# ADVERTISING PROJECT

**USING** 



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# **EXECUTIVE SUMMARY**

### **Advertising Project:**

#### Introduction:

In this project, we will be working with advertising data set, indicating whether a particular internet user clicked on an Advertisement on a company website. We will try to create a model that will predict whether they will click on an ad based off the features of that user.

The dataset contains 1000 data points collected from an Advertising Train Dataset and 200 from Advertising Test Dataset all together it has 1200 observations.

#### **Data Set Variables Information:**

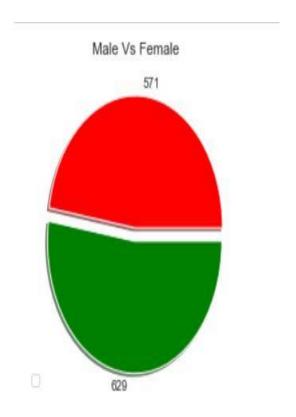
- **❖** Daily Time Spent on Site': consumer time on site in minutes
- ❖ 'Age': customer age in years
- ❖ 'Area Income': Avg. Income of geographical area of consumer
- ❖ 'Daily Internet Usage': Avg. minutes a day consumer is on the internet
- ❖ 'Ad Topic Line': Headline of the advertisement
- **❖** 'City': City of consumer
- **❖** 'Male': Whether or not consumer was male
- **❖** 'Country': Country of consumer
- ❖ 'Timestamp': Time at which consumer clicked on Ad or closed window
- ❖ Clicked on Ad': 0 or 1 indicated clicking on Ad

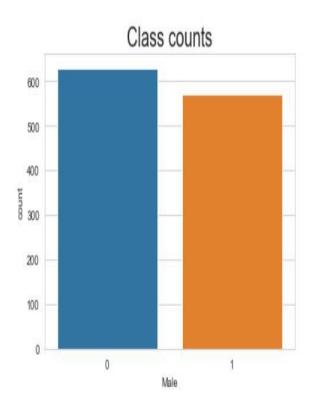
```
ad_data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1200 entries, 0 to 1199
Data columns (total 11 columns):
# Column
                           Non-Null Count Dtype
                           Ad Topic Line
                          1200 non-null object
0
                           1200 non-null int64
1
   Age
   Area Income
                           1200 non-null float64
2
3
   City
                          1200 non-null object
   Clicked on Ad
4
                           1000 non-null float64
5
   Country
                         1200 non-null object
   Daily Internet Usage 1200 non-null float64
7
   Daily Time Spent on Site 1200 non-null float64
   Male
                           1200 non-null int64
   Timestamp
9
                           1200 non-null object
10 source
                           1200 non-null
                                        object
dtypes: float64(4), int64(2), object(5)
memory usage: 103.2+ KB
```

#### **OBJECTIVE OF THE PROJECT:**

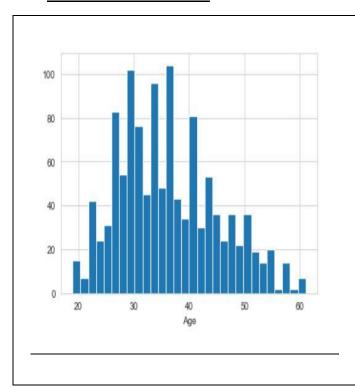
- > Analyzing Advertising Dataset.
- > Analyzing different model in Classification such as Logistic Regression, Random Forest Model, Support Vector Machine, KNN & Decision Tree.
- > Splitting the dataset into Training set and Test sets, Scaling the dataset
- > Training and model on Test sets.
- > Evaluating Accuracy, Confusion Matrix, and Micro Recall on classification mode.
- > Performing K-Fold on Logistic Regression models.

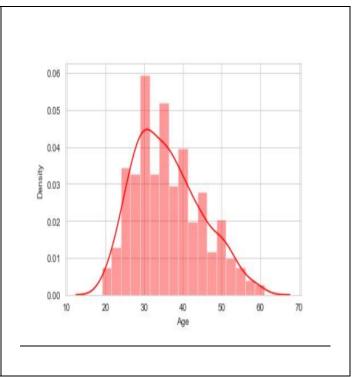
#### **ANALYSIS OF GENDER:**

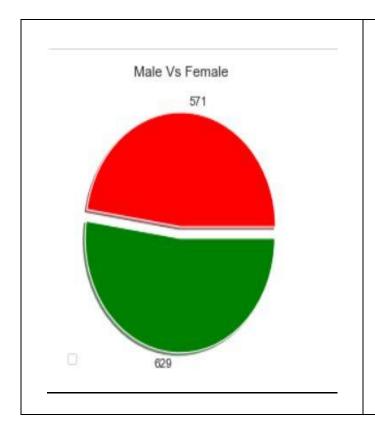




# ANALYSIS OF AGE:

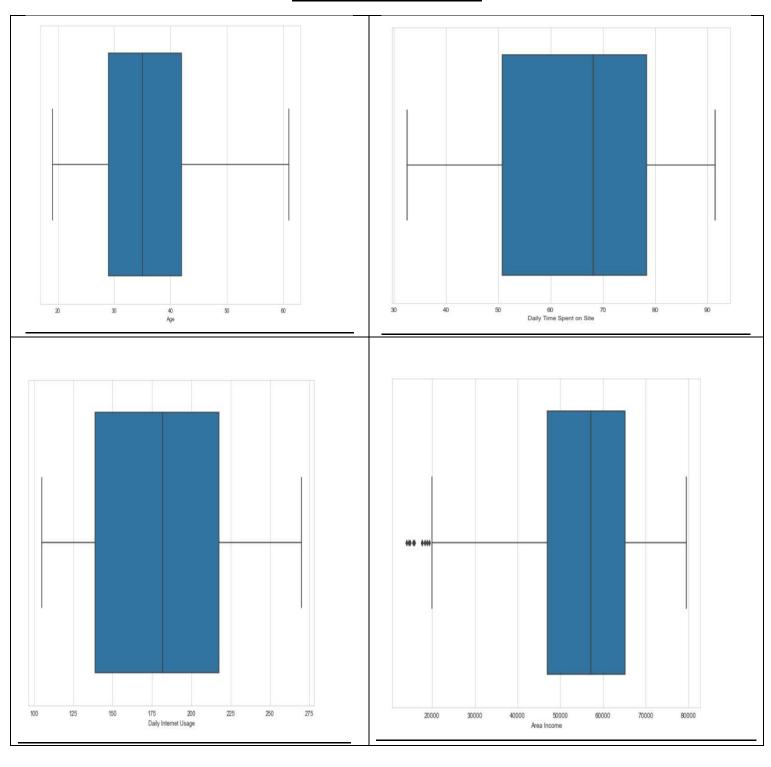




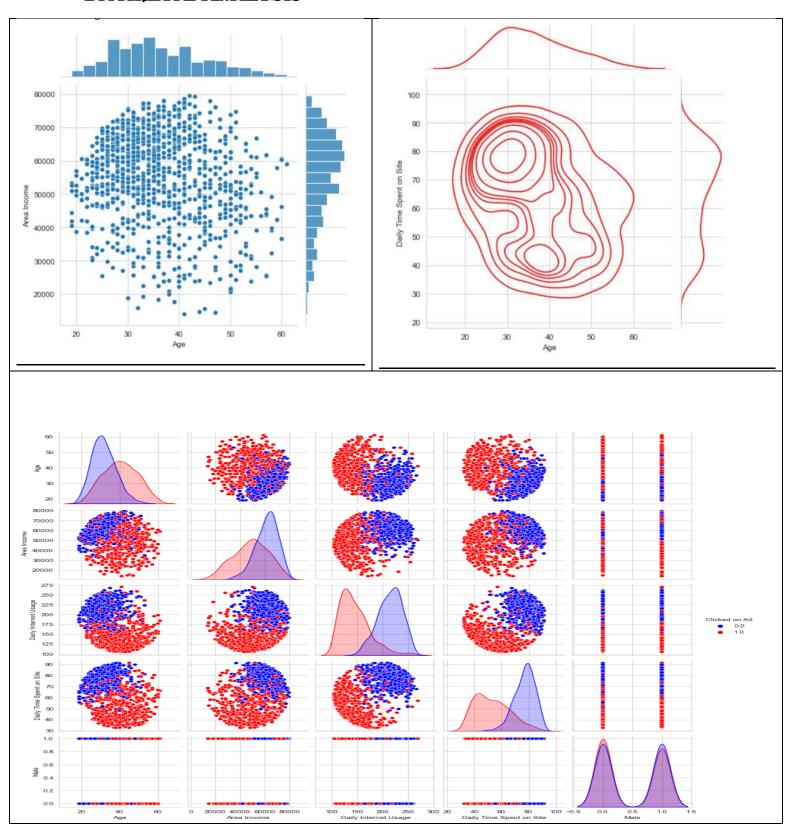




# **OUTLIER ANALYSIS**



# **BIVARIATE ANALYSIS**



# **CLASSIFICATION:**

# ACCURACY SCORE OF LOGISTIC REGRESSION MODEL

Create a classific	cation report	for the mo	odel.		
from sklearn.m	etrics <mark>impo</mark>	rt classi	fication_re	eport, confusion_matrix	
print(classifi	cation_repo	rt(y_test	,y_pred))		from sklearn.metrics import accuracy score
	precision	recall	f1-score	support	<pre>print(accuracy_score(y_test, y_pred))</pre>
0	0.97	0.97	0.97	164	
1	0.96	0.96	0.96	136	0.966666666666667
accuracy			0.97	300	
macro avg	0.97	0.97	0.97	300	
weighted avg	0.97	0.97	0.97	300	

# **ACCURACY SCORE OF KNN**

Classification I	Report					
<b>from</b> sklearn. print (classi				eport	Accuracy	
	precision	recall	f1-score	support		
	0.60	0.60	0.60	4.64	<pre>accuracy_score(y_test, pred)</pre>	
0 1	0.69	0.68	0.69	164	/(20) (0.5457)	
1	0.62	0.64	0.63	136	0.66	
accuracy			0.66	300		
macro avg	0.66	0.66	0.66	300		
weighted avg	0.66	0.66	0.66	300		

### **ACCURACY SCORE OF RANDOM FOREST**

# **Classification Report**

from sklearn.metrics import classification\_report
print (classification\_report(y\_test,rfc\_pred))

	precision	recall	f1-score	support	
0	0.96	0.95	0.95	164	
1	0.94	0.95	0.95	136	
accuracy			0.95	300	
macro avg	0.95	0.95	0.95	300	
weighted avg	0.95	0.95	0.95	300	

#### Accuracy

accuracy\_score(y\_test, rfc\_pred)

0.95

# **ACCURACY SCORE OF SVM**

#### **Classification Report**

print(classification\_report(y\_test,predictions))

	precision	recall	f1-score	support
	•			
0	0.97	0.96	0.97	164
1	0.96	0.96	0.96	136
accuracy			0.96	300
macro avg	0.96	0.96	0.96	300
weighted avg	0.96	0.96	0.96	300

#### Accuracy

accuracy score(y test, predictions)

0.96333333333333334

#### **ACCURACY SCORE OF DECISION TREE**

rint(classification_report(y_test,predictions_dtree))					
р	recision	recall	f1-score	support	Accuracy
0	0.95	0.90	0.92	164	accuracy company test prodictions dtrop
1	0.88	0.95	0.91	136	accuracy_score(y_test, predictions_dtree
accuracy			0.92	300	0.92
nacro avg	0.92	0.92	0.92	300	
ighted avg	0.92	0.92	0.92	300	

#### **CONCLUSION:**

- **❖** During our analysis, we noticed that most of the Advertisements are clicked by Internet users who ages are between 25 − 45
- **❖** Most the internet user who clicked the advertisements were females approximately 55%
- **❖** We have built five different model whose accuracy score are as follows:

**Logistic Regressions** : 97%

K Nearest Neighbors : 66%

Random Forest : 95%

SVM : 96.33%

Decision Tree : 92%

**❖** We conclude that the Logistic Regression is the best model followed by SVM.

# THANK YOU...