```
# Import necessary libraries
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeRegressor
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score

# Load the dataset
df = pd.read_csv('100_Sales.csv') # Change the filename according to the file you have
df.head()
```

	Region	Country	Item_Type	Sales_Channel	Order_Priority	Ship_Date	Unit_Cost	Total_Revenue	Total_Profit	Unnamed: 9	Unnamed: 10
0	Australia and Oceania	Tuvalu	Baby Food	Offline	Н	27/06/2010	159.42	2533654.00	951410.50	NaN	NaN
1	Central America and the Caribbean	Grenada	Cereal	Online	С	15/09/2012	117.11	576782.80	248406.36	NaN	NaN
2	Europe	Russia	Office Supplies	Offline	L	05/08/2014	524.96	1158502.59	224598.75	NaN	NaN
3	Sub_Saharan Africa	Sao Tome and Principe	Fruits	Online	С	07/05/2014	6.92	75591.66	19525.82	NaN	NaN
4	Sub Saharan		Office								

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Check for missing values
print("\nMissing Values Before Handling:\n", df.isnull().sum())

```
# Drop columns with all missing values
df.dropna(axis=1, how='all', inplace=True)
# Handle missing values for numerical columns, let's fill with the mean value
df['Unit Cost'].fillna(df['Unit Cost'].mean(), inplace=True)
df['Total Profit'].fillna(df['Total Profit'].mean(), inplace=True)
# After handling missing values, verify again
print("\nMissing Values After Handling:\n", df.isnull().sum())
\rightarrow
     Missing Values Before Handling:
      Region
                          0
     Country
                         0
     Item Type
                         0
     Sales Channel
     Order_Priority
                         0
     Ship Date
     Unit Cost
     Total_Revenue
     Total Profit
                         0
     Unnamed: 9
                       100
     Unnamed: 10
                       100
     dtype: int64
     Missing Values After Handling:
      Region
                        0
     Country
                       0
     Item_Type
                       0
     Sales Channel
     Order Priority
     Ship Date
     Unit Cost
                       0
     Total Revenue
                       0
     Total Profit
                       0
     dtype: int64
     <ipython-input-41-205c2df17bf5>:8: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignm
     The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting value.
     For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].me
       df['Unit Cost'].fillna(df['Unit Cost'].mean(), inplace=True)
```

<ipython-input-41-205c2df17bf5>:9: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignm The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting value.

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

df['Total Profit'].fillna(df['Total Profit'].mean(), inplace=True)

Drop 'Unnamed' columns if they exist df = df.loc[:, ~df.columns.str.contains('^Unnamed')]

Display the dataset again after dropping unwanted columns df.head()

Region		Country	<pre>Item_Type</pre>	Sales_Channel	Order_Priority	Ship_Date	Unit_Cost	Total_Revenue	Total_Profit	\blacksquare
0	Australia and Oceania	Tuvalu	Baby Food	Offline	Н	27/06/2010	159.42	2533654.00	951410.50	ıl.
1	Central America and the Caribbean	Grenada	Cereal	Online	С	15/09/2012	117.11	576782.80	248406.36	
2	Europe	Russia	Office Supplies	Offline	L	05/08/2014	524.96	1158502.59	224598.75	
3	Sub_Saharan Africa	Sao Tome and Principe	Fruits	Online	С	07/05/2014	6.92	75591.66	19525.82	
4	Sub_Saharan Africa	Rwanda	Office Supplies	Offline	L	02/06/2013	524.96	3296425.02	639077.50	
	1 2 3	 Australia and Oceania Central America and the Caribbean Europe Sub_Saharan Africa 	 Australia and Oceania Tuvalu Central America and the Caribbean Grenada Europe Russia Sub_Saharan Africa Sao Tome and Principe 	 Australia and Oceania Tuvalu Baby Food Central America and the Caribbean Grenada Cereal Europe Russia Office Supplies Sub_Saharan Africa Principe Office 	 Australia and Oceania Tuvalu Baby Food Offline Central America and the Caribbean Grenada Cereal Online Europe Russia Office Supplies Sub_Saharan Africa Principe Offline 	0 Australia and Oceania Tuvalu Baby Food Offline H 1 Central America and the Caribbean Grenada Cereal Online C 2 Europe Russia Office Supplies Offline L 3 Sub_Saharan Africa Sao Tome and Principe Fruits Online C 4 Sub_Saharan Africa Rwanda Office Offline L	0Australia and OceaniaTuvaluBaby FoodOfflineH 27/06/20101Central America and the CaribbeanGrenadaCerealOnlineC 15/09/20122EuropeRussiaOffice SuppliesOfflineL 05/08/20143Sub_Saharan AfricaSao Tome and PrincipeFruitsOnlineC 07/05/20144Sub_Saharan AfricaRwandaOfficeOfflineL 02/06/2013	0Australia and OceaniaTuvaluBaby FoodOfflineH 27/06/2010159.421Central America and the CaribbeanGrenadaCerealOnlineC 15/09/2012117.112EuropeRussiaOfflice SuppliesOfflineL 05/08/2014524.963Sub_Saharan AfricaSao Tome and PrincipeFruitsOnlineC 07/05/20146.924Sub_Saharan AfricaBwandaOffliceOfflineL 02/06/2013524.96	0 Australia and Oceania Tuvalu Baby Food Offline H 27/06/2010 159.42 2533654.00 1 Central America and the Caribbean Grenada Cereal Online C 15/09/2012 117.11 576782.80 2 Europe Russia Office Supplies Offline L 05/08/2014 524.96 1158502.59 3 Sub_Saharan Africa Sao Tome and Principe Fruits Online C 07/05/2014 6.92 75591.66 4 Sub_Saharan Africa Rwanda Offlice Offline L 02/06/2013 524.96 3296425.02	0 Australia and Oceania Tuvalu Baby Food Offline H 27/06/2010 159.42 2533654.00 951410.50 1 Central America and the Caribbean Grenada Cereal Online C 15/09/2012 117.11 576782.80 248406.36 2 Europe Russia Office Supplies Offline L 05/08/2014 524.96 1158502.59 224598.75 3 Sub_Saharan Africa Sao Tome and Principe Fruits Online C 07/05/2014 6.92 75591.66 19525.82 4 Sub_Saharan Africa Pwanda Office Offline L 02/06/2013 524.96 3296425.02 639077.50

Next steps:

Generate code with df

View recommended plots

New interactive sheet

```
# One-hot encode categorical columns
df encoded = pd.get dummies(df, drop first=True)
```

Display the encoded dataset df encoded.head()

 \rightarrow

•	Unit_Cost	Total_Revenue	Total_Profit	Region_Australia and Oceania	Region_Central America and the Caribbean	Region_Europe	Region_Middle East and North Africa	Region_North America	Region_Sub_Sahar Afri
0	159.42	2533654.00	951410.50	True	False	False	False	False	Fa
1	117.11	576782.80	248406.36	False	True	False	False	False	Fa
2	524.96	1158502.59	224598.75	False	False	True	False	False	Fa
3	6.92	75591.66	19525.82	False	False	False	False	False	Tı
4	524.96	3296425.02	639077.50	False	False	False	False	False	Tı

5 rows × 197 columns

```
# Define target variable (update the column name based on your dataset)
y = df encoded['Total Revenue'] # Target variable: Change as required
X = df encoded.drop(columns=['Total Revenue']) # Features: Drop target column
# Split the dataset into training and testing sets
X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
# Verify the split
print("X_train shape:", X_train.shape)
print("y_train shape:", y_train.shape)
print("X_test shape:", X_test.shape)
print("y test shape:", y test.shape)
→ X_train shape: (80, 196)
    y train shape: (80,)
    X test shape: (20, 196)
    y test shape: (20,)
# Initialize the Decision Tree Regressor model
dt model = DecisionTreeRegressor(random state=42)
# Train the model
```

```
dt model.fit(X train, y train)
# Make predictions
y pred dt = dt model.predict(X test)
# Fvaluate the model
print("Decision Tree - Mean Absolute Error:", mean absolute error(y test, y pred dt))
print("Decision Tree - Mean Squared Error:", mean squared error(y test, y pred dt))
print("Decision Tree - R2 Score:", r2 score(y test, y pred dt))
Decision Tree - Mean Absolute Error: 450258.1825000001
    Decision Tree - Mean Squared Error: 457717223225.0508
    Decision Tree - R2 Score: 0.7932414302893978
# Initialize the Random Forest Regressor model
rf model = RandomForestRegressor(random state=42)
# Train the model
rf model.fit(X train, y train)
# Make predictions
y pred rf = rf model.predict(X test)
# Evaluate the model
print("Random Forest - Mean Absolute Error:", mean absolute error(y test, y pred rf))
print("Random Forest - Mean Squared Error:", mean squared error(y test, y pred rf))
print("Random Forest - R2 Score:", r2 score(y test, y pred rf))
Random Forest - Mean Absolute Error: 373385.7374149999
     Random Forest - Mean Squared Error: 350122119657.1751
     Random Forest - R2 Score: 0.8418439485971261
# Compare models based on R2 Score or other metrics
print(f"Decision Tree R2 Score: {r2 score(y test, y pred dt)}")
print(f"Random Forest R2 Score: {r2_score(y_test, y_pred_rf)}")
→ Decision Tree R2 Score: 0.7932414302893978
     Random Forest R2 Score: 0.8418439485971261
```

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