

Survival of patients with heart failure

Raj Sharma

2023-11-11

Source of data: UCI machine learning repository:<https://archive.ics.uci.edu/dataset/519/heart+failure+clinical+records>

Upload date of data: 2nd April 2020

Owner of the data: The original dataset version was collected by Tanvir Ahmad, Assia Munir, Sajjad Haider Bhatti, Muhammad Aftab, and Muhammad Ali Raza in 2015. Current dataset has slight difference from original dataset in terms of names of columns which was done by D. Chicco and Giuseppe Jurman in their article published on BMC Medical Informatics and Decision Making. Link:<https://bmcmmedinformdecismak.biomedcentral.com/articles/10.1186/s12911-020-1023-5#article-info>

Objective of data collection: Dataset was collected for analyzing the survival pattern of heart failure patients who were admitted to Institute of Cardiology and Allied hospital Faisalabad-Pakistan during April-December (2015). And also rank the features which corresponds to the most important risk factors.

Analysis of data by other researchers: D.Chicco and Giuseppe Jurman found in their analysis that serum creatinine and ejection fraction plays a very important role in understanding if a patient will survive after heart failure or not. This discovery has the potential to impact on clinical practice, becoming a new supporting tool for physicians when predicting if a heart failure patient will survive or not.

Description of data set: This dataset was collected during April–December 2015. It contains the medical records of 299 heart failure patients collected at the Faisalabad Institute of Cardiology and at the Allied Hospital in Faisalabad (Punjab, Pakistan). The patients consisted of 105 women and 194 men, and their ages range between 40 and 95 years old. All 299 patients had left ventricular systolic dysfunction and had previous heart failures that put them in classes III or IV of New York Heart Association (NYHA) classification of the stages of heart failure. The dataset contains 13 features, which report clinical, body, and lifestyle information.

Description of columns:

```
column_des<-read.csv("C:/Users/rajsh/OneDrive/Documents/Rproject/Book1.csv")
print(column_des)
```

```
##           Feature
## 1           Age
## 2         Anaemia
## 3   High blood pressure
## 4 Creatinine phosphokinase(CPK)
## 5           Diabetes
## 6     Ejection fraction
## 7             Sex
## 8         Platelets
## 9     Serum creatinine
## 10    Serum sodium
```

```

## 11          Smoking
## 12          Time
## 13      (target) death event
##
##          Explanation          Measurement
## 1          Age of the patient          Years
## 2          Decrease of red blood cells or hemoglobin          Boolean
## 3          If a patient has hypertension          Boolean
## 4          Level of the CPK enzyme in the blood          mcg/L
## 5          If the patient has diabetes          Boolean
## 6  Percentage of blood leaving the heart at each contraction          Percentage
## 7          Woman or man          Binary
## 8          Platelets in the blood          kiloplatelets/mL
## 9          Level of creatinine in the blood          mg/dL
## 10         Level of sodium in the blood          mEq/L
## 11         If the patient smokes          Boolean
## 12         Follow-up period          Days
## 13         If the patient died during the follow-up period          Boolean
##
##          Range
## 1          [40,..., 95]
## 2          0, 1
## 3          0, 1
## 4          [23,..., 7861]
## 5          0, 1
## 6          [14,..., 80]
## 7          0, 1
## 8          [25.01,..., 850.00]
## 9          [0.50,..., 9.40]
## 10         [114,..., 148]
## 11         0, 1
## 12         [4,...,285]
## 13         0, 1

```

The hospital physician considered a patient having **anaemia** if haematocrit levels were lower than 36%. The **creatinine phosphokinase (CPK)** states the level of the CPK enzyme in blood. When a muscle tissue gets damaged, CPK flows into the blood. Therefore, high levels of CPK in the blood of a patient might indicate a heart failure or injury. It is measured in microgram per litre (mcg/L). The **ejection fraction** states the percentage of how much blood the left ventricle pumps out with each contraction. The **serum creatinine** is a waste product generated by creatine, when a muscle breaks down. Doctors focus on serum creatinine in blood to check kidney function. If a patient has high levels of serum creatinine, it may indicate renal dysfunction. A normal range for serum creatinine is approximately 0.6 to 1.3 milligrams per deciliter (mg/dL) for adult males and 0.5 to 1.2 mg/dL for adult females. The **serum sodium test** is a routine blood exam that indicates if a patient has normal levels of sodium in the blood. An abnormally low level of sodium (<125 mEq/L) in the blood might be caused by heart failure. mEq/L represents milliequivalents per liter. The **death event** feature, that we use as the target in our binary classification study, states if the patient died or survived before the end of the follow-up period, that was 130 days on average.

Purpose of my project: Carrying out EDA to look for important factors for predicting survival of patients from heart failure.

```

#loading necessary libraries
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(moments)
```

```
#Loading dataset
```

```
df<-read.csv("C:/Users/rajsh/OneDrive/Documents/Rproject/heart_failure_clinical_records_dataset.csv")
head(df)
```

```
##   age anaemia creatinine_phosphokinase diabetes ejection_fraction
## 1  75      0                582      0             20
## 2  55      0                7861     0             38
## 3  65      0                146      0             20
## 4  50      1                111      0             20
## 5  65      1                160      1             20
## 6  90      1                 47      0             40
##   high_blood_pressure platelets serum_creatinine serum_sodium sex smoking time
## 1                   1   265.000             1.9         130    1      0      4
## 2                   0   263.358             1.1         136    1      0      6
## 3                   0   162.000             1.3         129    1      1      7
## 4                   0   210.000             1.9         137    1      0      7
## 5                   0   327.000             2.7         116    0      0      8
## 6                   1   204.000             2.1         132    1      1      8
##   DEATH_EVENT
## 1           1
## 2           1
## 3           1
## 4           1
## 5           1
## 6           1
```

```
#Structure of data
```

```
cat("Shape of the dataset is:",dim(df),"\n")
```

```
## Shape of the dataset is: 299 13
```

```
str(df)
```

```
## 'data.frame':   299 obs. of  13 variables:
##  $ age          : num  75 55 65 50 65 90 75 60 65 80 ...
##  $ anaemia      : int   0 0 0 1 1 1 1 0 1 ...
##  $ creatinine_phosphokinase: int  582 7861 146 111 160 47 246 315 157 123 ...
##  $ diabetes     : int   0 0 0 0 1 0 0 1 0 0 ...
```

```
## $ ejection_fraction      : int  20 38 20 20 20 40 15 60 65 35 ...
## $ high_blood_pressure    : int   1 0 0 0 0 1 0 0 0 1 ...
## $ platelets              : num  265 263 162 210 327 ...
## $ serum_creatinine       : num   1.9 1.1 1.3 1.9 2.7 2.1 1.2 1.1 1.5 9.4 ...
## $ serum_sodium           : int  130 136 129 137 116 132 137 131 138 133 ...
## $ sex                    : int   1 1 1 1 0 1 1 1 0 1 ...
## $ smoking                : int   0 0 1 0 0 1 0 1 0 1 ...
## $ time                   : int   4 6 7 7 8 8 10 10 10 10 ...
## $ DEATH_EVENT            : int   1 1 1 1 1 1 1 1 1 1 ...
```

Since age is measured in years and platelets are count so they are integer columns, thus converting them into integer. By description of dataset we know that anaemia, high blood pressure, diabetes, sex and smoking are categorical columns which are encoded with 0 and 1.

```
df$age <- as.integer(df$age)
str(df)
```

```
## 'data.frame':    299 obs. of  13 variables:
## $ age              : int   75 55 65 50 65 90 75 60 65 80 ...
## $ anaemia          : int    0 0 0 1 1 1 1 1 0 1 ...
## $ creatinine_phosphokinase : int  582 7861 146 111 160 47 246 315 157 123 ...
## $ diabetes         : int    0 0 0 0 1 0 0 1 0 0 ...
## $ ejection_fraction : int   20 38 20 20 20 40 15 60 65 35 ...
## $ high_blood_pressure : int    1 0 0 0 0 1 0 0 0 1 ...
## $ platelets         : num  265 263 162 210 327 ...
## $ serum_creatinine   : num   1.9 1.1 1.3 1.9 2.7 2.1 1.2 1.1 1.5 9.4 ...
## $ serum_sodium       : int  130 136 129 137 116 132 137 131 138 133 ...
## $ sex               : int    1 1 1 1 0 1 1 1 0 1 ...
## $ smoking           : int    0 0 1 0 0 1 0 1 0 1 ...
## $ time              : int    4 6 7 7 8 8 10 10 10 10 ...
## $ DEATH_EVENT       : int    1 1 1 1 1 1 1 1 1 1 ...
```

```
#Getting basic descriptive statistic summary of categorical columns
#For Anaemia
```

```
Anaemia_df <- data.frame(
  Anaemia = c("Present", "Absent"),
  Full_sampleNumber = c(sum(df$anaemia), 299 - sum(df$anaemia)),
  Full_samplePercentage = c(sum(df$anaemia) * 100 / nrow(df), (299 - sum(df$anaemia)) * 100 / nrow(df)),
  Dead_patientsNumber = c(sum(df[df$DEATH_EVENT == 1, ]$anaemia), sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$anaemia)),
  Dead_patientsPercentage = c(sum(df[df$DEATH_EVENT == 1, ]$anaemia) * 100 / sum(df$DEATH_EVENT), (sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$anaemia)) * 100 / sum(df$DEATH_EVENT)),
  Survived_patientsNumber = c(sum(df[df$DEATH_EVENT == 0, ]$anaemia), ((299 - sum(df$DEATH_EVENT)) - sum(df[df$DEATH_EVENT == 1, ]$anaemia))),
  Survived_patientsPercentage = c(sum(df[df$DEATH_EVENT == 0, ]$anaemia) * 100 / (299 - sum(df$DEATH_EVENT)), ((299 - sum(df$DEATH_EVENT)) - sum(df[df$DEATH_EVENT == 1, ]$anaemia)) * 100 / (299 - sum(df$DEATH_EVENT)))
)
print(Anaemia_df)
```

```
##   Anaemia Full_sampleNumber Full_samplePercentage Dead_patientsNumber
## 1 Present              129              43.14381              46
## 2 Absent               170              56.85619              50
##   Dead_patientsPercentage Survived_patientsNumber Survived_patientsPercentage
## 1              47.91667              83              40.8867
## 2              52.08333              120              59.1133
```

#For High blood pressure

```
High_blood_pressure_df <- data.frame(
  High_blood_pressure = c("Present", "Absent"),
  Full_sampleNumber = c(sum(df$high_blood_pressure), 299 - sum(df$high_blood_pressure)),
  Full_samplePercentage = c(sum(df$high_blood_pressure) * 100 / nrow(df), (299 - sum(df$high_blood_pressure)) * 100 / nrow(df)),
  Dead_patientsNumber = c(sum(df[df$DEATH_EVENT == 1, ]$high_blood_pressure), sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$high_blood_pressure)),
  Dead_patientsPercentage = c(sum(df[df$DEATH_EVENT == 1, ]$high_blood_pressure) * 100 / sum(df$DEATH_EVENT), (sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$high_blood_pressure)) * 100 / sum(df$DEATH_EVENT)),
  Survived_patientsNumber = c(sum(df[df$DEATH_EVENT == 0, ]$high_blood_pressure), (299 - sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$high_blood_pressure))),
  Survived_patientsPercentage = c(sum(df[df$DEATH_EVENT == 0, ]$high_blood_pressure) * 100 / (299 - sum(df$DEATH_EVENT)), ((299 - sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$high_blood_pressure)) * 100 / (299 - sum(df$DEATH_EVENT))))
)
print(High_blood_pressure_df)
```

```
##   High_blood_pressure Full_sampleNumber Full_samplePercentage
## 1          Present             105             35.11706
## 2          Absent             194             64.88294
##   Dead_patientsNumber Dead_patientsPercentage Survived_patientsNumber
## 1                  39              40.625              66
## 2                  57              59.375             137
##   Survived_patientsPercentage
## 1              32.51232
## 2              67.48768
```

For Diabetes

```
Diabetes_df <- data.frame(
  Diabetes = c("Present", "Absent"),
  Full_sampleNumber = c(sum(df$diabetes), 299 - sum(df$diabetes)),
  Full_samplePercentage = c(sum(df$diabetes) * 100 / nrow(df), (299 - sum(df$diabetes)) * 100 / nrow(df)),
  Dead_patientsNumber = c(sum(df[df$DEATH_EVENT == 1, ]$diabetes), sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$diabetes)),
  Dead_patientsPercentage = c(sum(df[df$DEATH_EVENT == 1, ]$diabetes) * 100 / sum(df$DEATH_EVENT), (sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$diabetes)) * 100 / sum(df$DEATH_EVENT)),
  Survived_patientsNumber = c(sum(df[df$DEATH_EVENT == 0, ]$diabetes), ((299 - sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$diabetes))),
  Survived_patientsPercentage = c(sum(df[df$DEATH_EVENT == 0, ]$diabetes) * 100 / (299 - sum(df$DEATH_EVENT)), ((299 - sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$diabetes)) * 100 / (299 - sum(df$DEATH_EVENT))))
)
print(Diabetes_df)
```

```
##   Diabetes Full_sampleNumber Full_samplePercentage Dead_patientsNumber
## 1 Present             125             41.80602              40
## 2 Absent             174             58.19398              56
##   Dead_patientsPercentage Survived_patientsNumber Survived_patientsPercentage
## 1              41.66667              85              41.87192
## 2              58.33333             118              58.12808
```

For Sex

```
Sex_df <- data.frame(
  Sex = c("Man", "Female"),
  Full_sampleNumber = c(sum(df$sex), 299 - sum(df$sex)),
  Full_samplePercentage = c(sum(df$sex) * 100 / nrow(df), (299 - sum(df$sex)) * 100 / nrow(df)),
  Dead_patientsNumber = c(sum(df[df$DEATH_EVENT == 1, ]$sex), sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$sex)),
  Dead_patientsPercentage = c(sum(df[df$DEATH_EVENT == 1, ]$sex) * 100 / sum(df$DEATH_EVENT), (sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$sex)) * 100 / sum(df$DEATH_EVENT)),
  Survived_patientsNumber = c(sum(df[df$DEATH_EVENT == 0, ]$sex), ((299 - sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$sex))),
  Survived_patientsPercentage = c(sum(df[df$DEATH_EVENT == 0, ]$sex) * 100 / (299 - sum(df$DEATH_EVENT)), ((299 - sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$sex)) * 100 / (299 - sum(df$DEATH_EVENT))))
)
print(Sex_df)
```

```
##      Sex Full_sampleNumber Full_samplePercentage Dead_patientsNumber
## 1    Man                194                64.88294                62
## 2 Female                105                35.11706                34
##      Dead_patientsPercentage Survived_patientsNumber Survived_patientsPercentage
## 1                64.58333                132                65.02463
## 2                35.41667                71                34.97537
```

For Smoking

```
Smoking_df <- data.frame(
  Smoking = c("Present", "Absent"),
  Full_sampleNumber = c(sum(df$smoking), 299 - sum(df$smoking)),
  Full_samplePercentage = c(sum(df$smoking) * 100 / nrow(df), (299 - sum(df$smoking)) * 100 / nrow(df)),
  Dead_patientsNumber = c(sum(df[df$DEATH_EVENT == 1, ]$smoking), sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$smoking)),
  Dead_patientsPercentage = c(sum(df[df$DEATH_EVENT == 1, ]$smoking) * 100 / sum(df$DEATH_EVENT), (sum(df$DEATH_EVENT) - sum(df[df$DEATH_EVENT == 1, ]$smoking)) * 100 / sum(df$DEATH_EVENT)),
  Survived_patientsNumber = c(sum(df[df$DEATH_EVENT == 0, ]$smoking), ((299 - sum(df$DEATH_EVENT)) - sum(df[df$DEATH_EVENT == 1, ]$smoking))),
  Survived_patientsPercentage = c(sum(df[df$DEATH_EVENT == 0, ]$smoking) * 100 / (299 - sum(df$DEATH_EVENT)), ((299 - sum(df$DEATH_EVENT)) - sum(df[df$DEATH_EVENT == 1, ]$smoking)) * 100 / (299 - sum(df$DEATH_EVENT)))
)
print(Smoking_df)
```

```
##      Smoking Full_sampleNumber Full_samplePercentage Dead_patientsNumber
## 1 Present                96                32.10702                30
## 2 Absent                203                67.89298                66
##      Dead_patientsPercentage Survived_patientsNumber Survived_patientsPercentage
## 1                31.25                66                32.51232
## 2                68.75                137                67.48768
```

We can clearly observe that: 1) Out of 299 patients 129 were found to have anaemia which is about 43% of total population. Only 64.34% patient who had anaemia could survive till follow-up period. 2) Out of 299 patients 105 were found to have high blood pressure which is about 35% of total population. Only 62.85% patient who had high blood pressure could survive till follow-up period. 3) Out of 299 patients 125 were found to have diabetes which is about 42% of total population. Only 68% patient who had diabetes could survive till follow-up period. 4) Male constituted 64.88% and Female constituted 35.12% of total population. 5) Out of 299 patients 96 were found to have diabetes which is about 32% of total population. Only 68.75% patient who had diabetes could survive till follow-up period.

Univariate analysis

#Getting basic descriptive statistic summary of age column
cat("For age", "\n")

For age

```
summary(df$age)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      40.00   51.00   60.00   60.83   70.00   95.00
```

```
cat("Standard deviation:", sd(df$age), "\n")
```

Standard deviation: 11.895

```
cat("Skewness:",skewness(df$age),"\n")
```

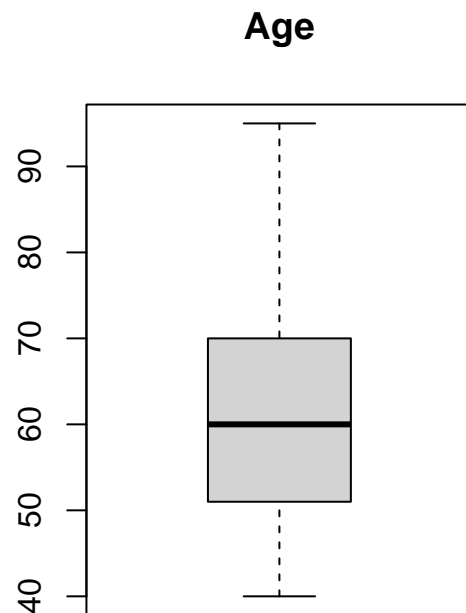
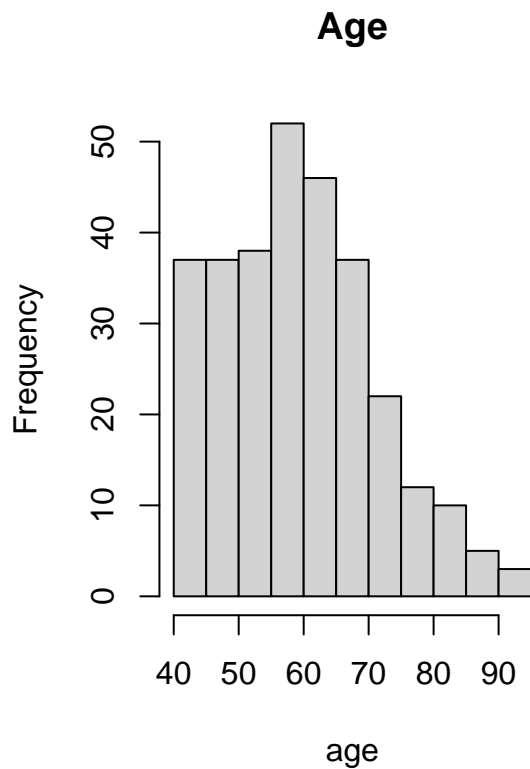
```
## Skewness: 0.4220415
```

```
cat("Kurtosis:",kurtosis(df$age),"\n","\n")
```

```
## Kurtosis: 2.798664
```

```
##
```

```
#For age  
par(mfrow=c(1,2))  
hist(df$age,main="Age",xlab="age")  
boxplot(df$age,main="Age")
```



```
par(mfrow=c(1,1))
```

Age of patients lie between range 40-95 years with average age of 60.83 years and standard deviation of 11.895. Most of the patients have age 55-65 years.

```
#Getting basic descriptive statistic summary of Creatinine phosphokinase  
cat("For creatinine_phosphokinase","\n")
```

```
## For creatinine_phosphokinase
```

```
summary(df$creatinine_phosphokinase)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      23.0  116.5   250.0   581.8  582.0  7861.0
```

```
cat("Standard deviation:",sd(df$creatinine_phosphokinase),"\n")
```

```
## Standard deviation: 970.2879
```

```
cat("Skewness:",skewness(df$creatinine_phosphokinase),"\n")
```

```
## Skewness: 4.440689
```

```
cat("Kurtosis:",kurtosis(df$creatinine_phosphokinase),"\n","\n")
```

```
## Kurtosis: 27.71046
```

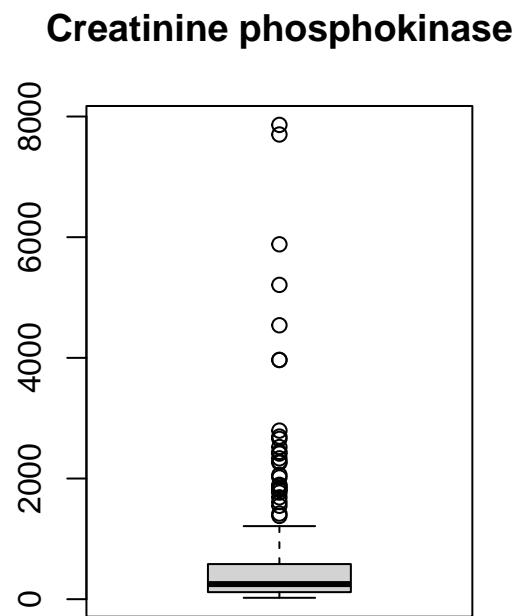
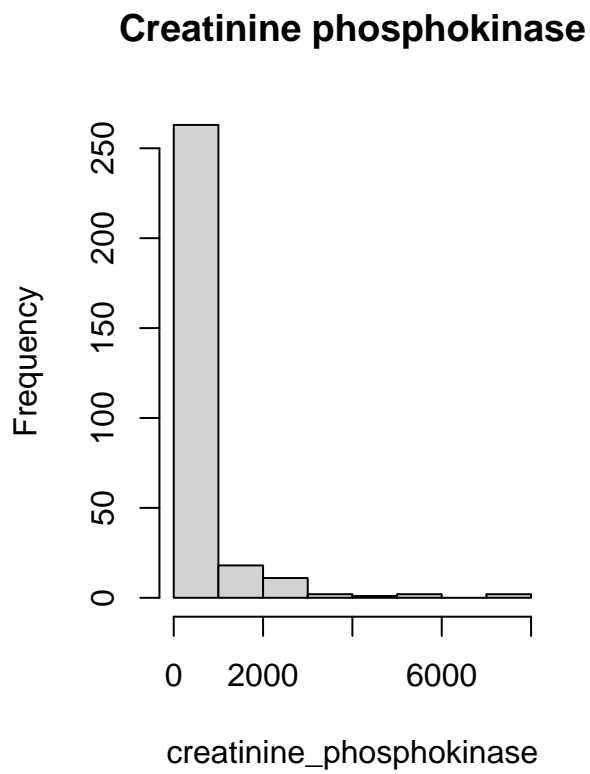
```
##
```

```
#For Creatinine phosphokinase
```

```
par(mfrow=c(1,2))
```

```
hist(df$creatinine_phosphokinase,main="Creatinine phosphokinase",xlab="creatinine_phosphokinase")
```

```
boxplot(df$creatinine_phosphokinase,main="Creatinine phosphokinase")
```




```
par(mfrow=c(1,1))
```

Creatinine phosphokinase column is highly right skewed & peaked with range of 23-7861, standard deviation of 970.28, Skewness of 4.44, kurtosis of 27.7. From the box-plot we can see that it has many outliers.

```
#Getting basic descriptive statistic summary of Ejection fraction  
cat("For ejection_fraction", "\n")
```

```
## For ejection_fraction
```

```
summary(df$ejection_fraction)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
##      14.00   30.00   38.00   38.08   45.00   80.00
```

```
cat("Standard deviation:", sd(df$ejection_fraction), "\n")
```

```
## Standard deviation: 11.83484
```

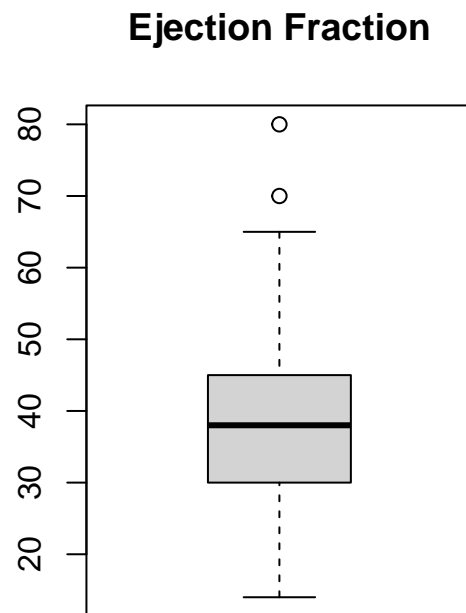
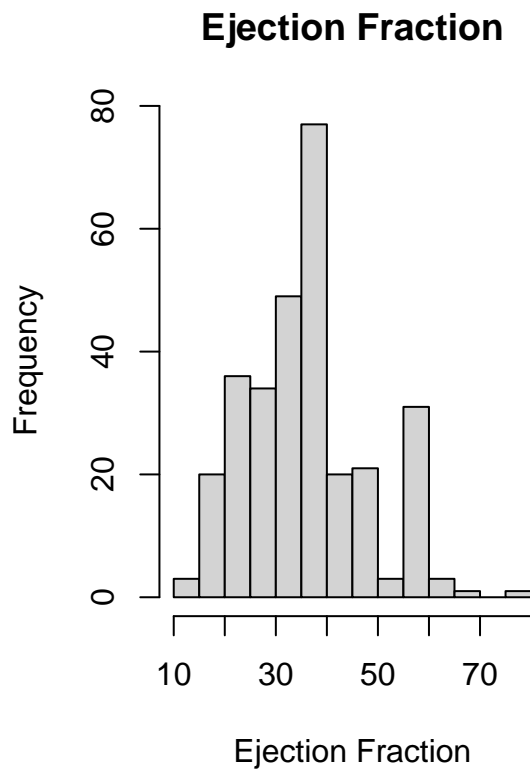
```
cat("Skewness:", skewness(df$ejection_fraction), "\n")
```

```
## Skewness: 0.5525927
```

```
cat("Kurtosis:", kurtosis(df$ejection_fraction), "\n", "\n")
```

```
## Kurtosis: 3.02072  
##
```

```
#For Ejection fraction  
par(mfrow=c(1,2))  
hist(df$ejection_fraction, main="Ejection Fraction", xlab="Ejection Fraction")  
boxplot(df$ejection_fraction, main="Ejection Fraction")
```



```
par(mfrow=c(1,1))
```

Ejection fraction column has range of 14-80 ,mean of 38.08 ,standard deviation of 11.83, Skewness of 0.55 ,kurtosis of 3.02.

```
#Getting basic descriptive statistic summary of Platelets
cat("For platelets","\n")
```

```
## For platelets
```

```
summary(df$platelets)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      25.1  211.5   262.0   262.7  303.5   850.0
```

```
cat("Standard deviation:",sd(df$platelets),"\n")
```

```
## Standard deviation: 98.39959
```

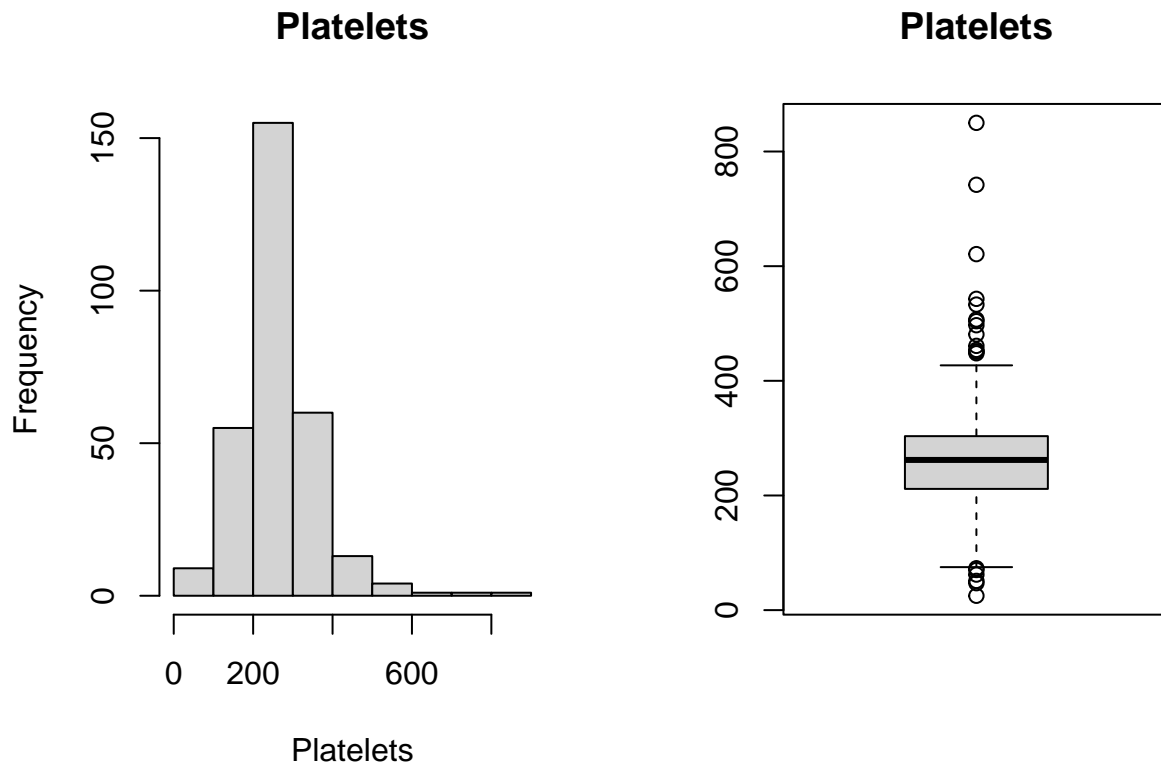
```
cat("Skewness:",skewness(df$platelets),"\n")
```

```
## Skewness: 1.425942
```

```
cat("Kurtosis:",kurtosis(df$platelets),"\n","\n")
```

```
## Kurtosis: 8.950815
##
```

```
#For Platelets
par(mfrow=c(1,2))
hist(df$platelets, main="Platelets", xlab="Platelets")
boxplot(df$platelets, main="Platelets")
```



```
par(mfrow=c(1,1))
```

Platelets column is high peaked column with range of 25100-850000, mean of 263358, standard deviation of 97804.24, Skewness of 1.45, kurtosis of 9.08. From the box-plot we can see that it has many outliers.

```
#Getting basic descriptive statistic summary of Serum creatinine
cat("For serum_creatinine","\n")
```

```
## For serum_creatinine
```

```
summary(df$serum_creatinine)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    0.500   0.900   1.100   1.394   1.400   9.400
```

```
cat("Standard deviation:",sd(df$serum_creatinine),"\n")
```

```
## Standard deviation: 1.03451
```

```
cat("Skewness:",skewness(df$serum_creatinine),"\n")
```

```
## Skewness: 4.43361
```

```
cat("Kurtosis:",kurtosis(df$serum_creatinine),"\n","\n")
```

```
## Kurtosis: 28.37835
```

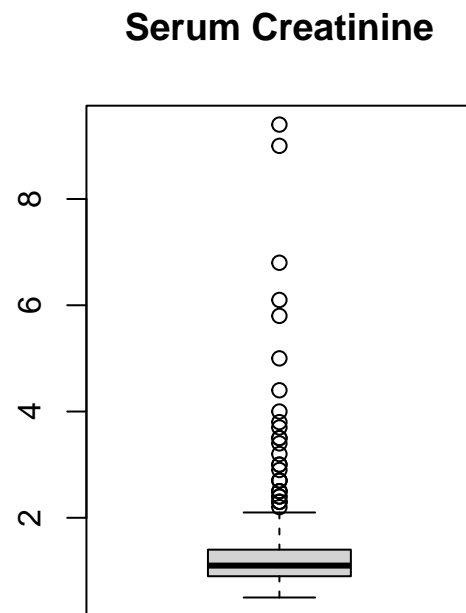
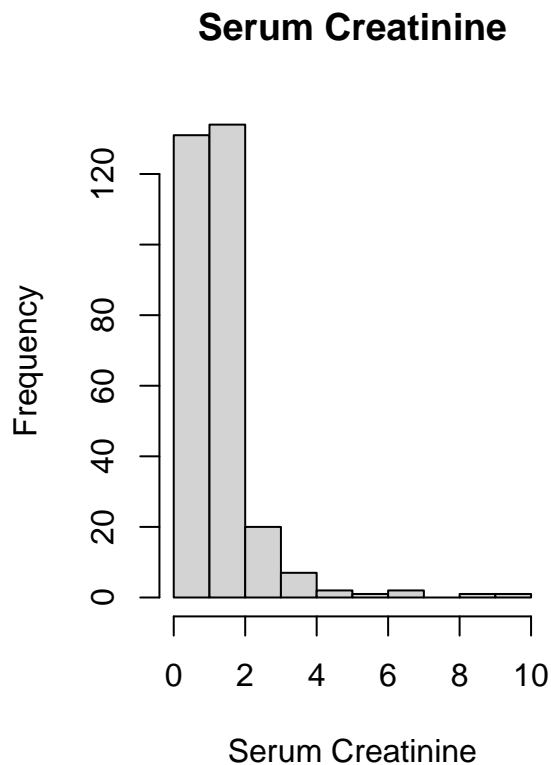
```
##
```

```
#For Serum creatinine
```

```
par(mfrow=c(1,2))
```

```
hist(df$serum_creatinine, main="Serum Creatinine", xlab="Serum Creatinine")
```

```
boxplot(df$serum_creatinine, main="Serum Creatinine")
```



```
par(mfrow=c(1,1))
```

Serum creatinine is highly right skewed & peaked with range of 0.5-9.4, standard deviation of 1.03, Skewness of 4.43, kurtosis of 28.37. From the box-plot we can see that it has many outliers.

```
#Getting basic descriptive statistic summary of Serum sodium  
cat("For serum_sodium","\n")
```

```
## For serum_sodium
```

```
summary(df$serum_sodium)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
##    113.0   134.0   137.0   136.6   140.0   148.0
```

```
cat("Standard deviation:",sd(df$serum_sodium),"\n")
```

```
## Standard deviation: 4.412477
```

```
cat("Skewness:",skewness(df$serum_sodium),"\n")
```

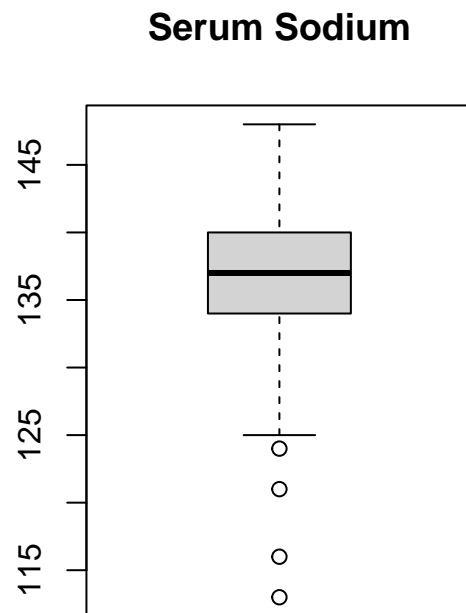
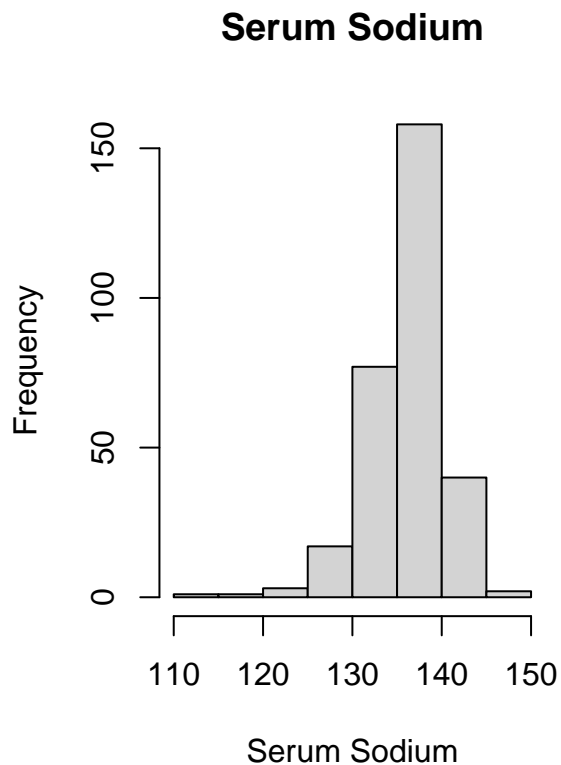
```
## Skewness: -1.04287
```

```
cat("Kurtosis:",kurtosis(df$serum_sodium),"\n","\n")
```

```
## Kurtosis: 7.031142
```

```
##
```

```
#For Serum sodium  
par(mfrow=c(1,2))  
hist(df$serum_sodium, main="Serum Sodium", xlab="Serum Sodium")  
boxplot(df$serum_sodium, main="Serum Sodium")
```



```
par(mfrow=c(1,1))
```

Serum sodium has range of 113-148, mean of 136.6, standard deviation of 4.41,Skewness of -1.04,kurtosis of 7.03. It is left skewed.

```
#Getting basic descriptive statistic summary of Death event
cat("For time","\n")
```

```
## For time
```

```
summary(df$time)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      4.0   73.0   115.0   130.3   203.0   285.0
```

```
cat("Standard deviation:",sd(df$time),"\n")
```

```
## Standard deviation: 77.61421
```

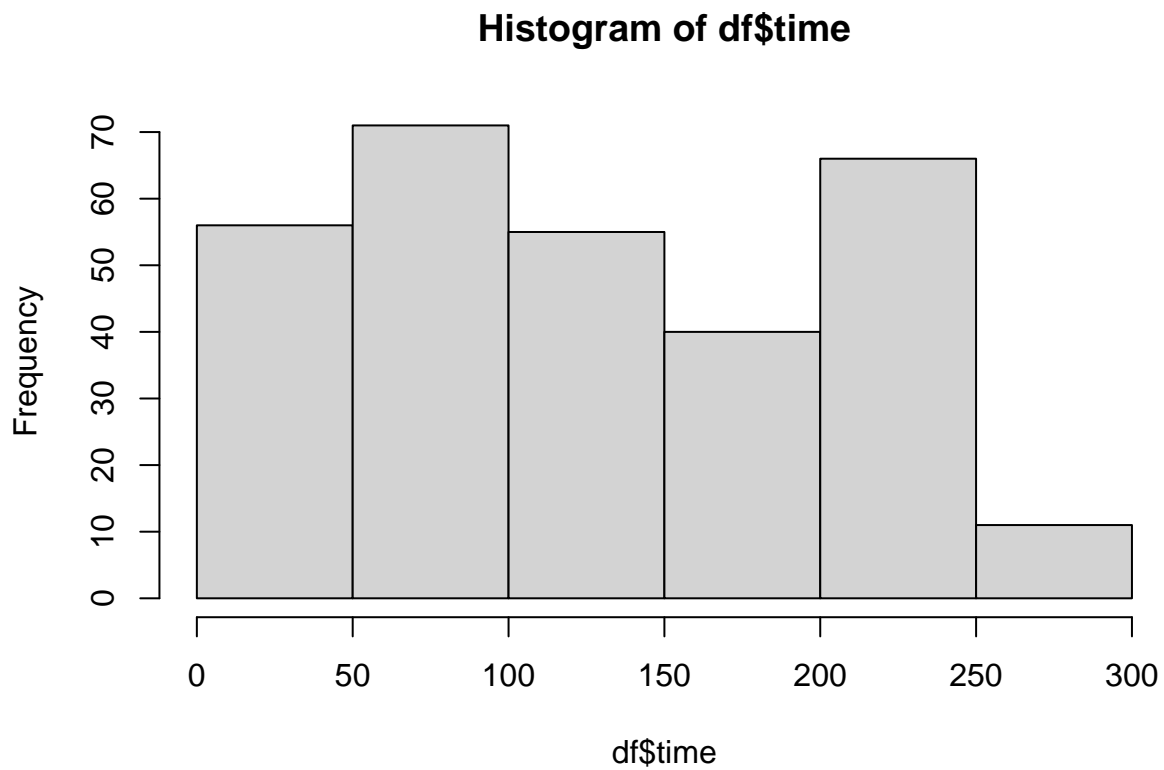
```
cat("Skewness:",skewness(df$time),"\n")
```

```
## Skewness: 0.1271606
```

```
cat("Kurtosis:",kurtosis(df$time),"\\n")
```

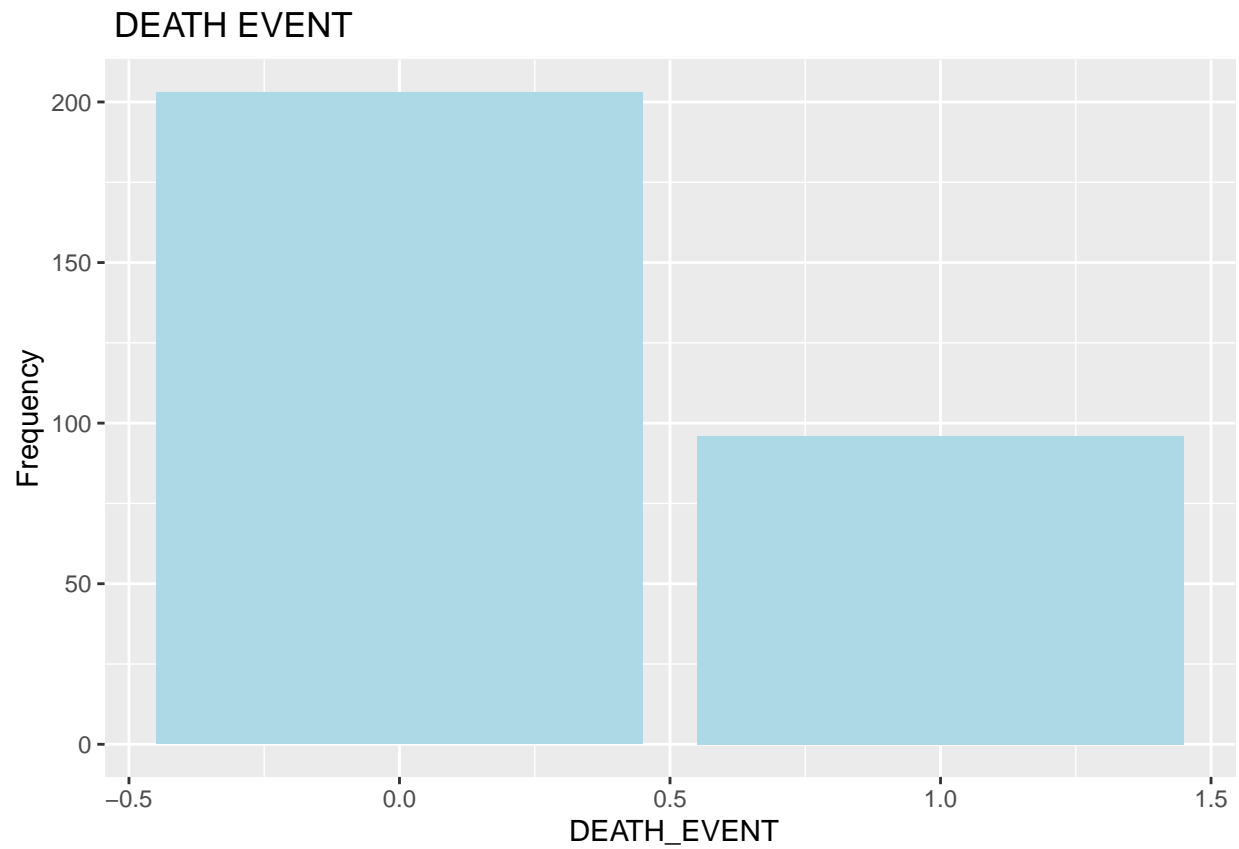
```
## Kurtosis: 1.788126
```

```
hist(df$time)
```



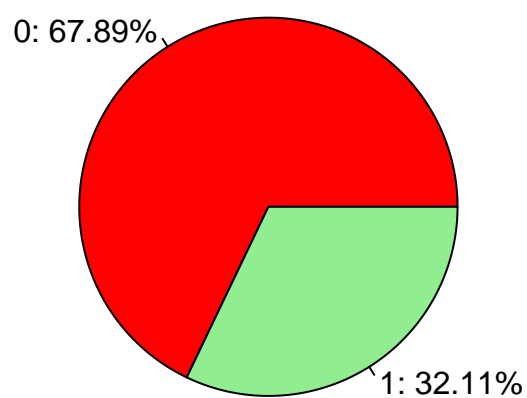
```
#For Death event
```

```
ggplot(df, aes(x = DEATH_EVENT)) +geom_bar(fill = "lightblue") +labs(title = " DEATH EVENT", x = " DEA
```



```
pie(table(df$ DEATH_EVENT), labels = paste0(names(table(df$ DEATH_EVENT)), ":", round(table(df$ DEATH_EVENT), 1)))
```

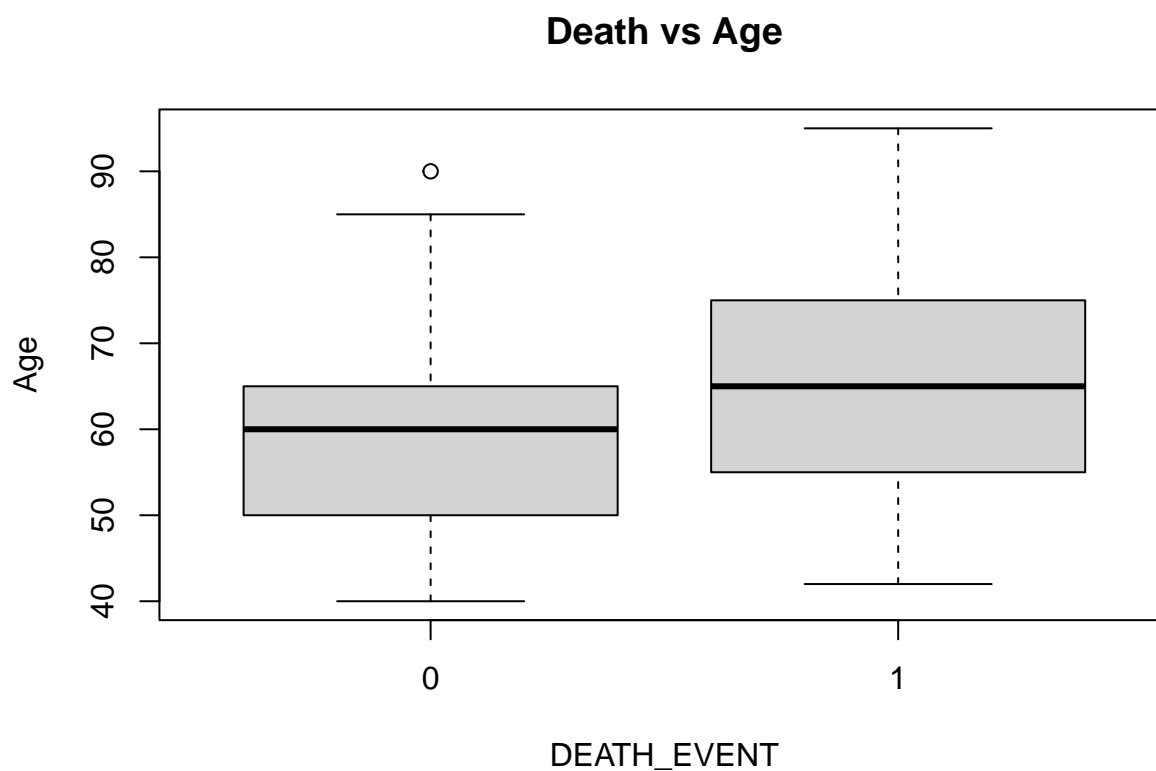

DEATH_EVENT Distribution



Among 299 patients 67.89% died before follow-up period.

Bivariate analysis

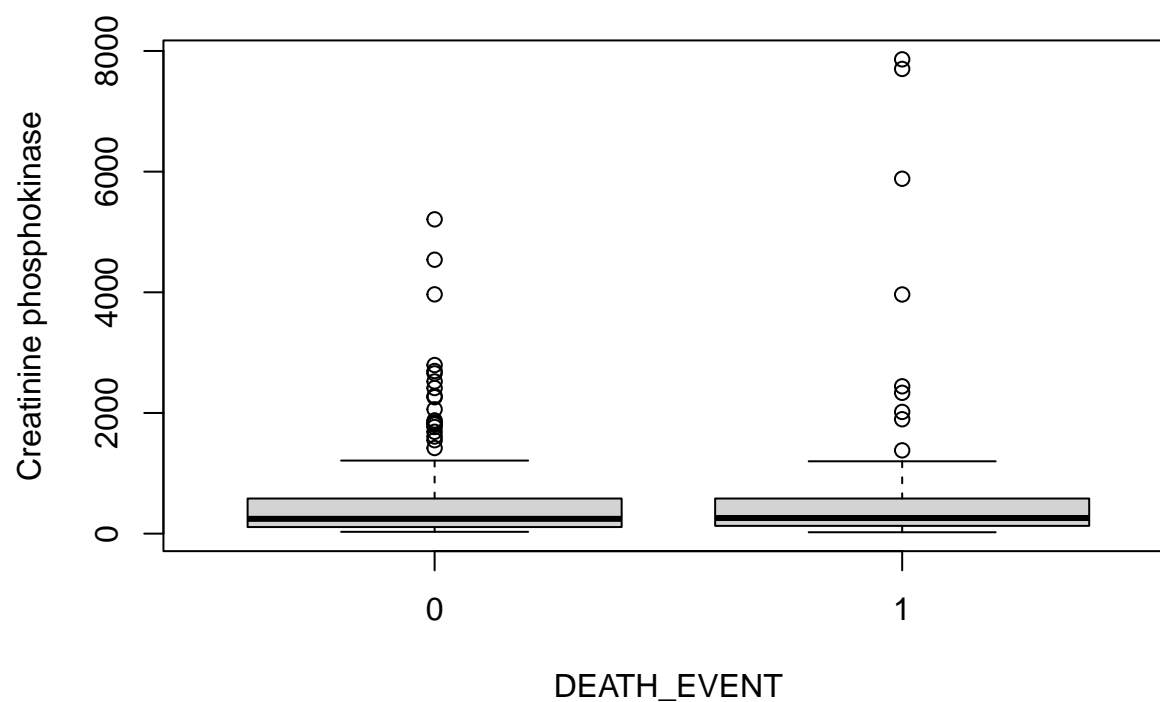
```
#Death vs Age  
boxplot(df$age ~ df$DEATH_EVENT, main = "Death vs Age", xlab = "DEATH_EVENT", ylab = "Age")
```



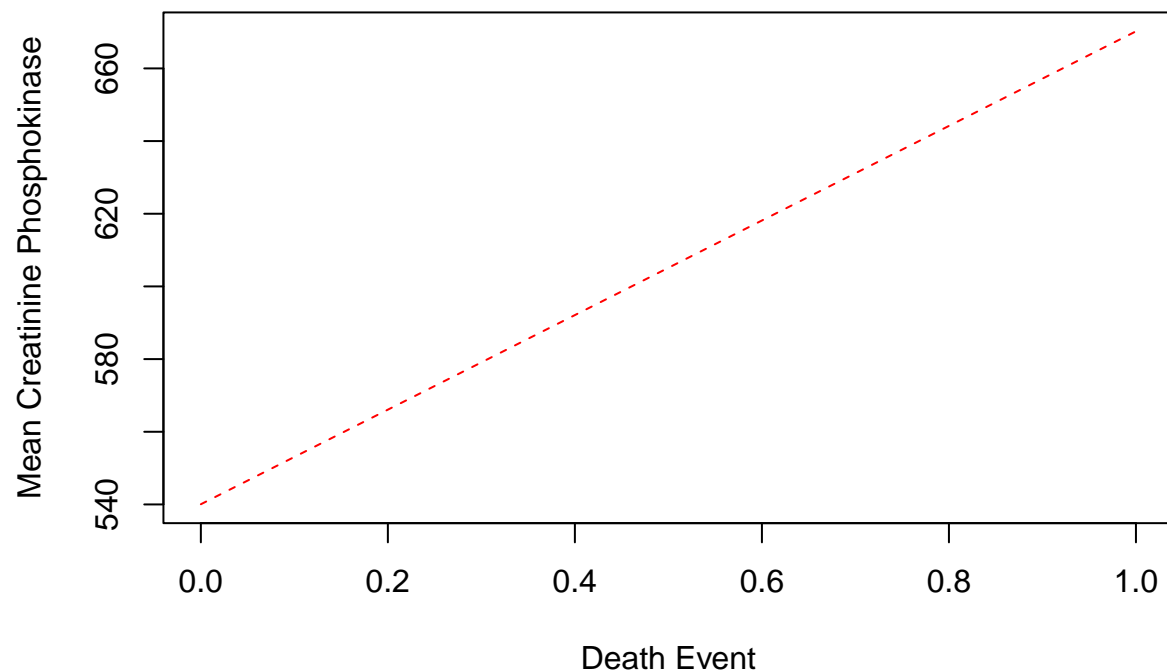
From death vs age plot we see that young age group people have more chances of survival.

```
#Death vs Creatinine phosphokinase  
boxplot(df$creatinine_phosphokinase ~ df$DEATH_EVENT, main = "Death vs Creatinine phosphokinase", xlab = "DEATH_EVENT", ylab = "creatinine_phosphokinase")
```

Death vs Creatinine phosphokinase



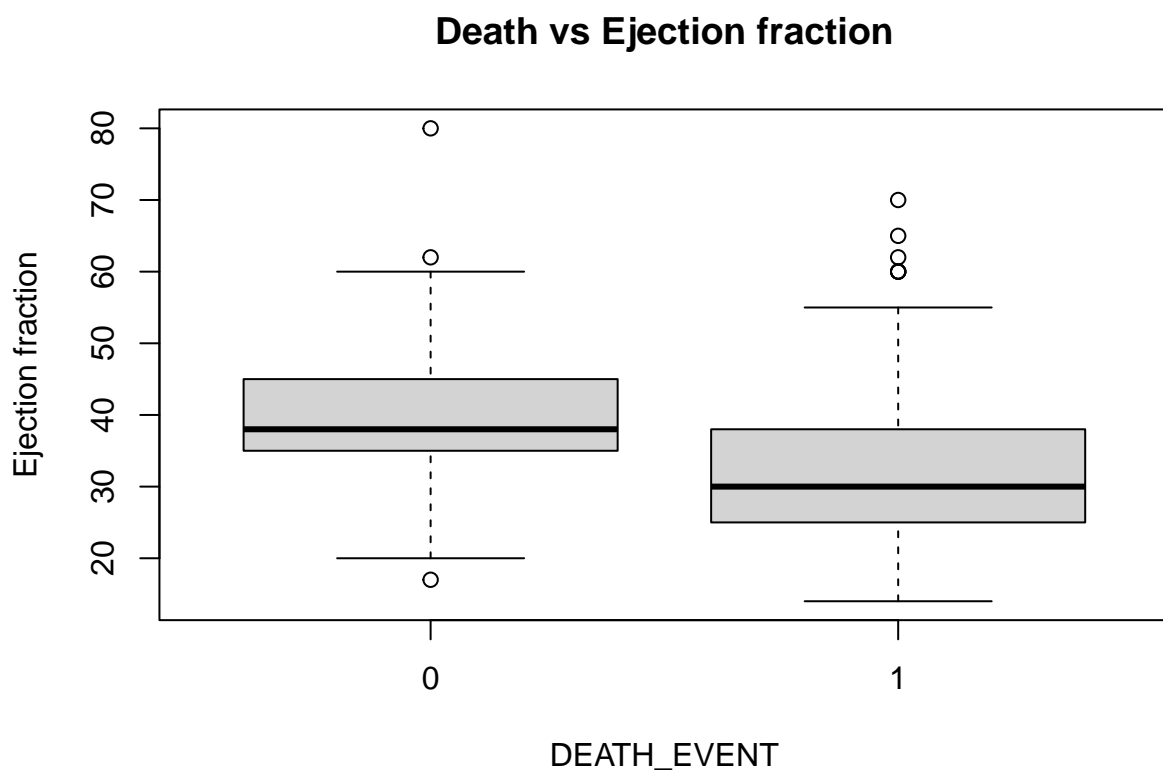
```
df_summary <- df %>%  
  group_by(DEATH_EVENT) %>%  
  summarize(mean_cpk = mean(creatinine_phosphokinase))  
  
# Draw a line plot with the summary data  
plot(df_summary$DEATH_EVENT, df_summary$mean_cpk, type = "l",  
      xlab = "Death Event", ylab = "Mean Creatinine Phosphokinase", col="red", lty=2)
```



From the line plot we can see that on an average the patients who died had high value of CPK. We might consider it as potential factor to predict survival of patients. But since the range of CPK was too high than its mean so we cannot say with high certainty that it is a good factor to predict survival.

```
#Death vs Ejection fraction
```

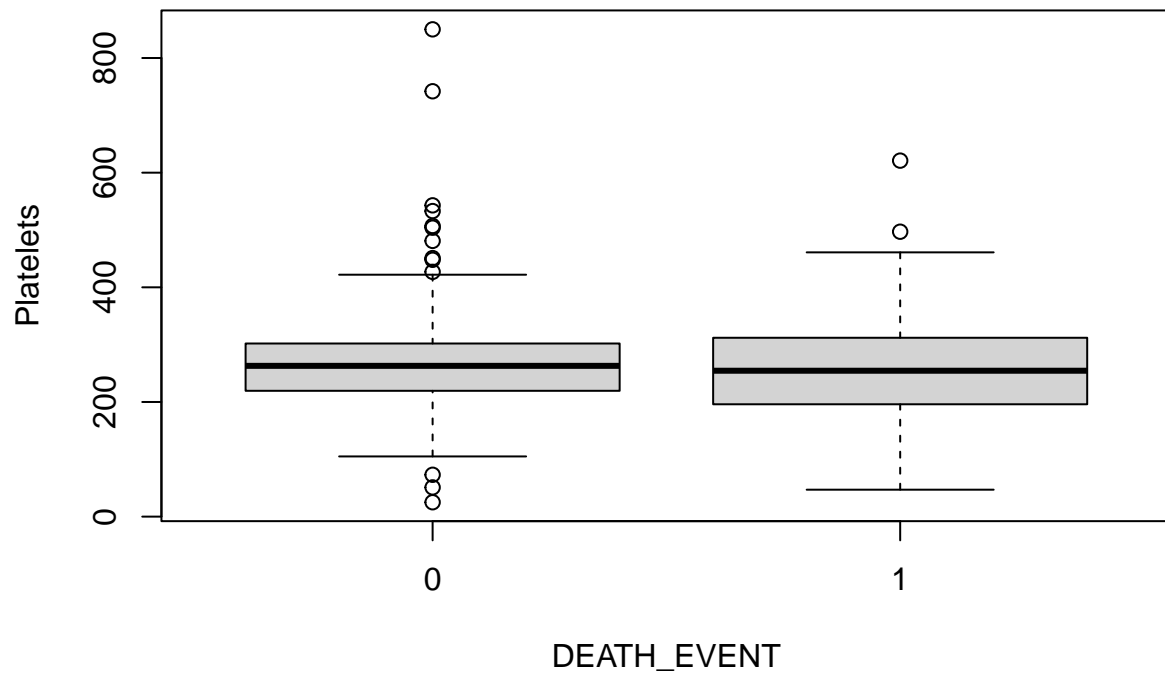
```
boxplot(df$ejection_fraction ~ df$DEATH_EVENT, main = "Death vs Ejection fraction", xlab = "DEATH_EVENT")
```



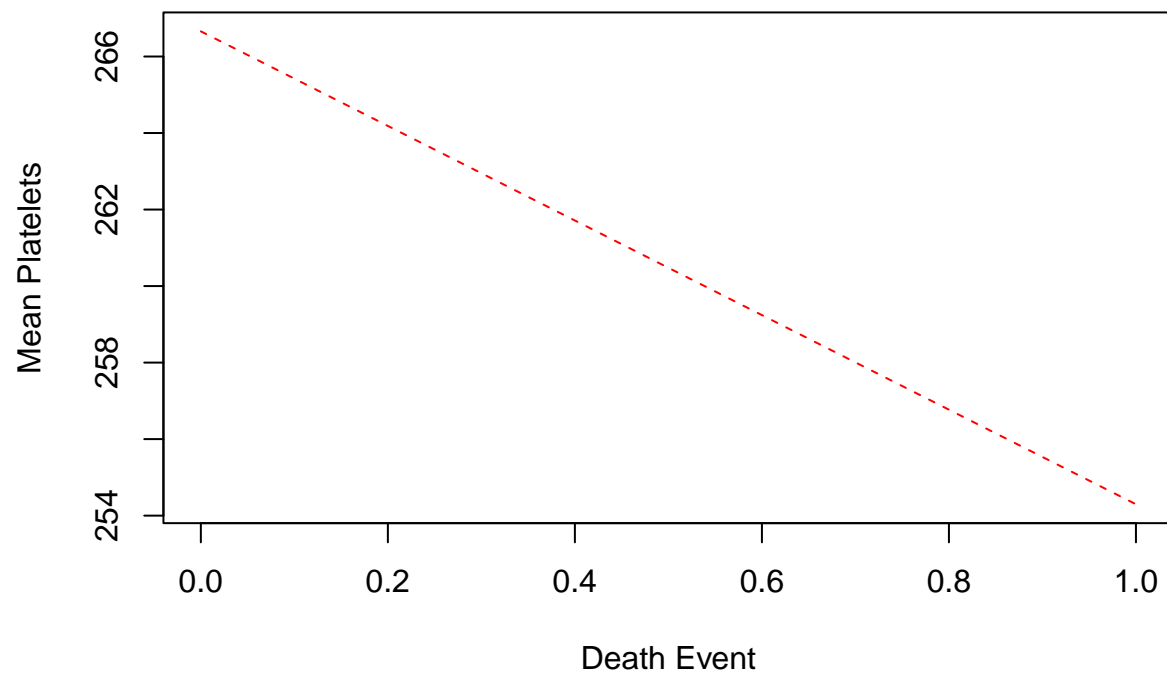
Ejection fraction vs death plot reveals that the people who died had low value of ejection fraction i.e. left ventricle does not pump effectively with each contraction.

```
#Death vs Platelets  
boxplot(df$platelets ~ df$DEATH_EVENT, main = "Death vs Platelets", xlab = "DEATH_EVENT", ylab = "Platelets")
```

Death vs Platelets



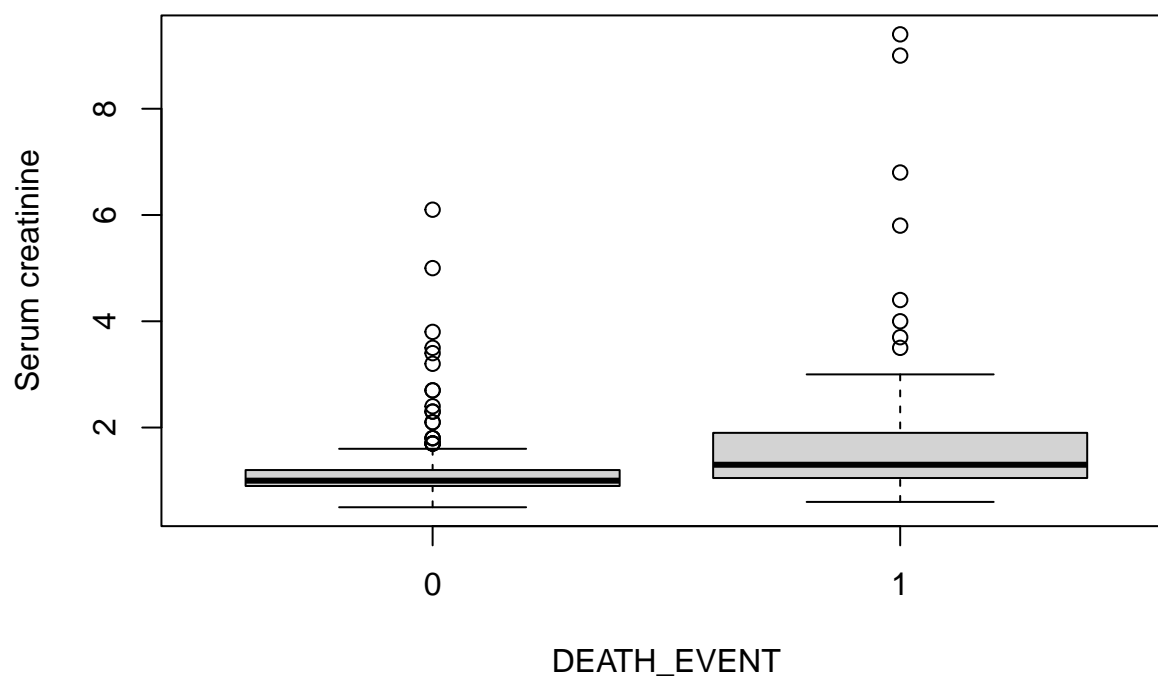
```
df_summary <- df %>%  
  group_by(DEATH_EVENT) %>%  
  summarize(mean_plat = mean(platelets))  
  
# Draw a line plot with the summary data  
plot(df_summary$DEATH_EVENT, df_summary$mean_plat, type = "l",  
      xlab = "Death Event", ylab = "Mean Platelets", col="red", lty=2)
```



On Average plateletes count remained same for patients who survived and who died. Since there is very slight change in average of platelets count.

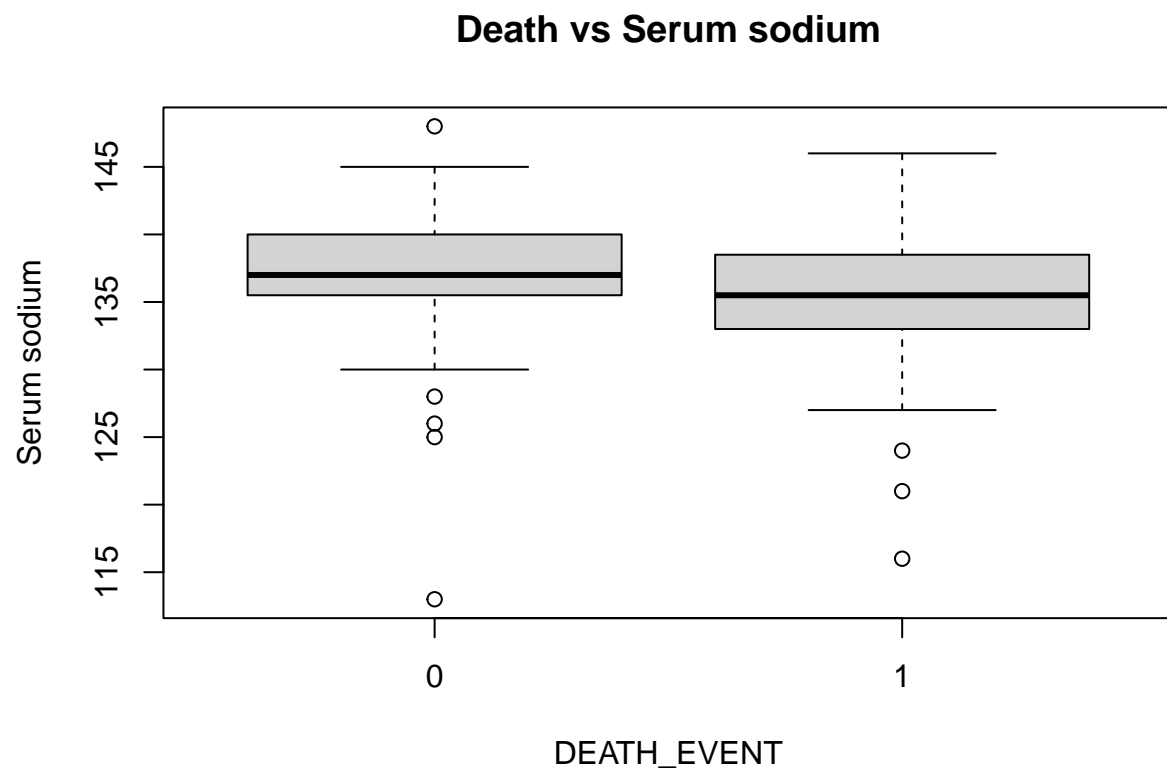
```
#Death vs Serum creatinine  
boxplot(df$serum_creatinine ~ df$DEATH_EVENT, main = "Death vs Serum creatinine", xlab = "DEATH_EVENT",
```

Death vs Serum creatinine



High value for Serum creatinine was observed for patients who died telling that it can be potential factor for predicting survival of patients.

```
#Death vs Serum sodium  
boxplot(df$serum_sodium ~ df$DEATH_EVENT, main = "Death vs Serum sodium", xlab = "DEATH_EVENT", ylab = "Serum sodium")
```

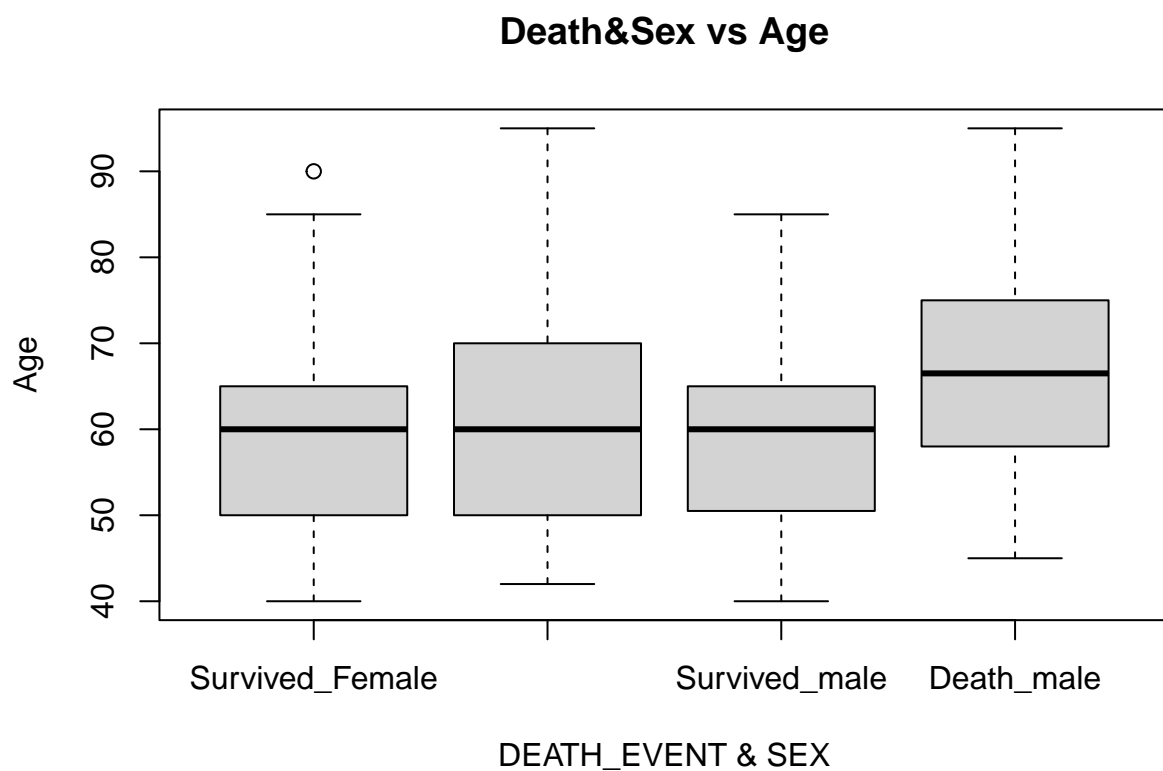



Low value for Serum sodium was observed for patients who died telling that it can be potential factor for predicting survival of patients.

#Analysis of survival and sex of patients taken together

#Death&Sex vs Age

```
boxplot(df$age ~ interaction(df$DEATH_EVENT,df$sex), main = "Death&Sex vs Age", xlab = "DEATH_EVENT & S
```



Among all patients male patients who were younger had high chances of survival since among those males who died older ones were majority. And same is observed for females.

```
#Death&Sex vs Serum Creatinine
boxplot(df$serum_creatinine ~ interaction(df$DEATH_EVENT, df$sex), main = "Death&Sex vs Serum Creatinine")
```

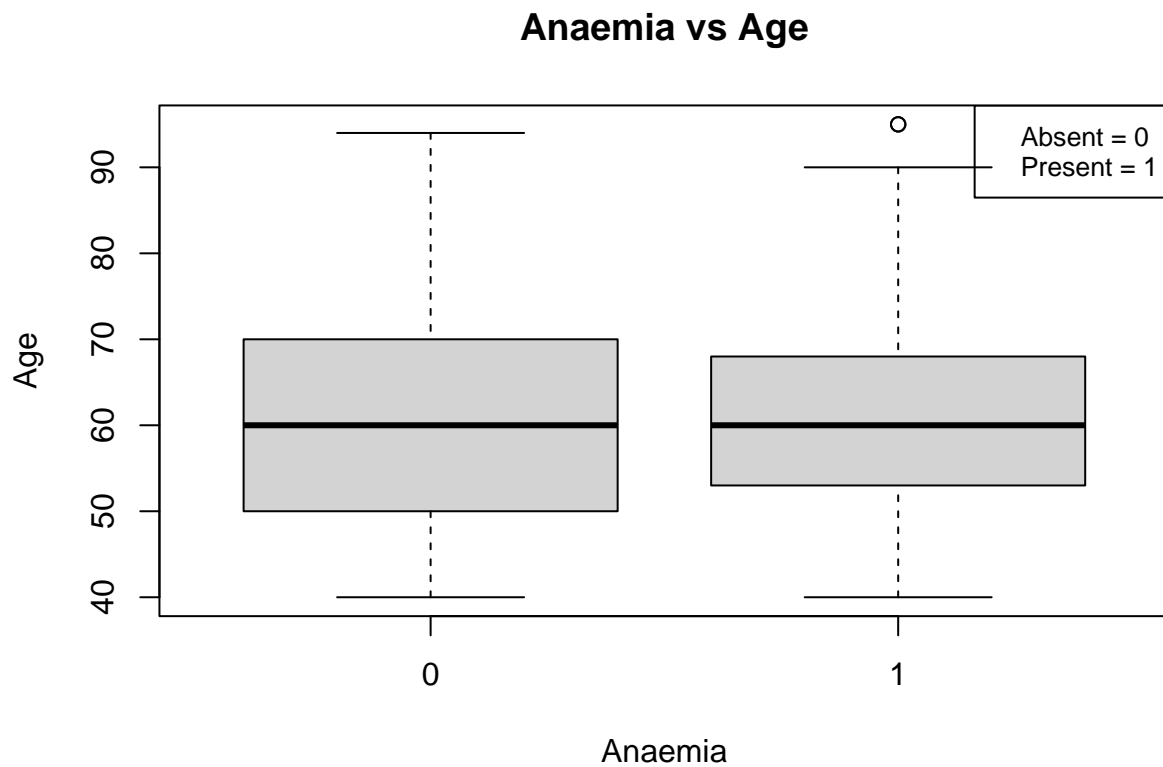
Death&Sex vs Serum Creatinine



In both males and females the patients who had high level of Serum Creatinine died before follow up period.
#Relationship between different feature columns

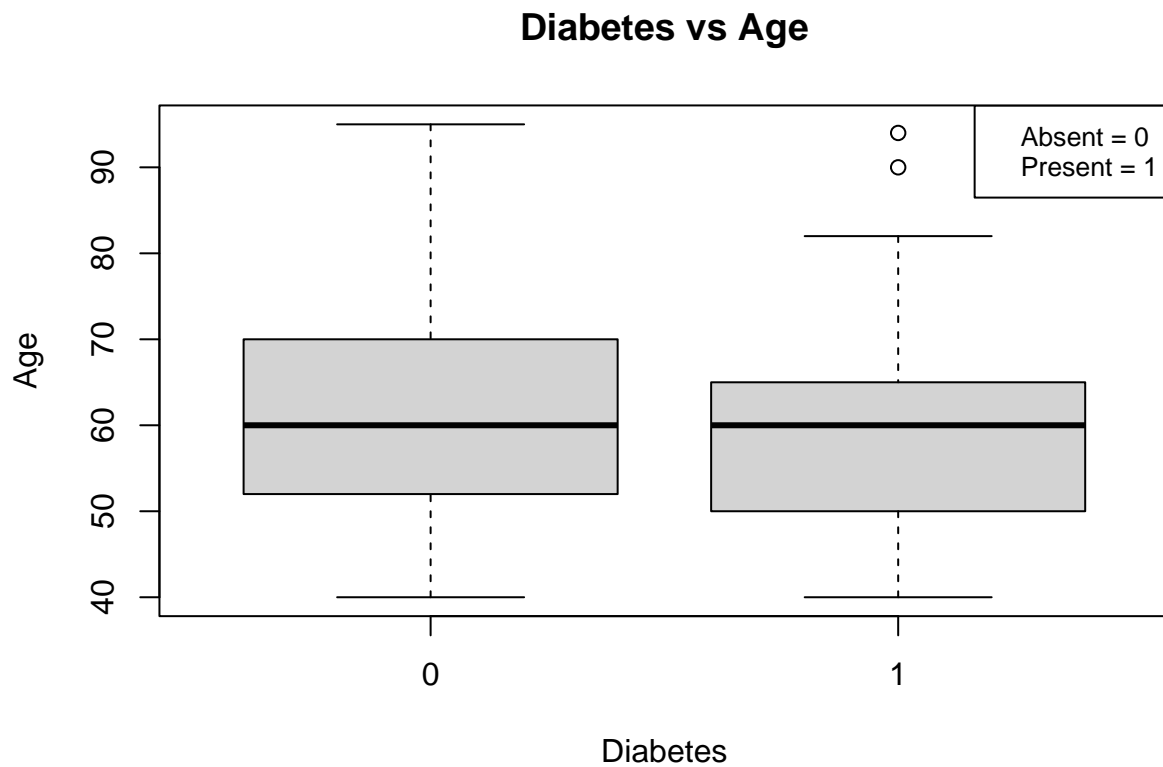
#Anaemia vs Age

```
boxplot(df$age ~ df$anaemia, main = "Anaemia vs Age", xlab = "Anaemia", ylab = "Age")  
legend("topright", legend = c("Absent = 0", "Present = 1"), cex=0.8)
```



On average age remained same for patients who had anaemia and who didn't have anaemia.

```
#Diabetes vs Age
boxplot(df$age ~ df$diabetes, main = "Diabetes vs Age", xlab = "Diabetes", ylab = "Age")
legend("topright", legend = c("Absent = 0", "Present = 1"), cex=0.8)
```



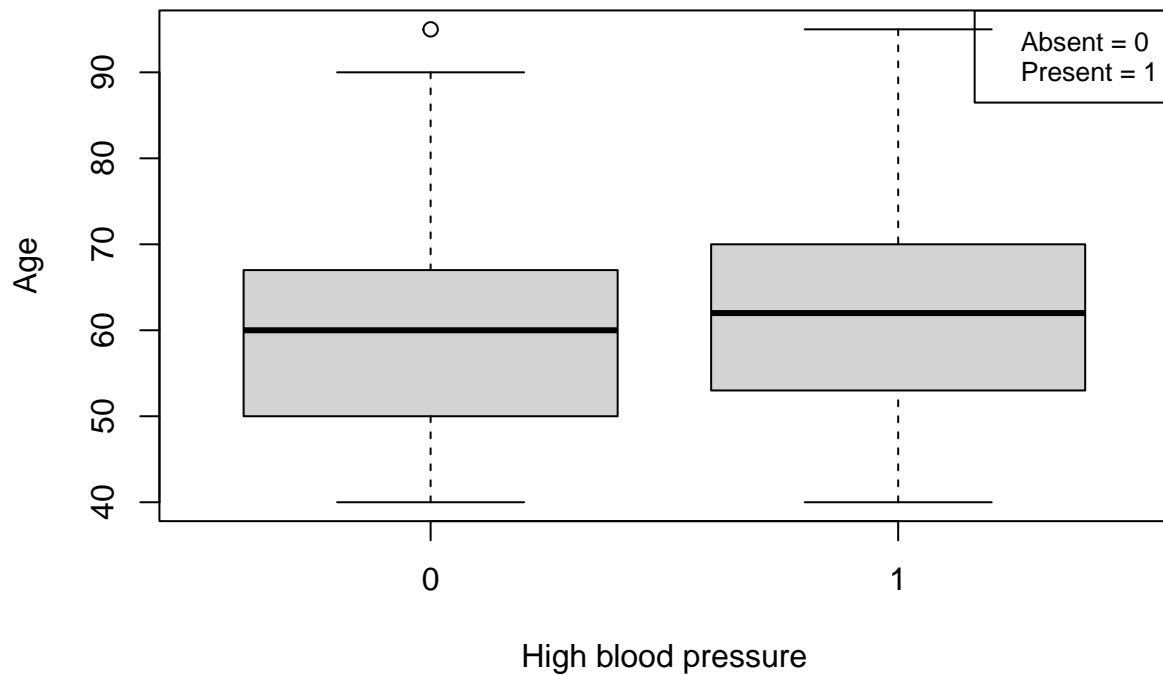
On average age remained same for patients who had diabetes and who didn't have diabetes.

```
#High blood pressure vs Age
```

```
boxplot(df$age ~ df$high_blood_pressure, main = "High blood pressures vs Age", xlab = "High blood pressure",
```

```
legend("topright", legend = c("Absent = 0", "Present = 1"), cex=0.8)
```

High blood pressures vs Age

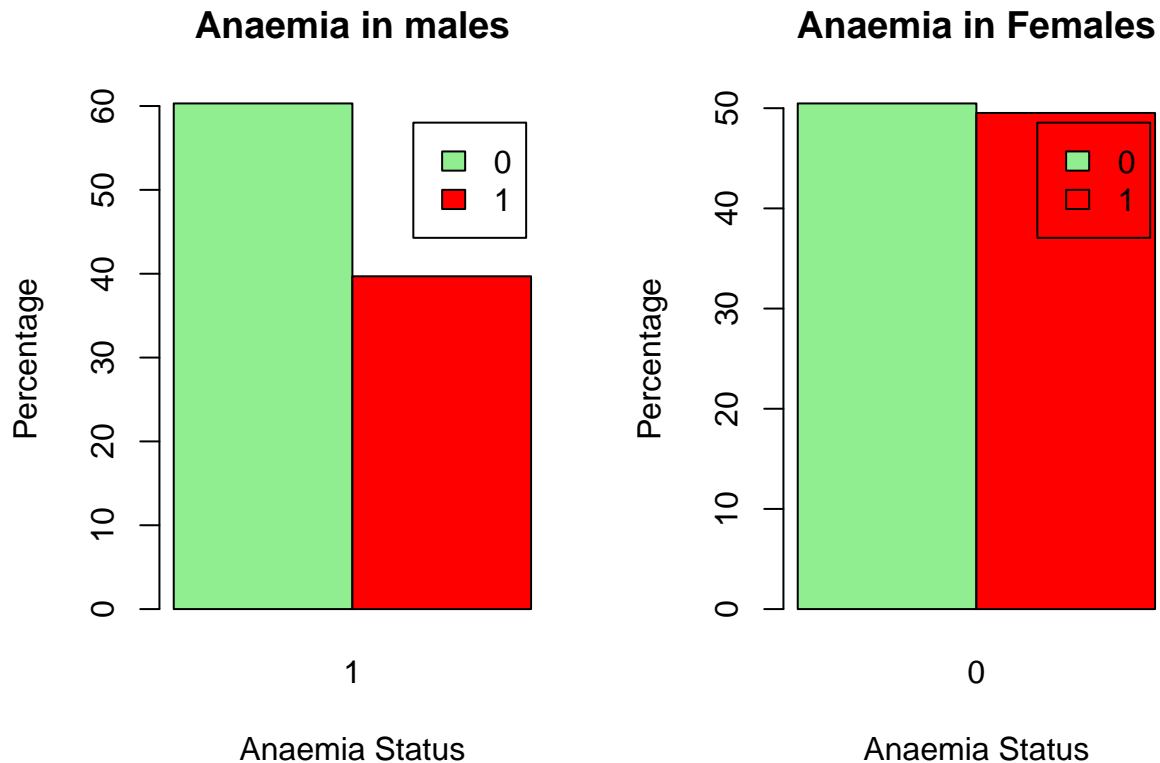


Patients who had high blood pressure were slightly aged than those who didn't had high blood pressure.

```
#Anaemia & Sex
df_males <- subset(df, sex == 1)
df_females <- subset(df, sex == 0)

# Create a table for anaemia and sex
x <- table(df_males$anaemia, df_males$sex)
y <- table(df_females$anaemia, df_females$sex)
x_percentage <- prop.table(x, margin = 2) * 100
y_percentage <- prop.table(y, margin = 2) * 100 # Calculate percentages across columns (sex)

par(mfrow=c(1,2))
# Barplot for males with percentages
barplot(x_percentage, beside = TRUE, col = c("lightgreen", "red"),
        legend = rownames(y_percentage), main = "Anaemia in males",
        xlab = "Anaemia Status", ylab = "Percentage")
# Barplot for females with percentages
barplot(y_percentage, beside = TRUE, col = c("lightgreen", "red"),
        legend = rownames(y_percentage), main = "Anaemia in Females",
        xlab = "Anaemia Status", ylab = "Percentage")
```

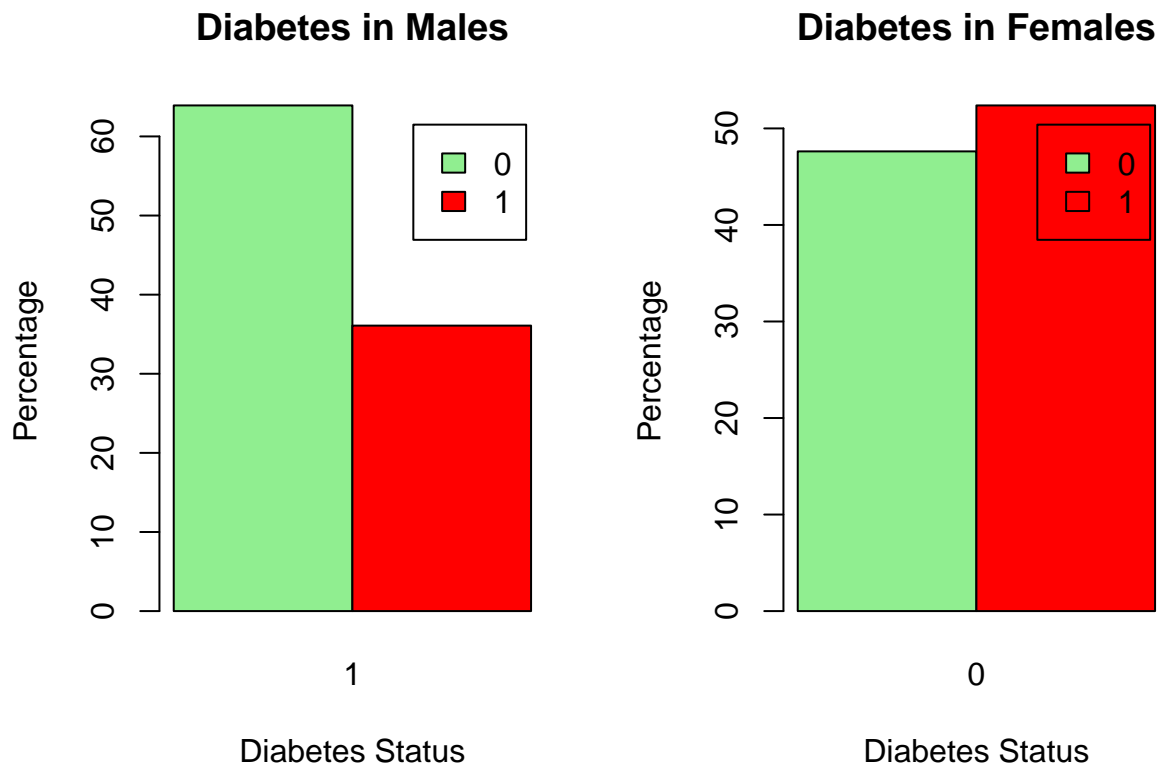


Among male patients around 40% had anaemia and among female around 50% had anaemia. That may mean that females are more prone to anaemia.

```
df_males <- subset(df, sex == 1)
df_females <- subset(df, sex == 0)

# Create a table for diabetes and sex
x <- table(df_males$diabetes, df_males$sex)
y <- table(df_females$diabetes, df_females$sex)
x_percentage <- prop.table(x, margin = 2) * 100
y_percentage <- prop.table(y, margin = 2) * 100 # Calculate percentages across columns (sex)

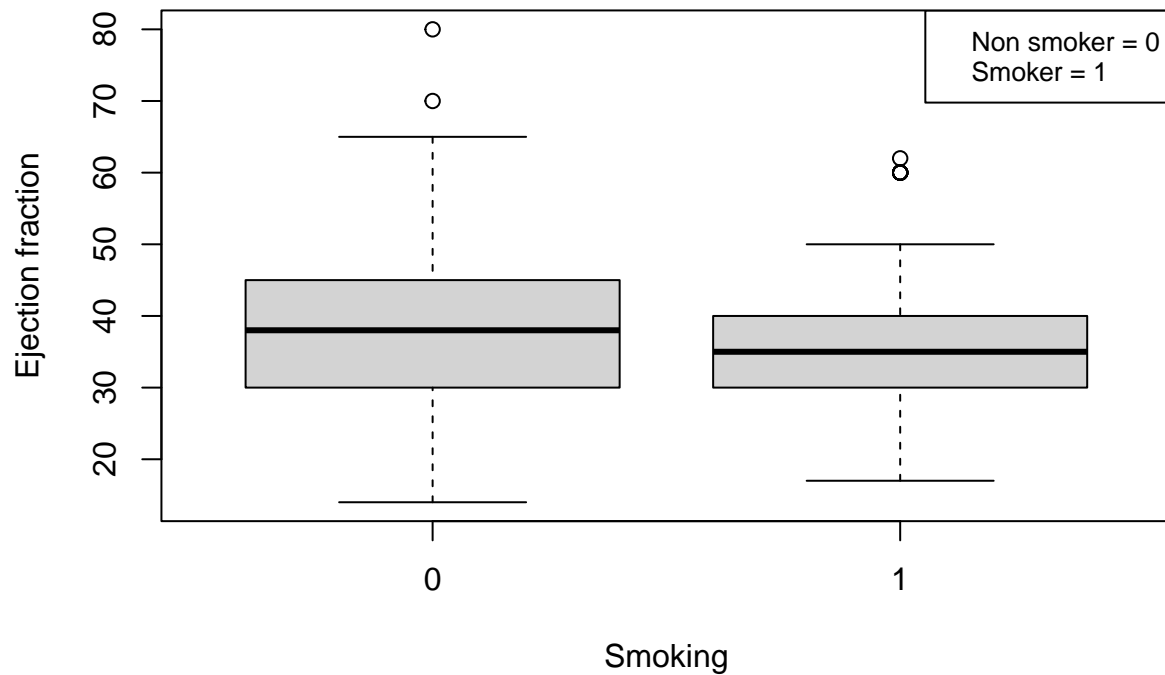
par(mfrow=c(1,2))
# Barplot for males with percentages
barplot(x_percentage, beside = TRUE, col = c("lightgreen", "red"),
        legend = rownames(y_percentage), main = "Diabetes in Males",
        xlab = "Diabetes Status", ylab = "Percentage")
# Barplot for females with percentages
barplot(y_percentage, beside = TRUE, col = c("lightgreen", "red"),
        legend = rownames(y_percentage), main = "Diabetes in Females",
        xlab = "Diabetes Status", ylab = "Percentage")
```



Among male patients around 35% had diabetes and among female around 52% had diabetes. That may mean that females are more prone to diabetes.

```
#Smoking vs Ejection fraction
boxplot(df$ejecction_fraction ~ df$smoking, main = "Smoking vs Ejection fraction", xlab = "Smoking", ylab = "Ejection fraction", legend = c("Non smoker = 0", "Smoker = 1"), cex=0.8)
```

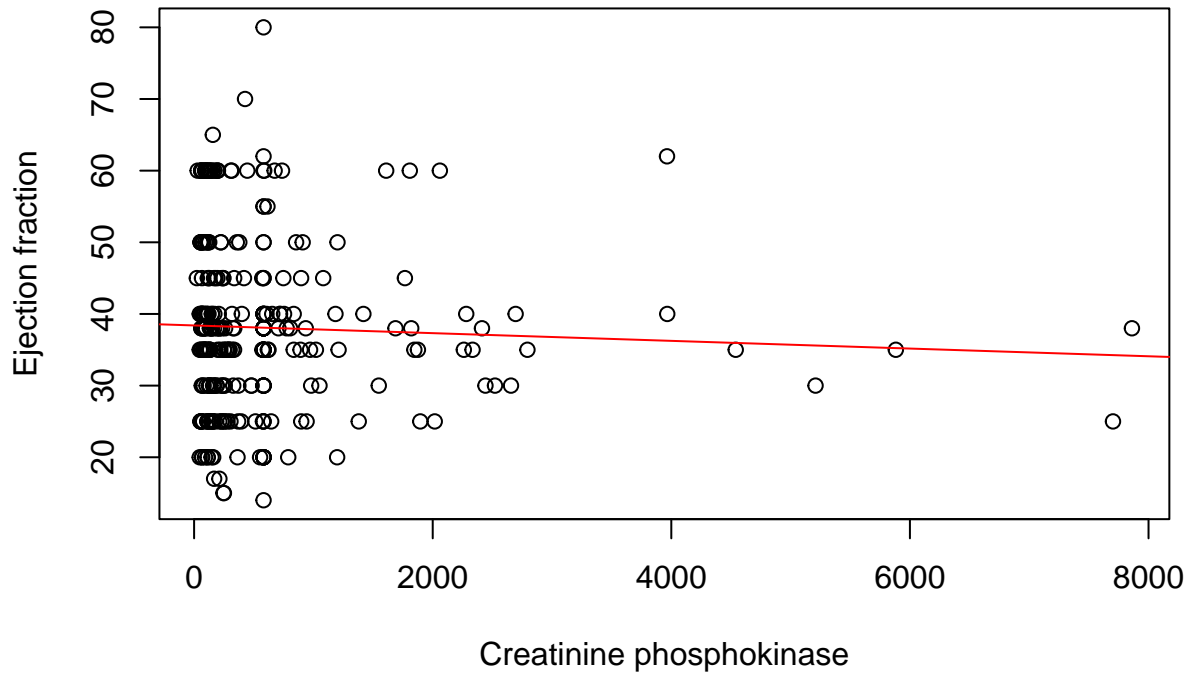

Smoking vs Ejection fraction



Smoker patients had low value of ejection fraction, meaning that smoking may cause decrease in the pumping capacity of heart.

```
#Creatinine phosphokinase vs Ejection fraction  
plot(x=df$creatinine_phosphokinase,y=df$ejection_fraction,main="Creatinine phosphokinase vs Ejection fr  
abline(lm(df$ejection_fraction~df$creatinine_phosphokinase),col="red")
```

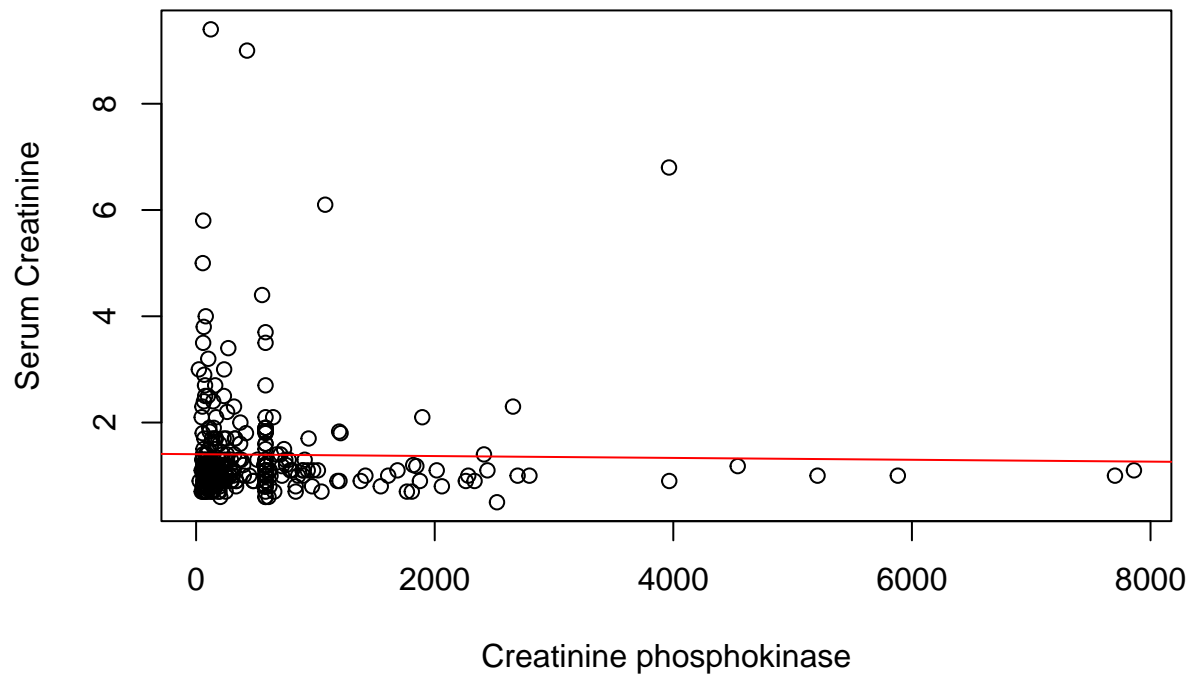
Creatinine phosphokinase vs Ejection fraction



We can see from the graph that as the CPK level increases ejection fraction decreases representing that blood pumping capacity of heart decrease whenever heart injury occurs.

```
#Creatinine phosphokinase vs Serum Creatinine  
plot(x=df$creatinine_phosphokinase,y=df$serum_creatinine,main="Creatinine phosphokinase vs Serum Creatinine",col="green")  
abline(lm(df$serum_creatinine~df$creatinine_phosphokinase),col="red")
```

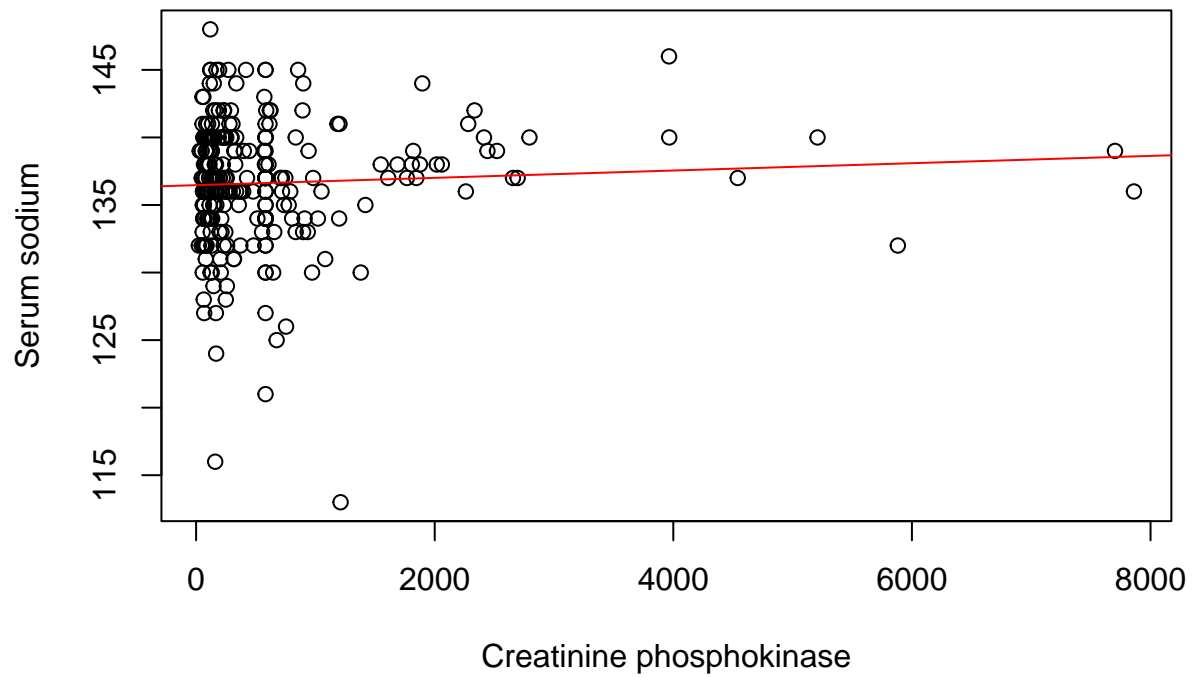
Creatinine phosphokinase vs Serum Creatinine



```
#Creatinine phosphokinase vs Serum sodium
```

```
plot(x=df$creatinine_phosphokinase,y=df$serum_sodium,main="Creatinine phosphokinase vs Serum sodium",xlab="Creatinine phosphokinase",ylab="Serum sodium",col="black",pch="o",lty="n")  
abline(lm(df$serum_sodium~df$creatinine_phosphokinase),col="red")
```

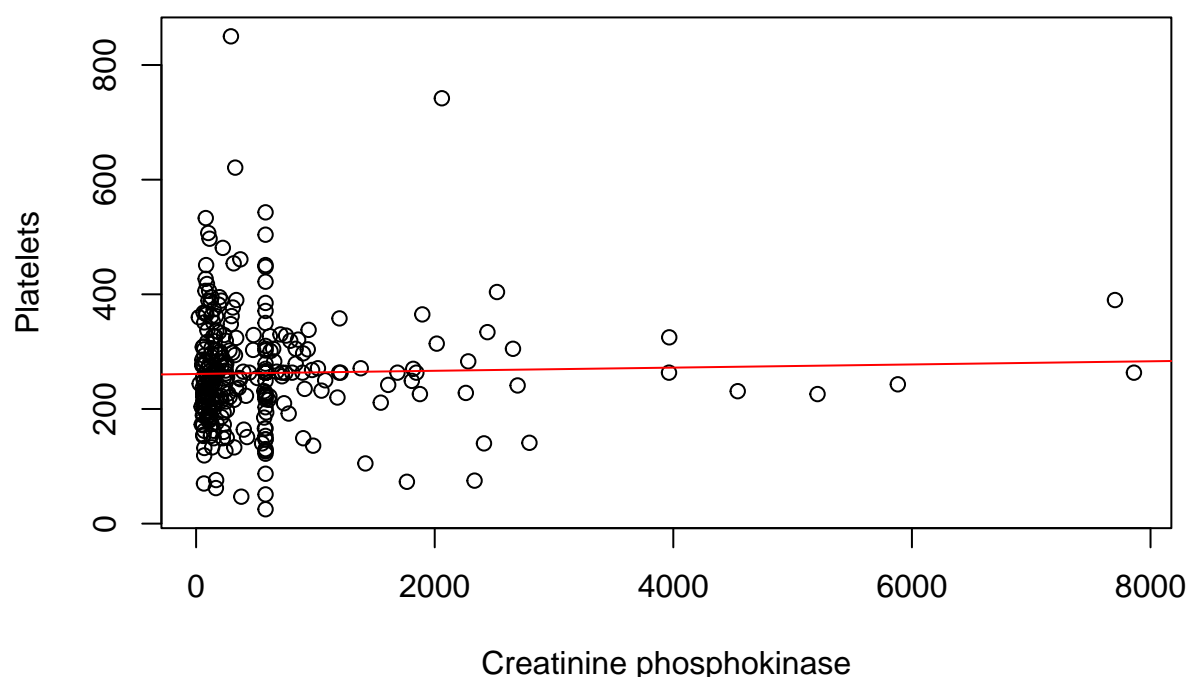
Creatinine phosphokinase vs Serum sodium



```
#Creatinine phosphokinase vs Platelets
```

```
plot(x=df$creatinine_phosphokinase,y=df$platelets,main="Creatinine phosphokinase vs Platelets",xlab="Cr  
abline(lm(df$platelets~df$creatinine_phosphokinase),col="red")
```

Creatinine phosphokinase vs Platelets



Key findings: 1) Ejection fraction, Serum sodium and Serum creatinine can be considered as important factors to predict survival of patients. 2) Females are more prone to anaemia and diabetes than males. 3) Smoking can cause decrease in pumping capacity of heart. 4) Presence of CPK enzyme may be considered for detection of heart injury.

Splitting Dataset into training and testing data

```
train_indices <- sample(1:nrow(df), 0.75 * nrow(df), replace = F)
train_data <- df[train_indices, ]
test_data <- df[-train_indices, ]
train_data
```

| ## | age | anaemia | creatinine_phosphokinase | diabetes | ejection_fraction |
|--------|-----|---------|--------------------------|----------|-------------------|
| ## 111 | 85 | 0 | 129 | 0 | 60 |
| ## 48 | 60 | 0 | 582 | 1 | 38 |
| ## 229 | 65 | 0 | 56 | 0 | 25 |
| ## 259 | 45 | 1 | 66 | 1 | 25 |
| ## 78 | 42 | 0 | 102 | 1 | 40 |
| ## 170 | 70 | 0 | 835 | 0 | 35 |
| ## 138 | 68 | 1 | 646 | 0 | 25 |
| ## 226 | 75 | 0 | 675 | 1 | 60 |
| ## 236 | 77 | 1 | 109 | 0 | 50 |
| ## 230 | 72 | 0 | 211 | 0 | 25 |
| ## 206 | 50 | 1 | 167 | 1 | 45 |
| ## 285 | 50 | 1 | 54 | 0 | 40 |
| ## 185 | 58 | 1 | 145 | 0 | 25 |

| | | | | | |
|--------|----|---|------|---|----|
| ## 197 | 45 | 0 | 582 | 1 | 38 |
| ## 40 | 60 | 0 | 235 | 1 | 38 |
| ## 98 | 70 | 1 | 59 | 0 | 60 |
| ## 205 | 78 | 1 | 64 | 0 | 40 |
| ## 280 | 55 | 0 | 84 | 1 | 38 |
| ## 270 | 40 | 0 | 582 | 1 | 35 |
| ## 9 | 65 | 0 | 157 | 0 | 65 |
| ## 210 | 49 | 0 | 972 | 1 | 35 |
| ## 265 | 61 | 0 | 582 | 1 | 38 |
| ## 103 | 80 | 0 | 898 | 0 | 25 |
| ## 80 | 55 | 0 | 336 | 0 | 45 |
| ## 36 | 69 | 0 | 582 | 1 | 35 |
| ## 261 | 55 | 0 | 66 | 0 | 40 |
| ## 11 | 75 | 1 | 81 | 0 | 38 |
| ## 212 | 50 | 0 | 582 | 0 | 62 |
| ## 153 | 50 | 0 | 115 | 0 | 45 |
| ## 288 | 45 | 0 | 582 | 1 | 55 |
| ## 41 | 70 | 0 | 582 | 0 | 20 |
| ## 188 | 60 | 0 | 1896 | 1 | 25 |
| ## 174 | 50 | 1 | 115 | 0 | 20 |
| ## 14 | 50 | 1 | 168 | 0 | 38 |
| ## 239 | 65 | 1 | 720 | 1 | 40 |
| ## 156 | 60 | 1 | 231 | 1 | 25 |
| ## 152 | 62 | 0 | 30 | 1 | 60 |
| ## 44 | 72 | 0 | 127 | 1 | 50 |
| ## 86 | 51 | 0 | 78 | 0 | 50 |
| ## 211 | 70 | 0 | 212 | 1 | 17 |
| ## 220 | 55 | 0 | 582 | 1 | 35 |
| ## 282 | 70 | 0 | 582 | 0 | 40 |
| ## 108 | 45 | 1 | 1876 | 1 | 35 |
| ## 17 | 87 | 1 | 149 | 0 | 38 |
| ## 70 | 65 | 0 | 113 | 1 | 25 |
| ## 268 | 56 | 1 | 135 | 1 | 38 |
| ## 25 | 75 | 0 | 582 | 1 | 30 |
| ## 100 | 63 | 1 | 61 | 1 | 40 |
| ## 297 | 45 | 0 | 2060 | 1 | 60 |
| ## 207 | 40 | 1 | 101 | 0 | 40 |
| ## 298 | 45 | 0 | 2413 | 0 | 38 |
| ## 74 | 65 | 0 | 224 | 1 | 50 |
| ## 159 | 85 | 1 | 910 | 0 | 50 |
| ## 42 | 50 | 0 | 124 | 1 | 30 |
| ## 296 | 55 | 0 | 1820 | 0 | 38 |
| ## 53 | 60 | 0 | 3964 | 1 | 62 |
| ## 1 | 75 | 0 | 582 | 0 | 20 |
| ## 221 | 73 | 0 | 582 | 0 | 20 |
| ## 187 | 50 | 0 | 582 | 0 | 50 |
| ## 99 | 60 | 1 | 156 | 1 | 25 |
| ## 96 | 58 | 1 | 133 | 0 | 60 |
| ## 73 | 85 | 0 | 5882 | 0 | 35 |
| ## 45 | 60 | 1 | 588 | 1 | 60 |
| ## 72 | 58 | 0 | 582 | 1 | 35 |
| ## 178 | 49 | 1 | 69 | 0 | 50 |
| ## 274 | 42 | 0 | 64 | 0 | 40 |
| ## 284 | 65 | 0 | 1688 | 0 | 38 |

| | | | | | |
|--------|----|---|------|---|----|
| ## 233 | 40 | 1 | 129 | 0 | 35 |
| ## 271 | 44 | 0 | 582 | 1 | 30 |
| ## 64 | 45 | 0 | 582 | 0 | 35 |
| ## 281 | 70 | 0 | 2695 | 1 | 40 |
| ## 235 | 53 | 1 | 582 | 0 | 45 |
| ## 147 | 52 | 0 | 132 | 0 | 30 |
| ## 124 | 60 | 1 | 582 | 0 | 30 |
| ## 33 | 50 | 1 | 249 | 1 | 35 |
| ## 276 | 45 | 0 | 582 | 0 | 38 |
| ## 175 | 65 | 0 | 198 | 1 | 35 |
| ## 149 | 75 | 1 | 582 | 0 | 30 |
| ## 47 | 51 | 0 | 1380 | 0 | 25 |
| ## 266 | 50 | 1 | 298 | 0 | 35 |
| ## 137 | 65 | 1 | 59 | 1 | 60 |
| ## 267 | 55 | 0 | 1199 | 0 | 20 |
| ## 57 | 70 | 1 | 75 | 0 | 35 |
| ## 102 | 75 | 0 | 582 | 0 | 45 |
| ## 262 | 62 | 1 | 655 | 0 | 40 |
| ## 208 | 85 | 0 | 212 | 0 | 38 |
| ## 49 | 80 | 1 | 553 | 0 | 20 |
| ## 141 | 80 | 0 | 805 | 0 | 38 |
| ## 232 | 70 | 0 | 93 | 0 | 35 |
| ## 37 | 90 | 1 | 60 | 1 | 50 |
| ## 68 | 72 | 1 | 110 | 0 | 25 |
| ## 77 | 70 | 0 | 92 | 0 | 60 |
| ## 143 | 50 | 0 | 482 | 1 | 30 |
| ## 51 | 68 | 1 | 577 | 0 | 25 |
| ## 290 | 90 | 1 | 337 | 0 | 38 |
| ## 162 | 45 | 1 | 130 | 0 | 35 |
| ## 81 | 70 | 0 | 69 | 0 | 40 |
| ## 157 | 52 | 1 | 58 | 0 | 35 |
| ## 249 | 40 | 0 | 624 | 0 | 35 |
| ## 58 | 60 | 1 | 607 | 0 | 40 |
| ## 248 | 64 | 0 | 143 | 0 | 25 |
| ## 244 | 73 | 1 | 1185 | 0 | 40 |
| ## 5 | 65 | 1 | 160 | 1 | 20 |
| ## 139 | 62 | 0 | 281 | 1 | 35 |
| ## 2 | 55 | 0 | 7861 | 0 | 38 |
| ## 129 | 61 | 0 | 248 | 0 | 30 |
| ## 105 | 60 | 0 | 53 | 0 | 50 |
| ## 134 | 63 | 0 | 193 | 0 | 60 |
| ## 200 | 60 | 0 | 1211 | 1 | 35 |
| ## 264 | 68 | 1 | 157 | 1 | 60 |
| ## 287 | 60 | 0 | 253 | 0 | 35 |
| ## 255 | 53 | 1 | 446 | 0 | 60 |
| ## 84 | 79 | 1 | 55 | 0 | 50 |
| ## 299 | 50 | 0 | 196 | 0 | 45 |
| ## 112 | 55 | 0 | 60 | 0 | 35 |
| ## 18 | 45 | 0 | 582 | 0 | 14 |
| ## 136 | 75 | 0 | 582 | 0 | 40 |
| ## 158 | 50 | 0 | 250 | 0 | 25 |
| ## 195 | 45 | 0 | 582 | 0 | 20 |
| ## 28 | 70 | 0 | 122 | 1 | 45 |
| ## 250 | 53 | 0 | 207 | 1 | 40 |

| | | | | | |
|--------|----|---|------|---|----|
| ## 180 | 55 | 0 | 835 | 0 | 40 |
| ## 12 | 62 | 0 | 231 | 0 | 25 |
| ## 75 | 69 | 0 | 582 | 0 | 20 |
| ## 161 | 66 | 1 | 72 | 0 | 40 |
| ## 272 | 51 | 0 | 582 | 1 | 40 |
| ## 140 | 50 | 0 | 1548 | 0 | 30 |
| ## 198 | 65 | 0 | 167 | 0 | 30 |
| ## 31 | 94 | 0 | 582 | 1 | 38 |
| ## 62 | 50 | 0 | 318 | 0 | 40 |
| ## 173 | 70 | 1 | 171 | 0 | 60 |
| ## 181 | 40 | 0 | 478 | 1 | 30 |
| ## 107 | 55 | 0 | 748 | 0 | 45 |
| ## 209 | 60 | 1 | 2281 | 1 | 40 |
| ## 121 | 60 | 1 | 737 | 0 | 60 |
| ## 21 | 65 | 1 | 52 | 0 | 25 |
| ## 55 | 60 | 1 | 260 | 1 | 38 |
| ## 128 | 58 | 1 | 200 | 1 | 60 |
| ## 204 | 60 | 0 | 59 | 0 | 25 |
| ## 16 | 82 | 1 | 379 | 0 | 50 |
| ## 3 | 65 | 0 | 146 | 0 | 20 |
| ## 177 | 69 | 0 | 1419 | 0 | 40 |
| ## 85 | 59 | 1 | 280 | 1 | 25 |
| ## 6 | 90 | 1 | 47 | 0 | 40 |
| ## 63 | 55 | 0 | 109 | 0 | 35 |
| ## 241 | 70 | 0 | 81 | 1 | 35 |
| ## 127 | 46 | 0 | 168 | 1 | 17 |
| ## 258 | 58 | 0 | 132 | 1 | 38 |
| ## 247 | 55 | 0 | 2017 | 0 | 25 |
| ## 256 | 52 | 1 | 191 | 1 | 30 |
| ## 172 | 52 | 0 | 3966 | 0 | 40 |
| ## 123 | 60 | 0 | 96 | 1 | 38 |
| ## 283 | 42 | 0 | 64 | 0 | 30 |
| ## 231 | 60 | 0 | 166 | 0 | 30 |
| ## 148 | 64 | 0 | 1610 | 0 | 60 |
| ## 186 | 60 | 1 | 104 | 1 | 30 |
| ## 183 | 65 | 0 | 395 | 1 | 25 |
| ## 67 | 42 | 1 | 250 | 1 | 15 |
| ## 279 | 50 | 1 | 1051 | 1 | 30 |
| ## 234 | 53 | 1 | 707 | 0 | 38 |
| ## 66 | 60 | 0 | 68 | 0 | 20 |
| ## 93 | 42 | 0 | 582 | 0 | 60 |
| ## 39 | 60 | 0 | 2656 | 1 | 30 |
| ## 146 | 50 | 0 | 185 | 0 | 30 |
| ## 192 | 64 | 1 | 62 | 0 | 60 |
| ## 245 | 54 | 0 | 582 | 1 | 38 |
| ## 46 | 50 | 0 | 582 | 1 | 38 |
| ## 27 | 95 | 1 | 112 | 0 | 40 |
| ## 20 | 48 | 1 | 582 | 1 | 55 |
| ## 110 | 45 | 0 | 292 | 1 | 35 |
| ## 109 | 63 | 0 | 936 | 0 | 38 |
| ## 60 | 72 | 0 | 364 | 1 | 20 |
| ## 131 | 53 | 1 | 1808 | 0 | 60 |
| ## 165 | 45 | 0 | 2442 | 1 | 30 |
| ## 160 | 59 | 1 | 129 | 0 | 45 |

| | | | | | | |
|----|---|----|---------|------|-----|-----|
| ## | 130 | 53 | 1 | 270 | 1 | 35 |
| ## | 257 | 65 | 0 | 326 | 0 | 38 |
| ## | 224 | 47 | 0 | 582 | 0 | 25 |
| ## | 168 | 59 | 0 | 66 | 1 | 20 |
| ## | 218 | 54 | 1 | 427 | 0 | 70 |
| ## | 71 | 41 | 0 | 148 | 0 | 40 |
| ## | 89 | 44 | 0 | 84 | 1 | 40 |
| ## | 227 | 58 | 1 | 57 | 0 | 25 |
| ## | 125 | 60 | 0 | 582 | 0 | 40 |
| ## | 194 | 73 | 1 | 231 | 1 | 30 |
| ## | 213 | 78 | 0 | 224 | 0 | 50 |
| ## | 166 | 80 | 0 | 776 | 1 | 38 |
| ## | 199 | 50 | 1 | 582 | 1 | 20 |
| ## | 294 | 63 | 1 | 103 | 1 | 35 |
| ## | 69 | 70 | 0 | 161 | 0 | 25 |
| ## | 29 | 58 | 1 | 60 | 0 | 38 |
| ## | 133 | 46 | 0 | 719 | 0 | 40 |
| ## | 118 | 85 | 1 | 102 | 0 | 60 |
| ## | 24 | 53 | 0 | 63 | 1 | 60 |
| ## | 13 | 45 | 1 | 981 | 0 | 30 |
| ## | 94 | 60 | 1 | 154 | 0 | 25 |
| ## | 26 | 80 | 0 | 148 | 1 | 38 |
| ## | 50 | 57 | 1 | 129 | 0 | 30 |
| ## | 176 | 60 | 1 | 95 | 0 | 60 |
| ## | 151 | 72 | 0 | 233 | 0 | 45 |
| ## | 115 | 60 | 1 | 754 | 1 | 40 |
| ## | 76 | 60 | 1 | 47 | 0 | 20 |
| ## | 182 | 59 | 1 | 176 | 1 | 25 |
| ## | 286 | 55 | 1 | 170 | 1 | 40 |
| ## | 92 | 60 | 0 | 897 | 1 | 45 |
| ## | 32 | 85 | 0 | 23 | 0 | 45 |
| ## | 38 | 82 | 1 | 855 | 1 | 50 |
| ## | 120 | 86 | 0 | 582 | 0 | 38 |
| ## | 116 | 58 | 1 | 400 | 0 | 40 |
| ## | 242 | 65 | 0 | 582 | 1 | 30 |
| ## | 190 | 40 | 0 | 244 | 0 | 45 |
| ## | 223 | 42 | 1 | 86 | 0 | 35 |
| ## | 201 | 63 | 1 | 1767 | 0 | 45 |
| ## | 10 | 80 | 1 | 123 | 0 | 35 |
| ## | 167 | 53 | 0 | 196 | 0 | 60 |
| ## | 228 | 55 | 1 | 2794 | 0 | 35 |
| ## | 88 | 65 | 1 | 68 | 1 | 60 |
| ## | 135 | 81 | 0 | 4540 | 0 | 35 |
| ## | 19 | 70 | 1 | 125 | 0 | 25 |
| ## | 15 | 49 | 1 | 80 | 0 | 30 |
| ## | 101 | 65 | 1 | 305 | 0 | 25 |
| ## | 292 | 60 | 0 | 320 | 0 | 35 |
| ## | 132 | 60 | 1 | 1082 | 1 | 45 |
| ## | 196 | 77 | 1 | 418 | 0 | 45 |
| ## | high_blood_pressure platelets serum_creatinine serum_sodium sex smoking | | | | | |
| ## | 111 | 0 | 306.000 | 1.20 | 132 | 1 1 |
| ## | 48 | 1 | 451.000 | 0.60 | 138 | 1 1 |
| ## | 229 | 0 | 237.000 | 5.00 | 130 | 0 0 |
| ## | 259 | 0 | 233.000 | 0.80 | 135 | 1 0 |

| | | | | | | |
|--------|---|---------|------|-----|---|---|
| ## 78 | 0 | 237.000 | 1.20 | 140 | 1 | 0 |
| ## 170 | 1 | 305.000 | 0.80 | 133 | 0 | 0 |
| ## 138 | 0 | 305.000 | 2.10 | 130 | 1 | 0 |
| ## 226 | 0 | 265.000 | 1.40 | 125 | 0 | 0 |
| ## 236 | 1 | 406.000 | 1.10 | 137 | 1 | 0 |
| ## 230 | 0 | 274.000 | 1.20 | 134 | 0 | 0 |
| ## 206 | 0 | 362.000 | 1.00 | 136 | 0 | 0 |
| ## 285 | 0 | 279.000 | 0.80 | 141 | 1 | 0 |
| ## 185 | 0 | 219.000 | 1.20 | 137 | 1 | 1 |
| ## 197 | 1 | 263.358 | 1.18 | 137 | 0 | 0 |
| ## 40 | 0 | 329.000 | 3.00 | 142 | 0 | 0 |
| ## 98 | 0 | 255.000 | 1.10 | 136 | 0 | 0 |
| ## 205 | 0 | 277.000 | 0.70 | 137 | 1 | 1 |
| ## 280 | 0 | 451.000 | 1.30 | 136 | 0 | 0 |
| ## 270 | 0 | 222.000 | 1.00 | 132 | 1 | 0 |
| ## 9 | 0 | 263.358 | 1.50 | 138 | 0 | 0 |
| ## 210 | 1 | 268.000 | 0.80 | 130 | 0 | 0 |
| ## 265 | 0 | 147.000 | 1.20 | 141 | 1 | 0 |
| ## 103 | 0 | 149.000 | 1.10 | 144 | 1 | 1 |
| ## 80 | 1 | 324.000 | 0.90 | 140 | 0 | 0 |
| ## 36 | 0 | 228.000 | 3.50 | 134 | 1 | 0 |
| ## 261 | 0 | 203.000 | 1.00 | 138 | 1 | 0 |
| ## 11 | 1 | 368.000 | 4.00 | 131 | 1 | 1 |
| ## 212 | 1 | 147.000 | 0.80 | 140 | 1 | 1 |
| ## 153 | 1 | 184.000 | 0.90 | 134 | 1 | 1 |
| ## 288 | 0 | 543.000 | 1.00 | 132 | 0 | 0 |
| ## 41 | 1 | 263.358 | 1.83 | 134 | 1 | 1 |
| ## 188 | 0 | 365.000 | 2.10 | 144 | 0 | 0 |
| ## 174 | 0 | 189.000 | 0.80 | 139 | 1 | 0 |
| ## 14 | 1 | 76.000 | 1.10 | 137 | 1 | 0 |
| ## 239 | 0 | 257.000 | 1.00 | 136 | 0 | 0 |
| ## 156 | 0 | 194.000 | 1.70 | 140 | 1 | 0 |
| ## 152 | 1 | 244.000 | 0.90 | 139 | 1 | 0 |
| ## 44 | 1 | 218.000 | 1.00 | 134 | 1 | 0 |
| ## 86 | 0 | 406.000 | 0.70 | 140 | 1 | 0 |
| ## 211 | 1 | 389.000 | 1.00 | 136 | 1 | 1 |
| ## 220 | 1 | 371.000 | 0.70 | 140 | 0 | 0 |
| ## 282 | 0 | 51.000 | 2.70 | 136 | 1 | 1 |
| ## 108 | 0 | 226.000 | 0.90 | 138 | 1 | 0 |
| ## 17 | 0 | 262.000 | 0.90 | 140 | 1 | 0 |
| ## 70 | 0 | 497.000 | 1.83 | 135 | 1 | 0 |
| ## 268 | 0 | 133.000 | 1.70 | 140 | 1 | 0 |
| ## 25 | 1 | 263.358 | 1.83 | 134 | 0 | 0 |
| ## 100 | 0 | 221.000 | 1.10 | 140 | 0 | 0 |
| ## 297 | 0 | 742.000 | 0.80 | 138 | 0 | 0 |
| ## 207 | 0 | 226.000 | 0.80 | 141 | 0 | 0 |
| ## 298 | 0 | 140.000 | 1.40 | 140 | 1 | 1 |
| ## 74 | 0 | 149.000 | 1.30 | 137 | 1 | 1 |
| ## 159 | 0 | 235.000 | 1.30 | 134 | 1 | 0 |
| ## 42 | 1 | 153.000 | 1.20 | 136 | 0 | 1 |
| ## 296 | 0 | 270.000 | 1.20 | 139 | 0 | 0 |
| ## 53 | 0 | 263.358 | 6.80 | 146 | 0 | 0 |
| ## 1 | 1 | 265.000 | 1.90 | 130 | 1 | 0 |
| ## 221 | 0 | 263.358 | 1.83 | 134 | 1 | 0 |

| | | | | | | |
|--------|---|---------|------|-----|---|---|
| ## 187 | 0 | 153.000 | 0.60 | 134 | 0 | 0 |
| ## 99 | 1 | 318.000 | 1.20 | 137 | 0 | 0 |
| ## 96 | 1 | 219.000 | 1.00 | 141 | 1 | 0 |
| ## 73 | 0 | 243.000 | 1.00 | 132 | 1 | 1 |
| ## 45 | 0 | 194.000 | 1.10 | 142 | 0 | 0 |
| ## 72 | 0 | 122.000 | 0.90 | 139 | 1 | 1 |
| ## 178 | 0 | 132.000 | 1.00 | 140 | 0 | 0 |
| ## 274 | 0 | 189.000 | 0.70 | 140 | 1 | 0 |
| ## 284 | 0 | 263.358 | 1.10 | 138 | 1 | 1 |
| ## 233 | 0 | 255.000 | 0.90 | 137 | 1 | 0 |
| ## 271 | 1 | 263.358 | 1.60 | 130 | 1 | 1 |
| ## 64 | 0 | 385.000 | 1.00 | 145 | 1 | 0 |
| ## 281 | 0 | 241.000 | 1.00 | 137 | 1 | 0 |
| ## 235 | 0 | 305.000 | 1.10 | 137 | 1 | 1 |
| ## 147 | 0 | 218.000 | 0.70 | 136 | 1 | 1 |
| ## 124 | 1 | 127.000 | 0.90 | 145 | 0 | 0 |
| ## 33 | 1 | 319.000 | 1.00 | 128 | 0 | 0 |
| ## 276 | 1 | 422.000 | 0.80 | 137 | 0 | 0 |
| ## 175 | 1 | 281.000 | 0.90 | 137 | 1 | 1 |
| ## 149 | 0 | 225.000 | 1.83 | 134 | 1 | 0 |
| ## 47 | 1 | 271.000 | 0.90 | 130 | 1 | 0 |
| ## 266 | 0 | 362.000 | 0.90 | 140 | 1 | 1 |
| ## 137 | 0 | 172.000 | 0.90 | 137 | 0 | 0 |
| ## 267 | 0 | 263.358 | 1.83 | 134 | 1 | 1 |
| ## 57 | 0 | 223.000 | 2.70 | 138 | 1 | 1 |
| ## 102 | 1 | 263.358 | 1.18 | 137 | 1 | 0 |
| ## 262 | 0 | 283.000 | 0.70 | 133 | 0 | 0 |
| ## 208 | 0 | 186.000 | 0.90 | 136 | 1 | 0 |
| ## 49 | 1 | 140.000 | 4.40 | 133 | 1 | 0 |
| ## 141 | 0 | 263.358 | 1.10 | 134 | 1 | 0 |
| ## 232 | 0 | 185.000 | 1.10 | 134 | 1 | 1 |
| ## 37 | 0 | 226.000 | 1.00 | 134 | 1 | 0 |
| ## 68 | 0 | 274.000 | 1.00 | 140 | 1 | 1 |
| ## 77 | 1 | 317.000 | 0.80 | 140 | 0 | 1 |
| ## 143 | 0 | 329.000 | 0.90 | 132 | 0 | 0 |
| ## 51 | 1 | 166.000 | 1.00 | 138 | 1 | 0 |
| ## 290 | 0 | 390.000 | 0.90 | 144 | 0 | 0 |
| ## 162 | 0 | 174.000 | 0.80 | 139 | 1 | 1 |
| ## 81 | 0 | 293.000 | 1.70 | 136 | 0 | 0 |
| ## 157 | 0 | 277.000 | 1.40 | 136 | 0 | 0 |
| ## 249 | 0 | 301.000 | 1.00 | 142 | 1 | 1 |
| ## 58 | 0 | 216.000 | 0.60 | 138 | 1 | 1 |
| ## 248 | 0 | 246.000 | 2.40 | 135 | 1 | 0 |
| ## 244 | 1 | 220.000 | 0.90 | 141 | 0 | 0 |
| ## 5 | 0 | 327.000 | 2.70 | 116 | 0 | 0 |
| ## 139 | 0 | 221.000 | 1.00 | 136 | 0 | 0 |
| ## 2 | 0 | 263.358 | 1.10 | 136 | 1 | 0 |
| ## 129 | 1 | 267.000 | 0.70 | 136 | 1 | 1 |
| ## 105 | 1 | 286.000 | 2.30 | 143 | 0 | 0 |
| ## 134 | 1 | 295.000 | 1.30 | 145 | 1 | 1 |
| ## 200 | 0 | 263.358 | 1.80 | 113 | 1 | 1 |
| ## 264 | 0 | 208.000 | 1.00 | 140 | 0 | 0 |
| ## 287 | 0 | 279.000 | 1.70 | 140 | 1 | 0 |
| ## 255 | 1 | 263.358 | 1.00 | 139 | 1 | 0 |

| | | | | | | |
|--------|---|---------|------|-----|---|---|
| ## 84 | 1 | 172.000 | 1.80 | 133 | 1 | 0 |
| ## 299 | 0 | 395.000 | 1.60 | 136 | 1 | 1 |
| ## 112 | 0 | 228.000 | 1.20 | 135 | 1 | 1 |
| ## 18 | 0 | 166.000 | 0.80 | 127 | 1 | 0 |
| ## 136 | 0 | 263.358 | 1.18 | 137 | 1 | 0 |
| ## 158 | 0 | 262.000 | 1.00 | 136 | 1 | 1 |
| ## 195 | 1 | 126.000 | 1.60 | 135 | 1 | 0 |
| ## 28 | 1 | 284.000 | 1.30 | 136 | 1 | 1 |
| ## 250 | 0 | 223.000 | 1.20 | 130 | 0 | 0 |
| ## 180 | 0 | 279.000 | 0.70 | 140 | 1 | 1 |
| ## 12 | 1 | 253.000 | 0.90 | 140 | 1 | 1 |
| ## 75 | 0 | 266.000 | 1.20 | 134 | 1 | 1 |
| ## 161 | 1 | 242.000 | 1.20 | 134 | 1 | 0 |
| ## 272 | 0 | 221.000 | 0.90 | 134 | 0 | 0 |
| ## 140 | 1 | 211.000 | 0.80 | 138 | 1 | 0 |
| ## 198 | 0 | 259.000 | 0.80 | 138 | 0 | 0 |
| ## 31 | 1 | 263.358 | 1.83 | 134 | 1 | 0 |
| ## 62 | 1 | 216.000 | 2.30 | 131 | 0 | 0 |
| ## 173 | 1 | 176.000 | 1.10 | 145 | 1 | 1 |
| ## 181 | 0 | 303.000 | 0.90 | 136 | 1 | 0 |
| ## 107 | 0 | 263.000 | 1.30 | 137 | 1 | 0 |
| ## 209 | 0 | 283.000 | 1.00 | 141 | 0 | 0 |
| ## 121 | 1 | 210.000 | 1.50 | 135 | 1 | 1 |
| ## 21 | 1 | 276.000 | 1.30 | 137 | 0 | 0 |
| ## 55 | 0 | 255.000 | 2.20 | 132 | 0 | 1 |
| ## 128 | 0 | 300.000 | 0.80 | 137 | 0 | 0 |
| ## 204 | 1 | 212.000 | 3.50 | 136 | 1 | 1 |
| ## 16 | 0 | 47.000 | 1.30 | 136 | 1 | 0 |
| ## 3 | 0 | 162.000 | 1.30 | 129 | 1 | 1 |
| ## 177 | 0 | 105.000 | 1.00 | 135 | 1 | 1 |
| ## 85 | 1 | 302.000 | 1.00 | 141 | 0 | 0 |
| ## 6 | 1 | 204.000 | 2.10 | 132 | 1 | 1 |
| ## 63 | 0 | 254.000 | 1.10 | 139 | 1 | 1 |
| ## 241 | 1 | 533.000 | 1.30 | 139 | 0 | 0 |
| ## 127 | 1 | 271.000 | 2.10 | 124 | 0 | 0 |
| ## 258 | 1 | 253.000 | 1.00 | 139 | 1 | 0 |
| ## 247 | 0 | 314.000 | 1.10 | 138 | 1 | 0 |
| ## 256 | 1 | 334.000 | 1.00 | 142 | 1 | 1 |
| ## 172 | 0 | 325.000 | 0.90 | 140 | 1 | 1 |
| ## 123 | 0 | 228.000 | 0.75 | 140 | 0 | 0 |
| ## 283 | 0 | 215.000 | 3.80 | 128 | 1 | 1 |
| ## 231 | 0 | 62.000 | 1.70 | 127 | 0 | 0 |
| ## 148 | 0 | 242.000 | 1.00 | 137 | 1 | 0 |
| ## 186 | 0 | 389.000 | 1.50 | 136 | 1 | 0 |
| ## 183 | 0 | 265.000 | 1.20 | 136 | 1 | 1 |
| ## 67 | 0 | 213.000 | 1.30 | 136 | 0 | 0 |
| ## 279 | 0 | 232.000 | 0.70 | 136 | 0 | 0 |
| ## 234 | 0 | 330.000 | 1.40 | 137 | 1 | 1 |
| ## 66 | 0 | 119.000 | 2.90 | 127 | 1 | 1 |
| ## 93 | 0 | 263.358 | 1.18 | 137 | 0 | 0 |
| ## 39 | 0 | 305.000 | 2.30 | 137 | 1 | 0 |
| ## 146 | 0 | 266.000 | 0.70 | 141 | 1 | 1 |
| ## 192 | 0 | 309.000 | 1.50 | 135 | 0 | 0 |
| ## 245 | 0 | 264.000 | 1.80 | 134 | 1 | 0 |

| | | | | | | |
|--------|---|---------|------|-----|---|---|
| ## 46 | 0 | 310.000 | 1.90 | 135 | 1 | 1 |
| ## 27 | 1 | 196.000 | 1.00 | 138 | 0 | 0 |
| ## 20 | 0 | 87.000 | 1.90 | 121 | 0 | 0 |
| ## 110 | 0 | 850.000 | 1.30 | 142 | 1 | 1 |
| ## 109 | 0 | 304.000 | 1.10 | 133 | 1 | 1 |
| ## 60 | 1 | 254.000 | 1.30 | 136 | 1 | 1 |
| ## 131 | 1 | 249.000 | 0.70 | 138 | 1 | 1 |
| ## 165 | 0 | 334.000 | 1.10 | 139 | 1 | 0 |
| ## 160 | 1 | 362.000 | 1.10 | 139 | 1 | 1 |
| ## 130 | 0 | 227.000 | 3.40 | 145 | 1 | 0 |
| ## 257 | 0 | 294.000 | 1.70 | 139 | 0 | 0 |
| ## 224 | 0 | 130.000 | 0.80 | 134 | 1 | 0 |
| ## 168 | 0 | 70.000 | 2.40 | 134 | 1 | 0 |
| ## 218 | 1 | 151.000 | 9.00 | 137 | 0 | 0 |
| ## 71 | 0 | 374.000 | 0.80 | 140 | 1 | 1 |
| ## 89 | 1 | 235.000 | 0.70 | 139 | 1 | 0 |
| ## 227 | 0 | 189.000 | 1.30 | 132 | 1 | 1 |
| ## 125 | 0 | 217.000 | 3.70 | 134 | 1 | 0 |
| ## 194 | 0 | 160.000 | 1.18 | 142 | 1 | 1 |
| ## 213 | 0 | 481.000 | 1.40 | 138 | 1 | 1 |
| ## 166 | 1 | 192.000 | 1.30 | 135 | 0 | 0 |
| ## 199 | 1 | 279.000 | 1.00 | 134 | 0 | 0 |
| ## 294 | 0 | 179.000 | 0.90 | 136 | 1 | 1 |
| ## 69 | 0 | 244.000 | 1.20 | 142 | 0 | 0 |
| ## 29 | 0 | 153.000 | 5.80 | 134 | 1 | 0 |
| ## 133 | 1 | 263.358 | 1.18 | 137 | 0 | 0 |
| ## 118 | 0 | 507.000 | 3.20 | 138 | 0 | 0 |
| ## 24 | 0 | 368.000 | 0.80 | 135 | 1 | 0 |
| ## 13 | 0 | 136.000 | 1.10 | 137 | 1 | 0 |
| ## 94 | 0 | 210.000 | 1.70 | 135 | 1 | 0 |
| ## 26 | 0 | 149.000 | 1.90 | 144 | 1 | 1 |
| ## 50 | 0 | 395.000 | 1.00 | 140 | 0 | 0 |
| ## 176 | 0 | 337.000 | 1.00 | 138 | 1 | 1 |
| ## 151 | 1 | 235.000 | 2.50 | 135 | 0 | 0 |
| ## 115 | 1 | 328.000 | 1.20 | 126 | 1 | 0 |
| ## 76 | 0 | 204.000 | 0.70 | 139 | 1 | 1 |
| ## 182 | 0 | 221.000 | 1.00 | 136 | 1 | 1 |
| ## 286 | 0 | 336.000 | 1.20 | 135 | 1 | 0 |
| ## 92 | 0 | 297.000 | 1.00 | 133 | 1 | 0 |
| ## 32 | 0 | 360.000 | 3.00 | 132 | 1 | 0 |
| ## 38 | 1 | 321.000 | 1.00 | 145 | 0 | 0 |
| ## 120 | 0 | 263.358 | 1.83 | 134 | 0 | 0 |
| ## 116 | 0 | 164.000 | 1.00 | 139 | 0 | 0 |
| ## 242 | 0 | 249.000 | 1.30 | 136 | 1 | 1 |
| ## 190 | 1 | 275.000 | 0.90 | 140 | 0 | 0 |
| ## 223 | 0 | 365.000 | 1.10 | 139 | 1 | 1 |
| ## 201 | 0 | 73.000 | 0.70 | 137 | 1 | 0 |
| ## 10 | 1 | 388.000 | 9.40 | 133 | 1 | 1 |
| ## 167 | 0 | 220.000 | 0.70 | 133 | 1 | 1 |
| ## 228 | 1 | 141.000 | 1.00 | 140 | 1 | 0 |
| ## 88 | 1 | 304.000 | 0.80 | 140 | 1 | 0 |
| ## 135 | 0 | 231.000 | 1.18 | 137 | 1 | 1 |
| ## 19 | 1 | 237.000 | 1.00 | 140 | 0 | 0 |
| ## 15 | 1 | 427.000 | 1.00 | 138 | 0 | 0 |

| | | | | | | |
|--------|------|-------------|------|-----|---|---|
| ## 101 | 0 | 298.000 | 1.10 | 141 | 1 | 0 |
| ## 292 | 0 | 133.000 | 1.40 | 139 | 1 | 0 |
| ## 132 | 0 | 250.000 | 6.10 | 131 | 1 | 0 |
| ## 196 | 0 | 223.000 | 1.80 | 145 | 1 | 0 |
| ## | time | DEATH_EVENT | | | | |
| ## 111 | 90 | 1 | | | | |
| ## 48 | 40 | 1 | | | | |
| ## 229 | 207 | 0 | | | | |
| ## 259 | 230 | 0 | | | | |
| ## 78 | 74 | 0 | | | | |
| ## 170 | 145 | 0 | | | | |
| ## 138 | 108 | 0 | | | | |
| ## 226 | 205 | 0 | | | | |
| ## 236 | 209 | 0 | | | | |
| ## 230 | 207 | 0 | | | | |
| ## 206 | 187 | 0 | | | | |
| ## 285 | 250 | 0 | | | | |
| ## 185 | 170 | 1 | | | | |
| ## 197 | 185 | 0 | | | | |
| ## 40 | 30 | 1 | | | | |
| ## 98 | 85 | 0 | | | | |
| ## 205 | 187 | 0 | | | | |
| ## 280 | 246 | 0 | | | | |
| ## 270 | 244 | 0 | | | | |
| ## 9 | 10 | 1 | | | | |
| ## 210 | 187 | 0 | | | | |
| ## 265 | 237 | 0 | | | | |
| ## 103 | 87 | 0 | | | | |
| ## 80 | 74 | 0 | | | | |
| ## 36 | 30 | 1 | | | | |
| ## 261 | 233 | 0 | | | | |
| ## 11 | 10 | 1 | | | | |
| ## 212 | 192 | 0 | | | | |
| ## 153 | 118 | 0 | | | | |
| ## 288 | 250 | 0 | | | | |
| ## 41 | 31 | 1 | | | | |
| ## 188 | 172 | 1 | | | | |
| ## 174 | 146 | 0 | | | | |
| ## 14 | 11 | 1 | | | | |
| ## 239 | 210 | 0 | | | | |
| ## 156 | 120 | 0 | | | | |
| ## 152 | 117 | 0 | | | | |
| ## 44 | 33 | 0 | | | | |
| ## 86 | 79 | 0 | | | | |
| ## 211 | 188 | 0 | | | | |
| ## 220 | 197 | 0 | | | | |
| ## 282 | 250 | 0 | | | | |
| ## 108 | 88 | 0 | | | | |
| ## 17 | 14 | 1 | | | | |
| ## 70 | 67 | 1 | | | | |
| ## 268 | 244 | 0 | | | | |
| ## 25 | 23 | 1 | | | | |
| ## 100 | 86 | 0 | | | | |
| ## 297 | 278 | 0 | | | | |

| | | | |
|----|-----|-----|---|
| ## | 207 | 187 | 0 |
| ## | 298 | 280 | 0 |
| ## | 74 | 72 | 0 |
| ## | 159 | 121 | 0 |
| ## | 42 | 32 | 1 |
| ## | 296 | 271 | 0 |
| ## | 53 | 43 | 1 |
| ## | 1 | 4 | 1 |
| ## | 221 | 198 | 1 |
| ## | 187 | 172 | 1 |
| ## | 99 | 85 | 0 |
| ## | 96 | 83 | 0 |
| ## | 73 | 72 | 1 |
| ## | 45 | 33 | 1 |
| ## | 72 | 71 | 0 |
| ## | 178 | 147 | 0 |
| ## | 274 | 245 | 0 |
| ## | 284 | 250 | 0 |
| ## | 233 | 209 | 0 |
| ## | 271 | 244 | 0 |
| ## | 64 | 61 | 1 |
| ## | 281 | 247 | 0 |
| ## | 235 | 209 | 0 |
| ## | 147 | 112 | 0 |
| ## | 124 | 95 | 0 |
| ## | 33 | 28 | 1 |
| ## | 276 | 245 | 0 |
| ## | 175 | 146 | 0 |
| ## | 149 | 113 | 1 |
| ## | 47 | 38 | 1 |
| ## | 266 | 240 | 0 |
| ## | 137 | 107 | 0 |
| ## | 267 | 241 | 1 |
| ## | 57 | 54 | 0 |
| ## | 102 | 87 | 0 |
| ## | 262 | 233 | 0 |
| ## | 208 | 187 | 0 |
| ## | 49 | 41 | 1 |
| ## | 141 | 109 | 1 |
| ## | 232 | 208 | 0 |
| ## | 37 | 30 | 1 |
| ## | 68 | 65 | 1 |
| ## | 77 | 74 | 0 |
| ## | 143 | 109 | 0 |
| ## | 51 | 43 | 1 |
| ## | 290 | 256 | 0 |
| ## | 162 | 121 | 0 |
| ## | 81 | 75 | 0 |
| ## | 157 | 120 | 0 |
| ## | 249 | 214 | 0 |
| ## | 58 | 54 | 0 |
| ## | 248 | 214 | 0 |
| ## | 244 | 213 | 0 |
| ## | 5 | 8 | 1 |

| | | | |
|----|-----|-----|---|
| ## | 139 | 108 | 0 |
| ## | 2 | 6 | 1 |
| ## | 129 | 104 | 0 |
| ## | 105 | 87 | 0 |
| ## | 134 | 107 | 0 |
| ## | 200 | 186 | 0 |
| ## | 264 | 237 | 0 |
| ## | 287 | 250 | 0 |
| ## | 255 | 215 | 0 |
| ## | 84 | 78 | 0 |
| ## | 299 | 285 | 0 |
| ## | 112 | 90 | 0 |
| ## | 18 | 14 | 1 |
| ## | 136 | 107 | 0 |
| ## | 158 | 120 | 0 |
| ## | 195 | 180 | 1 |
| ## | 28 | 26 | 1 |
| ## | 250 | 214 | 0 |
| ## | 180 | 147 | 0 |
| ## | 12 | 10 | 1 |
| ## | 75 | 73 | 1 |
| ## | 161 | 121 | 0 |
| ## | 272 | 244 | 0 |
| ## | 140 | 108 | 0 |
| ## | 198 | 186 | 0 |
| ## | 31 | 27 | 1 |
| ## | 62 | 60 | 1 |
| ## | 173 | 146 | 0 |
| ## | 181 | 148 | 0 |
| ## | 107 | 88 | 0 |
| ## | 209 | 187 | 0 |
| ## | 121 | 95 | 0 |
| ## | 21 | 16 | 0 |
| ## | 55 | 45 | 1 |
| ## | 128 | 104 | 0 |
| ## | 204 | 187 | 0 |
| ## | 16 | 13 | 1 |
| ## | 3 | 7 | 1 |
| ## | 177 | 147 | 0 |
| ## | 85 | 78 | 1 |
| ## | 6 | 8 | 1 |
| ## | 63 | 60 | 0 |
| ## | 241 | 212 | 0 |
| ## | 127 | 100 | 1 |
| ## | 258 | 230 | 0 |
| ## | 247 | 214 | 1 |
| ## | 256 | 216 | 0 |
| ## | 172 | 146 | 0 |
| ## | 123 | 95 | 0 |
| ## | 283 | 250 | 0 |
| ## | 231 | 207 | 1 |
| ## | 148 | 113 | 0 |
| ## | 186 | 171 | 1 |
| ## | 183 | 154 | 1 |

| | | |
|--------|-----|---|
| ## 67 | 65 | 1 |
| ## 279 | 246 | 0 |
| ## 234 | 209 | 0 |
| ## 66 | 64 | 1 |
| ## 93 | 82 | 0 |
| ## 39 | 30 | 0 |
| ## 146 | 112 | 0 |
| ## 192 | 174 | 0 |
| ## 245 | 213 | 0 |
| ## 46 | 35 | 1 |
| ## 27 | 24 | 1 |
| ## 20 | 15 | 1 |
| ## 110 | 88 | 0 |
| ## 109 | 88 | 0 |
| ## 60 | 59 | 1 |
| ## 131 | 106 | 0 |
| ## 165 | 129 | 1 |
| ## 160 | 121 | 0 |
| ## 130 | 105 | 0 |
| ## 257 | 220 | 0 |
| ## 224 | 201 | 0 |
| ## 168 | 135 | 1 |
| ## 218 | 196 | 1 |
| ## 71 | 68 | 0 |
| ## 89 | 79 | 0 |
| ## 227 | 205 | 0 |
| ## 125 | 96 | 1 |
| ## 194 | 180 | 0 |
| ## 213 | 192 | 0 |
| ## 166 | 130 | 1 |
| ## 199 | 186 | 0 |
| ## 294 | 270 | 0 |
| ## 69 | 66 | 1 |
| ## 29 | 26 | 1 |
| ## 133 | 107 | 0 |
| ## 118 | 94 | 0 |
| ## 24 | 22 | 0 |
| ## 13 | 11 | 1 |
| ## 94 | 82 | 1 |
| ## 26 | 23 | 1 |
| ## 50 | 42 | 1 |
| ## 176 | 146 | 0 |
| ## 151 | 115 | 1 |
| ## 115 | 91 | 0 |
| ## 76 | 73 | 1 |
| ## 182 | 150 | 1 |
| ## 286 | 250 | 0 |
| ## 92 | 80 | 0 |
| ## 32 | 28 | 1 |
| ## 38 | 30 | 1 |
| ## 120 | 95 | 1 |
| ## 116 | 91 | 0 |
| ## 242 | 212 | 0 |
| ## 190 | 174 | 0 |

| | | | |
|----|-----|-----|---|
| ## | 223 | 201 | 0 |
| ## | 201 | 186 | 0 |
| ## | 10 | 10 | 1 |
| ## | 167 | 134 | 0 |
| ## | 228 | 206 | 0 |
| ## | 88 | 79 | 0 |
| ## | 135 | 107 | 0 |
| ## | 19 | 15 | 1 |
| ## | 15 | 12 | 0 |
| ## | 101 | 87 | 0 |
| ## | 292 | 258 | 0 |
| ## | 132 | 107 | 0 |
| ## | 196 | 180 | 1 |

test_data

| ## | age | anaemia | creatinine_phosphokinase | diabetes | ejection_fraction |
|--------|-----|---------|--------------------------|----------|-------------------|
| ## 4 | 50 | 1 | 111 | 0 | 20 |
| ## 7 | 75 | 1 | 246 | 0 | 15 |
| ## 8 | 60 | 1 | 315 | 1 | 60 |
| ## 22 | 65 | 1 | 128 | 1 | 30 |
| ## 23 | 68 | 1 | 220 | 0 | 35 |
| ## 30 | 82 | 0 | 70 | 1 | 30 |
| ## 34 | 50 | 1 | 159 | 1 | 30 |
| ## 35 | 65 | 0 | 94 | 1 | 50 |
| ## 43 | 70 | 0 | 571 | 1 | 45 |
| ## 52 | 53 | 1 | 91 | 0 | 20 |
| ## 54 | 70 | 1 | 69 | 1 | 50 |
| ## 56 | 95 | 1 | 371 | 0 | 30 |
| ## 59 | 49 | 0 | 789 | 0 | 20 |
| ## 61 | 45 | 0 | 7702 | 1 | 25 |
| ## 65 | 45 | 0 | 582 | 0 | 80 |
| ## 79 | 75 | 1 | 203 | 1 | 38 |
| ## 82 | 67 | 0 | 582 | 0 | 50 |
| ## 83 | 60 | 1 | 76 | 1 | 25 |
| ## 87 | 55 | 0 | 47 | 0 | 35 |
| ## 90 | 57 | 1 | 115 | 0 | 25 |
| ## 91 | 70 | 0 | 66 | 1 | 45 |
| ## 95 | 58 | 0 | 144 | 1 | 38 |
| ## 97 | 63 | 1 | 514 | 1 | 25 |
| ## 104 | 42 | 0 | 5209 | 0 | 30 |
| ## 106 | 72 | 1 | 328 | 0 | 30 |
| ## 113 | 50 | 0 | 369 | 1 | 25 |
| ## 114 | 70 | 1 | 143 | 0 | 60 |
| ## 117 | 60 | 1 | 96 | 1 | 60 |
| ## 119 | 65 | 1 | 113 | 1 | 60 |
| ## 122 | 66 | 1 | 68 | 1 | 38 |
| ## 126 | 43 | 1 | 358 | 0 | 50 |
| ## 142 | 46 | 1 | 291 | 0 | 35 |
| ## 144 | 61 | 1 | 84 | 0 | 40 |
| ## 145 | 72 | 1 | 943 | 0 | 25 |
| ## 150 | 60 | 0 | 2261 | 0 | 35 |
| ## 154 | 50 | 0 | 1846 | 1 | 35 |
| ## 155 | 65 | 1 | 335 | 0 | 35 |

| | | | | | | |
|----|---|----|---------|------|-----|-----|
| ## | 163 | 63 | 1 | 582 | 0 | 40 |
| ## | 164 | 50 | 1 | 2334 | 1 | 35 |
| ## | 169 | 65 | 0 | 582 | 1 | 40 |
| ## | 171 | 51 | 1 | 582 | 1 | 35 |
| ## | 179 | 63 | 1 | 122 | 1 | 60 |
| ## | 184 | 75 | 0 | 99 | 0 | 38 |
| ## | 189 | 60 | 1 | 151 | 1 | 40 |
| ## | 191 | 80 | 0 | 582 | 1 | 35 |
| ## | 193 | 50 | 1 | 121 | 1 | 40 |
| ## | 202 | 45 | 0 | 308 | 1 | 60 |
| ## | 203 | 70 | 0 | 97 | 0 | 60 |
| ## | 214 | 48 | 1 | 131 | 1 | 30 |
| ## | 215 | 65 | 1 | 135 | 0 | 35 |
| ## | 216 | 73 | 0 | 582 | 0 | 35 |
| ## | 217 | 70 | 0 | 1202 | 0 | 50 |
| ## | 219 | 68 | 1 | 1021 | 1 | 35 |
| ## | 222 | 65 | 0 | 118 | 0 | 50 |
| ## | 225 | 58 | 0 | 582 | 1 | 25 |
| ## | 237 | 75 | 0 | 119 | 0 | 50 |
| ## | 238 | 70 | 0 | 232 | 0 | 30 |
| ## | 240 | 55 | 1 | 180 | 0 | 45 |
| ## | 243 | 40 | 0 | 90 | 0 | 35 |
| ## | 246 | 61 | 1 | 80 | 1 | 38 |
| ## | 251 | 50 | 0 | 2522 | 0 | 30 |
| ## | 252 | 55 | 0 | 572 | 1 | 35 |
| ## | 253 | 50 | 0 | 245 | 0 | 45 |
| ## | 254 | 70 | 0 | 88 | 1 | 35 |
| ## | 260 | 53 | 0 | 56 | 0 | 50 |
| ## | 263 | 65 | 1 | 258 | 1 | 25 |
| ## | 269 | 45 | 0 | 582 | 1 | 38 |
| ## | 273 | 67 | 0 | 213 | 0 | 38 |
| ## | 275 | 60 | 1 | 257 | 1 | 30 |
| ## | 277 | 70 | 0 | 618 | 0 | 35 |
| ## | 278 | 70 | 0 | 582 | 1 | 38 |
| ## | 289 | 65 | 0 | 892 | 1 | 35 |
| ## | 291 | 45 | 0 | 615 | 1 | 55 |
| ## | 293 | 52 | 0 | 190 | 1 | 38 |
| ## | 295 | 62 | 0 | 61 | 1 | 38 |
| ## | high_blood_pressure platelets serum_creatinine serum_sodium sex smoking | | | | | |
| ## | 4 | 0 | 210.000 | 1.90 | 137 | 1 0 |
| ## | 7 | 0 | 127.000 | 1.20 | 137 | 1 0 |
| ## | 8 | 0 | 454.000 | 1.10 | 131 | 1 1 |
| ## | 22 | 1 | 297.000 | 1.60 | 136 | 0 0 |
| ## | 23 | 1 | 289.000 | 0.90 | 140 | 1 1 |
| ## | 30 | 0 | 200.000 | 1.20 | 132 | 1 1 |
| ## | 34 | 0 | 302.000 | 1.20 | 138 | 0 0 |
| ## | 35 | 1 | 188.000 | 1.00 | 140 | 1 0 |
| ## | 43 | 1 | 185.000 | 1.20 | 139 | 1 1 |
| ## | 52 | 1 | 418.000 | 1.40 | 139 | 0 0 |
| ## | 54 | 1 | 351.000 | 1.00 | 134 | 0 0 |
| ## | 56 | 0 | 461.000 | 2.00 | 132 | 1 0 |
| ## | 59 | 1 | 319.000 | 1.10 | 136 | 1 1 |
| ## | 61 | 1 | 390.000 | 1.00 | 139 | 1 0 |
| ## | 65 | 0 | 263.358 | 1.18 | 137 | 0 0 |

| | | | | | | |
|--------|---|---------|------|-----|---|---|
| ## 79 | 1 | 283.000 | 0.60 | 131 | 1 | 1 |
| ## 82 | 0 | 263.358 | 1.18 | 137 | 1 | 1 |
| ## 83 | 0 | 196.000 | 2.50 | 132 | 0 | 0 |
| ## 87 | 1 | 173.000 | 1.10 | 137 | 1 | 0 |
| ## 90 | 1 | 181.000 | 1.10 | 144 | 1 | 0 |
| ## 91 | 0 | 249.000 | 0.80 | 136 | 1 | 1 |
| ## 95 | 1 | 327.000 | 0.70 | 142 | 0 | 0 |
| ## 97 | 1 | 254.000 | 1.30 | 134 | 1 | 0 |
| ## 104 | 0 | 226.000 | 1.00 | 140 | 1 | 1 |
| ## 106 | 1 | 621.000 | 1.70 | 138 | 0 | 1 |
| ## 113 | 0 | 252.000 | 1.60 | 136 | 1 | 0 |
| ## 114 | 0 | 351.000 | 1.30 | 137 | 0 | 0 |
| ## 117 | 1 | 271.000 | 0.70 | 136 | 0 | 0 |
| ## 119 | 1 | 203.000 | 0.90 | 140 | 0 | 0 |
| ## 122 | 1 | 162.000 | 1.00 | 136 | 0 | 0 |
| ## 126 | 0 | 237.000 | 1.30 | 135 | 0 | 0 |
| ## 142 | 0 | 348.000 | 0.90 | 140 | 0 | 0 |
| ## 144 | 1 | 229.000 | 0.90 | 141 | 0 | 0 |
| ## 145 | 1 | 338.000 | 1.70 | 139 | 1 | 1 |
| ## 150 | 1 | 228.000 | 0.90 | 136 | 1 | 0 |
| ## 154 | 0 | 263.358 | 1.18 | 137 | 1 | 1 |
| ## 155 | 1 | 235.000 | 0.80 | 136 | 0 | 0 |
| ## 163 | 0 | 448.000 | 0.90 | 137 | 1 | 1 |
| ## 164 | 0 | 75.000 | 0.90 | 142 | 0 | 0 |
| ## 169 | 0 | 270.000 | 1.00 | 138 | 0 | 0 |
| ## 171 | 0 | 263.358 | 1.50 | 136 | 1 | 1 |
| ## 179 | 0 | 267.000 | 1.20 | 145 | 1 | 0 |
| ## 184 | 1 | 224.000 | 2.50 | 134 | 1 | 0 |
| ## 189 | 1 | 201.000 | 1.00 | 136 | 0 | 0 |
| ## 191 | 0 | 350.000 | 2.10 | 134 | 1 | 0 |
| ## 193 | 0 | 260.000 | 0.70 | 130 | 1 | 0 |
| ## 202 | 1 | 377.000 | 1.00 | 136 | 1 | 0 |
| ## 203 | 1 | 220.000 | 0.90 | 138 | 1 | 0 |
| ## 214 | 1 | 244.000 | 1.60 | 130 | 0 | 0 |
| ## 215 | 1 | 290.000 | 0.80 | 134 | 1 | 0 |
| ## 216 | 1 | 203.000 | 1.30 | 134 | 1 | 0 |
| ## 217 | 1 | 358.000 | 0.90 | 141 | 0 | 0 |
| ## 219 | 0 | 271.000 | 1.10 | 134 | 1 | 0 |
| ## 222 | 0 | 194.000 | 1.10 | 145 | 1 | 1 |
| ## 225 | 0 | 504.000 | 1.00 | 138 | 1 | 0 |
| ## 237 | 1 | 248.000 | 1.10 | 148 | 1 | 0 |
| ## 238 | 0 | 173.000 | 1.20 | 132 | 1 | 0 |
| ## 240 | 0 | 263.358 | 1.18 | 137 | 1 | 1 |
| ## 243 | 0 | 255.000 | 1.10 | 136 | 1 | 1 |
| ## 246 | 0 | 282.000 | 1.40 | 137 | 1 | 0 |
| ## 251 | 1 | 404.000 | 0.50 | 139 | 0 | 0 |
| ## 252 | 0 | 231.000 | 0.80 | 143 | 0 | 0 |
| ## 253 | 1 | 274.000 | 1.00 | 133 | 1 | 0 |
| ## 254 | 1 | 236.000 | 1.20 | 132 | 0 | 0 |
| ## 260 | 0 | 308.000 | 0.70 | 135 | 1 | 1 |
| ## 263 | 0 | 198.000 | 1.40 | 129 | 1 | 0 |
| ## 269 | 0 | 302.000 | 0.90 | 140 | 0 | 0 |
| ## 273 | 0 | 215.000 | 1.20 | 133 | 0 | 0 |
| ## 275 | 0 | 150.000 | 1.00 | 137 | 1 | 1 |

| | | | | | | |
|--------|------|-------------|------|-----|---|---|
| ## 277 | 0 | 327.000 | 1.10 | 142 | 0 | 0 |
| ## 278 | 0 | 25.100 | 1.10 | 140 | 1 | 0 |
| ## 289 | 0 | 263.358 | 1.10 | 142 | 0 | 0 |
| ## 291 | 0 | 222.000 | 0.80 | 141 | 0 | 0 |
| ## 293 | 0 | 382.000 | 1.00 | 140 | 1 | 1 |
| ## 295 | 1 | 155.000 | 1.10 | 143 | 1 | 1 |
| ## | time | DEATH_EVENT | | | | |
| ## 4 | 7 | 1 | | | | |
| ## 7 | 10 | 1 | | | | |
| ## 8 | 10 | 1 | | | | |
| ## 22 | 20 | 1 | | | | |
| ## 23 | 20 | 1 | | | | |
| ## 30 | 26 | 1 | | | | |
| ## 34 | 29 | 0 | | | | |
| ## 35 | 29 | 1 | | | | |
| ## 43 | 33 | 1 | | | | |
| ## 52 | 43 | 1 | | | | |
| ## 54 | 44 | 1 | | | | |
| ## 56 | 50 | 1 | | | | |
| ## 59 | 55 | 1 | | | | |
| ## 61 | 60 | 1 | | | | |
| ## 65 | 63 | 0 | | | | |
| ## 79 | 74 | 0 | | | | |
| ## 82 | 76 | 0 | | | | |
| ## 83 | 77 | 1 | | | | |
| ## 87 | 79 | 0 | | | | |
| ## 90 | 79 | 0 | | | | |
| ## 91 | 80 | 0 | | | | |
| ## 95 | 83 | 0 | | | | |
| ## 97 | 83 | 0 | | | | |
| ## 104 | 87 | 0 | | | | |
| ## 106 | 88 | 1 | | | | |
| ## 113 | 90 | 0 | | | | |
| ## 114 | 90 | 1 | | | | |
| ## 117 | 94 | 0 | | | | |
| ## 119 | 94 | 0 | | | | |
| ## 122 | 95 | 0 | | | | |
| ## 126 | 97 | 0 | | | | |
| ## 142 | 109 | 0 | | | | |
| ## 144 | 110 | 0 | | | | |
| ## 145 | 111 | 1 | | | | |
| ## 150 | 115 | 0 | | | | |
| ## 154 | 119 | 0 | | | | |
| ## 155 | 120 | 0 | | | | |
| ## 163 | 123 | 0 | | | | |
| ## 164 | 126 | 1 | | | | |
| ## 169 | 140 | 0 | | | | |
| ## 171 | 145 | 0 | | | | |
| ## 179 | 147 | 0 | | | | |
| ## 184 | 162 | 1 | | | | |
| ## 189 | 172 | 0 | | | | |
| ## 191 | 174 | 0 | | | | |
| ## 193 | 175 | 0 | | | | |
| ## 202 | 186 | 0 | | | | |

```
## 203 186 0
## 214 193 1
## 215 194 0
## 216 195 0
## 217 196 0
## 219 197 0
## 222 200 0
## 225 205 0
## 237 209 0
## 238 210 0
## 240 211 0
## 243 212 0
## 246 213 0
## 251 214 0
## 252 215 0
## 253 215 0
## 254 215 0
## 260 231 0
## 263 235 1
## 269 244 0
## 273 245 0
## 275 245 0
## 277 245 0
## 278 246 0
## 289 256 0
## 291 257 0
## 293 258 0
## 295 270 0
```

Model development (logistic)

```
logistic_model = glm(DEATH_EVENT~., data=train_data, family="binomial")
summary(logistic_model)
```

```
##
## Call:
## glm(formula = DEATH_EVENT ~ ., family = "binomial", data = train_data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    7.9632799   6.0737111   1.311   0.1898
## age             0.0434821   0.0172850   2.516   0.0119 *
## anaemia        -0.3808572   0.4272709  -0.891   0.3727
## creatinine_phosphokinase 0.0001882 0.0002380   0.791   0.4292
## diabetes       0.2101722   0.4112369   0.511   0.6093
## ejection_fraction -0.0825491 0.0190516 -4.333 1.47e-05 ***
## high_blood_pressure -0.2561613 0.4318960  -0.593   0.5531
## platelets      -0.0031861 0.0023648  -1.347   0.1779
## serum_creatinine  0.6122125 0.1903677   3.216   0.0013 **
## serum_sodium    -0.0416713 0.0432340  -0.964   0.3351
## sex            -0.4533968 0.4677745  -0.969   0.3324
## smoking        -0.2713557 0.4774305  -0.568   0.5698
## time          -0.0209163 0.0034518 -6.059 1.37e-09 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 285.62  on 223  degrees of freedom
## Residual deviance: 166.87  on 211  degrees of freedom
## AIC: 192.87
##
## Number of Fisher Scoring iterations: 6
```

Applying this model to test data

```
pred_prob <- predict(logistic_model, newdata = test_data, type = "response")
#Setting threshold probability to 0.5
threshold <- 0.5
test_predictions <- ifelse(pred_prob > threshold, 1, 0)
test_predictions
```

```
##  4  7  8 22 23 30 34 35 43 52 54 56 59 61 65 79 82 83 87 90
##  1  1  0  1  1  1  1  0  1  1  0  1  1  1  0  0  0  1  0  0
## 91 95 97 104 106 113 114 117 119 122 126 142 144 145 150 154 155 163 164 169
##  0  0  1  0  0  1  0  0  0  0  0  0  0  0  0  0  0  0  0  0
## 171 179 184 189 191 193 202 203 214 215 216 217 219 222 225 237 238 240 243 246
##  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
## 251 252 253 254 260 263 269 273 275 277 278 289 291 293 295
##  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
```

Model evaluation

```
library(caret)
```

```
## Loading required package: lattice
```

```
# Create a confusion matrix
conf_matrix <- confusionMatrix(factor(test_predictions), factor(test_data$DEATH_EVENT))

# Display the confusion matrix and accuracy
conf_matrix
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction  0  1
##           0 51 10
##           1  3 11
##
##           Accuracy : 0.8267
##           95% CI : (0.7219, 0.9043)
##           No Information Rate : 0.72
##           P-Value [Acc > NIR] : 0.02306
```

```
##
##           Kappa : 0.5214
##
## Mcnemar's Test P-Value : 0.09609
##
##           Sensitivity : 0.9444
##           Specificity : 0.5238
##           Pos Pred Value : 0.8361
##           Neg Pred Value : 0.7857
##           Prevalence : 0.7200
##           Detection Rate : 0.6800
##           Detection Prevalence : 0.8133
##           Balanced Accuracy : 0.7341
##
##           'Positive' Class : 0
##
```

```
print(paste("Accuracy:", conf_matrix$overall["Accuracy"]))
```

```
## [1] "Accuracy: 0.826666666666667"
```

#Here we obtained an accuracy of 0.85 which is good for predicting survival of patients.

Drawing ROC curve

```
library(pROC)
```

```
## Type 'citation("pROC")' for a citation.
```

```
##
```

```
## Attaching package: 'pROC'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      cov, smooth, var
```

```
# Create ROC curve
```

```
roc_curve <- roc(test_data$DEATH_EVENT, pred_prob)
```

```
## Setting levels: control = 0, case = 1
```

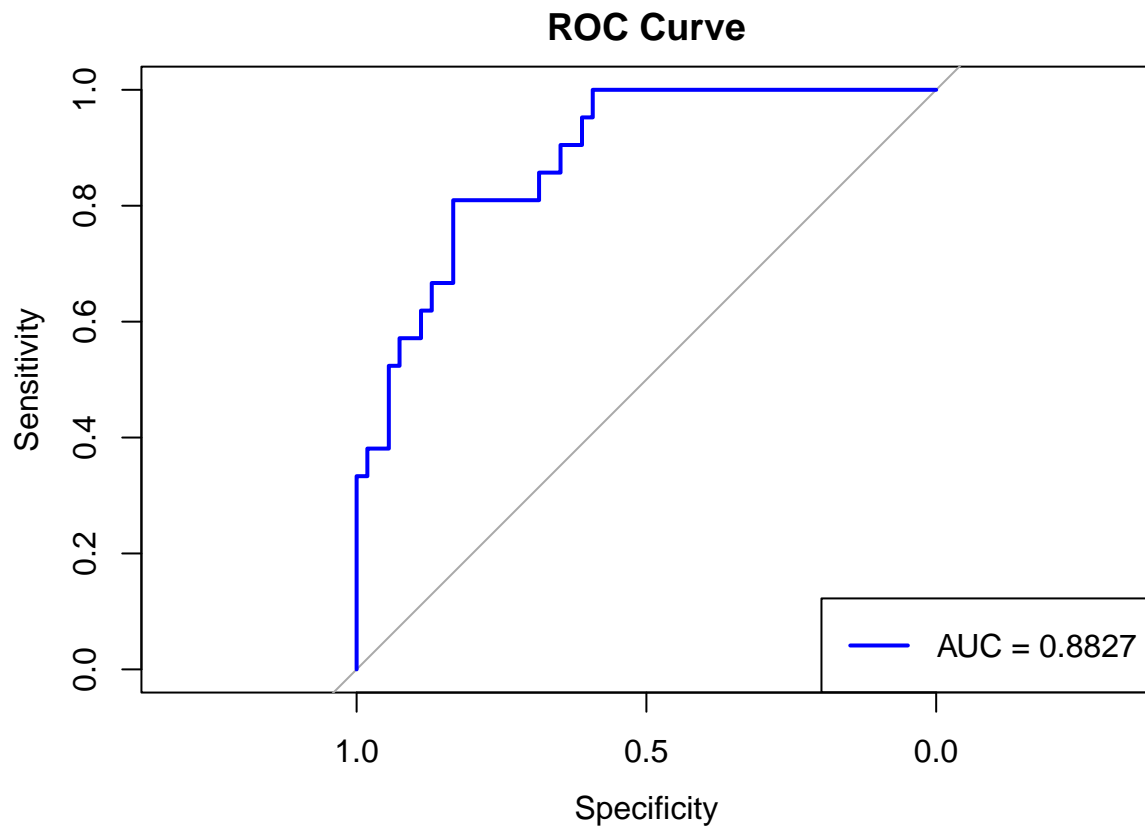
```
## Setting direction: controls < cases
```

```
# Plot ROC curve
```

```
plot(roc_curve, main = "ROC Curve", col = "blue", lwd = 2)
```

```
# Add a legend
```

```
legend("bottomright", legend = paste("AUC =", round(auc(roc_curve), 4)), col = "blue", lwd = 2)
```

Again AUC score is 0.8843 which represent good measure of the discriminative ability of a classification model.