```
[1]: !pip install scikit-learn
      Requirement already satisfied: scikit-learn in c:\users\vinatha\appdata\local\programs\python\python313\lib\site-packages (1.7.2)
      Requirement already satisfied: numpy>=1.22.0 in c:\users\vinatha\appdata\local\programs\python\python313\lib\site-packages (from scikit-learn) (2.3.3)
      Requirement already satisfied: scipy>=1.8.0 in c:\users\vinatha\appdata\local\programs\python\python313\lib\site-packages (from scikit-learn) (1.16.2)
      Requirement already satisfied: joblib>=1.2.0 in c:\users\vinatha\appdata\local\programs\python\python313\lib\site-packages (from scikit-learn) (1.5.2)
Requirement already satisfied: threadpoolctl>=3.1.0 in c:\users\vinatha\appdata\local\programs\python\python313\lib\site-packages (from scikit-learn)
      (3.6.0)
[2]: import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
      import seaborn as sns
      from sklearn.model_selection import train_test_split
      from sklearn.tree import DecisionTreeClassifier, plot_tree
      from sklearn.metrics import classification_report, accuracy_score, confusion_matrix
[8]: df = pd.read_csv("bank.csv", delimiter=";")
[9]: print(df.head())
     print(df.info())
                      job
                           marital education default balance housing loan
               unemployed married
         30
                                      primary
                                                   no
                                                           1787
                                                                      no
                                                                          no
                 services married secondary
         33
                                                    no
                                                            4789
      1
                                                                     yes yes
               management
                            single
                                     tertiary
                                                     no
                                                                     yes
                                                                           no
          30
               management
                           married
                                      tertiary
                                                     no
                                                            1476
                                                                     yes yes
         59 blue-collar married secondary
                                                                           no
                                                                     yes
                                                    no
          contact day month duration campaign pdays previous poutcome
        cellular 19 oct
                                     79
                                                                     unknown no
                                                       -1
                                                                  0
                                                1
         cellular
                    11
                         may
                                    220
                                                      339
                                                                     failure no
                                                                  1 failure no
         cellular
                    16
                         apr
                                    185
                                                     330
          unknown
                                                                     unknown
                         jun
         unknown
                     5
                         mav
                                    226
                                                      -1
                                                                  e unknown no
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 4521 entries, 0 to 4520
      Data columns (total 17 columns):
      # Column
                      Non-Null Count Dtype
      0
           age
                      4521 non-null
                                       int64
           job
                      4521 non-null
                                       object
           marital
                      4521 non-null
                                       object
           education 4521 non-null
                                       object
           default
                      4521 non-null
                                       object
           balance
                      4521 non-null
                                       int64
           housing
                      4521 non-null
                                       object
                      4521 non-null
                                       object
       8
           contact
                      4521 non-null
                                       object
                      4521 non-null
                                       int64
           dav
           month
                      4521 non-null
                                       object
       11
          duration
                      4521 non-null
                                       int64
           campaign
                      4521 non-null
                                       int64
       12
       13
           pdays
                      4521 non-null
                                       int64
       14
           previous
                      4521 non-null
                                       int64
       15
                      4521 non-null
                                       object
           poutcome
      16 y
                      4521 non-null
      dtypes: int64(7), object(10)
      memory usage: 600.6+ KB
      None
```

```
[10]: df_encoded = pd.get_dummies(df, drop_first=True)
[11]: \quad \textbf{X = df\_encoded.drop("y\_yes", axis=1)} \quad \text{\# target column "y" (yes/no) becomes y\_yes after encoding}
       y = df_encoded["y_yes"]
[12]: X_train, X_test, y_train, y_test = train_test_split( X, y, test_size=0.2, random_state=42 )
[13]: clf = DecisionTreeClassifier(criterion="entropy", max_depth=5, random_state=42)
[14]: clf.fit(X_train, y_train)
[14]: 
• DecisionTreeClassifier
        ▶ Parameters
[18]: y_pred = clf.predict(X_test)
                                                                                                                                                              ◎ ↑ ↓ 占 ♀ ▮
       print("Accuracy:", accuracy_score(y_test, y_pred))
print("\nclassification Report:\n", classification_report(y_test, y_pred))
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.xlabel("predicted")
        plt.ylabel("Actual")
        plt.show()
        Accuracy: 0.8994475138121547
        Classification Report:
                         precision
                                        recall f1-score support
               False
                              0.93
                                          0.96
                True
                              0.55
                                         0.43
                                                     0.48
                                                                   98
            accuracy
                                                     0.90
                                                                   905
       macro avg
weighted avg
                                                                  905
905
                              0.74
                                          0.69
                                                      0.71
                              0.89
                                         0.90
                                                     0.89
                                                                                          700
                                                                                         600
                                                               35
           0 -
                                                                                         - 500
                                                                                         400
                                                                                         - 300
                              56
                                                               42
                                                                                        - 200
                                                                                        - 100
                               Ó
                                           Predicted
```

