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# PROJECT : CUSTOMER SHOPPING BEHAVIOR ANALYSIS USING PANDAS
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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("\nCOUNT OF GENDERS:")
print(df["Gender"].value_counts())

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


print("\nDROPPING 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

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    Shopping_Frequency
0                  4
...
1                  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     ...     ...
96     Cust_96    48  Female       37034           31002
97     Cust_97    28  Male        27696           14714
98     Cust_98    37  Female       84130           39055
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 3  
1 2 Cust\_2 37 Male 28353 5  
... 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

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

[100 rows x 7 columns]

REPLACED Male->M, Female->F:

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

Shopping\_Frequency

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

[100 rows x 7 columns]

ADDING COLUMN (Name Contains 'u'):

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

Shopping\_Frequency Contains\_u

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

[100 rows x 8 columns]

MEAN INCOME BY GENDER:

Gender

Gender	Mean_Income
Female	54900.125000
Male	55464.433333

Name: Monthly\_Income, dtype: float64

MAX VALUES GROUPED BY AGE:

Age	Customer_ID	Customer_Name	Gender	Monthly_Income	Monthly_Spending
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	47927
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	Shopping_Frequency	Contains_u
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
47	89	Cust_89	Female	28778	31002
48	86	Cust_86	Female	82927	58829
49	71	Cust_71	Male	72994	39196
50	8	Cust_8	Male	54829	28291
51	38	Cust_38	Male	29516	5782
52	6	Cust_100	Male	24694	35450
53	43	Cust_43	Female	15523	23920
54	60	Cust_60	Male	88402	34887
55	9	Cust_20	Female	44346	6674
56	29	Cust_29	Female	19555	14204
57	26	Cust_26	Female	16017	25596

Age	Shopping_Frequency	Contains_u
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

Age	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

ands + Code + Text ▶ Run all ✓ RAM Disk

```

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           ...     ...
97           ...     ...
98           ...     ...
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           ...     ...
96           ...     ...
97           ...     ...
98           ...     ...
99           ...     ...

[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           ...     ...
97           ...     ...
98           ...     ...
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           ...     ...
96           ...     ...
97           ...     ...
98           ...     ...
99           ...     ...

[100 rows x 8 columns]

```

**Monthly Spending Line Plot**

**Age vs Mo**

