



DIVE INTO CODE

MACHINE LEARNING
GRADUATION ASSIGNMENT

PROJECT

ELECTRICITY POWER
CONSUMPTION

SELF INTRODUCTION

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LEVEL: YEAR 3

COURSE: DIVE INTO CODE (DIC)

PROJECT: ELECTRICITY POWER CONSUMPTION

ELECTRICITY POWER CONSUMPTION

Electricity power consumption has been growing rapidly in spite of serious supply constraints caused by delays in implementing new capacity additions, lower utilization of existing capacity due to maintenance problems, coal and oil availability, and draught conditions (in 1979-80), which simultaneously reduced hydro energy availability and increased agricultural demand for lift irrigation.

ENERGY CONSUMPTION IN SIERRA LEONE

The most important measure in the energy balance of Sierra Leone is the total consumption of 279.00 m kWh of electric energy per year. Per capita this is an average of **35 kWh**.

Sierra Leone can provide itself completely with self-produced energy. The total production of all electric energy producing facilities is 300 m kWh, also 108% of own requirements. The rest of the self-produced energy is either exported into other countries or unused. Along with pure consumptions the production, imports and exports play an important role. Other energy sources such as natural gas or crude oil are also used.

ENERGY BALANCE

Electricity	total	Sierra Leone per capita	USA per capita
Own consumption	279.00 m kWh	34.98 kWh	11,842.76 kWh
Production	300.00 m kWh	37.61 kWh	12,428.52 kWh

CARBON FOOTPRINT

	CO2 emissions in 2018	Sierra Leone per capita	USA per capita total
Total	1.02 m t	0.13 t	15.12 t

Development of CO2 emissions from 1979 to 2018 in million tons



Production capacities per energy source

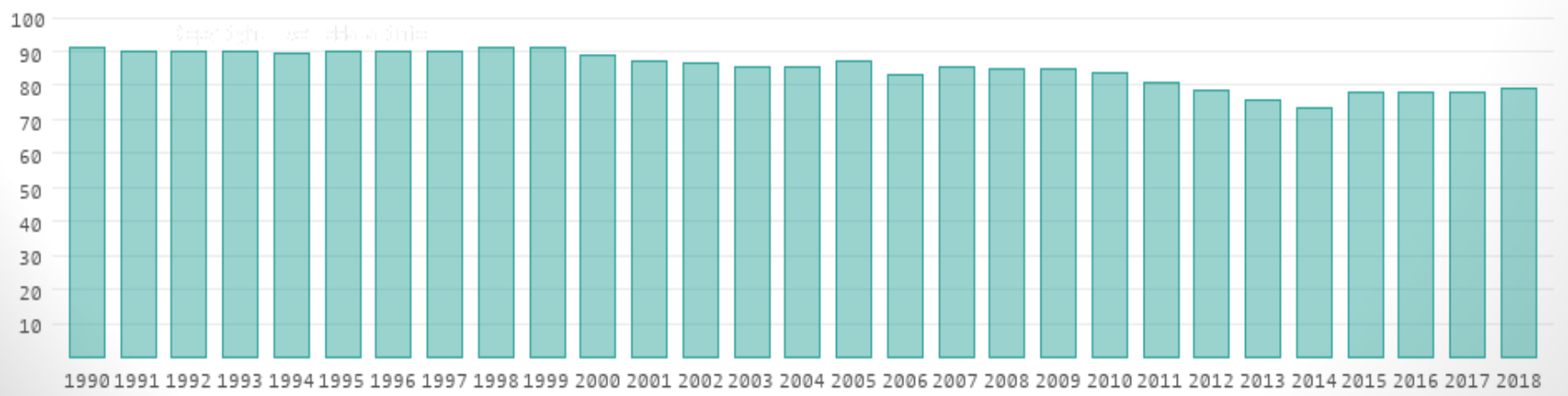
The given production capacities for electric energy have a theoretical value, which could only be obtainable under ideal conditions. They are measuring the generatable amount of energy, that would be reached under permanent and full use of all capacities of all power plants. In practice this isn't possible, because e.g. solar collectors are less efficient under clouds. Also wind- and water-power plants are not always operating under full load. All these values are only useful in relation to other energy sources or countries.

Energy source	total in Sierra Leone	percentage in Sierra Leone	percentage USA	per capita in Sierra Leone	per capita USA
Fossil fuels	228.28 m kWh	23,0 %	70,0 %	28.62 kWh	20,230.06 kWh
Nuclear power	0.00 kWh	0,0 %	9,0 %	0.00 kWh	2,601.01 kWh
Water power	506.18 m kWh	51,0 %	7,0 %	63.45 kWh	2,023.01 kWh
Renewable energy	258.05 m kWh	26,0 %	14,0 %	32.35 kWh	4,046.01 kWh
Total production capacity	992.51 m kWh	100,0 %	100,0 %	124.42 kWh	28,900.09 kWh
Actual total production	300.00 m kWh	30.2 %	43.0 %	37.61 kWh	12,428.52 kWh

Usage of renewable energies

Renewable energies include wind, solar, biomass and geothermal energy sources. This means all energy sources that renew themselves within a short time or are permanently available. Energy from hydropower is only partly a renewable energy. This is certainly the case with river or tidal power plants. Otherwise, numerous dams or reservoirs also produce mixed forms, e.g. by pumping water into their reservoirs at night and recovering energy from them during the day when there is an increased demand for electricity. Since it is not possible to clearly determine the amount of generated energy, all energies from hydropower are displayed separately.

In 2018, renewable energies accounted for around 79.6 percent of actual total consumption in Sierra Leone. The following chart shows the percentage share from 1990 to 2018:



DATASET CODE OF ELECTRICITY POWER CONSUMPTION

```
In 1  import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

```
/kaggle/input/electricity-consumption/train.csv
/kaggle/input/electricity-consumption/test.csv
```

```
In 2  !pip install DataScienceHelper
```

```
Collecting DataScienceHelper
```

```
  Downloading datasciencehelper-1.5.2.tar.gz (6.4 kB)
```

```
Building wheels for collected packages: DataScienceHelper
```

```
  Building wheel for DataScienceHelper (setup.py) ... done
```

```
  Created wheel for DataScienceHelper: filename=datasciencehelper-1.5.2-py3-none-any.whl size=5727
sha256=6aaeddf275658cdbca7b7f0564ea1b409583a584d2286b32b46d052c0b3399ae
```

```
  Stored in directory: /root/.cache/pip/wheels/49/5e/d2/84a664218a270ce173c1d02086f556367a00002afe0
aae6409
```

```
Successfully built DataScienceHelper
```

```
Installing collected packages: DataScienceHelper
```

```
Successfully installed DataScienceHelper-1.5.2
```

In 3 `!pip install --upgrade pip`

```
Collecting pip
  Downloading pip-20.2.2-py2.py3-none-any.whl (1.5 MB)
  | 1.5 MB 402 kB/s eta 0:00:01
Installing collected packages: pip
  Attempting uninstall: pip
    Found existing installation: pip 20.2.1
    Uninstalling pip-20.2.1:
      Successfully uninstalled pip-20.2.1
  Successfully installed pip-20.2.2
```

In 4

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import DataScienceHelper as dsh
import plotly.express as px
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
```

```
%matplotlib inline
```

In 5

```
import time
from datetime import datetime
import re
from math import *
```



```
In 6 data = pd.read_csv("/kaggle/input/electricity-consumption/train.csv")
data.head()
```

Out 6

ID	datetime	temperature	var1	pressure	windspeed	var2	electricity_consumption	
0	0	2013-07-01 00:00:00	-11.4	-17.1	1003.0	571.910	A	216.0
1	1	2013-07-01 01:00:00	-12.1	-19.3	996.0	575.040	A	210.0
2	2	2013-07-01 02:00:00	-12.9	-20.0	1000.0	578.435	A	225.0
3	3	2013-07-01 03:00:00	-11.4	-17.1	995.0	582.580	A	216.0
4	4	2013-07-01 04:00:00	-11.4	-19.3	1005.0	586.600	A	222.0

In 7 data.tail()

Out 7

ID	datetime	temperature	var1	pressure	windspeed	var2	electricity_consumption	
26491	34891	2017-06-23 19:00:00	-0.7	-15.0	1009.0	51.685	A	225.0
26492	34892	2017-06-23 20:00:00	-2.9	-11.4	1005.0	56.105	A	213.0
26493	34893	2017-06-23 21:00:00	-1.4	-12.9	995.0	61.275	A	213.0
26494	34894	2017-06-23 22:00:00	-2.9	-11.4	996.0	67.210	A	210.0
26495	34895	2017-06-23 23:00:00	-2.1	-11.4	1009.0	71.880	A	210.0

In 8 data.isnull().sum()

Out 8

```
ID      0
Datetime  0
temperature  0
var1      0
pressure  0
windspeed  0
var2      0
electricity_consumption  0
dtype: int64
```

In 9 data.describe()

Out 9	ID	temperature	var1	pressure	windspeed	electricity_consumption	
	count	26496.000000	26496.000000	26496.000000	26496.000000	26496.000000	26496.000000
	mean	17455.500000	5.098989	-1.916233	986.450615	23.959956	298.359601
	std	10122.873673	8.682860	10.424860	12.002647	48.280321	108.020555
	min	0.000000	-17.100000	-32.900000	953.000000	1.075000	174.000000
	25%	8717.750000	-2.900000	-10.700000	978.000000	3.155000	219.000000
	50%	17435.500000	6.400000	-1.400000	986.000000	6.545000	267.000000
	75%	26177.250000	12.100000	7.900000	995.000000	22.260000	342.000000
	max	34895.000000	23.600000	18.600000	1024.000000	586.600000	1386.000000

In 10 data.count()

Out 10	ID 26496	datetime	26496
	temperature		26496
	var1		26496
	pressure		26496
	windspeed		26496
	var2		26496
	electricity_consumption		26496
	dtype:	int64	

In 11 data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 26496 entries, 0 to 26495
Data columns (total 8 columns):
#      Column                                Non-Null Count  Dtype
---  -
0      ID                                26496 non-null  int64
1      datetime                          26496 non-null  object
2      temperature                      26496 non-null  float64
3      var1                             26496 non-null  float64
4      pressure                         26496 non-null  float64
5      windspeed                       26496 non-null  float64
6      var2                             26496 non-null  object
7      electricity_consumption  26496 non-null  float64
dtypes: float64(5), int64(1), object(2)
memory usage: 1.6+ MB
```

In 12 data.memory_usage()

Out 12

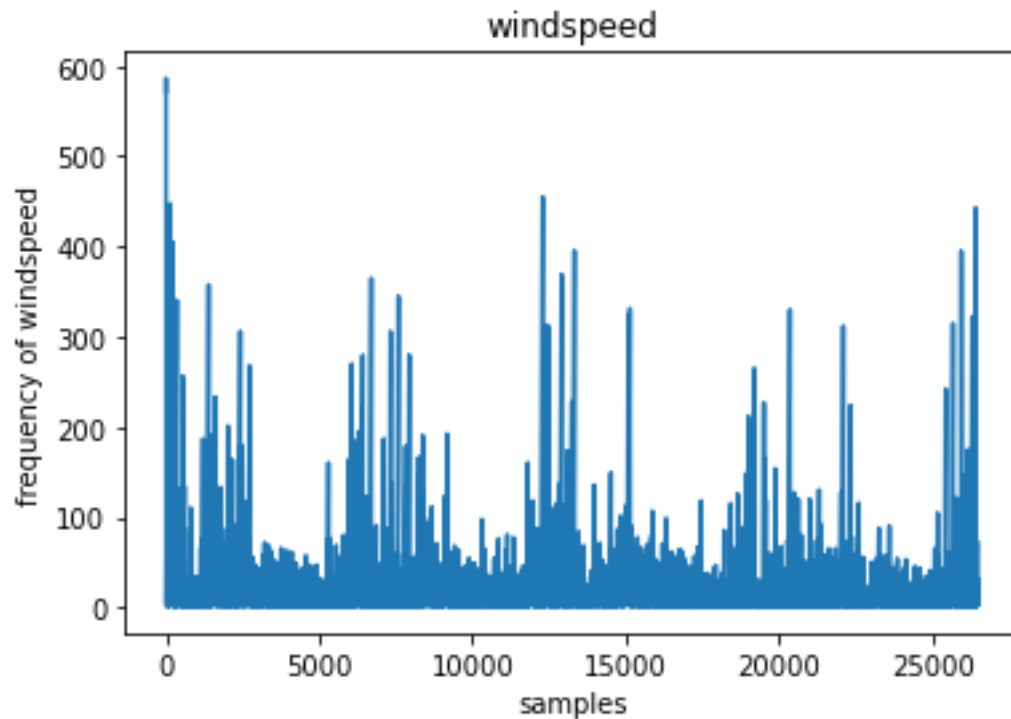
```
Index          128
ID            211968
datetime       211968
temperature    211968
var1           211968
pressure       211968
windspeed      211968
var2           211968
electricity_consumption 211968
dtype: int64
```

In 13 data.windspeed.value_counts()

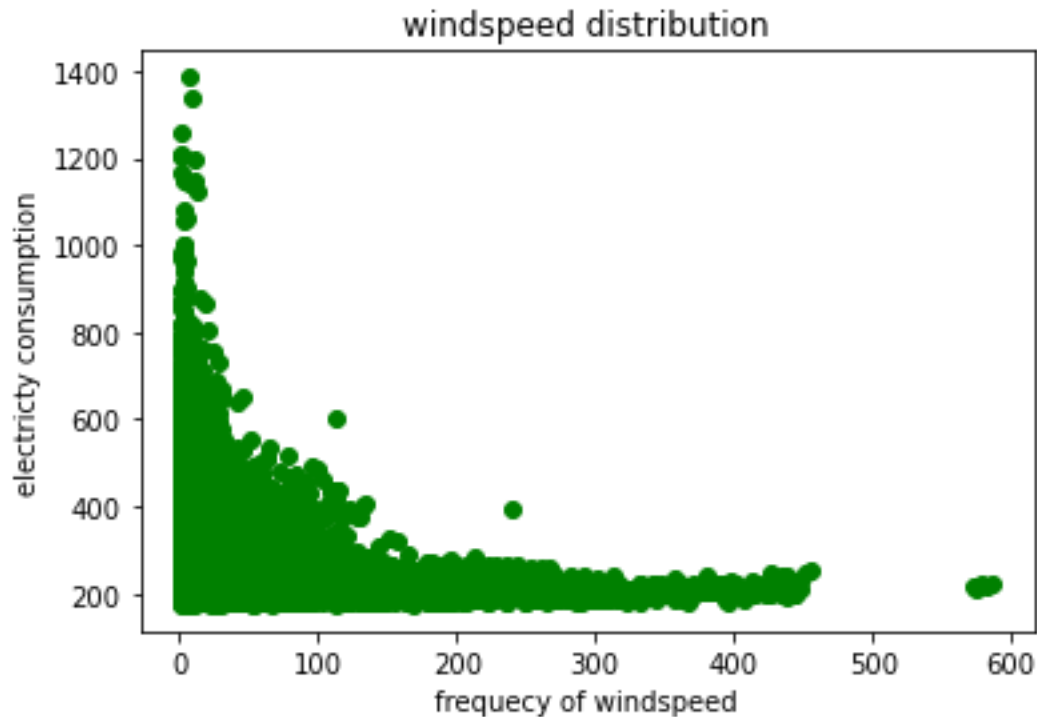
Out 13

```
2.265          380
1.890          369
2.015          359
2.390          354
2.140          347
...
318.210         1
123.435         1
282.485         1
160.465         1
27.825          1
Name: windspeed, Length: 5603, dtype: int64
```

```
In 14  plott = data.windspeed  
plt.plot(plott)  
plt.xlabel("samples")  
plt.ylabel("frequency of windspeed")  
plt.title("windspeed")  
plt.show()
```



```
In 15 plt.scatter(data.windspeed,data.electricity_consumption,c='green')  
plt.xlabel("frequency of windspeed")  
plt.ylabel("electricity consumption")  
plt.title("windspeed distribution")  
plt.show()
```



```
In16  average = round(data.windspeed.mean(),3)
      max_windspeed = round(max(data.windspeed),3)
      min_windspeed = round(min(data.windspeed),3)
      print(f'The average windspeed is : {average} ')
      print(f'The maximum windspeed is : {max_windspeed}')
      print(f'The minimum windspeed is : {min_windspeed}')
```

The average windspeed is : 23.96

The maximum windspeed is : 586.6

The minimum windspeed is : 1.075

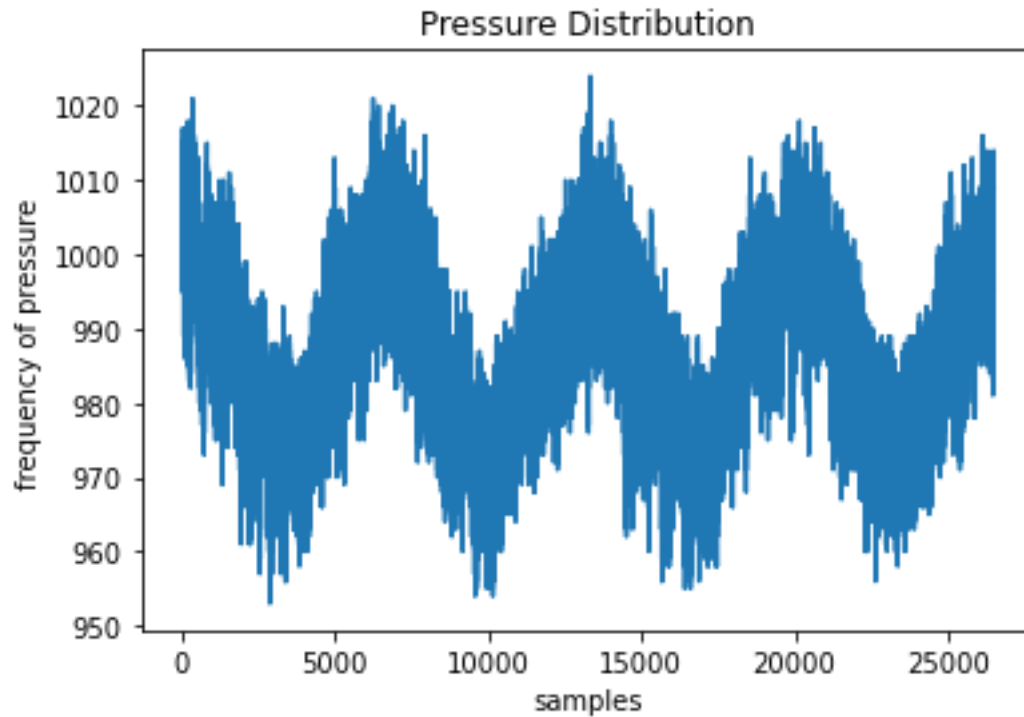
```
In17  avg_pressure = round(data.pressure.mean(),3)
      max_pressure = round(data.pressure.max(),3)
      min_pressure = round(data.pressure.min(),3)
      print(f'The average pressure is : {avg_pressure}')
      print(f'The maximum pressure is : {max_pressure}')
      print(f'The minimum pressure is : {min_pressure}')
```

The average pressure is : 986.451

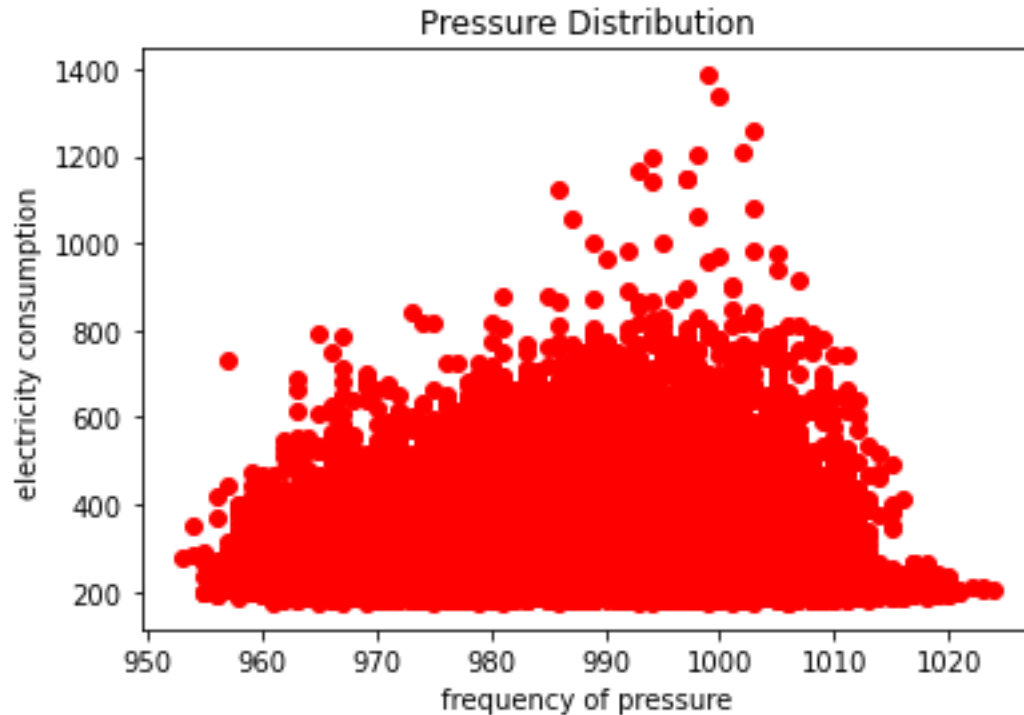
The maximum pressure is : 1024.0

The minimum pressure is : 953.0

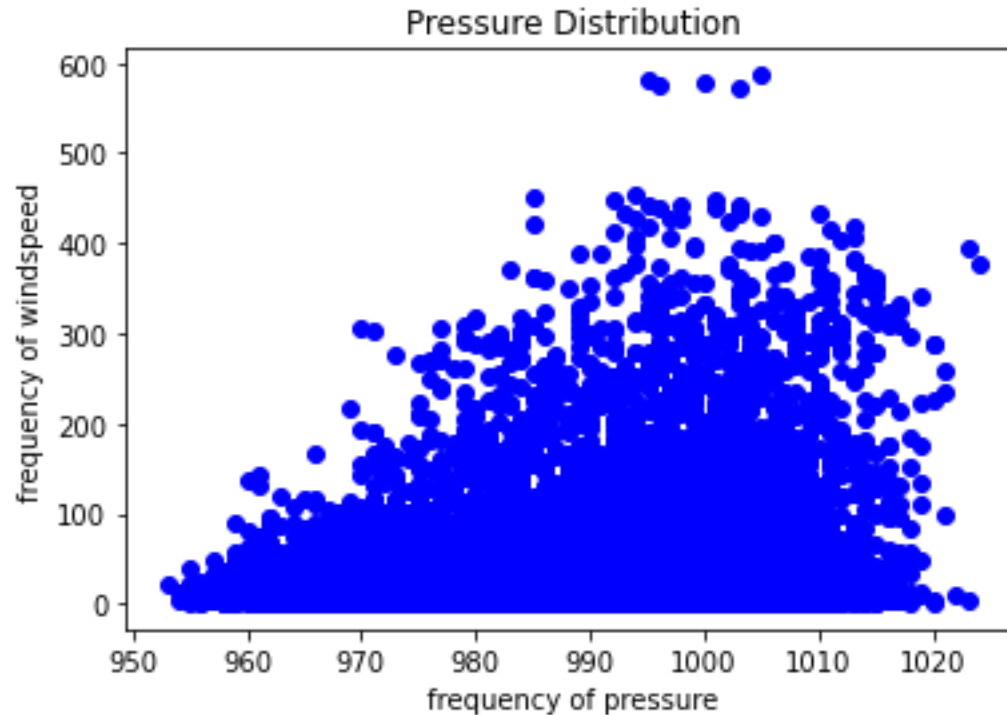

```
In18 plt.plot(data.pressure)
plt.xlabel("samples")
plt.ylabel("frequency of pressure")
plt.title("Pressure Distribution")
plt.show()
```



```
In19 plt.scatter(data.pressure,data.electricity_consumption,c='red')  
plt.xlabel("frequency of pressure")  
plt.ylabel("electricity consumption")  
plt.title("Pressure Distribution")  
plt.show()
```

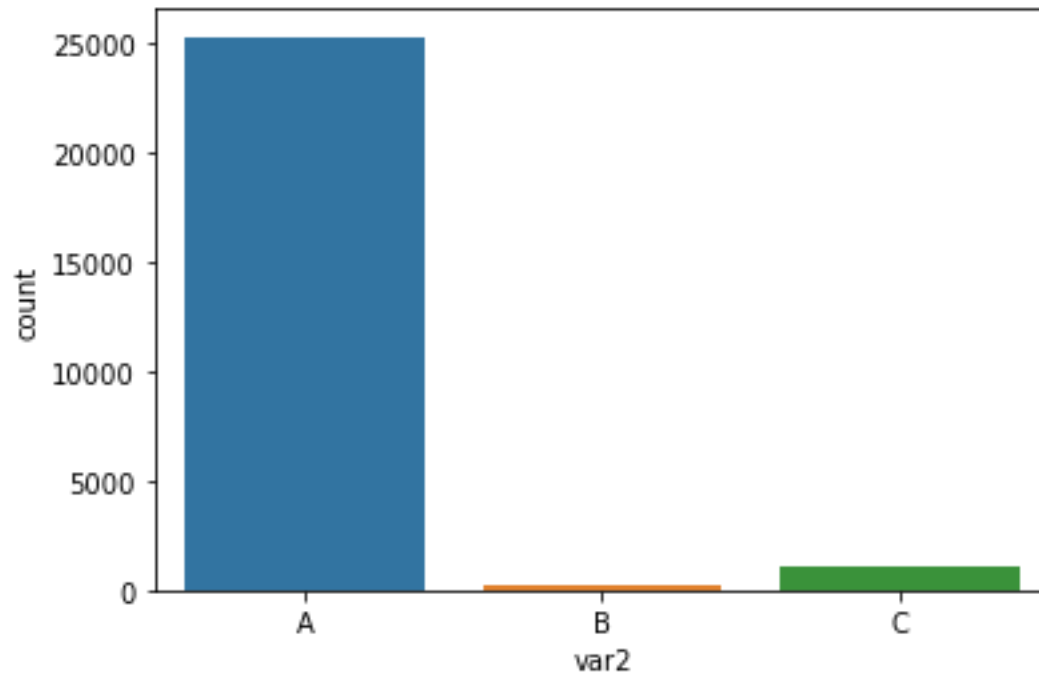


```
In20 plt.scatter(data.pressure,data.windspeed,c='blue')  
plt.xlabel("frequency of pressure")  
plt.ylabel("frequency of windspeed")  
plt.title("Pressure Distribution")  
plt.show()
```



In 21 `sns.countplot(x='var2',data = data)`

Out 21 `<matplotlib.axes._subplots.AxesSubplot at 0x7f66e65e2f90>`



In 43

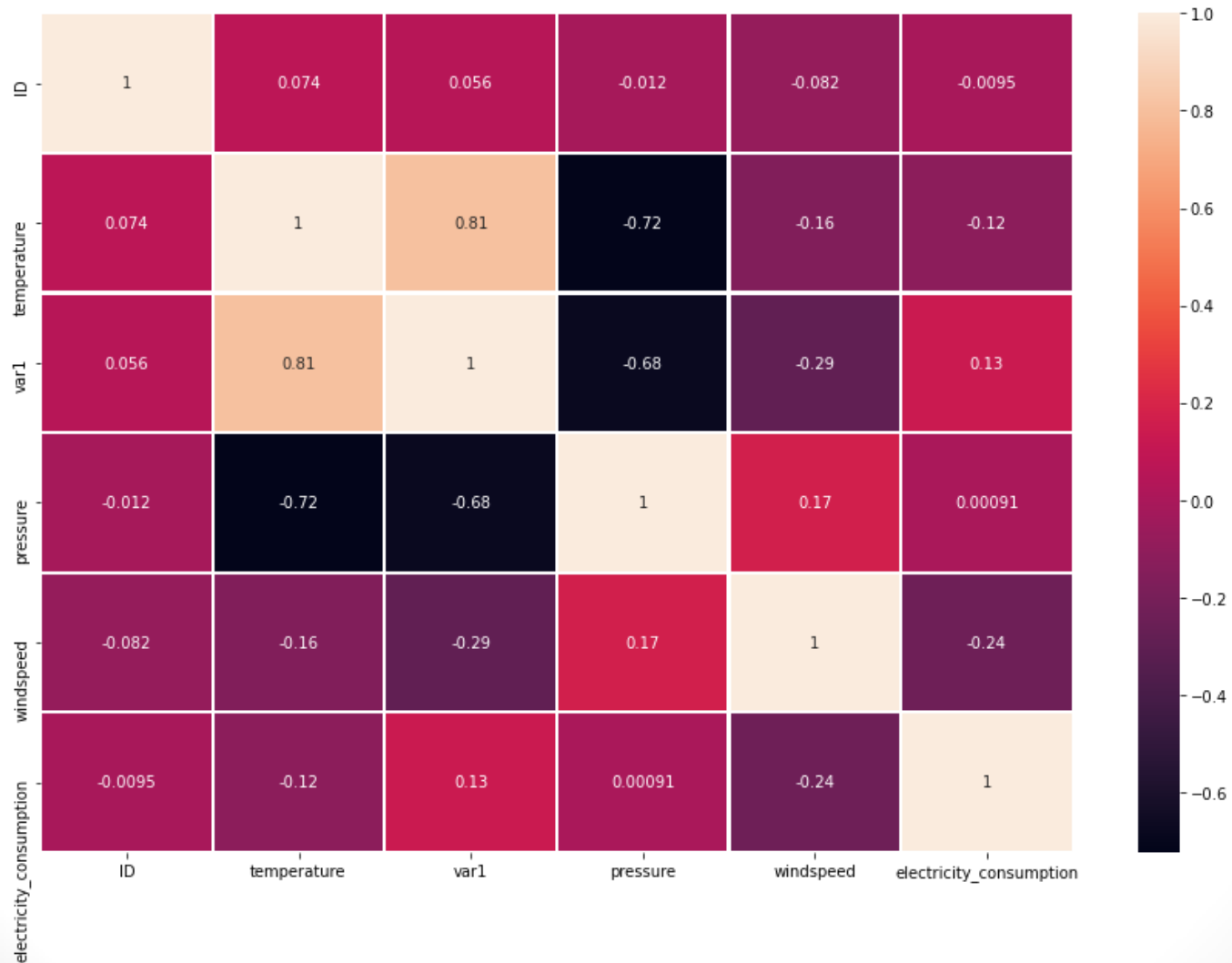
```
fig,ax = plt.subplots(figsize = (15,10))
```

```
corr = data.corr()
```

```
sns.heatmap(corr,xticklabels = corr.columns,annot = True,yticklabels = corr.columns,linewidth =1.2)
```

Out 43

<matplotlib.axes._subplots.AxesSubplot at 0x7f66e6c07e50>



```
In 23  corr[abs(corr['electricity_consumption']) > 0.1]['electricity_consumption']
```

```
Out 23  temperature          -0.117254  
        var1              0.133914  
        windspeed         -0.238883  
        electricity_consumption 1.000000  
        Name: electricity_consumption, dtype: float64
```

```
In 24  data.var1.value_counts()
```

```
Out 24  10.0      836  
        8.6      809  
        10.7     797  
        7.9      779  
        9.3      770  
        ...  
        -29.3     3  
        17.9     2  
        -32.9     1  
        18.6     1  
        -32.1     1  
        Name: var1, Length: 71, dtype: int64
```

```
In 25 data.var2.value_counts()
```

```
Out 25 A      25239  
      C      1040  
      B       217  
      Name: var2, dtype: int64
```

Reference

This is what i have gathered so far for my Graduation Assignment project title ELECTRICITY POWER CONSUMPTION, also gathered some assistance material on KAGGLE to add some important materials on my project



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END OF SESSION