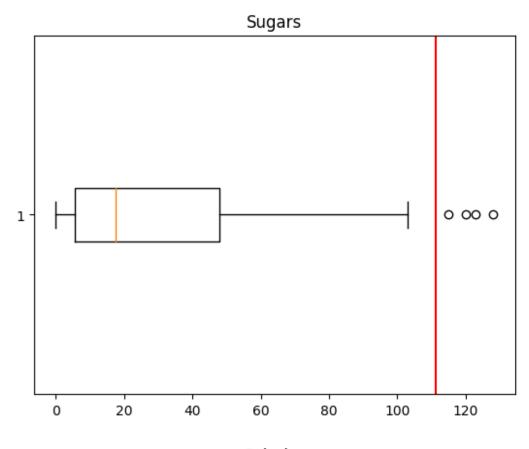
## **Explorando bases**

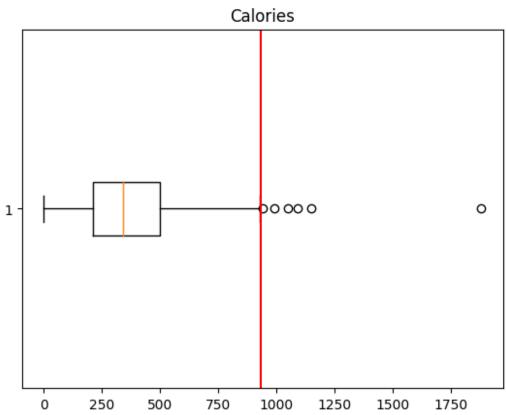
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import scipy.stats as stats
df =
pd.read csv('/Users/quillermocepeda/C:C++/Implementacion IA a01284015/
Entregables/Regresion logistica/mc-donalds-menu-1.csv',sep=',')
print(df.head())
#eliminate a column
df = df.drop(['Category'],axis=1)
print(df.head())
#leave only the column "Sugars" and "calories"
df = df[['Sugars','Calories']]
# 1. Realiza pruebas de normalidad univariada de las variables
(selecciona entre los métodos vistos en clase)
s q1 = np.quantile(df['Sugars'],0.25) #cuartil 1
c q1 = np.quantile(df['Calories'], 0.25)#cuartil 1
s q3 = np.quantile(df['Sugars'], 0.75) #cuartil 3
c_q3 = np.quantile(df['Calories'],0.75)#cuartil 3
#Rango intercuartilico
s_{iqr} = s_{q3} - s_{q1}
c iqr = c q3 - c q1
#Boxplot horizontal de sugars con límite de datos atípicos o extremos
plt.boxplot(df['Sugars'], vert=False)
#plot a vertical line at the limit of the outliers
plt.axvline(x=s q3+1.5*s iqr,color='red')
plt.title('Sugars')
plt.show()
#Boxplot horizontal de calories con límite de datos atípicos o
extremos
plt.boxplot(df['Calories'], vert=False)
plt.axvline(x=c q3+1.5*c iqr,color='red')
plt.title('Calories')
plt.show()
#Quitar los datos atípicos después de haber graficado los boxplots
s \lim = s q3 + 1.5*s iqr
c \lim = c q3 + 1.5*c iqr
df = df[df['Sugars']<s lim]</pre>
df = df[df['Calories'] < c lim]</pre>
```

Category				Item	Servi	ng Si	ze	
Calories 0 Breakfast			Egg McMu	ffin	4.8 oz	(136	g)	
300 \			hite Del					
1 Breakfast 250		Egg w	nite pet	ignt	4.8 02	(133	g)	
2 Breakfast 370		Saus	age McMu	ffin	3.9 oz	(111	g)	
3 Breakfast	Sausag	ge McMuf	fin with	Egg	5.7 oz	(161	g)	
450 4 Breakfast Sa 400	usage McMu1	fin wit	h Egg Wh	ites	5.7 oz	(161	g)	
Calories from	Fat Total	. Fat T	otal Fat	(% Da	ily Val	.ue)	Satur	rated
Fat 0	120	13.0				20		
5.0 \ 1	70	8.0				12		
3.0								
2 8.0	200	23.0				35		
3	250	28.0				43		
10.0 4	210	23.0				35		
8.0								
Saturated Fat 0 1 2 3	(% Daily \	/alue) 25 15 42 52 42	Trans Fa 0. 0. 0. 0.	0 0 0	Carbo	ohydra		\
Carbohydrates 0 1 2 3 4	(% Daily \		Dietary	_	\			
Dietary Fiber Value)	(% Daily \		_		n Vita	amin A	\ (% [	Daily
0 10 \		17	3	1	/			
1		17	3	1	8			
6 2		17	2	1	4			
8		17	2	2	1			
15		Δ,	_	_	_			

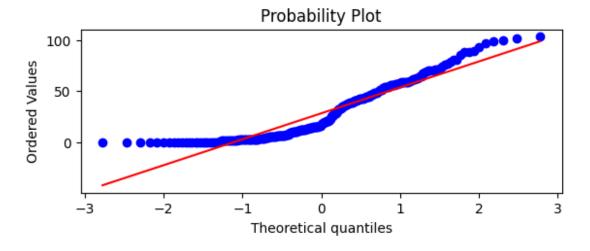
4 6	17	2 23	1								
Vitamin C (% Daily Value)	ue) Calcium	n (% Daily Va	alue) Ir	on (% Daily							
0	0		25								
15 1	0		25								
8 2											
2 10	0		25								
3	0		30								
15 4	0		25								
10	O		23								
[5 rows x 24 columns]	_										
0 Egg 1 Egg Whit 2 Sausage 3 Sausage McMuffir 4 Sausage McMuffin with E	te Delight e McMuffin n with Egg	4.8 oz (136 4.8 oz (135 3.9 oz (111 5.7 oz (161	g) g) g) g)	ries 300 \ 250 370 450 400							
Calories from Fat Total Fat Total Fat (% Daily Value) Saturated											
Fat 120	13.0		20	9							
5.0 \	0.0		1.	2							
1 3.0	8.0		12	Z							
2 200	23.0		3!	5							
8.0 3 250	28.0		43	3							
10.0 4 210	23.0		3!	5							
8.0	23.0		J.	,							
Saturated Fat (% Daily Carbohydrates	Value) Tra	ns Fat Cho	lesterol								
0	25	0.0	260								
31 \ 1	15	0.0	25								
30	13										
2 29	42	0.0	45								
3	52	0.0	285								
30	42	0.0	ΕO								
4 30	42	0.0	50								

0 1 2 3 4	Carbohydı	rates	(% Da	aily		e) 10 10 10 10 10	Dieta	ary	Fiber 4 4 4	\  -  -  -					
	Dietary I lue)	Fiber	(% Da	aily			Sugar		Prote		Vita	amin	A (%	o Dail	у
0 10	\					17		3		17					
1	\					17		3		18					
6 2 8 3						17		2		1 4					
8						17		2		14					
						17		2		21					
15						17		2		21					
4 6						17		2		21					
				_											
\/a	Vitamin ( lue)	C (% [	Daily	Valı	ıe)	Calc	cium (	(% [	Daily	Val	ue)	Iron	(%	Daily	
0 0	tue)				0						25				
15															
1					0						25				
8					0						25				
10															
3 15					0						30				
4					0						25				
10															
[5	rows x 23	3 colu	umns]												





```
# 2. Grafica los datos y su respectivo QQPlot: qqnorm(datos) y
qqline(datos) para cada variable
#QQplot de sugars
plt.subplot(2,1,1)
plt.title('QQplot Sugars')
stats.probplot(df['Sugars'],dist='norm',plot=plt)
plt.show()
```



```
# Calcula el coeficiente de sesgo y el coeficiente de curtosis de cada
variable.
sesgo sugars = df['Sugars'].skew()
sesgo_calories = df['Calories'].skew()
curtosis_sugars = df['Sugars'].kurtosis()
curtosis calories = df['Calories'].kurtosis()
print("Sesgo de sugars: ",sesgo_sugars)
print("Sesgo de calories: ",sesgo_calories)
print("Curtosis de sugars: ",curtosis_sugars)
print("Curtosis de calories: ",curtosis_calories)
Sesgo de sugars: 0.7980002703131476
Sesgo de calories: 0.22544161373962351
Curtosis de sugars: -0.3368579444441888
Curtosis de calories: -0.476368782403791
# Compara las medidas de media, mediana y rango medio de cada
variable.
m sugars = np.mean(df['Sugars'])
m calories = np.mean(df['Calories'])
mediana sugars = np.median(df['Sugars'])
mediana_calories = np.median(df['Calories'])
```

```
rm sugars = np.mean(df['Sugars']) - np.median(df['Sugars'])
rm calories = np.mean(df['Calories']) - np.median(df['Calories'])
print("Media de sugars: ",m sugars)
print("Media de calories: ",m calories)
print("Mediana de sugars: ",mediana_sugars)
print("Mediana de calories: ",mediana_calories)
print("Rango medio de sugars: ",rm sugars)
print("Rango medio de calories: ",rm calories)
Media de sugars: 28.376
Media de calories: 340.8
Mediana de sugars: 17.0
Mediana de calories: 330.0
Rango medio de sugars: 11.37600000000001
Rango medio de calories: 10.80000000000011
# Realiza el histograma y su distribución teórica de probabilidad
(sugerencia, adapta el código:
# hist(datos, freq=FALSE)
# lines(density(datos),col="red")
# curve(dnorm(x, mean=mean(datos, sd=sd(datos)), from=-6, to=6,
add=TRUE, col="blue", lwd=2)
# Comenta los gráficos y los resultados obtenidos con vías a
interpretar normalidad de los datos
plt.hist(df['Sugars'],density=True)
plt.axvline(df['Sugars'].mean(), color='k', linestyle='dashed',
linewidth=1)
#curva dnorm
x = np.linspace(df['Sugars'].min(),df['Sugars'].max(),100)
mean = df['Sugars'].mean()
sd = df['Sugars'].std()
y = \frac{1}{(sd*np.sqrt(2*np.pi))*np.exp(-(x-mean)**2/(2*sd**2))}
plt.plot(x,y,color='red')
plt.show()
plt.hist(df['Calories'],density=True)
plt.axvline(df['Calories'].mean(), color='k', linestyle='dashed',
linewidth=1)
#curva dnorm
x = np.linspace(df['Calories'].min(),df['Calories'].max(),100)
mean = df['Calories'].mean()
sd = df['Calories'].std()
y = \frac{1}{(sd*np.sqrt(2*np.pi))*np.exp(-(x-mean)**2/(2*sd**2))}
plt.plot(x,y,color='red')
plt.show()
#Se puede apreciar claramente que la distribución normal se ajusta
mejor a la variable Calories,
```

# mientras que la variable Sugars pareciera que tiene un sesgo a la izquierda

