Credit score approval prediction

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Handling nominal variable with dummy variable

Use library 'fastDummies' to handle the nominal variable \$OCCUPATION TYPE.

```
library(fastDummies)
library(janitor)
dummy_var <- function(data){</pre>
  # Since the library 'fastDummies' tansforms all factor variable into dummy variables,
  # we will convert our target "y" (factor) into a character variable
  # to avoid it being transformed to dummy variable.
  data$y <- as.numeric(as.character(data$y))</pre>
  # transform all factor variables to dummy variables,
  # and removes the original variables that were used to generate the dummy variables.
  data_dummy <- fastDummies::dummy_cols(data, remove_selected_columns=TRUE)</pre>
  # column name convention fix (mlr3 name convention - space to underscore)
  data_dummy <- clean_names(data_dummy)</pre>
  data_dummy <- as.data.frame(sapply(data_dummy, as.numeric))</pre>
  data dummy$y <- as.factor(data dummy$y)</pre>
  dummy_var <- data_dummy</pre>
# ----- handle missing data
# ----- OCCUPATION_TYPE
dl_dummy_data <- read.csv2(".../credit_card_prediction/dl_na_data.csv", header = TRUE)</pre>
dl_dummy_data <- dummy_var(dl_dummy_data)</pre>
mf_dummy_data <- read.csv2(".../../credit_card_prediction/mf_na_data.csv", header = TRUE)</pre>
mf dummy data <- dummy var(mf dummy data)</pre>
mice_dummy_data <- read.csv2(".../..redit_card_prediction/mice_na_data.csv", header = TRUE)
mice_dummy_data <- dummy_var(mice_dummy_data)</pre>
# row, column, NA
cat("dl dummy data\t", dim(dl dummy data), any(is.na(dl dummy data)), "\n")
cat("mf_dummy_data\t", dim(mf_dummy_data), any(is.na(mf_dummy_data)), "\n")
cat("mice_dummy_data\t", dim(mice_dummy_data), any(is.na(mice_dummy_data)), "\n")
```

Load all data for training (one-hot, dummy, IV)

```
library(mlr3)
# function to load data into task and define target
dataToTask <- function(path, id, sep=';', header=TRUE){</pre>
  dt <- read.csv2(path, sep = sep, header = header)</pre>
  dt <- as.data.frame(sapply(dt, as.numeric))</pre>
  dt$y <- as.factor(dt$y)</pre>
  dataToTask <- TaskClassif$new(id = id, backend = dt, target = "y")</pre>
dl dummy task <-
  dataToTask(".../credit_card_prediction/dummy_data/dl_dummy_data.csv", "dl_dummy")
dl oh task <-
  dataToTask("../../credit card prediction/oh data/dl oh data.csv", "dl oh")
dl iv task <-
  dataToTask(".../credit_card_prediction/iv_data/dl_iv_data.csv", "dl_iv")
mf_dummy_task <-
  dataToTask(".../credit_card_prediction/dummy_data/mf_dummy_data.csv", "mf_dummy")
mf oh task <-
  dataToTask(".../credit_card_prediction/oh_data/mf_oh_data.csv", "mf_oh")
mf_iv_task <-
  dataToTask("../../credit_card_prediction/iv_data/mf_iv_data.csv", "mf_iv")
mice_dummy_task <-
  dataToTask(".../credit card prediction/dummy data/mice dummy data.csv", "mice dummy")
mice_oh_task <-
  dataToTask(".../credit_card_prediction/oh_data/mice_oh_data.csv", "mice_oh")
mice_iv_task <-
 dataToTask("../../credit card prediction/iv data/mice iv data.csv", "mice iv")
# combine all tasks into one list
dl <- list(dummy=dl_dummy_task, oh=dl_oh_task, iv=dl_iv_task)</pre>
mf <- list(dummy=mf_dummy_task, oh=mf_oh_task, iv=mf_iv_task)</pre>
mice <- list(dummy=mice_dummy_task, oh=mice_oh_task, iv=mice_iv_task)</pre>
# tasks[["<type>"]][["<code>"]], tasks$<type>$<code>
# ex. tasks[["dl"]][["dummy"]], tasks$dl$dummy
tasks <- list(dl=dl, mf=mf, mice=mice)</pre>
# remove unused variables (save memory)
```

KNN

mf_oh 36457 55
mf_iv 36457 33
mice_dummy 36457 47
mice_oh 36457 55
mice_iv 36457 33

```
library(mlr3)
library("mlr3learners")
# train one model with fixed seed
train_model <- function(task, learner, resampling){</pre>
  set.seed(2020)
  print(task$id)
  train_model <- resample(task, learner, resampling, store_models = TRUE)</pre>
  print(train_model)
# train all task with one learner
train_all <- function(tasks, learner, resampling){</pre>
  models <- list()
  miss_name <- c('dl', 'mf', 'mice')</pre>
  code_name <- c('dummy', 'oh', 'iv')</pre>
  for(missing in miss_name){
    for(coding in code_name){
      name <- pasteO(missing, "_", coding)</pre>
      task <- tasks[[missing]][[coding]]</pre>
      models[[name]] <- train_model(task, learner, resampling)</pre>
    }
  }
  train_all <- models
# evaluate multiple models with AUC
```

```
evaluate_models <- function(models){</pre>
  for(m in models){
   name <- m$task$id
    auc <- m$aggregate(msr("classif.auc"))[[1]]</pre>
   max_auc <- max(m$score(msr("classif.auc"))[,9])</pre>
   print(sprintf("%10s: %.4f (max: %.4f)", name, auc, max_auc))
    \#cat(pasteO(name, ": ", auc, "\t(max: ", max_auc, ")\n"))
  }
}
resampling = rsmp("cv", folds = 5)
learner <- lrn("classif.kknn", id = "knn", predict_type = "prob", k=15, distance=2, scale=FALSE)</pre>
models <- train_all(tasks, learner, resampling)</pre>
## [1] "dl_dummy"
## INFO [19:38:36.288] Applying learner 'knn' on task 'dl_dummy' (iter 1/5)
## INFO [19:38:38.973] Applying learner 'knn' on task 'dl_dummy' (iter 2/5)
## INFO [19:38:40.348] Applying learner 'knn' on task 'dl_dummy' (iter 3/5)
## INFO [19:38:41.550] Applying learner 'knn' on task 'dl_dummy' (iter 4/5)
## INFO [19:38:42.864] Applying learner 'knn' on task 'dl_dummy' (iter 5/5)
## <ResampleResult> of 5 iterations
## * Task: dl dummy
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "dl oh"
## INFO [19:38:44.121] Applying learner 'knn' on task 'dl oh' (iter 1/5)
## INFO [19:38:45.248] Applying learner 'knn' on task 'dl_oh' (iter 2/5)
## INFO [19:38:46.365] Applying learner 'knn' on task 'dl_oh' (iter 3/5)
## INFO [19:38:47.512] Applying learner 'knn' on task 'dl_oh' (iter 4/5)
## INFO [19:38:48.814] Applying learner 'knn' on task 'dl_oh' (iter 5/5)
## <ResampleResult> of 5 iterations
## * Task: dl oh
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "dl_iv"
## INFO [19:38:49.965] Applying learner 'knn' on task 'dl_iv' (iter 1/5)
## INFO [19:38:50.937] Applying learner 'knn' on task 'dl_iv' (iter 2/5)
## INFO [19:38:51.896] Applying learner 'knn' on task 'dl_iv' (iter 3/5)
## INFO [19:38:52.867] Applying learner 'knn' on task 'dl_iv' (iter 4/5)
## INFO [19:38:53.831] Applying learner 'knn' on task 'dl_iv' (iter 5/5)
## <ResampleResult> of 5 iterations
## * Task: dl iv
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "mf_dummy"
## INFO [19:38:54.873] Applying learner 'knn' on task 'mf_dummy' (iter 1/5)
## INFO [19:38:57.725] Applying learner 'knn' on task 'mf_dummy' (iter 2/5)
## INFO [19:39:00.443] Applying learner 'knn' on task 'mf_dummy' (iter 3/5)
       [19:39:03.290] Applying learner 'knn' on task 'mf_dummy' (iter 4/5)
## INFO [19:39:05.886] Applying learner 'knn' on task 'mf_dummy' (iter 5/5)
## <ResampleResult> of 5 iterations
```

```
## * Task: mf_dummy
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "mf oh"
## INFO [19:39:08.561] Applying learner 'knn' on task 'mf_oh' (iter 1/5)
## INFO [19:39:11.312] Applying learner 'knn' on task 'mf_oh' (iter 2/5)
## INFO [19:39:14.057] Applying learner 'knn' on task 'mf_oh' (iter 3/5)
## INFO [19:39:16.833] Applying learner 'knn' on task 'mf_oh' (iter 4/5)
## INFO [19:39:19.592] Applying learner 'knn' on task 'mf_oh' (iter 5/5)
## <ResampleResult> of 5 iterations
## * Task: mf_oh
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "mf_iv"
## INFO [19:39:22.591] Applying learner 'knn' on task 'mf_iv' (iter 1/5)
## INFO [19:39:25.024] Applying learner 'knn' on task 'mf_iv' (iter 2/5)
## INFO [19:39:27.419] Applying learner 'knn' on task 'mf_iv' (iter 3/5)
## INFO [19:39:29.777] Applying learner 'knn' on task 'mf_iv' (iter 4/5)
## INFO [19:39:32.550] Applying learner 'knn' on task 'mf_iv' (iter 5/5)
## <ResampleResult> of 5 iterations
## * Task: mf_iv
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "mice_dummy"
## INFO [19:39:34.973] Applying learner 'knn' on task 'mice_dummy' (iter 1/5)
## INFO [19:39:37.580] Applying learner 'knn' on task 'mice_dummy' (iter 2/5)
## INFO [19:39:40.219] Applying learner 'knn' on task 'mice_dummy' (iter 3/5)
        [19:39:42.844] Applying learner 'knn' on task 'mice_dummy' (iter 4/5)
       [19:39:45.486] Applying learner 'knn' on task 'mice_dummy' (iter 5/5)
## <ResampleResult> of 5 iterations
## * Task: mice_dummy
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "mice_oh"
## INFO [19:39:48.153] Applying learner 'knn' on task 'mice_oh' (iter 1/5)
## INFO [19:39:50.801] Applying learner 'knn' on task 'mice_oh' (iter 2/5)
## INFO [19:39:53.515] Applying learner 'knn' on task 'mice_oh' (iter 3/5)
## INFO [19:39:56.443] Applying learner 'knn' on task 'mice_oh' (iter 4/5)
## INFO [19:39:59.181] Applying learner 'knn' on task 'mice_oh' (iter 5/5)
## <ResampleResult> of 5 iterations
## * Task: mice_oh
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
## [1] "mice_iv"
## INFO [19:40:01.975] Applying learner 'knn' on task 'mice_iv' (iter 1/5)
## INFO [19:40:04.301] Applying learner 'knn' on task 'mice_iv' (iter 2/5)
## INFO [19:40:06.608] Applying learner 'knn' on task 'mice_iv' (iter 3/5)
## INFO [19:40:08.925] Applying learner 'knn' on task 'mice_iv' (iter 4/5)
## INFO [19:40:11.465] Applying learner 'knn' on task 'mice_iv' (iter 5/5)
```

```
## <ResampleResult> of 5 iterations
## * Task: mice_iv
## * Learner: knn
## * Warnings: 0 in 0 iterations
## * Errors: 0 in 0 iterations
evaluate_models(models)
## [1] " dl_dummy: 0.7445 (max: 0.7839)"
## [1] "
         dl_oh: 0.7445 (max: 0.7839)"
## [1] "
            dl_iv: 0.7445 (max: 0.7841)"
## [1] " mf_dummy: 0.7520 (max: 0.7884)"
## [1] "
         mf_oh: 0.7520 (max: 0.7884)"
## [1] " mf_iv: 0.7520 (max: 0.7885)"
## [1] "mice_dummy: 0.7519 (max: 0.7884)"
## [1] " mice_oh: 0.7519 (max: 0.7884)"
## [1] " mice_iv: 0.7520 (max: 0.7886)"
```