

```
import pandas as pd
import numpy as np
```

```
df = pd.read_csv("Telco-Customer-Churn.csv")
df.head()
```

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure
0	7590-VHVEG	Female	0	Yes	No	1
1	5575-GNVDE	Male	0	No	No	34
2	3668-QPYBK	Male	0	No	No	2
3	7795-CF0CW	Male	0	No	No	45
4	9237-HQITU	Female	0	No	No	2

	MultipleLines	InternetService	OnlineSecurity	...
0	No phone service	DSL	No	...
1	No	DSL	Yes	...
2	No	DSL	Yes	...
3	No phone service	DSL	Yes	...
4	No	Fiber optic	No	...

	TechSupport	StreamingTV	StreamingMovies	Contract
0	No	No	No	Month-to-month
1	No	No	No	One year
2	No	No	No	Month-to-month
3	Yes	No	No	One year
4	No	No	No	Month-to-month

	PaymentMethod	MonthlyCharges	TotalCharges	Churn
0	Electronic check	29.85	29.85	No
1	Mailed check	56.95	1889.5	No
2	Mailed check	53.85	108.15	Yes
3	Bank transfer (automatic)	42.30	1840.75	No

4	Electronic check	70.70	151.65	Yes
---	------------------	-------	--------	-----

[5 rows x 21 columns]

```
df.info()
df.describe()
df.isnull().sum()
df['Churn'].value_counts()
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 7043 entries, 0 to 7042

Data columns (total 21 columns):

#	Column	Non-Null Count	Dtype
0	customerID	7043 non-null	object
1	gender	7043 non-null	object
2	SeniorCitizen	7043 non-null	int64
3	Partner	7043 non-null	object
4	Dependents	7043 non-null	object
5	tenure	7043 non-null	int64
6	PhoneService	7043 non-null	object
7	MultipleLines	7043 non-null	object
8	InternetService	7043 non-null	object
9	OnlineSecurity	7043 non-null	object
10	OnlineBackup	7043 non-null	object
11	DeviceProtection	7043 non-null	object
12	TechSupport	7043 non-null	object
13	StreamingTV	7043 non-null	object
14	StreamingMovies	7043 non-null	object
15	Contract	7043 non-null	object
16	PaperlessBilling	7043 non-null	object
17	PaymentMethod	7043 non-null	object
18	MonthlyCharges	7043 non-null	float64
19	TotalCharges	7043 non-null	object
20	Churn	7043 non-null	object

dtypes: float64(1), int64(2), object(18)

memory usage: 1.1+ MB

Churn

No 5174

Yes 1869

Name: count, dtype: int64

```
df.describe(include='all')
```

	customerID	gender	SeniorCitizen	Partner	Dependents
tenure \					
count	7043	7043	7043.000000	7043	7043
7043.000000					
unique	7043	2	NaN	2	2

NaN					
top	3186-AJIEK	Male	NaN	No	No
NaN					
freq	1	3555	NaN	3641	4933
NaN					
mean	NaN	NaN	0.162147	NaN	NaN
32.371149					
std	NaN	NaN	0.368612	NaN	NaN
24.559481					
min	NaN	NaN	0.000000	NaN	NaN
0.000000					
25%	NaN	NaN	0.000000	NaN	NaN
9.000000					
50%	NaN	NaN	0.000000	NaN	NaN
29.000000					
75%	NaN	NaN	0.000000	NaN	NaN
55.000000					
max	NaN	NaN	1.000000	NaN	NaN
72.000000					
	PhoneService	MultipleLines	InternetService	OnlineSecurity	...
\					
count	7043	7043	7043	7043	...
unique	2	3	3	3	...
top	Yes	No	Fiber optic	No	...
freq	6361	3390	3096	3498	...
mean	NaN	NaN	NaN	NaN	...
std	NaN	NaN	NaN	NaN	...
min	NaN	NaN	NaN	NaN	...
25%	NaN	NaN	NaN	NaN	...
50%	NaN	NaN	NaN	NaN	...
75%	NaN	NaN	NaN	NaN	...
max	NaN	NaN	NaN	NaN	...
	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	\
count	7043	7043	7043	7043	
unique	3	3	3	3	
top	No	No	No	No	
freq	3095	3473	2810	2785	

mean	NaN	NaN	NaN	NaN
std	NaN	NaN	NaN	NaN
min	NaN	NaN	NaN	NaN
25%	NaN	NaN	NaN	NaN
50%	NaN	NaN	NaN	NaN
75%	NaN	NaN	NaN	NaN
max	NaN	NaN	NaN	NaN

	Contract	PaperlessBilling	PaymentMethod
MonthlyCharges \			
count	7043	7043	7043
7043.000000			
unique	3	2	4
NaN			
top	Month-to-month	Yes	Electronic check
NaN			
freq	3875	4171	2365
NaN			
mean	NaN	NaN	NaN
64.761692			
std	NaN	NaN	NaN
30.090047			
min	NaN	NaN	NaN
18.250000			
25%	NaN	NaN	NaN
35.500000			
50%	NaN	NaN	NaN
70.350000			
75%	NaN	NaN	NaN
89.850000			
max	NaN	NaN	NaN
118.750000			

	TotalCharges	Churn
count	7043	7043
unique	6531	2
top	20.2	No
freq	11	5174
mean	NaN	NaN
std	NaN	NaN
min	NaN	NaN
25%	NaN	NaN
50%	NaN	NaN
75%	NaN	NaN
max	NaN	NaN

[11 rows x 21 columns]

df.isnull().sum()

```
customerID      0
gender          0
SeniorCitizen   0
Partner         0
Dependents      0
tenure          0
PhoneService    0
MultipleLines   0
InternetService 0
OnlineSecurity  0
OnlineBackup    0
DeviceProtection 0
TechSupport     0
StreamingTV     0
StreamingMovies 0
Contract        0
PaperlessBilling 0
PaymentMethod   0
MonthlyCharges  0
TotalCharges    0
Churn           0
dtype: int64
```

```
print(df.columns.tolist())
```

```
['customerID', 'gender', 'SeniorCitizen', 'Partner', 'Dependents',
 'tenure', 'PhoneService', 'MultipleLines', 'InternetService',
 'OnlineSecurity', 'OnlineBackup', 'DeviceProtection', 'TechSupport',
 'StreamingTV', 'StreamingMovies', 'Contract', 'PaperlessBilling',
 'PaymentMethod', 'MonthlyCharges', 'TotalCharges', 'Churn']
```

```
df['Churn'].value_counts()
```

```
Churn
No      5174
Yes     1869
Name: count, dtype: int64
```

```
df.head()
```

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure
0	7590-VHVEG	Female	0	Yes	No	1
1	5575-GNVDE	Male	0	No	No	34
2	3668-QPYBK	Male	0	No	No	2
3	7795-CF0CW	Male	0	No	No	45
4	9237-HQITU	Female	0	No	No	2

Yes

	MultipleLines	InternetService	OnlineSecurity	...
DeviceProtection \				
0	No phone service	DSL	No	...
No				
1	No	DSL	Yes	...
Yes				
2	No	DSL	Yes	...
No				
3	No phone service	DSL	Yes	...
Yes				
4	No	Fiber optic	No	...
No				

	TechSupport	StreamingTV	StreamingMovies	Contract
PaperlessBilling \				
0	No	No	No	Month-to-month
Yes				
1	No	No	No	One year
No				
2	No	No	No	Month-to-month
Yes				
3	Yes	No	No	One year
No				
4	No	No	No	Month-to-month
Yes				

	PaymentMethod	MonthlyCharges	TotalCharges	Churn
0	Electronic check	29.85	29.85	No
1	Mailed check	56.95	1889.5	No
2	Mailed check	53.85	108.15	Yes
3	Bank transfer (automatic)	42.30	1840.75	No
4	Electronic check	70.70	151.65	Yes

[5 rows x 21 columns]

```
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df.info()
df.describe()
df.head()
df.tail()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 21 columns):
#   Column              Non-Null Count  Dtype
---  -
```

0	customerID	7043	non-null	object
1	gender	7043	non-null	object
2	SeniorCitizen	7043	non-null	int64
3	Partner	7043	non-null	object
4	Dependents	7043	non-null	object
5	tenure	7043	non-null	int64
6	PhoneService	7043	non-null	object
7	MultipleLines	7043	non-null	object
8	InternetService	7043	non-null	object
9	OnlineSecurity	7043	non-null	object
10	OnlineBackup	7043	non-null	object
11	DeviceProtection	7043	non-null	object
12	TechSupport	7043	non-null	object
13	StreamingTV	7043	non-null	object
14	StreamingMovies	7043	non-null	object
15	Contract	7043	non-null	object
16	PaperlessBilling	7043	non-null	object
17	PaymentMethod	7043	non-null	object
18	MonthlyCharges	7043	non-null	float64
19	TotalCharges	7043	non-null	object
20	Churn	7043	non-null	object

dtypes: float64(1), int64(2), object(18)

memory usage: 1.1+ MB

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	\
7038	6840-RESVB	Male	0	Yes	Yes	24	
7039	2234-XADUH	Female	0	Yes	Yes	72	
7040	4801-JJAZL	Female	0	Yes	Yes	11	
7041	8361-LTMKD	Male	1	Yes	No	4	
7042	3186-AJIEK	Male	0	No	No	66	

	PhoneService	MultipleLines	InternetService
OnlineSecurity ... \			
7038	Yes	Yes	DSL
Yes ...			
7039	Yes	Yes	Fiber optic
No ...			
7040	No	No phone service	DSL
Yes ...			
7041	Yes	Yes	Fiber optic
No ...			
7042	Yes	No	Fiber optic
Yes ...			

	DeviceProtection	TechSupport	StreamingTV	StreamingMovies
Contract \				
7038	Yes	Yes	Yes	Yes
One year				
7039	Yes	No	Yes	Yes
One year				

7040	No	No	No	No	Month-
to-month					
7041	No	No	No	No	Month-
to-month					
7042	Yes	Yes	Yes	Yes	
Two year					

	PaperlessBilling	PaymentMethod	MonthlyCharges
TotalCharges \			
7038	Yes	Mailed check	84.80
1990.5			
7039	Yes	Credit card (automatic)	103.20
7362.9			
7040	Yes	Electronic check	29.60
346.45			
7041	Yes	Mailed check	74.40
306.6			
7042	Yes	Bank transfer (automatic)	105.65
6844.5			

	Churn
7038	No
7039	No
7040	No
7041	Yes
7042	No

[5 rows x 21 columns]

