Alcohol data consumption Analysis

Team 3: RuntimeTerror Anvi, Aruna Atreyi, Gauri, Riddhi, Vaibhavi

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
df1 = pd.read excel('/student-por.xlsx')
df2 = pd.read_excel('/student-por.xlsx')
# Add subject column
df1['subject'] = 'Math'
df2['subject'] = 'Portuguese'
# Combine datasets
df = pd.concat([df1, df2], ignore_index=True)
# Drop irrelevant columns if they exist
cols_to_drop = ['nursery', 'school'] # Add more if needed
df.drop(columns=[col for col in cols_to_drop if col in df.columns], inplace=True)
# Create Sum/60
df['Sum/60'] = df['G1'] + df['G2'] + df['G3']
df1.info()
<<class 'pandas.core.frame.DataFrame'>
    RangeIndex: 649 entries, 0 to 648
    Data columns (total 34 columns):
                     Non-Null Count Dtype
     # Column
                    649 non-null
     0
         school
                                     obiect
     1
         sex
                    649 non-null
                                     object
         age
                     649 non-null
         address
                     649 non-null
                                     object
                    649 non-null
         famsize
                                     object
         Pstatus
                     649 non-null
                                     object
                     649 non-null
         Medu
                                     int64
         Fedu
                     649 non-null
                                     int64
                     649 non-null
                                     obiect
         Miob
         Fjob
                     649 non-null
                                     object
     10
         reason
                     649 non-null
                                     object
         guardian
                     649 non-null
                                     object
         traveltime 649 non-null
                                     int64
     13 studytime 649 non-null
                     649 non-null
         failures
                                     int64
     15 schoolsup 649 non-null
                                     object
     16
         famsup
                     649 non-null
                                     object
                     649 non-null
     17 paid
                                     obiect
         activities 649 non-null
     18
                                     object
                     649 non-null
     19
         nursery
                                     object
                     649 non-null
      20
         higher
                                     object
      21 internet
                     649 non-null
                                     object
         romantic
                     649 non-null
                                     object
      23 famrel
                     649 non-null
                                     int64
         freetime
                     649 non-null
                                     int64
                     649 non-null
         goout
     26
         Dalc
                     649 non-null
                                     int64
                     649 non-null
                                     int64
         Walc
      27
     28 health
                     649 non-null
                                     int64
                     649 non-null
                                     int64
      29
         absences
                     649 non-null
                                     int64
     30 G1
                     649 non-null
     31 G2
                                     int64
     32 G3
                     649 non-null
                                     int64
     33 subject
                     649 non-null
                                     object
    dtypes: int64(16), object(18)
    memory usage: 172.5+ KB
```

<<cl><class 'pandas.core.frame.DataFrame'>
RangeIndex: 649 entries, 0 to 648
Data columns (total 34 columns):

#	Column	Non-Null Count	Dtype
0	school	649 non-null	object
1	sex	649 non-null	object
2	age	649 non-null	int64
3	address	649 non-null	object
4	famsize	649 non-null	object
5	Pstatus	649 non-null	object
6	Medu	649 non-null	int64
7	Fedu	649 non-null	int64
8	Mjob	649 non-null	object
9	Fjob	649 non-null	object
10	reason	649 non-null	object
11	guardian	649 non-null	object
12	traveltime	649 non-null	int64
13	studytime	649 non-null	int64
14	failures	649 non-null	int64
15	schoolsup	649 non-null	object
16	famsup	649 non-null	object
17	paid	649 non-null	object
18	activities	649 non-null	object
19	nursery	649 non-null	object
20	higher	649 non-null	object
21	internet	649 non-null	object
22	romantic	649 non-null	object
23	famrel	649 non-null	int64
24	freetime	649 non-null	int64
25	goout	649 non-null	int64
26	Dalc	649 non-null	int64
27	Walc	649 non-null	int64
28	health	649 non-null	int64
29	absences	649 non-null	int64
30	G1	649 non-null	int64
31	G2	649 non-null	int64
32	G3	649 non-null	int64
33	subject	649 non-null	object
dtype	es: int64(16), object(18)	

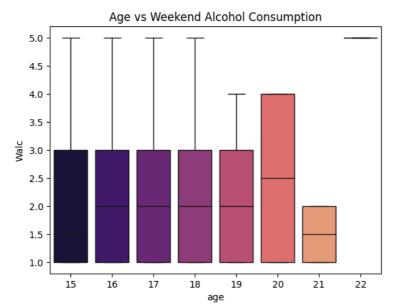
dtypes: int64(16), object(18)
memory usage: 172.5+ KB

df1.describe()

→ *		age	Medu	Fedu	traveltime	studytime	failures	famrel	freetime	goout	Dalc	Wal
	count	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.00000
	mean	16.744222	2.514638	2.306626	1.568567	1.930663	0.221880	3.930663	3.180277	3.184900	1.502311	2.28043
	std	1.218138	1.134552	1.099931	0.748660	0.829510	0.593235	0.955717	1.051093	1.175766	0.924834	1.28438
	min	15.000000	0.000000	0.000000	1.000000	1.000000	0.000000	1.000000	1.000000	1.000000	1.000000	1.00000
	25%	16.000000	2.000000	1.000000	1.000000	1.000000	0.000000	4.000000	3.000000	2.000000	1.000000	1.00000
	50%	17.000000	2.000000	2.000000	1.000000	2.000000	0.000000	4.000000	3.000000	3.000000	1.000000	2.00000
	75%	18.000000	4.000000	3.000000	2.000000	2.000000	0.000000	5.000000	4.000000	4.000000	2.000000	3.00000
	may	22 በበበበበበ	4 000000	4 000000	4 በበበበበበ	4 000000	3 000000	5 000000	5 000000	5 000000	5 000000	5 00000

df2.describe()

	age	Medu	Fedu	traveltime	studytime	failures	famrel	freetime	goout	Dalc	Wal
count	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.000000	649.00000
mean	16.744222	2.514638	2.306626	1.568567	1.930663	0.221880	3.930663	3.180277	3.184900	1.502311	2.28043
std	1.218138	1.134552	1.099931	0.748660	0.829510	0.593235	0.955717	1.051093	1.175766	0.924834	1.28438
min	15.000000	0.000000	0.000000	1.000000	1.000000	0.000000	1.000000	1.000000	1.000000	1.000000	1.00000
25%	16.000000	2.000000	1.000000	1.000000	1.000000	0.000000	4.000000	3.000000	2.000000	1.000000	1.00000
50%	17.000000	2.000000	2.000000	1.000000	2.000000	0.000000	4.000000	3.000000	3.000000	1.000000	2.00000
75%	18.000000	4.000000	3.000000	2.000000	2.000000	0.000000	5.000000	4.000000	4.000000	2.000000	3.00000
may	22 በበበበበበ	4 000000	4 000000	4 000000	4 000000	3 000000	5 000000	5 000000	5 000000	5 000000	5 00000



The above plot shows us how age impacts alcohol consumption in teenagers (per week) shown in the form of a box plot.

- 1. from the above representation we can tell that student of age group 20 have the most weekly alcohol consumption.
- 2. age groups 15-19 have the same alcohol consumption.

(basis of quantity of alc consumed)

