eda-laboratory

November 13, 2024

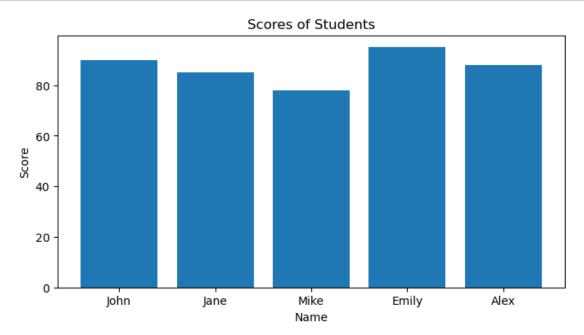
1 Exp 1: Install the Data Analysis and Visualizatio tool: R/ Python/ Tableau Public/ Power BI

```
[1]: import pandas as pd
     import matplotlib.pyplot as plt
[2]: # Sample data
     data = {
     'Name': ['John', 'Jane', 'Mike', 'Emily', 'Alex'],
     'Age': [25, 30, 22, 28, 35],
     'Score': [90, 85, 78, 95, 88]
     }
[3]: # Create a DataFrame from the data
     df = pd.DataFrame(data)
[4]: # Display the DataFrame
     print("DataFrame:")
     print(df)
    DataFrame:
        Name Age
                   Score
    0
        John
               25
                      90
    1
        Jane
               30
                      85
    2
       Mike
               22
                      78
    3 Emily
               28
                      95
        Alex
               35
                      88
[5]: # Data Analysis
     print("\nData Analysis:")
     print("Mean Age:", df['Age'].mean())
     print("Maximum Score:", df['Score'].max())
     print("Minimum Score:", df['Score'].min())
    Data Analysis:
```

Mean Age: 28.0

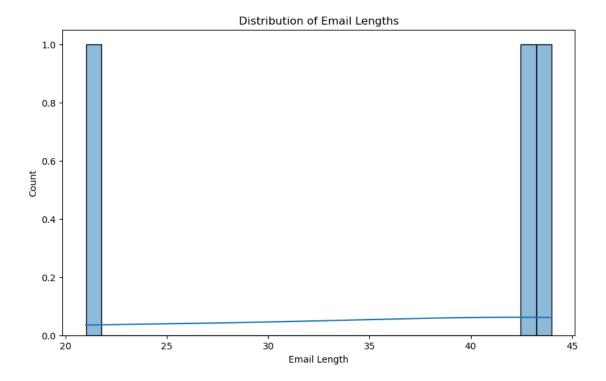
Maximum Score: 95 Minimum Score: 78

```
[6]: # Data Visualization
plt.figure(figsize=(8, 4))
plt.bar(df['Name'], df['Score'])
plt.xlabel('Name')
plt.ylabel('Score')
plt.title('Scores of Students')
plt.show()
```

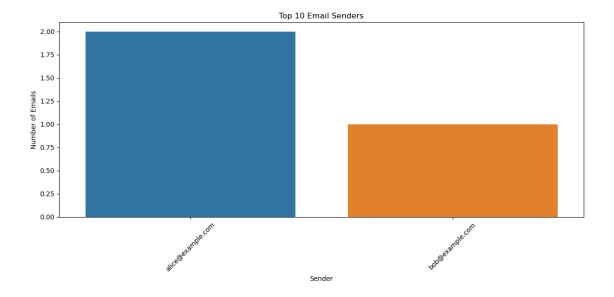


2 Perform EDA with Datasets like Email Dataset

```
[9]: df = pd.DataFrame(data)
     df['timestamp'] = pd.to_datetime(df['timestamp'])
     df.to_csv('emails.csv', index=False)
     print("CSV file created successfully.")
     CSV file created successfully.
[10]: import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
     C:\Users\JAWAHAR A S\anaconda3\lib\site-packages\scipy\__init__.py:155:
     UserWarning: A NumPy version >=1.18.5 and <1.25.0 is required for this version
     of SciPy (detected version 1.26.4
       warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"
[11]: df = pd.read_csv("emails.csv")
     print(df.info())
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 3 entries, 0 to 2
     Data columns (total 5 columns):
                    Non-Null Count Dtype
          Column
                    _____
     --- -----
         sender
                    3 non-null
                                    object
      0
      1
         receiver
                    3 non-null
                                    object
      2
          subject
                     3 non-null
                                    object
      3
         timestamp 3 non-null
                                    object
                     3 non-null
                                    object
          content
     dtypes: object(5)
     memory usage: 248.0+ bytes
[12]: df['timestamp'] = pd.to_datetime(df['timestamp'])
     df.dropna(inplace=True)
     df['email_length'] = df['content'].apply(len)
     plt.figure(figsize=(10, 6))
     sns.histplot(data=df, x='email_length', bins=30, kde=True)
     plt.xlabel('Email Length')
     plt.ylabel('Count')
     plt.title('Distribution of Email Lengths')
     plt.show()
```

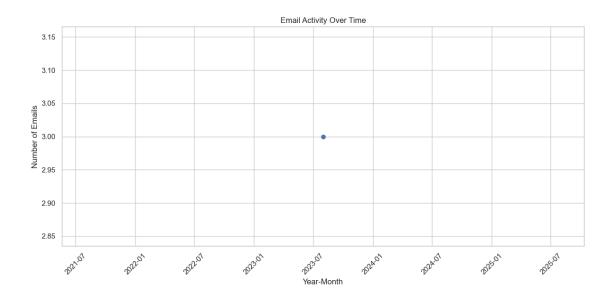


```
[13]: top_senders = df['sender'].value_counts()[:10]
    top_receivers = df['receiver'].value_counts()[:10]
    plt.figure(figsize=(12, 6))
    sns.barplot(x=top_senders.index, y=top_senders.values)
    plt.xticks(rotation=45)
    plt.xlabel('Sender')
    plt.ylabel('Number of Emails')
    plt.title('Top 10 Email Senders')
    plt.tight_layout()
    plt.show()
```



```
[14]: import matplotlib.pyplot as plt
      import pandas as pd
      import seaborn as sns
      df['timestamp'] = pd.to_datetime(df['timestamp'], errors='coerce')
      df['year_month'] = df['timestamp'].dt.to_period('M')
      email_activity = df.groupby('year_month').size()
      print(email_activity)
      email_activity = email_activity.reset_index()
      email_activity['year_month'] = email_activity['year_month'].dt.to_timestamp()
      sns.set(style="whitegrid")
      plt.figure(figsize=(12, 6))
      plt.plot(email_activity['year_month'], email_activity[0], marker='o', color='b')
      plt.xlabel('Year-Month')
      plt.ylabel('Number of Emails')
     plt.title('Email Activity Over Time')
      plt.xticks(rotation=45)
      plt.tight_layout()
      plt.show()
```

year_month 2023-08 3 Freq: M, dtype: int64



3 Exp 3: Working with Numpy Arrays, Pandas Data Frame, Basic Plots using Matplotlib

3.0.1 1) Numpy Arrays

```
[15]: import numpy as np
      arr=np.array([[1,2,3],[4,2,5]])
      print("Array is of type:",type(arr))
      print("No of dimensions:",arr.ndim)
      print("Shape of array:",arr.shape)
      print("Size of array:",arr.size)
      print("Array stores elements of type:",arr.dtype)
     Array is of type: <class 'numpy.ndarray'>
     No of dimensions: 2
     Shape of array: (2, 3)
     Size of array: 6
     Array stores elements of type: int32
[16]: import numpy as np
      a=np.array([[1,2,3],[3,4,5],[4,5,6]])
      print(a)
      print("After Slicing")
      print(a[1:])
     [[1 2 3]
      [3 4 5]
      [4 5 6]]
```

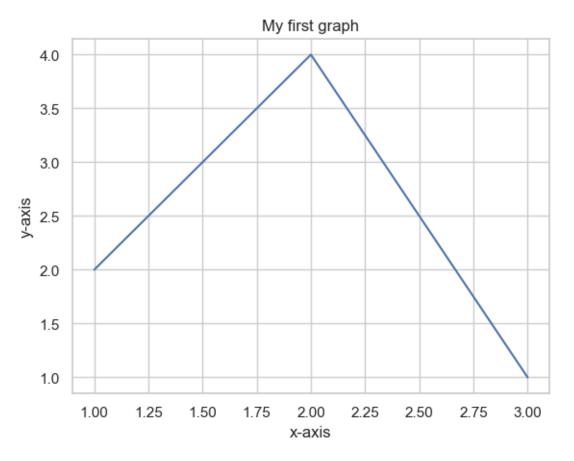
```
[4 5 6]]
[17]: import numpy as np
      a=np.array([[1,2,3],[3,4,5],[4,5,6]])
      print('Our arrayis:')
      print(a)
      print('The items in the second column are:')
      print(a[...,1])
      print('\n')
      print('The items in the second row are:')
      print(a[1,...])
      print('\n')
      print('The items column 1 onwards are:')
      print(a[...,1])
     Our arrayis:
     [[1 2 3]
      [3 4 5]
      [4 5 6]]
     The items in the second column are:
     [2 4 5]
     The items in the second row are:
     [3 4 5]
     The items column 1 onwards are:
     [2 4 5]
     3.0.2 2) Pandas Dataframes
[18]: import pandas as pd
      data = {'Name': ['Alice', 'Bob', 'Charlie'],
              'Age': [25, 30, 35],
              'City': ['New York', 'Los Angeles', 'Chicago']}
      df = pd.DataFrame(data)
      print(df)
      print("\nAge Column:")
      print(df['Age'])
```

After Slicing [[3 4 5]

```
print("\nRow at index 1:")
print(df.iloc[1])
df['Country'] = ['USA', 'USA', 'USA']
print("\nDataFrame with new column 'Country':")
print(df)
print("\nBasic Statistics:")
print(df.describe())
      Name Age
                        City
0
     Alice
             25
                    New York
1
       Bob
             30 Los Angeles
2 Charlie
                     Chicago
             35
Age Column:
     25
1
     30
2
     35
Name: Age, dtype: int64
Row at index 1:
Name
                Bob
Age
                 30
        Los Angeles
City
Name: 1, dtype: object
DataFrame with new column 'Country':
      Name Age
                        City Country
     Alice
                    New York
                                 USA
0
             25
       Bob
             30 Los Angeles
                                 USA
2 Charlie
             35
                     Chicago
                                 USA
Basic Statistics:
        Age
        3.0
count
mean
       30.0
std
        5.0
min
       25.0
25%
       27.5
50%
       30.0
75%
       32.5
       35.0
max
```

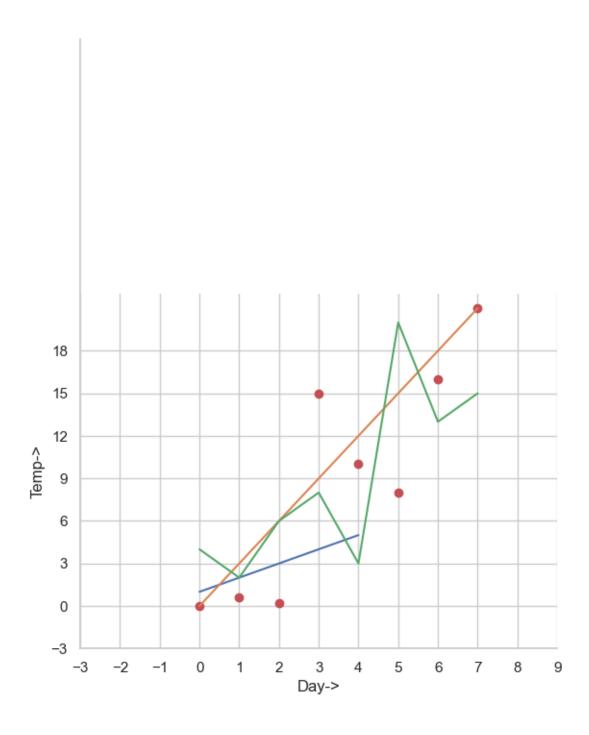
3.0.3 3) Basic Plot using Matplotlib

```
[19]: import matplotlib.pyplot as plt
    x=[1,2,3]
    y=[2,4,1]
    plt.plot(x,y)
    plt.xlabel('x-axis')
    plt.ylabel('y-axis')
    plt.title('My first graph')
    plt.show()
```



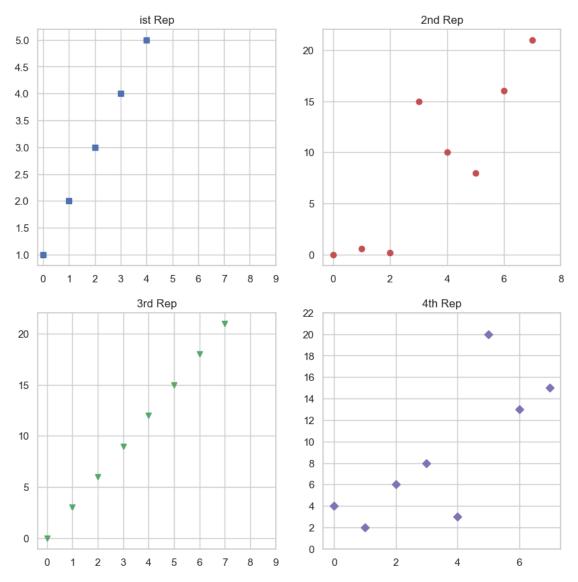
```
[20]: import matplotlib.pyplot as plt
a=[1,2,3,4,5]
b=[0,0.6,0.2,15,10,8,16,21]
plt.plot(a)
plt.plot(b,"or")
plt.plot(list(range(0,22,3)))
plt.xlabel('Day->')
plt.ylabel('Temp->')
c=[4,2,6,8,3,20,13,15]
```

```
plt.plot(c,label='4th Rep')
      ax=plt.gca()
      ax.spines['right'].set_visible(False)
      ax.spines['top'].set_visible(False)
      ax.spines['left'].set_bounds(-3,40)
      plt.xticks(list(range(-3,10)))
      plt.yticks(list(range(-3,20,3)))
[20]: ([<matplotlib.axis.YTick at 0x1c22a9f5be0>,
        <matplotlib.axis.YTick at 0x1c22a9f5460>,
        <matplotlib.axis.YTick at 0x1c22a9ef3d0>,
        <matplotlib.axis.YTick at 0x1c22a68ce20>,
        <matplotlib.axis.YTick at 0x1c22a6975b0>,
        <matplotlib.axis.YTick at 0x1c22a68c970>,
        <matplotlib.axis.YTick at 0x1c22a67d250>,
        <matplotlib.axis.YTick at 0x1c22a697340>],
       [Text(0, 0, ''),
        Text(0, 0, '')])
```



```
[21]: import matplotlib.pyplot as plt
a=[1,2,3,4,5]
b=[0,0.6,0.2,15,10,8,16,21]
c=[4,2,6,8,3,20,13,15]
fig=plt.figure(figsize=(10,10))
sub1=plt.subplot(2,2,1)
sub2=plt.subplot(2,2,2)
```

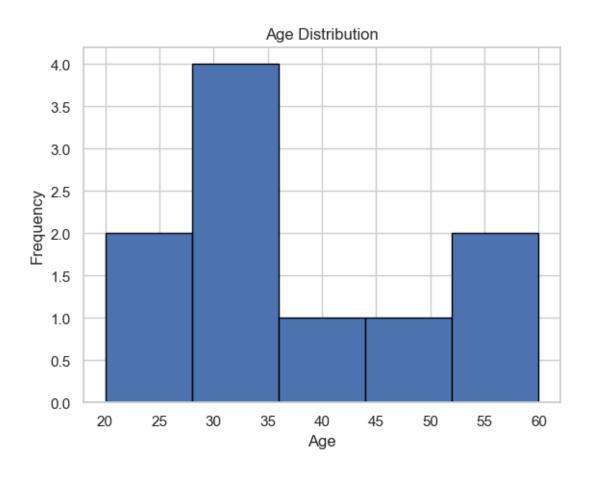
```
sub3=plt.subplot(2,2,3)
sub4=plt.subplot(2,2,4)
sub1.plot(a,'sb')
sub1.set_xticks(list(range(0,10,1)))
sub1.set_title('ist Rep')
sub2.plot(b,'or')
sub2.set_xticks(list(range(0,10,2)))
sub2.set_title('2nd Rep')
sub3.plot(list(range(0,22,3)),'vg')
sub3.set_xticks(list(range(0,10,1)))
sub3.set_title('3rd Rep')
sub4.plot(c,'Dm')
sub4.set_yticks(list(range(0,24,2)))
sub4.set_title('4th Rep')
plt.show()
```