```
df.isnull().sum()
```

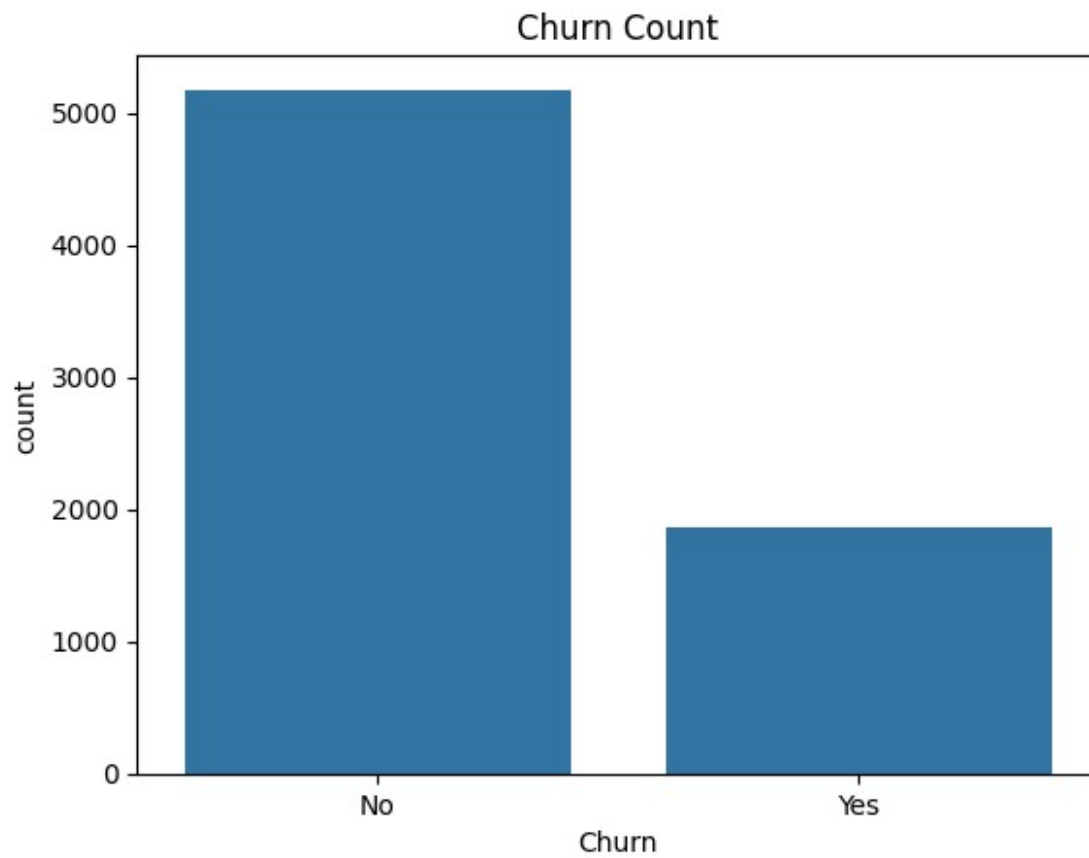
customerID	0
gender	0
SeniorCitizen	0
Partner	0
Dependents	0
tenure	0
PhoneService	0
MultipleLines	0
InternetService	0
OnlineSecurity	0
OnlineBackup	0
DeviceProtection	0
TechSupport	0
StreamingTV	0
StreamingMovies	0
Contract	0
PaperlessBilling	0
PaymentMethod	0
MonthlyCharges	0


```
TotalCharges      0
Churn              0
dtype: int64

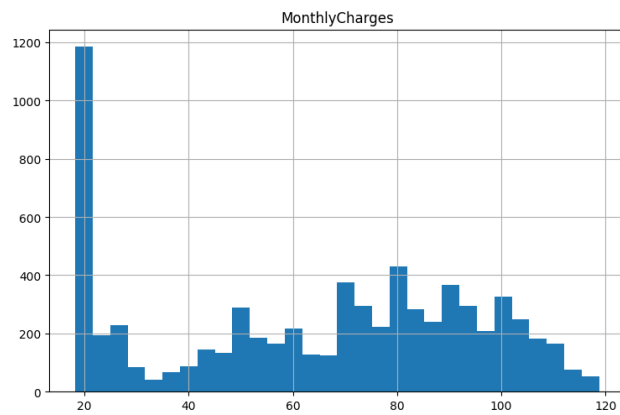
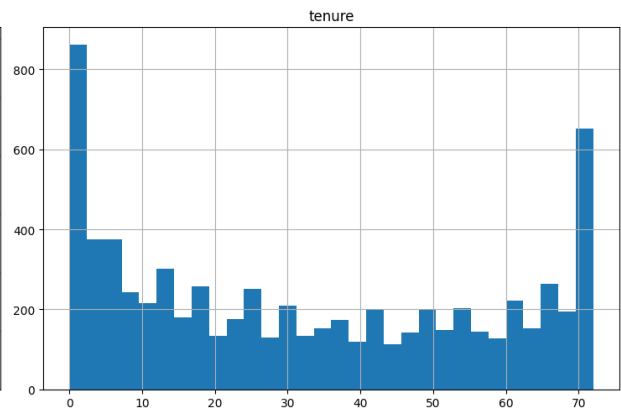
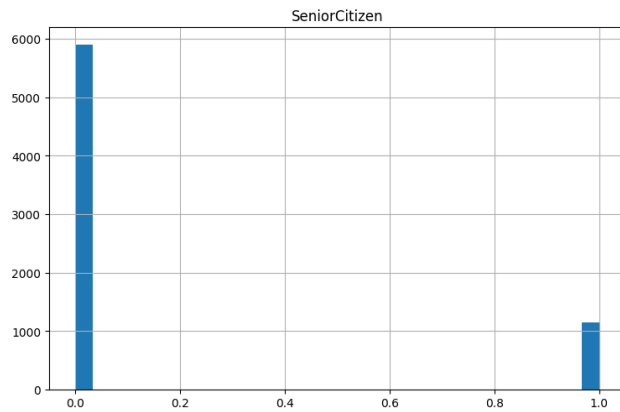
(df == '').sum()

