

```
In [ ]:
#*****
#                               Breast_Cancer_Prediction_Project
#*****
```

```
In [ ]:
#This project is made using CSV File called Data
#This CSV file is downloaded from https://www.kaggle.com/uciml/breast-cancer-wisconsin-data
#Project by:Anubha Sharma-Data Science
#Submitted to:Jyotika
```

```
In [2]:
#1.Import Libraries
import numpy
import pandas as pd
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
```

Matplotlib is building the font cache; this may take a moment.

```
In [4]:
df=pd.read_csv("data.csv")
df.head()
print(df)
```

| | id | diagnosis | radius_mean | texture_mean | perimeter_mean | area_mean | \ |
|-----|-----------------|------------------|----------------|--------------|----------------|-----------|---|
| 0 | 842302 | M | 17.99 | 10.38 | 122.80 | 1001.0 | |
| 1 | 842517 | M | 20.57 | 17.77 | 132.90 | 1326.0 | |
| 2 | 84300903 | M | 19.69 | 21.25 | 130.00 | 1203.0 | |
| 3 | 84348301 | M | 11.42 | 20.38 | 77.58 | 386.1 | |
| 4 | 84358402 | M | 20.29 | 14.34 | 135.10 | 1297.0 | |
| .. | ... | ... | ... | ... | ... | ... | |
| 564 | 926424 | M | 21.56 | 22.39 | 142.00 | 1479.0 | |
| 565 | 926682 | M | 20.13 | 28.25 | 131.20 | 1261.0 | |
| 566 | 926954 | M | 16.60 | 28.08 | 108.30 | 858.1 | |
| 567 | 927241 | M | 20.60 | 29.33 | 140.10 | 1265.0 | |
| 568 | 92751 | B | 7.76 | 24.54 | 47.92 | 181.0 | |
| | smoothness_mean | compactness_mean | concavity_mean | concave | points_mean | \ | |
| 0 | 0.11840 | 0.27760 | 0.30010 | | 0.14710 | | |
| 1 | 0.08474 | 0.07864 | 0.08690 | | 0.07017 | | |
| 2 | 0.10960 | 0.15990 | 0.19740 | | 0.12790 | | |
| 3 | 0.14250 | 0.28390 | 0.24140 | | 0.10520 | | |
| 4 | 0.10030 | 0.13280 | 0.19800 | | 0.10430 | | |

```

..      ...      ...      ...      ...
564      0.11100      0.11590      0.24390      0.13890
565      0.09780      0.10340      0.14400      0.09791
566      0.08455      0.10230      0.09251      0.05302
567      0.11780      0.27700      0.35140      0.15200
568      0.05263      0.04362      0.00000      0.00000

... texture_worst perimeter_worst area_worst smoothness_worst \
0 ...      17.33      184.60      2019.0      0.16220
1 ...      23.41      158.80      1956.0      0.12380
2 ...      25.53      152.50      1709.0      0.14440
3 ...      26.50      98.87      567.7      0.20980
4 ...      16.67      152.20      1575.0      0.13740
..      ...      ...      ...      ...
564 ...      26.40      166.10      2027.0      0.14100
565 ...      38.25      155.00      1731.0      0.11660
566 ...      34.12      126.70      1124.0      0.11390
567 ...      39.42      184.60      1821.0      0.16500
568 ...      30.37      59.16      268.6      0.08996

compactness_worst concavity_worst concave points_worst symmetry_worst \
0      0.66560      0.7119      0.2654      0.4601
1      0.18660      0.2416      0.1860      0.2750
2      0.42450      0.4504      0.2430      0.3613
3      0.86630      0.6869      0.2575      0.6638
4      0.20500      0.4000      0.1625      0.2364
..      ...      ...      ...      ...
564      0.21130      0.4107      0.2216      0.2060
565      0.19220      0.3215      0.1628      0.2572
566      0.30940      0.3403      0.1418      0.2218
567      0.86810      0.9387      0.2650      0.4087
568      0.06444      0.0000      0.0000      0.2871

fractal_dimension_worst Unnamed: 32
0      0.11890      NaN
1      0.08902      NaN
2      0.08758      NaN
3      0.17300      NaN
4      0.07678      NaN
..      ...      ...
564      0.07115      NaN
565      0.06637      NaN
566      0.07820      NaN
567      0.12400      NaN
568      0.07039      NaN

