

In [ ]:

```
# Run this cell to mount your drive to this notebook in order to read the datasets
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

In [ ]:

```
import pandas as pd
import numpy as np

import warnings
warnings.filterwarnings("ignore")
```

## Read Dataset

In [ ]:

```
# Put the folder path where the datasets are located
PATH = "drive/My Drive/Colab Notebooks/Project #1/"
```

In [ ]:

```
# Read the train and test set with read_csv() method of pandas
train = pd.read_csv(PATH + "train.csv")
test = pd.read_csv(PATH + "test.csv")
```

In [ ]:

```
# Display the training dataframe
train.head(5)
```

Out[ ]:

Unnamed: 0		text	label
0	0	I came here and left a review before but last ...	1
1	1	Had a very nice first visit here. The owner Te...	4
2	2	This is a gorgeous and very clean hotel. We h...	4
3	3	The gym is dirty. I have given up. Locker ro...	1
4	4	The food here is delicious, fast, and consiste...	5

## Preprocess Dataset

In [ ]:

```
import re
import nltk
nltk.download('stopwords')
nltk.download('punkt')
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.tokenize import word_tokenize
from nltk.tokenize.treebank import TreebankWordDetokenizer
```

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data] Package punkt is already up-to-date!
```

In [ ]:

```
# Define a function to perform preprocessing. This function can perform things like lower casing, stemming, removing stopwords, etc.
```

```
def preprocess(text: str):

    text = text.lower()
    stop_words = set(stopwords.words("english"))
    stop_words.add('\n\t')
    stop_words.add('\m')
    stop_words.add('\s')
    stop_words.add('\ve')
    stop_words.add('\ll')
    stop_words.add('\re')
    word_tokens = word_tokenize(text)

    stemmer = PorterStemmer()

    # remove the stopwords and stem the words by using PorterStemmer
    clean_text = []
    for word in word_tokens:
        if word not in stop_words:
            if len(word) > 1:
                clean_word = stemmer.stem(word)
                clean_text.append(clean_word)

    return " ".join(clean_text)
```

In [ ]:

```
# Apply your preprocessing function to your text fields.
```

```
train.text = train.text.apply(preprocess)
test.text = test.text.apply(preprocess)
```

```
train.shape, test.shape
```

Out[ ]:

```
((18000, 3), (2000, 3))
```

In [ ]:

```
train.head(5)
```

Out[ ]:

Unnamed: 0		text	label
0	0	came left review last time get food poison unl...	1
1	1	nice first visit owner ted friendli start rest...	4
2	2	gorgeou clean hotel room west wing first chore...	4
3	3	gym dirti given locker room total dirti manag ...	1
4	4	food delici fast consist everi singl time gene...	5

In [ ]:

```
# Create your binary and multiclass datasets
```

```
# For binary dataset, get rid of the class 3 in the dataset and map class 1 and 2 to 0, and class 4 and 5 to 1
```

```
train_binary = train
test_binary = test
```

```
# Function for mapping values; 1 & 2 --> 0, 4 & 5 --> 1
target_mapping_binary = {1: 0,
                          2: 0,
```

```

4: 1,
5: 1}

# Turn train data into binary format
train_binary = train_binary.drop(train_binary[train_binary.label == 3].index)
train_binary["label"] = train_binary["label"].map(lambda x: target_mapping_binary[x])

# Turn train data into binary format
test_binary = test_binary.drop(test_binary[test_binary.label == 3].index)
test_binary["label"] = test_binary["label"].map(lambda x: target_mapping_binary[x])

# -----
# For multiclass dataset, make sure your classes starts from 0 and goes until 4. (5->4, 4
->3, 3->2, 2->1, 1->0)
train_multi = train
test_multi = test

# Function for mapping values; 5 --> 4, 4 --> 3, 3 --> 2, 2 --> 1, 1 --> 0
target_mapping_multi = {1: 0,
                        2: 1,
                        3: 2,
                        4: 3,
                        5: 4}

# Turn train data into multi class format
train_multi["label"] = train_multi["label"].map(lambda x: target_mapping_multi[x])

# Turn test data into multi class format
test_multi["label"] = test_multi["label"].map(lambda x: target_mapping_multi[x])

# -----
# Print binary and multiclass train and test dataframes
train_binary.head(5), test_binary.head(5), train_multi.head(5), test_multi.head(5)

```

Out [ ]:

```

(   Unnamed: 0      text  label
0         0  came left review last time get food poison unl...    0
1         1  nice first visit owner ted friendli start rest...    1
2         2  gorgeou clean hotel room west wing first chore...    1
3         3  gym dirti given locker room total dirti manag ...    0
4         4  food delici fast consist everi singl time gene...    1,
   Unnamed: 0      text  label
0         0  stay weekend made stay pleasant locat great sp...    1
1         1  forev call upon delici design whenev need tast...    1
2         2  person order homicid boneless regular absolut ...    1
3         3  eat pretti much everytim go tarpon spring staf...    1
5         5  great hidden tavern grill wonder old place ear...    1,
   Unnamed: 0      text  label
0         0  came left review last time get food poison unl...    0
1         1  nice first visit owner ted friendli start rest...    3
2         2  gorgeou clean hotel room west wing first chore...    3
3         3  gym dirti given locker room total dirti manag ...    0
4         4  food delici fast consist everi singl time gene...    4,
   Unnamed: 0      text  label
0         0  stay weekend made stay pleasant locat great sp...    4
1         1  forev call upon delici design whenev need tast...    4
2         2  person order homicid boneless regular absolut ...    3
3         3  eat pretti much everytim go tarpon spring staf...    3
4         4  time never go want huge chang -- usual trim ge...    2)

```

In [ ]:

```

# Train Binary
train_binary_x = train_binary["text"]
train_binary_y = train_binary["label"]

# Train Multiclass
train_multi_x = train_multi["text"]

```

```
train_multi_y = train_multi["label"]

# -----

# Test Binary
test_multi_x = test_multi["text"]
test_multi_y = test_multi["label"]

# Test Multiclass
test_binary_x = test_binary["text"]
test_binary_y = test_binary["label"]
```

# Models

## Non-Neural Models

In [ ]:

```
from sklearn.model_selection import GridSearchCV
from sklearn.base import TransformerMixin
from sklearn.pipeline import Pipeline
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.naive_bayes import GaussianNB
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import f1_score, confusion_matrix, accuracy_score
```

## Naive Bayes

### Initializing the Pipeline and GridSearchCV Parameters

In [ ]:

```
from numpy.lib.function_base import average
# https://scikit-learn.org/stable/modules/generated/sklearn.naive\_bayes.GaussianNB.html

# Create a class for converting sparse matrix output of TfidfVectorizer to dense matrix
# for feeding into GaussianNB

class DenseTransformer(TransformerMixin):

    def fit(self, X, y=None, **fit_params):
        return self

    def transform(self, X, y=None, **fit_params):
        return X.todense()

# Initiate the pipeline with required components. You can use Pipeline class of sklearn ->
# https://scikit-learn.org/stable/modules/generated/sklearn.pipeline.Pipeline.html
# There will be three components; 1) TfidfVectorizer 2) DenseTransformer 3) Naive Bayes c
# lassifier.

NB_pipeline = Pipeline([
    ('vectorizer_tfidf', TfidfVectorizer()),
    ('to_dense', DenseTransformer()),
    ('classifier_NB', GaussianNB())
])

# Set the hyperparameter space that will be scanned with GridSearchCV.

grid_params_NB = {
    'vectorizer_tfidf__ngram_range': [(1,1), (1,2), (1,3)],
    'vectorizer_tfidf__min_df': [100, 500, 1000]
```

```
}
```

## Binary

### Train GridSearchCV with Binary Classification Dataset

In [ ]:

```
%%time
# Train Binary Classification
# Initialize and run the GridSearchCV to scan the hyperparameter and find the best hyperp
arameter set that will maximize the scoring option for binary classification.
grid_search_NB_Binary = GridSearchCV(
    NB_pipeline,
    grid_params_NB,
    cv=5,
    verbose=1,
    scoring='f1_macro')

# Fit binary dataset to the grid_search_NB
grid_search_NB_Binary.fit(train_binary_x, train_binary_y)
print(grid_search_NB_Binary.cv_results_)
```

```
Fitting 5 folds for each of 9 candidates, totalling 45 fits
{'mean_fit_time': array([1.09800296, 1.75072818, 3.05007677, 0.54481544, 1.73659186,
    2.93997054, 0.5206573 , 1.6194633 , 2.89493666]), 'std_fit_time': array([0.5615949
    2, 0.01766732, 0.0280571 , 0.01164694, 0.16291979,
    0.04022072, 0.0072968 , 0.02007778, 0.01669117]), 'mean_score_time': array([0.2059
    7191, 0.29165244, 0.41788597, 0.12075896, 0.24511776,
    0.37006927, 0.11626983, 0.23530316, 0.36091919]), 'std_score_time': array([0.05365
    202, 0.00945326, 0.00629129, 0.00287836, 0.00649322,
    0.00543085, 0.00482167, 0.0047043 , 0.01068762]), 'param_vectorizer_tfidf__min_df'
: masked_array(data=[100, 100, 100, 500, 500, 500, 1000, 1000, 1000],
    mask=[False, False, False, False, False, False, False, False, False,
    False],
    fill_value='?',
    dtype=object), 'param_vectorizer_tfidf__ngram_range': masked_array(data=[(1,
    1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3), (1, 1),
    (1, 2), (1, 3)],
    mask=[False, False, False, False, False, False, False, False, False,
    False],
    fill_value='?',
    dtype=object), 'params': [{'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf
__ngram_range': (1, 1)}, {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range
': (1, 2)}, {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'
vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'vectorizer_t
fidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'vectorizer_tfidf__min_df'
: 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'vectorizer_tfidf__min_df': 1000, 'vect
orizer_tfidf__ngram_range': (1, 1)}, {'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf
__ngram_range': (1, 2)}, {'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_rang
e': (1, 3)}], 'split0_test_score': array([0.85461441, 0.86293644, 0.86293644, 0.81748355,
    0.81921786,
    0.81921786, 0.7744612 , 0.7744612 , 0.7744612 ]), 'split1_test_score': array([0.84
    628728, 0.84975492, 0.84975492, 0.81259754, 0.81052186,
    0.81052186, 0.7680797 , 0.7680797 , 0.7680797 ]), 'split2_test_score': array([0.86
    011795, 0.871912 , 0.87052052, 0.81633638, 0.81529274,
    0.81529274, 0.76741493, 0.76741493, 0.76741493]), 'split3_test_score': array([0.85
    907466, 0.85941704, 0.85941704, 0.80735624, 0.80701032,
    0.80701032, 0.7726474 , 0.7726474 , 0.7726474 ]), 'split4_test_score': array([0.85
    109272, 0.85629905, 0.85837498, 0.8114798 , 0.81043622,
    0.81043622, 0.75586703, 0.75586703, 0.75586703]), 'mean_test_score': array([0.8542
    374 , 0.86006389, 0.86020078, 0.8130507 , 0.8124958 ,
    0.8124958 , 0.76769405, 0.76769405, 0.76769405]), 'std_test_score': array([0.00512
    28 , 0.00734162, 0.00673861, 0.00362066, 0.00427172,
    0.00427172, 0.00648745, 0.00648745, 0.00648745]), 'rank_test_score': array([3, 2,
    1, 4, 5, 5, 7, 7, 7], dtype=int32)}
CPU times: user 1min 34s, sys: 2.08 s, total: 1min 36s
Wall time: 1min 38s
```

## Report the Results

In [ ]:

```
# Report the standart deviation of split scores for each hyperparameter group.
# Report the mean and standart deviation of scores for each parameter group.
for i in range(9):
    print("Parameter Group ", i+1)
    print("Parameters: ", grid_search_NB_Binary.cv_results_['params'][i])
    print("Mean Test Score: ", grid_search_NB_Binary.cv_results_['mean_test_score'][i])
    print("Standard Deviation Test Score: ", grid_search_NB_Binary.cv_results_['std_test_score'][i])
    print("Min: ", min(grid_search_NB_Binary.cv_results_["split0_test_score"][i], grid_search_NB_Binary.cv_results_["split1_test_score"][i],
                      grid_search_NB_Binary.cv_results_["split2_test_score"][i], grid_search_NB_Binary.cv_results_["split3_test_score"][i],
                      grid_search_NB_Binary.cv_results_["split4_test_score"][i]))
    print("Max: ", max(grid_search_NB_Binary.cv_results_["split0_test_score"][i], grid_search_NB_Binary.cv_results_["split1_test_score"][i],
                      grid_search_NB_Binary.cv_results_["split2_test_score"][i], grid_search_NB_Binary.cv_results_["split3_test_score"][i],
                      grid_search_NB_Binary.cv_results_["split4_test_score"][i]))
    print("\n")

# Show the best parameter set for given dataset and hyperparameter space.
print("Best Parameters: ", grid_search_NB_Binary.best_params_)
```

```
Parameter Group 1
Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}
Mean Test Score: 0.8542374043833949
Standard Deviation Test Score: 0.005122799448803231
Min: 0.8462872819793779
Max: 0.8601179471665417
```

```
Parameter Group 2
Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}
Mean Test Score: 0.8600638910720917
Standard Deviation Test Score: 0.007341615443068069
Min: 0.8497549243583071
Max: 0.8719120033225389
```

```
Parameter Group 3
Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
Mean Test Score: 0.860200781216163
Standard Deviation Test Score: 0.0067386139688290665
Min: 0.8497549243583071
Max: 0.8705205202491479
```

```
Parameter Group 4
Parameters: {'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}
Mean Test Score: 0.8130507007904967
Standard Deviation Test Score: 0.0036206581525370014
Min: 0.8073562350927301
Max: 0.8174835487027747
```

```
Parameter Group 5
Parameters: {'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}
Mean Test Score: 0.812495799919392
Standard Deviation Test Score: 0.0042717176005138404
Min: 0.807010315041532
Max: 0.8192178647456403
```

```
Parameter Group 6
Parameters: {'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}
Mean Test Score: 0.812495799919392
```

```
Mean Test Score: 0.7676940517256192
Standard Deviation Test Score: 0.0042717176005138404
Min: 0.807010315041532
Max: 0.8192178647456403
```

Parameter Group 7

```
Parameters: {'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}
Mean Test Score: 0.7676940517256192
Standard Deviation Test Score: 0.006487450690960797
Min: 0.7558670285249693
Max: 0.7744612014994665
```

Parameter Group 8

```
Parameters: {'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}
Mean Test Score: 0.7676940517256192
Standard Deviation Test Score: 0.006487450690960797
Min: 0.7558670285249693
Max: 0.7744612014994665
```

Parameter Group 9

```
Parameters: {'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}
Mean Test Score: 0.7676940517256192
Standard Deviation Test Score: 0.006487450690960797
Min: 0.7558670285249693
Max: 0.7744612014994665
```

```
Best Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
```

### ***Setting Pipeline with Best Parameters and Making Prediction***

In [ ]:

```
# Building the pipeline with the best parameter group and reporting Conf. Mat. and Results on the Test Set
# Create your Pipeline object with the best parameter set.
NB_Binary_BestParams_Pipeline = Pipeline([
    ('vectorizer_tfidf', TfidfVectorizer(ngram_range=(1,2), min_df=100)),
    ('to_dense', DenseTransformer()),
    ('classifier_NB', GaussianNB())
])

# Fit your pipeline on training set.
BestParams_NB_Binary_Model = NB_Binary_BestParams_Pipeline.fit(train_binary_x, train_binary_y)

# Make prediction
y_Predict_NB_Binary = BestParams_NB_Binary_Model.predict(test_binary_x)
```

### ***Reporting the Scores and Confusion Matrix for Binary Classification***

In [ ]:

```
# Report the F1 and Accuracy and Confusion Matrix scores for binary classification
Accuracy_NB_Binary_Test_BestParams = accuracy_score(test_binary_y, y_Predict_NB_Binary)
F1_NB_Binary_Test_BestParams = f1_score(test_binary_y, y_Predict_NB_Binary, average='macro')

print('F1 Macro for Binary Classification with Best Parameters:\n', F1_NB_Binary_Test_BestParams)
print('\nAccuracy Score for Binary Classification with Best Parameters:\n', Accuracy_NB_Binary_Test_BestParams)
print('\nConfusion Matrix:')

```

```
confusion_matrix(test_binary_y, y_Predict_NB_Binary)
```

F1 Macro for Binary Classification with Best Parameters:  
0.8687712178561673

Accuracy Score for Binary Classification with Best Parameters:  
0.8688010043942247

Confusion Matrix:

Out[ ]:

```
array([[680, 93],
       [116, 704]])
```

## Multi

### Train GridSearchCV with Multiclass Classification Dataset

In [ ]:

```
%%time
# Train Multiclass Classification
# Initialize and run the GridSearchCV to scan the hyperparameter and find the best hyperp
arameter set that will maximize the scoring option for multiclass classification.
grid_search_NB_Multi = GridSearchCV(
    NB_pipeline,
    grid_params_NB,
    cv=5,
    verbose=1,
    scoring='f1_macro')

# Fit multiclass dataset to the grid_search_NB
grid_search_NB_Multi.fit(train_multi_x, train_multi_y)
print(grid_search_NB_Multi.cv_results_)
```

```
Fitting 5 folds for each of 9 candidates, totalling 45 fits
{'mean_fit_time': array([0.8925344 , 2.20540662, 4.24301386, 0.68350182, 2.00145941,
        3.86936245, 0.65147934, 2.23390021, 3.70643892]), 'std_fit_time': array([0.0376216
8, 0.02255642, 0.7811531 , 0.00725528, 0.03718182,
        0.31773042, 0.01558628, 0.48130737, 0.06185982]), 'mean_score_time': array([0.3099
1936, 0.47574139, 0.6322103 , 0.17763729, 0.33265028,
        0.49877515, 0.15736046, 0.32141318, 0.48121061]), 'std_score_time': array([0.00729
149, 0.01026245, 0.01448725, 0.00877463, 0.01173344,
        0.02406921, 0.0065055 , 0.01280232, 0.0181068 ]), 'param_vectorizer_tfidf__min_df'
: masked_array(data=[100, 100, 100, 500, 500, 500, 1000, 1000, 1000],
        mask=[False, False, False, False, False, False, False, False, False],
        fill_value='?',
        dtype=object), 'param_vectorizer_tfidf__ngram_range': masked_array(data=[(1,
1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3), (1, 1),
        (1, 2), (1, 3)],
        mask=[False, False, False, False, False, False, False, False, False],
        fill_value='?',
        dtype=object), 'params': [{'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf
__ngram_range': (1, 1)}, {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range
': (1, 2)}, {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}, {
'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'vectorizer_t
fidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'vectorizer_tfidf__min_df'
: 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'vectorizer_tfidf__min_df': 1000, 'vect
orizer_tfidf__ngram_range': (1, 1)}, {'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf
__ngram_range': (1, 2)}, {'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_rang
e': (1, 3)}], 'split0_test_score': array([0.45021319, 0.46183375, 0.46271782, 0.46003148,
0.46343308,
        0.46343308, 0.42595559, 0.42595559, 0.42595559]), 'split1_test_score': array([0.44
855055, 0.45256875, 0.45488544, 0.45125412, 0.4505941 ,
        0.4505941 , 0.42370347, 0.42370347, 0.42370347]), 'split2_test_score': array([0.44
647 , 0.46691843, 0.46784517, 0.4624363 , 0.46055602,
        0.46055602, 0.42002265, 0.42002265, 0.42002265]), 'split3_test_score': array([0.42
595559, 0.42595559, 0.42595559, 0.42595559, 0.42595559,
        0.42595559, 0.42595559, 0.42595559, 0.42595559])}
```



```

0.44033802, 0.42092203, 0.42092203, 0.42092203]], 'split3_test_score': array([0.44
053961, 0.44441133, 0.44392444, 0.43499259, 0.43469922,
0.43469922, 0.42012134, 0.42012134, 0.42012134]), 'split4_test_score': array([0.44
673308, 0.4548611 , 0.45428664, 0.4518841 , 0.45617818,
0.45617818, 0.41349404, 0.41349404, 0.41349404]), 'mean_test_score': array([0.4445
0129, 0.45611867, 0.4567319 , 0.45211972, 0.45309212,
0.45309212, 0.42083942, 0.42083942, 0.42083942]), 'std_test_score': array([0.00711
066, 0.00775296, 0.00815789, 0.00962467, 0.01016571,
0.01016571, 0.00421557, 0.00421557, 0.00421557]), 'rank_test_score': array([6, 2,
1, 5, 3, 3, 7, 7, 7], dtype=int32)}
CPU times: user 2min 1s, sys: 3.49 s, total: 2min 4s
Wall time: 2min 6s

```

## Report the Results

In [ ]:

```

# Report the standart deviation of split scores for each hyperparameter group.
# Report the max, min, mean, and standard deviation of scores for each parameter group.
for i in range(9):
    print("Parameter Group ", i+1)
    print("Parameters: ", grid_search_NB_Multi.cv_results_['params'][i])
    print("Mean Test Score: ", grid_search_NB_Multi.cv_results_['mean_test_score'][i])
    print("Standard Deviation Test Score: ", grid_search_NB_Multi.cv_results_['std_test_sco
re'][i])
    print("Min: ", min(grid_search_NB_Multi.cv_results_["split0_test_score"][i], grid_sear
ch_NB_Multi.cv_results_["split1_test_score"][i],
                        grid_search_NB_Multi.cv_results_["split2_test_score"][i], grid_sear
ch_NB_Multi.cv_results_["split3_test_score"][i],
                        grid_search_NB_Multi.cv_results_["split4_test_score"][i]))
    print("Max: ", max(grid_search_NB_Multi.cv_results_["split0_test_score"][i], grid_sear
ch_NB_Multi.cv_results_["split1_test_score"][i],
                        grid_search_NB_Multi.cv_results_["split2_test_score"][i], grid_sear
ch_NB_Multi.cv_results_["split3_test_score"][i],
                        grid_search_NB_Multi.cv_results_["split4_test_score"][i]))

    print("\n")

# Show the best parameter set for given dataset and hyperparameter space.
print("Best Parameters: ", grid_search_NB_Multi.best_params_)

```

```

Parameter Group 1
Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}
Mean Test Score: 0.4445012872862179
Standard Deviation Test Score: 0.007110661446049768
Min: 0.4305396111761393
Max: 0.4502131932296366

```

```

Parameter Group 2
Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}
Mean Test Score: 0.4561186691168338
Standard Deviation Test Score: 0.007752963984642025
Min: 0.4444113254577638
Max: 0.4669184291625168

```

```

Parameter Group 3
Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
Mean Test Score: 0.4567319032363632
Standard Deviation Test Score: 0.00815789056184797
Min: 0.4439244395806406
Max: 0.467845173574052

```

```

Parameter Group 4
Parameters: {'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}
Mean Test Score: 0.4521197195064187
Standard Deviation Test Score: 0.009624668608540174
Min: 0.4349925882462452
Max: 0.46243629865916047

```

Parameter Group 5  
Parameters: {'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 2)}  
Mean Test Score: 0.453092120097463  
Standard Deviation Test Score: 0.010165705629250173  
Min: 0.4346992227334148  
Max: 0.46343308487899926

Parameter Group 6  
Parameters: {'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}  
Mean Test Score: 0.453092120097463  
Standard Deviation Test Score: 0.010165705629250173  
Min: 0.4346992227334148  
Max: 0.46343308487899926

Parameter Group 7  
Parameters: {'vectorizer\_tfidf\_\_min\_df': 1000, 'vectorizer\_tfidf\_\_ngram\_range': (1, 1)}  
Mean Test Score: 0.4208394190327584  
Standard Deviation Test Score: 0.004215570211321058  
Min: 0.4134940422065131  
Max: 0.42595559202702066

Parameter Group 8  
Parameters: {'vectorizer\_tfidf\_\_min\_df': 1000, 'vectorizer\_tfidf\_\_ngram\_range': (1, 2)}  
Mean Test Score: 0.4208394190327584  
Standard Deviation Test Score: 0.004215570211321058  
Min: 0.4134940422065131  
Max: 0.42595559202702066

Parameter Group 9  
Parameters: {'vectorizer\_tfidf\_\_min\_df': 1000, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}  
Mean Test Score: 0.4208394190327584  
Standard Deviation Test Score: 0.004215570211321058  
Min: 0.4134940422065131  
Max: 0.42595559202702066

Best Parameters: {'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}

### ***Setting Pipeline with Best Parameters and Making Prediction***

In [ ]:

```
# Building the pipeline with the best parameter group and reporting Conf. Mat. and Results on the Test Set
# Create your Pipeline object with the best parameter set.
NB_Multi_BestParams_Pipeline = Pipeline([
    ('vectorizer_tfidf', TfidfVectorizer(ngram_range=(1,3), min_df=100)),
    ('to_dense', DenseTransformer()),
    ('classifier_NB', GaussianNB())
])

# Fit your pipeline on training set.
BestParams_NB_Multi_Model = NB_Multi_BestParams_Pipeline.fit(train_multi_x, train_multi_y)

# Make prediction
y_Predict_NB_Multi = BestParams_NB_Multi_Model.predict(test_multi_x)
```

### ***Reporting the Scores and Confusion Matrix for Multiclass Classification***

```
In [ ]:
```

```
# Report the F1 and Accuracy scores and Confusion Matrix for multiclass classification
Accuracy_NB_Multi_Test_BestParams = accuracy_score(test_multi_y, y_Predict_NB_Multi)
F1_NB_Multi_Test_BestParams = f1_score(test_multi_y, y_Predict_NB_Multi, average='macro'
)

print('F1 Macro for Multiclass Classification with Best Parameters:\n', F1_NB_Multi_Test_
BestParams)
print('\nAccuracy Score for Multiclass Classification with Best Parameters:\n', Accuracy_
NB_Multi_Test_BestParams)
print('\nConfusion Matrix:')
confusion_matrix(test_multi_y, y_Predict_NB_Multi)
```

```
F1 Macro for Multiclass Classification with Best Parameters:
0.47526559258349055
```

```
Accuracy Score for Multiclass Classification with Best Parameters:
0.4915
```

```
Confusion Matrix:
```

```
Out[ ]:
```

```
array([[264,  91,  18,   5,  15],
       [ 96, 156,  73,  21,  34],
       [ 46,  78, 116,  95,  72],
       [ 20,  32,  60, 133, 134],
       [ 26,  13,  19,  69, 314]])
```

## Logistic Regression

### Initializing the Pipeline and GridSearchCV Parameters

```
In [ ]:
```

```
# https://scikit-learn.org/stable/modules/generated/sklearn.linear\_model.LogisticRegression.html

# Initiate the pipeline with required components.You can use Pipeline class of sklearn ->
https://scikit-learn.org/stable/modules/generated/sklearn.pipeline.Pipeline.html
# There will be two components; 1) Word weighting 2) Logistic Regression classifier.
LR_pipeline = Pipeline([
    ('vectorizer_tfidf', TfidfVectorizer()),
    ('classifier_LR', LogisticRegression(random_state=22))
])

#Set the hyperparameter space that will be scanned.
grid_params_LR = {
    'vectorizer_tfidf__ngram_range': [(1,1), (1,2), (1,3)],
    'vectorizer_tfidf__min_df': [100, 500, 1000],
    'classifier_LR__l1_ratio': [0.0, 0.5, 1.0]
}
```

## Binary

### Train GridSearchCV with Binary Classification Dataset

```
In [ ]:
```

```
%time
# Initialize and run the GridSearchCV to scan the hyperparameter and find the best hyperp
arameter set that will maximize the scoring option for binary classification.
grid_search_LR_Binary = GridSearchCV(
    LR_pipeline,
    grid_params_LR,
    cv = 5,
```

```
verbose = 1,  
scoring = 'f1_macro')
```

```
# Fit binary dataset to the grid_search_LR
```

```
grid_search_LR_Binary.fit(train_binary_x, train_binary_y)
```

```
print(grid_search_LR_Binary.cv_results_)
```

Fitting 5 folds for each of 27 candidates, totalling 135 fits

```
{'mean_fit_time': array([0.64763546, 1.63877077, 2.78068924, 0.58505406, 1.59826751,  
3.35907249, 0.52542706, 1.53013873, 2.96164875, 1.4102571 ,  
1.79597039, 3.49111981, 0.67839761, 3.974506 , 4.45213137,  
1.02990761, 3.28383846, 2.81683698, 0.66542273, 1.67287664,  
2.96072183, 0.58706584, 2.85808315, 4.9504385 , 0.54023175,  
1.63738976, 2.87119451]), 'std_fit_time': array([0.01533409, 0.0306092 , 0.0297744  
1, 0.00894899, 0.02273817,  
0.83727759, 0.01219222, 0.01661217, 0.43576528, 0.21391453,  
0.32555714, 0.70638149, 0.15033251, 1.21727029, 0.807752 ,  
0.03470546, 1.03823062, 0.13780498, 0.09514831, 0.01063843,  
0.03839308, 0.02070404, 1.09406188, 1.82007664, 0.01343624,  
0.00593452, 0.04047213]), 'mean_score_time': array([0.11841493, 0.240558 , 0.3617  
5218, 0.11270795, 0.22927389,  
0.41183224, 0.10938854, 0.22329588, 0.41932635, 0.33311381,  
0.23290405, 0.5167573 , 0.1124516 , 0.68172483, 0.6966393 ,  
0.18614492, 0.5622035 , 0.33796072, 0.11769724, 0.24112377,  
0.37650003, 0.11184721, 0.51228557, 0.49288759, 0.11513581,  
0.23832445, 0.35250783]), 'std_score_time': array([0.00349636, 0.00725373, 0.01663  
5 , 0.00222708, 0.00711243,  
0.14283899, 0.00450281, 0.01102923, 0.15950218, 0.07990045,  
0.005834 , 0.2158938 , 0.00699544, 0.36634805, 0.18056895,  
0.06495789, 0.30761905, 0.0121466 , 0.00348684, 0.00393361,  
0.01504942, 0.00248724, 0.22135959, 0.17194235, 0.00831188,  
0.01566716, 0.00928222]), 'param_classifier_LR__l1_ratio': masked_array(data=[0.0,  
0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.5, 0.5,  
0.5, 0.5, 0.5, 0.5, 0.5, 0.5, 0.5, 1.0, 1.0, 1.0, 1.0,  
1.0, 1.0, 1.0, 1.0, 1.0],  
mask=[False, False, False, False, False, False, False, False, False,  
False, False, False, False, False, False, False, False, False,  
False, False, False, False, False, False, False],  
fill_value='?',  
dtype=object), 'param_vectorizer_tfidf__min_df': masked_array(data=[100, 100,  
100, 500, 500, 500, 1000, 1000, 1000, 100,  
100, 100, 500, 500, 500, 1000, 1000, 1000, 100, 100,  
100, 500, 500, 500, 1000, 1000, 1000],  
mask=[False, False, False, False, False, False, False, False, False,  
False, False, False, False, False, False, False, False, False,  
False, False, False, False, False, False, False, False, False],  
fill_value='?',  
dtype=object), 'param_vectorizer_tfidf__ngram_range': masked_array(data=[(1,  
1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3), (1, 1),  
(1, 2), (1, 3), (1, 1), (1, 2), (1, 3), (1, 1), (1, 2),  
(1, 3), (1, 1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3),  
(1, 1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3)],  
mask=[False, False, False, False, False, False, False, False, False,  
False, False, False, False, False, False, False, False, False,  
False, False, False, False, False, False, False, False, False],  
fill_value='?',  
dtype=object), 'params': [{'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__  
min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.0,  
'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_L  
R__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1,  
3)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__  
ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500,  
'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tf  
idf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio':  
0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classi  
fier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range  
': (1, 2)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorize  
r_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_d  
f': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.5, 'vect
```

```

orizer_tfidf__min_df':100, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}], 'split0_test_score': array([0.9049106 , 0.90422279, 0.90422279, 0.85976277, 0.85872336, 0.85872336, 0.79519811, 0.79519811, 0.79519811, 0.79519811, 0.9049106 , 0.90422279, 0.90422279, 0.85976277, 0.85872336, 0.85872336, 0.79519811, 0.79519811, 0.79519811]), 'split1_test_score': array([0.90214205, 0.90144517, 0.90144517, 0.85071495, 0.85003469, 0.85003469, 0.79534594, 0.79534594, 0.79534594, 0.79534594, 0.90214205, 0.90144517, 0.90144517, 0.85071495, 0.85003469, 0.85003469, 0.79534594, 0.79534594, 0.79534594, 0.79534594, 0.90214205, 0.90144517, 0.90144517, 0.85071495, 0.85003469, 0.85003469, 0.79534594, 0.79534594]), 'split2_test_score': array([0.91009789, 0.91357064, 0.91357064, 0.85802713, 0.85907058, 0.85907058, 0.79324492, 0.79324492, 0.79324492, 0.79324492, 0.91009789, 0.91357064, 0.91357064, 0.85802713, 0.85907058, 0.85907058, 0.79324492, 0.79324492, 0.79324492, 0.79324492, 0.91009789, 0.91357064, 0.91357064, 0.85802713, 0.85907058, 0.85907058, 0.79324492, 0.79324492]), 'split3_test_score': array([0.9031519 , 0.90592995, 0.90592995, 0.84967302, 0.85036737, 0.85036737, 0.80452286, 0.80452286, 0.80452286, 0.80452286, 0.9031519 , 0.90592995, 0.90592995, 0.84967302, 0.85036737, 0.85036737, 0.80452286, 0.80452286, 0.80452286, 0.80452286, 0.9031519 , 0.90592995, 0.90592995, 0.84967302, 0.85036737, 0.85036737, 0.80452286, 0.80452286]), 'split4_test_score': array([0.89933293, 0.90141788, 0.90141788, 0.84712273, 0.84849893, 0.84849893, 0.79571791, 0.79571791, 0.79571791, 0.79571791, 0.89933293, 0.90141788, 0.90141788, 0.84712273, 0.84849893, 0.84849893, 0.79571791, 0.79571791, 0.79571791, 0.79571791, 0.89933293, 0.90141788, 0.90141788, 0.84712273, 0.84849893, 0.84849893, 0.79571791, 0.79571791]), 'mean_test_score': array([0.90392707, 0.90531729, 0.90524782, 0.85306012, 0.85333899, 0.85333899, 0.79680595, 0.79680595, 0.79680595, 0.79680595, 0.90392707, 0.90531729, 0.90524782, 0.85306012, 0.85333899, 0.85333899, 0.79680595, 0.79680595, 0.79680595, 0.79680595, 0.90392707, 0.90531729, 0.90524782, 0.85306012, 0.85333899, 0.85333899, 0.79680595, 0.79680595]), 'std_test_score': array([0.00357602, 0.00446967, 0.00446967, 0.00493602, 0.00458296, 0.00458296, 0.00395303, 0.00395303, 0.00395303, 0.00395303, 0.00357602, 0.00446967, 0.00446967, 0.00493602, 0.00458296, 0.00458296, 0.00395303, 0.00395303, 0.00395303, 0.00395303, 0.00357602, 0.00446967, 0.00446967, 0.00493602, 0.00458296, 0.00458296, 0.00395303, 0.00395303]), 'rank_test_score': array([ 7,  1,  4, 16, 10, 10, 19, 19, 19,  7,  1,  4, 16, 10, 10, 19, 19, 19,  7,  1,  4, 16, 10, 10, 19, 19, 19], dtype=int32)}
CPU times: user 5min 4s, sys: 4.69 s, total: 5min 9s
Wall time: 5min 38s

```

## Report the Results

```

# Report the standard deviation of split scores for each hyperparameter group.
for i in range(27):
    print("Parameter Group ", i+1)
    print("Parameters: ", grid_search_LR_Binary.cv_results_['params'][i])
    print("Mean Test Score: ", grid_search_LR_Binary.cv_results_['mean_test_score'][i])
    print("Standard Deviation Test Score: ", grid_search_LR_Binary.cv_results_['std_test_score'][i])
    print("Min: ", min(grid_search_LR_Binary.cv_results_["split0_test_score"][i], grid_search_LR_Binary.cv_results_["split1_test_score"][i],
        grid_search_LR_Binary.cv_results_["split2_test_score"][i], grid_search_LR_Binary.cv_results_["split3_test_score"][i],
        grid_search_LR_Binary.cv_results_["split4_test_score"][i]))
    print("Max: ", max(grid_search_LR_Binary.cv_results_["split0_test_score"][i], grid_search_LR_Binary.cv_results_["split1_test_score"][i],
        grid_search_LR_Binary.cv_results_["split2_test_score"][i], grid_search_LR_Binary.cv_results_["split3_test_score"][i],
        grid_search_LR_Binary.cv_results_["split4_test_score"][i]))
    print("\n")

# Show the best parameter set for given dataset and hyperparameter space.
print("Best Parameters: ", grid_search_LR_Binary.best_params_)

```

```

Parameter Group 1
Parameters: {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}
Mean Test Score: 0.9039270734505316
Standard Deviation Test Score: 0.0035760217266048873
Min: 0.8993329265636055
Max: 0.9100978866625582

```

```

Parameter Group 2
Parameters: {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}
Mean Test Score: 0.9053172864559327
Standard Deviation Test Score: 0.004469668842880747
Min: 0.9014178791573693
Max: 0.9135706352037654

```

```

Parameter Group 3
Parameters: {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
Mean Test Score: 0.9052478213686159
Standard Deviation Test Score: 0.004341727994197243
Min: 0.9014178791573693
Max: 0.9132233097671819

```

```

Parameter Group 4
Parameters: {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}
Mean Test Score: 0.8530601176631698
Standard Deviation Test Score: 0.004936023300777656
Min: 0.8471227272946544
Max: 0.85976277196431

```

```

Parameter Group 5
Parameters: {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}
Mean Test Score: 0.8533389853094251
Standard Deviation Test Score: 0.0045829631254465574
Min: 0.8484989270469089
Max: 0.8590705800139762

```

```

Parameter Group 6
Parameters: {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}

```

Mean Test Score: 0.8533389853094251  
Standard Deviation Test Score: 0.0045829631254465574  
Min: 0.8484989270469089  
Max: 0.8590705800139762

#### Parameter Group 7

Parameters: {'classifier\_LR\_l1\_ratio': 0.0, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 1)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

#### Parameter Group 8

Parameters: {'classifier\_LR\_l1\_ratio': 0.0, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

#### Parameter Group 9

Parameters: {'classifier\_LR\_l1\_ratio': 0.0, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 3)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

#### Parameter Group 10

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 100, 'vectorizer\_tfidf\_ngram\_range': (1, 1)}  
Mean Test Score: 0.9039270734505316  
Standard Deviation Test Score: 0.0035760217266048873  
Min: 0.8993329265636055  
Max: 0.9100978866625582

#### Parameter Group 11

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 100, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}  
Mean Test Score: 0.9053172864559327  
Standard Deviation Test Score: 0.004469668842880747  
Min: 0.9014178791573693  
Max: 0.9135706352037654

#### Parameter Group 12

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 100, 'vectorizer\_tfidf\_ngram\_range': (1, 3)}  
Mean Test Score: 0.9052478213686159  
Standard Deviation Test Score: 0.004341727994197243  
Min: 0.9014178791573693  
Max: 0.9132233097671819

#### Parameter Group 13

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 500, 'vectorizer\_tfidf\_ngram\_range': (1, 1)}  
Mean Test Score: 0.8530601176631698  
Standard Deviation Test Score: 0.004936023300777656  
Min: 0.8471227272946544  
Max: 0.85976277196431

#### Parameter Group 14

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 500, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}

Mean Test Score: 0.8533389853094251  
Standard Deviation Test Score: 0.0045829631254465574  
Min: 0.8484989270469089  
Max: 0.8590705800139762

Parameter Group 15

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 500, 'vectorizer\_tfidf\_ngram\_range': (1, 3)}  
Mean Test Score: 0.8533389853094251  
Standard Deviation Test Score: 0.0045829631254465574  
Min: 0.8484989270469089  
Max: 0.8590705800139762

Parameter Group 16

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 1)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

Parameter Group 17

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

Parameter Group 18

Parameters: {'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 3)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

Parameter Group 19

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 100, 'vectorizer\_tfidf\_ngram\_range': (1, 1)}  
Mean Test Score: 0.9039270734505316  
Standard Deviation Test Score: 0.0035760217266048873  
Min: 0.8993329265636055  
Max: 0.9100978866625582

Parameter Group 20

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 100, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}  
Mean Test Score: 0.9053172864559327  
Standard Deviation Test Score: 0.004469668842880747  
Min: 0.9014178791573693  
Max: 0.9135706352037654

Parameter Group 21

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 100, 'vectorizer\_tfidf\_ngram\_range': (1, 3)}  
Mean Test Score: 0.9052478213686159  
Standard Deviation Test Score: 0.004341727994197243  
Min: 0.9014178791573693  
Max: 0.9132233097671819

Parameter Group 22

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 500, 'vectorizer\_tfidf\_ngram\_range': (1, 1)}



Mean Test Score: 0.8530601176631698  
Standard Deviation Test Score: 0.004936023300777656  
Min: 0.8471227272946544  
Max: 0.85976277196431

#### Parameter Group 23

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 500, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}  
Mean Test Score: 0.8533389853094251  
Standard Deviation Test Score: 0.0045829631254465574  
Min: 0.8484989270469089  
Max: 0.8590705800139762

#### Parameter Group 24

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 500, 'vectorizer\_tfidf\_ngram\_range': (1, 3)}  
Mean Test Score: 0.8533389853094251  
Standard Deviation Test Score: 0.0045829631254465574  
Min: 0.8484989270469089  
Max: 0.8590705800139762

#### Parameter Group 25

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 1)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

#### Parameter Group 26

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

#### Parameter Group 27

Parameters: {'classifier\_LR\_l1\_ratio': 1.0, 'vectorizer\_tfidf\_min\_df': 1000, 'vectorizer\_tfidf\_ngram\_range': (1, 3)}  
Mean Test Score: 0.7968059486549559  
Standard Deviation Test Score: 0.003953032507028959  
Min: 0.7932449196080784  
Max: 0.8045228649374805

Best Parameters: {'classifier\_LR\_l1\_ratio': 0.0, 'vectorizer\_tfidf\_min\_df': 100, 'vectorizer\_tfidf\_ngram\_range': (1, 2)}

### **Setting Pipeline with Best Parameters and Making Prediction**

In [ ]:

```
# Building the pipeline with the best parameter group and reporting Conf. Mat. and Results on the Test Set #
#Create your pipeline object with the best parameter set.
LR_Binary_BestParams_Pipeline = Pipeline([
    ('vectorizer_tfidf', TfidfVectorizer(ngram_range=(1,2), min_df=100)),
    ('classifier_LR', LogisticRegression(random_state=22, l1_ratio=0.0))
])

#Fit your pipeline on training set.
BestParams_LR_Binary_Model = LR_Binary_BestParams_Pipeline.fit(train_binary_x, train_binary_y)
```

```
# Make prediction
y_Predict_LR_Binary = BestParams_LR_Binary_Model.predict(test_binary_x)
```

## Reporting the Scores and Confusion Matrix for Binary Classification

In [ ]:

```
# Report the F1 and Accuracy scores and Confusion Matrix for binary classification
F1_LR_Binary_Test_BestParams = f1_score(test_binary_y, y_Predict_LR_Binary, average='macro')
Accuracy_LR_Binary_Test_BestParams = accuracy_score(test_binary_y, y_Predict_LR_Binary)

print('F1 Macro for Binary Classification with Best Parameters:\n', F1_LR_Binary_Test_BestParams)
print('\nAccuracy Score for Binary Classification with Best Parameters:\n', Accuracy_LR_Binary_Test_BestParams)
print('\nConfusion Matrix:')
confusion_matrix(test_binary_y, y_Predict_LR_Binary)
```

```
F1 Macro for Binary Classification with Best Parameters:
0.908329917386643
```

```
Accuracy Score for Binary Classification with Best Parameters:
0.9083490269930948
```

```
Confusion Matrix:
```

Out[ ]:

```
array([[712, 61],
       [ 85, 735]])
```

## Multi

### Train GridSearchCV with Multiclass Classification Dataset

In [ ]:

```
%%time
# Initialize and run the GridSearchCV to scan the hyperparameter and find the best hyperparameter set that will maximize the scoring option for binary classification.
grid_search_LR_Multi = GridSearchCV(
    LR_pipeline,
    grid_params_LR,
    cv = 5,
    verbose = 1,
    scoring = 'f1_macro')

# Fit binary dataset to the grid_search_LR
grid_search_LR_Multi.fit(train_multi_x, train_multi_y)
print(grid_search_LR_Multi.cv_results_)
```

```
Fitting 5 folds for each of 27 candidates, totalling 135 fits
{'mean_fit_time': array([2.3475028 , 3.76614108, 5.45782189, 1.97816172, 3.30320883,
 4.98713923, 1.70531092, 2.97863789, 4.60367241, 2.32767959,
 3.75832238, 5.35163207, 1.97474499, 3.30628633, 5.01560545,
 1.68397498, 3.01344953, 4.6164567 , 2.3203784 , 3.76328592,
 5.34449029, 1.97809834, 3.30582199, 4.97688971, 1.68936954,
 2.95064788, 4.51127062]), 'std_fit_time': array([0.024817 , 0.04857463, 0.0760942
7, 0.02067737, 0.01599347,
 0.09984373, 0.06942438, 0.05752073, 0.06879812, 0.02670958,
 0.04684159, 0.05589676, 0.03618616, 0.02216639, 0.03910853,
 0.06087461, 0.06356805, 0.07989687, 0.03335171, 0.08438838,
 0.10390594, 0.04203685, 0.02367419, 0.06156619, 0.07284807,
 0.04790349, 0.08726635]), 'mean_score_time': array([0.15404725, 0.31617384, 0.4995
0376, 0.14722528, 0.30519214,
 0.48424697, 0.14863591, 0.30320339, 0.45824723, 0.15800629,
```

```

0.31749897, 0.47871656, 0.15224004, 0.30582347, 0.47528672,
0.14570847, 0.30886745, 0.44694757, 0.15144529, 0.32836251,
0.47278247, 0.14960179, 0.30877309, 0.46080723, 0.14675159,
0.30065303, 0.43681145]], 'std_score_time': array([0.00773174, 0.0135232 , 0.02196
088, 0.00169811, 0.00878922,
0.01157899, 0.00520049, 0.00791397, 0.00644879, 0.0041849 ,
0.0125637 , 0.0157137 , 0.00571025, 0.00666378, 0.00991919,
0.00760981, 0.01597545, 0.0128681 , 0.00393809, 0.0046802 ,
0.01948283, 0.0066974 , 0.00976837, 0.00367236, 0.00662449,
0.01286064, 0.0078815 ]), 'param_classifier_LR__l1_ratio': masked_array(data=[0.0,
0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.5, 0.5,
0.5, 0.5, 0.5, 0.5, 0.5, 0.5, 0.5, 1.0, 1.0, 1.0, 1.0,
1.0, 1.0, 1.0, 1.0, 1.0],
mask=[False, False, False, False, False, False, False, False, False,
False, False, False, False, False, False, False, False, False,
False, False, False, False, False, False, False],
fill_value='?',
dtype=object), 'param_vectorizer_tfidf__min_df': masked_array(data=[100, 100,
100, 500, 500, 500, 1000, 1000, 1000, 100,
100, 100, 500, 500, 500, 1000, 1000, 1000, 100, 100,
100, 500, 500, 500, 1000, 1000, 1000],
mask=[False, False, False, False, False, False, False, False, False,
False, False, False, False, False, False, False, False, False,
False, False, False, False, False, False, False, False, False],
fill_value='?',
dtype=object), 'param_vectorizer_tfidf__ngram_range': masked_array(data=[(1,
1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3), (1, 1),
(1, 2), (1, 3), (1, 1), (1, 2), (1, 3), (1, 1), (1, 2),
(1, 3), (1, 1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3),
(1, 1), (1, 2), (1, 3), (1, 1), (1, 2), (1, 3)],
mask=[False, False, False, False, False, False, False, False, False,
False, False, False, False, False, False, False, False, False,
False, False, False, False, False, False, False, False, False],
fill_value='?',
dtype=object), 'params': [{'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}, {'classifier_LR__l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}], 'split0_test_score': array([0.53097175, 0.53593137, 0.53565442, 0.51490157, 0.51360355,

```

```

0.51360355, 0.46294246, 0.46294246, 0.46294246, 0.53097175,
0.53593137, 0.53565442, 0.51490157, 0.51360355, 0.51360355,
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0.53565442, 0.51490157, 0.51360355, 0.51360355, 0.46294246,
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322738, 0.51409587, 0.51371277,
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0.54342473, 0.54322738, 0.51409587, 0.51371277, 0.51371277,
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0.54322738, 0.51409587, 0.51371277, 0.51371277, 0.47186335,
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0.46719957, 0.46719957, 0.46719957, 0.53770449, 0.5392327 ,
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0.46719957, 0.46719957]), 'split3_test_score': array([0.52430152, 0.52842867, 0.52
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0.46342602, 0.46342602, 0.46342602, 0.53327816, 0.53631211,
0.53641665, 0.50687869, 0.50583035, 0.50583035, 0.46342602,
0.46342602, 0.46342602]), 'std_test_score': array([0.00515161, 0.00499084, 0.00509
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, 19,  7,  4,  1, 10, 13, 13, 19, 19,
19,  7,  4,  1, 10, 13, 13, 19, 19, 19], dtype=int32))
CPU times: user 8min 33s, sys: 7.71 s, total: 8min 41s
Wall time: 8min 39s

```

## Report the Results

In [ ]:

```

# Report the standart deviation of split scores for each hyperparameter group.
for i in range(27):
    print("Parameter Group ", i+1)
    print(grid_search_LR_Multi.cv_results_['params'][i])
    print(grid_search_LR_Multi.cv_results_['mean_test_score'][i])
    print(grid_search_LR_Multi.cv_results_['std_test_score'][i])
    print("Min: ", min(grid_search_LR_Multi.cv_results_["split0_test_score"][i], grid_sear
ch_LR_Multi.cv_results_["split1_test_score"][i],
                    grid_search_LR_Multi.cv_results_["split2_test_score"][i], grid_sear
ch_LR_Multi.cv_results_["split3_test_score"][i],
                    grid_search_LR_Multi.cv_results_["split4_test_score"][i]))
    print("Max: ", max(grid_search_LR_Multi.cv_results_["split0_test_score"][i], grid_sear
ch_LR_Multi.cv_results_["split1_test_score"][i],
                    grid_search_LR_Multi.cv_results_["split2_test_score"][i], grid_sear
ch_LR_Multi.cv_results_["split3_test_score"][i],
                    grid_search_LR_Multi.cv_results_["split4_test_score"][i]))
    print("\n")

# Show the best parameter set for given dataset and hyperparameter space.

```

```
print("Best Parameters: ", grid_search_LR_Multi.best_params_)
```

```
Parameter Group 1  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}  
0.533278157858257  
0.005151607577326998  
Min: 0.5243015182286054  
Max: 0.5380659333066168
```

```
Parameter Group 2  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}  
0.536312109663449  
0.0049908379275617675  
Min: 0.5284286664610763  
Max: 0.5434247275415489
```

```
Parameter Group 3  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}  
0.5364166486996111  
0.005090054209611724  
Min: 0.5280977399670365  
Max: 0.5432273768095992
```

```
Parameter Group 4  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}  
0.506878694432378  
0.009500533533121237  
Min: 0.48907890289351696  
Max: 0.5149015726124225
```

```
Parameter Group 5  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}  
0.5058303519343349  
0.01057041888035273  
Min: 0.48540494509154614  
Max: 0.5137127724584536
```

```
Parameter Group 6  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}  
0.5058303519343349  
0.01057041888035273  
Min: 0.48540494509154614  
Max: 0.5137127724584536
```

```
Parameter Group 7  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}  
0.4634260154004709  
0.00589814145367564  
Min: 0.4544388137370744  
Max: 0.47186335462833107
```

```
Parameter Group 8  
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}  
0.4634260154004709  
0.00589814145367564  
Min: 0.4544388137370744  
Max: 0.47186335462833107
```

Parameter Group 9  
{'classifier\_LR\_l1\_ratio': 0.0, 'vectorizer\_tfidf\_\_min\_df': 1000, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}  
0.4634260154004709  
0.00589814145367564  
Min: 0.4544388137370744  
Max: 0.47186335462833107

Parameter Group 10  
{'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 1)}  
0.533278157858257  
0.005151607577326998  
Min: 0.5243015182286054  
Max: 0.5380659333066168

Parameter Group 11  
{'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 2)}  
0.536312109663449  
0.0049908379275617675  
Min: 0.5284286664610763  
Max: 0.5434247275415489

Parameter Group 12  
{'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}  
0.5364166486996111  
0.005090054209611724  
Min: 0.5280977399670365  
Max: 0.5432273768095992

Parameter Group 13  
{'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 1)}  
0.506878694432378  
0.009500533533121237  
Min: 0.48907890289351696  
Max: 0.5149015726124225

Parameter Group 14  
{'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 2)}  
0.5058303519343349  
0.01057041888035273  
Min: 0.48540494509154614  
Max: 0.5137127724584536

Parameter Group 15  
{'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}  
0.5058303519343349  
0.01057041888035273  
Min: 0.48540494509154614  
Max: 0.5137127724584536

Parameter Group 16  
{'classifier\_LR\_l1\_ratio': 0.5, 'vectorizer\_tfidf\_\_min\_df': 1000, 'vectorizer\_tfidf\_\_ngram\_range': (1, 1)}  
0.4634260154004709  
0.00589814145367564  
Min: 0.4544388137370744  
Max: 0.47186335462833107

Parameter Group 17

```
{'classifier_LR_l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}  
0.4634260154004709  
0.00589814145367564  
Min: 0.4544388137370744  
Max: 0.47186335462833107
```

Parameter Group 18

```
{'classifier_LR_l1_ratio': 0.5, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}  
0.4634260154004709  
0.00589814145367564  
Min: 0.4544388137370744  
Max: 0.47186335462833107
```

Parameter Group 19

```
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 1)}  
0.533278157858257  
0.005151607577326998  
Min: 0.5243015182286054  
Max: 0.5380659333066168
```

Parameter Group 20

```
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 2)}  
0.536312109663449  
0.0049908379275617675  
Min: 0.5284286664610763  
Max: 0.5434247275415489
```

Parameter Group 21

```
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}  
0.5364166486996111  
0.005090054209611724  
Min: 0.5280977399670365  
Max: 0.5432273768095992
```

Parameter Group 22

```
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}  
0.506878694432378  
0.009500533533121237  
Min: 0.48907890289351696  
Max: 0.5149015726124225
```

Parameter Group 23

```
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}  
0.5058303519343349  
0.01057041888035273  
Min: 0.48540494509154614  
Max: 0.5137127724584536
```

Parameter Group 24

```
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 3)}  
0.5058303519343349  
0.01057041888035273  
Min: 0.48540494509154614  
Max: 0.5137127724584536
```

```
Parameter Group 25
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 1)}
0.4634260154004709
0.00589814145367564
Min: 0.4544388137370744
Max: 0.47186335462833107
```

```
Parameter Group 26
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 2)}
0.4634260154004709
0.00589814145367564
Min: 0.4544388137370744
Max: 0.47186335462833107
```

```
Parameter Group 27
{'classifier_LR_l1_ratio': 1.0, 'vectorizer_tfidf__min_df': 1000, 'vectorizer_tfidf__ngram_range': (1, 3)}
0.4634260154004709
0.00589814145367564
Min: 0.4544388137370744
Max: 0.47186335462833107
```

```
Best Parameters: {'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
```

### ***Setting Pipeline with Best Parameters and Making Prediction***

In [ ]:

```
# Building the pipeline with the best parameter group and reporting Conf. Mat. and Results on the Test Set
#Create your pipeline object with the best parameter set.
LR_Multi_BestParams_Pipeline = Pipeline([
    ('vectorizer_tfidf', TfidfVectorizer(ngram_range=(1,2), min_df=100)),
    ('classifier_LR', LogisticRegression(random_state=22, l1_ratio=0.0))
])

#Fit your pipeline on training set.
BestParams_LR_Multi_Model = LR_Multi_BestParams_Pipeline.fit(train_multi_x, train_multi_y)

# Make prediction
y_Predict_LR_Multi = BestParams_LR_Multi_Model.predict(test_multi_x)
```

### ***Reporting the Scores and Confusion Matrix for Multiclass Classification***

In [ ]:

```
# Report the F1 and Accuracy scores and Confusion Matrix for multiclass classification
F1_LR_Multi_Test_BestParams = f1_score(test_multi_y, y_Predict_LR_Multi, average='macro')
Accuracy_LR_Multi_Test_BestParams = accuracy_score(test_multi_y, y_Predict_LR_Multi)

print('F1 Macro for Multiclass Classification with Best Parameters:\n', F1_LR_Multi_Test_BestParams)
print('\nAccuracy Score for Multiclass Classification with Best Parameters:\n', Accuracy_LR_Multi_Test_BestParams)
print('\nConfusion Matrix:')
confusion_matrix(test_multi_y, y_Predict_LR_Multi)
```

F1 Macro for Multiclass Classification with Best Parameters:



0.5537120257977514

Accuracy Score for Multiclass Classification with Best Parameters:  
0.557

Confusion Matrix:

Out[ ]:

```
array([[278,  89,  14,   6,   6],
       [ 83, 189,  79,  18,  11],
       [ 26,  83, 187,  91,  20],
       [ 12,  27,  72, 171,  97],
       [  9,  11,  21, 111, 289]])
```

## Neural Models

Helpful Resource About the Whole CNN Sentence Classification Process:  
<https://cnvrg.io/cnn-sentence-classification/>

## Convolutional Neural Network (CNN)

### Libraries

In [ ]:

```
import pandas as pd
import numpy as np
import nltk, re
import tensorflow as tf
tf.config.experimental_run_functions_eagerly(True)
from sklearn.model_selection import train_test_split
from numpy import array, asarray, zeros

from nltk.stem import PorterStemmer
from nltk.tokenize import sent_tokenize
import nltk
nltk.download('punkt')

import keras
from keras.preprocessing.text import Tokenizer
from keras.preprocessing.sequence import pad_sequences

from keras.models import Sequential
from keras.layers.convolutional import Conv1D, MaxPooling1D
from keras.layers import GlobalMaxPooling1D
from keras.layers import Dense, Flatten, Embedding, Input, Dropout
from keras.callbacks import ModelCheckpoint

from tensorflow.keras.utils import to_categorical

from gensim.models import Word2Vec
import gensim.downloader as api
import string
```

WARNING:tensorflow:From <ipython-input-32-80fc57a2d553>:5: experimental\_run\_functions\_eagerly (from tensorflow.python.eager.def\_function) is deprecated and will be removed in a future version.

Instructions for updating:

Use `tf.config.run\_functions\_eagerly` instead of the experimental version.

[nltk\_data] Downloading package punkt to /root/nltk\_data...

[nltk\_data] Package punkt is already up-to-date!

### Preparing Datasets and Creating the Validation Datasets

In [ ]:

```

# Create a validation set from train set
# Please use random_state of 22 and test_size of 0.1
# =====

# CNN Validation Binary
x_train_binary_cnn, x_val_binary_cnn, y_train_binary_cnn, y_val_binary_cnn = train_test_split(train_binary["text"], train_binary["label"], test_size=0.1, random_state= 22)

x_test_binary_cnn, y_test_binary_cnn = test_binary["text"], test_binary["label"]

# CNN Validation Multiclass
x_train_multi_cnn, x_val_multi_cnn, y_train_multi_cnn, y_val_multi_cnn = train_test_split(train_multi["text"], train_multi["label"], test_size=0.1, random_state= 22)

x_test_multi_cnn, y_test_multi_cnn = test_multi["text"], test_multi["label"]

# Arranging y-values of the multiclass datasets using "to_categorical"
y_train_multi_cnn_p = to_categorical(y_train_multi_cnn)
y_test_multi_cnn_p = to_categorical(y_test_multi_cnn)
y_val_multi_cnn_p = to_categorical(y_val_multi_cnn)

```

## Length of the Longest Sentences

In [ ]:

```

# Find the length of the longest sentences
# =====

ultimate_dataset = train["text"].append(test["text"], ignore_index = True)

# Length of the longest sentences
max_sent_len = max(len(i.split()) for i in ultimate_dataset)

# General variable for the "Length of the Longest Sentences"
MAX_LENGTH_SENT = max_sent_len
print("Length of the Longest Sentences: ", MAX_LENGTH_SENT)

```

Length of the Longest Sentences: 495

## Tokenization

In [ ]:

```

# Initialize and apply "Tokenizers" to the data
# =====

# Binary
# -----

tokenizer_keras_binary = Tokenizer(filters='!"$%&()*+,-./:;<=>?@[\\]^_`{|}~\t\n\'')
tokenizer_keras_binary.fit_on_texts(list(x_train_binary_cnn))

# Train
x_train_binary_cnn_tok = tokenizer_keras_binary.texts_to_sequences(x_train_binary_cnn)

# Test
x_test_binary_cnn_tok = tokenizer_keras_binary.texts_to_sequences(x_test_binary_cnn)

# Validation
x_val_binary_cnn_tok = tokenizer_keras_binary.texts_to_sequences(x_val_binary_cnn)

# Multiclass
# -----

```

```
tokenizer_keras_multi = Tokenizer(filters='!"$%&()*+,-./:;<=>?@[\\]^_`{|}~\t\n\'')
tokenizer_keras_multi.fit_on_texts(list(x_train_multi_cnn))

# Train
x_train_multi_cnn_tok = tokenizer_keras_multi.texts_to_sequences(x_train_multi_cnn)

# Test
x_test_multi_cnn_tok = tokenizer_keras_multi.texts_to_sequences(x_test_multi_cnn)

# Validation
x_val_multi_cnn_tok = tokenizer_keras_multi.texts_to_sequences(x_val_multi_cnn)
```

## Padding

### **Binary**

In [ ]:

```
# Train
#-----
padded_train_binary_x = pad_sequences(x_train_binary_cnn_tok, padding='post', maxlen=MAX_LENGTH_SENT)

# Test
#-----
padded_test_binary_x = pad_sequences(x_test_binary_cnn_tok, padding='post', maxlen=MAX_LENGTH_SENT)

# Validation
#-----
padded_val_binary_x = pad_sequences(x_val_binary_cnn_tok, padding='post', maxlen=MAX_LENGTH_SENT)
```

### **Multiclass**

In [ ]:

```
# Train
#-----
padded_train_multi_x = pad_sequences(x_train_multi_cnn_tok, padding='post', maxlen=MAX_LENGTH_SENT)
# -----
# -----

# Test
#-----
padded_test_multi_x = pad_sequences(x_test_multi_cnn_tok, padding='post', maxlen=MAX_LENGTH_SENT)
# -----
# -----

# Validation
#-----
padded_val_multi_x = pad_sequences(x_val_multi_cnn_tok, padding='post', maxlen=MAX_LENGTH_SENT)
# -----
# -----
```

## Word Embeddings

### **Script for Getting the F1 Score of the Models**

Resource: [https://aakashgoel12.medium.com/how-to-add-user-defined-function-get-f1-score-in-keras-metrics-3013f979ce0d#:~:text=You%20will%20get%20training%20and%20validation%20F1%20score%20after%20each%20epoch%20of%20the%20model](https://aakashgoel12.medium.com/how-to-add-user-defined-function-get-f1-score-in-keras-metrics-3013f979ce0d#:~:text=You%20will%20get%20training%20and%20validation%20F1%20score%20after%20each%20epoch%20of%20the%20model,https://aakashgoel12.medium.com/how-to-add-user-defined-function-get-f1-score-in-keras-metrics-3013f979ce0d#:~:text=You%20will%20get%20training%20and%20validation%20F1%20score%20after%20each%20epoch%20of%20the%20model)

In [ ]:

```
import keras.backend as K

def get_f1(y_true, y_pred): #taken from old keras source code
    true_positives = K.sum(K.round(K.clip(y_true * y_pred, 0, 1)))
    possible_positives = K.sum(K.round(K.clip(y_true, 0, 1)))
    predicted_positives = K.sum(K.round(K.clip(y_pred, 0, 1)))
    precision = true_positives / (predicted_positives + K.epsilon())
    recall = true_positives / (possible_positives + K.epsilon())
    f1_val = 2*(precision*recall)/(precision+recall+K.epsilon())
    return f1_val
```

## Creating Embedding Layers

In [ ]:

```
# Embedding Layer For Randomly Initialized Word Embeddings
# =====

# Binary
random_emb_layer_binary = Embedding(input_dim=len(tokenizer_keras_binary.word_index)+1, output_dim=100, input_length=MAX_LENGTH_SENT)
# Multiclass
random_emb_layer_multi = Embedding(input_dim=len(tokenizer_keras_multi.word_index)+1, output_dim=100, input_length=MAX_LENGTH_SENT)

# Word2Vec Embedding Layer For Word Embeddings From Scratch
# =====

# Binary
my_binary_embedding = Word2Vec(sentences=[i.split() for i in x_train_binary_cnn], min_count=1)
my_binary_emb_layer = Embedding(input_dim=len(my_binary_embedding.wv.vocab), output_dim=100, input_length=MAX_LENGTH_SENT, weights=[my_binary_embedding.wv.vectors])
# Multiclass
my_multi_embedding = Word2Vec(sentences=[i.split() for i in x_train_multi_cnn], min_count=1)
my_multi_emb_layer = Embedding(input_dim=len(my_multi_embedding.wv.vocab), output_dim=100, input_length=MAX_LENGTH_SENT, weights=[my_multi_embedding.wv.vectors])

# Word2Vec Embedding Layer For Pretrained Word Embeddings
# =====

# Binary
pretrained_binary_embedding = Word2Vec(sentences=api.load('text8'), min_count=1)
pretrained_binary_emb_layer = Embedding(input_dim=len(pretrained_binary_embedding.wv.vocab), output_dim=100, input_length=MAX_LENGTH_SENT, weights=[pretrained_binary_embedding.wv.vectors])
# Multiclass
pretrained_multi_embedding = Word2Vec(sentences=api.load('text8'), min_count=1)
pretrained_multi_emb_layer = Embedding(input_dim=len(pretrained_multi_embedding.wv.vocab), output_dim=100, input_length=MAX_LENGTH_SENT, weights=[pretrained_multi_embedding.wv.vectors])

[=====] 100.0% 31.6/31.6MB downloaded
```

## Parameter Vectors for Running the Model Functions

In [ ]:

```
# General
filter_size_options = [16, 32]
kernel_size_options = [4, 8]
```

```
hidden_layer_options =[10, 20]
```

```
# Binary
```

```
operation_names_binary = ['Random Binary Model', 'From Scratch Binary Model', 'Pretrained Binary Model']
```

```
embedding_layer_binary = [random_emb_layer_binary, my_binary_emb_layer, pretrained_binary_emb_layer]
```

```
# Multiclass
```

```
operation_names_multi = ['Random Multiclass Model', 'From Scratch Multiclass Model', 'Pretrained Multiclass Model']
```

```
embedding_layer_multi = [random_emb_layer_multi, my_multi_emb_layer, pretrained_multi_emb_layer]
```

## Binary

## MODEL FUNCTION

```
In [ ]:
```

```
def CNN_Model_Function_Binary(operation_binary, emb_layer_binary, filter_size, kernel_size, hidden_layer_size):

    print(operation_binary)
    print("=====")
    print("Filter Size: ", filter_size)
    print("Kernel Size: ", kernel_size)
    print("Hidden Layer Size: ", hidden_layer_size)
    print("----- \n")

    # Define Model
    model_binary = Sequential()
    model_binary.add(emb_layer_binary)
    model_binary.add(Conv1D(filters=filter_size, kernel_size=kernel_size, activation='relu'))
    model_binary.add(GlobalMaxPooling1D())
    model_binary.add(Flatten())
    model_binary.add(Dense(hidden_layer_size, activation='relu'))
    model_binary.add(Dense(1, activation='sigmoid'))
    print(model_binary.summary())

    # Compile model
    model_binary.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy', get_f1])

    print("\n----- \n")

    # Fit
    model_binary.fit(padded_train_binary_x, y_train_binary_cnn, validation_data=(padded_val_binary_x, y_val_binary_cnn), epochs=10, verbose=0)

    # Getting score metrics
    scores_model_binary = model_binary.evaluate(padded_test_binary_x, y_test_binary_cnn, verbose=0)

    print("Accuracy Score: %.2f%%" % (scores_model_binary[1]*100))
    print("F1 Score: %.2f%%" % (scores_model_binary[2]*100))
    print('Confusion Matrix:')
    y_predict_model_binary = model_binary.predict(padded_test_binary_x)
    y_predict_model_binary = y_predict_model_binary.astype(int).tolist()
    print(confusion_matrix(y_test_binary_cnn, y_predict_model_binary))

    print("\n ----- \n")
```

## CALL FUNCTION IN LOOP

```
In [ ]:
```

```
# CALL IN LOOP
# Binary
#operation_names_binary = ['Random Binary Model', 'From Scratch Binary Model', 'Pretrained Binary Model']
#embedding_layer_binary = [random_emb_layer_binary, my_binary_emb_layer, pretrained_binary_emb_layer]

# Random Binary Model
for a in range(len(filter_size_options)):
    for b in range(len(kernel_size_options)):
        for c in range(len(hidden_layer_options)):
            CNN_Model_Function_Binary('Random Binary Model', random_emb_layer_binary, filter_size_options[a], kernel_size_options[b], hidden_layer_options[c])
```

Random Binary Model

=====

Filter Size: 16

Kernel Size: 4

Hidden Layer Size: 10

-----

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
embedding (Embedding)	(None, 495, 100)	2023000
conv1d (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d (GlobalMaxPooling1D)	(None, 16)	0
flatten (Flatten)	(None, 16)	0
dense (Dense)	(None, 10)	170
dense_1 (Dense)	(None, 1)	11

=====

Total params: 2,029,597

Trainable params: 2,029,597

Non-trainable params: 0

None

-----

Accuracy Score: 90.46%

F1 Score: 90.27%

Confusion Matrix:

```
[[773  0]
 [810 10]]
```

-----

-----

Random Binary Model

=====

Filter Size: 16

Kernel Size: 4

Hidden Layer Size: 20

-----

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
=====		
embedding (Embedding)	(None, 495, 100)	2023000
conv1d_1 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_1 (GlobalMaxPooling1D)	(None, 16)	0

global_max_pooling1d_1 (GlobalMaxPooling1D)	(None, 16)	0
flatten_1 (Flatten)	(None, 16)	0
dense_2 (Dense)	(None, 20)	340
dense_3 (Dense)	(None, 1)	21

=====  
Total params: 2,029,777  
Trainable params: 2,029,777  
Non-trainable params: 0

None

-----  
  
Accuracy Score: 89.08%  
F1 Score: 89.15%  
Confusion Matrix:  
[[772 1]  
 [652 168]]

-----  
Random Binary Model

=====  
Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 10

-----  
Model: "sequential\_2"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 495, 100)	2023000
conv1d_2 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_2 (GlobalMaxPooling1D)	(None, 16)	0
flatten_2 (Flatten)	(None, 16)	0
dense_4 (Dense)	(None, 10)	170
dense_5 (Dense)	(None, 1)	11

=====  
Total params: 2,035,997  
Trainable params: 2,035,997  
Non-trainable params: 0

None

-----  
  
Accuracy Score: 88.26%  
F1 Score: 88.14%  
Confusion Matrix:  
[[773 0]  
 [604 216]]

-----  
Random Binary Model

=====  
Filter Size: 16  
Kernel Size: 8

kernel Size: 8  
Hidden Layer Size: 20  
-----

Model: "sequential\_3"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 495, 100)	2023000
conv1d_3 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_3 (GlobalMaxPooling1D)	(None, 16)	0
flatten_3 (Flatten)	(None, 16)	0
dense_6 (Dense)	(None, 20)	340
dense_7 (Dense)	(None, 1)	21

=====  
Total params: 2,036,177  
Trainable params: 2,036,177  
Non-trainable params: 0

-----  
None

-----  
Accuracy Score: 88.45%  
F1 Score: 88.39%  
Confusion Matrix:  
[[772 1]  
 [565 255]]

-----  
Random Binary Model

=====  
Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 10  
-----

Model: "sequential\_4"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 495, 100)	2023000
conv1d_4 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_4 (GlobalMaxPooling1D)	(None, 32)	0
flatten_4 (Flatten)	(None, 32)	0
dense_8 (Dense)	(None, 10)	330
dense_9 (Dense)	(None, 1)	11

=====  
Total params: 2,036,173  
Trainable params: 2,036,173  
Non-trainable params: 0

-----  
None

-----  
Accuracy Score: 87.26%



Accuracy Score: 87.26%  
F1 Score: 87.17%  
Confusion Matrix:  
[[772 1]  
[600 220]]

Random Binary Model

Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 20

Model: "sequential\_5"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 495, 100)	2023000
conv1d_5 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_5 (GlobalMaxPooling1D)	(None, 32)	0
flatten_5 (Flatten)	(None, 32)	0
dense_10 (Dense)	(None, 20)	660
dense_11 (Dense)	(None, 1)	21

Total params: 2,036,513  
Trainable params: 2,036,513  
Non-trainable params: 0

None

Accuracy Score: 87.26%  
F1 Score: 87.18%  
Confusion Matrix:  
[[770 3]  
[555 265]]

Random Binary Model

Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 10

Model: "sequential\_6"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 495, 100)	2023000
conv1d_6 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_6 (GlobalMaxPooling1D)	(None, 32)	0
flatten_6 (Flatten)	(None, 32)	0
dense_12 (Dense)	(None, 10)	330

dense\_13 (Dense) (None, 1) 11

```
=====  
Total params: 2,048,973  
Trainable params: 2,048,973  
Non-trainable params: 0
```

None

```
-----  
  
Accuracy Score: 88.14%  
F1 Score: 88.10%  
Confusion Matrix:  
[[771  2]  
 [500 320]]
```

Random Binary Model

```
=====  
Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 20
```

Model: "sequential\_7"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 495, 100)	2023000
conv1d_7 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_7 (GlobalMaxPooling1D)	(None, 32)	0
flatten_7 (Flatten)	(None, 32)	0
dense_14 (Dense)	(None, 20)	660
dense_15 (Dense)	(None, 1)	21

```
=====  
Total params: 2,049,313  
Trainable params: 2,049,313  
Non-trainable params: 0
```

None

```
-----  
  
Accuracy Score: 88.26%  
F1 Score: 88.20%  
Confusion Matrix:  
[[772  1]  
 [533 287]]
```

In [ ]:

```
# From Scratch Binary Model  
for a in range(len(filter_size_options)):  
    for b in range(len(kernel_size_options)):  
        for c in range(len(hidden_layer_options)):  
            CNN_Model_Function_Binary('From Scratch Binary Model', my_binary_emb_layer, filter_size_options[a], kernel_size_options[b], hidden_layer_options[c])
```

From Scratch Binary Model

=====

Filter Size: 16  
Kernel Size: 4  
Hidden Layer Size: 10

-----

Model: "sequential\_8"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_8 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_8 (GlobalMaxPooling1D)	(None, 16)	0
flatten_8 (Flatten)	(None, 16)	0
dense_16 (Dense)	(None, 10)	170
dense_17 (Dense)	(None, 1)	11

=====

Total params: 2,529,497  
Trainable params: 2,529,497  
Non-trainable params: 0

---

None

-----

Accuracy Score: 86.82%  
F1 Score: 86.61%  
Confusion Matrix:  
[[773 0]  
 [752 68]]

-----

From Scratch Binary Model

=====

Filter Size: 16  
Kernel Size: 4  
Hidden Layer Size: 20

-----

Model: "sequential\_9"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_9 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_9 (GlobalMaxPooling1D)	(None, 16)	0
flatten_9 (Flatten)	(None, 16)	0
dense_18 (Dense)	(None, 20)	340
dense_19 (Dense)	(None, 1)	21

=====

Total params: 2,529,677  
Trainable params: 2,529,677  
Non-trainable params: 0

---

None

-----

Accuracy Score: 89.58%  
F1 Score: 89.58%  
Confusion Matrix:  
[[773 0]  
 [670 150]]

-----

-----

From Scratch Binary Model  
=====

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 10

-----

Model: "sequential\_10"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_10 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_10 (GlobalMaxPooling1D)	(None, 16)	0
flatten_10 (Flatten)	(None, 16)	0
dense_20 (Dense)	(None, 10)	170
dense_21 (Dense)	(None, 1)	11

=====

Total params: 2,535,897  
Trainable params: 2,535,897  
Non-trainable params: 0

None

-----

-----

Accuracy Score: 89.39%  
F1 Score: 89.40%  
Confusion Matrix:  
[[773 0]  
 [803 17]]

-----

-----

From Scratch Binary Model  
=====

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 20

-----

Model: "sequential\_11"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_11 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_11 (GlobalMaxPooling1D)	(None, 16)	0

flatten_11 (Flatten)	(None, 16)	0
dense_22 (Dense)	(None, 20)	340
dense_23 (Dense)	(None, 1)	21

=====  
Total params: 2,536,077  
Trainable params: 2,536,077  
Non-trainable params: 0

None

-----  
  
Accuracy Score: 88.32%  
F1 Score: 88.33%  
Confusion Matrix:  
[[771 2]  
 [569 251]]

-----  
From Scratch Binary Model

=====  
Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 10

-----  
Model: "sequential\_12"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_12 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_12 (GlobalMaxPooling1D)	(None, 32)	0
flatten_12 (Flatten)	(None, 32)	0
dense_24 (Dense)	(None, 10)	330
dense_25 (Dense)	(None, 1)	11

=====  
Total params: 2,536,073  
Trainable params: 2,536,073  
Non-trainable params: 0

None

-----  
  
Accuracy Score: 87.70%  
F1 Score: 87.86%  
Confusion Matrix:  
[[773 0]  
 [738 82]]

-----  
From Scratch Binary Model

=====  
Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 20

Model: "sequential\_13"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_13 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_13 (GlobalMaxPooling1D)	(None, 32)	0
flatten_13 (Flatten)	(None, 32)	0
dense_26 (Dense)	(None, 20)	660
dense_27 (Dense)	(None, 1)	21

Total params: 2,536,413  
Trainable params: 2,536,413  
Non-trainable params: 0

None

Accuracy Score: 87.01%  
F1 Score: 87.02%  
Confusion Matrix:  
[[771 2]  
 [512 308]]

From Scratch Binary Model

Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 10

Model: "sequential\_14"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_14 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_14 (GlobalMaxPooling1D)	(None, 32)	0
flatten_14 (Flatten)	(None, 32)	0
dense_28 (Dense)	(None, 10)	330
dense_29 (Dense)	(None, 1)	11

Total params: 2,548,873  
Trainable params: 2,548,873  
Non-trainable params: 0

None

Accuracy Score: 88.32%  
F1 Score: 88.36%  
Confusion Matrix:

Confusion Matrix:

```
[[772   1]
 [608 212]]
```

From Scratch Binary Model

```
=====
Filter Size:  32
Kernel Size:  8
Hidden Layer Size:  20
```

Model: "sequential\_15"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 495, 100)	2522900
conv1d_15 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_15 (GlobalMaxPooling1D)	(None, 32)	0
flatten_15 (Flatten)	(None, 32)	0
dense_30 (Dense)	(None, 20)	660
dense_31 (Dense)	(None, 1)	21

```
=====
Total params: 2,549,213
Trainable params: 2,549,213
Non-trainable params: 0
```

None

```
-----
Accuracy Score: 88.39%
F1 Score: 88.37%
Confusion Matrix:
[[769   4]
 [504 316]]
```

In [ ]:

```
# Pretrained Binary Model
for a in range(len(filter_size_options)):
    for b in range(len(kernel_size_options)):
        for c in range(len(hidden_layer_options)):
            CNN_Model_Function_Binary('Pretrained Binary Model', pretrained_binary_emb_layer,
filter_size_options[a], kernel_size_options[b], hidden_layer_options[c])
```

Pretrained Binary Model

```
=====
Filter Size:  16
Kernel Size:  4
Hidden Layer Size:  10
```

Model: "sequential\_16"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_16 (Conv1D)	(None, 488, 16)	6416

conv1d_16 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_16 (GlobalMaxPooling1D)	(None, 16)	0
flatten_16 (Flatten)	(None, 16)	0
dense_32 (Dense)	(None, 10)	170
dense_33 (Dense)	(None, 1)	11

=====  
Total params: 25,391,997  
Trainable params: 25,391,997  
Non-trainable params: 0

None

-----  
Accuracy Score: 85.06%  
F1 Score: 85.00%  
Confusion Matrix:  
[[773 0]  
 [772 48]]

-----  
Pretrained Binary Model

=====  
Filter Size: 16  
Kernel Size: 4  
Hidden Layer Size: 20

-----  
Model: "sequential\_17"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_17 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_17 (GlobalMaxPooling1D)	(None, 16)	0
flatten_17 (Flatten)	(None, 16)	0
dense_34 (Dense)	(None, 20)	340
dense_35 (Dense)	(None, 1)	21

=====  
Total params: 25,392,177  
Trainable params: 25,392,177  
Non-trainable params: 0

None

-----  
Accuracy Score: 87.57%  
F1 Score: 87.39%  
Confusion Matrix:  
[[771 2]  
 [665 155]]

-----  
Pretrained Binary Model



=====

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 10

-----

Model: "sequential\_18"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_18 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_18 (GlobalMaxPooling1D)	(None, 16)	0
flatten_18 (Flatten)	(None, 16)	0
dense_36 (Dense)	(None, 10)	170
dense_37 (Dense)	(None, 1)	11

=====

Total params: 25,398,397  
Trainable params: 25,398,397  
Non-trainable params: 0

None

-----

Accuracy Score: 87.95%  
F1 Score: 87.78%  
Confusion Matrix:  
[[772 1]  
 [662 158]]

-----

Pretrained Binary Model

=====

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 20

-----

Model: "sequential\_19"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_19 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_19 (GlobalMaxPooling1D)	(None, 16)	0
flatten_19 (Flatten)	(None, 16)	0
dense_38 (Dense)	(None, 20)	340
dense_39 (Dense)	(None, 1)	21

=====

Total params: 25,398,577  
Trainable params: 25,398,577  
Non-trainable params: 0

None

Accuracy Score: 87.95%  
F1 Score: 87.90%  
Confusion Matrix:  
[[771 2]  
[532 288]]

Pretrained Binary Model  
=====

Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 10

Model: "sequential\_20"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_20 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_20 (GlobalMaxPooling1D)	(None, 32)	0
flatten_20 (Flatten)	(None, 32)	0
dense_40 (Dense)	(None, 10)	330
dense_41 (Dense)	(None, 1)	11

=====  
Total params: 25,398,573  
Trainable params: 25,398,573  
Non-trainable params: 0

None

Accuracy Score: 90.14%  
F1 Score: 90.13%  
Confusion Matrix:  
[[773 0]  
[804 16]]

Pretrained Binary Model  
=====

Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 20

Model: "sequential\_21"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_21 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_21 (GlobalMaxPooling1D)	(None, 32)	0
flatten_21 (Flatten)	(None, 32)	0

dense_42 (Dense)	(None, 20)	660
dense_43 (Dense)	(None, 1)	21

=====  
Total params: 25,398,913  
Trainable params: 25,398,913  
Non-trainable params: 0

None

-----  
  
Accuracy Score: 89.14%  
F1 Score: 89.05%  
Confusion Matrix:  
[[772 1]  
 [585 235]]

-----  
Pretrained Binary Model

=====  
Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 10

-----  
Model: "sequential\_22"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_22 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_22 (GlobalMaxPooling1D)	(None, 32)	0
flatten_22 (Flatten)	(None, 32)	0
dense_44 (Dense)	(None, 10)	330
dense_45 (Dense)	(None, 1)	11

=====  
Total params: 25,411,373  
Trainable params: 25,411,373  
Non-trainable params: 0

None

-----  
  
Accuracy Score: 89.77%  
F1 Score: 89.69%  
Confusion Matrix:  
[[773 0]  
 [820 0]]

-----  
Pretrained Binary Model

=====  
Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 20

-----  
Model: "sequential\_22"

Model: "sequential\_23"

Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 495, 100)	25385400
conv1d_23 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_23 (GlobalMaxPooling1D)	(None, 32)	0
flatten_23 (Flatten)	(None, 32)	0
dense_46 (Dense)	(None, 20)	660
dense_47 (Dense)	(None, 1)	21

Total params: 25,411,713  
Trainable params: 25,411,713  
Non-trainable params: 0

None

Accuracy Score: 89.33%  
F1 Score: 89.30%  
Confusion Matrix:  
[[771 2]  
 [489 331]]

## Multiclass

## MODEL FUNCTION

In [ ]:

```
def CNN_Model_Function_Multi(operation_multi, emb_layer_multi, filter_size, kernel_size,
hidden_layer_size):

    print(operation_multi)
    print("=====")
    print("Filter Size: ", filter_size)
    print("Kernel Size: ", kernel_size)
    print("Hidden Layer Size: ", hidden_layer_size)
    print("----- \n")

    # Define Model
    model_multi = Sequential()
    model_multi.add(emb_layer_multi)
    model_multi.add(Conv1D(filters=filter_size, kernel_size=kernel_size, activation='relu'
))
    model_multi.add(GlobalMaxPooling1D())
    model_multi.add(Flatten())
    model_multi.add(Dense(hidden_layer_size, activation='relu'))
    model_multi.add(Dense(5, activation='softmax'))
    print(model_multi.summary())

    # Compile model
    model_multi.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy', get_f1])

    print("\n----- \n")
```

```
# Fit
model_multi.fit(padded_train_multi_x, y_train_multi_cnn_p, validation_data=(padded_val_multi_x, y_val_multi_cnn_p), epochs=10, verbose=0)

# Getting score metrics
scores_model_multi = model_multi.evaluate(padded_test_multi_x, y_test_multi_cnn_p, verbose=0)

print("Accuracy Score: %.2f%%" % (scores_model_multi[1]*100))
print("F1 Score: %.2f%%" % (scores_model_multi[2]*100))
print('Confusion Matrix:')
y_predict_model_multi = model_multi.predict(padded_test_multi_x)
y_predict_model_multi = np.argmax(y_predict_model_multi, axis=1)
print(confusion_matrix(y_test_multi_cnn, y_predict_model_multi))

print("\n -----
----- \n")
```

**CALL FUNCTION IN LOOP**

In [ ]:

```
# CALL IN LOOP
# Multiclass
#operation_names_multi = ['Random Multiclass Model', 'From Scratch Multiclass Model', 'Pretrained Multiclass Model']
#embedding_layer_multi = [random_emb_layer_multi, my_multi_emb_layer, pretrained_multi_emb_layer]

# Random Multiclass Model
for a in range(len(filter_size_options)):
    for b in range(len(kernel_size_options)):
        for c in range(len(hidden_layer_options)):
            CNN_Model_Function_Multi('Random Multiclass Model', random_emb_layer_multi, filter_size_options[a], kernel_size_options[b], hidden_layer_options[c])
```

Random Multiclass Model  
=====

Filter Size: 16  
Kernel Size: 4  
Hidden Layer Size: 10  
-----

Model: "sequential\_24"

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 495, 100)	2222100
conv1d_24 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_24 (GlobalMaxPooling1D)	(None, 16)	0
flatten_24 (Flatten)	(None, 16)	0
dense_48 (Dense)	(None, 10)	170
dense_49 (Dense)	(None, 5)	55

=====

Total params: 2,228,741  
Trainable params: 2,228,741  
Non-trainable params: 0

---

None

-----

Accuracy Score: 49.60%  
F1 Score: 49.85%

Confusion Matrix:  
[[244 107 25 7 10]  
[ 82 164 89 33 12]  
[ 23 100 159 100 25]  
[ 10 25 86 167 91]  
[ 8 16 33 126 258]]

Random Multiclass Model

Filter Size: 16  
Kernel Size: 4  
Hidden Layer Size: 20

Model: "sequential\_25"

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 495, 100)	2222100
conv1d_25 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_25 (GlobalMaxPooling1D)	(None, 16)	0
flatten_25 (Flatten)	(None, 16)	0
dense_50 (Dense)	(None, 20)	340
dense_51 (Dense)	(None, 5)	105

Total params: 2,228,961  
Trainable params: 2,228,961  
Non-trainable params: 0

None

Accuracy Score: 47.15%  
F1 Score: 47.05%  
Confusion Matrix:  
[[218 141 13 7 14]  
[ 79 179 89 23 10]  
[ 20 112 157 101 17]  
[ 15 24 81 174 85]  
[ 15 10 45 156 215]]

Random Multiclass Model

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 10

Model: "sequential\_26"

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 495, 100)	2222100
conv1d_26 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_26 (GlobalMaxPooling1D)	(None, 16)	0

flatten_26 (Flatten)	(None, 16)	0
dense_52 (Dense)	(None, 10)	170
dense_53 (Dense)	(None, 5)	55

=====  
Total params: 2,235,141  
Trainable params: 2,235,141  
Non-trainable params: 0

---

None

-----  
  
Accuracy Score: 49.10%  
F1 Score: 49.13%  
Confusion Matrix:

```
[[241 112 25 5 10]
 [ 90 160 92 27 11]
 [ 27 87 168 102 23]
 [ 27 11 91 178 72]
 [ 15 9 40 142 235]]
```

-----  
Random Multiclass Model

=====  
Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 20

-----  
Model: "sequential\_27"

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 495, 100)	2222100
conv1d_27 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_27 (GlobalMaxPooling1D)	(None, 16)	0
flatten_27 (Flatten)	(None, 16)	0
dense_54 (Dense)	(None, 20)	340
dense_55 (Dense)	(None, 5)	105

=====  
Total params: 2,235,361  
Trainable params: 2,235,361  
Non-trainable params: 0

---

None

-----  
  
Accuracy Score: 48.55%  
F1 Score: 48.64%  
Confusion Matrix:

```
[[220 132 21 4 16]
 [ 71 190 84 23 12]
 [ 20 94 164 102 27]
 [ 15 30 82 166 86]
 [ 19 13 40 138 231]]
```

```
Random Multiclass Model
=====
Filter Size:  32
Kernel Size:  4
Hidden Layer Size:  10
-----

Model: "sequential_28"

Layer (type)                Output Shape                Param #
=====
embedding_1 (Embedding)      (None, 495, 100)           2222100
conv1d_28 (Conv1D)           (None, 492, 32)            12832
global_max_pooling1d_28 (Gl  (None, 32)                  0
obalMaxPooling1D)
flatten_28 (Flatten)         (None, 32)                  0
dense_56 (Dense)             (None, 10)                  330
dense_57 (Dense)             (None, 5)                   55

=====
Total params: 2,235,317
Trainable params: 2,235,317
Non-trainable params: 0

None
-----
```

```
Accuracy Score: 48.20%
F1 Score: 48.21%
Confusion Matrix:
[[234 111  32   4  12]
 [ 82 156 104  19  19]
 [ 31  77 164 103  32]
 [ 18  27  85 167  82]
 [ 16  13  40 129 243]]
-----
-----
```

```
Random Multiclass Model
=====
Filter Size:  32
Kernel Size:  4
Hidden Layer Size:  20
-----

Model: "sequential_29"

Layer (type)                Output Shape                Param #
=====
embedding_1 (Embedding)      (None, 495, 100)           2222100
conv1d_29 (Conv1D)           (None, 492, 32)            12832
global_max_pooling1d_29 (Gl  (None, 32)                  0
obalMaxPooling1D)
flatten_29 (Flatten)         (None, 32)                  0
dense_58 (Dense)             (None, 20)                  660
dense_59 (Dense)             (None, 5)                   105

=====
Total params: 2,235,697
```



Trainable params: 2,235,697  
Non-trainable params: 0

None

-----  
Accuracy Score: 47.00%  
F1 Score: 47.06%  
Confusion Matrix:  
[[224 124 27 7 11]  
 [ 75 177 91 21 16]  
 [ 21 97 150 105 34]  
 [ 18 32 80 159 90]  
 [ 19 16 47 129 230]]

-----  
Random Multiclass Model

=====

Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 10

-----  
Model: "sequential\_30"

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 495, 100)	2222100
conv1d_30 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_30 (GlobalMaxPooling1D)	(None, 32)	0
flatten_30 (Flatten)	(None, 32)	0
dense_60 (Dense)	(None, 10)	330
dense_61 (Dense)	(None, 5)	55

=====

Total params: 2,248,117  
Trainable params: 2,248,117  
Non-trainable params: 0

-----  
None

-----  
Accuracy Score: 48.65%  
F1 Score: 48.64%  
Confusion Matrix:  
[[237 113 28 9 6]  
 [ 91 157 96 22 14]  
 [ 36 73 165 106 27]  
 [ 25 20 74 183 77]  
 [ 23 12 33 142 231]]

-----  
Random Multiclass Model

=====

Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 20

-----  
Model: "sequential\_31"

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 495, 100)	2222100
conv1d_31 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_31 (GlobalMaxPooling1D)	(None, 32)	0
flatten_31 (Flatten)	(None, 32)	0
dense_62 (Dense)	(None, 20)	660
dense_63 (Dense)	(None, 5)	105

Total params: 2,248,497  
 Trainable params: 2,248,497  
 Non-trainable params: 0

None

Accuracy Score: 47.05%

F1 Score: 47.17%

Confusion Matrix:

```

[[221 126 25 8 13]
 [ 86 167 90 23 14]
 [ 33 87 150 104 33]
 [ 18 25 77 171 88]
 [ 18 17 35 139 232]]
  
```

In [ ]:

```

# From Scratch Multiclass Model
for a in range(len(filter_size_options)):
    for b in range(len(kernel_size_options)):
        for c in range(len(hidden_layer_options)):
            CNN_Model_Function_Multi('From Scratch Multiclass Model', my_multi_emb_layer, filter_size_options[a], kernel_size_options[b], hidden_layer_options[c])
  
```

From Scratch Multiclass Model

Filter Size: 16

Kernel Size: 4

Hidden Layer Size: 10

Model: "sequential\_32"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_32 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_32 (GlobalMaxPooling1D)	(None, 16)	0
flatten_32 (Flatten)	(None, 16)	0
dense_64 (Dense)	(None, 10)	170
dense_65 (Dense)	(None, 5)	55

Total params: 2,849,641  
Trainable params: 2,849,641  
Non-trainable params: 0

None

-----  
Accuracy Score: 46.90%  
F1 Score: 46.49%  
Confusion Matrix:  
[[257 89 27 5 15]  
 [111 151 79 21 18]  
 [ 30 112 137 93 35]  
 [ 11 44 62 146 116]  
 [ 11 33 36 114 247]]

-----  
From Scratch Multiclass Model  
=====

Filter Size: 16  
Kernel Size: 4  
Hidden Layer Size: 20

Model: "sequential\_33"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_33 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_33 (GlobalMaxPooling1D)	(None, 16)	0
flatten_33 (Flatten)	(None, 16)	0
dense_66 (Dense)	(None, 20)	340
dense_67 (Dense)	(None, 5)	105

=====

Total params: 2,849,861  
Trainable params: 2,849,861  
Non-trainable params: 0

None

-----  
Accuracy Score: 49.80%  
F1 Score: 49.69%  
Confusion Matrix:  
[[221 126 32 3 11]  
 [ 77 180 85 24 14]  
 [ 18 93 188 74 34]  
 [ 6 29 92 156 96]  
 [ 6 22 45 117 251]]

-----  
From Scratch Multiclass Model  
=====

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 10

Model: "sequential\_34"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_34 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_34 (GlobalMaxPooling1D)	(None, 16)	0
flatten_34 (Flatten)	(None, 16)	0
dense_68 (Dense)	(None, 10)	170
dense_69 (Dense)	(None, 5)	55

Total params: 2,856,041  
Trainable params: 2,856,041  
Non-trainable params: 0

None

Accuracy Score: 47.35%  
F1 Score: 47.48%  
Confusion Matrix:  
[[217 139 14 5 18]  
 [ 89 188 64 17 22]  
 [ 22 108 142 100 35]  
 [ 2 37 71 143 126]  
 [ 9 30 27 118 257]]

From Scratch Multiclass Model

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 20

Model: "sequential\_35"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_35 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_35 (GlobalMaxPooling1D)	(None, 16)	0
flatten_35 (Flatten)	(None, 16)	0
dense_70 (Dense)	(None, 20)	340
dense_71 (Dense)	(None, 5)	105

Total params: 2,856,261  
Trainable params: 2,856,261  
Non-trainable params: 0

None

Accuracy Score: 47.00%

F1 Score: 46.98%  
Confusion Matrix:  
[[220 131 24 7 11]  
[ 89 179 75 30 7]  
[ 23 104 148 102 30]  
[ 9 47 69 171 83]  
[ 16 29 32 142 222]]

-----  
-----  
  
From Scratch Multiclass Model  
=====

Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 10

-----

Model: "sequential\_36"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_36 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_36 (GlobalMaxPooling1D)	(None, 32)	0
flatten_36 (Flatten)	(None, 32)	0
dense_72 (Dense)	(None, 10)	330
dense_73 (Dense)	(None, 5)	55

=====

Total params: 2,856,217  
Trainable params: 2,856,217  
Non-trainable params: 0

-----  
-----  
  
None

-----  
-----  
  
Accuracy Score: 47.25%  
F1 Score: 47.25%  
Confusion Matrix:  
[[214 132 26 5 16]  
[ 81 184 71 28 16]  
[ 23 100 149 104 31]  
[ 10 32 85 159 93]  
[ 12 18 40 132 239]]

-----  
-----  
  
From Scratch Multiclass Model  
=====

Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 20

-----

Model: "sequential\_37"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_37 (Conv1D)	(None, 492, 32)	12832
global max pooling1d 37 (GlobalMaxPooling1D)	(None, 32)	0

```
GlobalMaxPooling1D)

flatten_37 (Flatten)          (None, 32)          0

dense_74 (Dense)              (None, 20)          660

dense_75 (Dense)              (None, 5)           105
```

```
=====
Total params: 2,856,597
Trainable params: 2,856,597
Non-trainable params: 0
```

---

None

-----

Accuracy Score: 48.00%

F1 Score: 48.20%

Confusion Matrix:

```
[[230 121  23   6  13]
 [ 84 168  83  24  21]
 [ 29 100 162  86  30]
 [ 17  38  87 150  87]
 [ 13  21  37 120 250]]
```

-----

From Scratch Multiclass Model

=====

Filter Size: 32

Kernel Size: 8

Hidden Layer Size: 10

-----

Model: "sequential\_38"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_38 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_38 (GlobalMaxPooling1D)	(None, 32)	0
flatten_38 (Flatten)	(None, 32)	0
dense_76 (Dense)	(None, 10)	330
dense_77 (Dense)	(None, 5)	55

=====

Total params: 2,869,017

Trainable params: 2,869,017

Non-trainable params: 0

---

None

-----

Accuracy Score: 45.50%

F1 Score: 45.63%

Confusion Matrix:

```
[[213 126  29  12  13]
 [ 92 161  84  23  20]
 [ 30 101 127 104  45]
 [ 17  34  69 154 105]
 [ 16  20  23 127 255]]
```

-----

From Scratch Multiclass Model

=====

Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 20

-----

Model: "sequential\_39"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 495, 100)	2843000
conv1d_39 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_39 (GlobalMaxPooling1D)	(None, 32)	0
flatten_39 (Flatten)	(None, 32)	0
dense_78 (Dense)	(None, 20)	660
dense_79 (Dense)	(None, 5)	105

=====

Total params: 2,869,397  
Trainable params: 2,869,397  
Non-trainable params: 0

None

-----

Accuracy Score: 46.10%  
F1 Score: 46.10%  
Confusion Matrix:  
[[217 134 21 8 13]  
 [ 82 186 66 25 21]  
 [ 31 110 130 102 34]  
 [ 16 37 71 167 88]  
 [ 17 21 25 156 222]]

-----

-----

In [ ]:

```
# Pretrained Multiclass Model
for a in range(len(filter_size_options)):
    for b in range(len(kernel_size_options)):
        for c in range(len(hidden_layer_options)):
            CNN_Model_Function_Multi('Pretrained Multiclass Model', pretrained_multi_emb_layer
, filter_size_options[a], kernel_size_options[b], hidden_layer_options[c])
```

Pretrained Multiclass Model

=====

Filter Size: 16  
Kernel Size: 4  
Hidden Layer Size: 10

-----

Model: "sequential\_40"

Layer (type)	Output Shape	Param #
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_40 (Conv1D)	(None, 492, 16)	6416

global_max_pooling1d_40 (GlobalMaxPooling1D)	(None, 16)	0
flatten_40 (Flatten)	(None, 16)	0
dense_80 (Dense)	(None, 10)	170
dense_81 (Dense)	(None, 5)	55

=====  
Total params: 25,392,041  
Trainable params: 25,392,041  
Non-trainable params: 0

---

None

-----  
Accuracy Score: 43.05%

F1 Score: 41.65%

Confusion Matrix:

```
[[226  97  53   4  13]
 [ 91 138 107  24  20]
 [ 51  88 166  72  30]
 [ 19  38 117 121  84]
 [ 25  19  65 122 210]]
```

-----  
Pretrained Multiclass Model

=====  
Filter Size: 16

Kernel Size: 4

Hidden Layer Size: 20  
-----

Model: "sequential\_41"

Layer (type)	Output Shape	Param #
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_41 (Conv1D)	(None, 492, 16)	6416
global_max_pooling1d_41 (GlobalMaxPooling1D)	(None, 16)	0
flatten_41 (Flatten)	(None, 16)	0
dense_82 (Dense)	(None, 20)	340
dense_83 (Dense)	(None, 5)	105

=====  
Total params: 25,392,261  
Trainable params: 25,392,261  
Non-trainable params: 0

---

None

-----  
Accuracy Score: 46.10%

F1 Score: 44.85%

Confusion Matrix:

```
[[270  63  41  10   9]
 [122 115  99  29  15]
 [ 40  84 146  91  46]
 [ 13  41  80 121 124]
 [ 12  15  54  90 270]]
```



-----  
-----  
Pretrained Multiclass Model  
=====

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 10  
-----

Model: "sequential\_42"

Layer (type)	Output Shape	Param #
=====		
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_42 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_42 (GlobalMaxPooling1D)	(None, 16)	0
flatten_42 (Flatten)	(None, 16)	0
dense_84 (Dense)	(None, 10)	170
dense_85 (Dense)	(None, 5)	55

=====

Total params: 25,398,441  
Trainable params: 25,398,441  
Non-trainable params: 0

-----  
None

-----  
-----  
Accuracy Score: 44.85%  
F1 Score: 44.92%  
Confusion Matrix:  
[[198 146 35 10 4]  
 [ 80 169 86 37 8]  
 [ 20 107 161 96 23]  
 [ 8 31 91 164 85]  
 [ 8 29 36 163 205]]

-----  
-----  
Pretrained Multiclass Model  
=====

Filter Size: 16  
Kernel Size: 8  
Hidden Layer Size: 20  
-----

Model: "sequential\_43"

Layer (type)	Output Shape	Param #
=====		
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_43 (Conv1D)	(None, 488, 16)	12816
global_max_pooling1d_43 (GlobalMaxPooling1D)	(None, 16)	0
flatten_43 (Flatten)	(None, 16)	0
dense_86 (Dense)	(None, 20)	340
dense_87 (Dense)	(None, 5)	105

=====  
Total params: 25,398,661  
Trainable params: 25,398,661  
Non-trainable params: 0

-----  
None  
-----  
  
Accuracy Score: 47.50%  
F1 Score: 47.29%  
Confusion Matrix:  
[[219 124 30 14 6]  
 [ 68 180 97 24 11]  
 [ 24 88 180 93 22]  
 [ 7 28 102 167 75]  
 [ 9 17 61 150 204]]  
  
-----  
-----

Pretrained Multiclass Model  
=====  
Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 10  
-----

Model: "sequential\_44"

Layer (type)	Output Shape	Param #
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_44 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_44 (GlobalMaxPooling1D)	(None, 32)	0
flatten_44 (Flatten)	(None, 32)	0
dense_88 (Dense)	(None, 10)	330
dense_89 (Dense)	(None, 5)	55

=====  
Total params: 25,398,617  
Trainable params: 25,398,617  
Non-trainable params: 0  
  
-----  
None  
-----

-----  
  
Accuracy Score: 47.70%  
F1 Score: 47.90%  
Confusion Matrix:  
[[223 134 27 6 3]  
 [ 69 192 89 25 5]  
 [ 13 113 174 90 17]  
 [ 3 33 102 152 89]  
 [ 6 17 40 165 213]]  
  
-----  
-----

Pretrained Multiclass Model  
=====  
Filter Size: 32  
Kernel Size: 4  
Hidden Layer Size: 20  
-----

Model: "sequential\_45"

Layer (type)	Output Shape	Param #
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_45 (Conv1D)	(None, 492, 32)	12832
global_max_pooling1d_45 (GlobalMaxPooling1D)	(None, 32)	0
flatten_45 (Flatten)	(None, 32)	0
dense_90 (Dense)	(None, 20)	660
dense_91 (Dense)	(None, 5)	105

Total params: 25,398,997  
Trainable params: 25,398,997  
Non-trainable params: 0

None

-----

Accuracy Score: 47.95%  
F1 Score: 47.86%  
Confusion Matrix:  
[[232 114 29 6 12]  
 [ 79 171 73 39 18]  
 [ 22 98 153 105 29]  
 [ 8 33 81 172 85]  
 [ 14 16 38 142 231]]

-----  
-----

Pretrained Multiclass Model

Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 10

Model: "sequential\_46"

Layer (type)	Output Shape	Param #
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_46 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_46 (GlobalMaxPooling1D)	(None, 32)	0
flatten_46 (Flatten)	(None, 32)	0
dense_92 (Dense)	(None, 10)	330
dense_93 (Dense)	(None, 5)	55

Total params: 25,411,417  
Trainable params: 25,411,417  
Non-trainable params: 0

None

-----

Accuracy Score: 47.00%  
F1 Score: 46.93%  
Confusion Matrix:  
[[231 125 23 9 5]  
[ 81 184 84 23 8]  
[ 25 103 167 94 18]  
[ 7 39 103 156 74]  
[ 15 17 53 154 202]]

-----  
-----  
  
Pretrained Multiclass Model  
=====

Filter Size: 32  
Kernel Size: 8  
Hidden Layer Size: 20

-----

Model: "sequential\_47"

Layer (type)	Output Shape	Param #
embedding_5 (Embedding)	(None, 495, 100)	25385400
conv1d_47 (Conv1D)	(None, 488, 32)	25632
global_max_pooling1d_47 (GlobalMaxPooling1D)	(None, 32)	0
flatten_47 (Flatten)	(None, 32)	0
dense_94 (Dense)	(None, 20)	660
dense_95 (Dense)	(None, 5)	105

=====

Total params: 25,411,797  
Trainable params: 25,411,797  
Non-trainable params: 0

---

None

-----  
-----

Accuracy Score: 46.75%  
F1 Score: 46.75%  
Confusion Matrix:  
[[222 129 21 13 8]  
[ 83 170 84 33 10]  
[ 27 100 151 108 21]  
[ 12 32 78 174 83]  
[ 13 16 30 164 218]]

# My Report

## About the Dataset

Train and Test dataset used in this project is a customer reviews data collection from Yelp. Customer reviews labeled based on their sentiments. Both datasets consist of two columns

- **text:** Consists of raw text of customer reviews.
- **label:** Consists of the multiclass label of sentiments of a given customer review. There are 5 levels of sentiments [1,2,3,4,5].

## Preprocessing and Preparing the Dataset

## Preprocessing

**After reading the dataset successfully, dataset needs to be preprocessed in order to obtain better accuracies from the models. Operations carried out in this step are:**

- **Lowercasing**

**Lowercased all the customer comments in the “text” column of the train and test dataset**

- **Stop-word Removal**

**After tokenizing the texts using `word_tokenize()` function, some additional stop-words are added to the existed English stop-words list and then these words removed from the datasets.**

**Additional stop-words:** ['n't', 'm', 's', 've', 'll', 're']

- **Stemming**

**The customer comments in the datasets reduced to their stems using the Porter Stemmer.**

**After finalizing all these operation to preprocess the texts, words put together using the join() function.**

**Original Text:**

	Unnamed: 0	text	label
0	0	I came here and left a review before but last ...	1
1	1	Had a very nice first visit here. The owner Te...	4
2	2	This is a gorgeous and very clean hotel. We h...	4
3	3	The gym is dirty. I have given up. Locker ro...	1
4	4	The food here is delicious, fast, and consiste...	5

### Preprocessed Text:

	Unnamed: 0	text	label
0	0	came left review last time get food poison unl...	1
1	1	nice first visit owner ted friendli start rest...	4
2	2	gorgeou clean hotel room west wing first chore...	4
3	3	gym dirti given locker room total dirti manag ...	1
4	4	food delici fast consist everi singl time gene...	5

## Preparing the Binary and Multiclass Datasets

**For this project, two different datasets created out of this original Train and Test datasets.**

## Binary Dataset

**Get rid of the class 3 in the dataset and map class 1 and 2 to 0, and class 4 and 5 to 1.**

***Train:***

	Unnamed: 0		text	label
0	0		came left review last time get food poison unl...	0
1	1		nice first visit owner ted friendli start rest...	1
2	2		gorgeou clean hotel room west wing first chore...	1
3	3		gym dirti given locker room total dirti manag ...	0
4	4		food delici fast consist everi singl time gene...	1,

### ***Test:***

	Unnamed: 0		text	label
0	0		stay weekend made stay pleasant locat great sp...	1
1	1		forev call upon delici design whenev need tast...	1
2	2		person order homicid boneless regular absolut ...	1
3	3		eat pretti much everytim go tarpon spring staf...	1
5	5		great hidden tavern grill wonder old place ear...	1,

### **Multiclass Dataset:**

Shift classes to the range 0 to 4. [5 → 4, 4 → 3, 3 → 2, 2 → 1, 1 → 0]

### ***Train:***

	Unnamed: 0		text	label
0	0		came left review last time get food poison unl...	0
1	1		nice first visit owner ted friendli start rest...	3
2	2		gorgeou clean hotel room west wing first chore...	3
3	3		gym dirti given locker room total dirti manag ...	0
4	4		food delici fast consist everi singl time gene...	4,

### ***Test:***

	Unnamed: 0		text	label
0	0		stay weekend made stay pleasant locat great sp...	4
1	1		forev call upon delici design whenev need tast...	4
2	2		person order homicid boneless regular absolut ...	3
3	3		eat pretti much everytim go tarpon spring staf...	3
4	4		time never go want huge chang -- usual trim ge...	2)

## **Text Classification with Naïve Bayes (Non-Neural Model)**

### **Initializing the Pipeline and GridSearchCV Parameters**

#### ***Pipeline***

The pipeline which will be passed to GridSearchCV consist of three main components which are TfidfVectorizer, DenseTransformer, and Naïve Bayes classifier. Sklearn's TfidfVectorizer is used for representing data as vectors. The class DenseTransformer is used for converting sparse matrix output of TfidfVectorizer to dense matrix for feeding into classifier. Lastly the classifier that will be used in this part of the project will be GaussianNB.

This pipeline will be applied both for Binary and Multiclass datasets.

```
18 NB_pipeline = Pipeline([
19     ('vectorizer_tfidf', TfidfVectorizer()),
20     ('to_dense', DenseTransformer()),
21     ('classifier_NB', GaussianNB())
22 ])
```

### **GridSearchCV Parameters**

The parameters that will be passed to the GridSearchCV are TfidfVectorizer's "ngram\_range" and "min\_df". N-grams will be up to 3-grams and, since, for a term to be valid it should exist in at least N documents, therefore the min\_df values determined as 100, 500, and 1000.

This GridSearchCV parameters will be applied both for Binary and Multiclass datasets.

```
27 grid_params_NB = {
28     'vectorizer_tfidf_ngram_range': [(1,1), (1,2), (1,3)],
29     'vectorizer_tfidf_min_df': [100, 500, 1000]
30 }
```

### Train GridSearchCV with Binary Classification Dataset

GridSearchCV will be trained separately for Binary and Multiclass datasets.

#### *Binary*

```
4 grid_search_NB_Binary = GridSearchCV(
5     NB_pipeline,
6     grid_params_NB,
7     cv=5,
8     verbose=1,
9     scoring='f1_macro')
10
11
12 # Fit binary dataset to the grid_search_NB
13 grid_search_NB_Binary.fit(train_binary_x, train_binary_y)
14 print(grid_search_NB_Binary.cv_results_)

Fitting 5 folds for each of 9 candidates, totalling 45 fits
```

#### *Multiclass*

```
4 grid_search_NB_Multi = GridSearchCV(
5     NB_pipeline,
6     grid_params_NB,
7     cv=5,
8     verbose=1,
9     scoring='f1_macro')
10
11
12 # Fit multiclass dataset to the grid_search_NB
13 grid_search_NB_Multi.fit(train_multi_x, train_multi_y)
14 print(grid_search_NB_Multi.cv_results_)

Fitting 5 folds for each of 9 candidates, totalling 45 fits
```

### Report the Results

Mean Test Score, Standard Deviation Test Score, Min - Max Split Test Scores will be reported for each 9 hyperparameter group. Among these hyperparameter groups, best parameter set for given dataset and hyperparameter space will be selected. In the next step, the best parameters reported for both Binary and Multiclass datasets will be used for setting the pipeline again and making predictions.

#### *Binary*

##### *Parameter Groups (First 5)*

```
Parameter Group 1
Parameters: {'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 1)}
Mean Test Score: 0.8542374043833949
Standard Deviation Test Score: 0.005122799448803231
Min: 0.8462872819793779
Max: 0.8601170471665417
```

Max: 0.8601179471663417

#### Parameter Group 2

Parameters: {'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 2)}  
Mean Test Score: 0.8600638910720917  
Standard Deviation Test Score: 0.007341615443068069  
Min: 0.8497549243583071  
Max: 0.8719120033225389

#### Parameter Group 3

Parameters: {'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}  
Mean Test Score: 0.860200781216163  
Standard Deviation Test Score: 0.0067386139688290665  
Min: 0.8497549243583071  
Max: 0.8705205202491479

#### Parameter Group 4

Parameters: {'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 1)}  
Mean Test Score: 0.8130507007904967  
Standard Deviation Test Score: 0.0036206581525370014  
Min: 0.8073562350927301  
Max: 0.8174835487027747

#### Parameter Group 5

Parameters: {'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 2)}  
Mean Test Score: 0.812495799919392  
Standard Deviation Test Score: 0.0042717176005138404  
Min: 0.807010315041532  
Max: 0.8192178647456403

### **Best Parameters**

```
Best Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
```

### **Multiclass**

#### **Parameter Groups (First 5)**

##### Parameter Group 1

Parameters: {'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 1)}  
Mean Test Score: 0.4445012872862179  
Standard Deviation Test Score: 0.007110661446049768  
Min: 0.4305396111761393  
Max: 0.4502131932296366

##### Parameter Group 2

Parameters: {'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 2)}  
Mean Test Score: 0.4561186691168338  
Standard Deviation Test Score: 0.007752963984642025  
Min: 0.4444113254577638  
Max: 0.4669184291625168

##### Parameter Group 3

Parameters: {'vectorizer\_tfidf\_\_min\_df': 100, 'vectorizer\_tfidf\_\_ngram\_range': (1, 3)}  
Mean Test Score: 0.4567319032363632  
Standard Deviation Test Score: 0.00815789056184797  
Min: 0.4439244395806406  
Max: 0.467845173574052

##### Parameter Group 4

Parameters: {'vectorizer\_tfidf\_\_min\_df': 500, 'vectorizer\_tfidf\_\_ngram\_range': (1, 1)}  
Mean Test Score: 0.4521197195064187



```
Standard Deviation Test Score: 0.009624668608540174
Min: 0.4349925882462452
Max: 0.46243629865916047
```

```
Parameter Group 5
Parameters: {'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}
Mean Test Score: 0.453092120097463
Standard Deviation Test Score: 0.010165705629250173
Min: 0.4346992227334148
Max: 0.46343308487899926
```

### ***Best Parameters***

```
Best Parameters: {'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
```

### **Setting Pipeline with Best Parameters / Making Predictions / Reporting the Scores and Confusion Matrix**

Best parameters obtained for both Binary and Multiclass datasets will be passed to the pipeline again. These new pipelines will be fitted to the training sets of the each datasets and predictions will be made.

#### ***Binary***

Obtained F1-Macro and Accuracy Scores and Confusion Matrix for the Binary classification.

```
F1 Macro for Binary Classification with Best Parameters:
0.8687712178561673

Accuracy Score for Binary Classification with Best Parameters:
0.8688010043942247

Confusion Matrix:
array([[680, 93],
       [116, 704]])
```

#### ***Multiclass***

Obtained F1-Macro and Accuracy Scores and Confusion Matrix for the Multiclass classification.

```
F1 Macro for Multiclass Classification with Best Parameters:
0.47526559258349055

Accuracy Score for Multiclass Classification with Best Parameters:
0.4915

Confusion Matrix:
array([[264, 91, 18, 5, 15],
       [ 96, 156, 73, 21, 34],
       [ 46, 78, 116, 95, 72],
       [ 20, 32, 60, 133, 134],
       [ 26, 13, 19, 69, 314]])
```

## **Text Classification with Logistic Regression (Non-Neural Model)**

### **Initializing the Pipeline and GridSearchCV Parameters**

#### ***Pipeline***

The pipeline which will be passed to GridSearchCV consist of two main components which are TfidfVectorizer, and Logistic Regression classifier. Sklearn's TfidfVectorizer is used for representing data as vectors. Lastly the classifier that will be used in this part of the project will be LogisticRegression with random state 22.

This pipeline will be applied both for Binary and Multiclass datasets.

```
5 LR_pipeline = Pipeline([
6     ('vectorizer_tfidf', TfidfVectorizer()),
```

```

7     ('classifier_LR', LogisticRegression(random_state=22))
8 ])

```

## GridSearchCV Parameters

The parameters that will be passed to the GridSearchCV are TfidfVectorizer's "ngram\_range" and "min\_df"; and LogisticRegression's "l1\_ratio". N-grams will be up to 3-grams and, since, for a term to be valid it should exist in at least N documents, therefore the min\_df values determined as 100, 500, and 1000. Lastly, "l1\_ratio" will be used to try different values [0.0, 0.5, 0.1] for regularization distribution between L1 and L2 regularization. For  $0 < l1\_ratio < 1$ , the penalty is a combination of L1 and L2.

- l1\_ratio = 0 means using the penalty L2.
- l1\_ratio = 1 means using the penalty L1.

This GridSearchCV parameters will be applied both for Binary and Multiclass datasets.

```

12 grid_params_LR = {
13     'vectorizer_tfidf_ngram_range': [(1,1), (1,2), (1,3)],
14     'vectorizer_tfidf_min_df': [100, 500, 1000],
15     'classifier_LR_l1_ratio': [0.0, 0.5, 1.0]
16 }

```

## Train GridSearchCV with Binary Classification Dataset

GridSearchCV will be trained separately for Binary and Multiclass datasets.

### Binary

```

3 grid_search_LR_Binary = GridSearchCV(
4     LR_pipeline,
5     grid_params_LR,
6     cv = 5,
7     verbose = 1,
8     scoring = 'f1_macro')
9
10 # Fit binary dataset to the grid_search_LR
11 grid_search_LR_Binary.fit(train_binary_x, train_binary_y)
12 print(grid_search_LR_Binary.cv_results_)

```

Fitting 5 folds for each of 27 candidates, totalling 135 fits

### Multiclass

```

3 grid_search_LR_Multi = GridSearchCV(
4     LR_pipeline,
5     grid_params_LR,
6     cv = 5,
7     verbose = 1,
8     scoring = 'f1_macro')
9
10 # Fit binary dataset to the grid_search_LR
11 grid_search_LR_Multi.fit(train_multi_x, train_multi_y)
12 print(grid_search_LR_Multi.cv_results_)

```

Fitting 5 folds for each of 27 candidates, totalling 135 fits

## Report the Results

Mean Test Score and Standard Deviation Test Score will be reported for each 27 hyperparameter group. Among these hyperparameter groups, best parameter set for given dataset and hyperparameter space will be selected. In the next step, the best parameters reported for both Binary and Multiclass datasets will be used for setting the pipeline again and making predictions.

Parameter Groups (First 5)

```
Parameter Group 1
Parameters: {'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 1)}
Mean Test Score: 0.9039270734505316
Standard Deviation Test Score: 0.0035760217266048873
Min: 0.8993329265636055
Max: 0.9100978866625582

Parameter Group 2
Parameters: {'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 2)}
Mean Test Score: 0.9053172864559327
Standard Deviation Test Score: 0.004469668842880747
Min: 0.9014178791573693
Max: 0.9135706352037654

Parameter Group 3
Parameters: {'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 3)}
Mean Test Score: 0.9052478213686159
Standard Deviation Test Score: 0.004341727994197243
Min: 0.9014178791573693
Max: 0.9132233097671819

Parameter Group 4
Parameters: {'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 500, 'vectorizer_tfidf_ngram_range': (1, 1)}
Mean Test Score: 0.8530601176631698
Standard Deviation Test Score: 0.004936023300777656
Min: 0.8471227272946544
Max: 0.85976277196431

Parameter Group 5
Parameters: {'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 500, 'vectorizer_tfidf_ngram_range': (1, 2)}
Mean Test Score: 0.8533389853094251
Standard Deviation Test Score: 0.0045829631254465574
Min: 0.8484989270469089
Max: 0.8590705800139762
```

Best Parameters

```
Best Parameters: {'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 2)}
```

Multiclass

Parameter Groups (First 5)

```
Parameter Group 1
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 1)}
0.533278157858257
0.005151607577326998
Min: 0.5243015182286054
Max: 0.5380659333066168

Parameter Group 2
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 2)}
0.536312109663449
0.0049908379275617675
Min: 0.5284286664610763
Max: 0.5434247275415489

Parameter Group 3
{'classifier_LR_l1_ratio': 0.0, 'vectorizer_tfidf_min_df': 100, 'vectorizer_tfidf_ngram_range': (1, 3)}
0.5364166486996111
0.005090054209611724
Min: 0.5280977399670365
Max: 0.5432273768095992
```

```

Parameter Group  4
{'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 1)}
0.506878694432378
0.009500533533121237
Min:  0.48907890289351696
Max:  0.5149015726124225

Parameter Group  5
{'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 500, 'vectorizer_tfidf__ngram_range': (1, 2)}
0.5058303519343349
0.01057041888035273
Min:  0.48540494509154614
Max:  0.5137127724584536

```

### ***Best Parameters***

```
Best Parameters: {'classifier_LR__l1_ratio': 0.0, 'vectorizer_tfidf__min_df': 100, 'vectorizer_tfidf__ngram_range': (1, 3)}
```

## **Setting Pipeline with Best Parameters / Making Predictions / Reporting the Scores and Confusion Matrix**

Best parameters obtained for both Binary and Multiclass datasets will be passed to the pipeline again. These new pipelines will be fitted to the training sets of the each datasets and predictions will be made.

### ***Binary***

Obtained F1-Macro and Accuracy Scores and Confusion Matrix for the Binary classification.

```

F1 Macro for Binary Classification with Best Parameters:
0.908329917386643

Accuracy Score for Binary Classification with Best Parameters:
0.9083490269930948

Confusion Matrix:
array([[712,  61],
       [ 85, 735]])

```

### ***Multiclass***

Obtained F1-Macro and Accuracy Scores and Confusion Matrix for the Multiclass classification.

```

F1 Macro for Multiclass Classification with Best Parameters:
0.5537120257977514

Accuracy Score for Multiclass Classification with Best Parameters:
0.557

Confusion Matrix:
array([[278,  89,  14,   6,   6],
       [ 83, 189,  79,  18,  11],
       [ 26,  83, 187,  91,  20],
       [ 12,  27,  72, 171,  97],
       [  9,  11,  21, 111, 289]])

```

## **Text Classification with Convolutional Neural Network (CNN) (Neural Model)**

### **Preparing Datasets and Creating the Validation Datasets**

A Validation dataset for Binary and Multiclass datasets is created using `train_test_split` function with `random_state=22` and `test_size = 0.1`.

Additionally, y-values (label) of the train - test - validation multiclass datasets are arranged using `to_categorical` function.

## Length of the Longest Sentences

Length of the longest sentences is found 495. This information will be useful in the padding and creating embedding layers stages of the CNN part of this project.

## Tokenization

In this stage, firstly two different keras Tokenizer() is initialized; one for handling binary datasets, other one for handling multiclass datasets. Secondly, fit\_on\_text() operation applied on both binary and multiclass train datasets. Lastly, both binary and multiclass train - test - validation datasets tokenized using the keras tokenizers text\_to\_sequence() function.

## Padding

Both binary and multiclass train - test - validation datasets padded using the pad\_sequence() function which takes the related x-value dataset as its parameter besides padding setted to 'post' and maxlen setted to the length of the longest sentences (MAX\_LENGTH\_SENT).

## Word Embeddings

### *Script for Getting the F1 Score of the Models*

Since, there is no default method written for obtaining the F1 Score for CNN Models. A script obtained from Internet is used in this project.

**Link of the resource (Also added to the References section):**

[https://aakashgoel12.medium.com/how-to-add-user-defined-function-get-f1-score-in-keras-metrics-3013f979ce0d#:~:text=You%20will%20get%20training%20and%20validation%20F1%20score%20after%20each%20epoch](https://aakashgoel12.medium.com/how-to-add-user-defined-function-get-f1-score-in-keras-metrics-3013f979ce0d#:~:text=You%20will%20get%20training%20and%20validation%20F1%20score%20after%20each%20epoch,)

### *Creating Embedding Layers*

Two types of layers (Binary and Multiclass) created for each word embedding style.

- Embedding Layer For Randomly Initialized Word Embeddings

"input\_dim" setted as the word count+1 which is obtained from the word\_index() function of the keras tokenizer and adding 1 to it. "output\_dim" is setted as 100 and "input\_length" is setted as the length of the longest sentences (MAX\_LENGTH\_SENT).

- Word2Vec Embedding Layer For Word Embeddings From Scratch

Word2Vec is used for embedding the "From Scratch" CNN Word Embedding Model. "sentences" is obtained by applying split() method for each entry in the x\_train datasets. "min\_count" is equal to 1.

For the embedding layer, "input\_dim" is obtained by calling Word2Vec's "vocab" method on the embedding and taking the length of it. "output\_dim" is setted as 100 and "input\_length" is setted as the length of the longest sentences (MAX\_LENGTH\_SENT). Lastly, the weights obtained by applying Word2Vec's "vectors" method on the embeddings.

- Word2Vec Embedding Layer For Pretrained Word Embeddings

Word2Vec is used for embedding the "Pretrained" CNN Word Embedding Model. "sentences" is obtained by utilizing the api.load function to download the pretrained model 'text8'. "min\_count" is equal to 1.

The code for this part is written by using the resorce (Also added to the Reference section):

[https://radimrehurek.com/gensim/auto\\_examples/howtos/run\\_downloader\\_api.html](https://radimrehurek.com/gensim/auto_examples/howtos/run_downloader_api.html)

Lastly, the weights obtained by applying Word2Vec's "vectors" method on the embeddings.

### *Parameter Vectors for Running the Model Functions*

To find out the model which provides the best results, different parameters are tested on the models. These parameters are Filter Size, Kernel Size, and Hidden Layer Size.

parameters are Filter Size, Kernel Size, and Hidden Layer Size.

- **General Parameters:**  
filter\_size\_options = [16, 32]  
kernel\_size\_options = [4, 8]  
hidden\_layer\_options = [10, 20]

Additional, it is preferred to give "Operation Name" as a parameter to the functions due to indicating the functionality of the models.

- **For Binary Model:**  
operation\_names\_binary = ['Random Binary Model', 'From Scratch Binary Model', 'Pretrained Binary Model']  
embedding\_layer\_binary = [random\_emb\_layer\_binary, my\_binary\_emb\_layer, pretrained\_binary\_emb\_layer]
- **For Multiclass Model:**  
operation\_names\_multi = ['Random Multiclass Model', 'From Scratch Multiclass Model', 'Pretrained Multiclass Model']  
embedding\_layer\_multi = [random\_emb\_layer\_multi, my\_multi\_emb\_layer, pretrained\_multi\_emb\_layer]

### *Scores Obtained From the Binary Model Function*

### *Scores Obtained From the Multiclass Model Function*

## **Conclusion**

For Binary Classification, results obtained from each model were very good, usually both F1 and Accuracy Scores are between 90 - 85% .

- Naive Bayes: 86% F1 Score | 86% Accuracy Score
- Logistic Regression: 90% F1 Score | 90% Accuracy Score
- CNN:
  - Randomly Initialized Word Embedding Model:  
90 - 87% F1-Accuracy Score (Best Result: Filter Size: 16, Kernel Size: 4, Hidden Layer Size: 10)
  - "From Scratch" Word Embedding Model:  
89 - 86% F1-Accuracy Score (Best Result: Filter Size: 16, Kernel Size: 4, Hidden Layer Size: 20)
  - Pretrained Word Embedding Model:  
90-85% F1-Accuracy Score (Best Result: Filter Size: 32, Kernel Size: 4, Hidden Layer Size: 10)

For Multiclass Classification, results obtained from each model are not good, usually both F1 and Accuracy Scores are between 50 - 45% .

- Naive Bayes: 47% F1 Score | 49% Accuracy Score
- Logistic Regression: 55% F1 Score | 55% Accuracy Score
- CNN:
  - Randomly Initialized Word Embedding Model:  
49 - 47% F1-Accuracy Score (Best Result: Filter Size: 16, Kernel Size: 8, Hidden Layer Size: 10)
  - "From Scratch" Word Embedding Model:  
49 - 45% F1-Accuracy Score (Best Result: Filter Size: 16, Kernel Size: 4, Hidden Layer Size: 20)
  - Pretrained Word Embedding Model:  
47 - 41% F1 Score | 47 - 43 Accuracy Score (Best Result: Filter Size: 32, Kernel Size: 4, Hidden Layer Size: 20)

Overall, it can be concluded that:

- Binary Classification provides better accuracy and F1 Scores compared to Multiclass Classification. (
- For both Binary and Multiclass Classification, Logistic Regression do a very good job.
- In case o CNN, it also provide accurate results as good as Logistic Regression. When the complexity and

compilation (execution) time of the CNN considered, Logistic Regression Model is a better option for this project.

- Among the CNN Word Embedding Models, Randomly Initialized Word Embedding Models provide the best results.

## References

- Script for Getting F1 Scores of the Models:  
<https://aakashgoel12.medium.com/how-to-add-user-defined-function-get-f1-score-in-keras-metrics-3013f979ce0d#:~:text=You%20will%20get%20training%20and%20validation%20F1%20score%20after%20ea>
- Pretrained CNN Word Embedding Model Reference:  
[https://radimrehurek.com/gensim/auto\\_examples/howtos/run\\_downloader\\_api.html](https://radimrehurek.com/gensim/auto_examples/howtos/run_downloader_api.html)