customerID        0
gender            0
SeniorCitizen     0
Partner           0
Dependents        0
tenure            0
PhoneService      0
MultipleLines     0
InternetService   0
OnlineSecurity    0
OnlineBackup      0
DeviceProtection  0
TechSupport       0
StreamingTV       0
StreamingMovies   0
Contract          0
PaperlessBilling  0
PaymentMethod     0
MonthlyCharges    0
TotalCharges      0
Churn             0
dtype: int64

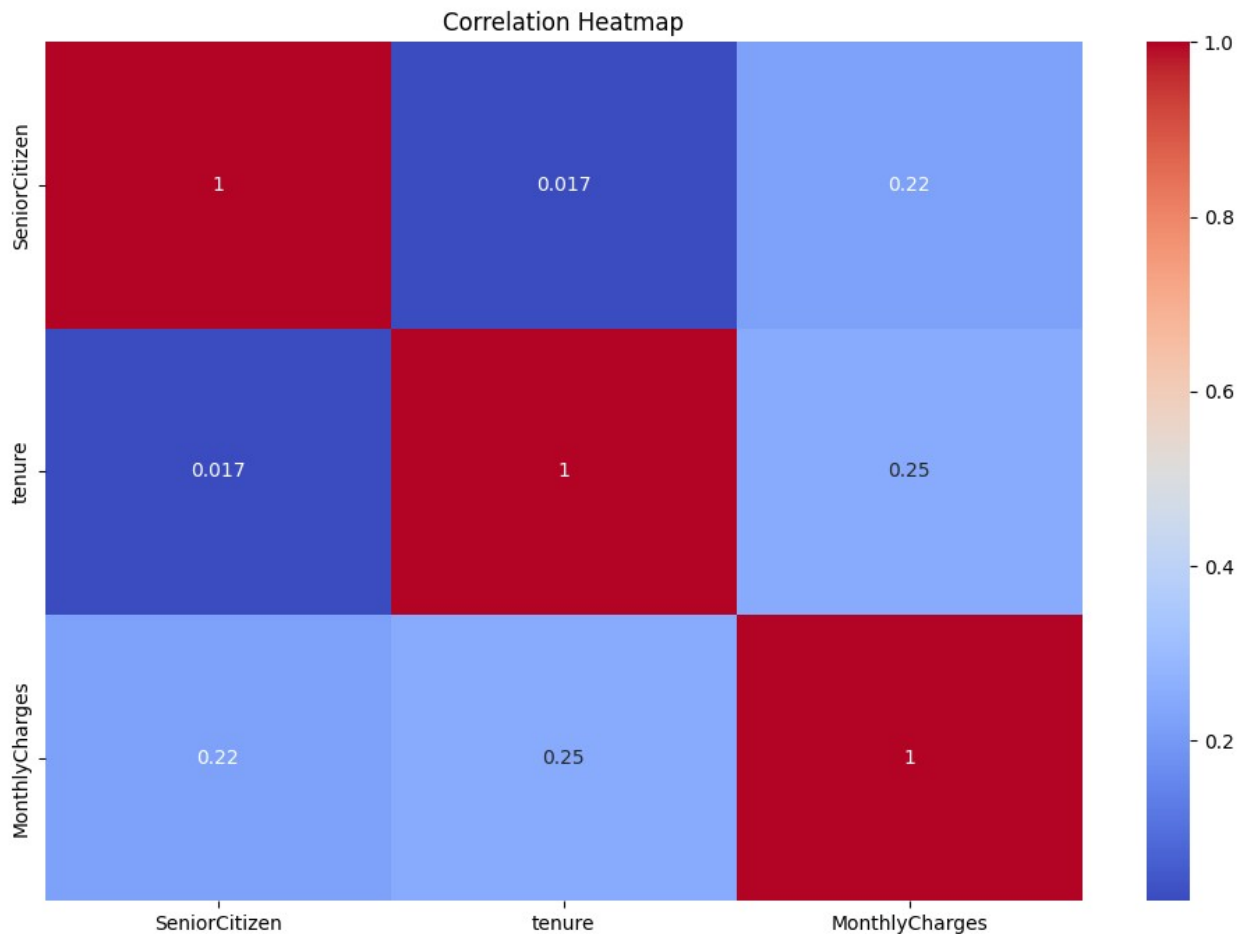
df['Churn'].value_counts()
sns.countplot(data=df, x='Churn')
plt.title('Churn Count')
plt.show()
```



