SUMMER ANALYTICS 2024 Week-1 Assignment

Data Grand Prix!

name

buick skylark 320

plymouth satellite

amc rebel sst

ford torino

70 usa chevrolet chevelle malibu

70 usa

70 usa

70 usa

70 usa

Welcome to your first assignment of Summer Analytics 2025! We hope you are excited to implement and test everything you have learnt up until now. The dataset which you'll use includes information about cars.

We've got an interesting set of questions for you to get a basic understanding of pandas and data visualization libraries. GOOD LUCK!

Let's get started with importing numpy, pandas, seaborn and matplotlib!

Note - matplotlib should be imported with the command :

import matplotlib.pyplot as plt

So lets get started!! Buckle up your belts for this exciting ride!! 1) Start by importing all important libraries For eg, "import numpy as np" import numpy as np import pandas as pd 2) Read the csv file and assign it to a variable . In [8]: df = pd.read_csv("Cars.csv") df.head()

1 15.0 350.0 165.0 3693 **2** 18.0 8 318.0 150.0 3436

304.0

302.0

307.0

0 18.0

3 16.0

4 17.0

Expected Output - (398, 9)

mpg cylinders displacement horsepower weight acceleration model_year origin

130.0 3504

150.0 3433

140.0 3449

12.0

11.5

11.0

12.0

10.5

mpg cylinders displacement horsepower weight acceleration model_year origin

130.0 3504

165.0 3693

12.0

11.5

11.0

8) Create a column which contains the horsepower divided by weightas its metric and make this new column the index.

12.0

11.5

11.0

12.0

10.5

Checkpoint!! Congratulations on making it this far. You are really keeping up in Data Grand Prix. Now starts the real race i.e. graded questions of the quiz.

Try to retrieve some information from the data and answer the questions below . BEST OF LUCK !!

70 usa

70 usa

70 usa

70 usa

name

buick skylark 320

plymouth satellite

amc rebel sst

ford torino

70 usa chevrolet chevelle malibu

70 usa

70 usa

70 usa

70 usa

 $[18.0,\ 15.0,\ 16.0,\ 17.0,\ 14.0,\ 24.0,\ 22.0,\ 21.0,\ 27.0,\ 26.0,\ 25.0,\ 10.0,\ 11.0,\ 9.0,\ 28.0,\ 19.0,\ 12.0,\ 33.0,\ 17.5,\ 15.5,\ 14.5,\ 22.5,\ 24.5,\ 18.5,\ 29.5,\ 26.5,\ 16.5,\ 31.5,\ 36.0,\ 25.5,\ 33.5,\ 20.5,\ 30.5,\ 21.5,\ 43.1,\ 36.1,\$ 32.8, 39.4, 19.9, 19.4, 20.2, 19.2, 25.1, 20.6, 20.8, 18.6, 18.1, 17.7, 27.5, 27.2, 30.9, 21.1, 23.2, 23.8, 23.9, 20.3, 21.6, 16.2, 19.8, 22.3, 17.6, 18.2, 16.9, 31.9, 34.1, 35.7, 27.4, 25.4, 34.2, 34.5, 31.8, 37.3, 28.4, 28.8, 26.8, 41.5, 38.1, 32.1, 37.2, 26.4,

24.3, 19.1, 34.3, 29.8, 31.3, 37.0, 32.2, 46.6, 27.9, 40.8, 44.3, 43.4, 36.4, 44.6, 40.9, 33.8, 32.7, 23.7, 23.7, 39.1, 39.0, 35.1, 32.3, 37.7, 34.4, 29.9, 33.7, 32.9, 31.6, 28.1, 30.7, 24.2, 22.4, 34.0, 38.0, 44.0]

Congratulations on coming this far! Since we were having so much fun playing with this dataset, let's move towards finish line by attempting some Ungraded questions!

3) Display shape of dataframe

In [10]: df.shape Out[10]: (398, 9)

In [12]: print(df.columns) list(df.columns) Index(['mpg', 'cylinders', 'displacement', 'horsepower', 'weight', 'acceleration', 'model_year', 'origin', 'name'],

dtype='object')

chevrolet chevelle malibu 18.0

buick skylark 320 15.0

df = pd.read_csv('Cars.csv')

print(list(unique_mpg))

df.head()

hp_per_weight

0.037100 18.0

0.044679 15.0

0.043655 18.0

0.043694 16.0

0.040591 17.0

In [21]: df.reset_index(inplace=True)

Out[18]:

unique_mpg = df['mpg'].drop_duplicates()

In [18]: df['hp_per_weight'] = df['horsepower'] / df['weight'] df.set_index('hp_per_weight', inplace=True)

'cylinders', 'displacement', 'horsepower', 'weight',

df.head()

Out[14]:

Out[12]: ['mpg',

4) Print all columns of dataframe

Return an array containing names of all the columns.

'acceleration', 'model_year', 'origin', 'name'] 6) Set the 'name' column as the index of dataframe In [14]: df.set_index('name', inplace=True)

318.0 150.0 3436 plymouth satellite 18.0 amc rebel sst 16.0 150.0 3433 ford torino 17.0 140.0 3449 7) Print a list of all the unique mpg values In [16]: import pandas as pd

307.0

350.0

mpg cylinders displacement horsepower weight acceleration model_year origin

130.0 3504

165.0 3693

150.0 3436

150.0 3433

140.0 3449

307.0

350.0

318.0

302.0

df['horsepower'] = pd.to_numeric(df['horsepower'], errors='coerce') max_hp_row = df.loc[df['horsepower'].idxmax()] print("Car with highest horsepower:", max_hp_row['name']) Car with highest horsepower: pontiac grand prix

2. How many cars have mpg \geq 35?

print("Number of cars with mpg ≥ 35:", count)

In [23]: count = df[df['mpg'] >= 35].shape[0]

In [27]: japan_cars = df[df['origin'] == 'japan']

dtype='object')

Year with the highest average mpg: 80

In [37]: print(df.columns)

1. What is name of car that has the highest horsepower?

GRADED Questions (To be answered in the quiz)

Number of cars with mpg \geq 35: 36 3. What is the most common origin for cars with horsepower > 100 and weight < 3000? In [25]: df['horsepower'] = pd.to_numeric(df['horsepower'], errors='coerce')

filtered_df = df[(df['horsepower'] > 100) & (df['weight'] < 3000)]</pre> filtered_df = df[(df['horsepower'] > 100) & (df['weight'] < 3000)]</pre> most_common_origin = filtered_df['origin'].value_counts().idxmax()

Most common origin for cars with horsepower > 100 and weight < 3000: usa

avg_mpg_by_year = df.groupby('model_year')['mpg'].mean()

print("Year with the highest average mpg:", year_with_highest_mpg)

Index(['hp_per_weight', 'mpg', 'cylinders', 'displacement', 'horsepower', 'weight', 'acceleration', 'model_year', 'origin', 'name'],

year_with_highest_mpg = avg_mpg_by_year.idxmax()

max_ratio = high_mpg_cars['hp_to_weight'].max()

horsepower weight hp_to_weight 113.0 2234 0.050582

import matplotlib.pyplot as plt

import seaborn as sns

plt.figure(figsize=(12, 6))

origin -•- europe ─• japan -- usa

35

MPG 30

₹ 25

Color by: origin

Size by: mpg

In [5]: import pandas as pd

import seaborn as sns

In [7]: df = pd.read_csv('Cars.csv')

Create scatter plot

x='horsepower', y='weight', hue='origin', size='mpg', palette='deep',

sizes=(20, 200),

alpha=0.7

Add titles

plt.grid(**True**) plt.tight_layout()

plt.show()

Out[21]: 0 70

70 70

Step 2: Filter only those cars

filtered_df = df[df['name'].isin(repeated_names)]

sns.scatterplot(data=df,

plt.figure(figsize=(12, 6))

Hue order = ['japan', 'europe', 'usa']

Add meaningful plot titles and axis titles.

import matplotlib.pyplot as plt

In [9]: # Convert origin values to lowercase (just in case) df['origin'] = df['origin'].str.lower()

hue_order=['japan', 'europe', 'usa'],

plt.legend(title='Origin', loc='upper right')

plt.xlabel('Horsepower', fontsize=12) plt.ylabel('Weight', fontsize=12)

plt.title('Horsepower vs Weight of Cars\n(Sized by MPG, Colored by Origin)', fontsize=14)

In [7]: import pandas as pd

In [15]: print(df.columns)

mean_acceleration = round(japan_cars['acceleration'].mean(), 2) print("Mean acceleration of cars from Japan:", mean_acceleration) Mean acceleration of cars from Japan: 16.17 5. Which year had the highest average mpg?

print("Most common origin for cars with horsepower > 100 and weight < 3000:", most_common_origin)</pre>

4. What is the mean acceleration of cars from Japan? (rounded to 2 decimals)

Find the car (or cars) with the best ratio of horsepower to weight among all cars that also have above-median mpg. In [39]: median_mpg = df['mpg'].median() high_mpg_cars = df[df['mpg'] > median_mpg].copy()

high_mpg_cars['hp_to_weight'] = high_mpg_cars['horsepower'] / high_mpg_cars['weight']

best_cars = high_mpg_cars[high_mpg_cars['hp_to_weight'] == max_ratio]

print(best_cars[['horsepower', 'weight', 'hp_to_weight']])

Index(['mpg', 'cylinders', 'displacement', 'horsepower', 'weight',

plt.title('Evolution of Average MPG Over the Years by Origin')

Note: These questions are UNGRADED, and are given as an extra exercise.

df = pd.read_csv('Cars.csv') In [11]: df = pd.read_csv('Cars.csv', index_col='name') In [19]: print(df.columns)

Design a multi-line plot using Matplotlib or Seaborn that shows the evolution of average mpg over the years, separately for each origin

avg_mpg = df.groupby(['model_year', 'origin'])['mpg'].mean().reset_index() Index(['mpg', 'cylinders', 'displacement', 'horsepower', 'weight', 'acceleration', 'model_year', 'origin'], dtype='object')

'acceleration', 'model_year', 'origin'], dtype='object') In [17]: avg_mpg = df.groupby(['model_year', 'origin'])['mpg'].mean().reset_index() import matplotlib.pyplot as plt import seaborn as sns

sns.lineplot(data=avg_mpg, x='model_year', y='mpg', hue='origin', marker='o')

plt.xlabel('Model Year') plt.ylabel('Average MPG') plt.grid(**True**) plt.show() Evolution of Average MPG Over the Years by Origin

20 15 74 78 76 Model Year Create a Seaborn scatterplot (or PairGrid) where: X = horsepower Y = weight

5000 origin japan europe 4500 usa mpg 16 4000 24 32 **4**0 Weight 0058 3000 2500 1500 100 125 150 175 200 225 Horsepower We define a "consistent" car model as one that was produced over multiple years and had very low variation in mpg across those years (standard deviation < 1.0). Tasks: Identify car names that appear in more than one model_year. For each such name, compute the standard deviation of mpg across years. Return the car(s) with the lowest variation in mpg, among those with at least 2 appearances and std(mpg) < 1.0.

Horsepower vs Weight of Cars (Sized by MPG, Colored by Origin)

Origin

Report the model name(s), number of appearances, and the average mpg. Bonus: Sort the result by number of appearances (descending), then mpg (descending). In [15]: import pandas as pd df = pd.read_csv('Cars.csv') print(df.columns.tolist()) ['mpg', 'cylinders', 'displacement', 'horsepower', 'weight', 'acceleration', 'model_year', 'origin', 'name'] In [19]: df.columns = df.columns.str.strip().str.lower().str.replace(' ', '_') In [21]: df['model_year']

393 82 82 394 395 82 396 82 397 82 Name: model_year, Length: 398, dtype: int64 In [23]: import pandas as pd # Load data df = pd.read_csv('Cars.csv') # Normalize column names df.columns = df.columns.str.strip().str.lower().str.replace(' ', '_') # Step 1: Find car names appearing in multiple model years name_years = df.groupby('name')['model_year'].nunique() repeated_names = name_years[name_years > 1].index

Step 3: Compute statistics stats = filtered_df.groupby('name').agg(appearances=('mpg', 'count'), avg_mpg=('mpg', 'mean')).reset_index() # Step 4: Filter those with very low variation consistent_cars = stats[stats['mpg_std'] < 1.0]</pre> # Step 5: Sort by appearances (desc), then avg_mpg (desc) consistent_sorted = consistent_cars.sort_values(by=['appearances', 'avg_mpg'], ascending=[False, False] # Display final result print(consistent_sorted[['name', 'appearances', 'avg_mpg']]) name appearances avg_mpg 26 ford galaxie 500 3 14.333333 3 14.333333 41 plymouth fury iii 49 toyota corolla 1200 2 31.500000 35 mazda 626 2 31.450000 53 volkswagen rabbit 2 29.250000 2 27.000000 21 datsun pl510 45 saab 991e 2 24.500000 51 toyota mark ii 2 19.500000 22 2 18.850000 dodge aspen 10 chevrolet chevelle malibu 2 17.500000 2 14.500000 amc matador (sw) 28 ford gran torino (sw) 2 13.500000 29 ford ltd 2 13.500000

