Tathagata Bardhan Importing the dataset rm(list=ls()) `cardio\_train(project)` <- read.csv("C:/Users/user/OneDrive/Desktop/project-baan bapat/cardio\_train(project).csv" data=`cardio\_train(project)` head(data) ## id age gender height weight ap\_hi ap\_lo cholesterol gluc smoke alco active ## 1 0 18393 2 168 62 110 80 1 1 ## cardio ## 1 ## 2 ## 3 1 ## 4 ## 5 ## 6 Data cleaning **Null value detection** sapply(data, function(x) sum(is.na(x))) gender height weight ap\_hi 0 ## Θ 0 0 ap\_lo cholesterol gluc 0 0 0 smoke active alco ## 0 0 ## cardio ## So there are no null values in the dataset Identifying duplicate rows library(tidyverse) ## -- Attaching packages ----- tidyverse 1.3.0 --## v ggplot2 3.3.3 v purrr 0.3.4 ## v tibble 3.0.4 v dplyr 1.0.3 ## v tidyr 1.1.2 v stringr 1.4.0## v readr 1.4.0 v forcats 0.5.0 ## -- Conflicts ------ tidyverse\_conflicts() --## x dplyr::filter() masks stats::filter() ## x dplyr::lag() masks stats::lag() df=data %>% distinct(age, gender, height, weight, ap\_hi, ap\_lo, cholesterol, gluc, smoke, alco, active, cardio, . keep\_all=TRUE) dim(df) ## [1] 69976 Although the id is unique for all the patients, there are 24 duplicate rows based on all other columns Overview and summary statistics of the data str(df) ## 'data.frame': 69976 obs. of 13 variables: ## \$ id : int 0 1 2 3 4 8 9 12 13 14 ... ## \$ age : int 18393 20228 18857 17623 17474 21914 22113 22584 17668 19834 ... ## \$ gender : int 2 1 1 2 1 1 1 2 1 1 ... ## \$ height : int 168 156 165 169 156 151 157 178 158 164 ... ## \$ weight : num 62 85 64 82 56 67 93 95 71 68 ... ## \$ ap\_hi : int 110 140 130 150 100 120 130 130 110 110 ... ## \$ ap\_lo : int 80 90 70 100 60 80 80 90 70 60 ... ## \$ cholesterol: int 1 3 3 1 1 2 3 3 1 1 ... ## \$ gluc : int 1 1 1 1 1 2 1 3 1 1 ... ## \$ smoke : int 00000000000... ## \$ alco : int 00000000000... ## \$ active : int 1 1 0 1 0 0 1 1 1 0 ... ## \$ cardio : int 0 1 1 1 0 0 0 1 0 0 ... We require gender, cholesterol, gluc, smoke, alco, active, cardio variables to be Factors but these are continuous here. So we need explicit conversion df[,c(3,8:13)]=lapply(df[,c(3,8:13)], function(x) as.factor(x))str(df) ## 'data.frame': 69976 obs. of 13 variables: ## \$ id : int 0 1 2 3 4 8 9 12 13 14 ... : int 18393 20228 18857 17623 17474 21914 22113 22584 17668 19834 ... ## \$ age ## \$ gender : Factor w/ 2 levels "1", "2": 2 1 1 2 1 1 1 2 1 1 ... ## \$ height : int 168 156 165 169 156 151 157 178 158 164 ... ## \$ weight : num 62 85 64 82 56 67 93 95 71 68 ... : int 110 140 130 150 100 120 130 130 110 110 ... ## \$ cholesterol: Factor w/ 3 levels "1", "2", "3": 1 3 3 1 1 2 3 3 1 1 ... ## \$ gluc : Factor w/ 3 levels "1", "2", "3": 1 1 1 1 1 2 1 3 1 1 ... ## \$ smoke : Factor w/ 2 levels "0", "1": 1 1 1 1 1 1 1 1 1 1 ... ## \$ alco : Factor w/ 2 levels "0", "1": 1 1 1 1 1 1 1 1 1 1 ... ## \$ active : Factor w/ 2 levels "0", "1": 2 2 1 2 1 1 2 2 2 1 ... ## \$ cardio : Factor w/ 2 levels "0", "1": 1 2 2 2 1 1 1 2 1 1 ... summary(df[,-c(1)]) gender height weight ap\_hi ## Min. :10798 1:45509 Min. :55.0 Min. :10.00 Min. :-150.0 ## 1st Qu.:17664 2:24467 1st Qu.:159.0 1st Qu.: 65.00 1st Qu.: 120.0 cholesterol gluc smoke alco active ap\_lo ## Min. : -70.00 1:52361 1:59455 0:63807 0:66212 0:13735 ## 1st Qu.: 80.00 2: 9549 2: 5190 1: 6169 1: 3764 1:56241 Median: 80.00 3: 8066 3: 5331 ## Mean : 96.64 ## 3rd Qu.: 90.00 :11000.00 Max. cardio 0:35004 1:34972 ## ## ## ## Thus the values for each variable are of much different scales from each other and so we may need to standardize these later Univariate Analysis For variable 'Age' par(mfrow=c(1,3))library(robustbase) adjbox(df\$age, main="Adjusted Boxplot of Age") hist(df\$age, main="Histogram of Age", prob=TRUE) plot(density(df\$age), main="Density Plot of Age") **Adjusted Boxplot of Age Density Plot of Age** Histogram of Age 0.00015 22000 20000 18000 Density 0.0000.0 10000 14000 18000 22000 16000 22000 10000 df\$age N = 69976 Bandwidth = 238.5 • From the summary statistics of Age we see that here we have people aged from 29 years to 65 years approx and more precisely majority of the patients are between 49 years to 60 years • The distribution of Age is slightly negatively skewed · Adjusted boxplot is preferred here, which shows no presence of outliers For variable 'Height' par(mfrow=c(1,3))adjbox(df\$height, main="Adjusted Boxplot of Height") hist(df\$height, main="Histogram of Height", prob=TRUE) adjbox(df\$height~df\$cardio,names=c("Without CVD","With CVD"),main="Height vs CVD") **Adjusted Boxplot of Height Histogram of Height** Height vs CVD 0.04 200 200 0.03 Density 150 150 0.02 100 100 0.01 00 20 100 150 200 Without CVD With CVD df\$height · Heights are more or less symmetrically distributed • Most of the patients are of height 160 to 170 cm • On an average, patients with or without CVD have more or less same height. • Thus height is not an important factor in detecting CVD • So, we are bothered about the outliers present in variable Height For variable 'Weight' par(mfrow=c(1,3)) adjbox(df\$weight, main="Adjusted Boxplot of Weight") hist(df\$weight, main="Histogram of Weight", prob=TRUE) plot(density(df\$weight), main="Density Plot of Weight") **Density Plot of Weight Adjusted Boxplot of Weight Histogram of Weight** 0.030 0.025 0.03 150 0.020 0.015 100 0.010 0.01 50 0.005 0.000 0.00 100 100 150 50 150 df\$weight N = 69976 Bandwidth = 1.226 · Here Weight is slightly positively skewed • Majority of people in our data have weight 60 to 80 kg • From the summary statistics we can argue that Weight of 10 kgs is not possible for a person 29 years of Age and above. • So, we suspect that outliers beyond the lower whisker of the boxplot of Weight are only the meaningful ones that should be removed. df\$weight[df\$weight<adjboxStats(df\$weight)\$fence[1]]=adjboxStats(df\$weight)\$fence[1]</pre> For pulse pressure par(mfrow=c(1,3)) adjbox(df\$ap\_hi-df\$ap\_lo,main="Adjusted Boxplot of pulse pressure") hist(df\ap\_hi-df\ap\_lo,xlim=c(-100,500),main="Histogram of pulse pressure",prob=TRUE) plot(density(df\$ap\_hi-df\$ap\_lo),xlim=c(-100,100),main="Density Plot of pulse pressure") Adjusted Boxplot of pulse pressure Histogram of pulse pressure **Density Plot of pulse pressure** 5e-04 4e-04 3e-04 5000 Density Density 0.2 2e-04 0 -5000 -10000 0e+00 -100 0 100 300 -50 0 df\$ap\_hi - df\$ap\_lo N = 69976 Bandwidth = 0.7214 **Bivariate Analysis** Comparison of 'Age' with CVD par(mfrow=c(1,2)) adjbox(df\$age~df\$cardio, names=c("Without CVD", "With CVD"), main="Age vs CVD") plot(density(df\$age[df\$cardio==1]), main="Age dist.(with CVD)") Age vs CVD Age dist.