```
df.hist(bins=30, figsize=(15,10))  
plt.tight_layout()  
plt.show()
```



```
plt.figure(figsize=(12,8))
sns.heatmap(df.corr(numeric_only=True), annot=True, cmap='coolwarm')
plt.title("Correlation Heatmap")
plt.show()
```



```
df['TotalCharges'] = df['TotalCharges'].replace(' ', np.nan)
df['TotalCharges'] = df['TotalCharges'].astype(float)
df['TotalCharges'].isnull().sum() # check again

np.int64(11)

df = df.dropna()

from sklearn.preprocessing import LabelEncoder

le = LabelEncoder()
binary_cols = ['gender', 'Partner', 'Dependents', 'PhoneService',
               'PaperlessBilling', 'Churn']

for col in binary_cols:
    df[col] = le.fit_transform(df[col])

df = pd.get_dummies(df, columns=['InternetService', 'OnlineSecurity',
                                'OnlineBackup',
                                'DeviceProtection', 'TechSupport',
                                'StreamingTV'],
```

```

        'StreamingMovies', 'Contract',
        'PaymentMethod'], drop_first=True)

def tenure_group(tenure):
    if tenure <= 12:
        return '0-1 year'
    elif tenure <= 24:
        return '1-2 years'
    elif tenure <= 48:
        return '2-4 years'
    elif tenure <= 60:
        return '4-5 years'
    else:
        return '5+ years'

df['tenure_group'] = df['tenure'].apply(tenure_group)
df = pd.get_dummies(df, columns=['tenure_group'], drop_first=True)

df.info()
df.head()

```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 7032 entries, 0 to 7042
```

```
Data columns (total 35 columns):
```

#	Column	Non-Null Count	Dtype
0	customerID	7032 non-null	object
1	gender	7032 non-null	int64
2	SeniorCitizen	7032 non-null	int64
3	Partner	7032 non-null	int64
4	Dependents	7032 non-null	int64
5	tenure	7032 non-null	int64
6	PhoneService	7032 non-null	int64
7	MultipleLines	7032 non-null	object
8	PaperlessBilling	7032 non-null	int64
9	MonthlyCharges	7032 non-null	float64
10	TotalCharges	7032 non-null	float64
11	Churn	7032 non-null	int64
12	InternetService_Fiber optic	7032 non-null	bool
13	InternetService_No	7032 non-null	bool
14	OnlineSecurity_No internet service	7032 non-null	bool
15	OnlineSecurity_Yes	7032 non-null	bool
16	OnlineBackup_No internet service	7032 non-null	bool
17	OnlineBackup_Yes	7032 non-null	bool
18	DeviceProtection_No internet service	7032 non-null	bool
19	DeviceProtection_Yes	7032 non-null	bool
20	TechSupport_No internet service	7032 non-null	bool
21	TechSupport_Yes	7032 non-null	bool
22	StreamingTV_No internet service	7032 non-null	bool
23	StreamingTV_Yes	7032 non-null	bool

24	StreamingMovies_No internet service	7032	non-null	bool
25	StreamingMovies_Yes	7032	non-null	bool
26	Contract_One year	7032	non-null	bool
27	Contract_Two year	7032	non-null	bool
28	PaymentMethod_Credit card (automatic)	7032	non-null	bool
29	PaymentMethod_Electronic check	7032	non-null	bool
30	PaymentMethod_Mailed check	7032	non-null	bool
31	tenure_group_1-2 years	7032	non-null	bool
32	tenure_group_2-4 years	7032	non-null	bool
33	tenure_group_4-5 years	7032	non-null	bool
34	tenure_group_5+ years	7032	non-null	bool

dtypes: bool(23), float64(2), int64(8), object(2)

memory usage: 872.1+ KB

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	\
0	7590-VHVEG	0	0	1	0	1	
1	5575-GNVDE	1	0	0	0	34	
2	3668-QPYBK	1	0	0	0	2	
3	7795-CF0CW	1	0	0	0	45	
4	9237-HQITU	0	0	0	0	2	

	PhoneService	MultipleLines	PaperlessBilling
MonthlyCharges	...	\	
0	0	No phone service	1
29.85	...		
1	1	No	0
56.95	...		
2	1	No	1
53.85	...		
3	0	No phone service	0
42.30	...		
4	1	No	1
70.70	...		

	StreamingMovies_Yes	Contract_One year	Contract_Two year	\
0	False	False	False	
1	False	True	False	
2	False	False	False	
3	False	True	False	
4	False	False	False	

	PaymentMethod_Credit card (automatic)	PaymentMethod_Electronic check	\
0	False		
True			
1	False		
False			
2	False		
False			
3	False		

False		
4		False
True		

	PaymentMethod_Mailed check	tenure_group_1-2 years	tenure_group_2-4 years \
0	False		False
False			
1	True		False
True			
2	True		False
False			
3	False		False
True			
4	False		False
False			

	tenure_group_4-5 years	tenure_group_5+ years
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False

[5 rows x 35 columns]

```
X = df.drop(['customerID', 'Churn'], axis=1) # Features
y = df['Churn'] # Target
```

```
from sklearn.model_selection import train_test_split
```

