

```
In [11]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import ttest_ind
```

```
In [31]: # Load the dataset (Assuming a CSV file named 'ev_data.csv')
df=pd.read_csv("D:/Bhavesh/Internshala Data Analytics/C5 PYTHON/Project/FEV-data-Excel.xlsx - Auta elektryczne.csv")
df.head()
```

Out[31]:

	Car full name	Make	Model	Minimal price (gross) [PLN]	Engine power [KM]	Maximum torque [Nm]	Type of brakes	Drive type	Battery capacity [kWh]	Range (WLTP) [km]	...	Permissible gross weight [kg]	Maximum load capacity [kg]	Number of seats	Number of doors	Tire size [in]	Maximum speed [kph]	Boot capacity (VDA) [l]	Acceleration 0-100 kph [s]	Maximum DC charging power [kW]	mean - Energy consumption [kWh/100 km]
0	Audi e-tron 55 quattro	Audi	e-tron 55 quattro	345700	360	664	disc (front + rear)	4WD	95.0	438	...	3130.0	640.0	5	5	19	200	660.0	5.7	150	24.45
1	Audi e-tron 50 quattro	Audi	e-tron 50 quattro	308400	313	540	disc (front + rear)	4WD	71.0	340	...	3040.0	670.0	5	5	19	190	660.0	6.8	150	23.80
2	Audi e-tron S quattro	Audi	e-tron S quattro	414900	503	973	disc (front + rear)	4WD	95.0	364	...	3130.0	565.0	5	5	20	210	660.0	4.5	150	27.55
3	Audi Sportback 50 quattro	Audi	e-tron Sportback 50 quattro	319700	313	540	disc (front + rear)	4WD	71.0	346	...	3040.0	640.0	5	5	19	190	615.0	6.8	150	23.30
4	Audi Sportback 55 quattro	Audi	e-tron Sportback 55 quattro	357000	360	664	disc (front + rear)	4WD	95.0	447	...	3130.0	670.0	5	5	19	200	615.0	5.7	150	23.85

5 rows × 25 columns

```
In [15]: #Task 1 Part a
filtered_ews = df[(df['Minimal price (gross) [PLN]'] <= 350000) & (df['Range (WLTP) [km]'] >= 400)]
print(filtered_ews[['Make', 'Model', 'Minimal price (gross) [PLN]', 'Range (WLTP) [km]']])
```

	Make	Model	Minimal price (gross) [PLN]	\
0	Audi	e-tron 55 quattro	345700	
8	BMW	iX3	282900	
15	Hyundai	Kona electric 64kWh	178400	
18	Kia	e-Niro 64kWh	167990	
20	Kia	e-Soul 64kWh	160990	
22	Mercedes-Benz	EQC	334700	
39	Tesla	Model 3 Standard Range Plus	195490	
40	Tesla	Model 3 Long Range	235490	
41	Tesla	Model 3 Performance	260490	
47	Volkswagen	ID.3 Pro Performance	155890	
48	Volkswagen	ID.3 Pro S	179990	
49	Volkswagen	ID.4 1st	202390	

	Range (WLTP) [km]
0	438
8	460
15	449
18	455
20	452
22	414
39	430
40	580
41	567
47	425
48	549
49	500

```
In [102]: #Part b
grouped_ews = filtered_ews.groupby('Make').size()
print(grouped_ews)
```

Make	
Audi	1
BMW	1
Hyundai	1
Kia	2
Mercedes-Benz	1
Tesla	3
Volkswagen	3
dtype:	int64

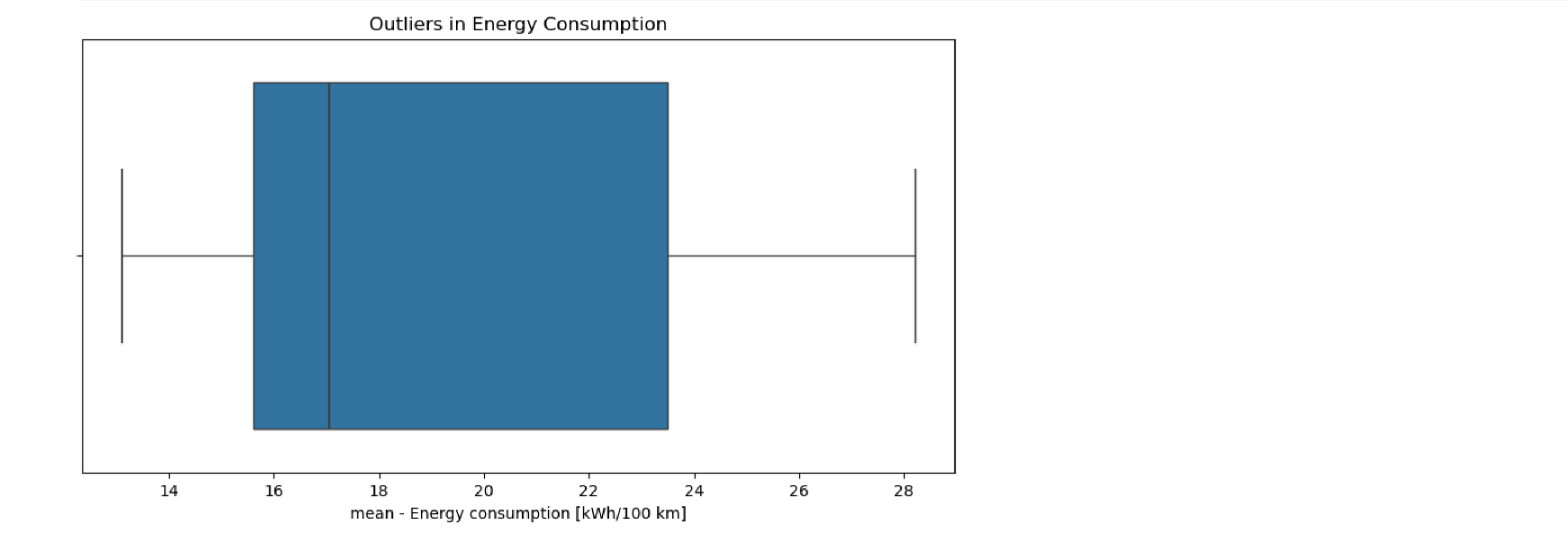
```
In [126]: #Part c
avg_battery_capacity = filtered_ews.groupby('Make')['Battery capacity [kWh]'].mean()
print(avg_battery_capacity)
```

Make	
Audi	95.000000
BMW	80.000000
Hyundai	64.000000
Kia	64.000000
Mercedes-Benz	80.000000
Tesla	68.000000
Volkswagen	70.666667
Name: Battery capacity [kWh], dtype: float64	

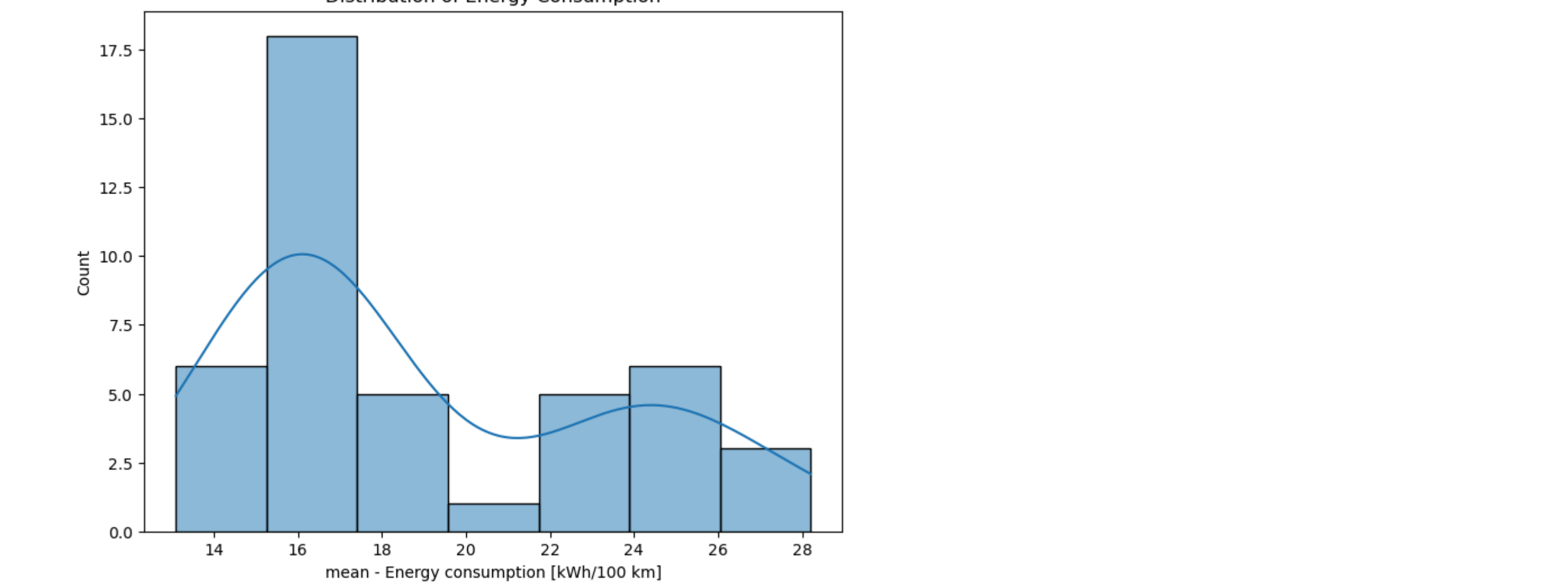
```
In [124]: #Task 2
q1 = df['mean - Energy consumption [kWh/100 km]'].quantile(0.25)
q3 = df['mean - Energy consumption [kWh/100 km]'].quantile(0.75)
iqr = q3 - q1
lower_bound = q1 - 1.5 * iqr
upper_bound = q3 + 1.5 * iqr
outliers = df[(df['mean - Energy consumption [kWh/100 km]'] < lower_bound) | (df['mean - Energy consumption [kWh/100 km]'] > upper_bound)]
print(outliers[['Make', 'Model', 'mean - Energy consumption [kWh/100 km]']])

Empty DataFrame
Columns: [Make, Model, mean - Energy consumption [kWh/100 km]]
Index: []
```

```
In [136]: #Task 2 Plotting Boxplot
plt.figure(figsize=(10,5))
sns.boxplot(x=df['mean - Energy consumption [kWh/100 km]'])
plt.title("Outliers in Energy Consumption")
plt.show()
```



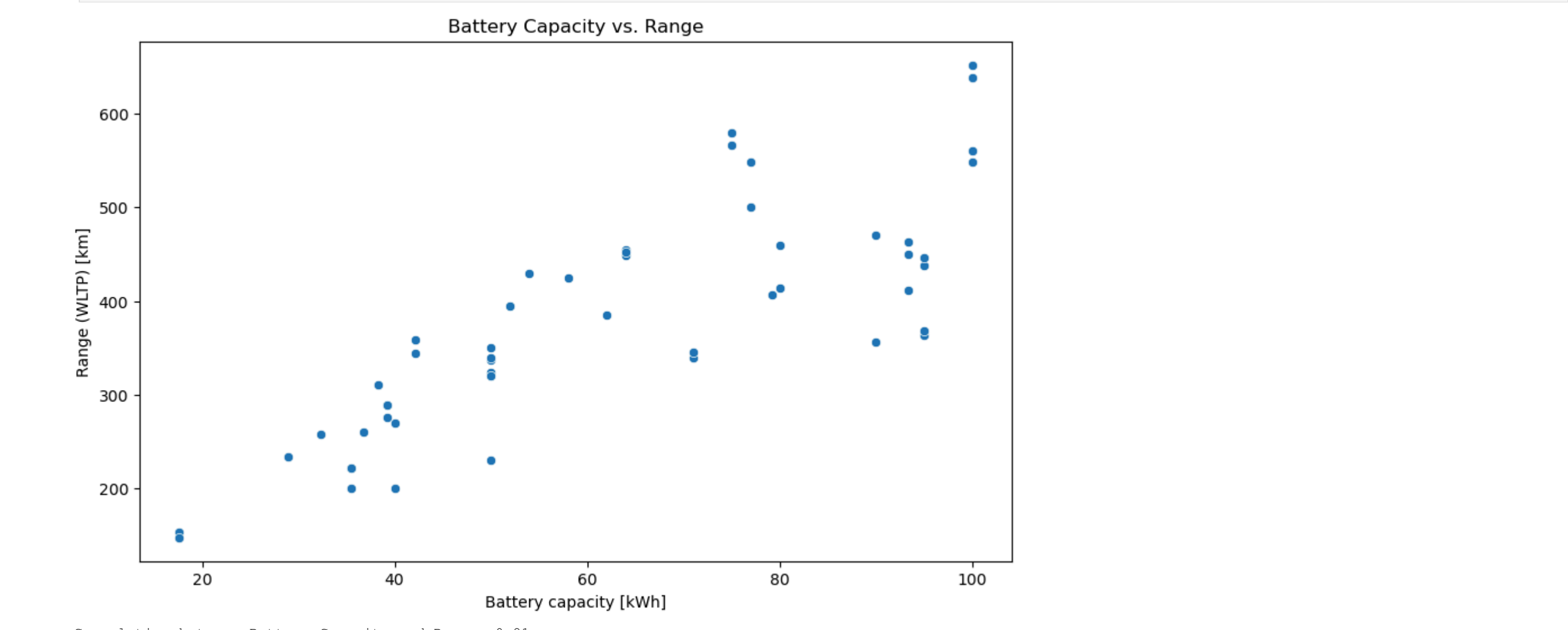
```
In [138]: #Task 2 Plotting Histogram
plt.figure(figsize=(8, 6))
sns.histplot(df['mean - Energy consumption [kWh/100 km]'], kde=True)
plt.title("Distribution of Energy Consumption")
plt.show()
```



```
In [158]: #Task 3
def analyze_battery_range_relationship(df: pd.DataFrame):
    plt.figure(figsize=(10, 6))
    sns.scatterplot(x=df['Battery capacity [kWh]'], y=df['Range (WLTP) [km]'])
    plt.xlabel("Battery capacity [kWh]")
    plt.ylabel("Range (WLTP) [km]")
    plt.title("Battery Capacity vs. Range")
    plt.show()

    correlation = df['Battery capacity [kWh]'].corr(df['Range (WLTP) [km]'])
    return correlation
```

```
if __name__ == "__main__":
    df = pd.read_csv("D:/Bhavesh/Internshala Data Analytics/C5 PYTHON/Project/FEV-data-Excel.xlsx - Auta elektryczne.csv")
    correlation = analyze_battery_range_relationship(df)
    print(f"\nCorrelation between Battery Capacity and Range: {correlation:.2f}")
```



Correlation between Battery Capacity and Range: 0.81

```
In [23]: # Task 4
class EVRecommendation:
    def __init__(self, df):
        self.df = df

    def recommend_ev(self, budget, min_range, min_battery_capacity):
        filtered_ews = self.df[
            (self.df['Minimal price (gross) [PLN]'] <= budget) &
            (self.df['Range (WLTP) [km]'] >= min_range) &
            (self.df['Battery capacity [kWh]'] >= min_battery_capacity)
        ]

        if filtered_ews.empty:
            return "No EVs match your criteria. Try adjusting your budget or requirements."

        top_ews = filtered_ews.sort_values(by=['Range (WLTP) [km]', 'Battery capacity [kWh]'], ascending=[False, False]).head(3)

        return top_ews[['Car full name', 'Minimal price (gross) [PLN]', 'Range (WLTP) [km]', 'Battery capacity [kWh]']]

ev_recommender = EVRecommendation(df)

budget = float(input("Enter your budget in PLN: "))
min_range = float(input("Enter the minimum range you need (in km): "))
min_battery_capacity = float(input("Enter the minimum battery capacity (in kWh): "))

recommended_ews = ev_recommender.recommend_ev(budget, min_range, min_battery_capacity)

print("\nTop Recommended EVs:")
print(recommended_ews)
```

	Car full name	Minimal price (gross) [PLN]	Range (WLTP) [km]	\
40	Tesla Model 3 Long Range	235490	580	
41	Tesla Model 3 Performance	260490	567	
48	Volkswagen ID.3 Pro S	179990	549	

	Battery capacity [kWh]
40	75.0
41	75.0
48	77.0

```
In [162]: #Task 5
import scipy.stats as stats

tesla_power = df[df['Make'] == 'Tesla']['Engine power [kW]']
audi_power = df[df['Make'] == 'Audi']['Engine power [kW]']

t_stat, p_value = stats.ttest_ind(tesla_power, audi_power, equal_var=False)

print(f"***50")
print("Hypothesis Testing: Tesla vs. Audi Engine Power")
print(f"***50")
print(f"T-statistic: {t_stat:.4f}")
print(f"P-value: {p_value:.4f}")
print(f"***50")

alpha = 0.05
if p_value < alpha:
    print("Conclusion: We reject the Null Hypothesis (H0).")
    print("There is a significant difference in the average engine power between Tesla and Audi EVs.")
else:
    print("Conclusion: We fail to reject the Null Hypothesis (H0).")
    print("There is NO significant difference in the average engine power between Tesla and Audi EVs.")

print(f"***50")

if p_value < alpha:
    if tesla_power.mean() > audi_power.mean():
        print("Insight: Tesla's EVs have significantly higher average engine power than Audi.")
        print("Recommendation: Tesla can market its EVs as high-performance vehicles, targeting customers who value power.")
    else:
        print("Insight: Audi's EVs have significantly higher average engine power than Tesla.")
        print("Recommendation: Audi should highlight its power advantage in marketing and product positioning.")
else:
    print("Insight: There is no significant difference in the average engine power of Tesla and Audi EVs.")
    print("Recommendation: Both brands are competing closely in performance. Customers should focus on other factors like range, price, and charging speed when making pu")

print(f"***50")

=====
Hypothesis Testing: Tesla vs. Audi Engine Power
=====
T-statistic: 1.7940
P-value: 0.1068
=====
Conclusion: We fail to reject the Null Hypothesis (H0).
There is NO significant difference in the average engine power between Tesla and Audi EVs.
=====
Insight: There is no significant difference in the average engine power of Tesla and Audi EVs.
Recommendation: Both brands are competing closely in performance. Customers should focus on other factors like range, price, and charging speed when making purchasing dec
```

