

AD Click Through Rate Analysis

November 23, 2023

1 AD Click Through Rate Analysis

Name: Keerthan Kumar C Manipal Institute of Technology Click-through rate is the ratio of users who clicked on an ad/link to the number of total users who left impressions on the ad/link. In simple words, $\text{clicks} \div \text{impressions} = \text{CTR}$. Analyzing the click-through rate helps companies determine the types of people most likely to click on their ads. A high CTR gives validation to your advertising strategies.

Below are all the features in the dataset:

Daily Time Spent on Site: the daily timespan of the user on the website;

Age: the age of the user;

Area Income: the average income in the area of the user;

Daily Internet Usage: the daily internet usage of the user;

Ad Topic Line: the title of the ad;

City: the city of the user;

Gender: the gender of the user;

Country: the country of the user;

Timestamp: the time when the user visited the website;

Clicked on Ad: 1 if the user clicked on the ad, otherwise 0;

2 Importing Libraries

```
[38]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

3 EDA

```
[39]: df=pd.read_csv("ad_10000records.csv")
df
```

```
[39]:      Daily Time Spent on Site   Age   Area Income   Daily Internet Usage \
0          62.26  32.0    69481.85          172.83
1          41.73  31.0    61840.26          207.17
2          44.40  30.0    57877.15          172.83
3          59.88  28.0    56180.93          207.17
4          49.21  30.0    54324.73          201.58
...
9995        41.73  31.0    61840.26          207.17
9996        41.73  28.0    51501.38          120.49
9997        55.60  39.0    38067.08          124.44
9998        46.61  50.0    43974.49          123.13
9999        46.61  43.0    60575.99          198.45
```

```
      Ad Topic Line      City  Gender \
0  Decentralized real-time circuit  Lisafort  Male
1  Optional full-range projection  West Angelabury  Male
2  Total 5thgeneration standardization  Reyesfurt  Female
3  Balanced empowering success  New Michael  Female
4  Total 5thgeneration standardization  West Richard  Female
...
9995  Profound executive flexibility  West Angelabury  Male
9996  Managed zero tolerance concept  Kennedyfurt  Male
9997  Intuitive exuding service-desk  North Randy  Female
9998  Realigned content-based leverage  North Samantha  Female
9999  Optimized upward-trending productivity  Port Jeffrey  Male
```

```
      Country      Timestamp  Clicked on Ad
0  Svalbard & Jan Mayen Islands  2016-06-09 21:43:05  0
1  Singapore  2016-01-16 17:56:05  0
2  Guadeloupe  2016-06-29 10:50:45  0
3  Zambia  2016-06-21 14:32:32  0
4  Qatar  2016-07-21 10:54:35  1
...
9995  Singapore  2016-01-03 03:22:15  1
9996  Luxembourg  2016-05-28 12:20:15  0
9997  Egypt  2016-01-05 11:53:17  0
9998  Malawi  2016-04-04 07:07:46  1
9999  Northern Mariana Islands  2016-04-03 21:13:46  1
```

[10000 rows x 10 columns]

```
[40]: df.shape
```

```
[40]: (10000, 10)
```

```
[41]: df.isna().sum()
```

```
[41]: Daily Time Spent on Site    0
      Age                      0
      Area Income              0
      Daily Internet Usage      0
      Ad Topic Line            0
      City                    0
      Gender                  0
      Country                 0
      Timestamp               0
      Clicked on Ad           0
      dtype: int64
```

```
[42]: df.describe()
```

```
[42]:
```

	Daily Time Spent on Site	Age	Area Income \
count	10000.000000	10000.000000	10000.000000
mean	61.660757	35.940100	53840.047721
std	15.704142	8.572973	13343.708718
min	32.600000	19.000000	13996.500000
25%	48.860000	29.000000	44052.302500
50%	59.590000	35.000000	56180.930000
75%	76.580000	42.000000	61840.260000
max	90.970000	60.000000	79332.330000

	Daily Internet Usage	Clicked on Ad
count	10000.000000	10000.000000
mean	177.759831	0.491700
std	40.820951	0.499956
min	105.220000	0.000000
25%	140.150000	0.000000
50%	178.920000	0.000000
75%	212.670000	1.000000
max	269.960000	1.000000

