```
In [109]:
           import pandas as pd
           import numpy as np
           from sklearn.model_selection import train_test_split
           from sklearn.linear model import LogisticRegression
           from sklearn.tree import DecisionTreeClassifier
           from sklearn.ensemble import RandomForestClassifier
           from sklearn.neighbors import KNeighborsClassifier
           from sklearn.svm import SVC
           from sklearn.naive bayes import GaussianNB,MultinomialNB,BernoulliNB
           from sklearn.metrics import confusion_matrix,classification_report
           from sklearn.preprocessing import LabelEncoder,MinMaxScaler,OneHotEncoder
           from sklearn.compose import ColumnTransformer
           from sklearn.pipeline import Pipeline
           from xgboost.sklearn import XGBClassifier
  In [4]: | df = pd.read_csv('EDA_Customer.csv')
  In [7]: | df.drop(columns=['Unnamed: 0'],inplace=True)
  In [8]: | df
  Out[8]:
                  CustomerID Age Gender ContractType_In_days MonthlyCharges
                                                                             TotalCharges TechSupport Intel
               0
                       1083
                            79.0
                                    Male
                                                         365
                                                                   90.038513
                                                                              3511.502019
                                                                                                  No
               1
                        1117 60.0
                                  Female
                                                         365
                                                                   80.590894
                                                                              2901.272196
                                                                                                  No
               2
                       3833 84.0
                                  Female
                                                         365
                                                                   43.042067
                                                                              1549.514395
                                                                                                  No
               3
                       1976 69.0
                                                         365
                                                                   51.930032
                                                                              2232.991377
                                    Male
                                                                                                  No
                       3132 49.0
                                                         365
                                                                  101.524194
               4
                                    Male
                                                                              913.717747
                                                                                                 Yes
                                                                                                  ...
            4359
                       2133 39.0
                                                         365
                                                                   30.017101
                                    Male
                                                                               210.119705
                                                                                                  No
            4360
                       1514 54.0
                                    Male
                                                         365
                                                                   57.803077
                                                                              462.424613
                                                                                                  Νo
                       2716 45.0
                                                         730
            4361
                                    Male
                                                                  103.314530
                                                                              826.516243
                                                                                                  No
            4362
                        756 21.0 Female
                                                         730
                                                                  103.105344
                                                                               103.105344
                                                                                                 Yes
            4363
                       3284 85.0
                                                         730
                                                                   36.907180
                                                                              1660.823112
                                                                                                 Yes
                                    Male
           4364 rows × 12 columns
```

In [11]: X = df.iloc[:,1:11]

In [12]: X Out[12]: Gender ContractType\_In\_days MonthlyCharges TotalCharges TechSupport InternetService 1 Age 0 79.0 Male 365 90.038513 3511.502019 No No 1 60.0 Female 365 80.590894 2901.272196 No Fiber optic 365 2 84.0 Female 43.042067 1549.514395 No No 3 69.0 365 Male 51.930032 2232.991377 No No 49.0 Male 365 101.524194 913.717747 DSL Yes **4359** 39.0 Male 365 30.017101 210.119705 Fiber optic No **4360** 54.0 DSL Male 365 57.803077 462.424613 No **4361** 45.0 730 DSL Male 103.314530 826.516243 No **4362** 21.0 Female 730 103.105344 103.105344 Yes Fiber optic **4363** 85.0 730 Fiber optic 36.907180 1660.823112 Male Yes 4364 rows × 10 columns y = df.iloc[:,11] In [14]: In [28]: encode = LabelEncoder() y = pd.Series(encode.fit\_transform(y)) y.value\_counts() In [30]: Out[30]: 1 2233 2131

