→ Preprocessing Training Data

import numpy as np
import pandas as pd
import seaborn as sns
df=pd.read_csv('/content/churn-bigml-80.csv')
df

	State	Account length		International plan	Voice mail plan	Number vmail messages	day	Total day calls	Total day charge	Total eve minutes	Total eve calls	To1 € char
0	KS	128	415	No	Yes	25	265.1	110	45.07	197.4	99	16
1	ОН	107	415	No	Yes	26	161.6	123	27.47	195.5	103	16
2	NJ	137	415	No	No	0	243.4	114	41.38	121.2	110	10
3	ОН	84	408	Yes	No	0	299.4	71	50.90	61.9	88	5
4	OK	75	415	Yes	No	0	166.7	113	28.34	148.3	122	12
2661	SC	79	415	No	No	0	134.7	98	22.90	189.7	68	16
2662	AZ	192	415	No	Yes	36	156.2	77	26.55	215.5	126	18
2663	WV	68	415	No	No	0	231.1	57	39.29	153.4	55	13
2664	RI	28	510	No	No	0	180.8	109	30.74	288.8	58	24
2665	TN	74	415	No	Yes	25	234.4	113	39.85	265.9	82	22

2666 rows × 20 columns



#First 5 observation print
df.head()

	State	Account length		International plan	Voice mail plan	Number vmail messages	day	Total day calls	Total day charge	Total eve minutes	eve	Total eve charge
0	KS	128	415	No	Yes	25	265.1	110	45.07	197.4	99	16.78
1	ОН	107	415	No	Yes	26	161.6	123	27.47	195.5	103	16.62
2	NJ	137	415	No	No	0	243.4	114	41.38	121.2	110	10.30
3	ОН	84	408	Yes	No	0	299.4	71	50.90	61.9	88	5.26
4	ОК	75	415	Yes	No	0	166.7	113	28.34	148.3	122	12.61



#Last 5 observation print
df.tail()

	State	Account length		International plan	Voice mail plan	Number vmail messages	day	Total day calls	Total day charge	Total eve minutes	Total eve calls	To1 • char
2661	SC	79	415	No	No	0	134.7	98	22.90	189.7	68	16
2662	AZ	192	415	No	Yes	36	156.2	77	26.55	215.5	126	18
2663	WV	68	415	No	No	0	231.1	57	39.29	153.4	55	13
2664	RI	28	510	No	No	0	180.8	109	30.74	288.8	58	24
2665	TN	74	415	No	Yes	25	234.4	113	39.85	265.9	82	22



UT

CO

СТ

ΜI

VT

ID

NC

ΤX

60

59

59

58

57

56

56

55

```
3/10/23, 11:31 AM
    #Column heading print
    df.columns
         'Total day calls', 'Total day charge', 'Total eve minutes',
'Total eve calls', 'Total eve charge', 'Total night minutes',
'Total night calls', 'Total night charge', 'Total intl minutes',
'Total intl calls', 'Total intl charge', 'Customer service calls',
                dtype='object')
    #Each column types
    df.dtypes
          State
                                       object
          Account length
                                        int64
          Area code
                                        int64
          International plan
                                       object
          Voice mail plan
                                       object
          Number vmail messages
                                        int64
          Total day minutes
                                      float64
          Total day calls
                                       int64
          Total day charge
                                      float64
                                      float64
          Total eve minutes
                                       int64
          Total eve calls
                                      float64
          Total eve charge
          Total night minutes
                                     float64
          Total night calls
                                       int64
          Total night charge
                                      float64
          Total intl minutes
                                      float64
          Total intl calls
                                       int64
          Total intl charge
                                      float64
          Customer service calls
                                       int64
          Churn
                                         bool
          dtype: object
    #To find Missing values
    df.isna().sum()
          State
          Account length
                                      0
          Area code
                                      0
          International plan
                                      0
          Voice mail plan
                                      0
          Number vmail messages
                                      0
          Total day minutes
                                      0
          Total day calls
          Total day charge
                                      0
          Total eve minutes
          Total eve calls
          Total eve charge
          Total night minutes
                                      0
          Total night calls
                                      0
          Total night charge
                                      0
          Total intl minutes
                                      a
          Total intl calls
                                      0
          Total intl charge
                                      0
          Customer service calls
                                      0
          Churn
                                      0
          dtype: int64
    #Each state counts
    df['State'].value_counts()
          WV
                88
          MN
                70
          NY
                68
          VA
                67
          ΑL
                66
          ОН
                66
                66
          OR
                62
          NV
                61
          WI
                61
          MD
                60
```

https://colab.research.google.com/drive/1x nHGmvafcm9PvCUl3ZCMdxrSteHAclK#scrollTo=h7cg0k5vGaik&printMode=true

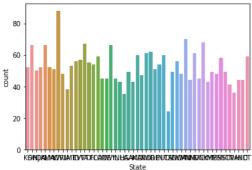
```
MT
      53
OK
      52
MA
      52
KS
      52
MO
      51
DE
      51
NJ
      50
SC
SD
      49
      49
ME
      49
GA
      49
RΙ
      48
MS
WA
      48
AR
      47
ΙL
      45
DC
      45
ΑZ
      45
NE
      45
ΗI
      44
NM
      44
ND
      44
      43
ΚY
      43
NH
      43
TN
      41
IΑ
      38
PA
      36
ΙΔ
      35
CA
      24
```

Name: State, dtype: int64

#Each state count Graph
sns.countplot('State',data=df)

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

<AxesSubplot:xlabel='State', ylabel='count'>



#Each International plan count
df['International plan'].value_counts()

No 2396 Yes 270

Name: International plan, dtype: int64

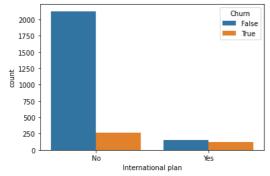
#Each International plan count graph
sns.countplot('International plan',data=df)

 $/usr/local/lib/python 3.9/dist-packages/seaborn/_decorators.py: 36: \ Future Warning: Pass the following variable of the property of the pro$

#How Each International plan affect Churn graph
sns.countplot('International plan',data=df,hue='Churn')

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

<AxesSubplot:xlabel='International plan', ylabel='count'>



#Each voice plan count
df['Voice mail plan'].value_counts()

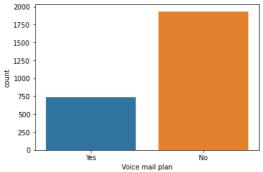
No 1933 Yes 733

Name: Voice mail plan, dtype: int64

#Each voice plan count graph
sns.countplot('Voice mail plan',data=df)

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

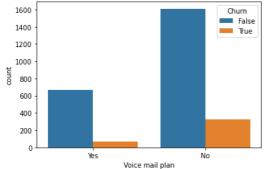
<AxesSubplot:xlabel='Voice mail plan', ylabel='count'>



#How Each Voice mail plan affect Churn graph
sns.countplot('Voice mail plan',data=df,hue='Churn')

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

<AxesSubplot:xlabel='Voice mail plan', ylabel='count'>

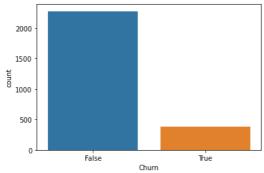


#Each Churn count
df['Churn'].value_counts()

False 2278 True 388 Name: Churn, dtype: int64

#Each Churn counts graph sns.countplot('Churn',data=df)

> /usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(
> <AxesSubplot:xlabel='Churn', ylabel='count'>

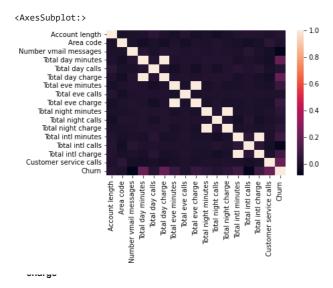


Correlation

df.corr()



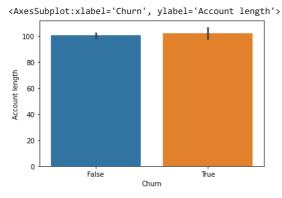
#HeatMap-correlation showing
sns.heatmap(df.corr())



BARPLOT

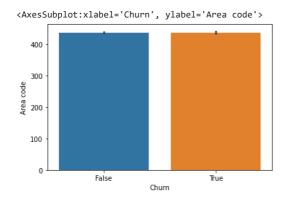
minutes

#Barplot Graph Of churn & Account length
sns.barplot(x='Churn',y='Account length',data=df)



OL:---- 0.047700 0.004040 0.006474 0.405600 0.040000 0.405600 0.070000 0.004500 0.070000

#Barplot Graph Of churn & area code
sns.barplot(x='Churn',y='Area code',data=df)



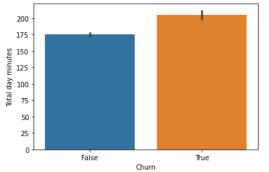
#Barplot Graph Of churn & Number vmail messages
sns.barplot(x='Churn',y='Number vmail messages',data=df)

<AxesSubplot:xlabel='Churn', ylabel='Number vmail messages'>



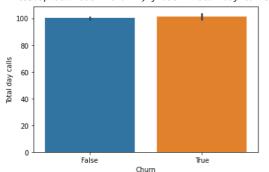
#Barplot Graph Of churn & Total day minutes
sns.barplot(x='Churn',y='Total day minutes',data=df)

<AxesSubplot:xlabel='Churn', ylabel='Total day minutes'>



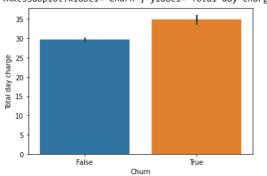
#Barplot Graph Of churn & Total day calls
sns.barplot(x='Churn',y='Total day calls',data=df)

