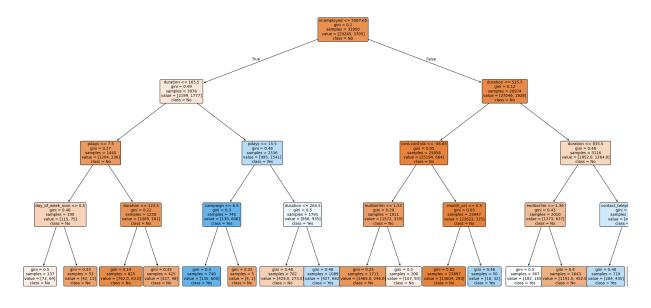
```
In [23]: import zipfile
import os
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.tree import DecisionTreeClassifier, plot tree
from sklearn.metrics import accuracy_score, confusion_matrix, classificat:
from IPython.display import display, HTML
      Step 1: Extract Outer ZIP
outer zip path = r"C:\Users\ahire\Downloads\bank+marketing.zip"
outer extract path = "outer extracted"
if not os.path.exists(outer extract path):
    with zipfile.ZipFile(outer_zip_path, 'r') as outer_zip:
        outer zip.extractall(outer extract path)
      Step 2: Extract Inner ZIP
inner zip path = os.path.join(outer extract path, "bank-additional.zip")
inner extract path = "inner extracted"
if not os.path.exists(inner extract path):
    with zipfile.ZipFile(inner zip path, 'r') as inner zip:
        inner zip.extractall(inner extract path)
      Step 3: Load CSV
csv path = os.path.join(inner extract path, "bank-additional", "bank-addi"
df = pd.read csv(csv path, sep=';')
      Step 4: Preprocessing
df['y'] = df['y'].map(\{'yes': 1, 'no': 0\})
df encoded = pd.get dummies(df, drop first=True)
X = df_encoded.drop('y', axis=1)
y = df encoded['y']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, |
      Step 5: Train Decision Tree
clf = DecisionTreeClassifier(max depth=4, random state=42)
clf.fit(X_train, y_train)
      Step 6: Visualize Decision Tree
fig = plt.figure(figsize=(24, 12), facecolor='white')
ax = fig.add subplot(1, 1, 1)
plot_tree(
    clf,
    filled=True,
    feature_names=X.columns,
    class names=["No", "Yes"],
    rounded=True,
    precision=2,
    fontsize=10,
    ax=ax
)
ax.set title("Decision Tree: Predicting Customer Subscription", fontsize=
plt.tight_layout()
plt.show()
      Step 7: Evaluate Model
v pred = clf.predict(X test)
```

Decision Tree: Predicting Customer Subscription



Accuracy: 0.9124

Confusion Matrix

	Predicted: No	Predicted: Yes	
Actual: No	7052	251	
Actual: Yes	471	464	

Classifi	cation Report precision		f1-score	support
No	0.94	0.97	0.95	7303
Yes	0.65	0.50	0.56	935
accuracy			0.91	8238
macro avg	0.79	0.73	0.76	8238
weighted avg	0.90	0.91	0.91	8238