

# Heart Disease Analysis

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```
library(ggplot2)
library(dplyr)

## 
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
## 
##     filter, lag

## The following objects are masked from 'package:base':
## 
##     intersect, setdiff, setequal, union

library(MASS)

## 
## Attaching package: 'MASS'

## The following object is masked from 'package:dplyr':
## 
##     select

library(ca)
library(CCA)

## Loading required package: fda

## Loading required package: splines

## Loading required package: fds

## Loading required package: rainbow

## Loading required package: pcaPP

## Loading required package: RCurl
```

```

## Loading required package: deSolve

##
## Attaching package: 'fda'

## The following object is masked from 'package:graphics':
##      matplot

## Loading required package: fields

## Loading required package: spam

## Spam version 2.11-1 (2025-01-20) is loaded.
## Type 'help( Spam)' or 'demo( spam)' for a short introduction
## and overview of this package.
## Help for individual functions is also obtained by adding the
## suffix '.spam' to the function name, e.g. 'help( chol.spam)'.

##
## Attaching package: 'spam'

## The following objects are masked from 'package:base':
##      backsolve, forwardsolve

## Loading required package: viridisLite

## Loading required package: RColorBrewer

##
## Try help(fields) to get started.

library(sem)

library(corrplot)

## corrplot 0.95 loaded

library(MVA)

## Loading required package: HSAUR2

## Loading required package: tools

library(kableExtra)

##
## Attaching package: 'kableExtra'

## The following object is masked from 'package:dplyr':
##      group_rows

```

```
# Loading Data
data <- read.csv("C:/Users/Ty Campbell/Desktop/data/Heart_Disease_Prediction.csv")
head(data)
```

```
##   Age Sex Chest.pain.type BP Cholesterol FBS.over.120 EKG.results Max.HR
## 1 70   1             4 130          322            0           2        109
## 2 67   0             3 115          564            0           2        160
## 3 57   1             2 124          261            0           0        141
## 4 64   1             4 128          263            0           0        105
## 5 74   0             2 120          269            0           2        121
## 6 65   1             4 120          177            0           0        140
##   Exercise.angina ST.depression Slope.of.ST Number.of.vessels.fluro Thallium
## 1                 0            2.4            2                  3        3
## 2                 0            1.6            2                  0        7
## 3                 0            0.3            1                  0        7
## 4                 1            0.2            2                  1        7
## 5                 1            0.2            1                  1        3
## 6                 0            0.4            1                  0        7
##   Heart.Disease
## 1     Presence
## 2     Absence
## 3     Presence
## 4     Absence
## 5     Absence
## 6     Absence
```

```
# checking data types and null values
dim(data)
```

```
## [1] 270 14
```

```
colSums(is.na(data))
```

```
##               Age                      Sex          Chest.pain.type
##               0                      0                      0
##               BP                     Cholesterol          FBS.over.120
##               0                      0                      0
##               EKG.results          Max.HR          Exercise.angina
##               0                      0                      0
##               ST.depression      Slope.of.ST Number.of.vessels.fluro
##               0                      0                      0
##               Thallium          Heart.Disease
##               0                      0
```

```
str(data)
```

```
## 'data.frame': 270 obs. of 14 variables:
## $ Age : int 70 67 57 64 74 65 56 59 60 63 ...
## $ Sex : int 1 0 1 1 0 1 1 1 0 ...
## $ Chest.pain.type : int 4 3 2 4 2 4 3 4 4 4 ...
## $ BP : int 130 115 124 128 120 120 130 110 140 150 ...
```

```

## $ Cholesterol      : int 322 564 261 263 269 177 256 239 293 407 ...
## $ FBS.over.120     : int 0 0 0 0 0 1 0 0 0 ...
## $ EKG.results       : int 2 2 0 0 2 0 2 2 2 ...
## $ Max.HR            : int 109 160 141 105 121 140 142 142 170 154 ...
## $ Exercise.angina    : int 0 0 0 1 1 0 1 1 0 0 ...
## $ ST.depression      : num 2.4 1.6 0.3 0.2 0.2 0.4 0.6 1.2 1.2 4 ...
## $ Slope.of.ST         : int 2 2 1 2 1 1 2 2 2 ...
## $ Number.of.vessels.fluro: int 3 0 0 1 1 0 1 1 2 3 ...
## $ Thallium             : int 3 7 7 7 3 7 6 7 7 7 ...
## $ Heart.Disease        : chr "Presence" "Absence" "Presence" "Absence" ...

# preprocessing and standardization
data$Heart.Disease <- ifelse(data$Heart.Disease == "Presence", 1, 0)

# standardize
df <- data %>%
  mutate(across(-Heart.Disease, scale))

head(df)

##          Age      Sex Chest.pain.type      BP Cholesterol FBS.over.120
## 1 1.7089201 0.6882217 0.8693133 -0.07527007 1.3996132 -0.4162558
## 2 1.3795779 -1.4476387 -0.1832185 -0.91506006 6.0817107 -0.4162558
## 3 0.2817705 0.6882217 -1.2357503 -0.41118607 0.2194151 -0.4162558
## 4 1.0502357 0.6882217 0.8693133 -0.18724207 0.2581101 -0.4162558
## 5 2.1480430 -1.4476387 -1.2357503 -0.63513007 0.3741952 -0.4162558
## 6 1.1600164 0.6882217 0.8693133 -0.63513007 -1.4057758 -0.4162558
##   EKG.results      Max.HR Exercise.angina ST.depression Slope.of.ST
## 1 0.9798441 -1.7559473 -0.6999225 1.1788233 0.6751655
## 2 0.9798441 0.4455818 -0.6999225 0.4802613 0.6751655
## 3 -1.0243824 -0.3745957 -0.6999225 -0.6549018 -0.9524656
## 4 -1.0243824 -1.9286162 1.4234380 -0.7422221 0.6751655
## 5 0.9798441 -1.2379404 1.4234380 -0.7422221 -0.9524656
## 6 -1.0243824 -0.4177629 -0.6999225 -0.5675816 -0.9524656
##   Number.of.vessels.fluro      Thallium Heart.Disease
## 1 2.4680989 -0.8740826 1
## 2 -0.7102161 1.1870729 0
## 3 -0.7102161 1.1870729 1
## 4 0.3492223 1.1870729 0
## 5 0.3492223 -0.8740826 0
## 6 -0.7102161 1.1870729 0

# data partitioning

n <- nrow(df)
idx <- sample(1:n, size = .7 * n)
train <- df[idx,]
valid <- df[-idx,]
dim(train)

## [1] 189 14

```

```

dim(valid)

## [1] 81 14

# logit model
logit_model <- glm(Heart.Disease ~ ., data = train, family = "binomial")
logit_predict <- predict(logit_model, newdata = valid, type = "response")
logit_class <- ifelse(logit_predict > .5, 1, 0)
table(Predicted = logit_class, Actual = valid$Heart.Disease)

##           Actual
## Predicted  0   1
##          0 41  9
##          1  6 25

mean(logit_class == valid$Heart.Disease)

