

```
In [26]: import os, pandas as pd, numpy as np, matplotlib.pyplot as plt, pickle as pkl
from sklearn import model_selection as modele, linear_model as lm, metrics as mes
import matplotlib as plt
import seaborn as sb
```

```
In [27]: os.getcwd()
```

```
Out[27]: 'C:\\Users\\HPPC\\regession logistic'
```

```
In [28]: os.chdir("C:\\Users\\HPPC\\regession logistic")
```

```
In [29]: data = pd.read_excel("null.xlsx")
data.head()
```

```
Out[29]:
```

	AGE	SEXE	TDT	PAR	CHOLESTEROL	GAJ	ECG	FCMAX	ANGINE	DEPRESSION	PEN
0	40	homme	AA	140	289	0	Normal	172	Non	0.0	Ascenda
1	49	femme	DNA	160	180	0	Normal	156	Non	1.0	PI
2	37	homme	AA	130	283	0	ST	98	Non	0.0	Ascenda
3	48	femme	ASY	138	214	0	Normal	108	Oui	1.5	PI
4	54	homme	DNA	150	195	0	Normal	122	Non	0.0	Ascenda

```
In [30]: df = data.copy()
print(df)
```

	AGE	SEXE	TDT	PAR	CHOLESTEROL	GAJ	ECG	FCMAX	ANGINE	\
0	40	homme	AA	140	289	0	Normal	172	Non	
1	49	femme	DNA	160	180	0	Normal	156	Non	
2	37	homme	AA	130	283	0	ST	98	Non	
3	48	femme	ASY	138	214	0	Normal	108	Oui	
4	54	homme	DNA	150	195	0	Normal	122	Non	
..	
913	45	homme	AT	110	264	0	Normal	132	Non	
914	68	homme	ASY	144	193	1	Normal	141	Non	
915	57	homme	ASY	130	131	0	Normal	115	Oui	
916	57	femme	AA	130	236	0	LVH	174	Non	
917	38	homme	DNA	138	175	0	Normal	173	Non	

	DEPRESSION	PENTE	CŒUR
0	0.0	Ascendant	0
1	1.0	Plat	1
2	0.0	Ascendant	0
3	1.5	Plat	1
4	0.0	Ascendant	0
..
913	1.2	Plat	1
914	3.4	Plat	1
915	1.2	Plat	1
916	0.0	Plat	1
917	0.0	Ascendant	0

[918 rows x 12 columns]

```
In [31]: df.duplicated().sum()
```

```
Out[31]: 0
```

```
In [32]: df.nunique()
```

```
Out[32]: AGE          50
SEX     2
TDT     4
PAR     67
CHOLESTEROL  222
GAJ     2
ECG     3
FCMAX   119
ANGINE   2
DEPRESSION  53
PENTE   3
CŒUR    2
dtype: int64
```

```
In [33]: df.isna().sum()
```

```
Out[33]: AGE          0
SEX     0
TDT     0
PAR     0
CHOLESTEROL  0
GAJ     0
ECG     0
FCMAX   0
ANGINE   0
DEPRESSION  0
PENTE   0
CŒUR    0
dtype: int64
```

```
In [38]: quant = ['AGE', 'PAR', 'CHOLESTEROL', 'FCMAX', 'DEPRESSION ']  
#df.rename('DEPRESSION')
```

```
In [52]: def recoder(df):  
    for i in df.select_dtypes('object').columns:  
        df[i]=df[i].astype('category').cat.codes  
    return(df)  
recoder(df)
```

```
Out[52]:
```

	AGE	SEXE	TDT	PAR	CHOLESTEROL	GAJ	ECG	FCMAX	ANGINE	DEPRESSION	PENTE	CO
0	0.75	1	0	1.06	1.45	0	1	1.26	0	0.00	0	
1	0.92	0	3	1.21	0.91	0	1	1.14	0	1.13	2	
2	0.69	1	0	0.98	1.42	0	2	0.72	0	0.00	0	
3	0.90	0	1	1.04	1.08	0	1	0.79	1	1.69	2	
4	1.01	1	3	1.13	0.98	0	1	0.89	0	0.00	0	
...	
913	0.84	1	2	0.83	1.33	0	1	0.96	0	1.35	2	

