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# -----
# PROJECT : CUSTOMER SHOPPING BEHAVIOR ANALYSIS USING PANDAS
# -----

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

# STEP 1: CREATE DATASET
data = {
    "Customer_ID": range(1, 101),
    "Customer_Name": ["Cust_" + str(i) for i in range(1, 101)],
    "Age": np.random.randint(18, 60, 100),
    "Gender": np.random.choice(["Male", "Female"], 100),
    "Monthly_Income": np.random.randint(15000, 90000, 100),
    "Monthly_Spending": np.random.randint(5000, 60000, 100),
    "Shopping_Frequency": np.random.randint(1, 20, 100)
}

df = pd.DataFrame(data)

# DISPLAY FULL DATASET
print("FULL DATASET:")
print(df)

# -----
# STEP 2: BASIC INSPECTION
# -----

print("\nHEAD:")
print(df.head())

print("\nTAIL:")
print(df.tail())

print("\nINFO:")
print(df.info())

print("\nDESCRIBE:")
print(df.describe())

# -----
# STEP 3: SHAPE, COLUMNS, DTYPES
# -----

print("\nSHAPE:")
print(df.shape)

print("\nCOLUMNS:")
print(df.columns)

print("\nDTYPES:")
print(df.dtypes)

# -----
# STEP 4: DROP, RENAME, SORT, FILLNA, UNIQUE
# -----

print("\nAFTER DROPPING Customer_ID:")
df_dropped = df.drop(columns=["Customer_ID"])
print(df_dropped)

print("\nAFTER RENAMING Customer_Name TO Full_Name:")
df_renamed = df.rename(columns={"Customer_Name": "Full_Name"})
print(df_renamed)

print("\nSORTED BY Monthly_Spending:")
df_sorted = df.sort_values(by="Monthly_Spending")
print(df_sorted)

print("\nFILL NA WITH 0:")
df_filled = df.fillna(0)
print(df_filled)

print("\nUNIQUE ROWS (DROP DUPLICATES):")
df_unique = df.drop_duplicates()
print(df_unique)

# -----
# STEP 5: REPLACE & STRING OPERATIONS
# -----

print("\nREPLACED Male->M, Female->F:")
df_replaced = df.replace({"Male": "M", "Female": "F"})
print(df_replaced)

print("\nADDING COLUMN (Name Contains 'u'):")
df["Contains_u"] = df["Customer_Name"].str.contains("u")
print(df)

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# -----
# STEP 6: GROUPBY OPERATIONS (MEAN, MAX, MIN, COUNT)
# -----

print("\nMEAN INCOME BY GENDER:")
print(df.groupby("Gender")["Monthly_Income"].mean())

print("\nMAX VALUES GROUPED BY AGE:")
print(df.groupby("Age").max())

print("\nMIN VALUES GROUPED BY AGE:")
print(df.groupby("Age").min())

print("\nCOUNTOF GENDERS:")
print(df["Gender"].value_counts())

# -----
# STEP 7: DATA CLEANING
# -----

print("\nDRIPPING NA VALUES:")
df_cleaned = df.dropna()
print(df_cleaned)

print("\nFILL NA WITH 0 AGAIN:")
df_filled_again = df.fillna(0)
print(df_filled_again)

print("\nREPLACE NaN STRING WITH 0:")
df_replaced_nan = df.replace({"NaN": 0})
print(df_replaced_nan)

# -----
# STEP 8: VISUALIZATIONS
# -----

plt.figure(figsize=(15,4))
plt.plot(df["Customer_ID"], df["Monthly_Spending"], marker="o")
plt.title("Monthly Spending Line Plot")
plt.xlabel("Customer ID")
plt.ylabel("Monthly Spending")
plt.show()

df.plot(x="Age", y="Monthly_Income", kind="bar", figsize=(20,5))
plt.title("Age vs Monthly Income (Bar Plot)")
plt.show()

df["Monthly_Spending"].plot(kind="hist", edgecolor="black")
plt.title("Monthly Spending Histogram")
plt.show()

df.plot(x="Monthly_Income", y="Monthly_Spending", kind="scatter")
plt.title("Income vs Spending (Scatter Plot)")
plt.show()

```

... FULL DATASET:

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

	Shopping_Frequency
0	4
1	18
2	10
3	14
4	19
..	...
95	4
96	11
97	7
98	8
99	3

[100 rows x 7 columns]

HEAD:

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505

	Shopping_Frequency
0	4

```

Shopping_Frequency
0      4
...    18
2      10
3      14
4      19

TAIL:
  Customer_ID Customer_Name Age Gender Monthly_Income Monthly_Spending \
95          96      Cust_96  48  Female         37034          31002
96          97      Cust_97  28   Male         27696          14714
97          98      Cust_98  37  Female         84130          39055
98          99      Cust_99  24   Male         87063          29934
99         100      Cust_100  53   Male         73577          35450

Shopping_Frequency
95      4
96     11
97      7
98      8
99      3

INFO:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 7 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Customer_ID            100 non-null   int64
1   Customer_Name          100 non-null   object
2   Age                    100 non-null   int64
3   Gender                 100 non-null   object
4   Monthly_Income         100 non-null   int64
5   Monthly_Spending       100 non-null   int64
6   Shopping_Frequency     100 non-null   int64
dtypes: int64(5), object(2)
memory usage: 5.6+ KB
None

DESCRIBE:
  Customer_ID      Age  Monthly_Income  Monthly_Spending \
count  100.000000  100.000000      100.000000      100.000000
mean    50.500000   37.650000    55238.710000    33064.440000
std     29.011492   12.220512    24212.266542    15649.490557
min      1.000000   18.000000    15523.000000     5101.000000
25%     25.750000   27.000000    31330.750000    19971.750000
50%     50.500000   36.500000    55126.500000    32730.000000
75%     75.250000   46.250000    80566.250000    47375.750000
max     100.000000   59.000000    89923.000000    59619.000000

Shopping_Frequency
count    100.000000
mean      10.560000
std        5.693022
min         1.000000
25%         6.000000
50%        11.000000
75%        16.000000
max        19.000000

SHAPE:
(100, 7)

COLUMNS:
Index(['Customer_ID', 'Customer_Name', 'Age', 'Gender', 'Monthly_Income',
      'Monthly_Spending', 'Shopping_Frequency'],
      dtype='object')

DTYPES:
Customer_ID      int64
Customer_Name    object
Age              int64
Gender           object
Monthly_Income   int64
Monthly_Spending int64
Shopping_Frequency int64
dtype: object

AFTER DROPPING Customer_ID:
  Customer_Name Age Gender Monthly_Income Monthly_Spending \
0      Cust_1  31  Female      86306          31313
1      Cust_2  37   Male      28353           5955
2      Cust_3  38   Male      51011          53346
3      Cust_4  45  Female      80228          25510
4      Cust_5  24   Male      46576          32505
..      ...   ...   ...         ...         ...
95     Cust_96  48  Female      37034          31002
96     Cust_97  28   Male      27696          14714
97     Cust_98  37  Female      84130          39055
98     Cust_99  24   Male      87063          29934
99     Cust_100 53   Male      73577          35450

Shopping_Frequency
0      4
1     18
2     10
3     14
4     19
..     ...
95      4

```

95 4  
96 11  
97 7  
98 8  
99 3

[100 rows x 6 columns]

AFTER RENAMING Customer\_Name TO Full\_Name:

	Customer_ID	Full_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

	Shopping_Frequency
0	4
1	18
2	10
3	14
4	19
..	...
95	4
96	11
97	7
98	8
99	3

[100 rows x 7 columns]

SORTED BY Monthly\_Spending:

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
72	73	Cust_73	46	Male	55715	5101
71	72	Cust_72	52	Male	82251	5782
1	2	Cust_2	37	Male	28353	5955
33	34	Cust_34	46	Female	24646	6301
8	9	Cust_9	57	Female	82959	6674
..	...	...	...	...	...	...
87	88	Cust_88	32	Male	27806	58120
85	86	Cust_86	49	Female	82927	58829
94	95	Cust_95	28	Male	80543	59115
19	20	Cust_20	57	Male	44346	59598
29	30	Cust_30	47	Female	16535	59619

	Shopping_Frequency
72	16
71	2
1	18
33	6
8	5
..	...
87	7
85	19
94	16
19	11
29	9

[100 rows x 7 columns]

FILL NA WITH 0:

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

	Shopping_Frequency
0	4
1	18
2	10
3	14
4	19
..	...
95	4
96	11
97	7
98	8
99	3

[100 rows x 7 columns]

UNIQUE ROWS (DROP DUPLICATES):

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510

0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
...	2	3	Cust_3	38	Male	51011
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

Shopping_Frequency	
0	4
1	18
2	10
3	14
4	19
..	...
95	4
96	11
97	7
98	8
99	3

[100 rows x 7 columns]

REPLACED Male->M, Female->F:

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	F	86306	31313
1	2	Cust_2	37	M	28353	5955
2	3	Cust_3	38	M	51011	53346
3	4	Cust_4	45	F	80228	25510
4	5	Cust_5	24	M	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	F	37034	31002
96	97	Cust_97	28	M	27696	14714
97	98	Cust_98	37	F	84130	39055
98	99	Cust_99	24	M	87063	29934
99	100	Cust_100	53	M	73577	35450

Shopping_Frequency	
0	4
1	18
2	10
3	14
4	19
..	...
95	4
96	11
97	7
98	8
99	3

[100 rows x 7 columns]

ADDING COLUMN (Name Contains 'u'):

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

	Shopping_Frequency	Contains_u
0	4	True
1	18	True
2	10	True
3	14	True
4	19	True
..	...	...
95	4	True
96	11	True
97	7	True
98	8	True
99	3	True

[100 rows x 8 columns]

MEAN INCOME BY GENDER:

```
Gender
Female    54900.125000
Male      55464.433333
Name: Monthly_Income, dtype: float64
```

MAX VALUES GROUPED BY AGE:

	Customer_ID	Customer_Name	Gender	Monthly_Income	Monthly_Spending \
Age					
18	92	Cust_92	Male	73426	48866
19	77	Cust_77	Male	84531	48885
20	36	Cust_36	Male	63296	30243
21	42	Cust_42	Male	68366	52756
22	70	Cust_70	Male	62339	56743
23	74	Cust_74	Female	82174	24127

23	74	Cust_74	Female	82174	24127
24	99	Cust_99	Male	87063	35898
...	25	31	Cust_31	Male	89061
26	32	Cust_32	Female	46109	42258
27	87	Cust_87	Male	73655	49248
28	97	Cust_97	Male	80543	59115
30	94	Cust_94	Male	84071	31731
31	78	Cust_78	Male	86306	31313
32	93	Cust_93	Male	49691	58120
34	69	Cust_69	Male	84392	54623
35	91	Cust_91	Male	79271	43774
36	82	Cust_82	Male	85724	38006
37	98	Cust_98	Male	84130	39055
38	3	Cust_3	Male	51011	53346
39	66	Cust_66	Male	44463	50958
42	51	Cust_51	Male	86616	49536
43	76	Cust_76	Male	85741	54823
44	84	Cust_84	Male	89923	54032
45	90	Cust_90	Male	89866	44113
46	73	Cust_73	Male	64598	37502
47	30	Cust_30	Female	16535	59619
48	96	Cust_96	Male	37034	50025
49	86	Cust_86	Female	82927	58829
50	71	Cust_71	Male	72994	39196
51	8	Cust_8	Male	54829	28291
52	72	Cust_72	Male	82251	32955
53	100	Cust_6	Male	86135	53894
54	81	Cust_81	Male	88298	35796
56	60	Cust_60	Male	88402	34887
57	50	Cust_9	Male	82959	59598
58	75	Cust_75	Male	57933	15757
59	67	Cust_67	Male	65768	50526

Shopping\_Frequency Contains\_u

Age		
18	14	True
19	19	True
20	17	True
21	19	True
22	14	True
23	15	True
24	19	True
25	12	True
26	13	True
27	17	True
28	19	True
30	19	True
31	10	True
32	19	True
34	19	True
35	12	True
36	16	True
37	18	True
38	10	True
39	16	True
42	17	True
43	18	True
44	19	True
45	14	True
46	16	True
47	9	True
48	11	True
49	19	True
50	8	True
51	8	True
52	17	True
53	17	True
54	16	True
56	18	True
57	11	True
58	17	True
59	10	True

MIN VALUES GROUPED BY AGE:

Age	Customer_ID	Customer_Name	Gender	Monthly_Income	Monthly_Spending	\
18	7	Cust_13	Female	25219	19917	
19	25	Cust_25	Female	42956	31169	
20	36	Cust_36	Male	63296	30243	
21	40	Cust_40	Female	56604	34612	
22	46	Cust_46	Female	20061	7420	
23	74	Cust_74	Female	82174	24127	
24	5	Cust_23	Female	30490	19290	
25	10	Cust_10	Female	69477	27674	
26	27	Cust_27	Female	29528	31755	
27	14	Cust_14	Female	23878	14417	
28	18	Cust_18	Female	17293	14714	
30	12	Cust_12	Female	38829	12472	
31	1	Cust_1	Female	76134	22670	
32	54	Cust_54	Male	17371	19990	
34	17	Cust_17	Female	47411	13077	
35	16	Cust_16	Female	24514	7054	
36	11	Cust_11	Male	57010	12922	
37	2	Cust_2	Female	28353	5955	
38	3	Cust_3	Male	51011	53346	
39	37	Cust_37	Female	19968	9007	
42	21	Cust_21	Female	34240	12157	
43	15	Cust_15	Female	30372	47135	
44	39	Cust_39	Female	55424	41361	
45	4	Cust_33	Female	20270	15597	

45	4	Cust_33	Female	20270	15597
46	19	Cust_19	Female	24646	5101
...	30	Cust_30	Female	16535	59619
48	89	Cust_89	Female	28778	31002
49	86	Cust_86	Female	82927	58829
50	71	Cust_71	Male	72994	39196
51	8	Cust_8	Male	54829	28291
52	38	Cust_38	Male	29516	5782
53	6	Cust_100	Male	24694	35450
54	43	Cust_43	Female	15523	23920
56	60	Cust_60	Male	88402	34887
57	9	Cust_20	Female	44346	6674
58	29	Cust_29	Female	19555	14204
59	26	Cust_26	Female	16017	25596

	Shopping_Frequency	Contains_u
Age		
18	2	True
19	18	True
20	17	True
21	5	True
22	6	True
23	15	True
24	6	True
25	3	True
26	6	True
27	1	True
28	1	True
30	1	True
31	4	True
32	7	True
34	11	True
35	1	True
36	8	True
37	7	True
38	10	True
39	7	True
42	11	True
43	15	True
44	7	True
45	1	True
46	4	True
47	9	True
48	4	True
49	19	True
50	8	True
51	8	True
52	2	True
53	2	True
54	7	True
56	18	True
57	5	True
58	4	True
59	2	True

COUNT OF GENDERS:  
Gender  
Male 60  
Female 40  
Name: count, dtype: int64

DROPPING NA VALUES:						
	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

	Shopping_Frequency	Contains_u
0	4	True
1	18	True
2	10	True
3	14	True
4	19	True
..	...	...
95	4	True
96	11	True
97	7	True
98	8	True
99	3	True

[100 rows x 8 columns]

FILL NA WITH 0 AGAIN:						
	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending \
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714



```
96      11      True
97      7      True
... 98      8      True
99      3      True
```

[100 rows x 8 columns]

FILL NA WITH 0 AGAIN:

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

	Shopping_Frequency	Contains_u
0	4	True
1	18	True
2	10	True
3	14	True
4	19	True
..	...	...
95	4	True
96	11	True
97	7	True
98	8	True
99	3	True

[100 rows x 8 columns]

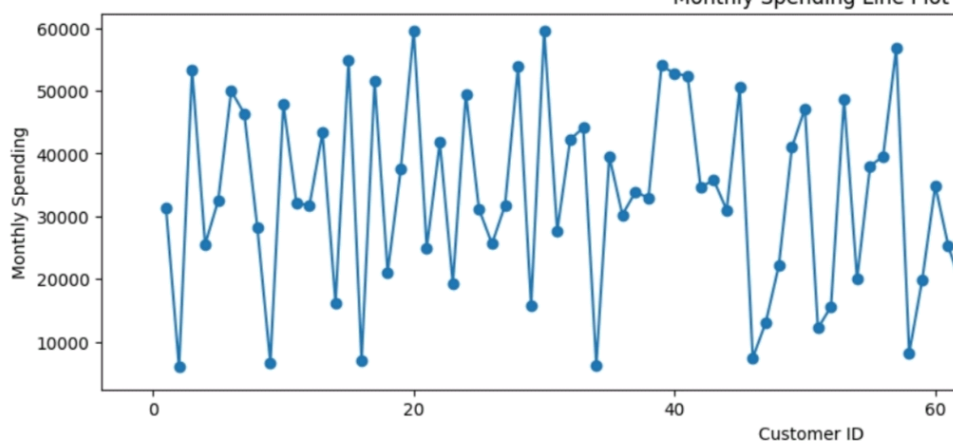
REPLACE NaN STRING WITH 0:

	Customer_ID	Customer_Name	Age	Gender	Monthly_Income	Monthly_Spending
0	1	Cust_1	31	Female	86306	31313
1	2	Cust_2	37	Male	28353	5955
2	3	Cust_3	38	Male	51011	53346
3	4	Cust_4	45	Female	80228	25510
4	5	Cust_5	24	Male	46576	32505
..	...	...	...	...	...	...
95	96	Cust_96	48	Female	37034	31002
96	97	Cust_97	28	Male	27696	14714
97	98	Cust_98	37	Female	84130	39055
98	99	Cust_99	24	Male	87063	29934
99	100	Cust_100	53	Male	73577	35450

	Shopping_Frequency	Contains_u
0	4	True
1	18	True
2	10	True
3	14	True
4	19	True
..	...	...
95	4	True
96	11	True
97	7	True
98	8	True
99	3	True

[100 rows x 8 columns]

Monthly Spending Line Plot



Age vs Mo