```

[569 rows x 33 columns]

In [32]:

```
df.describe()
```

Out[32]:

| | id | diagnosis | radius_mean | texture_mean | perimeter_mean | area_mean | smoothness_mean | compactness_mean | concavity_m |
|-------|--------------|------------|-------------|--------------|----------------|-------------|-----------------|------------------|-------------|
| count | 5.690000e+02 | 569.000000 | 569.000000 | 569.000000 | 569.000000 | 569.000000 | 569.000000 | 569.000000 | 569.000000 |
| mean | 3.037183e+07 | 0.372583 | 14.127292 | 19.289649 | 91.969033 | 654.889104 | 0.096360 | 0.104341 | 0.086100 |
| std | 1.250206e+08 | 0.483918 | 3.524049 | 4.301036 | 24.298981 | 351.914129 | 0.014064 | 0.052813 | 0.079000 |
| min | 8.670000e+03 | 0.000000 | 6.981000 | 9.710000 | 43.790000 | 143.500000 | 0.052630 | 0.019380 | 0.000000 |
| 25% | 8.692180e+05 | 0.000000 | 11.700000 | 16.170000 | 75.170000 | 420.300000 | 0.086370 | 0.064920 | 0.029000 |
| 50% | 9.060240e+05 | 0.000000 | 13.370000 | 18.840000 | 86.240000 | 551.100000 | 0.095870 | 0.092630 | 0.061000 |
| 75% | 8.813129e+06 | 1.000000 | 15.780000 | 21.800000 | 104.100000 | 782.700000 | 0.105300 | 0.130400 | 0.130000 |
| max | 9.113205e+08 | 1.000000 | 28.110000 | 39.280000 | 188.500000 | 2501.000000 | 0.163400 | 0.345400 | 0.426000 |

8 rows × 33 columns



In [34]:

```
df.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    int64
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  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
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  fractal_dimension_worst 569 non-null    float64
32  Unnamed: 32            0 non-null      float64

```

dtypes: float64(31), int64(2)

memory usage: 146.8 KB

In [31]:

```

#*****
#1.Training Model
#Here 1 stands for M
#Here 2 stands for B
#*****

from sklearn.preprocessing import LabelEncoder
labelencoder_Y = LabelEncoder()
df.iloc[:,1]=labelencoder_Y.fit_transform(df.iloc[:,1].values)
df.head()

```

Out[31]:

| | id | diagnosis | radius_mean | texture_mean | perimeter_mean | area_mean | smoothness_mean | compactness_mean | concavity_mean | point |
|---|----------|-----------|-------------|--------------|----------------|-----------|-----------------|------------------|----------------|-------|
| 0 | 842302 | 1 | 17.99 | 10.38 | 122.80 | 1001.0 | 0.11840 | 0.27760 | 0.3001 | |
| 1 | 842517 | 1 | 20.57 | 17.77 | 132.90 | 1326.0 | 0.08474 | 0.07864 | 0.0869 | |
| 2 | 84300903 | 1 | 19.69 | 21.25 | 130.00 | 1203.0 | 0.10960 | 0.15990 | 0.1974 | |
| 3 | 84348301 | 1 | 11.42 | 20.38 | 77.58 | 386.1 | 0.14250 | 0.28390 | 0.2414 | |
| 4 | 84358402 | 1 | 20.29 | 14.34 | 135.10 | 1297.0 | 0.10030 | 0.13280 | 0.1980 | |

