

Title:**Predicting Product Returns Using Machine Learning****Submitted By:****Name:** Samarth Shukla**Roll No.:** 202401100300212**Course:** Intro to AI**Instructor:** Shivansh Sir**Date:** 13/05/2025**Problem Statement:**

The goal of this project is to predict whether a product will be returned based on purchase-related features such as amount spent, review score, and delivery time using machine learning.

In e-commerce, product returns lead to financial losses and logistical challenges. Being able to **predict the likelihood of a return** helps companies:

- Reduce unnecessary shipment costs.
- Improve inventory planning.
- Identify potentially unsatisfied customers early.

This project uses a dataset containing features like:

- Purchase amount
- Customer review score
- Days taken for delivery
- Return status (Yes/No)

The objective is to train a machine learning classifier that can predict the return status using these numeric inputs.

c. Methodology**1. Data Upload & Exploration:**

- The dataset (`product_return.csv`) was uploaded in Google Colab.
- It was found to have no missing values.
- All features were numerical except the target column (`returned`), which was categorical.

2. Preprocessing:

- The target variable was label-encoded (Yes = 1, No = 0).
- No additional text or categorical features were present.

3. Model Selection:

- A **Random Forest Classifier** was chosen for its high accuracy and ability to handle numeric data well.

4. Training and Evaluation:

- Data was split into 80% training and 20% testing.
- Evaluation metrics included **confusion matrix** and **classification report**.

5. Prediction:

- A custom function was implemented to predict returns based on new input data.

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Commands + Code + Text

Files

Analyze your files with code written by Gemini Upload

sample_data

product_return (1).csv

product_return (2).csv

product_return.csv

```
from google.colab import files
uploaded = files.upload()

product_return.csv (text/csv) - 4315 bytes, last modified: 5/13/2025 - 100% done
Saving product_return.csv to product_return (2).csv

[5] import pandas as pd
import io

# Replace the filename with the exact name you see in uploaded.keys()
filename = list(uploaded.keys())[0]
df = pd.read_csv(io.BytesIO(uploaded[filename]))

# View the first few rows
df.head()
```

	purchase_amount	review_score	days_to_delivery	returned
0	687.011818	3.778615	4	no
1	325.972093	2.458683	1	yes
2	685.382724	3.954024	7	no
3	291.100577	3.666468	14	yes
4	209.806672	1.478248	2	no

Next steps: Generate code with df View recommended plots New interactive sheet

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```
print("Columns:", df.columns.tolist())
print("\nData types:\n", df.dtypes)
print("\nMissing values:\n", df.isnull().sum())

Columns: ['purchase_amount', 'review_score', 'days_to_delivery', 'returned']

Data types:
purchase_amount    float64
review_score        float64
days_to_delivery   int64
returned            object
dtype: object

Missing values:
purchase_amount    0
review_score        0
days_to_delivery   0
returned            0
dtype: int64

[7] from sklearn.preprocessing import LabelEncoder

# Define features and target
X = df[['purchase_amount', 'review_score', 'days_to_delivery']].values
y = df['returned'].values

# Encode target variable
label_encoder = LabelEncoder()
y = label_encoder.fit_transform(y)

[8] from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
```

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```
[8] from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier

# Split data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Train model
clf = RandomForestClassifier(n_estimators=100, random_state=42)
clf.fit(X_train, y_train)
```

RandomForestClassifier

```
[9] from sklearn.metrics import classification_report, confusion_matrix

y_pred = clf.predict(X_test)

print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test, y_pred, target_names=label_encoder.classes_))
```

Confusion Matrix:

```
[[3 6]
 [3 8]]
```

Classification Report:

	precision	recall	f1-score	support
no	0.50	0.33	0.40	9
yes	0.57	0.73	0.64	11
accuracy			0.55	20
macro avg	0.54	0.53	0.52	20

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yes	0.57	0.73	0.64	11
accuracy			0.55	20
macro avg	0.54	0.53	0.52	20
weighted avg	0.54	0.55	0.53	20

```
[10] def predict_return(purchase_amount, review_score, days_to_delivery):
    sample = [[purchase_amount, review_score, days_to_delivery]]
    pred = clf.predict(sample)[0]
    return label_encoder.inverse_transform([pred])[0]

# Example usage:
# predict_return(120.0, 2.5, 7)
```

Start coding or generate with AI.

Variables Terminal

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