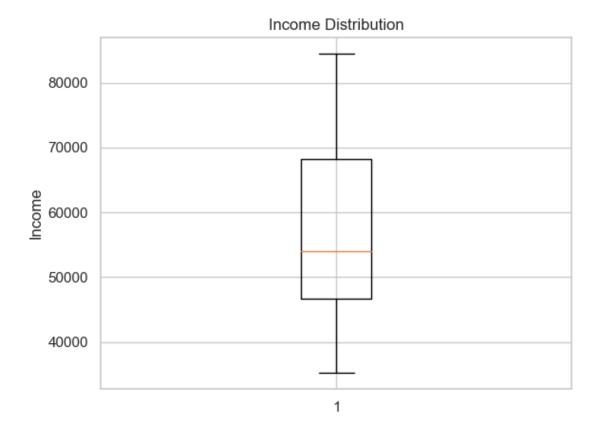


4 Exp 4: Explore various variable and row filters in R for Cleaning Data

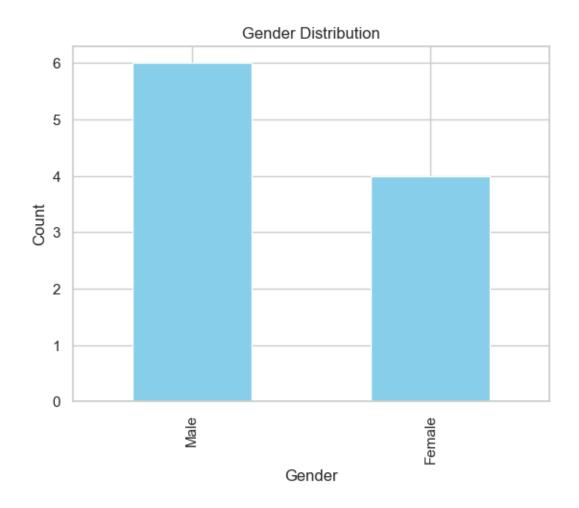
```
[22]: import pandas as pd
     import numpy as np
[23]: data = {
      'ID': range(1, 11),
      'Age': np.random.randint(18, 65, size=10),
      'Income': np.random.randint(30000, 90000, size=10),
      'Gender': ['Male', 'Female', 'Male', 'Female', 'Male', 'Female', 'Male',
      'Education': ['High School', 'Bachelor', 'Master', 'PhD', 'Bachelor', 'Master',
       ⇔'Bachelor','PhD','High School', 'Master']
     }
[24]: df = pd.DataFrame(data)
     print(df.head())
     print(df.describe())
        ID
           Age Income
                         Gender
                                   Education
             30
                                High School
     0
         1
                  53149
                           Male
     1
         2
             30
                  84492
                       Female
                                    Bachelor
     2
         3
             60
                  46930
                                      Master
                           Male
     3
         4
             43
                  68991
                        Female
                                         PhD
     4
         5
             34
                  71345
                           Male
                                    Bachelor
                  ID
                            Age
                                       Income
                     10.000000
     count
            10.00000
                                    10.000000
             5.50000 37.900000
                                56429.700000
     mean
     std
             3.02765 13.253511
                                15976.774748
     min
             1.00000 20.000000
                                35237.000000
     25%
             3.25000 30.000000 46673.500000
     50%
             5.50000 33.000000 53941.000000
     75%
             7.75000 46.750000 68260.500000
            10.00000
                     60.000000 84492.000000
     max
[25]: print(df.isnull().sum())
     print(df['Gender'].unique())
     print(df['Education'].unique())
     selected_columns = df[['Age', 'Income']]
     print(selected_columns.head())
     filtered_data = df[df['Age'] > 30]
     print(filtered_data.head())
```

```
filtered_rows = df[(df['Gender'] == 'Male') & (df['Education'] == 'Master')]
      print(filtered_rows.head())
     ID
                  0
                  0
     Age
                  0
     Income
     Gender
                  0
     Education
                  0
     dtype: int64
     ['Male' 'Female']
     ['High School' 'Bachelor' 'Master' 'PhD']
        Age Income
         30
              53149
     0
     1
         30
              84492
     2
         60
              46930
     3
         43
              68991
         34
              71345
        ID Age Income Gender Education
     2
         3
             60
                  46930
                           Male
                                   Master
                                       PhD
     3
         4
             43
                  68991 Female
     4
         5
             34
                           Male Bachelor
                  71345
         7
     6
             48
                  35237
                           Male Bachelor
     7
                  36763 Female
                                       PhD
         8
             56
            Age Income Gender Education
        ID
     2
         3
                  46930
                          Male
                                  Master
             60
                                  Master
     9
       10
             20
                  54733
                          Male
[26]: import matplotlib.pyplot as plt
[27]: plt.hist(df['Age'], bins=5, edgecolor='black')
      plt.title('Age Distribution')
      plt.xlabel('Age')
      plt.ylabel('Frequency')
      plt.show()
      plt.boxplot(df['Income'])
      plt.title('Income Distribution')
      plt.ylabel('Income')
      plt.show()
```

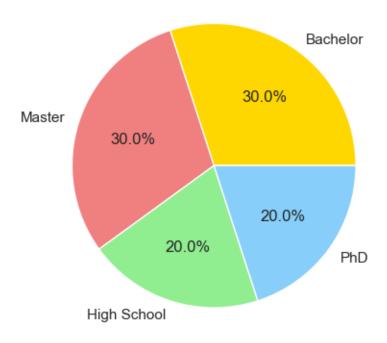




```
[28]: gender_counts = df['Gender'].value_counts()
  gender_counts.plot(kind='bar', color='skyblue')
  plt.title('Gender Distribution')
  plt.xlabel('Gender')
  plt.ylabel('Count')
  plt.show()
```



Education Distribution

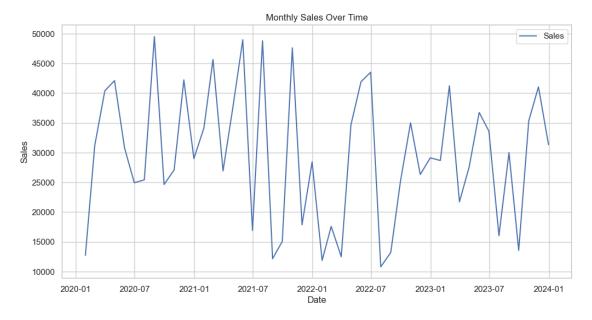


5 Exp 5: Perform Time Series Analysis and apply the various Visualization techniques

```
[30]: import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
      import statsmodels.api as sm
[31]: np.random.seed(0)
      start_date = '2020-01-01'
      end_date = '2023-12-31'
      date_range = pd.date_range(start=start_date, end=end_date, freq='M')
      sales_data = np.random.randint(10000, 50000, size=len(date_range))
      df = pd.DataFrame({'Date': date_range, 'Sales': sales_data})
      df.set_index('Date', inplace=True)
      print(df.describe())
      plt.figure(figsize=(12, 6))
      plt.title('Monthly Sales Over Time')
      plt.xlabel('Date')
      plt.ylabel('Sales')
```

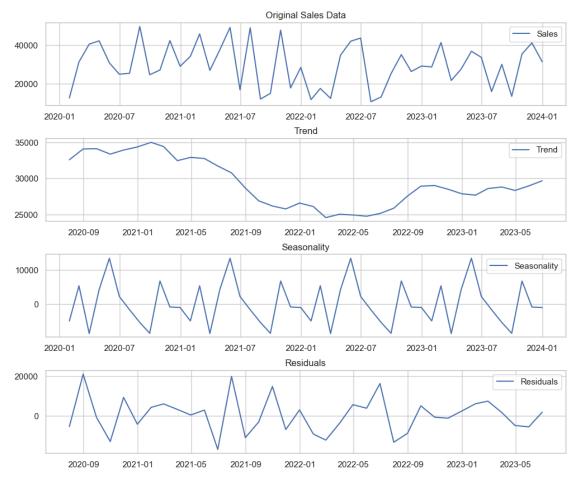
```
plt.plot(df.index, df['Sales'], label='Sales')
plt.legend()
plt.grid(True)
plt.show()
```

```
Sales
          48.000000
count
       29559.562500
mean
std
       11393.671539
       10797.000000
min
25%
       20761.500000
50%
       29056.000000
75%
       38202.500000
       49512.000000
max
```



```
[32]: from statsmodels.tsa.seasonal import seasonal_decompose
  decomposition = seasonal_decompose(df['Sales'], model='additive',period=12)
  trend = decomposition.trend
  seasonal = decomposition.seasonal
  residual = decomposition.resid
  plt.figure(figsize=(12, 10))
  plt.subplots_adjust(hspace=0.4)
  plt.subplot(4, 1, 1)
  plt.title('Original Sales Data')
  plt.plot(df.index, df['Sales'], label='Sales')
  plt.legend()
  plt.subplot(4, 1, 2)
```

