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Report pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

# 1. Data Loading and Initial Exploration
df_cities=pd.read_csv("../ICG Input Files/Dataset/vla_cities.csv")
df_respondents=pd.read_csv("../ICG Input Files/Dataset/ida_respondents.csv")
df_fact_survey_responses=pd.read_csv("../ICG Input Files/Dataset/fact_survey_responses.csv")

# 2. Data Cleaning and Preprocessing
df_cities.head(5)

Out[11]:
  City_ID  City  Tier
0  CT111  Delhi  Tier 1
1  CT112  Mumbai  Tier 1
2  CT113  Bangalore  Tier 1
3  CT114  Chennai  Tier 1
4  CT115  Kolkata  Tier 2

In [12]: df_cities.isna().sum()

Out[12]:
City_ID    0
City        0
Tier        0
dtype: int64

In [13]: df_cities.dtypes

Out[13]:
City_ID    object
City        object
Tier        object
dtype: object

In [14]: df_respondents.head(5)

Out[14]:
  Respondent_ID  Name  Age  Gender  City_ID
0              0  Anirudh Isaac  15-18  Female  CT117
1              1  Trisha Rishi  19-30  Male  CT118
2              2  Yuvraj Visk  15-18  Male  CT116
3              3  Pranay Chand  31-45  Female  CT113
4              4  Mohanlal Joshi  19-30  Female  CT120

In [15]: df_respondents.isna().sum()

Out[15]:
Respondent_ID    0
Name              0
Age              0
Gender           0
City_ID          0
dtype: int64

In [16]: df_respondents.dtypes

Out[16]:
Respondent_ID  int64
Name           object
Age            object
Gender         object
City_ID        object
dtype: object

In [22]: df_respondents["Respondent_ID"] = df_respondents["Respondent_ID"].astype(str)

In [23]: df_respondents.dtypes

Out[23]:
Respondent_ID  object
Name           object
Age            object
Gender         object
City_ID        object
dtype: object

In [24]: df_fact_survey_responses.head(5)
df_fact_survey_responses.columns

Out[24]:
Index(['Respondent_ID', 'Respondent_ID', 'Consume_Frequency', 'Consume_Time',
      'Consume_Reason', 'Heard_Before', 'Brand_Perception', 'General_Perception',
      'Tried_Before', 'Taste_Experience', 'Marketing_Channels', 'Packaging_Preference',
      'Purchase_Location', 'Typical_Consumption_Situations'],
      dtype='object')

In [26]: df_fact_survey_responses.dtypes

Out[26]:
Respondent_ID    int64
Respondent_ID    int64
Consume_Frequency  object
Consume_Time      object
Consume_Reason    object
Consume_Reason    object
Brand_Perception  object
General_Perception  object
Tried_Before      object
Taste_Experience  int64
Reasons_for_preventing_trying  object
Current_Brands    object
Reasons_for_choosing_brands  object
Improvements_desired  object
Ingredients_expected  object
Health_concerns    object
Interest_in_natural_or_organic  object
Marketing_channels  object
Packaging_preference  object
Listed_edition_packaging  object
Price_range        object
Purchase_location  object
Typical_Consumption_Situations  object
dtype: object

In [27]: df_fact_survey_responses["Respondent_ID"] = df_fact_survey_responses["Respondent_ID"].astype(str)

# 3. Demographic Insights
a. Who prefers energy drink more? (male/female/non-binary)

In [29]: df_age_gender = df_fact_survey_responses[df_fact_survey_responses["Respondent_ID"]

Out[29]:
  Respondent_ID  Respondent_ID  Consume_Frequency  Consume_Time  Consume_Reason  Heard_Before  Brand_Perception  General_Perception  Tried_Before  Taste_Experience  ...  Marketing_Channels  Packaging_Preference
0  103001         120031         2-3 times a week  Throughout the day  Increased energy and focus  Yes  Neutral  Not sure  No  5  ...  TV commercials  Compact and portable cans
1  103002         120032         2-3 times a month  Throughout the day  To boost performance  No  Neutral  Not sure  No  5  ...  Print media  Compact and portable cans
2  103003         120033         Randomly  Before exercise  Increased energy and focus  No  Neutral  Not sure  No  2  ...  Online ads  Innovative bottle design
3  103004         120034         2-3 times a week  To stay awake during work/study  To boost performance  No  Positive  Dangerous  Yes  5  ...  Online ads  Compact and portable cans
4  103005         120035         Daily  To stay awake during work/study  Increased energy and focus  Yes  Neutral  Effective  Yes  5  ...  Online ads  Compact and portable cans

5 rows x 27 columns

In [30]: df_age_gender.groupby("Gender").size().plot(kind="pie", autopct="%1.1f%%")
plt.title("Gender Distribution")

Out[30]: Text(0.5, 1.0, 'Gender Distribution')

Gender Distribution
Female
34.5%
Non-binary
5.1%
Male
60.4%

Answer: Male prefers energy drink more followed by Female and Non-binary

b) Which age group prefers energy drinks more?

In [31]: df_age_group = df_fact_survey_responses.groupby("Age").size().sort_values(ascending=False).plot(kind="bar", color="green")
plt.title("Energy Drink Preference based on Age Group")

Out[31]: Text(0.5, 1.0, 'Energy Drink Preference based on Age Group')

Energy Drink Preference based on Age Group
Age
65+
46-65
15-18
31-45
19-30
0 1000 2000 3000 4000 5000

Answer: Age Group 19-30 prefer energy drink more followed by 31-45, 15-18, 46-65, 65+

c) Which type of marketing reaches the most Youth (15-30)?

In [32]: df_filtered = df_fact_survey_responses[df_fact_survey_responses["Respondent_ID"]

Out[32]:
Marketing_Channels  Online ads  3373
Other               762
Outdoor billboards  762
Print media         446
TV commercials      2795
dtype: int64

In [36]: g = g.plot(kind="pie", autopct="%1.1f%%")

Out[36]: <Axes: >

Online ads
40.1%
TV commercials
25.5%
Print media
10.0%
Outdoor billboards
10.0%
Other
6.4%

Answer: Online ads reaches most youth

# 4. Consumer Preferences
a. What are the preferred ingredients of energy drinks among respondents?

In [35]: df_ingredients_exp = df_fact_survey_responses.groupby("Ingredients_Expected").size().sort_values(ascending=False)
df_ingredients_exp

Out[35]:
Ingredients_Expected  Caffeine  3895
                    Vitamins  2134
                    Sugar    2027
                    Guarana  1553
dtype: int64

In [36]: df_ingredients_exp.plot(kind="bar", x="Ingredients_Expected", y="Count", color="purple", legend=False)
plt.title("Distribution of Preferred Ingredients")
plt.xlabel("Ingredients_Expected")
plt.ylabel("Count")

Out[36]: Text(0, 8.5, 'Count')

Distribution of Preferred Ingredients
Count
4000
3500
3000
2500
2000
1500
1000
500
0
Caffeine  Vitamins  Sugar  Guarana
Ingredients Expected

Answer: Caffeine and Vitamins are the most preferred ingredients

What packaging preferences do respondents have for energy drinks?

In [37]: df_packaging_pref = df_fact_survey_responses.groupby("Packaging_Preference").size().sort_values(ascending=False)
df_packaging_pref

Out[37]:
Packaging_Preference  Compact and portable cans  3084
                    Innovative bottle design  3087
                    Collectible packaging  1581
                    Eco-friendly design  583
                    Other  485
dtype: int64

In [38]: df_packaging_pref.plot(kind="bar", color="purple", y="Packaging_Preference")
plt.title("Distribution of Packaging Preference")
plt.xlabel("Packaging_Preference")
plt.ylabel("Packaging_Preference")

Out[38]: Text(0, 8.5, 'Packaging_Preference')

Distribution of Packaging Preference
Packaging Preference
Eco-friendly design
Collectible packaging
Innovative bottle design
Compact and portable cans
0 500 1000 1500 2000 2500 3000 3500 4000
Count

Answer: People prefer Compact and portable cans and Innovative bottle design

# 5. Competition Analysis
a. Who are the current market leaders?

In [39]: df_fact_survey_responses.groupby("Current_Brands").size().sort_values(ascending=False).reset_index(name="Count")

Out[39]:
  Current_Brands  count
0  Cola-Coke  2538
1  Bepsi  2112
2  Gangster  1854
3  Blue Bull  1058
4  CodeX  980
5  Sky 9  979
6  Others  479

Answer: Current Market Leaders are Cola-Coke, Bepsi, Gangster

b. What are the primary reasons consumers prefer those brands over ours?

In [39]: df_filter_brands = df_fact_survey_responses[df_fact_survey_responses["Current_Brands"]

Out[39]:
  Reasons_for_choosing_brands  count
0  Brand reputation  1993
1  Taste/Flavor preference  1548
2  Availability  1447
3  Effectiveness  1297
4  Other  1277

People mostly prefer other brands because of its brand reputation followed by their Taste/Flavor preference, Availability, Effectiveness, Other reasons

# 6. Marketing Channels and Brand Awareness
a. Which marketing channel can be used to reach more customers?

In [40]: df_fact_survey_responses.groupby("Marketing_Channels").size().plot(kind="pie", autopct="%1.1f%%")

Out[40]: <Axes: >

Online ads
40.2%
TV commercials
26.9%
Print media
12.3%
Outdoor billboards
12.3%
Other
8.4%

Answer: Online ads and TV commercials can be used to reach customers

b. How effective are different marketing strategies and channels in reaching our customers?

In [41]: df_reach_of_marketing_channels = df_fact_survey_responses.groupby("Marketing_Channels").size().reset_index(name="Total Count of Marketing Channels")
df_reach_of_marketing_channels.sort_values(by="Total Count of Marketing Channels", ascending=False)

Out[41]:
  Marketing_Channels  Total Count of Marketing Channels
0  Online ads  4020
4  TV commercials  2688
2  Outdoor billboards  1226
1  Other  1225
3  Print media  841

In [42]: df_reach_of_marketing_channels.groupby(["Age", "Marketing_Channels"]).size().reset_index(name="Count of Marketing Channels")
df_reach_of_marketing_channels.sort_values(by="Count of Marketing Channels", ascending=False)

Out[42]:
  Age  Marketing_Channels  Count
0  15-18  Online ads  188
1  15-18  TV commercials  1487
2  15-18  Outdoor billboards  1487
3  15-18  Other  1487
4  15-18  Print media  1487
5  15-18  Online ads  1487
6  15-18  TV commercials  1487
7  15-18  Outdoor billboards  1487
8  15-18  Other  1487
9  15-18  Print media  1487
10 15-18  Online ads  1487
11 15-18  TV commercials  1487
12 15-18  Outdoor billboards  1487
13 15-18  Other  1487
14 15-18  Print media  1487
15 15-18  Online ads  1487
16 15-18  TV commercials  1487
17 15-18  Outdoor billboards  1487
18 15-18  Other  1487
19 15-18  Print media  1487
20 15-18  Online ads  1487
21 15-18  TV commercials  1487
22 15-18  Outdoor billboards  1487
23 15-18  Other  1487
24 15-18  Print media  1487
25 15-18  Online ads  1487
26 15-18  TV commercials  1487
27 15-18  Outdoor billboards  1487
28 15-18  Other  1487
29 15-18  Print media  1487
30 15-18  Online ads  1487
31 15-18  TV commercials  1487
32 15-18  Outdoor billboards  1487
33 15-18  Other  1487
34 15-18  Print media  1487
35 15-18  Online ads  1487
36 15-18  TV commercials  1487
37 15-18  Outdoor billboards  1487
38 15-18  Other  1487
39 15-18  Print media  1487
40 15-18  Online ads  1487
41 15-18  TV commercials  1487
42 15-18  Outdoor billboards  1487
43 15-18  Other  1487
44 15-18  Print media  1487
45 15-18  Online ads  1487
46 15-18  TV commercials  1487
47 15-18  Outdoor billboards  1487
48 15-18  Other  1487
49 15-18  Print media  1487
50 15-18  Online ads  1487
51 15-18  TV commercials  1487
52 15-18  Outdoor billboards  1487
53 15-18  Other  1487
54 15-18  Print media  1487
55 15-18  Online ads  1487
56 15-18  TV commercials  1487
57 15-18  Outdoor billboards  1487
58 15-18  Other  1487
59 15-18  Print media  1487
60 15-18  Online ads  1487
61 15-18 
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