

```

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, GridSearchCV
from sklearn.preprocessing import LabelEncoder # only for 2 classes if more than 2 classes then use OneHotEncoder
from sklearn.tree import DecisionTreeClassifier, plot_tree
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix, roc_auc_score

```

```

df = pd.read_csv("Telco-Customer-Churn.csv") # Ensure file is downloaded from the given link
print("Dataset Shape:", df.shape)
print(df.head())

```

→ Dataset Shape: (7043, 21)

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	... DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges	Churn
0	7590-VHVEG	Female	0	Yes	No	1	No	No phone service	DSL	No	...	No	No	No	Month-to-month	Yes	Electronic check	29.85	29.85	No
1	5575-GNVDE	Male	0	No	No	34	Yes	No	DSL	Yes	...	Yes	Yes	Yes	One year	No	Mailed check	56.95	1889.5	No
2	3668-QPYBK	Male	0	No	No	2	Yes	No	DSL	Yes	...	No	No	No	Month-to-month	Yes	Mailed check	53.85	108.15	Yes
3	7795-CFOCW	Male	0	No	No	45	No	No	DSL	Yes	...	Yes	No	No	One year	No	Bank transfer (automatic)	42.30	1840.75	No
4	9237-HQITU	Female	0	No	No	2	Yes	No	Fiber optic	No	...	No	No	No	Month-to-month	Yes	Electronic check	70.70	151.65	Yes

[5 rows x 21 columns]

```
df.drop('customerID', axis=1, inplace=True)
```

df

→

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	
0	Female	0	Yes	No	1	No	No	No phone service	DSL	No	Yes
1	Male	0	No	No	34	Yes	No	No	DSL	Yes	No
2	Male	0	No	No	2	Yes	No	No	DSL	Yes	Yes
3	Male	0	No	No	45	No	No	No phone service	DSL	Yes	No
4	Female	0	No	No	2	Yes	No	Fiber optic	No	No	No
...
7038	Male	0	Yes	Yes	24	Yes	Yes	No	DSL	Yes	No
7039	Female	0	Yes	Yes	72	Yes	Yes	Fiber optic	No	No	Yes
7040	Female	0	Yes	Yes	11	No	No	No phone service	DSL	Yes	No
7041	Male	1	Yes	No	4	Yes	Yes	Fiber optic	No	No	No
7042	Male	0	No	No	66	Yes	No	Fiber optic	No	Yes	No

7043 rows x 20 columns

Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)

```
df['TotalCharges'] = pd.to_numeric(df['TotalCharges'], errors='coerce') #coerce is convert data in NaN if not convert in numeric
```

df

		gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup
0	Female		0	Yes	No	1	No	No phone service	DSL	No	Yes
1	Male		0	No	No	34	Yes	No	DSL	Yes	No
2	Male		0	No	No	2	Yes	No	DSL	Yes	Yes
3	Male		0	No	No	45	No	No phone service	DSL	Yes	No
4	Female		0	No	No	2	Yes	No	Fiber optic	No	No
...
7038	Male		0	Yes	Yes	24	Yes	Yes	DSL	Yes	No
7039	Female		0	Yes	Yes	72	Yes	Yes	Fiber optic	No	Yes
7040	Female		0	Yes	Yes	11	No	No phone service	DSL	Yes	No
7041	Male		1	Yes	No	4	Yes	Yes	Fiber optic	No	No
7042	Male		0	No	No	66	Yes	No	Fiber optic	Yes	No

7043 rows × 20 columns

Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)

```
df = df.dropna()
```

df

		gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup
0	Female		0	Yes	No	1	No	No phone service	DSL	No	Yes
1	Male		0	No	No	34	Yes	No	DSL	Yes	No
2	Male		0	No	No	2	Yes	No	DSL	Yes	Yes
3	Male		0	No	No	45	No	No phone service	DSL	Yes	No
4	Female		0	No	No	2	Yes	No	Fiber optic	No	No
...
7038	Male		0	Yes	Yes	24	Yes	Yes	DSL	Yes	No
7039	Female		0	Yes	Yes	72	Yes	Yes	Fiber optic	No	Yes
7040	Female		0	Yes	Yes	11	No	No phone service	DSL	Yes	No
7041	Male		1	Yes	No	4	Yes	Yes	Fiber optic	No	No
7042	Male		0	No	No	66	Yes	No	Fiber optic	Yes	No

7032 rows × 20 columns

Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)

```
for col in df.columns:  
    if df[col].dtype == 'object':
```

```

if df[col].nunique() == 2:
    df[col] = LabelEncoder().fit_transform(df[col])
else:
    df = pd.get_dummies(df, columns=[col])

→ /tmp/ipython-input-2470158310.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus
df[col] = LabelEncoder().fit_transform(df[col])
/tmp/ipython-input-2470158310.py:4: SettingWithCopyWarning:
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df[col] = LabelEncoder().fit_transform(df[col])
/tmp/ipython-input-2470158310.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus
df[col] = LabelEncoder().fit_transform(df[col])

```

df

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	PaperlessBilling	MonthlyCharges	TotalCharges	Churn	...	
0	0	0	1	0	1	0	1	29.85	29.85	0	...	
1	1	0	0	0	34	1	0	56.95	1889.50	0	...	
2	1	0	0	0	2	1	1	53.85	108.15	1	...	
3	1	0	0	0	45	0	0	42.30	1840.75	0	...	
4	0	0	0	0	2	1	1	70.70	151.65	1	...	
...	
7038	1	0	1	1	24	1	1	84.80	1990.50	0	...	
7039	0	0	1	1	72	1	1	103.20	7362.90	0	...	
7040	0	0	1	1	11	0	1	29.60	346.45	0	...	
7041	1	1	1	0	4	1	1	74.40	306.60	1	...	
7042	1	0	0	0	66	1	1	105.65	6844.50	0	...	

7032 rows × 41 columns

```

X = df[[
    'Contract_Month-to-month',
    'InternetService_Fiber optic',
    'TotalCharges',
    'tenure',
    'MonthlyCharges',
    'TechSupport_No',
    'PaymentMethod_Electronic check',
    'Contract_One year',
    'StreamingMovies_Yes',
    'MultipleLines_No phone service'
]]

```

X

	Contract_Month-to-month	InternetService_Fiber optic	TotalCharges	tenure	MonthlyCharges	TechSupport_No	PaymentMethod_Electronic check	Contra check
0	True	False	29.85	1	29.85	True		True
1	False	False	1889.50	34	56.95	True		False
2	True	False	108.15	2	53.85	True		False
3	False	False	1840.75	45	42.30	False		False
4	True	True	151.65	2	70.70	True		True
...
7038	False	False	1990.50	24	84.80	False		False
7039	False	True	7362.90	72	103.20	True		False
7040	True	False	346.45	11	29.60	True		True
7041	True	True	306.60	4	74.40	True		False
7042	False	True	6844.50	66	105.65	False		False

7032 rows × 10 columns

Next steps: [Generate code with X](#) [View recommended plots](#) [New interactive sheet](#)

```
y = df['Churn']
```

y

	Churn
0	0
1	0
2	1
3	0
4	1
...	...
7038	0
7039	0
7040	0
7041	1
7042	0

7032 rows × 1 columns

dtype: int64

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42, stratify=y)
```

```
dt_model = DecisionTreeClassifier(random_state=42, max_depth=5)
dt_model.fit(X_train, y_train)
```

Decision Tree Performance:
i *?*
 DecisionTreeClassifier(max_depth=5, random_state=42)

```
y_pred_dt = dt_model.predict(X_test)
```

```
print("\nDecision Tree Performance:")
print("Accuracy:", accuracy_score(y_test, y_pred_dt))
print("ROC-AUC:", roc_auc_score(y_test, dt_model.predict_proba(X_test)[:, 1]))
print("\nClassification Report:\n", classification_report(y_test, y_pred_dt))
```

Decision Tree Performance:
 Accuracy: 0.7903340440653873
 ROC-AUC: 0.828923596191975

```
Classification Report:
precision    recall    f1-score   support
```

```

      0      0.86      0.86      0.86      1033
      1      0.61      0.61      0.61      374

accuracy                         0.79      1407
macro avg                        0.73      0.73      0.73      1407
weighted avg                      0.79      0.79      0.79      1407

dt_importance = pd.Series(dt_model.feature_importances_, index=X.columns).sort_values(ascending=False)

#### Random Forest

rf_model = RandomForestClassifier(n_estimators=100, max_depth=8, random_state=42)
rf_model.fit(X_train, y_train)

RandomForestClassifier(max_depth=8, random_state=42)

y_pred_rf = rf_model.predict(X_test)

print("\nRandom Forest Performance:")
print("Accuracy:", accuracy_score(y_test, y_pred_rf))
print("ROC-AUC:", roc_auc_score(y_test, rf_model.predict_proba(X_test)[:, 1]))
print("\nClassification Report:\n", classification_report(y_test, y_pred_rf))

Random Forest Performance:
Accuracy: 0.7874911158493249
ROC-AUC: 0.8350192316652084

Classification Report:
precision    recall    f1-score   support
      0      0.83      0.89      0.86      1033
      1      0.63      0.50      0.56      374

accuracy                         0.79      1407
macro avg                        0.73      0.70      0.71      1407
weighted avg                      0.78      0.79      0.78      1407

rf_importance = pd.Series(rf_model.feature_importances_, index=X.columns).sort_values(ascending=False)

