```
In [4]: %pip install numpy
%pip install pandas
%pip install scikit-learn
%pip install matplotlib
%pip install seaborn
```

```
Collecting numpy
  Downloading numpy-2.3.1-cp312-cp312-macosx 14 0 arm64.whl.metadata (62 kB)
Downloading numpy-2.3.1-cp312-cp312-macosx 14 0 arm64.whl (5.1 MB)
                                      _____ 5.1/5.1 MB 5.5 MB/s eta 0:00:00a
0:00:01
Installing collected packages: numpy
Successfully installed numpy-2.3.1
[notice] A new release of pip is available: 24.2 -> 25.1.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.
Collecting pandas
  Downloading pandas-2.3.1-cp312-cp312-macosx 11 0 arm64.whl.metadata (91 k
B)
Requirement already satisfied: numpy>=1.26.0 in /Users/arun/myenv/lib/python
3.12/site-packages (from pandas) (2.3.1)
Requirement already satisfied: python-dateutil>=2.8.2 in /Users/arun/myenv/l
ib/python3.12/site-packages (from pandas) (2.9.0.post0)
Collecting pytz>=2020.1 (from pandas)
  Downloading pytz-2025.2-py2.py3-none-any.whl.metadata (22 kB)
Collecting tzdata>=2022.7 (from pandas)
  Downloading tzdata-2025.2-py2.py3-none-any.whl.metadata (1.4 kB)
Requirement already satisfied: six>=1.5 in /Users/arun/myenv/lib/python3.12/
site-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
Downloading pandas-2.3.1-cp312-cp312-macosx 11 0 arm64.whl (10.7 MB)
                                      ----- 10.7/10.7 MB 5.3 MB/s eta 0:00:0
0 0:00:01
Downloading pytz-2025.2-py2.py3-none-any.whl (509 kB)
Downloading tzdata-2025.2-py2.py3-none-any.whl (347 kB)
Installing collected packages: pytz, tzdata, pandas
Successfully installed pandas-2.3.1 pytz-2025.2 tzdata-2025.2
[notice] A new release of pip is available: 24.2 -> 25.1.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.
Collecting scikit-learn
  Using cached scikit learn-1.7.1-cp312-cp312-macosx 12 0 arm64.whl.metadata
(11 kB)
Requirement already satisfied: numpy>=1.22.0 in /Users/arun/myenv/lib/python
3.12/site-packages (from scikit-learn) (2.3.1)
Collecting scipy>=1.8.0 (from scikit-learn)
  Using cached scipy-1.16.0-cp312-cp312-macosx_14_0_arm64.whl.metadata (61 k
B)
Collecting joblib>=1.2.0 (from scikit-learn)
  Using cached joblib-1.5.1-py3-none-any.whl.metadata (5.6 kB)
Collecting threadpoolctl>=3.1.0 (from scikit-learn)
  Using cached threadpoolctl-3.6.0-py3-none-any.whl.metadata (13 kB)
Using cached scikit learn-1.7.1-cp312-cp312-macosx 12 0 arm64.whl (8.6 MB)
Using cached joblib-1.5.1-py3-none-any.whl (307 kB)
Using cached scipy-1.16.0-cp312-cp312-macosx 14 0 arm64.whl (20.8 MB)
Using cached threadpoolctl-3.6.0-py3-none-any.whl (18 kB)
Installing collected packages: threadpoolctl, scipy, joblib, scikit-learn
Successfully installed joblib-1.5.1 scikit-learn-1.7.1 scipy-1.16.0 threadpo
olctl-3.6.0
```

[notice] A new release of pip is available: 24.2 -> 25.1.1

```
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.
Collecting matplotlib
  Downloading matplotlib-3.10.3-cp312-cp312-macosx 11 0 arm64.whl.metadata
(11 kB)
Collecting contourpy>=1.0.1 (from matplotlib)
  Downloading contourpy-1.3.2-cp312-cp312-macosx 11 0 arm64.whl.metadata (5.
5 kB)
Collecting cycler>=0.10 (from matplotlib)
  Using cached cycler-0.12.1-py3-none-any.whl.metadata (3.8 kB)
Collecting fonttools>=4.22.0 (from matplotlib)
  Downloading fonttools-4.59.0-cp312-cp312-macosx 10 13 universal2.whl.metad
ata (107 kB)
Collecting kiwisolver>=1.3.1 (from matplotlib)
  Using cached kiwisolver-1.4.8-cp312-cp312-macosx 11 0 arm64.whl.metadata
(6.2 kB)
Requirement already satisfied: numpy>=1.23 in /Users/arun/myenv/lib/python3.
12/site-packages (from matplotlib) (2.3.1)
Requirement already satisfied: packaging>=20.0 in /Users/arun/myenv/lib/pyth
on3.12/site-packages (from matplotlib) (25.0)
Collecting pillow>=8 (from matplotlib)
  Downloading pillow-11.3.0-cp312-cp312-macosx 11 0 arm64.whl.metadata (9.0
kB)
Collecting pyparsing>=2.3.1 (from matplotlib)
  Downloading pyparsing-3.2.3-py3-none-any.whl.metadata (5.0 kB)
Requirement already satisfied: python-dateutil>=2.7 in /Users/arun/myenv/li
b/python3.12/site-packages (from matplotlib) (2.9.0.post0)
Requirement already satisfied: six>=1.5 in /Users/arun/myenv/lib/python3.12/
site-packages (from python-dateutil>=2.7->matplotlib) (1.17.0)
Downloading matplotlib-3.10.3-cp312-cp312-macosx 11 0 arm64.whl (8.1 MB)
                                  8.1/8.1 MB 7.6 MB/s eta 0:00:00a
0:00:01
Downloading contourpy-1.3.2-cp312-cp312-macosx 11 0 arm64.whl (255 kB)
Using cached cycler-0.12.1-py3-none-any.whl (8.3 kB)
Downloading fonttools-4.59.0-cp312-cp312-macosx 10 13 universal2.whl (2.8 M
B)
                                  2.8/2.8 MB 8.4 MB/s eta 0:00:00a
0:00:01
Using cached kiwisolver-1.4.8-cp312-cp312-macosx 11 0 arm64.whl (65 kB)
Downloading pillow-11.3.0-cp312-cp312-macosx_11_0_arm64.whl (4.7 MB)
                                   4.7/4.7 MB 7.7 MB/s eta 0:00:00a
0:00:01
Downloading pyparsing-3.2.3-py3-none-any.whl (111 kB)
Installing collected packages: pyparsing, pillow, kiwisolver, fonttools, cyc
ler, contourpy, matplotlib
Successfully installed contourpy-1.3.2 cycler-0.12.1 fonttools-4.59.0 kiwiso
lver-1.4.8 matplotlib-3.10.3 pillow-11.3.0 pyparsing-3.2.3
[notice] A new release of pip is available: 24.2 -> 25.1.1
[notice] To update, run: pip install --upgrade pip
Note: you may need to restart the kernel to use updated packages.
Collecting seaborn
  Using cached seaborn-0.13.2-py3-none-any.whl.metadata (5.4 kB)
Requirement already satisfied: numpy!=1.24.0,>=1.20 in /Users/arun/myenv/li
b/python3.12/site-packages (from seaborn) (2.3.1)
Requirement already satisfied: pandas>=1.2 in /Users/arun/myenv/lib/python3.
```

```
12/site-packages (from seaborn) (2.3.1)
       Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in /Users/arun/myenv/
       lib/python3.12/site-packages (from seaborn) (3.10.3)
      Requirement already satisfied: contourpy>=1.0.1 in /Users/arun/myenv/lib/pyt
      hon3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.3.2)
      Requirement already satisfied: cycler>=0.10 in /Users/arun/myenv/lib/python
       3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (0.12.1)
      Requirement already satisfied: fonttools>=4.22.0 in /Users/arun/myenv/lib/py
       thon3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (4.59.0)
      Requirement already satisfied: kiwisolver>=1.3.1 in /Users/arun/myenv/lib/py
       thon3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (1.4.8)
      Requirement already satisfied: packaging>=20.0 in /Users/arun/myenv/lib/pyth
       on3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (25.0)
      Requirement already satisfied: pillow>=8 in /Users/arun/myenv/lib/python3.1
      2/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (11.3.0)
      Requirement already satisfied: pyparsing>=2.3.1 in /Users/arun/myenv/lib/pyt
       hon3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (3.2.3)
      Requirement already satisfied: python-dateutil>=2.7 in /Users/arun/myenv/li
       b/python3.12/site-packages (from matplotlib!=3.6.1,>=3.4->seaborn) (2.9.0.po
       st0)
      Requirement already satisfied: pytz>=2020.1 in /Users/arun/myenv/lib/python
       3.12/site-packages (from pandas>=1.2->seaborn) (2025.2)
      Requirement already satisfied: tzdata>=2022.7 in /Users/arun/myenv/lib/pytho
       n3.12/site-packages (from pandas>=1.2->seaborn) (2025.2)
      Requirement already satisfied: six>=1.5 in /Users/arun/myenv/lib/python3.12/
       site-packages (from python-dateutil>=2.7->matplotlib!=3.6.1,>=3.4->seaborn)
       (1.17.0)
      Using cached seaborn-0.13.2-py3-none-any.whl (294 kB)
       Installing collected packages: seaborn
       Successfully installed seaborn-0.13.2
       [notice] A new release of pip is available: 24.2 -> 25.1.1
       [notice] To update, run: pip install --upgrade pip
      Note: you may need to restart the kernel to use updated packages.
In [5]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        sns.set_theme(style='whitegrid')
In [7]: # Loading Data
        df = pd.read_csv("Databel - Data.csv")
        df.head(3)
```

			_		_	
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				/	-	_

		Customer ID	Churn Label	Account Length (in months)			Intl Calls		Intl Active	Intl Plan	Extra International Charges	
	0	4444- BZPU	No	1	3	8.0	0.0	0.0	No	no	0.0	
	1	5676- PTZX	No	33	179	431.3	0.0	0.0	No	no	0.0	
	2	8532 - ZEKQ	No	44	82	217.6	0.0	0.0	No	yes	0.0	

3 rows × 29 columns

Data Check

In [8]: df.info()

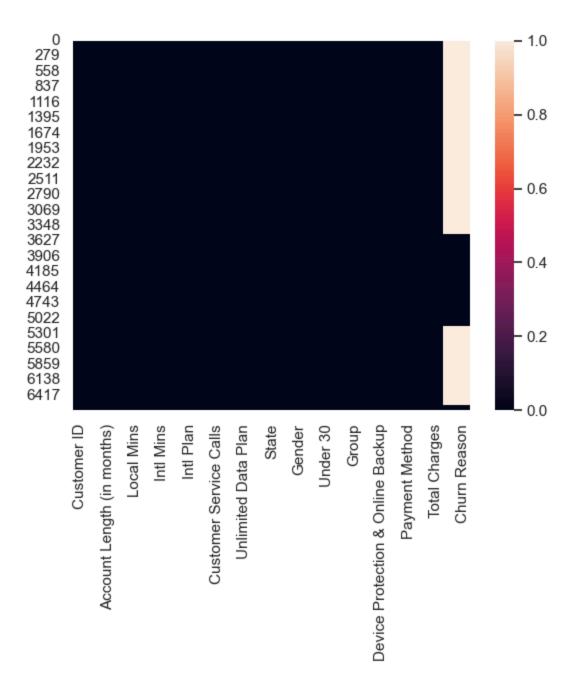
<class 'pandas.core.frame.DataFrame'> RangeIndex: 6687 entries, 0 to 6686 Data columns (total 29 columns):

