Logistic Regression in R

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Definition

Logistic regression is a statistical analysis method to predict a binary outcome, such as yes or no, based on prior observations of a data set.

A logistic regression model predicts a dependent data variable by analyzing the relationship between one or more existing independent variables. For example, in health, Logistic regression can also be used in the following areas:

- To identify risk factors and plan preventive measures;
- In drug research to tease apart the effectiveness of medicines on health outcomes across age, gender and ethnicity;

About the project

I used data of 318 individuals with and without diabetes type 2. The aim is to examine the relationship between age, gender, BMI, diet type, smoking status and family history of disease and build a model for predicting diabetes risk.

(i) Importing data into R and loading packages

```
Diabetes <- rio::import(here::here("diabetes.csv"))
if(!require(pacman)) install.packages("pacman")</pre>
```

Loading required package: pacman

```
pacman::p_load( tidyverse, janitor, plotrix, gtsummary, survival )
```

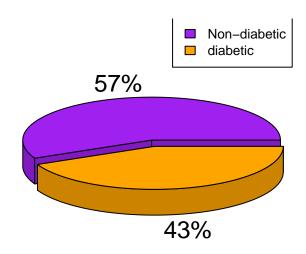
(ii) Data manipulation and exploration

```
labels = c("Yes","No"))
Diabetes <- Diabetes %>%
  mutate(BMI = weight/height^2)
Diabetes <- Diabetes %>%
  select(age, gender, smoking, weight, diabetes, veg, familiy_history, BMI) %%
  mutate(Agegroup=
           ifelse(age < 65, "< 65", "> =65"),
    BMIclass = case_when(
    BMI < 18.5~ "Underweight",
    BMI >=18.5 & BMI < 24.9 ~ "Normal weight",
   TRUE ~ "Obese"
  ))
dim(Diabetes)
## [1] 318 10
summary(Diabetes)
##
         age
                       gender
                                        smoking
                                                      weight
## Min. :17.00
                   male :127
                                                  Min. : 39.30
                                 smokers
                                            :154
                                                  1st Qu.: 61.00
## 1st Qu.:54.00
                   female:191
                                Non-smokers:164
## Median :65.00
                                                  Median : 67.00
## Mean
         :63.29
                                                  Mean
                                                        : 72.01
   3rd Qu.:75.00
                                                   3rd Qu.: 78.00
##
                                                         :186.00
## Max. :89.00
                                                  Max.
##
           diabetes
                                           familiy_history
                                                                BMI
                                   veg
## Non-diabetic:181
                                                           Min. :13.60
                                    :165
                                           Yes: 81
                       vegeterian
## Diabetic :137
                      Non-vegeterian:153
                                           No :237
                                                           1st Qu.:21.08
##
                                                           Median :23.88
##
                                                           Mean :25.01
##
                                                           3rd Qu.:27.40
##
                                                           Max. :57.52
##
                        BMIclass
      Agegroup
## Length:318
                      Length:318
## Class :character Class :character
## Mode :character Mode :character
##
##
##
table1 <- table(Diabetes$diabetes)</pre>
piepercent <-paste0(round(100 * table1/sum(table1)), "%")</pre>
plotrix::pie3D(table1,radius = 1.0,
               explode = 0.05,
              labels = piepercent,
```

main = "Diabetic Vs Non-diabetic compositions",

col = c("purple", "orange"),

Diabetic Vs Non-diabetic compositions



(iii) Descriptive statistics

Table 1: Table1: Descriptive statistics

Characteristic	N = 318
age	65 (54, 75)
BMI	23.9 (21.1, 27.4)
diabetes	
Non-diabetic	181~(57%)
Diabetic	137 (43%)

Characteristic	N = 318
Agegroup	
< 65	152 (48%)
> =65	166 (52%)
gender	
male	127 (40%)
female	191 (60%)
BMIclass	
Normal weight	164 (52%)
Obese	131 (41%)
Underweight	23~(7.2%)
smoking	
smokers	154 (48%)
Non-smokers	164 (52%)
veg	
vegeterian	165~(52%)
Non-vegeterian	153 (48%)
familiy_history	, ,
Yes	81~(25%)
No	237 (75%)

The minimum age was 17yrs and maximum was 89yrs. The median age was 65 yrs and mean age was 63 yrs. 127 were males and 191 were females. 137 were diabetic and 181 were not.

(iv) Testing for significance between diabetic and non diabetic

Table 2: Table2: Statistical significance difference by diabetes status

Characteristic	Non-diabetic, $N = 181$	Diabetic , $N = 137$	p-value
mean age,yrs (sd) median BMI (IQR)	57(16) 23.3 (20.7, 26.2)	71(12) 24.4 (21.3, 28.5)	<0.001 0.001
Agegroup			< 0.001
< 65	115 (64%)	37~(27%)	
> =65	66~(36%)	100 (73%)	

Characteristic	Non-diabetic, $N = 181$	Diabetic , $N = 137$	p-value
gender			0.4
male	68 (38%)	59 (43%)	
female	113~(62%)	78 (57%)	
BMIclass	,	, ,	0.009
Normal weight	100 (55%)	64 (47%)	
Obese	$63\ (35\%)^{'}$	68 (50%)	
Underweight	18 (9.9%)	5 (3.6%)	
smoking status	` ,	,	0.055
smokers	79 (44%)	75 (55%)	
Non-smokers	102(56%)	62(45%)	
Family history	` /	,	0.020
Yes	37 (20%)	44 (32%)	
No	144 (80%)	93 (68%)	
Diet type	,	, ,	0.002
vegeterian	108 (60%)	57 (42%)	
Non-vegeterian	73 (40%)	80 (58%)	

There was significant difference in mean age, mean BMI, smoking status, diet type and family history of the disease between the diabetic and non diabetic individuals. (All p < 0.05)

(vi) Univariate analysis of diabetes risk factors

```
Diabetes %>%
  select(diabetes,Agegroup,gender,BMIclass,smoking,familiy_history,veg) %>%
  tbl_uvregression(
    method = glm,
    y = diabetes,
    method.args = list(family = binomial),
    exponentiate = TRUE) %>%
  modify_caption("Table3: Univariate analysis of diabetes risk factors") %>%
  bold_p() %>%
  italicize_levels() %>%
  bold_labels()
```

Table 3: Table3: Univariate analysis of diabetes risk factors

Characteristic	N	OR	95% CI	p-value
Agegroup	318			
< 65			_	
> =65		4.71	2.93, 7.71	< 0.001
gender	318			
male			_	
female		0.80	0.51, 1.25	0.3
BMIclass	318			
Normal weight			_	
Obese		1.69	1.06, 2.69	0.027
Underweight		0.43	0.14, 1.15	0.12
smoking	318			
smokers				

Characteristic	N	OR	95% CI	p-value
Non-smokers		0.64	0.41, 1.00	0.050
familiy_history	318			
Yes				
No		0.54	0.33, 0.90	0.019
veg	318			
vegeterian				
Non-vegeterian		2.08	$1.33,\ 3.27$	0.002

(vi) Multivariate analysis of diabetes risk factors

```
multivariate <- glm(
   diabetes~ Agegroup+BMIclass+smoking+familiy_history+veg,
   data = Diabetes, family=binomial)

tbl_regression(multivariate, exponentiate = TRUE) %>%
   bold_p() %>%
   bold_labels() %>%
   italicize_levels() %>%
   modify_caption("Table4: Multivariate analysis of diabetes risk factors")
```

Table 4: Table4: Multivariate analysis of diabetes risk factors

Characteristic	OR	95% CI	p-value
Agegroup			
< 65		_	
> =65	5.40	3.24, 9.22	< 0.001
BMIclass			
Normal weight		_	
Obese	1.85	1.10, 3.13	0.020
Underweight	0.36	0.10, 1.05	0.076
smoking			
smokers		_	
Non-smokers	0.55	0.33, 0.92	0.024
familiy_history			
Yes	_	_	
No	0.55	0.30, 0.97	0.039
veg			
vegeterian		_	
Non-vegeterian	1.96	1.19, 3.25	0.008

All the risk factors that were significant risk factors for diabetes in univariate analysis were also significant in multivariate analysis. The factors were: (i) $age(>=65yrs)[OR=5.4(3.2-9.2;\ p<0.001)]$ (ii) Obesity[OR=1.85(1.1-3.1:;\ p=0.02)] (iii) Non smoking[OR=0.55(0.33-0.92);\ P=0.02] (iv) Non vegetarian[OR=1.96(1.19-3.25);\ P=0.008] (v) No family history of disease[OR=0.55(0.3-30.97);\ p=0.04]