	AGE	SEXE	TDT	PAR	CHOLESTEROL	GAJ	ECG	FCMAX	ANGINE	DEPRESSION	PENTE	CQ
914	1.27	1	1	1.09	0.97	1	1	1.03	0	3.83	2	
915	1.07	1	1	0.98	0.66	0	1	0.84	1	1.35	2	
916	1.07	0	0	0.98	1.19	0	0	1.27	0	0.00	2	
917	0.71	1	3	1.04	0.88	0	1	1.26	0	0.00	0	

918 rows × 12 columns



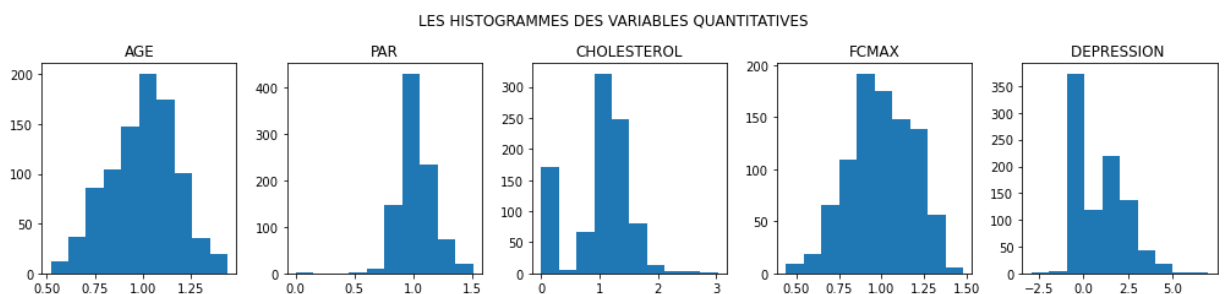
```
In [53]: df[quant] = round(df[quant]/df[quant].mean(), 2)
```

```
In [54]: df.head()
```

	AGE	SEXE	TDT	PAR	CHOLESTEROL	GAJ	ECG	FCMAX	ANGINE	DEPRESSION	PENTE	CŒUR
0	0.75	1	0	1.06	1.45	0	1	1.26	0	0.00	0	
1	0.92	0	3	1.21	0.91	0	1	1.14	0	1.13	2	
2	0.69	1	0	0.98	1.42	0	2	0.72	0	0.00	0	
3	0.90	0	1	1.04	1.08	0	1	0.79	1	1.69	2	
4	1.01	1	3	1.13	0.98	0	1	0.89	0	0.00	0	



