# Clicked Ads Analyst Using Machine Learning

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**Business Understanding** 

#### **Problem Statement**

"A company in Indonesia wants to know the effectiveness of an advertisement that they run, this is important for the company so they can find out how much achievement the advertisements marketed they can attract customers to see advertisements.

By processing historical advertisement data and finding insights and patterns that occur, it can help companies determine marketing targets, the focus of this case is to create a machine learning classification model that functions to determine the right target customers "

## Goals

Create a machine learning model that can detect potential users to convert or interested in an advertisement. So that we can optimize the cost of advertising on digital platforms.

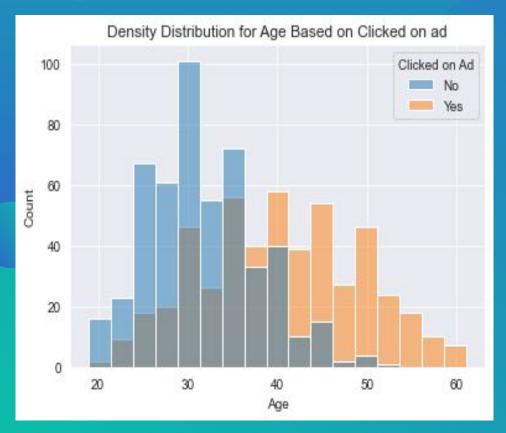


**Exploratory Data Analyst** 

# Data Overview

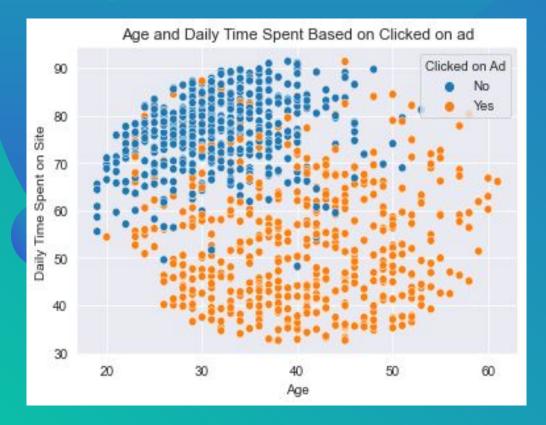
Variable	Description		
Daily Time Spent On Site	Length stay on site in minutes		
Age	User Age		
Area Income	Income User		
Daily Internet Usage	Daily internet usage		
Male	User Gander		
Tlmestamp	Time user visit the website		
Clicked on Ad	Ads clicked by user		
City	City user		
Province	Province		
Category	Product Category		





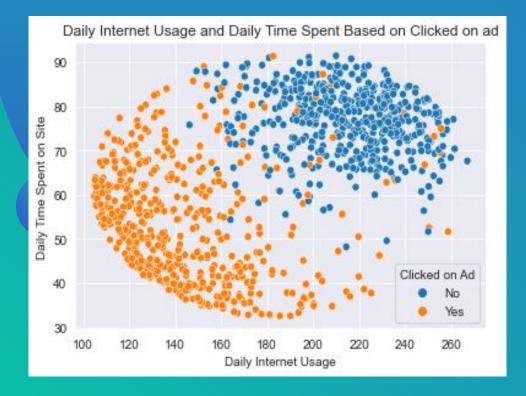
Based on the picture beside, it can be seen that users who are over 40 years old tend to like advertisements compared to users who are under 40 years old.





Users under the age of 40 who spend more time on the website are less likely to be interested in advertising.





Users who have daily internet usage are less likely to like advertisements compared to those who have longer daily internet usage.





Based on the plot, there are 2 segments that define users, namely low income users and high income users.

Low Income users tend to be interested in advertisements while High Income users tend not to be interested in advertisements.



**Model Machine Learning** 

## **Data Preparation**

**Data Split Train** Feature Encoding Test **Data Cleansing Data Scalling** 

## **Data Modelling**

Model	Accuracy (Train)	Accuracy (Test)	Precision (Train)	Precision (Test)	Recall (Train)	Recall (Test)
Decision Tree	1.00	0.64	1.00	0.62	1.00	0.78
Random Forest	0.97	0.93	0.97	0.93	0.97	0.94
XGBoost	0.99	0.94	0.99	0.95	0.98	0.94

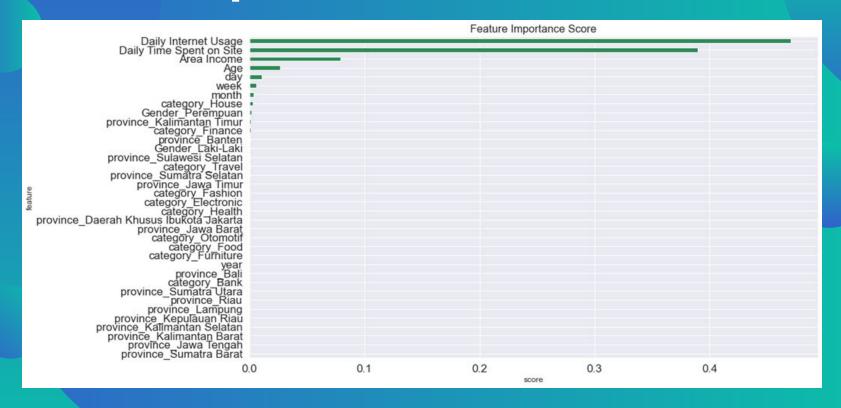
Based on the results obtained after creating the model, XGBoost is the model chosen because it has the Recall Train and Recall Test values with the least difference compared to other models.

## **Confusion Metrics**



For this case, we used Recall..

#### **Feature Importance**



Features Daily Internet Usage, Daily Time Spent on Site, Area Income, Age have the highest score of feature importance compared to other features.



Recommendation

#### Based on Data, we can conclude that:

- 1. The data we get has 2 user segments, namely the upper class user segment and the lower class user segment.
- The upper class has the criteria of frequently using the internet, frequently visiting the website of a product, having a relatively young age and high income.
- The lower class has the opposite nature.
- 2. Users with lower economic class tend to be more easily attracted to click products on digital ads.
- 3. Users who are very heavy in using the internet are even more difficult to be given advertising because they may already be used to digital ads.
- 4. Parents are a potential market for the digital market.

#### Business takeaway:

- 1. We can use a more unique method (soft selling) so that it is less visible to advertise by users.
- 2. Use mainstream content (simple but a topic of conversation) in order to attract users with lower class segments.