```
# Q2: Which gender drinks more?
sns.countplot(data=df, x='Walc', hue='sex', palette='Set2')
plt.title("Gender-wise Alcohol Consumption Distribution")
plt.xlabel("Weekend Alcohol Consumption")
plt.ylabel("Count")
plt.show()
```



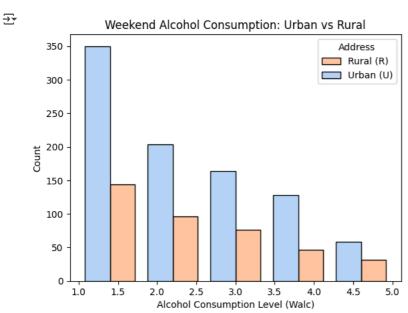
The above countplot shows the frequency of alcohol consumption betweeen male and female.

- 1. we can see that count of females descrease as quantity of alcohol consumed increases on a weekly basis.
- 2. on the other hand, count of men increases where there is more alcohol consumption

thus, we can say that among teens males drink more alcohol than females (if we consider quantity of alcohol consumed) and more females drink than male (if the count of students is considered)

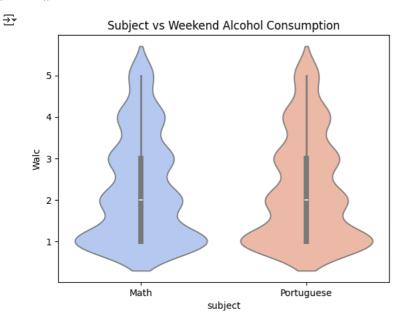
```
# Q3: Urban/rural area's influence on alcoholism
sns.histplot(data=df, x='Walc', hue='address', multiple='dodge', bins=5, shrink=0.8, palette='pastel')
plt.title("Weekend Alcohol Consumption: Urban vs Rural")
plt.xlabel("Alcohol Consumption Level (Walc)")
```

```
plt.ylabel("Count")
plt.legend(title='Address', labels=['Rural (R)', 'Urban (U)'])
plt.show()
```



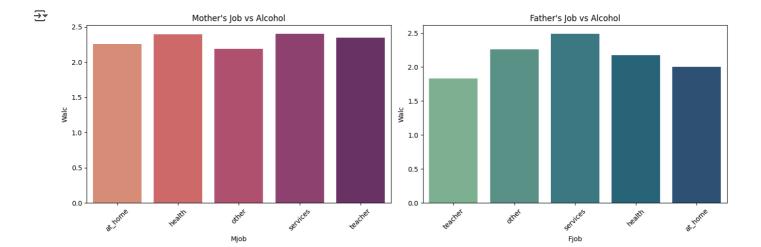
This histogram shows us the weekly alcohol consumption between urban students and rural students. From the plot we can see that urban students drink more alcohol weekly.

```
# Q4: Subject analysis - Which subject's students drink more?
df['subject'] = ['Math'] * len(df1) + ['Portuguese'] * len(df2)
sns.violinplot(data=df, x='subject', y='Dalc', palette='coolwarm')
plt.title("Subject vs Weekend Alcohol Consumption")
plt.show()
```



The above violin plot shows amount of alcohol consumed per subject. We infer that there is not much disparity in levels of alcohol consumed. Hence we can say that subjects don't have much influence over alcohol consumption

```
# Q5: How does the type of parental job influence alcohol consumption?
fig, axs = plt.subplots(1, 2, figsize=(14, 5))
sns.barplot(data=df, x='Mjob', y='Walc', ci=None, ax=axs[0], palette='flare')
axs[0].set_title("Mother's Job vs Alcohol")
axs[0].tick_params(axis='x', rotation=45)
sns.barplot(data=df, x='Fjob', y='Walc', ci=None, ax=axs[1], palette='crest')
axs[1].set_title("Father's Job vs Alcohol")
axs[1].tick_params(axis='x', rotation=45)
plt.tight_layout()
plt.show()
```

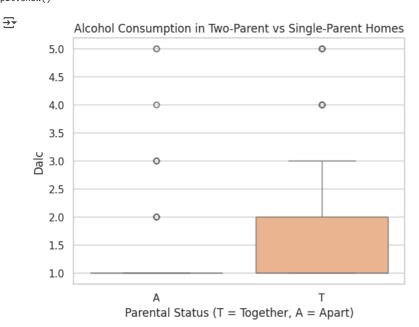


Overall Insights: Service sector jobs (for both parents) seem to correlate with higher alcohol consumption among students.

Parental involvement or presence at home (especially mothers) appears to relate to slightly lower alcohol consumption.

There is a more noticeable variation in alcohol consumption based on father's job than mother's, possibly indicating stronger correlation or influence.

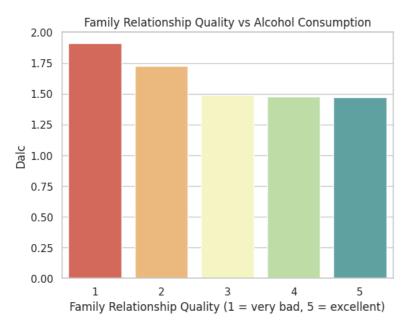
```
# Q6: Alcohol difference between two-parent and single-parent families
sns.boxplot(data=df, x='Pstatus', y='Dalc', palette='pastel')
plt.title("Alcohol Consumption in Two-Parent vs Single-Parent Homes")
plt.xlabel("Parental Status (T = Together, A = Apart)")
plt.show()
```



Students from single parent home generally have a slightly higher median alcohol consumption than those from 2 parent homes. There are more outliers with high alcohol use in single parent home suggesting a possible impact of family structure on behaviour.

```
# Q7: What kind of relationship does the child have with family members?
sns.barplot(data=df, x='famrel', y='Dalc', ci=None, palette='Spectral')
plt.title("Family Relationship Quality vs Alcohol Consumption")
plt.xlabel("Family Relationship Quality (1 = very bad, 5 = excellent)")
plt.show()
```





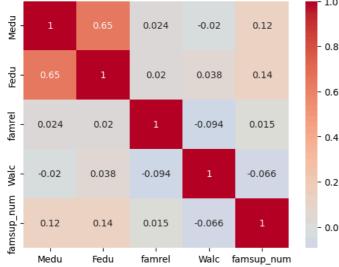
Student with bad family relations tend to have a higher daily alcohol consumption level.

```
# Q8: Family background's influence on alcoholism (correlation heatmap)
corr_data = df[['Medu', 'Fedu', 'famrel', 'Walc']].copy()
corr_data['famsup_num'] = df['famsup'].map({'yes': 1, 'no': 0}) # Convert to numeric

corr = corr_data.corr()
sns.heatmap(corr, annot=True, cmap='coolwarm', center=0)
plt.title("Correlation between Family Background & Alcohol Consumption")
plt.show()
```

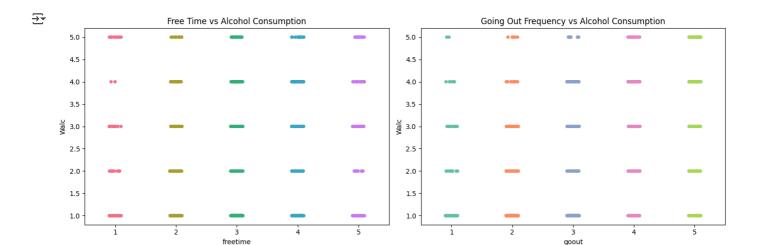






The heat map shows the correlation between family background factors and alcohol consumption. A strong positive correlation exists between mothers and fathers education, while alcohol consumption on weekends has very weak or negative correlation with family background variables suggesting minimal influence.

```
# Q9: Does free time or going out relate to alcohol?
fig, axs = plt.subplots(1, 2, figsize=(14, 5))
sns.stripplot(data=df, x='freetime', y='Walc', ax=axs[0], palette='husl', jitter=True)
axs[0].set_title("Free Time vs Alcohol Consumption")
sns.stripplot(data=df, x='goout', y='Walc', ax=axs[1], palette='Set2', jitter=True)
axs[1].set_title("Going Out Frequency vs Alcohol Consumption")
plt.tight_layout()
plt.show()
```



These plots explore how free time and going out frequency relate to alcohol consumption. The left plot shows no strong pattern between free time and alcohol use. However, the right plot indiciates a clear trend: Students who go out more frequenty tend to have high alcohol consumption levels.

```
# Q10: Alcohol consumption vs academic performance
sns.set_theme(style="whitegrid")
palette = sns.color_palette("coolwarm", as_cmap=True)
\ensuremath{\text{\#}}\xspace Lmplot with improved aesthetics and gender hue
sns.lmplot(
    data=df.
    x='Walc'
    y='Sum/60',
                                     # highlight by gender
    hue='sex',
    palette='Set2',
                                     # bright and readable colors
    height=6,
    aspect=1.2,
    markers=["o", "s"],
    scatter_kws={'alpha': 0.7, 's': 60}, # transparency & point size
    line_kws={"linewidth": 2}
plt.title("Impact of Alcohol Consumption on Academic Performance by Gender", fontsize=14)
plt.xlabel("Weekend Alcohol Consumption (Walc)", fontsize=12)
plt.ylabel("Total Grades (Sum/60)", fontsize=12)
plt.tight_layout()
plt.show()
```