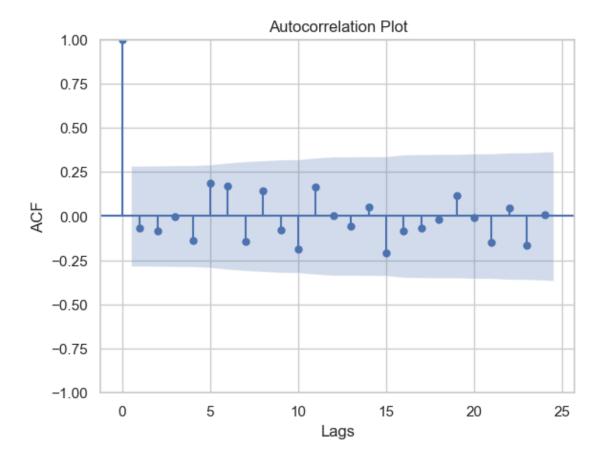
```
plt.title('Trend')
plt.plot(trend, label='Trend')
plt.legend()
plt.subplot(4, 1, 3)
plt.title('Seasonality')
plt.plot(seasonal, label='Seasonality')
plt.legend()
plt.subplot(4, 1, 4)
plt.title('Residuals')
plt.plot(residual, label='Residuals')
plt.legend()
plt.show()
```



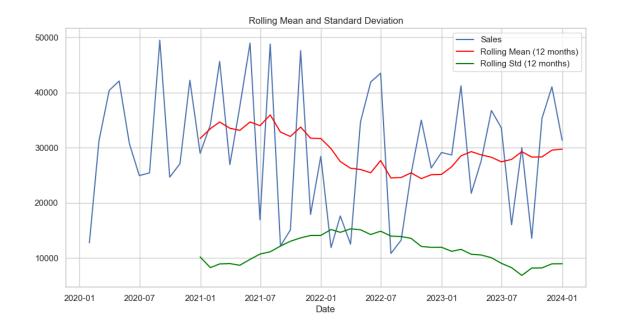
```
[33]: from statsmodels.graphics.tsaplots import plot_acf
plt.figure(figsize=(12, 4))
plot_acf(df['Sales'], lags=24)
plt.title('Autocorrelation Plot')
plt.xlabel('Lags')
```

```
plt.ylabel('ACF')
plt.grid(True)
plt.show()
```

<Figure size 1200x400 with 0 Axes>



```
[34]: rolling_mean = df['Sales'].rolling(window=12).mean()
    rolling_std = df['Sales'].rolling(window=12).std()
    plt.figure(figsize=(12, 6))
    plt.title('Rolling Mean and Standard Deviation')
    plt.xlabel('Date')
    plt.plot(df.index, df['Sales'], label='Sales')
    plt.plot(rolling_mean, label='Rolling Mean (12 months)', color='red')
    plt.plot(rolling_std, label='Rolling Std (12 months)', color='green')
    plt.legend()
    plt.grid(True)
    plt.show()
```



6 Exp 6: Perform Data Analysis and Representation on a map using various map datasets with mouse rollover effect, user interaction

```
[35]: import pandas as pd
      import numpy as np
[36]: # Generate random data
      np.random.seed(42)
      num_points = 100
      latitude = np.random.uniform(37.5, 38.5, num_points)
      longitude = np.random.uniform(-123, -121, num_points)
      value = np.random.randint(1, 100, num_points)
[37]: # Create a DataFrame
      df = pd.DataFrame({'Latitude': latitude, 'Longitude': longitude, 'Value':
       ⇒value})
       # Save the DataFrame to a CSV file
      df.to_csv('map_data.csv', index=False)
[38]: pip install folium
     Requirement already satisfied: folium in c:\users\jawahar a
     s\anaconda3\lib\site-packages (0.18.0)
     Requirement already satisfied: jinja2>=2.9 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from folium) (3.1.4)
```

```
s\anaconda3\lib\site-packages (from folium) (2.32.3)
     Requirement already satisfied: branca>=0.6.0 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from folium) (0.8.0)
     Requirement already satisfied: numpy in c:\users\jawahar a s\anaconda3\lib\site-
     packages (from folium) (1.26.4)
     Requirement already satisfied: xyzservices in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from folium) (2024.9.0)
     Requirement already satisfied: MarkupSafe>=2.0 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from jinja2>=2.9->folium) (2.0.1)
     Requirement already satisfied: certifi>=2017.4.17 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from requests->folium) (2022.9.14)
     Requirement already satisfied: idna<4,>=2.5 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from requests->folium) (2.10)
     Requirement already satisfied: charset-normalizer<4,>=2 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from requests->folium) (2.0.4)
     Requirement already satisfied: urllib3<3,>=1.21.1 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from requests->folium) (1.26.11)
     Note: you may need to restart the kernel to use updated packages.
[39]: import folium
[40]: # Load the generated CSV file
      df = pd.read_csv('map_data.csv')
[41]: #Create a base map
      base_map = folium.Map(location=[df['Latitude'].mean(), df['Longitude'].mean()],
       ⇒zoom start=10)
[42]: # Add data points to the map
      for index, row in df.iterrows():
          popup text = f"Value: {row['Value']}"
          folium.Marker([row['Latitude'], row['Longitude']], popup=popup_text).
       →add_to(base_map)
      # Save the map as an HTML file
      base_map.save('interactive_map.html')
[43]: # Import necessary libraries
      import pandas as pd
      import folium
      # Load the data
      df = pd.read csv('map data.csv')
      #Display basic statistics
      print("Basic Statistics:")
      print(df.describe())
      # Create a base map
```

Requirement already satisfied: requests in c:\users\jawahar a

Basic Statistics:

```
Longitude
                                    Value
        Latitude
count 100.000000 100.000000 100.000000
        37.970181 -122.004337
                                50.670000
mean
        0.297489
                     0.586223
                                28.640247
std
        37.505522 -122.986096
min
                               2.000000
25%
       37.693201 -122.515991
                                27.000000
50%
       37.964142 -121.988750 47.500000
75%
        38.230203 -121.467633
                               72.750000
       38.486887 -121.028699
                               99.000000
max
```

