

Import libraries

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report
```

Load the dataset (directly from GitHub or Kaggle link)

```
url = "https://raw.githubusercontent.com/datasciencedojo/datasets/master/titanic.csv"
data = pd.read_csv(url)
data
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S
...
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN	S
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42	S
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN	S
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148	C
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	NaN	Q

891 rows × 12 columns

Next steps:

[Generate code with data](#)

☒ [View recommended plots](#)

[New interactive sheet](#)

Data Preprocessing

Fill missing values

```
data['Age'].fillna(data['Age'].median(), inplace=True)
data['Embarked'].fillna('S', inplace=True)
print(data)
```

	PassengerId	Survived	Pclass	\							
0	1	0	3								
1	2	1	1								
2	3	1	3								
3	4	1	1								
4	5	0	3								
..								
886	887	0	2								
887	888	1	1								
888	889	0	3								
889	890	1	1								
890	891	0	3								
		Name	Sex	Age	SibSp	\					
0		Braund, Mr. Owen Harris	male	22.0	1						
1	Cumings, Mrs. John Bradley (Florence Briggs Th...		female	38.0	1						
2	Heikkinen, Miss. Laina		female	26.0	0						
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)		female	35.0	1						
4	Allen, Mr. William Henry		male	35.0	0						
..						
886	Montvila, Rev. Juozas		male	27.0	0						
887	Graham, Miss. Margaret Edith		female	19.0	0						
888	Johnston, Miss. Catherine Helen "Carrie"		female	28.0	1						
889	Behr, Mr. Karl Howell		male	26.0	0						
890	Dooley, Mr. Patrick		male	32.0	0						
	Parch	Ticket	Fare	Cabin	Embarked						
0	0	A/5 21171	7.2500	NaN	S						
1	0	PC 17599	71.2833	C85	C						
2	0	STON/O2. 3101282	7.9250	NaN	S						
3	0	113803	53.1000	C123	S						
4	0	373450	8.0500	NaN	S						
..						
886	0	211536	13.0000	NaN	S						
887	0	112053	30.0000	B42	S						
888	2	W./C. 6607	23.4500	NaN	S						
889	0	111369	30.0000	C148	C						
890	0	370376	7.7500	NaN	Q						

[891 rows x 12 columns]

<ipython-input-13-9bf9d701e241>:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method. The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the ope

```
data['Age'].fillna(data['Age'].median(), inplace=True)
```

✓ Drop irrelevant features

```
data = data.drop(['cabin', 'name', 'ticket'], axis=1, errors='ignore')
data.head()
```

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	
0	1	0	3	male	22.0	1	0	7.2500	S	
1	2	1	1	female	38.0	1	0	71.2833	C	
2	3	1	3	female	26.0	0	0	7.9250	S	
3	4	1	1	female	35.0	1	0	53.1000	S	
4	5	0	3	male	35.0	0	0	8.0500	S	

Next steps:

[Generate code with data](#)[View recommended plots](#)[New interactive sheet](#)

✓ One-hot encoding for categorical variables

```
print("Columns before processing:", data.columns)
data.columns = data.columns.str.lower()
data['age'].fillna(data['age'].median(), inplace=True)
columns_to_drop = ['cabin', 'name', 'ticket']
for col in columns_to_drop:
    if col in data.columns:
        data.drop(col, axis=1, inplace=True)
if 'sex' in data.columns and 'embarked' in data.columns:
    data = pd.get_dummies(data, columns=['sex', 'embarked'], drop_first=True)
else:
    print("Columns 'sex' and 'embarked' are not found in the dataset.")
print(data)
```

```
Columns before processing: Index(['passengerid', 'survived', 'pclass', 'age', 'sibsp', 'parch', 'fare',
    'sex_male', 'embarked_q', 'embarked_s'],
    dtype='object')
Columns 'sex' and 'embarked' are not found in the dataset.
```

	passengerid	survived	pclass	age	sibsp	parch	fare	sex_male \
0	1	0	3	22.0	1	0	7.2500	True
1	2	1	1	38.0	1	0	71.2833	False
2	3	1	3	26.0	0	0	7.9250	False

3	4	1	1	35.0	1	0	53.1000	False
4	5	0	3	35.0	0	0	8.0500	True
..
886	887	0	2	27.0	0	0	13.0000	True
887	888	1	1	19.0	0	0	30.0000	False
888	889	0	3	28.0	1	2	23.4500	False
889	890	1	1	26.0	0	0	30.0000	True
890	891	0	3	32.0	0	0	7.7500	True

	embarked_q	embarked_s
0	False	True
1	False	False
2	False	True
3	False	True
4	False	True
..
886	False	True
887	False	True
888	False	True
889	False	False
890	True	False

[891 rows x 10 columns]

<ipython-input-30-48250e006ad7>:3: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method. The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation.

```
data['age'].fillna(data['age'].median(), inplace=True)
```

✖ Feature selection

```
X = data.drop('survived', axis=1)
y = data['survived']
```

✖ Train-test split

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

✖ Train the model

```
model = RandomForestClassifier(random_state=42)
model.fit(X_train, y_train)
```



```
▼ RandomForestClassifier ⓘ ?
RandomForestClassifier(random_state=42)
```

▼ Make predictions and evaluate

```
y_pred = model.predict(X_test)
print("Accuracy:", accuracy_score(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```



```
Accuracy: 0.8212290502793296
```

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

     0       0.83         0.88         0.85         105
     1       0.81         0.74         0.77          74

   accuracy          0.82
  macro avg          0.82
weighted avg          0.82
```

▼ Feature Importance

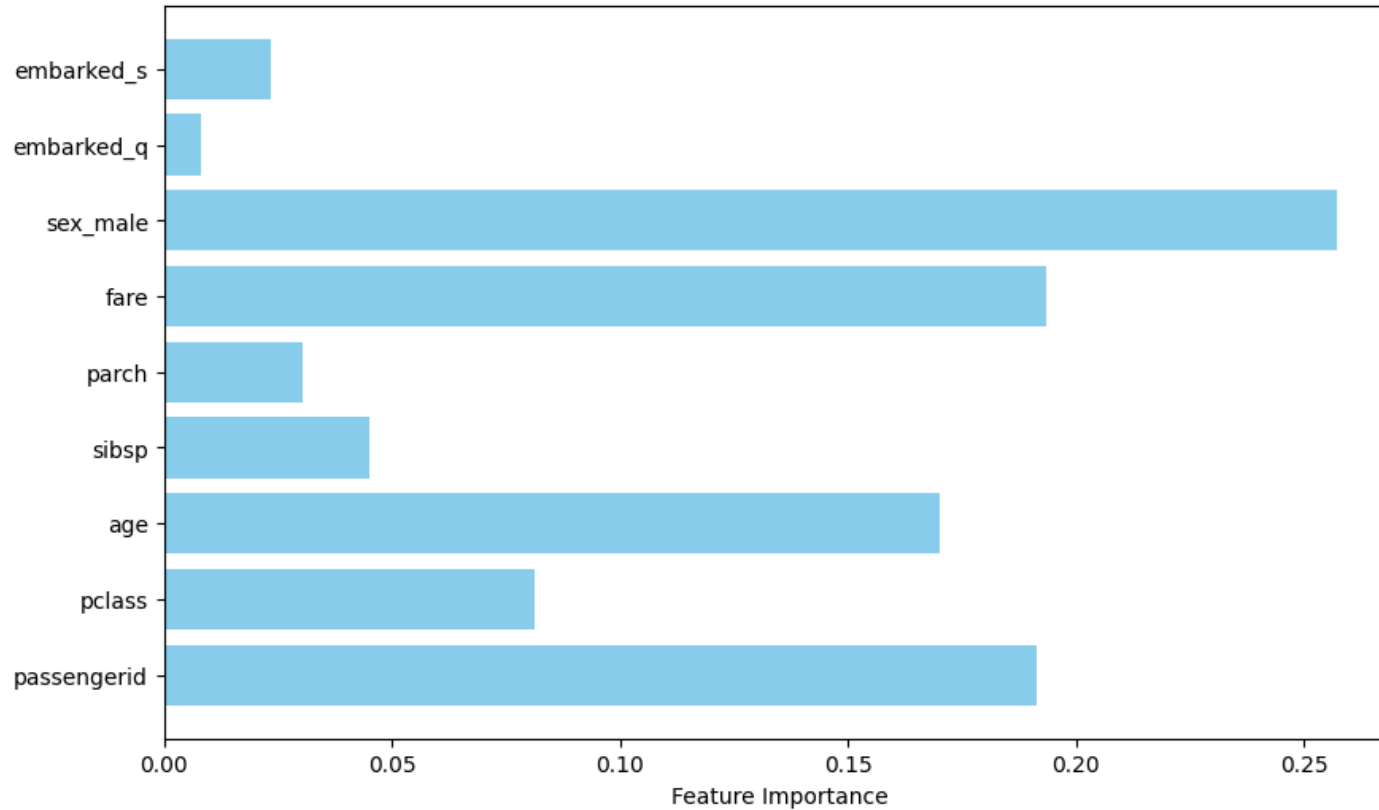
```
import matplotlib.pyplot as plt
```

▼ Plot feature importance

```
feature_importances = model.feature_importances_
features = X.columns
plt.figure(figsize=(10, 6))
plt.barh(features, feature_importances, color='skyblue')
plt.xlabel("Feature Importance")
plt.title("Feature Importance for Titanic Dataset")
plt.show()
```



Feature Importance for Titanic Dataset



✕ Insights from the Predictions

The model's predictions determine whether each passenger in the test dataset survived or not. The evaluation metrics (accuracy, precision, recall) help assess how well the model performs this task. The feature importance plot provides insights into which factors were most influential in predicting survival, such as:

Gender (Sex_male): Being female significantly increases the likelihood of survival.

Passenger class (Pclass): Higher classes have better survival chances.

Age: Younger passengers, particularly children, are more likely to survive.

Double-click (or enter) to edit

