

Assignment-8.R

nidhi

2020-11-03

```
#Loading Packages
```

```
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(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 3.6.3
```

```
library(cowplot)
```

```
## Warning: package 'cowplot' was built under R version 3.6.3
```

```
library(regclass)
```

```
## Warning: package 'regclass' was built under R version 3.6.3
```

```
## Loading required package: bestglm
```

```
## Warning: package 'bestglm' was built under R version 3.6.3
```

```
## Loading required package: leaps
```

```
## Warning: package 'leaps' was built under R version 3.6.3
```

```
## Loading required package: VGAM
```

```
## Warning: package 'VGAM' was built under R version 3.6.3
```

```
## Loading required package: stats4
```

```
## Loading required package: splines
```

```
## Loading required package: rpart
```

```
## Loading required package: randomForest
```

```
## Warning: package 'randomForest' was built under R version 3.6.3
## randomForest 4.6-14
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:ggplot2':
##
##     margin
## The following object is masked from 'package:dplyr':
##
##     combine
## Important regclass change from 1.3:
## All functions that had a . in the name now have an _
## all.correlations -> all_correlations, cor.demo -> cor_demo, etc.

library(caret)

## Warning: package 'caret' was built under R version 3.6.3
## Loading required package: lattice
## Warning: package 'lattice' was built under R version 3.6.3
##
## Attaching package: 'lattice'
## The following object is masked from 'package:regclass':
##
##     qq
##
## Attaching package: 'caret'
## The following object is masked from 'package:VGAM':
##
##     predictors

library(e1071)

## Warning: package 'e1071' was built under R version 3.6.3

library(pROC)

## Warning: package 'pROC' was built under R version 3.6.3
## Type 'citation("pROC")' for a citation.
```

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

## The following objects are masked from 'package:stats':
##
##      cov, smooth, var

#Loading dataset
rawdata <-
read.csv("C:/Users/nidhi/OneDrive/Desktop/MVA/heart_failure_clinical_records_
dataset.csv")
View(rawdata)

#Identifying different columns names
names(rawdata)

## [1] "age" "anaemia"
## [3] "creatinine_phosphokinase" "diabetes"
## [5] "ejection_fraction" "high_blood_pressure"
## [7] "platelets" "serum_creatinine"
## [9] "serum_sodium" "sex"
## [11] "smoking" "time"
## [13] "DEATH_EVENT"

#Data Summary
str(rawdata)

## '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 265000 263358 162000 210000 327000 ...
## $ 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 : Factor w/ 2 levels "Female","male": 2 2 2 2 1
2 2 2 1 2 ...
## $ 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 : Factor w/ 2 levels "Death","No Death": 2 2 2
2 2 2 2 2 2 ...

summary(rawdata)

## age anaemia creatinine_phosphokinase
## Min. :40.00 Min. :0.0000 Min. : 23.0
```

```
## 1st Qu.:51.00 1st Qu.:0.0000 1st Qu.: 116.5
## Median :60.00 Median :0.0000 Median : 250.0
## Mean :60.83 Mean :0.4314 Mean : 581.8
## 3rd Qu.:70.00 3rd Qu.:1.0000 3rd Qu.: 582.0
## Max. :95.00 Max. :1.0000 Max. :7861.0
## diabetes ejection_fraction high_blood_pressure platelets
## Min. :0.0000 Min. :14.00 Min. :0.0000 Min. : 25100
## 1st Qu.:0.0000 1st Qu.:30.00 1st Qu.:0.0000 1st Qu.:212500
## Median :0.0000 Median :38.00 Median :0.0000 Median :262000
## Mean :0.4181 Mean :38.08 Mean :0.3512 Mean :263358
## 3rd Qu.:1.0000 3rd Qu.:45.00 3rd Qu.:1.0000 3rd Qu.:303500
## Max. :1.0000 Max. :80.00 Max. :1.0000 Max. :850000
## serum_creatinine serum_sodium sex smoking
## Min. :0.500 Min. :113.0 Female:105 Min. :0.0000
## 1st Qu.:0.900 1st Qu.:134.0 male :194 1st Qu.:0.0000
## Median :1.100 Median :137.0 Median :0.0000
## Mean :1.394 Mean :136.6 Mean :0.3211
## 3rd Qu.:1.400 3rd Qu.:140.0 3rd Qu.:1.0000
## Max. :9.400 Max. :148.0 Max. :1.0000
## time DEATH_EVENT
## Min. : 4.0 Death :203
## 1st Qu.: 73.0 No Death: 96
## Median :115.0
## Mean :130.3
## 3rd Qu.:203.0
## Max. :285.0
```

head(rawdata)

```
## 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
## 1 1 265000 1.9 130 male
## 2 0 263358 1.1 136 male
## 3 0 162000 1.3 129 male
## 4 0 210000 1.9 137 male
## 5 0 327000 2.7 116 Female
## 6 1 204000 2.1 132 male
## smoking time DEATH_EVENT
## 1 0 4 No Death
## 2 0 6 No Death
## 3 1 7 No Death
## 4 0 7 No Death
## 5 0 8 No Death
## 6 1 8 No Death
```

```

dim(rawdata)

## [1] 299 13

#Data Cleaning

#Checking for missing values
is.null(rawdata)

## [1] FALSE

##The "FALSE" output shows there is no missing data in the dataset.

#Transforming data (Converting 0,1's to meaningful form)

data <- rawdata %>%
  mutate(anaemia = ifelse(anaemia ==1, "Yes", "No"),
         high_blood_pressure = ifelse(high_blood_pressure ==1, "Yes", "No"),
         diabetes = ifelse(diabetes ==1, "Yes", "No"),
         smoking =ifelse(smoking ==1,"Yes","No"),
         DEATH_EVENT=ifelse(DEATH_EVENT=="No Death", "Survived", "Death"))
  ) %>%
  mutate_if(is.character, as.factor) %>%
  dplyr::select(age, anaemia, creatinine_phosphokinase, diabetes,
ejection_fraction, high_blood_pressure, platelets,serum_creatinine,
serum_sodium, sex, smoking, time, DEATH_EVENT)

View(data)
str(data)

## 'data.frame': 299 obs. of 13 variables:
## $ age : num 75 55 65 50 65 90 75 60 65 80 ...
## $ anaemia : Factor w/ 2 levels "No","Yes": 1 1 1 2 2 2 2
2 1 2 ...
## $ creatinine_phosphokinase: int 582 7861 146 111 160 47 246 315 157 123
...
## $ diabetes : Factor w/ 2 levels "No","Yes": 1 1 1 1 2 1 1
2 1 1 ...
## $ ejection_fraction : int 20 38 20 20 20 40 15 60 65 35 ...
## $ high_blood_pressure : Factor w/ 2 levels "No","Yes": 2 1 1 1 1 2 1
1 1 2 ...
## $ platelets : num 265000 263358 162000 210000 327000 ...
## $ 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 : Factor w/ 2 levels "Female","male": 2 2 2 2 2 1
2 2 2 1 2 ...
## $ smoking : Factor w/ 2 levels "No","Yes": 1 1 2 1 1 2 1
2 1 2 ...
## $ time : int 4 6 7 7 8 8 10 10 10 10 ...

```

```
## $ DEATH_EVENT          : Factor w/ 2 levels "Death","Survived": 2 2 2
2 2 2 2 2 2 2 ...
```

```
summary(data)
```

```
##      age      anaemia  creatinine_phosphokinase  diabetes
## Min.   :40.00    No :170    Min.   : 23.0          No :174
## 1st Qu.:51.00    Yes:129    1st Qu.: 116.5        Yes:125
## Median :60.00
## Mean   :60.83
## 3rd Qu.:70.00
## Max.   :95.00
## Max.   :7861.0
## ejection_fraction  high_blood_pressure  platelets      serum_creatinine
## Min.   :14.00      No :194      Min.   : 25100    Min.   :0.500
## 1st Qu.:30.00      Yes:105     1st Qu.:212500   1st Qu.:0.900
## Median :38.00
## Mean   :38.08
## 3rd Qu.:45.00
## Max.   :80.00
## Max.   :850000    Max.   :9.400
## serum_sodium      sex      smoking      time      DEATH_EVENT
## Min.   :113.0      Female:105    No :203    Min.   : 4.0    Death   :203
## 1st Qu.:134.0      male :194     Yes: 96    1st Qu.: 73.0   Survived: 96
## Median :137.0
## Mean   :136.6
## 3rd Qu.:140.0
## Max.   :148.0
## Max.   :285.0
```

```
dataset<-data
attach(dataset)
```

```
#Simple logistic model 1
```

```
#Implementing simple logistic model using our independent variable sex to
predict patient's death or survival
```

```
xtabs(~ DEATH_EVENT + sex, data=dataset)
```

```
##           sex
## DEATH_EVENT Female male
##      Death      71  132
##      Survived   34   62
```

```
logistic_simple <- glm(DEATH_EVENT ~ sex, data=dataset, family="binomial")
summary(logistic_simple)
```

```
##
## Call:
## glm(formula = DEATH_EVENT ~ sex, family = "binomial", data = dataset)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.8846  -0.8776  -0.8776   1.5017   1.5105
##
```

```

## Coefficients:
##             Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.73632    0.20856  -3.531 0.000415 ***
## sexmale     -0.01935    0.25923  -0.075 0.940504
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 375.35  on 298  degrees of freedom
## Residual deviance: 375.34  on 297  degrees of freedom
## AIC: 379.34
##
## Number of Fisher Scoring iterations: 4

#The intercept is the log(odds) a female that can survive.
#sexmale is the log(odds ratio) that tells us if a sample has sex=male, the
odds of survival are on a log scale 0.019 times lesser than if a sample has
sex=female
#However, sex is not a significant predictor for death event

#Calculating the overall "Pseudo R-squared" and its p-value
ll.null <- logistic_simple$null.deviance/-2
ll.proposed <- logistic_simple$deviance/-2
ll.null

## [1] -187.6744

ll.proposed

## [1] -187.6716

## McFadden's Pseudo R^2 = [ LL(Null) - LL(Proposed) ] / LL(Null)
(ll.null - ll.proposed) / ll.null

## [1] 1.483129e-05

## chi-square value = 2*(LL(Proposed) - LL(Null))
## p-value = 1 - pchisq(chi-square value, df = 2-1)
1 - pchisq(2*(ll.proposed - ll.null), df=1)

