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# === AI Loan Approver (v2 SAFE) ===
# Paste into a fresh Google Colab cell and run.
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
# Reproducibility
SEED = 42
rng = np.random.default_rng(SEED)
# 1) Generate synthetic data (balanced classes by design)
n = 800
income
                 = rng.normal(3000, 1200, n).clip(500, 12000)
                                                                   # monthly USD
                                                                  # FICO-like
                 = rng.normal(650, 80, n).clip(300, 850)
credit score
debt_to_income = rng.uniform(0.05, 0.6, n)
                                                                   # ratio
employment_years = rng.integers(0, 15, n)
                                                                   # vears
loan_amount
                 = rng.normal(8000, 5000, n).clip(1000, 30000)
                                                                   # USD
# Hidden scoring rule -> then we threshold at the median so both classes exist
score = (
   0.003 * income
   + 0.01 * (credit_score - 600)
    - 1.5 * debt_to_income
   + 0.05 * employment_years
    - 0.00003 * loan_amount
    + rng.normal(0, 0.30, n)
)
threshold = np.median(score)
y = (score > threshold).astype(int) # roughly 50/50 (1=Approve, 0=Reject)
X = pd.DataFrame({
    "income": income,
    "credit_score": credit_score,
    "debt_to_income": debt_to_income,
    "employment_years": employment_years,
    "loan_amount": loan_amount
})
# 2) Train/test split (stratify keeps the 0/1 balance)
Xtr, Xte, ytr, yte = train_test_split(
   X, y, test_size=0.25, random_state=SEED, stratify=y
)
# 3) Train model
model = RandomForestClassifier(n_estimators=200, random_state=SEED)
model.fit(Xtr, ytr)
# 4) Evaluate (labels specified so reports never crash even if a class is rare)
pred = model.predict(Xte)
print("Accuracy:", round(accuracy_score(yte, pred), 3))
print("\nClassification Report:\n",
      classification_report(
         yte, pred,
          labels=[0, 1],
         target_names=["Reject", "Approve"],
          zero_division=0
      ))
print("\nConfusion Matrix:\n",
      confusion_matrix(yte, pred, labels=[0, 1]))
# 5) Simple predictor function
def approve_loan(income, credit_score, dti, years, loan_amount):
    x = np.array([[income, credit_score, dti, years, loan_amount]])
    p = model.predict_proba(x)[0, 1]
    return ("Approve" if p >= 0.5 else "Reject", round(float(p), 2))
# Example call
print("\nExample:", approve_loan(4000, 700, 0.25, 3, 10000))
Accuracy: 0.945
```

Classification	Report: precision	recall	f1-score	support
Reject	0.96	0.93	0.94	100
Approve	0.93	0.96	0.95	100
accuracy			0.94	200
macro avg	0.95	0.95	0.94	200
weighted avg	0.95	0.94	0.94	200

Confusion Matrix: [[93 7] [4 96]]

Example: ('Approve', 0.99)
/usr/local/lib/python3.12/dist-packages/sklearn/utils/validation.py:2739: UserWarning: X does not have valid feature names, but Ran warnings.warn(

Start coding or <u>generate</u> with AI.