

Abstract :

This project will be based on classification problem to predict customer online shopping intention. we aim to help companies to be successful in a highly competitive eCommerce environment by predict whether the customer, visiting web pages of an online shop, will end up with a purchase or not. The dataset contains 12330 observation and 18 features divided into 80% training and 20% testing.

I have used three classification models which are logistic regression, Random Forest and decision tree .The results show The random forest outperform the other model in accuracy , precision, recall and F1 Score with 90%91%, 98%,94% respectively for not purchase . While for purchasing precision is is 84%, recall is 57% ,F1 score is 68%.

Design:

The dataset consists of 10 numerical and 8 categorical attributes. I used five attributes (Month , operating system, Browser , Visitor type , Weekend, Revenue) in Eda part and in model training. The 'Revenue' attribute is used as class label. The dataset is clean, there are no missing values, but the dataset is unbalanced. There is a risk of bias, so the analysis must take the unbalanced dataset into consideration.

Algorithms:

I have used Random Forest, Logistic Regression and decision, the results showing in the figure below:

	precision	recall	f1-score	support
False	0.91	0.98	0.94	985
True	0.84	0.57	0.68	215
accuracy			0.90	1200
macro avg	0.88	0.77	0.81	1200
weighted avg	0.90	0.90	0.90	1200

Figure 1 : Random forest performance

precision, recall and F1 Score with 91%, 98%,94% respectively for not purchase . While for purchasing the precision is 84%, recall is 57% ,F1 score is 68% .

	precision	recall	f1-score	support
False	0.91	0.91	0.91	985
True	0.58	0.59	0.58	215
accuracy			0.85	1200
macro avg	0.75	0.75	0.75	1200
weighted avg	0.85	0.85	0.85	1200

Figure 2 : Decision Tree performance

precision, recall and F1 Score with 91%,91%, 91%respectively for not purchase . While for purchasing precision is 58%, recall is 59% ,F1 score is 58% .

	precision	recall	f1-score	support
False	0.86	0.99	0.92	985
True	0.82	0.28	0.42	215
accuracy			0.86	1200
macro avg	0.84	0.64	0.67	1200
weighted avg	0.86	0.86	0.83	1200

Figure 3 : Logistic Regression performance

Tool :

- Numpy and Pandas for data manipulation
- Scikit-learn for modeling, confusion matrix and feature extraction
- Matplotlib and Seaborn for plotting

