

MACHINE LEARNING END-TO-END PROJECT

CUSTOMER CHURN PREDICTION

TELECOM

WPS Office tel_churn.csv (Read-only) first_telc.csv Go Premium

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		A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U
1		gender	SeniorCitize	Partner	Dependents	tenure	PhoneServic	MultipleLine	InternetSen	OnlineSecur	OnlineBack	DeviceProt	TechSuppor	StreamingTV	StreamingM	Contract	PaperlessBill	PaymentMe	MonthlyChu	TotalCharges		
2	0	Female	0	Yes	No	1	No	No phone si DSL	No	Yes	No	No	No	No	Month-to-n Yes	Electronic c	29.85	29.85				
3	1	Male	0	No	No	34	Yes	No	DSL	Yes	No	Yes	No	No	One year	No	Mailed chec	56.95	1889.5			
4	3	Male	0	No	No	45	No	No phone si DSL	Yes	No	Yes	Yes	No	No	One year	No	Bank transfi	42.3	1840.75			
5	6	Male	0	No	Yes	22	Yes	Fiber optic	No	Yes	No	No	Yes	No	Month-to-n Yes	Credit card i	89.1	1949.4				
6	11	Male	0	No	No	16	Yes	No	No	No internet	Two year	No	Credit card i	18.95	326.8							
7	5	Female	0	No	No	8	Yes	Fiber optic	No	No	Yes	No	Yes	Yes	Month-to-n Yes	Electronic c	99.65	820.5				
8	4	Female	0	No	No	2	Yes	No	Fiber optic	No	No	No	No	No	Month-to-n Yes	Electronic c	70.7	151.65				
9	2	Male	0	No	No	2	Yes	No	DSL	Yes	Yes	No	No	No	Month-to-n Yes	Mailed chec	53.85	108.15				
10	7	Female	0	No	No	10	No	No phone si DSL	Yes	No	No	No	No	No	Month-to-n No	Mailed chec	29.75	301.9				
11	8	Female	0	Yes	No	28	Yes	Fiber optic	No	No	Yes	Yes	Yes	Yes	Month-to-n Yes	Electronic c	104.8	3046.05				
12	9	Male	0	No	Yes	62	Yes	No	DSL	Yes	Yes	No	No	No	One year	Bank transfi	56.15	3487.95				
13	10	Male	0	Yes	Yes	13	Yes	No	DSL	Yes	No	No	No	No	Month-to-n Yes	Mailed chec	49.95	587.45				
14	12	Male	0	Yes	No	58	Yes	Fiber optic	No	No	Yes	No	Yes	Yes	One year	No	Credit card i	100.35	5681.1			
15	13	Male	0	No	No	49	Yes	Fiber optic	No	Yes	No	Yes	No	Yes	Month-to-n Yes	Bank transfi	103.7	5036.3				
16	14	Male	0	No	No	25	Yes	No	Fiber optic	Yes	No	Yes	Yes	Yes	Month-to-n Yes	Electronic c	105.5	2686.05				
17	15	Female	0	Yes	Yes	69	Yes	Fiber optic	Yes	Yes	Yes	Yes	Yes	Yes	Two year	No	Credit card i	113.25	7895.15			
18	16	Female	0	No	No	52	Yes	No	No	No internet	One year	No	Mailed chec	20.65	1022.95							
19	17	Male	0	No	Yes	71	Yes	Fiber optic	Yes	No	Yes	No	Yes	Yes	Two year	No	Bank transfi	106.7	7382.25			
20	19	Female	0	No	No	21	Yes	No	Fiber optic	No	Yes	Yes	No	Yes	Month-to-n Yes	Electronic c	90.05	1862.9				
21	21	Male	0	Yes	No	12	Yes	No	No	No internet	One year	No	Bank transfi	19.8	202.25							
22	25	Female	0	No	No	30	Yes	No	DSL	Yes	Yes	No	No	No	Month-to-n Yes	Bank transfi	55.3	1530.6				
23	26	Male	0	Yes	Yes	47	Yes	Fiber optic	No	Yes	No	No	Yes	Yes	Month-to-n Yes	Electronic c	99.35	4749.15				
24	28	Male	0	Yes	No	72	Yes	DSL	Yes	Yes	Yes	Yes	Yes	Yes	Two year	Yes	Credit card i	90.25	6369.45			
25	29	Female	0	No	Yes	17	Yes	No	DSL	No	No	No	No	Yes	Month-to-n Yes	Mailed chec	64.7	1093.1				
26	32	Female	0	Yes	Yes	27	Yes	No	DSL	Yes	Yes	Yes	Yes	No	One year	No	Mailed chec	66.15	1874.45			
27	36	Male	0	No	No	5	Yes	No	Fiber optic	No	No	No	No	No	Month-to-n Yes	Electronic c	69.7	316.9				
28	37	Female	0	No	No	46	Yes	No	Fiber optic	No	No	Yes	No	No	Month-to-n Yes	Credit card i	74.8	3548.3				
29	39	Female	0	No	No	11	Yes	Fiber optic	No	No	Yes	No	Yes	Yes	Month-to-n Yes	Bank transfi	97.85	1105.4				
30	41	Female	0	Yes	Yes	70	Yes	DSL	Yes	Yes	No	No	Yes	No	Two year	Yes	Credit card i	69.2	4872.35			

