

# **Google Ads Hourly Analysis**

Date: 7 June 2023

Project Start Date - End Date	<ul> <li>Start Date – 07 -06 -2023</li> <li>End Date – 07 -06 2023</li> </ul>
Objectives	<ul> <li>To analyses how many people who clicked on the advertisement enrolled in our course</li> <li>General exploratory analyses</li> <li>General descriptive analyses</li> </ul>
Milestones accomplished the week of Start Date - End Date:	<ul> <li>Descriptive analyses</li> <li>Exploratory analyses</li> <li>Classification of data with respect to term</li> </ul>

### **Contact Information**

This project is performed for educational purpose of under the guidance of Siddhivinayak Sir.

Project Manager

Name: Siddhivinayak Phulwadkar

Mobile: 9028965955

Email:

siddhivinayakphulwadkar@gmail.com

Client Project Champion

Name : Arti Sukhadev Patil Mobile: 8623845944

Email: parti5863@gmail.com

### **Project Abstract**

The dataset is about showing advertisement to students for course enrollment. Our main objective was to understand on which time students are clicked on our ads and enrolled our course. Problem statement is classify as we are looking for preferred timing in a day where we can do marketing and we will get definitely sales. For this dataset we have applied decision tree algorithm and performed exploratory and descriptive analysis

# Google Ads Hourly Analysis

# Importing libraries

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

# Importing dataset

file=pd.read\_excel("C:/Users/Administrator/Desktop/Marketing Data Google Ads 6th june.xlsx")

```
file.head()
```

	Sr no	Impressions	Clicks	Sales Unit	Time Preferred
0	00:00:00	258647	7759.41	54.31587	1
1	00:30:00	219974	8798.96	61.59272	1
2	01:00:00	1096	10.96	0.07672	0
3	01:30:00	1481	14.81	0.10367	0
4	02:00:00	1794	17.94	0.12558	0

# Preprocessing the dataset

```
dataset=file.drop("Sr no",axis=1)
```

#### dataset.head()

	Impressions	Clicks	Sales Unit	Time Preferred
0	258647	7759.41	54.31587	1
1	219974	8798.96	61.59272	1
2	1096	10.96	0.07672	0
3	1481	14.81	0.10367	0
4	1794	17.94	0.12558	0

#### # descriptive analysis

#### dataset.sum()

Impressions 2.207097e+06 Clicks 8.431078e+04 Sales Unit 5.901755e+02 Time Preferred 3.300000e+01

dtype: float64

#Impression are no visible Ads to customers #Total Impressions are 2.207097e+06 #Clicks indicates total no of customers who clicked on our Ads #Total no of Clicks Ad 8.431078e+04

#sales unit indicates total no of purchased made in particular 30 mi of slot #total no of sales unit for entire day 5.901755e+02

#time preferred is the coloumn which tells us preferred time slot for showing advertisement to the customers

#### dataset.mean()

Impressions 45981.187500 Clicks 1756.474554 Sales Unit 12.295322 Time Preferred 0.687500

dtype: float64

```
#the average no of Impression are 45981.187500
#the average no of Clicks are 1756.474554
#the average no of Sales unit are 12.295322
```

### dataset.isnull().sum()

Impressions 0
Clicks 0
Sales Unit 0
Time Preferred 0
dtype: int64

```
x=dataset.iloc[:,:-1].values
y=dataset.iloc[:,-1].values
```

```
Х
array([[2.58647000e+05, 7.75941000e+03, 5.43158700e+01],
       [2.19974000e+05, 8.79896000e+03, 6.15927200e+01],
      [1.09600000e+03, 1.09600000e+01, 7.67200000e-02],
      [1.48100000e+03, 1.48100000e+01, 1.03670000e-01],
      [1.79400000e+03, 1.79400000e+01, 1.25580000e-01],
      [2.15600000e+03, 2.15600000e+01, 1.50920000e-01],
      [4.13000000e+02, 4.13000000e+00, 2.89100000e-02],
      [1.44000000e+02, 1.44000000e+00, 1.00800000e-02],
      [1.39000000e+02, 1.39000000e+00, 9.73000000e-03],
      [1.91000000e+02, 1.91000000e+00, 1.33700000e-02],
      [3.96000000e+02, 3.96000000e+00, 2.77200000e-02],
      [2.74000000e+02, 2.74000000e+00, 1.91800000e-02],
      [3.15200000e+03, 3.15200000e+01, 2.20640000e-01],
      [3.08900000e+03, 3.08900000e+01, 2.16230000e-01],
      [2.13500000e+03, 2.13500000e+01, 1.49450000e-01],
      [1.84980000e+04, 3.69960000e+02, 2.58972000e+00],
      [1.78430000e+04, 5.35290000e+02, 3.74703000e+00],
      [1.27410000e+04, 4.45935000e+02, 3.12154500e+00],
      [1.89730000e+04, 6.83028000e+02, 4.78119600e+00],
      [1.17460000e+04, 4.69840000e+02, 3.28888000e+00],
      [1.09670000e+04, 2.19340000e+02, 1.53538000e+00],
      [8.74200000e+03, 4.19616000e+02, 2.93731200e+00],
      [4.05900000e+03, 4.87080000e+01, 3.40956000e-01],
      [2.05650000e+04, 4.11300000e+02, 2.87910000e+00],
      [2.44860000e+04, 7.10094000e+02, 4.97065800e+00],
      [2.04910000e+04, 6.76203000e+02, 4.73342100e+00],
      [1.29161000e+05, 1.23994560e+03, 8.67961920e+00],
      [1.01236000e+05, 2.02472000e+03, 1.41730400e+01],
      [1.45966000e+05, 2.91932000e+03, 2.04352400e+01],
array([1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1,
      1, 1, 1, 1], dtype=int64)
```

## Training the model

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.05, random_state =0)
```

```
print(x train)
[[1.78600000e+03 5.35800000e+01 3.75060000e-01]
 [1.96730000e+04 5.90190000e+02 4.13133000e+00]
 [3.96130000e+04 1.98065000e+03 1.38645500e+01]
 [1.74910000e+04 5.77203000e+02 4.04042100e+00]
 [1.12940000e+04 1.12940000e+02 7.90580000e-01]
 [1.44000000e+02 1.44000000e+00 1.00800000e-02]
 [3.96000000e+02 3.96000000e+00 2.77200000e-02]
 [2.74000000e+02 2.74000000e+00 1.91800000e-02]
 [1.47410000e+04 4.42230000e+02 3.09561000e+00]
 [1.24360000e+04 4.22824000e+02 2.95976800e+00]
 [1.01236000e+05 2.02472000e+03 1.41730400e+01]
 [2.94484000e+05 2.29697520e+04 1.60788264e+02]
 [1.09600000e+03 1.09600000e+01 7.67200000e-02]
 [2.14526000e+05 1.28715600e+04 9.01009200e+01]
 [1.89730000e+04 6.83028000e+02 4.78119600e+00]
 [1.84980000e+04 3.69960000e+02 2.58972000e+00]
 [1.45966000e+05 2.91932000e+03 2.04352400e+01]
 [4.05900000e+03 4.87080000e+01 3.40956000e-01]
 [1.78430000e+04 5.35290000e+02 3.74703000e+00]
 [1.59770000e+04 1.59770000e+02 1.11839000e+00]
 [1.09670000e+04 2.19340000e+02 1.53538000e+00]
 [1.96640000e+04 1.96640000e+02 1.37648000e+00]
 [1.39000000e+02 1.39000000e+00 9.73000000e-03]
 [3.08900000e+03 3.08900000e+01 2.16230000e-01]
 [2.04910000e+04 6.76203000e+02 4.73342100e+00]
 [2.15600000e+03 2.15600000e+01 1.50920000e-01]
 [1.27410000e+04 4.45935000e+02 3.12154500e+00]
 [1.19640000e+04 3.70884000e+02 2.59618800e+00]
 [2.13500000e+03 2.13500000e+01 1.49450000e-01]
print(y_train)
1 1 0 1 1 0 1 1
print(x_test)
[[1.7860000e+04 5.3580000e+02 3.7506000e+00]
 [1.7940000e+03 1.7940000e+01 1.2558000e-01]
 [1.2916100e+05 1.2399456e+03 8.6796192e+00]]
```

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```
print(y_test)
```

[1 0 1]

```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
```

#### print(x\_train)

```
[[-0.56125976 -0.42277215 -0.42277215]
 [-0.33285405 -0.29533013 -0.29533013]
[-0.07823282 0.03489671 0.03489671]
 [-0.36071681 -0.29841448 -0.29841448]
[-0.4398486 -0.40867447 -0.40867447]
 [-0.58222707 -0.43515512 -0.43515512]
 [-0.57900918 -0.43455663 -0.43455663]
 [-0.58056705 -0.43484638 -0.43484638]
 [-0.39583258 -0.33046984 -0.33046984]
 [-0.42526598 -0.33507867 -0.33507867]
                           0.0453631 ]
 0.70865404 0.0453631
 [ 3.17630916 5.01969665 5.01969665]
 [-0.57007063 -0.43289417 -0.43289417]
 [ 2.15529591  2.62142983  2.62142983]
 [-0.34179261 -0.2732816 -0.2732816 ]
 [-0.34785806 -0.34763358 -0.34763358]
 [ 1.27982794  0.25782584  0.25782584]
 [-0.53223498 -0.42392922 -0.42392922]
 [-0.35622199 -0.30836859 -0.30836859]
 [-0.38004964 -0.39755259 -0.39755259]
 [-0.44402418 -0.38340503 -0.38340503]
 [-0.33296897 -0.38879616 -0.38879616]
 [-0.58229091 -0.435167
                         -0.435167
 [-0.54462127 -0.4281609 -0.4281609 ]
 [-0.3224087 -0.2749025 -0.2749025 ]
 [-0.55653509 -0.43037673 -0.43037673]
 [-0.42137132 -0.32958993 -0.32958993]
 [-0.43129312 -0.34741414 -0.34741414]
[-0.55680325 -0.4304266 -0.4304266 ]
```

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```
print(x_test)

[[-0.35600491 -0.30824747 -0.30824747]
  [-0.56115761 -0.43123646 -0.43123646]
  [ 1.06523868 -0.14101664 -0.14101664]]

from sklearn.tree import DecisionTreeClassifier
  classifier = DecisionTreeClassifier(criterion = 'gini', random_state = 0)
  classifier.fit(x_train, y_train)
```

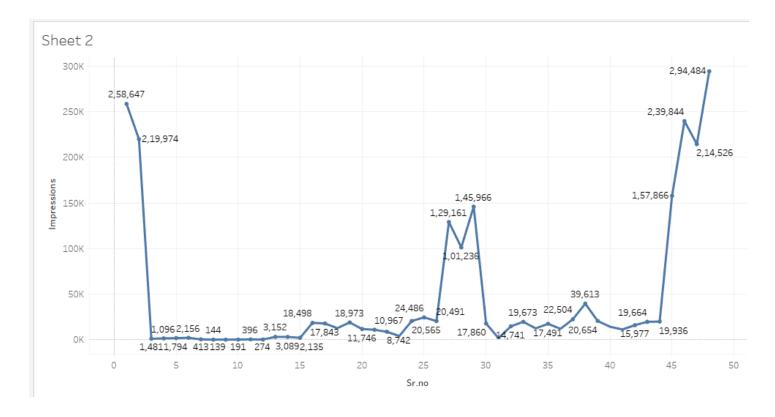
DecisionTreeClassifier

DecisionTreeClassifier(random\_state=0)

```
print(classifier.predict(sc.transform([[20491,676.2030,4.733421]])))
```

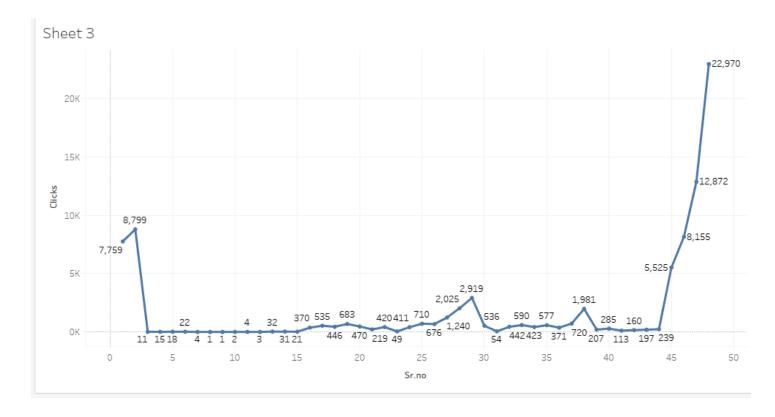
[1]

### **Data Visualisation**



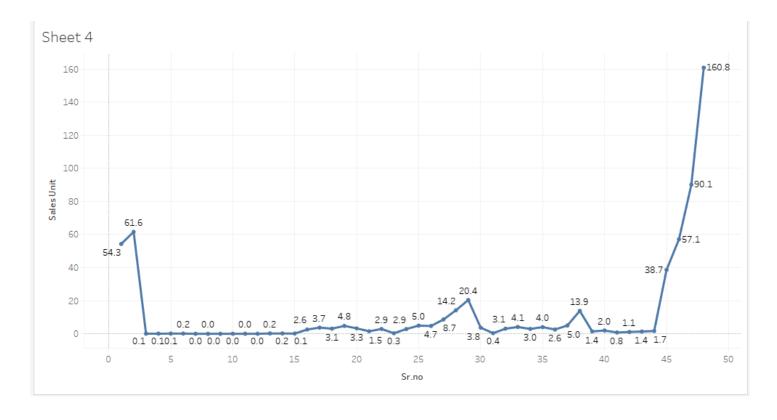
### Insights of the Impressions Graph

- Impressions are higher in the night at 12:00 am to 12:30 am.
- Impressions are going to down after 12:30 am because at night less amount of people are awake
- Impressions are less all most same in time period 1:00 am to 5:30 am
- From morning 6:00 am to afternoon 12:30 pm Impressions are little bit high because many people are wake up
- As campare to morning impressions afternoon and evening Impressions are high from 1pm to 9:30 pm
- Then 10:00 pm to 11:30 pm Impressions are going to high.



# Insights of the Clicks Graph

- Clicks are high at night 12:00 am to 12:30 am because after 11:00 pm every is going to seats on their career path. And searching for such courses.
- Then Clicks are all most going to down after 1:00 am to 7:00 am because their lot of people are sleeping
- After 7:30 am to 12:30 pm Clicks are going to high but in little bit amount because most of the people are busy in their work
- Then 1:00 pm to 2:00 pm Clicks are slightly higher than previous time because most of the people are in relaxing time
- After 2:30 pm to 9:30 pm Clicks are going down
- From 10:00 pm to 11:30 pm Clicks are going to high as campare to other times because every one is searching new courses for build up their career



### Insights of the Sales Unit Graph

- Sales unit are **high from 12:00 am to 12:30 am** because mainly people are usually seats on their career path and searching new courses and **enrolled the courses**
- From 1:00 am to 7:00 am their no one is enrolled for course
- From 7:30 am to 9:30 pm the sale is all most same but few people are enrolled
- Then sale is **going to high from 10:00 pm to 11:30 pm** than other time because more people are engaged with their online sites for **searching new things for career**

### Conclusion

#### #Conclusion

#There are 33 no of preferred time slot of 30 min in a entire day where it is preferable to show the Advertisement

```
#In late night 12 am to 12:30 am,
#In morning 7:30 am to 10:30 am,
#In late morning 11:30 am to 12:00 pm,
#In Early afternoon 12:00 pm to 2:30 pm,
#In Mid afternoon 3:30 pm to 4 pm,
#In Late afternoon 4 pm to 6 pm,
#In Evening 6 pm to 9 pm,
#In Late evening 9 pm to 11:30 pm
#These are the timing where copmany has received the sales
```

#Using above algorithm we have classify that we have successfully train the data and classified into 2 categerious #this timing is preferred and not preferred

#In this way we have used decision tree algorithm to classify time slots which are preferable for conversion campaign #This will help makketing to take decision about remarketing campaign