```
[43]: df.dtypes
```

```
[43]: Daily Time Spent on Site    float64
      Age                      float64
      Area Income              float64
      Daily Internet Usage      float64
      Ad Topic Line            object
      City                    object
      Gender                  object
      Country                 object
```

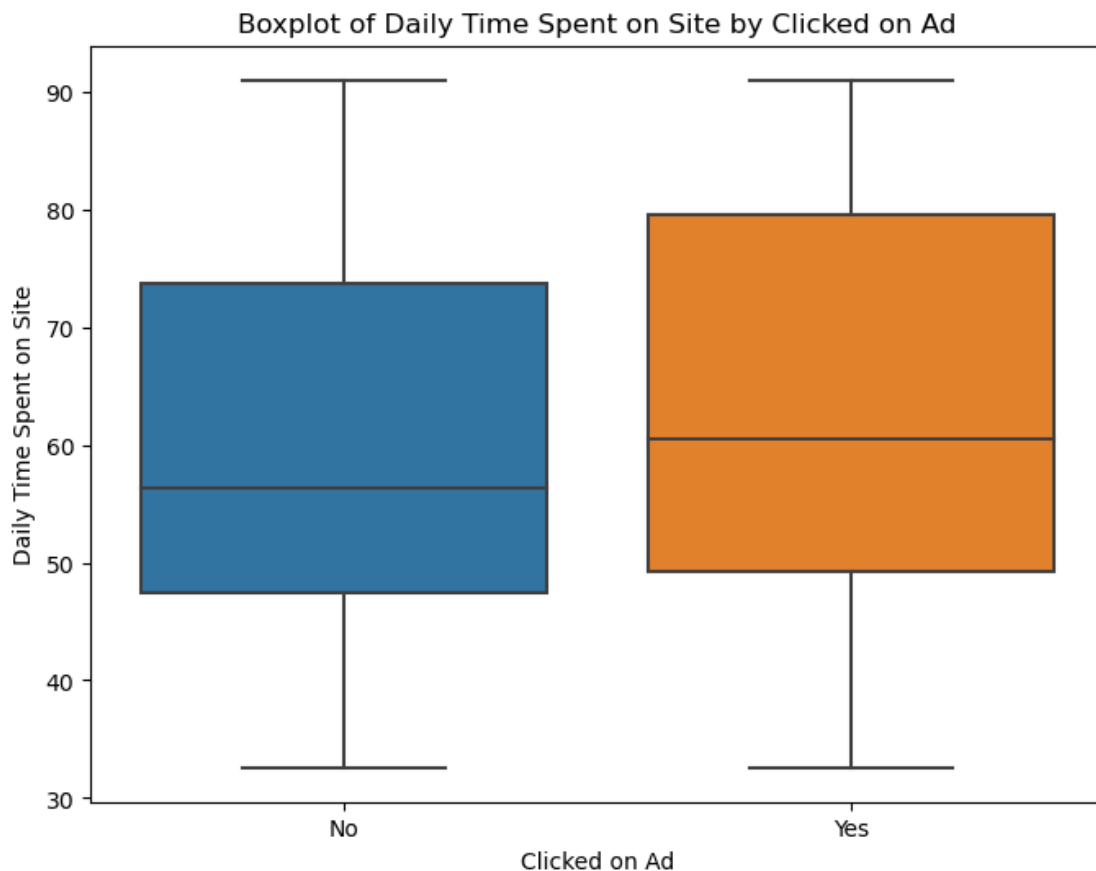
```
Timestamp                object
Clicked on Ad             int64
dtype: object
```

```
[45]: df["Clicked on Ad"] = df["Clicked on Ad"].map({1: "Yes", 0: "No"})
```

4 Click Through Rate Analysis

```
[46]: def boxplot_compare(df, col1):
    plt.figure(figsize=(8, 6))
    sns.boxplot(x='Clicked on Ad', y=col1, data=df)
    plt.title(f'Boxplot of {col1} by Clicked on Ad')
    plt.xlabel('Clicked on Ad')
    plt.ylabel(col1)
    plt.show()
```

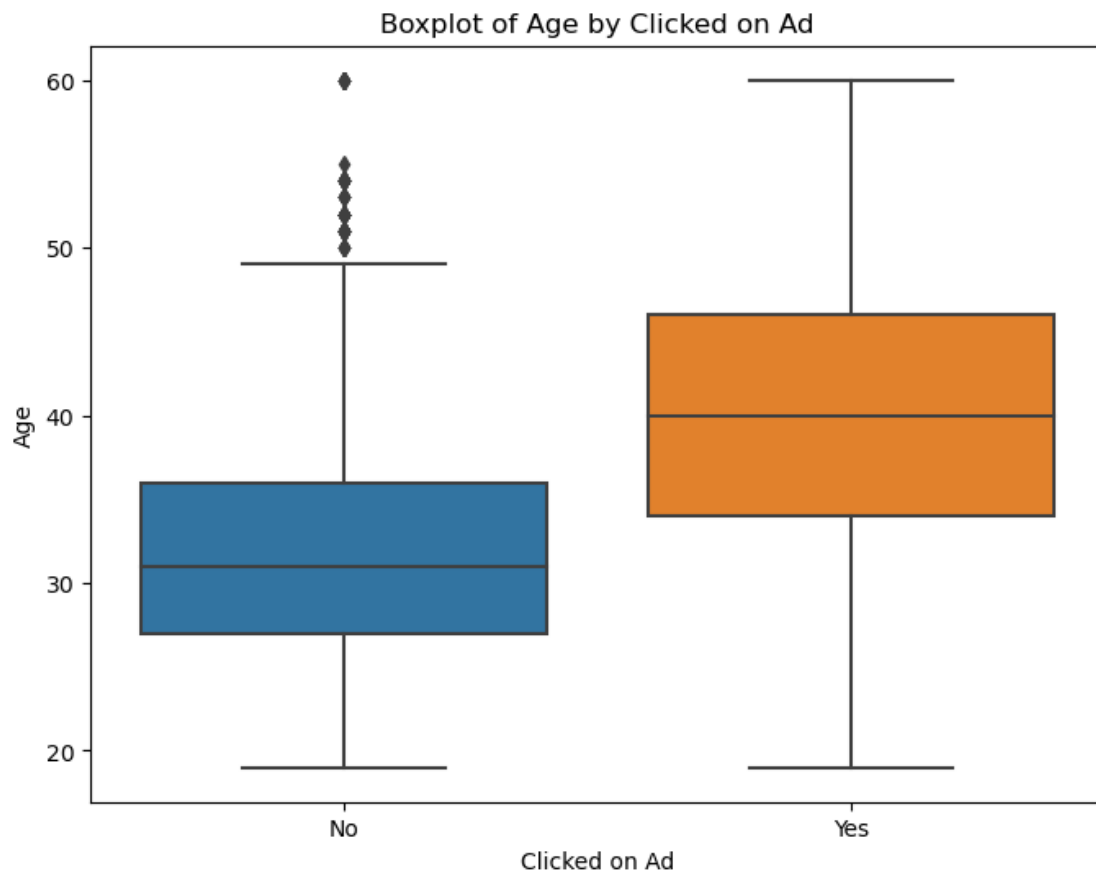
```
[47]: boxplot_compare(df, "Daily Time Spent on Site")
```



From the above graph, we can see that the users who spend more time on the website click more

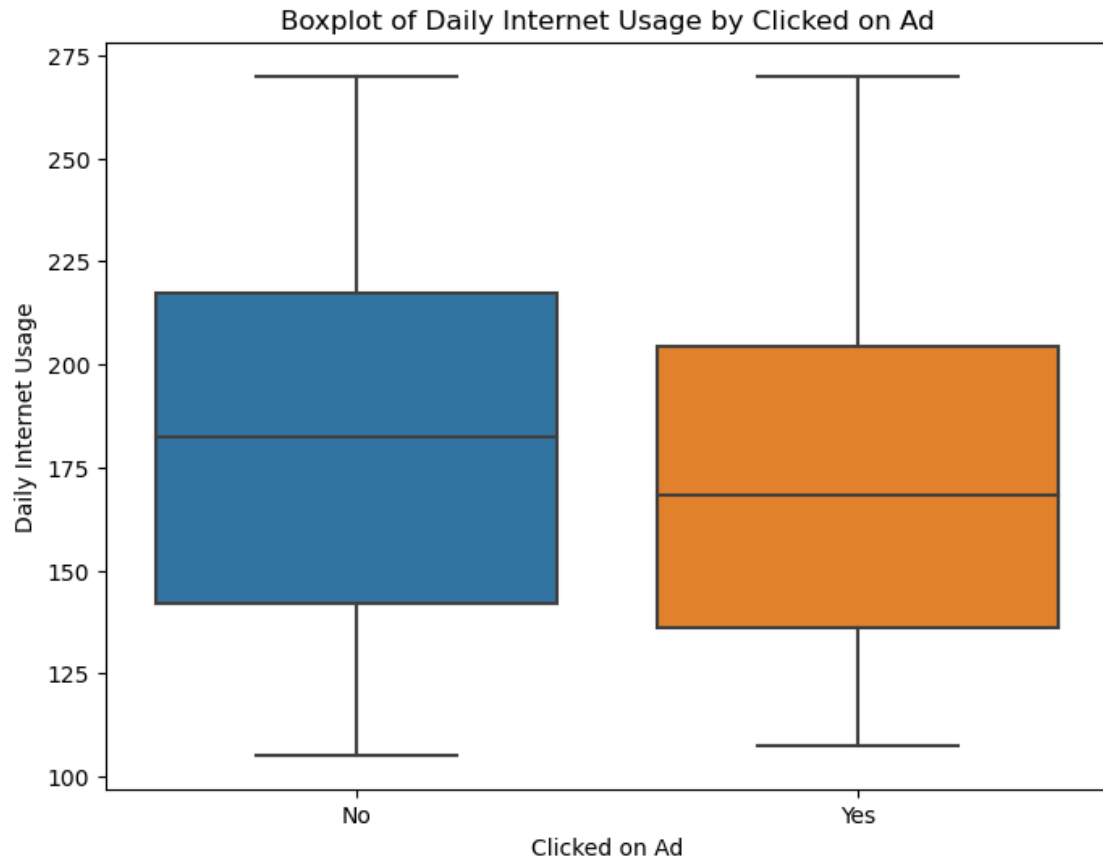
on ads.

```
[48]: boxplot_compare(df, "Age")
```



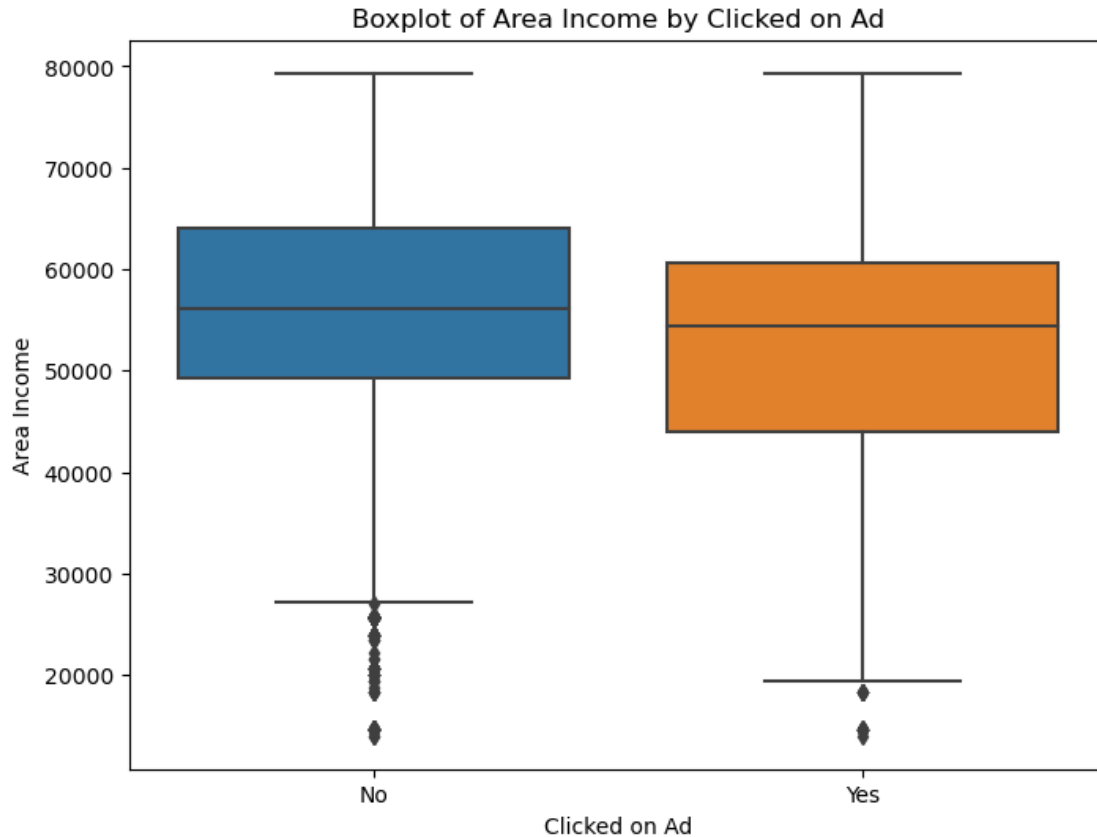
From the above graph, we can see that users around 40 years click more on ads compared to users around 27-36 years old.

```
[49]: boxplot_compare(df, "Daily Internet Usage")
```



From the above graph, we can see that the users with high internet usage click less on ads compared to the users with low internet usage.

```
[50]: boxplot_compare(df, "Area Income")
```



There's not much difference, but people from high-income areas click less on ads.

```
[54]: a=pd.crosstab(df["Gender"],df["Clicked on Ad"])
      a["Gender CTR"]=(a["Yes"]/(a["Yes"]+a["No"]))*100
      a
```

```
[54]: Clicked on Ad    No   Yes  Gender CTR
      Gender
      Female         2609  2767   51.469494
      Male          2474  2150   46.496540
```

We See that Female generally Have a High CTR than Men(we know why :)

5 Calculating CTR of Ads

Now let's calculate the overall Ads click-through rate. Here we need to calculate the ratio of users who clicked on the ad to users who left an impression on the ad.

```
[60]: x=df["Clicked on Ad"].value_counts()[1]
      ctr=(x/df.shape[0])*100
      print("CTR is:",ctr)
```

CTR is: 49.17

6 Click Through Rate Prediction Model

Now let's move on to training a Machine Learning model to predict click-through rate. I'll start by dividing the data into training and testing sets:

```
[55]: df["Gender"] = df["Gender"].map({"Male": 1,
                                       "Female": 0})

x=df.iloc[:,0:7]
x=x.drop(['Ad Topic Line','City'],axis=1)
y=df.iloc[:,9]

from sklearn.model_selection import train_test_split
xtrain,xtest,ytrain,ytest=train_test_split(x,y,
                                           test_size=0.2,
                                           random_state=4)
```

Now let's train the model using the random forest classification algorithm:

```
[56]: from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier()
model.fit(x, y)
```

```
[56]: RandomForestClassifier()
```

Now let's have a look at the accuracy of the model:

```
[61]: from sklearn.metrics import accuracy_score
y_pred=model.predict(xtest)
print("The accuracy is: ",(accuracy_score(ytest,y_pred))*100)
```

The accuracy is: 95.7

7 Now let's test the model by making predictions:

```
[58]: print("Ads Click Through Rate Prediction : ")
a = float(input("Daily Time Spent on Site: "))
b = float(input("Age: "))
c = float(input("Area Income: "))
d = float(input("Daily Internet Usage: "))
e = input("Gender (Male = 1, Female = 0) : ")

features = np.array([[a, b, c, d, e]])
print("Will the user click on ad = ", model.predict(features))
```

Ads Click Through Rate Prediction :
Daily Time Spent on Site: 70.20


```
Age: 35
Area Income: 40000
Daily Internet Usage: 220
Gender (Male = 1, Female = 0) : 1
Will the user click on ad = ['No']

C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarning: X
does not have valid feature names, but RandomForestClassifier was fitted with
feature names
  warnings.warn(
```

8 Summary

Ads CTR means predicting whether the user will click on the ad. In this task, we need to train a Machine Learning model to find relationships between the characteristics of all the users who click on ads.