## [1] 0.9405237

1 - pchisq((logistic_simple$null.deviance - logistic_simple$deviance), df=1)

## [1] 0.9405237

#Checking what logistic regression is predicting, given that a patient is
either female or male (and no other data about them).
predicted.data <-
data.frame(probability.of.death_event=logistic_simple$fitted.values, sex=datas

```

```
et$sex)
predicted.data
```

##	probability.of.death_event	sex
## 1	0.3195876	male
## 2	0.3195876	male
## 3	0.3195876	male
## 4	0.3195876	male
## 5	0.3238095	Female
## 6	0.3195876	male
## 7	0.3195876	male
## 8	0.3195876	male
## 9	0.3238095	Female
## 10	0.3195876	male
## 11	0.3195876	male
## 12	0.3195876	male
## 13	0.3195876	male
## 14	0.3195876	male
## 15	0.3238095	Female
## 16	0.3195876	male
## 17	0.3195876	male
## 18	0.3195876	male
## 19	0.3238095	Female
## 20	0.3238095	Female
## 21	0.3238095	Female
## 22	0.3238095	Female
## 23	0.3195876	male
## 24	0.3195876	male
## 25	0.3238095	Female
## 26	0.3195876	male
## 27	0.3238095	Female
## 28	0.3195876	male
## 29	0.3195876	male
## 30	0.3195876	male
## 31	0.3195876	male
## 32	0.3195876	male
## 33	0.3238095	Female
## 34	0.3238095	Female
## 35	0.3195876	male
## 36	0.3195876	male
## 37	0.3195876	male
## 38	0.3238095	Female
## 39	0.3195876	male
## 40	0.3238095	Female
## 41	0.3195876	male
## 42	0.3238095	Female
## 43	0.3195876	male
## 44	0.3195876	male
## 45	0.3238095	Female
## 46	0.3195876	male

## 47	0.3195876	male
## 48	0.3195876	male
## 49	0.3195876	male
## 50	0.3238095	Female
## 51	0.3195876	male
## 52	0.3238095	Female
## 53	0.3238095	Female
## 54	0.3238095	Female
## 55	0.3238095	Female
## 56	0.3195876	male
## 57	0.3195876	male
## 58	0.3195876	male
## 59	0.3195876	male
## 60	0.3195876	male
## 61	0.3195876	male
## 62	0.3238095	Female
## 63	0.3195876	male
## 64	0.3195876	male
## 65	0.3238095	Female
## 66	0.3195876	male
## 67	0.3238095	Female
## 68	0.3195876	male
## 69	0.3238095	Female
## 70	0.3195876	male
## 71	0.3195876	male
## 72	0.3195876	male
## 73	0.3195876	male
## 74	0.3195876	male
## 75	0.3195876	male
## 76	0.3195876	male
## 77	0.3238095	Female
## 78	0.3195876	male
## 79	0.3195876	male
## 80	0.3238095	Female
## 81	0.3238095	Female
## 82	0.3195876	male
## 83	0.3238095	Female
## 84	0.3195876	male
## 85	0.3238095	Female
## 86	0.3195876	male
## 87	0.3195876	male
## 88	0.3195876	male
## 89	0.3195876	male
## 90	0.3195876	male
## 91	0.3195876	male
## 92	0.3195876	male
## 93	0.3238095	Female
## 94	0.3195876	male
## 95	0.3238095	Female
## 96	0.3195876	male

## 97	0.3195876	male
## 98	0.3238095	Female
## 99	0.3238095	Female
## 100	0.3238095	Female
## 101	0.3195876	male
## 102	0.3195876	male
## 103	0.3195876	male
## 104	0.3195876	male
## 105	0.3238095	Female
## 106	0.3238095	Female
## 107	0.3195876	male
## 108	0.3195876	male
## 109	0.3195876	male
## 110	0.3195876	male
## 111	0.3195876	male
## 112	0.3195876	male
## 113	0.3195876	male
## 114	0.3238095	Female
## 115	0.3195876	male
## 116	0.3238095	Female
## 117	0.3238095	Female
## 118	0.3238095	Female
## 119	0.3238095	Female
## 120	0.3238095	Female
## 121	0.3195876	male
## 122	0.3238095	Female
## 123	0.3238095	Female
## 124	0.3238095	Female
## 125	0.3195876	male
## 126	0.3238095	Female
## 127	0.3238095	Female
## 128	0.3238095	Female
## 129	0.3195876	male
## 130	0.3195876	male
## 131	0.3195876	male
## 132	0.3195876	male
## 133	0.3238095	Female
## 134	0.3195876	male
## 135	0.3195876	male
## 136	0.3195876	male
## 137	0.3238095	Female
## 138	0.3195876	male
## 139	0.3238095	Female
## 140	0.3195876	male
## 141	0.3195876	male
## 142	0.3238095	Female
## 143	0.3238095	Female
## 144	0.3238095	Female
## 145	0.3195876	male
## 146	0.3195876	male

## 147	0.3195876	male
## 148	0.3195876	male
## 149	0.3195876	male
## 150	0.3195876	male
## 151	0.3238095	Female
## 152	0.3195876	male
## 153	0.3195876	male
## 154	0.3195876	male
## 155	0.3238095	Female
## 156	0.3195876	male
## 157	0.3238095	Female
## 158	0.3195876	male
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## 160	0.3195876	male
## 161	0.3195876	male
## 162	0.3195876	male
## 163	0.3195876	male
## 164	0.3238095	Female
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## 171	0.3195876	male
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## 175	0.3195876	male
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## 177	0.3195876	male
## 178	0.3238095	Female
## 179	0.3195876	male
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## 183	0.3195876	male
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## 186	0.3195876	male
## 187	0.3238095	Female
## 188	0.3238095	Female
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## 190	0.3238095	Female
## 191	0.3195876	male
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## 197	0.3238095	Female
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## 200	0.3195876	male
## 201	0.3195876	male
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## 206	0.3238095	Female
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## 226	0.3238095	Female
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## 228	0.3195876	male
## 229	0.3238095	Female
## 230	0.3238095	Female
## 231	0.3238095	Female
## 232	0.3195876	male
## 233	0.3195876	male
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## 239	0.3238095	Female
## 240	0.3195876	male
## 241	0.3238095	Female
## 242	0.3195876	male
## 243	0.3195876	male
## 244	0.3238095	Female
## 245	0.3195876	male
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## 247	0.3195876	male
## 248	0.3195876	male
## 249	0.3195876	male
## 250	0.3238095	Female
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## 269	0.3238095	Female
## 270	0.3195876	male
## 271	0.3195876	male
## 272	0.3238095	Female
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## 274	0.3195876	male
## 275	0.3195876	male
## 276	0.3238095	Female
## 277	0.3238095	Female
## 278	0.3195876	male
## 279	0.3238095	Female
## 280	0.3238095	Female
## 281	0.3195876	male
## 282	0.3195876	male
## 283	0.3195876	male
## 284	0.3195876	male
## 285	0.3195876	male
## 286	0.3195876	male
## 287	0.3195876	male
## 288	0.3238095	Female
## 289	0.3238095	Female
## 290	0.3238095	Female
## 291	0.3238095	Female
## 292	0.3195876	male
## 293	0.3195876	male
## 294	0.3195876	male
## 295	0.3195876	male
## 296	0.3238095	Female

```
## 297          0.3238095 Female
## 298          0.3195876  male
## 299          0.3195876  male

## There are only two probabilities (one for females and one for males),so we can summarize the predicted probabilities.
xtabs(~ probability.of.death_event + sex, data=predicted.data)

##                sex
## probability.of.death_event Female male
##          0.31958762886608          0  194
##          0.323809523809605        105   0

#Simple logistic model 2
#Implementing simple logistic model using our independent variable age to predict patient's death or survival
logistic_simple2 <- glm(DEATH_EVENT ~ age, data=dataset, family="binomial")
summary(logistic_simple2)

##
## Call:
## glm(formula = DEATH_EVENT ~ age, family = "binomial", data = dataset)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.4276  -0.8993  -0.6922   1.2344   1.9251
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -3.65433    0.70662  -5.172 2.32e-07 ***
## age          0.04695    0.01107   4.241 2.23e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 375.35  on 298  degrees of freedom
## Residual deviance: 355.99  on 297  degrees of freedom
## AIC: 359.99
##
## Number of Fisher Scoring iterations: 4

#Age is a significant predictor for death event
#For every one year increase in Age the odds of survival on a log scale is expected to increase by 0.047

#Calculating the overall "Pseudo R-squared" and its p-value
ll.null_age <- logistic_simple2$null.deviance/-2
ll.proposed_age <- logistic_simple2$deviance/-2
ll.null_age
```

```
## [1] -187.6744

ll.proposed_age

## [1] -177.9964

## McFadden's Pseudo R^2 = [ LL(Null) - LL(Proposed) ] / LL(Null)
(ll.null_age - ll.proposed_age) / ll.null_age

## [1] 0.051568

## chi-square value = 2*(LL(Proposed) - LL(Null))
## p-value = 1 - pchisq(chi-square value, df = 2-1)
1 - pchisq(2*(ll.proposed_age - ll.null_age), df=1)

## [1] 1.084786e-05

1 - pchisq((logistic_simple2$null.deviance - logistic_simple2$deviance),
df=1)

## [1] 1.084786e-05

#Checking what logistic regression is predicting with Age as the predictor
predicted.data_age <-
data.frame(probability.of.death_event=logistic_simple2$fitted.values, age=data
set$age)
predicted.data_age