Name: count, dtype: int64

```
In [37]:
Out[37]:
                  Age
                       Gender ContractType_In_days MonthlyCharges
                                                                   TotalCharges TechSupport InternetService 1
               0 79.0
                          Male
                                               365
                                                         90.038513
                                                                     3511.502019
                                                                                         No
                                                                                                        No
               1
                  60.0
                       Female
                                               365
                                                         80.590894
                                                                    2901.272196
                                                                                         No
                                                                                                  Fiber optic
               2
                  84.0
                       Female
                                               365
                                                         43.042067
                                                                    1549.514395
                                                                                         No
                                                                                                        No
               3
                  69.0
                         Male
                                               365
                                                         51.930032
                                                                    2232.991377
                                                                                         No
                                                                                                        No
                  49.0
                                               365
                                                         101.524194
                                                                     913.717747
                                                                                                       DSL
                         Male
                                                                                         Yes
            4359 39.0
                                                         30.017101
                                                                      210.119705
                                                                                                  Fiber optic
                         Male
                                               365
                                                                                         Nο
            4360 54.0
                                               365
                                                                                                       DSL
                         Male
                                                         57.803077
                                                                     462.424613
                                                                                         No
                                                                                                       DSL
            4361 45.0
                         Male
                                               730
                                                         103.314530
                                                                     826.516243
                                                                                         No
            4362 21.0
                       Female
                                               730
                                                         103.105344
                                                                      103.105344
                                                                                         Yes
                                                                                                  Fiber optic
            4363 85.0
                                               730
                                                                                                  Fiber optic
                         Male
                                                         36.907180
                                                                     1660.823112
                                                                                         Yes
           4364 rows × 10 columns
           X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.1,random_state=42
  In [ ]:
           step1 = ColumnTransformer(transformers=[('cols',
                                                         OneHotEncoder(sparse_output=False,drop='first
                                                          ,[1,2,5,6,8,9])],remainder='passthrough')
           step2 = MinMaxScaler()
           step3 = LogisticRegression()
           step4 = RandomForestClassifier()
           step5 = BernoulliNB()
           step6 = XGBClassifier(objective='binary:logistic', use_label_encoder=False, eval_metr
           step7 = SVC()
In [190]:
           pipe1 = Pipeline([
                ('step1', step1),
                ('step2', step2),
```

('step3', step3)

])

```
In [191]:
          pipe2 = Pipeline([
               ('step1', step1),
               ('step2', step2),
               ('step4', step4)
           ])
In [192]: pipe3 = Pipeline([
               ('step1', step1),
               ('step2',step2),
               ('step5',step5)
           ])
In [193]: |pipe4 = Pipeline([
               ('step1', step1),
               ('step2', step2),
               ('step6', step6)
           ])
In [194]: |pipe5 = Pipeline([
               ('step1', step1),
               ('step2', step2),
               ('step7', step7)
           ])
In [195]: pipe1.fit(X_train,y_train)
Out[195]:
                          Pipeline
                 step1: ColumnTransformer
                    cols
                                  remainder
              OneHotEncoder
                                passthrough
                      ▶ MinMaxScaler
                   ▶ LogisticRegression
In [196]:
          pipe2.fit(X_train,y_train)
Out[196]:
                          Pipeline
                 step1: ColumnTransformer
                    cols
                                  remainder
              ▶ OneHotEncoder
                                ▶ passthrough
                      MinMaxScaler
                 RandomForestClassifier
```

## Modelling and Evaluation - Jupyter Notebook pipe3.fit(X\_train,y\_train) In [197]: Out[197]: Pipeline step1: ColumnTransformer cols remainder OneHot Encoder passthrough

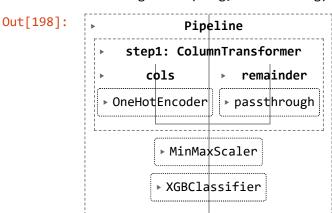
## In [198]: pipe4.fit(X\_train,y\_train)

C:\Users\Mangukiya Ansh\anaconda3\Lib\site-packages\xgboost\core.py:158: UserWarnin g: [17:44:19] WARNING: C:\buildkite-agent\builds\buildkite-windows-cpu-autoscaling-g roup-i-0015a694724fa8361-1\xgboost\xgboost-ci-windows\src\learner.cc:740: Parameters: { "use\_label\_encoder" } are not used.

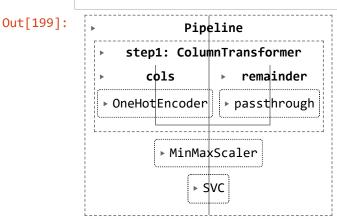
warnings.warn(smsg, UserWarning)

MinMaxScaler

▶ BerndulliNB



In [199]: pipe5.fit(X\_train,y\_train)



```
In [200]: pipe1.score(X_test,y_test)
Out[200]: 0.49
In [201]: pipe2.score(X_test,y_test)
Out[201]: 0.495
In [202]: pipe3.score(X_test,y_test)
Out[202]: 0.51
In [203]: pipe4.score(X_test,y_test)
Out[203]: 0.51
In [204]: pipe5.score(X_test,y_test)
Out[204]: 0.485
In [206]: y_pred = pipe3.predict(X_test)
In [209]: print(classification_report(y_test,y_pred))
                         precision
                                      recall f1-score
                                                         support
                      0
                                        0.40
                              0.54
                                                  0.46
                                                             104
                      1
                              0.49
                                        0.62
                                                  0.55
                                                              96
                                                  0.51
                                                             200
              accuracy
                              0.52
                                        0.51
                                                  0.51
                                                             200
             macro avg
          weighted avg
                              0.52
                                        0.51
                                                  0.50
                                                             200
In [210]: metrics = confusion_matrix(y_pred,y_test)
```

In [213]: import seaborn as sns
sns.heatmap(metrics)

Out[213]: <Axes: >