<AxesSubplot:xlabel='Churn', ylabel='Total day calls'>



#Barplot Graph Of churn & Total day charge
sns.barplot(x='Churn',y='Total day charge',data=df)

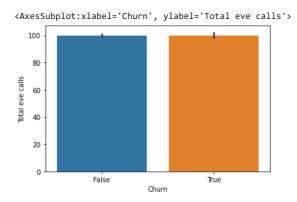
<AxesSubplot:xlabel='Churn', ylabel='Total day charge'>



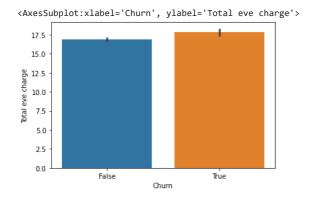
#barplot Graph Of Churn & Total eve minutes
sns.barplot(x='Churn',y='Total eve minutes',data=df)

<AxesSubplot:xlabel='Churn', ylabel='Total eve minutes'>

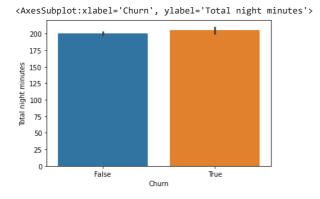
#barplot Graph Of Churn & Total eve calls
sns.barplot(x='Churn',y='Total eve calls',data=df)



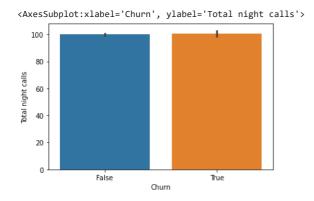
#barplot Graph Of Churn & Total eve charge
sns.barplot(x='Churn',y='Total eve charge',data=df)



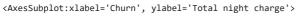
#Barplot Graph Of churn & Total night minutes
sns.barplot(x='Churn',y='Total night minutes',data=df)

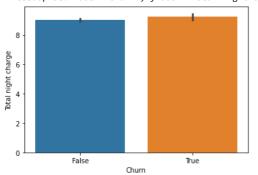


#Barplot Graph Of churn & Total night calls
sns.barplot(x='Churn',y='Total night calls',data=df)



#barplot Graph Of Churn & Total night charge
sns.barplot(x='Churn',y='Total night charge',data=df)





#barplot Graph Of Churn & Total intl minutes sns.barplot(x='Churn',y='Total intl minutes',data=df)

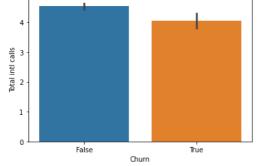
<AxesSubplot:xlabel='Churn', ylabel='Total intl minutes'> 10 Total intl minutes 6 4 2

Churn

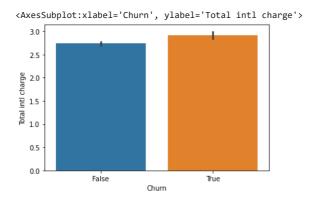
#barplot Graph Of churn & Total intl calls sns.barplot(x='Churn',y='Total intl calls',data=df)

False

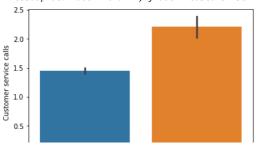
<AxesSubplot:xlabel='Churn', ylabel='Total intl calls'>



#barplot Graph Of churn & Total intl charge sns.barplot(x='Churn',y='Total intl charge',data=df)



#barplot Graph Of churn & Customer service calls sns.barplot(x='Churn',y='Customer service calls',data=df) <AxesSubplot:xlabel='Churn', ylabel='Customer service calls'>



Encoding string to numeric using getdummies

num

df1=pd.get_dummies(df[['State','International plan','Voice mail plan']],drop_first=True)
df1

	State_AL	State_AK	State_AZ	State_CA	State_CO	State_CT	State_DC	State_DE	State_FL	State_G
0	0	0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	0	0	1
2	0	0	0	0	0	0	0	0	0	1
3	0	0	0	0	0	0	0	0	0	1
4	0	0	0	0	0	0	0	0	0	1
2661	0	0	0	0	0	0	0	0	0	1
2662	0	0	1	0	0	0	0	0	0	1
2663	0	0	0	0	0	0	0	0	0	1
2664	0	0	0	0	0	0	0	0	0	1
2665	0	0	0	0	0	0	0	0	0	1

2666 rows × 52 columns



concatination-combining

dfe=pd.concat([df,df1],axis=1)
dfe

	State	Account length		International plan	Voice mail plan	Number vmail messages	day	Total day calls	Total day charge	Total eve minutes	•••	State_
0	KS	128	415	No	Yes	25	265.1	110	45.07	197.4		
1	ОН	107	415	No	Yes	26	161.6	123	27.47	195.5		
2	NJ	137	415	No	No	0	243.4	114	41.38	121.2		
3	ОН	84	408	Yes	No	0	299.4	71	50.90	61.9		
4	OK	75	415	Yes	No	0	166.7	113	28.34	148.3		
2661	SC	79	415	No	No	0	134.7	98	22.90	189.7		
2662	AZ	192	415	No	Yes	36	156.2	77	26.55	215.5		
2663	WV	68	415	No	No	0	231.1	57	39.29	153.4		
2664	RI	28	510	No	No	0	180.8	109	30.74	288.8		
2665	TN	74	415	No	Yes	25	234.4	113	39.85	265.9		

2666 rows × 72 columns



#Dropping unwanted columns
dfe.drop(['State','International plan','Voice mail plan'],axis=1,inplace=True)
dfe

	Account length		Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	Total eve calls	Total eve charge	Total night minutes	 State_TX	S
0	128	415	25	265.1	110	45.07	197.4	99	16.78	244.7	 0	
1	107	415	26	161.6	123	27.47	195.5	103	16.62	254.4	 0	
2	137	415	0	243.4	114	41.38	121.2	110	10.30	162.6	 0	
3	84	408	0	299.4	71	50.90	61.9	88	5.26	196.9	 0	
4	75	415	0	166.7	113	28.34	148.3	122	12.61	186.9	 0	
2661	79	415	0	134.7	98	22.90	189.7	68	16.12	221.4	 0	
2662	192	415	36	156.2	77	26.55	215.5	126	18.32	279.1	 0	
2663	68	415	0	231.1	57	39.29	153.4	55	13.04	191.3	 0	
2664	28	510	0	180.8	109	30.74	288.8	58	24.55	191.9	 0	
2665	74	415	25	234.4	113	39.85	265.9	82	22.60	241.4	 0	

2666 rows × 69 columns



 $\mbox{\tt\#TO}$ find types of columns after encoding,concation,dropping $\mbox{\tt dfe.dtypes}$

Account length int64 Area code int64 Number vmail messages int64 Total day minutes float64 Total day calls int64 uint8 State_WI State_WV uint8 State_WY uint8 International plan_Yes uint8 Voice mail plan_Yes uint8 Length: 69, dtype: object

#To find missing values after encoding,concation,dropping
dfe.isna().sum()

Account length 0 Area code 0 Number vmail messages Total day minutes 0 Total day calls 0 State_WI 0 State_WV 0 State_WY International plan Yes 0 Voice mail plan_Yes Length: 69, dtype: int64

#Seperate x_train
x_train=dfe.drop(['Churn'],axis=1)
x train

	Account length		Number vmail messages	day	day	Total day charge	Total eve minutes	Total eve calls	eve	_	 State_TX	S
0	128	415	25	265.1	110	45.07	197.4	99	16.78	244.7	 0	
1	107	415	26	161.6	123	27.47	195.5	103	16.62	254.4	 0	
2	137	415	0	243.4	114	41.38	121.2	110	10.30	162.6	 0	
3	84	408	0	299.4	71	50.90	61.9	88	5.26	196.9	 0	
4	75	415	0	166.7	113	28.34	148.3	122	12.61	186.9	 0	

```
#Seperate y_train
y_train=dfe['Churn']
y_train
```

0 False
1 False
2 False
3 False
4 False
...
2661 False
2662 False
2664 False
2664 False
2665 False

Name: Churn, Length: 2666, dtype: bool

Preprocessing Testing Data

import numpy as np
import pandas as pd
import seaborn as sns
dt=pd.read_csv('/content/churn-bigml-20.csv')
dt

	State	Account length		International plan	Voice mail plan	Number vmail messages	Total day minutes	day	Total day charge	Total eve minutes	Total eve calls	Tota e\ char{
0	LA	117	408	No	No	0	184.5	97	31.37	351.6	80	29.8
1	IN	65	415	No	No	0	129.1	137	21.95	228.5	83	19.4
2	NY	161	415	No	No	0	332.9	67	56.59	317.8	97	27.0
3	SC	111	415	No	No	0	110.4	103	18.77	137.3	102	11.€
4	Н	49	510	No	No	0	119.3	117	20.28	215.1	109	18.2
662	WI	114	415	No	Yes	26	137.1	88	23.31	155.7	125	13.2
663	AL	106	408	No	Yes	29	83.6	131	14.21	203.9	131	17.3
664	VT	60	415	No	No	0	193.9	118	32.96	85.0	110	7.2
665	WV	159	415	No	No	0	169.8	114	28.87	197.7	105	16.8
666	СТ	184	510	Yes	No	0	213.8	105	36.35	159.6	84	13.5

667 rows × 20 columns



#First 5 observation print
dt.head()

	State	Account length	Area code	International plan	Voice mail plan	Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	Total eve calls	Total eve charge
0	LA		408	No	No	0				351.6		
1	IN	65	415	No	No	0	129.1	137	21.95	228.5	83	19.42