5 rows × 33 columns

In [15]:

```

#####
# Models/ Algorithms
#####
def models(X_train,Y_train):

    #####
    #Logistic regression
    #####
    from sklearn.linear_model import LogisticRegression
    log=LogisticRegression(random_state=0)
    log.fit(X_train,Y_train)

    #####
    #Decision Tree
    #####
    from sklearn.tree import DecisionTreeClassifier
    tree=DecisionTreeClassifier(random_state=0,criterion="entropy")
    tree.fit(X_train,Y_train)

    #####
    #Random Forest
    #####
    from sklearn.ensemble import RandomForestClassifier
    forest=RandomForestClassifier(random_state=0,criterion="entropy",n_estimators=10)
    forest.fit(X_train,Y_train)

    print('[0]logistic regression accuracy:',log.score(X_train,Y_train))
    print('[1]Decision tree accuracy:',tree.score(X_train,Y_train))
    print('[2]Random forest accuracy:',forest.score(X_train,Y_train))

    return log,tree,forest

```

In [21]:

```
model=models(X_train,Y_train)
```

```

[0]logistic regression accuracy: 0.9912087912087912
[1]Decision tree accuracy: 1.0
[2]Random forest accuracy: 0.9978021978021978

```

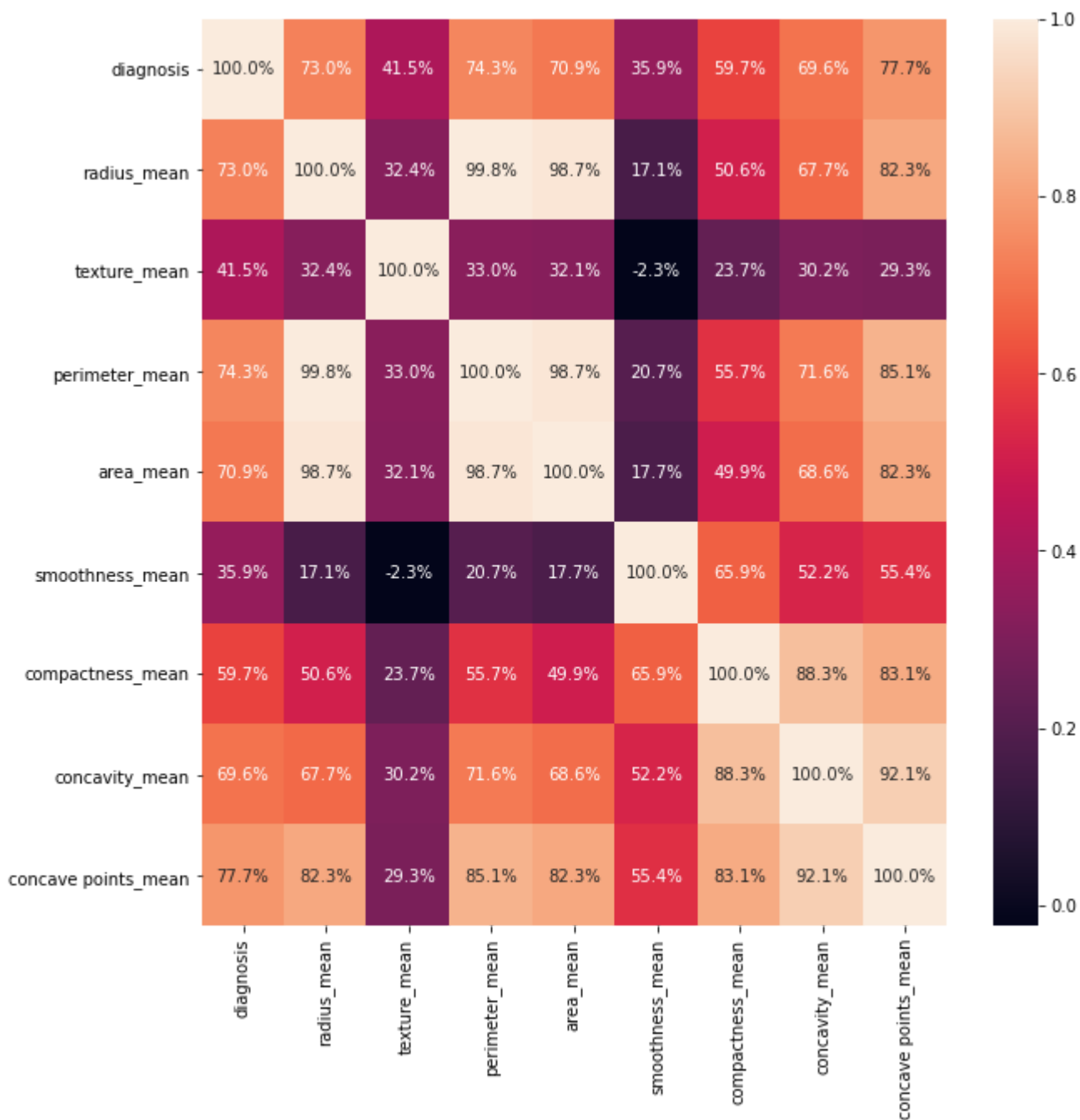
In [20]:

```

plt.figure(figsize=(10,10))
sns.heatmap(df.iloc[:,1:10].corr(),annot=True,fmt=".01%")

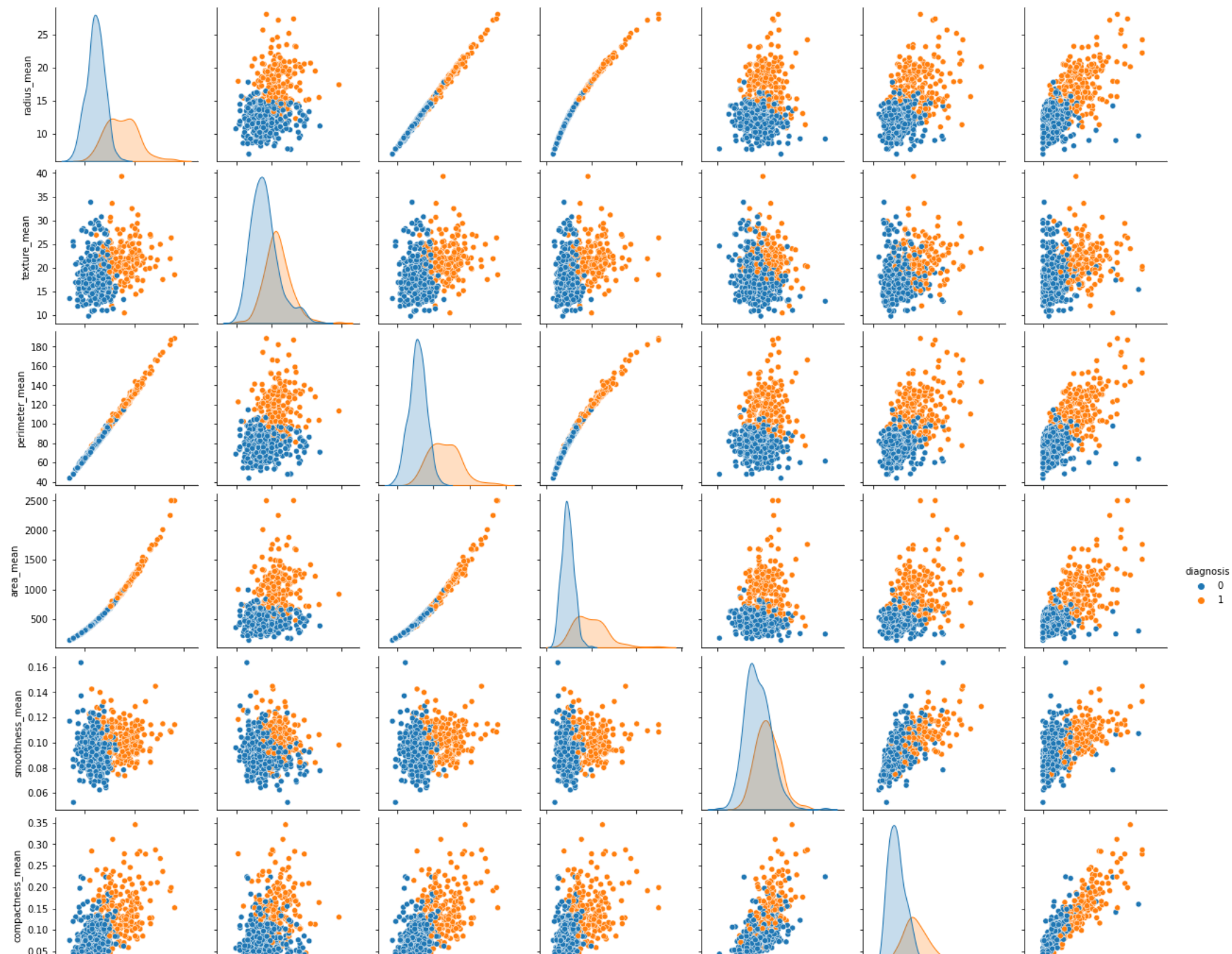
```

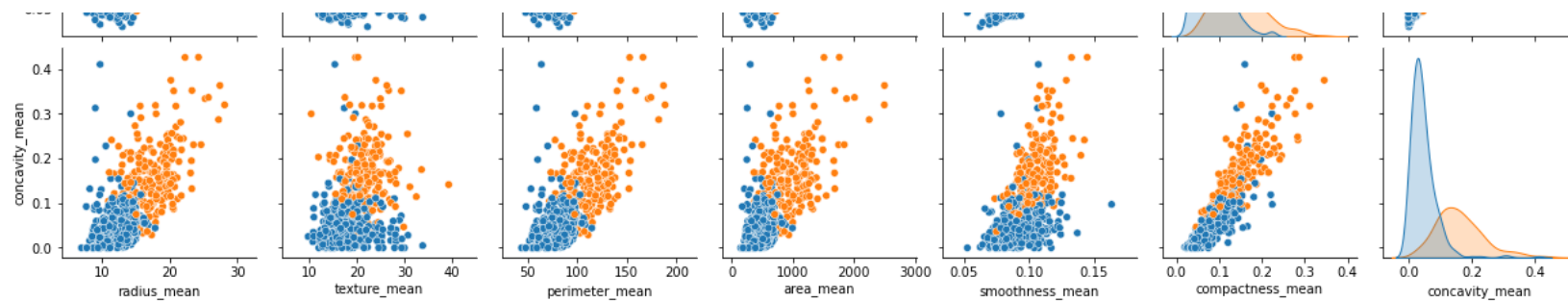
Out[20]: <AxesSubplot:>



```
In [33]: sns.pairplot(df.iloc[:,1:9],hue="diagnosis")
```

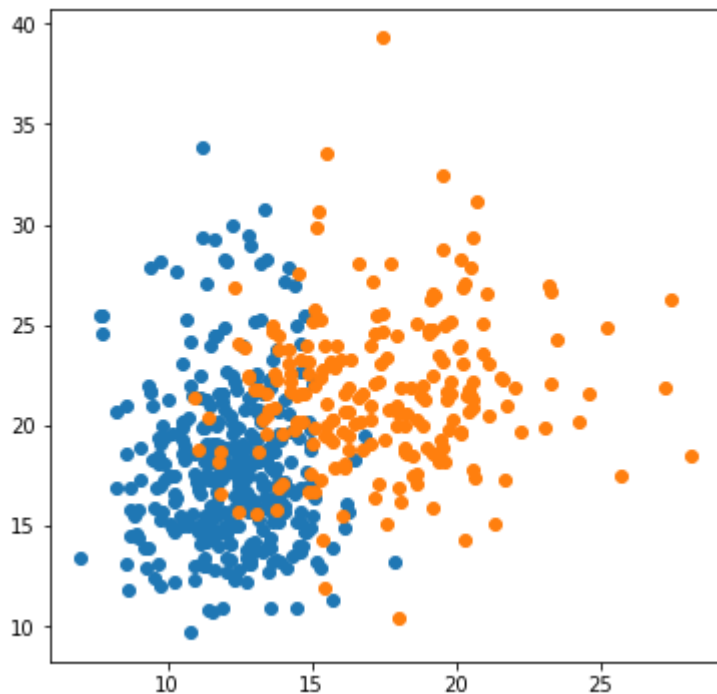
```
Out[33]: <seaborn.axisgrid.PairGrid at 0x29bf4a19280>
```





```
In [39]: plt.figure(figsize=(6,6))
plt.scatter(X[Y == 0][:, 0], X[Y == 0][:, 1], label = '0')
plt.scatter(X[Y == 1][:, 0], X[Y == 1][:, 1], label = '1')
```

```
Out[39]: <matplotlib.collections.PathCollection at 0x29bf7330490>
```



```
In [23]: #*****
# Testing the models
#*****
```



```

from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report

for i in range(len(model)):
    print("Model",i)
    print(classification_report(Y_test,model[i].predict(X_test)))
    print('Accuracy : ',accuracy_score(Y_test,model[i].predict(X_test)))

```

Model 0

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.96 | 0.99 | 0.97 | 67 |
| 1 | 0.98 | 0.94 | 0.96 | 47 |
| accuracy | | | 0.96 | 114 |
| macro avg | 0.97 | 0.96 | 0.96 | 114 |
| weighted avg | 0.97 | 0.96 | 0.96 | 114 |

Accuracy : 0.9649122807017544

Model 1

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.94 | 0.96 | 0.95 | 67 |
| 1 | 0.93 | 0.91 | 0.92 | 47 |
| accuracy | | | 0.94 | 114 |
| macro avg | 0.94 | 0.94 | 0.94 | 114 |
| weighted avg | 0.94 | 0.94 | 0.94 | 114 |

Accuracy : 0.9385964912280702

Model 2

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.96 | 1.00 | 0.98 | 67 |
| 1 | 1.00 | 0.94 | 0.97 | 47 |
| accuracy | | | 0.97 | 114 |
| macro avg | 0.98 | 0.97 | 0.97 | 114 |
| weighted avg | 0.97 | 0.97 | 0.97 | 114 |

Accuracy : 0.9736842105263158

In [18]:

```

#*****
#Accuracy
#*****
pred=model[2].predict(X_test)
print('Predicted values:')

```

```
print(pred)
print('Actual values:')
print(Y_test)
```

Predicted values:

```
[1 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 1 1 1 1 1 0 0 1 0 0 1 0 1 0 1 0 1 0
 1 0 1 0 0 0 0 0 1 0 0 0 1 1 1 1 0 0 0 0 0 0 1 1 1 0 0 1 0 1 1 1 0 0 1 0 0
 1 0 0 0 0 0 1 1 1 0 1 0 0 0 1 1 0 1 0 1 0 0 1 0 0 0 0 0 0 0 1 0 1 0 1 1 0
 1 1 0]
```

Actual values:

```
[1 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 1 1 1 1 1 0 0 1 0 0 1 0 1 0 1 0 1 0
 1 0 1 1 0 1 0 0 1 0 0 0 1 1 1 1 0 0 0 0 0 0 1 1 1 0 0 1 0 1 1 1 0 0 1 0 1
 1 0 0 0 0 0 1 1 1 0 1 0 0 0 1 1 0 1 0 1 0 0 1 0 0 0 0 0 0 0 1 0 1 0 1 1 0
 1 1 0]
```

In []:

```
#####
#
#                               End of the Project
#####
```