Final Model - readmission_60d

Eduardo Yuki Yada

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
max_auc_loss <- params$max_auc_loss # Max accepted loss of AUC for reducing num of features
repeats <- params$repeats
Hmisc::list.tree(params)

## params = list 5 (952 bytes)
## . max_auc_loss = double 1= 0.01
## . outcome_column = character 1= readmission_60d
## . k = double 1= 10
## . grid_size = double 1= 50
## . repeats = double 1= 2</pre>
```

Imports

```
library(tidyverse)
library(yaml)
library(tidymodels)
library(usemodels)
library(vip)
library(kableExtra)
library(SHAPforxgboost)
library(xgboost)
library(Matrix)
library(mltools)
library(bonsai)
library(lightgbm)
library(pROC)
library(caret)
library(themis)
source("aux_functions.R")
select <- dplyr::select</pre>
```

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[columns_list$outcome_columns] <- lapply(df[columns_list$outcome_columns], factor)</pre>
```

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)
eligible_columns <- df_names %>%
  filter(momento.aquisicao == 'Admissão t0') %>%
  .$variable.name
exception_columns <- c('death_intraop', 'death_intraop_1', 'disch_outcomes_t0')</pre>
correlated_columns = c('year_procedure_1', # com year_adm_t0
                       'age_surgery_1', # com age
                       'admission_t0', # com admission_pre_t0_count
                       'atb', # com meds_antimicrobianos
                       'classe_meds_cardio_qtde', # com classe_meds_qtde
                       'suporte_hemod', # com proced_invasivos_qtde,
                       'radiografia', # com exames_imagem_qtde
                       'ecg' # com metodos_graficos_qtde
eligible_features <- eligible_columns %>%
  base::intersect(c(columns_list$categorical_columns, columns_list$numerical_columns)) %>%
  setdiff(c(exception_columns, correlated_columns))
features = base::intersect(eligible_features, features_list)
```

Starting features:

```
gluedown::md_order(features, seq = TRUE, pad = TRUE)
```

- 1. age
- 2. education_level
- 3. underlying heart disease
- $4.\ \, heart_disease$
- 5. nyha basal
- 6. prior_mi
- 7. heart_failure

- 8. af
- 9. cardiac arrest
- 10. transplant
- 11. valvopathy
- 12. diabetes
- 13. hemodialysis
- 14. comorbidities_count
- 15. procedure_type_1
- $16. \text{ reop_type_1}$
- 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
- $44.\ \mathrm{meds_cardiovasc_qtde}$
- 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
- $57. \text{ cve_desf}$
- 58. transfusao
- 59. interconsulta
- 60. equipe_multiprof
- 61. holter
- 62. teste_esforco
- 63. espiro_ergoespiro
- 64. tilt_teste
- $65.\ \mathrm{metodos_graficos_qtde}$
- 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
```

Train test split (70%/30%)

Feature Selection

```
custom_dummy_names <- function(var, lvl, ordinal = FALSE) {</pre>
  dummy_names(var, lvl, ordinal = FALSE, sep = "__")
}
model_fit_wf <- function(df_train, features, outcome_column, hyperparameters){</pre>
 model_recipe <-</pre>
    recipe(formula = sprintf("%s ~ .", outcome_column) %>% as.formula,
           data = df_train %>% select(all_of(c(features, outcome_column)))) %>%
    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(), naming = custom_dummy_names)
  model_spec <-
    do.call(boost_tree, hyperparameters) %>%
    set_engine("lightgbm") %>%
    set_mode("classification")
  model_workflow <-</pre>
    workflow() %>%
    add_recipe(model_recipe) %>%
    add_model(model_spec)
```

```
model_fit_rs <- model_workflow %>%
    fit_resamples(df_folds)
 model_fit <- model_workflow %>%
    fit(df_train)
  model_auc <- validation(model_fit, df_test, plot = F)</pre>
  raw_model <- parsnip::extract_fit_engine(model_fit)</pre>
  feature_importance <- lgb.importance(raw_model, percentage = TRUE) %>%
    separate(Feature, c("Feature","value"), "__", fill = 'right') %>%
    group_by(Feature) %>%
    summarise(Gain = sum(Gain),
              Cover = sum(Cover),
              Frequency = sum(Frequency)) %>%
    ungroup() %>%
    arrange(desc(Gain))
  cv_results <- collect_metrics(model_fit_rs) %>% filter(.metric == 'roc_auc')
  return(
    list(
      cv_auc = cv_results$mean,
      cv_auc_std_err = cv_results$std_err,
      importance = feature_importance,
      auc = as.numeric(model_auc$auc),
      auc_lower = model_auc$ci[1],
      auc_upper = model_auc$ci[3]
  )
}
hyperparameters <- readRDS(</pre>
  sprintf(
    "./auxiliar/model_selection/hyperparameters/lightgbm_%s.rds",
    outcome_column
  )
)
hyperparameters$sample_size <- 1
full_model <- model_fit_wf(df_train, features, outcome_column, hyperparameters)
sprintf('Full Model CV Train AUC: %.3f' ,full_model$cv_auc)
## [1] "Full Model CV Train AUC: 0.711"
sprintf('Full Model Test AUC: %.3f' ,full_model$auc)
## [1] "Full Model Test AUC: 0.672"
Features with zero importance on the initial model:
unimportant_features <- setdiff(features, full_model$importance$Feature)
unimportant_features %>%
  gluedown::md_order()
  1. transplant
  2. procedure_type_1
  3. dialysis_t0
  4. cec
```

```
5. transplante_cardiaco
  6. cir toracica
  7. teste_esforco
  8. espiro_ergoespiro
  9. tilt_teste
 10. biopsia
 11. arteriografia
 12. pet ct
 13. ressonancia
trimmed_features <- full_model$importance$Feature</pre>
trimmed_model <- model_fit_wf(df_train, trimmed_features,</pre>
                                outcome_column, hyperparameters)
sprintf('Trimmed Model CV Train AUC: %.3f', trimmed_model$cv_auc)
## [1] "Trimmed Model CV Train AUC: 0.712"
sprintf('Trimmed Model Test AUC: %.3f' ,trimmed_model$auc)
## [1] "Trimmed Model Test AUC: 0.672"
selection_results <- tibble::tribble(</pre>
  ~`Tested Feature`, ~`Dropped`, ~`Number of Features`, ~`CV AUC`, ~`CV AUC Std Error`, ~`Total AUC Loss`, ~`Ins
  'None', TRUE, length(features), full_model$cv_auc, full_model$cv_auc_std_err, 0, 0
)
whitelist <- c()
if (full_model$cv_auc - trimmed_model$cv_auc < max_auc_loss) {</pre>
  current_features <- trimmed_features</pre>
  current_model <- trimmed_model</pre>
  current_auc_loss <- full_model$cv_auc - current_model$cv_auc</pre>
  instant_auc_loss <- full_model$cv_auc - current_model$cv_auc
  selection_results <- selection_results %>%
    add_row(`Tested Feature` = 'All unimportant',
             `Dropped` = TRUE,
             `Number of Features` = length(trimmed_features),
             `CV AUC` = current_model$cv_auc,
            `CV AUC Std Error` = current_model$cv_auc_std_err,
             `Total AUC Loss` = current_auc_loss,
             `Instant AUC Loss` = instant_auc_loss)
} else {
  current_features <- features</pre>
  current_model <- full_model</pre>
  current_auc_loss <- 0</pre>
}
while (current_auc_loss < max_auc_loss & mean(current_features %in% whitelist) < 1) {</pre>
  zero_importance_features <-
    setdiff(current_features, current_model$importance$Feature) %>%
    setdiff(whitelist)
  if (length(zero_importance_features) > 0) {
    current_least_important <- zero_importance_features[1]</pre>
    current_least_important <-</pre>
      tail(setdiff(current_model$importance$Feature, whitelist), 1)
  }
  test_features <-
    setdiff(current_features, current_least_important)
  current_model <-</pre>
    model_fit_wf(df_train, test_features, outcome_column, hyperparameters)
```

```
instant_auc_loss <-</pre>
   tail(selection_results %>% filter(Dropped) %>% .$`CV AUC`, n = 1) - current_model$cv_auc
  if (instant_auc_loss < max_auc_loss / 5 &</pre>
     current_auc_loss < max_auc_loss) {</pre>
   dropped <- TRUE
   current_features <- test_features</pre>
   current_auc_loss <- full_model$cv_auc - current_model$cv_auc</pre>
 } else {
   dropped <- FALSE
   whitelist <- c(whitelist, current_least_important)</pre>
 }
  selection_results <- selection_results %>%
   add row(
     `Tested Feature` = current_least_important,
     `Dropped` = dropped,
     `Number of Features` = length(test_features),
     `CV AUC` = current_model$cv_auc,
     `CV AUC Std Error` = current_model$cv_auc_std_err,
     `Total AUC Loss` = current_auc_loss,
      `Instant AUC Loss` = instant_auc_loss
 print(c(
   length(current_features),
   round(current_auc_loss, 4),
   round(instant_auc_loss, 4),
   current_least_important
 ))
}
## [1] "70"
                    "-0.0011"
                                 "2e-04"
                                               "reop_type_1"
                                "4e-04"
## [1] "69"
                    "-7e-04"
                                                  "angioplastia"
                 "-0.001" "-3e-04" "angio_tc"
## [1] "68"
                              "-0.0011"
## [1] "67"
                                                       "-1e-04"
                                                                               "cateter_venoso_central"
                  "-9e-04" "2e-04" "valvopathy"
## [1] "66"
                 "-7e-04" "2e-04" "prior_mi"
## [1] "65"
## [1] "64"
                           "-0.001"
                                       "-3e-04"
                                                                   "antiplaquetario ev"
## [1] "63"
                 "-0.0012" "-2e-04"
                                      "insulina"
## [1] "62"
               "-6e-04" "6e-04" "holter"
## [1] "61"
                                     "6e-04"
                                                     "heart_failure"
                      "3e-04"
## [1] "60"
                                      "2e-04"
                                                       "cardiac_arrest"
## [1] "59"
                              "4e-04"
                                                       "1e-04"
                                                                               "analises_clinicas_qtde"
## [1] "58"
                   ## [1] "57"
                     "-9e-04"
                                     "-8e-04"
                                                       "betabloqueador"
                 "-8e-04" "1e-04" "cve_desf"
## [1] "56"
## [1] "55"
                    "-3e-04" "5e-04"
                                                  "hemodialysis"
## [1] "54"
                  "-1e-04"
                             "2e-04" "antiviral"
## [1] "53"
                   "1e-04"
                              "2e-04"
                                          "tomografia"
                   "-4e-04"
                                "-5e-04"
                                              "antifungico"
## [1] "52"
                "-2e-04" "2e-04" "cultura"
## [1] "51"
                           "7e-04"
                                                                      "ventilacao_mecanica"
## [1] "50"
                                                "9e-04"
                  "-6e-04" "-0.0012" "ultrassom"
## [1] "49"
                                       "-1e-04"
## [1] "48"
                      "-7e-04"
                                                       "insuf_cardiaca"
                    "-0.001" "-3e-04"
                                              "cateterismo"
## [1] "47"
## [1] "46"
                         "-9e-04"
                                               "1e-04"
                                                                   "procedure_type_new"
## [1] "45"
                      "-4e-04"
                                      "5e-04"
                                                      "ecocardiograma"
                 "-0.0022" "-0.0018" "digoxina"
## [1] "44"
                      "-0.0018" "4e-04"
## [1] "43"
                                                   "interconsulta"
## [1] "42"
                           "-0.0012"
                                                 "6e-04"
                                                                      "histopatologia_qtde"
```

```
## [1] "41" "-0.0017" "-5e-04" "ic
## [1] "40" "-7e-04" "0.001" "af"
              "-0.0017" "-5e-04" "icp"
                                 "-3e-04" "eletrofisiologia"
"-2e-04" "cintilografia"
                 "-0.001"
"-0.0012"
## [1] "39"
## [1] "38"
               "-0.0012" "0" "citologia" "-0.001" "2e-04" "transfusa
## [1] "37"
## [1] "36"
                                        "transfusao"
## [1] "35"
                              "-0.0014"
                                                      "-4e-04"
## [4] "outros_proced_cirurgicos"
                                           "0.0011"
## [1] "34"
                        "-3e-04"
                                                               "cied_final_group_1"
              "-4e-04" "-1e-04" "aco"
## [1] "33"
              "-4e-04" "-0.0015" "-0.0012" "heart_disease" "-3e-04" "0.0012" "nyha_basal"
## [1] "32"
## [1] "31"
            "-8e-04" "-5e-04" "bic"
## [1] "30"
                "-0.0015"
                                     "-6e-04"
## [1] "29"
                                                      "espironolactona"
                "-0.0027" "-0.0012" "angio_rm"
## [1] "28"
## [1] "27"
                          "-0.0022"
                                                "5e-04"
                                                                        "admission_pre_t0_180d"
                              "-0.0019"
## [1] "26"
                                                      "3e-04"
## [4] "underlying_heart_disease"
                "-0.0025" "-5e-04" "diabetes"
## [1] "25"
                 "-0.002" "5e-04" "bloq_calcio"
## [1] "24"
                                      "-2e-04" "equipe_multiprof"
## [1] "23"
                    "-0.0022"
               "-0.0029" "-7e-04" "diuretico" "anticonvulsivante"
## [1] "22"
## [1] "21"
                  "-0.0036" "-7e-04" "psicofarmacos" "-8e-04"
## [1] "20"
## [1] "19"
                                                                           "admission_t0_emergency"
                   "-0.0026" "0.0018" "cied_final_1"
## [1] "18"
                         0025" "1e-04" "antiarritmico" "-0.0035" "-0.001"
## [1] "17"
                   "-0.0025"
                                                                "comorbidities_count"
## [1] "16"
                                           "-4e-04"
                                                             "exames_imagem_qtde"
                        "-0.0039"
## [1] "15"
             "-0.0051" "-0.0012" "dva"
## [1] "14"
## [1] "13"
                                               "0"
                          "-0.0051"
                                                                     "meds_cardiovasc_qtde"
                         "-0.0057"
                                               "-6e-04"
## [1] "12"
                                                                     "meds_antimicrobianos"
                    "-0.0057" "0.0026" "vasodilatador" "-0.0049" "7e-04" "education
## [1] "12"
                                                  "education_level"
## [1] "11"
                                                "0.0013"
## [1] "10"
                          "-0.0037"
                                                                        "proced_invasivos_qtde"
               "-0.003" "7e-04" "estatina"
## [1] "9"
               "-0.0053" "-0.0024" "icu_t0"
## [1] "8"
## [1] "7"
                          "-0.0056"
                                                 "-2e-04"
                                                                        "metodos_graficos_qtde"
                "-0.0038" "0.0017" "ieca_bra"
## [1] "6"
                "-0.0038" "0.0101" "classe_meds_qtde"
## [1] "6"
               "-0.0051" "-0.0013" "age"
## [1] "5"
                                                  "0.0152"
                  "-0.0051"
## [1] "5"
                                                                           "admission_pre_t0_count"
                              "0.001" "laboratorio"
                   "-0.0041"
## [1] "4"
                   "-0.0041" "0.0676"
## [1] "4"
                                              "hospital_stay"
selection_results %>%
 rename(Features = `Number of Features`) %>%
 niceFormatting(digits = 4, label = 1)
```

Table 1:

Tested Feature	Dropped	Features	CV AUC	CV AUC Std Error	Total AUC Loss	Instant AUC Loss
None	TRUE	84	0.7105	0.0088	0.0000	0.0000
All unimportant	TRUE	71	0.7118	0.0086	-0.0013	-0.0013
reop_type_1	TRUE	70	0.7116	0.0085	-0.0011	0.0002
angioplastia	TRUE	69	0.7112	0.0084	-0.0007	0.0004
angio_tc	TRUE	68	0.7115	0.0084	-0.0010	-0.0003
cateter_venoso_central	TRUE	67	0.7116	0.0084	-0.0011	-0.0001
valvopathy	TRUE	66	0.7114	0.0084	-0.0009	0.0002
prior_mi	TRUE	65	0.7112	0.0084	-0.0007	0.0002
antiplaquetario_ev	TRUE	64	0.7115	0.0084	-0.0010	-0.0003

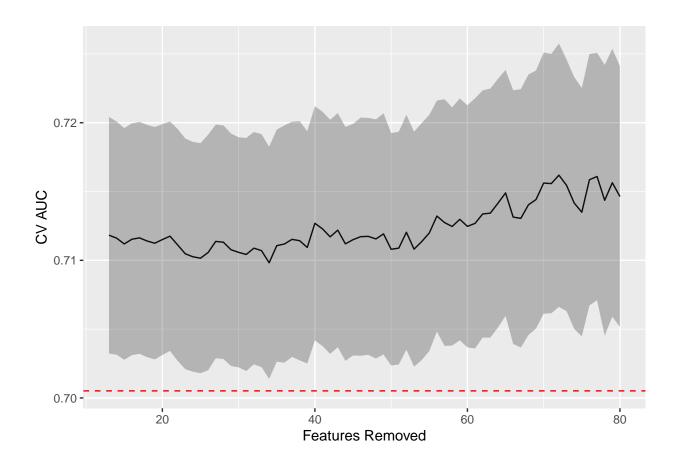
Table 1: (continued)

Tested Feature	Dropped	Features	CV AUC	CV AUC Std Error	Total AUC Loss	Instant AUC Loss
insulina	TRUE	63	0.7118	0.0083	-0.0012	-0.0002
holter	TRUE	62	0.7111	0.0084	-0.0006	0.0006
heart failure	TRUE	61	0.7105	0.0084	0.0000	0.0006
cardiac_arrest	TRUE	60	0.7103	0.0084	0.0003	0.0002
analises_clinicas_qtde	TRUE	59	0.7102	0.0084	0.0004	0.0001
endoscopia	TRUE	58	0.7106	0.0086	-0.0001	-0.0004
betabloqueador	TRUE	57	0.7114	0.0085	-0.0009	-0.0008
cve desf	TRUE	56	0.7113	0.0085	-0.0008	0.0001
hemodialysis	TRUE	55	0.7108	0.0084	-0.0003	0.0005
antiviral	TRUE	54	0.7106	0.0084	-0.0001	0.0002
tomografia	TRUE	53	0.7104	0.0085	0.0001	0.0002
antifungico	TRUE	52	0.7109	0.0084	-0.0004	-0.0005
cultura	TRUE	51	0.7107	0.0085	-0.0002	0.0002
ventilacao mecanica	TRUE	50	0.7098	0.0084	0.0002	0.0002
ultrassom	TRUE	49	0.7111	0.0084	-0.0006	-0.0012
insuf cardiaca	TRUE	48	0.7111	0.0084	-0.0007	-0.0012
_						
cateterismo	TRUE	47	0.7115	0.0085	-0.0010	-0.0003
procedure_type_new	TRUE	46	0.7114	0.0087	-0.0009	0.0001
ecocardiograma	TRUE	45	0.7109	0.0084	-0.0004	0.0005
digoxina	TRUE	44	0.7127	0.0085	-0.0022	-0.0018
interconsulta	TRUE	43	0.7123	0.0085	-0.0018	0.0004
histopatologia_qtde	TRUE	42	0.7117	0.0085	-0.0012	0.0006
icp	TRUE	41	0.7122	0.0085	-0.0017	-0.0005
af	TRUE	40	0.7112	0.0085	-0.0007	0.0010
eletrofisiologia	TRUE	39	0.7115	0.0084	-0.0010	-0.0003
cintilografia	TRUE	38	0.7117	0.0087	-0.0012	-0.0002
citologia	TRUE	37	0.7117	0.0086	-0.0012	0.0000
transfusao	TRUE	36	0.7116	0.0087	-0.0010	0.0002
outros_proced_cirurgicos	TRUE	35	0.7119	0.0088	-0.0014	-0.0004
cied final group 1	TRUE	34	0.7108	0.0084	-0.0003	0.0011
aco	TRUE	33	0.7109	0.0085	-0.0004	-0.0001
heart disease	TRUE	32	0.7120	0.0085	-0.0015	-0.0012
nyha_basal	TRUE	31	0.7108	0.0085	-0.0003	0.0012
bic	TRUE	30	0.7114	0.0086	-0.0008	-0.0005
espironolactona	TRUE	29	0.7120	0.0086	-0.0015	-0.0006
angio_rm	TRUE	28	0.7132	0.0084	-0.0027	-0.0012
admission_pre_t0_180d	TRUE	27	0.7127	0.0090	-0.0022	0.0005
underlying heart disease	TRUE	26	0.7127	0.0090	-0.0022	0.0003
diabetes	TRUE					
		25	0.7130	0.0088	-0.0025	-0.0005
bloq_calcio	TRUE	24	0.7125	0.0088	-0.0020	0.0005
equipe_multiprof	TRUE	23	0.7127	0.0091	-0.0022	-0.0002
diuretico	TRUE	22	0.7134	0.0090	-0.0029	-0.0007
anticonvulsivante	TRUE	21	0.7134	0.0090	-0.0029	-0.0001
psicofarmacos	TRUE	20	0.7141	0.0090	-0.0036	-0.0007
admission_t0_emergency	TRUE	19	0.7149	0.0089	-0.0044	-0.0008
cied_final_1	TRUE	18	0.7131	0.0092	-0.0026	0.0018
antiarritmico	TRUE	17	0.7130	0.0094	-0.0025	0.0001
comorbidities_count	TRUE	16	0.7140	0.0095	-0.0035	-0.0010
exames_imagem_qtde	TRUE	15	0.7144	0.0094	-0.0039	-0.0004
dva	TRUE	14	0.7156	0.0095	-0.0051	-0.0012
$meds_cardiovasc_qtde$	TRUE	13	0.7156	0.0094	-0.0051	0.0000
meds antimicrobianos	TRUE	12	0.7162	0.0096	-0.0057	-0.0006
vasodilatador	FALSE	11	0.7136	0.0093	-0.0057	0.0026
vasodilavadoi	THESE		0.1100	0.0000	0.0001	0.002

Table 1: (continued)

Tested Feature	Dropped	Features	CV AUC	CV AUC Std Error	Total AUC Loss	Instant AUC Loss
education_level	TRUE	11	0.7154	0.0091	-0.0049	0.0007
$proced_invasivos_qtde$	TRUE	10	0.7142	0.0092	-0.0037	0.0013
estatina	TRUE	9	0.7135	0.0090	-0.0030	0.0007
icu_t0	TRUE	8	0.7159	0.0091	-0.0053	-0.0024
metodos_graficos_qtde	TRUE	7	0.7161	0.0090	-0.0056	-0.0002
ieca_bra	TRUE	6	0.7144	0.0098	-0.0038	0.0017
$classe_meds_qtde$	FALSE	5	0.7042	0.0095	-0.0038	0.0101
age	TRUE	5	0.7156	0.0097	-0.0051	-0.0013
$admission_pre_t0_count$	FALSE	4	0.7004	0.0093	-0.0051	0.0152
laboratorio	TRUE	4	0.7146	0.0095	-0.0041	0.0010
hospital_stay	FALSE	3	0.6470	0.0091	-0.0041	0.0676

```
selected_features <- current_features</pre>
con <- file(sprintf('./auxiliar/final_model/selected_features/%s.yaml', outcome_column), "w")</pre>
write_yaml(selected_features, con)
close(con)
feature_selected_model <- model_fit_wf(df_train, selected_features,</pre>
                                        outcome_column, hyperparameters)
sprintf('Selected Model CV Train AUC: %.3f', feature_selected_model$cv_auc)
## [1] "Selected Model CV Train AUC: 0.715"
sprintf('Selected Model Test AUC: %.3f', feature_selected_model$auc)
## [1] "Selected Model Test AUC: 0.670"
selection_results <- selection_results %>%
 filter(`Number of Features` < length(features)) %>%
 mutate(`Features Removed` = length(features) - `Number of Features`,
         `CV AUC Low` = `CV AUC` - `CV AUC Std Error`,
         `CV AUC High` = `CV AUC` + `CV AUC Std Error`)
selection_results %>%
  filter(Dropped) %>%
  ggplot(aes(x = `Features Removed`, y = `CV AUC`,
             ymin = `CV AUC Low`, ymax = `CV AUC High`)) +
  geom_line() +
  geom_ribbon(alpha = .3) +
 geom_hline(yintercept = full_model$cv_auc - max_auc_loss,
             linetype = "dashed", color = "red")
```



Hyperparameter tuning

Selected features:

```
gluedown::md_order(selected_features, seq = TRUE, pad = TRUE)
```

- 1. hospital_stay
- $2. \ admission_pre_t0_count$
- 3. classe_meds_qtde
- 4. vasodilatador

Standard

```
lightgbm_recipe <-</pre>
  recipe(formula = sprintf("%s ~ .", outcome_column) %>% as.formula,
         data = df_train %>% select(all_of(c(selected_features, outcome_column)))) %>%
  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_tuning <- function(recipe) {</pre>
  lightgbm_spec <- boost_tree(</pre>
    trees = tune(),
    min_n = tune(),
    tree_depth = tune(),
    learn_rate = tune(),
    sample_size = 1.0
  ) %>%
    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(recipe) %>%
    add_model(lightgbm_spec)
  lightgbm_tune <-
    lightgbm_workflow %>%
    tune_grid(resamples = df_folds,
              grid = lightgbm_grid)
 lightgbm_tune %>%
    show best("roc auc") %>%
    niceFormatting(digits = 5, label = 4)
  best_lightgbm <- lightgbm_tune %>%
    select_best("roc_auc")
  autoplot(lightgbm_tune, metric = "roc_auc")
  final_lightgbm_workflow <-</pre>
    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)</pre>
  lightgbm_auc <- validation(final_lightgbm_fit, df_test)</pre>
  lightgbm_parameters <- lightgbm_tune %>%
    show_best("roc_auc", n = 1) \%
    select(trees, min_n, tree_depth, learn_rate) %>%
    as.list
  return(list(auc = as.numeric(lightgbm_auc$auc),
              auc_lower = lightgbm_auc$ci[1],
              auc_upper = lightgbm_auc$ci[3],
              parameters = lightgbm_parameters,
              fit = final_lightgbm_fit))
}
standard_results <- lightgbm_tuning(lightgbm_recipe)</pre>
```

```
Sensitivity

AUC: 0.675 (0.638–0.712)

1.0 0.5 0.0

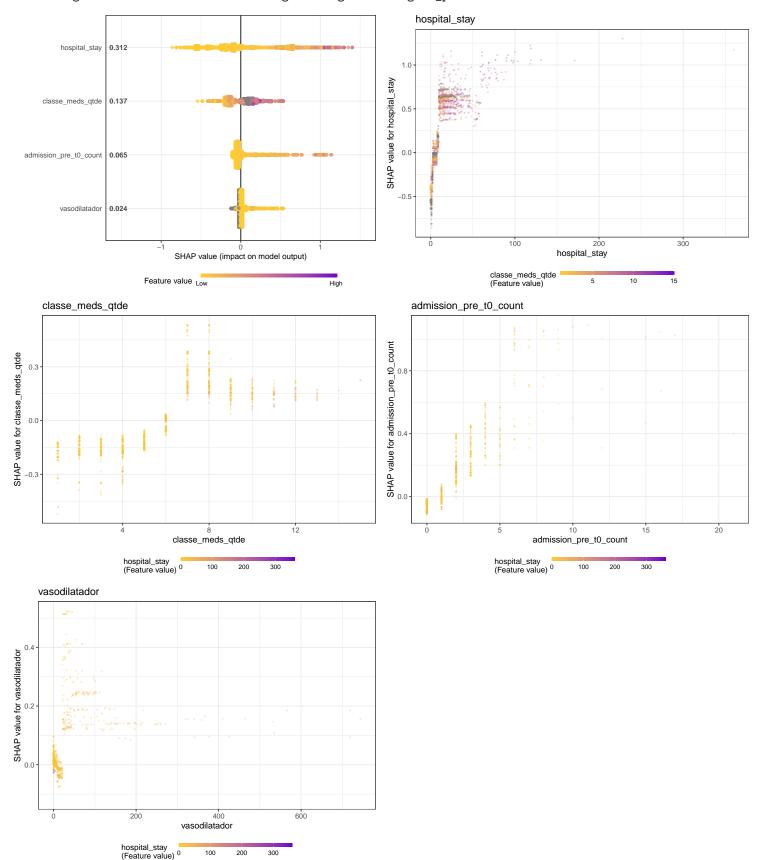
Specificity
```

```
## [1] "Optimal Threshold: 0.07"
   Confusion Matrix and Statistics
##
##
       reference
##
   data
           0
##
      0 3460
              121
##
      1 1016 133
##
                   Accuracy : 0.7596
##
                     95% CI: (0.7472, 0.7717)
##
##
       No Information Rate: 0.9463
       P-Value [Acc > NIR] : 1
##
##
##
                      Kappa : 0.1114
##
##
    Mcnemar's Test P-Value : <2e-16
##
##
                Sensitivity: 0.7730
##
                Specificity: 0.5236
##
            Pos Pred Value: 0.9662
##
            Neg Pred Value: 0.1158
##
                 Prevalence: 0.9463
##
            Detection Rate: 0.7315
##
      Detection Prevalence : 0.7571
##
         Balanced Accuracy : 0.6483
##
##
           'Positive' Class: 0
##
final_lightgbm_fit <- standard_results$fit</pre>
{\tt lightgbm\_parameters} \ {\tt <-} \ {\tt standard\_results\$parameters}
saveRDS(
  lightgbm_parameters,
```

SHAP values

```
lightgbm_model <- parsnip::extract_fit_engine(final_lightgbm_fit)</pre>
trained_rec <- prep(lightgbm_recipe, training = df_train)</pre>
df_train_baked <- bake(trained_rec, new_data = df_train)</pre>
df_test_baked <- bake(trained_rec, new_data = df_test)</pre>
matrix_train <- as.matrix(df_train_baked %>% select(-all_of(outcome_column)))
matrix_test <- as.matrix(df_test_baked %>% select(-all_of(outcome_column)))
n_plots <- min(6, length(selected_features))</pre>
plotted <- 0
shap.plot.summary.wrap1(model = lightgbm_model, X = matrix_train,
                         top_n = n_plots, dilute = F)
shap <- shap.prep(lightgbm_model, X_train = matrix_test)</pre>
for (x in shap.importance(shap, names_only = TRUE)) {
  p <- shap.plot.dependence(</pre>
    shap,
    x = x,
    color_feature = "auto",
    smooth = FALSE,
    jitter_width = 0.01,
    alpha = 0.3
    labs(title = x)
  if (plotted < n_plots) {</pre>
    print(p)
    plotted <- plotted + 1
  ggsave(sprintf("./auxiliar/final_model/shap_plots/%s.png", x),
         plot = p,
         dpi = 300)
}
## Saving 6.5 x 5 in image
## Warning: Removed 1455 rows containing missing values (geom_point).
## Saving 6.5 x 5 in image
```

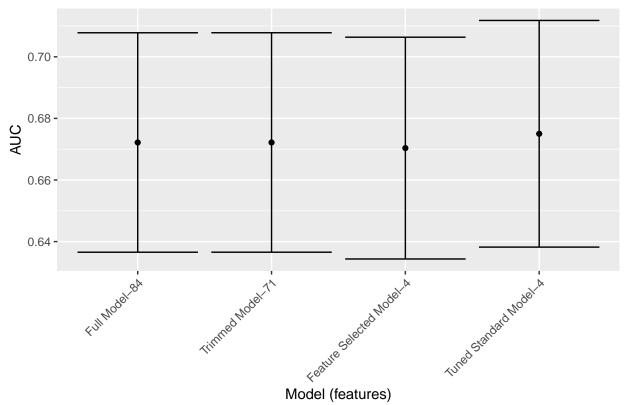
- ## Warning: Removed 1455 rows containing missing values (geom_point).
- ## Saving 6.5 x 5 in image
- ## Warning: Removed 1041 rows containing missing values (geom_point).
- ## Saving 6.5 x 5 in image
- ## Warning: Removed 1041 rows containing missing values (geom_point).



```
## $num_iterations
## [1] 135
##
## $learning_rate
## [1] 0.008090184
## $max_depth
## [1] 4
##
## $feature_fraction
## [1] 1
##
## $min_data_in_leaf
## [1] 33
##
## $min_gain_to_split
## [1] 0
##
## $bagging_fraction
## [1] 1
##
## $num_class
## [1] 1
##
## $objective
## [1] "binary"
##
## $num_threads
## $num_threads$num_threads
## [1] 0
##
##
## $nthread
## [1] 8
##
## $seed
## [1] 1591
## $deterministic
## [1] TRUE
##
## $verbose
## [1] -1
##
## $metric
## list()
## $interaction_constraints
## list()
## $feature_pre_filter
## [1] FALSE
```

Models Comparison

readmission_60d



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