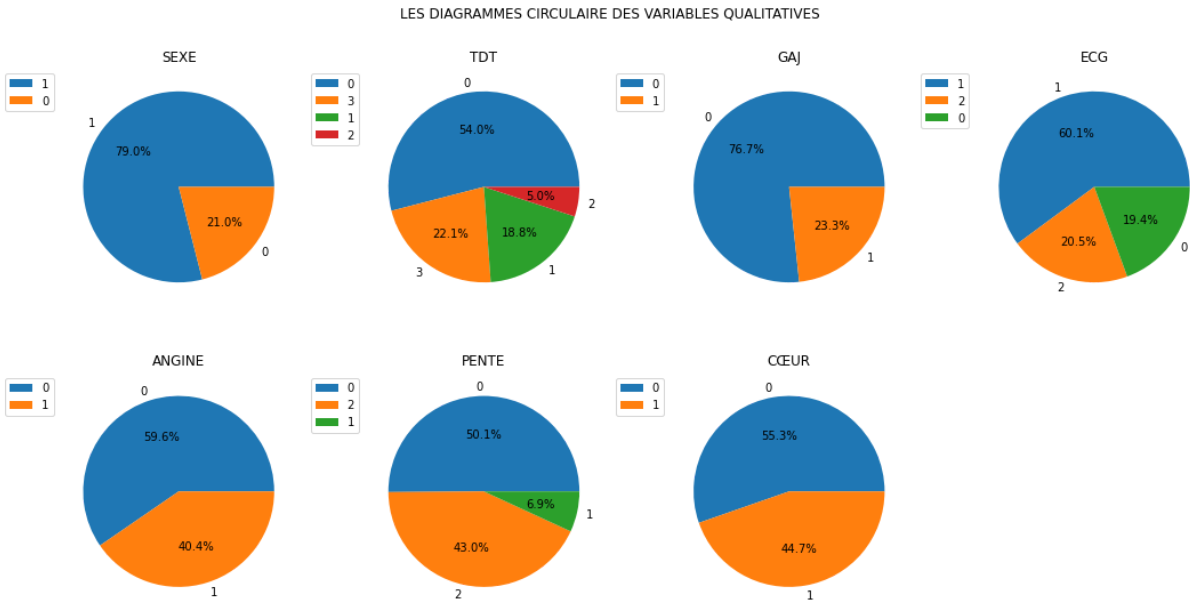
```
In [55]: pl.figure(tight_layout = True, figsize = (14, 6))
pl.suptitle("LES HISTOGRAMMES DES VARIABLES QUANTITATIVES")
for y,x in enumerate(quant):
    pl.subplot(2,5,y+1)
    pl.hist(df[x])
    pl.title(f"{quant[y]}")
pl.show()
```



```
In [56]: qual = ['SEXE', 'TDT', 'GAJ', 'ECG', 'ANGINE', 'PENTE', 'CŒUR'] # la liste des varia

pl.figure(tight_layout = True, figsize = (15,8))
pl.suptitle("LES DIAGRAMMES CIRCULAIRE DES VARIABLES QUALITATIVES")
for y,x in enumerate(qual):
    eff = df[x].value_counts()
    modalite = df[x].unique()
    pl.subplot(2,4,y+1)
    pl.pie(eff, labels = modalite, autopct = '%1.1f%')
    pl.legend(bbox_to_anchor = (0, 1))
```

```
pl.title(f"{qual[y]}")  
pl.show()
```



```
In [57]: X = df.iloc[:, 0:11]  
X
```

Out[57]:

	AGE	SEXE	TDT	PAR	CHOLESTEROL	GAJ	ECG	FCMAX	ANGINE	DEPRESSION	PENTE
0	0.75	1	0	1.06	1.45	0	1	1.26	0	0.00	0
1	0.92	0	3	1.21	0.91	0	1	1.14	0	1.13	2
2	0.69	1	0	0.98	1.42	0	2	0.72	0	0.00	0
3	0.90	0	1	1.04	1.08	0	1	0.79	1	1.69	2
4	1.01	1	3	1.13	0.98	0	1	0.89	0	0.00	0
...
913	0.84	1	2	0.83	1.33	0	1	0.96	0	1.35	2
914	1.27	1	1	1.09	0.97	1	1	1.03	0	3.83	2
915	1.07	1	1	0.98	0.66	0	1	0.84	1	1.35	2
916	1.07	0	0	0.98	1.19	0	0	1.27	0	0.00	2
917	0.71	1	3	1.04	0.88	0	1	1.26	0	0.00	0

918 rows × 11 columns

```
In [58]: y = data.iloc[:,11:]  
y
```

Out[58]:

	CŒUR
0	0
1	1
2	0
3	1

CŒUR	
4	0
...	...
913	1
914	1
915	1
916	1
917	0

918 rows × 1 columns

```
In [59]: X_train, X_test, Y_train, Y_test = modele.train_test_split(X, y, test_size = 0.2, ra
```

```
In [60]: mod = lm.LogisticRegression(penalty = 'none', solver = 'newton-cg')
mod.fit(X_train, Y_train)
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py:63: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

return f(*args, **kwargs)

```
Out[60]: LogisticRegression(penalty='none', solver='newton-cg')
```

```
In [61]: mod.predict_proba(X_test)
```

```
Out[61]: array([[0.35496203, 0.64503797],
 [0.16960133, 0.83039867],
 [0.03457301, 0.96542699],
 [0.0764487 , 0.9235513 ],
 [0.17768705, 0.82231295],
 [0.74672317, 0.25327683],
 [0.29597548, 0.70402452],
 [0.51735064, 0.48264936],
 [0.96541828, 0.03458172],
 [0.2826058 , 0.7173942 ],
 [0.88994162, 0.11005838],
 [0.6539992 , 0.3460008 ],
 [0.40417279, 0.59582721],
 [0.16705882, 0.83294118],
 [0.09203056, 0.90796944],
 [0.88406903, 0.11593097],
 [0.07255202, 0.92744798],
 [0.79787133, 0.20212867],
 [0.04776453, 0.95223547],
 [0.21131067, 0.78868933],
 [0.07855208, 0.92144792],
 [0.96654432, 0.03345568],
 [0.03879931, 0.96120069],
 [0.96237084, 0.03762916],
 [0.41115221, 0.58884779],
 [0.02625307, 0.97374693],
 [0.86081418, 0.13918582],
 [0.184074 , 0.815926 ],
 [0.14306526, 0.85693474],
```

[0.84833901, 0.15166099],
[0.98822374, 0.01177626],
[0.93968456, 0.06031544],
[0.69221562, 0.30778438],
[0.96131221, 0.03868779],
[0.10723281, 0.89276719],
[0.06521403, 0.93478597],
[0.09615314, 0.90384686],
[0.04487176, 0.95512824],
[0.94696335, 0.05303665],
[0.61414904, 0.38585096],
[0.22452249, 0.77547751],
[0.2325561 , 0.7674439],
[0.08042256, 0.91957744],
[0.78349444, 0.21650556],
[0.96391422, 0.03608578],
[0.986358 , 0.013642],
[0.11334199, 0.88665801],
[0.81843 , 0.18157],
[0.95221882, 0.04778118],
[0.13528833, 0.86471167],
[0.85314986, 0.14685014],
[0.07228734, 0.92771266],
[0.01298134, 0.98701866],
[0.79992191, 0.20007809],
[0.339827 , 0.660173],
[0.85369522, 0.14630478],
[0.07269278, 0.92730722],
[0.91619274, 0.08380726],
[0.33051036, 0.66948964],
[0.83005714, 0.16994286],
[0.01856838, 0.98143162],
[0.43551455, 0.56448545],
[0.24693029, 0.75306971],
[0.5685169 , 0.4314831],
[0.96279401, 0.03720599],
[0.07861632, 0.92138368],
[0.10312646, 0.89687354],
[0.18763544, 0.81236456],
[0.9679706 , 0.0320294],
[0.01794743, 0.98205257],
[0.03016454, 0.96983546],
[0.9105003 , 0.0894997],
[0.13276655, 0.86723345],
[0.07152434, 0.92847566],
[0.93459766, 0.06540234],
[0.10696642, 0.89303358],
[0.90419903, 0.09580097],
[0.16606083, 0.83393917],
[0.03020698, 0.96979302],
[0.33841854, 0.66158146],
[0.02984471, 0.97015529],
[0.05997388, 0.94002612],
[0.08251973, 0.91748027],
[0.15846023, 0.84153977],
[0.1623932 , 0.8376068],
[0.03220352, 0.96779648],
[0.67115575, 0.32884425],
[0.56227941, 0.43772059],
[0.89738474, 0.10261526],
[0.71542579, 0.28457421],
[0.03055251, 0.96944749],
[0.05390453, 0.94609547],
[0.92583897, 0.07416103],

[0.44620256, 0.55379744],
[0.07400792, 0.92599208],
[0.15027603, 0.84972397],
[0.07246277, 0.92753723],
[0.1857801 , 0.8142199],
[0.13377028, 0.86622972],
[0.63356425, 0.36643575],
[0.28735637, 0.71264363],
[0.9659129 , 0.0340871],
[0.1756883 , 0.8243117],
[0.17112415, 0.82887585],
[0.08141389, 0.91858611],
[0.92355839, 0.07644161],
[0.90402068, 0.09597932],
[0.54537251, 0.45462749],
[0.04300625, 0.95699375],
[0.81612043, 0.18387957],
[0.84740312, 0.15259688],
[0.03079608, 0.96920392],
[0.93610009, 0.06389991],
[0.80066341, 0.19933659],
[0.98075761, 0.01924239],
[0.94266284, 0.05733716],
[0.1665684 , 0.8334316],
[0.98523093, 0.01476907],
[0.63669803, 0.36330197],
[0.12532174, 0.87467826],
[0.18458031, 0.81541969],
[0.44485531, 0.55514469],
[0.06414396, 0.93585604],
[0.96693085, 0.03306915],
[0.00488167, 0.99511833],
[0.76680806, 0.23319194],
[0.05064536, 0.94935464],
[0.84542875, 0.15457125],
[0.0481394 , 0.9518606],
[0.17078444, 0.82921556],
[0.59743877, 0.40256123],
[0.08829912, 0.91170088],
[0.8876137 , 0.1123863],
[0.03362888, 0.96637112],
[0.02557452, 0.97442548],
[0.24947139, 0.75052861],
[0.02407659, 0.97592341],
[0.06909475, 0.93090525],
[0.81325681, 0.18674319],
[0.08863516, 0.91136484],
[0.02686139, 0.97313861],
[0.83116316, 0.16883684],
[0.84021365, 0.15978635],
[0.01628384, 0.98371616],
[0.2188689 , 0.7811311],
[0.20984284, 0.79015716],
[0.24260794, 0.75739206],
[0.11317876, 0.88682124],
[0.44319422, 0.55680578],
[0.0312486 , 0.9687514],
[0.0358787 , 0.9641213],
[0.65569386, 0.34430614],
[0.89579027, 0.10420973],
[0.90985105, 0.09014895],
[0.92803873, 0.07196127],
[0.04143907, 0.95856093],
[0.98723148, 0.01276852],

```
[0.04003012, 0.95996988],
[0.13481194, 0.86518806],
[0.04486161, 0.95513839],
[0.30865514, 0.69134486],
[0.42271784, 0.57728216],
[0.08754011, 0.91245989],
[0.61294864, 0.38705136],
[0.05356502, 0.94643498],
[0.58391665, 0.41608335],
[0.96259529, 0.03740471],
[0.03546116, 0.96453884],
[0.0365794 , 0.9634206 ],
[0.04124809, 0.95875191],
[0.73310092, 0.26689908],
[0.27857483, 0.72142517],
[0.08148662, 0.91851338],
[0.17840631, 0.82159369],
[0.12022803, 0.87977197],
[0.05141357, 0.94858643],
[0.86555268, 0.13444732],
[0.81765148, 0.18234852],
[0.03583028, 0.96416972],
[0.54436461, 0.45563539],
[0.48625252, 0.51374748],
[0.97215613, 0.02784387],
[0.87130014, 0.12869986],
[0.58628658, 0.41371342]])
```

```
In [62]: pred = mod.predict(X_test)

pred
```

```
Out[62]: array([1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0,
        1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0,
        0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1,
        1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
        0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0,
        0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1,
        0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0,
        0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1,
        0, 0, 1, 0, 1, 0, 0, 0], dtype=int64)
```

```
In [63]: mes.confusion_matrix(Y_test, pred)
```

```
Out[63]: array([[ 66,   8],
        [  8, 102]], dtype=int64)
```

```
In [64]: mes.accuracy_score(Y_test, pred)
```

```
Out[64]: 0.9130434782608695
```

```
In [65]: mes.recall_score(Y_test, pred)
```

```
Out[65]: 0.9272727272727272
```

```
In [66]: mes.precision_score(Y_test, pred)
```

```
Out[66]: 0.9272727272727272
```


In [67]: *# sauvegarde du modele de prediction mod*

```
pk1.dump(mod, open('mod.pk1', 'wb' ))
```

In []: