DAV Assignment

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Roll Number: 23B3323

Links to original Google Colab files

- Q1, Q2: https://colab.research.google.com/drive/1go3ZNHivZT6uYTUN9luy9T-yaRoqV94w?usp=sharing
- Q3, Q4:
 https://colab.research.google.com/drive/1L18iM5lkv9_nUPDT9nUbdyQvi32cq5xQ?

 usp=sharing
- Q6: https://colab.research.google.com/drive/1pN89cFP62pxCKvwYNxKW2sA0tXLIIg5-?
 <a href="http
- Q7: https://colab.research.google.com/drive/1qAgQQws0h3xzMxZT3PfRz5oAecLzyXVP?

 usp=sharing

Links to Google Colab files without the debugging outputs

- Q1, Q2: https://colab.research.google.com/drive/13GkunQLL1eVTN2JLmBuxlCXDcY9XwmDJ? usp=sharing
- Q3, Q4:
 https://colab.research.google.com/drive/1qEjCFV74wzXWvD064UJhlmfvezKZTDpZ?

 usp=sharing
- Q6: https://colab.research.google.com/drive/1pN89cFP62pxCKvwYNxKW2sA0tXLIIg5-?
 <a href="http
- Q7: https://colab.research.google.com/drive/1q5kq3tOLcAulCc7Z4eqQ-hWlur-7VpZR?

 usp=sharing

Link to my GitHub repository

https://github.com/PanavShah1/DAV_assignment

Have uploaded all of the datasets (csv files) to the repository to access in the ipynb code

Predictions for Q3 & Q7

Q3 Edited Consumer Test Dataset -

https://drive.google.com/file/d/1ZaSdC6wvsJ2Mk4rxwhQ-Z6T59pMyNbPk/view?usp=sharing

Q7 Edited Email Test Dataset -

https://drive.google.com/file/d/1IBFMYjsE7XOsz3dxl0I5O3mJvRI6JXUb/view?usp=sharing

DAV Assignment Q1 and Q2

You are the head of the sales and marketing division of "The Renewables", a company that sells renewable energy solutions to people. Recognizing India's vast potential for renewable energy growth, you decide to extend your operations into this dynamic market. To make informed decisions, you need to understand the country's energy landscape, including its actual and installed energy capacities. For this you turned to the NDAP for comprehensive data on India's energy sector, and obtained this dataset from there:

Q1: Ask 10 reasonably involved questions and try to answer them by analyzing the Dataset.

→ *		ROWID	Country	State LGD Code	State	Actual energy generated	Category of Plant	Type of fuel used	Installed Capacity	Ge P
	0	1	India	22	Chhattisgarh	0.0	THERMAL	COAL	0.0	
	1	2	India	2	Himachal Pradesh	0.0	HYDRO	HYDRO	0.0	
	2	3	India	27	Maharashtra	0.0	THERMAL	COAL	0.0	
	3	4	India	9	Uttar Pradesh	0.0	THERMAL	COAL	0.0	
	4	5	India	22	Chhattisgarh	0.0	THERMAL	COAL	0.0	

Next steps: Generate code with df View recommended plots

states = df['State'].unique()
plants = df['Category of Plant'].unique()
fuels = df['Type of fuel used'].unique()
sectors = df['Sector of power plant'].unique()
years = df['YearCode'].unique()

Q1: How many Thermal, Hydro and Nuclear Plants are there in India?

```
✓ A1:
```

```
num_thermal = len(df.loc[df['Category of Plant'] == 'THERMAL'])
num_hydro = len(df.loc[df['Category of Plant'] == 'HYDRO'])
num_nuclear = len(df.loc[df['Category of Plant'] == 'NUCLEAR'])
print(f'There are {num_thermal} Thermal Plants, {num_hydro} Hydro Plants and {num_nuclear} Nuclear Plants in India')
```

There are 26427 Thermal Plants, 19465 Hydro Plants and 783 Nuclear Plants in India

Q2: How many plants use the different types of fuels

```
num_fuels = {fuel: 0 for fuel in fuels}

for fuel in num_fuels:
    condition = len(df.loc[df['Type of fuel used'] == fuel])
    num_fuels[fuel] = condition
```

✓ A2:

```
for fuel in num_fuels:
    print(f"There are {num_fuels[fuel]} plants using {fuel}")

→ There are 17380 plants using COAL
    There are 19465 plants using HYDRO
    There are 986 plants using LIGNITE
    There are 4795 plants using GAS
    There are 783 plants using NUCLEAR
    There are 137 plants using NULTI FUEL
    There are 1220 plants using NATURAL GAS
    There are 20 plants using THERMAL
    There are 1177 plants using DIESEL
```

Q3: What fuels do the Thermal, Hydro and Nuclear plants use

There are 267 plants using HIGH SPEED DIESEL

There are 445 plants using NAPTHA

```
plant_data = {plant: {fuel: 0 for fuel in num_fuels} for plant in plants}

for plant in plants:
    for fuel in num_fuels:
        condition = (df['Category of Plant'] == plant) & (df['Type of fuel used'] == fuel)
        length = len(df.loc[condition])
        plant_data[plant][fuel] = length
```

✓ A3:

```
for plant in plants:
    for fuel in num_fuels:
        if plant_data[plant][fuel] == 0:
            continue
        print(f"There are {plant_data[plant][fuel]} {plant} plants using {fuel}")
    print()
```

```
There are 17380 THERMAL plants using COAL
There are 986 THERMAL plants using LIGNITE
There are 4795 THERMAL plants using GAS
There are 137 THERMAL plants using MULTI FUEL
There are 1220 THERMAL plants using NATURAL GAS
There are 20 THERMAL plants using THERMAL
There are 1177 THERMAL plants using DIESEL
There are 267 THERMAL plants using HIGH SPEED DIESEL
There are 445 THERMAL plants using NAPTHA

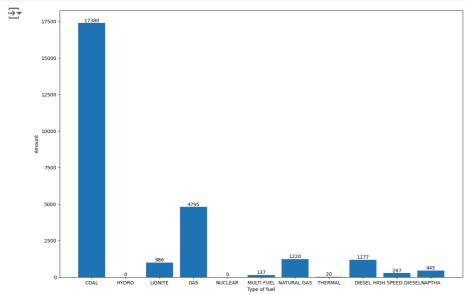
There are 19465 HYDRO plants using HYDRO
There are 783 NUCLEAR plants using NUCLEAR
```

Q4: Graph the various fuels used in a Coal Plant

```
x = fuels
y = [0 for i in range(len(fuels))]
for i, fuel in enumerate(list(fuels)):
    y[i] = plant_data['THERMAL'][fuel]
```

✓ A4:

```
fig = plt.figure(figsize=(15, 10))
plt.bar(x, y)
plt.xlabel("Type of fuel")
plt.ylabel("Amount")
for i, value in enumerate(y):
    plt.text(i, value + 1, str(value), ha='center', va='bottom', fontsize=10)
```



→ Q5: What is the plant distribution in the states of India

```
num_states = {state: len(df.loc[df['State'] == state]) for state in states }
print("State: Number of plants\n")
for state in sorted(num_states):
    print(f"{state}: {num_states[state]}")
```

→ State: Number of plants

Andaman And Nicobar Islands: 107

Andhra Pradesh: 1785 Arunachal Pradesh: 18

Assam: 666 Bihar: 486

Chhattisgarh: 2476 Delhi: 678 Goa: 107 Gujarat: 3443 Haryana: 690

Himachal Pradesh: 1776 Jammu Kashmir: 1103 Jharkhand: 1196 Karnataka: 2545 Kerala: 1819 Ladakh: 107

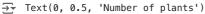
Ladakh: 107
Madhya Pradesh: 2218
Maharashtra: 4417
Manipur: 107
Meghalaya: 468
Mizoram: 3
New Delhi: 1
Odisha: 1643
Puducherry: 107
Punjab: 1471
Rajasthan: 1532
Sikkim: 412
Tamil Nadu: 4478
Telangana: 2191
Trinura: 827

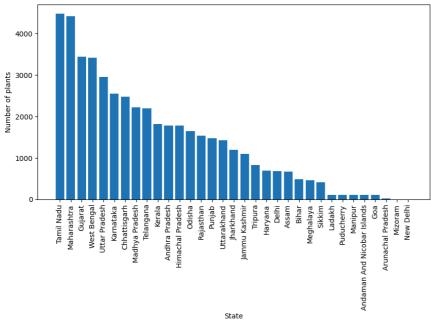
Tripura: 827 Uttar Pradesh: 2959 Uttarakhand: 1427 West Bengal: 3412

 $num_states = \{k: \ v \ for \ k, \ v \ in \ sorted(num_states.items(), \ key=lambda \ item: \ item[1], \ reverse=True)\}$

✓ A5:

```
fig = plt.figure(figsize=(10, 5))
plt.bar(num_states.keys(), num_states.values())
plt.xticks(rotation='vertical')
plt.xlabel("State")
plt.ylabel("Number of plants")
```





Q6: What is the distribution of the Thermal, Hydro and Nuclear plants in each state

```
state_data = {state: {plant: 0 for plant in plants} for state in states}

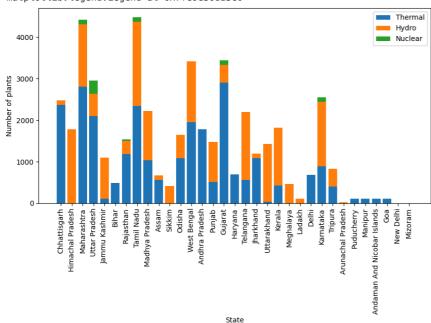
for state in states:
    for plant in plants:
        condition = (df['State'] == state) & (df['Category of Plant'] == plant)
        length = len(df.loc[condition])
        state_data[state][plant] = length
```

A6:

```
categories = list(state_data.keys())
thermal_data = [state_data[category]['THERMAL'] for category in categories]
hydro_data = [state_data[category]['HYDRO'] for category in categories]
nuclear_data = [state_data[category]['NUCLEAR'] for category in categories]

fig, ax = plt.subplots(figsize=(10, 5))
ax.bar(categories, thermal_data, label='Thermal')
ax.bar(categories, hydro_data, bottom=thermal_data, label='Hydro')
ax.bar(categories, nuclear_data, bottom=[x + y for x, y in zip(thermal_data, hydro_data)], label='Nuclear')
plt.xticks(rotation='vertical')
plt.xlabel("State")
plt.ylabel("Number of plants")
plt.legend()
```

<matplotlib.legend.Legend at 0x7fe9e36da3e0>



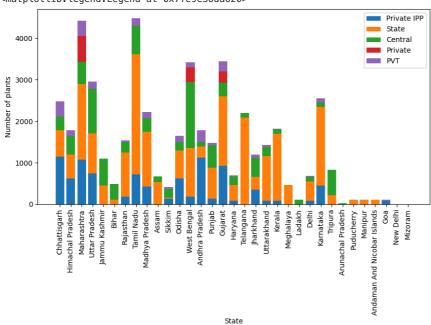
Q7: What's the distribution of the sector of power plants in the states

```
state_sector_data = {state: {sector: 0 for sector in sectors} for state in states}

for state in states:
    for sector in sectors:
        condition = (df['State'] == state) & (df['Sector of power plant'] == sector)
        length = len(df.loc[condition])
        state_sector_data[state][sector] = length
```

```
categories = list(state_sector_data.keys())
private_ipp = [state_sector_data[category]['PRIVATE IPP'] for category in categories]
state = [state_sector_data[category]['STATE'] for category in categories]
central = [state_sector_data[category]['CENTRAL'] for category in categories]
private = [state_sector_data[category]['PRIVATE'] for category in categories]
pvt = [state_sector_data[category]['PVT'] for category in categories]
fig, ax = plt.subplots(figsize=(10, 5))
ax.bar(categories, private_ipp, label='Private IPP')
ax.bar(categories, state, bottom=private_ipp, label='State')
ax.bar(categories, central, bottom=[x + y for x, y in zip(private_ipp, state)], label='Central')
ax.bar(categories,\ private,\ bottom=[x\ +\ y\ +\ z\ for\ x,\ y,\ z\ in\ zip(private\_ipp,\ state,\ central)],\ label='Private')
ax.bar(categories, pvt, bottom=[x + y + z + w for x, y, z, w in zip(private_ipp, state, central, private)], label='PVT')
plt.xticks(rotation='vertical')
plt.xlabel("State")
plt.ylabel("Number of plants")
plt.legend()
```

<matplotlib.legend.Legend at 0x7fe9e36da620>

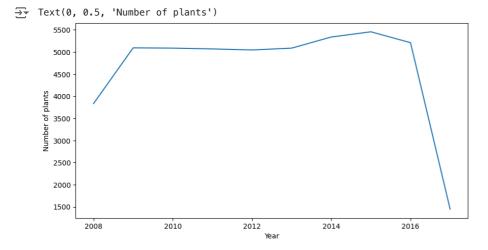


Q8: When were the power plants commissioned

```
year_data = {year: len(df.loc[df['YearCode'] == year]) for year in years}
year_data = dict(sorted(year_data.items(), key=lambda item: item[0]))
```

✓ A8:

```
x = list(year_data.keys())
y = list(year_data.values())
fig = plt.figure(figsize=(10, 5))
plt.plot(x, y)
plt.xlabel("Year")
plt.ylabel("Number of plants")
```



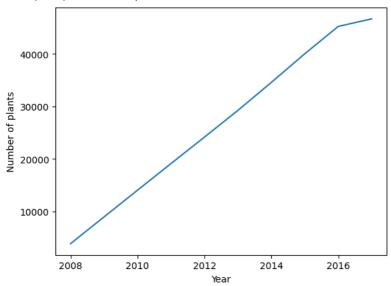
Q9: How many plants were running in that year

✓ A9:

```
def cumulative_sum(nums):
    return [sum(nums[:i+1]) for i in range(len(nums))]

y1 = cumulative_sum(y)
plt.plot(x, y1)
plt.xlabel("Year")
plt.ylabel("Number of plants")
```

→ Text(0, 0.5, 'Number of plants')



Q10: Which are the top coal companies

```
companies = df['Name of coal company'].unique()
company_data = {company: len(df.loc[df['Name of coal company'] == company]) for company in companies}
sorted_company_data = {k: v for k, v in sorted(company_data.items(), key=lambda item: item[1], reverse=True)}
```

✓ A10:

```
for i, company in enumerate(sorted_company_data):
    if i == 10:
        break
    print(f"{i+1}: {company} with {sorted_company_data[company]} coal plants")

1 : TNGDCL with 2889 coal plants
    2 : NTPC Ltd. with 2651 coal plants
    3 : NHPC with 2045 coal plants
    4 : APGENCO with 2002 coal plants
    5 : KPCL with 1894 coal plants
    6 : MAHAGENCO with 1819 coal plants
    7 : KSEB with 1605 coal plants
    8 : MPPGCL with 1320 coal plants
    9 : GSECL with 1100 coal plants
    10 : RRVUNL with 1070 coal plants
```

Q11: What is the amount of installed capacity of energy per state

```
installed_capacity = {state: 0 for state in states}
for state in states:
  installed_capacity[state] = df.loc[df['State'] == state]['Installed Capacity'].sum().round(2)
installed_capacity = {k: v for k, v in sorted(installed_capacity.items(), key=lambda item: item[1], reverse=True)}
```

✓ A11:

```
fig = plt.figure(figsize=(10, 5))
plt.bar(list(installed_capacity.keys()), list(installed_capacity.values()))
plt.xlabel("State")
plt.ylabel("Installed Capacity")
plt.xticks(rotation='vertical')
```

```
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Text(0, 0, 'Maharashtra'),
Text(1, 0, 'Gujarat'),
                  'Uttar Pradesh'),
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 Text(3, 0,
                   'West Bengal'),
                  'Tamil Nadu'),
 Text(4, 0,
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                  'Madhya Pradesh'),
                  'Chhattisgarh'),
 Text(6, 0,
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Text(8, 0, 'Andhra Prade
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                    'Punjab'),
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                    'Himachal Pradesh'),
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                    'Uttarakhand'),
                    'Kerala'),
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'Tripura'),
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                    'Assam'),
'Sikkim'),
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 Text(23, 0,
 Text(24, 0,
                    'Meghalaya'),
 Text(25, 0,
                     'Andaman And Nicobar Islands'),
 Text(26, 0,
                    'Goa'),
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                    'Manipur'),
 Text(28, 0,
                    'Puducherry'),
                    'Ladakh'),
 Text(29, 0,
 Text(30, 0, 'Arunachal Pradesh'),
 Text(31, 0, 'New Delhi'),
Text(32, 0, 'Mizoram')])
Installed Capacity
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                                  Karnataka -
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Uttar Pradesh -
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Punjab -
Rajasthan -
                     West Bengal -
Tamil Nadu -
                                                     Haryana -
Jharkhand -
                                                                        Kerala -
                           Madhya Pradesh
                                                              Bihar
                               Chhattisgarh
                                                           Himachal Pradesh
                                                                 Jammu Kashmir
                                                                    Uttarakhand
                                                                                             Goa
```

- Q2: Seeing high levels of pollution in the country, you get curious about the trends in pollution and try to correlate it with the dataset for power plants you are given. How would you go about doing
- that? Is the current dataset sufficient to identify pollution trends? If not, what additional data would you need, and where would you obtain it? List the potential sources for acquiring this necessary data.
 - · Pollution is caused by various sources like combustion devices, motor vehicles, industrial facilites, forest fires, etc.
 - But if we look into the distribution of pollution caused by these sources we find that **75.80% of green house gas emmisions is due to Energy Production**.
 - Source https://www.downtoearth.org.in/blog/pollution/5-rise-in-india-s-ghg-emissions-since-2016-driven-by-energy-industrial-sectors-94076
 - · So it is fair to approximate the trends in pollution to correspond to the amount of emissions released by these energy plants.
 - The current dataset can give us an approximate of the amount of GHG emissions which correlates to the amount of pollution in India.
 - · Although if additional data about IPPU, Agriculture, Waste and Vehicle emissions is given, we can formulate a better approximation.

The amount of GHG emissions produced by the energy plants are:

- Thermal Plants 584 g CO2/kWh
- Hydro Plants 23 g CO2/kWh
- Nuclear Plants 4 g CO2/kWh

Sources

- <a href="https://teriin.org/index.php/research-paper/assessment-greenhouse-gas-emissions-coal-and-natural-gas-thermal-power-plants-using#:~:text=The%20total%20GHG%20emission%20from,CO2%20eg%2FkWh%20electricity%20generation
- https://www.hydropower.org/factsheets/greenhouse-gas-emissions#:~:text=The%20results%20published%20in%20Water,consistent%20with%20the%20IPCC%20findings
- https://www.carbonbrief.org/solar-wind-nuclear-amazingly-low-carbon-footprints/
- Let's calculate the distibution of GHG emmisions by the States of India

```
def assign_ghg_emission(category):
    if category == 'THERMAL':
        return 584
    elif category == 'HYDRO':
        return 23
    else:
        return 4
df['GHG Emission'] = df['Category of Plant'].apply(assign_ghg_emission)
df.head(5)
```

•		ROWID	Country	State LGD Code	State	Actual energy generated	Category of Plant	Type of fuel used	Installed Capacity	Ge P
	0	1	India	22	Chhattisgarh	0.0	THERMAL	COAL	0.0	
	1	2	India	2	Himachal Pradesh	0.0	HYDRO	HYDRO	0.0	
	2	3	India	27	Maharashtra	0.0	THERMAL	COAL	0.0	
	3	4	India	9	Uttar Pradesh	0.0	THERMAL	COAL	0.0	
	4	5	India	22	Chhattisgarh	0.0	THERMAL	COAL	0.0	

5 rows x 21 columns

```
emission_states = {state: df.loc[df['State'] == state]['GHG Emission'].sum() for state in states}
emission_states = {k: v for k, v in sorted(emission_states.items(), key=lambda item: item[1], reverse=True)}
```

for emission_state in emission_states:
 print(f"{emission_state} produces {emission_states[emission_state]} g CO2/kWh")

→ Gujarat produces 1708544 g CO2/kWh Maharashtra produces 1677090 g CO2/kWh Tamil Nadu produces 1412579 g CO2/kWh Chhattisgarh produces 1385957 g CO2/kWh Uttar Pradesh produces 1241741 g CO2/kWh West Bengal produces 1174670 g CO2/kWh Andhra Pradesh produces 1042440 g C02/kWh Rajasthan produces 694887 g CO2/kWh Odisha produces 644791 g CO2/kWh Jharkhand produces 638437 g CO2/kWh Madhya Pradesh produces 635015 g CO2/kWh Karnataka produces 554670 g CO2/kWh Haryana produces 402960 g CO2/kWh Delhi produces 395952 g CO2/kWh Telangana produces 362870 g CO2/kWh Assam produces 328917 g CO2/kWh Punjab produces 318821 g CO2/kWh Bihar produces 283824 g CO2/kWh Kerala produces 281945 g CO2/kWh Tripura produces 245104 g CO2/kWh Jammu Kashmir produces 85396 g CO2/kWh Puducherry produces 62488 g CO2/kWh Manipur produces 62488 g CO2/kWh Andaman And Nicobar Islands produces 62488 g CO2/kWh Goa produces 62488 g CO2/kWh Uttarakhand produces 53017 g CO2/kWh Himachal Pradesh produces 40848 g CO2/kWh Meghalaya produces 10764 g CO2/kWh Sikkim produces 9476 g CO2/kWh Ladakh produces 2461 g CO2/kWh New Delhi produces 584 g CO2/kWh Arunachal Pradesh produces 414 g CO2/kWh Mizoram produces 69 g CO2/kWh

```
fig = plt.figure(figsize=(10, 5))
plt.bar(list(emission_states.keys()), list(emission_states.values()))
plt.xlabel("State")
plt.ylabel("GHG Emission")
plt.xticks(rotation='vertical')
```

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Text(0, 0, 'Gujarat'),
Text(1, 0, 'Maharashtra'),
 Text(2, 0, 'Tamil Nadu'),
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                   'Manipur'),
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 Text(28, 0,
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  Text(29, 0,
                   'Ladakh'),
 Text(30, 0,
                   'New Delhi'),
 Text(31, 0, 'Arunachal Pradesh'),
Text(32, 0, 'Mizoram')])
  1.6
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GHG Emission
   1.0
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   0.6
   0.4
   0.2
   0.0
                                                                                        Meghalaya – Sikkim – Ladakh – New Delhi – Arunachal Pradesh – Mizoram –
                               Rajasthan
Odisha
Darkhand
Madhya Pradesh
Karnataka
Haryana
Delhi
Telangana
Assam
Punjab
Bihar
Rerala
                    Chhattisgarh -
Uttar Pradesh -
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Himachal Pradesh -
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Andaman And Nicobar Islands
                                                                     Jammu Kashmir
                                                                                 Goa
                                                                        Puducherry
```

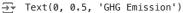
State

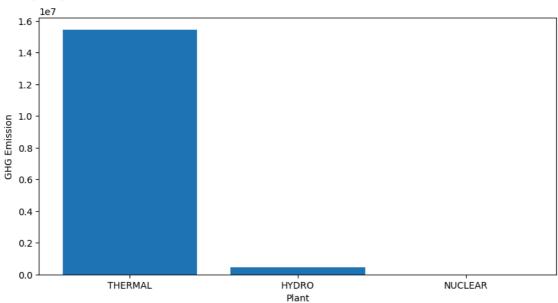
Now let's see which source of energy is worse for the country and by how much

```
emission_plant = {plant: df.loc[df['Category of Plant'] == plant]['GHG Emission'].sum() for plant in plants}
emission_plant
```

→ {'THERMAL': 15433368, 'HYDRO': 447695, 'NUCLEAR': 3132}

```
fig = plt.figure(figsize=(10, 5))
plt.bar(list(emission_plant.keys()), list(emission_plant.values()))
plt.xlabel("Plant")
plt.ylabel("GHG Emission")
```





emission_plant['THERMAL'] / (emission_plant['THERMAL'] + emission_plant['HYDRO'] + emission_plant['NUCLEAR']) * 100

→ 97.16178880956825

✓ Conclusion

- 1. The top five states which produce the most pollution are Gujrat, Maharashtra, Tamil Nadu, Chattisgarh and Uttar Pradesh
- 2. Clearly from the bar graph we can see that the production of energy from thermal plants creates a lot more pollution than the production of energy from hydro or nuclear sources. To be precise, it releases 97.16% of the total emissions released by energy production sources

DAV Assignment Q3 and Q4

After extensive research, suppose you decided to enter the Indian Market with your existing products (P, Q, R, S and T). Your team has determined that the behavior of the new market in India closely mirrors that of your existing market. In your current market, the sales team categorizes all customers into four segments (A, B, C, D). They then tailor their outreach and communication strategies to each specific segment. This approach has proven to be highly effective for them.

They plan to use the same strategy for the Indian markets and have identified about 2500 new potential trial customers. For this they have an existing database of about 8000 customers which they have already classified.

Q3: Implement methods to assist your team in identifying and predicting the appropriate group (A, B, C, or D) for each customer in the test data.