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		A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U
1		SeniorCitize	MonthlyChu	TotalCharge	Churn	gender_F	gender_M	Partner_No	Partner_Yes	Dependents	Dependents	PhoneServic	PhoneServic	MultipleLine	MultipleLine	MultipleLine	MultipleLine	InternetSen	InternetSen	InternetSen	OnlineSecur	OnlineSe
2	0	0	29.85	29.85	0	1	0	0	0	1	1	0	1	0	0	1	0	1	0	0	0	1
3	1	0	56.95	1889.5	0	0	1	1	0	1	0	0	1	1	0	0	1	0	0	0	0	0
4	2	0	53.85	108.15	1	0	1	1	0	1	0	0	0	1	1	0	0	1	0	0	0	0
5	3	0	42.3	1840.75	0	0	1	1	0	1	0	0	1	0	0	1	0	1	0	0	0	0
6	4	0	70.7	151.65	1	1	0	1	0	1	0	0	1	1	0	0	0	0	1	0	0	1
7	5	0	99.65	820.5	1	1	0	1	0	1	0	0	1	0	0	1	0	1	0	1	0	1
8	6	0	89.1	1949.4	0	0	1	1	0	0	0	1	0	1	0	0	1	0	1	0	1	1
9	7	0	29.75	301.9	0	1	0	1	0	1	0	1	0	1	0	0	1	0	1	0	0	0
10	8	0	104.8	3046.05	1	1	0	0	1	1	0	0	1	0	0	1	0	1	0	1	0	1
11	9	0	56.15	3487.95	0	0	1	1	0	0	1	0	1	1	0	0	1	0	1	0	0	0
12	10	0	49.95	587.45	0	0	1	0	1	0	1	0	1	1	0	0	1	0	1	0	0	0
13	11	0	18.95	326.8	0	0	1	1	0	1	0	0	1	1	0	0	0	0	1	0	1	0
14	12	0	100.35	5681.1	0	0	1	0	1	1	0	0	0	1	0	0	1	0	1	0	1	1
15	13	0	103.7	5036.3	1	0	1	1	0	1	0	0	1	0	0	1	0	1	0	1	0	1
16	14	0	105.5	2686.05	0	0	1	1	0	1	0	0	1	1	0	0	1	0	1	0	1	0
17	15	0	113.25	7895.15	0	1	0	0	1	0	1	0	1	0	0	1	0	1	0	1	0	0
18	16	0	20.65	1022.95	0	1	0	1	0	1	0	0	1	1	0	0	0	0	1	0	1	0
19	17	0	106.7	7382.25	0	0	1	1	0	0	1	0	1	0	0	1	0	1	0	1	0	0
20	18	0	55.2	528.35	1	1	0	0	1	0	1	0	1	1	0	0	1	0	1	0	1	1
21	19	0	90.05	1862.9	0	1	0	1	0	1	0	0	1	1	1	0	0	0	1	0	1	1
22	20	1	39.65	39.65	1	0	1	1	0	1	0	1	0	0	1	0	1	0	1	0	1	1
23	21	0	19.8	202.25	0	0	1	0	1	1	0	0	1	1	0	0	0	0	1	0	1	0
24	22	0	20.15	20.15	1	0	1	1	0	1	0	0	1	1	0	0	0	0	1	0	1	0
25	23	0	59.9	3505.1	0	1	0	0	1	1	0	0	1	0	0	1	1	0	1	0	1	1
26	24	0	59.6	2970.3	0	0	1	0	1	0	1	0	1	1	0	0	1	1	0	0	0	0
27	25	0	55.3	1530.6	0	1	0	1	0	1	0	0	1	1	0	0	0	1	0	0	0	0
28	26	0	99.35	4749.15	1	0	1	0	1	0	1	0	1	0	0	1	0	1	0	1	0	1
29	27	0	30.2	30.2	1	0	1	0	1	0	1	0	1	1	0	0	1	0	1	0	0	1
30	28	0	90.25	6369.45	0	0	1	0	1	1	0	0	1	0	0	1	0	1	0	0	0	0

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localhost:8888/notebooks/Project%20Churn%20Analysis.ipynb

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In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.ticker as mtick
import matplotlib.pyplot as plt
%matplotlib inline
```

In [4]:

```
telco_base_data = pd.read_csv("C:/Users/DELL/Downloads/WA_Fn-UseC_-Telco-Customer-Churn.csv")
```

In [5]:

```
telco_base_data.head()
```

Out[5]:

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	TechSupport	StreamingTV	StreamingMovies	Contract	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges	Churn
0	VtVEG	Female	0	Yes	No	1	No	No phone service	DSL	No	...	No	Yes	Yes	Month-to-month	Electronic check	10.0	10.0	Yes	
1	GNVDE	Male	0	No	No	34	Yes	No	DSL	Yes	...	Yes	Yes	Yes	One year	Billing by month	10.0	10.0	No	
2	QPYBK	Male	0	No	No	2	Yes	No	DSL	Yes	...	No	Yes	Yes	Month-to-month	Electronic check	10.0	10.0	No	
3	7795-CFOCW	Male	0	No	No	45	No	No phone service	DSL	Yes	...	Yes	Yes	Yes	Month-to-month	Check-by-mail	10.0	10.0	No	
4	9237-HQITU	Female	0	No	No	2	Yes	No	Fiber optic	No	...	No	Yes	Yes	Month-to-month	Check-by-mail	10.0	10.0	No	

5 rows × 21 columns

In [6]:

```
telco_base_data.shape
```

Out[6]:

```
(7043, 21)
```

In [7]:

```
telco_base_data.columns.values
```

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In [7]:

```
telco_base_data.columns.values
```

Out[7]:

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

In [8]:

```
telco_base_data.dtypes
```

Out[8]:

```
customerID    object
gender        object
SeniorCitizen  int64
Partner        object
Dependents     object
tenure         int64
PhoneService   object
MultipleLines  object
InternetService object
OnlineSecurity object
OnlineBackup    object
DeviceProtection object
TechSupport    object
StreamingTV    object
StreamingMovies object
Contract       object
PaperlessBilling object
PaymentMethod   object
MonthlyCharges float64
TotalCharges   object
Churn          object
dtype: object
```

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In [8]:

```
PaperlessBilling    object
PaymentMethod       object
MonthlyCharges      float64
TotalCharges        object
Churn               object
dtype: object
```

In [9]:

```
telco_base_data.describe()
```

Out[9]:

	SeniorCitizen	tenure	MonthlyCharges
count	7043.000000	7043.000000	7043.000000
mean	0.162147	32.371149	64.761692
std	0.368612	24.559481	30.090047
min	0.000000	0.000000	18.250000
25%	0.000000	9.000000	35.500000
50%	0.000000	29.000000	70.350000
75%	0.000000	55.000000	89.850000
max	1.000000	72.000000	118.750000

In [10]:

```
telco_base_data['Churn'].value_counts().plot(kind='barh', figsize=(8, 6))
plt.xlabel("Count", labelpad=14)
plt.ylabel("Target Variable", labelpad=14)
plt.title("Count of TARGET Variable per category", y=1.02);
```

Count of TARGET Variable per category

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In [9]: telco_base_data['Churn'].value_counts().plot(kind='barh', figsize=(5, 3))
 plt.xlabel("Count", labelpad=14)
 plt.ylabel("Target Variable", labelpad=14)
 plt.title("Count of TARGET Variable per category", y=1.02);

Count of TARGET Variable per category

target variable

Count

In [10]: 100*telco_base_data['Churn'].value_counts()/len(telco_base_data['Churn'])

Out[10]: No 73.463013
 Yes 26.536987
 Name: Churn, dtype: float64

In [12]: telco_base_data['Churn'].value_counts()

Out[12]: No 5174
 Yes 7346

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In [13]: telco_base_data.info(verbose = True)