#Last 5 observation print
dt.tail()

	State	Account length		International plan	Voice mail plan	Number vmail messages	day	Total day calls	Total day charge	Total eve minutes	Total eve calls	Tota e\ char{
662	WI	114	415	No	Yes	26	137.1	88	23.31	155.7	125	13.2
663	AL	106	408	No	Yes	29	83.6	131	14.21	203.9	131	17.3
664	VT	60	415	No	No	0	193.9	118	32.96	85.0	110	7.2
665	WV	159	415	No	No	0	169.8	114	28.87	197.7	105	16.8
666	СТ	184	510	Yes	No	0	213.8	105	36.35	159.6	84	13.5



#Each column types dt.dtypes

```
State
                          object
Account length
                          int64
Area code
                           int64
International plan
                          object
Voice mail plan
                          object
Number vmail messages
                          int64
Total day minutes
                         float64
Total day calls
                          int64
                         float64
Total day charge
Total eve minutes
                         float64
Total eve calls
                          int64
                         float64
Total eve charge
Total night minutes
                        float64
                          int64
Total night calls
                         float64
Total night charge
Total intl minutes
                        float64
Total intl calls
                          int64
Total intl charge
                         float64
Customer service calls
                          int64
                            bool
dtype: object
```

#Column heading print dt.columns

#To find missing values dt.isna().sum()

State	0
Account length	0
Area code	0
International plan	0
Voice mail plan	0
Number vmail messages	0
Total day minutes	0
Total day calls	0
Total day charge	0
Total eve minutes	0
Total eve calls	0
Total eve charge	0
Total night minutes	0
Total night calls	0
Total night charge	0
Total intl minutes	0

```
Total intl calls 0
Total intl charge 0
Customer service calls 0
Churn 0
dtype: int64
```

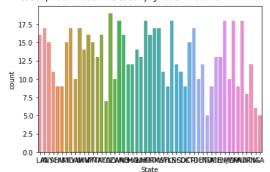
#Each state counts

```
dt['State'].value_counts()
```

```
ND
      18
WV
      18
NJ
KS
      18
NM
      18
WA
      18
RI
      17
ID
      17
MS
      17
WI
      17
\mathsf{T}\mathsf{X}
      17
IN
      17
LA
      16
      16
OR
      16
ΚY
      16
VT
      16
NY
      15
ΜI
      15
MT
      15
СТ
      15
AL
      14
MN
      14
MA
      13
ΙL
      13
ME
      13
NH
      13
TN
      12
UT
      12
NC
      12
MO
      12
ОН
      12
SD
      11
WY
SC
      11
CA
      10
DE
      10
VA
      10
MD
      10
FΙ
       9
ΑK
       9
DC
       9
ΗI
       9
РΑ
AR
CO
IΑ
        6
NV
        5
GΑ
Name: State, dtype: int64
```

#Each state counts graph
sns.countplot('State',data=dt)

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(
<AxesSubplot:xlabel='State', ylabel='count'>

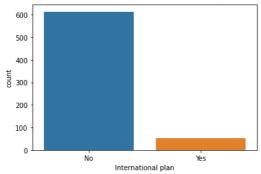


Yes 53 Name: International plan, dtype: int64

#Each International plan counts graph
sns.countplot('International plan',data=dt)

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

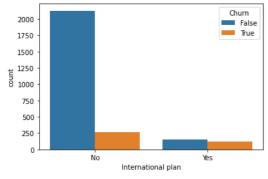
<AxesSubplot:xlabel='International plan', ylabel='count'>



#How Each International plan affect Churn graph
sns.countplot('International plan',data=df,hue='Churn')

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

<AxesSubplot:xlabel='International plan', ylabel='count'>



#Each Voice mail plan counts
dt['Voice mail plan'].value_counts()

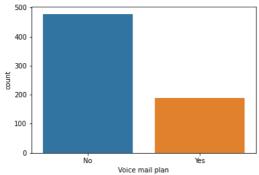
No 478 Yes 189

Name: Voice mail plan, dtype: int64

#Each Voice mail plan counts graph
sns.countplot('Voice mail plan',data=dt)

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

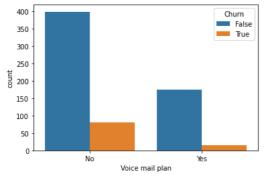
<AxesSubplot:xlabel='Voice mail plan', ylabel='count'>



#How Each Voice mail plan affect Churn graph
sns.countplot('Voice mail plan',data=dt,hue='Churn')

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

<AxesSubplot:xlabel='Voice mail plan', ylabel='count'>



#Each churn counts
dt['Churn'].value_counts()

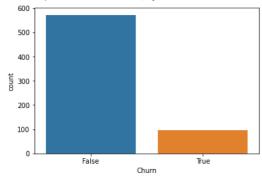
False 572 True 95

Name: Churn, dtype: int64

#Each Churn counts graph
sns.countplot('Churn',data=dt)