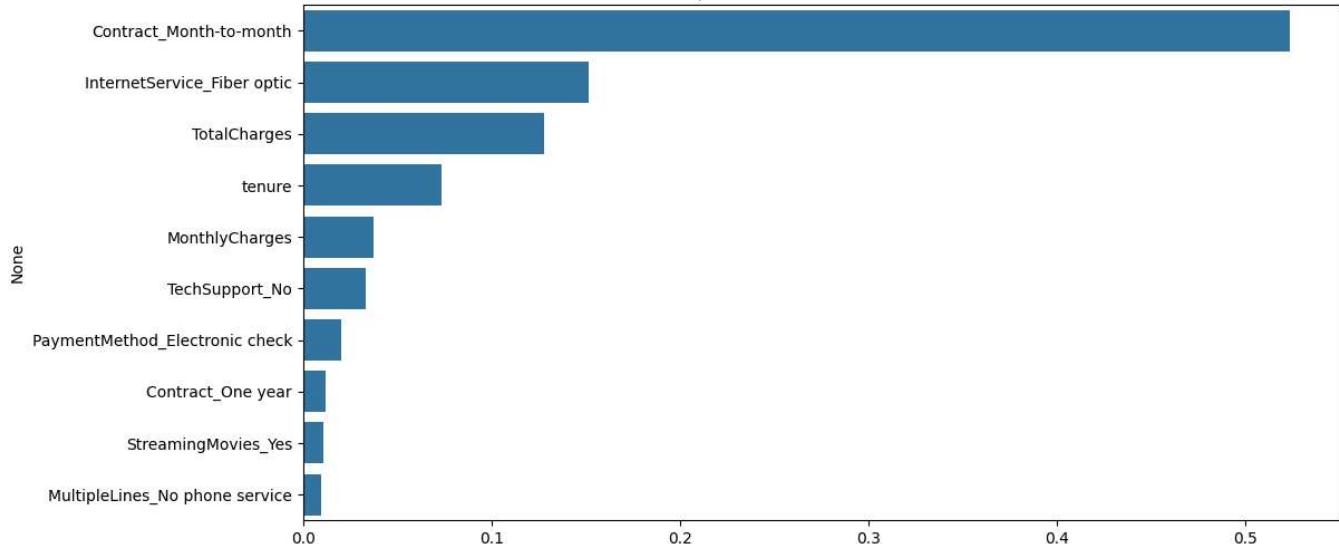
plt.figure(figsize=(12, 6))
sns.barplot(x=dt_importance[:10].values, y=dt_importance[:10].index)
plt.title("Top 10 Features - Decision Tree")
plt.show()

plt.figure(figsize=(12, 6))
sns.barplot(x=rf_importance[:10].values, y=rf_importance[:10].index)
plt.title("Top 10 Features - Random Forest")
plt.show()

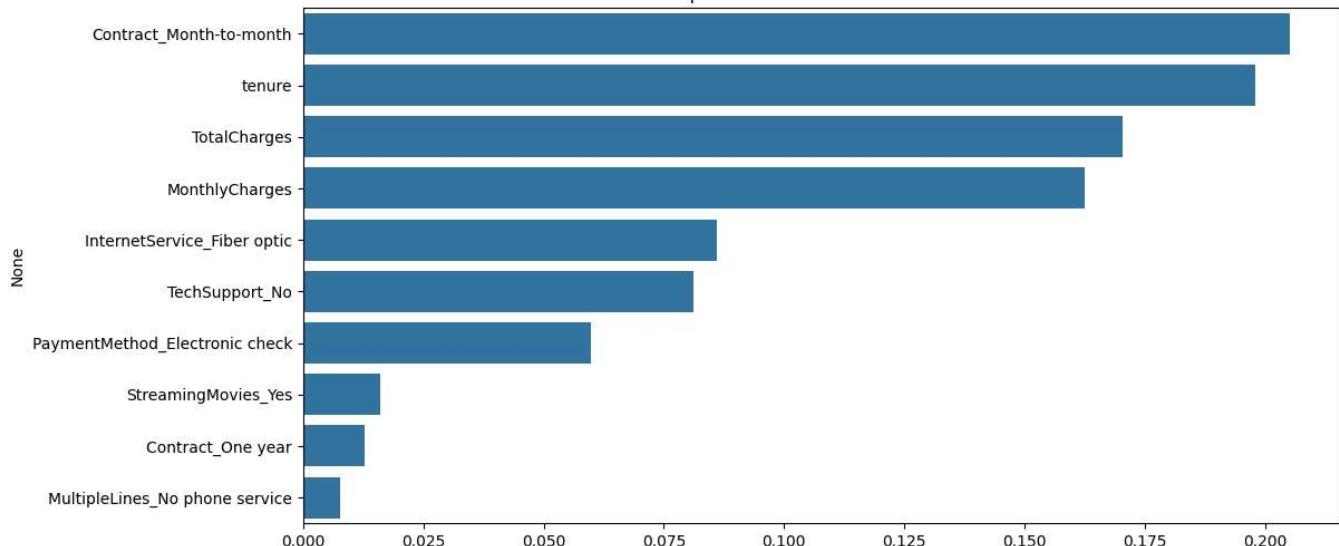
```

[

Top 10 Features - Decision Tree



Top 10 Features - Random Forest