[43]: <folium.folium.Map at 0x1c22b7588e0>

7 Exp 7: Build Cartographic Visualization for Multiple Datasets involving various countries of the world, states and Districts in India

```
[44]: pip install geopandas
```

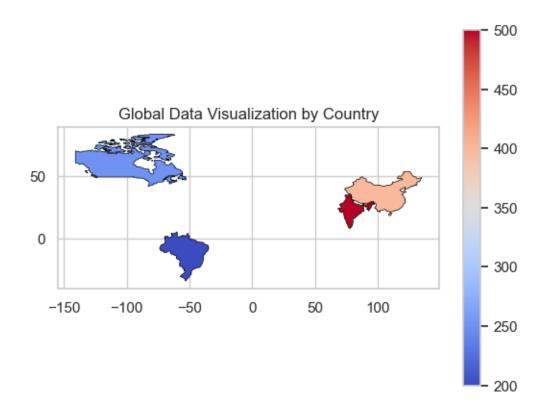
```
Requirement already satisfied: geopandas in c:\users\jawahar a
s\anaconda3\lib\site-packages (1.0.1)
Requirement already satisfied: packaging in c:\users\jawahar a
s\anaconda3\lib\site-packages (from geopandas) (21.3)
Requirement already satisfied: pandas>=1.4.0 in c:\users\jawahar a
s\anaconda3\lib\site-packages (from geopandas) (1.4.4)
Requirement already satisfied: numpy>=1.22 in c:\users\jawahar a
s\anaconda3\lib\site-packages (from geopandas) (1.26.4)
Requirement already satisfied: pyogrio>=0.7.2 in c:\users\jawahar a
s\anaconda3\lib\site-packages (from geopandas) (0.10.0)
Requirement already satisfied: shapely>=2.0.0 in c:\users\jawahar a
s\anaconda3\lib\site-packages (from geopandas) (2.0.6)
Requirement already satisfied: pyproj>=3.3.0 in c:\users\jawahar a
s\anaconda3\lib\site-packages (from geopandas) (3.6.1)
Requirement already satisfied: pytz>=2020.1 in c:\users\jawahar a
s\anaconda3\lib\site-packages (from pandas>=1.4.0->geopandas) (2022.1)
Requirement already satisfied: python-dateutil>=2.8.1 in c:\users\jawahar a
```

```
s\anaconda3\lib\site-packages (from pandas>=1.4.0->geopandas) (2.8.2)
     Requirement already satisfied: certifi in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from pyogrio>=0.7.2->geopandas) (2022.9.14)
     Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from packaging->geopandas) (3.0.9)
     Requirement already satisfied: six>=1.5 in c:\users\jawahar a
     s\anaconda3\lib\site-packages (from python-
     dateutil>=2.8.1->pandas>=1.4.0->geopandas) (1.16.0)
     Note: you may need to restart the kernel to use updated packages.
[45]: import pandas as pd
     import geopandas as gpd
     import matplotlib.pyplot as plt
[46]: world_data = pd.DataFrame({'Country': ['USA', 'Canada', 'India', _
      'Value': [100, 150, 200, 80, 120]})
     india_states_data = pd.DataFrame({'State': ['Maharashtra', 'Karnataka', 'Tamil_
      →Nadu', 'Uttar Pradesh', 'Gujarat'],
                                       'Value': [50, 75, 60, 40, 30]})
     india_districts_data = pd.DataFrame({'District': ['Mumbai', 'Bengaluru', _
      'Value': [20, 30, 25, 15, 10]})
[48]: import pandas as pd
     import geopandas as gpd
     import matplotlib.pyplot as plt
     # Sample data for visualization
     world_data = pd.DataFrame({
          'Country': ['United States', 'Canada', 'India', 'Brazil', 'China'],
          'Value': [300, 250, 500, 200, 400]
     })
     india_states_data = pd.DataFrame({
         'State': ['Maharashtra', 'Karnataka', 'Tamil Nadu', 'Uttar Pradesh',
      'Value': [100, 80, 90, 50, 70]
     })
     india_districts_data = pd.DataFrame({
          'District': ['Mumbai', 'Bengaluru', 'Chennai', 'Lucknow', 'Ahmedabad'],
          'Value': [50, 40, 45, 30, 35]
     })
     # Paths to the local JSON files
```

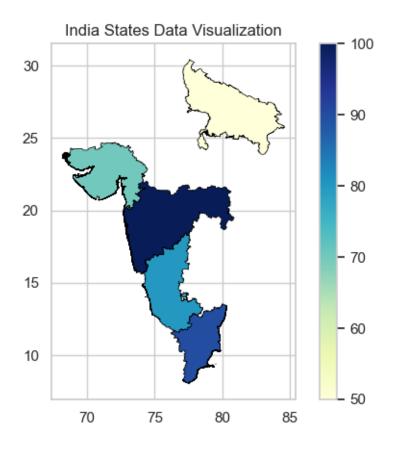
```
world_map_path = 'countries.geo.json'
                                                    # Replace with the actual file
       ⇒path for world map
      india_states_map_path = 'india_state.geojson' # Replace with the actual file_
      ⇔path for India states
      india_districts_map_path = 'india_district.geojson' # Replace with the actual_
      ⇔file path for India districts
      # Load GeoJSON maps
      world_map = gpd.read_file(world_map_path)
      india_states_map = gpd.read_file(india_states_map_path)
      india_districts_map = gpd.read_file(india_districts_map_path)
      # Merge data with maps
      world_data_geo = world_map.merge(world_data, how='left', left_on='name',_

¬right_on='Country')
      india_states_data_geo = india_states_map.merge(india_states_data, how='left',__
      ⇔left_on='NAME_1', right_on='State')
      india_districts_data_geo = india_districts_map.merge(india_districts_data,__
       ⇔how='left', left_on='NAME_2', right_on='District')
[49]: # Plot World Map Visualization
      plt.figure(figsize=(12, 8))
      world_data_geo.plot(column='Value', cmap='coolwarm', legend=True, __
      ⇔edgecolor='black', linewidth=0.5)
      plt.title('Global Data Visualization by Country')
      plt.show()
```

<Figure size 1200x800 with 0 Axes>



<Figure size 1000x800 with 0 Axes>



```
[51]: # Plot India Districts Map Visualization

plt.figure(figsize=(10, 8))

india_districts_data_geo.plot(column='Value', cmap='YlOrRd', legend=True,

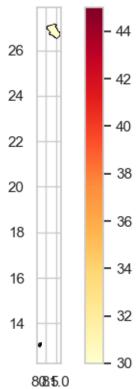
dedgecolor='black', linewidth=0.5)

plt.title('India Districts Data Visualization')

plt.show()
```

<Figure size 1000x800 with 0 Axes>





8 Exp 8: Perform EDA on Wine quality dataset

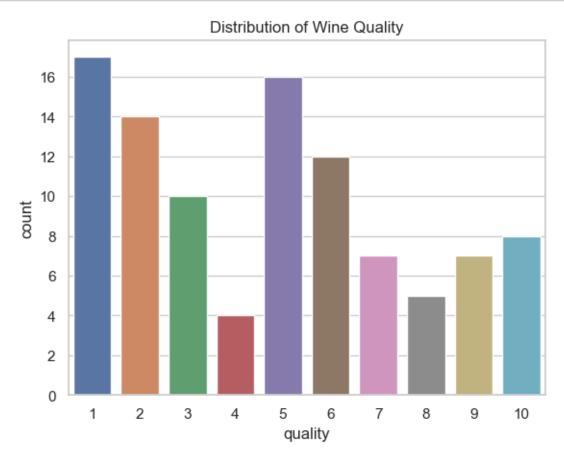
```
[53]: import pandas as pd
  import numpy as np

[54]: np.random.seed(42)
  data = {
    'fixed acidity': np.random.uniform(5, 15, 100),
    'volatile acidity': np.random.uniform(0.1, 1.5, 100),
    'citric acid': np.random.uniform(0, 1, 100),
    'residual sugar': np.random.uniform(0, 10, 100),
    'alcohol': np.random.uniform(8, 15, 100),
    'quality': np.random.randint(1, 11, 100)
  }

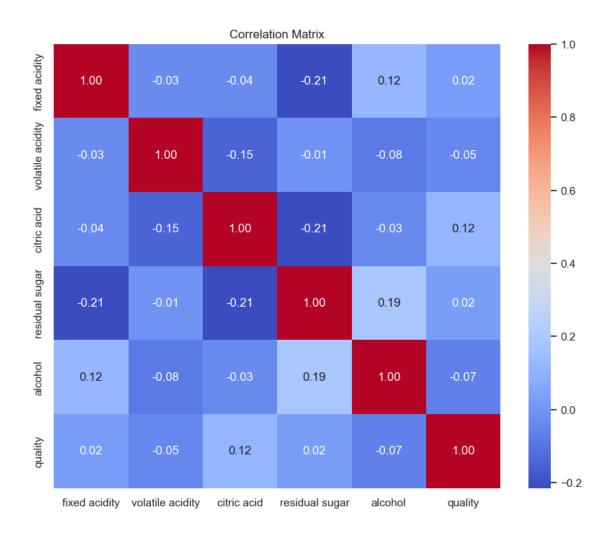
[55]: wine_df = pd.DataFrame(data)
    wine_df.to_csv('wine_quality_dataset.csv', index=False)
[56]: import matplotlib.pyplot as plt
  import seaborn as sns
```

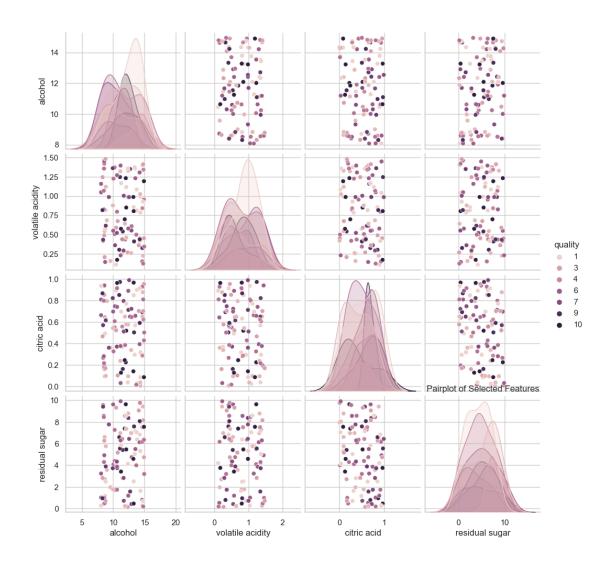
```
[57]: # Load the dataset
      wine_df = pd.read_csv('wine_quality_dataset.csv')
      print(wine_df.info())
      print(wine_df.describe())
      print(wine_df.isnull().sum())
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 100 entries, 0 to 99
     Data columns (total 6 columns):
          Column
                             Non-Null Count
                                             Dtype
          -----
                             _____
      0
          fixed acidity
                             100 non-null
                                             float64
      1
          volatile acidity 100 non-null
                                             float64
      2
          citric acid
                             100 non-null
                                             float64
      3
          residual sugar
                             100 non-null
                                             float64
      4
          alcohol
                             100 non-null
                                             float64
          quality
                             100 non-null
                                              int64
     dtypes: float64(5), int64(1)
     memory usage: 4.8 KB
     None
            fixed acidity volatile acidity citric acid residual sugar
               100.000000
                                  100.000000
                                                100.000000
                                                                100.000000
     count
                 9.701807
                                    0.796964
                                                  0.517601
                                                                  4.911489
     mean
     std
                 2.974894
                                    0.410356
                                                  0.293426
                                                                  2.934522
     min
                 5.055221
                                    0.109733
                                                 0.005062
                                                                  0.143935
     25%
                 6.932008
                                    0.438806
                                                 0.276880
                                                                  2.496149
     50%
                 9.641425
                                    0.807875
                                                 0.562555
                                                                  5.097183
                                                 0.752367
     75%
                12.302031
                                                                  7.357775
                                    1.172657
     max
                14.868869
                                    1.479911
                                                 0.990054
                                                                  9.905051
               alcohol
                            quality
                        100.000000
     count
            100.000000
     mean
             11.612321
                           4.750000
     std
              2.230206
                           2.900279
              8.075864
                           1.000000
     min
     25%
              9.847544
                           2.000000
     50%
             11.677796
                           5.000000
     75%
                           7.000000
             13.578189
             14.950754
                          10.000000
     max
     fixed acidity
                          0
     volatile acidity
                          0
     citric acid
                          0
     residual sugar
                          0
     alcohol
                          0
     quality
                          0
     dtype: int64
```

```
[58]: sns.countplot(x='quality', data=wine_df)
plt.title('Distribution of Wine Quality')
plt.show()
```

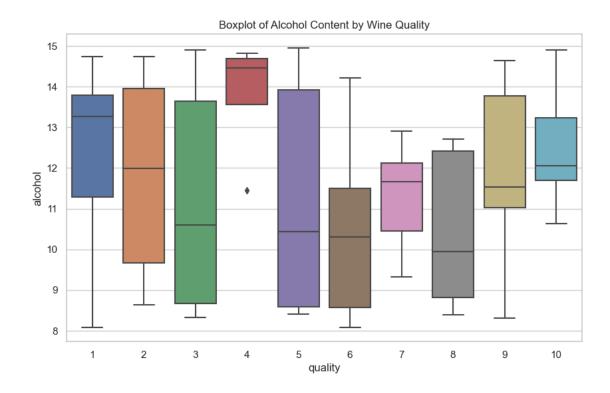


```
[59]: correlation_matrix = wine_df.corr()
   plt.figure(figsize=(10, 8))
   sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
   plt.title('Correlation Matrix')
   plt.show()
```





```
[61]: plt.figure(figsize=(10, 6))
    sns.boxplot(x='quality', y='alcohol', data=wine_df)
    plt.title('Boxplot of Alcohol Content by Wine Quality')
    plt.show()
```



9 Exp 9: Use a Case Study on a Dataset and apply the various EDA and Visualization techniqes and present an analysis report

```
[62]: import pandas as pd
      import numpy as np
      import seaborn as sns
      import matplotlib.pyplot as plt
[63]: np.random.seed(42)
      employee_id = range(1, 101)
      age = np.random.randint(22, 60, size=100)
      gender = np.random.choice(['Male', 'Female'], size=100)
      department = np.random.choice(['HR', 'Sales', 'IT', 'Marketing'], size=100)
      years_of_experience = np.random.randint(1, 15, size=100)
      performance_rating = np.random.randint(1, 6, size=100)
      salary = np.random.randint(40000, 120000, size=100)
      employee_data = pd.DataFrame({'EmployeeID': employee_id,'Age':age,'Gender':u
       ⇒gender, 'Department':department,'YearsOfExperience':⊔
       ⇔years_of_experience, 'PerformanceRating': performance_rating, 'Salary':
       ⇒salary})
      employee_data.to_csv('employee_data.csv', index=False)
      employee_data = pd.read_csv('employee_data.csv')
```

```
[64]: print(employee_data.info())
    print('\n')
    print(employee_data.describe())
    print('\n')
    print(employee_data.isnull().sum())
    print('\n')
    department_distribution = employee_data['Department'].value_counts()
    print(department_distribution)
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	EmployeeID	100 non-null	int64
1	Age	100 non-null	int64
2	Gender	100 non-null	object
3	Department	100 non-null	object
4	YearsOfExperience	100 non-null	int64
5	PerformanceRating	100 non-null	int64
6	Salary	100 non-null	int64

dtypes: int64(5), object(2)

memory usage: 5.6+ KB

None

	EmployeeID	Age	YearsOfExperience	PerformanceRating	١
aat	100.000000	100.000000	100.000000	100.00000	`
count	100.000000	100.000000	100.000000	100.00000	
mean	50.500000	40.060000	6.970000	2.96000	
std	29.011492	10.688255	4.381907	1.44194	
min	1.000000	22.000000	1.000000	1.00000	
25%	25.750000	30.000000	3.000000	2.00000	
50%	50.500000	41.500000	6.000000	3.00000	
75%	75.250000	48.000000	11.000000	4.00000	
max	100.000000	59.000000	14.000000	5.00000	

Salary 100.000000 count 81430.860000 mean 22497.541898 std min 40412.000000 25% 62624.750000 50% 80856.500000 75% 100741.500000 119605.000000 max

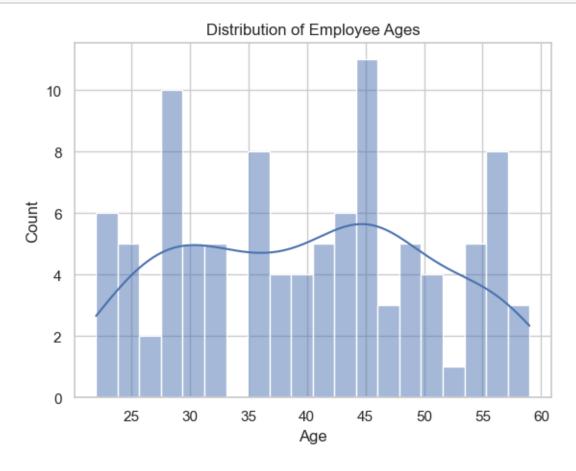
```
EmployeeID 0
Age 0
Gender 0
Department 0
YearsOfExperience 0
PerformanceRating 0
Salary 0
```

dtype: int64

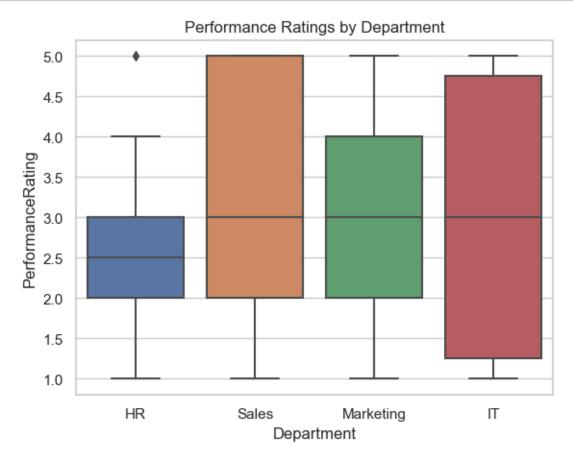
Marketing 31 IT 26 HR 22 Sales 21

Name: Department, dtype: int64

```
[65]: import matplotlib.pyplot as plt
import seaborn as sns
sns.histplot(employee_data['Age'], bins=20, kde=True)
plt.title('Distribution of Employee Ages')
plt.show()
```



```
[66]: sns.boxplot(x='Department', y='PerformanceRating', data=employee_data)
plt.title('Performance Ratings by Department')
plt.show()
```



[]: