Positivity Assumption Evaluation

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Approach

The goal of this document is to select the number of splits to use for the continuous variables contained in W with the objective of removing a minimal amount of observations due to violations of the positivity assumption. There are eight covariates W in the data which are discussed.

- $1. \ satisfaction_level$
- 2. last evaluation
- 3. number_project
- 4. average monthy hours
- 5. time_spend_company
- 6. Work_accident
- 7. promotion_last_5years
- 8. sales

There are three continuous numeric variables. Quantiles are used to split the variables into buckets of size k = 2, 3, ..., 10.

- 1. satisfaction level
- 2. last evaluation
- 4. average_montly_hours

There are two discrete numeric variables. Quantiles are used to split them into buckets of k=2,3.

- 3. number_project
- 5. time_spend_company

There are three variables that are not transformed. This because they are already binary or becuase they have no natural ordering:

- 6. Work accident
- 7. promotion_last_5years
- 8. sales

Define functions

```
get_data <- function(dir){
    df <- read.csv(paste0(dir, 'HR_comma_sep_2.csv'))
    df <- df %>% filter(salary != 'medium')
    return(df)
}

split_var <- function(x, k){
    x_split <- cut(x, breaks = quantile(x, probs = seq(0, 1, 1/k)), include.lowest = TRUE)
    if(any(is.na(x_split))) stop('There are NA values in')
    return(factor(x_split, labels = 1:k))
}

count_drop_outs <- function(df, k_disc = 2, k_cont = 2){
    # Count the total number of observations removed</pre>
```

```
# when we remove strata with positivity violations
  n pos viol <- df %>%
  mutate(
    satisfaction_level_s = split_var(satisfaction_level, k_cont)
    , last_evaluation_s = split_var(last_evaluation, k_cont)
    , number_project_s = split_var(number_project, k_disc)
    , average_montly_hours_s = split_var(average_montly_hours, k_cont)
    , time_spend_company_s = split_var(time_spend_company, k_disc)
  ) %>%
  group_by(
    satisfaction_level_s
    , last_evaluation_s
    , number_project_s
    , average_montly_hours_s
    , time_spend_company_s
    , Work_accident
    , promotion_last_5years
    , sales
  ) %>%
  summarize(
    sal_h = sum(salary == 'high'),
   sal_l = sum(salary == 'low'),
   class_size = n()
  ) %>% ungroup() %>%
  filter(sal_h == 0 | sal_l == 0) %>%
  summarize(
    tot = sum(sal_h) + sum(sal_l)
  )
  return(unlist(n_pos_viol)[1])
format_df <- function(m){</pre>
  df <- as.data.frame(m)</pre>
  colnames(df) <- c('k_discrete', 'k_continuous', 'num_dropped')</pre>
  df$k_discrete <- factor(df$k_discrete)</pre>
  df$pct_dropped <- df$num_dropped / nrow(d)</pre>
  return(df)
}
count_over_k <- function(df){</pre>
  m \leftarrow matrix(nrow = 2*9, ncol = 3)
  i <- 1
  for (k_disc in 2:3){
    for (k_cont in 2:10){
      m[i,1:2] \leftarrow c(k_disc, k_cont)
      m[i,3] <- count_drop_outs(df, k_disc, k_cont)</pre>
      i \leftarrow i + 1
    }
  }
  return(m)
```

Run evaluation

```
library(dplyr)
library(ggplot2)

DIR <- '/Users/josiahdavis/Documents/Berkeley/PH252D/data/' # <= UPDATE AS NEEDED

# Read in data
d <- get_data(DIR)

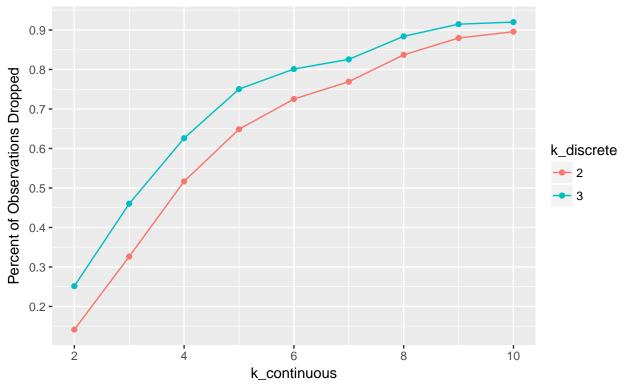
# Count the results over a range of k values
raw_results <- count_over_k(d)

# Format results to make it easier to visualize
results <- format_df(raw_results)

# Create the data visusalizeion
visualize(results)</pre>
```

Evaluation of the Positivity Assumption

% of dropped observation for different numbers of buckets k



Other