```

depths = range(1, 21)
dt_accuracies = []
rf_accuracies = []

for d in depths:
    dt_temp = DecisionTreeClassifier(max_depth=d, random_state=42)
    dt_temp.fit(X_train, y_train)
    dt_accuracies.append(accuracy_score(y_test, dt_temp.predict(X_test)))

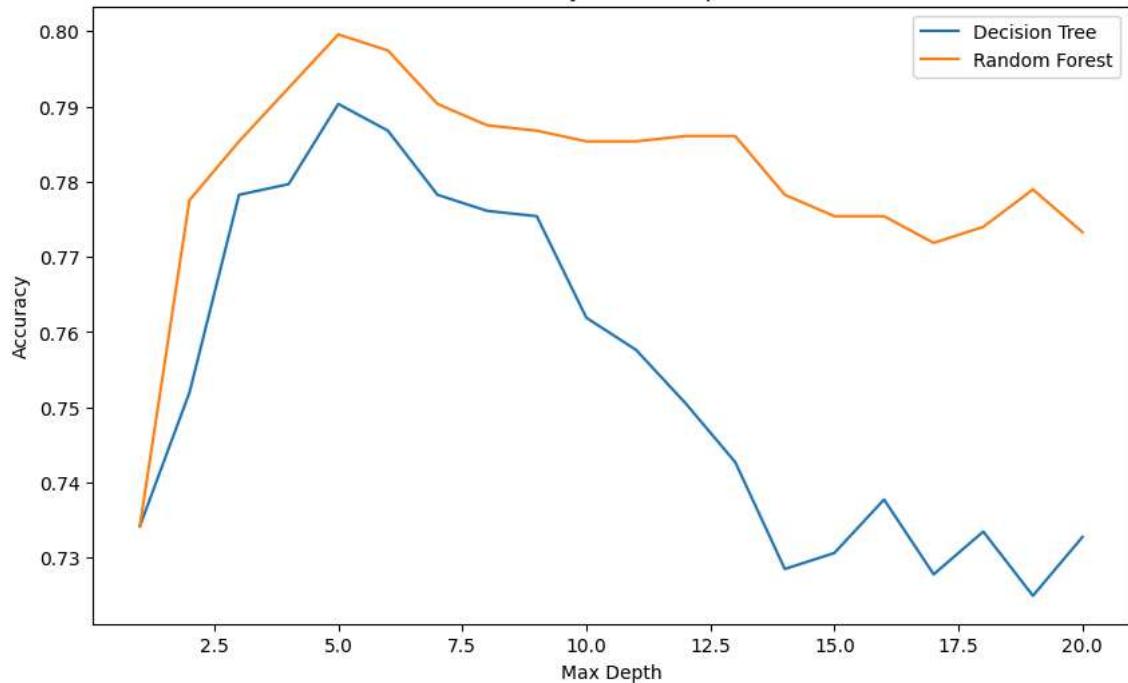
    rf_temp = RandomForestClassifier(n_estimators=100, max_depth=d, random_state=42)
    rf_temp.fit(X_train, y_train)
    rf_accuracies.append(accuracy_score(y_test, rf_temp.predict(X_test)))

plt.figure(figsize=(10, 6))
plt.plot(depths, dt_accuracies, label='Decision Tree')
plt.plot(depths, rf_accuracies, label='Random Forest')
plt.xlabel("Max Depth")
plt.ylabel("Accuracy")
plt.title("Accuracy vs Max Depth")
plt.legend()
plt.show()

```

[]

Accuracy vs Max Depth



```

estimators = [10, 50, 100, 200, 300]
rf_scores = []

for n in estimators:
    rf_temp = RandomForestClassifier(n_estimators=n, max_depth=8, random_state=42)
    rf_temp.fit(X_train, y_train)
    rf_scores.append(accuracy_score(y_test, rf_temp.predict(X_test)))

plt.figure(figsize=(10, 6))
plt.plot(estimators, rf_scores, marker='o')
plt.xlabel("Number of Trees")
plt.ylabel("Accuracy")
plt.title("Random Forest Accuracy vs Number of Trees")
plt.show()

```

[]

Random Forest Accuracy vs Number of Trees