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

```
df_encoded = pd.get_dummies(df, drop_first=True)
```

```
# 1. Imports
```

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

```
# 2. Data (assumes df is already loaded and cleaned)
```

```
# Convert categorical variables (one-hot encode)
```

```
df_encoded = pd.get_dummies(df, drop_first=True)
```

```
# 3. Feature and Target Separation
```

```
X = df_encoded.drop('Churn', axis=1)
```

```
y = df_encoded['Churn']
```

4. Train-test Split

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

5. Model Training

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

6. Predictions

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

7. Evaluation

```
print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))  
print("\nClassification Report:\n", classification_report(y_test,  
y_pred))
```

Confusion Matrix:

```
[[930 103]  
 [198 176]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.82	0.90	0.86	1033
1	0.63	0.47	0.54	374
accuracy			0.79	1407
macro avg	0.73	0.69	0.70	1407
weighted avg	0.77	0.79	0.78	1407

```
from sklearn.ensemble import RandomForestClassifier
```

```
rf = RandomForestClassifier()  
rf.fit(X_train, y_train)
```

RandomForestClassifier()

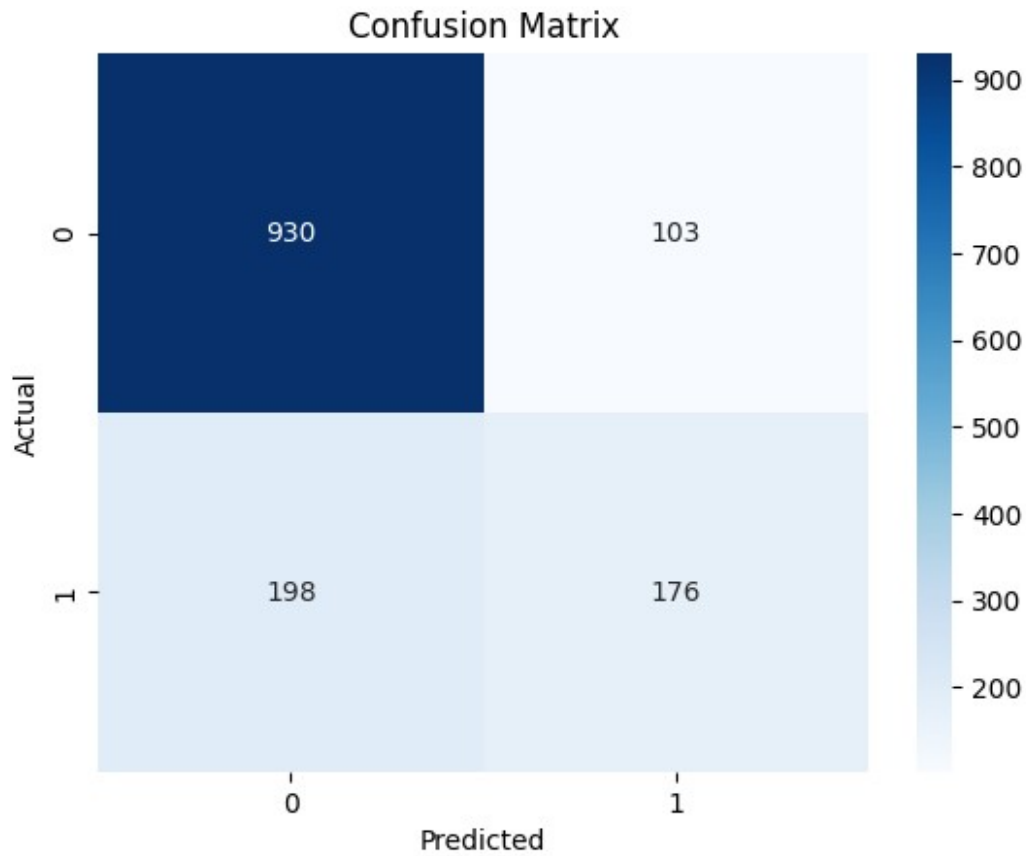
```
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay  
import seaborn as sns  
import matplotlib.pyplot as plt
```

Get confusion matrix