#	Column	Non-l	Null Count	Dtype						
0	Customer ID	6687	non-null	object						
1	Churn Label	6687	non-null	object						
2	Account Length (in months)	6687	non-null	int64						
3	Local Calls	6687	non-null	int64						
4	Local Mins	6687	non-null	float64						
5	Intl Calls	6687	non-null	float64						
6	Intl Mins	6687	non-null	float64						
7	Intl Active	6687	non-null	object						
8	Intl Plan	6687	non-null	object						
9	Extra International Charges	6687	non-null	float64						
10	Customer Service Calls	6687	non-null	int64						
11	Avg Monthly GB Download	6687	non-null	int64						
12	Unlimited Data Plan	6687	non-null	object						
13	Extra Data Charges	6687	non-null	int64						
14	State	6687	non-null	object						
15	Phone Number	6687	non-null	object						
16	Gender	6687	non-null	object						
17	Age	6687	non-null	int64						
18	Under 30	6687	non-null	object						
19	Senior	6687	non-null	object						
20	Group	6687	non-null	object						
21	Number of Customers in Group	6687	non-null	int64						
22	Device Protection & Online Backup	6687	non-null	object						
23	Contract Type	6687	non-null	object						
24	Payment Method	6687	non-null	object						
25	Monthly Charge	6687	non-null	int64						
26	Total Charges	6687	non-null	int64						
27	Churn Category	1769	non-null	object						
28	Churn Reason		non-null	object						
dtyp	es: float64(4), int64(9), object(16)								
memo	memory usage: 1.5+ MB									

memory usage: 1.5+ MB

```
In [9]: # Heatmap for check Null Values
       sns.heatmap(df.isnull())
        print("No Null Values in DataSet")
```

No Null Values in DataSet



There is only null values in Churn Reason

```
In [10]: import warnings
warnings.filterwarnings('ignore')
print(f"Columns : {df.shape[0]} \nRows : {df.shape[1]}")

Columns : 6687
Rows : 29
In [11]: df.describe()
```

0	[11]	-
	1 1 1 1	

	Account Length (in months)	Local Calls	Local Mins	Intl Calls	Intl Mins	Ex Internatio Charç
count	6687.000000	6687.000000	6687.000000	6687.000000	6687.000000	6687.0000
mean	32.337820	130.974129	322.752864	51.097524	130.070624	33.6417
std	24.595689	121.893966	288.619931	103.592369	243.527828	76.3468
min	1.000000	1.000000	4.000000	0.000000	0.000000	0.0000
25%	9.000000	31.000000	76.900000	0.000000	0.000000	0.0000
50%	29.000000	98.000000	250.500000	0.000000	0.000000	0.0000
75%	55.000000	199.000000	498.050000	52.000000	140.400000	16.4000
max	77.000000	918.000000	1234.200000	1120.000000	1372.500000	585.8000

In [12]: df.describe(include="object").transpose()

			_			_	
\cap		ı±	П	1	\neg	1	-
	1		- 1	- 1	_/	-	-

	count	unique	top	freq
Customer ID	6687	6687	4444-BZPU	1
Churn Label	6687	2	No	4891
Intl Active	6687	2	No	4116
Intl Plan	6687	2	no	6036
Unlimited Data Plan	6687	2	Yes	4494
State	6687	51	WV	213
Phone Number	6687	6677	359-9794	2
Gender	6687	3	Male	3379
Under 30	6687	2	No	5400
Senior	6687	2	No	5460
Group	6687	2	No	5166
Device Protection & Online Backup	6687	2	No	4393
Contract Type	6687	3	Month-to-Month	3411
Payment Method	6687	3	Direct Debit	3702
Churn Category	1769	5	Competitor	805
Churn Reason	1769	20	Competitor made better offer	303

Out[13]:		Customer ID	Churn Label	Account Length (in months)				Intl Mins	Intl Active	Intl Plan	Extra International Charges
	0	4444- BZPU	No	1	3	8.0	0.0	0.0	No	no	0.0
	1	5676 - PTZX	No	33	179	431.3	0.0	0.0	No	no	0.0
	2	8532 - ZEKQ	No	44	82	217.6	0.0	0.0	No	yes	0.0

3 rows × 29 columns

EDA

```
In [14]: | print(f"Total Number of Customers: {df['Customer ID'].nunique()}")
         print(f"Total NUmber of Churned Customers: {df['Churn Label'].value_counts()
         print(f"Rate of Churned Customers : {(df['Churn Label'].value_counts()[1]/df
        Total Number of Customers: 6687
        Total NUmber of Churned Customers: 1796
        Rate of Churned Customers: 26.86 %
In [16]: reason=df[df['Churn Label'] == 'Yes'][['Churn Label','Churn Reason','Churn (
In [17]: #Find Churn Reason
         churn_reason=reason['Churn Reason'].value_counts().reset_index()
         churn reason.head(3)
Out[17]:
                         Churn Reason count
             Competitor made better offer
                                        303
          1 Competitor had better devices
                                        297
          2
                                        203
                Attitude of support person
In [25]: ## Create a barplot which Will Show the Churn Reason
         sns.barplot(x='count',y='Churn Reason',data=churn_reason,palette='Blues_r')
         plt.title('Count of Churn Reason',fontweight='bold',fontsize=15)
         plt.ylabel('Churn Reason', fontweight='bold', fontsize=15)
         plt.xlabel('Count', fontweight='bold', fontsize=15)
         plt.show()
```

Count of Churn Reason Competitor made better offer Competitor had better devices Attitude of support person Don't know Competitor offered more data Competitor offered higher download speeds Attitude of service provider Churn Reason Price too high Product dissatisfaction Network reliability Long distance charges Service dissatisfaction Moved Extra data charges Limited range of services Poor expertise of online support Lack of affordable download/upload speed Lack of self-service on Website Poor expertise of phone support

50

100

150

Count

200

250

300

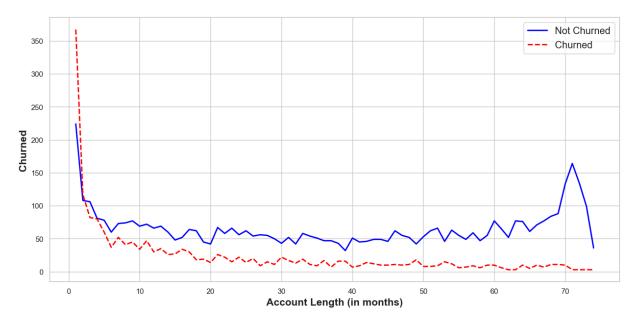
In [19]: churned=df[df['Churn Label'] == 'Yes'].groupby('Account Length (in months)')
 retained=df[df['Churn Label'] == 'No'].groupby('Account Length (in months)')
 merged_df = pd.merge(churned, retained, on='Account Length (in months)', how
 merged_df.head()

Deceased

0

Out[19]:		Account Length (in months)	Churned	Not Churned
	0	1	367	224
	1	2	117	108
	2	3	82	106
	3	4	80	81
	4	5	60	78

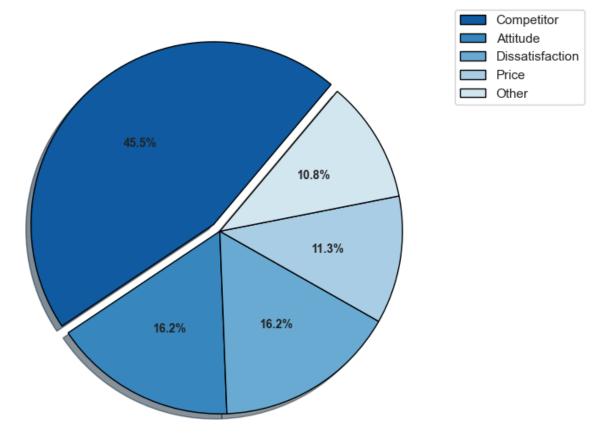
```
In [20]: plt.figure(figsize=(15,7))
    sns.lineplot(x='Account Length (in months)',y='Not Churned',data=merged_df,l
    sns.lineplot(x='Account Length (in months)',y='Churned',data=merged_df,label
    plt.ylabel('Churned',fontweight='bold',fontsize=15)
    plt.xlabel('Account Length (in months)',fontweight='bold',fontsize=15)
    plt.legend(fontsize='large')
    plt.show()
```



```
In [ ]: reason['Churn Category'].value_counts()
    labels=reason['Churn Category'].value_counts().index
```

```
In [27]: import matplotlib.pyplot as plt
         # Example data
         \exp = [0.05, 0, 0, 0, 0]
         col = ['#FF9999','#66B3FF','#99FF99','#FFCC99','#FFD700']
         # Prepare data for Values and label
         values = reason['Churn Category'].value counts()
         labels = values.index
         exp = [0.05] + [0] * (len(values) - 1) # explode only the first slice
         # Generate professional blue palette
         col = sns.color_palette("Blues_r", len(values))
         # Plot pie chart
         plt.figure(figsize=(7, 7))
         plt.pie(x=values.values,
                 explode=exp,
                 autopct="%1.1f%",
                 shadow=True,
                 radius=1,
                 colors=col,
                 textprops={"fontsize": 10, "fontweight": 'bold'},
                 wedgeprops={"linewidth": 1, "edgecolor": 'black'},
                 rotatelabels=False,
                 startangle=50)
         plt.legend(labels, bbox_to_anchor=(1, 1))
         plt.title("Churn Categories", fontsize=15, fontweight='bold')
         plt.show()
```

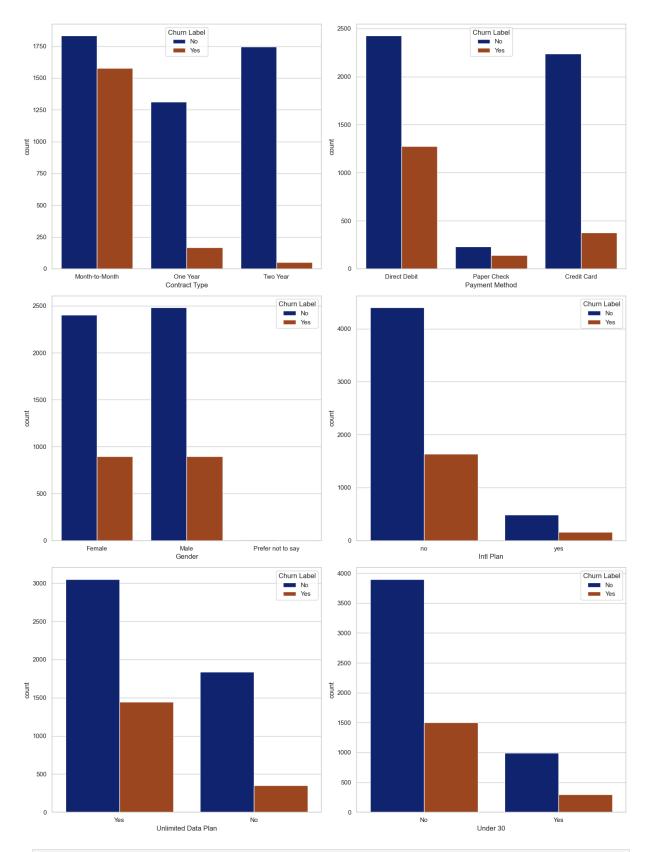
Churn Categories



```
In [48]: # Set a dark professional palette
with sns.color_palette("dark"):
    # Create a 3x2 grid of subplots
    fig, axes = plt.subplots(nrows=3, ncols=2, figsize=(15, 20))

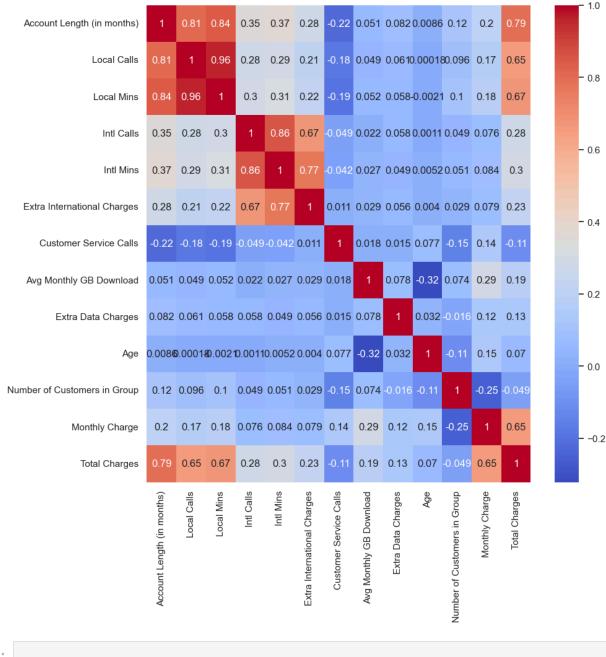
# Plot 6 categorical features against Churn Label
sns.countplot(x="Contract Type", data=df, hue="Churn Label", ax=axes[0, sns.countplot(x="Payment Method", data=df, hue="Churn Label", ax=axes[1, 0])
sns.countplot(x="Gender", data=df, hue="Churn Label", ax=axes[1, 0])
sns.countplot(x="Intl Plan", data=df, hue="Churn Label", ax=axes[1, 1])
sns.countplot(x="Unlimited Data Plan", data=df, hue="Churn Label", ax=axes[2, 1])

# Set overall spacing
plt.tight_layout()
plt.show()
```



```
In [30]: plt.figure(figsize=(10,10))
    sns.heatmap(cor,annot=True,cmap='coolwarm')
```

