Utilization of algorithms, Dynamic programming, Optimization

Firstly, we are importing pandas and numpy module. Pandas is a fast, powerful, flexible and easy to use open source data analysis and manipulation tool. Numpy refers to numeric python.

In [1]:

import pandas as pd

import numpy as np

Reading the dataset

We are reading the csv file and storing it as dataframe

In [2]:

df=pd.read_csv("Dataset - Sheet3.csv")

Data pre-processing

head() method - Returns the first 5 rows of the dataframe.

In [3]:

df.head()

Out[3]:

Date	Day	Time	Empty level in cm(Total size=100cm)				
0	03/01/21		Monday	7:00 AM	80		
1	03/01/21		Monday	9:00 AM	60		
2	03/01/21		Monday	11:00 AM	60		
3	03/01/21		Monday	1:00 PM	50		
4	03/01/21		Monday	3:00 PM	90		

tail() method - Returns the last 5 rows of the dataframe.

In [4]:

df.tail()

Out[4]:

Date	Day	Time	Empty level in cm(Total	size=100cm)		
239	03/28/21		Sunday 3:00 PM	60		
240	03/28/	21	Sunday 4:00 PM	40		
241	03/28/	21	Sunday 5:00 PM	35		
242	03/28/	21	Sunday 6:00 PM	100		
243	03/28/	21	Sunday 7:00 PM	80		
Columns attribute return the column labels of the given Data						

aframe.

In [5]: df.columns Out[5]: Index(['Date', 'Day', 'Time', 'Empty level in cm(Total size=100cm)'], dtype='object') dtypes attributes returns a Series with the data type of each column. In [6]:

df.dtypes Out[6]:

Date object Day object Time object

Empty level in cm(Total size=100cm) int64

dtype: object

describe() method is used to view some basic statistical details like percentile, mean, std etc. of a data frame

In [7]:

df.describe()

Out[7]:

Empty level in cm(Total size=100cm)

```
count 244.000000
mean 67.860656
std 24.060731
min 15.000000
25% 50.000000
50% 70.000000
75% 90.000000
max 100.000000
```

info() function is used to print a concise summary of a DataFrame. This method prints information about a DataFrame including the index dtype and column dtypes, non-null values and memory usage.

In [8]: df.info()

RangeIndex: 244 entries, 0 to 243

Data columns (total 4 columns):

Column Non-Null Count Dtype
--- ---
O Date 244 non-null object

1 Day 244 non-null object

2 Time 244 non-null object

3 Empty level in cm(Total size=100cm) 244 non-null int64

dtypes: int64(1), object(3) memory usage: 7.8+ KB

Using the below snippet code we are converting the datatypes of the columns(object-->int).

In [9]:
from sklearn.preprocessing import LabelEncoder
category= ['Date','Day','Time']
encoder= LabelEncoder()

for i in category:

df[i] = encoder.fit_transform(df[i])

df.dtypes

Out[9]:

Date int32

Day int32

Time int32

Empty level in cm(Total size=100cm) int64

dtype: object

corr() is used to find the pairwise correlation of all columns in the dataframe. Any nan values are automatically excluded. For any non-numeric data type columns in the dataframe it is ignored. Note: The correlation of a variable with itself is 1.

In [10]:

df.corr()

Out[10]:

Date	Day	Time	Empty level in	cm(Total size=10	00cm)		
Date	1.0000	000	-0.056907	-0.020373	-0.032041		
Day	-0.056	907	1.000000	0.018148	-0.104016		
Time	-0.020	373	0.018148	1.000000	0.200872		
Empty level in cm(Total size=100cm)			ıl size=100cm)	-0.032041	-0.104016	0.200872	1.000000

From the above result we can conclude that all variables are independent of each other. So the idea of linear regression also fails

In [11]:

X=df[['Date','Day','Time']].values

X[0:5] #we are converting dataframe into numpy arrays.

^{*}values attribute returns the numpy representation of the given DataFrame. 'Date','Day','Time' are taken into X and these independent variables

Out[11]:

[0, 1, 5]])

*values attribute returns the numpy representation of the given DataFrame. 'Empty level in cm(Total size=100cm)' ar