```
cm = confusion_matrix(y_test, y_pred)
```

Plot using seaborn

```
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")  
plt.title("Confusion Matrix")  
plt.xlabel("Predicted")  
plt.ylabel("Actual")  
plt.show()
```

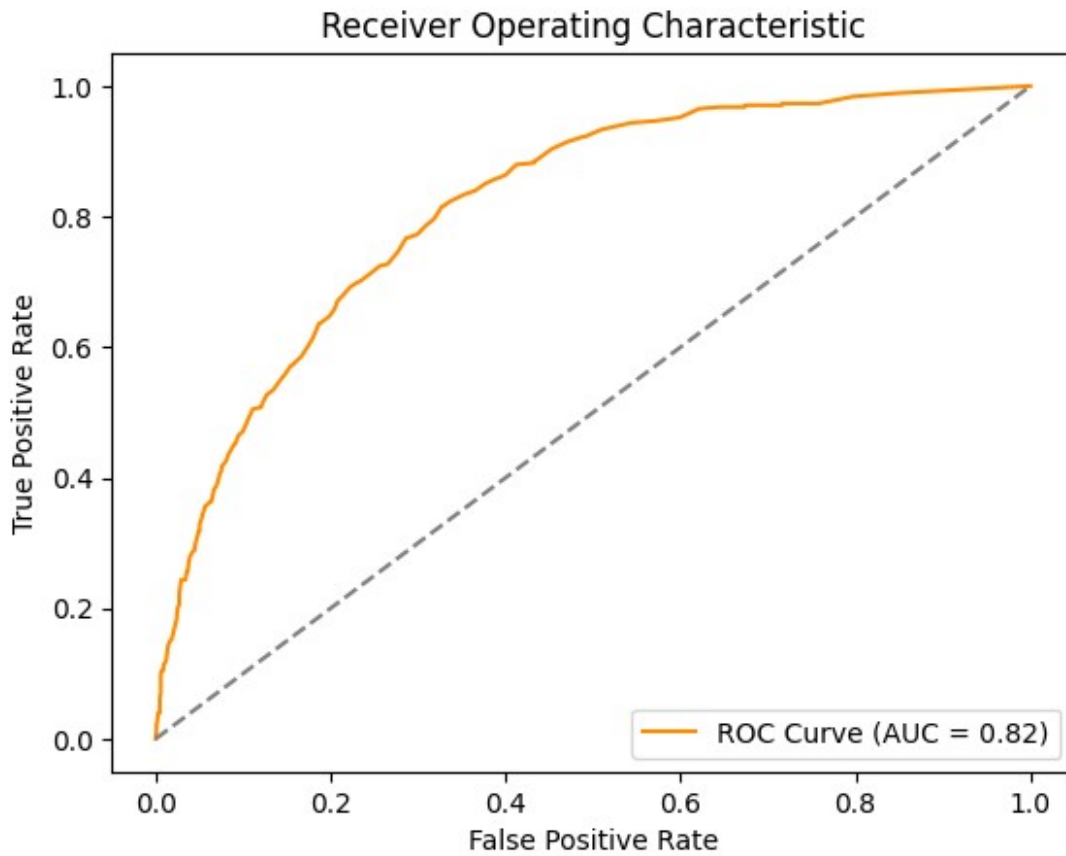



```
from sklearn.metrics import roc_curve, auc

# Probabilities for ROC
y_prob = model.predict_proba(X_test)[: , 1]

# ROC values
fpr, tpr, thresholds = roc_curve(y_test, y_prob)
roc_auc = auc(fpr, tpr)

# Plot ROC
plt.figure()
plt.plot(fpr, tpr, color='darkorange', label=f'ROC Curve (AUC = {roc_auc:.2f})')
plt.plot([0, 1], [0, 1], linestyle='--', color='gray')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('Receiver Operating Characteristic')
plt.legend(loc="lower right")
plt.show()
```

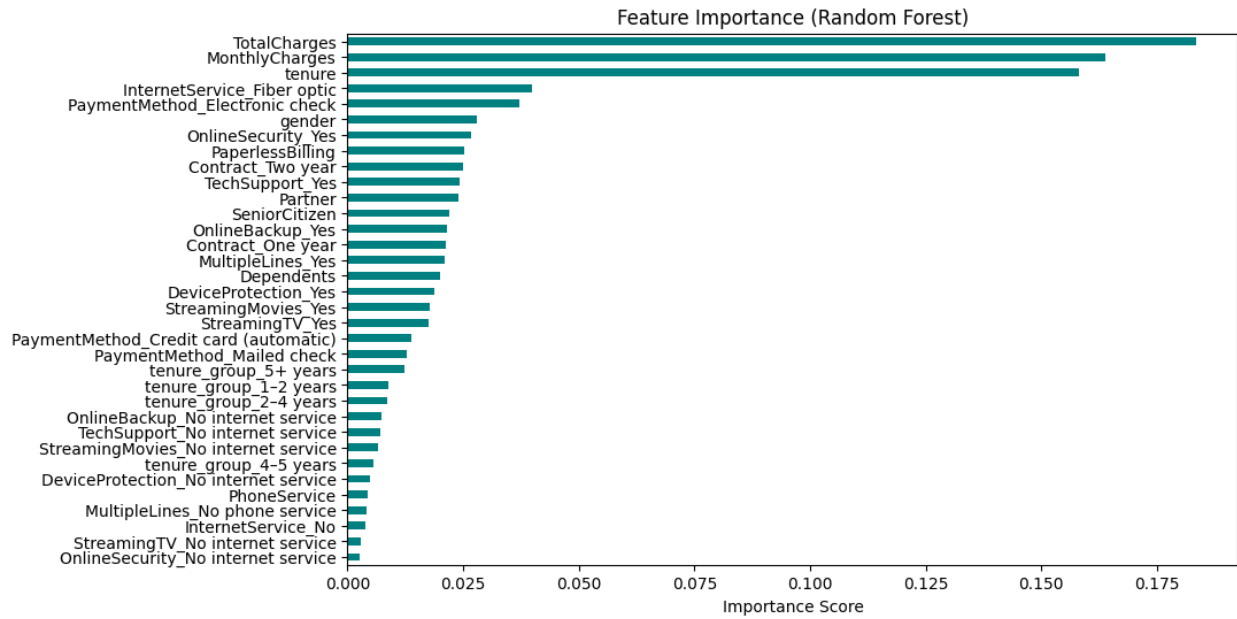