```
# Import libraries
import requests
from pathlib import Path
import zipfile
import os
request = requests.get("https://github.com/PanavShah1/DAV_assignment/blob/main/Consumer_Dataset.csv?raw=true")
with open("Consumer_Dataset.csv", "wb") as f:
  f.write(request.content)
  print("csv file downloaded")
request = requests.get("https://github.com/PanavShah1/DAV_assignment/blob/main/Consumer%20Test%20Dataset.csv?raw=true")
with open("Consumer_Test_Dataset.csv", "wb") as f:
  f.write(request.content)
  print("csv file downloaded")
    csv file downloaded
     csv file downloaded
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
df = pd.read_csv("Consumer_Dataset.csv")
df.head(5)
₹
        Unnamed:
                  Gender Age Ever_Married Family_Size Profession Graduated Wor
     0
               0
                     Male
                                          No
                                                       4.0
                                                              Healthcare
                                                                               No
     1
                           38
                                                       3.0
                                                               Engineer
                                                                               Yes
               1
                   Female
                                         Yes
     2
               2
                   Female
                                         Yes
                                                       1.0
                                                               Engineer
                                                                               Yes
     3
               3
                     Male
                           67
                                         Yes
                                                       2.0
                                                                Lawyer
                                                                               Yes
      4
                   Female
                            40
                                         Yes
                                                       6.0 Entertainment
                                                                               Yes
 Next steps:
             Generate code with df
                                    View recommended plots
```

Get the dataset ready

```
professions = df['Profession'].unique()
renewable = df['Preferred_Renewable'].unique()
avg_work_experience = df['Work_Experience'].mean()
avg_family_size = df['Family_Size'].mean()
```

```
df = df.replace(to_replace="Female", value=1)
df = df.replace(to_replace="Yes", value=1)
df = df.replace(to_replace="No", value=0)
df['Work_Experience'] = df['Work_Experience'].fillna(avg_work_experience)
df = df.replace(to_replace="Low", value=0)
df = df.replace(to_replace="Average", value=1)
df = df.replace(to_replace="High", value=2)
df['Family_Size'] = df['Family_Size'].fillna(avg_family_size)
df = df.replace(to_replace="A", value=0)
df = df.replace(to_replace="B", value=1)
df = df.replace(to_replace="C", value=2)
df = df.replace(to_replace="D", value=3)
df.head(5)
\rightarrow
          Unnamed:
                     Gender Age Ever_Married Family_Size Profession Graduated Wor
      0
                  0
                            0
                                22
                                                 0.0
                                                                 40
                                                                         Healthcare
                                                                                             0.0
                                                 1.0
                                                                 3.0
                                                                           Engineer
                                                                                             1.0
      2
                  2
                                                                           Engineer
                            1
                                67
                                                 1.0
                                                                 1 0
                                                                                             1.0
      3
                  3
                            0
                                67
                                                 1.0
                                                                 2.0
                                                                            Lawyer
                                                                                             1.0
       4
                  4
                                 40
                                                 1.0
                                                                 6.0 Entertainment
                                                                                             1.0
 Next steps: Generate code with df
                                           View recommended plots
for i, source in enumerate(professions):
  df['Profession'][df['Profession'] == source] = i
avg_profession = df['Profession'].mean()
df['Profession'] = df['Profession'].fillna(avg_profession)
avg_profession
<ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-df">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-df</a>['Profession'] [df['Profession'] == source] = i
      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
        df['Profession'][df['Profession'] == source] = i
      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame
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      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
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        df['Profession'][df['Profession'] == source] = i
      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: \frac{https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html \#returning-a-df['Profession'][df['Profession'] == source] = i
      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame
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      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
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      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame
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        df['Profession'][df['Profession'] == source] = i
      <ipython-input-248-2b2318068495>:2: SettingWithCopyWarning:
      A value is trying to be set on a copy of a slice from a DataFrame
      See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
        df['Profession'][df['Profession'] == source] = i
      3.2775679758308156
```

df = df.replace(to_replace="Male", value=0)

```
avg_married = df['Ever_Married'].mean()
df['Ever_Married'] = df['Ever_Married'].fillna(avg_married)
avg married
→ 0.5856458123107972
for i. source in enumerate(renewable):
  df['Preferred_Renewable'][df['Preferred_Renewable'] == source] = i
avg_preferred_renewable = df['Preferred_Renewable'].mean()
df['Preferred_Renewable'] = df['Preferred_Renewable'].fillna(avg_preferred_renewable)
avg_preferred_renewable
    <ipython-input-250-cd5c1d51057a>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df['Preferred_Renewable'][df['Preferred_Renewable'] == source] = i
     <ipython-input-250-cd5c1d51057a>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df['Preferred_Renewable'][df['Preferred_Renewable'] == source] = i
     <ipython-input-250-cd5c1d51057a>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-df">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-df</a>['Preferred_Renewable'] == source] = i
     <ipython-input-250-cd5c1d51057a>:2: SettingWithCopyWarning:
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     <ipython-input-250-cd5c1d51057a>:2: SettingWithCopyWarning:
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     See the caveats in the documentation: \frac{https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html \#returning-adf['Preferred\_Renewable'] == source] = i
     <ipython-input-250-cd5c1d51057a>:2: SettingWithCopyWarning:
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       df['Preferred_Renewable'][df['Preferred_Renewable'] == source] = i
     <ipython-input-250-cd5c1d51057a>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: \frac{https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html \#returning-adf['Preferred\_Renewable'] == source] = i
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     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df['Preferred_Renewable'][df['Preferred_Renewable'] == source] = i
     1.4197947947947949
avg_graduated = df['Graduated'].mean()
df['Graduated'] = df['Graduated'].fillna(avg_graduated)
avg_graduated
→ 0.6217772215269086
df = df.drop('Unnamed: 0', axis=1)
df.head(5)
\overline{2}
         Gender Age Ever Married Family Size Profession Graduated Work Experien
      0
                    22
                                    0.0
                                                   4.0
                                                                 0.0
                                                                              0.0
                                                                                             1.0000
                                                                                             2.6416
      1
               1
                   38
                                   1.0
                                                   3.0
                                                                 1.0
                                                                              1.0
                    67
                                    1.0
                                                   1.0
                                                                 1.0
                                                                              1.0
                                                                                             1.0000
               1
      3
               0
                   67
                                    1.0
                                                   20
                                                                 20
                                                                              1.0
                                                                                             0.0000
                                    1.0
                                                   6.0
                                                                 3.0
                                                                              1.0
                                                                                             2.6416
                   40
 Next steps: Generate code with df
                                         View recommended plots
data = df.to_numpy()
→ array([[ 0., 22., 0., ..., 0., 0., 3.],
```

[1., 38., 1., ..., 1., 0., 0.], [1., 67., 1., ..., 0., 1., 1.],

```
[ 1., 33., 0., ..., 0., 1., 3.],
[ 1., 27., 0., ..., 0., 1., 1.],
            [ 0., 37., 1., ..., 1., 0., 1.]])
data.shape

→ (8068, 10)

X = data[:, 0:9]
y = data[:, 9]
X.shape, y.shape
→ ((8068, 9), (8068,))
import torch
from torch import nn
device = "cuda" if torch.cuda.is_available() else "cpu"
device
→ 'cpu'
X = torch.from_numpy(X)
y = torch.from_numpy(y)
X.shape, v.shape
(torch.Size([8068, 9]), torch.Size([8068]))
X = X.type(torch.float32)
y = y.type(torch.long)
X.dtype, y.dtype
→ (torch.float32, torch.int64)
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.1, random_state=42)
X_train.shape, X_test.shape, y_train.shape, y_test.shape
→ (torch.Size([7261, 9]),
     torch.Size([807, 9]),
      torch.Size([7261]),
      torch.Size([807]))
from torch.utils.data import Dataset
class ConsumerDataset(Dataset):
  def __init__(self, X, y):
    self_X = X
    self.y = y
  def __len__(self):
    return len(self.X)
  def __getitem__(self, idx):
    return self.X[idx], self.y[idx]
train_dataset = ConsumerDataset(X_train, y_train)
test_dataset = ConsumerDataset(X_test, y_test)
train_dataset, test_dataset
    (<__main__.ConsumerDataset at 0x78e875a6ffd0>,
      <__main__.ConsumerDataset at 0x78e875a6d7b0>)
from torch.utils.data import DataLoader
BATCH_SIZE = 16
train_dataloader = DataLoader(dataset=train_dataset,
                              batch_size=BATCH_SIZE,
                              shuffle=True)
test_dataloader = DataLoader(dataset=test_dataset,
                             batch_size=BATCH_SIZE,
                              shuffle=False)
train_dataloader, test_dataloader
(<torch.utils.data.dataloader.DataLoader at 0x78e875a6d750>,
```

<torch.utils.data.dataloader.DataLoader at 0x78e875a6f5e0>)

```
from torch import nn
class ModelV0(nn.Module):
 def __init__(self,
               input_shape: int,
               hidden_layers: int,
              output_shape: int = 4):
    super().__init__()
    self.layer_stack = nn.Sequential(
       nn.Linear(in_features=input_shape,
                 out_features=hidden_layers),
        nn.ReLU(),
       nn.Linear(in_features=hidden_layers,
                  out_features=hidden_layers),
       # nn.ReLU(),
       # nn.Linear(in_features=hidden_layers,
                    out_features=hidden_layers),
       nn.ReLU(),
       nn.Linear(in_features=hidden_layers,
                  out_features=output_shape)
   )
 def forward(self, x):
    return self.layer_stack(x)
model_0 = ModelV0(input_shape=X.shape[1], hidden_layers=30, output_shape=4)
model_0
→ ModelV0(
      (layer_stack): Sequential(
        (0): Linear(in_features=9, out_features=30, bias=True)
         (1): ReLU()
        (2): Linear(in_features=30, out_features=30, bias=True)
        (3): ReLU()
        (4): Linear(in_features=30, out_features=4, bias=True)
model_0.state_dict().keys()

→ odict_keys(['layer_stack.0.weight', 'layer_stack.0.bias', 'layer_stack.2.weight', 'layer_stack.2.bias',
     'layer_stack.4.weight', 'layer_stack.4.bias'])
def train_step(model: torch.nn.Module,
               dataloader: torch.utils.data.DataLoader,
               loss_fn: torch.nn.Module,
               optimizer: torch.optim.Optimizer,
              device=device):
 net_loss, net_acc = 0, 0
 model.train()
  for batch, (X, y) in enumerate(dataloader):
   X, y = X.to(device), y.to(device)
   y_pred = model(X)
   loss = loss_fn(y_pred, y)
   net_loss += loss
   net_acc += (y_pred.argmax(dim=1) == y).sum().item() / len(y)
   optimizer.zero_grad()
   loss.backward()
   optimizer.step()
 net_loss /= len(dataloader)
 net_acc /= len(dataloader)
 return net_loss, net_acc
```

```
def test_step(model: torch.nn.Module,
              dataloader: torch.utils.data.DataLoader,
              loss fn: torch.nn.Module,
              device=device):
  net_loss, net_acc = 0, 0
  model.eval()
  with torch.inference_mode():
    for batch, (X, y) in enumerate(dataloader):
      X, y = X.to(device), y.to(device)
      y_pred = model(X)
      net_loss += loss_fn(y_pred, y)
      net_acc += (y_pred.argmax(dim=1) == y).sum().item() / len(y)
    net_loss = net_loss / len(dataloader)
    net_acc = net_acc / len(dataloader)
    return net_loss, net_acc
from tqdm.auto import tqdm
def train(model: torch.nn.Module,
          train_dataloader: torch.utils.data.DataLoader,
          test_dataloader: torch.utils.data.DataLoader,
          optimizer: torch.optim.Optimizer,
          loss_fn: torch.nn.Module,
          epochs: int = 5,
          device = device):
  results = {"train_loss": [],
             "train_acc": [],
             "test_loss": [],
             "test_acc": []}
  for epoch in range(epochs):
    train_loss, train_acc = train_step(model=model,
                                       dataloader=train_dataloader,
                                       loss_fn=loss_fn,
                                       optimizer=optimizer,
                                       device=device)
    test_loss, test_acc = test_step(model=model,
                                     dataloader=test_dataloader,
                                     loss_fn=loss_fn,
                                     device=device)
    print(f"Epoch: {epoch} | Train loss: {train_loss:.4f} | Train acc: {train_acc:.4f} | Test loss: {test_loss:4f} | Test acc
    results["train_loss"].append(train_loss)
    results["train_acc"].append(train_acc)
    results["test_loss"].append(test_loss)
    results["test_acc"].append(test_acc)
  return results
NUM EPOCHS = 50
model_0 = ModelV0(input_shape=X.shape[1],
                  hidden layers=100,
                  output_shape=4)
loss_fn = nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model_0.parameters(), lr=0.001)
results = train(model=model 0,
                train_dataloader=train_dataloader,
                test_dataloader=test_dataloader,
                 optimizer=optimizer,
                 loss_fn=loss_fn,
                epochs=NUM_EPOCHS)

    Epoch: 0 | Train loss: 1.3058 | Train acc: 0.3889 | Test loss: 1.228549 | Test acc: 0.4039

     Epoch: 1 | Train loss: 1.2180 | Train acc: 0.4496 | Test loss: 1.174514 | Test acc: 0.4680
     Epoch: 2 |
               Train loss: 1.1847 | Train acc: 0.4659 |
                                                          Test loss: 1.232061 | Test acc: 0.4597
     Epoch: 3 |
               Train loss: 1.1680
                                    | Train acc: 0.4749
                                                           Test loss: 1.142212 |
                                                                                  Test acc: 0.4762
     Epoch: 4 | Train loss: 1.1550 | Train acc: 0.4782 |
                                                           Test loss: 1.124821 | Test acc: 0.4977
     Epoch: 5 |
               Train loss: 1.1469
                                    | Train acc: 0.4827
                                                           Test loss: 1.133963 |
                                                                                  Test acc: 0.4876
     Epoch: 6 | Train loss: 1.1410 | Train acc: 0.4898 |
                                                           Test loss: 1.125877 |
                                                                                  Test acc: 0.4912
     Epoch: 7 I
               Train loss: 1.1301 | Train acc: 0.4887
                                                           Test loss: 1.136982 |
                                                                                  Test acc: 0.4944
     Epoch: 8 | Train loss: 1.1274 | Train acc: 0.4910
                                                           Test loss: 1.137922
                                                                                  Test acc: 0.4940
     Epoch: 9 | Train loss: 1.1185 | Train acc: 0.4987 | Test loss: 1.101650 | Test acc: 0.4998
     Epoch: 10 | Train loss: 1.1084 | Train acc: 0.4958 | Test loss: 1.101140 | Test acc: 0.5026
     Epoch: 11 | Train loss: 1.1036 | Train acc: 0.4976 | Test loss: 1.101348 | Test acc: 0.5039
Epoch: 12 | Train loss: 1.0972 | Train acc: 0.5048 | Test loss: 1.124134 | Test acc: 0.4891
```

```
Epoch: 16
                Train loss: 1.0781
                                     Train acc: 0.5163
                                                          Test loss: 1.086476 |
                                                                                Test acc: 0.5030
    Epoch: 17
                Train loss: 1.0751
                                     Train acc: 0.5101
                                                          Test loss: 1.099423 |
                                                                                Test acc: 0.4928
    Epoch: 18
                Train loss: 1.0691
                                     Train acc: 0.5150 |
                                                          Test loss: 1.092672 |
                                                                                Test acc: 0.4977
                                                          Test loss: 1.071761 |
    Epoch: 19
                Train loss: 1.0696
                                     Train acc: 0.5176
                                                                                Test acc: 0.5054
    Epoch: 20
                Train loss: 1.0632
                                    | Train acc: 0.5187
                                                          Test loss: 1.094769 | Test acc: 0.4974
                                                          Test loss: 1.080727
    Epoch: 21
                Train loss: 1.0669
                                     Train acc: 0.5199
                                                                                Test acc: 0.5014
    Epoch: 22
                Train loss: 1.0603
                                    | Train acc: 0.5200
                                                          Test loss: 1.087723 | Test acc: 0.5124
    Epoch: 23
                Train loss: 1.0586
                                     Train acc: 0.5246 |
                                                          Test loss: 1.086855 |
                                                                                Test acc: 0.5348
    Epoch: 24
                Train loss: 1.0550
                                     Train acc: 0.5277
                                                          Test loss: 1.095753 |
                                                                               Test acc: 0.5201
                                     Train acc: 0.5294 |
                                                          Test loss: 1.089631 |
    Epoch: 25
                Train loss: 1.0513 |
                                                                                Test acc: 0.5014
    Epoch: 26
                Train loss: 1.0512
                                     Train acc: 0.5304
                                                          Test loss: 1.075911 |
                                                                                Test acc: 0.5226
    Epoch: 27
                Train loss: 1.0476
                                     Train acc: 0.5259
                                                          Test loss: 1.085704 |
                                                                                Test acc: 0.5161
    Epoch: 28
                Train loss: 1.0425
                                     Train acc: 0.5317
                                                          Test loss: 1.102374 |
                                                                                Test acc: 0.5312
    Epoch: 29
                Train loss: 1.0424 |
                                     Train acc: 0.5391
                                                          Test loss: 1.106436 | Test acc: 0.4928
    Epoch: 30
                Train loss: 1.0411
                                     Train acc: 0.5322
                                                          Test loss: 1.118480 |
                                                                                Test acc: 0.5103
    Epoch: 31
                Train loss: 1.0395 | Train acc: 0.5324 |
                                                          Test loss: 1.087901 | Test acc: 0.5112
                Train loss: 1.0375
                                                          Test loss: 1.106410 |
    Epoch: 32
                                     Train acc: 0.5314
                                                                                Test acc: 0.5124
    Epoch: 33
                Train loss: 1.0352
                                   | Train acc: 0.5398 |
                                                         Test loss: 1.112103 | Test acc: 0.4916
                Train loss: 1.0321 |
                                     Train acc: 0.5374 |
                                                          Test loss: 1.098401 |
                                                                                Test acc: 0.5140
    Epoch: 34
                Train loss: 1.0276 |
                                                          Test loss: 1.108210 |
    Epoch: 35
                                     Train acc: 0.5440 |
                                                                               Test acc: 0.4919
    Epoch: 36 |
                Train loss: 1.0277 |
                                     Train acc: 0.5431 |
                                                          Test loss: 1.113733 |
                                                                               Test acc: 0.5165
    Epoch: 37
                Train loss: 1.0249
                                     Train acc: 0.5399
                                                          Test loss: 1.101434 |
                                                                                Test acc: 0.5088
    Epoch: 38
                Train loss: 1.0232
                                     Train acc: 0.5402
                                                          Test loss: 1.106688
                                                                                Test acc: 0.4981
    Epoch: 39
                Train loss: 1.0215
                                     Train acc: 0.5462
                                                          Test loss: 1.100720 |
                                                                                Test acc: 0.5385
                Train loss: 1.0195 | Train acc: 0.5437
    Epoch: 40
                                                          Test loss: 1.112374 | Test acc: 0.5137
    Epoch: 41
                Train loss: 1.0167
                                     Train acc: 0.5421
                                                          Test loss: 1.108079
                                                                                Test acc: 0.5137
    Epoch: 42
                Train loss: 1.0140 | Train acc: 0.5505
                                                          Test loss: 1.120486 | Test acc: 0.5112
    Epoch: 43
                Train loss: 1.0124
                                     Train acc: 0.5527
                                                          Test loss: 1.143582 |
                                                                                Test acc: 0.4977
    Epoch: 44
                Train loss: 1.0134 |
                                                         Test loss: 1.133351 | Test acc: 0.4876
                                     Train acc: 0.5418 |
    Epoch: 45
                Train loss: 1.0069
                                     Train acc: 0.5533 |
                                                          Test loss: 1.102942 |
                                                                                Test acc: 0.5152
    Epoch: 46
                Train loss: 1.0048
                                     Train acc: 0.5490
                                                          Test loss: 1.119493 |
                                                                                Test acc: 0.5075
                Train loss: 1.0022 |
    Fnoch: 47
                                     Train acc: 0.5517 |
                                                          Test loss: 1.124027 |
                                                                                Test acc: 0.5116
    Epoch: 48
                Train loss: 0.9985
                                     Train acc: 0.5546
                                                          Test loss: 1.111841 |
                                                                                Test acc: 0.5189
    Epoch: 49 | Train loss: 0.9993 | Train acc: 0.5532 | Test loss: 1.123729 | Test acc: 0.5075
from sklearn import tree
clf = tree.DecisionTreeClassifier()
clf = clf.fit(X_train, y_train)
clf.score(X_test, y_test)
→ 0.4510532837670384
from sklearn.model_selection import GridSearchCV
param grid = {
    'max_depth': [None, 10, 20, 30, 40, 50],
    'min_samples_split': [2, 10, 20],
    'min_samples_leaf': [1, 5, 10],
    'max_features': [None, 'sqrt', 'log2']
grid_search = GridSearchCV(estimator=tree.DecisionTreeClassifier(), param_grid=param_grid, cv=5, n_jobs=-1)
grid_search.fit(X_train, y_train)
print(f"Best parameters found: {grid_search.best_params_}")
best model = grid search.best estimator
best_model.score(X_test, y_test)
    Best parameters found: {'max_depth': 10, 'max_features': 'sqrt', 'min_samples_leaf': 10, 'min_samples_split': 20}
    0.48079306071871125
from sklearn.ensemble import RandomForestClassifier
rf_clf = RandomForestClassifier(n_estimators=100, random_state=42)
rf_clf.fit(X_train, y_train)
rf_clf.score(X_test, y_test)
→ 0.46964064436183395
from sklearn.model_selection import cross_val_score
cross_val_scores = cross_val_score(tree.DecisionTreeClassifier(), X_train, y_train, cv=5)
print(f"Cross-validation scores: {cross_val_scores}")
print(f"Average cross-validation score: {cross_val_scores.mean()}")
    Cross-validation scores: [0.44872677 0.45247934 0.43801653 0.44214876 0.43181818]
    Average cross-validation score: 0.44263791642256256
```

Epoch: 13 | Train loss: 1.0945 | Train acc: 0.5012 | Test loss: 1.103670 | Test acc: 0.5100

Train acc: 0.5048 |

Test loss: 1.081837 |

Test loss: 1.095448 | Test acc: 0.5084

Test acc: 0.5165

Train loss: 1.0866 | Train acc: 0.5077 |

Train loss: 1.0850

Epoch: 14

Epoch: 15

}

```
clf = tree.DecisionTreeClassifier(class_weight='balanced')
clf.fit(X_train, y_train)
clf.score(X_test, y_test)
→ 0.44114002478314746
from sklearn.linear_model import LogisticRegression
lr = LogisticRegression()
lr.fit(X_train, y_train)
lr.score(X_test, y_test)
🚁 /usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: lbfgs failed to conve
    STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
    Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
      n_iter_i = _check_optimize_result(
    0.4510532837670384
```

Make predictions

Using the manually created linear regression model

```
df1 = pd.read_csv("Consumer_Test_Dataset.csv")
df2 = df1.copy()
df1.head(5)
```

→ *		Unnamed: 0	Gender	Age	Ever_Married	Family_Size	Profession	Graduated	Wor
	0	0	Female	36	Yes	1.0	Engineer	Yes	
	1	1	Male	37	Yes	4.0	Healthcare	Yes	
	2	2	Female	69	Yes	1.0	NaN	No	
	3	3	Male	59	Yes	2.0	Executive	No	
	4	4	Female	19	No	4.0	Marketing	No	

Next steps: Generate code with df1

• View recommended plots

```
df1 = df1.replace(to_replace="Male", value=0)
df1 = df1.replace(to_replace="Female", value=1)
df1 = df1.replace(to_replace="Yes", value=1)
df1 = df1.replace(to_replace="No", value=0)
df1['Work_Experience'] = df1['Work_Experience'].fillna(avg_work_experience)
df1 = df1.replace(to_replace="Low", value=0)
df1 = df1.replace(to_replace="Average", value=1)
df1 = df1.replace(to_replace="High", value=2)
df1['Family_Size'] = df1['Family_Size'].fillna(avg_family_size)

df1 = df1.replace(to_replace="A", value=0)
df1 = df1.replace(to_replace="B", value=1)
df1 = df1.replace(to_replace="C", value=2)
df1 = df1.replace(to_replace="C", value=3)
df1
```

```
Unnamed:
                       Gender Age Ever Married Family Size Profession Graduated
       0
                    0
                                                                        Engineer
                                                                                         1.0
                                 36
                                                 1.0
                                                               1.0
                                                                      Healthcare
                             0
                                 37
                                                 1.0
                                                               4.0
                                                                                         1.0
       1
                    1
       2
                    2
                                 69
                                                 1.0
                                                               1.0
                                                                            NaN
                                                                                         0.0
       3
                    3
                             0
                                 59
                                                 1.0
                                                               20
                                                                        Executive
                                                                                         0.0
       4
                    4
                                 19
                                                 0.0
                                                               4.0
                                                                        Marketing
                                                                                         0.0
      2622
                 2622
                                 29
                                                 0.0
                                                               4.0
                                                                      Healthcare
                                                                                         0.0
                             0
      2623
                 2623
                                 35
                                                 0.0
                                                               1.0
                                                                                          1.0
                                                                          Doctor
      2624
                 2624
                                 53
                                                 0.0
                                                               2.0 Entertainment
                                                                                         1.0
      2625
                 2625
                                                 1.0
                                                               5.0
                                                                        Executive
                                                                                          1.0
     2626
                 2626
                                 43
                                                 0.0
                                                               3.0
                                                                      Healthcare
                                                                                         1.0
     2627 rows × 10 columns
 Next steps:
             Generate code with df1
                                         View recommended plots
for i, source in enumerate(professions):
  df1['Profession'][df1['Profession'] == source] = i
avg_profession = df1['Profession'].mean()
df1['Profession'] = df1['Profession'].fillna(avg_profession)
avg_profession
    <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
    See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-df1['Profession'] [df1['Profession'] == source] = i</a>
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Profession'][df1['Profession'] == source] = i
     <ipython-input-279-e4db59ec5021>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Profession'][df1['Profession'] == source] = i
     3.3151796060254926
```

 \rightarrow

```
avg_married = df1['Ever_Married'].mean()
df1['Ever_Married'] = df1['Ever_Married'].fillna(avg_married)
avg married
→ 0.5898331393092744
for i, source in enumerate(renewable):
  df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i
avg_preferred_renewable = df1['Preferred_Renewable'].mean()
df1['Preferred_Renewable'] = df1['Preferred_Renewable'].fillna(avg_preferred_renewable)
avg_preferred_renewable
<ipython-input-281-880a2e04fb0b>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i
     <ipython-input-281-880a2e04fb0b>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
    See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i</a>
     <ipython-input-281-880a2e04fb0b>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i
     <ipython-input-281-880a2e04fb0b>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i
     <ipython-input-281-880a2e04fb0b>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i
     <ipython-input-281-880a2e04fb0b>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i
     <ipvthon-input-281-880a2e04fb0b>:2: SettingWithCopvWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-</a>
       df1['Preferred_Renewable'][df1['Preferred_Renewable'] == source] = i
     <ipython-input-281-880a2e04fb0b>:2: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df1['Preferred Renewable'][df1['Preferred Renewable'] == source] = i
     1.40616570327553
avg graduated = df1['Graduated'].mean()
df1['Graduated'] = df1['Graduated'].fillna(avg_graduated)
avg_graduated
→ 0.6154437187860161
df1 = df1.drop('Unnamed: 0', axis=1)
df1.head(5)
\overline{\Rightarrow}
         Gender Age Ever_Married Family_Size Profession Graduated Work_Experien
      0
                   36
                                                                                         0.0000
               1
                                  1.0
                                                           1 00000
                                                                           1.0
                                                          0.00000
                                                                                         8.0000
               0
                   37
                                  1.0
                                                 4.0
                                                                           1.0
      2
                                                 1.0
                                                          3.31518
                                                                           0.0
                                                                                         0.0000
               1
                   69
                                  1.0
      3
               0
                                                 2.0
                                                          5.00000
                                                                                         11.0000
                   59
                                  1.0
                                                                           0.0
      4
               1
                   19
                                  0.0
                                                 40
                                                          8 00000
                                                                           0.0
                                                                                         2 6416
 Next steps: Generate code with df1

    View recommended plots

data1 = df1.to_numpy()
data1
                           , 36.
\rightarrow array([[ 1.
             [ 0.
                          , 37.
                                                         , ..., 8.
                                          ],
               1.
                              1.
                           , 1.
```

ſ 1.

1.

```
, 53.
             [ 1.
                                                      , ..., 2.64166321,
                                           0.
               0.
                            1.
                          , 47.
             [ 0.
                                           1.
                                                              1.
               2.
                            0.
                          , 43.
             [ 1.
                                           0.
                                        j])
data1.shape
→ (2627, 9)
X1 = data1[:, 0:9]
X1.shape
<del>→</del> (2627, 9)
X1 = torch.from_numpy(X1)
X1.shape
→ torch.Size([2627, 9])
X1 = X1.type(torch.float32)
predictions = model_0(X1).argmax(dim=1).tolist()
for i, prediction in enumerate(predictions):
  if prediction == 0:
    predictions[i] = 'A'
  elif prediction == 1:
    predictions[i] = 'B'
  elif prediction == 2:
    predictions[i] = 'C'
  elif prediction == 3:
    predictions[i] = 'D'
  else:
    predictions[i] = 'Error'
predictions[:5]
df2.head(5)
\rightarrow
        Unnamed:
                   Gender Age Ever_Married Family_Size Profession Graduated Work_Experience Energy_Consumption Preferred_
                0
      0
                                                                 Engineer
                0
                   Female
                            36
                                           Yes
                                                         1.0
                                                                                 Yes
                                                                                                    0.0
                                                                                                                        Low
      1
                1
                      Male
                            37
                                           Yes
                                                         4.0
                                                                Healthcare
                                                                                 Yes
                                                                                                    8.0
                                                                                                                     Average
                2
                                                                    NaN
                                                                                                    0.0
                   Female
                            69
                                           Yes
                                                         1.0
                                                                                  No
                                                                                                                        Low
      3
                3
                      Male
                             59
                                           Yes
                                                         2.0
                                                                Executive
                                                                                  No
                                                                                                   11.0
                                                                                                                        High
      4
                                                         4.0
                                                                Marketing
                                                                                                   NaN
                                                                                                                        Low
                4
                   Female
                             19
                                           No
                                                                                  No
                                       View recommended plots
             Generate code with df2
 Next steps:
df2['Group'] = predictions
df2.head(5)
\overline{\mathbf{T}}
                   Gender Age Ever_Married Family_Size Profession Graduated Work_Experience Energy_Consumption Preferred_
      0
                0
                   Female
                            36
                                           Yes
                                                         1.0
                                                                 Engineer
                                                                                 Yes
                                                                                                    0.0
                                                                                                                        Low
      1
                1
                      Male
                            37
                                           Yes
                                                         4.0
                                                                Healthcare
                                                                                 Yes
                                                                                                    8.0
                                                                                                                     Average
      2
                2
                    Female
                            69
                                           Yes
                                                         1.0
                                                                     NaN
                                                                                  No
                                                                                                    0.0
                                                                                                                        Low
      3
                3
                      Male
                            59
                                           Yes
                                                         2.0
                                                                Executive
                                                                                  No
                                                                                                   11.0
                                                                                                                        High
      4
                             19
                                                                                                   NaN
                   Female
                                           No
                                                         4.0
                                                                Marketing
                                                                                  No
                                                                                                                        Low
 Next steps:
              Generate code with df2
                                       View recommended plots
```

],

, 1.

with open("Edited_Consumer_Test_Dataset.csv", "wb") as f:
 f.write(df2.to_csv(index=False).encode("utf-8"))

print("csv file created")

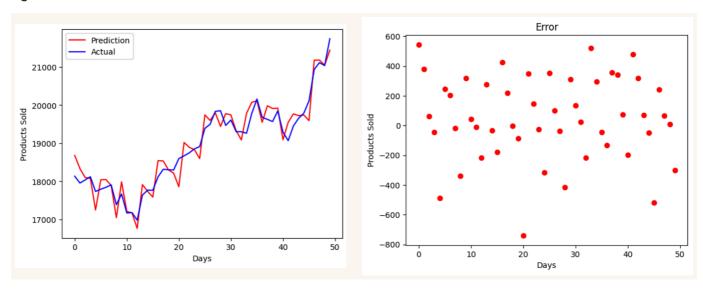
0.

Q4: In the above case your team already had data from which customers were classified in groups. How will you predict the classes if the groups and it's data weren't available? State the method you would have used.

Instead of deploying a Deep Learning model which predicts the group for the customers, I would try to find some pattern among the customers and using that write code to distribute them in the four classes

DAV Assignment Q6

After launching your renewable energy products in the Indian market, you notice a positive response from customers. Sales are increasing steadily, and the demand for your products is reaching new heights. To ensure your company can meet this growing demand, you need to predict the number of goods that can be produced over time. This forecast will help you make informed decisions about scaling operations and supply chain management. Given that your company relies on renewable energy sources, daily production can vary based on factors like weather, season, and other variables. Your team has developed a model that accounts for these factors, but its performance on test data has been disappointing, as shown in the accompanying figure.



Q6: What do you think might be causing the poor performance of the model? To improve the model's accuracy, what steps would you take? Provide a detailed justification for each of your proposed methods.

Causes

- 1. Model Complexity The model used might be too simple to capture the underlying patterns in the data.
- 2. Data Quality and Quantity The data might have some noise or missing values. Also, the amount of data available might not be sufficient to train a robust model.
- 3. Data Preprocessing The data may not have been cleaned, reduced and transform sufficiently.
- 4. Incorrect Hyperparameter Tuning The hyperparameters may not have been tuned to achive optimal performance.

Steps to take

1. Increase the Model Complexity

- Instead of a simple linear regression or a basic time series model, consider using more advanced models like random forests, decision tree or logistic regression models.
- Use advanced time series models like LSTM.

2. Encorporate More Features

- Add more features to the dataset.
- Normalise or transform the features to improve the model's performance.

3. Improve data quality

- Handle missing values appropriately. Assign the average value of the column to the null field.
- Augment existing data to create more data.

4. Model tuning

 Use techniques like grid search or random search to find the optimal hyperparameters.

DAV Assignment Q7

You have been provided with a dataset containing emails categorized as either spam or non-spam and another dataset with emails not yet categorized.

Q7: Develop a model to predict whether each email is spam or not, and use it to classify the uncategorized emails.

```
# Import libraries
import requests
from pathlib import Path
import zipfile
import os
request = requests.get("https://github.com/PanavShah1/DAV_assignment/blob/main/Email_Dataset.csv?raw=true")
with open("Email_Dataset.csv", "wb") as f:
  f.write(request.content)
  print("csv file downloaded")

→ csv file downloaded
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
df = pd.read_csv("Email_Dataset.csv")
df.head(5)
\overline{2}
        Unnamed: 0 CATEGORY
                                                                          MESSAGE
                                                                                     ▦
     0
               1000
                         Spam
                                          \n\nThe Internet's Online Pharmacy\n\n\nViag...
                                                                                     ıı.
                                      -----=_NextPart_000_00B0_35C58D0E.D7267B06\n\...
               1001
      1
                         Spam
      2
               1002
                         Spam
                                         <a href="http-equiv="Cont...">html>\n\n\n<head>\n\n<meta http-equiv="Cont...</a>
      3
               1003
                                      -----=_NextPart_000_00E4_86E61E0A.B5488E11\n\...
                         Spam
               1004
                         Spam BARRISTER ADEWALE COKER CHAMBERS\n\nLegal Prac...
 Next steps:
             Generate code with df
                                     View recommended plots
import string
def remove_angle_brackets(text):
  return pd.Series(text).str.replace(r'<(.|\n)*>', '', regex=True)
def remove_new_line(text):
  return pd.Series(text).str.replace(r'\n', '', regex=True)
punctuation_list = string.punctuation
def remove_punctuations(text):
    temp = str.maketrans('', '', punctuation_list)
    return text.translate(temp)
df['MESSAGE'] = df['MESSAGE'].apply(remove_angle_brackets)
df['MESSAGE'] = df['MESSAGE'].apply(remove_new_line)
df['MESSAGE'] = df['MESSAGE'].apply(remove_punctuations)
df.drop('Unnamed: 0', axis=1, inplace=True)
df.replace('Spam', value=1, inplace=True)
df.replace('Not Spam', value=0, inplace=True)
df.head(10)
```

```
CATEGORY
                                                             MESSAGE
                                                                         0
                              The Internets Online PharmacyViagra Xenical ...
      1
                1
                      NextPart00000B035C58D0ED7267B06ContentType tex...
      2
                1
                                            httpxentcommailmanlistinfofork
                       NextPart00000E486E61E0AB5488E11ContentType tex...
      3
                1
      4
                   BARRISTER ADEWALE COKER CHAMBERSLegal Practiti...
      5
                1
                      DeathToSpamDeathToSpamDeathToSpamThis sfnet em...
      6
                1
      7
                1
      8
                               Just sent you a note with the wrong link Tthe ...
      9
                             This is a multipart message in MIME formatNext...
 Next steps:
              Generate code with df
                                      View recommended plots
X = df['MESSAGE']
y = df['CATEGORY']
X.shape, y.shape
→ ((4000,), (4000,))
from \ sklearn.model\_selection \ import \ train\_test\_split
X_{\text{train}}, X_{\text{test}}, y_{\text{train}}, y_{\text{test}} = train_test_split(X, y, test_size=0.2, random_state=42)
X_train.shape, X_test.shape, y_train.shape, y_test.shape

→ ((3200,), (800,), (3200,), (800,))
from \ sklearn.feature\_extraction.text \ import \ TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=1, stop_words='english', lowercase=True)
X_train_vectorized = vectorizer.fit_transform(X_train)
X_test_vectorized = vectorizer.transform(X_test)
X_{\text{train\_vectorized.shape}}, X_{\text{test\_vectorized.shape}}
→ ((3200, 59378), (800, 59378))
y_train = y_train.astype('int')
y_test = y_test.astype('int')
from sklearn.linear_model import LogisticRegression
model = LogisticRegression()
model.fit(X_train_vectorized, y_train)
     ▼ LogisticRegression
     LogisticRegression()
train_predictions = model.predict(X_train_vectorized)
train_predictions[:10]
→ array([0, 0, 0, 0, 1, 0, 0, 0, 0])
train_accuracy = model.score(X_train_vectorized, y_train)
train_accuracy
→ 0.9134375
test_predictions = model.predict(X_test_vectorized)
test_predictions[:10]
\Rightarrow array([1, 0, 1, 0, 0, 0, 0, 1, 0, 1])
test_accuracy = model.score(X_test_vectorized, y_test)
test accuracy
→ 0.8825
```

₹

Predict

```
request = requests.get ("https://github.com/PanavShah1/DAV\_assignment/blob/main/Email%20Test%20Data.csv?raw=true") (assignment/blob/main/Email%20Test%20Data.csv?raw=true") (blob/main/Email%20Test%20Data.csv?raw=true") (blob/main/Email%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20Test%20
with open("Email_Test_Data.csv", "wb") as f:
    f.write(request.content)
    print("csv file downloaded")
 → csv file downloaded
df1 = pd.read_csv("Email_Test_Data.csv")
df1.head(5)
 <del>_</del>
                   Unnamed: 0
                                                                                                                                    MESSAGE
                                                                                                                                                            \blacksquare
            0
                                                              Dear Homeowner,\n\n \n\nInterest Rates are at ...
             1
                                        1
                                                      ATTENTION: This is a MUST for ALL Computer Use...
            2
                                        2
                                                               This is a multi-part message in MIME format.\n...
             3
                                        3 IMPORTANT INFORMATION:\n\n\nThe new domain n...
                                                                 This is the bottom line. If you can GIVE AWAY...
             4
                            Generate code with df1
                                                                                    View recommended plots
  Next steps:
from copy import deepcopy
df2 = deepcopy(df1)
df1['MESSAGE'] = df1['MESSAGE'].apply(remove_angle_brackets)
df1['MESSAGE'] = df1['MESSAGE'].apply(remove_new_line)
df1['MESSAGE'] = df1['MESSAGE'].apply(remove_punctuations)
df1.head(5)
 ₹
                   Unnamed: 0
                                                                                                                                       MESSAGE
                                                                                                                                                                \overline{\blacksquare}
             0
                                                                 Dear Homeowner Interest Rates are at their low...
                                        1
                                                         ATTENTION This is a MUST for ALL Computer User...
             1
                                                                This is a multipart message in MIME formatNext...
                                        3 IMPORTANT INFORMATIONThe new domain names are ...
             3
             4
                                                                    This is the bottom line If you can GIVE AWAY ...
  Next steps: Generate code with df1
                                                                                    View recommended plots
X1 = df1['MESSAGE']
X1.shape
 → (1000,)
X1_vectorized = vectorizer.transform(X1)
spam_predictions_binary = model.predict(X1_vectorized)
spam_predictions = [0 for i in range(len(spam_predictions_binary))]
for i, x in enumerate(spam_predictions_binary):
    if x == 1:
        spam_predictions[i] = 'Spam'
    else:
         spam_predictions[i] = 'Not Spam'
spam_predictions[:10]
           ['Spam',
              'Not Spam',
             'Spam',
              'Not Spam',
              'Spam',
              'Not Spam',
              'Not Spam',
              'Not Spam'
              'Not Spam']
```

This is the bottom line. If you can GIVE AWAY... Not Spam

Next steps: Generate code with df2 View recommended plots

with open('Edited_Email_Test_Data.csv', 'wb') as f:
 f.write(df2.to_csv(index=False).encode('utf-8'))
 print("csv file created")

→ csv file created