```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
from sklearn.preprocessing import OneHotEncoder
import pickle
from sklearn.neural_network import MLPClassifier
from sklearn.metrics import classification_report, accuracy_score
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
url = 'https://archive.ics.uci.edu/ml/machine-learning-databases/molecular-biology/promote
names = ['Class', 'id', 'Sequence']
data = pd.read_csv(url, names = names)
```

data.head(10)

from google.colab import drive
drive.mount('/content/drive')

Sequence	id	Class		
\t\ttactagcaatacgcttgcgttcggtggttaagtatgtat	S10	+	0	
\t\ttgctatcctgacagttgtcacgctgattggtgtcgttacaat	AMPC	+	1	
\t\tgtactagagaactagtgcattagcttatttttttgttatcat	AROH	+	2	
\taattgtgatgtgtatcgaagtgtgttgcggagtagatgttagaa	DEOP2	+	3	
\ttcgataattaactattgacgaaaagctgaaaaccactagaatgc	LEU1_TRNA	+	4	
\taggggcaaggaggatggaaagaggttgccgtataaagaaactag	MALEFG	+	5	
\t\tcaggggtggaggatttaagccatctcctgatgacgcatagt	MALK	+	6	
\t\tttctacaaaacacttgatactgtatgagcatacagtataat	RECA	+	7	
\t\tcgacttaatatactgcgacaggacgtccgttctgtgtaaatc	RPOB	+	8	
\ttttaaatttcctcttgtcaggccggaataactccctataatgc	RRNAB_P1	+	9	

data.shape

(106, 3)

data.dtypes

Class object id object Sequence object dtype: object

```
# Refining and structuring the data
# Build our dataset using custom pandas dataframe
classes = data.loc[:,'Class']
classes.head()
print()
print(classes.value_counts())
         53
    +
         53
    Name: Class, dtype: int64
# generate list of DNA sequence
sequence = list(data.loc[:, 'Sequence'])
sequence[-1]
     #Remove tab from each sequence
dic = \{\}
for i, seq in enumerate(sequence):
   nucleotides = list(seq)
   nucleotides = [char for char in nucleotides if char != '\t']
   #append class assignment
   nucleotides.append(classes[i])
   dic[i] = nucleotides
list(dic[0])
    ['t',
      'a',
      't',
      'a',
      't',
      'a',
```

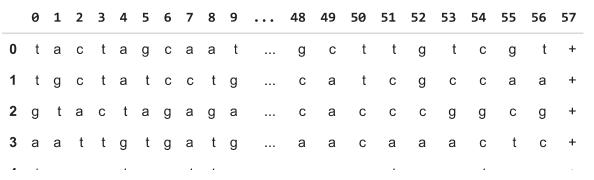
```
g ,
't',
't',
'a',
't',
'a',
't',
't',
'a',
't',
'a',
'a',
't',
'c',
'c',
'g',
'c',
't',
'g',
't',
't',
'+']
```

Convert Dict object into dataframe
df = pd.DataFrame(dic)
df.head()

	0	1	2	3	4	5	6	7	8	9	• • •	96	97	98	99	100	101	102	103	104	105
0	t	t	g	а	t	а	С	t	С	t		С	С	t	а	g	С	g	С	С	t
1	а	g	t	а	С	g	а	t	g	t		С	g	а	g	а	С	t	g	t	а
2	С	С	а	t	g	g	g	t	а	t		g	С	t	а	g	t	а	С	С	а
3	t	t	С	t	а	g	g	С	С	t		а	t	g	g	а	С	t	g	g	С
4	а	а	t	g	t	g	g	t	t	а		g	а	а	g	g	а	t	а	t	а

5 rows × 106 columns

```
# transpose dataframe into correct format
df = df.transpose()
df.head()
```



Rename the 57th column as it is our classes
df.rename(columns = {57:'Class'}, inplace = True)

df.head()

	0	1	2	3	4	5	6	7	8	9	• • •	48	49	50	51	52	53	54	55	56	Class
0	t	а	С	t	а	g	С	а	а	t		g	С	t	t	g	t	С	g	t	+
1	t	g	С	t	а	t	С	С	t	g		С	а	t	С	g	С	С	а	а	+
2	g	t	а	С	t	а	g	а	g	а		С	а	С	С	С	g	g	С	g	+
3	а	а	t	t	g	t	g	а	t	g		а	а	С	а	а	а	С	t	С	+
4	t	С	g	а	t	а	а	t	t	а		С	С	g	t	g	g	t	а	g	+

5 rows × 58 columns

```
temp = df.copy(deep=True)
temp = temp.drop(['Class'], axis = 1)
```

temp.head()

	0	1	2	3	4	5	6	7	8	9	• • •	47	48	49	50	51	52	53	54	55	56
0	t	а	С	t	а	g	С	а	а	t		g	g	С	t	t	g	t	С	g	t
1	t	g	С	t	а	t	С	С	t	g		g	С	а	t	С	g	С	С	а	а
2	g	t	а	С	t	а	g	а	g	а		С	С	а	С	С	С	g	g	С	g
3	а	а	t	t	g	t	g	а	t	g		t	а	а	С	а	а	а	С	t	С
4	t	С	g	а	t	а	а	t	t	а		t	С	С	g	t	g	g	t	а	g

5 rows × 57 columns

```
# Encoding using one-hot encoder:
```

```
enc = OneHotEncoder(handle_unknown='ignore')
enc.fit(temp)
print(enc.categories_)
df1 = enc.transform(temp).toarray()
del temp
# df1[1:3]
```

```
# Saving the one-hot encoder
with open("drive/MyDrive/EColi-encoder.pickle", "wb") as f:
    pickle.dump(enc, f)
# Loading the file later:
# encoder = pickle.load(f)
# data = encoder.transform(df).toarray()
df new = pd.DataFrame(df1)
df_new.head()
                 1
                          3
                               4
                                    5
                                         6
                                              7
                                                   8
                                                                  218
                                                                       219
                                                                             220
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                                                                                        222
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     5 rows × 228 columns
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# Fixing the classes column:
df["Class"] = df["Class"].replace(to_replace =["+"], value =1)
df["Class"] = df["Class"].replace(to_replace =["-"], value =0)
df_new["Classes"] = df['Class']
df_new.head()
                          3
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                                              7
                                                   8
                                                                                  222
                                                                                        223
           0
                 1
                     2
                               4
                                         6
                                                        9
                                                                 219
                                                                       220
                                                                            221
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      5 rows × 229 columns
                                                                                                - ◀
#Encoding - Alternative
numerical_df = pd.get_dummies(df)
```

numerical_df.head()

[array(['a', 'c', 'g', 't'], dtype=object), array(['a', 'c', 'g', 't'], dtype=object)

	Class	0_a	0_c	0_g	0_t	1_a	1_c	1_g	1_t	2_a	• • •	54_g	54_t	55_a	55_c
0	1	0	0	0	1	1	0	0	0	0		0	0	0	0
1	1	0	0	0	1	0	0	1	0	0		0	0	1	0
2	1	0	0	1	0	0	0	0	1	1		1	0	0	1
3	1	1	0	0	0	1	0	0	0	0		0	0	0	0
4	1	0	0	0	1	0	1	0	0	0		0	1	1	0
5 rc	5 rows × 229 columns														

Training and Testing the Classification Algorithms

```
y = df_new['Classes'].values# numerical_df['Class'].values
X = df_new.drop(['Classes'], axis = 1).values# numerical_df.drop(['Class'], axis = 1).valu
#define a seed for reproducibility
seed = 1
# Splitting data into training and testing data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25, random_state =
model = MLPClassifier(hidden_layer_sizes=(150,100,50), max_iter=300,activation = 'relu',so
model.fit(X_train, y_train)
print(model.score(X_train, y_train))
1.0
```

#Predicting y for X_val
y_pred = model.predict(X_test)
model.score(X_test, y_test)

0.9259259259259

Model evaluation
print(classification_report(y_test, y_pred))

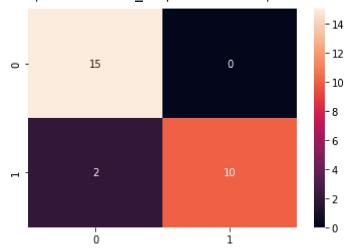
	precision	recall	f1-score	support
0	1.00	0.88	0.94	17
1	0.83	1.00	0.91	10
accuracy macro avg weighted avg	0.92 0.94	0.94 0.93	0.93 0.92 0.93	27 27 27

[#]Importing Confusion Matrix

[#]Comparing the predictions against the actual observations in y_val

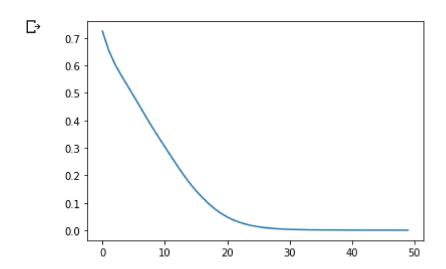
```
cm = confusion_matrix(y_pred, y_test)
sns.heatmap(cm, annot=True)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f60b2bf0990>



Plotting graph for MLPClassifier

```
loss_values = model.loss_curve_
plt.plot(loss_values)
plt.show()
```



save the model to disk

```
filename = 'drive/MyDrive/E-Coli_model.pickle'
pickle.dump(model, open(filename, 'wb'))
```

```
0 1 2 3 4 5 6 7 8 9 ...
                                      48 49
                                             50
                                                51 52
                                                        53 54 55
                                                                  56 Class
                                                                    t
                                                                          1
       tactagcaa
                                       g
                                           С
                                                  t
                                                     g
                                                         t
                                                            С
                                                                g
                                                                          1
        tgctatcc
                                       С
                                          а
                                              t
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                                                         С
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                                                                а
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    5 rows × 58 columns
encoder = pickle.load(open("drive/MyDrive/EColi-encoder.pickle", 'rb'))
data_test = encoder.transform(df_test).toarray()
print(model.predict(data_test))
    [1]
type(model.predict(data_test)[0])
    numpy.int64
# load the model from disk
# loaded_model = pickle.load(open(filename, 'rb'))
# result = loaded_model.score(X_test, Y_test)
```