##      probability.of.death_event      age
## 1          0.4667769 75.000
## 2          0.2550066 55.000
## 3          0.3537520 65.000
## 4          0.2130169 50.000
## 5          0.3537520 65.000
## 6          0.6390307 90.000
## 7          0.4667769 75.000
## 8          0.3020960 60.000
## 9          0.3537520 65.000
## 10         0.5253933 80.000
## 11         0.4667769 75.000
## 12         0.3222531 62.000
## 13         0.1763050 45.000
## 14         0.2130169 50.000
## 15         0.2052522 49.000
## 16         0.5487349 82.000
## 17         0.6059477 87.000
## 18         0.1763050 45.000
## 19         0.4090636 70.000
## 20         0.1976995 48.000
## 21         0.3537520 65.000
## 22         0.3537520 65.000
## 23         0.3865737 68.000
```

## 24	0.2375814	53.000
## 25	0.4667769	75.000
## 26	0.5253933	80.000
## 27	0.6912370	95.000
## 28	0.4090636	70.000
## 29	0.2826740	58.000
## 30	0.5487349	82.000
## 31	0.6811277	94.000
## 32	0.5833183	85.000
## 33	0.2130169	50.000
## 34	0.2130169	50.000
## 35	0.3537520	65.000
## 36	0.3977647	69.000
## 37	0.6390307	90.000
## 38	0.5487349	82.000
## 39	0.3020960	60.000
## 40	0.3020960	60.000
## 41	0.4090636	70.000
## 42	0.2130169	50.000
## 43	0.4090636	70.000
## 44	0.4319405	72.000
## 45	0.3020960	60.000
## 46	0.2130169	50.000
## 47	0.2209936	51.000
## 48	0.3020960	60.000
## 49	0.5253933	80.000
## 50	0.2732518	57.000
## 51	0.3865737	68.000
## 52	0.2375814	53.000
## 53	0.3020960	60.000
## 54	0.4090636	70.000
## 55	0.3020960	60.000
## 56	0.6912370	95.000
## 57	0.4090636	70.000
## 58	0.3020960	60.000
## 59	0.2052522	49.000
## 60	0.4319405	72.000
## 61	0.1763050	45.000
## 62	0.2130169	50.000
## 63	0.2550066	55.000
## 64	0.1763050	45.000
## 65	0.1763050	45.000
## 66	0.3020960	60.000
## 67	0.1567734	42.000
## 68	0.4319405	72.000
## 69	0.4090636	70.000
## 70	0.3537520	65.000
## 71	0.1506664	41.000
## 72	0.2826740	58.000
## 73	0.5833183	85.000

## 74	0.3537520	65.000
## 75	0.3977647	69.000
## 76	0.3020960	60.000
## 77	0.4090636	70.000
## 78	0.1567734	42.000
## 79	0.4667769	75.000
## 80	0.2550066	55.000
## 81	0.4090636	70.000
## 82	0.3755013	67.000
## 83	0.3020960	60.000
## 84	0.5136743	79.000
## 85	0.2922903	59.000
## 86	0.2209936	51.000
## 87	0.2550066	55.000
## 88	0.3537520	65.000
## 89	0.1695902	44.000
## 90	0.2732518	57.000
## 91	0.4090636	70.000
## 92	0.3020960	60.000
## 93	0.1567734	42.000
## 94	0.3020960	60.000
## 95	0.2826740	58.000
## 96	0.2826740	58.000
## 97	0.3325916	63.000
## 98	0.4090636	70.000
## 99	0.3020960	60.000
## 100	0.3325916	63.000
## 101	0.3537520	65.000
## 102	0.4667769	75.000
## 103	0.5253933	80.000
## 104	0.1567734	42.000
## 105	0.3020960	60.000
## 106	0.4319405	72.000
## 107	0.2550066	55.000
## 108	0.1763050	45.000
## 109	0.3325916	63.000
## 110	0.1763050	45.000
## 111	0.5833183	85.000
## 112	0.2550066	55.000
## 113	0.2130169	50.000
## 114	0.4090636	70.000
## 115	0.3020960	60.000
## 116	0.2826740	58.000
## 117	0.3020960	60.000
## 118	0.5833183	85.000
## 119	0.3537520	65.000
## 120	0.5946833	86.000
## 121	0.3020960	60.000
## 122	0.3645575	66.000
## 123	0.3020960	60.000

## 124	0.3020960	60.000
## 125	0.3020960	60.000
## 126	0.1630804	43.000
## 127	0.1832271	46.000
## 128	0.2826740	58.000
## 129	0.3120856	61.000
## 130	0.2375814	53.000
## 131	0.2375814	53.000
## 132	0.3020960	60.000
## 133	0.1832271	46.000
## 134	0.3325916	63.000
## 135	0.5370845	81.000
## 136	0.4667769	75.000
## 137	0.3537520	65.000
## 138	0.3865737	68.000
## 139	0.3222531	62.000
## 140	0.2130169	50.000
## 141	0.5253933	80.000
## 142	0.1832271	46.000
## 143	0.2130169	50.000
## 144	0.3120856	61.000
## 145	0.4319405	72.000
## 146	0.2130169	50.000
## 147	0.2291821	52.000
## 148	0.3430939	64.000
## 149	0.4667769	75.000
## 150	0.3020960	60.000
## 151	0.4319405	72.000
## 152	0.3222531	62.000
## 153	0.2130169	50.000
## 154	0.2130169	50.000
## 155	0.3537520	65.000
## 156	0.3020960	60.000
## 157	0.2291821	52.000
## 158	0.2130169	50.000
## 159	0.5833183	85.000
## 160	0.2922903	59.000
## 161	0.3645575	66.000
## 162	0.1763050	45.000
## 163	0.3325916	63.000
## 164	0.2130169	50.000
## 165	0.1763050	45.000
## 166	0.5253933	80.000
## 167	0.2375814	53.000
## 168	0.2922903	59.000
## 169	0.3537520	65.000
## 170	0.4090636	70.000
## 171	0.2209936	51.000
## 172	0.2291821	52.000
## 173	0.4090636	70.000

## 174	0.2130169	50.000
## 175	0.3537520	65.000
## 176	0.3020960	60.000
## 177	0.3977647	69.000
## 178	0.2052522	49.000
## 179	0.3325916	63.000
## 180	0.2550066	55.000
## 181	0.1447564	40.000
## 182	0.2922903	59.000
## 183	0.3537520	65.000
## 184	0.4667769	75.000
## 185	0.2826740	58.000
## 186	0.3087390	60.667
## 187	0.2130169	50.000
## 188	0.3020960	60.000
## 189	0.3087390	60.667
## 190	0.1447564	40.000
## 191	0.5253933	80.000
## 192	0.3430939	64.000
## 193	0.2130169	50.000
## 194	0.4434953	73.000
## 195	0.1763050	45.000
## 196	0.4902041	77.000
## 197	0.1763050	45.000
## 198	0.3537520	65.000
## 199	0.2130169	50.000
## 200	0.3020960	60.000
## 201	0.3325916	63.000
## 202	0.1763050	45.000
## 203	0.4090636	70.000
## 204	0.3020960	60.000
## 205	0.5019403	78.000
## 206	0.2130169	50.000
## 207	0.1447564	40.000
## 208	0.5833183	85.000
## 209	0.3020960	60.000
## 210	0.2052522	49.000
## 211	0.4090636	70.000
## 212	0.2130169	50.000
## 213	0.5019403	78.000
## 214	0.1976995	48.000
## 215	0.3537520	65.000
## 216	0.4434953	73.000
## 217	0.4090636	70.000
## 218	0.2461902	54.000
## 219	0.3865737	68.000
## 220	0.2550066	55.000
## 221	0.4434953	73.000
## 222	0.3537520	65.000
## 223	0.1567734	42.000

## 224	0.1903581	47.000
## 225	0.2826740	58.000
## 226	0.4667769	75.000
## 227	0.2826740	58.000
## 228	0.2550066	55.000
## 229	0.3537520	65.000
## 230	0.4319405	72.000
## 231	0.3020960	60.000
## 232	0.4090636	70.000
## 233	0.1447564	40.000
## 234	0.2375814	53.000
## 235	0.2375814	53.000
## 236	0.4902041	77.000
## 237	0.4667769	75.000
## 238	0.4090636	70.000
## 239	0.3537520	65.000
## 240	0.2550066	55.000
## 241	0.4090636	70.000
## 242	0.3537520	65.000
## 243	0.1447564	40.000
## 244	0.4434953	73.000
## 245	0.2461902	54.000
## 246	0.3120856	61.000
## 247	0.2550066	55.000
## 248	0.3430939	64.000
## 249	0.1447564	40.000
## 250	0.2375814	53.000
## 251	0.2130169	50.000
## 252	0.2550066	55.000
## 253	0.2130169	50.000
## 254	0.4090636	70.000
## 255	0.2375814	53.000
## 256	0.2291821	52.000
## 257	0.3537520	65.000
## 258	0.2826740	58.000
## 259	0.1763050	45.000
## 260	0.2375814	53.000
## 261	0.2550066	55.000
## 262	0.3222531	62.000
## 263	0.3537520	65.000
## 264	0.3865737	68.000
## 265	0.3120856	61.000
## 266	0.2130169	50.000
## 267	0.2550066	55.000
## 268	0.2640282	56.000
## 269	0.1763050	45.000
## 270	0.1447564	40.000
## 271	0.1695902	44.000
## 272	0.2209936	51.000
## 273	0.3755013	67.000

```
## 274          0.1567734 42.000
## 275          0.3020960 60.000
## 276          0.1763050 45.000
## 277          0.4090636 70.000
## 278          0.4090636 70.000
## 279          0.2130169 50.000
## 280          0.2550066 55.000
## 281          0.4090636 70.000
## 282          0.4090636 70.000
## 283          0.1567734 42.000
## 284          0.3537520 65.000
## 285          0.2130169 50.000
## 286          0.2550066 55.000
## 287          0.3020960 60.000
## 288          0.1763050 45.000
## 289          0.3537520 65.000
## 290          0.6390307 90.000
## 291          0.1763050 45.000
## 292          0.3020960 60.000
## 293          0.2291821 52.000
## 294          0.3325916 63.000
## 295          0.3222531 62.000
## 296          0.2550066 55.000
## 297          0.1763050 45.000
## 298          0.1763050 45.000
## 299          0.2130169 50.000
```

#Implementing multiple logistic model using all the variables

```
logistic <- glm(DEATH_EVENT ~ ., data=dataset, family="binomial")
summary(logistic)
```

```
##
```

```
## Call:
```

```
## glm(formula = DEATH_EVENT ~ ., family = "binomial", data = dataset)
```

```
##
```

```
## Deviance Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -2.1848 -0.5706 -0.2401  0.4466  2.6668
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   1.018e+01  5.657e+00  1.801 0.071774 .
## age           4.742e-02  1.580e-02  3.001 0.002690 **
## anaemiaYes    -7.470e-03  3.605e-01 -0.021 0.983467
## creatinine_phosphokinase 2.222e-04  1.779e-04  1.249 0.211684
## diabetesYes   1.451e-01  3.512e-01  0.413 0.679380
## ejection_fraction -7.666e-02  1.633e-02 -4.695 2.67e-06 ***
## high_blood_pressureYes -1.027e-01  3.587e-01 -0.286 0.774688
## platelets     -1.200e-06  1.889e-06 -0.635 0.525404
## serum_creatinine 6.661e-01  1.815e-01  3.670 0.000242 ***
```

```
## serum_sodium          -6.698e-02  3.974e-02  -1.686 0.091855 .
## sexmale                -5.337e-01  4.139e-01  -1.289 0.197299
## smokingYes             -1.349e-02  4.126e-01  -0.033 0.973915
## time                   -2.104e-02  3.014e-03  -6.981 2.92e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 375.35  on 298  degrees of freedom
## Residual deviance: 219.55  on 286  degrees of freedom
## AIC: 245.55
##
## Number of Fisher Scoring iterations: 6
```

#Based on the summary, we can say age, ejection_fraction, serum_creatinine and time are significant variables for predicting death event
#For every one year increase in Age the odds of survival on a log scale is expected to increase by 0.047, holding all the other variables constant
#For every one percent increase in ejection_fraction the odds of survival on a log scale is expected to decrease by 0.076, holding all the other variables constant
#For every one mg/dL increase in serum_creatinine the odds of survival on a log scale is expected to increase by 0.66, holding all the other variables constant
#For every one day increase in follow-up period the odds of survival on a log scale is expected to decrease by 0.021, holding all the other variables constant

#Calculating the overall "Pseudo R-squared" and its p-value

```
ll.null2 <- logistic$null.deviance/-2
```

```
ll.proposed2 <- logistic$deviance/-2
```

```
## McFadden's Pseudo R^2 = [ LL(Null) - LL(Proposed) ] / LL(Null)
(ll.null2 - ll.proposed2) / ll.null2
```

```
## [1] 0.4150663
```

The p-value for the R^2

```
1 - pchisq(2*(ll.proposed2 - ll.null2), df=(length(logistic$coefficients)-1))
```

```
## [1] 0
```

#Plot the data

```
predicted.data2 <-
```

```
data.frame(probability.of.death_event=logistic$fitted.values, DEATH_EVENT=data
set$DEATH_EVENT)
```

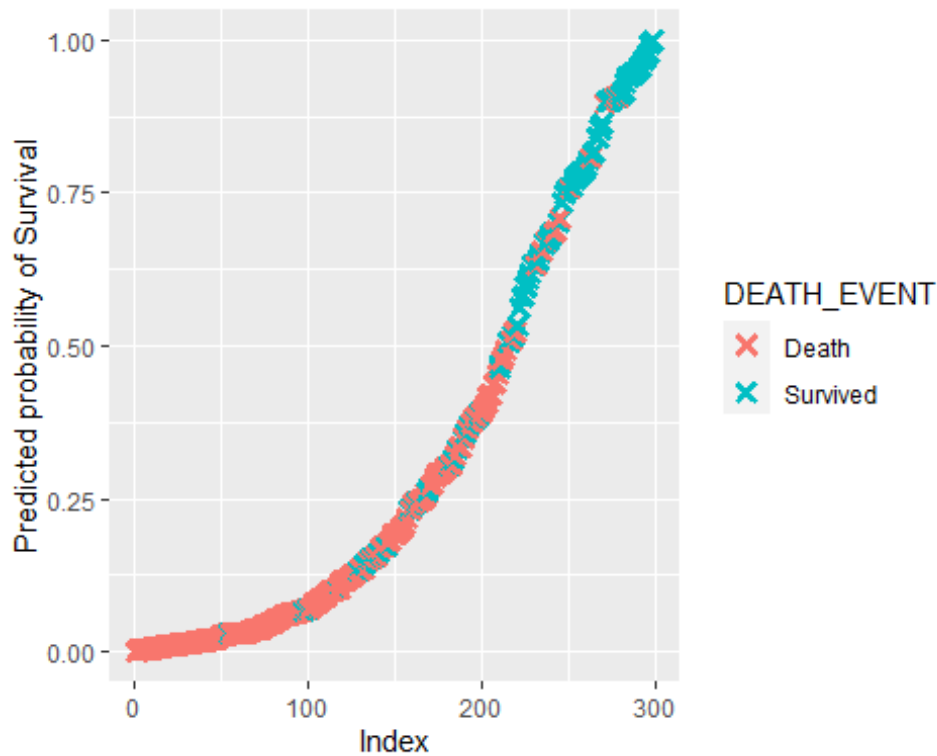
```
predicted.data2 <-
```

```
predicted.data2[order(predicted.data2$probability.of.death_event,
decreasing=FALSE),]
```

```
predicted.data2$rank <- 1:nrow(predicted.data2)
```

```
ggplot(data=predicted.data2, aes(x=rank, y=probability.of.death_event)) +
```

```
geom_point(aes(color=DEATH_EVENT), alpha=1, shape=4, stroke=2) +
xlab("Index") +
ylab("Predicted probability of Survival")
```



#Confusion Matrix

```
confusion_matrix(logistic)
```

```
##               Predicted Death Predicted Survived Total
## Actual Death           187           16      203
## Actual Survived         27           69       96
## Total                   214           85      299
```

```
pdata <- predict(logistic,newdata=dataset,type="response" )
pdata
```

```
##           1           2           3           4           5
## 0.9793073820 0.9067149740 0.9570132585 0.8998995179 0.9953911997
##           6           7           8           9          10
## 0.9470249475 0.9665948236 0.3304746994 0.3889735944 0.9993527922
##          11          12          13          14          15
## 0.9685325921 0.7888549332 0.7021269470 0.5074851512 0.6844173509
##          16          17          18          19          20
## 0.7702631162 0.8179328418 0.9148003004 0.8997178303 0.7777244922
##          21          22          23          24          25
## 0.9066767238 0.9011333759 0.6393118753 0.1557585354 0.9550761551
##          26          27          28          29          30
## 0.8393907390 0.9020743771 0.6095748828 0.9747028378 0.9280890356
```

##	31	32	33	34	35
##	0.9385809020	0.9147338133	0.7471777699	0.7334826829	0.3581656449
##	36	37	38	39	40
##	0.9509908151	0.7359553845	0.5993106534	0.9080717450	0.8551062002
##	41	42	43	44	45
##	0.9385259276	0.7586802972	0.5619806901	0.5094117878	0.2675407498
##	46	47	48	49	50
##	0.6188645962	0.7547723020	0.3810741199	0.9927237792	0.6318486141
##	51	52	53	54	55
##	0.7851493399	0.7842762388	0.9444024848	0.5162619559	0.8415566938
##	56	57	58	59	60
##	0.9435669099	0.8075643832	0.3328296287	0.6419562917	0.8643877667
##	61	62	63	64	65
##	0.7769259139	0.6666495055	0.3639674646	0.1735805857	0.0237291876
##	66	67	68	69	70
##	0.9478586912	0.7946040072	0.6779812327	0.7737882333	0.7608804713
##	71	72	73	74	75
##	0.1019430790	0.4080523162	0.9097212170	0.2844243526	0.8137059736
##	76	77	78	79	80
##	0.5811662333	0.1220379662	0.1576972850	0.5223566391	0.1964405102
##	81	82	83	84	85
##	0.6332520973	0.2324247683	0.9034577751	0.4772643585	0.5941891923
##	86	87	88	89	90
##	0.0569575440	0.3036732295	0.0636962347	0.1134280994	0.3919251701
##	91	92	93	94	95
##	0.2907571900	0.2908714510	0.0614628641	0.6723733875	0.2520900088
##	96	97	98	99	100
##	0.0445136022	0.6704862295	0.1734651449	0.6538056760	0.3863843064
##	101	102	103	104	105
##	0.4943149263	0.3202369931	0.6882833309	0.4191760614	0.2470874833
##	106	107	108	109	110
##	0.6319411124	0.1821007909	0.2477920851	0.3840529029	0.0981006911
##	111	112	113	114	115
##	0.2304142893	0.3026544365	0.5231874578	0.1548905226	0.4215619668
##	116	117	118	119	120
##	0.3087609160	0.0787251124	0.4831822056	0.0934640215	0.8033632326
##	121	122	123	124	125
##	0.0864589139	0.4537855903	0.2958473090	0.3740274799	0.6830300680
##	126	127	128	129	130
##	0.1151429894	0.8589178599	0.0644979297	0.2750398670	0.4376849402
##	131	132	133	134	135
##	0.0303635301	0.8967572521	0.1677982416	0.0297029145	0.7078888336
##	136	137	138	139	140
##	0.3345708565	0.0982498578	0.7591174037	0.4177665248	0.2187830430
##	141	142	143	144	145
##	0.4645497176	0.1481755844	0.3959698154	0.1802048594	0.5781942743
##	146	147	148	149	150
##	0.1296345117	0.1933096070	0.0609660520	0.6514053794	0.2707977142
##	151	152	153	154	155
##	0.5047835199	0.0315179140	0.0692556725	0.2092043894	0.3024749905

##	156	157	158	159	160
##	0.4417145641	0.2570049451	0.2431913414	0.2902174211	0.0657354759
##	161	162	163	164	165
##	0.1989961836	0.0814535296	0.1168817669	0.2452025047	0.1794601269
##	166	167	168	169	170
##	0.5343327748	0.0192403470	0.6843262164	0.2022509509	0.2913151203
##	171	172	173	174	175
##	0.1373227695	0.0893313081	0.0183791751	0.1722622760	0.1343270196
##	176	177	178	179	180
##	0.0154090848	0.1919585844	0.0362649306	0.0159363280	0.0485935443
##	181	182	183	184	185
##	0.0766684346	0.2369616329	0.3031166637	0.3155436536	0.1517264800
##	186	187	188	189	190
##	0.1434887697	0.0285556070	0.3488701344	0.0961830070	0.0149776716
##	191	192	193	194	195
##	0.3492802398	0.0307853801	0.0424645873	0.1522858366	0.1628268752
##	196	197	198	199	200
##	0.0671893288	0.0464337149	0.1266277312	0.2023944130	0.4046519018
##	201	202	203	204	205
##	0.0415688917	0.0040153990	0.0106405152	0.3796646692	0.0629106092
##	206	207	208	209	210
##	0.0284131170	0.0165001172	0.1355694216	0.0798109326	0.0872443260
##	211	212	213	214	215
##	0.2454660433	0.0030820518	0.0330756155	0.1454391737	0.0520072769
##	216	217	218	219	220
##	0.1191573382	0.0275590066	0.4691565071	0.1037258914	0.0384074871
##	221	222	223	224	225
##	0.3704402368	0.0108993974	0.0135508102	0.0608954966	0.0608506722
##	226	227	228	229	230
##	0.0668953402	0.1150933760	0.0419386004	0.7933814662	0.2438555978
##	231	232	233	234	235
##	0.2615818011	0.0726026980	0.0121871592	0.0253174306	0.0123110291
##	236	237	238	239	240
##	0.0189728557	0.0101710435	0.1249018695	0.0646063860	0.0131484281
##	241	242	243	244	245
##	0.0651459275	0.0903769651	0.0137518690	0.0542663536	0.0473135723
##	246	247	248	249	250
##	0.0363141287	0.0691433689	0.1913641876	0.0088378852	0.0543461698
##	251	252	253	254	255
##	0.0372647778	0.0303225759	0.0102365430	0.1227338442	0.0026485134
##	256	257	258	259	260
##	0.0196392330	0.0609922401	0.0143742576	0.0263795292	0.0041658398
##	261	262	263	264	265
##	0.0108066400	0.0296381264	0.1324948593	0.0069219640	0.0197666677
##	266	267	268	269	270
##	0.0075506559	0.1029223514	0.0183755862	0.0100017625	0.0116898012
##	271	272	273	274	275
##	0.0294131610	0.0186204951	0.0462272307	0.0033243461	0.0302263065
##	276	277	278	279	280
##	0.0075890060	0.0332975887	0.0287535899	0.0301428442	0.0194953196

```
##          281          282          283          284          285
## 0.0339186983 0.0682835446 0.0981659815 0.0197277171 0.0038837855
##          286          287          288          289          290
## 0.0105702565 0.0184985489 0.0032980392 0.0277978553 0.0357715039
##          291          292          293          294          295
## 0.0020211380 0.0165845423 0.0053715227 0.0133320619 0.0067397574
##          296          297          298          299
## 0.0140111896 0.0008008297 0.0059965251 0.0026973899
```

```
dataset$DEATH_EVENT
```

```
## [1] Survived Survived Survived Survived Survived Survived Survived
## [8] Survived Survived Survived Survived Survived Survived Survived
## [15] Death    Survived Survived Survived Survived Survived Death
## [22] Survived Survived Death    Survived Survived Survived Survived
## [29] Survived Survived Survived Survived Survived Death    Survived
## [36] Survived Survived Survived Death    Survived Survived Survived
## [43] Survived Death    Survived Survived Survived Survived Survived
## [50] Survived Survived Survived Survived Survived Survived Survived
## [57] Death    Death    Survived Survived Survived Survived Death
## [64] Survived Death    Survived Survived Survived Survived Survived
## [71] Death    Death    Survived Death    Survived Survived Death
## [78] Death    Death    Death    Death    Death    Survived Death
## [85] Survived Death    Death    Death    Death    Death    Death    Death
## [92] Death    Death    Survived Death    Death    Death    Death    Death
## [99] Death    Death    Death    Death    Death    Death    Death    Death
## [106] Survived Death    Death    Death    Death    Death    Survived Death
## [113] Death    Survived Death    Death    Death    Death    Death    Death
## [120] Survived Death    Death    Death    Death    Death    Survived Death
## [127] Survived Death    Death    Death    Death    Death    Death    Death
## [134] Death    Death    Death    Death    Death    Death    Death    Death
## [141] Survived Death    Death    Death    Death    Survived Death    Death
## [148] Death    Survived Death    Survived Death    Death    Death    Death
## [155] Death    Death    Death    Death    Death    Death    Death    Death
## [162] Death    Death    Survived Survived Survived Death    Survived
## [169] Death    Death    Death    Death    Death    Death    Death    Death
## [176] Death    Death    Death    Death    Death    Death    Death    Survived
## [183] Survived Survived Survived Survived Survived Survived Death
## [190] Death    Death    Death    Death    Death    Death    Survived Survived
## [197] Death    Death    Death    Death    Death    Death    Death    Death
## [204] Death    Death    Death    Death    Death    Death    Death    Death
## [211] Death    Death    Death    Survived Death    Death    Death    Death
## [218] Survived Death    Death    Survived Death    Death    Death    Death
## [225] Death    Death    Death    Death    Death    Death    Death    Survived
## [232] Death    Death    Death    Death    Death    Death    Death    Death
## [239] Death    Death    Death    Death    Death    Death    Death    Death
## [246] Death    Survived Death    Death    Death    Death    Death    Death
## [253] Death    Death    Death    Death    Death    Death    Death    Death
## [260] Death    Death    Death    Survived Death    Death    Death    Death
## [267] Survived Death    Death    Death    Death    Death    Death    Death
```

```

## [274] Death    Death    Death    Death    Death    Death    Death
## [281] Death    Death    Death    Death    Death    Death    Death
## [288] Death    Death    Death    Death    Death    Death    Death
## [295] Death    Death    Death    Death    Death
## Levels: Death Survived

pdataF <- as.factor(ifelse(test=as.numeric(pdata>0.5) == 0, yes="Death",
no="Survived"))

confusionMatrix(pdataF, dataset$DEATH_EVENT)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction Death Survived
##   Death      187      27
##   Survived    16      69
##
##              Accuracy : 0.8562
##              95% CI : (0.8112, 0.8939)
##   No Information Rate : 0.6789
##   P-Value [Acc > NIR] : 1.627e-12
##
##              Kappa : 0.6599
##
##  Mcnemar's Test P-Value : 0.1273
##
##              Sensitivity : 0.9212
##              Specificity : 0.7188
##              Pos Pred Value : 0.8738
##              Neg Pred Value : 0.8118
##              Prevalence : 0.6789
##              Detection Rate : 0.6254
##              Detection Prevalence : 0.7157
##              Balanced Accuracy : 0.8200
##
##              'Positive' Class : Death
##
#ROC curve
roc(dataset$DEATH_EVENT, logistic$fitted.values, plot=TRUE)

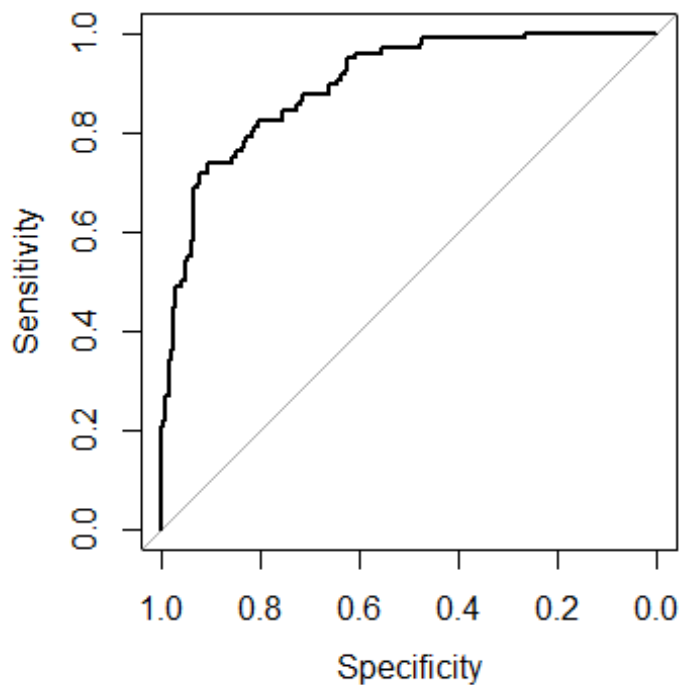
## Setting levels: control = Death, case = Survived
## Setting direction: controls < cases
##
## Call:
## roc.default(response = dataset$DEATH_EVENT, predictor =
logistic$fitted.values,      plot = TRUE)
##

```

```
## Data: logistic$fitted.values in 203 controls (dataset$DEATH_EVENT Death) <
96 cases (dataset$DEATH_EVENT Survived).
## Area under the curve: 0.8972

par(pty = "s")
roc(dataset$DEATH_EVENT,logistic$fitted.values,plot=TRUE)

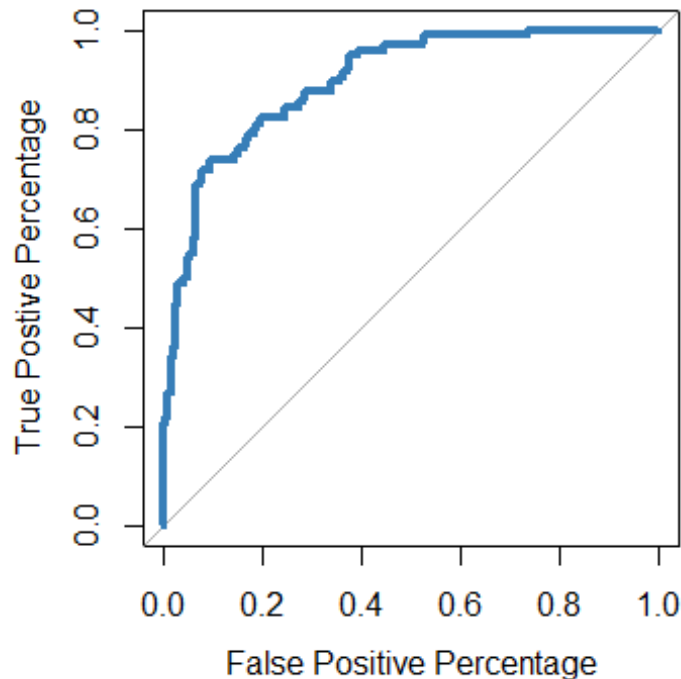
## Setting levels: control = Death, case = Survived
## Setting direction: controls < cases
```



```
##
## Call:
## roc.default(response = dataset$DEATH_EVENT, predictor =
logistic$fitted.values,      plot = TRUE)
##
## Data: logistic$fitted.values in 203 controls (dataset$DEATH_EVENT Death) <
96 cases (dataset$DEATH_EVENT Survived).
## Area under the curve: 0.8972

roc(dataset$DEATH_EVENT,logistic$fitted.values,plot=TRUE, legacy.axes=TRUE,
xlab="False Positive Percentage", ylab="True Positive Percentage",
col="#377eb8", lwd=4)

## Setting levels: control = Death, case = Survived
## Setting direction: controls < cases
```



```
##
## Call:
## roc.default(response = dataset$DEATH_EVENT, predictor =
logistic$fitted.values,      plot = TRUE, legacy.axes = TRUE, xlab = "False
Positive Percentage",      ylab = "True Postive Percentage", col = "#377eb8",
lwd = 4)
##
## Data: logistic$fitted.values in 203 controls (dataset$DEATH_EVENT Death) <
96 cases (dataset$DEATH_EVENT Survived).
## Area under the curve: 0.8972

roc.info <- roc(dataset$DEATH_EVENT, logistic$fitted.values,
legacy.axes=TRUE)

## Setting levels: control = Death, case = Survived
## Setting direction: controls < cases

str(roc.info)

## List of 15
## $ percent          : logi FALSE
## $ sensitivities     : num [1:300] 1 1 1 1 1 1 1 1 1 1 ...
## $ specificities     : num [1:300] 0 0.00493 0.00985 0.01478 0.0197 ...
## $ thresholds        : num [1:300] -Inf 0.00141 0.00233 0.00267 0.00289
...
## $ direction        : chr "<"
## $ cases             : Named num [1:96] 0.979 0.907 0.957 0.9 0.995 ...
```

```

##   ..- attr(*, "names")= chr [1:96] "1" "2" "3" "4" ...
##   $ controls           : Named num [1:203] 0.684 0.907 0.156 0.733 0.908 ...
##   ..- attr(*, "names")= chr [1:203] "15" "21" "24" "34" ...
##   $ fun.sesp           :function (thresholds, controls, cases, direction)
##   $ auc                 : 'auc' num 0.897
##   ..- attr(*, "partial.auc")= logi FALSE
##   ..- attr(*, "percent")= logi FALSE
##   ..- attr(*, "roc")=List of 15
##   .. ..$ percent       : logi FALSE
##   .. ..$ sensitivities  : num [1:300] 1 1 1 1 1 1 1 1 1 1 ...
##   .. ..$ specificities  : num [1:300] 0 0.00493 0.00985 0.01478 0.0197
##   ...
##   .. ..$ thresholds    : num [1:300] -Inf 0.00141 0.00233 0.00267
##   0.00289 ...
##   .. ..$ direction     : chr "<"
##   .. ..$ cases         : Named num [1:96] 0.979 0.907 0.957 0.9 0.995
##   ...
##   .. .. ..- attr(*, "names")= chr [1:96] "1" "2" "3" "4" ...
##   .. .. ..$ controls   : Named num [1:203] 0.684 0.907 0.156 0.733
##   0.908 ...
##   .. .. ..- attr(*, "names")= chr [1:203] "15" "21" "24" "34" ...
##   .. .. ..$ fun.sesp   :function (thresholds, controls, cases,
##   direction)
##   .. .. ..$ auc        : 'auc' num 0.897
##   .. .. ..- attr(*, "partial.auc")= logi FALSE
##   .. .. ..- attr(*, "percent")= logi FALSE
##   .. .. ..- attr(*, "roc")=List of 8
##   .. .. .. ..$ percent  : logi FALSE
##   .. .. .. ..$ sensitivities: num [1:300] 1 1 1 1 1 1 1 1 1 1 ...
##   .. .. .. ..$ specificities: num [1:300] 0 0.00493 0.00985 0.01478 0.0197
##   ...
##   .. .. .. ..$ thresholds : num [1:300] -Inf 0.00141 0.00233 0.00267
##   0.00289 ...
##   .. .. .. ..$ direction  : chr "<"
##   .. .. .. ..$ cases     : Named num [1:96] 0.979 0.907 0.957 0.9 0.995
##   ...
##   .. .. .. ..- attr(*, "names")= chr [1:96] "1" "2" "3" "4" ...
##   .. .. .. ..$ controls   : Named num [1:203] 0.684 0.907 0.156 0.733
##   0.908 ...
##   .. .. .. ..- attr(*, "names")= chr [1:203] "15" "21" "24" "34" ...
##   .. .. .. ..$ fun.sesp   :function (thresholds, controls, cases,
##   direction)
##   .. .. .. ..- attr(*, "class")= chr "roc"
##   .. .. ..$ call          : language roc.default(response =
##   dataset$DEATH_EVENT, predictor = logistic$fitted.values, legacy.axes =
##   TRUE)
##   .. .. ..$ original.predictor: Named num [1:299] 0.979 0.907 0.957 0.9 0.995
##   ...
##   .. .. ..- attr(*, "names")= chr [1:299] "1" "2" "3" "4" ...
##   .. .. ..$ original.response : Factor w/ 2 levels "Death","Survived": 2 2 2

```

```

2 2 2 2 2 2 2 ...
## .. ..$ predictor          : Named num [1:299] 0.979 0.907 0.957 0.9 0.995
...
## .. ..$- attr(*, "names")= chr [1:299] "1" "2" "3" "4" ...
## .. ..$ response           : Factor w/ 2 levels "Death","Survived": 2 2 2
2 2 2 2 2 2 2 ...
## .. ..$ levels             : chr [1:2] "Death" "Survived"
## .. ..$- attr(*, "class")= chr "roc"
## $ call                     : language roc.default(response =
dataset$DEATH_EVENT, predictor = logistic$fitted.values,      legacy.axes =
TRUE)
## $ original.predictor: Named num [1:299] 0.979 0.907 0.957 0.9 0.995 ...
## ..$- attr(*, "names")= chr [1:299] "1" "2" "3" "4" ...
## $ original.response : Factor w/ 2 levels "Death","Survived": 2 2 2 2 2 2
2 2 2 2 2 ...
## $ predictor              : Named num [1:299] 0.979 0.907 0.957 0.9 0.995 ...
## ..$- attr(*, "names")= chr [1:299] "1" "2" "3" "4" ...
## $ response               : Factor w/ 2 levels "Death","Survived": 2 2 2 2 2 2
2 2 2 2 2 ...
## $ levels                 : chr [1:2] "Death" "Survived"
## - attr(*, "class")= chr "roc"

roc.df <- data.frame(tpp=roc.info$sensitivities*100, ## tpp = true positive
percentage
                    fpp=(1 - roc.info$specificities)*100, ## fpp = false
positive precentage
                    thresholds=roc.info$thresholds)
roc.df

##          tpp          fpp thresholds
## 1  100.000000 100.000000      -Inf
## 2  100.000000  99.5073892 0.001410984
## 3  100.000000  99.0147783 0.002334826
## 4  100.000000  98.5221675 0.002672952
## 5  100.000000  98.0295567 0.002889721
## 6  100.000000  97.5369458 0.003190045
## 7  100.000000  97.0443350 0.003311193
## 8  100.000000  96.5517241 0.003604066
## 9  100.000000  96.0591133 0.003949592
## 10 100.000000  95.5665025 0.004090619
## 11 100.000000  95.0738916 0.004768681
## 12 100.000000  94.5812808 0.005684024
## 13 100.000000  94.0886700 0.006368141
## 14 100.000000  93.5960591 0.006830861
## 15 100.000000  93.1034483 0.007236310
## 16 100.000000  92.6108374 0.007569831
## 17 100.000000  92.1182266 0.008213446
## 18 100.000000  91.6256158 0.009419824
## 19 100.000000  91.1330049 0.010086403
## 20 100.000000  90.6403941 0.010203793

```

## 21	100.000000	90.1477833	0.010403400
## 22	100.000000	89.6551724	0.010605386
## 23	100.000000	89.1625616	0.010723578
## 24	100.000000	88.6699507	0.010853019
## 25	100.000000	88.1773399	0.011294599
## 26	100.000000	87.6847291	0.011938480
## 27	100.000000	87.1921182	0.012249094
## 28	100.000000	86.6995074	0.012729729
## 29	100.000000	86.2068966	0.013240245
## 30	100.000000	85.7142857	0.013441436
## 31	100.000000	85.2216749	0.013651340
## 32	100.000000	84.7290640	0.013881529
## 33	100.000000	84.2364532	0.014192724
## 34	100.000000	83.7438424	0.014675965
## 35	100.000000	83.2512315	0.015193378
## 36	100.000000	82.7586207	0.015672706
## 37	100.000000	82.2660099	0.016218223
## 38	100.000000	81.7733990	0.016542330
## 39	100.000000	81.2807882	0.017480064
## 40	100.000000	80.7881773	0.018377381
## 41	100.000000	80.2955665	0.018438862
## 42	100.000000	79.8029557	0.018559522
## 43	100.000000	79.3103448	0.018796675
## 44	100.000000	78.8177340	0.019106601
## 45	100.000000	78.3251232	0.019367833
## 46	100.000000	77.8325123	0.019567276
## 47	100.000000	77.3399015	0.019683475
## 48	100.000000	76.8472906	0.019747192
## 49	100.000000	76.3546798	0.021747928
## 50	100.000000	75.8620690	0.024523309
## 51	100.000000	75.3694581	0.025848480
## 52	100.000000	74.8768473	0.026969268
## 53	100.000000	74.3842365	0.027678431
## 54	100.000000	73.8916256	0.028105486
## 55	100.000000	73.3990148	0.028484362
## 56	98.958333	73.3990148	0.028654598
## 57	98.958333	72.9064039	0.029083375
## 58	98.958333	72.4137931	0.029525644
## 59	98.958333	71.9211823	0.029670520
## 60	98.958333	71.4285714	0.029922879
## 61	98.958333	70.9359606	0.030184575
## 62	98.958333	70.4433498	0.030274441
## 63	98.958333	69.9507389	0.030343053
## 64	98.958333	69.4581281	0.030574455
## 65	98.958333	68.9655172	0.031151647
## 66	98.958333	68.4729064	0.032296765
## 67	98.958333	67.9802956	0.033186602
## 68	98.958333	67.4876847	0.033608143
## 69	98.958333	66.9950739	0.034845101
## 70	98.958333	66.5024631	0.036018217

## 71	98.958333	66.0098522	0.036289530
## 72	98.958333	65.5172414	0.036789453
## 73	98.958333	65.0246305	0.037836132
## 74	98.958333	64.5320197	0.039988189
## 75	98.958333	64.0394089	0.041753746
## 76	98.958333	63.5467980	0.042201594
## 77	98.958333	63.0541872	0.043489095
## 78	98.958333	62.5615764	0.045370416
## 79	98.958333	62.0689655	0.046330473
## 80	98.958333	61.5763547	0.046873644
## 81	98.958333	61.0837438	0.047953558
## 82	98.958333	60.5911330	0.050300411
## 83	98.958333	60.0985222	0.053136815
## 84	98.958333	59.6059113	0.054306262
## 85	98.958333	59.1133005	0.055651857
## 86	98.958333	58.6206897	0.058904108
## 87	98.958333	58.1280788	0.060873084
## 88	98.958333	57.6354680	0.060930774
## 89	98.958333	57.1428571	0.060979146
## 90	98.958333	56.6502463	0.061227552
## 91	98.958333	56.1576355	0.062186737
## 92	98.958333	55.6650246	0.063303422
## 93	98.958333	55.1724138	0.064097082
## 94	98.958333	54.6798030	0.064552158
## 95	98.958333	54.1871921	0.064876157
## 96	98.958333	53.6945813	0.065440702
## 97	98.958333	53.2019704	0.066315408
## 98	98.958333	52.7093596	0.067042335
## 99	97.916667	52.7093596	0.067736437
## 100	97.916667	52.2167488	0.068713457
## 101	96.875000	52.2167488	0.069199521
## 102	96.875000	51.7241379	0.070929185
## 103	96.875000	51.2315271	0.074635566
## 104	96.875000	50.7389163	0.077696774
## 105	96.875000	50.2463054	0.079268022
## 106	96.875000	49.7536946	0.080632231
## 107	96.875000	49.2610837	0.083956222
## 108	96.875000	48.7684729	0.086851620
## 109	96.875000	48.2758621	0.088287817
## 110	96.875000	47.7832512	0.089854137
## 111	96.875000	47.2906404	0.091920493
## 112	96.875000	46.7980296	0.094823514
## 113	96.875000	46.3054187	0.097141849
## 114	96.875000	45.8128079	0.098133336
## 115	96.875000	45.3201970	0.098207920
## 116	96.875000	44.8275862	0.100096468
## 117	96.875000	44.3349754	0.102432715
## 118	95.833333	44.3349754	0.103324121
## 119	95.833333	43.8423645	0.108576995
## 120	95.833333	43.3497537	0.114260738

## 121	95.833333	42.8571429	0.115118183
## 122	95.833333	42.3645320	0.116012378
## 123	95.833333	41.8719212	0.118019553
## 124	95.833333	41.3793103	0.120597652
## 125	95.833333	40.8866995	0.122385905
## 126	95.833333	40.3940887	0.123817857
## 127	95.833333	39.9014778	0.125764800
## 128	95.833333	39.4088670	0.128131121
## 129	95.833333	38.9162562	0.131064686
## 130	94.791667	38.9162562	0.133410939
## 131	94.791667	38.4236453	0.134948221
## 132	94.791667	37.9310345	0.136446096
## 133	94.791667	37.4384236	0.140405770
## 134	93.750000	37.4384236	0.144463972
## 135	92.708333	37.4384236	0.146807379
## 136	92.708333	36.9458128	0.149951032
## 137	91.666667	36.9458128	0.152006158
## 138	91.666667	36.4532020	0.153588180
## 139	90.625000	36.4532020	0.155324529
## 140	90.625000	35.9605911	0.156727910
## 141	90.625000	35.4679803	0.160262080
## 142	89.583333	35.4679803	0.165312558
## 143	89.583333	34.9753695	0.170030259
## 144	89.583333	34.4827586	0.172863710
## 145	89.583333	33.9901478	0.173522865
## 146	88.541667	33.9901478	0.176520356
## 147	87.500000	33.9901478	0.179832493
## 148	87.500000	33.4975369	0.181152825
## 149	87.500000	33.0049261	0.186732489
## 150	87.500000	32.5123153	0.191661386
## 151	87.500000	32.0197044	0.192634096
## 152	87.500000	31.5270936	0.194875059
## 153	87.500000	31.0344828	0.197718347
## 154	87.500000	30.5418719	0.200623567
## 155	87.500000	30.0492611	0.202322682
## 156	87.500000	29.5566502	0.205799401
## 157	87.500000	29.0640394	0.213993716
## 158	87.500000	28.5714286	0.224598666
## 159	86.458333	28.5714286	0.231419529
## 160	86.458333	28.0788177	0.234693201
## 161	85.416667	28.0788177	0.240076487
## 162	85.416667	27.5862069	0.243523470
## 163	85.416667	27.0935961	0.244529051
## 164	84.375000	27.0935961	0.245334274
## 165	84.375000	26.6009852	0.246276763
## 166	84.375000	26.1083744	0.247439784
## 167	84.375000	25.6157635	0.249941047
## 168	84.375000	25.1231527	0.254547477
## 169	84.375000	24.6305419	0.259293373
## 170	83.333333	24.6305419	0.264561275

## 171	82.291667	24.6305419	0.269169232
## 172	82.291667	24.1379310	0.272918791
## 173	82.291667	23.6453202	0.279732110
## 174	82.291667	23.1527094	0.287320887
## 175	82.291667	22.6600985	0.290487306
## 176	82.291667	22.1674877	0.290814320
## 177	82.291667	21.6748768	0.291093286
## 178	82.291667	21.1822660	0.293581215
## 179	82.291667	20.6896552	0.299161150
## 180	82.291667	20.1970443	0.302564713
## 181	82.291667	19.7044335	0.302885550
## 182	81.250000	19.7044335	0.303394947
## 183	81.250000	19.2118227	0.306217073
## 184	81.250000	18.7192118	0.312152285
## 185	80.208333	18.7192118	0.317890323
## 186	80.208333	18.2266010	0.325355846
## 187	79.166667	18.2266010	0.331652164
## 188	79.166667	17.7339901	0.333700243
## 189	79.166667	17.2413793	0.341720495
## 190	78.125000	17.2413793	0.349075187
## 191	78.125000	16.7487685	0.353722942
## 192	77.083333	16.7487685	0.361066555
## 193	77.083333	16.2561576	0.367203851
## 194	76.041667	16.2561576	0.372233858
## 195	76.041667	15.7635468	0.376846075
## 196	76.041667	15.2709360	0.380369395
## 197	75.000000	15.2709360	0.382563511
## 198	75.000000	14.7783251	0.385218605
## 199	75.000000	14.2857143	0.387678950
## 200	73.958333	14.2857143	0.390449382
## 201	73.958333	13.7931034	0.393947493
## 202	73.958333	13.3004926	0.400310859
## 203	73.958333	12.8078818	0.406352109
## 204	73.958333	12.3152709	0.412909420
## 205	73.958333	11.8226601	0.418471293
## 206	73.958333	11.3300493	0.420369014
## 207	73.958333	10.8374384	0.429623454
## 208	73.958333	10.3448276	0.439699752
## 209	73.958333	9.8522167	0.447750077
## 210	73.958333	9.3596059	0.459167654
## 211	72.916667	9.3596059	0.466853112
## 212	71.875000	9.3596059	0.473210433
## 213	71.875000	8.8669951	0.480223282
## 214	71.875000	8.3743842	0.488748566
## 215	71.875000	7.8817734	0.499549223
## 216	70.833333	7.8817734	0.506134336
## 217	69.791667	7.8817734	0.508448469
## 218	69.791667	7.3891626	0.512836872
## 219	68.750000	7.3891626	0.519309297
## 220	68.750000	6.8965517	0.522772048

## 221	68.750000	6.4039409	0.528760116
## 222	67.708333	6.4039409	0.548156732
## 223	66.666667	6.4039409	0.570087482
## 224	65.625000	6.4039409	0.579680254
## 225	64.583333	6.4039409	0.587677713
## 226	63.541667	6.4039409	0.596749923
## 227	62.500000	6.4039409	0.604442768
## 228	61.458333	6.4039409	0.614219739
## 229	60.416667	6.4039409	0.625356605
## 230	59.375000	6.4039409	0.631894863
## 231	58.333333	6.4039409	0.632596605
## 232	58.333333	5.9113300	0.636281986
## 233	57.291667	5.9113300	0.640634083
## 234	56.250000	5.9113300	0.646680836
## 235	55.208333	5.9113300	0.652605528
## 236	55.208333	5.4187192	0.660227591
## 237	54.166667	5.4187192	0.668567868
## 238	54.166667	4.9261084	0.671429809
## 239	53.125000	4.9261084	0.675177310
## 240	52.083333	4.9261084	0.680505650
## 241	51.041667	4.9261084	0.683678142
## 242	50.000000	4.9261084	0.684371784
## 243	50.000000	4.4334975	0.686350341
## 244	50.000000	3.9408867	0.695205139
## 245	48.958333	3.9408867	0.705007890
## 246	48.958333	3.4482759	0.720685758
## 247	48.958333	2.9556650	0.734719034
## 248	47.916667	2.9556650	0.741566577
## 249	46.875000	2.9556650	0.750975036
## 250	45.833333	2.9556650	0.756726300
## 251	44.791667	2.9556650	0.758898850
## 252	44.791667	2.4630542	0.759998937
## 253	43.750000	2.4630542	0.765571794
## 254	42.708333	2.4630542	0.772025675
## 255	41.666667	2.4630542	0.775357074
## 256	40.625000	2.4630542	0.777325203
## 257	39.583333	2.4630542	0.781000366
## 258	38.541667	2.4630542	0.784712789
## 259	37.500000	2.4630542	0.787002137
## 260	36.458333	2.4630542	0.791118200
## 261	36.458333	1.9704433	0.793992737
## 262	35.416667	1.9704433	0.798983620
## 263	34.375000	1.9704433	0.805463808
## 264	34.375000	1.4778325	0.810635178
## 265	33.333333	1.4778325	0.815819408
## 266	32.291667	1.4778325	0.828661790
## 267	31.250000	1.4778325	0.840473716
## 268	30.208333	1.4778325	0.848331447
## 269	29.166667	1.4778325	0.857012030
## 270	28.125000	1.4778325	0.861652813

```
## 271 27.083333 1.4778325 0.880572509
## 272 27.083333 0.9852217 0.898237541
## 273 26.041667 0.9852217 0.899808674
## 274 25.000000 0.9852217 0.900516447
## 275 23.958333 0.9852217 0.901603877
## 276 22.916667 0.9852217 0.902766076
## 277 21.875000 0.9852217 0.905067249
## 278 21.875000 0.4926108 0.906695849
## 279 20.833333 0.4926108 0.907393360
## 280 20.833333 0.0000000 0.908896481
## 281 19.791667 0.0000000 0.912227515
## 282 18.750000 0.0000000 0.914767057
## 283 17.708333 0.0000000 0.921444668
## 284 16.666667 0.0000000 0.933307482
## 285 15.625000 0.0000000 0.938553415
## 286 14.583333 0.0000000 0.941073906
## 287 13.541667 0.0000000 0.943984697
## 288 12.500000 0.0000000 0.945713716
## 289 11.458333 0.0000000 0.947441819
## 290 10.416667 0.0000000 0.949424753
## 291 9.375000 0.0000000 0.953033485
## 292 8.333333 0.0000000 0.956044707
## 293 7.291667 0.0000000 0.961804041
## 294 6.250000 0.0000000 0.967563708
## 295 5.208333 0.0000000 0.971617715
## 296 4.166667 0.0000000 0.977005110
## 297 3.125000 0.0000000 0.986015581
## 298 2.083333 0.0000000 0.994057489
## 299 1.041667 0.0000000 0.997371996
## 300 0.000000 0.0000000 Inf
```

`head(roc.df)`

```
##      tpp      fpp thresholds
## 1 100 100.00000      -Inf
## 2 100  99.50739 0.001410984
## 3 100  99.01478 0.002334826
## 4 100  98.52217 0.002672952
## 5 100  98.02956 0.002889721
## 6 100  97.53695 0.003190045
```

`tail(roc.df)`

```
##      tpp      fpp thresholds
## 295 5.208333  0  0.9716177
## 296 4.166667  0  0.9770051
## 297 3.125000  0  0.9860156
## 298 2.083333  0  0.9940575
## 299 1.041667  0  0.9973720
## 300 0.000000  0      Inf
```

```
#Setting thresholds between TPP 60% and 80%  
roc.df[roc.df$tp > 60 & roc.df$tp < 80,]
```

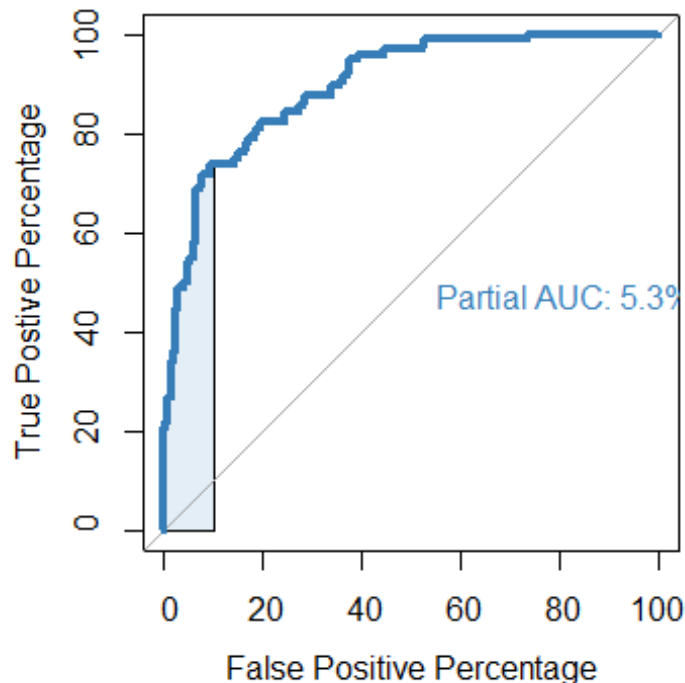
```
##          tp          fp thresholds  
## 187 79.16667 18.226601 0.3316522  
## 188 79.16667 17.733990 0.3337002  
## 189 79.16667 17.241379 0.3417205  
## 190 78.12500 17.241379 0.3490752  
## 191 78.12500 16.748768 0.3537229  
## 192 77.08333 16.748768 0.3610666  
## 193 77.08333 16.256158 0.3672039  
## 194 76.04167 16.256158 0.3722339  
## 195 76.04167 15.763547 0.3768461  
## 196 76.04167 15.270936 0.3803694  
## 197 75.00000 15.270936 0.3825635  
## 198 75.00000 14.778325 0.3852186  
## 199 75.00000 14.285714 0.3876790  
## 200 73.95833 14.285714 0.3904494  
## 201 73.95833 13.793103 0.3939475  
## 202 73.95833 13.300493 0.4003109  
## 203 73.95833 12.807882 0.4063521  
## 204 73.95833 12.315271 0.4129094  
## 205 73.95833 11.822660 0.4184713  
## 206 73.95833 11.330049 0.4203690  
## 207 73.95833 10.837438 0.4296235  
## 208 73.95833 10.344828 0.4396998  
## 209 73.95833 9.852217 0.4477501  
## 210 73.95833 9.359606 0.4591677  
## 211 72.91667 9.359606 0.4668531  
## 212 71.87500 9.359606 0.4732104  
## 213 71.87500 8.866995 0.4802233  
## 214 71.87500 8.374384 0.4887486  
## 215 71.87500 7.881773 0.4995492  
## 216 70.83333 7.881773 0.5061343  
## 217 69.79167 7.881773 0.5084485  
## 218 69.79167 7.389163 0.5128369  
## 219 68.75000 7.389163 0.5193093  
## 220 68.75000 6.896552 0.5227720  
## 221 68.75000 6.403941 0.5287601  
## 222 67.70833 6.403941 0.5481567  
## 223 66.66667 6.403941 0.5700875  
## 224 65.62500 6.403941 0.5796803  
## 225 64.58333 6.403941 0.5876777  
## 226 63.54167 6.403941 0.5967499  
## 227 62.50000 6.403941 0.6044428  
## 228 61.45833 6.403941 0.6142197  
## 229 60.41667 6.403941 0.6253566
```

```
roc(dataset$DEATH_EVENT, logistic$fitted.values, plot=TRUE, legacy.axes=TRUE,  
xlab="False Positive Percentage", ylab="True Positive Percentage",
```

```
col="#377eb8", lwd=4, percent=TRUE, print.auc=TRUE, partial.auc=c(100, 90),
auc.polygon = TRUE, auc.polygon.col = "#377eb822", print.auc.x=45)
```

```
## Setting levels: control = Death, case = Survived
```

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



```
##
```

```
## Call:
```

```
## roc.default(response = dataset$DEATH_EVENT, predictor =
logistic$fitted.values,      percent = TRUE, plot = TRUE, legacy.axes = TRUE,
xlab = "False Positive Percentage",      ylab = "True Postive Percentage", col
= "#377eb8", lwd = 4,      print.auc = TRUE, partial.auc = c(100, 90),
auc.polygon = TRUE,      auc.polygon.col = "#377eb822", print.auc.x = 45)
##
```

```
## Data: logistic$fitted.values in 203 controls (dataset$DEATH_EVENT Death) <
96 cases (dataset$DEATH_EVENT Survived).
```

```
## Partial area under the curve (specificity 100%-90%): 5.271%
```

```
# Plotting and comparing two ROC's (one using Age as single predictor and
other using all variables as predictors)
```

```
roc(dataset$DEATH_EVENT, logistic_simple2$fitted.values, plot=TRUE,
legacy.axes=TRUE, percent=TRUE, xlab="False Positive Percentage", ylab="True
Postive Percentage", col="#377eb8", lwd=4, print.auc=TRUE)
```

```
## Setting levels: control = Death, case = Survived
```

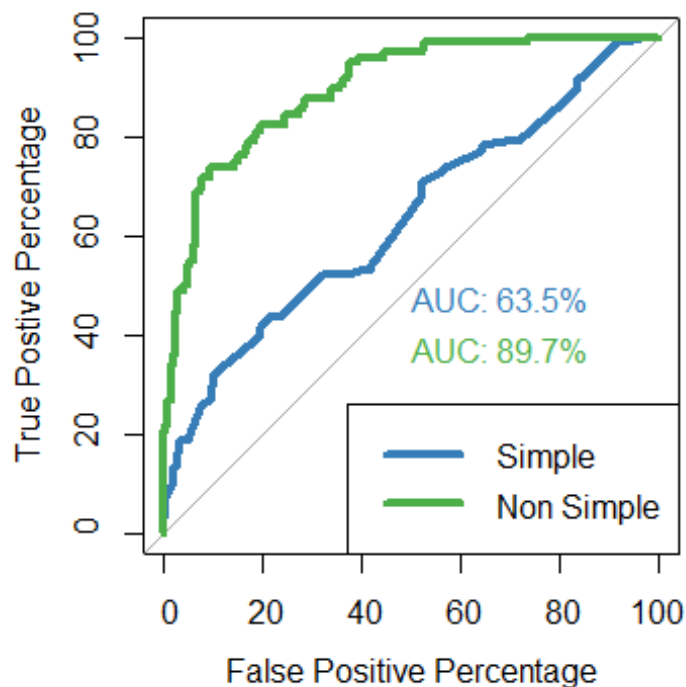
```
## Setting direction: controls < cases
```

```
##
## Call:
## roc.default(response = dataset$DEATH_EVENT, predictor =
logistic_simple2$fitted.values, percent = TRUE, plot = TRUE, legacy.axes
= TRUE, xlab = "False Positive Percentage", ylab = "True Postive
Percentage", col = "#377eb8", lwd = 4, print.auc = TRUE)
##
## Data: logistic_simple2$fitted.values in 203 controls (dataset$DEATH_EVENT
Death) < 96 cases (dataset$DEATH_EVENT Survived).
## Area under the curve: 63.46%

plot.roc(dataset$DEATH_EVENT, logistic$fitted.values, percent=TRUE,
col="#4daf4a", lwd=4, print.auc=TRUE, add=TRUE, print.auc.y=40)

## Setting levels: control = Death, case = Survived
## Setting direction: controls < cases

legend("bottomright", legend=c("Simple", "Non Simple"), col=c("#377eb8",
"#4daf4a"), lwd=4)
```



#Based on the plot, we can say the model using all variables as predictors is better