```
In [8]:
```

```
import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
import pickle
from preprocessor class import Preprocessor
from functions import f1_metric
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import OneHotEncoder
from sklearn.model selection import train test split
from sklearn.feature extraction.text import TfidfVectorizer
import tensorflow as tf
from tensorflow import keras
from keras.models import Sequential
from keras.layers import Dense, Input, Dropout, LeakyReLU
from keras.callbacks import EarlyStopping
from keras.utils import to categorical
from keras.regularizers import 12
from keras.optimizers import Adam
```

Neural Network

This part of the program is largely experimental. I wanted to test the use of a neural network to see if a simple network would perform well on the data. Ultimately, a sequential Dense network didn't do better than my other models and adding additional hidden layers, more epochs, and different transformations didn't help much.

In the interest of time, I decided it was best to focus on the working models instead of looking to improve the neural network, but I am leaving the notebook for the network here so when I have more time, I can experiment further and see if I can get the network to match or exceed the scores of the boosting, bagging, and ensemble models in my final_models.ipynb notebook.

```
In [4]:
```

```
data = pd.read_csv('Data/prepared_text_data_sugar.csv')
data
```

Out[4]:

	sugar_class	text
0	5	Cookie Dough Blizzard Cake, 10 in Cookie Dou
1	5	Reeses Peanut Butter Cups Blizzard Cake, 10
2	5	Chocolate Xtreme Blizzard Cake, 10 in Chocol
3	5	Oreo Blizzard Cake, 10 in Oreo Blizzard Cake
4	5	DQ Round Cake, 10 in DQ Round Cake, 10 in DQ
•••		
52926	1	6 Nuggets 6 Nuggets 6 Nuggets, Tenders Entrees
52927	1	Breast, Bonafide Spicy Chicken Breast, Bonafi
52928	1	Thigh, Bonafide Spicy Chicken Thigh, Bonafide
52929	1	Leg, Bonafide Spicy Chicken Leg, Bonafide Spi
52930	1	Black Pepper, for MTO Shnack Wrapz Black Pepp

52931 rows × 2 columns

```
X = data['text']
y = data['sugar class']
#One-Hot Encoding target variable, train/test split
ohe = OneHotEncoder(drop='first', sparse=False)
X_train_raw, X_test_raw, y_train_raw, y_test_raw = train_test_split(X, y, test_size = 0.
2, random state = 200)
y_train_raw = y_train_raw.values.reshape(-1, 1)
y_test_raw = y_test_raw.values.reshape(-1, 1)
y_train = ohe.fit_transform(y_train_raw - 1)
y test = ohe.transform(y test raw - 1)
#Pre-processing and vectorizing text
processor = Preprocessor()
X train transformed = processor.fit transform(X train raw)
X test transformed = processor.transform(X test raw)
vector pipe = Pipeline([('tfidf', TfidfVectorizer())])
X train vector = vector pipe.fit transform(X train transformed)
X test vector = vector pipe.transform(X test transformed)
#Returning independent variables to pd.Dataframe
X train = pd.DataFrame(X train vector.toarray(), columns = vector pipe['tfidf'].get feat
X test = pd.DataFrame(X test vector.toarray(), columns = vector pipe['tfidf'].get featur
e names())
```

In [6]:

```
#Ensuring X_train and X_test are proper data types
X_train_array = X_train.values if isinstance(X_train, pd.DataFrame) else X_train
X_test_array = X_test.values if isinstance(X_test, pd.DataFrame) else X_test
y_train_reshaped = np.argmax(y_train, axis = 1)
trainCallback = EarlyStopping(monitor='loss', min_delta = 1e-6, patience = 5)
reg = 12(0.0001)
opt = Adam(learning_rate=0.001, beta_1=0.9, beta_2=0.999, epsilon=1e-07)
```

In [10]:

```
model = Sequential()
model.add(Dense(1800, activation = LeakyReLU(), input_shape = (X_train_array.shape[1],),
kernel_regularizer = reg))
model.add(Dense(4, activation = 'softmax'))
model.compile(loss = 'categorical_crossentropy', optimizer = opt, metrics = f1_metric)
model.fit(X_train_array, y_train, epochs = 15, callbacks=[trainCallback], batch_size= 12
8, validation_split = 0.2)
```

```
Epoch 1/15
627 - val loss: 13.9435 - val f1 metric: 0.5062
Epoch 2/15
4824 - val loss: 34.9437 - val f1 metric: 0.4615
Epoch 3/15
836 - val loss: 59.1820 - val f1 metric: 0.3115
Epoch 4/15
997 - val loss: 85.0725 - val_f1_metric: 0.4864
Epoch 5/15
704 - val loss: 113.6235 - val_f1_metric: 0.5096
Epoch 6/15
4634 - val loss: 144.1213 - val f1 metric: 0.4857
```

Out[10]: