

```
In [1]: import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier, plot_tree
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
import matplotlib.pyplot as plt

# Load dataset (assuming bank-full.csv is downloaded locally)
df = pd.read_csv("C:/Users/user/Desktop/bank-full.csv", sep=';')
df.head()
```

Out[1]:

	age	job	marital	education	default	balance	housing	loan	contact	duration
--	-----	-----	---------	-----------	---------	---------	---------	------	---------	----------

0	58	management	married	tertiary	no	2143	yes	no	unknown	
1	44	technician	single	secondary	no	29	yes	no	unknown	
2	33	entrepreneur	married	secondary	no	2	yes	yes	unknown	
3	47	blue-collar	married	unknown	no	1506	yes	no	unknown	
4	33	unknown	single	unknown	no	1	no	no	unknown	



```
In [2]: print(df.head())
print(df.info())
print(df['y'].value_counts()) # target variable distribution
```

	age	job	marital	education	default	balance	housing	loan	\
0	58	management	married	tertiary	no	2143	yes	no	
1	44	technician	single	secondary	no	29	yes	no	
2	33	entrepreneur	married	secondary	no	2	yes	yes	
3	47	blue-collar	married	unknown	no	1506	yes	no	
4	33	unknown	single	unknown	no	1	no	no	

	contact	day	month	duration	campaign	pdays	previous	poutcome	y
0	unknown	5	may	261	1	-1	0	unknown	no
1	unknown	5	may	151	1	-1	0	unknown	no
2	unknown	5	may	76	1	-1	0	unknown	no
3	unknown	5	may	92	1	-1	0	unknown	no
4	unknown	5	may	198	1	-1	0	unknown	no

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 45211 entries, 0 to 45210
```

```
Data columns (total 17 columns):
```

#	Column	Non-Null Count	Dtype
0	age	45211 non-null	int64
1	job	45211 non-null	object
2	marital	45211 non-null	object
3	education	45211 non-null	object
4	default	45211 non-null	object
5	balance	45211 non-null	int64
6	housing	45211 non-null	object
7	loan	45211 non-null	object
8	contact	45211 non-null	object
9	day	45211 non-null	int64
10	month	45211 non-null	object
11	duration	45211 non-null	int64
12	campaign	45211 non-null	int64
13	pdays	45211 non-null	int64
14	previous	45211 non-null	int64
15	poutcome	45211 non-null	object
16	y	45211 non-null	object

```
dtypes: int64(7), object(10)
```

```
memory usage: 5.9+ MB
```

```
None
```

```
y
```

```
no      39922
```

```
yes      5289
```

```
Name: count, dtype: int64
```

```
In [3]: #DATA PROPROCESSING
```

```
In [4]: #Encode categorical variables
#Most features are categorical, so use one-hot encoding or Label encoding.
```

```
In [5]: # Binary target: yes=1, no=0
df['y'] = df['y'].map({'yes': 1, 'no': 0})

# One-hot encode categorical variables
categorical_cols = df.select_dtypes(include=['object']).columns
df_encoded = pd.get_dummies(df, columns=categorical_cols, drop_first=True)
```

```
In [6]: #Split data into features and target
```

```
In [7]: X = df_encoded.drop('y', axis=1)
        y = df_encoded['y']
```

```
In [8]: # Split into train/test sets
```

```
In [9]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_
```

```
In [10]: #DT CLASSIFIER
```

```
In [11]: clf = DecisionTreeClassifier(random_state=42, max_depth=5) # max_depth to preve
        clf.fit(X_train, y_train)
```

```
Out[11]: ▼ DecisionTreeClassifier
         DecisionTreeClassifier(max_depth=5, random_state=42)
```

```
In [12]: #EVALUATE MODEL
```

```
In [13]: y_pred = clf.predict(X_test)

        print("Accuracy:", accuracy_score(y_test, y_pred))
        print("\nConfusion Matrix:\n", confusion_matrix(y_test, y_pred))
        print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

Accuracy: 0.9002506635210852

Confusion Matrix:

```
[[11733  244]
 [ 1109  478]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.91	0.98	0.95	11977
1	0.66	0.30	0.41	1587
accuracy			0.90	13564
macro avg	0.79	0.64	0.68	13564
weighted avg	0.88	0.90	0.88	13564

```
In [14]: #SEE WHAT FEATIRES MOST INFLUENCE PREDICTION
```

```
In [15]: import numpy as np

        importance = clf.feature_importances_
        indices = np.argsort(importance)[::-1]
        features = X.columns[indices]

        for i in range(10): # Top 10
            print(f"{features[i]}: {importance[indices[i]]:.4f}")
```

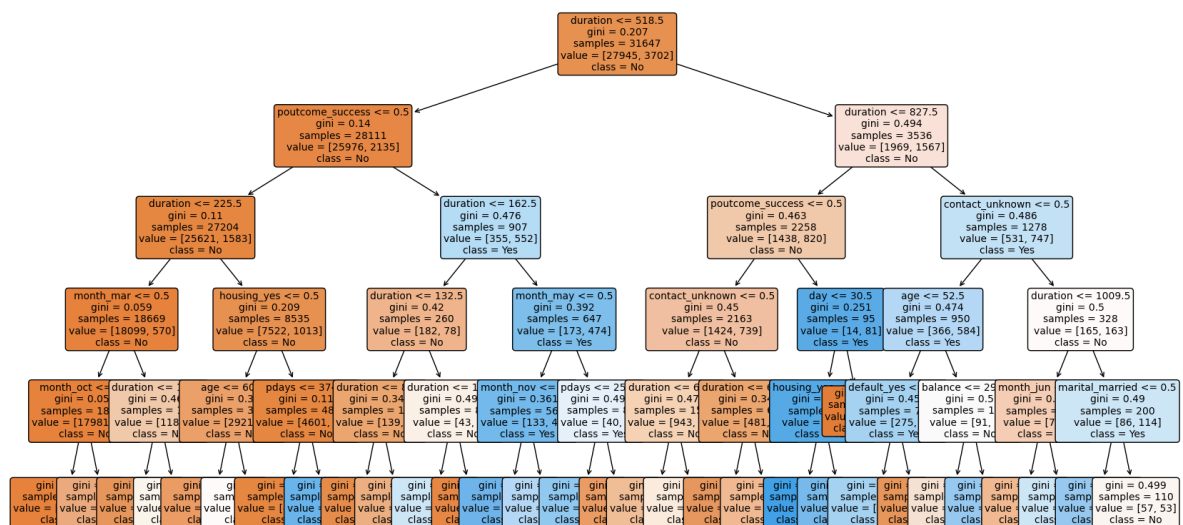
```

duration: 0.5611
poutcome_success: 0.2859
housing_yes: 0.0423
age: 0.0250
pdays: 0.0210
month_mar: 0.0208
contact_unknown: 0.0159
month_oct: 0.0137
month_may: 0.0040
marital_married: 0.0019

```

In [16]: `#VISUALIZE THE DT CLASSIFIER`

In [17]: `plt.figure(figsize=(20,10))  
plot_tree(clf, feature_names=X.columns, class_names=['No', 'Yes'], filled=True,  
plt.show())`



In [18]: `#PREDICT WHETHER A NEW CLIENT WILL SUBSCRIBE OR NOT`

```

In [ ]: import pandas as pd

# New client data as a dictionary
new_client = {
    'age': 35,
    'job': 'technician',
    'marital': 'single',
    'education': 'tertiary',
    'default': 'no',
    'balance': 1500,
    'housing': 'yes',
    'loan': 'no',
    'contact': 'cellular',
    'day': 12,
    'month': 'may',
    'campaign': 1,
    'pdays': -1,
    'previous': 0,
    'poutcome': 'unknown'
}

# Convert to DataFrame
new_df = pd.DataFrame([new_client])

```

```
# Drop columns not used in training (e.g., duration)
# One-hot encode to match training
new_df_encoded = pd.get_dummies(new_df)

# Align columns with training set
# Add missing columns (set to 0) or drop extra columns
for col in X_train.columns:
    if col not in new_df_encoded:
        new_df_encoded[col] = 0
new_df_encoded = new_df_encoded[X_train.columns]

# Make prediction
prediction = model.predict(new_df_encoded)
probability = model.predict_proba(new_df_encoded)

# Output
print("Subscribed?" , "Yes" if prediction[0] == 1 else "No")
print("Probability of subscription:", probability[0][1])
```