Customer Purchase Prediction & Effect of Micro-Numerosity

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In [1]:
         ▶ # Step 1 : import library
            import pandas as pd
In [2]:
         ₩ # Step 2 : import data
            purchase = pd.read csv('https://github.com/YBIFoundation/Dataset/raw/ma
In [3]:
         purchase.head()
   Out[3]:
               Customer ID Age Gender Education
                                               Review Purchased
             0
                     1021
                           30 Female
                                                             No
                                         School Average
             1
                     1022
                           68 Female
                                           UG
                                                  Poor
                                                             No
             2
                                           PG
                     1023
                           70 Female
                                                 Good
                                                             No
                                           PG
             3
                     1024
                           72 Female
                                                 Good
                                                             No
                     1025
                           16 Female
                                           UG Average
                                                             No
In [4]:
         purchase.info()
            <class 'pandas.core.frame.DataFrame'>
            RangeIndex: 50 entries, 0 to 49
            Data columns (total 6 columns):
             #
                 Column
                              Non-Null Count Dtype
            ---
             0
                 Customer ID 50 non-null
                                              int64
             1
                              50 non-null
                                              int64
                 Age
                 Gender
             2
                              50 non-null
                                              object
             3
                 Education
                              50 non-null
                                              object
                                              object
             4
                 Review
                              50 non-null
             5
                              50 non-null
                                              object
                 Purchased
            dtypes: int64(2), object(4)
            memory usage: 2.5+ KB
```

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▶ purchase.describe()
In [5]:
    Out[5]:
                    Customer ID
                                     Age
                       50.00000 50.000000
              count
                      1045.50000 54.160000
               mean
                       14.57738 25.658161
                std
                      1021.00000 15.000000
                min
               25%
                      1033.25000 30.250000
               50%
                     1045.50000 57.000000
                     1057.75000 74.000000
               75%
                     1070.00000 98.000000
               max
          ▶ # Step 3 : define target (y) and features (X)
In [6]:
In [7]:
          purchase.columns
    Out[7]: Index(['Customer ID', 'Age', 'Gender', 'Education', 'Review', 'Purchas
             ed'], dtype='object')
In [8]:
          y = purchase['Purchased']

X = purchase.drop(['Purchased', 'Customer ID'], axis=1)

In [9]:
In [10]:
           ▶ # encoding categorical variable
             X.replace({'Review':{'Poor':0,'Average':1,'Good':2}},inplace=True)
             X.replace({'Education':{'School':0,'UG':1,'PG':2}},inplace=True)
             X.replace({'Gender':{'Male': 0, 'Female':1}}, inplace=True)
           # display first 5 rows
In [11]:
             X.head()
   Out[11]:
                 Age Gender Education Review
              0
                  30
                                    0
                           1
              1
                  68
                          1
                                    1
                                           0
              2
                  70
                                    2
                                           2
                          1
                                    2
                                           2
              3
                  72
                          1
```

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```
from sklearn.model_selection import train_test_split
           X_train, X_test, y_train, y_test = train_test_split(X,y, train_size=0.8
In [13]: ▶ # check shape of train and test sample
           X_train.shape, X_test.shape, y_train.shape, y_test.shape
   Out[13]: ((40, 4), (10, 4), (40,), (10,))
         # Step 5 : select model
In [14]:
           from sklearn.ensemble import RandomForestClassifier
           model = RandomForestClassifier()
         # Step 6 : train or fit model
In [15]:
           model.fit(X train,y train)
   Out[15]: RandomForestClassifier()
In [16]:
         # Step 7 : predict model
           y_pred = model.predict(X_test)
Out[17]: array(['No', 'Yes', 'No', 'Yes', 'Yes', 'Yes', 'No', 'Ye
           s'],
                 dtype=object)
In [18]: 

# Step 8 : model accuracy
           from sklearn.metrics import confusion_matrix, accuracy_score, classific
Out[19]: array([[2, 1],
                  [3, 4]], dtype=int64)

    accuracy_score(y_test,y_pred)

In [20]:
   Out[20]: 0.6
In [22]:
         print(classification_report(y_test,y_pred))
                                   recall f1-score
                        precision
                                                    support
                            0.40
                                     0.67
                                              0.50
                                                          3
                    No
                   Yes
                            0.80
                                     0.57
                                              0.67
                                                          7
                                              0.60
                                                         10
               accuracy
                                              0.58
                            0.60
                                     0.62
                                                         10
              macro avg
           weighted avg
                            0.68
                                     0.60
                                              0.62
                                                         10
```

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