Data Visualization with Python

Task	Syntax	Description	Example
Load CSV data	<pre>pd.read_csv('fil ename.csv')</pre>	Read data from a CSV file into a Pandas DataFrame	<pre>df_can=pd.read_c sv('data.csv')</pre>
Handling Missing Values	df.dropna()	Drop rows with missing values	df_can.dropna()
	<pre>df.fillna(value)</pre>	Fill missing values with a specified value	<pre>df_can.fillna(0)</pre>
Removing Duplicates	<pre>df.drop_duplicat es() df.rename(column</pre>	Remove duplicate rows Rename one or more	<pre>df_can.drop_dupl icates() df_can.rename(co</pre>
Renaming Columns	<pre>s={'old_name': 'new_name'})</pre>	columns	<pre>lumns={'Age': 'Years'})</pre>
Selecting Columns	<pre>df['column_name'] or df.column name</pre>	Select a single column	<pre>df_can.Age or df_can['Age]'</pre>
	df[['col1', 'col2']]	Select multiple columns	<pre>df_can[['Name', 'Age']]</pre>
Filtering Rows	<pre>df[df['column'] > value]</pre>	Filter rows based on a condition	<pre>df_can[df_can['A ge'] > 30]</pre>
Applying Functions to Columns	<pre>df['column'].app ly(function_name)</pre>	Apply a function to transform values in a column	<pre>df_can['Age'].ap ply(lambda x: x + 1) df_can['Total']</pre>
Creating New Columns	<pre>df['new_column'] = expression</pre>	Create a new column with values derived from existing ones	= df_can['Quantity '] *
Grouping and Aggregating	<pre>df.groupby('colu mn').agg({'col1' : 'sum', 'col2': 'mean'})</pre>	Group rows by a column and apply aggregate functions	<pre>df_can['Price'] df_can.groupby(' Category').agg({ 'Total': 'mean'})</pre>
Sorting Rows	<pre>df.sort_values(' column', ascending=True/F alse)</pre>	Sort rows based on a column	<pre>df_can.sort_valu es('Date', ascending=True)</pre>
Displaying First n Rows	df.head(n)	Show the first n rows of the DataFrame	df_can.head(3)
Displaying Last n Rows	df.tail(n)	Show the last n rows of the DataFrame	<pre>df_can.tail(3)</pre>
Checking for Null Values	<pre>df.isnull()</pre>	Check for null values in the DataFrame	<pre>df_can.isnull()</pre>
Selecting Rows by Index	<pre>df.iloc[index]</pre>	Select rows based on integer index	<pre>df_can.iloc[3]</pre>
	<pre>df.iloc[start:en d]</pre>	Select rows in a specified range	df_can.iloc[2:5]

Selecting Rows by Label	<pre>df.loc[label]</pre>	Select rows based on label/index name	<pre>df_can.loc['Labe 1']</pre>
Label	<pre>df.loc[start:end]</pre>	Select rows in a specified label/index range	<pre>df_can.loc['Age' :'Quantity']</pre>
Summary Statistics	df.describe()	Generates descriptive statistics for numerical columns	<pre>df_can.describe()</pre>

Cheat Sheet: Plot Libraries

Library	Main Purpose	Key Features	Programm ing Language	Level of Customiza tion	Dashboard Capabilitie s	Types of Plots Possible Line plots,
Matplotlib	General- purpose plotting	Comprehen sive plot types and variety of customizati on options	Python	High	Requires additional components and customizati on	scatter plots, bar charts, histograms, pie charts, box plots, heatmaps, etc.
Pandas	Fundament ally used for data manipulatio n but also has plotting functionalit y	Easy to plot directly on Panda data structures	Python	Medium	Can be combined with web frameworks for creating dashboards	Line plots, scatter plots, bar charts, histograms, pie charts, box plots, etc.
Seaborn	Statistical data visualizatio n	Stylish, specialized statistical plot types	Python	Medium	Can be combined with other libraries to display plots on dashboards	Heatmaps, violin plots, scatter plots, bar plots, count plots, etc.
Plotly	Interactive data visualizatio n	interactive web-based visualizatio ns	Python, R, JavaScript	High	Dash framework is dedicated for building interactive dashboards	Line plots, scatter plots, bar charts, pie charts, 3D plots, choropleth maps, etc.
Folium	Geospatial data visualizatio n	Interactive, customizabl e maps	Python	Medium	For incorporati ng maps into dashboards, it can be	Choropleth maps, point maps, heatmaps, etc.

integrated with other frameworks /libraries

PyWaffle Wa

Plotting Waffle charts

Waffle charts

Python

Low