Bank Marketing Customer Purchase Prediction

(INSIGHTS)

Project predicting customer purchase behavior using the Bank Marketing dataset, with EDA and a Decision Tree model. Here are the key insights:

1) Dataset Overview:

- a) Contains 45,211 records and 17 features, with both numeric and categorical data.
- b) No missing values in the dataset.

2) Exploratory Data Analysis (EDA):

- a) Initial exploration included summary statistics, data type checks, and null value assessments.
- b) Class distribution analysis showed an imbalance in the target variable ("y").
- c) Correlation heatmap highlighted the relationships between features, guiding feature selection.

3) Data Preprocessing:

- a) Categorical variables (e.g., job, marital, education) were encoded using Label Encoding.
- b) Features like "duration" and "campaign" displayed high variance, potentially impacting predictions.

4) Model Building:

- a) Split data into training and test sets (70%-30% split).
- b) Trained a Decision Tree Classifier with a max depth of 5 to prevent overfitting.

5) Model Evaluation:

- a) Accuracy: 89.5% on the test set.
- b) Confusion Matrix: Showed the model performed well in identifying "No" (did not purchase) cases but struggled with "Yes" (purchased) cases.
- c) Classification Report:
 - i) Precision: 92% for "No", 58% for "Yes".
 - ii) Recall: 96% for "No", 40% for "Yes".
 - iii) F1-Score: 94% for "No", 47% for "Yes".
- d) The imbalance in target classes impacted recall for the "Yes" class.

6) **Decision Tree Visualization:**

a) The decision tree plot showed key splits, with features like "duration" and "pdays" playing major roles in the decision-making process.

7) Insights:

- a) Imbalance in the dataset suggests the need for techniques like SMOTE or cost-sensitive learning.
- b) Duration was a strong predictor of purchase behavior.
- c) Model performance could improve with hyperparameter tuning or ensemble methods like Random Forest or Gradient Boosting.