(with CVD) 24000 0.00015 20000 0.00010 Density 16000 12000 0.0000.0 Without CVD With CVD 14000 18000 22000 N = 34972 Bandwidth = 257.2 • Patients who are suffering from CVD are, on an average, older than patients who donot suffer from CVD • More precisely patients having age between 20,000-22,000 days i.e. between 55 to 60 years approximately suffer mostly from CVD Comparison of 'Weight' with CVD par(mfrow=c(1,2))adjbox(df\$weight~df\$cardio,names=c("Without CVD","With CVD"),main="Weight vs CVD") plot(density(df\$weight[df\$cardio==1]), main="Weight dist.(with CVD)") Weight dist.(with CVD) Weight vs CVD 200 0.030 150 0.020 Density 0.010 100 0.000 Without CVD With CVD 50 100 150 200 N = 34972 Bandwidth = 1.575 • Thus, on an average, people who suffer from CVD have more weight than those who don't • Infact heart diseases are mostly occuring for patients whose weights are within 65-80 kgs Comparison of 'Pulse pressure' with CVD par(mfrow=c(1,2)) adjbox((df\$ap\_hi-df\$ap\_lo)~df\$cardio,ylim=c(-100,100),names=c("Without CVD","With CVD"),main="Pulse pressure vs plot(density(df\$ap\_hi[df\$cardio==1]-df\$ap\_lo[df\$cardio==1]),xlim=c(-100,100),main="Pulse pressure(with CVD)") Pulse pressure(with CVD) Pulse pressure vs CVD 0.15 100 50 10 Density 0 -50 -100 Without CVD With CVD -50 100 0 50 N = 34972 Bandwidth = 1.657 • On an average, people suffering from CVD have higher Pulse pressure • Many people suffering from CVD have Pulse pressure higher than normal range i.e. greater than 60 mm Hg So, high pulse pressure can be a good indication of CVD Comparison of CVD with 'Cholesterol level' library(gmodels)  $\label{lem:constable} CrossTable(df\cholesterol, df\cardio, prop. c = F, prop. r = F, prop. t = F, prop. chisq = F, chisq = T)$ ## ## Cell Contents ## ## Total Observations in Table: 69976 ## ## | df\$cardio ## df\$cholesterol | 0 | 1 | Row Total | -----|-----|------| 1 | 29313 | 23048 | 52361 | -----|-----|------| 2 | 3799 | 5750 | 9549 | ## -----|-----| 3 | 1892 | 6174 | 8066 | -----|-----|------| Column Total | 35004 | 34972 | 69976 | ## -----|-----| ## ## Statistics for All Table Factors ## ## Pearson's Chi-squared test ##  $Chi^2 = 3421.399$  d.f. = 2 p = 0 ## ## ## Chi-sq test of Association shows dependence of presence of CVD on Cholesterol level par(mfrow=c(1,1))mytable=table(df\$cardio,df\$cholesterol) mytable2=prop.table(mytable, 1) barplot(mytable2, beside = TRUE, legend.text = c("Without CVD", "With CVD"), xlab = "Cholesterol level")o. ■ Without CVD ☐ With CVD 0.0 2 3 1 Cholesterol level • Patients having Cholesterol level well above normal suffer very much from CVD • Patients having Cholesterol in normal level are much less sufferers of CVD Comparison of CVD with 'Glucose level'  $CrossTable(df\gluc, df\cardio, prop.c = F, prop.r = F, prop.t = F, prop.chisq = F, chisq = T)$ ## ## Cell Contents |-----| ## -----| ## ## ## Total Observations in Table: 69976 ## ## | df\$cardio df\$gluc | 0 | 1 | Row Total | -----|------| 1 | 30877 | 28578 | 59455 | ## -----| 2 | 2112 | 3078 | ## -----|-----| 3 | 2015 | 3316 | 5331 | ## -----| ## Column Total | 35004 | 34972 | 69976 | ## -----| ## Statistics for All Table Factors ## ## Pearson's Chi-squared test ## -----##  $Chi^2 = 586.1834$  d.f. = 2 p = 5.150929e-128 ## ## Chi-sq test of Association tells that presence of CVD depends upon Glucose level