```
In [186]:
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn import metrics
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
```

Load the data

```
In [187]:
```

```
data=pd.read_csv('breast_cancer.csv')
data
```

Out[187]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	C
0	842302	М	17.99	10.38	122.80	1001.0	0.11840	0.27760	
1	842517	М	20.57	17.77	132.90	1326.0	0.08474	0.07864	
2	84300903	М	19.69	21.25	130.00	1203.0	0.10960	0.15990	
3	84348301	М	11.42	20.38	77.58	386.1	0.14250	0.28390	
4	84358402	М	20.29	14.34	135.10	1297.0	0.10030	0.13280	
564	926424	М	21.56	22.39	142.00	1479.0	0.11100	0.11590	
565	926682	М	20.13	28.25	131.20	1261.0	0.09780	0.10340	
566	926954	М	16.60	28.08	108.30	858.1	0.08455	0.10230	
567	927241	М	20.60	29.33	140.10	1265.0	0.11780	0.27700	
568	92751	В	7.76	24.54	47.92	181.0	0.05263	0.04362	

569 rows × 33 columns

1

In [188]:

```
data.info()
```

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

#	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
 1/ compactness_se
18 concavity_se
                                             569 non-null float64
                                            569 non-null float64
19 concave points_se 569 non-null float64
20 symmetry_se 569 non-null float64
21 fractal_dimension_se 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
 26 smoothness worst
                                            569 non-null
                                                                   float64
                                                                  float64
float64
float64
 27 compactness_worst
                                           569 non-null
 28 concavity_worst
                                            569 non-null
 29 concave points_worst 569 non-null 30 symmetry_worst 569 non-null
                                                                    float64
 31 fractal_dimension_worst 569 non-null
                                                                     float64
 32 Unnamed: 32
                                            0 non-null
                                                                     float64
dtypes: float64(31), int64(1), object(1)
memory usage: 146.8+ KB
In [189]:
data.describe()
```

Out[189]:

	id	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	conc
count	5.690000e+02	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	
mean	3.037183e+07	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	
std	1.250206e+08	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	
min	8.670000e+03	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	
25%	8.692180e+05	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	
50%	9.060240e+05	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	
75%	8.813129e+06	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	
max	9.113205e+08	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	

8 rows × 32 columns

In [190]:

data.head()

Out[190]:

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

5 rows × 33 columns

In [191]:

data.tail()

Out[191]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	CO
564	926424	М	21.56	22.39	142.00	1479.0	0.11100	0.11590	
565	926682	М	20.13	28.25	131.20	1261.0	0.09780	0.10340	
566	926954	М	16.60	28.08	108.30	858.1	0.08455	0.10230	
567	927241	М	20.60	29.33	140.10	1265.0	0.11780	0.27700	
568	92751	В	7.76	24.54	47.92	181.0	0.05263	0.04362	

 $5 \text{ rows} \times 33 \text{ columns}$

1

In [192]:

data.shape

Out[192]:

(569, 33)

In [193]:

data.describe().T

Out[193]:

	count	mean	std	min	25%	50%	75%	
id	569.0	3.037183e+07	1.250206e+08	8670.000000	869218.000000	906024.000000	8.813129e+06	9.1132
radius_mean	569.0	1.412729e+01	3.524049e+00	6.981000	11.700000	13.370000	1.578000e+01	2.811(
texture_mean	569.0	1.928965e+01	4.301036e+00	9.710000	16.170000	18.840000	2.180000e+01	3.9280
perimeter_mean	569.0	9.196903e+01	2.429898e+01	43.790000	75.170000	86.240000	1.041000e+02	1.8850
area_mean	569.0	6.548891e+02	3.519141e+02	143.500000	420.300000	551.100000	7.827000e+02	2.5010
smoothness_mean	569.0	9.636028e-02	1.406413e-02	0.052630	0.086370	0.095870	1.053000e-01	1.634
compactness_mean	569.0	1.043410e-01	5.281276e-02	0.019380	0.064920	0.092630	1.304000e-01	3.454
concavity_mean	569.0	8.879932e-02	7.971981e-02	0.000000	0.029560	0.061540	1.307000e-01	4.268
concave points_mean	569.0	4.891915e-02	3.880284e-02	0.000000	0.020310	0.033500	7.400000e-02	2.012
symmetry_mean	569.0	1.811619e-01	2.741428e-02	0.106000	0.161900	0.179200	1.957000e-01	3.040
fractal_dimension_mean	569.0	6.279761e-02	7.060363e-03	0.049960	0.057700	0.061540	6.612000e-02	9.744
radius_se	569.0	4.051721e-01	2.773127e-01	0.111500	0.232400	0.324200	4.789000e-01	2.8730
texture_se	569.0	1.216853e+00	5.516484e-01	0.360200	0.833900	1.108000	1.474000e+00	4.8850
perimeter_se	569.0	2.866059e+00	2.021855e+00	0.757000	1.606000	2.287000	3.357000e+00	2.1980
area_se	569.0	4.033708e+01	4.549101e+01	6.802000	17.850000	24.530000	4.519000e+01	5.4220
smoothness_se	569.0	7.040979e-03	3.002518e-03	0.001713	0.005169	0.006380	8.146000e-03	3.113
compactness_se	569.0	2.547814e-02	1.790818e-02	0.002252	0.013080	0.020450	3.245000e-02	1.354
concavity_se	569.0	3.189372e-02	3.018606e-02	0.000000	0.015090	0.025890	4.205000e-02	3.960
concave points_se	569.0	1.179614e-02	6.170285e-03	0.000000	0.007638	0.010930	1.471000e-02	5.279
symmetry_se	569.0	2.054230e-02	8.266372e-03	0.007882	0.015160	0.018730	2.348000e-02	7.895
fractal_dimension_se	569.0	3.794904e-03	2.646071e-03	0.000895	0.002248	0.003187	4.558000e-03	2.984
radius_worst	569.0	1.626919e+01	4.833242e+00	7.930000	13.010000	14.970000	1.879000e+01	3.6040
texture_worst	569.0	2.567722e+01	6.146258e+00	12.020000	21.080000	25.410000	2.972000e+01	4.9540
perimeter_worst	569.0	1.072612e+02	3.360254e+01	50.410000	84.110000	97.660000	1.254000e+02	2.5120
area_worst	569.0	8.805831e+02	5.693570e+02	185.200000	515.300000	686.500000	1.084000e+03	4.2540
emonthness worst	569 N	1 323686e-01	2 283243e-N2	N N7117N	N 1166NN	0 131300	1 4600000-01	2 226

	JJJ.J			0.0	0.1.10000	00.000	11-100000 01 EIEE
compactness_worst	569.0	mean 2.542650e-01	std 1.573365e-01	min 0.027290	25% 0.147200	50% 0.211900	75% 3.391000e-01 1.058
concavity_worst	569.0	2.721885e-01	2.086243e-01	0.000000	0.114500	0.226700	3.829000e-01 1.252
concave points_worst	569.0	1.146062e-01	6.573234e-02	0.000000	0.064930	0.099930	1.614000e-01 2.910
symmetry_worst	569.0	2.900756e-01	6.186747e-02	0.156500	0.250400	0.282200	3.179000e-01 6.638
fractal_dimension_worst	569.0	8.394582e-02	1.806127e-02	0.055040	0.071460	0.080040	9.208000e-02 2.07
Unnamed: 32	0.0	NaN	NaN	NaN	NaN	NaN	NaN
4							

Clean and Prepare the data

```
In [194]:
data.diagnosis.unique()
Out[194]:
array(['M', 'B'], dtype=object)
In [195]:
data['diagnosis'].value counts()
Out[195]:
     357
     212
Name: diagnosis, dtype: int64
In [196]:
data.drop('id',axis=1,inplace=True)
In [197]:
data.drop('Unnamed: 32',axis=1,inplace=True)
In [198]:
data.head()
Out[198]:
```

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mea
0	М	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.300
1	М	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.086
2	М	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.197
3	М	11.42	20.38	77.58	386.1	0.14250	0.28390	0.241
4	М	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.198

5 rows × 31 columns

```
In [199]:
data['diagnosis']=data['diagnosis'].map({'M':1,'B':0})
```

```
In [200]:
data.head()
```

Out[200]:

diagnosis radius_mean texture_mean perimeter_mean area_mean smoothness_mean compactness_mean concavity_mea 0 1 17.99 10.38 122.80 1001.0 0.11840 0.27760 0.300 0.086 1 1 20.57 17.77 132.90 1326.0 0.08474 0.07864 2 19.69 21.25 130.00 1203.0 0.10960 0.15990 0.197 0.241 3 1 11.42 20.38 77.58 386.1 0.14250 0.28390 20.29 14.34 135.10 1297.0 0.10030 0.13280 0.198

5 rows × 31 columns

In [201]:

```
data.isnull().sum()
```

Out[201]:

```
diagnosis
                            0
                            0
radius_mean
texture mean
                            0
perimeter mean
                            0
area mean
                            0
smoothness mean
                            0
compactness mean
                            0
concavity_mean
concave points mean
                            0
symmetry_mean
                            0
fractal dimension mean
                            0
radius se
                            0
                            0
texture se
                            0
perimeter se
                            0
area se
smoothness se
                            0
compactness se
concavity_se
concave points_se
                            0
symmetry_se
fractal dimension se
                            0
radius worst
                            0
texture worst
                            0
                            0
perimeter worst
area worst
                            0
smoothness worst
                            0
compactness worst
                            0
concavity_worst
concave points worst
                            0
symmetry_worst
                            0
fractal_dimension_worst
dtype: int64
```

In [202]:

data.corr()

Out[202]:

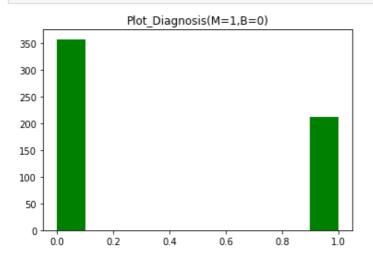
	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness
diagnosis	1.000000	0.730029	0.415185	0.742636	0.708984	0.358560	0.
radius_mean	0.730029	1.000000	0.323782	0.997855	0.987357	0.170581	0.
texture_mean	0.415185	0.323782	1.000000	0.329533	0.321086	-0.023389	0.
perimeter_mean	0.742636	0.997855	0.329533	1.000000	0.986507	0.207278	0.
area_mean	0.708984	0.987357	0.321086	0.986507	1.000000	0.177028	0.
smoothness mean	0.358560	0 170581	-0 023389	N 2N7278	N 177N2R	1 000000	n

JVVIIVVIIVVII	V	·········	V.V.	0.20,2.0	0		٧.
compactness_mean	djagnosia	radiyş5000420	texture299992	perimeter559999	ar 6 34 98582	smoothneas	compactness
concavity_mean	0.696360	0.676764	0.302418	0.716136	0.685983	0.521984	0.
concave points_mean	0.776614	0.822529	0.293464	0.850977	0.823269	0.553695	0.
symmetry_mean	0.330499	0.147741	0.071401	0.183027	0.151293	0.557775	0.
fractal_dimension_mean	-0.012838	-0.311631	-0.076437	-0.261477	-0.283110	0.584792	0.
radius_se	0.567134	0.679090	0.275869	0.691765	0.732562	0.301467	0.
texture_se	-0.008303	-0.097317	0.386358	-0.086761	-0.066280	0.068406	0.
perimeter_se	0.556141	0.674172	0.281673	0.693135	0.726628	0.296092	0.
area_se	0.548236	0.735864	0.259845	0.744983	0.800086	0.246552	0.
smoothness_se	-0.067016	-0.222600	0.006614	-0.202694	-0.166777	0.332375	0.
compactness_se	0.292999	0.206000	0.191975	0.250744	0.212583	0.318943	0.
concavity_se	0.253730	0.194204	0.143293	0.228082	0.207660	0.248396	0.
concave points_se	0.408042	0.376169	0.163851	0.407217	0.372320	0.380676	0.
symmetry_se	-0.006522	-0.104321	0.009127	-0.081629	-0.072497	0.200774	0.
fractal_dimension_se	0.077972	-0.042641	0.054458	-0.005523	-0.019887	0.283607	0.
radius_worst	0.776454	0.969539	0.352573	0.969476	0.962746	0.213120	0.
texture_worst	0.456903	0.297008	0.912045	0.303038	0.287489	0.036072	0.
perimeter_worst	0.782914	0.965137	0.358040	0.970387	0.959120	0.238853	0.
area_worst	0.733825	0.941082	0.343546	0.941550	0.959213	0.206718	0.
smoothness_worst	0.421465	0.119616	0.077503	0.150549	0.123523	0.805324	0.
compactness_worst	0.590998	0.413463	0.277830	0.455774	0.390410	0.472468	0.
concavity_worst	0.659610	0.526911	0.301025	0.563879	0.512606	0.434926	0.
concave points_worst	0.793566	0.744214	0.295316	0.771241	0.722017	0.503053	0.
symmetry_worst	0.416294	0.163953	0.105008	0.189115	0.143570	0.394309	0.
fractal_dimension_worst	0.323872	0.007066	0.119205	0.051019	0.003738	0.499316	0.

31 rows × 31 columns

In [204]:

```
plt.hist(data['diagnosis'],color='g')
plt.title('Plot_Diagnosis(M=1,B=0)')
plt.show()
```



Building Model

```
________.
X=data.drop(['diagnosis'],axis=1)
In [206]:
Y=data['diagnosis']
In [207]:
from sklearn.model selection import train test split
In [208]:
X train, X test, y train, y test = train test split(
  X, Y, test size=0.33, random state=42)
Feature scaling
In [214]:
from sklearn.preprocessing import StandardScaler
In [216]:
ss = StandardScaler()
x train = ss.fit transform(X train)
x test = ss.transform(X test)
Models and find the Best one
Logistic Regression
In [217]:
from sklearn.linear model import LogisticRegression
In [220]:
lr=LogisticRegression()
In [221]:
model=lr.fit(X_train,y_train)
/home/anjali/.local/lib/python3.6/site-packages/sklearn/linear model/ logistic.py:765: Co
nvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as shown in:
   https://scikit-learn.org/stable/modules/preprocessing.html
```

```
0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0,
       1, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1,
       0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0,
       1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1,
       0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 0,
       0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
       1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0])
In [227]:
from sklearn.metrics import confusion matrix
In [229]:
cm=confusion matrix(y test,prediction)
Out[229]:
array([[118,
             3],
      [ 3, 64]])
In [230]:
from sklearn.metrics import accuracy score
In [231]:
accuracy score(y test, prediction)
Out[231]:
0.9680851063829787
Decision Tree
In [232]:
from sklearn.tree import DecisionTreeClassifier
In [234]:
dec=DecisionTreeClassifier()
model1=dec.fit(X_train,y_train)
In [235]:
model1
Out[235]:
DecisionTreeClassifier()
In [236]:
prediction=model1.predict(X test)
In [237]:
prediction
Out[237]:
array([0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0,
       1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0,
       0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0,
       1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1,
       0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0,
       1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1,
       1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0,
```

 \cap \cap \cap 1

∩ 1

 \cap

```
1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0])
In [238]:
cnn=confusion matrix(y test,prediction)
In [239]:
cnn
Out[239]:
array([[104, 17],
   [ 4, 63]])
In [240]:
accuracy score(y test,prediction)
Out[240]:
0.8882978723404256
In [241]:
from sklearn.metrics import classification report
In [242]:
print(classification_report(y_test,prediction))
           precision recall f1-score support
         0
               0.96 0.86
                             0.91 121
                       0.94
         1
                0.79
                                0.86
                                          67
                                0.89
0.88
0.89
                                         188
  accuracy
macro avg 0.88 0.90 weighted avg 0.90 0.89
                                         188
                                         188
In [ ]:
```