
    file = sprintf(
        "./auxiliar/model_selection/hyperparameters/lightgbm_%s.rds",
        outcome_column
    )
)
```

Minutes to run: 3.683

Histogram of preds



Minutes to run:

0.008

GLM

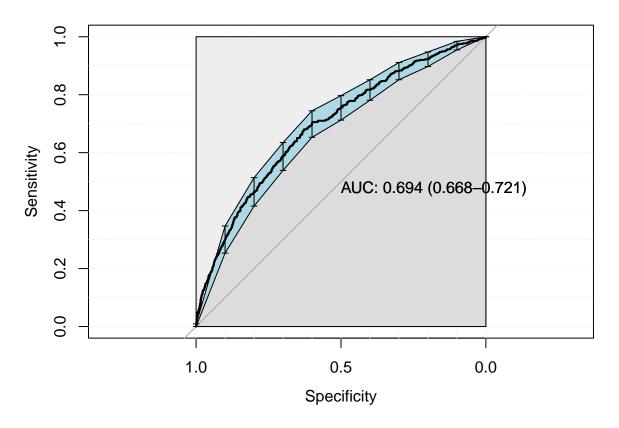
```
glmnet_recipe <-
    recipe(formula = sprintf("%s ~ .", outcome_column) %>% as.formula, data = df_train) %>%
    step_novel(all_nominal_predictors()) %>%
    step_unknown(all_nominal_predictors()) %>%
    step_other(all_nominal_predictors(), threshold = 0.05, other = ".merged") %>%
    step_dummy(all_nominal_predictors())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

```
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>
```



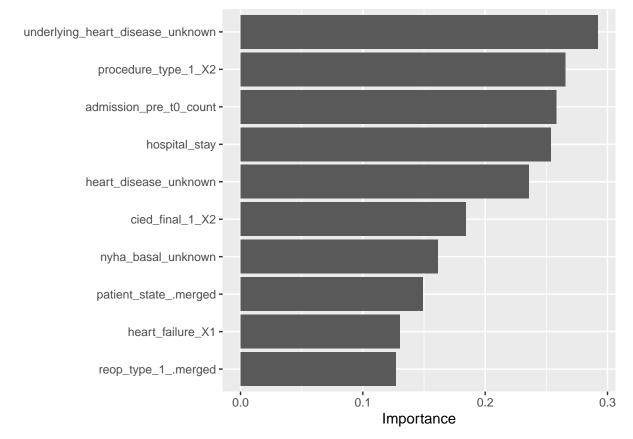
```
## [1] "Optimal Threshold: 0.08"
##
  Confusion Matrix and Statistics
##
##
       reference
##
           0
  data
                1
##
      0 2696
              140
      1 1596 298
##
##
##
                  Accuracy: 0.633
##
                    95% CI : (0.6191, 0.6467)
##
       No Information Rate: 0.9074
##
       P-Value [Acc > NIR] : 1
##
##
                     Kappa : 0.1238
##
##
    Mcnemar's Test P-Value : <2e-16
##
               Sensitivity: 0.6281
##
##
               Specificity: 0.6804
```

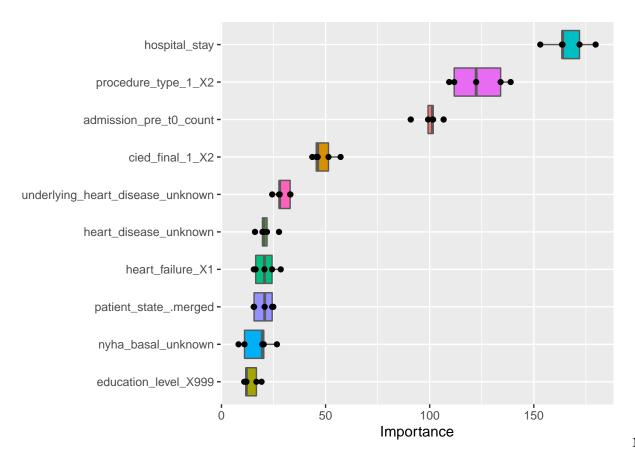
Pos Pred Value : 0.9506

Neg Pred Value : 0.1573

##

##





Minutes to run:

2.041

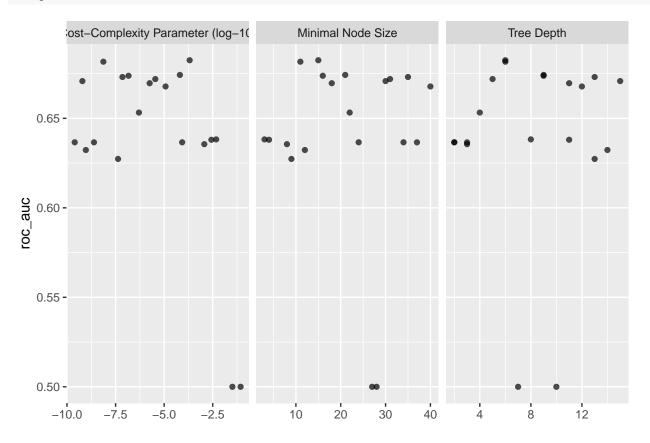
Decision Tree

```
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
                                                             <dbl> <int>
                                                                           <dbl> <chr>
##
                <dbl>
                            <int> <int> <chr>
                                                 <chr>>
             3.50e- 6
                                                             0.904
                                                                      10 0.00288 Preprocessor1_Model01
##
    1
                                5
                                     31 accuracy binary
                                     31 roc_auc binary
##
    2
             3.50e- 6
                               5
                                                             0.672
                                                                      10 0.0123 Preprocessor1_Model01
                                                                      10 0.00326 Preprocessor1_Model02
##
    3
             1.79e- 6
                                     18 accuracy binary
                                                             0.893
                               11
             1.79e- 6
                                     18 roc_auc binary
                                                             0.670
                                                                      10 0.0113 Preprocessor1_Model02
                               11
             2.69e- 3
                                                             0.903
##
    5
                               11
                                      4 accuracy binary
                                                                      10 0.00307 Preprocessor1_Model03
##
    6
             2.69e- 3
                               11
                                      4 roc_auc binary
                                                             0.638
                                                                      10 0.00802 Preprocessor1_Model03
                                3
##
    7
             1.16e- 3
                                      8 accuracy binary
                                                             0.905
                                                                      10 0.00285 Preprocessor1_Model04
    8
             1.16e- 3
                                3
                                     8 roc_auc binary
                                                             0.636
                                                                      10 0.00817 Preprocessor1_Model04
                                2
             2.50e-10
                                                             0.906
                                                                      10 0.00278 Preprocessor1_Model05
##
    9
                                     24 accuracy binary
##
  10
             2.50e-10
                                2
                                     24 roc_auc binary
                                                             0.637
                                                                      10 0.00798 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>
##
       0.000205
                               6
                                    15 roc_auc binary
                                                                    10 0.00944 Preprocessor1_Model06
  1
                                                           0.682
##
  2
       0.0000000745
                               6
                                    11 roc_auc binary
                                                           0.682
                                                                    10 0.00919 Preprocessor1_Model17
##
  3
       0.0000645
                              9
                                    21 roc_auc binary
                                                           0.674
                                                                    10 0.00824 Preprocessor1_Model08
##
  4
       0.00000147
                               9
                                    16 roc_auc binary
                                                           0.674
                                                                    10 0.0103 Preprocessor1_Model16
                                                           0.673
## 5
       0.000000714
                              13
                                    35 roc_auc binary
                                                                    10 0.0110 Preprocessor1_Model19
```

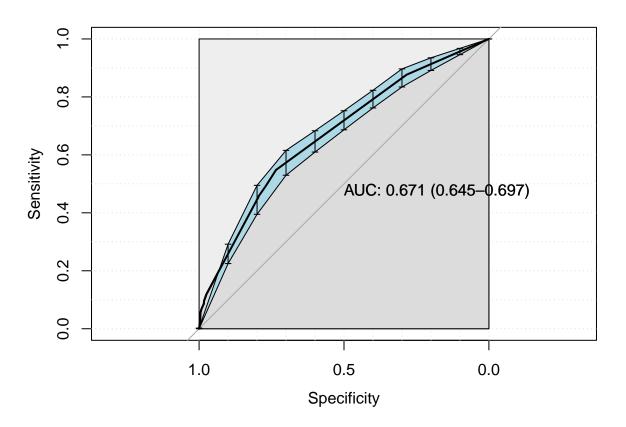
```
best_tree <- tree_tune %>%
    select_best("roc_auc")

final_tree_workflow <-
    tree_workflow %>%
    finalize_workflow(best_tree)

last_tree_fit <-
    final_tree_workflow %>%
    last_fit(df_split)

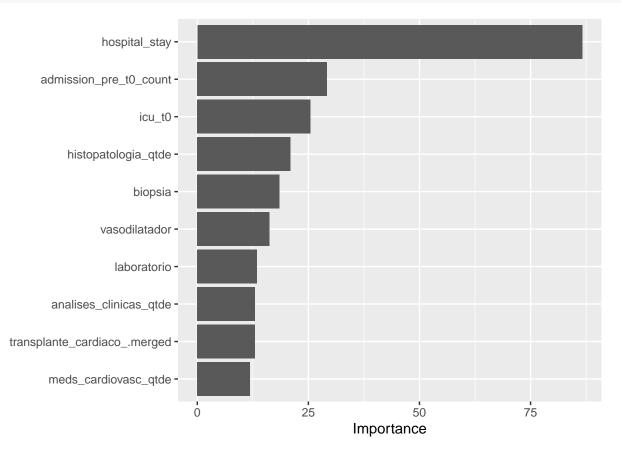
final_tree_fit <- extract_workflow(last_tree_fit)

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



```
## [1] "Optimal Threshold: 0.09"
## Confusion Matrix and Statistics
##
##
      reference
         0
##
  data
             1
      0 3153 198
##
##
      1 1139 240
##
##
                  Accuracy : 0.7173
##
                    95% CI : (0.7043, 0.7301)
##
      No Information Rate: 0.9074
      P-Value [Acc > NIR] : 1
##
##
##
                     Kappa : 0.1438
##
   Mcnemar's Test P-Value : <2e-16
##
##
##
               Sensitivity: 0.7346
```

```
##
           Neg Pred Value: 0.1740
##
                Prevalence: 0.9074
##
            Detection Rate: 0.6666
##
      Detection Prevalence: 0.7085
##
         Balanced Accuracy: 0.6413
##
##
          'Positive' Class: 0
##
extract_vip(final_tree_fit, pred_wrapper = predict,
            reference_class = "0", use_matrix = FALSE,
           method = 'model')
```



Minutes to run: 5.319

##

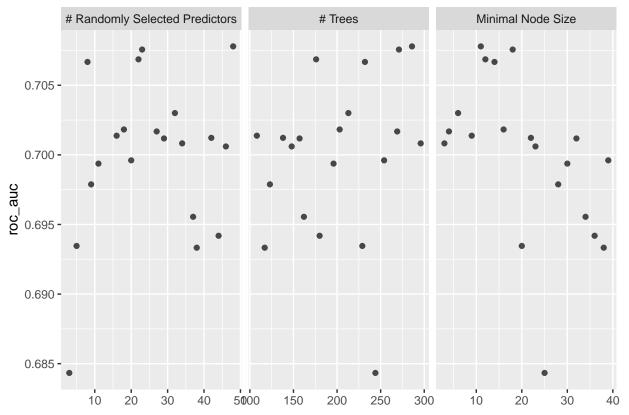
##

Specificity: 0.5479

Pos Pred Value: 0.9409

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>
                  3 accuracy binary 0.906 10 0.00274 Preprocessor1_Model01 3 roc_auc binary 0.701 10 0.0140 Preprocessor1 Model01
       34 296
## 1
        34
             296
##
   2
      3 244 25 accuracy binary
                                           0.906 10 0.00281 Preprocessor1_Model02
## 3
## 4
        3 244 25 roc_auc binary
                                           0.684 10 0.0113 Preprocessor1_Model02
                                           0.906 10 0.00264 Preprocessor1_Model03 0.707 10 0.0136 Preprocessor1_Model03
       22 176
##
   5
                   12 accuracy binary
## 6
       22 176
                  12 roc_auc binary
## 7
        8 232
                   14 accuracy binary
                                           0.906 10 0.00273 Preprocessor1_Model04
                                           ## 8
        8
             232
                   14 roc_auc binary
                                                 10 0.00264 Preprocessor1_Model05
## 9
        29
             157
                    32 accuracy binary
                                           0.906
## 10
        29
             157
                    32 roc_auc binary
                                           0.701
                                                 10 0.0129 Preprocessor1_Model05
## # ... with 30 more rows
## # i Use 'print(n = ...)' to see more rows
autoplot(rf_tune, metric = "roc_auc")
```

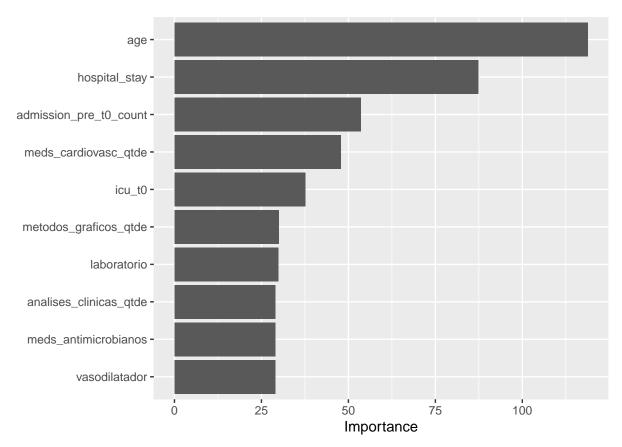


```
rf_tune %>%
  show_best("roc_auc")
## # A tibble: 5 x 9
##
      mtry trees min_n .metric .estimator mean
                                                      n std_err .config
##
     <int> <int> <int> <chr>
                                <chr>
                                            <dbl> <int>
                                                           <dbl> <chr>
## 1
        48
             286
                     11 roc_auc binary
                                            0.708
                                                     10 0.0125 Preprocessor1_Model19
                     18 roc_auc binary
                                                     10 0.0131 Preprocessor1_Model12
## 2
        23
             271
                                            0.708
## 3
        22
             176
                     12 roc_auc binary
                                            0.707
                                                     10 0.0136 Preprocessor1_Model03
                                                     10 0.0126 Preprocessor1_Model04
## 4
         8
             232
                     14 roc_auc binary
                                            0.707
## 5
        32
             213
                      6 roc_auc binary
                                            0.703
                                                     10 0.0142 Preprocessor1_Model16
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>
```

```
0.8
Sensitivity
                                                       AUC: 0.677 (0.651-0.704)
     0.4
     0.0
                                                     0.5
                            1.0
                                                                               0.0
                                                 Specificity
```

[1] "Optimal Threshold: 0.05"

```
Confusion Matrix and Statistics
##
##
       reference
##
  data
           0
##
      0 2511
              149
##
      1 1781 289
##
                  Accuracy: 0.592
##
                    95% CI: (0.5778, 0.606)
##
##
       No Information Rate: 0.9074
       P-Value [Acc > NIR] : 1
##
##
##
                     Kappa : 0.0916
##
##
    Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity: 0.5850
##
               Specificity: 0.6598
##
            Pos Pred Value : 0.9440
##
            Neg Pred Value: 0.1396
##
                Prevalence: 0.9074
##
            Detection Rate: 0.5309
##
      Detection Prevalence: 0.5624
##
         Balanced Accuracy : 0.6224
##
##
          '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')
```

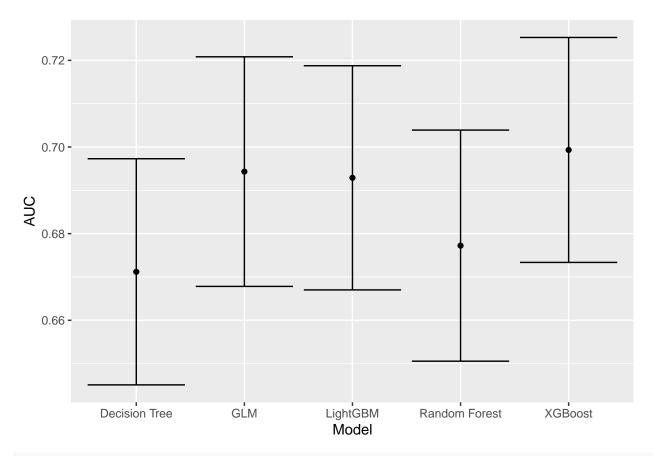


```
# extract_vip(final_rf_fit, pred_wrapper = predict,
# reference_class = "1", use_matrix = FALSE,
# method = 'permute')
```

Minutes to run: 69.791

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()
```



saveRDS(df_auc, sprintf("./auxiliar/model_selection/performance/%s.RData", outcome_column))

Minutes to run: 0.003