# → 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	Total day minutes	Total day calls	Total day charge	T min
0	KS	128	415	No	Yes	25	265.1	110	45.07	1
1	ОН	107	415	No	Yes	26	161.6	123	27.47	1
2	NJ	137	415	No	No	0	243.4	114	41.38	1
3	ОН	84	408	Yes	No	0	299.4	71	50.90	
4	OK	75	415	Yes	No	0	166.7	113	28.34	1
2661	SC	79	415	No	No	0	134.7	98	22.90	1
2662	AZ	192	415	No	Yes	36	156.2	77	26.55	2
2663	WV	68	415	No	No	0	231.1	57	39.29	1
2664	RI	28	510	No	No	0	180.8	109	30.74	2
2665	TN	74	415	No	Yes	25	234.4	113	39.85	2

2666 rows × 20 columns

#First 5 observation print
df.head()

	State	Account length		International plan	Voice mail plan	vmail	,	day	day	Tota ev minute
0	KS	128	415	No	Yes	25	265.1	110	45.07	197.
1	ОН	107	415	No	Yes	26	161.6	123	27.47	195.
2	NJ	137	415	No	No	0	243.4	114	41.38	121.
3	ОН	84	408	Yes	No	0	299.4	71	50.90	61.

#Last 5 observation print
df.tail()

	State	Account length		International plan	Voice mail plan	Number vmail messages	Total day minutes	Total day calls	Total day charge	T min
2661	SC	79	415	No	No	0	134.7	98	22.90	1
2662	AZ	192	415	No	Yes	36	156.2	77	26.55	2
2663	WV	68	415	No	No	0	231.1	57	39.29	1
2664	RI	28	510	No	No	0	180.8	109	30.74	2
2665	TN	74	415	No	Yes	25	234.4	113	39.85	2
4										•

#Column heading print
df.columns

```
Index(['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', 'Total night minutes',
    'Total night calls', 'Total night charge', 'Total intl minutes',
    'Total intl calls', 'Total intl charge', 'Customer service calls',
    'Churn'],
    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 float64 Total day charge Total eve minutes float64 Total eve calls int64 Total eve charge float64 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 0 Account length Area code 0 International plan 0 Voice mail plan 0 Number vmail messages 0 Total day minutes Total day calls Total day charge 0 Total eve minutes Total eve calls 0 Total eve charge 0 Total night minutes Total night calls Total night charge 0 Total intl minutes Total intl calls 0 Total intl charge 0 Customer service calls 0 Churn dtype: int64

# Finding Count Of Each String Value Columns and plotting Graphs

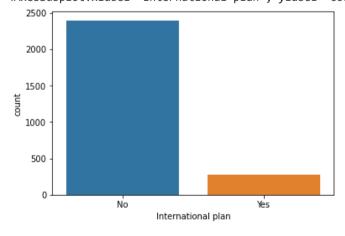
```
#state counts
df['State'].value_counts()
          88
    WV
    MN
          70
    NY
          68
    VA
          67
    ΑL
          66
    OH
          66
    WY
          66
    OR
          62
          61
    NV
    WI
          61
    MD
          60
    UT
          60
    CO
          59
          59
    CT
          58
    ΜI
    VT
          57
    ID
          56
    NC
          56
    TX
          55
          54
    FL
    IN
          54
          53
    MT
          52
    OK
          52
    MA
    KS
          52
    МО
          51
          51
    DE
    NJ
          50
    SC
          49
    SD
          49
    ME
          49
    GΑ
          49
    RΙ
          48
          48
    MS
    WA
          48
          47
    AR
    ΙL
          45
    DC
          45
    ΑZ
          45
    NE
          45
    ΗI
          44
```

```
8/16/23, 10:36 AM
```

```
NM
           44
     ND
           44
    ΑK
          43
    ΚY
          43
     NH
          43
     TN
          41
    IΑ
           38
           36
     РΑ
     LA
           35
     CA
           24
     Name: State, dtype: int64
#International plan count
df['International plan'].value_counts()
    No
           2396
     Yes
            270
    Name: International plan, dtype: int64
```

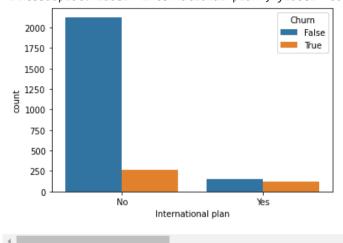
#International plan count graph
sns.countplot('International plan',data=df)

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



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

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



#voice mail plan count
df['Voice mail plan'].value\_counts()

No 1933 Yes 733

Name: Voice mail plan, dtype: int64

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

```
/usr/local/lib/python3.9/dist-packages/seaborn/ decorators.py:36: FutureWarning: Pass
       warnings.warn(
     <AxesSubplot:xlabel='Voice mail plan', ylabel='count'>
#How Voice mail plan affect Churn(Output_Variables) graph
sns.countplot('Voice mail plan',data=df,hue='Churn')
     /usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass 1
       warnings.warn(
     <AxesSubplot:xlabel='Voice mail plan', ylabel='count'>
       1600
                                                Churn
                                               False
       1400
                                               True
       1200
       1000
        800
        600
        400
        200
                     Yes
                                           No
                            Voice mail plan
#Churn count
df['Churn'].value counts()
     False
             2278
              388
     True
     Name: Churn, dtype: int64
#Churn counts graph
sns.countplot('Churn',data=df)
```

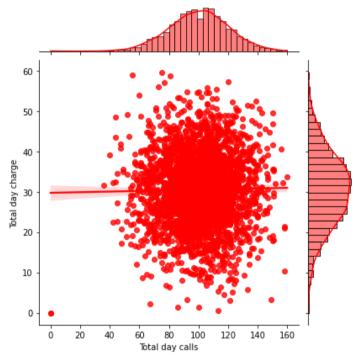
	Account length	Area code	Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	T(
Account length	1.000000	-0.008620	-0.002996	0.002847	0.038862	0.002843	-0.015923	0.018
Area code	-0.008620	1.000000	-0.000584	-0.023134	-0.009629	-0.023130	0.000679	-0.018
Number vmail messages	-0.002996	-0.000584	1.000000	0.019027	-0.009622	0.019027	0.011401	0.00{
Total day minutes	0.002847	-0.023134	0.019027	1.000000	0.016780	1.000000	0.003999	0.009
Total day calls	0.038862	-0.009629	-0.009622	0.016780	1.000000	0.016787	-0.026003	0.006
Total day charge	0.002843	-0.023130	0.019027	1.000000	0.016787	1.000000	0.004008	0.009
Total eve minutes	-0.015923	0.000679	0.011401	0.003999	-0.026003	0.004008	1.000000	-0.007
Total eve calls	0.018552	-0.018602	0.005131	0.009059	0.006473	0.009056	-0.007654	1.000

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



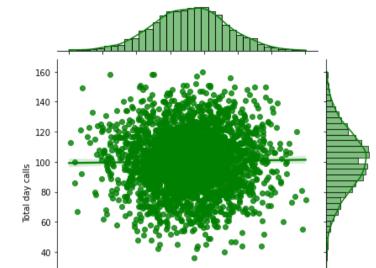
#How Total day calls affect Total day charge
sns.jointplot(x='Total day calls',y='Total day charge',data=df,kind='reg',color='red')

<seaborn.axisgrid.JointGrid at 0x7f76183a40d0>



#How Total day mins affect Total day calls
sns.jointplot(x='Total day minutes',y='Total day calls',data=df,kind='reg',color='green')





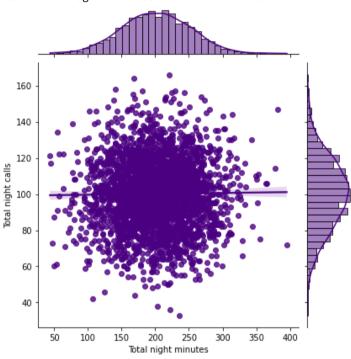
#How Total night calls affect Total night charge
sns.jointplot(x='Total night calls',y='Total night charge',data=df,kind='reg',color='black')

<seaborn.axisgrid.JointGrid at 0x7f76186aec10>

#How Total night minutes affect Total night calls
sns.jointplot(x='Total night minutes',y='Total night calls',data=df,kind='reg',color='indigo')

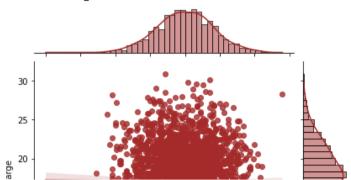


<seaborn.axisgrid.JointGrid at 0x7f76180329d0>



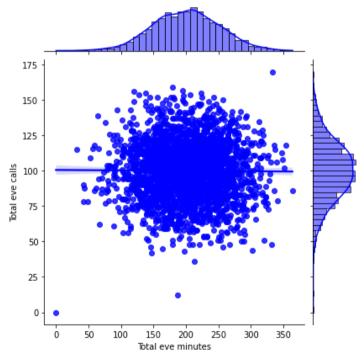
#How Total eve calls affect Total eve charge
sns.jointplot(x='Total eve calls',y='Total eve charge',data=df,kind='reg',color='brown')

<seaborn.axisgrid.JointGrid at 0x7f7617f7f2e0>



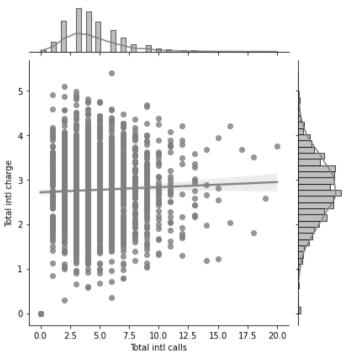
#How Total eve minutes affect Total eve calls
sns.jointplot(x='Total eve minutes',y='Total eve calls',data=df,kind='reg',color='blue')

<seaborn.axisgrid.JointGrid at 0x7f7617c9f370>



#How Total intl calls affect Total intl charge
sns.jointplot(x='Total intl calls',y='Total intl charge',data=df,kind='reg',color='grey')

<seaborn.axisgrid.JointGrid at 0x7f7618491580>



# **Encoding string to numeric using getdummies**

df1=pd.get\_dummies(df[['State','International plan','Voice mail plan']],drop\_first=True)
df1

	State_AL	State_AR	State_AZ	State_CA	State_CO	State_CT	State_DC	State_DE	S
0	0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	0	
2	0	0	0	0	0	0	0	0	
3	0	0	0	0	0	0	0	0	
4	0	0	0	0	0	0	0	0	

# concatination-combining

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

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

2666 rows × 72 columns

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

	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
0	128	415	25	265.1	110	45.07	197.4	99	16.78	244.7
1	107	415	26	161.6	123	27.47	195.5	103	16.62	254.4
2	137	415	0	243.4	114	41.38	121.2	110	10.30	162.6
3	84	408	0	299.4	71	50.90	61.9	88	5.26	196.9
4	75	415	0	166.7	113	28.34	148.3	122	12.61	186.9
2661	79	415	0	134.7	98	22.90	189.7	68	16.12	221.4
2662	192	415	36	156.2	77	26.55	215.5	126	18.32	279.1
2663	68	415	0	231.1	57	39.29	153.4	55	13.04	191.3
2664	28	510	0	180.8	109	30.74	288.8	58	24.55	191.9
2665	74	415	25	234.4	113	39.85	265.9	82	22.60	241.4
2666 rd	ows × 69 co	lumns								
4										<b>&gt;</b>

#To find types of columns after encoding,concation,dropping
dfe.dtypes

Account length	int64
Area code	int64
Number vmail messages	int64
Total day minutes	float64
Total day calls	int64
-	

x\_train

```
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
    Total day calls
    State_WI
    State WV
    State WY
    International plan_Yes
    Voice mail plan Yes
    Length: 69, dtype: int64
#Seperate x train
x_train=dfe.drop(['Churn'],axis=1)
```

	Account length		Number vmail messages	day	Total day calls	Total day charge	Total eve minutes	Total eve calls	Total eve charge	Total night minutes
0	128	415	25	265.1	110	45.07	197.4	99	16.78	244.7
1	107	415	26	161.6	123	27.47	195.5	103	16.62	254.4
2	137	415	0	243.4	114	41.38	121.2	110	10.30	162.6
3	84	408	0	299.4	71	50.90	61.9	88	5.26	196.9

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

```
0
       False
       False
1
       False
3
        False
       False
2661
       False
       False
2662
2663
       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
dft=pd.read_csv('/content/churn-bigml-20.csv')
dft
```

	State	Account length	Area code	International plan	Voice mail plan	Number vmail messages	Total day minutes	Total day calls	Total day charge	To minu
0	LA	117	408	No	No	0	184.5	97	31.37	35
1	IN	65	415	No	No	0	129.1	137	21.95	22
2	NY	161	415	No	No	0	332.9	67	56.59	31
3	SC	111	415	No	No	0	110.4	103	18.77	13
4	НІ	49	510	No	No	0	119.3	117	20.28	21
662	WI	114	415	No	Yes	26	137.1	88	23.31	15
663	AL	106	408	No	Yes	29	83.6	131	14.21	20
664	VT	60	415	No	No	0	193.9	118	32.96	8
665	WV	159	415	No	No	0	169.8	114	28.87	19
222	^ <del>-</del>	404	F40	¥7	K I =	^	040.0	405	22.25	41

#First 5 observation print
dft.head()

	State	Account length		International plan	Voice mail plan	Number vmail messages	day	Total day calls	Total day charge	Tota ev minute
0	LA	117	408	No	No	0	184.5	97	31.37	351.
1	IN	65	415	No	No	0	129.1	137	21.95	228.
2	NY	161	415	No	No	0	332.9	67	56.59	317.
3	SC	111	415	No	No	0	110.4	103	18.77	137.
4	HI	49	510	No	No	0	119.3	117	20.28	215.
4										<b>&gt;</b>

#Last 5 observation print
dft.tail()

	State	Account length		International plan	Voice mail plan	Number vmail messages	Total day minutes	Total day calls	Total day charge	To minu
662	WI	114	415	No	Yes	26	137.1	88	23.31	15
663	AL	106	408	No	Yes	29	83.6	131	14.21	20
664	VT	60	415	No	No	0	193.9	118	32.96	8
665	WV	159	415	No	No	0	169.8	114	28.87	19
666	СТ	184	510	Yes	No	0	213.8	105	36.35	15
4										<b>•</b>

# #Each column types dft.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
Total eve minutes	float64
Total eve calls	int64
Total eve charge	float64
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
dtvpe: object	

dtype: object

#Column heading print
dft.columns

```
Index(['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', 'Total night minutes',
            'Total night calls', 'Total night charge', 'Total intl minutes',
            'Total intl calls', 'Total intl charge', 'Customer service calls',
            'Churn'],
           dtype='object')
#To find missing values
dft.isna().sum()
     State
     Account length
     Area code
     International plan
                               0
     Voice mail plan
     Number vmail messages
     Total day minutes
     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
                               0
     Total intl calls
     Total intl charge
                               0
     Customer service calls
     Churn
     dtype: int64
```

#### Finding Count Of Each String Value Columns and plotting Graphs

```
#State Counts
dft['State'].value_counts()

AZ 19
ND 18
WV 18
NJ 18
KS 18
```

18 NM WA 18 RΙ 17 ID 17 MS 17 WI 17 TX 17 IN 17 LA 16 NE 16 OR 16 ΚY 16 VT 16 NY 15 ΜI 15 15 MT СТ 15 ΑL 14 MN 14 MA 13 13 ΙL ME 13 NH 13 12 TN UT 12 NC 12 MO 12 ОН 12 SD 11 WY 11 SC 11 CA 10 DE 10 VA 10 MD 10 FL 9 ΑK 9 DC 9 ΗI 9 9 PA 9 OK 8 AR 7 CO IΑ 6 NV 5 GΑ 5

Name: State, dtype: int64

```
#International plan counts
dft['International plan'].value_counts()
     No
            614
             53
     Yes
     Name: International plan, dtype: int64
#International plan counts graph
sns.countplot('International plan',data=dft)
     /usr/local/lib/python3.9/dist-packages/seaborn/ decorators.py:36: FutureWarning: Pass
       warnings.warn(
     <AxesSubplot:xlabel='International plan', ylabel='count'>
        600
       500
        400
      300 gnit
       200
       100
         0
                     No
                                           Yes
                           International plan
```

#How International plan affect Churn(Output\_Variables) graph
sns.countplot('International plan',data=dft,hue='Churn')

```
/usr/local/lib/python3.9/dist-packages/seaborn/ decorators.py:36: FutureWarning: Pass
       warnings.warn(
     <AxesSubplot:xlabel='International plan', ylabel='count'>
                                                Churn
        500
                                               False
                                                True
#Voice mail plan counts
dft['Voice mail plan'].value counts()
            478
     No
     Yes
           189
     Name: Voice mail plan, dtype: int64
#Voice mail plan counts graph
sns.countplot('Voice mail plan',data=dft)
     /usr/local/lib/python3.9/dist-packages/seaborn/ decorators.py:36: FutureWarning: Pass
       warnings.warn(
     <AxesSubplot:xlabel='Voice mail plan', ylabel='count'>
        500
        400
        300
       200
       100
                     No
                                          Yes
                           Voice mail plan
```

#How Voice mail plan affect Churn(Output\_Variables) graph
sns.countplot('Voice mail plan',data=dft,hue='Churn')

```
/usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass
       warnings.warn(
     <AxesSubplot:xlabel='Voice mail plan', ylabel='count'>
        400
                                                Churn
                                                False
        350
                                                True
        300
        250
      100 Z00
        150
        100
        50
#churn counts
dft['Churn'].value_counts()
              572
     False
     True
               95
     Name: Churn, dtype: int64
#Churn Counts graph
sns.countplot('Churn',data=dft)
     /usr/local/lib/python3.9/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass 1
       warnings.warn(
     <AxesSubplot:xlabel='Churn', ylabel='count'>
        600
        500
        400
        300
        200
        100
                    False
                                           True
                               Churn
```

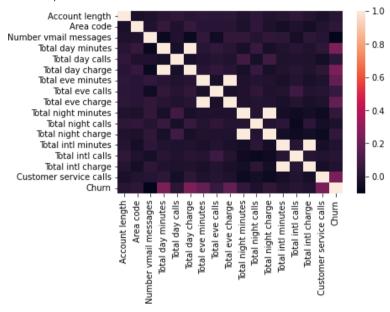
# Correlation

dft.corr()

	Account length	Area code	Number vmail messages	Total day minutes	Total day calls	Total day charge	Total eve minutes	T(
Account length	1.000000	-0.026327	-0.011993	0.017833	0.035703	0.017839	0.027043	0.02
Area code	-0.026327	1.000000	-0.006907	0.051507	-0.008972	0.051492	0.017160	0.017
Number vmail messages	-0.011993	-0.006907	1.000000	-0.069172	-0.009952	-0.069187	0.040865	-0.05
Total day	0.017833	0.051507	-0.069172	1.000000	-0.032306	1.000000	0.017987	0.043

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

#### <AxesSubplot:>

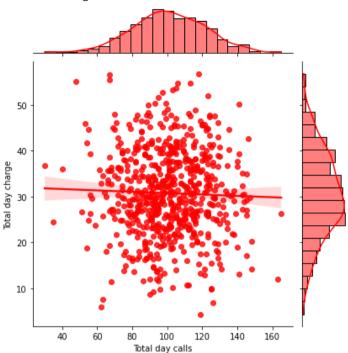


Total intl

#### **JOINTPLOTGRAPH**

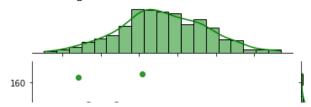
#How Total day calls affect Total day charge sns.jointplot(x='Total day calls',y='Total day charge',data=dft,kind='reg',color='red')





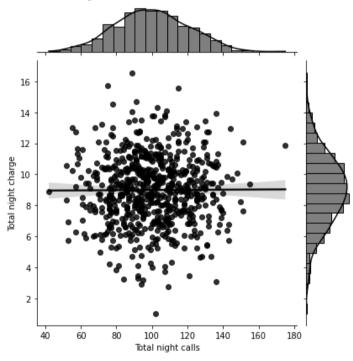
#How Total day mins affect Total day calls
sns.jointplot(x='Total day minutes',y='Total day calls',data=dft,kind='reg',color='green')

<seaborn.axisgrid.JointGrid at 0x7f7617370fa0>



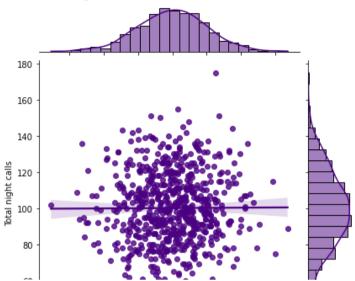
#How Total night calls affect Total night charge
sns.jointplot(x='Total night calls',y='Total night charge',data=dft,kind='reg',color='black')

<seaborn.axisgrid.JointGrid at 0x7f76174c89a0>



#How Total night minutes affect Total night calls
sns.jointplot(x='Total night minutes',y='Total night calls',data=dft,kind='reg',color='indigo')





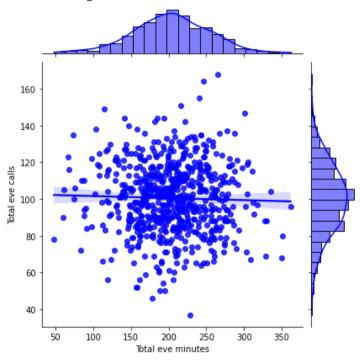
#How Total eve calls affect Total eve charge
sns.jointplot(x='Total eve calls',y='Total eve charge',data=dft,kind='reg',color='brown')

<seaborn.axisgrid.JointGrid at 0x7f7616e94c40>

\_

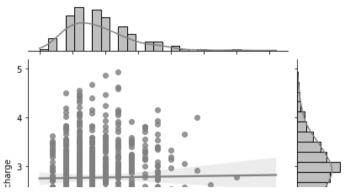
#How Total eve minutes affect Total eve calls
sns.jointplot(x='Total eve minutes',y='Total eve calls',data=dft,kind='reg',color='blue')

<seaborn.axisgrid.JointGrid at 0x7f7616d90b80>



#How Total eve minutes affect Total eve calls
sns.jointplot(x='Total intl calls',y='Total intl charge',data=dft,kind='reg',color='grey')





# **Encoding string to numeric using getdummies**

dft1=pd.get\_dummies(dft[['State','International plan','Voice mail plan']],drop\_first=True)
dft1

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

667 rows × 52 columns

# concatination-combining

dfte=pd.concat([dft,dft1],axis=1)
dfte

	State	Account length	Area code	International plan	Voice mail plan	Number vmail messages	Total day minutes	Total day calls	Total day charge	To minu
0	LA	117	408	No	No	0	184.5	97	31.37	35
1	IN	65	415	No	No	0	129.1	137	21.95	22
2	NY	161	415	No	No	0	332.9	67	56.59	31
3	SC	111	415	No	No	0	110.4	103	18.77	13
4	HI	49	510	No	No	0	119.3	117	20.28	21
662	WI	114	415	No	Yes	26	137.1	88	23.31	15
663	AL	106	408	No	Yes	29	83.6	131	14.21	20
664	VT	60	415	No	No	0	193.9	118	32.96	8
665	WV	159	415	No	No	0	169.8	114	28.87	19
666	СТ	184	510	Yes	No	0	213.8	105	36.35	15
667 r	667 rows × 72 columns									
4										•

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

	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
0	117	408	0	184.5	97	31.37	351.6	80	29.89	215.8
1	65	415	0	129.1	137	21.95	228.5	83	19.42	208.8
2	161	415	0	332.9	67	56.59	317.8	97	27.01	160.6
3	111	415	0	110.4	103	18.77	137.3	102	11.67	189.6
4	49	510	0	119.3	117	20.28	215.1	109	18.28	178.7
662	114	415	26	137.1	88	23.31	155.7	125	13.23	247.6
663	106	408	29	83.6	131	14.21	203.9	131	17.33	229.5
664	60	415	0	193.9	118	32.96	85.0	110	7.23	210.1
665	159	415	0	169.8	114	28.87	197.7	105	16.80	193.7
666	184	510	0	213.8	105	36.35	159.6	84	13.57	139.2
667 rd	ows × 69 cc	lumns								

#To find types of columns after encoding,concation,dropping
dfte.dtypes

Account length	int6
Area code	int6
Number vmail messages	int6
Total day minutes	float6
Total day calls	int6
State_WI	uint
State_WV	uint
State_WY	uint
International plan_Yes	uint
Voice mail plan_Yes	uint
Length: 69, dtype: object	

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

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

667 rows × 68 columns

```
#Seperate y_test
y_test=dfte['Churn']
y_test
     0
           False
    1
           True
     2
           True
     3
           False
           False
           . . .
     662
           False
     663
           False
     664
           False
     665
           False
     666
           False
     Name: Churn, Length: 667, dtype: bool
```

from sklearn.preprocessing import MinMaxScaler

#### **Normalization using Minmaxscaler**

```
x_test
```

1.

]])

#### **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**

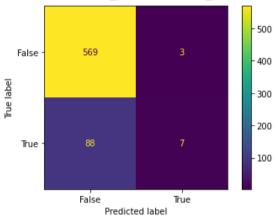
# Accuracy score, Classification\_report

KNeighborsCla [[566 6] [88 7]] ***********************************	******** 76162							
	precision	recall	f1-score	support				
False	0.87	0.99	0.92	572				
True	0.54	0.07	0.13	95				
accuracy			0.86	667				
macro avg	0.70	0.53	0.53	667				
weighted avg	0.82	0.86	0.81	667				
MultinomialNB [[569 3] [91 4]] **********	*****	*****	******					
0.85907046476 *******		******	******					
	precision recall f1-score supp							
False	0.86	0.99	0.92	572				
True	0.57	0.04	0.08	95				
accuracy			0.86	667				
macro avg	0.72	0.52	0.50	667				
weighted avg	0.82	0.86	0.80	667				
SVC() [[569 3] [ 88 7]] ***********************************								
*******								
	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				

# **Dislay Confusion metrics**

from sklearn.metrics.\_plot.confusion\_matrix import ConfusionMatrixDisplay
from sklearn.metrics import confusion\_matrix,ConfusionMatrixDisplay
cm=['False','True']
cmd=ConfusionMatrixDisplay(result,display\_labels=cm)
cmd.plot()

<sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x7f7616abc4f0>



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