Data_splitting

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Importing data

Algorithm for splitting variables into training and testing set evenly

```
# My goal is to use quantiles to split data evenly
split <- function(variable, p) {</pre>
  #Case for categorical variable or small number of levels of continuosu variable
  if (length(unique(variable)) < 10){</pre>
    unique <- unique(variable)</pre>
    unique_list <- list()</pre>
    training_store <- vector()</pre>
    testing store <- vector()</pre>
    for (i in 1:length(unique)) {
      unique_list[[i]] <- variable[variable %in% unique[i]]</pre>
      #To make sure all variables of training/testing have the same number
      ifelse((i \% 2) == 1 && (length(variable)*p)\%1 > 0.5,
             index <- sample(1:length(unique_list[[i]]),</pre>
                       ceiling(length(unique_list[[i]])*p),replace=FALSE),
             index <- sample(1:length(unique_list[[i]]),</pre>
                       floor(length(unique_list[[i]])*p),replace=FALSE))
      #To make sure all variables of training/testing have the same number
      ifelse((i \% 2) == 1 && (length(variable)*p)\%1 < 0.5,
              index <- sample(1:length(unique_list[[i]]),</pre>
                       floor(length(unique_list[[i]])*p),replace=FALSE),
             index <- sample(1:length(unique list[[i]]),</pre>
                       ceiling(length(unique_list[[i]])*p),replace=FALSE))
      training_store <- c(training_store, unique_list[[i]][index])</pre>
      testing_store <- c(testing_store, unique_list[[i]][-index])</pre>
    split <- list(training = training_store, testing = testing_store)</pre>
  }
```

```
else {
 #Case for continuous variable
 variable <- sort(variable)</pre>
 #Using quantiles to split data evenly
 smallest <- min(variable)</pre>
 first quantile <- variable[round(length(variable)*0.25, 0)]
 second_quantile <- variable[round(length(variable)*0.5, 0)]</pre>
 third_quantile <- variable[round(length(variable)*0.75, 0)]
 largest <- max(variable)</pre>
 summary <- c(smallest, first_quantile, second_quantile, third_quantile, largest)</pre>
 training_data <- list()</pre>
 testing_data <- list()</pre>
 training_store <- vector()</pre>
 testing_store <- vector()</pre>
 for(i in 1:(length(summary)-1)){
   #To make sure all variables of training/testing have the same nummber
   ifelse(i == (length(summary)-1),
             Q_data <- variable[variable>=summary[i] & variable<=summary[i+1]],</pre>
             Q_data <- variable[variable>=summary[i] & variable<summary[i+1]])</pre>
   #To make sure all variables of training/testing have the same nummber
   ifelse((i %% 2) == 1,
             index <- sample(1:length(Q_data), ceiling(length(Q_data)*p), replace=FALSE),</pre>
             index <- sample(1:length(Q_data), floor(length(Q_data)*p), replace=FALSE))</pre>
   training_data[[i]] <- Q_data[index]</pre>
   testing_data[[i]] <- Q_data[-index]</pre>
   training_store <- c(training_store, training_data[[i]])</pre>
   testing_store <- c(testing_store, testing_data[[i]])</pre>
 split <- list(training = training_store, testing = testing_store)</pre>
return(split)
```

```
#Let's check if the algorithm works for continuous variable
split1 <- split(data$radius_mean, 0.8)</pre>
data_frame(mean = mean(split1$training), sd = sd(split1$training),
          proportion = length(split1$training)/nrow(data))
## # A tibble: 1 x 3
     mean
           sd proportion
     <dbl> <dbl>
                     <dbl>
## 1 14.2 3.56
                     0.800
data_frame(mean = mean(split1$testing), sd = sd(split1$testing),
          proportion = length(split1$testing)/nrow(data))
## # A tibble: 1 x 3
##
     mean
             sd proportion
     <dbl> <dbl> <dbl>
## 1 14.0 3.40
                     0.200
#Looks good
#Let's check if the algorithm works for categorical variable
split2 <- split(data$diagnosis, 0.8)</pre>
data frame(mean = mean(split2$training), sd = sd(split2$training),
          proportion = length(split2$training)/length(data$radius_mean))
## # A tibble: 1 x 3
     mean
           sd proportion
##
     <dbl> <dbl>
                 <dbl>
## 1 0.371 0.484
                     0.800
data_frame(mean = mean(split2$testing), sd = sd(split2$testing),
          proportion = length(split2$testing)/length(data$radius_mean))
## # A tibble: 1 x 3
             sd proportion
     mean
     <dbl> <dbl>
                     <dbl>
## 1 0.377 0.487
                     0.200
#Looks good
```

Applying algorithm to get the training/testing data frame of entire data

```
data_split <- function(data,split, p){
  data_split <- map(data, split, p)
  training_list <- list()</pre>
```

```
testing_list <- list()
for (i in 1:length(data_split)) {

   training_list[[i]] <- data_split[[i]]$training
   testing_list[[i]] <- data_split[[i]]$testing
}

names(training_list) <- names(data_split)
names(testing_list) <- names(data_split)

training <- dplyr::bind_rows(training_list) %>% data.frame()
   testing <- dplyr::bind_rows(testing_list) %>% data.frame()

return(list(training, testing))
}
```

Combining result together to make it reader-frindly

```
trainging_result <-
    skimr::skim_without_charts(data_split(data,split, 0.8)[1]) %>% data.frame() %>%
    select(2,5,6) %>%
    rename(training_mean = numeric.mean, training_sd = numeric.sd) %>%
    mutate_if(is.numeric, ~round(.x, digits = 3)) %>%
    mutate_if(is.numeric, ~format(.x, scientific = FALSE))

testing_result <-
    skimr::skim_without_charts(data_split(data,split, 0.8)[2]) %>%
    select(2,5,6) %>%
    rename(testing_mean = numeric.mean, testing_sd = numeric.sd) %>%
    mutate_if(is.numeric, ~round(.x, digits = 3)) %>%
    mutate_if(is.numeric, ~format(.x, scientific = FALSE))

trainging_result %>% left_join(testing_result, by = "skim_variable") %>% kable()
```

skim_variable	training_mean	training_sd	testing_mean	testing_sd
id	27573499.703	112500444.331	20935774.711	86027240.237
diagnosis	0.371	0.484	0.377	0.487
radius_mean	14.111	3.484	13.992	3.278
texture_mean	19.261	4.177	19.102	4.410
smoothness_mean	0.097	0.014	0.096	0.015
compactness_mean	0.104	0.052	0.103	0.050
concave_points_mean	0.049	0.039	0.048	0.036
symmetry_mean	0.181	0.027	0.181	0.029
fractal_dimension_mean	0.063	0.007	0.063	0.008
radius_se	0.399	0.260	0.401	0.239
texture_se	1.212	0.545	1.230	0.551
smoothness_se	0.007	0.003	0.007	0.004
compactness_se	0.026	0.018	0.024	0.015
concavity_se	0.032	0.032	0.030	0.021
concave_points_se	0.012	0.006	0.011	0.005
symmetry_se	0.021	0.008	0.021	0.009
fractal_dimension_se	0.004	0.003	0.004	0.002
radius_worst	16.251	4.844	16.306	4.711
smoothness_worst	0.132	0.023	0.133	0.023
compactness_worst	0.254	0.156	0.251	0.158
concavity_worst	0.273	0.209	0.261	0.200
symmetry_worst	0.290	0.061	0.287	0.055
fractal_dimension_worst	0.084	0.018	0.083	0.016

training_nrow	testing_nrow	eighty_percent_data_nrow	twenty_percent_data_nrow
455	114	455.2	113.8