

FDS_Project(CustomerChurnPrediction)

October 28, 2025

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sn
```

```
[2]: df=pd.read_csv('Telco-Customer-Churn.csv')
```

```
[3]: df.head(5)
```

```
[3]:  customerID  gender  SeniorCitizen  Partner  Dependents  tenure  PhoneService  \
0  7590-VHVEG  Female                0      Yes           No         1           No
1  5575-GNVDE   Male                0      No            No        34           Yes
2  3668-QPYBK   Male                0      No            No         2           Yes
3  7795-CFOCW   Male                0      No            No        45           No
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]

```
[4]: df=df.dropna()
```

```
[5]: df
```

```
[5]:
```

	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-CFOCW	Male	0	No	No	45	
4	9237-HQITU	Female	0	No	No	2	
...	
7038	6840-RESVB	Male	0	Yes	Yes	24	
7039	2234-XADUH	Female	0	Yes	Yes	72	
7040	4801-JZAZL	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	...	\
0	No	No phone service	DSL	No	...	
1	Yes	No	DSL	Yes	...	
2	Yes	No	DSL	Yes	...	
3	No	No phone service	DSL	Yes	...	
4	Yes	No	Fiber optic	No	...	
...	
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	\
0	No	No	No	No	Month-to-month	
1	Yes	No	No	No	One year	
2	No	No	No	No	Month-to-month	
3	Yes	Yes	No	No	One year	
4	No	No	No	No	Month-to-month	
...	
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	\
0	Yes	Electronic check	29.85	29.85	

1	No	Mailed check	56.95	1889.5
2	Yes	Mailed check	53.85	108.15
3	No	Bank transfer (automatic)	42.30	1840.75
4	Yes	Electronic check	70.70	151.65
...
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	
0	No
1	No
2	Yes
3	No
4	Yes
...	...
7038	No
7039	No
7040	No
7041	Yes
7042	No

[7043 rows x 21 columns]

```
[6]: df.drop_duplicates(inplace=True)
```

```
[7]: df.head(5)
```

```
[7]:
```

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	\
0	7590-VHVEG	Female	0	Yes	No	1	No	
1	5575-GNVDE	Male	0	No	No	34	Yes	
2	3668-QPYBK	Male	0	No	No	2	Yes	
3	7795-CFOCW	Male	0	No	No	45	No	
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]

```
[8]: df.to_csv('Telco-Customer-Churn.csv', index=False)
```

```
[9]: target=df['Churn']
```

```
[10]: df=df.drop('Churn',axis=1)
```

```
[11]: df.head(5)
```

```
[11]:
```

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

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

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

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

4	Yes	Electronic check	70.70	151.65
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```
[12]: from sklearn.linear_model import LogisticRegression
```

```
[13]: model1=LogisticRegression()
```

```
[14]: from sklearn import tree
```

```
[15]: model2=tree.DecisionTreeClassifier(random_state=42)
```

```
[16]: from sklearn.ensemble import RandomForestClassifier
```

```
[17]: model3=RandomForestClassifier(n_estimators=80)
```

```
[18]: df.drop('customerID', axis=1, inplace=True)
```

```
[19]: target=target.replace({'Yes':1,'No':0})
```

```
C:\Users\aswin\AppData\Local\Temp\ipykernel_33836\3723651564.py:1:
FutureWarning: Downcasting behavior in `replace` is deprecated and will be
removed in a future version. To retain the old behavior, explicitly call
`result.infer_objects(copy=False)`. To opt-in to the future behavior, set
`pd.set_option('future.no_silent_downcasting', True)`
target=target.replace({'Yes':1,'No':0})
```

```
[20]: df = pd.get_dummies(df, drop_first=True)
```

```
[21]: df.head(5)
```

```
[21]: SeniorCitizen  tenure  MonthlyCharges  gender_Male  Partner_Yes  \
0                0        1           29.85         False         True
1                0       34           56.95          True         False
2                0        2           53.85          True         False
3                0       45           42.30          True         False
4                0        2           70.70         False         False