```
import pandas as pd
import matplotlib.pyplot as plt

# Feature importance from RandomForest
importances = model.feature_importances_

# Create a Series with feature names and importance values
feat_imp = pd.Series(importances, index=X.columns)

# Plot
feat_imp.sort_values().plot(kind='barh', figsize=(10,6), color='teal')
plt.title("Feature Importance (Random Forest)")
plt.xlabel("Importance Score")
plt.show()
```



```
df.head(15)
```

	gender	SeniorCitizen	Partner	Dependents	tenure	
PhoneService \						
0	0	0	1	0	1	0
1	1	0	0	0	34	1
2	1	0	0	0	2	1
3	1	0	0	0	45	0
4	0	0	0	0	2	1
5	0	0	0	0	8	1
6	1	0	0	1	22	1
7	0	0	0	0	10	0
8	0	0	1	0	28	1
9	1	0	0	1	62	1
10	1	0	1	1	13	1
11	1	0	0	0	16	1
12	1	0	1	0	58	1
13	1	0	0	0	49	1

14	1	0	0	0	25	1
----	---	---	---	---	----	---

	MultipleLines	PaperlessBilling	MonthlyCharges
--	---------------	------------------	----------------

TotalCharges	...	\
--------------	-----	---

0	No phone service	1	29.85
---	------------------	---	-------

29.85	...
-------	-----

1	No	0	56.95
---	----	---	-------

1889.50	...
---------	-----

2	No	1	53.85
---	----	---	-------

108.15	...
--------	-----

3	No phone service	0	42.30
---	------------------	---	-------

1840.75	...
---------	-----

4	No	1	70.70
---	----	---	-------

151.65	...
--------	-----

5	Yes	1	99.65
---	-----	---	-------

820.50	...
--------	-----

6	Yes	1	89.10
---	-----	---	-------

1949.40	...
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7	No phone service	0	29.75
---	------------------	---	-------

301.90	...
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8	Yes	1	104.80
---	-----	---	--------

3046.05	...
---------	-----

9	No	0	56.15
---	----	---	-------

3487.95	...
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10	No	1	49.95
----	----	---	-------

587.45	...
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11	No	0	18.95
----	----	---	-------

326.80	...
--------	-----

12	Yes	0	100.35
----	-----	---	--------

5681.10	...
---------	-----

13	Yes	1	103.70
----	-----	---	--------

5036.30	...
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14	No	1	105.50
----	----	---	--------

2686.05	...
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	StreamingMovies_Yes	Contract_One year	Contract_Two year	\
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0	False	False	False
---	-------	-------	-------

1	False	True	False
---	-------	------	-------

2	False	False	False
---	-------	-------	-------

3	False	True	False
---	-------	------	-------

4	False	False	False
---	-------	-------	-------

5	True	False	False
---	------	-------	-------

6	False	False	False
---	-------	-------	-------

7	False	False	False
---	-------	-------	-------

8	True	False	False
---	------	-------	-------

9	False	True	False
---	-------	------	-------

10	False	False	False
----	-------	-------	-------

11	False	False	True
----	-------	-------	------

12	True	True	False
----	------	------	-------

13	True	False	False
14	True	False	False
PaymentMethod_Credit card (automatic) PaymentMethod_Electronic			
check \			
0		False	
True			
1		False	
False			
2		False	
False			
3		False	
False			
4		False	
True			
5		False	
True			
6		True	
False			
7		False	
False			
8		False	
True			
9		False	
False			
10		False	
False			
11		True	
False			
12		True	
False			
13		False	
False			
14		False	
True			
PaymentMethod_Mailed check tenure_group_1-2 years \			
0	False	False	
1	True	False	
2	True	False	
3	False	False	
4	False	False	
5	False	False	
6	False	True	
7	True	False	
8	False	False	
9	False	False	
10	True	True	
11	False	True	

12	False	False
13	False	False
14	False	False

	tenure_group_2–4 years	tenure_group_4–5 years	tenure_group_5+ years
0	False	False	
False			
1	True	False	
False			
2	False	False	
False			
3	True	False	
False			
4	False	False	
False			
5	False	False	
False			
6	False	False	
False			
7	False	False	
False			
8	True	False	
False			
9	False	False	
True			
10	False	False	
False			
11	False	False	
False			
12	False	True	
False			
13	False	True	
False			
14	True	False	
False			

[15 rows x 34 columns]

Customer Churn Analysis - Final Report (Week 4)

Model Performance:

The Random Forest model achieved approximately **79% accuracy**.

Confusion Matrix Results:

- Precision (Churn=1): ~63%
- Recall (Churn=1): ~47%

Key Drivers of Churn:

- Month-to-month contracts increase churn likelihood.
- Lack of Tech Support / Online Security strongly correlates with churn.
- Customers using Electronic Check payments churn more often.
- Customers with tenure less than 1 year are at highest risk.
- High monthly charges without additional value increase churn.

Business Recommendations:

- Promote long-term contracts (annual/two-year plans).
- Bundle security & tech support services at reduced prices.
- Encourage customers to switch from electronic check to auto-pay methods.
- Focus retention programs on new customers (first-year tenure).
- Offer loyalty rewards or discounts for high-charge customers.

Conclusion:

By implementing these strategies, the telecom company can significantly reduce churn and improve customer lifetime value.