

```

import pandas as pd
import matplotlib.pyplot as plt
from collections import Counter
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier, plot_tree
from sklearn.metrics import accuracy_score, precision_score,
recall_score, f1_score
from math import log2

df = pd.read_csv('EPL.csv')

df = df[['HomeTeam', 'AwayTeam', 'FullTimeResult', 'HalfTimeResult']]
df = df.apply(lambda x: pd.factorize(x)[0])

def entropy(data):
    total = len(data)
    counts = Counter(data)
    return -sum((count/total) * log2(count/total) for count in
counts.values())

target_entropy = entropy(df['HalfTimeResult'])
print(f"Dataset Entropy(half): {target_entropy:.4f}")
target_entropy = entropy(df['FullTimeResult'])
print(f"Dataset Entropy(full): {target_entropy:.4f}")

Dataset Entropy(half): 1.5688
Dataset Entropy(full): 1.5392

def information_gain(df, feature, target):
    total_entropy = entropy(df[target])
    values = df[feature].unique()
    weighted_entropy = sum(
        (len(df[df[feature] == v]) / len(df)) * entropy(df[df[feature]
== v][target])
        for v in values
    )
    return total_entropy - weighted_entropy

info_gains = {feature: information_gain(df, feature, 'FullTimeResult')}
for feature in df.columns[:-1]:
    best_feature = max(info_gains, key=info_gains.get)
    print("\nInformation Gain for Each Feature:")
    for feature, gain in info_gains.items():
        print(f"{feature}: {gain:.4f}")
    print(f"\nBest Feature for Splitting: {best_feature}")

Information Gain for Each Feature:
HomeTeam: 0.0930
AwayTeam: 0.0715
FullTimeResult: 1.5392

```

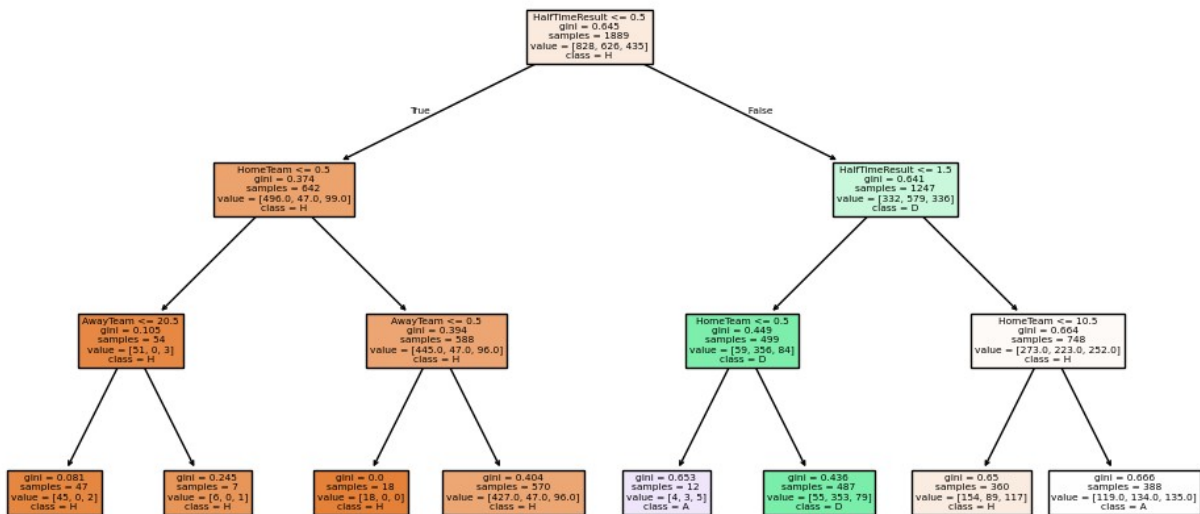
Best Feature for Splitting: FullTimeResult

```
X = df.drop(columns=['FullTimeResult'])
y = df['FullTimeResult']
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.1, random_state=42)

# Initialize and fit the CART model (using Gini impurity)
cart_tree = DecisionTreeClassifier(criterion="gini", max_depth=3,
min_samples_split=10, min_samples_leaf=5)
cart_tree.fit(X_train, y_train)

DecisionTreeClassifier(max_depth=3, min_samples_leaf=5,
min_samples_split=10)

plt.figure(figsize=(12, 6))
plot_tree(cart_tree, feature_names=X.columns, class_names=['H', 'D',
'A'], filled=True)
plt.show()
```



```
importances = cart_tree.feature_importances_
feature_importance = pd.DataFrame({'Feature': X.columns, 'Importance':
importances})
feature_importance = feature_importance.sort_values(by='Importance',
ascending=False)

print("\nTop Important Features:")
print(feature_importance.head(10))
```

Top Important Features:

	Feature	Importance
2	HalfTimeResult	0.951697
0	HomeTeam	0.041522
1	AwayTeam	0.006781

```
y_pred = cart_tree.predict(X_test)
```

```
accuracy = accuracy_score(y_test, y_pred)
```

```
precision = precision_score(y_test, y_pred, average='weighted')
```

```
recall = recall_score(y_test, y_pred, average='weighted')
```

```
f1 = f1_score(y_test, y_pred, average='weighted')
```

```
print(f"Accuracy: {accuracy:.4f}")
```

```
print(f"Precision: {precision:.4f}")
```

```
print(f"Recall: {recall:.4f}")
```

```
print(f"F1 Score: {f1:.4f}")
```

Accuracy: 0.6190

Precision: 0.6129

Recall: 0.6190

F1 Score: 0.6129

```
new_result = {  
    'Attribute': 'Cart',  
    'Accuracy': accuracy,  
    'Precision': precision,  
    'Recall': recall,  
    'F1 Score': f1  
}
```

```
df_existing = pd.read_csv('result.csv')
```

```
df_new = pd.DataFrame([new_result])
```

```
df_combined = pd.concat([df_existing, df_new], ignore_index=True)
```

```
df_combined.to_csv('result.csv', index=False)
```