    Dependents_Yes  PhoneService_Yes  MultipleLines_No phone service  \
0             False              False                        True
1             False              True                        False
2             False              True                        False
3             False              False                        True
4             False              True                        False

    MultipleLines_Yes  InternetService_Fiber optic  ...  TotalCharges_995.35  \
0             False                        False  ...                False
1             False                        False  ...                False
2             False                        False  ...                False
3             False                        False  ...                False
```

4	False	True	...	False
---	-------	------	-----	-------

	TotalCharges_996.45	TotalCharges_996.85	TotalCharges_996.95	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	

	TotalCharges_997.65	TotalCharges_997.75	TotalCharges_998.1	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	

	TotalCharges_999.45	TotalCharges_999.8	TotalCharges_999.9
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False

[5 rows x 6559 columns]

```
[22]: from sklearn.model_selection import train_test_split
      x_train,x_test,y_train,y_test=train_test_split(df,target,test_size=0.5)
```

```
[23]: model1.fit(x_train,y_train)
```

C:\Users\aswin\anaconda3\Lib\site-packages\sklearn\linear_model_logistic.py:465: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. OF ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

```
n_iter_i = _check_optimize_result(
```

```
[23]: LogisticRegression()
```

```
[24]: model1.score(x_test,y_test)
```

```
[24]: 0.7955706984667802
```

```

[25]: model2.fit(x_train,y_train)

[25]: DecisionTreeClassifier(random_state=42)

[26]: model2.score(x_test,y_test)

[26]: 0.7637705848949461

[27]: model3.fit(x_train,y_train)

[27]: RandomForestClassifier(n_estimators=80)

[28]: model3.score(x_test,y_test)

[28]: 0.7842135150482681

[29]: from sklearn.ensemble import VotingClassifier

[30]: c_modelslist = [('logreg', model1), ('dtree', model2), ('rforest', model3)]

[31]: c_model = VotingClassifier(estimators=c_modelslist, voting='soft', n_jobs=-1 )

[32]: c_model.fit(x_train,y_train)

[32]: VotingClassifier(estimators=[('logreg', LogisticRegression()),
                                ('dtree', DecisionTreeClassifier(random_state=42)),
                                ('rforest',
                                 RandomForestClassifier(n_estimators=80))],
                                n_jobs=-1, voting='soft')

[33]: c_model.score(x_test,y_test)

[33]: 0.778534923339012

[34]: c_model.predict(x_test)

[34]: array([0, 0, 0, ..., 0, 1, 0])

[35]: test=np.array([[0.0, -1.277, 1.341, -0.992, 0.0, 0.0, 0.0, 1.0, 0.0, 0.0, 1.0,
    ↪ 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 1.
    ↪ 0, 0.0, 1.0, 0.0]])

[36]: test = pd.DataFrame(test, columns=x_train.columns[:test.shape[1]])

[37]: test = test.reindex(columns=x_train.columns, fill_value=0)

[38]: c_model.predict(test)

[38]: array([0])

```

```
[39]: test_input = np.array([[
    0,      # SeniorCitizen = No (0)
    1,      # tenure = very low
    99.90,  # MonthlyCharges = very high
    1,      # gender_Male
    0,      # Partner_Yes
    0,      # Dependents_Yes
    1,      # PhoneService_Yes
    0,      # MultipleLines_No phone service
    1,      # MultipleLines_Yes
    1,      # InternetService_Fiber optic
    1,0,0,1,0,0,1,0,0,0,0,0,0,0,0,0,0,0
]])
```

```
[40]: test = pd.DataFrame(test, columns=x_train.columns[:test.shape[1]])
```

```
[41]: test = test.reindex(columns=x_train.columns, fill_value=0)
```

```
[42]: c_model.predict(test)
```

```
[42]: array([0])
```

```
[43]: churned_customer = df[target == 1].sample(1)
```

```
[44]: churned_customer_array = churned_customer.values.reshape(1, -1)
```

```
[45]: c_model.predict(churned_customer_array)
```

```
C:\Users\aswin\anaconda3\Lib\site-packages\sklearn\utils\validation.py:2739:
```

```
UserWarning: X does not have valid feature names, but LogisticRegression was fitted with feature names
```

```
warnings.warn(
```

```
C:\Users\aswin\anaconda3\Lib\site-packages\sklearn\utils\validation.py:2739:
```

```
UserWarning: X does not have valid feature names, but DecisionTreeClassifier was fitted with feature names
```

```
warnings.warn(
```

```
C:\Users\aswin\anaconda3\Lib\site-packages\sklearn\utils\validation.py:2739:
```

```
UserWarning: X does not have valid feature names, but RandomForestClassifier was fitted with feature names
```

```
warnings.warn(
```

```
[45]: array([0])
```

```
[46]: y_predict=c_model.predict(x_test)
```

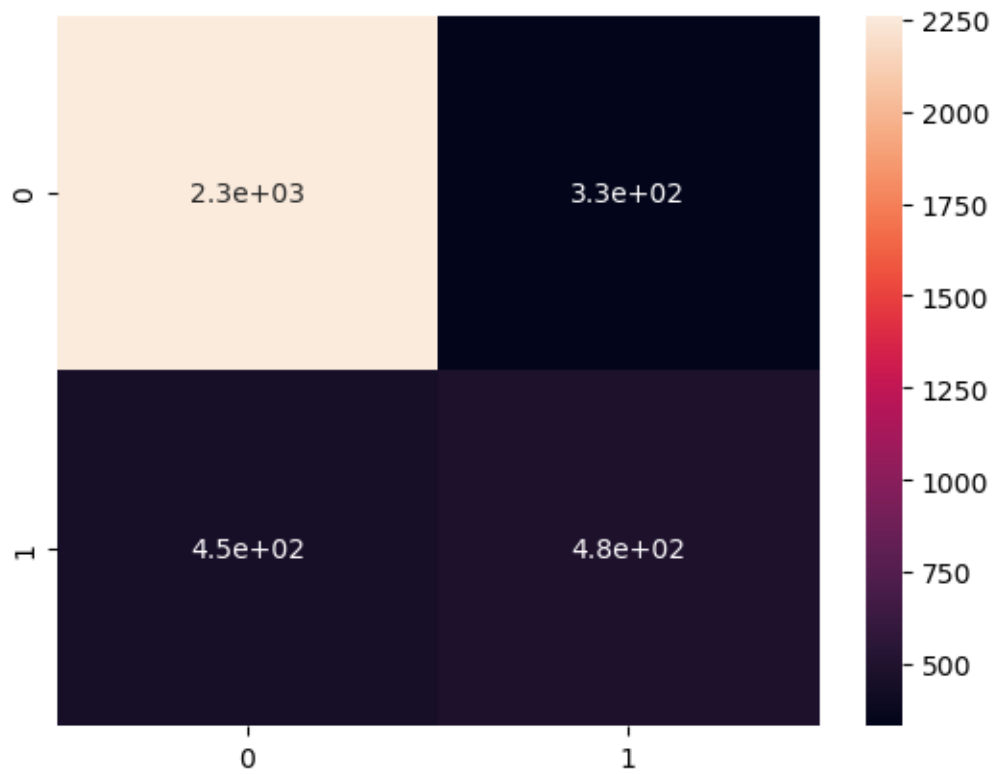
```
[47]: from sklearn.metrics import confusion_matrix
```

```
[48]: visual=confusion_matrix(y_test,y_predict)
```



```
[49]: sn.heatmap(visual,annot=True)
```

```
[49]: <Axes: >
```



```
[ ]:
```