## **Libraries Required**

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
```

#### **Dataset Import**

```
churn_set=pd.read_excel('/content/customer_churn_large_dataset.xlsx')
```

# **Data Preprocessing**

Number of rows and column

```
churn_set.shape
(100000, 8)
```

Showing first five rows of the data set

```
churn set.head()
   CustomerID Age Gender
                               Location
                                          Subscription Length Months
0
            1
                63
                      Male Los Angeles
                                                                   17
1
            2
                62 Female
                               New York
                                                                   1
2
            3
                                                                    5
                24 Female Los Angeles
3
            4
                                                                   3
                36 Female
                                  Miami
4
            5
                46 Female
                                  Miami
                                                                   19
   Monthly Bill Total Usage GB Churn
          73.36
0
                            236
                                      0
                                      0
1
          48.76
                            172
2
          85.47
                            460
                                      0
3
                                      1
          97.94
                            297
4
          58.14
                                      0
                            266
```

Check wheather there is null value in dataset or not

```
Monthly_Bill 0
Total_Usage_GB 0
Churn 0
dtype: int64
```

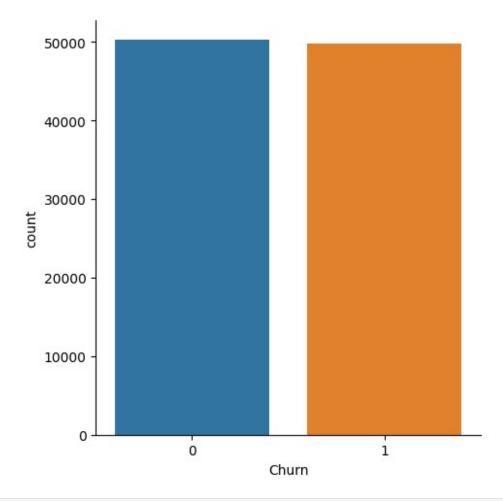
### **Data Analysis and Visualization**

Statistical properties of the dataset

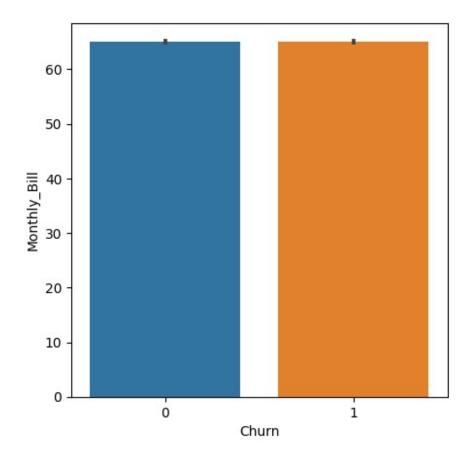
```
churn set.describe()
          CustomerID
                                       Subscription Length Months
                                  Age
       100000.000000
                       100000.000000
count
                                                     100000.000000
        50000.500000
                            44.027020
                                                          12.490100
mean
std
        28867.657797
                            15.280283
                                                           6.926461
min
             1.000000
                            18.000000
                                                           1.000000
25%
        25000.750000
                            31.000000
                                                           6.000000
        50000.500000
                            44.000000
                                                          12.000000
50%
75%
        75000.250000
                            57.000000
                                                          19.000000
       100000.000000
                                                          24.000000
max
                            70.000000
        Monthly Bill
                       Total Usage GB
                                                 Churn
       100000.000000
                        100000.000000
                                         100000.000000
count
mean
           65.053197
                            274.393650
                                              0.497790
std
           20.230696
                            130.463063
                                              0.499998
                             50.000000
min
           30.000000
                                              0.000000
25%
           47.540000
                            161.000000
                                              0.000000
                            274.000000
50%
           65.010000
                                              0.000000
75%
           82.640000
                            387.000000
                                              1.000000
          100.000000
                            500.000000
                                              1.000000
max
```

Graph plot (To see which properties effect our label)

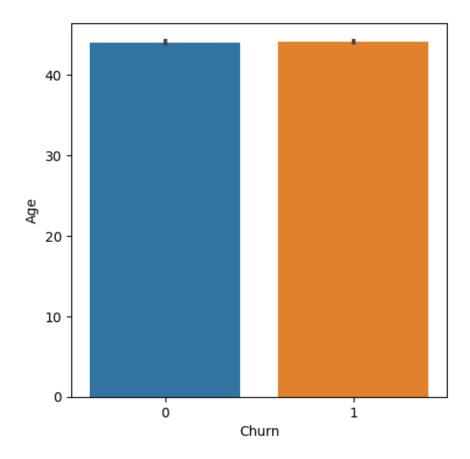
```
sns.catplot(x='Churn',data=churn_set,kind='count')
<seaborn.axisgrid.FacetGrid at 0x785cca2b2c50>
```



```
plot=plt.figure(figsize=(5,5))
sns.barplot(x='Churn',y='Monthly_Bill',data=churn_set)
<Axes: xlabel='Churn', ylabel='Monthly_Bill'>
```

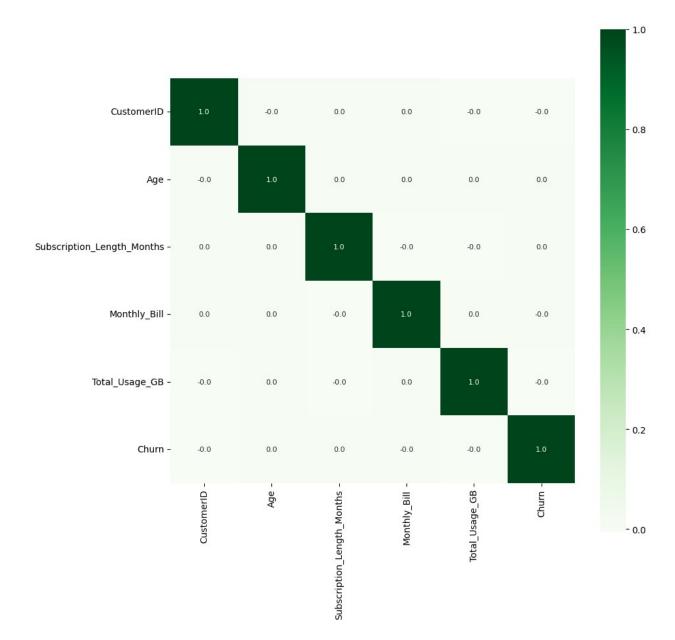


```
plot=plt.figure(figsize=(5,5))
sns.barplot(x='Churn',y='Age',data=churn_set)
<Axes: xlabel='Churn', ylabel='Age'>
```



## Correlation

```
correlation=churn_set.corr()
plt.figure(figsize=(10,10))
sns.heatmap(correlation,cbar=True,square=True,fmt='.1f',annot=True,ann
ot_kws={'size':8},cmap='Greens')
<ipython-input-10-580e05bd6ddd>:1: FutureWarning: The default value of
numeric_only in DataFrame.corr is deprecated. In a future version, it
will default to False. Select only valid columns or specify the value
of numeric_only to silence this warning.
    correlation=churn_set.corr()
```



# **Data Preprocessing**

Removing Label for Churn Analysis

```
X = churn_set.drop('Churn',axis=1)
print(X)
       CustomerID Age
                                   Location
                        Gender
Subscription_Length_Months
                                 Los Angeles
0
                          Male
                    63
17
1
                2
                    62
                        Female
                                    New York
1
```

```
2
                 3
                      24
                          Female Los Angeles
5
                 4
                      36
                          Female
                                          Miami
3
4
                 5
                      46
                          Female
                                          Miami
19
. . .
99995
             99996
                      33
                            Male
                                       Houston
23
99996
             99997
                         Female
                                      New York
                      62
19
99997
             99998
                      64
                            Male
                                       Chicago
17
99998
             99999
                      51
                          Female
                                      New York
20
99999
            100000
                      27 Female Los Angeles
19
       Monthly_Bill
                       Total Usage GB
               73.36
                                   <del>2</del>36
0
               48.76
1
                                   172
2
               85.47
                                   460
3
               97.94
                                   297
4
               58.14
                                   266
                                   . . .
99995
               55.13
                                   226
99996
               61.65
                                   351
99997
               96.11
                                   251
               49.25
                                   434
99998
99999
               76.57
                                   173
[100000 rows x 7 columns]
churn_set["Gender"].value_counts()
Female
           50216
           49784
Male
Name: Gender, dtype: int64
```

#### Label Binarization

```
4 0
...
99995 1
99996 0
99997 1
99998 1
99999 1
Name: Churn, Length: 100000, dtype: int64
```

## **Model Training and Testing**

Splitting data into training and testing

```
X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.2,rando
m_state=1)
print(Y.shape,Y_train.shape,Y_test.shape)
(100000,) (80000,) (20000,)
```

#### Model: Random Forest Classifier

```
from sklearn.ensemble import RandomForestRegressor
model=RandomForestRegressor()
model.fit(X_train,Y_train)
model.score(X test, Y test)
ValueError
                                          Traceback (most recent call
last)
<ipython-input-19-86405439abb2> in <cell line: 2>()
      1 model=RandomForestRegressor()
----> 2 model.fit(X train,Y train)
      3 model.score(X test,Y test)
/usr/local/lib/python3.10/dist-packages/sklearn/ensemble/ forest.py in
fit(self, X, y, sample_weight)
    343
                if issparse(y):
    344
                    raise ValueError("sparse multilabel-indicator for
y is not supported.")
              X, y = self. validate data(
--> 345
    346
                    X, y, multi output=True, accept sparse="csc",
dtype=DTYPE
    347
                )
/usr/local/lib/python3.10/dist-packages/sklearn/base.py in
validate data(self, X, y, reset, validate separately, **check params)
                        y = check array(y, input name="y",
    582
```

```
**check_y_params)
    583
                    else:
--> 584
                        X, y = \text{check } X y(X, y, **\text{check params})
    585
                    out = X, y
    586
/usr/local/lib/python3.10/dist-packages/sklearn/utils/validation.py in
check X y(X, y, accept sparse, accept large sparse, dtype, order,
copy, force_all_finite, ensure_2d, allow_nd, multi_output,
ensure min samples, ensure min features, y numeric, estimator)
   1104
   1105
-> 1106
            X = check array(
   1107
                Χ,
   1108
                accept sparse=accept sparse,
/usr/local/lib/python3.10/dist-packages/sklearn/utils/validation.py in
check array(array, accept sparse, accept large sparse, dtype, order,
copy, force all finite, ensure 2d, allow nd, ensure min samples,
ensure min features, estimator, input name)
    877
                            array = xp.astype(array, dtype,
copy=False)
    878
                        else:
--> 879
                            array = asarray with order(array,
order=order, dtype=dtype, xp=xp)
    880
                    except ComplexWarning as complex warning:
    881
                        raise ValueError(
/usr/local/lib/python3.10/dist-packages/sklearn/utils/ array api.py in
asarray with order(array, dtype, order, copy, xp)
            if xp.__name__ in {"numpy", "numpy.array_api"}:
    183
                # Use NumPy API to support order
    184
--> 185
                array = numpy.asarray(array, order=order, dtype=dtype)
    186
                return xp.asarray(array, copy=copy)
    187
            else:
/usr/local/lib/python3.10/dist-packages/pandas/core/generic.py in
array (self, dtype)
   2068
   2069
            def array (self, dtype: npt.DTypeLike | None = None) ->
np.ndarray:
-> 2070
                return np.asarray(self. values, dtype=dtype)
   2071
   2072
            def array wrap (
ValueError: could not convert string to float: 'Male'
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
categorical features=["Gender","Location"]
```

```
one hot=OneHotEncoder()
transformer=ColumnTransformer([("one hot",
                                  one hot,
                                  categorical features)],
                                  remainder='passthrough')
transformed X=transformer.fit transform(X)
transformed X
                                                 73.36, 236.
array([[
                   1.
                           0.
                                        17.
                           0.
                   0.
                                         1.
                                                 48.76, 172.
                                                               ],
                   0.
                                         5.
                           0.
                                                 85.47, 460.
          1.
                                                               ],
                                                 96.11, 251.
          0.
                   1.
                            1.
                                        17.
                                                               ],
                   0.
                                                 49.25, 434.
          1.
                           0.
                                        20.
                   0.
                           0.
                                        19. ,
                                                 76.57, 173.
                                                               ]])
X.head()
                                            Subscription Length Months
   CustomerID
               Age Gender
                                 Location
0
            1
                 63
                       Male
                             Los Angeles
                                                                     17
            2
                                 New York
                                                                      1
1
                 62
                     Female
                                                                      5
2
            3
                 24
                     Female Los Angeles
                                                                      3
3
            4
                 36
                     Female
                                    Miami
4
            5
                                    Miami
                 46
                     Female
                                                                     19
   Monthly_Bill
                Total Usage GB
          73.36
0
                              236
1
          48.76
                              172
2
          85.47
                              460
3
          97.94
                              297
          58.14
                              266
pd.DataFrame(transformed X)
        0
                        3
                                                                       10
                              4
                                   5
                                        6
                                                   7
11
0
       0.0
            1.0
                  0.0
                      0.0
                            1.0
                                 0.0
                                       0.0
                                                  1.0 63.0
                                                             17.0
                                                                    73.36
236.0
1
       1.0
            0.0
                  0.0
                       0.0
                            0.0
                                  0.0
                                       1.0
                                                  2.0
                                                       62.0
                                                               1.0
                                                                    48.76
172.0
       1.0
            0.0
                  0.0
                       0.0
                            1.0
                                  0.0
                                       0.0
                                                  3.0
                                                       24.0
                                                               5.0
                                                                    85.47
460.0
3
       1.0
            0.0
                  0.0
                       0.0
                            0.0
                                  1.0
                                       0.0
                                                  4.0
                                                       36.0
                                                               3.0
                                                                    97.94
297.0
       1.0
            0.0
                  0.0
                       0.0
                            0.0
                                  1.0
                                       0.0
                                                  5.0
                                                       46.0
                                                              19.0
                                                                    58.14
266.0
. . .
99995
       0.0 1.0
                  0.0 1.0 0.0 0.0
                                       0.0
                                              99996.0 33.0
                                                              23.0 55.13
226.0
```

99996	1.0	0.0	0.0	0.0	0.0	0.0	1.0	99997.0	62.0	19.0	61.65
351.0											
99997	0.0	1.0	1.0	0.0	0.0	0.0	0.0	99998.0	64.0	17.0	96.11
251.0											
99998	1.0	0.0	0.0	0.0	0.0	0.0	1.0	99999.0	51.0	20.0	49.25
434.0											
99999	1.0	0.0	0.0	0.0	1.0	0.0	0.0	100000.0	27.0	19.0	76.57
173.0											

# [100000 rows x 12 columns]

dummies=pd.get\_dummies(churn\_set[["Gender","Location"]])
dummies

	Gender_Female	Gender_Male	Location_Chicago	Location_Houston
0	0	1	0	0
1	1	0	0	0
2	1	0	0	0
3	1	0	0	0
4	1	0	0	0
99995	0	1	0	1
99996	1	0	0	0
99997	0	1	1	0
99998	1	0	0	0
99999	1	0	0	0

	Location_Los Angeles	Location_Miami	Location_New York
0	1	_ 0	Θ
1	Θ	0	1
2	1	0	Θ
3	Θ	1	Θ
4	0	1	0
99995	Θ	0	Θ
99996	Θ	0	1
99997	0	0	Θ
99998	Θ	0	1
99999	1	0	0

```
[100000 rows x 7 columns]
np.random.seed(24)
X_train,X_test,Y_train_Train,Y_Test = train_test_split(transformed_X,
                                                        test size=0.2)
model.fit(X train,Y train)
RandomForestRegressor()
model.score(X test,Y test)
-0.03346709386837565
X.head()
   CustomerID Age Gender
                                          Subscription Length Months
                               Location
0
            1
                63
                      Male Los Angeles
                                                                  17
            2
                               New York
1
                62 Female
                                                                   1
                24 Female Los Angeles
2
            3
                                                                   5
3
            4
                36 Female
                                  Miami
                                                                   3
4
            5
                                  Miami
                46 Female
                                                                  19
   Monthly Bill Total Usage GB
                            236
0
          73.36
1
          48.76
                            172
2
          85.47
                            460
3
          97.94
                            297
4
          58.14
                            266
```

## **Predictive System**

```
input_data=(2,62,1,0,0,0,0,1,0,1,48.76,172)
input_data_as_numpy_array = np.asarray(input_data)
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
prediction = model.predict(input_data_reshaped)
print(prediction)
if(prediction[0]==1):
    print('Churned')
else:
    print('Not Churned')
[0.53]
Not Churned
```