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No.02	
	Visualizing using Graphs
	Aim:
	To develop a Python program for
	analysing and visualising data using
	graphical representations. The objective
	Ps to use various types of plots
	to understand data distribution.
	relationships and patterns within a
	dataset effectively.
	Procedure:
	step of: Import the necessary librarie
	for data analysis and visualisation,
	such as Pandas, Matplotleb, Seaborn
	and tools to load datasets.
	Step 02: Load a structured dataset
	surbable for resualisation and convert
	9t 90to a tabular format using
	Pandas for easy handling.
	Step 03: Explore the dataset by
-	viewing 9ts structure checking for
	missing values, and generating basic
	statested summaries.

Step 04: Create line plots to observe trends or charges in numerical values across records.

step 05: Use scatter plots to study relationships between two numerical attributes, possibly cooked by category.

step of: Draw bar plots to display the frequency distribution of categorical data.

step of: Plot histograms to understand the distribution of a numerical feature across intervals.

Step 08: (nenerate bac plots to compare the distribution and spread of values across different categories.

step 09: Use advanced visualisations such as count plots, violin plots, KDE plots, and pair plots to analyse patterns, relationships and class-wise comparisons in the datasets.

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Output:

Kclass 'pandas.core.frame. DataFrame'>

Range Index: 569 entries, 0 to 568

Data columns (total 31 columns):

mean radius mean texture

mean perimeter mean area

mean Smoothness

mean Compactness

mean Concavity

mean concave points

mean Symmetry

mean fractal dimension mean error

texture error

perimeter error

rorrs error

smoothness error

Compactness error

Concavity error

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worst radius

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step 10: Display a heatmap to show the correlation between different features. highlighting the strength and direction of relationships.

Program 2a:

Propost matplotlib.pyplot as plt
Propost pandas as pd
from skilearn.dabasets Propost load_breast_
cancer

cancer = load_breast_cancer()

data=pd. DataFrame (cancerdata, cdumns= cancer, feature names) data[target'] = cancer, target

print (data. info ())
print (data. head(1))
print (data. describe ())
print (data. isnull). sum())

pt. figure (figsize = (10,6))

pt. plot (data.index, data [mean radius'],

label = 'Mean Radius']

pt. title ('Line Plot of Mean Radius')

pt. xlabel ('Index')

pt. ylabel ('Mean Radius')

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	ory usage:	135.7. KB			
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50%	13.37 0000	18.840000	86.240000	551.10000
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max	28.110000	39,280000	188.50000	2501.00000
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Std	0-014064	0.052813	0.079720	0-03 2203
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mase	0.163400	0.345400	0.426300	The state of the s
	mean symmetry	mean fro		Worst exture \
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Std	0.027414	0.007060		6.146258
min	0.106000	0.049960		12.020000
25%	0-161600	0.057700		7.030000
50%.	0.179200	0.061540		25.410000

Page No.:2-3
plt. title ('Histogram of Mean Area') plt. xlabel ('Mean Area') plt. ylabel ('Frequency') plt. grid (True) plt. show()
pit. figure (figsize = (10,6)) pit. boxplot (Idata [data ['target'] == 0] ['mean radius'], data [data ['target'] ==1] ['mean radius'], labels = ['Malignant', 'Benign']) pit. title ('Bar Plot of Mean Radius by Target Class') pit. rdabel ('Target Class') pit. ylabel ('Mean Radius')
plt. grid (Toue) plt. Show()

Page No.	2-8				
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max	0.304000	0.097440	49.54 0000		
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	Perimeter	area smoothness	Compactness \		
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count	569.000000	569.000000	569.000000		
mean	0.745188	0.114606	0.290076		
Std	0.208624	0.65732	0.061867		
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	Page No.: 2.9
Program	n ab'.
impost.	seaborn as sns
Impost	pandas as pd
The second secon	matplotlib.pyplot as plt
	sklearn.datasets Propost load-breast-c
Cancer	load_breast_cancer()
data=po	1. DabaFrame (cancer. data, columna: canc
	feature-names)
datalit	carget'I = cancer.barget
	lata. info(s)
printle	data-head (1))
print(d	ata.describe()
printle	data. isnuil (). sum())
plt.figu	re (figsize = (6,4))
shs.cou	sobplot (x = 'target', daba = daba,
	palette : coolwarm
plt. Litle	('Count Plot of Target Classe
plt. xlat	sel ('Target Class')
pik.ylal	bel ('Count')
	cks (bicks = [0.1] labels = [Maligna
-	Benigh'I)
plt.shoc	
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max 0.207500	1.000000
La rous x 31 columns.	J
mean radius	0
mean texture	0
mean perimeter	6
mean area	0
mean Smoothness	ь
mean compactness	. 0
mean concavity	0
mean Concave points	0
mean symmetry	0
mean fractal dimension	0
radius error	0
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perimeter error	0
area error	0
smoothness exxor	0
Compactness error	0
concavity error	0
concave points error	0
Symmetry error	0
fractal dimension error	0
wordt radius	0
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worst perimeter	0
	9

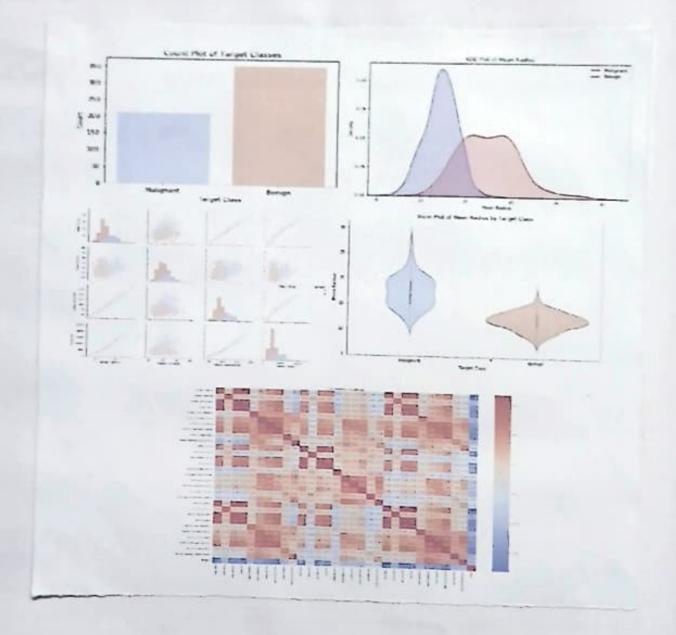
```
plt.figure (figsize = (10,6))
sns. Adeplot (data = data [data ['target'] ==0]
           I'mean radius'I shade = True,
                     label = Malignant , color='r)
sns.kdeplot(data=data [data [target] == 1]
          [mean radius'], Shade = True,
                  label = Benign', color = b')
plt. Little ('KDE Plot of Mean Radius')
plt.xlabel (Mean Radius)
plt. ylabel ('Density')
pit.legend ()
plt. show()
plt. figure (figsize = (10.6))
sns. violinplot (x = target, y = mean radius,
     data=data, palette = 'coolwarm')
pittitle ('Violin plot of Mean Radius by
                        Target Class')
pit. xlabel ('Target Class')
pit. ylabel (Mean Radius')
pit.xticks (ticks = Lo, 1], labels = ['Malignant',
                           Benign'])
pit.show()
sns. paraplot (data, vass: Emean sadius! mean
          texture, mean perimeter, mean areal
                        .hue - barget' palette =
                                       (coolwarmi)
plt. Litle ('Pary Plot')
```

Page No.:31 wast area worst smoothness worst compactness worst concavity worst concave points worst symmetry worst fractal demension bargeb dtype: int 64 Output 2a:

Page No.;33
 pit.show()
 plt. figure (fig size= (20,20))
 sns.heatmap (data, corre), annot = True, frot= '. 2f', cmap= 'coolwarm')
ptt. Bitle ('Correlation Heatmap')
plk.show()

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Output ab:



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Section 1000	
	Result:
	The program was successfully
	implemented to visualise data using graphs, helps to understand
	graphs, helps to understand
0	distributions, compare features, and
9	identify patterns for better analysis.
7,19,0	