



Experiment-3.3

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Subject Name: Advanced Python Programming Subject Code: 23CSH-623

1. Aim: Breast Cancer case study using Python Libraries

2. Objective: A breast cancer case study using Python typically involves the analysis of breast cancer data to build a predictive model for classification (malignant or benign tumors). In this example, I'll use the Breast Cancer Wisconsin (Diagnostic) dataset, which is commonly used for such studies. We'll use the scikit-learn library for machine learning and other libraries for data analysis and visualization. This case study covers loading the dataset, exploring and visualizing the data, preprocessing, and building a simple Logistic Regression model for breast cancer classification. further explore more advanced models, hyperparameter tuning, and feature selection based on your specific requirements.

3. Algorithm/ Steps for Experiment:

- 1. Import necessary libraries for data manipulation, analysis, visualization, and machine learning.
- 2. Load the Breast Cancer dataset (e.g., Breast Cancer Wisconsin dataset.
- 3. Visualize relationships between features using pairplots and correlation heatmaps.
- 4. Split the data into features (X) and target (y).
- 5. Split the data into training and testing sets.
- 6. Standardize features using StandardScaler.
- 7. Choose a machine learning model (e.g., Logistic Regression).
- 8. Train the model using the training set.
- 9. Make predictions on the test set.
- 10. Evaluate the model's performance using accuracy, confusion matrix, and classification report.







4. Code for Experiment and output:

In [38]:	print(df.columns)										
	Inde	'area_ 'conca 'area_ 'fract 'peri1 'co1pa 'sy11e	_lean', 'slo ave points_1 _se', 'colpa cal_dilensio Leter_worst' actness_wors	othness_1ea ean', 'sy11 ctness_se', n_se', 'rad , 'area_wor t', 'concav	etry_1ean', 'concavity_! ius_worst', st', 's1ooth	ness_1ean 'radius_se se', 'cond 'texture_v ness_worst 'concave p	, 'concavity_1 e', 'perileter_ cave points_se' worst',	se',			
In [39]:	▶ df										
Out[39	9]:										
		diagnosis	radius_1ean	texture_1ean	peri1eter_1ean	area_1ean	s1oothness_1ean	co1pactness_1ean	concavity_1ean	concave points_1ean	sy11etry_1ean
	0		radius_1ean	texture_1ean 10.38	peri1eter_1ean 122.80	area_1ean 1001.0	s1oothness_1ean 0.11840	co1pactness_1ean 0.27760	concavity_1ean 0.30010		D (2.00)
	0		(20)		5 50	- IP			2070	points_1ean	D (2.00)
		1	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.30010	0.14710	0.2419 0.1812
	1	1 1	17.99 20.57	10.38 17.77	122.80 132.90	1001.0 1326.0	0.11840 0.08474	0.27760 0.07864	0.30010 0.08690	0.14710 0.07017	0.2419 0.1812
	1	1 1 1	17.99 20.57 19.69	10.38 17.77 21.25	122.80 132.90 130.00	1001.0 1326.0 1203.0	0.11840 0.08474 0.10960	0.27760 0.07864 0.15990	0.30010 0.08690 0.19740	0.14710 0.07017 0.12790	0.2419 0.1812 0.2069 0.2597
	1 2 3	1 1 1 1	17.99 20.57 19.69 11.42	10.38 17.77 21.25 20.38	122.80 132.90 130.00 77.58	1001.0 1326.0 1203.0 386.1	0.11840 0.08474 0.10960 0.14250	0.27760 0.07864 0.15990 0.28390	0.30010 0.08690 0.19740 0.24140	0.14710 0.07017 0.12790 0.10520	0.2419 0.1812 0.2069 0.2597
	1 2 3 4	1 1 1 1	17.99 20.57 19.69 11.42 20.29	10.38 17.77 21.25 20.38 14.34	122.80 132.90 130.00 77.58 135.10	1001.0 1326.0 1203.0 386.1 1297.0	0.11840 0.08474 0.10960 0.14250 0.10030	0.27760 0.07864 0.15990 0.28390 0.13280	0.30010 0.08690 0.19740 0.24140 0.19800	0.14710 0.07017 0.12790 0.10520 0.10430	0.2419 0.1812 0.2069 0.2597 0.1809







In [40]: M df.shape
Out[40]: (569, 27)

In [41]: M df.size
Out[41]: 15363

In [42]: M df.describe()

Out[42]:

	diagnosis	radius_1ean	texture_1ean	peri1eter_1ean	area_1ean	s1oothness_1ean	co1pactness_1ean	concavity_1ean	concave points_1ean	sy11etry_1ea
count	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.00000
mean	0.372583	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	0.088799	0.048919	0.18116
std	0.483918	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0.079720	0.038803	0.02741
min	0.000000	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0.000000	0.000000	0.10600
25%	0.000000	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0.029560	0.020310	0.16190
50%	0.000000	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0.061540	0.033500	0.17920
75%	1.000000	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0.130700	0.074000	0.19570
max	1.000000	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0.426800	0.201200	0.30400

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In [43]: ► df.info()

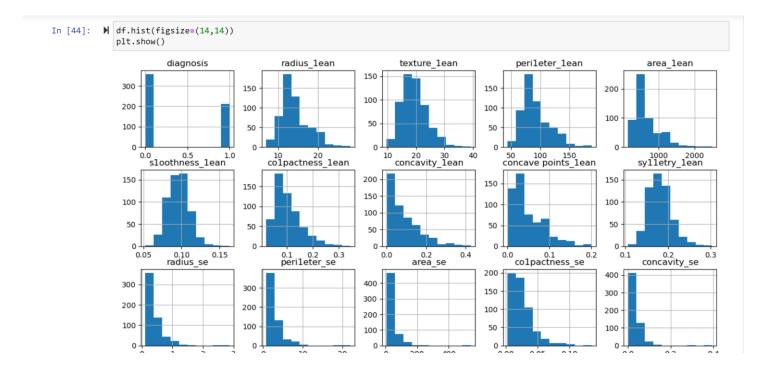
<class 'pandas.core.frame.DataFrame'> RangeIndex: 569 entries, 0 to 568 Data columns (total 27 columns):

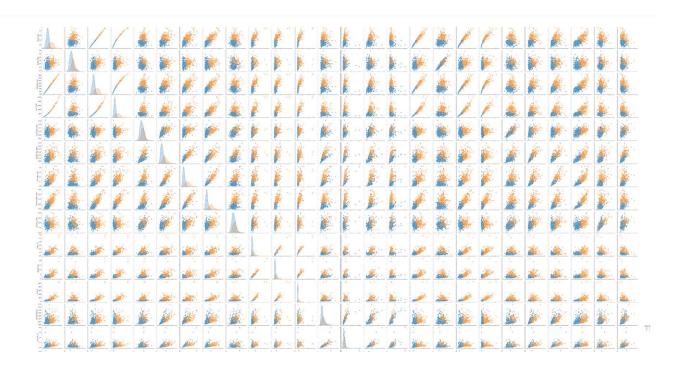
#	Column	Non-Null Count	Dtype
0	diagnosis	569 non-null	int64
1	radius_1ean	569 non-null	float64
2	texture_1ean	569 non-null	float64
3	peri1eter_1ean	569 non-null	float64
4	area_1ean	569 non-null	float64
5	s1oothness_1ean	569 non-null	float64
6	co1pactness_1ean	569 non-null	float64
7	concavity_1ean	569 non-null	float64
8	concave points_1ean	569 non-null	float64
9	sy11etry_1ean	569 non-null	float64
10	radius_se	569 non-null	float64
11	peri1eter_se	569 non-null	float64
12	area_se	569 non-null	float64
13	co1pactness_se	569 non-null	float64
14	concavity_se	569 non-null	float64
15	concave points_se	569 non-null	float64
16	<pre>fractal_di1ension_se</pre>	569 non-null	float64
17	radius_worst	569 non-null	float64
18	texture_worst	569 non-null	float64
19	peri1eter_worst	569 non-null	float64
20	area_worst	569 non-null	float64
21	s1oothness_worst	569 non-null	float64
^^		500 11	







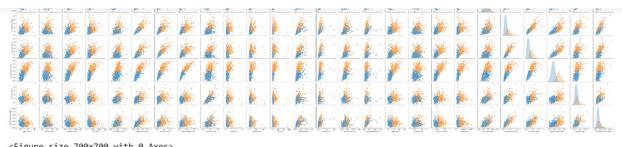












<Figure size 700x700 with 0 Axes>

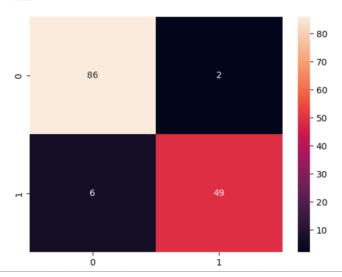
In [46]: ▶ #define X variables and our target(y) X = df.drop(['diagnosis'],axis=1).values y = df['diagnosis'].values #split Train and Test from sklearn.model_selection import train_test_split X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=101)

In [47]: ▶ from sklearn.svm import SVC svc_model = SVC() svc_model.fit(X_train, y_train)

SVC()

y_predict = svc_model.predict(X_test) cm = confusion_matrix(y_test, y_predict) sns.heatmap(cm, annot=True)

Out[48]: <Axes: >









Learning Outcomes:

- 1. Learnt how to import libraries in python program.
- 2. Learnt to load dataset in python.
- 3. Learnt how to train model in python.
- 4. Learnt how to evaluate model's performance using accuracy, confusion matrix, and classification report.