```
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   float64 
 20  Churn           7043 non-null   object  
dtypes: float64(1), int64(2), object(18)
memory usage: 1.1+ MB
```

In [27]: missing = pd.DataFrame((telco_base_data.isnull().sum())*100/telco_base_data.shape[0]).reset_index()

plt.figure(figsize=(16,5))

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In [17]: telco_data = telco_base_data.copy()

In [18]: telco_data.TotalCharges = pd.to_numeric(telco_data.TotalCharges, errors='coerce')
 telco_data.isnull().sum()

customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges	Churn
488	4472-LVYGI	Female	0	Yes	Yes	0	No	No phone service	DSL	Yes	...	Yes						11		0

In [19]: telco_data.loc[telco_data ['TotalCharges'].isnull() == True]

Out[19]:

customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges	Churn
488	4472-LVYGI	Female	0	Yes	Yes	0	No	No phone service	DSL	Yes	...	Yes						11		0

488 4472-LVYGI Female 0 Yes Yes 0 No No phone service DSL Yes ... Yes

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```
In [19]: telco_data.loc[telco_data['TotalCharges'].isnull() == True]
```

	customerID	gender	SeniorCitizen	Partner	Dependents	Tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	SMS
488	4472-LVYGI	Female	0	Yes	Yes	0	No	No phone service	DSL	Yes	...	Yes	Yes	0
753	3115-CZMZD	Male	0	No	Yes	0	Yes	No	No	No	...	No internet service	No internet service	0
936	5708-LVOEQ	Female	0	Yes	Yes	0	Yes	No	DSL	Yes	...	Yes	Yes	0
1082	4367-NUYAO	Male	0	Yes	Yes	0	Yes	Yes	No	No	...	No internet service	No internet service	0
1340	1371-DWPAZ	Female	0	Yes	Yes	0	No	No phone service	DSL	Yes	...	Yes	Yes	0
3331	7644-OMVMY	Male	0	Yes	Yes	0	Yes	No	No	No	...	No internet service	No internet service	0
3826	3213-VVOLG	Male	0	Yes	Yes	0	Yes	Yes	No	No	...	No internet service	No internet service	0
4380	2520-SGTT	Female	0	Yes	Yes	0	Yes	No	No	No	...	No internet service	No internet service	0
5218	2923-ARZLG	Male	0	Yes	Yes	0	Yes	No	No	No	...	No internet service	No internet service	0
6570	4075-WKNUU	Female	0	Yes	Yes	0	Yes	Yes	DSL	No	...	Yes	Yes	0
6754	2776-SEFEE	Male	0	No	Yes	0	Yes	Yes	DSL	Yes	...	No	No	0

11 rows × 21 columns

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```
In [22]: labels = ["{} - {}".format(i, i + 11) for i in range(1, 72, 12)]
telco_data['tenure_group'] = pd.cut(telco_data.tenure, range(1, 80, 12), right=False, labels=labels)
```

```
In [23]: telco_data['tenure_group'].value_counts()
```

Tenure Group	Count
1 - 12	2175
13 - 24	1407
25 - 36	1024
37 - 48	832
49 - 60	832
61 - 72	762

```
In [28]: telco_data.drop(columns=['customerID', 'tenure'], axis=1, inplace=True)
telco_data.head()
```

```
In [28]: gender SeniorCitizen Partner Dependents PhoneService MultipleLines InternetService OnlineSecurity OnlineBackup DeviceProtection TechSupport Str:
```

	gender	SeniorCitizen	Partner	Dependents	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	Str:
0	Female	0	Yes	No	No	No phone service	DSL	No	Yes	No	Yes	No
1	Male	0	No	No	Yes	No	DSL	Yes	No	Yes	No	No
2	Male	0	No	No	Yes	No	DSL	Yes	Yes	No	No	No
3	Male	0	No	No	No	No phone service	DSL	Yes	No	Yes	Yes	Yes
4	Female	0	No	No	Yes	No	Fiber optic	No	No	No	No	No

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```
In [29]: for i, predictor in enumerate(telco_data.drop(columns=['Churn', 'TotalCharges', 'MonthlyCharges'])):
    plt.figure(i)
    sns.countplot(data=telco_data, x=predictor, hue='Churn')
```

```
In [30]: telco_data['Churn'] = np.where(telco_data.Churn == 'Yes', 1, 0)
```

```
In [31]: telco_data.head()
```

```
In [31]: gender SeniorCitizen Partner Dependents PhoneService MultipleLines InternetService OnlineSecurity OnlineBackup DeviceProtection TechSupport Str:
```

	gender	SeniorCitizen	Partner	Dependents	PhoneService	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	DeviceProtection	TechSupport	Str:
0	Female	0	Yes	No	No	No phone service	DSL	No	Yes	No	No	No
1	Male	0	No	No	Yes	No	DSL	Yes	No	Yes	No	No

Jupyter Project Churn Analysis Last Checkpoint 08/19/2023 (unsaved changes)

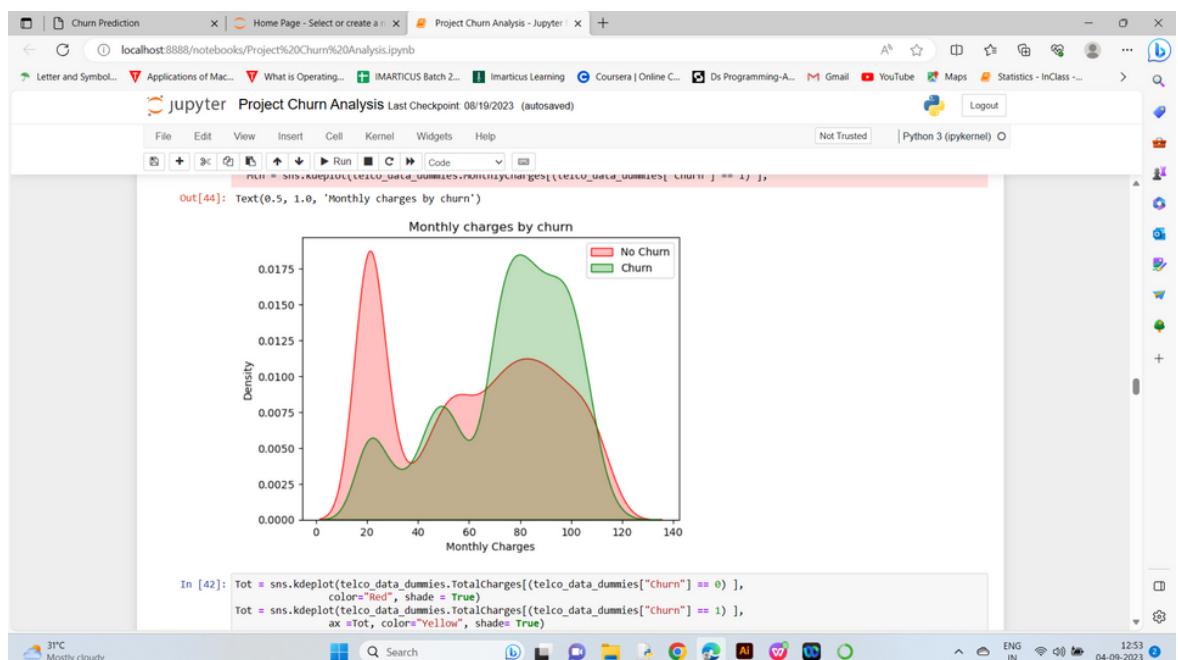
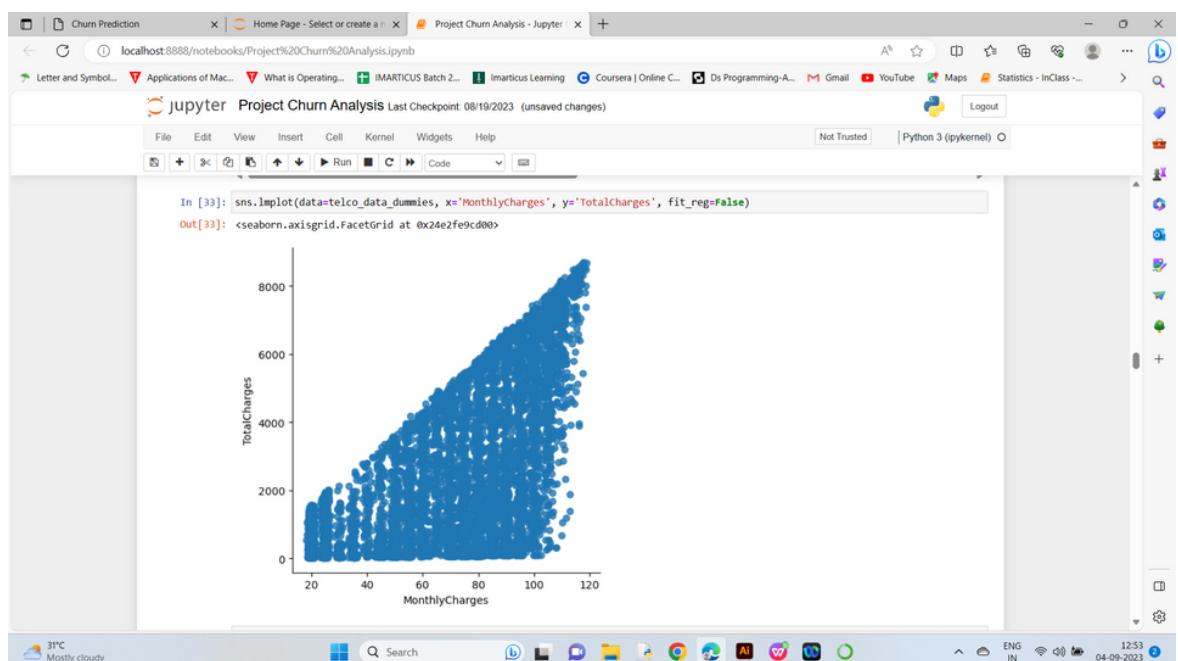
```
In [30]: telco_data['Churn'] = np.where(telco_data.Churn == 'Yes', 1, 0)

In [31]: telco_data.head()
Out[31]:
   gender SeniorCitizen Partner Dependents PhoneService MultipleLines InternetService OnlineSecurity OnlineBackup DeviceProtection TechSupport Stream
0  Female           0     Yes      No       No  No phone service    DSL      No    Yes    No    No
1  Male             0      No     No       Yes      No    DSL      Yes    No    Yes    No
2  Male             0      No     No       Yes      No    DSL      Yes    Yes    No    No
3  Male             0      No     No       No  No phone service    DSL      Yes    No    Yes    Yes
4  Female           0      No     No      Yes      No  Fiber optic    No    No    No    No
```

```
In [32]: telco_data_dummies = pd.get_dummies(telco_data)
telco_data_dummies.head()

Out[32]:
   SeniorCitizen  MonthlyCharges  TotalCharges  Churn  gender_Female  gender_Male  Partner_No  Partner_Yes  Dependents_No  Dependents_Yes ... PaymentMethod
0              0            29.85        29.85      0          1            0          0          1          1          1          0 ...
1              0            56.95       1889.50      0          0            1          1          0          1          0          0 ...
2              0            53.85        108.15      1          0            1          1          0          1          0          0 ...
3              0            42.30       1840.75      0          0            1          1          0          1          0          0 ...
4              0            70.70       151.65      1          1            0          1          0          1          0          0 ...
```

5 rows × 51 columns



A heatmap visualization showing the correlation matrix for the `telco_data_dummies` dataset. The x-axis and y-axis both list 20 categorical features: SeniorCitizen, gender_Female, Dependents_No, MultipleLines_No, InternetService_Fiber optic, OnlineSecurity_Yes, DeviceProtection_No, TechSupport_No, No internet service, StreamingTV_Yes, Contract_Month-to-month, PaperlessBilling_Yes, PaymentMethod_Mailed check, tenure_group_37 - 48, tenure_group_49 - 54, tenure_group_55 - 64, tenure_group_65 - 74, tenure_group_75 - 99, tenure_group_100 - 104, tenure_group_105 - 114, and tenure_group_115 - 124. The color scale ranges from -1.0 (dark red) to 1.0 (dark green). The heatmap shows a strong positive correlation (green) between features like SeniorCitizen, gender_Female, and Dependents_No, as well as between tenure groups. Conversely, there are negative correlations (red) between some tenure groups and other features.

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In [33]: `new_df1['target0']=telco_data.loc[telco_data['Churn']=="0"]
new_df1['target1']=telco_data.loc[telco_data['Churn']=="1"]`

In [46]: `def uniplot(df,col,title,hue=None):

 sns.set_style('whitegrid')
 sns.set_context('talk')
 plt.rcParams['axes.labelsize'] = 20
 plt.rcParams['axes.titlesize'] = 22
 plt.rcParams['axes.titlepad'] = 30

 temp = pd.Series(data = df)
 fig, ax = plt.subplots()
 width = len(df[col].unique()) + 7 + 4*len(temp.unique())
 fig.set_size_inches(width , 8)
 plt.xticks(rotation=45)
 plt.yscale('log')
 plt.title(title)
 ax = sns.countplot(data = df, x= col, order=df[col].value_counts().index,hue = hue,palette='bright')

 plt.show()`

In [55]: `uniplot(new_df1['target1'],col='Partner',title='Distribution of Gender for Churned Customers',hue='gender')
uniplot(new_df1['target0'],col='Partner',title='Distribution of Gender for Non Churned Customers',hue='gender')
uniplot(new_df1['target1'],col='PaymentMethod',title='Distribution of PaymentMethod for Churned Customers',hue='gender')
uniplot(new_df1['target1'],col='Contract',title='Distribution of Contract for Churned Customers',hue='gender')
uniplot(new_df1['target1'],col='TechSupport',title='Distribution of TechSupport for Churned Customers',hue='gender')`

Distribution of Gender for Churned Customers

MODEL BUILDING PART:

```
In [57]: import pandas as pd
from sklearn import metrics
from sklearn.model_selection import train_test_split
from sklearn.metrics import recall_score
from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix
from sklearn.tree import DecisionTreeClassifier
from imblearn.combine import SMOTETENN

In [60]: df=pd.read_csv("C:/Users/DELL/Downloads/tel_churn.csv")
df.head()

Out[60]:
```

	SeniorCitizen	MonthlyCharges	TotalCharges	Churn	gender_Female	gender_Male	Partner_No	Partner_Yes	Dependents_No	Dependents_Yes	PaymentMethod_transfer (auto)
0	0	29.85	29.85	0	1	0	0	0	1	1	...
1	1	0	56.95	1889.50	0	0	1	1	0	1	...
2	2	0	53.85	108.15	1	0	1	1	0	1	...
3	3	0	42.30	1840.75	0	0	1	1	0	1	...
4	4	0	70.70	151.65	1	1	0	1	0	1	...

5 rows × 12 columns

```
In [61]: df=df.drop('Unnamed: 0',axis=1)

In [62]: x=df.drop('Churn',axis=1)
```

```
Out[62]:
```

	SeniorCitizen	MonthlyCharges	TotalCharges	gender_Female	gender_Male	Partner_No	Partner_Yes	Dependents_No	Dependents_Yes	PhoneService_No
0	0	29.85	29.85	1	0	0	0	1	1	0
1	0	56.95	1889.50	0	1	1	0	1	0	0
2	0	53.85	108.15	0	1	1	0	1	0	0
3	0	42.30	1840.75	0	1	1	0	1	0	1
4	0	70.70	151.65	1	0	1	0	1	0	0
...
7027	0	84.80	1990.50	0	1	0	1	0	1	0
7028	0	103.20	7362.90	1	0	0	1	0	1	0
7029	0	29.60	346.45	1	0	0	1	0	1	1
7030	1	74.40	306.60	0	1	0	1	1	0	0
7031	0	105.65	6844.50	0	1	1	0	1	0	0

7032 rows × 10 columns

```
In [63]: y=df['Churn']
y
```

```
Out[63]:
```

	0	0	1	0	1	0	0	1	0	1
0	0	0	1	0	1	0	0	1	0	1
1	0	0	1	0	1	0	0	1	0	0
2	1	0	0	1	1	1	0	0	1	0
3	0	0	1	0	1	1	0	1	0	1
4	1	0	0	1	1	1	0	1	0	0
...
7027	0	0	1	0	1	0	1	0	1	0
7028	0	0	1	0	1	0	1	0	1	0
7029	0	0	1	0	1	0	1	0	1	1
7030	1	0	0	1	1	0	1	1	0	0
7031	0	0	0	1	1	1	0	1	0	0

```
7029 0
7030 1
7031 0
Name: Churn, Length: 7032, dtype: int64
```

```
In [64]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)

In [65]: model_dt=DecisionTreeClassifier(criterion = "gini",random_state = 100,max_depth=6, min_samples_leaf=8)

In [66]: model_dt.fit(x_train,y_train)

Out[66]: DecisionTreeClassifier(max_depth=6, min_samples_leaf=8, random_state=100)
In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
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```

```
In [67]: y_pred=model_dt.predict(x_test)
y_pred
```

```
Out[67]: array([0, 0, 0, ..., 0, 0, 0], dtype=int64)
```

```
In [68]: model_dt.score(x_test,y_test)

Out[68]: 0.7825159914712153
```

```
In [69]: print(classification_report(y_test, y_pred, labels=[0,1]))
```

	precision	recall	f1-score	support
0	0.82	0.91	0.86	1031
1	0.64	0.44	0.52	376
accuracy			0.78	1407
macro avg	0.73	0.67	0.69	1407

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```
In [68]: model_dt.score(x_test,y_test)
Out[68]: 0.7825159914712153

In [69]: print(classification_report(y_test, y_pred, labels=[0,1]))
         precision    recall   f1-score   support
          0       0.82      0.91      0.86     1031
          1       0.64      0.44      0.52      376

      accuracy                           0.78     1407
   macro avg       0.73      0.67      0.69     1407
weighted avg       0.77      0.78      0.77     1407

In [72]: sm = SMOTEENN()
X_resampled, y_resampled = sm.fit_resample(x,y)

In [73]: xr_train,xr_test,yr_train,yr_test=train_test_split(X_resampled, y_resampled,test_size=0.2)

In [74]: model_dt_smote=DecisionTreeClassifier(criterion = "gini",random_state = 100,max_depth=6, min_samples_leaf=8)

In [75]: model_dt_smote.fit(xr_train,yr_train)
yr_predict = model_dt_smote.predict(xr_test)
model_score_r = model_dt_smote.score(xr_test, yr_test)
print(model_score_r)
print(metrics.classification_report(yr_test, yr_predict))

0.9324894514767933
         precision    recall   f1-score   support
          0       0.95      0.90      0.93     553
          1       0.92      0.96      0.94     632

      accuracy                           0.93     1185
   macro avg       0.93      0.93      0.93     1185
weighted avg       0.93      0.93      0.93     1185
```

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```
In [75]: model_dt_smote.fit(xr_train,yr_train)
yr_predict = model_dt_smote.predict(xr_test)
model_score_r = model_dt_smote.score(xr_test, yr_test)
print(model_score_r)
print(metrics.classification_report(yr_test, yr_predict))

0.9324894514767933
         precision    recall   f1-score   support
          0       0.95      0.90      0.93     553
          1       0.92      0.96      0.94     632

      accuracy                           0.93     1185
   macro avg       0.93      0.93      0.93     1185
weighted avg       0.93      0.93      0.93     1185

In [76]: print(metrics.confusion_matrix(yr_test, yr_predict))
[[497 56]
 [24 693]]
```

RANDOM FOREST CLASSIFIER

```
In [77]: from sklearn.ensemble import RandomForestClassifier
In [78]: model_rf=RandomForestClassifier(n_estimators=100, criterion='gini', random_state = 100,max_depth=6, min_samples_leaf=8)
In [79]: model_rf.fit(x_train,y_train)
Out[79]: RandomForestClassifier(max_depth=6, min_samples_leaf=8, random_state=100)
```

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```
In [81]: model_rf.score(x_test,y_test)
Out[81]: 0.789623120113717

In [82]: print(classification_report(y_test, y_pred, labels=[0,1]))
         precision    recall   f1-score   support
          0       0.82      0.91      0.86     1031
          1       0.65      0.45      0.53      376

      accuracy                           0.79     1407
   macro avg       0.74      0.68      0.70     1407
weighted avg       0.78      0.79      0.78     1407

In [84]: sm = SMOTEENN()
X_resampled1, y_resampled1 = sm.fit_resample(x,y)

In [85]: xr_train1,xr_test1,yr_train1,yr_test1=train_test_split(X_resampled1, y_resampled1,test_size=0.2)

In [86]: model_rf_smote=RandomForestClassifier(n_estimators=100, criterion='gini', random_state = 100,max_depth=6, min_samples_leaf=8)
model_rf_smote.fit(xr_train1,yr_train1)

Out[86]: RandomForestClassifier(max_depth=6, min_samples_leaf=8, random_state=100)
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In [87]: yr_predict1 = model_rf_smote.predict(xr_test1)
In [88]: yr_predict1 = model_rf_smote.predict(xr_test1)

In [89]: print(model_rf_smote)
```

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Out[86]: RandomForestClassifier(max_depth=6, min_samples_leaf=8, random_state=100)
In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
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In [87]: yr_predict1 = model_rf_smote.predict(xr_test1)

In [88]: yr_predict1 = model_rf_smote.predict(xr_test1)

In [92]: print(model_rf_smote)
print(metrics.classification_report(yr_test1, yr_predict1))

```
RandomForestClassifier(max_depth=6, min_samples_leaf=8, random_state=100)
precision    recall   f1-score   support
          0       0.95      0.89      0.92      522
          1       0.92      0.96      0.94      652

accuracy                           0.93      1174
macro avg       0.94      0.93      0.93      1174
weighted avg    0.93      0.93      0.93      1174
```

In [93]: print(metrics.confusion_matrix(yr_test1, yr_predict1))

```
[ [467  55]
 [ 24 628]]
```

PCA - PERFORMANCE

In [94]: from sklearn.decomposition import PCA
pca = PCA(0.9)

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