/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following var warnings.warn(

<AxesSubplot:xlabel='Churn', ylabel='count'>

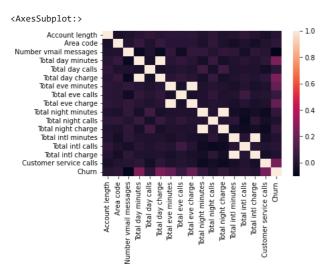


Correlation

dt.corr()

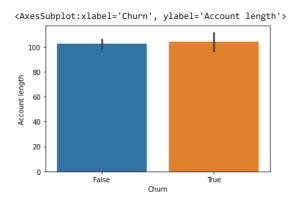
	Account length	Area code	Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	Total eve calls	Total eve charge	
Account length	1.000000	-0.026327	-0.011993	0.017833	0.035703	0.017839	0.027043	0.021237	0.027051	-
Area code	-0.026327	1.000000	-0.006907	0.051507	-0.008972	0.051492	0.017160	0.017783	0.017182	-
Number vmail messages	-0.011993	-0.006907	1.000000	-0.069172	-0.009952	-0.069187	0.040865	-0.051951	0.040876	
Total day minutes	0.017833	0.051507	-0.069172	1.000000	-0.032306	1.000000	0.017987	0.043219	0.017945	-
Total day calls	0.035703	-0.008972	-0.009952	-0.032306	1.000000	-0.032319	-0.004688	0.005851	-0.004664	
Total day charge	0.017839	0.051492	-0.069187	1.000000	-0.032319	1.000000	0.017983	0.043231	0.017941	-
Total eve minutes	0.027043	0.017160	0.040865	0.017987	-0.004688	0.017983	1.000000	-0.029077	1.000000	-
Total eve calls	0.021237	0.017783	-0.051951	0.043219	0.005851	0.043231	-0.029077	1.000000	-0.029089	-
Total eve charge	0.027051	0.017182	0.040876	0.017945	-0.004664	0.017941	1.000000	-0.029089	1.000000	-
Total night	-0.007527	-0.016832	0.039751	-0.031600	0.079536	-0.031613	-0.007705	-0.009856	-0.007691	

#Heatmap-correlation showing graph
sns.heatmap(dt.corr())



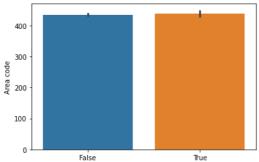
BARPLOT

#Barplot Graph Of churn & Account length
sns.barplot(x='Churn',y='Account length',data=dt)



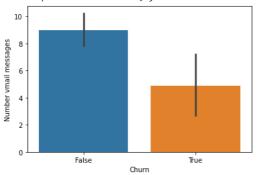
#Barplot Graph Of churn & area code
sns.barplot(x='Churn',y='Area code',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Area code'>



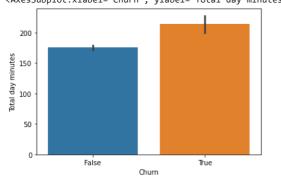
#Barplot Graph Of churn & Number vmail messages
sns.barplot(x='Churn',y='Number vmail messages',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Number vmail messages'>



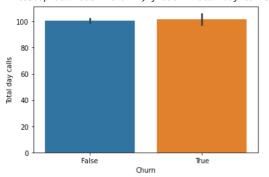
#Barplot Graph Of churn & Total day minutes
sns.barplot(x='Churn',y='Total day minutes',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Total day minutes'>



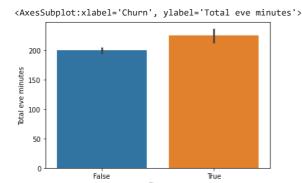
#Barplot Graph Of churn & Total day calls
sns.barplot(x='Churn',y='Total day calls',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Total day calls'>

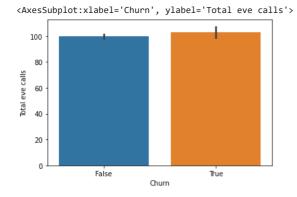


#Barplot Graph Of churn & Total day charge
sns.barplot(x='Churn',y='Total day charge',data=dt)

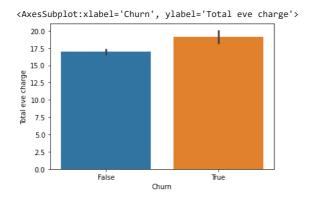
#barplot Graph Of Churn & Total eve minutes
sns.barplot(x='Churn',y='Total eve minutes',data=dt)



#barplot Graph Of Churn & Total eve calls
sns.barplot(x='Churn',y='Total eve calls',data=dte)



#barplot Graph Of Churn & Total eve charge
sns.barplot(x='Churn',y='Total eve charge',data=dt)



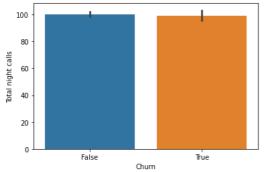
#Barplot Graph Of churn & Total night minutes
sns.barplot(x='Churn',y='Total night minutes',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Total night minutes'>



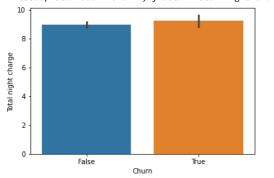
#Barplot Graph Of churn & Total night calls
sns.barplot(x='Churn',y='Total night calls',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Total night calls'>



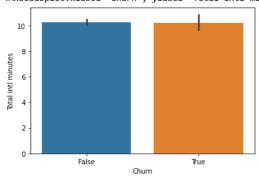
#barplot Graph Of Churn & Total night charge
sns.barplot(x='Churn',y='Total night charge',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Total night charge'>



#barplot Graph Of Churn & Total intl minutes
sns.barplot(x='Churn',y='Total intl minutes',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Total intl minutes'>



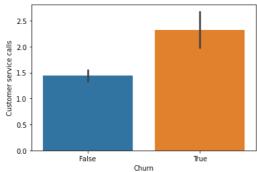
#barplot Graph Of churn & Total intl calls
sns.barplot(x='Churn',y='Total intl calls',data=dt)

#barplot Graph Of churn & Total intl charge
sns.barplot(x='Churn',y='Total intl charge',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Total intl charge'>
3.0
2.5
9 2.0
1.5
1.5
1.0
0.5
0.0
False
True
Churn

#barplot Graph Of churn & Customer service calls
sns.barplot(x='Churn',y='Customer service calls',data=dt)

<AxesSubplot:xlabel='Churn', ylabel='Customer service calls'>



Encoding string to numeric using getdummies

dt1=pd.get_dummies(dt[['State','International plan','Voice mail plan']],drop_first=True)
d+1

	State_AL	State_AR	State_AZ	State_CA	State_CO	State_CT	State_DC	State_DE	State_FL	State_GA
0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0
662	0	0	0	0	0	0	0	0	0	0
663	1	0	0	0	0	0	0	0	0	0
664	0	0	0	0	0	0	0	0	0	0
665	0	0	0	0	0	0	0	0	0	0
666	0	0	0	0	0	1	0	0	0	0

667 rows × 52 columns



concatination-combining

dte=pd.concat([dt,dt1],axis=1)
dte

	State	Account length		International plan	Voice mail plan	Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	•••	State_1
0	LA	117	408	No	No	0	184.5	97	31.37	351.6		
1	IN	65	415	No	No	0	129.1	137	21.95	228.5		
2	NY	161	415	No	No	0	332.9	67	56.59	317.8		
3	SC	111	415	No	No	0	110.4	103	18.77	137.3		
4	HI	49	510	No	No	0	119.3	117	20.28	215.1		
662	WI	114	415	No	Yes	26	137.1	88	23.31	155.7		
663	AL	106	408	No	Yes	29	83.6	131	14.21	203.9		
664	VT	60	415	No	No	0	193.9	118	32.96	85.0		
665	WV	159	415	No	No	0	169.8	114	28.87	197.7		
666	СТ	184	510	Yes	No	0	213.8	105	36.35	159.6		

667 rows × 72 columns



#Dropping unwanted columns
dte.drop(['State','International plan','Voice mail plan'],axis=1,inplace=True)

	Account length		Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	eve	Total eve charge	Total night minutes	 State_TX	St
0	117	408	0	184.5	97	31.37	351.6	80	29.89	215.8	 0	
1	65	415	0	129.1	137	21.95	228.5	83	19.42	208.8	 0	
2	161	415	0	332.9	67	56.59	317.8	97	27.01	160.6	 0	
3	111	415	0	110.4	103	18.77	137.3	102	11.67	189.6	 0	
4	49	510	0	119.3	117	20.28	215.1	109	18.28	178.7	 0	
				***			***				 	
662	114	415	26	137.1	88	23.31	155.7	125	13.23	247.6	 0	
663	106	408	29	83.6	131	14.21	203.9	131	17.33	229.5	 0	
664	4 60	415	0	193.9	118	32.96	85.0	110	7.23	210.1	 0	
665	159	415	0	169.8	114	28.87	197.7	105	16.80	193.7	 0	
666	184	510	0	213.8	105	36.35	159.6	84	13.57	139.2	 0	

667 rows × 69 columns



 $\mbox{\tt\#TO}$ find types of columns after encoding,concation,dropping $\mbox{\tt dte.dtypes}$

Account length Area code Number vmail messages Total day minutes Total day calls	int64 int64 int64 float64 int64
State_WI State_WV State_WY International plan_Yes Voice mail plan Yes	uint8 uint8 uint8 uint8 uint8
Length: 69, dtype: object	dinco

#To find missing values after encoding,concation,dropping
dte.isna().sum()

Account length	0
Area code	0
Number vmail messages	0
Total day minutes	0

```
Total day calls

...
State_WI 0
State_WV 0
State_WY 0
International plan_Yes 0
Voice mail plan_Yes 0
Length: 69, dtype: int64
```

#Seperate x_test

x_test=dte.drop(['Churn'],axis=1)

x_test

	Account length	Area code	Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	Total eve calls	Total eve charge	Total night minutes	 State_TX	St
0	117	408	0	184.5	97	31.37	351.6	80	29.89	215.8	 0	
1	65	415	0	129.1	137	21.95	228.5	83	19.42	208.8	 0	
2	161	415	0	332.9	67	56.59	317.8	97	27.01	160.6	 0	
3	111	415	0	110.4	103	18.77	137.3	102	11.67	189.6	 0	
4	49	510	0	119.3	117	20.28	215.1	109	18.28	178.7	 0	
662	114	415	26	137.1	88	23.31	155.7	125	13.23	247.6	 0	
663	106	408	29	83.6	131	14.21	203.9	131	17.33	229.5	 0	
664	60	415	0	193.9	118	32.96	85.0	110	7.23	210.1	 0	
665	159	415	0	169.8	114	28.87	197.7	105	16.80	193.7	 0	
666	184	510	0	213.8	105	36.35	159.6	84	13.57	139.2	 0	

667 rows × 68 columns



```
#Seperate y_test
y_test=dte['Churn']
y_test
           False
            True
     2
            True
           False
     3
           False
           False
     662
     663
           False
     664
           False
     665
           False
     666
           False
     Name: Churn, Length: 667, dtype: bool
```

Normalization using Minmaxscaler

```
from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
scaler.fit(x_train)
x_train=scaler.fit_transform(x_train)
x_test=scaler.fit_transform(x_test)
x_train
```

```
array([[0.52479339, 0.06862745, 0.5
                                        , ..., 0.
       1.
                ],
      [0.43801653, 0.06862745, 0.52
                                        , ..., 0.
                                                        , 0.
             ],
      [0.56198347, 0.06862745, 0.
                                       , ..., 0.
       0. ],
      [0.2768595 , 0.06862745, 0.
                                                        , 0.
               ],
                                                        , 0.
                                       , ..., 0.
      0. ],
[0.30165289, 0.06862745, 0.5
                                       , ..., 0.
                                                        , 0.
```

```
x_test
```

```
, 0.
                                                    , 0.
array([[0.5021645 , 0.
                                     , ..., 0.
      0.
      [0.27705628, 0.06862745, 0.
                                     , ..., 0.
                                                    , 0.
      0.
               ],
      [0.69264069, 0.06862745, 0.
                                                    , 0.
            ],
      [0.25541126, 0.06862745, 0.
                                                    , 0.
                                     , ..., 0.
       0.
              ],
                                                    , 0.
      [0.68398268, 0.06862745, 0.
       0.
           ],
                       , 0.
      [0.79220779, 1.
                                     , ..., 0.
                                                    , 1.
      0.
           ]])
```

Model creation

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive_bayes import MultinomialNB
from sklearn.svm import SVC
K_model=KNeighborsClassifier(n_neighbors=5)
nb_model=MultinomialNB()
sv_model=SVC()
lsb_model=[K_model,nb_model,sv_model]
```

Perfomance Evaluation

Accuracy score, Classification_report

```
from sklearn.metrics import confusion_matrix,accuracy_score,classification_report
for i in lsb_model:
 print(i)
 i.fit(x_train,y_train)
 y_pred=i.predict(x_test)
 result=confusion_matrix(y_test,y_pred)
 print(result)
 print(accuracy_score(y_test,y_pred))
 print(classification_report(y_test,y_pred))
   KNeighborsClassifier()
   0.8590704647676162
   **************
            precision recall f1-score support
                 0.87 0.99
        False
                               0.92
                                       572
                0.54
         True
                     0.07
                               0.13
                                        95
      accuracy
                               0.86
                                       667
                0.70
     macro avg
                       0.53
                               0.53
                                       667
   weighted avg
                 0.82
                        0.86
                               0.81
                                       667
   MultinomialNB()
   [[569 3]
[91 4]]
   [ 91 4]
   0.8590704647676162
           precision recall f1-score support
                 0.86 0.99
        False
                               0.92
                                       572
                0.57 0.04
         True
      accuracy
                               0.86
                                       667
             0.72
0.82
                        0.52
                               0.50
                                       667
     macro avg
   weighted avg
                        0.86
                               0.80
                                       667
   SVC()
         31
   [[569
    [ 88 7]]
   0.863568215892054
     ***************
             precision recall f1-score support
        False
                0.87
                        0.99
                               0.93
                                       572
```

True	0.70	0.07	0.13	95
accuracy			0.86	667
macro avg	0.78	0.53	0.53	667
weighted avg	0.84	0.86	0.81	667

✓ 0s completed at 11:30 AM