# Pre-processing measures

#### **Imports**

```
library(fairmodels)
library(tidymodels)
## -- Attaching packages ----- tidymodels 0.2.0 --
## v broom
               0.8.0 v recipes
                                         0.2.0
              0.1.1 v rsample
1.0.9 v tibble
3.3.6 v tidyr
1.0.0 v tune
## v dials
                                         0.1.1
## v dplyr
                                         3.1.7
## v ggplot2
                                        1.2.0
## v infer
                                        0.2.0
## v modeldata 0.1.1 v workflows 0.2.6 ## v parsnip 0.2.1 v workflowsets 0.2.1
## v purrr
                 0.3.4 v yardstick 0.0.9
## -- Conflicts ----- tidymodels conflicts() --
## x purrr::discard() masks scales::discard()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
## x recipes::step() masks stats::step()
## * Dig deeper into tidy modeling with R at https://www.tmwr.org
library(rpart)
## Attaching package: 'rpart'
## The following object is masked from 'package:dials':
##
##
      prune
library(discrim)
## Attaching package: 'discrim'
## The following object is masked from 'package:dials':
##
##
       smoothness
```

```
source("../scripts/metrics_on_dataset.R")
## -- Attaching packages -----
                                     ----- tidyverse 1.3.1 --
## v readr
          2.1.2
                   v forcats 0.5.1
## v stringr 1.4.0
## -- Conflicts ----- tidyverse conflicts() --
## x readr::col_factor() masks scales::col_factor()
## x purrr::discard() masks scales::discard()
## x dplyr::filter()
                    masks stats::filter()
## x stringr::fixed() masks recipes::fixed()
                 masks stats::lag()
masks yardstick::spec()
## x dplyr::lag()
## x readr::spec()
```

#### Data

```
df <- read_rds("../data/selection.rds") %>%
    select(-rating, -gender)

df_test <- read_rds("../data/selection_test.rds") %>%
    select(-rating, -gender)
```

## Massaging

```
nB_df <- df %>%
  mutate(accepted = as.factor(accepted),
         index = 1:1000)
fitted_Bayes_model <- naive_Bayes() %>%
  fit(accepted ~ nationality + test_score + english_cert +
        extracurricular,
      nB_df)
nB predictions <- predict(fitted Bayes model, nB df, type = "prob")
nB_df \leftarrow nB_df \%
  mutate(prob_FALSE = nB_predictions[[".pred_FALSE"]],
         prob_TRUE = nB_predictions[[".pred_TRUE"]])
prob_true = sum(nB_df$accepted == "TRUE") / nrow(nB_df)
non dutch acceptance = nrow(
  filter(nB_df, accepted == "TRUE", nationality == "Non_Dutch")) /
  nrow(filter(nB_df, nationality == "Non_Dutch"))
print(non_dutch_acceptance)
```

## [1] 0.111332

```
while (non_dutch_acceptance < round(prob_true, 3)) {</pre>
  # Flip highest Non-Dutch
  highest_non_dutch <- nB_df %>%
   filter(accepted == "FALSE", nationality == "Non_Dutch") %>%
   arrange(desc(prob_TRUE)) %>%
    slice head(n = 1) %>%
    select(index) %>%
   as.integer()
  # Flip lowest Dutch
  lowest_dutch <- nB_df %>%
   filter(accepted == "TRUE", nationality == "Dutch") %>%
   arrange(prob_TRUE) %>%
   slice_head(n = 1) \%
    select(index) %>%
   as.integer()
  nB_df \leftarrow nB_df \%
   mutate(
      accepted = replace(accepted, index == highest_non_dutch, "TRUE"),
      accepted = replace(accepted, index == lowest_dutch, "FALSE")
    )
 non_dutch_acceptance = nrow(filter(
   nB_df, accepted == "TRUE", nationality == "Non_Dutch")) /
   nrow(filter(nB_df, nationality == "Non_Dutch"))
  # print(non_dutch_acceptance)
nB_df \leftarrow nB_df \%
  select(-index,-prob_FALSE,-prob_TRUE)
model <- decision_tree(mode = "classification")</pre>
massaging_results <- all_metrics(nB_df, model, df_test)</pre>
print_all_metrics("Massaging", massaging_results)
## [1] "Massaging"
## [1] "Group fairness"
## # A tibble: 4 x 4
     nationality accepted total perc
##
                 <fct>
                          <int> <dbl>
     <fct>
## 1 Dutch
                 FALSE
                            416 83.7
                 TRUE
## 2 Dutch
                             81 16.3
## 3 Non_Dutch FALSE
                            421 83.7
                 TRUE
                             82 16.3
## 4 Non_Dutch
## [1] "Causal discrimination"
## [1] 0
## [1] "Unawareness"
## # A tibble: 4 x 4
   nationality predicted_accepted total perc
##
   <fct>
                <fct>
                                    <int> <dbl>
## 1 Dutch
               FALSE
                                      407 80.8
                                       97 19.2
## 2 Dutch
                 TRUE
## 3 Non_Dutch FALSE
                                      419 84.5
```

## Reweighting

## [1] "Unawareness"

```
weights <- reweight(df$nationality, as.numeric(df$accepted))</pre>
weighted_model <- rpart(accepted ~ nationality + test_score + english_cert +</pre>
                          extracurricular, df, weights)
weighted_df <- df_test %>%
  mutate(predicted = as.logical(predict(weighted_model, df_test)))
print("Reweighting")
## [1] "Reweighting"
print("Group fairness")
## [1] "Group fairness"
print(group_fairness(weighted_df, nationality, predicted)[[1]])
## # A tibble: 4 x 4
    nationality predicted total perc
##
##
     <fct>
                         <int> <dbl>
            <lgl>
## 1 Dutch
                 FALSE
                            379 75.2
## 2 Dutch
                             125 24.8
                 TRUE
## 3 Non_Dutch FALSE
                             453 91.3
## 4 Non_Dutch
                 TRUE
                              43 8.67
print("Causal Discrimination")
## [1] "Causal Discrimination"
pop_size <- nrow(df)</pre>
# Flip nationalities
inverted df <- df test %>%
  mutate(nationality = ifelse(nationality == "Dutch", "Non_Dutch", "Dutch"))
# Add prediction column
eval_df <- df_test %>%
  mutate(inverted_accepted = as.logical(predict(weighted_model, inverted_df)),
         different = accepted != inverted_accepted)
print(sum(eval_df$different)/pop_size)
## [1] 0.13
print("Unawareness")
```

```
weighted_unaware_model <- rpart(accepted ~ test_score + english_cert +</pre>
                                  extracurricular, df, weights)
weighted_unaware_df <- df_test %>%
  mutate(predicted = as.logical(predict(weighted_unaware_model, df_test)))
print(group_fairness(weighted_unaware_df, nationality, predicted)[[1]])
## # A tibble: 4 x 4
     nationality predicted total perc
     <fct>
                <lg1>
                           <int> <dbl>
## 1 Dutch
                             379 75.2
                FALSE
                             125 24.8
## 2 Dutch
                TRUE
## 3 Non Dutch
                             390 78.6
                FALSE
## 4 Non Dutch
                TRUE
                             106 21.4
```

### Resampling

```
resample_indexes <- resample(df$nationality, as.numeric(df$accepted))
resampled_df <- df[resample_indexes, ]
model <- decision_tree(mode = "classification")
resamping_results <- all_metrics(resampled_df, model, df_test)
print_all_metrics("Resampling", resamping_results)

## [1] "Resampling"
## [1] "Group fairness"</pre>
```

```
## # A tibble: 4 x 4
    nationality accepted total perc
                         <int> <dbl>
##
    <fct>
                <lgl>
## 1 Dutch
                FALSE
                           416 83.7
## 2 Dutch
                TRUE
                            81 16.3
## 3 Non_Dutch
                FALSE
                           421 83.7
## 4 Non_Dutch
                TRUE
                            82 16.3
## [1] "Causal discrimination"
## [1] 0.13
## [1] "Unawareness"
## # A tibble: 4 x 4
    nationality predicted_accepted total perc
##
                <fct>
                                   <int> <dbl>
## 1 Dutch
                                     401 79.6
                FALSE
## 2 Dutch
                TRUE
                                     103 20.4
## 3 Non_Dutch
                FALSE
                                    444 89.5
## 4 Non_Dutch
                TRUE
                                     52 10.5
```