Out[30]: <Axes: >



```
In []:
In [31]: values=[df['Customer ID'].nunique(),df['Churn Label'].value_counts()[1],(df]
In [32]: metric=["Total Number of Customers","Total NUmber of Churned Customers","Rat
In [33]: summary=pd.DataFrame({'metric':metric,'Values':values},index=range(1,4))
    summary
```

```
Out[33]:
                                                 Values
                                        metric
           1
                      Total Number of Customers
                                                6687.00
              Total NUmber of Churned Customers
                                                1796.00
          3
                      Rate of Churned Customers
                                                   26.86
In [34]: df.groupby('Contract Type')['Avg Monthly GB Download'].mean().reset_index(na
Out[34]:
               Contract Type Avg Monthly GB Download
              Month-to-Month
                                               7.154500
           1
                     One Year
                                               6.934415
           2
                     Two Year
                                              5.630495
In [35]: with sns.color palette("dark"):
               plt.figure(figsize=(11,7))
               sns.lmplot(y='Total Charges',x='Account Length (in months)',data=df,hue=
                       col='Contract Type')
         <Figure size 1100x700 with 0 Axes>
                                              Contract Type = One Ye
          5000
          4000
          2000
          -1000
                    Account Length (in months)
                                                                        Account Length (in months)
In [36]: print("Majorit of Our Customers")
          df['State'].value_counts().head().reset_index(name='No. of Customers')
         Majorit of Our Customers
Out[36]:
              State No. of Customers
                WV
                                  213
           1
                MN
                                  168
           2
                NY
                                  167
           3
                AL
                                  161
```

158

4

ОН

Out[37]:		State	Customer Churned
	0	WV	57
	1	ОН	55
	2	OR	48
	3	MD	46
	4	AL	46

- Majority of our customers from West Virginia US state(WV)
- Most of the customers is also churned from WV

Logistic Regression

:		Churn Label	Account Length (in months)					Intl Active	Intl Plan	Extra International Charges	Customer Service Calls
	0	No	1	3	8.0	0.0	0.0	No	no	0.0	0
	1	No	33	179	431.3	0.0	0.0	No	no	0.0	0
	2	No	44	82	217.6	0.0	0.0	No	yes	0.0	0

3 rows × 26 columns

```
In [49]: features=column_selection.drop('Churn Label',axis=1)
   X=pd.get_dummies(features,drop_first=True)
   y=df['Churn Label']
```

```
In [50]: X_train, X_test, y_train, y_test = train_test_split(
                X, y, test_size=0.5, random_state=42)
In [42]: logmodel=linear_model.LogisticRegression()
In [43]: logmodel.fit(X_train,y_train)
Out[43]:
         ▶ Parameters
In [51]: predictions=logmodel.predict(X_test)
In [45]: print(classification_report(y_test,predictions))
                                  recall f1-score
                     precision
                                                    support
                 No
                          0.89
                                   0.93
                                             0.91
                                                       2447
                          0.79
                                             0.73
                Yes
                                    0.68
                                                        897
                                             0.87
                                                       3344
           accuracy
          macro avg
                          0.84
                                   0.81
                                             0.82
                                                       3344
                          0.86
                                   0.87
                                             0.86
                                                       3344
       weighted avg
In [52]: print(confusion_matrix(y_test,predictions))
        [[2282 165]
         [ 283 614]]
In [47]: print(accuracy_score(y_test,predictions))
       0.8660287081339713
```

ACCURACY 86.60 %

In []: