```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
# from google.colab import files
# uploaded = files.upload()
# import io
dataset = pd.read_csv('/data.csv')
dataset.head()
```

id diagnosis radius\_mean texture\_mean perimeter\_mean area\_mean smoothness\_mean compactn 842302 17.99 10.38 122.80 1001.0 0.11840 M 842517 M 20.57 17.77 132.90 1326.0 0.08474 **2** 84300903 Μ 19.69 21.25 130.00 1203.0 0.10960 **3** 84348301 386.1 0.14250 20.38 77.58 M 11.42 0.10030 **4** 84358402 Μ 20.29 14.34 135.10 1297.0

5 rows × 33 columns

dataset.shape

(569, 33)

dataset.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 569 entries, 0 to 568 Data columns (total 33 columns):

Data	columns (total 33 column		Dhua				
#	Column	Non-Null Count	Dtype 				
0	id	569 non-null	int64				
1	diagnosis	569 non-null	object				
2	radius mean	569 non-null	float64				
3	texture mean	569 non-null	float64				
4	perimeter mean	569 non-null	float64				
5	area mean	569 non-null	float64				
6	smoothness_mean	569 non-null	float64				
7	compactness mean	569 non-null	float64				
8	concavity mean	569 non-null	float64				
9	concave points mean	569 non-null	float64				
10	symmetry mean	569 non-null	float64				
11	fractal dimension mean	569 non-null	float64				
12	radius se	569 non-null	float64				
13	texture_se	569 non-null	float64				
14	perimeter se	569 non-null	float64				
15	area se	569 non-null	float64				
16	smoothness se	569 non-null	float64				
17	compactness se	569 non-null	float64				
18	concavity se	569 non-null	float64				
19	concave points_se	569 non-null	float64				
20	symmetry_se	569 non-null	float64				
21	<pre>fractal_dimension_se</pre>	569 non-null	float64				
22	radius_worst	569 non-null	float64				
23	texture_worst	569 non-null	float64				
24	perimeter_worst	569 non-null	float64				
25	area_worst	569 non-null	float64				
26	smoothness_worst	569 non-null	float64				
27	compactness_worst	569 non-null	float64				
28	concavity_worst	569 non-null	float64				
29	concave points_worst	569 non-null	float64				
30	symmetry_worst	569 non-null	float64				
31	<pre>fractal_dimension_worst</pre>	569 non-null	float64				
32	Unnamed: 32	0 non-null	float64				
	es: float64(31), int64(1)	, object(1)					
memor	ry usage: 146.8+ KB						

dataset.describe()

		id	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactnes		
	count	5.690000e+02	569.000000	569.000000	569.000000	569.000000	569.000000	569		
	mean	3.037183e+07	14.127292	19.289649	91.969033	654.889104	0.096360	0		
	std	1.250206e+08	3.524049	4.301036	24.298981	351.914129	0.014064	0		
	min	8.670000e+03	6.981000	9.710000	43.790000	143.500000	0.052630	0		
	25%	8.692180e+05	11.700000	16.170000	75.170000	420.300000	0.086370	0		
	50%	9.060240e+05	13.370000	18.840000	86.240000	551.100000	0.095870	0		
	75%	8.813129e+06	15.780000	21.800000	104.100000	782.700000	0.105300	0		
	max	9.113205e+08	28.110000	39.280000	188.500000	2501.000000	0.163400	0		
8 rows × 32 columns										
datas	et.sele	ect_dtypes(inc	lude = 'objec	t').columns						
	Index(	['diagnosis'],	dtype='objec	t')						
datas	set.sele	ect_dtypes(inc	lude = ['int6	4','float64'])	.columns					
<pre>Index(['id', 'radius_mean', 'texture_mean', 'perimeter_mean', 'area_mean',</pre>										
datas	et.colu	ımns								
<pre>Index(['id', 'diagnosis', 'radius_mean', 'texture_mean', 'perimeter_mean',</pre>										
datas	et.isnu	ıll().values.a	ny()							
	True									
datas	set.isnu	ıll().values.s	um()							
	569									
datas	et.colu	ımns[dataset.i	snull().any()	]						
<pre>Index(['Unnamed: 32'], dtype='object')</pre>										
<pre>dataset = dataset.drop(columns = 'Unnamed: 32')</pre>										
dataset.shape										
(569, 32)										
<pre>dataset.isnull().values.any()</pre>										
False										
datas	<pre>dataset.select_dtypes(include = 'object')</pre>									

	diagnosis
0	М
1	M
2	М
3	M
4	M
564	M
565	M
566	M

dataset.select\_dtypes(include = 'object')

	diagnosis
0	М
1	M
2	M
3	М
4	М
564	М
565	М
566	М
567	М
568	В

569 rows × 1 columns

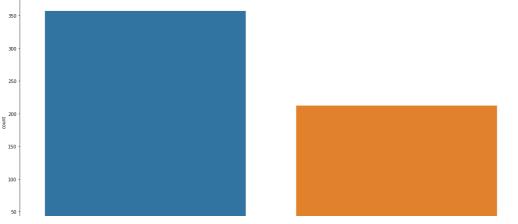
5 rows × 32 columns

dataset.head()

	id	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	С
0	842302	17.99	10.38	122.80	1001.0	0.11840	0.27760	
1	842517	20.57	17.77	132.90	1326.0	0.08474	0.07864	
2	84300903	19.69	21.25	130.00	1203.0	0.10960	0.15990	
3	84348301	11.42	20.38	77.58	386.1	0.14250	0.28390	
4	84358402	20.29	14.34	135.10	1297.0	0.10030	0.13280	

sns.countplot(dataset['diagnosis\_M'], label = 'count')

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f76c9f5bb20>



(dataset.diagnosis\_M == 0).sum()

357

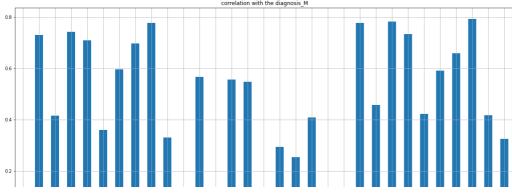
dataset\_2 = dataset.drop(columns = 'diagnosis\_M')

dataset\_2.corrwith(dataset['diagnosis\_M'])

id radius mean	0.039769 0.730029
texture mean	0.730023
perimeter mean	0.742636
area mean	0.742030
smoothness mean	0.358560
compactness_mean	0.596534
concavity mean	0.696360
concave points mean	0.776614
symmetry mean	0.330499
fractal dimension mean	-0.012838
radius se	0.567134
texture se	-0.008303
perimeter se	0.556141
. –	0.548236
area_se smoothness se	-0.067016
<del>-</del>	0.292999
compactness_se	0.253730
concavity_se	0.408042
concave points_se	-0.006522
symmetry_se	0.077972
fractal_dimension_se	0.776454
radius_worst	
texture_worst	0.456903
perimeter_worst	0.782914
area_worst	0.733825
smoothness_worst	0.421465
compactness_worst	0.590998
concavity_worst	0.659610
concave points_worst	0.793566
symmetry_worst	0.416294
<pre>fractal_dimension_worst dtype: float64</pre>	0.323872
acype. 110aco+	

dataset\_2.corrwith(dataset['diagnosis\_M']).plot.bar(figsize = (20,10), title = 'correlation with the diagnosis\_M', rot = 45, grid = True

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f76c98fe280>



corr = dataset.corr()

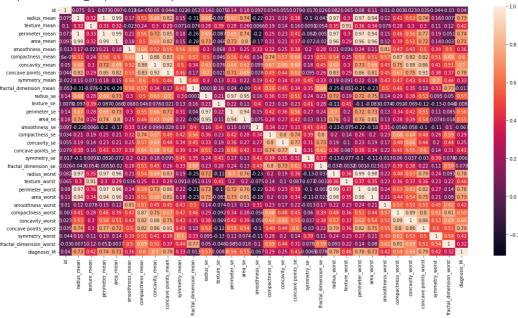
corr

	id	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean
id	1.000000	0.074626	0.099770	0.073159	0.096893	-0.012968
radius_mean	0.074626	1.000000	0.323782	0.997855	0.987357	0.170581
texture_mean	0.099770	0.323782	1.000000	0.329533	0.321086	-0.023389
perimeter_mean	0.073159	0.997855	0.329533	1.000000	0.986507	0.207278
area_mean	0.096893	0.987357	0.321086	0.986507	1.000000	0.177028
smoothness_mean	-0.012968	0.170581	-0.023389	0.207278	0.177028	1.000000
compactness_mean	0.000096	0.506124	0.236702	0.556936	0.498502	0.659123
concavity_mean	0.050080	0.676764	0.302418	0.716136	0.685983	0.521984
concave points_mean	0.044158	0.822529	0.293464	0.850977	0.823269	0.553695
symmetry_mean	-0.022114	0.147741	0.071401	0.183027	0.151293	0.557775
fractal_dimension_mean	-0.052511	-0.311631	-0.076437	-0.261477	-0.283110	0.584792
radius_se	0.143048	0.679090	0.275869	0.691765	0.732562	0.301467
texture_se	-0.007526	-0.097317	0.386358	-0.086761	-0.066280	0.068406
perimeter_se	0.137331	0.674172	0.281673	0.693135	0.726628	0.296092
area_se	0.177742	0.735864	0.259845	0.744983	0.800086	0.246552
smoothness_se	0.096781	-0.222600	0.006614	-0.202694	-0.166777	0.332375
compactness_se	0.033961	0.206000	0.191975	0.250744	0.212583	0.318943
concavity_se	0.055239	0.194204	0.143293	0.228082	0.207660	0.248396
concave points_se	0.078768	0.376169	0.163851	0.407217	0.372320	0.380676
symmetry_se	-0.017306	-0.104321	0.009127	-0.081629	-0.072497	0.200774
fractal_dimension_se	0.025725	-0.042641	0.054458	-0.005523	-0.019887	0.283607
radius_worst	0.082405	0.969539	0.352573	0.969476	0.962746	0.213120
texture_worst	0.064720	0.297008	0.912045	0.303038	0.287489	0.036072
perimeter_worst	0.079986	0.965137	0.358040	0.970387	0.959120	0.238853
area_worst	0.107187	0.941082	0.343546	0.941550	0.959213	0.206718
smoothness_worst	0.010338	0.119616	0.077503	0.150549	0.123523	0.805324
compactness_worst	-0.002968	0.413463	0.277830	0.455774	0.390410	0.472468
concavity_worst	0.023203	0.526911	0.301025	0.563879	0.512606	0.434926
concave points_worst	0.035174	0.744214	0.295316	0.771241	0.722017	0.503053
symmetry_worst	-0.044224	0.163953	0.105008	0.189115	0.143570	0.394309
fractal_dimension_worst	-0.029866	0.007066	0.119205	0.051019	0.003738	0.499316
diagnosis_M	0.039769	0.730029	0.415185	0.742636	0.708984	0.358560

32 rows × 32 columns

plt.figure(figsize = (20,10))
sns.heatmap(corr, annot = True)

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f76c9ee3940>



```
x = dataset.iloc[:,1:-1].values
y = dataset.iloc[:,-1].values
x.shape
     (569, 30)
y.shape
     (569,)
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y, test_size = 0.33, random_state = 10)
x_test.shape
     (188, 30)
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
x test
     array([[ 0.32939352, -1.70277545, 0.38002753, ..., 1.01205104,
             0.56500364, 0.94395709],
[-1.37094822, 0.62832296, -1.3674522 , ..., -1.41478814,
              -0.80492439, -0.56996616],
             [-0.16181015, -0.73322407, -0.20902864, ..., -0.7313608, 0.75268904, -0.39771335],
             [ 0.04077667, 0.12499127, 0.05777443, ..., 0.60721731,
              -0.30852281, 0.5261881 ],
             [\ 0.14900799,\ -1.05099569,\ 0.06500704,\ \ldots,\ -0.58178864,
             -0.98208538, -1.26960339],
[-0.39492376, 0.49611141, -0.41837261, ..., -0.94552094,
              -0.71195873, -0.14604528]])
```

```
from \ sklearn.linear\_model \ import \ LogisticRegression
from \ sklearn.metrics \ import \ f1\_score, \ accuracy\_score, \ confusion\_matrix, \ precision\_score, \ recall\_score, \ accuracy\_score, \ confusion\_matrix, \ precision\_score, \ recall\_score, \ accuracy\_score, 
logreg = LogisticRegression()
logreg.fit(x_train,y_train,)
                       LogisticRegression()
y_pred = logreg.predict(x_test)
acc = accuracy_score(y_test,y_pred)
f1 = f1_score(y_test,y_pred)
precision = precision_score(y_test,y_pred)
recall = recall_score(y_test,y_pred)
results = pd.DataFrame([['Logistic Regression', acc, f1, precision, recall]],
                                                                                                            columns = ['Model','Accuracy', 'F1', ' Precision', 'Recall'])
results
```

Model Accuracy F1 Precision Recall **0** Logistic Regression 0.973404 0.96124 0.953846 0.96875

```
cm = confusion_matrix(y_test,y_pred)
print(cm)
sns.heatmap(cm,annot = True)
     [[121 3]
[ 2 62]]
```

