

# Hyperspectral Imaging Data

## Absract

Remote sensing involves understanding activities on surface of the earth based on the reflected light from the surface. Devices which help us get this information are the cameras on board satellites, Surveillance planes/ UAVs/Drones. There are multiple methods of collecting information of activities of surface of the earth as mentioned in [https://en.wikipedia.org/wiki/Remote\\_sensing](https://en.wikipedia.org/wiki/Remote_sensing). Hyper spectral imaging is one of the method which help us peek into the unseen light by collecting signals of multiple narrow bands of wavelength beyond visible light spectrum.

In remote sensing, hyperspectral images help study surface of the earth. It captures unique information associated with specific activity or structures (natural or human) on the surface. This can help Agriculture, Minerals Exploration, Defense, Surveillance etc. These images can help monitor Earths surface activity.

Based on the image data we can segment the areas of our interest. With image and segmented data we can train a Deep Learning algorithm which can in future generate segmented image from a raw image.

## Problem

At the present we have limited data available to train a DL model to make segmentation of any given patch of earth's surface based on the hyperspectral image. For study purpose there are images collected through airbornes or satellites. The data set is available in the link - [http://www.ehu.eus/ccwintco/index.php/Hyperspectral\\_Remote\\_Sensing\\_Scenes#Indian\\_Pines](http://www.ehu.eus/ccwintco/index.php/Hyperspectral_Remote_Sensing_Scenes#Indian_Pines)

The scene was gathered by AVIRIS sensor over the Indian Pines test site in North-western Indiana and consists of 145x145 pixels and 224 spectral reflectance bands in the wavelength range 0.4–2.5  $10^{(-6)}$  meters. This scene is a subset of a larger one.

The ground truth available is designated into 16 classes and is not all mutually exclusive. The number of bands are reduced to 200 by removing bands covering the region of water absorption. (In the same link we can observe other patches of HS images, which can be explored for study)

Now we have to come up with system that should be able to segment the regions based on the available Hyper Spectral(HS) image data.

Deep Learning is one of the techniques which can predict classification of specific data based on the information associated with data. Deeplearning invovles different kinds of Neural Network architecture to be built and trained on the data.

Refer the blog in the link mentioned, which captures overview of simple to complex neural network architectures-<https://www.sabrepc.com/blog/Deep-Learning-and-AI/6-types-of-neural-networks-to-know-about>

## Module imports

In [1]:

```
import scipy.io
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sb
from sklearn.model_selection import train_test_split
import tensorflow as tf
from tensorflow.keras.layers import Input
from tensorflow.keras.layers import Dense, Dropout
from tensorflow.keras.layers import Flatten
from tensorflow.keras.losses import categorical_crossentropy
from tensorflow.keras import Model, initializers, regularizers
from tensorflow.keras.utils import plot_model, to_categorical
from sklearn.metrics import f1_score, roc_auc_score
from tensorflow.keras.callbacks import ModelCheckpoint
from tensorflow.keras.callbacks import EarlyStopping, ReduceLROnPlateau
from tensorflow.keras.callbacks import LearningRateScheduler
import time, os
from tensorflow.python.keras.callbacks import TensorBoard
from datetime import datetime
```

```

from datetime import datetime
from sklearn.preprocessing import MinMaxScaler, StandardScaler
from sklearn.utils import class_weight
from sklearn.metrics import confusion_matrix, cohen_kappa_score
from tensorflow.keras import optimizers

```

In [2]:

```

# Data Source : http://www.ehu.es/ccwintco/index.php/Hyperspectral_Remote_Sensing_Scenes#Indian_Pines
!wget wget --header="Host: www.ehu.es" --header="User-Agent: Mozilla/5.0 (Windows NT 10.0; Win64; x64)
AppleWebKit/537.36 (KHTML, like Gecko) Chrome/97.0.4692.71 Safari/537.36" --header="Accept: text/html,a
pplication/xhtml+xml,application/xml;q=0.9,image/avif,image/webp,image/apng,*/*;q=0.8,application/sign
e-d-exchange;v=b3;q=0.9" --header="Accept-Language: en-US,en;q=0.9" --header="Referer: http://www.ehu.es
/ccwintco/index.php/Hyperspectral_Remote_Sensing_Scenes" "http://www.ehu.es/ccwintco/uploads/6/67/Indi
an_pines_corrected.mat" -c -O 'Indian_pines_corrected.mat'
!unzip Indian_pines_corrected.mat

```

```

--2022-02-26 02:09:01-- http://wget/
Resolving wget (wget)... failed: Name or service not known.
wget: unable to resolve host address 'wget'
--2022-02-26 02:09:01-- http://www.ehu.es/ccwintco/uploads/6/67/Indian_pines_corrected.mat
Resolving www.ehu.es (www.ehu.es)... 158.227.0.65, 2001:720:1410::65
Connecting to www.ehu.es (www.ehu.es)|158.227.0.65|:80... connected.
HTTP request sent, awaiting response... 200 OK
Length: 5953527 (5.7M)
Saving to: 'Indian_pines_corrected.mat'

Indian_pines_correc 100%[=====>] 5.68M 742KB/s in 8.6s

2022-02-26 02:09:10 (677 KB/s) - 'Indian_pines_corrected.mat' saved [5953527/5953527]

FINISHED --2022-02-26 02:09:10--
Total wall clock time: 9.0s
Downloaded: 1 files, 5.7M in 8.6s (677 KB/s)
Archive: Indian_pines_corrected.mat
  End-of-central-directory signature not found. Either this file is not
  a zipfile, or it constitutes one disk of a multi-part archive. In the
  latter case the central directory and zipfile comment will be found on
  the last disk(s) of this archive.
unzip: cannot find zipfile directory in one of Indian_pines_corrected.mat or
      Indian_pines_corrected.mat.zip, and cannot find Indian_pines_corrected.mat.ZIP, period.

```

In [3]:

```

# mat = scipy.io.loadmat('HSI_Data\Indian_pines_corrected.mat')
mat = scipy.io.loadmat('Indian_pines_corrected.mat')

```

## Indian pines Image data

### Shape

In [4]:

```
mat['indian_pines_corrected'].shape
```

Out[4]:

```
(145, 145, 200)
```

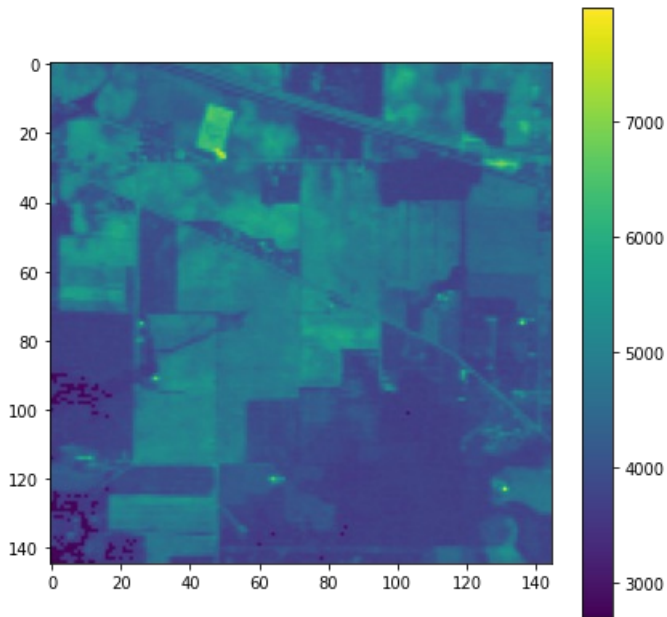
### Image

In [5]:

```

plt.figure(figsize=(7,7))
plt.imshow(mat['indian_pines_corrected'][:, :, 10])
plt.colorbar()
plt.show()

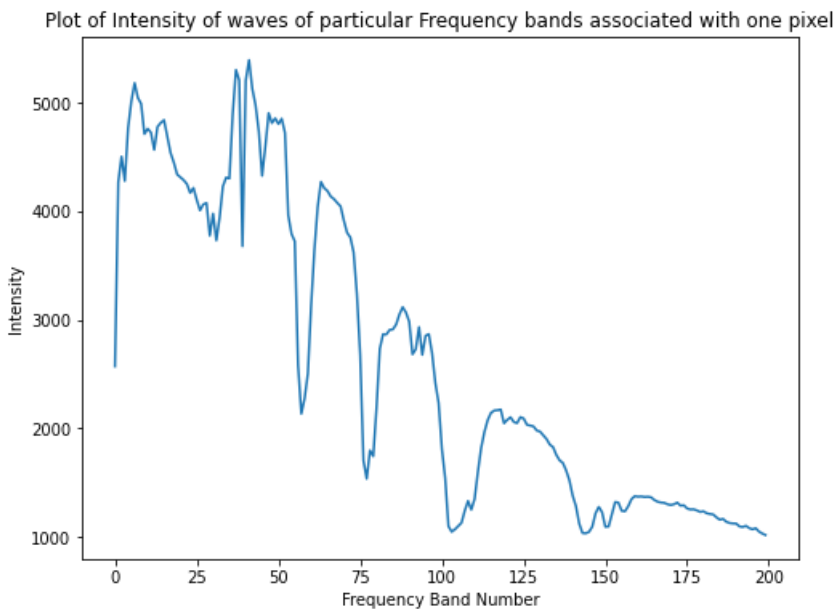
```



### Information in one Pixel

In [35]:

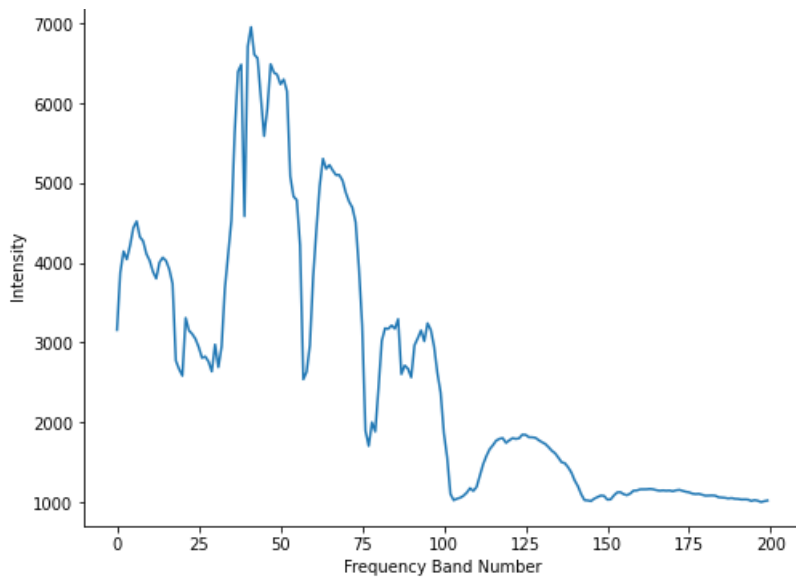
```
plt.figure(figsize=(8,6))
plt.plot(mat['indian_pines_corrected'][1,20,:])
plt.title('Plot of Intensity of waves of particular Frequency bands associated with one pixel')
plt.xlabel('Frequency Band Number')
plt.ylabel('Intensity')
plt.show()
```



In [36]:

```
plt.figure(figsize=(8,6))
plt.plot(mat['indian_pines_corrected'][20,30,:])
plt.title('Plot of Intensity of waves of particular Frequency bands associated with one pixel')
plt.xlabel('Frequency Band Number')
plt.ylabel('Intensity')
plt.show()
```

Plot of Intensity of waves of particular Frequency bands associated with one pixel



In [8]:

```
np.max(mat['indian_pines_corrected']), np.min(mat['indian_pines_corrected'])
```

Out[8]:

```
(9604, 955)
```

### Ground Truth image

In [9]:

```
# mat_gt = scipy.io.loadmat('HSI_Data\Indian_pines_gt.mat')
mat_gt = scipy.io.loadmat('Indian_pines_gt.mat')
```

In [10]:

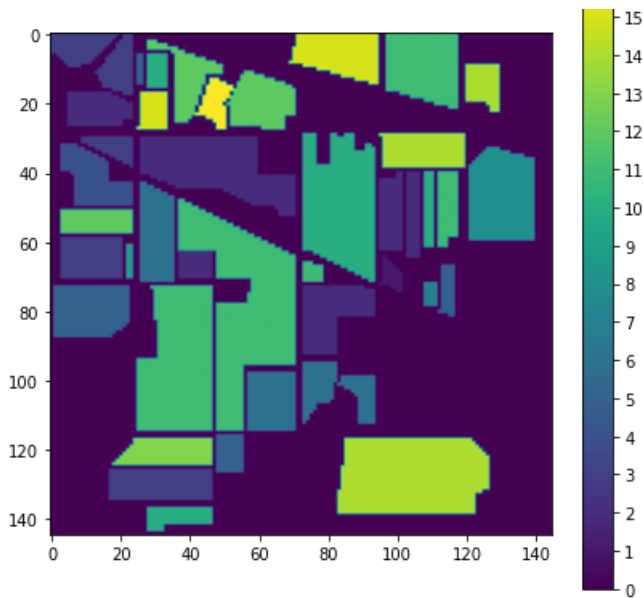
```
mat_gt['indian_pines_gt'][25:40,15:30]
```

Out[10]:

```
array([[ 2,  2,  2,  2,  2,  2,  0,  0,  0,  0,  0, 15, 15, 15, 15],
       [ 2,  2,  2,  2,  2,  2,  0,  0,  0,  0,  0, 15, 15, 15, 15],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 15, 15, 15, 15],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [ 3,  3,  3,  3,  3,  3,  3,  3,  3,  3,  0,  2,  2,  2,  2],
       [ 3,  3,  3,  3,  3,  3,  3,  3,  3,  3,  0,  2,  2,  2,  2],
       [ 3,  3,  3,  3,  3,  3,  3,  3,  3,  3,  0,  2,  2,  2,  2],
       [ 3,  3,  3,  3,  3,  3,  3,  3,  3,  3,  0,  2,  2,  2,  2],
       [ 3,  3,  3,  3,  3,  3,  3,  3,  3,  3,  0,  2,  2,  2,  2],
       [ 0,  0,  3,  3,  3,  3,  3,  3,  3,  3,  0,  2,  2,  2,  2],
       [ 4,  0,  0,  0,  3,  3,  3,  3,  3,  3,  0,  2,  2,  2,  2],
       [ 4,  4,  0,  0,  0,  0,  3,  3,  3,  0,  0,  2,  2,  2,  2],
       [ 4,  4,  0,  0,  0,  0,  0,  0,  3,  0,  0,  2,  2,  2,  2],
       [ 4,  4,  0,  0,  0,  0,  0,  0,  0,  0,  0,  2,  2,  2,  2]],
      dtype=uint8)
```

In [33]:

```
plt.figure(figsize=(7,7))
plt.imshow(mat_gt['indian_pines_gt'])
plt.colorbar(ticks=range(0,17))
plt.show()
```



In [12]:

```
mat_gt['indian_pines_gt'].shape
```

Out[12]:

```
(145, 145)
```

## Pixelwise classification using Neural Network

Reshaping of the image (3 dimension - x,y,l) into rows and columns. Each row represents the each pixel at location(x,y). Each column represents the third dimension(l) which captures the intensity of different frequency bands.

In [13]:

```
# https://panjeh.medium.com/convert-numpy-3d-array-to-2d-array-in-python-931a4cdf8b12
```

In [14]:

```
mat_rshp = mat['indian_pines_corrected'].reshape(-1,200)
```

In [15]:

```
mat_gt_rshp = mat_gt['indian_pines_gt'].reshape(-1,1)
```

In [16]:

```
px_data_full = pd.DataFrame(mat_rshp)
```

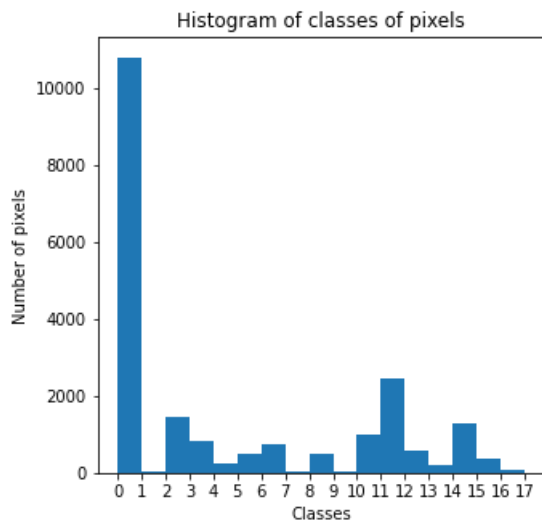
In [17]:

```
# px_class_data = pd.DataFrame(mat_gt_rshp)
```

In [18]:

```
# px_class_data.value_counts()
plt.figure(figsize=(5,5))
plt.hist(mat_gt_rshp,bins=range(0,18,1))
plt.xticks(range(0,18,1))
plt.title('Histogram of classes of pixels')
plt.xlabel('Classes')
plt.ylabel('Number of pixels')
```

```
plt.show()
```



## Reshape verification

In [19]:

```
# mat_rshp_chk = mat_rshp.reshape(145,145,200)
```

In [20]:

```
# plt.imshow(mat_rshp_chk[:, :, 20])  
# plt.show()
```

In [21]:

```
# mat_gt_rshp_chk = mat_gt_rshp.reshape(145,145)
```

In [22]:

```
# plt.imshow(mat_gt_rshp_chk)  
# plt.show()
```

## Pixelwise data

In [23]:

```
px_data = pd.DataFrame(mat_rshp)  
px_data
```

Out[23]:

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18
0	3172	4142	4506	4279	4782	5048	5213	5106	5053	4750	4816	4769	4610	4805	4828	4861	4767	4624	4549
1	2580	4266	4502	4426	4853	5249	5352	5353	5347	5065	5141	5100	4994	5172	5290	5289	5217	5053	5033
2	3687	4266	4421	4498	5019	5293	5438	5427	5383	5132	5227	5172	5097	5313	5411	5412	5341	5191	5140
3	2749	4258	4603	4493	4958	5234	5417	5355	5349	5096	5147	5078	5040	5237	5321	5344	5255	5121	5035
4	2746	4018	4675	4417	4886	5117	5215	5096	5098	4834	4853	4857	4734	4879	4976	4958	4885	4754	4647
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
21020	2561	3987	4011	4023	4201	4377	4418	4248	4180	3838	3762	3696	3594	3750	3806	3764	3655	2611	3199
21021	2726	4104	4024	3880	4210	4377	4413	4174	4229	3900	3786	3696	3623	3726	3758	3725	3614	2584	3174

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18
21022	3153	3864	4282	3889	4310	4372	4375	4208	4096	3878	3890	3683	3594	3714	3724	3700	2809	2564	3177
21023	3155	4104	4106	4027	4139	4318	4413	4174	4140	3933	3799	3627	2752	3667	3688	3628	2750	3245	3153
21024	3323	3860	4197	3952	4148	4279	4375	4225	3988	3866	3811	3639	3623	3686	3722	3673	2775	3315	3180

21025 rows × 200 columns



In [24]:

```
px_class = pd.DataFrame(mat_gt_rshp)
px_class
```

Out[24]:

	0
0	3
1	3
2	3
3	3
4	3
...	...
21020	0
21021	0
21022	0
21023	0
21024	0

21025 rows × 1 columns

In [25]:

```
# px_class[px_class[0]==0].index
```

In [26]:

```
# px_data = px_data.drop(px_class[px_class[0]==0].index,axis=0)
```

In [27]:

```
# px_class = px_class.drop(px_class[px_class[0]==0].index,axis=0)
```

In [28]:

```
px_class[0].describe()
```

Out[28]:

```
count    21025.000000
mean         4.224923
std         5.281972
min          0.000000
25%          0.000000
50%          0.000000
75%         10.000000
max         16.000000
Name: 0, dtype: float64
```

## Train and Test Split

In [29]:

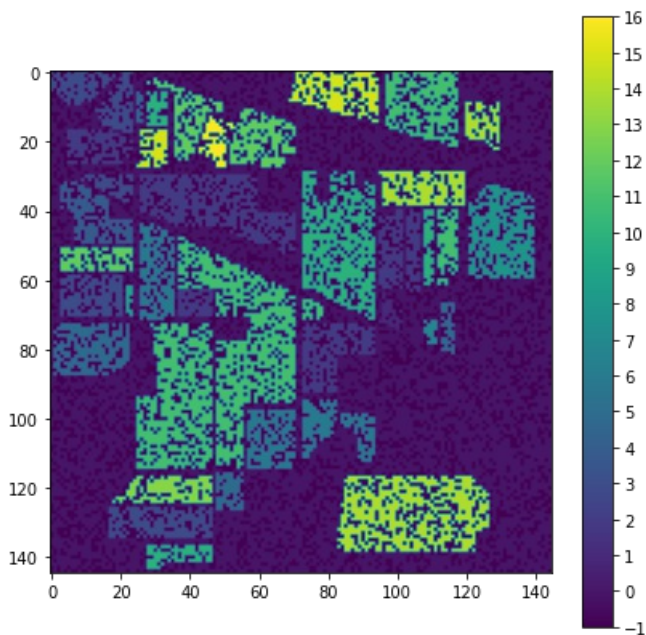
```
X_train, X_test, y_train, y_test = train_test_split(px_data, px_class, test_size=0.3, random_state=20, stratify=px_class)
```

In [30]:

```
PX_class_train = y_train.reindex(px_data_full.index, fill_value=-1)
```

In [34]:

```
plt.figure(figsize=(7,7))
plt.imshow(PX_class_train.to_numpy().reshape((145,145)))
plt.colorbar(ticks=range(-1,17,1))
plt.show()
```



In [ ]:

```
y_ctg_train = to_categorical(y_train)
y_ctg_test = to_categorical(y_test)
```

In [ ]:

```
# y_ctg_train = y_ctg_train[:,1:]
# y_ctg_test = y_ctg_test[:,1:]
```

## Feature Engineering

### Standardization

In [ ]:

```
# Scaler = MinMaxScaler()
Scaler = StandardScaler()
X_train_std = pd.DataFrame(Scaler.fit_transform(X_train), index = X_train.index)
X_test_std = pd.DataFrame(Scaler.transform(X_test), index = X_test.index)
```

In [ ]:



```
In [ ]:
```

```
y_train
```

```
Out[ ]:
```

	0
7674	8
14004	0
17623	0
12672	11
15734	6
...	...
3002	0
14264	11
2104	0
7618	10
15062	0

14717 rows × 1 columns

## Variance

```
In [ ]:
```

```
variance_data = X_train.var()
```

```
In [ ]:
```

```
variance_data
```

```
Out[ ]:
```

```
0      127297.253743
1       52361.818113
2       66493.685912
3       78627.459402
4      120039.036524
...
195      698.846583
196      885.827231
197      433.704496
198      129.637613
199       49.795652
Length: 200, dtype: float64
```

```
In [ ]:
```

```
min(variance_data)
```

```
Out[ ]:
```

```
49.79565233299031
```

```
In [ ]:
```

```
Feat_var_reject = variance_data[variance_data<0.01*variance_data.describe()['max']].index
Feat_var_reject
```

```
Out[ ]:
```

```
Int64Index([ 76,  77,  78,  79, 101, 102, 103, 104, 105, 106, 107, 109, 141,
            142, 143, 144, 145, 146, 147, 149, 150, 151, 152, 155, 156, 175,
            176, 177, 178, 179, 180, 181, 182, 183, 184, 185, 186, 187, 188,
            189, 190, 191, 192, 193, 194, 195, 196, 197, 198, 199],
           dtype='int64')
```

In [ ]:

```
X_train[Feat_var_reject]
```

Out[ ]:

	76	77	78	79	101	102	103	104	105	106	107	109	141	142	143	144	145	146	147
7674	1842	1676	1997	1884	1612	1095	1041	1060	1097	1112	1205	1233	1263	1126	1047	1025	1028	1059	1148
14004	1900	1686	1965	1874	1509	1090	1044	1052	1069	1065	1111	1137	1184	1080	1040	1015	1015	1045	1068
17623	1860	1678	1926	1821	1498	1059	1026	1046	1038	1050	1092	1104	1150	1063	1033	1010	1010	1018	1040
12672	1726	1552	1856	1762	1587	1120	1034	1070	1112	1133	1251	1278	1316	1154	1053	1040	1035	1116	1227
15734	1748	1579	1869	1753	1460	1085	1031	1038	1063	1060	1130	1144	1189	1088	1027	1011	1020	1051	1085
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
3002	1587	1437	1689	1596	1416	1066	1020	1040	1073	1079	1163	1177	1184	1090	1038	1030	1013	1059	1131
14264	1759	1585	1827	1778	1598	1087	1057	1084	1102	1128	1256	1260	1311	1138	1048	1035	1040	1101	1208
2104	1804	1613	1884	1755	1484	1084	1028	1048	1062	1079	1134	1162	1186	1093	1043	1010	1035	1048	1102
7618	1755	1619	1901	1839	1643	1102	1034	1086	1116	1147	1298	1306	1329	1163	1051	1049	1058	1131	1268
15062	1806	1604	1918	1802	1532	1094	1034	1058	1071	1080	1144	1152	1198	1085	1033	1006	1024	1034	1084

14717 rows × 50 columns



Note: Varince is not low (not close to zero). each feature might have distribution of their own. So we will not consider dropping features based on variance.

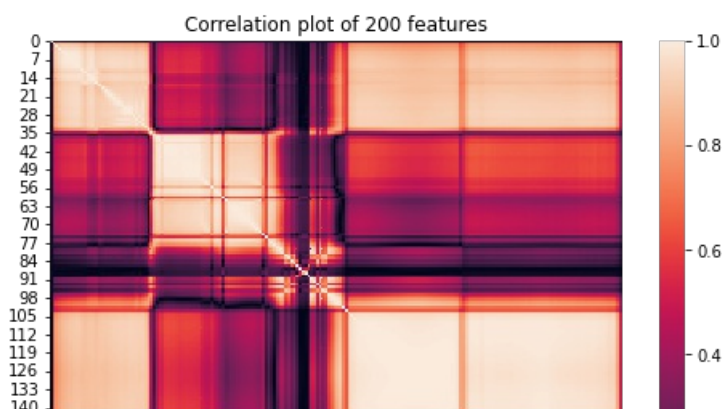
## Correlation

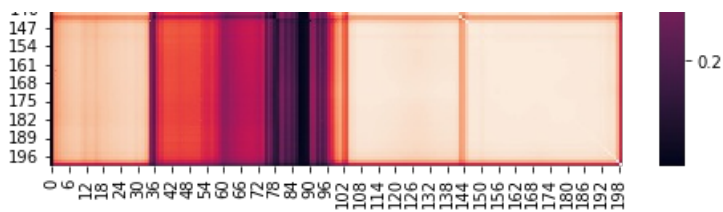
In [ ]:

```
# correlation evaluation of filtered dataset
Correlation_mat = X_train_std.corr('pearson')
```

In [ ]:

```
# Correlation heatmap on subset of features. Example plot as the feature size is large.
plt.figure(figsize=(8,6))
sb.heatmap(abs(Correlation_mat))#.iloc[40:60,40:60]),annot=True)
plt.title('Correlation plot of 200 features')
plt.show()
```





If there were no correlation among features, there would be high correlation (white colored) only on the diagonal of the above figure.

In [ ]:

```
def get_corr_features(corr_mat):
    '''This function returns feature list which have correlation above 0.99 for a given correlation matrix'''
    # list to store correlated features
    feature_correlated = []
    for i in range(len(corr_mat)):
        for j in range(i): #This loop repeats only till one step before j=i, Thus captures information from
            #lower triangle matrix.
            if((corr_mat.iloc[i,j]>=0.99) and (corr_mat.columns[i] not in feature_correlated)):
                # features with correlation above 0.99 added to list.
                feature_correlated.append(corr_mat.columns[i])
    return feature_correlated
```

In [ ]:

```
# Correlation matrix of filtered train data passed to get_corr_features function to get list of feature with high correlation.
feature_correlated = get_corr_features(Correlation_mat)
```

In [ ]:

```
len(feature_correlated)
```

Out[ ]:

105

In [ ]:

```
feature_correlated
```

Out[ ]:

```
[7,
 8,
 9,
15,
24,
27,
28,
38,
40,
41,
42,
43,
44,
45,
46,
47,
48,
49,
50,
51,
52,
58,
```

64,  
65,  
66,  
67,  
68,  
69,  
108,  
109,  
110,  
111,  
112,  
113,  
114,  
115,  
116,  
117,  
118,  
119,  
120,  
121,  
122,  
123,  
124,  
125,  
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127,  
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134,  
135,  
136,  
137,  
138,  
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141,  
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183,  
184,

```
185,  
186,  
187,  
188,  
189,  
190]
```

### Dropping features with high correlation

```
In [ ]:
```

```
X_train_std = X_train_std.drop(feature_correlated, axis=1)  
X_test_std = X_test_std.drop(feature_correlated, axis=1)
```

```
In [ ]:
```

```
X_train_std.shape, X_test_std.shape
```

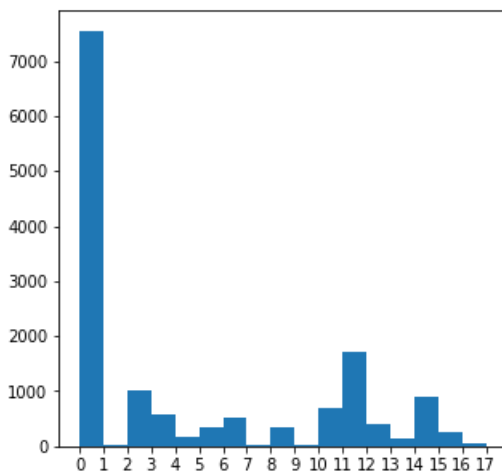
```
Out[ ]:
```

```
((14717, 95), (6308, 95))
```

### Class Weights Evaluation

```
In [ ]:
```

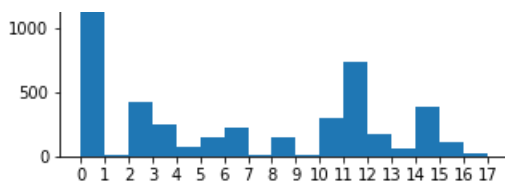
```
plt.figure(figsize=(5,5))  
plt.xticks(range(0,18,1))  
plt.hist(y_train.to_numpy(),bins=range(0,18,1))  
plt.show()
```



```
In [ ]:
```

```
plt.figure(figsize=(5,5))  
plt.xticks(range(0,18,1))  
plt.hist(y_test.to_numpy(),bins=range(0,18,1))  
plt.show()
```





In [ ]:

```
y_train.shape
```

Out[ ]:

```
(14717, 1)
```

In [ ]:

```
np.squeeze(y_train.to_numpy()).shape
```

Out[ ]:

```
(14717,)
```

In [ ]:

```
# https://datascience.stackexchange.com/questions/13490/how-to-set-class-weights-for-imbalanced-classes-in-keras
# wts = n_samples / (n_classes * np.bincount(y))
# class_weights_1 = class_weight.compute_class_weight(class_weight='balanced', classes=np.unique(y_train), y=np.squeeze(y_train.to_numpy()))
```

In [ ]:

```
# class_wts_1 = dict(enumerate(class_weights_1/np.sum(class_weights_1)))
# class_wts_1
```

In [ ]:

```
## Get the class weights for loss evaluation as the data has class imbalance
# https://medium.com/gumgum-tech/handling-class-imbalance-by-introducing-sample-weighting-in-the-loss-function-3bdebd8203b4
# https://neptune.ai/blog/keras-loss-functions
# https://keras.io/api/models/model_training_apis
class_count = y_train.value_counts().sort_index()
class_wts = 1/class_count
class_wts = class_wts/np.sum(class_wts)
class_wts = dict(enumerate(class_wts.to_list()))
class_wts
```

Out[ ]:

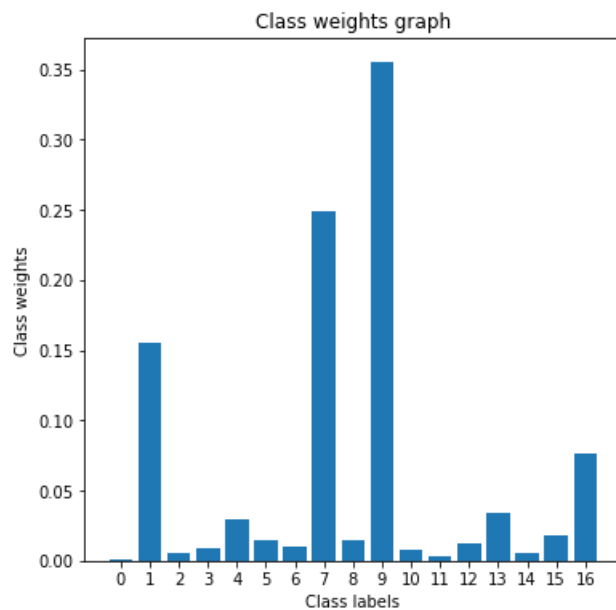
```
{0: 0.0006592993609901092,
 1: 0.15540922124838732,
 2: 0.0049730950799483946,
 3: 0.008559544027449904,
 4: 0.029958404096074667,
 5: 0.014713299053101758,
 6: 0.009732084305182768,
 7: 0.24865475399741974,
 8: 0.014845059940144461,
 9: 0.355221077139171,
10: 0.0073133751175711685,
11: 0.0028947002793646066,
12: 0.011983361638429868,
13: 0.03453538249964162,
14: 0.00561931647451796,
15: 0.018418870666475535,
16: 0.07650915507612915}
```

```
In [ ]:
```

```
# class_wts.values()
```

```
In [ ]:
```

```
plt.figure(figsize=(6,6))
plt.bar(range(0,17,1),list(class_wts.values()))
plt.xlabel('Class labels')
plt.xticks(range(0,17,1))
plt.ylabel('Class weights')
plt.title('Class weights graph')
plt.show()
```



```
In [ ]:
```

```
list(class_wts.values())
```

```
Out[ ]:
```

```
[0.0006592993609901092,
 0.15540922124838732,
 0.0049730950799483946,
 0.008559544027449904,
 0.029958404096074667,
 0.014713299053101758,
 0.009732084305182768,
 0.24865475399741974,
 0.014845059940144461,
 0.355221077139171,
 0.0073133751175711685,
 0.0028947002793646066,
 0.011983361638429868,
 0.03453538249964162,
 0.00561931647451796,
 0.018418870666475535,
 0.07650915507612915]
```

```
In [ ]:
```

```
np.argmax(list(class_wts.values()))
```

```
Out[ ]:
```

In [ ]:

```
np.argmax(list(class_wts.values()))
```

Out[ ]:

0

In [ ]:

```
sum(list(class_wts.values()))
```

Out[ ]:

1.0

In [ ]:

```
X_train_std.shape, y_ctg_train.shape
```

Out[ ]:

((14717, 95), (14717, 17))

In [ ]:

```
X_test_std.shape, y_ctg_test.shape
```

Out[ ]:

((6308, 95), (6308, 17))

In [ ]:

```
# y_ctg_train
```

## Neural network model

### F1 score calculation callback function definition

In [ ]:

```
class metric_calc(tf.keras.callbacks.Callback):

    def on_train_begin(self, logs={}):
        ## on begin of training, we are creating a instance variable called metrics
        ## it is a dict with keys [loss, acc, val_loss, val_acc]
        self.metrics={'micro_F1_train': [],
                      'micro_F1_val': [],
                      'val_acc': []}

    def on_epoch_end(self, epoch, logs={}):
        ## on end of each epoch, we will get logs and update the self.metrics dict

        # Training score
        y_true = y_train
        y_pred = self.model.predict(X_train_std)

        yp = np.argmax(y_pred,axis=1)

        score = f1_score(y_true,yp,average='micro')

        self.metrics['micro_F1_train'].append(score)

        print('\n\nmicro F1 train: ',score)
```



```

# Validation score
y_true_val = y_test
y_pred_val = self.model.predict(X_test_std)

yp_val = np.argmax(y_pred_val,axis=1)

score_val = f1_score(y_true_val,yp_val,average='micro')

self.metrics['micro_F1_val'].append(score_val)

print('\nmicro_F1_val: ',score_val)

self.metrics['val_acc'].append(logs.get('val_acc'))

#any NaN values(either weigths or loss) while training will terminate training
loss = logs.get('loss')
if loss is not None:
    if np.isnan(loss) or np.isinf(loss):
        print("Invalid loss and terminated at epoch {}".format(epoch))
        self.model.stop_training = True

```

## Confusion matrix

In [ ]:

```

# code reference: appliedaicourse.com case studies
def plot_confusion_matrix(test_y, predict_y):
    C = confusion_matrix(test_y, predict_y)
    print("Number of misclassified points ", (len(test_y)-np.trace(C))/len(test_y)*100)
    # C = 17x17 matrix, each cell (i,j) represents number of points of class i are predicted class j

    #Precision Matrix
    A =(C/C.sum(axis=0))
    #divid each element of the confusion matrix with the sum of elements in that column
    # C = [[1, 2],
    #      [3, 4]]
    # C.sum(axis = 0)  axis=0 corresonds to columns and axis=1 corresponds to rows in two dimensional
array
    # C.sum(axix =0) = [[4, 6]]
    # (C/C.sum(axis=0)) = [[1/4, 2/6],
    #                      [3/4, 4/6]]

    #Recall Matrix
    B =(((C.T)/(C.sum(axis=1))).T)
    #divid each element of the confusion matrix with the sum of elements in that row

    # C = [[1, 2],
    #      [3, 4]]
    # C.T = [[1, 3],
    #        [2, 4]]
    # C.sum(axis = 1)  axis=0 corresonds to columns and axis=1 corresponds to rows in two dimensional
array
    # C.sum(axix =1) = [[3, 7]]
    # ((C.T)/(C.sum(axis=1))) = [[1/3, 3/7]
    #                            [2/3, 4/7]]

    # ((C.T)/(C.sum(axis=1))).T = [[1/3, 2/3]
    #                             [3/7, 4/7]]
    # sum of row elements = 1

    labels = list(range(0,17,1))
    cmap=sb.light_palette("green")
    # representing C in heatmap format
    print("-"*50, "Confusion matrix", "-"*50)
    plt.figure(figsize=(16,8))
    sb.heatmap(C, annot=True, cmap=cmap, fmt=".1f", xticklabels=labels, yticklabels=labels)
    plt.xlabel('Predicted Class')
    plt.ylabel('Original Class')
    plt.show()

    # representing A in heatmap format
    print("-"*50, "Precision matrix", "-"*50)
    plt.figure(figsize=(16,8))

```

```

sb.heatmap(A, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=labels)
plt.xlabel('Predicted Class')
plt.ylabel('Original Class')
plt.show()
print("Sum of columns in precision matrix",A.sum(axis=0))

# representing B in heatmap format
print("-"*50, "Recall matrix" , "-"*50)
plt.figure(figsize=(16,8))
sb.heatmap(B, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=labels)
plt.xlabel('Predicted Class')
plt.ylabel('Original Class')
plt.show()
print("Sum of rows in recall matrix",B.sum(axis=1))

```

## 1- Four layered network with relu activation and SGD optimizer

### Model 1 Definition

In [ ]:

```
# del NeuNet_1
```

In [ ]:

```

In = Input(shape=(95,))
L1 = Dense(128,activation='relu',
           kernel_initializer=initializers.he_uniform())(In) #https://keras.io/api/layers/initializers/

L1_Drp = Dropout(0.4)(L1)

L2 = Dense(64,activation='relu',
           kernel_initializer=initializers.he_uniform())(L1_Drp)

L2_Drp = Dropout(0.3)(L2)

L3 = Dense(32,activation='relu',
           kernel_initializer=initializers.he_uniform())(L2_Drp)

L4 = Dense(20,activation='relu',
           kernel_initializer=initializers.he_uniform())(L3)

Out = Dense(17,activation='softmax',
           kernel_initializer=initializers.GlorotUniform())(L4)

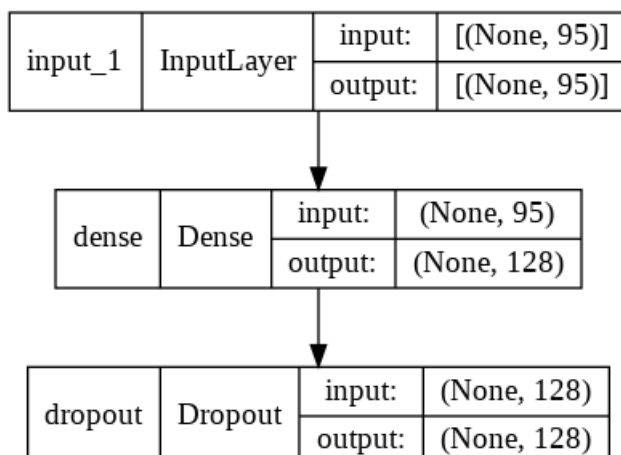
NeuNet_1 = Model(inputs=In, outputs = Out)

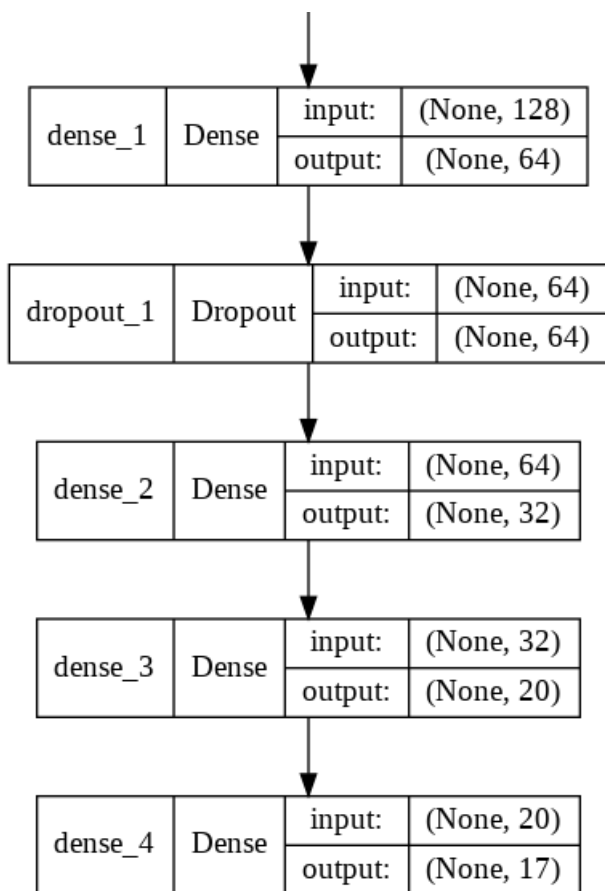
# cce = categorical_crossentropy()
NeuNet_1.compile(loss='categorical_crossentropy', optimizer='SGD', metrics=['accuracy'])

plot_model(NeuNet_1,show_layer_names=True, show_shapes=True)#,to_file='NeuNet_1.png')

```

Out[ ]:





In [ ]:

```
NeuNet_1.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[ (None, 95) ]	0
dense (Dense)	(None, 128)	12288
dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 64)	8256
dropout_1 (Dropout)	(None, 64)	0
dense_2 (Dense)	(None, 32)	2080
dense_3 (Dense)	(None, 20)	660
dense_4 (Dense)	(None, 17)	357

```

Total params: 23,641
Trainable params: 23,641
Non-trainable params: 0

```

### Model 1 Training

In [ ]:

```

logdir = os.path.join("logs", datetime.now().strftime("%Y%m%d-%H%M%S"))
print(datetime.now().strftime("%Y%m%d-%H%M%S"))
file_writer = tf.summary.create_file_writer(logdir + "/metrics")
# tensorboard = TensorBoard(log_dir=logdir)
tensorboard = TensorBoard(log_dir=logdir, histogram_freq=1, write_graph=True, write_grads=True)

```

```

tensorboard = tensorboard(log_dir=logdir, histogram_freq=1, write_graph=True, write_grads=True,

metric_calc_1 = metric_calc()

# Saving model at every epoch if validation accuracy is improved from previous epoch
filepath_m1="model1_save/weights-{epoch:02d}-{val_accuracy:.4f}.hdf5"
checkpoint_m1 = ModelCheckpoint(filepath=filepath_m1, monitor='val_accuracy', verbose=1,
                                save_best_only=True,
                                mode='auto')
earlystop_m1 = EarlyStopping(monitor='val_accuracy', min_delta=0.01, patience=20, verbose=1)
reduce_lr_m1 = ReduceLROnPlateau(monitor='val_loss', factor=0.9, patience=2, min_lr=0.001)

callback_list_1 = [metric_calc_1,
                   checkpoint_m1,
                   earlystop_m1,
                   reduce_lr_m1,
                   tensorboard]

# fit network
verbose_1, epochs_1, batch_size_1 = 1, 100, 10
start = time.time()
history_1 = NeuNet_1.fit(X_train_std, y_ctg_train,
                        class_weight = class_wts,
                        validation_data=(X_test_std,y_ctg_test),
                        epochs=epochs_1,
                        batch_size=batch_size_1,
                        verbose=verbose_1,
                        callbacks=callback_list_1)

# Passing the F1 score data of each epoch to tensor board
# https://stackoverflow.com/questions/58102016/how-visualize-in-tensorboard-a-metric-callback
for i in range(len(metric_calc_1.metrics['micro_F1_train'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_train', metric_calc_1.metrics['micro_F1_train'][i])
file_writer.flush()

for i in range(len(metric_calc_1.metrics['micro_F1_val'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_val', metric_calc_1.metrics['micro_F1_val'][i])
file_writer.flush()

stop = time.time()
print('Time Taken for training (sec): ',stop-start)

```

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WARNING:tensorflow:`write\_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.

Epoch 1/100

2/1472 [.....] - ETA: 5:55 - loss: 0.1137 - accuracy: 0.0500 WARNING:tensorflow:Callback method `on\_train\_batch\_begin` is slow compared to the batch time (batch time: 0.0042s vs `on\_train\_batch\_begin` time: 0.0223s). Check your callbacks.

WARNING:tensorflow:Callback method `on\_train\_batch\_end` is slow compared to the batch time (batch time: 0.0042s vs `on\_train\_batch\_end` time: 0.0190s). Check your callbacks.

1465/1472 [=====>.] - ETA: 0s - loss: 0.0206 - accuracy: 0.0422

micro\_F1\_train: 0.029421757151593394

micro\_F1\_val: 0.03138871274571972

Epoch 00001: val\_accuracy improved from -inf to 0.03139, saving model to model1\_save/weights-01-0.0314.hdf5

1472/1472 [=====>.] - 11s 7ms/step - loss: 0.0205 - accuracy: 0.0421 - val\_loss: 3.2352 - val\_accuracy: 0.0314 - lr: 0.0100

Epoch 2/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0185 - accuracy: 0.0435

micro\_F1\_train: 0.03981789766936196

micro\_F1\_val: 0.040424857324032976

Epoch 00002: val\_accuracy improved from 0.03139 to 0.04042, saving model to model1\_save/weights-02-0.0404.hdf5

1472/1472 [=====>.] - 10s 6ms/step - loss: 0.0185 - accuracy: 0.0434 - val\_loss: 3.0865 - val\_accuracy: 0.0404 - lr: 0.0100

Epoch 3/100

1460/1472 [=====>.] - ETA: 0s - loss: 0.0173 - accuracy: 0.0481

micro\_F1\_train: 0.0485832710470884

micro\_F1\_val: 0.04787571337983513

Epoch 00003: val\_accuracy improved from 0.04042 to 0.04788, saving model to model1\_save/weights-03-0.0479.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0173 - accuracy: 0.0482 - val\_loss: 3.0311 - val\_accuracy: 0.0479 - lr: 0.0100

Epoch 4/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0167 - accuracy: 0.0512

micro\_F1\_train: 0.05415505877556567

micro\_F1\_val: 0.05294863665187064

Epoch 00004: val\_accuracy improved from 0.04788 to 0.05295, saving model to model1\_save/weights-04-0.0529.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0167 - accuracy: 0.0512 - val\_loss: 3.0004 - val\_accuracy: 0.0529 - lr: 0.0100

Epoch 5/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0163 - accuracy: 0.0513

micro\_F1\_train: 0.05809607936400082

micro\_F1\_val: 0.056753329105897275

Epoch 00005: val\_accuracy improved from 0.05295 to 0.05675, saving model to model1\_save/weights-05-0.0568.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0164 - accuracy: 0.0514 - val\_loss: 2.9746 - val\_accuracy: 0.0568 - lr: 0.0100

Epoch 6/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0161 - accuracy: 0.0531

micro\_F1\_train: 0.06047428144322892

micro\_F1\_val: 0.059606848446417254

Epoch 00006: val\_accuracy improved from 0.05675 to 0.05961, saving model to model1\_save/weights-06-0.0596.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0161 - accuracy: 0.0533 - val\_loss: 2.9592 - val\_accuracy: 0.0596 - lr: 0.0100

Epoch 7/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0157 - accuracy: 0.0570

micro\_F1\_train: 0.06292043215329211

micro\_F1\_val: 0.06246036778693722

Epoch 00007: val\_accuracy improved from 0.05961 to 0.06246, saving model to model1\_save/weights-07-0.0625.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0157 - accuracy: 0.0570 - val\_loss: 2.9489 - val\_accuracy: 0.0625 - lr: 0.0100

Epoch 8/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0156 - accuracy: 0.0576

micro\_F1\_train: 0.06400761024665352

micro\_F1\_val: 0.06293595434369055

Epoch 00008: val\_accuracy improved from 0.06246 to 0.06294, saving model to model1\_save/weights-08-0.0629.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0156 - accuracy: 0.0578 - val\_loss: 2.9436 - val\_accuracy: 0.0629 - lr: 0.0100

Epoch 9/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0154 - accuracy: 0.0613

micro\_F1\_train: 0.06550248012502548

micro\_F1\_val: 0.06515535827520609

Epoch 00009: val\_accuracy improved from 0.06294 to 0.06516, saving model to model1\_save/weights-09-0.0652.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0154 - accuracy: 0.0613 - val\_loss: 2.9333 - val\_accuracy: 0.0652 - lr: 0.0100

Epoch 10/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0151 - accuracy: 0.0624

micro\_F1\_train: 0.06631786369504654

micro\_F1\_val: 0.06563094483195941

Epoch 00010: val\_accuracy improved from 0.06516 to 0.06563, saving model to model1\_save/weights-10-0.0656.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0151 - accuracy: 0.0623 - val\_loss: 2.9269 - val\_accuracy: 0.0656 - lr: 0.0100

Epoch 11/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0152 - accuracy: 0.0634

micro\_F1\_train: 0.06611401780254128

micro\_F1\_val: 0.06800887761572606

Epoch 00011: val\_accuracy improved from 0.06563 to 0.06801, saving model to model1\_save/weights-11-0.0680.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0152 - accuracy: 0.0635 - val\_loss: 2.9138 - val\_accuracy: 0.0680 - lr: 0.0100

Epoch 12/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0150 - accuracy: 0.0641

micro\_F1\_train: 0.06747299041924305

micro\_F1\_val: 0.06927710843373494

Epoch 00012: val\_accuracy improved from 0.06801 to 0.06928, saving model to model1\_save/weights-12-0.0693.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0150 - accuracy: 0.0641 - val\_loss: 2.9077 - val\_accuracy: 0.0693 - lr: 0.0100

Epoch 13/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0148 - accuracy: 0.0702

micro\_F1\_train: 0.07025888428348169

micro\_F1\_val: 0.07244768547875713

Epoch 00013: val\_accuracy improved from 0.06928 to 0.07245, saving model to model1\_save/weights-13-0.0724.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0148 - accuracy: 0.0702 - val\_loss: 2.9007 - val\_accuracy: 0.0724 - lr: 0.0100

Epoch 14/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0146 - accuracy: 0.0641

micro\_F1\_train: 0.07297682951688524

micro\_F1\_val: 0.07276474318325936

Epoch 00014: val\_accuracy improved from 0.07245 to 0.07276, saving model to model1\_save/weights-14-0.0728.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0146 - accuracy: 0.0641 - val\_loss: 2.8893 - val\_accuracy: 0.0728 - lr: 0.0100

Epoch 15/100

1460/1472 [=====>.] - ETA: 0s - loss: 0.0145 - accuracy: 0.0694

micro\_F1\_train: 0.07290888088605015

micro\_F1\_val: 0.07260621433100824

Epoch 00015: val\_accuracy did not improve from 0.07276

1472/1472 [=====] - 9s 6ms/step - loss: 0.0146 - accuracy: 0.0691 - val\_loss: 2.8785 - val\_accuracy: 0.0726 - lr: 0.0100

Epoch 16/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0145 - accuracy: 0.0725

micro\_F1\_train: 0.07297682951688524

micro\_F1\_val: 0.07403297400126824

Epoch 00016: val\_accuracy improved from 0.07276 to 0.07403, saving model to model1\_save/weights-16-0.0740.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0145 - accuracy: 0.0724 - val\_loss: 2.8714 - val\_accuracy: 0.0740 - lr: 0.0100

Epoch 17/100

1472/1472 [=====] - ETA: 0s - loss: 0.0144 - accuracy: 0.0736

micro\_F1\_train: 0.07583067201195896

micro\_F1\_val: 0.0767279644895371

Epoch 00017: val\_accuracy improved from 0.07403 to 0.07673, saving model to model1\_save/weights-17-0.0767.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0144 - accuracy: 0.0736 - val\_loss: 2.8599 - val\_accuracy: 0.0767 - lr: 0.0100

Epoch 18/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0142 - accuracy: 0.0795

micro\_F1\_train: 0.0773255418903309

micro\_F1\_val: 0.07894736842105263

Epoch 00018: val\_accuracy improved from 0.07673 to 0.07895, saving model to model1\_save/weights-18-0.0789.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0142 - accuracy: 0.0796 - val\_loss: 2.8479 - val\_accuracy: 0.0789 - lr: 0.0100

Epoch 19/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0141 - accuracy: 0.0786

micro\_F1\_train: 0.07949989807705375

micro\_F1\_val: 0.08037412809131261

Epoch 00019: val\_accuracy improved from 0.07895 to 0.08037, saving model to model1\_save/weights-19-0.0804.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0141 - accuracy: 0.0786 - val\_loss: 2.8364 - val\_accuracy: 0.0804 - lr: 0.0100

Epoch 20/100

1461/1472 [=====>.] - ETA: 0s - loss: 0.0141 - accuracy: 0.0765

micro\_F1\_train: 0.08167425426377659

micro\_F1\_val: 0.08275206087507926

Epoch 00020: val\_accuracy improved from 0.08037 to 0.08275, saving model to model1\_save/weights-20-0.0828.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0141 - accuracy: 0.0764 - val\_loss: 2.8243 - val\_accuracy: 0.0828 - lr: 0.0100

Epoch 21/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0140 - accuracy: 0.0853

micro\_F1\_train: 0.08262553509546783

micro\_F1\_val: 0.08370323398858592

Epoch 00021: val\_accuracy improved from 0.08275 to 0.08370, saving model to model1\_save/weights-21-0.0837.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0139 - accuracy: 0.0853 - val\_loss: 2.8107 - val\_accuracy: 0.0837 - lr: 0.0100

Epoch 22/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0138 - accuracy: 0.0857

micro\_F1\_train: 0.08357681592715907

micro\_F1\_val: 0.08639822447685479

Epoch 00022: val\_accuracy improved from 0.08370 to 0.08640, saving model to model1\_save/weights-22-0.0864.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0138 - accuracy: 0.0859 - val\_loss: 2.7995 - val\_accuracy: 0.0864 - lr: 0.0100

Epoch 23/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0137 - accuracy: 0.0907

micro\_F1\_train: 0.08663450431473806

micro\_F1\_val: 0.08877615726062144

Epoch 00023: val\_accuracy improved from 0.08640 to 0.08878, saving model to model1\_save/weights-23-0.0888.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0137 - accuracy: 0.0908 - val\_loss: 2.7869 - val\_accuracy: 0.0888 - lr: 0.0100

Epoch 24/100

1472/1472 [=====] - ETA: 0s - loss: 0.0136 - accuracy: 0.0915

micro\_F1\_train: 0.08785757966976965

micro\_F1\_val: 0.09051997463538364

Epoch 00024: val\_accuracy improved from 0.08878 to 0.09052, saving model to model1\_save/weights-24-0.0905.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0136 - accuracy: 0.0915 - val\_loss: 2.7717 - val\_accuracy: 0.0905 - lr: 0.0100

Epoch 25/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0136 - accuracy: 0.0919

micro\_F1\_train: 0.09009988448732759

micro\_F1\_val: 0.09273937856689918

Epoch 00025: val\_accuracy improved from 0.09052 to 0.09274, saving model to model1\_save/weights-25-0.0927.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0136 - accuracy: 0.0919 - val\_loss: 2.7600 - val\_accuracy: 0.0927 - lr: 0.0100

Epoch 26/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0135 - accuracy: 0.0977

micro\_F1\_train: 0.0917306516273697

micro\_F1\_val: 0.09590995561192138

Epoch 00026: val\_accuracy improved from 0.09274 to 0.09591, saving model to model1\_save/weights-26-0.0959.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0134 - accuracy: 0.0978 - val\_loss: 2.7437 - val\_accuracy: 0.0959 - lr: 0.0100

Epoch 27/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0134 - accuracy: 0.0964

micro\_F1\_train: 0.0941768023374329

micro\_F1\_val: 0.0963855421686747

Epoch 00027: val\_accuracy improved from 0.09591 to 0.09639, saving model to model1\_save/weights-27-0.0964.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0134 - accuracy: 0.0965 - val\_loss: 2.7324 - val\_accuracy: 0.0964 - lr: 0.0100

Epoch 28/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0133 - accuracy: 0.0967

micro\_F1\_train: 0.09641910715499083

micro\_F1\_val: 0.09844641724793912

Epoch 00028: val\_accuracy improved from 0.09639 to 0.09845, saving model to model1\_save/weights-28-0.0984.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0133 - accuracy: 0.0967 - val\_loss: 2.7198 - val\_accuracy: 0.0984 - lr: 0.0100

Epoch 29/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0133 - accuracy: 0.1029

micro\_F1\_train: 0.0971665420941768

micro\_F1\_val: 0.09892200380469246

Epoch 00029: val\_accuracy improved from 0.09845 to 0.09892, saving model to model1\_save/weights-29-0.0989.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0133 - accuracy: 0.1031 - val\_loss: 2.7076 - val\_accuracy: 0.0989 - lr: 0.0100

Epoch 30/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0132 - accuracy: 0.1046

micro\_F1\_train: 0.09798192566419786

micro\_F1\_val: 0.10098287888395688

Epoch 00030: val\_accuracy improved from 0.09892 to 0.10098, saving model to model1\_save/weights-30-0.1010.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0132 - accuracy: 0.1047 - val\_loss: 2.6948 - val\_accuracy: 0.1010 - lr: 0.0100

Epoch 31/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0130 - accuracy: 0.1029

micro\_F1\_train: 0.10056397363593124



micro\_F1\_val: 0.10304375396322131

Epoch 00031: val\_accuracy improved from 0.10098 to 0.10304, saving model to model1\_save/weights-31-0.1030.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0130 - accuracy: 0.1029 - val\_loss: 2.6780 - val\_accuracy: 0.1030 - lr: 0.0100

Epoch 32/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0130 - accuracy: 0.1113

micro\_F1\_train: 0.1002242304817558

micro\_F1\_val: 0.10336081166772351

Epoch 00032: val\_accuracy improved from 0.10304 to 0.10336, saving model to model1\_save/weights-32-0.1034.hdf5

1472/1472 [=====] - 12s 8ms/step - loss: 0.0130 - accuracy: 0.1112 - val\_loss: 2.6597 - val\_accuracy: 0.1034 - lr: 0.0100

Epoch 33/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0129 - accuracy: 0.1053

micro\_F1\_train: 0.10042807637426107

micro\_F1\_val: 0.10415345592897908

Epoch 00033: val\_accuracy improved from 0.10336 to 0.10415, saving model to model1\_save/weights-33-0.1042.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0129 - accuracy: 0.1053 - val\_loss: 2.6503 - val\_accuracy: 0.1042 - lr: 0.0100

Epoch 34/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0128 - accuracy: 0.1100

micro\_F1\_train: 0.1019908948834681

micro\_F1\_val: 0.10605580215599239

Epoch 00034: val\_accuracy improved from 0.10415 to 0.10606, saving model to model1\_save/weights-34-0.1061.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0128 - accuracy: 0.1101 - val\_loss: 2.6321 - val\_accuracy: 0.1061 - lr: 0.0100

Epoch 35/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0126 - accuracy: 0.1121

micro\_F1\_train: 0.10470884011687165

micro\_F1\_val: 0.10811667723525682

Epoch 00035: val\_accuracy improved from 0.10606 to 0.10812, saving model to model1\_save/weights-35-0.1081.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0126 - accuracy: 0.1122 - val\_loss: 2.6141 - val\_accuracy: 0.1081 - lr: 0.0100

Epoch 36/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0125 - accuracy: 0.1132

micro\_F1\_train: 0.10695114493442957

micro\_F1\_val: 0.1095434369055168

Epoch 00036: val\_accuracy improved from 0.10812 to 0.10954, saving model to model1\_save/weights-36-0.1095.hdf5

1472/1472 [=====] - 14s 9ms/step - loss: 0.0125 - accuracy: 0.1129 - val\_loss: 2.5957 - val\_accuracy: 0.1095 - lr: 0.0100

Epoch 37/100

1467/1472 [=====>.] - ETA: 0s - loss: 0.0124 - accuracy: 0.1186

micro\_F1\_train: 0.10966909016783312

micro\_F1\_val: 0.11303107165504121

Epoch 00037: val\_accuracy improved from 0.10954 to 0.11303, saving model to model1\_save/weights-37-0.1130.hdf5

1472/1472 [=====] - 14s 10ms/step - loss: 0.0124 - accuracy: 0.1186 - val\_loss: 2.5786 - val\_accuracy: 0.1130 - lr: 0.0100

Epoch 38/100

1472/1472 [=====] - ETA: 0s - loss: 0.0125 - accuracy: 0.1114

micro\_F1\_train: 0.11021267921451385

micro\_F1\_val: 0.11223842739378567

Epoch 00038: val\_accuracy did not improve from 0.11303

1472/1472 [=====] - 9s 6ms/step - loss: 0.0125 - accuracy: 0.1114 - val\_loss: 2.5641 - val\_accuracy: 0.1122 - lr: 0.0100

Epoch 39/100

1472/1472 [=====] - ETA: 0s - loss: 0.0123 - accuracy: 0.1182

micro\_F1\_train: 0.11225113813956648

micro\_F1\_val: 0.11334812935954344

Epoch 00039: val\_accuracy improved from 0.11303 to 0.11335, saving model to model1\_save/weights-39-0.1133.hdf5

1472/1472 [=====] - 14s 10ms/step - loss: 0.0123 - accuracy: 0.1182 - val\_loss: 2.5477 - val\_accuracy: 0.1133 - lr: 0.0100

Epoch 40/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0120 - accuracy: 0.1177

micro\_F1\_train: 0.1163960046205069

micro\_F1\_val: 0.11715282181357008

Epoch 00040: val\_accuracy improved from 0.11335 to 0.11715, saving model to model1\_save/weights-40-0.1172.hdf5

1472/1472 [=====] - 15s 10ms/step - loss: 0.0120 - accuracy: 0.1176 - val\_loss: 2.5277 - val\_accuracy: 0.1172 - lr: 0.0100

Epoch 41/100

1472/1472 [=====] - ETA: 0s - loss: 0.0121 - accuracy: 0.1236

micro\_F1\_train: 0.12101651151729292

micro\_F1\_val: 0.12175015852885225

Epoch 00041: val\_accuracy improved from 0.11715 to 0.12175, saving model to model1\_save/weights-41-0.1218.hdf5

1472/1472 [=====] - 16s 11ms/step - loss: 0.0121 - accuracy: 0.1236 - val\_loss: 2.5059 - val\_accuracy: 0.1218 - lr: 0.0100

Epoch 42/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0120 - accuracy: 0.1224

micro\_F1\_train: 0.12393830264320174

micro\_F1\_val: 0.12444514901712111

Epoch 00042: val\_accuracy improved from 0.12175 to 0.12445, saving model to model1\_save/weights-42-0.1244.hdf5

1472/1472 [=====] - 18s 12ms/step - loss: 0.0120 - accuracy: 0.1224 - val\_loss: 2.4902 - val\_accuracy: 0.1244 - lr: 0.0100

Epoch 43/100

1472/1472 [=====] - ETA: 0s - loss: 0.0120 - accuracy: 0.1277

micro\_F1\_train: 0.1248216348440579

micro\_F1\_val: 0.12682308180088775

Epoch 00043: val\_accuracy improved from 0.12445 to 0.12682, saving model to model1\_save/weights-43-0.1268.hdf5

1472/1472 [=====] - 17s 11ms/step - loss: 0.0120 - accuracy: 0.1277 - val\_loss: 2.4821 - val\_accuracy: 0.1268 - lr: 0.0100

Epoch 44/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0118 - accuracy: 0.1248

micro\_F1\_train: 0.12624855609159474

micro\_F1\_val: 0.12999365884590997

Epoch 00044: val\_accuracy improved from 0.12682 to 0.12999, saving model to model1\_save/weights-44-0.1300.hdf5

1472/1472 [=====] - 15s 10ms/step - loss: 0.0118 - accuracy: 0.1248 - val\_loss: 2.4689 - val\_accuracy: 0.1300 - lr: 0.0100

Epoch 45/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0118 - accuracy: 0.1289

micro\_F1\_train: 0.12849086090915268

micro F1 val: 0.13157894736842105

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Epoch 00045: val\_accuracy improved from 0.12999 to 0.13158, saving model to model1\_save/weights-45-0.1316.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0118 - accuracy: 0.1287 - val\_loss: 2.4516 - val\_accuracy: 0.1316 - lr: 0.0100  
Epoch 46/100  
1461/1472 [=====>.] - ETA: 0s - loss: 0.0117 - accuracy: 0.1266

micro\_F1\_train: 0.13229598423591765

micro\_F1\_val: 0.13474952441344323

Epoch 00046: val\_accuracy improved from 0.13158 to 0.13475, saving model to model1\_save/weights-46-0.1347.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0117 - accuracy: 0.1265 - val\_loss: 2.4355 - val\_accuracy: 0.1347 - lr: 0.0100  
Epoch 47/100  
1461/1472 [=====>.] - ETA: 0s - loss: 0.0115 - accuracy: 0.1307

micro\_F1\_train: 0.13358700822178432

micro\_F1\_val: 0.1355421686746988

Epoch 00047: val\_accuracy improved from 0.13475 to 0.13554, saving model to model1\_save/weights-47-0.1355.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0116 - accuracy: 0.1305 - val\_loss: 2.4162 - val\_accuracy: 0.1355 - lr: 0.0100  
Epoch 48/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0116 - accuracy: 0.1334

micro\_F1\_train: 0.13603315893184753

micro\_F1\_val: 0.13855421686746988

Epoch 00048: val\_accuracy improved from 0.13554 to 0.13855, saving model to model1\_save/weights-48-0.1386.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0115 - accuracy: 0.1333 - val\_loss: 2.4009 - val\_accuracy: 0.1386 - lr: 0.0100  
Epoch 49/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0115 - accuracy: 0.1363

micro\_F1\_train: 0.1390228986885914

micro\_F1\_val: 0.14124920735573873

Epoch 00049: val\_accuracy improved from 0.13855 to 0.14125, saving model to model1\_save/weights-49-0.1412.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0115 - accuracy: 0.1362 - val\_loss: 2.3851 - val\_accuracy: 0.1412 - lr: 0.0100  
Epoch 50/100  
1471/1472 [=====>.] - ETA: 0s - loss: 0.0114 - accuracy: 0.1375

micro\_F1\_train: 0.1391587959502616

micro\_F1\_val: 0.1404565630944832

Epoch 00050: val\_accuracy did not improve from 0.14125  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0114 - accuracy: 0.1374 - val\_loss: 2.3712 - val\_accuracy: 0.1405 - lr: 0.0100  
Epoch 51/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0113 - accuracy: 0.1388

micro\_F1\_train: 0.14561391587959502

micro\_F1\_val: 0.1469562460367787

Epoch 00051: val\_accuracy improved from 0.14125 to 0.14696, saving model to model1\_save/weights-51-0.1470.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0113 - accuracy: 0.1388 - val\_loss: 2.3564 - val\_accuracy: 0.1470 - lr: 0.0100  
Epoch 52/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0114 - accuracy: 0.1381

micro\_F1\_train: 0.14683699123462662

micro\_F1\_val: 0.14774889029803423

Epoch 00052: val\_accuracy improved from 0.14696 to 0.14775, saving model to model1\_save/weights-52-0.1477.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0114 - accuracy: 0.1382 - val\_loss: 2.3439 - val\_accuracy: 0.1477 - lr: 0.0100  
Epoch 53/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0111 - accuracy: 0.1410  
  
micro\_F1\_train: 0.1470408371271319  
  
micro\_F1\_val: 0.14806594800253647  
  
Epoch 00053: val\_accuracy improved from 0.14775 to 0.14807, saving model to model1\_save/weights-53-0.1481.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0111 - accuracy: 0.1409 - val\_loss: 2.3270 - val\_accuracy: 0.1481 - lr: 0.0100  
Epoch 54/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0111 - accuracy: 0.1419  
  
micro\_F1\_train: 0.15091390908473196  
  
micro\_F1\_val: 0.15060240963855423  
  
Epoch 00054: val\_accuracy improved from 0.14807 to 0.15060, saving model to model1\_save/weights-54-0.1506.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0111 - accuracy: 0.1419 - val\_loss: 2.3103 - val\_accuracy: 0.1506 - lr: 0.0100  
Epoch 55/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0111 - accuracy: 0.1481  
  
micro\_F1\_train: 0.1553985187198478  
  
micro\_F1\_val: 0.15298034242232086  
  
Epoch 00055: val\_accuracy improved from 0.15060 to 0.15298, saving model to model1\_save/weights-55-0.1530.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0111 - accuracy: 0.1482 - val\_loss: 2.2968 - val\_accuracy: 0.1530 - lr: 0.0100  
Epoch 56/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0109 - accuracy: 0.1477  
  
micro\_F1\_train: 0.15444723788815656  
  
micro\_F1\_val: 0.15171211160431197  
  
Epoch 00056: val\_accuracy did not improve from 0.15298  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0109 - accuracy: 0.1474 - val\_loss: 2.2846 - val\_accuracy: 0.1517 - lr: 0.0100  
Epoch 57/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0110 - accuracy: 0.1441  
  
micro\_F1\_train: 0.15533057008901272  
  
micro\_F1\_val: 0.15456563094483197  
  
Epoch 00057: val\_accuracy improved from 0.15298 to 0.15457, saving model to model1\_save/weights-57-0.1546.hdf5  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0110 - accuracy: 0.1443 - val\_loss: 2.2758 - val\_accuracy: 0.1546 - lr: 0.0100  
Epoch 58/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0109 - accuracy: 0.1449  
  
micro\_F1\_train: 0.15811646395325135  
  
micro\_F1\_val: 0.1566265060240964  
  
Epoch 00058: val\_accuracy improved from 0.15457 to 0.15663, saving model to model1\_save/weights-58-0.1566.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0110 - accuracy: 0.1449 - val\_loss: 2.2677 - val\_accuracy: 0.1566 - lr: 0.0100  
Epoch 59/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0109 - accuracy: 0.1526  
  
micro\_F1\_train: 0.16029082013997417  
  
micro\_F1\_val: 0.15726062143310082  
  
Epoch 00059: val accuracy improved from 0.15663 to 0.15726. saving model to model1\_save/weights-59-0.15

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Epoch 00059: val_accuracy improved from 0.1522 to 0.1573, saving model to model1_save/weights-59-0.16
73.hdf5
1472/1472 [=====] - 11s 8ms/step - loss: 0.0109 - accuracy: 0.1522 - val_loss:
2.2620 - val_accuracy: 0.1573 - lr: 0.0100
Epoch 60/100
1469/1472 [=====>.] - ETA: 0s - loss: 0.0109 - accuracy: 0.1499

micro_F1_train: 0.16423184072840932

micro_F1_val: 0.1621750158528852

Epoch 00060: val_accuracy improved from 0.15726 to 0.16218, saving model to model1_save/weights-60-0.16
22.hdf5
1472/1472 [=====] - 9s 6ms/step - loss: 0.0109 - accuracy: 0.1498 - val_loss:
2.2497 - val_accuracy: 0.1622 - lr: 0.0100
Epoch 61/100
1468/1472 [=====>.] - ETA: 0s - loss: 0.0107 - accuracy: 0.1512

micro_F1_train: 0.16470748114425493

micro_F1_val: 0.16296766011414077

Epoch 00061: val_accuracy improved from 0.16218 to 0.16297, saving model to model1_save/weights-61-0.16
30.hdf5
1472/1472 [=====] - 10s 7ms/step - loss: 0.0107 - accuracy: 0.1513 - val_loss:
2.2453 - val_accuracy: 0.1630 - lr: 0.0100
Epoch 62/100
1467/1472 [=====>.] - ETA: 0s - loss: 0.0107 - accuracy: 0.1519

micro_F1_train: 0.16715363185431814

micro_F1_val: 0.16471147748890297

Epoch 00062: val_accuracy improved from 0.16297 to 0.16471, saving model to model1_save/weights-62-0.16
47.hdf5
1472/1472 [=====] - 9s 6ms/step - loss: 0.0107 - accuracy: 0.1519 - val_loss:
2.2398 - val_accuracy: 0.1647 - lr: 0.0100
Epoch 63/100
1472/1472 [=====] - ETA: 0s - loss: 0.0106 - accuracy: 0.1486

micro_F1_train: 0.16654209417680232

micro_F1_val: 0.161857958148383

Epoch 00063: val_accuracy did not improve from 0.16471
1472/1472 [=====] - 12s 8ms/step - loss: 0.0106 - accuracy: 0.1486 - val_loss:
2.2291 - val_accuracy: 0.1619 - lr: 0.0100
Epoch 64/100
1472/1472 [=====] - ETA: 0s - loss: 0.0106 - accuracy: 0.1503

micro_F1_train: 0.16756132363932866

micro_F1_val: 0.16280913126188967

Epoch 00064: val_accuracy did not improve from 0.16471
1472/1472 [=====] - 14s 9ms/step - loss: 0.0106 - accuracy: 0.1503 - val_loss:
2.2246 - val_accuracy: 0.1628 - lr: 0.0100
Epoch 65/100
1466/1472 [=====>.] - ETA: 0s - loss: 0.0105 - accuracy: 0.1543

micro_F1_train: 0.17055106339607257

micro_F1_val: 0.16772352568167406

Epoch 00065: val_accuracy improved from 0.16471 to 0.16772, saving model to model1_save/weights-65-0.16
77.hdf5
1472/1472 [=====] - 11s 8ms/step - loss: 0.0106 - accuracy: 0.1542 - val_loss:
2.2137 - val_accuracy: 0.1677 - lr: 0.0100
Epoch 66/100
1468/1472 [=====>.] - ETA: 0s - loss: 0.0105 - accuracy: 0.1563

micro_F1_train: 0.17313311136780593

micro_F1_val: 0.1701014584654407

Epoch 00066: val_accuracy improved from 0.16772 to 0.17010, saving model to model1_save/weights-66-0.17
01.hdf5
1472/1472 [=====] - 12s 8ms/step - loss: 0.0105 - accuracy: 0.1566 - val_loss:
```

```
1472/1472 [=====] - 12s 8ms/step - loss: 0.0106 - accuracy: 0.1509 - val_loss: 2.2068 - val_accuracy: 0.1701 - lr: 0.0100
Epoch 67/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0106 - accuracy: 0.1509

micro_F1_train: 0.17544336481619896

micro_F1_val: 0.17089410272669625

Epoch 00067: val_accuracy improved from 0.17010 to 0.17089, saving model to model1_save/weights-67-0.1709.hdf5
1472/1472 [=====] - 15s 10ms/step - loss: 0.0106 - accuracy: 0.1513 - val_loss: 2.2047 - val_accuracy: 0.1709 - lr: 0.0100
Epoch 68/100
1468/1472 [=====>.] - ETA: 0s - loss: 0.0105 - accuracy: 0.1529

micro_F1_train: 0.17938438540463408

micro_F1_val: 0.17533291058972733

Epoch 00068: val_accuracy improved from 0.17089 to 0.17533, saving model to model1_save/weights-68-0.1753.hdf5
1472/1472 [=====] - 14s 10ms/step - loss: 0.0105 - accuracy: 0.1527 - val_loss: 2.1946 - val_accuracy: 0.1753 - lr: 0.0100
Epoch 69/100
1472/1472 [=====] - ETA: 0s - loss: 0.0104 - accuracy: 0.1585

micro_F1_train: 0.17591900523204457

micro_F1_val: 0.16978440076093848

Epoch 00069: val_accuracy did not improve from 0.17533
1472/1472 [=====] - 15s 10ms/step - loss: 0.0104 - accuracy: 0.1585 - val_loss: 2.1931 - val_accuracy: 0.1698 - lr: 0.0100
Epoch 70/100
1469/1472 [=====>.] - ETA: 0s - loss: 0.0104 - accuracy: 0.1609

micro_F1_train: 0.17782156689542702

micro_F1_val: 0.17200380469245402

Epoch 00070: val_accuracy did not improve from 0.17533
1472/1472 [=====] - 17s 12ms/step - loss: 0.0104 - accuracy: 0.1609 - val_loss: 2.1871 - val_accuracy: 0.1720 - lr: 0.0100
Epoch 71/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0104 - accuracy: 0.1557

micro_F1_train: 0.1782972073112727

micro_F1_val: 0.17247939124920736

Epoch 00071: val_accuracy did not improve from 0.17533
1472/1472 [=====] - 16s 11ms/step - loss: 0.0104 - accuracy: 0.1557 - val_loss: 2.1866 - val_accuracy: 0.1725 - lr: 0.0100
Epoch 72/100
1472/1472 [=====] - ETA: 0s - loss: 0.0102 - accuracy: 0.1608

micro_F1_train: 0.17591900523204457

micro_F1_val: 0.1716867469879518

Epoch 00072: val_accuracy did not improve from 0.17533
1472/1472 [=====] - 12s 8ms/step - loss: 0.0102 - accuracy: 0.1608 - val_loss: 2.1843 - val_accuracy: 0.1717 - lr: 0.0100
Epoch 73/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0103 - accuracy: 0.1610

micro_F1_train: 0.17965617992797442

micro_F1_val: 0.1751743817374762

Epoch 00073: val_accuracy did not improve from 0.17533
1472/1472 [=====] - 9s 6ms/step - loss: 0.0103 - accuracy: 0.1610 - val_loss: 2.1765 - val_accuracy: 0.1752 - lr: 0.0100
Epoch 74/100
1461/1472 [=====>.] - ETA: 0s - loss: 0.0102 - accuracy: 0.1532

micro_F1_train: 0.18774206699735
```

micro\_F1\_train: 0.17774200097155

micro\_F1\_val: 0.17945466074825622

Epoch 00074: val\_accuracy improved from 0.17533 to 0.17945, saving model to model1\_save/weights-74-0.1795.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0102 - accuracy: 0.1532 - val\_loss: 2.1708 - val\_accuracy: 0.1795 - lr: 0.0100  
Epoch 75/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0103 - accuracy: 0.1578

micro\_F1\_train: 0.18624719711897803

micro\_F1\_val: 0.17913760304375398

Epoch 00075: val\_accuracy did not improve from 0.17945  
1472/1472 [=====] - 12s 8ms/step - loss: 0.0103 - accuracy: 0.1579 - val\_loss: 2.1671 - val\_accuracy: 0.1791 - lr: 0.0100  
Epoch 76/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0100 - accuracy: 0.1601

micro\_F1\_train: 0.1906638581232588

micro\_F1\_val: 0.18611287254280276

Epoch 00076: val\_accuracy improved from 0.17945 to 0.18611, saving model to model1\_save/weights-76-0.1861.hdf5  
1472/1472 [=====] - 12s 8ms/step - loss: 0.0100 - accuracy: 0.1601 - val\_loss: 2.1587 - val\_accuracy: 0.1861 - lr: 0.0100  
Epoch 77/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0101 - accuracy: 0.1609

micro\_F1\_train: 0.19841000203845893

micro\_F1\_val: 0.19340519974635384

Epoch 00077: val\_accuracy improved from 0.18611 to 0.19341, saving model to model1\_save/weights-77-0.1934.hdf5  
1472/1472 [=====] - 14s 9ms/step - loss: 0.0101 - accuracy: 0.1610 - val\_loss: 2.1506 - val\_accuracy: 0.1934 - lr: 0.0100  
Epoch 78/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0100 - accuracy: 0.1670

micro\_F1\_train: 0.20126384453353266

micro\_F1\_val: 0.19546607482561829

Epoch 00078: val\_accuracy improved from 0.19341 to 0.19547, saving model to model1\_save/weights-78-0.1955.hdf5  
1472/1472 [=====] - 18s 12ms/step - loss: 0.0101 - accuracy: 0.1671 - val\_loss: 2.1462 - val\_accuracy: 0.1955 - lr: 0.0100  
Epoch 79/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0102 - accuracy: 0.1661

micro\_F1\_train: 0.19643949174424136

micro\_F1\_val: 0.1911857958148383

Epoch 00079: val\_accuracy did not improve from 0.19547  
1472/1472 [=====] - 14s 9ms/step - loss: 0.0102 - accuracy: 0.1661 - val\_loss: 2.1551 - val\_accuracy: 0.1912 - lr: 0.0100  
Epoch 80/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0101 - accuracy: 0.1615

micro\_F1\_train: 0.19474077597336414

micro\_F1\_val: 0.19007609384908053

Epoch 00080: val\_accuracy did not improve from 0.19547  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0101 - accuracy: 0.1614 - val\_loss: 2.1543 - val\_accuracy: 0.1901 - lr: 0.0100  
Epoch 81/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0100 - accuracy: 0.1656

micro\_F1\_train: 0.19154719032411494

micro\_F1\_val: 0.18738110336081168

Epoch 00081: val\_accuracy did not improve from 0.19547  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0101 - accuracy: 0.1655 - val\_loss: 2.1490 - val\_accuracy: 0.1874 - lr: 0.0090  
Epoch 82/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0101 - accuracy: 0.1591  
  
micro\_F1\_train: 0.19657538900591154  
  
micro\_F1\_val: 0.1932466708941027  
  
Epoch 00082: val\_accuracy did not improve from 0.19547  
1472/1472 [=====] - 18s 12ms/step - loss: 0.0101 - accuracy: 0.1589 - val\_loss: 2.1436 - val\_accuracy: 0.1932 - lr: 0.0090  
Epoch 83/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0099 - accuracy: 0.1668  
  
micro\_F1\_train: 0.19888564245430454  
  
micro\_F1\_val: 0.19483195941661383  
  
Epoch 00083: val\_accuracy did not improve from 0.19547  
1472/1472 [=====] - 13s 9ms/step - loss: 0.0099 - accuracy: 0.1670 - val\_loss: 2.1397 - val\_accuracy: 0.1948 - lr: 0.0090  
Epoch 84/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0101 - accuracy: 0.1688  
  
micro\_F1\_train: 0.2003805123326765  
  
micro\_F1\_val: 0.19657577679137603  
  
Epoch 00084: val\_accuracy improved from 0.19547 to 0.19658, saving model to model1\_save/weights-84-0.1966.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0101 - accuracy: 0.1693 - val\_loss: 2.1371 - val\_accuracy: 0.1966 - lr: 0.0090  
Epoch 85/100  
1471/1472 [=====>.] - ETA: 0s - loss: 0.0101 - accuracy: 0.1610  
  
micro\_F1\_train: 0.19752666983760278  
  
micro\_F1\_val: 0.1941978440076094  
  
Epoch 00085: val\_accuracy did not improve from 0.19658  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0101 - accuracy: 0.1610 - val\_loss: 2.1427 - val\_accuracy: 0.1942 - lr: 0.0090  
Epoch 86/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0099 - accuracy: 0.1655  
  
micro\_F1\_train: 0.201535639056873  
  
micro\_F1\_val: 0.19720989220038046  
  
Epoch 00086: val\_accuracy improved from 0.19658 to 0.19721, saving model to model1\_save/weights-86-0.1972.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0099 - accuracy: 0.1657 - val\_loss: 2.1301 - val\_accuracy: 0.1972 - lr: 0.0090  
Epoch 87/100  
1465/1472 [=====>.] - ETA: 0s - loss: 0.0098 - accuracy: 0.1653  
  
micro\_F1\_train: 0.19949718013182036  
  
micro\_F1\_val: 0.19800253646163601  
  
Epoch 00087: val\_accuracy improved from 0.19721 to 0.19800, saving model to model1\_save/weights-87-0.1980.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0098 - accuracy: 0.1655 - val\_loss: 2.1352 - val\_accuracy: 0.1980 - lr: 0.0090  
Epoch 88/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0098 - accuracy: 0.1643  
  
micro\_F1\_train: 0.20568050553781342  
  
micro\_F1\_val: 0.2010145846544071  
  
Epoch 00088: val\_accuracy improved from 0.19800 to 0.20101, saving model to model1\_save/weights-88-0.2010.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0098 - accuracy: 0.1644 - val\_loss: 2.1342 - val\_accuracy: 0.2010 - lr: 0.0090



```
2.1242 - val_accuracy: 0.2010 - lr: 0.0090
Epoch 89/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0099 - accuracy: 0.1630

micro_F1_train: 0.20568050553781342

micro_F1_val: 0.1999048826886493

Epoch 00089: val_accuracy did not improve from 0.20101
1472/1472 [=====] - 11s 7ms/step - loss: 0.0098 - accuracy: 0.1631 - val_loss:
2.1241 - val_accuracy: 0.1999 - lr: 0.0090
Epoch 90/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0099 - accuracy: 0.1620

micro_F1_train: 0.20119589590269757

micro_F1_val: 0.19863665187064045

Epoch 00090: val_accuracy did not improve from 0.20101
1472/1472 [=====] - 11s 7ms/step - loss: 0.0098 - accuracy: 0.1619 - val_loss:
2.1263 - val_accuracy: 0.1986 - lr: 0.0090
Epoch 91/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0097 - accuracy: 0.1684

micro_F1_train: 0.20853434803288712

micro_F1_val: 0.20434369055168042

Epoch 00091: val_accuracy improved from 0.20101 to 0.20434, saving model to model1_save/weights-91-0.20
43.hdf5
1472/1472 [=====] - 11s 8ms/step - loss: 0.0097 - accuracy: 0.1690 - val_loss:
2.1148 - val_accuracy: 0.2043 - lr: 0.0090
Epoch 92/100
1466/1472 [=====>.] - ETA: 0s - loss: 0.0097 - accuracy: 0.1755

micro_F1_train: 0.20404973839777127

micro_F1_val: 0.20228281547241597

Epoch 00092: val_accuracy did not improve from 0.20434
1472/1472 [=====] - 11s 8ms/step - loss: 0.0097 - accuracy: 0.1756 - val_loss:
2.1144 - val_accuracy: 0.2023 - lr: 0.0090
Epoch 93/100
1467/1472 [=====>.] - ETA: 0s - loss: 0.0097 - accuracy: 0.1691

micro_F1_train: 0.2081266562478766

micro_F1_val: 0.20481927710843376

Epoch 00093: val_accuracy improved from 0.20434 to 0.20482, saving model to model1_save/weights-93-0.20
48.hdf5
1472/1472 [=====] - 12s 8ms/step - loss: 0.0097 - accuracy: 0.1692 - val_loss:
2.1120 - val_accuracy: 0.2048 - lr: 0.0090
Epoch 94/100
1462/1472 [=====>.] - ETA: 0s - loss: 0.0097 - accuracy: 0.1699

micro_F1_train: 0.21159203642046612

micro_F1_val: 0.21020925808497146

Epoch 00094: val_accuracy improved from 0.20482 to 0.21021, saving model to model1_save/weights-94-0.21
02.hdf5
1472/1472 [=====] - 11s 7ms/step - loss: 0.0097 - accuracy: 0.1699 - val_loss:
2.1073 - val_accuracy: 0.2102 - lr: 0.0090
Epoch 95/100
1472/1472 [=====] - ETA: 0s - loss: 0.0097 - accuracy: 0.1650

micro_F1_train: 0.20669973500033975

micro_F1_val: 0.20656309448319593

Epoch 00095: val_accuracy did not improve from 0.21021
1472/1472 [=====] - 11s 8ms/step - loss: 0.0097 - accuracy: 0.1650 - val_loss:
2.1060 - val_accuracy: 0.2066 - lr: 0.0090
Epoch 96/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0097 - accuracy: 0.1645

micro_F1_train: 0.20867004500455725
```

```
micro_F1_train: 0.20867024529455735
```

```
micro_F1_val: 0.20672162333544702
```

```
Epoch 00096: val_accuracy did not improve from 0.21021
```

```
1472/1472 [=====] - 10s 6ms/step - loss: 0.0098 - accuracy: 0.1648 - val_loss: 2.1023 - val_accuracy: 0.2067 - lr: 0.0090
```

```
Epoch 97/100
```

```
1471/1472 [=====>.] - ETA: 0s - loss: 0.0097 - accuracy: 0.1664
```

```
micro_F1_train: 0.21668818373309776
```

```
micro_F1_val: 0.21496512365250475
```

```
Epoch 00097: val_accuracy improved from 0.21021 to 0.21497, saving model to model1_save/weights-97-0.2150.hdf5
```

```
1472/1472 [=====] - 10s 7ms/step - loss: 0.0097 - accuracy: 0.1663 - val_loss: 2.0933 - val_accuracy: 0.2150 - lr: 0.0090
```

```
Epoch 98/100
```

```
1465/1472 [=====>.] - ETA: 0s - loss: 0.0096 - accuracy: 0.1694
```

```
micro_F1_train: 0.2098933206495889
```

```
micro_F1_val: 0.20703868103994927
```

```
Epoch 00098: val_accuracy did not improve from 0.21497
```

```
1472/1472 [=====] - 10s 7ms/step - loss: 0.0096 - accuracy: 0.1695 - val_loss: 2.0963 - val_accuracy: 0.2070 - lr: 0.0090
```

```
Epoch 99/100
```

```
1469/1472 [=====>.] - ETA: 0s - loss: 0.0097 - accuracy: 0.1686
```

```
micro_F1_train: 0.21193177957464157
```

```
micro_F1_val: 0.2098922003804692
```

```
Epoch 00099: val_accuracy did not improve from 0.21497
```

```
1472/1472 [=====] - 10s 7ms/step - loss: 0.0097 - accuracy: 0.1687 - val_loss: 2.1020 - val_accuracy: 0.2099 - lr: 0.0090
```

```
Epoch 100/100
```

```
1462/1472 [=====>.] - ETA: 0s - loss: 0.0095 - accuracy: 0.1733
```

```
micro_F1_train: 0.21458177617721003
```

```
micro_F1_val: 0.215282181357007
```

```
Epoch 00100: val_accuracy improved from 0.21497 to 0.21528, saving model to model1_save/weights-100-0.2153.hdf5
```

```
1472/1472 [=====] - 10s 7ms/step - loss: 0.0095 - accuracy: 0.1733 - val_loss: 2.0998 - val_accuracy: 0.2153 - lr: 0.0081
```

```
Time Taken for training (sec): 1120.963470697403
```

```
In [ ]:
```

```
# http://localhost:6006/  
%load_ext tensorboard  
%tensorboard --logdir logs --host localhost
```

The tensorboard extension is already loaded. To reload it, use:

```
%reload_ext tensorboard
```

Reusing TensorBoard on port 6006 (pid 2268), started 0:22:00 ago. (Use '!kill 2268' to kill it.)

## Model 1 Predictions

```
In [ ]:
```

```
NeuNet_1.load_weights("/content/ml_weights-97-0.2150.hdf5")
```

```
In [ ]:
```

```
px_data_std = pd.DataFrame(Scaler.transform(px_data))
px_data_std = px_data_std.drop(feature_correlated,axis=1)
```

```
In [ ]:
```

```
y_pred_1 = NeuNet_1.predict(px_data_std)
```

```
In [ ]:
```

```
np.argmax(y_pred_1,axis=1)
```

```
Out[ ]:
```

```
array([10, 11,  3, ..., 14, 14, 14])
```

```
In [ ]:
```

```
y_pred_pd_1 = pd.DataFrame(np.argmax(y_pred_1,axis=1),index=px_data.index)
# y_pred_pd[0] = y_pred_pd[0]+1
```

```
In [ ]:
```

```
# y_pred_pd_1
```

```
In [ ]:
```

```
y_pred_pd_1.value_counts()
```

```
Out[ ]:
```

```
14    4694
11    3394
15    2198
 3    1900
13    1568
10    1436
 5    1179
16     865
12     752
 4     713
 1     694
 8     464
 2     383
 9     347
 0     205
 7     144
 6      89
dtype: int64
```

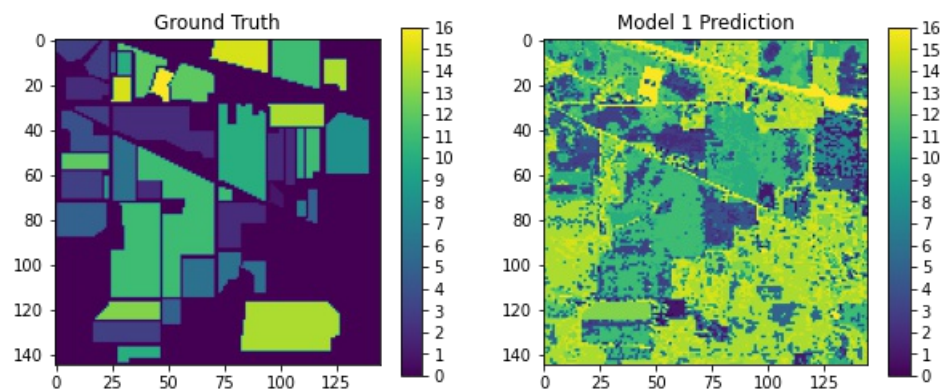
```
In [ ]:
```

```
# px_data.index
```

```
In [ ]:
```

```
figr,axis = plt.subplots(1,2,figsize=(10,10))
im0 = axis[0].imshow(mat_gt['indian_pines_gt'])#,cmap='jet')
axis[0].set_title('Ground Truth')
plt.colorbar(im0,ax=axis[0],shrink=0.4,aspect=16, ticks=range(0,17,1))

im1 = axis[1].imshow(y_pred_pd_1.to_numpy().reshape((145,145)))#,cmap='jet')
axis[1].set_title('Model 1 Prediction')
plt.colorbar(im1,ax=axis[1],shrink=0.4,aspect=16, ticks=range(0,17,1))
plt.savefig('NeuNet_1_e100.png')
plt.show()
```



## Observation

### Confusion Matrix

In [ ]:

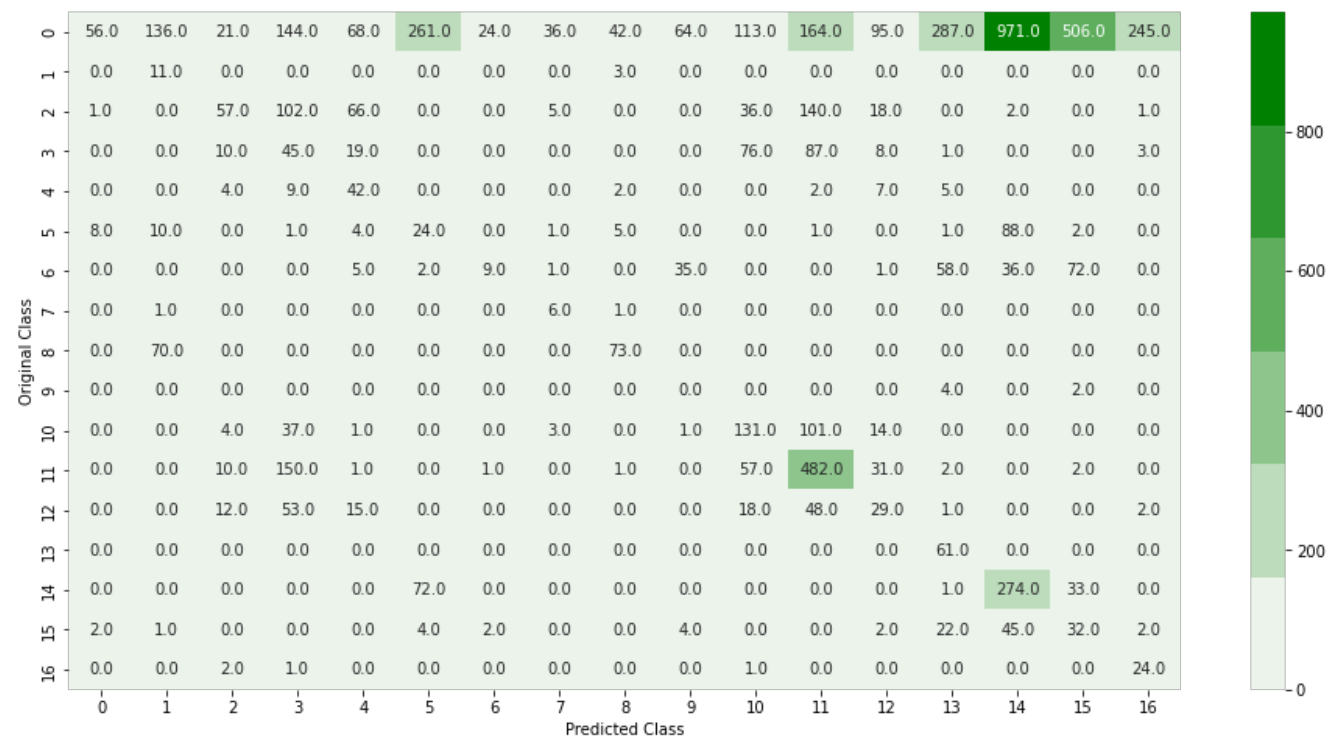
```
y_pred_1_test = NeuNet_1.predict(X_test_std)
y_pred_pd_1_test = pd.DataFrame(np.argmax(y_pred_1_test,axis=1),index=y_test.index)
```

In [ ]:

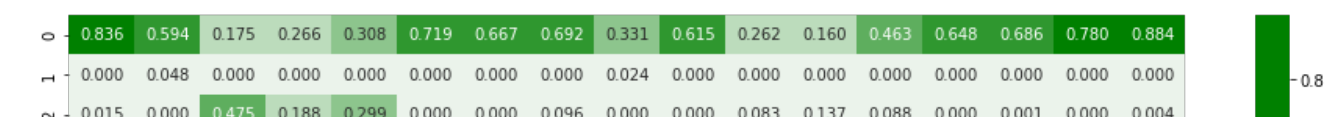
```
# plot_confusion_matrix(px_class,y_pred_pd_1)
plot_confusion_matrix(y_test,y_pred_pd_1_test)
```

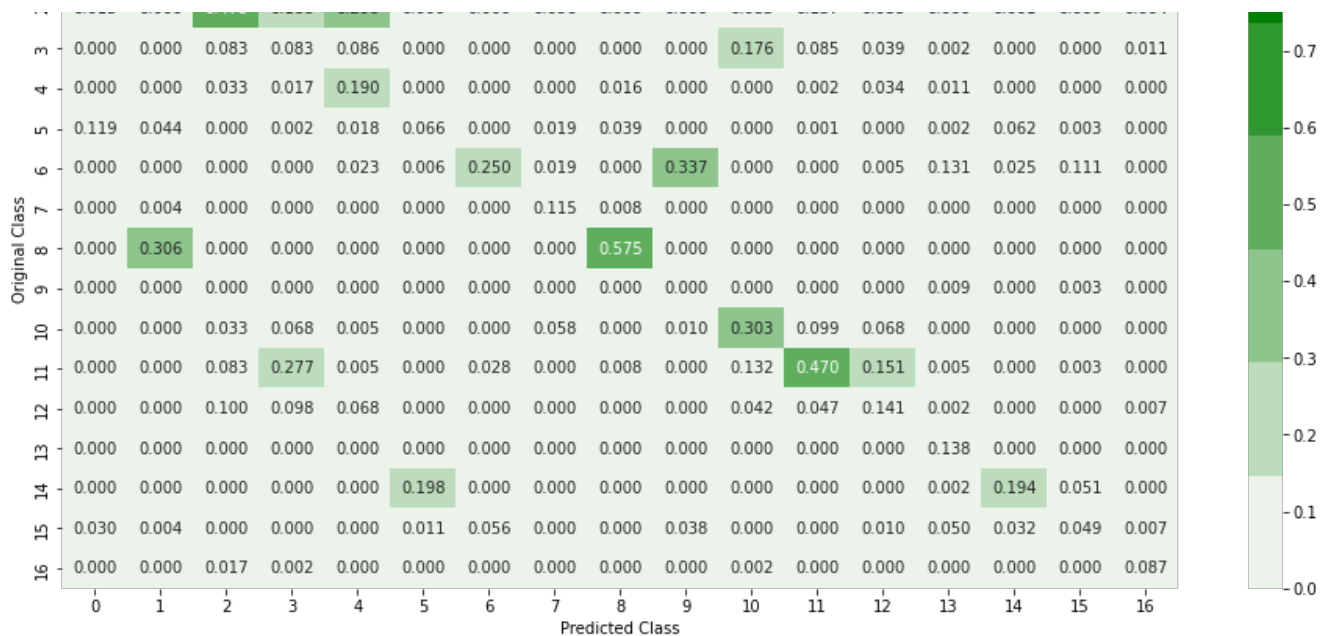
Number of misclassified points 78.50348763474952

----- Confusion matrix -----



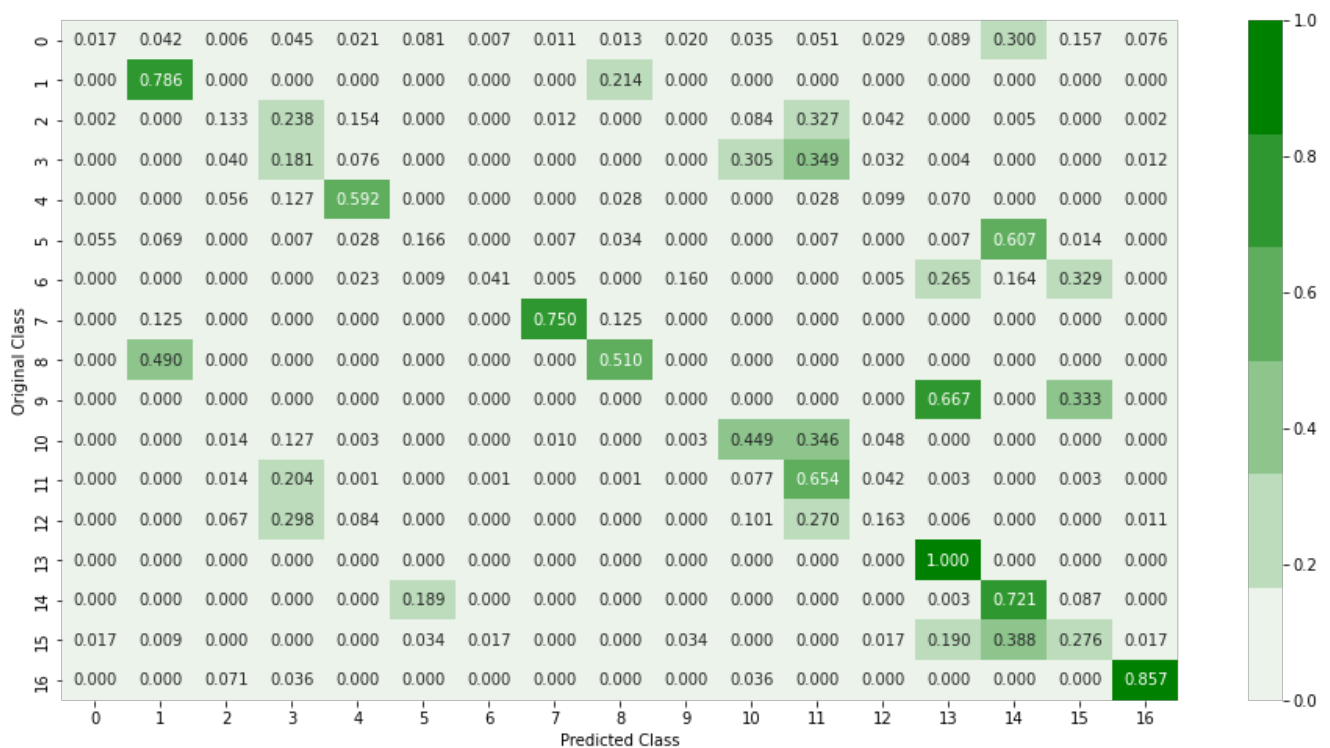
----- Precision matrix -----





Sum of columns in precision matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

Recall matrix



Sum of rows in recall matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

Model prediction doesnt match ground truth very well. From Recall matrix we can see that only 2 of the classes have been predicted high recall. Precisions are all low for all classes except for class 'zero'.

## 2- Four layered network with relu activation and ADAM optimizer

### Model 2 Definition

In [ ]:

```
# del NeuNet_2
```

In [ ]:

```
In = Input(shape=(95,))
L1 = Dense(128,activation='relu',
          kernel_initializer=initializers.he_uniform())(In) #https://keras.io/api/layers/initializers/

L1_Drp = Dropout(0.4)(L1)

L2 = Dense(64,activation='relu',
          kernel_initializer=initializers.he_uniform())(L1_Drp)

L2_Drp = Dropout(0.3)(L2)

L3 = Dense(32,activation='relu',
          kernel_initializer=initializers.he_uniform())(L2_Drp)

L4 = Dense(20,activation='relu',
          kernel_initializer=initializers.he_uniform())(L3)

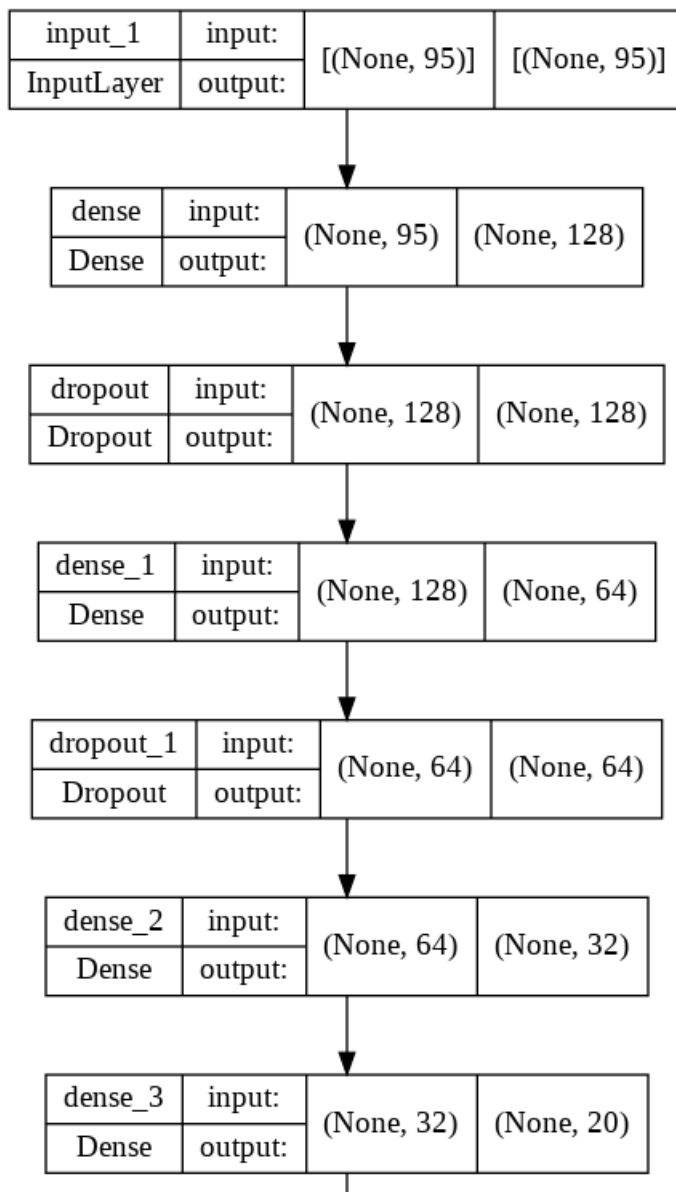
Out = Dense(17,activation='softmax',
          kernel_initializer=initializers.GlorotUniform())(L4)


NeuNet_2 = Model(inputs=In, outputs = Out)

# cce = categorical_crossentropy()
NeuNet_2.compile(loss='categorical_crossentropy', optimizer='Adam', metrics=['accuracy'])

plot_model(NeuNet_2,show_layer_names=True, show_shapes=True)#,to_file='NeuNet_2.png')
```

Out[ ]:





dense_4	input:	(None, 20)	(None, 17)
Dense	output:		

In [ ]:

```
NeuNet_2.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 95)]	0
dense (Dense)	(None, 128)	12288
dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 64)	8256
dropout_1 (Dropout)	(None, 64)	0
dense_2 (Dense)	(None, 32)	2080
dense_3 (Dense)	(None, 20)	660
dense_4 (Dense)	(None, 17)	357

```

=====
Total params: 23,641
Trainable params: 23,641
Non-trainable params: 0
=====

```

## Model 2 Training

In [ ]:

```

logdir = os.path.join("logs", datetime.now().strftime("%Y%m%d-%H%M%S"))
print(logdir)
file_writer = tf.summary.create_file_writer(logdir + "/metrics")
# tensorboard = TensorBoard(log_dir=logdir)
tensorboard = TensorBoard(log_dir=logdir, histogram_freq=1, write_graph=True, write_grads=True)

metric_calc_2 = metric_calc()

# Saving model at every epoch if validation accuracy is improved from previous epoch
filepath_m2="model2_save/weights-{epoch:02d}-{val_accuracy:.4f}.hdf5"
checkpoint_m2 = ModelCheckpoint(filepath=filepath_m2, monitor='val_accuracy', verbose=1,
                                save_best_only=True,
                                mode='auto')
earlystop_m2 = EarlyStopping(monitor='val_accuracy', min_delta=0.01, patience=20, verbose=1)
reduce_lr_m2 = ReduceLROnPlateau(monitor='val_loss', factor=0.9, patience=2, min_lr=0.001)

callback_list_2 = [metric_calc_2,
                   checkpoint_m2,
                   earlystop_m2,
                   reduce_lr_m2,
                   tensorboard]

# fit network
verbose_2, epochs_2, batch_size_2 = 1, 100, 10
start = time.time()
history_2 = NeuNet_2.fit(X_train_std, y_ctg_train,
                        class_weight = class_wts,
                        validation_data=(X_test_std,y_ctg_test),
                        epochs=epochs_2,
                        batch_size=batch_size_2,
                        verbose=verbose_2,
                        callbacks=callback_list_2)

```

```

# Passing the F1 score data of each epoch to tensor board
# https://stackoverflow.com/questions/58102016/how-visualize-in-tensorboard-a-metric-callback
for i in range(len(metric_calc_2.metrics['micro_F1_train'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_train', metric_calc_2.metrics['micro_F1_train'][i])
file_writer.flush()

for i in range(len(metric_calc_2.metrics['micro_F1_val'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_val', metric_calc_2.metrics['micro_F1_val'][i])
file_writer.flush()

stop = time.time()
print('Time Taken for training (sec): ', stop-start)

```

logs/20220131-122955

WARNING:tensorflow: `write\_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.

Epoch 1/100

2/1472 [.....] - ETA: 5:57 - loss: 0.0230 - accuracy: 0.1500 WARNING:tensorflow:Callback method `on\_train\_batch\_begin` is slow compared to the batch time (batch time: 0.0045s vs `on\_train\_batch\_begin` time: 0.0233s). Check your callbacks.

WARNING:tensorflow:Callback method `on\_train\_batch\_end` is slow compared to the batch time (batch time: 0.0045s vs `on\_train\_batch\_end` time: 0.0195s). Check your callbacks.

1463/1472 [=====>.] - ETA: 0s - loss: 0.0124 - accuracy: 0.1524

micro\_F1\_train: 0.23754841339947

micro\_F1\_val: 0.2382688649334179

Epoch 00001: val\_accuracy improved from -inf to 0.23827, saving model to model2\_save/weights-01-0.2383.hdf5

1472/1472 [=====] - 12s 8ms/step - loss: 0.0124 - accuracy: 0.1527 - val\_loss: 2.0437 - val\_accuracy: 0.2383 - lr: 0.0010

Epoch 2/100

1472/1472 [=====] - ETA: 0s - loss: 0.0092 - accuracy: 0.1820

micro\_F1\_train: 0.2711829856628389

micro\_F1\_val: 0.27060875079264424

Epoch 00002: val\_accuracy improved from 0.23827 to 0.27061, saving model to model2\_save/weights-02-0.2706.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0092 - accuracy: 0.1820 - val\_loss: 1.8567 - val\_accuracy: 0.2706 - lr: 0.0010

Epoch 3/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0081 - accuracy: 0.2037

micro\_F1\_train: 0.24454712237548412

micro\_F1\_val: 0.24381737476220672

Epoch 00003: val\_accuracy did not improve from 0.27061

1472/1472 [=====] - 10s 7ms/step - loss: 0.0081 - accuracy: 0.2038 - val\_loss: 1.8605 - val\_accuracy: 0.2438 - lr: 0.0010

Epoch 4/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0074 - accuracy: 0.2148

micro\_F1\_train: 0.2746483658354284

micro\_F1\_val: 0.27869372225745087

Epoch 00004: val\_accuracy improved from 0.27061 to 0.27869, saving model to model2\_save/weights-04-0.2787.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0074 - accuracy: 0.2146 - val\_loss: 1.8081 - val\_accuracy: 0.2787 - lr: 0.0010

Epoch 5/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0074 - accuracy: 0.2215

micro\_F1\_train: 0.2765509274988109

micro\_F1\_val: 0.2710843373493976

Epoch 00005: val\_accuracy did not improve from 0.27869

1472/1472 [=====] - 10s 7ms/step - loss: 0.0074 - accuracy: 0.2218 - val\_loss: 1.7603 - val\_accuracy: 0.2711 - lr: 0.0010



Epoch 6/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0067 - accuracy: 0.2359

micro\_F1\_train: 0.28348168784398997

micro\_F1\_val: 0.28233988585922637

Epoch 00006: val\_accuracy improved from 0.27869 to 0.28234, saving model to model2\_save/weights-06-0.2823.hdf5  
1472/1472 [=====] - 14s 10ms/step - loss: 0.0067 - accuracy: 0.2358 - val\_loss: 1.7780 - val\_accuracy: 0.2823 - lr: 0.0010

Epoch 7/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0066 - accuracy: 0.2540

micro\_F1\_train: 0.33750084935788544

micro\_F1\_val: 0.3333861762840837

Epoch 00007: val\_accuracy improved from 0.28234 to 0.33339, saving model to model2\_save/weights-07-0.3334.hdf5  
1472/1472 [=====] - 16s 11ms/step - loss: 0.0066 - accuracy: 0.2539 - val\_loss: 1.7090 - val\_accuracy: 0.3334 - lr: 0.0010

Epoch 8/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0062 - accuracy: 0.2718

micro\_F1\_train: 0.32180471563498

micro\_F1\_val: 0.3181674064679772

Epoch 00008: val\_accuracy did not improve from 0.33339  
1472/1472 [=====] - 15s 10ms/step - loss: 0.0062 - accuracy: 0.2715 - val\_loss: 1.6405 - val\_accuracy: 0.3182 - lr: 0.0010

Epoch 9/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0061 - accuracy: 0.2695

micro\_F1\_train: 0.3173880546306992

micro\_F1\_val: 0.31436271401395055

Epoch 00009: val\_accuracy did not improve from 0.33339  
1472/1472 [=====] - 13s 9ms/step - loss: 0.0061 - accuracy: 0.2696 - val\_loss: 1.6644 - val\_accuracy: 0.3144 - lr: 0.0010

Epoch 10/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0059 - accuracy: 0.2889

micro\_F1\_train: 0.3394034110212679

micro\_F1\_val: 0.3316423589093215

Epoch 00010: val\_accuracy did not improve from 0.33339  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0059 - accuracy: 0.2890 - val\_loss: 1.6379 - val\_accuracy: 0.3316 - lr: 0.0010

Epoch 11/100  
1465/1472 [=====>.] - ETA: 0s - loss: 0.0059 - accuracy: 0.2905

micro\_F1\_train: 0.3082829380987973

micro\_F1\_val: 0.30469245402663286

Epoch 00011: val\_accuracy did not improve from 0.33339  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0059 - accuracy: 0.2907 - val\_loss: 1.7137 - val\_accuracy: 0.3047 - lr: 0.0010

Epoch 12/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0057 - accuracy: 0.3025

micro\_F1\_train: 0.3263572739009309

micro\_F1\_val: 0.32228915662650603

Epoch 00012: val\_accuracy did not improve from 0.33339  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0057 - accuracy: 0.3023 - val\_loss: 1.5551 - val\_accuracy: 0.3223 - lr: 0.0010

Epoch 13/100  
1465/1472 [=====>.] - ETA: 0s - loss: 0.0056 - accuracy: 0.2993

micro\_F1\_train: 0.33743290072705034

micro\_F1\_val: 0.3298985415345593

Epoch 00013: val\_accuracy did not improve from 0.33339

1472/1472 [=====] - 17s 11ms/step - loss: 0.0056 - accuracy: 0.2995 - val\_loss : 1.5404 - val\_accuracy: 0.3299 - lr: 0.0010

Epoch 14/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0054 - accuracy: 0.3140

micro\_F1\_train: 0.3401508459604539

micro\_F1\_val: 0.3333861762840837

Epoch 00014: val\_accuracy did not improve from 0.33339

1472/1472 [=====] - 18s 12ms/step - loss: 0.0053 - accuracy: 0.3137 - val\_loss : 1.5316 - val\_accuracy: 0.3334 - lr: 0.0010

Epoch 15/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0056 - accuracy: 0.3155

micro\_F1\_train: 0.3558469796833594

micro\_F1\_val: 0.34923906150919465

Epoch 00015: val\_accuracy improved from 0.33339 to 0.34924, saving model to model2\_save/weights-15-0.3492.hdf5

1472/1472 [=====] - 14s 10ms/step - loss: 0.0056 - accuracy: 0.3155 - val\_loss : 1.5412 - val\_accuracy: 0.3492 - lr: 0.0010

Epoch 16/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0057 - accuracy: 0.3058

micro\_F1\_train: 0.3654277366311069

micro\_F1\_val: 0.3563728598604946

Epoch 00016: val\_accuracy improved from 0.34924 to 0.35637, saving model to model2\_save/weights-16-0.3564.hdf5

1472/1472 [=====] - 15s 11ms/step - loss: 0.0057 - accuracy: 0.3059 - val\_loss : 1.4910 - val\_accuracy: 0.3564 - lr: 0.0010

Epoch 17/100

1472/1472 [=====] - ETA: 0s - loss: 0.0051 - accuracy: 0.3223

micro\_F1\_train: 0.3849969423116124

micro\_F1\_val: 0.3780913126188966

Epoch 00017: val\_accuracy improved from 0.35637 to 0.37809, saving model to model2\_save/weights-17-0.3781.hdf5

1472/1472 [=====] - 15s 10ms/step - loss: 0.0051 - accuracy: 0.3223 - val\_loss : 1.4848 - val\_accuracy: 0.3781 - lr: 0.0010

Epoch 18/100

1472/1472 [=====] - ETA: 0s - loss: 0.0056 - accuracy: 0.3255

micro\_F1\_train: 0.35618672283753483

micro\_F1\_val: 0.3495561192136969

Epoch 00018: val\_accuracy did not improve from 0.37809

1472/1472 [=====] - 14s 10ms/step - loss: 0.0056 - accuracy: 0.3255 - val\_loss : 1.4967 - val\_accuracy: 0.3496 - lr: 0.0010

Epoch 19/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0053 - accuracy: 0.3197

micro\_F1\_train: 0.4064007610246654

micro\_F1\_val: 0.4058338617628409

Epoch 00019: val\_accuracy improved from 0.37809 to 0.40583, saving model to model2\_save/weights-19-0.4058.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0053 - accuracy: 0.3196 - val\_loss : 1.4312 - val\_accuracy: 0.4058 - lr: 0.0010

Epoch 20/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0055 - accuracy: 0.3392

micro\_F1\_train: 0.3650200448460964

micro\_F1\_val: 0.36699429296131897

Epoch 00020: val\_accuracy did not improve from 0.40583

```
1472/1472 [=====] - 10s 7ms/step - loss: 0.0055 - accuracy: 0.3391 - val_loss: 1.5724 - val_accuracy: 0.3670 - lr: 0.0010
Epoch 21/100
1469/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3473

micro_F1_train: 0.39627641503023714

micro_F1_val: 0.38950538998097656

Epoch 00021: val_accuracy did not improve from 0.40583
1472/1472 [=====] - 9s 6ms/step - loss: 0.0048 - accuracy: 0.3470 - val_loss: 1.4315 - val_accuracy: 0.3895 - lr: 0.0010
Epoch 22/100
1461/1472 [=====>.] - ETA: 0s - loss: 0.0046 - accuracy: 0.3434

micro_F1_train: 0.35605082557586465

micro_F1_val: 0.34369055168040585

Epoch 00022: val_accuracy did not improve from 0.40583
1472/1472 [=====] - 9s 6ms/step - loss: 0.0046 - accuracy: 0.3434 - val_loss: 1.5183 - val_accuracy: 0.3437 - lr: 0.0010
Epoch 23/100
1468/1472 [=====>.] - ETA: 0s - loss: 0.0053 - accuracy: 0.3256

micro_F1_train: 0.36787388734117005

micro_F1_val: 0.3611287254280279

Epoch 00023: val_accuracy did not improve from 0.40583
1472/1472 [=====] - 12s 8ms/step - loss: 0.0053 - accuracy: 0.3255 - val_loss: 1.4602 - val_accuracy: 0.3611 - lr: 0.0010
Epoch 24/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0055 - accuracy: 0.3218

micro_F1_train: 0.3920635999184616

micro_F1_val: 0.3853836398224477

Epoch 00024: val_accuracy did not improve from 0.40583
1472/1472 [=====] - 9s 6ms/step - loss: 0.0055 - accuracy: 0.3217 - val_loss: 1.4365 - val_accuracy: 0.3854 - lr: 0.0010
Epoch 25/100
1469/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3445

micro_F1_train: 0.4119725487531426

micro_F1_val: 0.4078947368421052

Epoch 00025: val_accuracy improved from 0.40583 to 0.40789, saving model to model2_save/weights-25-0.4079.hdf5
1472/1472 [=====] - 11s 7ms/step - loss: 0.0048 - accuracy: 0.3446 - val_loss: 1.3854 - val_accuracy: 0.4079 - lr: 0.0010
Epoch 26/100
1467/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3503

micro_F1_train: 0.37820207922810356

micro_F1_val: 0.3703233988585923

Epoch 00026: val_accuracy did not improve from 0.40789
1472/1472 [=====] - 11s 7ms/step - loss: 0.0047 - accuracy: 0.3504 - val_loss: 1.5111 - val_accuracy: 0.3703 - lr: 0.0010
Epoch 27/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3458

micro_F1_train: 0.35842902765509277

micro_F1_val: 0.35003170577045023

Epoch 00027: val_accuracy did not improve from 0.40789
1472/1472 [=====] - 10s 7ms/step - loss: 0.0048 - accuracy: 0.3458 - val_loss: 1.4912 - val_accuracy: 0.3500 - lr: 0.0010
Epoch 28/100
1472/1472 [=====] - ETA: 0s - loss: 0.0048 - accuracy: 0.3563

micro_F1_train: 0.3736495209621526
```

micro\_F1\_val: 0.3636651870640456

Epoch 00028: val\_accuracy did not improve from 0.40789

1472/1472 [=====] - 10s 7ms/step - loss: 0.0048 - accuracy: 0.3563 - val\_loss: 1.5120 - val\_accuracy: 0.3637 - lr: 0.0010

Epoch 29/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3538

micro\_F1\_train: 0.43364816198953593

micro\_F1\_val: 0.4292961318960051

Epoch 00029: val\_accuracy improved from 0.40789 to 0.42930, saving model to model2\_save/weights-29-0.4293.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0048 - accuracy: 0.3544 - val\_loss: 1.3837 - val\_accuracy: 0.4293 - lr: 0.0010

Epoch 30/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0046 - accuracy: 0.3667

micro\_F1\_train: 0.42155330570089006

micro\_F1\_val: 0.4085288522511097

Epoch 00030: val\_accuracy did not improve from 0.42930

1472/1472 [=====] - 9s 6ms/step - loss: 0.0046 - accuracy: 0.3668 - val\_loss: 1.3478 - val\_accuracy: 0.4085 - lr: 0.0010

Epoch 31/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3683

micro\_F1\_train: 0.35972005164095944

micro\_F1\_val: 0.3557387444514902

Epoch 00031: val\_accuracy did not improve from 0.42930

1472/1472 [=====] - 10s 7ms/step - loss: 0.0048 - accuracy: 0.3682 - val\_loss: 1.5477 - val\_accuracy: 0.3557 - lr: 0.0010

Epoch 32/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3582

micro\_F1\_train: 0.37759054155058774

micro\_F1\_val: 0.36731135066582116

Epoch 00032: val\_accuracy did not improve from 0.42930

1472/1472 [=====] - 9s 6ms/step - loss: 0.0048 - accuracy: 0.3582 - val\_loss: 1.5076 - val\_accuracy: 0.3673 - lr: 0.0010

Epoch 33/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0053 - accuracy: 0.3516

micro\_F1\_train: 0.40130461371203363

micro\_F1\_val: 0.39267596702599866

Epoch 00033: val\_accuracy did not improve from 0.42930

1472/1472 [=====] - 11s 7ms/step - loss: 0.0054 - accuracy: 0.3517 - val\_loss: 1.4726 - val\_accuracy: 0.3927 - lr: 0.0010

Epoch 34/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3583

micro\_F1\_train: 0.4133994700006795

micro\_F1\_val: 0.40932149651236527

Epoch 00034: val\_accuracy did not improve from 0.42930

1472/1472 [=====] - 10s 7ms/step - loss: 0.0047 - accuracy: 0.3585 - val\_loss: 1.3839 - val\_accuracy: 0.4093 - lr: 0.0010

Epoch 35/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0046 - accuracy: 0.3587

micro\_F1\_train: 0.40762383637969696

micro\_F1\_val: 0.40091946734305645

Epoch 00035: val\_accuracy did not improve from 0.42930

1472/1472 [=====] - 10s 7ms/step - loss: 0.0046 - accuracy: 0.3588 - val\_loss: 1.4080 - val accuracy: 0.4009 - lr: 0.0010

Epoch 36/100  
1472/1472 [=====] - ETA: 0s - loss: 0.0047 - accuracy: 0.3797

micro\_F1\_train: 0.4494122443432765

micro\_F1\_val: 0.44705136334812934

Epoch 00036: val\_accuracy improved from 0.42930 to 0.44705, saving model to model2\_save/weights-36-0.4471.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0047 - accuracy: 0.3797 - val\_loss: 1.3930 - val\_accuracy: 0.4471 - lr: 0.0010

Epoch 37/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3825

micro\_F1\_train: 0.4080315281647075

micro\_F1\_val: 0.40107799619530754

Epoch 00037: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0047 - accuracy: 0.3824 - val\_loss: 1.4483 - val\_accuracy: 0.4011 - lr: 0.0010

Epoch 38/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3719

micro\_F1\_train: 0.3994700006794863

micro\_F1\_val: 0.39505389980976535

Epoch 00038: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0045 - accuracy: 0.3717 - val\_loss: 1.5196 - val\_accuracy: 0.3951 - lr: 0.0010

Epoch 39/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3757

micro\_F1\_train: 0.4134674186315146

micro\_F1\_val: 0.40456563094483194

Epoch 00039: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0047 - accuracy: 0.3754 - val\_loss: 1.3823 - val\_accuracy: 0.4046 - lr: 0.0010

Epoch 40/100  
1465/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3788

micro\_F1\_train: 0.4231161242100972

micro\_F1\_val: 0.41597970830691183

Epoch 00040: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0047 - accuracy: 0.3789 - val\_loss: 1.3911 - val\_accuracy: 0.4160 - lr: 0.0010

Epoch 41/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3705

micro\_F1\_train: 0.37432900727050344

micro\_F1\_val: 0.3639822447685479

Epoch 00041: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0045 - accuracy: 0.3710 - val\_loss: 1.4946 - val\_accuracy: 0.3640 - lr: 0.0010

Epoch 42/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0049 - accuracy: 0.3554

micro\_F1\_train: 0.41122511381395666

micro\_F1\_val: 0.4053582752060875

Epoch 00042: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0049 - accuracy: 0.3559 - val\_loss: 1.4960 - val\_accuracy: 0.4054 - lr: 0.0010

Epoch 43/100  
1465/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3830

micro\_F1\_train: 0.4212815111775498

micro\_F1\_val: 0.4145529486366519

- -

Epoch 00043: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0048 - accuracy: 0.3826 - val\_loss: 1.3951 - val\_accuracy: 0.4146 - lr: 0.0010  
Epoch 44/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3866

micro\_F1\_train: 0.4064687096555004

micro\_F1\_val: 0.38966391883322765

Epoch 00044: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0045 - accuracy: 0.3866 - val\_loss: 1.4096 - val\_accuracy: 0.3897 - lr: 0.0010  
Epoch 45/100  
1472/1472 [=====] - ETA: 0s - loss: 0.0046 - accuracy: 0.3778

micro\_F1\_train: 0.41183665149147247

micro\_F1\_val: 0.40456563094483194

Epoch 00045: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0046 - accuracy: 0.3778 - val\_loss: 1.3966 - val\_accuracy: 0.4046 - lr: 0.0010  
Epoch 46/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.3699

micro\_F1\_train: 0.3883943738533668

micro\_F1\_val: 0.37888395688015225

Epoch 00046: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0044 - accuracy: 0.3696 - val\_loss: 1.4585 - val\_accuracy: 0.3789 - lr: 0.0010  
Epoch 47/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0050 - accuracy: 0.3661

micro\_F1\_train: 0.3610790242576612

micro\_F1\_val: 0.3530437539632213

Epoch 00047: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0050 - accuracy: 0.3660 - val\_loss: 1.7086 - val\_accuracy: 0.3530 - lr: 0.0010  
Epoch 48/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0050 - accuracy: 0.3791

micro\_F1\_train: 0.42141740843921993

micro\_F1\_val: 0.4085288522511097

Epoch 00048: val\_accuracy did not improve from 0.44705  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0050 - accuracy: 0.3793 - val\_loss: 1.3880 - val\_accuracy: 0.4085 - lr: 0.0010  
Epoch 49/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3746

micro\_F1\_train: 0.4534212135625467

micro\_F1\_val: 0.4475269499048827

Epoch 00049: val\_accuracy improved from 0.44705 to 0.44753, saving model to model2\_save/weights-49-0.4475.hdf5  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0047 - accuracy: 0.3746 - val\_loss: 1.3921 - val\_accuracy: 0.4475 - lr: 0.0010  
Epoch 50/100  
1471/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3761

micro\_F1\_train: 0.4205340762383638

micro\_F1\_val: 0.41407736207989854

Epoch 00050: val\_accuracy did not improve from 0.44753  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0047 - accuracy: 0.3760 - val\_loss: 1.4187 - val\_accuracy: 0.4141 - lr: 0.0010  
Epoch 51/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0050 - accuracy: 0.3822

micro\_F1\_train: 0.4580417204593327

micro\_F1\_val: 0.44467343056436265

Epoch 00051: val\_accuracy did not improve from 0.44753

1472/1472 [=====] - 10s 7ms/step - loss: 0.0050 - accuracy: 0.3816 - val\_loss: 1.3375 - val\_accuracy: 0.4447 - lr: 0.0010

Epoch 52/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.3806

micro\_F1\_train: 0.40966229530474957

micro\_F1\_val: 0.4018706404565631

Epoch 00052: val\_accuracy did not improve from 0.44753

1472/1472 [=====] - 10s 7ms/step - loss: 0.0044 - accuracy: 0.3809 - val\_loss: 1.3657 - val\_accuracy: 0.4019 - lr: 0.0010

Epoch 53/100

1472/1472 [=====] - ETA: 0s - loss: 0.0043 - accuracy: 0.3875

micro\_F1\_train: 0.4181558741591357

micro\_F1\_val: 0.41169942929613196

Epoch 00053: val\_accuracy did not improve from 0.44753

1472/1472 [=====] - 11s 7ms/step - loss: 0.0043 - accuracy: 0.3875 - val\_loss: 1.4696 - val\_accuracy: 0.4117 - lr: 0.0010

Epoch 54/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0043 - accuracy: 0.3818

micro\_F1\_train: 0.464360943126996

micro\_F1\_val: 0.46163601775523144

Epoch 00054: val\_accuracy improved from 0.44753 to 0.46164, saving model to model2\_save/weights-54-0.4616.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0043 - accuracy: 0.3820 - val\_loss: 1.3096 - val\_accuracy: 0.4616 - lr: 0.0010

Epoch 55/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0049 - accuracy: 0.3851

micro\_F1\_train: 0.39913025752531084

micro\_F1\_val: 0.38966391883322765

Epoch 00055: val\_accuracy did not improve from 0.46164

1472/1472 [=====] - 10s 7ms/step - loss: 0.0049 - accuracy: 0.3854 - val\_loss: 1.4138 - val\_accuracy: 0.3897 - lr: 0.0010

Epoch 56/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.3937

micro\_F1\_train: 0.411496908337297

micro\_F1\_val: 0.4021876981610653

Epoch 00056: val\_accuracy did not improve from 0.46164

1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.3934 - val\_loss: 1.3916 - val\_accuracy: 0.4022 - lr: 0.0010

Epoch 57/100

1472/1472 [=====] - ETA: 0s - loss: 0.0051 - accuracy: 0.3657

micro\_F1\_train: 0.4449955833389957

micro\_F1\_val: 0.43547875713379836

Epoch 00057: val\_accuracy did not improve from 0.46164

1472/1472 [=====] - 11s 7ms/step - loss: 0.0051 - accuracy: 0.3657 - val\_loss: 1.3446 - val\_accuracy: 0.4355 - lr: 0.0010

Epoch 58/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.3694

micro\_F1\_train: 0.44261738125976763

micro\_F1\_val: 0.43785668991756493

Epoch 00058: val accuracy did not improve from 0.46164

```
Epoch 59/100: val_accuracy did not improve from 0.46164
1472/1472 [=====] - 10s 7ms/step - loss: 0.0044 - accuracy: 0.3698 - val_loss:
1.3529 - val_accuracy: 0.4379 - lr: 0.0010
Epoch 59/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0046 - accuracy: 0.3878

micro_F1_train: 0.42196099748590066

micro_F1_val: 0.4151870640456563

Epoch 00059: val_accuracy did not improve from 0.46164
1472/1472 [=====] - 11s 7ms/step - loss: 0.0046 - accuracy: 0.3878 - val_loss:
1.3434 - val_accuracy: 0.4152 - lr: 0.0010
Epoch 60/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.3874

micro_F1_train: 0.4337161106203711

micro_F1_val: 0.43040583386176284

Epoch 00060: val_accuracy did not improve from 0.46164
1472/1472 [=====] - 11s 7ms/step - loss: 0.0044 - accuracy: 0.3873 - val_loss:
1.3440 - val_accuracy: 0.4304 - lr: 0.0010
Epoch 61/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.3843

micro_F1_train: 0.4333763674661955

micro_F1_val: 0.4246987951807229

Epoch 00061: val_accuracy did not improve from 0.46164
1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.3845 - val_loss:
1.3543 - val_accuracy: 0.4247 - lr: 0.0010
Epoch 62/100
1468/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.3888

micro_F1_train: 0.4271250934293674

micro_F1_val: 0.4153455928979074

Epoch 00062: val_accuracy did not improve from 0.46164
1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.3887 - val_loss:
1.3517 - val_accuracy: 0.4153 - lr: 0.0010
Epoch 63/100
1466/1472 [=====>.] - ETA: 0s - loss: 0.0051 - accuracy: 0.3772

micro_F1_train: 0.43806482299381666

micro_F1_val: 0.42945466074825617

Epoch 00063: val_accuracy did not improve from 0.46164
1472/1472 [=====] - 11s 8ms/step - loss: 0.0051 - accuracy: 0.3777 - val_loss:
1.3482 - val_accuracy: 0.4295 - lr: 0.0010
Epoch 64/100
1466/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.3926

micro_F1_train: 0.44920839845077126

micro_F1_val: 0.4456246036778694

Epoch 00064: val_accuracy did not improve from 0.46164
1472/1472 [=====] - 9s 6ms/step - loss: 0.0042 - accuracy: 0.3926 - val_loss:
1.3529 - val_accuracy: 0.4456 - lr: 0.0010
Epoch 65/100
1472/1472 [=====] - ETA: 0s - loss: 0.0042 - accuracy: 0.4035

micro_F1_train: 0.49860705306788067

micro_F1_val: 0.4936588459099556

Epoch 00065: val_accuracy improved from 0.46164 to 0.49366, saving model to model2_save/weights-65-0.49
37.hdf5
1472/1472 [=====] - 11s 8ms/step - loss: 0.0042 - accuracy: 0.4035 - val_loss:
1.2149 - val_accuracy: 0.4937 - lr: 0.0010
Epoch 66/100
1469/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.4031

micro_F1_train: 0.3989264116328057
```



micro\_F1\_train: 0.3922003804692454

micro\_F1\_val: 0.3922003804692454

Epoch 00066: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 10s 7ms/step - loss: 0.0048 - accuracy: 0.4031 - val\_loss: 1.4344 - val\_accuracy: 0.3922 - lr: 0.0010

Epoch 67/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0050 - accuracy: 0.3862

micro\_F1\_train: 0.4152340830332269

micro\_F1\_val: 0.40424857324032976

Epoch 00067: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 10s 7ms/step - loss: 0.0050 - accuracy: 0.3859 - val\_loss: 1.4106 - val\_accuracy: 0.4042 - lr: 0.0010

Epoch 68/100

1461/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.3949

micro\_F1\_train: 0.45274172725419587

micro\_F1\_val: 0.4430881420418516

Epoch 00068: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 10s 7ms/step - loss: 0.0041 - accuracy: 0.3951 - val\_loss: 1.3557 - val\_accuracy: 0.4431 - lr: 0.0010

Epoch 69/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3886

micro\_F1\_train: 0.43290072705034993

micro\_F1\_val: 0.4212111604311985

Epoch 00069: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 11s 7ms/step - loss: 0.0047 - accuracy: 0.3887 - val\_loss: 1.3916 - val\_accuracy: 0.4212 - lr: 0.0010

Epoch 70/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3835

micro\_F1\_train: 0.4408507168580554

micro\_F1\_val: 0.4329422954977806

Epoch 00070: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 10s 7ms/step - loss: 0.0047 - accuracy: 0.3840 - val\_loss: 1.4188 - val\_accuracy: 0.4329 - lr: 0.0010

Epoch 71/100

1472/1472 [=====] - ETA: 0s - loss: 0.0042 - accuracy: 0.4040

micro\_F1\_train: 0.43643405585377454

micro\_F1\_val: 0.4292961318960051

Epoch 00071: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 11s 7ms/step - loss: 0.0042 - accuracy: 0.4040 - val\_loss: 1.3792 - val\_accuracy: 0.4293 - lr: 0.0010

Epoch 72/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3839

micro\_F1\_train: 0.4518583950533397

micro\_F1\_val: 0.4394419784400761

Epoch 00072: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 10s 7ms/step - loss: 0.0045 - accuracy: 0.3841 - val\_loss: 1.3069 - val\_accuracy: 0.4394 - lr: 0.0010

Epoch 73/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.4007

micro\_F1\_train: 0.4305904736019569

micro\_F1\_val: 0.42105263157894735

Epoch 00073: val\_accuracy did not improve from 0.49366

1472/1472 [=====] - 9s 6ms/step - loss: 0.0041 - accuracy: 0.4010 - val\_loss: 1.3538 - val\_accuracy: 0.4211 - lr: 0.0010

Epoch 74/100

```
Epoch 74/100
1467/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.3905

micro_F1_train: 0.46938914180879254

micro_F1_val: 0.4575142675967026

Epoch 00074: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 11s 7ms/step - loss: 0.0042 - accuracy: 0.3912 - val_loss:
1.3082 - val_accuracy: 0.4575 - lr: 0.0010
Epoch 75/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0049 - accuracy: 0.3893

micro_F1_train: 0.42549432628932526

micro_F1_val: 0.4139188332276474

Epoch 00075: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 11s 8ms/step - loss: 0.0049 - accuracy: 0.3894 - val_loss:
1.3696 - val_accuracy: 0.4139 - lr: 0.0010
Epoch 76/100
1472/1472 [=====] - ETA: 0s - loss: 0.0041 - accuracy: 0.4119

micro_F1_train: 0.42997893592444114

micro_F1_val: 0.41867469879518066

Epoch 00076: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 11s 8ms/step - loss: 0.0041 - accuracy: 0.4119 - val_loss:
1.3972 - val_accuracy: 0.4187 - lr: 0.0010
Epoch 77/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.4018

micro_F1_train: 0.42284432968675684

micro_F1_val: 0.4126506024096386

Epoch 00077: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.4021 - val_loss:
1.3950 - val_accuracy: 0.4127 - lr: 0.0010
Epoch 78/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.3935

micro_F1_train: 0.40701229870218114

micro_F1_val: 0.39600507292327203

Epoch 00078: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.3935 - val_loss:
1.4341 - val_accuracy: 0.3960 - lr: 0.0010
Epoch 79/100
1469/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3973

micro_F1_train: 0.427600733845213

micro_F1_val: 0.41708941027266955

Epoch 00079: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 11s 7ms/step - loss: 0.0045 - accuracy: 0.3972 - val_loss:
1.4227 - val_accuracy: 0.4171 - lr: 0.0010
Epoch 80/100
1465/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3930

micro_F1_train: 0.4573622341509819

micro_F1_val: 0.4513316423589093

Epoch 00080: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 10s 6ms/step - loss: 0.0045 - accuracy: 0.3933 - val_loss:
1.3088 - val_accuracy: 0.4513 - lr: 0.0010
Epoch 81/100
1461/1472 [=====>.] - ETA: 0s - loss: 0.0043 - accuracy: 0.4003

micro_F1_train: 0.43473534008289727

micro_F1_val: 0.42675967025998734

Epoch 00081: val_accuracy did not improve from 0.49366
```

```
Epoch 00081: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 11s 7ms/step - loss: 0.0043 - accuracy: 0.3997 - val_loss:
1.3251 - val_accuracy: 0.4268 - lr: 0.0010
Epoch 82/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4205

micro_F1_train: 0.42284432968675684

micro_F1_val: 0.41550412175015855

Epoch 00082: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 10s 6ms/step - loss: 0.0037 - accuracy: 0.4204 - val_loss:
1.3754 - val_accuracy: 0.4155 - lr: 0.0010
Epoch 83/100
1465/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4188

micro_F1_train: 0.44071481959638514

micro_F1_val: 0.43896639188332276

Epoch 00083: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 10s 7ms/step - loss: 0.0040 - accuracy: 0.4193 - val_loss:
1.3575 - val_accuracy: 0.4390 - lr: 0.0010
Epoch 84/100
1467/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.4116

micro_F1_train: 0.44513148060066593

micro_F1_val: 0.4367469879518072

Epoch 00084: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 11s 8ms/step - loss: 0.0041 - accuracy: 0.4116 - val_loss:
1.3252 - val_accuracy: 0.4367 - lr: 0.0010
Epoch 85/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.4014

micro_F1_train: 0.47108785757966976

micro_F1_val: 0.46496512365250475

Epoch 00085: val_accuracy did not improve from 0.49366
1472/1472 [=====] - 10s 7ms/step - loss: 0.0044 - accuracy: 0.4013 - val_loss:
1.2943 - val_accuracy: 0.4650 - lr: 0.0010
Epoch 00085: early stopping
Time Taken for training (sec): 919.1604928970337
```

In [ ]:

```
# http://localhost:6006/
%load_ext tensorboard
%tensorboard --logdir logs --host localhost
```

The tensorboard extension is already loaded. To reload it, use:

```
%reload_ext tensorboard
```

Reusing TensorBoard on port 6006 (pid 2268), started 0:43:48 ago. (Use '!kill 2268' to kill it.)

## Model 2 Predictions

In [ ]:

```
NeuNet_2.load_weights("/content/m2_weights-65-0.4937.hdf5")
```

In [ ]:

```
y_pred_2 = NeuNet_2.predict(px_data_std)
```

In [ ]:

```
np.argmax(y_pred_2,axis=1)
```

```
Out[ ]:
```

```
array([ 3,  3,  3, ..., 14, 14, 14])
```

```
In [ ]:
```

```
y_pred_pd_2 = pd.DataFrame(np.argmax(y_pred_2,axis=1),index=px_data.index)  
# y_pred_pd[0] = y_pred_pd[0]+1
```

```
In [ ]:
```

```
# y_pred_pd_2
```

```
In [ ]:
```

```
y_pred_pd_2.value_counts()
```

```
Out[ ]:
```

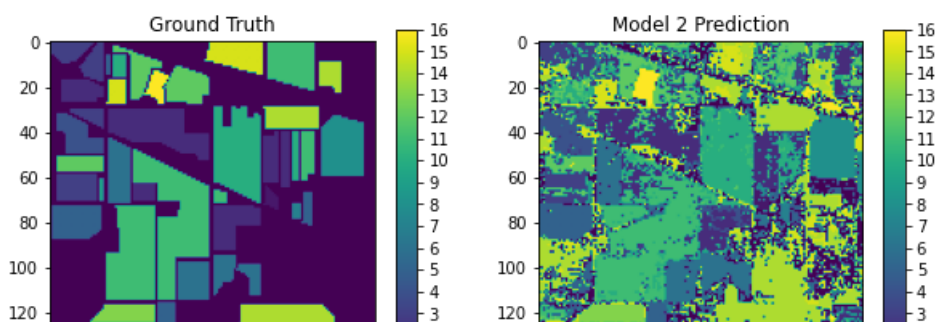
```
14    3795  
0     2726  
11    2244  
15    2170  
2     1747  
10    1594  
12    1408  
6     1347  
5      945  
3      794  
4      766  
8      691  
13     292  
16     247  
7     103  
1       82  
9       74  
dtype: int64
```

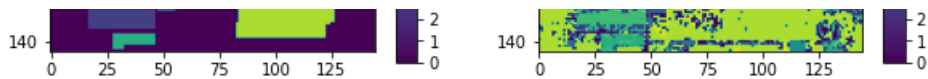
```
In [ ]:
```

```
# px_data.index
```

```
In [ ]:
```

```
figr,axis = plt.subplots(1,2,figsize=(10,10))  
im0 = axis[0].imshow(mat_gt['indian_pines_gt'])#,cmap='jet')  
axis[0].set_title('Ground Truth')  
plt.colorbar(im0,ax=axis[0],shrink=0.4,aspect=16, ticks=range(0,17,1))  
  
im1 = axis[1].imshow(y_pred_pd_2.to_numpy().reshape((145,145)))#,cmap='jet')  
axis[1].set_title('Model 2 Prediction')  
plt.colorbar(im1,ax=axis[1],shrink=0.4,aspect=16, ticks=range(0,17,1))  
plt.savefig('NeuNet_2_e100.png')  
plt.show()
```





## Observation

In [ ]:

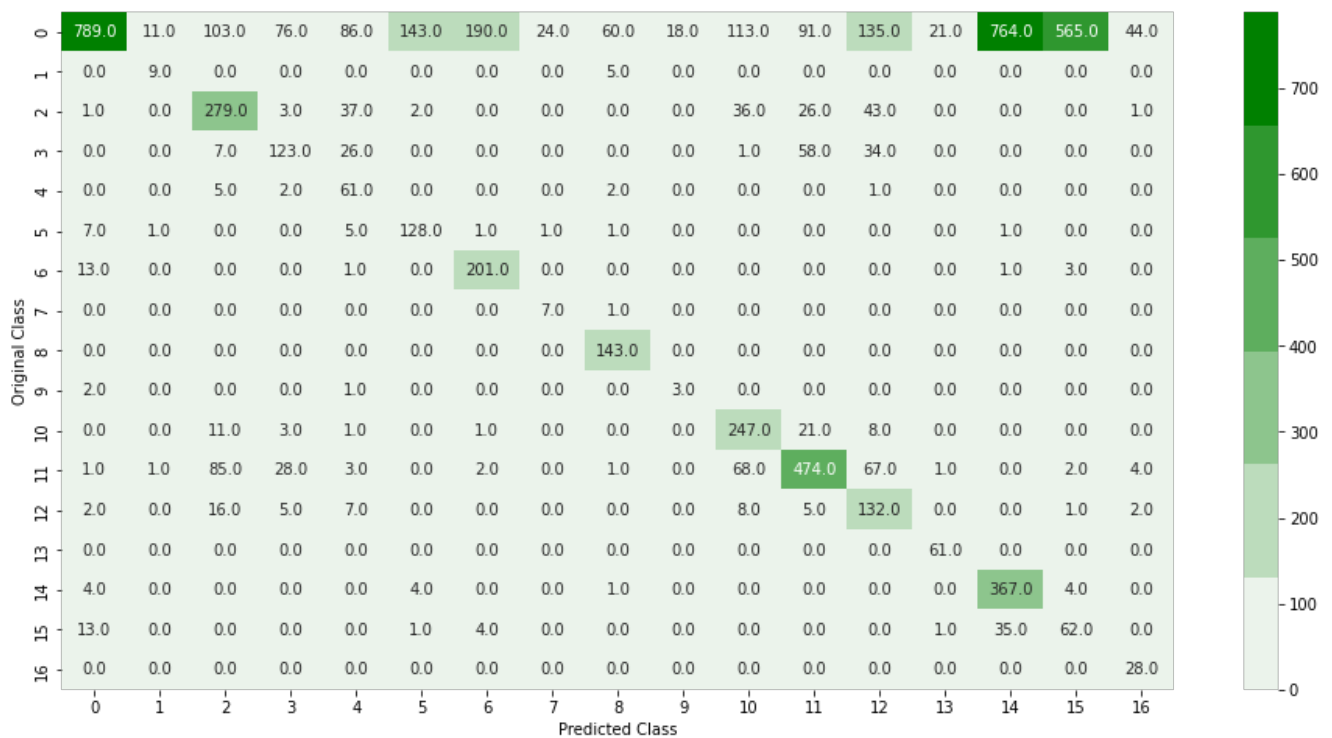
```
y_pred_2_test = NeuNet_2.predict(X_test_std)
y_pred_pd_2_test = pd.DataFrame(np.argmax(y_pred_2_test,axis=1),index=y_test.index)
```

In [ ]:

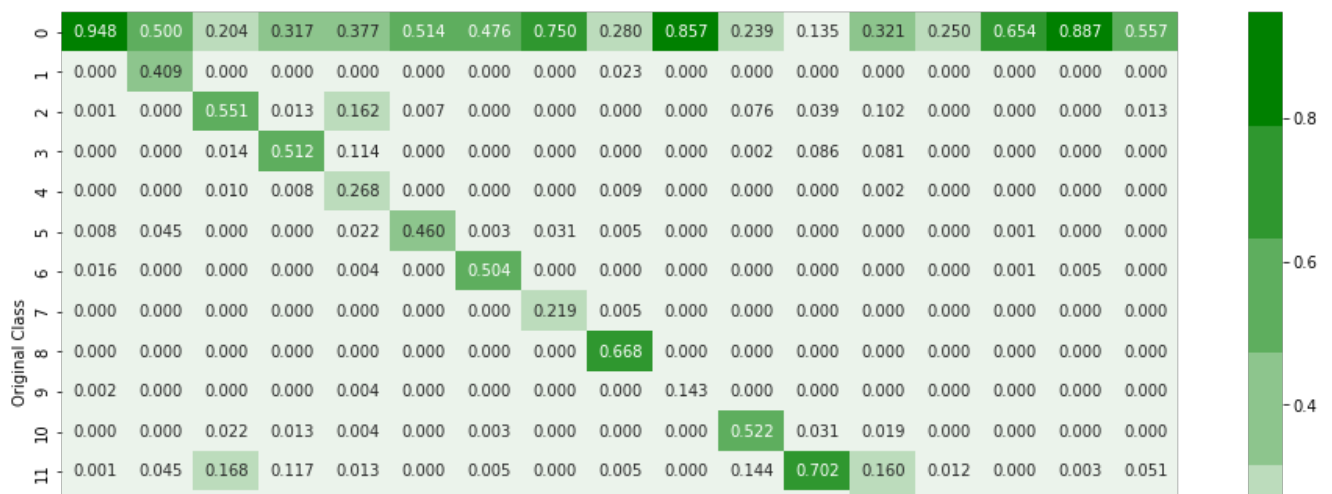
```
# plot_confusion_matrix(px_class,y_pred_pd_2)
plot_confusion_matrix(y_test,y_pred_pd_2_test)
```

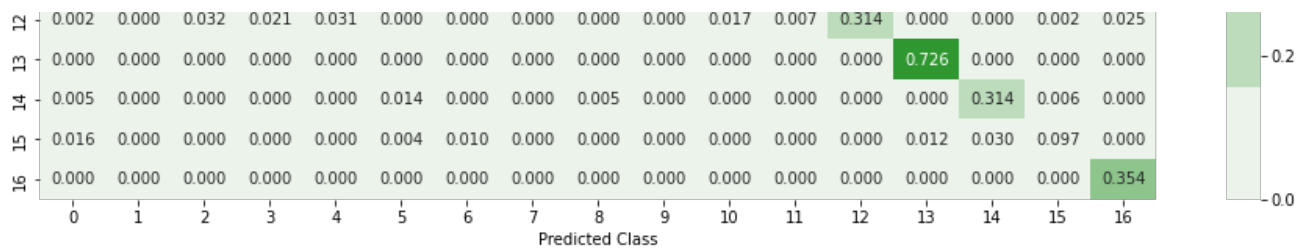
Number of misclassified points 50.634115409004444

----- Confusion matrix -----  
-----



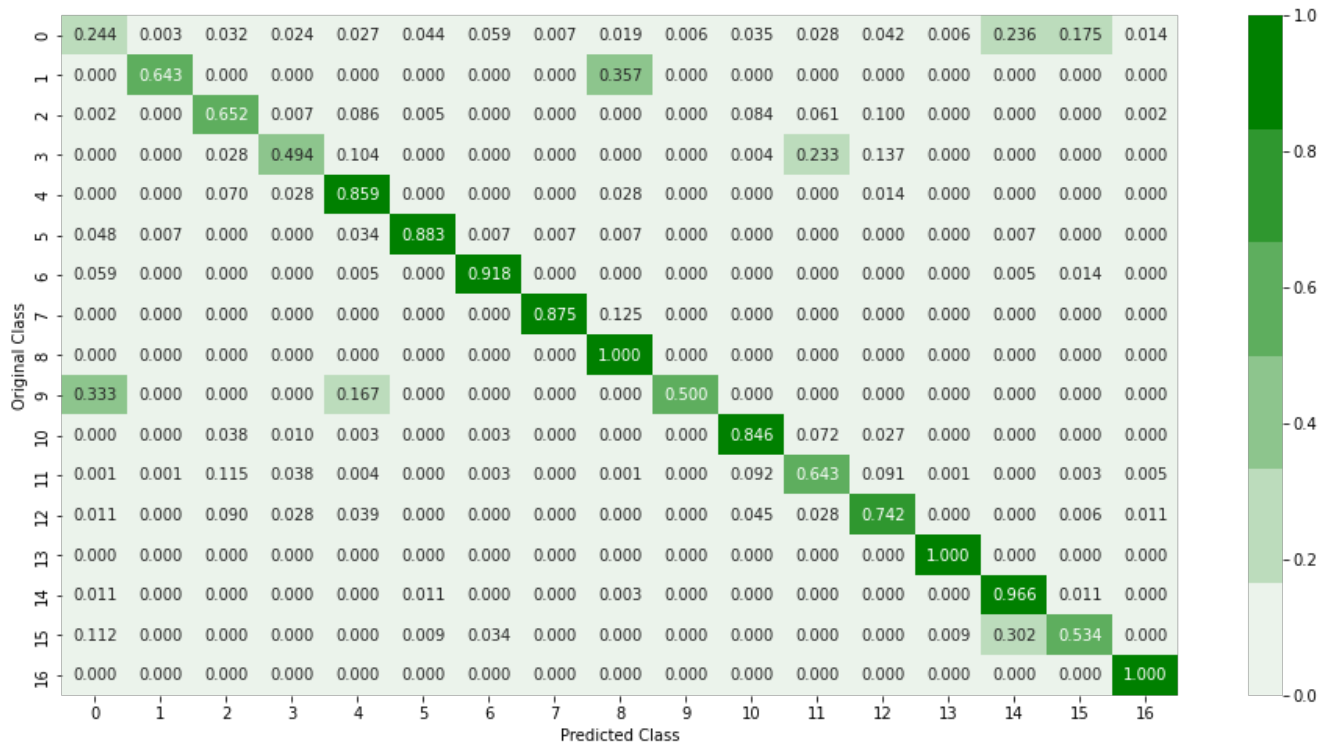
----- Precision matrix -----  
-----





Sum of columns in precision matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

Recall matrix



Sum of rows in recall matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

- From Recall matrix we can see that most of the classes have been predicted with high recall.
- Precisions are low for most classes except for class 'zero'.
- From precision matrix (first row), we observe that model finds there are similarity among class 'zero' and other classes.

### 3- Four layered network with tanh activation and SGD optimizer

#### Model 3 Definition

In [ ]:

```
# del NeuNet_3
```

In [ ]:

```
In = Input(shape=(95,))
L1 = Dense(128,activation='tanh',
          kernel_initializer=initializers.he_uniform())(In) #https://keras.io/api/layers/initializers/

L1_Drp = Dropout(0.4)(L1)

L2 = Dense(64,activation='tanh',
          kernel_initializer=initializers.he_uniform())(L1_Drp)

L2_Drp = Dropout(0.3)(L2)
```

```

L3 = Dense(32,activation='tanh',
          kernel_initializer=initializers.he_uniform()) (L2_Drp)

L4 = Dense(20,activation='tanh',
          kernel_initializer=initializers.he_uniform()) (L3)

Out = Dense(17,activation='softmax',
           kernel_initializer=initializers.GlorotUniform()) (L4)

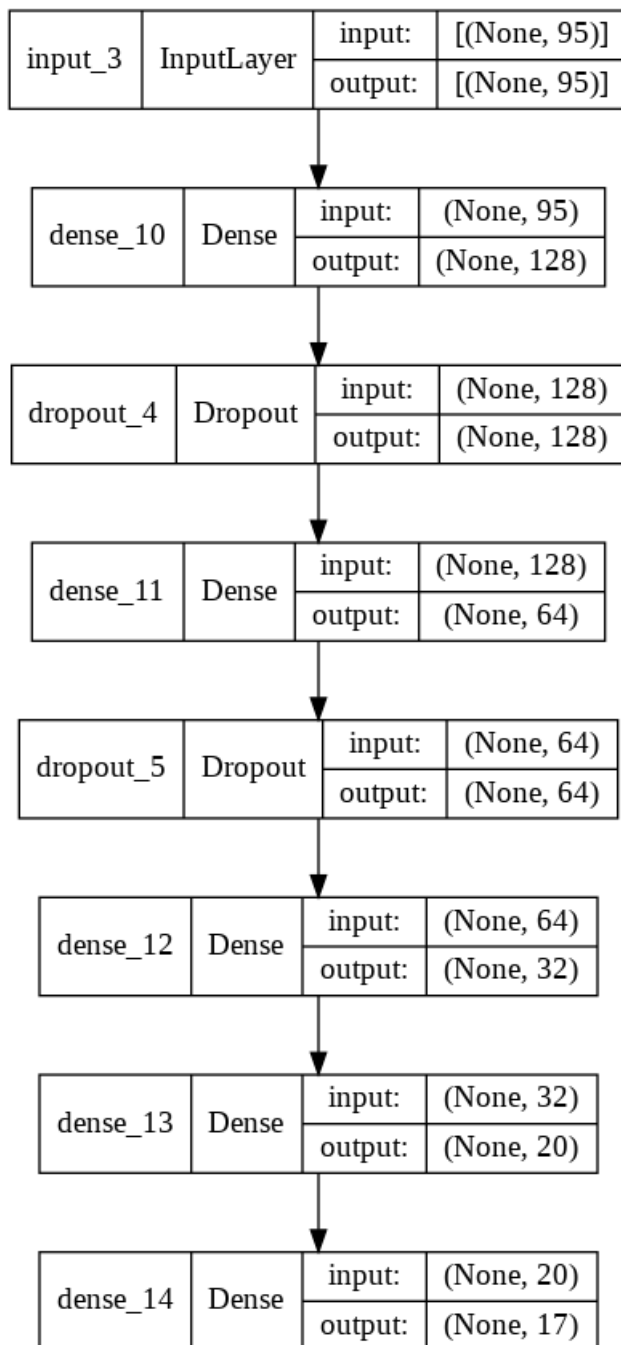
NeuNet_3 = Model(inputs=In, outputs = Out)

# cce = categorical_crossentropy()
NeuNet_3.compile(loss='categorical_crossentropy', optimizer='SGD', metrics=['accuracy'])

plot_model(NeuNet_3,show_layer_names=True, show_shapes=True)#,to_file='NeuNet_3.png')

```

Out[ ]:



In [ ]:

```
NeuNet_3.summary()
```

Model: "model\_2"

Layer (type)	Output Shape	Param #
input_3 (InputLayer)	[ (None, 95) ]	0
dense_10 (Dense)	(None, 128)	12288
dropout_4 (Dropout)	(None, 128)	0
dense_11 (Dense)	(None, 64)	8256
dropout_5 (Dropout)	(None, 64)	0
dense_12 (Dense)	(None, 32)	2080
dense_13 (Dense)	(None, 20)	660
dense_14 (Dense)	(None, 17)	357

Total params: 23,641  
Trainable params: 23,641  
Non-trainable params: 0

### Model 3 Training

In [ ]:

```
logdir = os.path.join("logs", datetime.now().strftime("%Y%m%d-%H%M%S"))
print(logdir)
file_writer = tf.summary.create_file_writer(logdir + "/metrics")
# tensorboard = TensorBoard(log_dir=logdir)
tensorboard = TensorBoard(log_dir=logdir, histogram_freq=1, write_graph=True, write_grads=True)

metric_calc_3 = metric_calc()

# Saving model at every epoch if validation accuracy is improved from previous epoch
filepath_m3="model3_save/weights-{epoch:02d}-{val_accuracy:.4f}.hdf5"
checkpoint_m3 = ModelCheckpoint(filepath=filepath_m3, monitor='val_accuracy', verbose=1,
                                save_best_only=True,
                                mode='auto')
earlystop_m3 = EarlyStopping(monitor='val_accuracy', min_delta=0.01, patience=20, verbose=1)
reduce_lr_m3 = ReduceLROnPlateau(monitor='val_loss', factor=0.9, patience=2, min_lr=0.001)

callback_list_3 = [metric_calc_3,
                   checkpoint_m3,
                   earlystop_m3,
                   reduce_lr_m3,
                   tensorboard]

# fit network
verbose_3, epochs_3, batch_size_3 = 1, 100, 10
start = time.time()
history_3 = NeuNet_3.fit(X_train_std, y_ctg_train,
                        class_weight = class_wts,
                        validation_data=(X_test_std, y_ctg_test),
                        epochs=epochs_3,
                        batch_size=batch_size_3,
                        verbose=verbose_3,
                        callbacks=callback_list_3)

# Passing the F1 score data of each epoch to tensor board
# https://stackoverflow.com/questions/58102016/how-visualize-in-tensorboard-a-metric-callback
for i in range(len(metric_calc_3.metrics['micro_F1_train'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_train', metric_calc_3.metrics['micro_F1_train'][i])
file_writer.flush ()

for i in range(len(metric_calc_3.metrics['micro_F1_val'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_val', metric_calc_3.metrics['micro_F1_val'][i])
file_writer.flush ()
```



```
stop = time.time()
print('Time Taken for training (sec): ',stop-start)
```

logs/20220131-130044

WARNING:tensorflow:write\_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.

Epoch 1/100

2/1472 [.....] - ETA: 5:40 - loss: 0.0167 - accuracy: 0.1500 WARNING:tensorflow:Callback method `on\_train\_batch\_begin` is slow compared to the batch time (batch time: 0.0041s vs `on\_train\_batch\_begin` time: 0.0226s). Check your callbacks.

WARNING:tensorflow:Callback method `on\_train\_batch\_end` is slow compared to the batch time (batch time: 0.0041s vs `on\_train\_batch\_end` time: 0.0171s). Check your callbacks.

1461/1472 [=====>.] - ETA: 0s - loss: 0.0171 - accuracy: 0.0643

micro\_F1\_train: 0.05802813073316573

micro\_F1\_val: 0.060875079264426125

Epoch 00001: val\_accuracy improved from -inf to 0.06088, saving model to model3\_save/weights-01-0.0609.hdf5

1472/1472 [=====>.] - 11s 7ms/step - loss: 0.0171 - accuracy: 0.0643 - val\_loss: 3.0213 - val\_accuracy: 0.0609 - lr: 0.0100

Epoch 2/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0169 - accuracy: 0.0724

micro\_F1\_train: 0.06427940476999389

micro\_F1\_val: 0.06800887761572606

Epoch 00002: val\_accuracy improved from 0.06088 to 0.06801, saving model to model3\_save/weights-02-0.0680.hdf5

1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0169 - accuracy: 0.0724 - val\_loss: 2.9864 - val\_accuracy: 0.0680 - lr: 0.0100

Epoch 3/100

1461/1472 [=====>.] - ETA: 0s - loss: 0.0168 - accuracy: 0.0743

micro\_F1\_train: 0.0713460623768431

micro\_F1\_val: 0.07688649334178821

Epoch 00003: val\_accuracy improved from 0.06801 to 0.07689, saving model to model3\_save/weights-03-0.0769.hdf5

1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0168 - accuracy: 0.0744 - val\_loss: 2.9584 - val\_accuracy: 0.0769 - lr: 0.0100

Epoch 4/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0165 - accuracy: 0.0762

micro\_F1\_train: 0.07725759325949583

micro\_F1\_val: 0.08608116677235257

Epoch 00004: val\_accuracy improved from 0.07689 to 0.08608, saving model to model3\_save/weights-04-0.0861.hdf5

1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0165 - accuracy: 0.0766 - val\_loss: 2.9263 - val\_accuracy: 0.0861 - lr: 0.0100

Epoch 5/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0164 - accuracy: 0.0827

micro\_F1\_train: 0.08330502140381871

micro\_F1\_val: 0.0911540900443881

Epoch 00005: val\_accuracy improved from 0.08608 to 0.09115, saving model to model3\_save/weights-05-0.0912.hdf5

1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0164 - accuracy: 0.0825 - val\_loss: 2.9009 - val\_accuracy: 0.0912 - lr: 0.0100

Epoch 6/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0161 - accuracy: 0.0853

micro\_F1\_train: 0.0881973228239451

micro\_F1\_val: 0.09559289790741915

Epoch 00006: val\_accuracy improved from 0.09115 to 0.09559, saving model to model3\_save/weights-06-0.0956.hdf5

1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0161 - accuracy: 0.0853 - val\_loss: 2.8728 - val\_accuracy: 0.0956 - lr: 0.0100

Epoch 7/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0159 - accuracy: 0.0882

micro\_F1\_train: 0.09213834341238024

micro\_F1\_val: 0.10082435003170578

Epoch 00007: val\_accuracy improved from 0.09559 to 0.10082, saving model to model3\_save/weights-07-0.1008.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0159 - accuracy: 0.0881 - val\_loss: 2.8477 - val\_accuracy: 0.1008 - lr: 0.0100

Epoch 8/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0157 - accuracy: 0.0905

micro\_F1\_train: 0.09798192566419786

micro\_F1\_val: 0.10732403297400127

Epoch 00008: val\_accuracy improved from 0.10082 to 0.10732, saving model to model3\_save/weights-08-0.1073.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0157 - accuracy: 0.0908 - val\_loss: 2.8253 - val\_accuracy: 0.1073 - lr: 0.0100

Epoch 9/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0156 - accuracy: 0.0948

micro\_F1\_train: 0.10416525107019094

micro\_F1\_val: 0.11176284083703233

Epoch 00009: val\_accuracy improved from 0.10732 to 0.11176, saving model to model3\_save/weights-09-0.1118.hdf5  
1472/1472 [=====] - 16s 11ms/step - loss: 0.0156 - accuracy: 0.0949 - val\_loss: 2.8003 - val\_accuracy: 0.1118 - lr: 0.0100

Epoch 10/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0153 - accuracy: 0.1031

micro\_F1\_train: 0.1091934497519875

micro\_F1\_val: 0.11905516804058339

Epoch 00010: val\_accuracy improved from 0.11176 to 0.11906, saving model to model3\_save/weights-10-0.1191.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0153 - accuracy: 0.1032 - val\_loss: 2.7759 - val\_accuracy: 0.1191 - lr: 0.0100

Epoch 11/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0152 - accuracy: 0.0984

micro\_F1\_train: 0.1145613915879595

micro\_F1\_val: 0.12492073557387444

Epoch 00011: val\_accuracy improved from 0.11906 to 0.12492, saving model to model3\_save/weights-11-0.1249.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0152 - accuracy: 0.0985 - val\_loss: 2.7594 - val\_accuracy: 0.1249 - lr: 0.0100

Epoch 12/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0151 - accuracy: 0.1000

micro\_F1\_train: 0.11931779574641571

micro\_F1\_val: 0.12698161065313887

Epoch 00012: val\_accuracy improved from 0.12492 to 0.12698, saving model to model3\_save/weights-12-0.1270.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0151 - accuracy: 0.1000 - val\_loss: 2.7410 - val\_accuracy: 0.1270 - lr: 0.0100

Epoch 13/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0150 - accuracy: 0.1078

micro\_F1\_train: 0.12570496704491405

micro\_F1\_val: 0.13157894736842105

Epoch 00013: val\_accuracy improved from 0.12698 to 0.13158, saving model to model3\_save/weights-13-0.1316.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0150 - accuracy: 0.1075 - val\_loss: 2.7219 - val\_accuracy: 0.1316 - lr: 0.0100

Epoch 14/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0148 - accuracy: 0.1069

micro\_F1\_train: 0.13059726846504044

micro\_F1\_val: 0.13474952441344323

Epoch 00014: val\_accuracy improved from 0.13158 to 0.13475, saving model to model3\_save/weights-14-0.1347.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0148 - accuracy: 0.1069 - val\_loss: 2.7055 - val\_accuracy: 0.1347 - lr: 0.0100

Epoch 15/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0146 - accuracy: 0.1089

micro\_F1\_train: 0.13453828905347556

micro\_F1\_val: 0.1388712745719721

Epoch 00015: val\_accuracy improved from 0.13475 to 0.13887, saving model to model3\_save/weights-15-0.1389.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0147 - accuracy: 0.1089 - val\_loss: 2.6893 - val\_accuracy: 0.1389 - lr: 0.0100

Epoch 16/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0145 - accuracy: 0.1133

micro\_F1\_train: 0.13895495005775635

micro\_F1\_val: 0.14267596702599875

Epoch 00016: val\_accuracy improved from 0.13887 to 0.14268, saving model to model3\_save/weights-16-0.1427.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0145 - accuracy: 0.1138 - val\_loss: 2.6744 - val\_accuracy: 0.1427 - lr: 0.0100

Epoch 17/100  
1461/1472 [=====>.] - ETA: 0s - loss: 0.0143 - accuracy: 0.1094

micro\_F1\_train: 0.14242033023034587

micro\_F1\_val: 0.14426125554850983

Epoch 00017: val\_accuracy improved from 0.14268 to 0.14426, saving model to model3\_save/weights-17-0.1443.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0143 - accuracy: 0.1093 - val\_loss: 2.6624 - val\_accuracy: 0.1443 - lr: 0.0100

Epoch 18/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0144 - accuracy: 0.1132

micro\_F1\_train: 0.14690493986546171

micro\_F1\_val: 0.14901712111604312

Epoch 00018: val\_accuracy improved from 0.14426 to 0.14902, saving model to model3\_save/weights-18-0.1490.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0144 - accuracy: 0.1132 - val\_loss: 2.6501 - val\_accuracy: 0.1490 - lr: 0.0100

Epoch 19/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0142 - accuracy: 0.1181

micro\_F1\_train: 0.14894339879051438

micro\_F1\_val: 0.15266328471781865

Epoch 00019: val\_accuracy improved from 0.14902 to 0.15266, saving model to model3\_save/weights-19-0.1527.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0142 - accuracy: 0.1178 - val\_loss: 2.6379 - val\_accuracy: 0.1527 - lr: 0.0100

Epoch 20/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0141 - accuracy: 0.1210

micro\_F1\_train: 0.15091390908473196

micro\_F1\_val: 0.1569435637285986

Epoch 00020: val\_accuracy improved from 0.15266 to 0.15694, saving model to model3\_save/weights-20-0.1569.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0140 - accuracy: 0.1212 - val\_loss: 2.6265 - val\_accuracy: 0.1569 - lr: 0.0100

Epoch 21/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0138 - accuracy: 0.1229

micro\_F1\_train: 0.15349595705646532

micro\_F1\_val: 0.15710209258084973

Epoch 00021: val\_accuracy improved from 0.15694 to 0.15710, saving model to model3\_save/weights-21-0.1571.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0138 - accuracy: 0.1230 - val\_loss: 2.6166 - val\_accuracy: 0.1571 - lr: 0.0100

Epoch 22/100  
1460/1472 [=====>.] - ETA: 0s - loss: 0.0139 - accuracy: 0.1235

micro\_F1\_train: 0.15397159747231093

micro\_F1\_val: 0.1586873811033608

Epoch 00022: val\_accuracy improved from 0.15710 to 0.15869, saving model to model3\_save/weights-22-0.1587.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0138 - accuracy: 0.1235 - val\_loss: 2.6087 - val\_accuracy: 0.1587 - lr: 0.0100

Epoch 23/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0136 - accuracy: 0.1251

micro\_F1\_train: 0.15512672419650744

micro\_F1\_val: 0.15900443880786302

Epoch 00023: val\_accuracy improved from 0.15869 to 0.15900, saving model to model3\_save/weights-23-0.1590.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0136 - accuracy: 0.1251 - val\_loss: 2.6010 - val\_accuracy: 0.1590 - lr: 0.0100

Epoch 24/100  
1472/1472 [=====] - ETA: 0s - loss: 0.0136 - accuracy: 0.1235

micro\_F1\_train: 0.15743697764490044

micro\_F1\_val: 0.16058972733037413

Epoch 00024: val\_accuracy improved from 0.15900 to 0.16059, saving model to model3\_save/weights-24-0.1606.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0136 - accuracy: 0.1235 - val\_loss: 2.5937 - val\_accuracy: 0.1606 - lr: 0.0100

Epoch 25/100  
1465/1472 [=====>.] - ETA: 0s - loss: 0.0135 - accuracy: 0.1264

micro\_F1\_train: 0.16090235781749

micro\_F1\_val: 0.1631261889663919

Epoch 00025: val\_accuracy improved from 0.16059 to 0.16313, saving model to model3\_save/weights-25-0.1631.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0135 - accuracy: 0.1262 - val\_loss: 2.5856 - val\_accuracy: 0.1631 - lr: 0.0100

Epoch 26/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0133 - accuracy: 0.1291

micro\_F1\_train: 0.1626690222192023

micro\_F1\_val: 0.16582117945466074

Epoch 00026: val\_accuracy improved from 0.16313 to 0.16582, saving model to model3\_save/weights-26-0.1658.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0133 - accuracy: 0.1293 - val\_loss: 2.5781 - val\_accuracy: 0.1658 - lr: 0.0100

Epoch 27/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0133 - accuracy: 0.1269

micro\_F1\_train: 0.1655908133451111

micro\_F1\_val: 0.16724793912492072

Epoch 00027: val\_accuracy improved from 0.16582 to 0.16725, saving model to model3\_save/weights-27-0.1672.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0133 - accuracy: 0.1271 - val\_loss: 2.5708 - val\_accuracy: 0.1672 - lr: 0.0100

Epoch 28/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0132 - accuracy: 0.1327

micro\_F1\_train: 0.16756132363932866

micro\_F1\_val: 0.16804058338617628

Epoch 00028: val\_accuracy improved from 0.16725 to 0.16804, saving model to model3\_save/weights-28-0.1680.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0132 - accuracy: 0.1327 - val\_loss: 2.5647 - val\_accuracy: 0.1680 - lr: 0.0100

Epoch 29/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0130 - accuracy: 0.1359

micro\_F1\_train: 0.16871645036352517

micro\_F1\_val: 0.1683576410906785

Epoch 00029: val\_accuracy improved from 0.16804 to 0.16836, saving model to model3\_save/weights-29-0.1684.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0130 - accuracy: 0.1360 - val\_loss: 2.5577 - val\_accuracy: 0.1684 - lr: 0.0100

Epoch 30/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0129 - accuracy: 0.1388

micro\_F1\_train: 0.1697356798260515

micro\_F1\_val: 0.16851616994292962

Epoch 00030: val\_accuracy improved from 0.16836 to 0.16852, saving model to model3\_save/weights-30-0.1685.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0130 - accuracy: 0.1387 - val\_loss: 2.5512 - val\_accuracy: 0.1685 - lr: 0.0100

Epoch 31/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0129 - accuracy: 0.1377

micro\_F1\_train: 0.17245362505945505

micro\_F1\_val: 0.1713696892834496

Epoch 00031: val\_accuracy improved from 0.16852 to 0.17137, saving model to model3\_save/weights-31-0.1714.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0129 - accuracy: 0.1377 - val\_loss: 2.5456 - val\_accuracy: 0.1714 - lr: 0.0100

Epoch 32/100  
1471/1472 [=====>.] - ETA: 0s - loss: 0.0128 - accuracy: 0.1346

micro\_F1\_train: 0.17422028946116735

micro\_F1\_val: 0.173430564362714

Epoch 00032: val\_accuracy improved from 0.17137 to 0.17343, saving model to model3\_save/weights-32-0.1734.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0128 - accuracy: 0.1347 - val\_loss: 2.5414 - val\_accuracy: 0.1734 - lr: 0.0100

Epoch 33/100  
1464/1472 [=====>.] - ETA: 0s - loss: 0.0126 - accuracy: 0.1412

micro\_F1\_train: 0.17578310797037439

micro\_F1\_val: 0.17438173747622068

Epoch 00033: val\_accuracy improved from 0.17343 to 0.17438, saving model to model3\_save/weights-33-0.1744.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0127 - accuracy: 0.1413 - val\_loss: 2.5371 - val\_accuracy: 0.1744 - lr: 0.0100

Epoch 34/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0127 - accuracy: 0.1366

micro\_F1\_train: 0.17782156689542702

micro\_F1\_val: 0.17707672796448953

Epoch 00034: val\_accuracy improved from 0.17438 to 0.17708, saving model to model3\_save/weights-34-0.1771.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0127 - accuracy: 0.1366 - val\_loss: 2.5331 - val\_accuracy: 0.1771 - lr: 0.0100

Epoch 35/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0126 - accuracy: 0.1440  
  
micro\_F1\_train: 0.1804715634979955  
  
micro\_F1\_val: 0.1797717184527584  
  
Epoch 00035: val\_accuracy improved from 0.17708 to 0.17977, saving model to model3\_save/weights-35-0.1798.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0126 - accuracy: 0.1441 - val\_loss: 2.5285 - val\_accuracy: 0.1798 - lr: 0.0100  
Epoch 36/100  
1465/1472 [=====>.] - ETA: 0s - loss: 0.0125 - accuracy: 0.1446  
  
micro\_F1\_train: 0.18203438200720257  
  
micro\_F1\_val: 0.18167406467977168  
  
Epoch 00036: val\_accuracy improved from 0.17977 to 0.18167, saving model to model3\_save/weights-36-0.1817.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0125 - accuracy: 0.1444 - val\_loss: 2.5236 - val\_accuracy: 0.1817 - lr: 0.0100  
Epoch 37/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0125 - accuracy: 0.1403  
  
micro\_F1\_train: 0.1833254059930692  
  
micro\_F1\_val: 0.18373493975903613  
  
Epoch 00037: val\_accuracy improved from 0.18167 to 0.18373, saving model to model3\_save/weights-37-0.1837.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0125 - accuracy: 0.1401 - val\_loss: 2.5197 - val\_accuracy: 0.1837 - lr: 0.0100  
Epoch 38/100  
1472/1472 [=====] - ETA: 0s - loss: 0.0124 - accuracy: 0.1442  
  
micro\_F1\_train: 0.1848202758714412  
  
micro\_F1\_val: 0.18516169942929614  
  
Epoch 00038: val\_accuracy improved from 0.18373 to 0.18516, saving model to model3\_save/weights-38-0.1852.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0124 - accuracy: 0.1442 - val\_loss: 2.5160 - val\_accuracy: 0.1852 - lr: 0.0100  
Epoch 39/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0122 - accuracy: 0.1486  
  
micro\_F1\_train: 0.18631514574981312  
  
micro\_F1\_val: 0.18753963221306277  
  
Epoch 00039: val\_accuracy improved from 0.18516 to 0.18754, saving model to model3\_save/weights-39-0.1875.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0123 - accuracy: 0.1489 - val\_loss: 2.5124 - val\_accuracy: 0.1875 - lr: 0.0100  
Epoch 40/100  
1460/1472 [=====>.] - ETA: 0s - loss: 0.0122 - accuracy: 0.1529  
  
micro\_F1\_train: 0.18679078616565875  
  
micro\_F1\_val: 0.18896639188332276  
  
Epoch 00040: val\_accuracy improved from 0.18754 to 0.18897, saving model to model3\_save/weights-40-0.1890.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0122 - accuracy: 0.1534 - val\_loss: 2.5098 - val\_accuracy: 0.1890 - lr: 0.0100  
Epoch 41/100  
1460/1472 [=====>.] - ETA: 0s - loss: 0.0122 - accuracy: 0.1524  
  
micro\_F1\_train: 0.18713052931983418  
  
micro\_F1\_val: 0.18896639188332276  
  
Epoch 00041: val\_accuracy did not improve from 0.18897  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0122 - accuracy: 0.1523 - val\_loss: 2.5067 - val\_accuracy: 0.1890 - lr: 0.0100  
Epoch 42/100

```
1466/1472 [=====>.] - ETA: 0s - loss: 0.0120 - accuracy: 0.1494

micro_F1_train: 0.18848950193653602

micro_F1_val: 0.1897590361445783

Epoch 00042: val_accuracy improved from 0.18897 to 0.18976, saving model to model3_save/weights-42-0.1898.hdf5
1472/1472 [=====] - 9s 6ms/step - loss: 0.0121 - accuracy: 0.1491 - val_loss: 2.5036 - val_accuracy: 0.1898 - lr: 0.0100
Epoch 43/100
1468/1472 [=====>.] - ETA: 0s - loss: 0.0121 - accuracy: 0.1515

micro_F1_train: 0.1895087313990623

micro_F1_val: 0.19261255548509829

Epoch 00043: val_accuracy improved from 0.18976 to 0.19261, saving model to model3_save/weights-43-0.1926.hdf5
1472/1472 [=====] - 10s 7ms/step - loss: 0.0121 - accuracy: 0.1515 - val_loss: 2.5005 - val_accuracy: 0.1926 - lr: 0.0100
Epoch 44/100
1461/1472 [=====>.] - ETA: 0s - loss: 0.0120 - accuracy: 0.1483

micro_F1_train: 0.1901202690765781

micro_F1_val: 0.19150285351934052

Epoch 00044: val_accuracy did not improve from 0.19261
1472/1472 [=====] - 10s 7ms/step - loss: 0.0120 - accuracy: 0.1483 - val_loss: 2.4975 - val_accuracy: 0.1915 - lr: 0.0100
Epoch 45/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0118 - accuracy: 0.1515

micro_F1_train: 0.19100360127743427

micro_F1_val: 0.19261255548509829

Epoch 00045: val_accuracy did not improve from 0.19261
1472/1472 [=====] - 11s 7ms/step - loss: 0.0119 - accuracy: 0.1513 - val_loss: 2.4954 - val_accuracy: 0.1926 - lr: 0.0100
Epoch 46/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0118 - accuracy: 0.1489

micro_F1_train: 0.19059590949242372

micro_F1_val: 0.1929296131896005

Epoch 00046: val_accuracy improved from 0.19261 to 0.19293, saving model to model3_save/weights-46-0.1929.hdf5
1472/1472 [=====] - 10s 7ms/step - loss: 0.0118 - accuracy: 0.1490 - val_loss: 2.4933 - val_accuracy: 0.1929 - lr: 0.0100
Epoch 47/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0117 - accuracy: 0.1542

micro_F1_train: 0.19079975538492897

micro_F1_val: 0.19308814204185162

Epoch 00047: val_accuracy improved from 0.19293 to 0.19309, saving model to model3_save/weights-47-0.1931.hdf5
1472/1472 [=====] - 10s 7ms/step - loss: 0.0117 - accuracy: 0.1538 - val_loss: 2.4912 - val_accuracy: 0.1931 - lr: 0.0100
Epoch 48/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0117 - accuracy: 0.1521

micro_F1_train: 0.19141129306244478

micro_F1_val: 0.19356372859860496

Epoch 00048: val_accuracy improved from 0.19309 to 0.19356, saving model to model3_save/weights-48-0.1936.hdf5
1472/1472 [=====] - 10s 7ms/step - loss: 0.0117 - accuracy: 0.1530 - val_loss: 2.4889 - val_accuracy: 0.1936 - lr: 0.0100
Epoch 49/100
1472/1472 [=====] - ETA: 0s - loss: 0.0116 - accuracy: 0.1565
```

micro\_F1\_train: 0.19215872800163078

micro\_F1\_val: 0.19435637285986046

Epoch 00049: val\_accuracy improved from 0.19356 to 0.19436, saving model to model3\_save/weights-49-0.1944.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0116 - accuracy: 0.1565 - val\_loss: 2.4866 - val\_accuracy: 0.1944 - lr: 0.0100

Epoch 50/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0116 - accuracy: 0.1579

micro\_F1\_train: 0.19243052252497111

micro\_F1\_val: 0.19435637285986046

Epoch 00050: val\_accuracy did not improve from 0.19436

1472/1472 [=====] - 9s 6ms/step - loss: 0.0116 - accuracy: 0.1578 - val\_loss: 2.4841 - val\_accuracy: 0.1944 - lr: 0.0100

Epoch 51/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0115 - accuracy: 0.1563

micro\_F1\_train: 0.1937215465108378

micro\_F1\_val: 0.19499048826886492

Epoch 00051: val\_accuracy improved from 0.19436 to 0.19499, saving model to model3\_save/weights-51-0.1950.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0115 - accuracy: 0.1559 - val\_loss: 2.4821 - val\_accuracy: 0.1950 - lr: 0.0100

Epoch 52/100

1472/1472 [=====] - ETA: 0s - loss: 0.0115 - accuracy: 0.1587

micro\_F1\_train: 0.1944010328191887

micro\_F1\_val: 0.19641724793912493

Epoch 00052: val\_accuracy improved from 0.19499 to 0.19642, saving model to model3\_save/weights-52-0.1964.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0115 - accuracy: 0.1587 - val\_loss: 2.4801 - val\_accuracy: 0.1964 - lr: 0.0100

Epoch 53/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0115 - accuracy: 0.1529

micro\_F1\_train: 0.19480872460419923

micro\_F1\_val: 0.199429296131896

Epoch 00053: val\_accuracy improved from 0.19642 to 0.19943, saving model to model3\_save/weights-53-0.1994.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0115 - accuracy: 0.1529 - val\_loss: 2.4795 - val\_accuracy: 0.1994 - lr: 0.0100

Epoch 54/100

1460/1472 [=====>.] - ETA: 0s - loss: 0.0114 - accuracy: 0.1589

micro\_F1\_train: 0.1941971869266834

micro\_F1\_val: 0.1992707672796449

Epoch 00054: val\_accuracy did not improve from 0.19943

1472/1472 [=====] - 9s 6ms/step - loss: 0.0114 - accuracy: 0.1589 - val\_loss: 2.4779 - val\_accuracy: 0.1993 - lr: 0.0100

Epoch 55/100

1472/1472 [=====] - ETA: 0s - loss: 0.0113 - accuracy: 0.1581

micro\_F1\_train: 0.19460487871169396

micro\_F1\_val: 0.19863665187064045

Epoch 00055: val\_accuracy did not improve from 0.19943

1472/1472 [=====] - 10s 6ms/step - loss: 0.0113 - accuracy: 0.1581 - val\_loss: 2.4769 - val\_accuracy: 0.1986 - lr: 0.0100

Epoch 56/100

1472/1472 [=====] - ETA: 0s - loss: 0.0113 - accuracy: 0.1606

micro\_F1\_train: 0.19460487871169396

micro\_F1\_val: 0.19879518072289157



- -

Epoch 00056: val\_accuracy did not improve from 0.19943  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0113 - accuracy: 0.1606 - val\_loss: 2.4751 - val\_accuracy: 0.1988 - lr: 0.0100  
Epoch 57/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0112 - accuracy: 0.1556

micro\_F1\_train: 0.1948766732350343

micro\_F1\_val: 0.19863665187064045

Epoch 00057: val\_accuracy did not improve from 0.19943  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0112 - accuracy: 0.1557 - val\_loss: 2.4740 - val\_accuracy: 0.1986 - lr: 0.0100  
Epoch 58/100  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0112 - accuracy: 0.1605

micro\_F1\_train: 0.19501257049670445

micro\_F1\_val: 0.20006341154090043

Epoch 00058: val\_accuracy improved from 0.19943 to 0.20006, saving model to model3\_save/weights-58-0.2001.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0112 - accuracy: 0.1605 - val\_loss: 2.4724 - val\_accuracy: 0.2001 - lr: 0.0100  
Epoch 59/100  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0111 - accuracy: 0.1575

micro\_F1\_train: 0.1960317999592308

micro\_F1\_val: 0.2010145846544071

Epoch 00059: val\_accuracy improved from 0.20006 to 0.20101, saving model to model3\_save/weights-59-0.2010.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0111 - accuracy: 0.1576 - val\_loss: 2.4712 - val\_accuracy: 0.2010 - lr: 0.0100  
Epoch 60/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0112 - accuracy: 0.1561

micro\_F1\_train: 0.1960317999592308

micro\_F1\_val: 0.20085605580215599

Epoch 00060: val\_accuracy did not improve from 0.20101  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0112 - accuracy: 0.1562 - val\_loss: 2.4704 - val\_accuracy: 0.2009 - lr: 0.0100  
Epoch 61/100  
1461/1472 [=====>.] - ETA: 0s - loss: 0.0110 - accuracy: 0.1613

micro\_F1\_train: 0.19657538900591154

micro\_F1\_val: 0.20117311350665817

Epoch 00061: val\_accuracy improved from 0.20101 to 0.20117, saving model to model3\_save/weights-61-0.2012.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0110 - accuracy: 0.1612 - val\_loss: 2.4693 - val\_accuracy: 0.2012 - lr: 0.0100  
Epoch 62/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0110 - accuracy: 0.1652

micro\_F1\_train: 0.19623564585173608

micro\_F1\_val: 0.2010145846544071

Epoch 00062: val\_accuracy did not improve from 0.20117  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0110 - accuracy: 0.1653 - val\_loss: 2.4677 - val\_accuracy: 0.2010 - lr: 0.0100  
Epoch 63/100  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0110 - accuracy: 0.1587

micro\_F1\_train: 0.19589590269756063

micro\_F1\_val: 0.20196575776791376

Epoch 00063: val\_accuracy improved from 0.20117 to 0.20197, saving model to model3\_save/weights-63-0.2020.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0110 - accuracy: 0.1589 - val\_loss:

2.4663 - val\_accuracy: 0.2020 - lr: 0.0100  
Epoch 00063: early stopping  
Time Taken for training (sec): 629.2495293617249

In [ ]:

```
# http://localhost:6006/  
%load_ext tensorboard  
%tensorboard --logdir logs --host localhost
```

The tensorboard extension is already loaded. To reload it, use:  
%reload\_ext tensorboard

Reusing TensorBoard on port 6006 (pid 2268), started 1:51:16 ago. (Use '!kill 2268' to kill it.)

### Model 3 Predictions

In [ ]:

```
NeuNet_3.load_weights("/content/m3_weights-63-0.2020.hdf5")
```

In [ ]:

```
y_pred_3 = NeuNet_3.predict(px_data_std)
```

In [ ]:

```
np.argmax(y_pred_3,axis=1)
```

Out[ ]:

```
array([10, 11, 11, ..., 14, 14, 14])
```

In [ ]:

```
y_pred_pd_3 = pd.DataFrame(np.argmax(y_pred_3,axis=1),index=px_data.index)  
# y_pred_pd[0] = y_pred_pd[0]+1
```

In [ ]:

```
# y_pred_pd_3
```

In [ ]:

```
y_pred_pd_3.value_counts()
```

Out[ ]:

```
14    4500  
11    3797  
10    2370  
15    2099  
6     1585  
12    1151  
9     1028  
16     882  
13     778  
1      577  
7      532  
4      504  
8      462  
3      435  
2      263  
5         0
```

```
5         62
dtype: int64
```

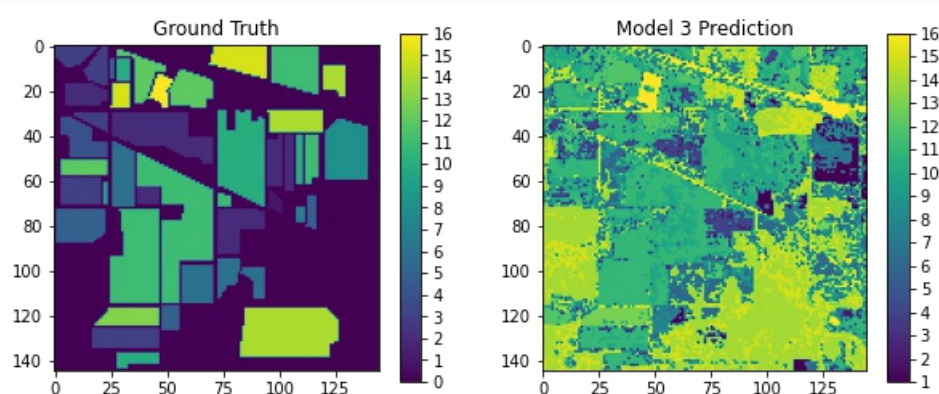
```
In [ ]:
```

```
# px_data.index
```

```
In [ ]:
```

```
figr,axis = plt.subplots(1,2,figsize=(10,10))
im0 = axis[0].imshow(mat_gt['indian_pines_gt'])#,cmap='jet')
axis[0].set_title('Ground Truth')
plt.colorbar(im0,ax=axis[0],shrink=0.4,aspect=16, ticks=range(0,17,1))

im1 = axis[1].imshow(y_pred_pd_3.to_numpy().reshape((145,145)))#,cmap='jet')
axis[1].set_title('Model 3 Prediction')
plt.colorbar(im1,ax=axis[1],shrink=0.4,aspect=16, ticks=range(0,17,1))
plt.savefig('NeuNet_3_e100.png')
plt.show()
```



### Observation

```
In [ ]:
```

```
y_pred_3_test = NeuNet_3.predict(X_test_std)
y_pred_pd_3_test = pd.DataFrame(np.argmax(y_pred_3_test,axis=1),index=y_test.index)
```

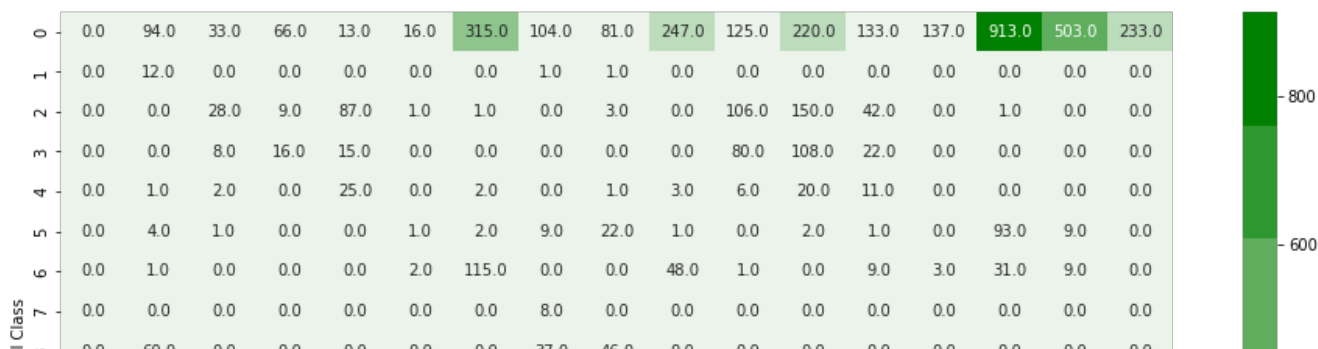
```
In [ ]:
```

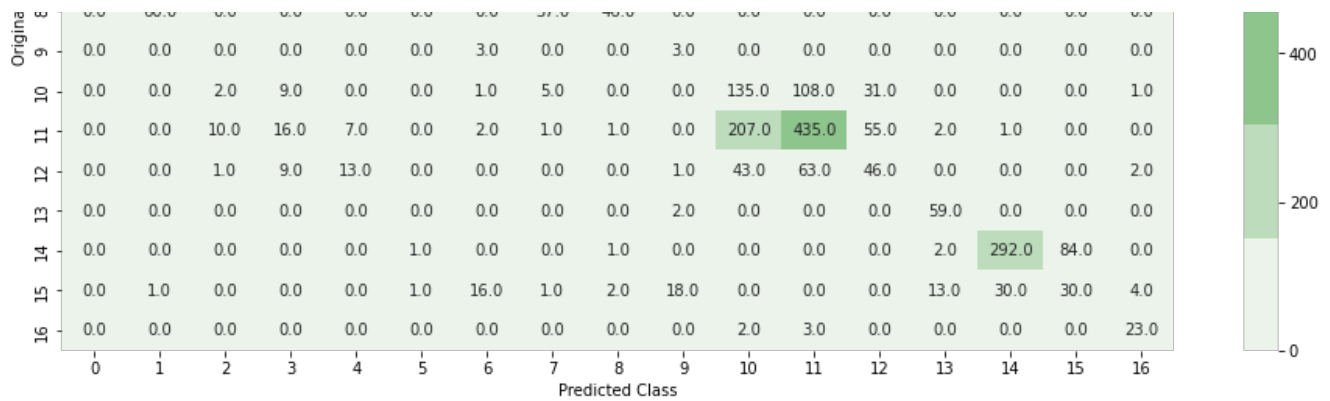
```
# plot_confusion_matrix(px_class,y_pred_pd_3)
plot_confusion_matrix(y_test,y_pred_pd_3_test)
```

Number of misclassified points 79.80342422320862

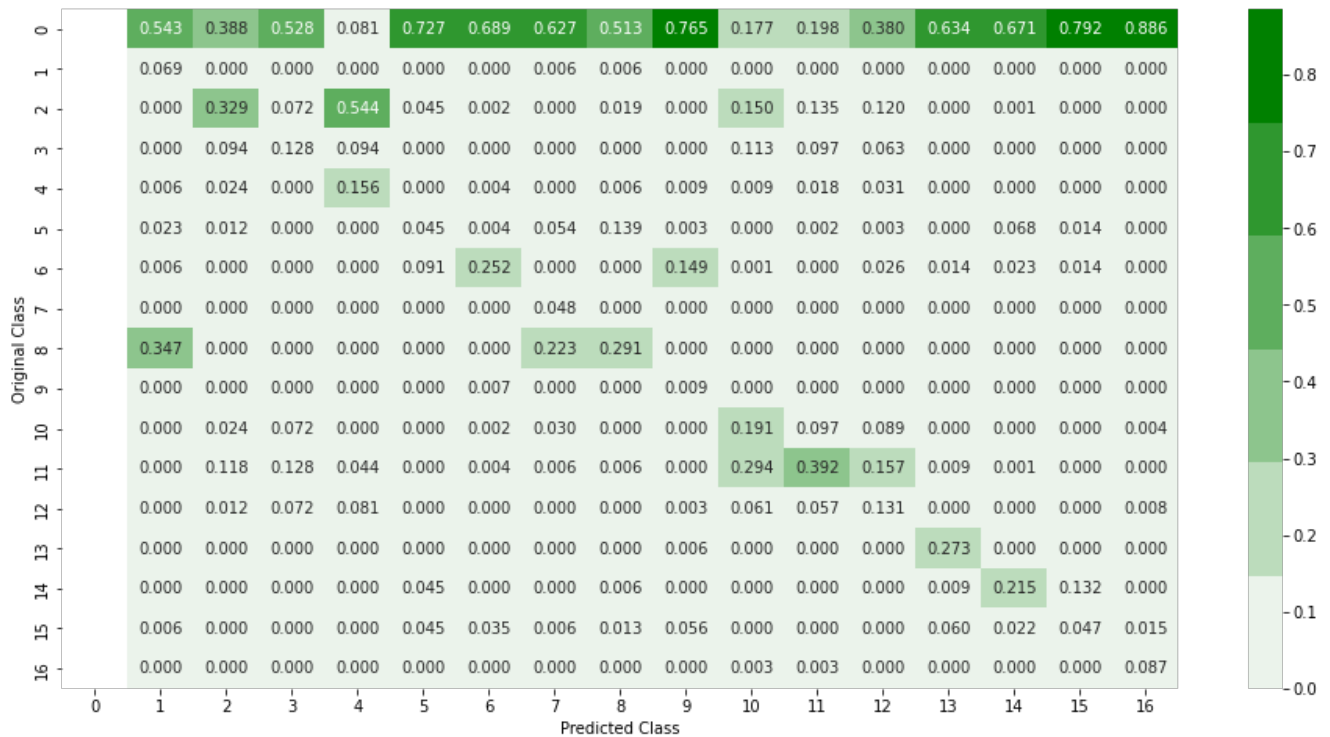
----- Confusion matrix -----  
-----

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:23: RuntimeWarning: invalid value encountered in true\_divide



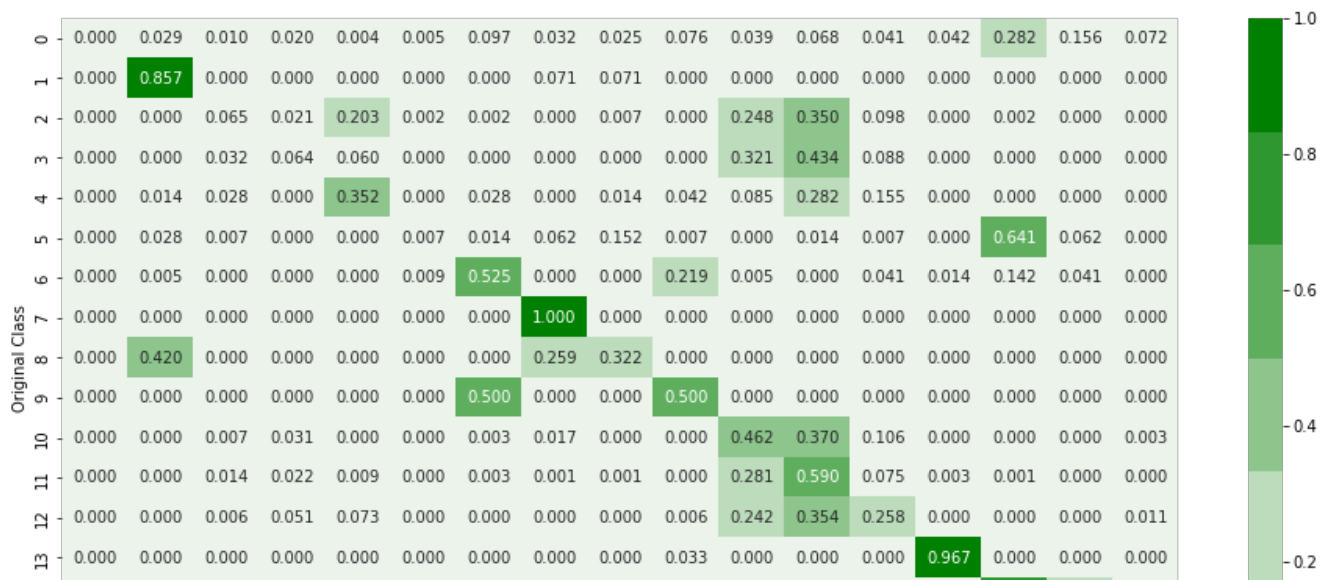


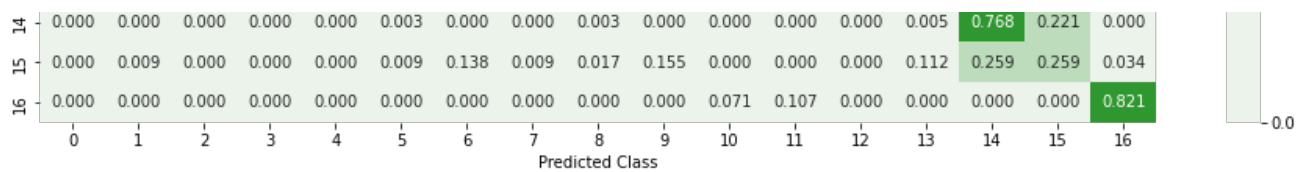
Precision matrix



Sum of columns in precision matrix [nan 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

Recall matrix





Sum of rows in recall matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

Model prediction doesnt match ground truth very well. Similar to Model 1 we can see that in recall matrix have only 2 of the classes have been predicted with high recall. Precisions are all low for all classes except for class 'zero'.

Model is not able to predict regions of class 'zero'. pixels of class 'zero' are classified into different class

#### 4- Four layered network with tanh activation and Adam optimizer

##### Model 4 Definition

In [ ]:

```
# del NeuNet_4
```

In [ ]:

```
In = Input(shape=(95,))
L1 = Dense(128,activation='tanh',
          kernel_initializer=initializers.he_uniform())(In) #https://keras.io/api/layers/initializers/

L1_Drp = Dropout(0.4)(L1)

L2 = Dense(64,activation='tanh',
          kernel_initializer=initializers.he_uniform())(L1_Drp)

L2_Drp = Dropout(0.3)(L2)

L3 = Dense(32,activation='tanh',
          kernel_initializer=initializers.he_uniform())(L2_Drp)

L4 = Dense(20,activation='tanh',
          kernel_initializer=initializers.he_uniform())(L3)

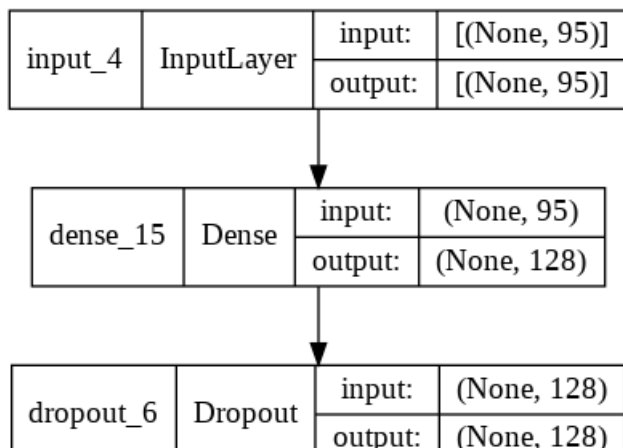
Out = Dense(17,activation='softmax',
          kernel_initializer=initializers.GlorotUniform())(L4)

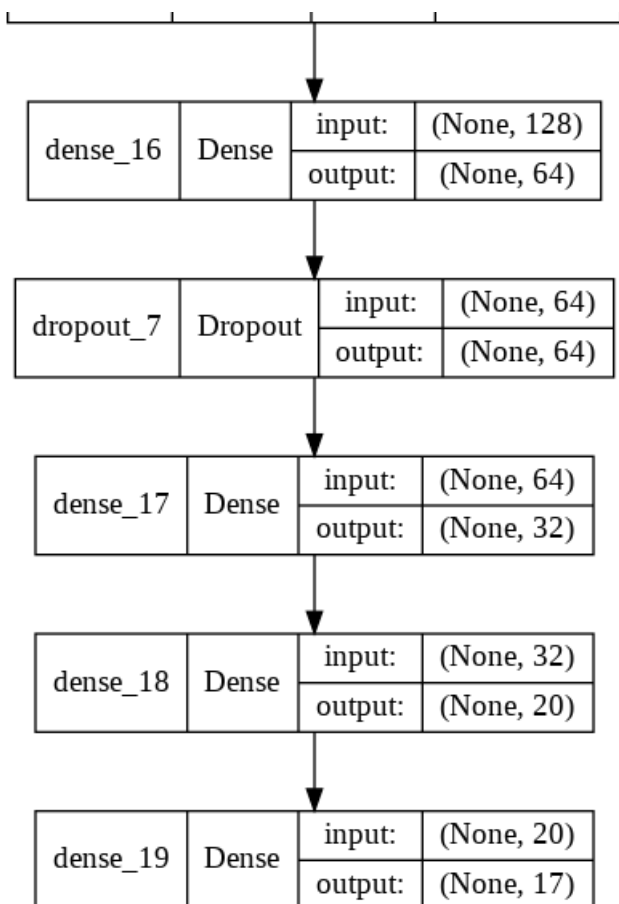
NeuNet_4 = Model(inputs=In, outputs = Out)

# cce = categorical_crossentropy()
NeuNet_4.compile(loss='categorical_crossentropy', optimizer='Adam', metrics=['accuracy'])

plot_model(NeuNet_4,show_layer_names=True, show_shapes=True)#,to_file='NeuNet_4.png')
```

Out[ ]:





In [ ]:

```
NeuNet_4.summary()
```

Model: "model\_3"

Layer (type)	Output Shape	Param #
input_4 (InputLayer)	[ (None, 95) ]	0
dense_15 (Dense)	(None, 128)	12288
dropout_6 (Dropout)	(None, 128)	0
dense_16 (Dense)	(None, 64)	8256
dropout_7 (Dropout)	(None, 64)	0
dense_17 (Dense)	(None, 32)	2080
dense_18 (Dense)	(None, 20)	660
dense_19 (Dense)	(None, 17)	357

```

Total params: 23,641
Trainable params: 23,641
Non-trainable params: 0

```

### Model 4 Training

In [ ]:

```

logdir = os.path.join("logs", datetime.now().strftime("%Y%m%d-%H%M%S"))
print(logdir)
file_writer = tf.summary.create_file_writer(logdir + "/metrics")
# tensorboard = TensorBoard(log_dir=logdir)

```

```

tensorboard = TensorBoard(log_dir=logdir, histogram_freq=1, write_graph=True, write_grads=True)

metric_calc_4 = metric_calc()

# Saving model at every epoch if validation accuracy is improved from previous epoch
filepath_m4="model_4_save/weights-{epoch:02d}-{val_accuracy:.4f}.hdf5"
checkpoint_m4 = ModelCheckpoint(filepath=filepath_m4, monitor='val_accuracy', verbose=1,
                                save_best_only=True,
                                mode='auto')
earlystop_m4 = EarlyStopping(monitor='val_accuracy', min_delta=0.01, patience=20, verbose=1)
reduce_lr_m4 = ReduceLROnPlateau(monitor='val_loss', factor=0.9, patience=2, min_lr=0.001)

callback_list_4 = [metric_calc_4,
                   checkpoint_m4,
                   earlystop_m4,
                   reduce_lr_m4,
                   tensorboard]

# fit network
verbose_4, epochs_4, batch_size_4 = 1, 100, 10
start = time.time()
history_4 = NeuNet_4.fit(X_train_std, y_ctg_train,
                        class_weight = class_wts,
                        validation_data=(X_test_std,y_ctg_test),
                        epochs=epochs_4,
                        batch_size=batch_size_4,
                        verbose=verbose_4,
                        callbacks=callback_list_4)

# Passing the F1 score data of each epoch to tensor board
# https://stackoverflow.com/questions/58102016/how-visualize-in-tensorboard-a-metric-callback
for i in range(len(metric_calc_4.metrics['micro_F1_train'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_train', metric_calc_4.metrics['micro_F1_train'][i])
file_writer.flush ()

for i in range(len(metric_calc_4.metrics['micro_F1_val'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_val', metric_calc_4.metrics['micro_F1_val'][i])
file_writer.flush ()

stop = time.time()
print('Time Taken for training (sec): ',stop-start)

```

logs/20220131-132844

WARNING:tensorflow:`write\_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.

Epoch 1/100

2/1472 [.....] - ETA: 5:34 - loss: 0.0208 - accuracy: 0.0500 WARNING:tensorflow:Callback method `on\_train\_batch\_begin` is slow compared to the batch time (batch time: 0.0036s vs `on\_train\_batch\_begin` time: 0.0220s). Check your callbacks.

WARNING:tensorflow:Callback method `on\_train\_batch\_end` is slow compared to the batch time (batch time: 0.0036s vs `on\_train\_batch\_end` time: 0.0182s). Check your callbacks.

1461/1472 [=====>.] - ETA: 0s - loss: 0.0115 - accuracy: 0.1502

micro\_F1\_train: 0.21845484813481006

micro\_F1\_val: 0.21623335447051364

Epoch 00001: val\_accuracy improved from -inf to 0.21623, saving model to model\_4\_save/weights-01-0.2162.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0116 - accuracy: 0.1503 - val\_loss: 2.3319 - val\_accuracy: 0.2162 - lr: 0.0010

Epoch 2/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0093 - accuracy: 0.1753

micro\_F1\_train: 0.1535639056873004

micro\_F1\_val: 0.1548826886493342

Epoch 00002: val\_accuracy did not improve from 0.21623

1472/1472 [=====] - 11s 8ms/step - loss: 0.0093 - accuracy: 0.1750 - val\_loss: 2.3473 - val\_accuracy: 0.1549 - lr: 0.0010

Epoch 3/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0084 - accuracy: 0.1810

micro\_F1\_train: 0.22531765984915403

micro\_F1\_val: 0.22114774889029803

Epoch 00003: val\_accuracy improved from 0.21623 to 0.22115, saving model to model\_4\_save/weights-03-0.2211.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0084 - accuracy: 0.1809 - val\_loss: 2.2647 - val\_accuracy: 0.2211 - lr: 0.0010

Epoch 4/100

1461/1472 [=====>.] - ETA: 0s - loss: 0.0077 - accuracy: 0.2034

micro\_F1\_train: 0.26160222871509137

micro\_F1\_val: 0.25586556753329104

Epoch 00004: val\_accuracy improved from 0.22115 to 0.25587, saving model to model\_4\_save/weights-04-0.2559.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0077 - accuracy: 0.2030 - val\_loss: 2.1440 - val\_accuracy: 0.2559 - lr: 0.0010

Epoch 5/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0075 - accuracy: 0.2177

micro\_F1\_train: 0.2238227899707821

micro\_F1\_val: 0.22352568167406467

Epoch 00005: val\_accuracy did not improve from 0.25587

1472/1472 [=====] - 10s 7ms/step - loss: 0.0075 - accuracy: 0.2177 - val\_loss: 2.0571 - val\_accuracy: 0.2235 - lr: 0.0010

Epoch 6/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0068 - accuracy: 0.2512

micro\_F1\_train: 0.2877624515866005

micro\_F1\_val: 0.2880469245402663

Epoch 00006: val\_accuracy improved from 0.25587 to 0.28805, saving model to model\_4\_save/weights-06-0.2880.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0068 - accuracy: 0.2511 - val\_loss: 2.0150 - val\_accuracy: 0.2880 - lr: 0.0010

Epoch 7/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0066 - accuracy: 0.2596

micro\_F1\_train: 0.28932527009580755

micro\_F1\_val: 0.2923272035510463

Epoch 00007: val\_accuracy improved from 0.28805 to 0.29233, saving model to model\_4\_save/weights-07-0.2923.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0066 - accuracy: 0.2598 - val\_loss: 1.9131 - val\_accuracy: 0.2923 - lr: 0.0010

Epoch 8/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0064 - accuracy: 0.2734

micro\_F1\_train: 0.31779574641570973

micro\_F1\_val: 0.31467977171845274

Epoch 00008: val\_accuracy improved from 0.29233 to 0.31468, saving model to model\_4\_save/weights-08-0.3147.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0064 - accuracy: 0.2736 - val\_loss: 1.9168 - val\_accuracy: 0.3147 - lr: 0.0010

Epoch 9/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0061 - accuracy: 0.2843

micro\_F1\_train: 0.32316368825168174

micro\_F1\_val: 0.31642358909321494

Epoch 00009: val\_accuracy improved from 0.31468 to 0.31642, saving model to model\_4\_save/weights-09-0.3164.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0061 - accuracy: 0.2842 - val\_loss: 1.8760 - val\_accuracy: 0.3164 - lr: 0.0010

Epoch 10/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0060 - accuracy: 0.2991

micro\_F1\_train: 0.30875857851464295

micro\_F1\_val: 0.3004121750158529



micro\_F1\_val: 0.300112170010029

Epoch 00010: val\_accuracy did not improve from 0.31642

1472/1472 [=====] - 10s 7ms/step - loss: 0.0060 - accuracy: 0.2991 - val\_loss: 1.8970 - val\_accuracy: 0.3004 - lr: 0.0010

Epoch 11/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0058 - accuracy: 0.3052

micro\_F1\_train: 0.3207175375416185

micro\_F1\_val: 0.325142675967026

Epoch 00011: val\_accuracy improved from 0.31642 to 0.32514, saving model to model\_4\_save/weights-11-0.3251.hdf5

1472/1472 [=====] - 13s 9ms/step - loss: 0.0058 - accuracy: 0.3055 - val\_loss: 1.8485 - val\_accuracy: 0.3251 - lr: 0.0010

Epoch 12/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0055 - accuracy: 0.3170

micro\_F1\_train: 0.30739960589794113

micro\_F1\_val: 0.30675332910589725

Epoch 00012: val\_accuracy did not improve from 0.32514

1472/1472 [=====] - 10s 7ms/step - loss: 0.0055 - accuracy: 0.3171 - val\_loss: 1.8111 - val\_accuracy: 0.3068 - lr: 0.0010

Epoch 13/100

1467/1472 [=====>.] - ETA: 0s - loss: 0.0056 - accuracy: 0.3202

micro\_F1\_train: 0.3555072365291839

micro\_F1\_val: 0.3557387444514902

Epoch 00013: val\_accuracy improved from 0.32514 to 0.35574, saving model to model\_4\_save/weights-13-0.3557.hdf5

1472/1472 [=====] - 11s 7ms/step - loss: 0.0056 - accuracy: 0.3208 - val\_loss: 1.7490 - val\_accuracy: 0.3557 - lr: 0.0010

Epoch 14/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0054 - accuracy: 0.3300

micro\_F1\_train: 0.3377046952503907

micro\_F1\_val: 0.3313253012048193

Epoch 00014: val\_accuracy did not improve from 0.35574

1472/1472 [=====] - 9s 6ms/step - loss: 0.0054 - accuracy: 0.3300 - val\_loss: 1.7275 - val\_accuracy: 0.3313 - lr: 0.0010

Epoch 15/100

1460/1472 [=====>.] - ETA: 0s - loss: 0.0052 - accuracy: 0.3276

micro\_F1\_train: 0.388530271115037

micro\_F1\_val: 0.38031071655041215

Epoch 00015: val\_accuracy improved from 0.35574 to 0.38031, saving model to model\_4\_save/weights-15-0.3803.hdf5

1472/1472 [=====] - 9s 6ms/step - loss: 0.0052 - accuracy: 0.3276 - val\_loss: 1.6355 - val\_accuracy: 0.3803 - lr: 0.0010

Epoch 16/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0052 - accuracy: 0.3331

micro\_F1\_train: 0.38601617177413877

micro\_F1\_val: 0.3769816106531389

Epoch 00016: val\_accuracy did not improve from 0.38031

1472/1472 [=====] - 11s 7ms/step - loss: 0.0052 - accuracy: 0.3337 - val\_loss: 1.7056 - val\_accuracy: 0.3770 - lr: 0.0010

Epoch 17/100

1461/1472 [=====>.] - ETA: 0s - loss: 0.0050 - accuracy: 0.3470

micro\_F1\_train: 0.38791873343752126

micro\_F1\_val: 0.3815789473684211

Epoch 00017: val\_accuracy improved from 0.38031 to 0.38158, saving model to model\_4\_save/weights-17-0.3816.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0050 - accuracy: 0.3469 - val\_loss:

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1472/1472 [=====] - 10s 7ms/step - loss: 0.0051 - accuracy: 0.3537 - val_loss: 1.6821 - val_accuracy: 0.3816 - lr: 0.0010
Epoch 18/100
1467/1472 [=====>.] - ETA: 0s - loss: 0.0051 - accuracy: 0.3537

micro_F1_train: 0.3788136169056193

micro_F1_val: 0.36794546607482564

Epoch 00018: val_accuracy did not improve from 0.38158
1472/1472 [=====] - 10s 7ms/step - loss: 0.0051 - accuracy: 0.3537 - val_loss: 1.7052 - val_accuracy: 0.3679 - lr: 0.0010
Epoch 19/100
1460/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3551

micro_F1_train: 0.39247129170347217

micro_F1_val: 0.3787254280279011

Epoch 00019: val_accuracy did not improve from 0.38158
1472/1472 [=====] - 10s 7ms/step - loss: 0.0048 - accuracy: 0.3552 - val_loss: 1.6593 - val_accuracy: 0.3787 - lr: 0.0010
Epoch 20/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3702

micro_F1_train: 0.3779982333355983

micro_F1_val: 0.3723842739378567

Epoch 00020: val_accuracy did not improve from 0.38158
1472/1472 [=====] - 9s 6ms/step - loss: 0.0048 - accuracy: 0.3703 - val_loss: 1.6900 - val_accuracy: 0.3724 - lr: 0.0010
Epoch 21/100
1472/1472 [=====] - ETA: 0s - loss: 0.0048 - accuracy: 0.3723

micro_F1_train: 0.4265135557518516

micro_F1_val: 0.40710209258084973

Epoch 00021: val_accuracy improved from 0.38158 to 0.40710, saving model to model_4_save/weights-21-0.4071.hdf5
1472/1472 [=====] - 9s 6ms/step - loss: 0.0048 - accuracy: 0.3723 - val_loss: 1.6024 - val_accuracy: 0.4071 - lr: 0.0010
Epoch 22/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0048 - accuracy: 0.3846

micro_F1_train: 0.4320173948494938

micro_F1_val: 0.42057704502219406

Epoch 00022: val_accuracy improved from 0.40710 to 0.42058, saving model to model_4_save/weights-22-0.4206.hdf5
1472/1472 [=====] - 11s 7ms/step - loss: 0.0048 - accuracy: 0.3849 - val_loss: 1.6121 - val_accuracy: 0.4206 - lr: 0.0010
Epoch 23/100
1461/1472 [=====>.] - ETA: 0s - loss: 0.0047 - accuracy: 0.3888

micro_F1_train: 0.441394305904736

micro_F1_val: 0.430564362714014

Epoch 00023: val_accuracy improved from 0.42058 to 0.43056, saving model to model_4_save/weights-23-0.4306.hdf5
1472/1472 [=====] - 11s 7ms/step - loss: 0.0048 - accuracy: 0.3887 - val_loss: 1.5678 - val_accuracy: 0.4306 - lr: 0.0010
Epoch 24/100
1462/1472 [=====>.] - ETA: 0s - loss: 0.0046 - accuracy: 0.3837

micro_F1_train: 0.43636610722293945

micro_F1_val: 0.4245402663284718

Epoch 00024: val_accuracy did not improve from 0.43056
1472/1472 [=====] - 10s 7ms/step - loss: 0.0046 - accuracy: 0.3847 - val_loss: 1.6017 - val_accuracy: 0.4245 - lr: 0.0010
Epoch 25/100
1468/1472 [=====>.] - ETA: 0s - loss: 0.0046 - accuracy: 0.3829
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micro\_F1\_train: 0.4169327988041041

micro\_F1\_val: 0.4039315155358275

Epoch 00025: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 9s 6ms/step - loss: 0.0046 - accuracy: 0.3829 - val\_loss: 1.5791 - val\_accuracy: 0.4039 - lr: 0.0010

Epoch 26/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0046 - accuracy: 0.3949

micro\_F1\_train: 0.422980226948427

micro\_F1\_val: 0.40868738110336084

Epoch 00026: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 10s 7ms/step - loss: 0.0046 - accuracy: 0.3953 - val\_loss: 1.5544 - val\_accuracy: 0.4087 - lr: 0.0010

Epoch 27/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.4034

micro\_F1\_train: 0.3976353876469389

micro\_F1\_val: 0.3869689283449588

Epoch 00027: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 9s 6ms/step - loss: 0.0045 - accuracy: 0.4032 - val\_loss: 1.6266 - val\_accuracy: 0.3870 - lr: 0.0010

Epoch 28/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.4008

micro\_F1\_train: 0.4361622613304342

micro\_F1\_val: 0.4237476220672162

Epoch 00028: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 10s 7ms/step - loss: 0.0044 - accuracy: 0.4005 - val\_loss: 1.5835 - val\_accuracy: 0.4237 - lr: 0.0010

Epoch 29/100

1467/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.4039

micro\_F1\_train: 0.430658422232792

micro\_F1\_val: 0.4239061509194673

Epoch 00029: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 10s 7ms/step - loss: 0.0045 - accuracy: 0.4042 - val\_loss: 1.5951 - val\_accuracy: 0.4239 - lr: 0.0010

Epoch 30/100

1471/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3949

micro\_F1\_train: 0.40144051097370387

micro\_F1\_val: 0.39283449587824987

Epoch 00030: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 11s 7ms/step - loss: 0.0045 - accuracy: 0.3948 - val\_loss: 1.6342 - val\_accuracy: 0.3928 - lr: 0.0010

Epoch 31/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.3911

micro\_F1\_train: 0.4216892029625603

micro\_F1\_val: 0.4120164870006341

Epoch 00031: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 9s 6ms/step - loss: 0.0044 - accuracy: 0.3914 - val\_loss: 1.5577 - val\_accuracy: 0.4120 - lr: 0.0010

Epoch 32/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0043 - accuracy: 0.3991

micro\_F1\_train: 0.43813277162465175

micro\_F1\_val: 0.41915028535193405

Epoch 00032: val\_accuracy did not improve from 0.43056

1472/1472 [=====] - 11s 7ms/step - loss: 0.0043 - accuracy: 0.3994 - val\_loss: 1.5430 - val\_accuracy: 0.4193 - lr: 0.0010

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1.5450 - val_accuracy: 0.4192 - lr: 0.0010
Epoch 33/100
1462/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.4013

micro_F1_train: 0.41951484677583745

micro_F1_val: 0.40076093849080535

Epoch 00033: val_accuracy did not improve from 0.43056
1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.4016 - val_loss:
1.6508 - val_accuracy: 0.4008 - lr: 0.0010
Epoch 34/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.4089

micro_F1_train: 0.4385404634096623

micro_F1_val: 0.4194673430564363

Epoch 00034: val_accuracy did not improve from 0.43056
1472/1472 [=====] - 10s 7ms/step - loss: 0.0044 - accuracy: 0.4088 - val_loss:
1.5422 - val_accuracy: 0.4195 - lr: 0.0010
Epoch 35/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0044 - accuracy: 0.4054

micro_F1_train: 0.432968675681185

micro_F1_val: 0.41708941027266955

Epoch 00035: val_accuracy did not improve from 0.43056
1472/1472 [=====] - 11s 7ms/step - loss: 0.0044 - accuracy: 0.4054 - val_loss:
1.5679 - val_accuracy: 0.4171 - lr: 0.0010
Epoch 36/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.4177

micro_F1_train: 0.42019433308418835

micro_F1_val: 0.40424857324032976

Epoch 00036: val_accuracy did not improve from 0.43056
1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.4179 - val_loss:
1.5954 - val_accuracy: 0.4042 - lr: 0.0010
Epoch 37/100
1472/1472 [=====] - ETA: 0s - loss: 0.0043 - accuracy: 0.4173

micro_F1_train: 0.4368417476387851

micro_F1_val: 0.4177235256816741

Epoch 00037: val_accuracy did not improve from 0.43056
1472/1472 [=====] - 10s 7ms/step - loss: 0.0043 - accuracy: 0.4173 - val_loss:
1.5475 - val_accuracy: 0.4177 - lr: 0.0010
Epoch 38/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4210

micro_F1_train: 0.45600326153428006

micro_F1_val: 0.4400760938490805

Epoch 00038: val_accuracy improved from 0.43056 to 0.44008, saving model to model_4_save/weights-38-0.4
401.hdf5
1472/1472 [=====] - 11s 7ms/step - loss: 0.0040 - accuracy: 0.4209 - val_loss:
1.5117 - val_accuracy: 0.4401 - lr: 0.0010
Epoch 39/100
1464/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4245

micro_F1_train: 0.45586736427260993

micro_F1_val: 0.43785668991756493

Epoch 00039: val_accuracy did not improve from 0.44008
1472/1472 [=====] - 10s 7ms/step - loss: 0.0040 - accuracy: 0.4250 - val_loss:
1.5081 - val_accuracy: 0.4379 - lr: 0.0010
Epoch 40/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.4169

micro_F1_train: 0.436026364068764

micro_F1_val: 0.41471147748800207
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micro\_F1\_val: 0.414/114//4889029/

Epoch 00040: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 11s 7ms/step - loss: 0.0041 - accuracy: 0.4170 - val\_loss: 1.5177 - val\_accuracy: 0.4147 - lr: 0.0010

Epoch 41/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0043 - accuracy: 0.4087

micro\_F1\_train: 0.4446558401848203

micro\_F1\_val: 0.4296131896005073

Epoch 00041: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 10s 7ms/step - loss: 0.0043 - accuracy: 0.4087 - val\_loss: 1.5088 - val\_accuracy: 0.4296 - lr: 0.0010

Epoch 42/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4224

micro\_F1\_train: 0.4373173880546308

micro\_F1\_val: 0.42136968928344964

Epoch 00042: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 10s 7ms/step - loss: 0.0040 - accuracy: 0.4224 - val\_loss: 1.5312 - val\_accuracy: 0.4214 - lr: 0.0010

Epoch 43/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.4089

micro\_F1\_train: 0.42549432628932526

micro\_F1\_val: 0.409004438807863

Epoch 00043: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 9s 6ms/step - loss: 0.0042 - accuracy: 0.4090 - val\_loss: 1.5749 - val\_accuracy: 0.4090 - lr: 0.0010

Epoch 44/100

1467/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.4153

micro\_F1\_train: 0.44492763470816066

micro\_F1\_val: 0.4266011414077362

Epoch 00044: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 10s 7ms/step - loss: 0.0041 - accuracy: 0.4152 - val\_loss: 1.5136 - val\_accuracy: 0.4266 - lr: 0.0010

Epoch 45/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.4219

micro\_F1\_train: 0.4427532785214378

micro\_F1\_val: 0.4281864299302473

Epoch 00045: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 9s 6ms/step - loss: 0.0041 - accuracy: 0.4219 - val\_loss: 1.5545 - val\_accuracy: 0.4282 - lr: 0.0010

Epoch 46/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4227

micro\_F1\_train: 0.43921994971801315

micro\_F1\_val: 0.42786937222574506

Epoch 00046: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 11s 7ms/step - loss: 0.0040 - accuracy: 0.4227 - val\_loss: 1.5323 - val\_accuracy: 0.4279 - lr: 0.0010

Epoch 47/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.4226

micro\_F1\_train: 0.44948019297411157

micro\_F1\_val: 0.4323081800887762

Epoch 00047: val\_accuracy did not improve from 0.44008

1472/1472 [=====] - 10s 7ms/step - loss: 0.0041 - accuracy: 0.4227 - val\_loss: 1.5054 - val\_accuracy: 0.4323 - lr: 0.0010

Epoch 48/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.4183

micro\_F1\_train: 0.4580417204593327

micro\_F1\_val: 0.44086873811033606

Epoch 00048: val\_accuracy improved from 0.44008 to 0.44087, saving model to model\_4\_save/weights-48-0.4409.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0041 - accuracy: 0.4184 - val\_loss: 1.4989 - val\_accuracy: 0.4409 - lr: 0.0010

Epoch 49/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4268

micro\_F1\_train: 0.45743018278181696

micro\_F1\_val: 0.442136968928345

Epoch 00049: val\_accuracy improved from 0.44087 to 0.44214, saving model to model\_4\_save/weights-49-0.4421.hdf5

1472/1472 [=====] - 10s 7ms/step - loss: 0.0040 - accuracy: 0.4268 - val\_loss: 1.4942 - val\_accuracy: 0.4421 - lr: 0.0010

Epoch 50/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4209

micro\_F1\_train: 0.45321736767004145

micro\_F1\_val: 0.43912492073557385

Epoch 00050: val\_accuracy did not improve from 0.44214

1472/1472 [=====] - 10s 7ms/step - loss: 0.0040 - accuracy: 0.4206 - val\_loss: 1.4868 - val\_accuracy: 0.4391 - lr: 0.0010

Epoch 51/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0039 - accuracy: 0.4280

micro\_F1\_train: 0.4455391723856764

micro\_F1\_val: 0.4315155358275206

Epoch 00051: val\_accuracy did not improve from 0.44214

1472/1472 [=====] - 10s 7ms/step - loss: 0.0039 - accuracy: 0.4285 - val\_loss: 1.5114 - val\_accuracy: 0.4315 - lr: 0.0010

Epoch 52/100

1467/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4224

micro\_F1\_train: 0.44846096351158526

micro\_F1\_val: 0.4340519974635384

Epoch 00052: val\_accuracy did not improve from 0.44214

1472/1472 [=====] - 9s 6ms/step - loss: 0.0038 - accuracy: 0.4224 - val\_loss: 1.5035 - val\_accuracy: 0.4341 - lr: 0.0010

Epoch 53/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4223

micro\_F1\_train: 0.454440443025073

micro\_F1\_val: 0.4331008243500317

Epoch 00053: val\_accuracy did not improve from 0.44214

1472/1472 [=====] - 11s 7ms/step - loss: 0.0040 - accuracy: 0.4221 - val\_loss: 1.5258 - val\_accuracy: 0.4331 - lr: 0.0010

Epoch 54/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4278

micro\_F1\_train: 0.42902765509274987

micro\_F1\_val: 0.4215282181357007

Epoch 00054: val\_accuracy did not improve from 0.44214

1472/1472 [=====] - 9s 6ms/step - loss: 0.0038 - accuracy: 0.4276 - val\_loss: 1.5114 - val\_accuracy: 0.4215 - lr: 0.0010

Epoch 55/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0039 - accuracy: 0.4280

micro\_F1\_train: 0.4456071210165115

micro\_F1\_val: 0.43088142041851624

Epoch 00055: val\_accuracy did not improve from 0.44214  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0039 - accuracy: 0.4281 - val\_loss: 1.5037 - val\_accuracy: 0.4309 - lr: 0.0010  
Epoch 56/100  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0039 - accuracy: 0.4360  
  
micro\_F1\_train: 0.4512468573758239  
  
micro\_F1\_val: 0.4334178820545339  
  
Epoch 00056: val\_accuracy did not improve from 0.44214  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0039 - accuracy: 0.4361 - val\_loss: 1.5401 - val\_accuracy: 0.4334 - lr: 0.0010  
Epoch 57/100  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0039 - accuracy: 0.4316  
  
micro\_F1\_train: 0.4708840116871645  
  
micro\_F1\_val: 0.45450221940393154  
  
Epoch 00057: val\_accuracy improved from 0.44214 to 0.45450, saving model to model\_4\_save/weights-57-0.4545.hdf5  
1472/1472 [=====] - 9s 6ms/step - loss: 0.0039 - accuracy: 0.4315 - val\_loss: 1.4401 - val\_accuracy: 0.4545 - lr: 0.0010  
Epoch 58/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4340  
  
micro\_F1\_train: 0.4342596996670517  
  
micro\_F1\_val: 0.4153455928979074  
  
Epoch 00058: val\_accuracy did not improve from 0.45450  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0038 - accuracy: 0.4339 - val\_loss: 1.4905 - val\_accuracy: 0.4153 - lr: 0.0010  
Epoch 59/100  
1472/1472 [=====] - ETA: 0s - loss: 0.0037 - accuracy: 0.4188  
  
micro\_F1\_train: 0.4500917306516274  
  
micro\_F1\_val: 0.4348446417247939  
  
Epoch 00059: val\_accuracy did not improve from 0.45450  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0037 - accuracy: 0.4188 - val\_loss: 1.5612 - val\_accuracy: 0.4348 - lr: 0.0010  
Epoch 60/100  
1472/1472 [=====] - ETA: 0s - loss: 0.0038 - accuracy: 0.4340  
  
micro\_F1\_train: 0.4509750628524835  
  
micro\_F1\_val: 0.435003170577045  
  
Epoch 00060: val\_accuracy did not improve from 0.45450  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0038 - accuracy: 0.4340 - val\_loss: 1.4701 - val\_accuracy: 0.4350 - lr: 0.0010  
Epoch 61/100  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4285  
  
micro\_F1\_train: 0.4546442889175783  
  
micro\_F1\_val: 0.439283449587825  
  
Epoch 00061: val\_accuracy did not improve from 0.45450  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0038 - accuracy: 0.4283 - val\_loss: 1.4457 - val\_accuracy: 0.4393 - lr: 0.0010  
Epoch 62/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0039 - accuracy: 0.4339  
  
micro\_F1\_train: 0.4431609703064483  
  
micro\_F1\_val: 0.43547875713379836  
  
Epoch 00062: val\_accuracy did not improve from 0.45450  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0039 - accuracy: 0.4341 - val\_loss: 1.5055 - val\_accuracy: 0.4355 - lr: 0.0010  
Epoch 63/100  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4310  
  
micro\_F1\_train: 0.4566166616661666  
  
micro\_F1\_val: 0.4566166616661666

micro\_F1\_train: 0.4563430046884555

micro\_F1\_val: 0.43912492073557385

Epoch 00063: val\_accuracy did not improve from 0.45450

1472/1472 [=====] - 11s 7ms/step - loss: 0.0038 - accuracy: 0.4312 - val\_loss: 1.4645 - val\_accuracy: 0.4391 - lr: 0.0010

Epoch 64/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4383

micro\_F1\_train: 0.46150710063192224

micro\_F1\_val: 0.4473684210526316

Epoch 00064: val\_accuracy did not improve from 0.45450

1472/1472 [=====] - 9s 6ms/step - loss: 0.0038 - accuracy: 0.4385 - val\_loss: 1.4640 - val\_accuracy: 0.4474 - lr: 0.0010

Epoch 65/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4364

micro\_F1\_train: 0.46103146021607666

micro\_F1\_val: 0.44720989220038043

Epoch 00065: val\_accuracy did not improve from 0.45450

1472/1472 [=====] - 10s 7ms/step - loss: 0.0037 - accuracy: 0.4362 - val\_loss: 1.4423 - val\_accuracy: 0.4472 - lr: 0.0010

Epoch 66/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4460

micro\_F1\_train: 0.4481212203574098

micro\_F1\_val: 0.43864933417882057

Epoch 00066: val\_accuracy did not improve from 0.45450

1472/1472 [=====] - 11s 8ms/step - loss: 0.0038 - accuracy: 0.4458 - val\_loss: 1.4839 - val\_accuracy: 0.4386 - lr: 0.0010

Epoch 67/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4343

micro\_F1\_train: 0.4634776109261398

micro\_F1\_val: 0.4495878249841471

Epoch 00067: val\_accuracy did not improve from 0.45450

1472/1472 [=====] - 10s 7ms/step - loss: 0.0038 - accuracy: 0.4344 - val\_loss: 1.4611 - val\_accuracy: 0.4496 - lr: 0.0010

Epoch 68/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4372

micro\_F1\_train: 0.4476455799415642

micro\_F1\_val: 0.4342105263157895

Epoch 00068: val\_accuracy did not improve from 0.45450

1472/1472 [=====] - 10s 6ms/step - loss: 0.0038 - accuracy: 0.4372 - val\_loss: 1.4917 - val\_accuracy: 0.4342 - lr: 0.0010

Epoch 69/100

1461/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4300

micro\_F1\_train: 0.4708840116871645

micro\_F1\_val: 0.45878249841471147

Epoch 00069: val\_accuracy improved from 0.45450 to 0.45878, saving model to model\_4\_save/weights-69-0.4588.hdf5

1472/1472 [=====] - 11s 8ms/step - loss: 0.0037 - accuracy: 0.4296 - val\_loss: 1.4593 - val\_accuracy: 0.4588 - lr: 0.0010

Epoch 70/100

1462/1472 [=====>.] - ETA: 0s - loss: 0.0039 - accuracy: 0.4392

micro\_F1\_train: 0.45790582319766254

micro\_F1\_val: 0.44150285351934054

Epoch 00070: val\_accuracy did not improve from 0.45878

1472/1472 [=====] - 10s 7ms/step - loss: 0.0039 - accuracy: 0.4392 - val\_loss:



