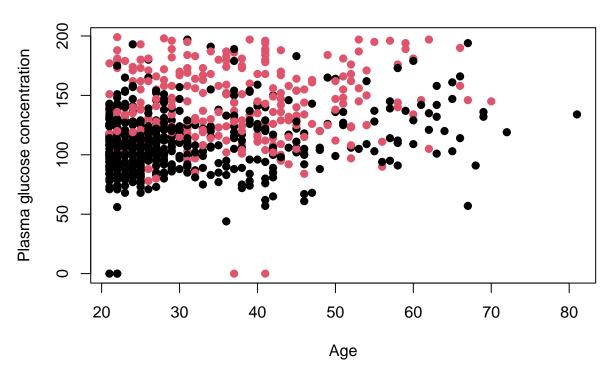
# Assignment-3.R

axelj

### 2022-11-17

```
#Assignment 3 lab1
library(readxl)
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(tidyr)
#Part 1
diabetes = read.csv('pima-indians-diabetes.csv')
diabetes1 = as.data.frame(diabetes)
x = diabetes1[[8]] #Vector of plasma glucose concentration levels
y = diabetes1[[2]] #Vector of ages
plot(x,y, col=as.factor(diabetes1$X1), pch=19,
    main="Plasma glucose concentration on Age",
    xlab="Age", ylab="Plasma glucose concentration")
```



```
#Part 2
set.seed(12345)
train=diabetes%>%select(X1, X148, X50)
m1=glm(as.factor(diabetes1$X1)~., train, family="binomial")
coef(m1)
## (Intercept)
                      X148
                                   X50
## -5.89785793 0.03558250 0.02450157
Prob=predict(m1, type="response")
Pred=ifelse(Prob>0.5, "1", "0")
table(train$X1, Pred)
     Pred
##
##
##
     0 436 64
     1 140 127
summary(m1)
##
## Call:
```

## glm(formula = as.factor(diabetes1\$X1) ~ ., family = "binomial",

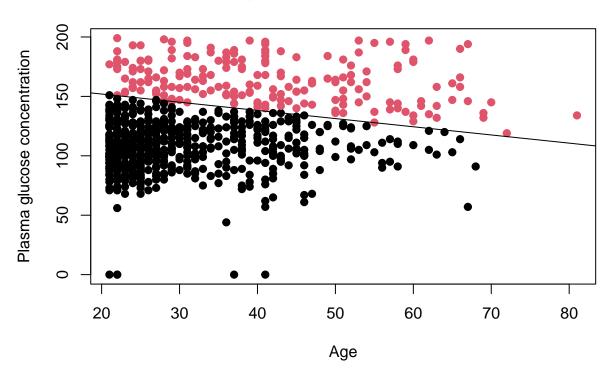
data = train)

##

```
##
## Deviance Residuals:
      Min
          1Q
                  Median
## -2.3303 -0.7775 -0.5095 0.8370
                                   3.1617
## Coefficients:
             Estimate Std. Error z value Pr(>|z|)
## (Intercept) -5.897858  0.462450 -12.75 < 2e-16 ***
## X148
            ## X50
             ## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
      Null deviance: 991.38 on 766 degrees of freedom
## Residual deviance: 796.49 on 764 degrees of freedom
## AIC: 802.49
## Number of Fisher Scoring iterations: 4
missclass=function(X,X1) {
 n=length(X)
 return(1-sum(diag(table(X,X1)))/n)
missclass(as.factor(diabetes1$X1), Pred)
## [1] 0.2659713
#Part 3
plot(x, y, col=as.factor(Pred), pch=19,
    main="Plasma glucose concentration on Age",
    xlab="Age", ylab="Plasma glucose concentration")
```

#The values in abline below are calculated by hand. Can be seen in the report

abline(165.7539767, -0.6886066)



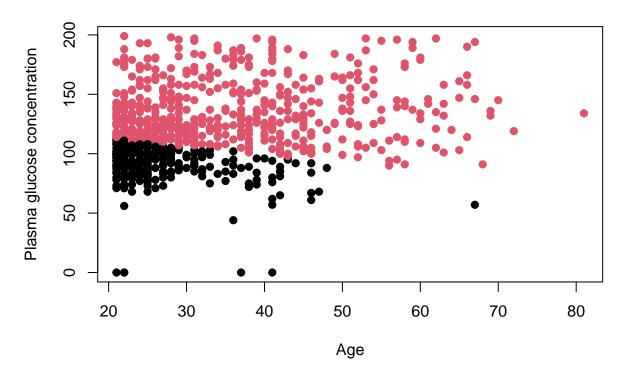
```
#Part 4
# r = 0.2
Pred=ifelse(Prob>0.2, "1", "0")
table(train$X1, Pred)

## Pred
## 0 1
## 0 238 262
## 1 25 242

summary(m1)

## ## Call:
```

```
## glm(formula = as.factor(diabetes1$X1) ~ ., family = "binomial",
##
      data = train)
##
## Deviance Residuals:
##
                1Q
                    Median
                                  ЗQ
      Min
                                         Max
                            0.8370
## -2.3303 -0.7775 -0.5095
##
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) -5.897858
                        0.462450 -12.75 < 2e-16 ***
## X148
              0.035582
                        0.003288
                                   10.82 < 2e-16 ***
```

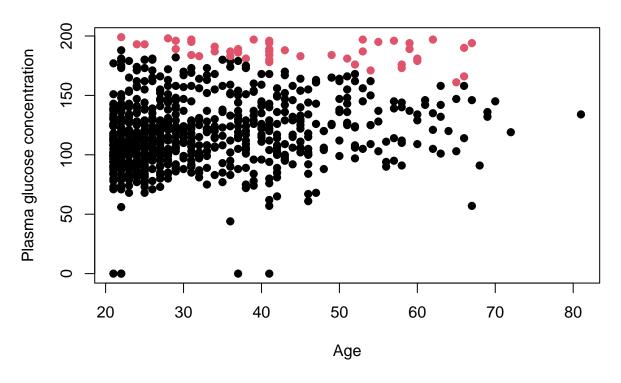


```
# r = 0.8
Pred=ifelse(Prob>0.8, "1", "0")
table(train$X1, Pred)
```

```
## Pred
## 0 1
## 0 490 10
## 1 231 36
```

#### summary(m1)

```
##
## Call:
## glm(formula = as.factor(diabetes1$X1) ~ ., family = "binomial",
##
      data = train)
##
## Deviance Residuals:
      Min
              1Q
                  Median
                               ЗQ
                                      Max
## -2.3303 -0.7775 -0.5095 0.8370
                                   3.1617
##
## Coefficients:
             Estimate Std. Error z value Pr(>|z|)
## (Intercept) -5.897858  0.462450  -12.75  < 2e-16 ***
        ## X148
## X50
             ## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
      Null deviance: 991.38 on 766 degrees of freedom
## Residual deviance: 796.49 on 764 degrees of freedom
## AIC: 802.49
## Number of Fisher Scoring iterations: 4
plot(x, y, col=as.factor(Pred), pch=19,
    main="Plasma glucose concentration on Age",
    xlab="Age", ylab="Plasma glucose concentration")
```



```
#Part 5
head(df)
## 1 function (x, df1, df2, ncp, log = FALSE)
## 3
         if (missing(ncp))
## 4
             .Call(C_df, x, df1, df2, log)
## 5
         else .Call(C_dnf, x, df1, df2, ncp, log)
## 6 }
z1 = c((diabetes1$X50)^4)
z2 = c(((diabetes1$X50)^3)*diabetes1$X148)
z3 = c(((diabetes1$X50)^2)*((diabetes1$X148)^2))
z4 = c(diabetes1$X50*((diabetes1$X148)^3))
z5 = c(diabetes X148^4)
X1 = diabetes1$X1
df = data.frame(z1, z2, z3, z4, z5, X1)
head(df)
##
          z1
                  z2
                           z3
                                                z5 X1
## 1 923521 2532235 6943225
                              19037875
                                          52200625
## 2 1048576 5996544 34292736 196111584 1121513121
## 3 194481 824229 3493161 14804349
                                          62742241
```

```
## 4 1185921 4923369 20439441 84854649 352275361 1
## 5 810000 3132000 12110400 46826880 181063936 0
## 6 456976 1370928 4112784 12338352 37015056 1
train=df%>%select(X1, z1, )
m1=glm(as.factor(diabetes1$X1)~., train, family="binomial")
coef(m1)
    (Intercept)
## -7.479194e-01 4.771323e-08
Prob=predict(m1, type="response")
Pred=ifelse(Prob>0.5, "1", "0")
table(train$X1, Pred)
##
     Pred
##
        0
            1
##
    0 483 17
    1 263
summary(m1)
##
## Call:
## glm(formula = as.factor(diabetes1$X1) ~ ., family = "binomial",
      data = train)
##
## Deviance Residuals:
                1Q Median
##
      Min
                                  3Q
                                          Max
## -1.7582 -0.8928 -0.8845 1.4337
                                       1.5028
##
## Coefficients:
                Estimate Std. Error z value Pr(>|z|)
## (Intercept) -7.479e-01 8.890e-02 -8.413 < 2e-16 ***
               4.771e-08 1.789e-08
                                     2.668 0.00764 **
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 991.38 on 766 degrees of freedom
## Residual deviance: 984.09 on 765 degrees of freedom
## AIC: 988.09
## Number of Fisher Scoring iterations: 4
```