## [1] 0.8148148

# logit performance
TP <- 21
TN <- 41
FP <- 10
FN <- 9

accuracy <- (TP + TN) / (TP + TN + FP + FN)
precision <- TP / (TP + FP)
recall <- TP / (TP + FN)
specificity <- TN / (TN + FP)

accuracy

## [1] 0.7654321

precision

## [1] 0.6774194

recall

## [1] 0.7

specificity

## [1] 0.8039216

# install.packages("xgboost")
library(xgboost)

## Warning: package 'xgboost' was built under R version 4.4.3

```

```

# Assume train and valid have Heart.Disease as 0/1
y_train <- train$Heart.Disease
y_valid <- valid$Heart.Disease

X_train <- model.matrix(Heart.Disease ~ . - 1, data = train)
X_valid <- model.matrix(Heart.Disease ~ . - 1, data = valid)

dtrain <- xgb.DMatrix(data = X_train, label = y_train)
dvalid <- xgb.DMatrix(data = X_valid, label = y_valid)

params <- list(
  objective = "binary:logistic",
  eval_metric = "logloss",
  eta = 0.05,
  max_depth = 4,
  subsample = 0.8,
  colsample_bytree = 0.8
)

watch <- list(train = dtrain, valid = dvalid)

xgb_fit <- xgb.train(
  params = params,
  data = dtrain,
  nrounds = 500,
  watchlist = watch,
  early_stopping_rounds = 25,
  verbose = 0
)

## Warning in throw_err_or_depr_msg("Parameter '", match_old, "' has been renamed
## to '", : Parameter 'watchlist' has been renamed to 'evals'. This warning will
## become an error in a future version.

p_valid <- predict(xgb_fit, dvalid)          # probabilities
pred_valid <- ifelse(p_valid > 0.5, 1, 0)    # class

table(Predicted = pred_valid, Actual = y_valid)

##           Actual
## Predicted   0   1
##             0 40   9
##             1   7 25

mean(pred_valid == y_valid)

## [1] 0.8024691

# random forest model
library(randomForest)

## Warning: package 'randomForest' was built under R version 4.4.3

```

```

## randomForest 4.7-1.2

## Type rfNews() to see new features/changes/bug fixes.

##
## Attaching package: 'randomForest'

## The following object is masked from 'package:dplyr':
##       combine

## The following object is masked from 'package:ggplot2':
##       margin

rf_model <- randomForest(as.factor(Heart.Disease) ~ ., data = train, ntree = 1000, mtry = 3, importance = TRUE)
rf_predict <- predict(rf_model, newdata = valid)

table(Predicted = rf_predict, Actual = valid$Heart.Disease)

##          Actual
## Predicted 0 1
##           0 44 9
##           1  3 25

mean(rf_predict == valid$Heart.Disease)

## [1] 0.8518519

```

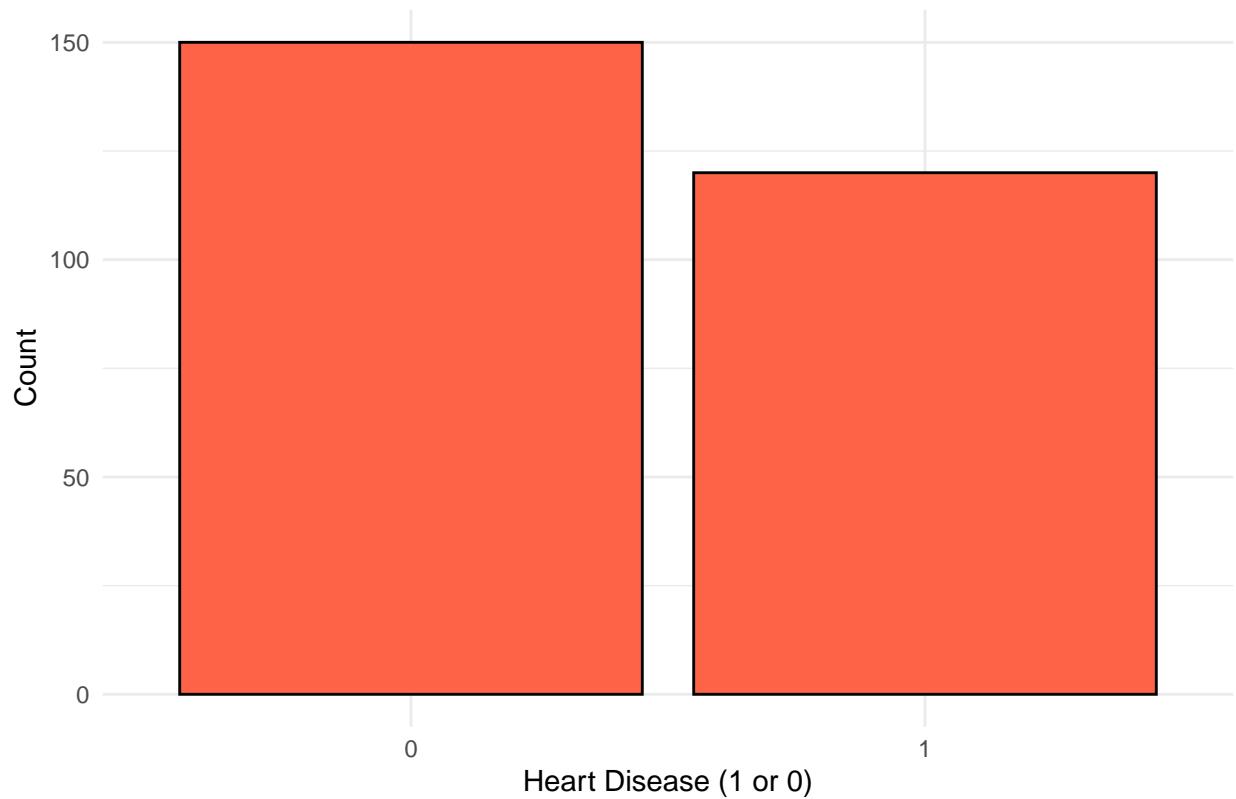
## Visualizations

```

# heart disease boxplot
ggplot(data, aes(x = factor(Heart.Disease))) +
  geom_bar(fill = "tomato", col = "black") +
  labs(
    title = "Heart Disease Distribution",
    x = "Heart Disease (1 or 0)",
    y = "Count"
  ) +
  theme_minimal() +
  theme(legend.position = "none")

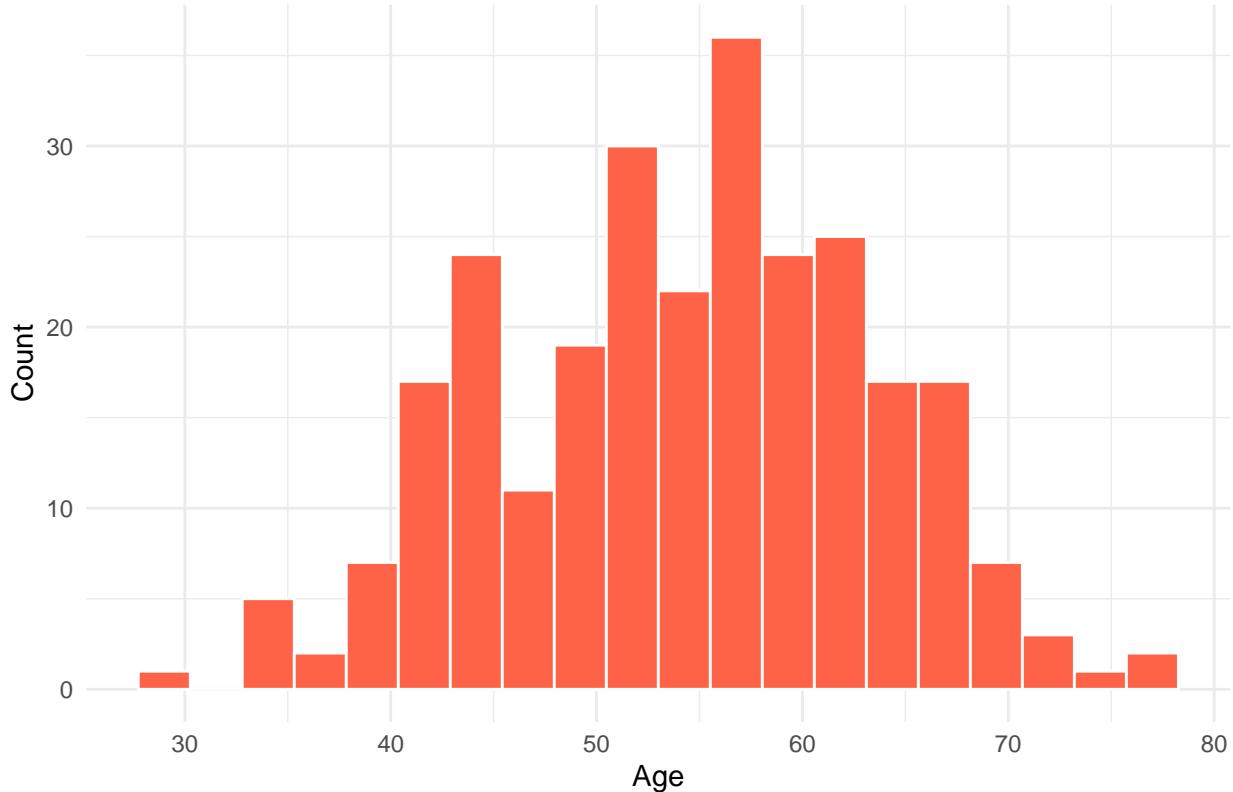
```

## Heart Disease Distribution



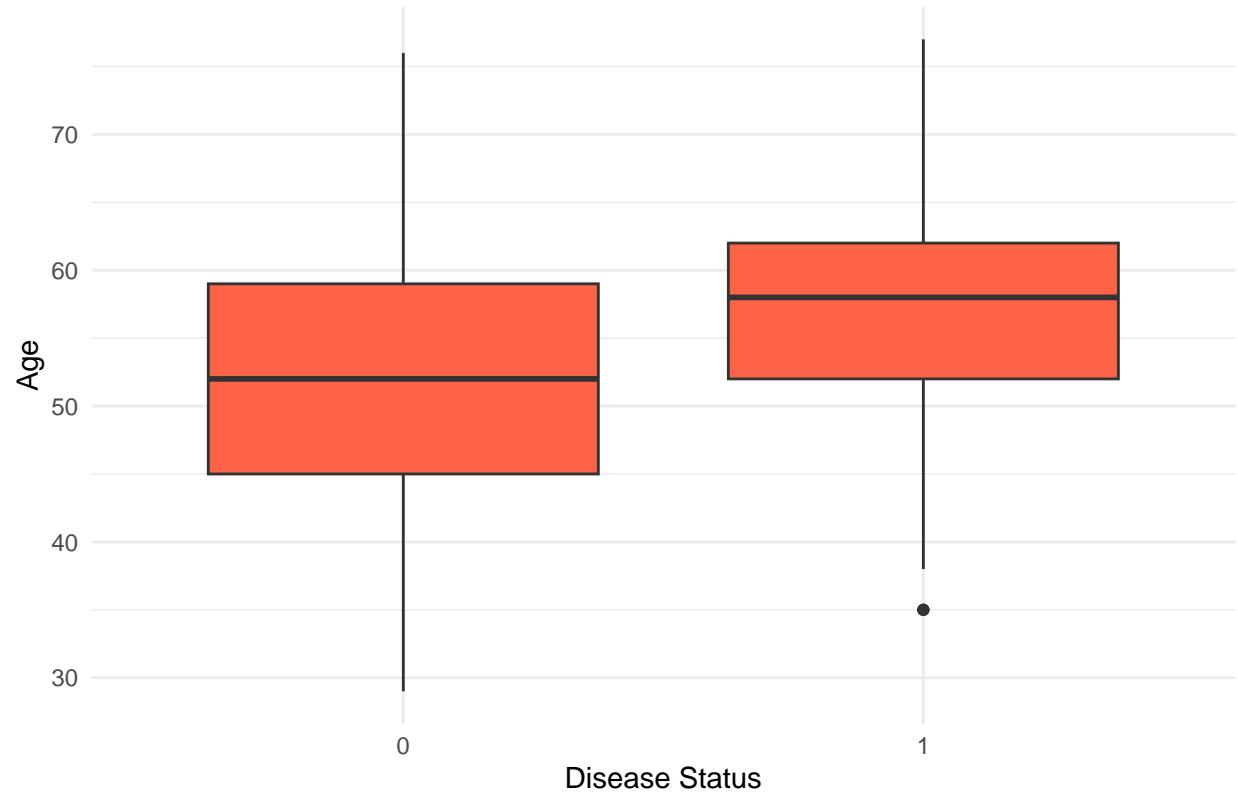
```
# age histogram
ggplot(data, aes(x = Age)) +
  geom_histogram(
    bins = 20,
    fill = "tomato",
    col = "white"
  ) +
  labs(
    title = "Age Histogram",
    x = "Age",
    y = "Count"
  ) +
  theme_minimal()
```

## Age Histogram



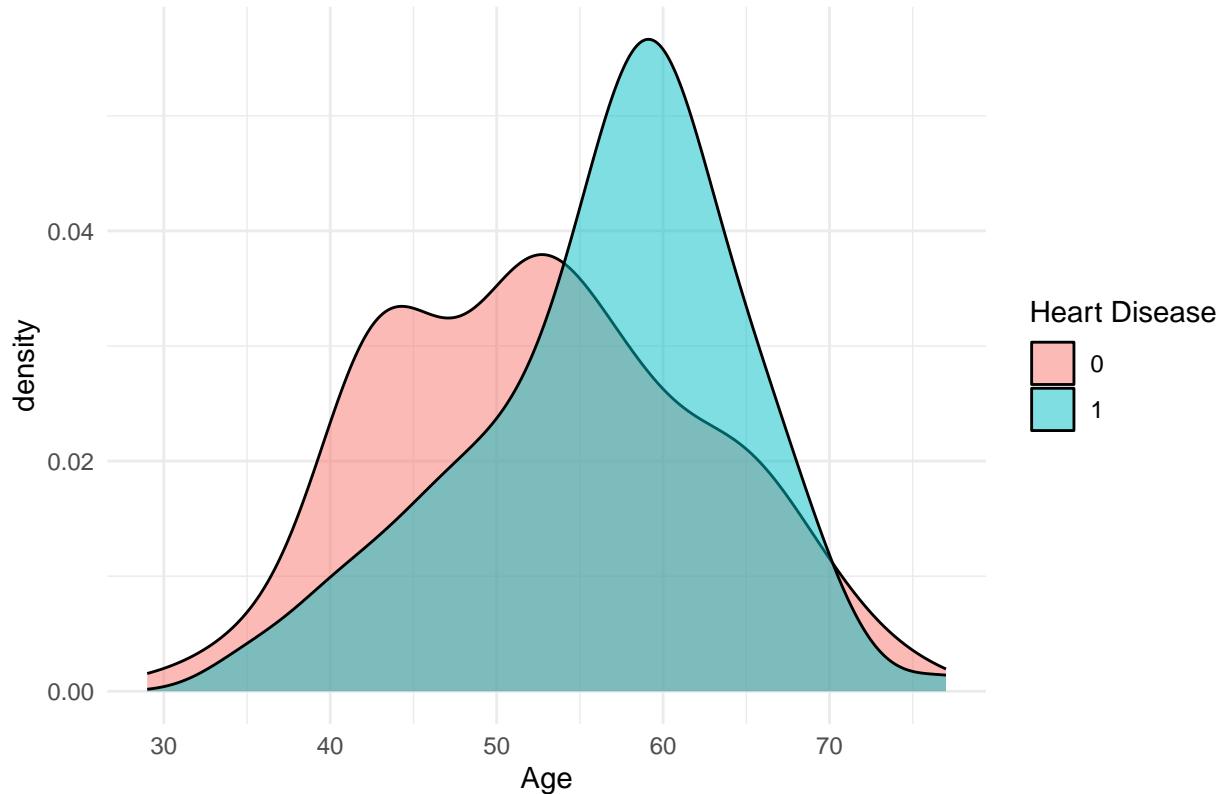
```
# boxplot of heart disease by age
ggplot(data, aes(x = factor(Heart.Disease), y = Age)) +
  geom_boxplot(
    fill = "tomato"
  ) +
  labs(
    title = "Heart Disease by Age",
    x = "Disease Status",
    y = "Age"
  ) +
  theme_minimal()
```

## Heart Disease by Age



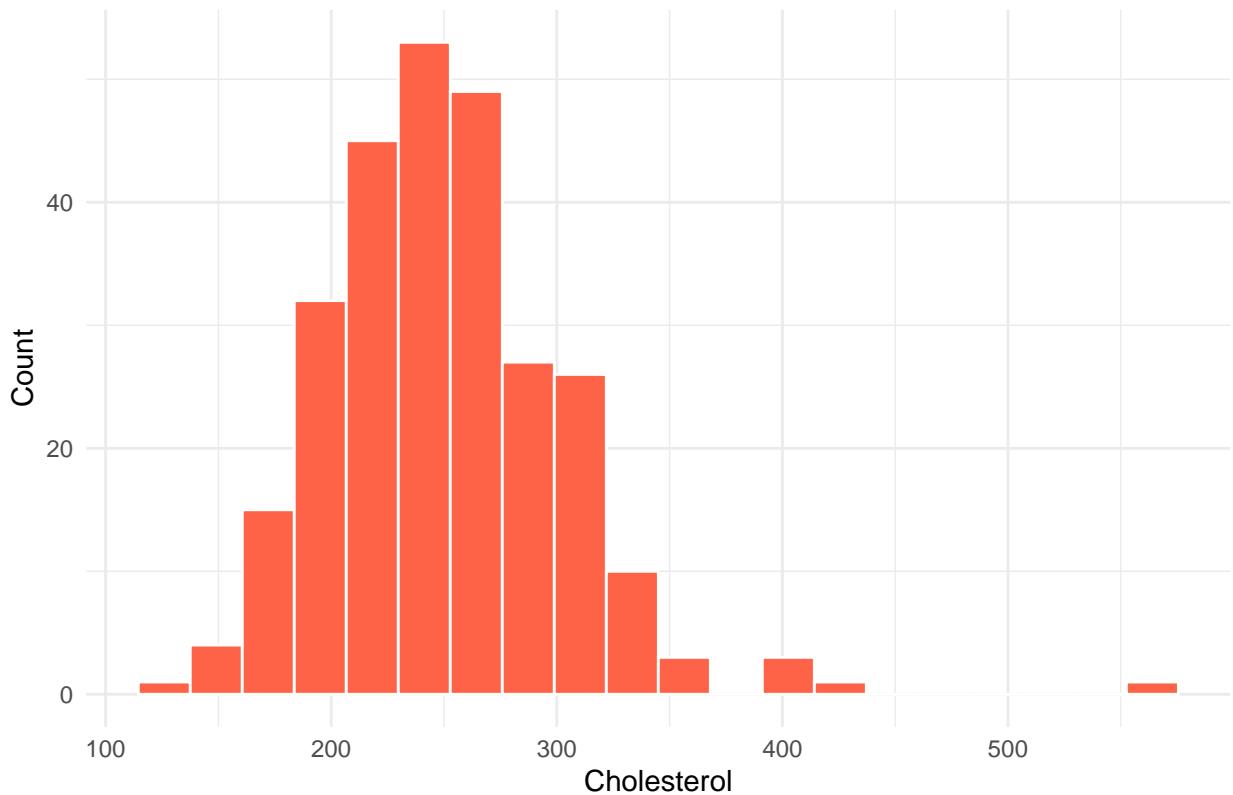
```
ggplot(data, aes(x = Age, fill = factor(Heart.Disease))) +  
  geom_density(  
    alpha = 0.5  
  ) +  
  labs(  
    title = "Density Plot",  
    x = "Age",  
    fill = "Heart Disease"  
  ) +  
  theme_minimal()
```

## Density Plot



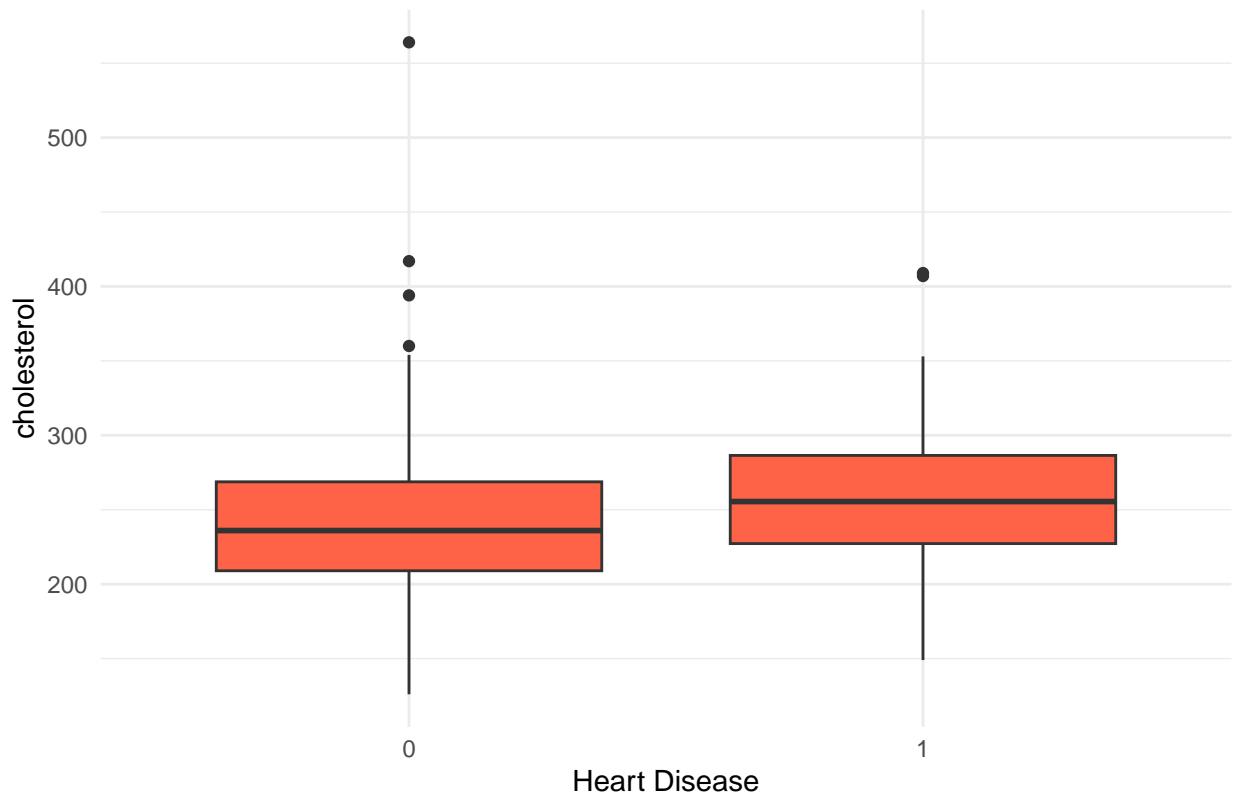
```
# cholesterol histogram
ggplot(data, aes(x = Cholesterol)) +
  geom_histogram(
    bins = 20,
    fill = "tomato",
    col = "white"
  ) +
  labs(
    title = "Histogram of Cholesterol",
    x = "Cholesterol",
    y = "Count"
  ) +
  theme_minimal()
```

### Histogram of Cholesterol

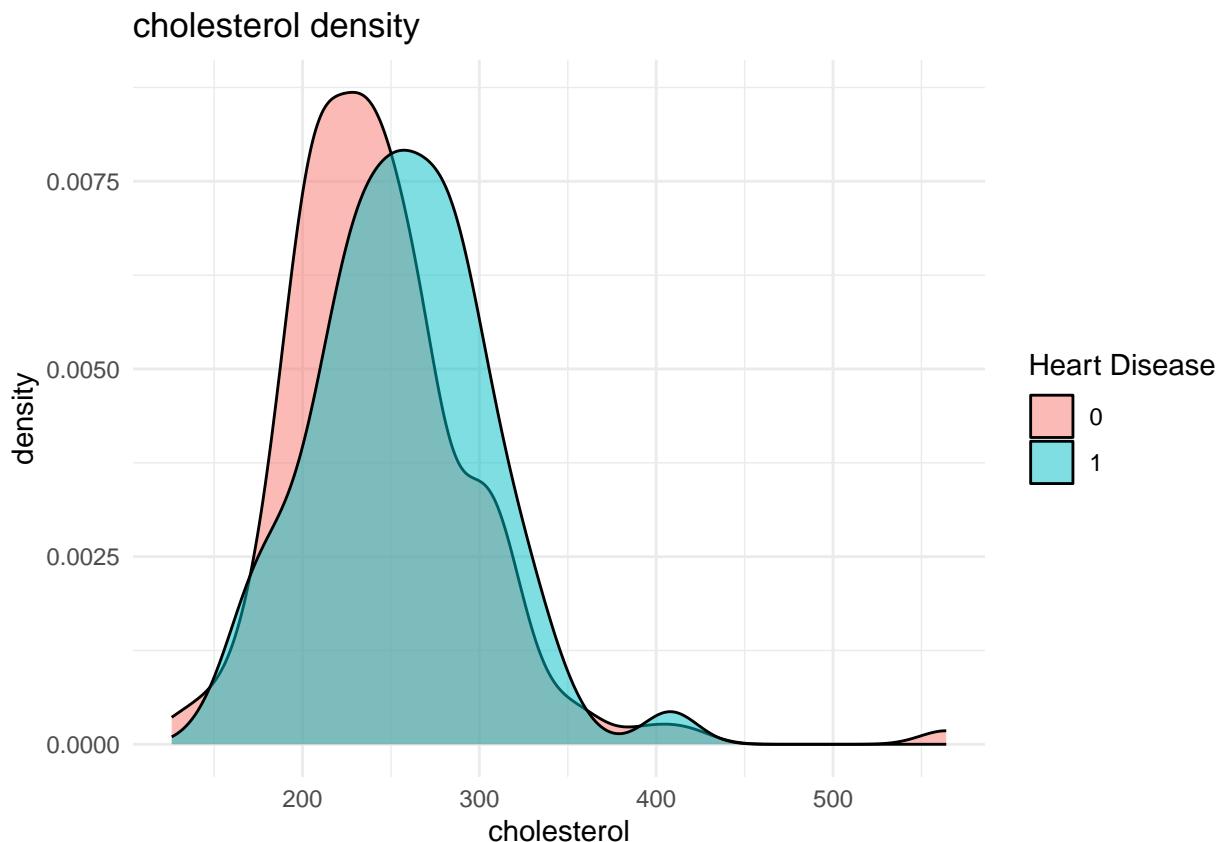


```
ggplot(data, aes(x = factor(Heart.Disease), y = Cholesterol)) +  
  geom_boxplot(  
    fill = "tomato"  
) +  
  labs(  
    title = "Cholesterol Boxplot",  
    x = "Heart Disease",  
    y = "cholesterol"  
) +  
  theme_minimal()
```

### Cholesterol Boxplot

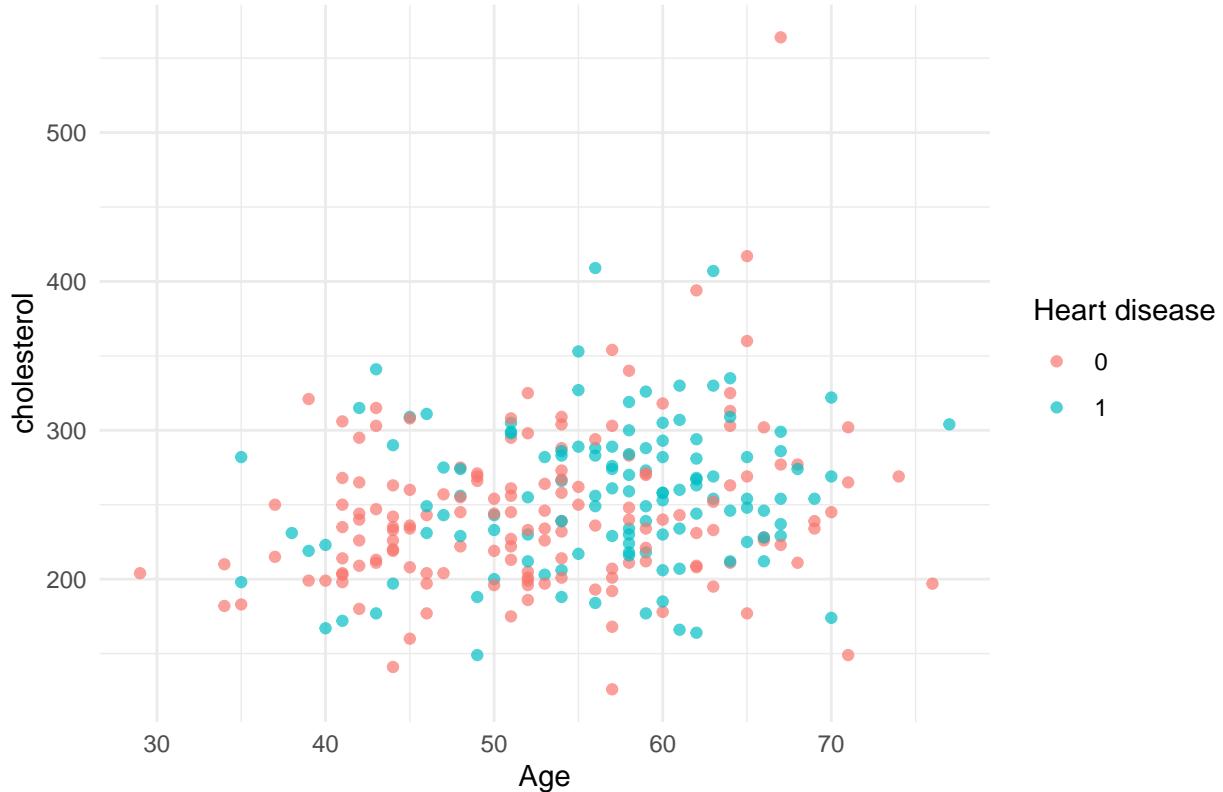


```
# cholesterol density plot
ggplot(data, aes(x = Cholesterol, fill = factor(Heart.Disease))) +
  geom_density(
    alpha = .5
  ) +
  labs(
    title = "cholesterol density",
    x = "cholesterol",
    y = "density",
    fill = "Heart Disease"
  ) +
  theme_minimal()
```



```
# scatter plot
ggplot(data, aes(x = Age, y = Cholesterol, color = factor(Heart.Disease))) +
  geom_point(alpha = 0.7) +
  labs(
    title = "Scatter",
    x = "Age",
    y = "cholesterol",
    color = "Heart disease"
  ) +
  theme_minimal()
```

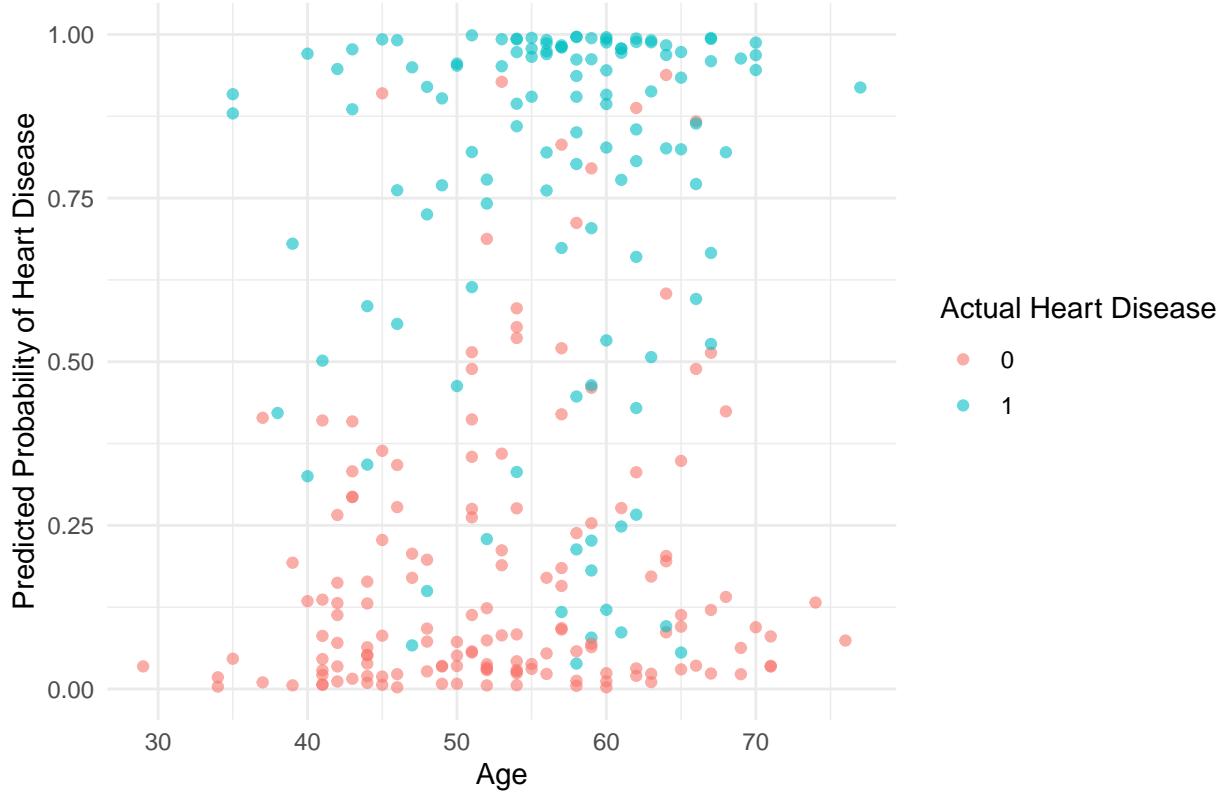
## Scatter



```
# logistic variable  
data$logit_prob <- predict(logit_model, newdata = df, type = "response")
```

```
# logistic regression plot  
ggplot(data, aes(x = Age, y = logit_prob, color = factor(Heart.Disease))) +  
  geom_point(alpha = 0.6) +  
  labs(  
    title = "Logistic Regression Predicted Risk vs Age",  
    x = "Age",  
    y = "Predicted Probability of Heart Disease",  
    color = "Actual Heart Disease"  
) +  
  theme_minimal()
```

## Logistic Regression Predicted Risk vs Age



```

library(dplyr)
multi_var <- data %>%
  dplyr::select(-Heart.Disease, -Sex)

multi_var <- scale(multi_var)
multi_var <- as.data.frame(multi_var)
head(multi_var)

##          Age Chest.pain.type          BP Cholesterol FBS.over.120 EKG.results
## 1 1.7089201      0.8693133 -0.07527007   1.3996132   -0.4162558    0.9798441
## 2 1.3795779     -0.1832185 -0.91506006   6.0817107   -0.4162558    0.9798441
## 3 0.2817705     -1.2357503 -0.41118607   0.2194151   -0.4162558   -1.0243824
## 4 1.0502357      0.8693133 -0.18724207   0.2581101   -0.4162558   -1.0243824
## 5 2.1480430     -1.2357503 -0.63513007   0.3741952   -0.4162558    0.9798441
## 6 1.1600164      0.8693133 -0.63513007  -1.4057758   -0.4162558   -1.0243824
##          Max.HR Exercise.angina ST.depression Slope.of.ST Number.of.vessels.fluro
## 1 -1.7559473     -0.6999225     1.1788233    0.6751655           2.4680989
## 2  0.4455818     -0.6999225     0.4802613    0.6751655          -0.7102161
## 3 -0.3745957     -0.6999225    -0.6549018   -0.9524656          -0.7102161
## 4 -1.9286162      1.4234380    -0.7422221    0.6751655           0.3492223
## 5 -1.2379404      1.4234380    -0.7422221   -0.9524656           0.3492223
## 6 -0.4177629     -0.6999225    -0.5675816   -0.9524656          -0.7102161
##          Thallium logit_prob
## 1 -0.8740826  1.4334779
## 2  1.1870729  0.1935783

```

```

## 3 1.1870729 -0.8428601
## 4 1.1870729  1.3041743
## 5 -0.8740826 -0.8047743
## 6 1.1870729 -0.2386462

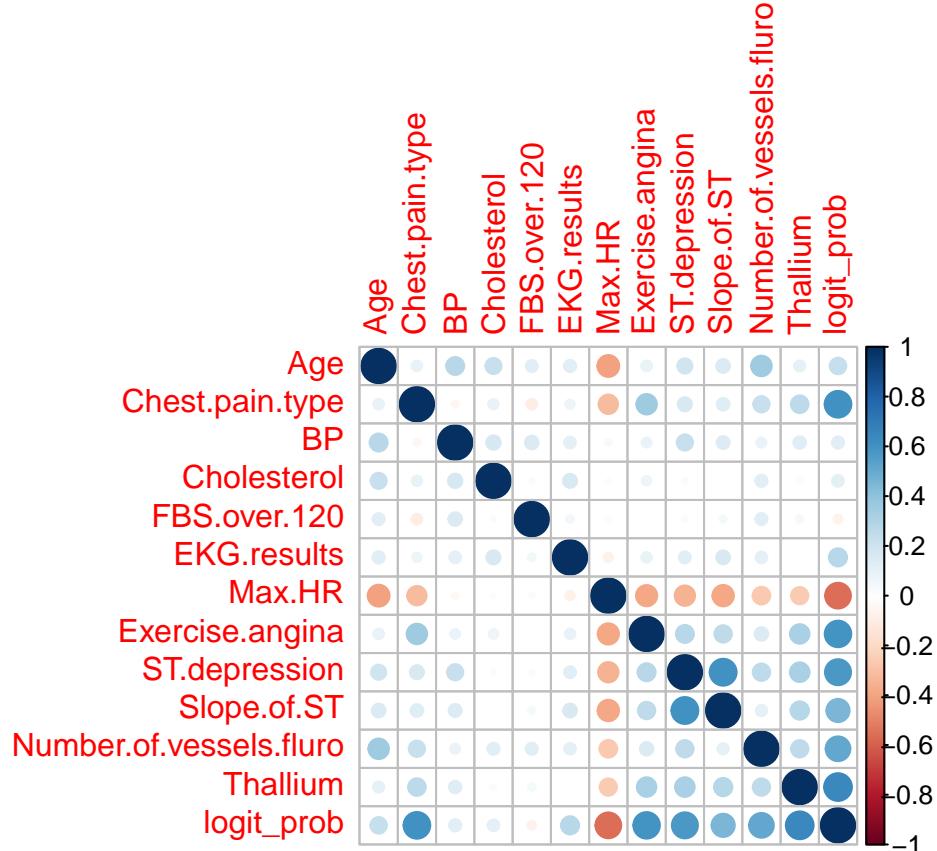
cor <- cor(multi_var)

round(cor,2)

##                                     Age Chest.pain.type      BP Cholesterol FBS.over.120
## Age                               1.00        0.10     0.27       0.22       0.12
## Chest.pain.type                  0.10        1.00    -0.04       0.09      -0.10
## BP                                0.27       -0.04     1.00       0.17       0.16
## Cholesterol                      0.22        0.09     0.17       1.00       0.03
## FBS.over.120                     0.12       -0.10     0.16       0.03       1.00
## EKG.results                      0.13        0.07     0.12       0.17       0.05
## Max.HR                            -0.40      -0.32    -0.04      -0.02       0.02
## Exercise.angina                 0.10        0.35     0.08       0.08       0.00
## ST.depression                     0.19        0.17     0.22       0.03      -0.03
## Slope.of.ST                      0.16        0.14     0.14      -0.01       0.04
## Number.of.vessels.fluro         0.36        0.23     0.09       0.13       0.12
## Thallium                          0.11        0.26     0.13       0.03       0.05
## logit_prob                        0.23        0.60     0.13       0.12      -0.07
##                                     EKG.results Max.HR Exercise.angina ST.depression
## Age                               0.13     -0.40       0.10       0.19
## Chest.pain.type                  0.07     -0.32       0.35       0.17
## BP                                0.12     -0.04       0.08       0.22
## Cholesterol                      0.17     -0.02       0.08       0.03
## FBS.over.120                     0.05     0.02        0.00      -0.03
## EKG.results                      1.00    -0.07       0.10       0.12
## Max.HR                            -0.07     1.00      -0.38      -0.35
## Exercise.angina                 0.10     -0.38       1.00       0.27
## ST.depression                     0.12     -0.35       0.27       1.00
## Slope.of.ST                      0.16     -0.39       0.26       0.61
## Number.of.vessels.fluro         0.11     -0.27       0.15       0.26
## Thallium                          0.01     -0.25       0.32       0.32
## logit_prob                        0.28     -0.56       0.59       0.57
##                                     Slope.of.ST Number.of.vessels.fluro Thallium logit_prob
## Age                               0.16        0.36     0.11       0.23
## Chest.pain.type                  0.14        0.23     0.26       0.60
## BP                                0.14        0.09     0.13       0.13
## Cholesterol                      -0.01       0.13     0.03       0.12
## FBS.over.120                     0.04        0.12     0.05      -0.07
## EKG.results                      0.16        0.11     0.01       0.28
## Max.HR                            -0.39      -0.27    -0.25      -0.56
## Exercise.angina                 0.26        0.15     0.32       0.59
## ST.depression                     0.61        0.26     0.32       0.57
## Slope.of.ST                      1.00        0.11     0.28       0.45
## Number.of.vessels.fluro         0.11        1.00     0.26       0.52
## Thallium                          0.28        0.26     1.00       0.65
## logit_prob                        0.45        0.52     0.65       1.00

```

```
library(corrplot)
corrplot(cor)
```



```
table(data$Heart.Disease)
```

```
##
##    0    1
## 150 120
```

```
pca <- prcomp(multi_var, .scale = TRUE)
```

```
## Warning: In prcomp.default(multi_var, .scale = TRUE) :
##   extra argument '.scale' will be disregarded
```

```
summary(pca)
```

```
## Importance of components:
##                 PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation     1.955  1.2312  1.10520  1.02386  0.98524  0.94899  0.89983
## Proportion of Variance 0.294  0.1166  0.09396  0.08064  0.07467  0.06928  0.06228
## Cumulative Proportion  0.294  0.4106  0.50457  0.58521  0.65988  0.72915  0.79144
##                           PC8      PC9      PC10     PC11     PC12     PC13
## Standard deviation     0.8440  0.7783  0.75807  0.63644  0.58758  0.26123
## Proportion of Variance 0.0548  0.0466  0.04421  0.03116  0.02656  0.00525
## Cumulative Proportion  0.8462  0.8928  0.93703  0.96819  0.99475  1.00000
```

```

print(pca$rotation[,1:2])

##                                PC1          PC2
## Age                  -0.22266906  0.451621477
## Chest.pain.type    -0.27712033 -0.285704464
## BP                  -0.13063745  0.478499061
## Cholesterol        -0.09138260  0.372059767
## FBS.over.120       -0.01533569  0.404902194
## EKG.results         -0.13923185  0.243921666
## Max.HR              0.34404285  0.062813586
## Exercise.angina    -0.31178970 -0.201893459
## ST.depression       -0.34526262 -0.001049262
## Slope.of.ST         -0.30912082 -0.020069828
## Number.of.vessels.fluro -0.27246869  0.186413384
## Thallium             -0.31298500 -0.139276852
## logit_prob           -0.47544309 -0.155946169

# kmeans cluster
km <- kmeans(multi_var, centers = 2, nstart = 25)
table(km$cluster)

## 
##   1   2
## 106 164

# pca scores
pc_scores <- as.data.frame(pca$x)
head(pc_scores)

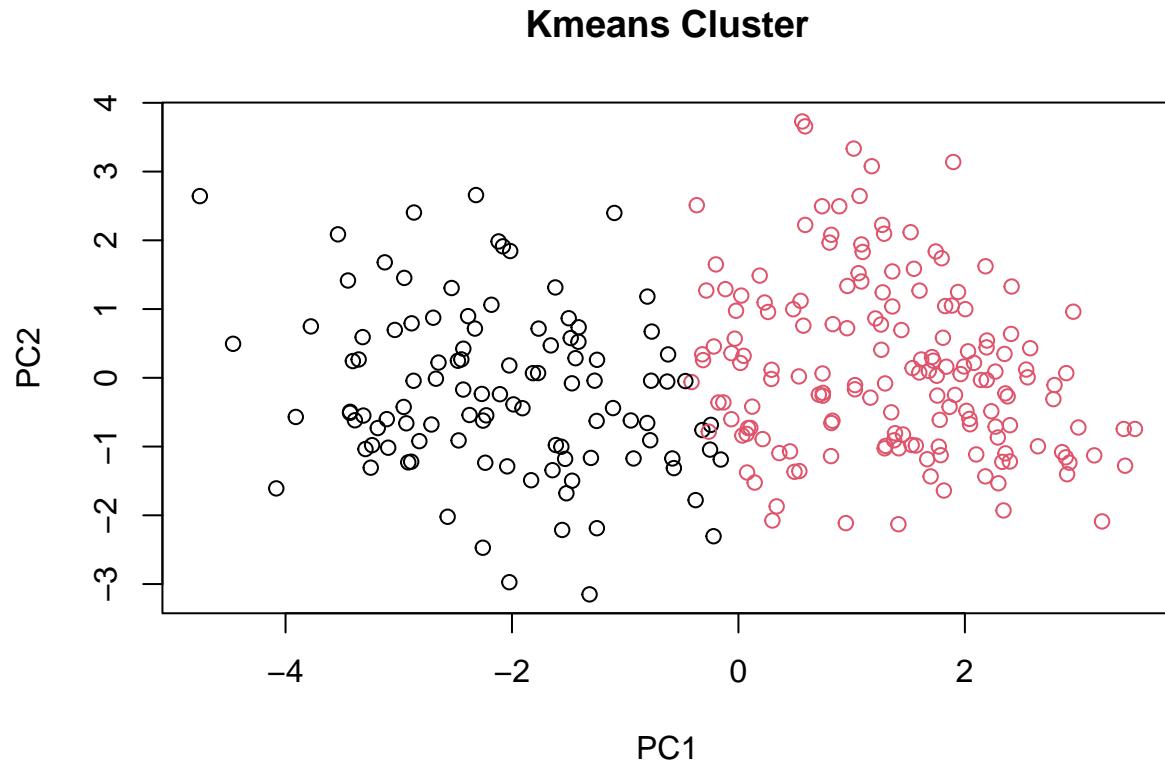
##          PC1         PC2         PC3         PC4         PC5         PC6
## 1 -2.95158308 1.45310956 -1.32439727  0.24811734 2.21800588  0.7973748
## 2 -1.09573504 2.39809077 -2.05371634  2.91448692 0.01041309 -1.6438160
## 3  1.29498512 -0.08191726 -0.03470948 -0.84284742 0.30437297 -1.0900887
## 4 -2.471414003 -0.91094329 -0.92604601 -1.06908881 0.63953279 -0.8828207
## 5  0.02490136 1.19603083 -1.29124080  0.02837445 1.87348889  0.5838256
## 6  0.36157154 -1.09556173 -0.43912665 -1.59544396 0.66991104 -0.4626737
##          PC7         PC8         PC9         PC10        PC11        PC12        PC13
## 1 -1.13304147 0.5939057  0.7905200  0.3466217 -0.6594097  0.1278234 -0.2758876
## 2 -0.61549719 4.1059172 -1.2020589 -1.6565643  0.5849665  0.2921236  0.0299745
## 3 -0.19220048 0.2867704 -1.6002607 -0.9825213 -0.2742787  0.4194717  0.2525976
## 4  1.11894013 0.1008178 -0.8172562 -0.4823359 -0.2682254 -0.8630407 -0.2476966
## 5  1.33487139 -0.4835677 -2.0079685  0.8752423  0.5628900  0.1881792  0.5023395
## 6  0.05320517 -0.9311517 -0.2155345 -1.7623136  0.7438717  0.4735880  0.1897508

# selecting first two principle components
pc_12 <- pc_scores %>%
  dplyr::select(PC1, PC2)

# adding to data
multi_var <- multi_var %>%
  bind_cols(pc_12)

```

```
# kmeans clustering
plot(pca$x[,1], pca$x[,2], col = km$cluster, main = "Kmeans Cluster", xlab = "PC1", ylab = "PC2")
```



```
table(data$Sex)
```

```
##  
##    0    1  
##   87 183
```

```
gender_summary <- data %>%  
  group_by(Sex) %>%  
  summarise(  
    proportion = round(mean(Heart.Disease == 1), 2)  
  )  
gender_summary
```

```
## # A tibble: 2 x 2  
##       Sex   proportion  
##   <int>      <dbl>  
## 1     0        0.23  
## 2     1        0.55
```

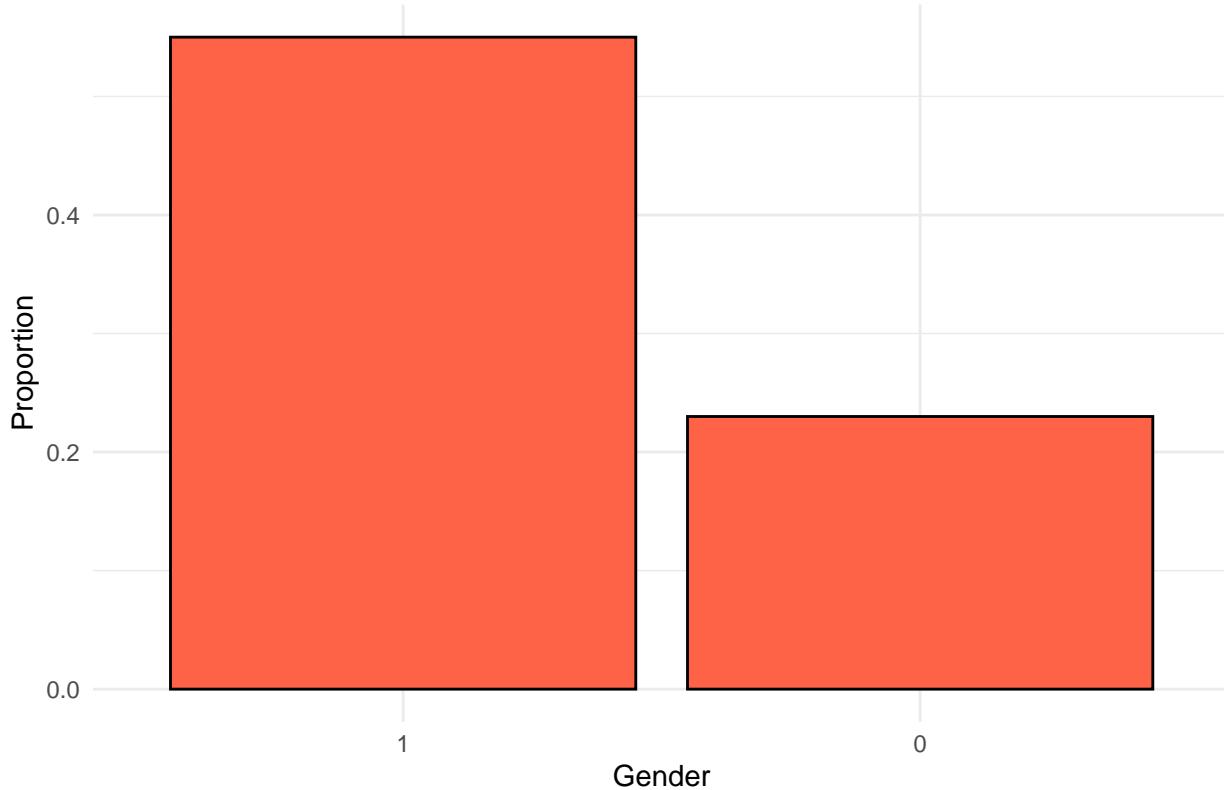
```
kable(gender_summary, caption = "Heart Disease by Gender") %>%
  kable_styling()
```

Table 1: Heart Disease by Gender

Sex	proportion
0	0.23
1	0.55

```
ggplot(gender_summary, aes(x = reorder(Sex, -proportion), y = proportion, fill = sex)) +
  geom_col(fill = "tomato", col = "black") +
  labs(
    title = "Proportion of Heart Disease for Each Gender",
    x = "Gender",
    y = "Proportion"
  ) +
  theme_minimal() +
  theme(legend.position = "none")
```

Proportion of Heart Disease for Each Gender



```
data <- data %>%
  mutate(HR_Cat = factor(ntile(Max.HR, 3), labels = c("Low", "Med", "High")))

ca_table <- table(data$Heart.Disease, data$HR_Cat)
ca_table
```

```

##          Low Med High
## 0    26   53   71
## 1    64   37   19

library(FactoMineR)

## Warning: package 'FactoMineR' was built under R version 4.4.3

ca <- CA(ca_table, graph = FALSE)
mosaicplot(ca_table, main="Heart Disease vs HR Category", color=TRUE)

```