par(mfrow=c(1,1))mytable=table(df\$cardio,df\$gluc) mytable3=prop.table(mytable, 1) barplot(mytable3, beside = TRUE, legend.text = c("Without CVD", "With CVD"), xlab="Glucose level") ■ Without CVD o. ☐ With CVD 0.0 1 2 3 Glucose level · Patients having Glucose level well above normal have more tendency to develop CVD · Patients having normal Glucose level suffers comparatively less from CVD Comparison of CVD with 'Activity' CrossTable(df\$active, df\$cardio, prop.c = F, prop.r = F, prop.t = F, prop.chisq = F, chisq = T) ## ## ## Cell Contents 1-----| ## ## |-----| ## ## ## Total Observations in Table: 69976 ## ## ## | df\$cardio ## df\$active | 0 | 1 | Row Total | ## -----| 0 | 6375 | 7360 | 13735 | ## -----|-----| 1 | 28629 | 27612 | 56241 | ## ## -----|-----| ## Column Total | 35004 | 34972 | 69976 | ## -----| ## Statistics for All Table Factors ## ## ## Pearson's Chi-squared test ##  $Chi^2 = 89.01457$  d.f. = 1 p = 3.919161e-21 ## ## Pearson's Chi-squared test with Yates' continuity correction ##  $Chi^2 = 88.83506$  d.f. = 1 p = 4.29142e-21 ## ## From the results of Chi-sq test of Association we see that presence of CVD depends on physical activity of patients par(mfrow=c(1,1))mytable=table(df\$cardio,df\$active) mytable4=prop.table(mytable, 1) barplot(mytable4, beside = TRUE, legend.text = c("Without CVD", "With CVD"), args.legend = list(x = "topleft"), xlab= "Activity", main="Comparison of CVD with Activity of patients") Comparison of CVD with Activity of patients  $\infty$ o. ■ Without CVD ☐ With CVD 9.0 0.4 0.0 0 Activity There are more CVD patients among those who are not active • There is a bit lesser tendency to develop CVD among the people who are active Comparison of CVD with 'Smoking' CrossTable(df\$smoke, df\$cardio, prop.c = F, prop.r = F, prop.t = F, prop.chisq = F, chisq = T) ## ## ## Cell Contents |-----| ## ## Total Observations in Table: 69976 ## | df\$cardio ## ## df\$smoke | 0 | 1 | Row Total | ## -----|-----| 0 | 31764 | 32043 | 63807 | -----|-----| 1 | 3240 | 2929 | 6169 | ## -----|-----| ## Column Total | 35004 | 34972 | 69976 | ## -----| ## ## Statistics for All Table Factors ## Pearson's Chi-squared test ## -----##  $Chi^2 = 16.88387$  d.f. = 1 p = 3.973787e-05 ## Pearson's Chi-squared test with Yates' continuity correction ## -----##  $Chi^2 = 16.77447$  d.f. = 1 p = 4.209577e-05 ## Chi-sq test of Association indicates dependence of CVD on smoking habits par(mfrow=c(1,1))mytable=table(df\$cardio,df\$smoke) mytable5=prop.table(mytable, 1) barplot(mytable5, beside = TRUE, legend.text = c("Without CVD", "With CVD"), xlab="Smoking habit", main="Comparison of CVD with Smoking habit of patients") Comparison of CVD with Smoking habit of patients ■ Without CVD 0.8 ☐ With CVD 0 Smoking habit Pairwise correlation coefficients round((cor(df[,-c(1,3,8:13)])),4) age height weight ap\_hi ap\_lo 1.0000 -0.0815 0.0533 0.0208 0.0176 ## height -0.0815 1.0000 0.2852 0.0055 0.0062 ## weight 0.0533 0.2852 1.0000 0.0306 0.0439 ## ap\_hi 0.0208 0.0055 0.0306 1.0000 0.0161 ## ap\_lo 0.0176 0.0062 0.0439 0.0161 1.0000 cor.test(df\$ap\_hi,df\$ap\_lo)\$p.value ## [1] 2.092353e-05 There is significant correlation between variables 'ap\_hi' and 'ap\_lo' which may cause the problem of multicollinearity later

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