```
1.4817 - val_accuracy: 0.4415 - lr: 0.0010
Epoch 71/100
1471/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4403

micro_F1_train: 0.4861045049942244

micro_F1_val: 0.4722574508560558

Epoch 00071: val_accuracy improved from 0.45878 to 0.47226, saving model to model_4_save/weights-71-0.4723.hdf5
1472/1472 [=====] - 10s 7ms/step - loss: 0.0037 - accuracy: 0.4402 - val_loss: 1.4383 - val_accuracy: 0.4723 - lr: 0.0010
Epoch 72/100
1467/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4514

micro_F1_train: 0.469728884962968

micro_F1_val: 0.4499048826886493

Epoch 00072: val_accuracy did not improve from 0.47226
1472/1472 [=====] - 10s 7ms/step - loss: 0.0036 - accuracy: 0.4512 - val_loss: 1.4417 - val_accuracy: 0.4499 - lr: 0.0010
Epoch 73/100
1465/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4433

micro_F1_train: 0.4709519603179996

micro_F1_val: 0.45624603677869374

Epoch 00073: val_accuracy did not improve from 0.47226
1472/1472 [=====] - 10s 7ms/step - loss: 0.0037 - accuracy: 0.4431 - val_loss: 1.3985 - val_accuracy: 0.4562 - lr: 0.0010
Epoch 74/100
1463/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4361

micro_F1_train: 0.44975198749745193

micro_F1_val: 0.4300887761572606

Epoch 00074: val_accuracy did not improve from 0.47226
1472/1472 [=====] - 10s 6ms/step - loss: 0.0039 - accuracy: 0.4362 - val_loss: 1.4969 - val_accuracy: 0.4301 - lr: 0.0010
Epoch 75/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4422

micro_F1_train: 0.4634096622953047

micro_F1_val: 0.4506975269499049

Epoch 00075: val_accuracy did not improve from 0.47226
1472/1472 [=====] - 10s 7ms/step - loss: 0.0040 - accuracy: 0.4423 - val_loss: 1.4356 - val_accuracy: 0.4507 - lr: 0.0010
Epoch 76/100
1470/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4388

micro_F1_train: 0.4601481280152205

micro_F1_val: 0.44720989220038043

Epoch 00076: val_accuracy did not improve from 0.47226
1472/1472 [=====] - 10s 7ms/step - loss: 0.0038 - accuracy: 0.4388 - val_loss: 1.4803 - val_accuracy: 0.4472 - lr: 0.0010
Epoch 77/100
1466/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4520

micro_F1_train: 0.4732622137663926

micro_F1_val: 0.45973367152821815

Epoch 00077: val_accuracy did not improve from 0.47226
1472/1472 [=====] - 10s 7ms/step - loss: 0.0038 - accuracy: 0.4521 - val_loss: 1.4542 - val_accuracy: 0.4597 - lr: 0.0010
Epoch 78/100
1462/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4380

micro_F1_train: 0.4295032955085955
```

micro\_F1\_val: 0.41502853519340516

Epoch 00078: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 10s 6ms/step - loss: 0.0037 - accuracy: 0.4382 - val\_loss: 1.5368 - val\_accuracy: 0.4150 - lr: 0.0010

Epoch 79/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4396

micro\_F1\_train: 0.4713596521030101

micro\_F1\_val: 0.4576727964489537

Epoch 00079: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 10s 7ms/step - loss: 0.0037 - accuracy: 0.4396 - val\_loss: 1.4292 - val\_accuracy: 0.4577 - lr: 0.0010

Epoch 80/100

1465/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4475

micro\_F1\_train: 0.46565196711286266

micro\_F1\_val: 0.4513316423589093

Epoch 00080: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 10s 7ms/step - loss: 0.0037 - accuracy: 0.4472 - val\_loss: 1.3940 - val\_accuracy: 0.4513 - lr: 0.0010

Epoch 81/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4443

micro\_F1\_train: 0.46109940884691175

micro\_F1\_val: 0.4481610653138871

Epoch 00081: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 11s 8ms/step - loss: 0.0037 - accuracy: 0.4446 - val\_loss: 1.4451 - val\_accuracy: 0.4482 - lr: 0.0010

Epoch 82/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4437

micro\_F1\_train: 0.4787660528640348

micro\_F1\_val: 0.46163601775523144

Epoch 00082: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 11s 8ms/step - loss: 0.0038 - accuracy: 0.4438 - val\_loss: 1.4157 - val\_accuracy: 0.4616 - lr: 0.0010

Epoch 83/100

1469/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4462

micro\_F1\_train: 0.46021607664605557

micro\_F1\_val: 0.44705136334812934

Epoch 00083: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 11s 8ms/step - loss: 0.0036 - accuracy: 0.4461 - val\_loss: 1.4151 - val\_accuracy: 0.4471 - lr: 0.0010

Epoch 84/100

1468/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4281

micro\_F1\_train: 0.46334171366446963

micro\_F1\_val: 0.452282815472416

Epoch 00084: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 11s 7ms/step - loss: 0.0037 - accuracy: 0.4281 - val\_loss: 1.4445 - val\_accuracy: 0.4523 - lr: 0.0010

Epoch 85/100

1472/1472 [=====] - ETA: 0s - loss: 0.0037 - accuracy: 0.4399

micro\_F1\_train: 0.45240198410002036

micro\_F1\_val: 0.44356372859860493

Epoch 00085: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 10s 6ms/step - loss: 0.0037 - accuracy: 0.4399 - val\_loss: 1.4524 - val\_accuracy: 0.4436 - lr: 0.0010

Epoch 86/100

1461/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4420

micro\_F1\_train: 0.465923761636203

micro\_F1\_val: 0.452282815472416

Epoch 00086: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 11s 7ms/step - loss: 0.0038 - accuracy: 0.4419 - val\_loss: 1.4724 - val\_accuracy: 0.4523 - lr: 0.0010

Epoch 87/100

1466/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4447

micro\_F1\_train: 0.4564789019501257

micro\_F1\_val: 0.4464172479391249

Epoch 00087: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 10s 6ms/step - loss: 0.0037 - accuracy: 0.4451 - val\_loss: 1.4322 - val\_accuracy: 0.4464 - lr: 0.0010

Epoch 88/100

1472/1472 [=====] - ETA: 0s - loss: 0.0037 - accuracy: 0.4436

micro\_F1\_train: 0.4756404158456207

micro\_F1\_val: 0.46337983512999364

Epoch 00088: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 10s 6ms/step - loss: 0.0037 - accuracy: 0.4436 - val\_loss: 1.4167 - val\_accuracy: 0.4634 - lr: 0.0010

Epoch 89/100

1464/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4524

micro\_F1\_train: 0.4640891486036556

micro\_F1\_val: 0.44673430564362715

Epoch 00089: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 11s 8ms/step - loss: 0.0037 - accuracy: 0.4525 - val\_loss: 1.4537 - val\_accuracy: 0.4467 - lr: 0.0010

Epoch 90/100

1463/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4504

micro\_F1\_train: 0.4608955629544065

micro\_F1\_val: 0.4445149017121116

Epoch 00090: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 10s 6ms/step - loss: 0.0036 - accuracy: 0.4506 - val\_loss: 1.4686 - val\_accuracy: 0.4445 - lr: 0.0010

Epoch 91/100

1470/1472 [=====>.] - ETA: 0s - loss: 0.0035 - accuracy: 0.4473

micro\_F1\_train: 0.48624040225589454

micro\_F1\_val: 0.47178186429930247

Epoch 00091: val\_accuracy did not improve from 0.47226

1472/1472 [=====] - 11s 8ms/step - loss: 0.0035 - accuracy: 0.4473 - val\_loss: 1.4211 - val\_accuracy: 0.4718 - lr: 0.0010

Epoch 00091: early stopping

Time Taken for training (sec): 927.6565854549408

In [ ]:

```
# http://localhost:6006/  
%load_ext tensorboard  
%tensorboard --logdir logs --host localhost
```

The tensorboard extension is already loaded. To reload it, use:

```
%reload_ext tensorboard
```

Reusing TensorBoard on port 6006 (pid 2268), started 1:51:23 ago. (Use '!kill 2268' to kill it.)

## Model 4 Predictions

In [ ]:

```
NeuNet_4.load_weights("/content/m4_weights-71-0.4723.hdf5")
```

In [ ]:

```
y_pred_4 = NeuNet_4.predict(px_data_std)
```

In [ ]:

```
np.argmax(y_pred_4,axis=1)
```

Out[ ]:

```
array([ 3,  3,  3, ..., 14, 14, 14])
```

In [ ]:

```
y_pred_pd_4 = pd.DataFrame(np.argmax(y_pred_4,axis=1),index=px_data.index)  
# y_pred_pd[0] = y_pred_pd[0]+1
```

In [ ]:

```
# y_pred_pd_4
```

In [ ]:

```
y_pred_pd_4.value_counts()
```

Out[ ]:

```
14    3583  
0     2459  
15    2396  
11    2357  
2     1825  
10    1559  
6     1341  
5     1322  
12    1211  
3      829  
8      648  
4      601  
13     317  
16     248  
1      136  
7       98  
9       95  
dtype: int64
```

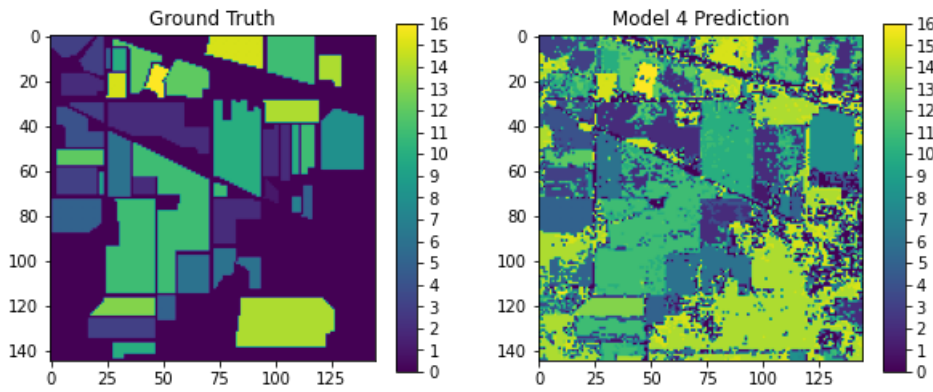
In [ ]:

```
# px_data.index
```

In [ ]:

```
figr,axis = plt.subplots(1,2,figsize=(10,10))  
im0 = axis[0].imshow(mat_gt['indian_pines_gt'])#,cmap='jet')  
axis[0].set_title('Ground Truth')  
plt.colorbar(im0,ax=axis[0],shrink=0.4,aspect=16, ticks=range(0,17,1))  
  
im1 = axis[1].imshow(y_pred_pd_4.to_numpy().reshape((145,145)))#,cmap='jet')  
axis[1].set_title('Model 4 Prediction')  
plt.colorbar(im1,ax=axis[1],shrink=0.4,aspect=16, ticks=range(0,17,1))
```

```
plt.savefig('NeuNet_4_e100.png')
plt.show()
```



### Observation:

Model prediction better compared to model 1 but fails to predict '0' class

In [ ]:

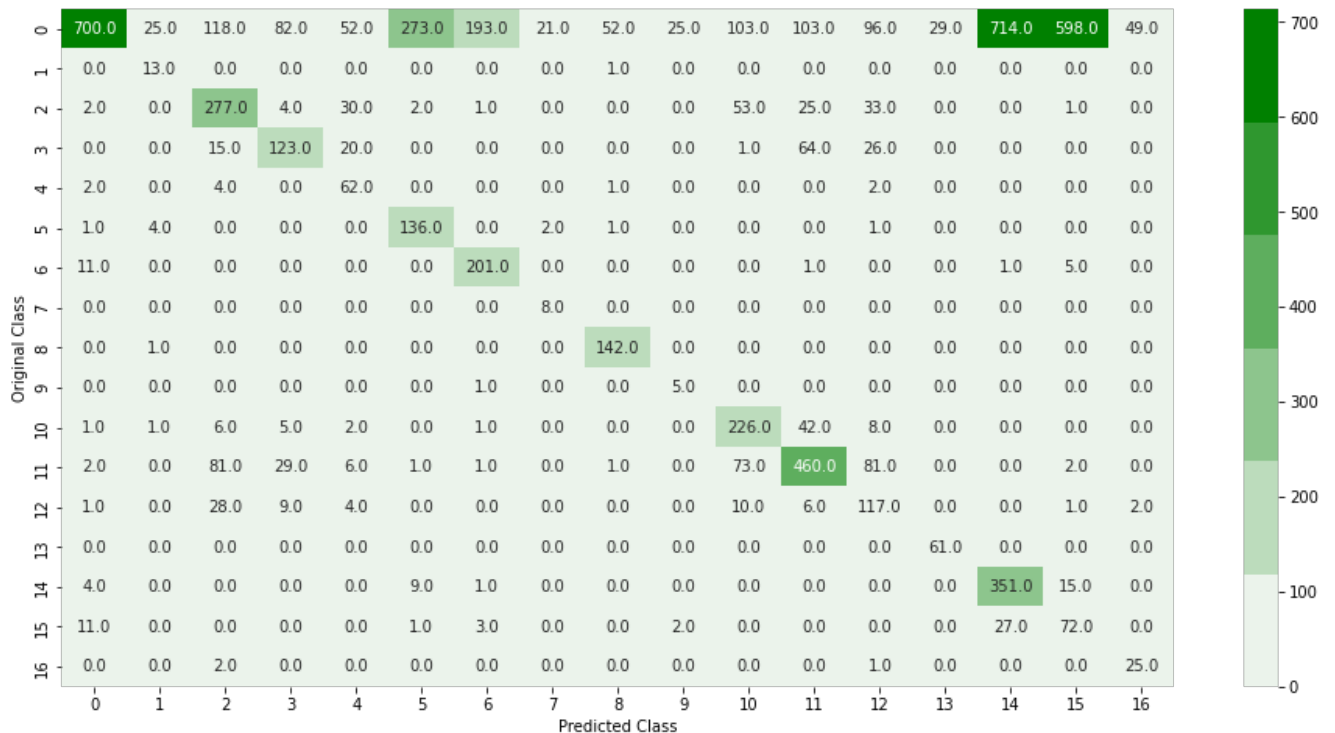
```
y_pred_4_test = NeuNet_4.predict(X_test_std)
y_pred_pd_4_test = pd.DataFrame(np.argmax(y_pred_4_test,axis=1),index=y_test.index)
```

In [ ]:

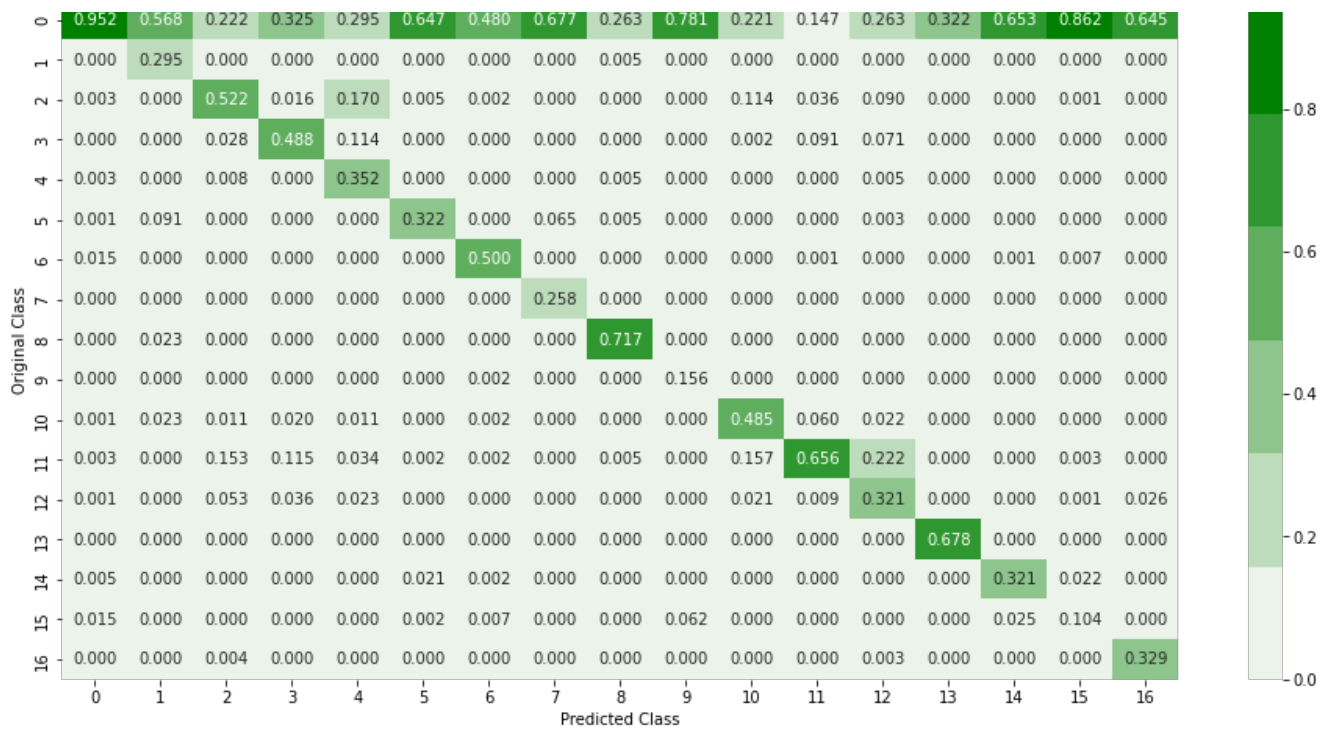
```
# plot_confusion_matrix(px_class,y_pred_pd_4)
plot_confusion_matrix(y_test,y_pred_pd_4_test)
```

Number of misclassified points 52.774254914394426

----- Confusion matrix -----  
-----

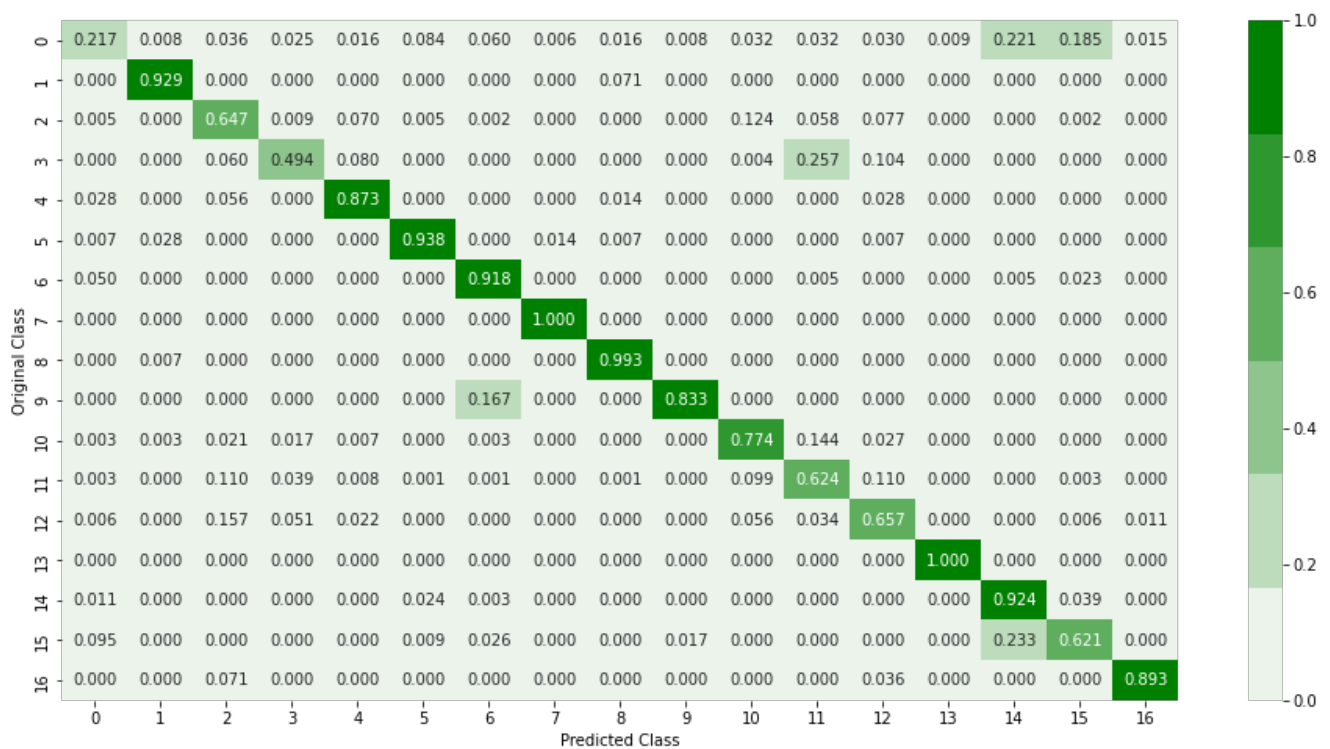


----- Precision matrix -----  
-----



Sum of columns in precision matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

Recall matrix



Sum of rows in recall matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

- From Recall matrix we can see that most of the classes have been predicted with high recall.
- Precisions are low for most classes except for class 'zero'.
- From precision matrix (first row), we observe that model finds there are similarity among class 'zero' and other classes.

## Retrain Model 2

Load saved model weights

## Retrain 1

In [ ]:

```
# NeuNet_2.load_weights("/content/m2_weights-65-0.4937.hdf5")
```

In [ ]:

```
# opt = optimizers.Adam(learning_rate=0.0009)
# NeuNet_2.compile(loss='categorical_crossentropy', optimizer = opt, metrics=['accuracy'])
```

logs/20220211-030126 WARNING:tensorflow.write\_grads will be ignored in TensorFlow 2.0 for the TensorBoard Callback.  
Epoch 1/50 2/1472 [.....] - ETA: 5:10 - loss: 0.0051 - accuracy: 0.5000 WARNING:tensorflow.Callback method on\_train\_batch\_begin is slow compared to the batch time (batch time: 0.0048s vs on\_train\_batch\_begin time: 0.0189s). Check your callbacks. WARNING:tensorflow.Callback method on\_train\_batch\_end is slow compared to the batch time (batch time: 0.0048s vs on\_train\_batch\_end time: 0.0175s). Check your callbacks. 1467/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.4080

micro\_F1\_train: 0.4197866412991778

micro\_F1\_val: 0.4085288522511097

Epoch 00001: val\_accuracy improved from -inf to 0.40853, saving model to model\_5\_save/weights-e65+01-0.4085.hdf5 1472/1472 [=====>.] - 13s 8ms/step - loss: 0.0045 - accuracy: 0.4080 - val\_loss: 1.3893 - val\_accuracy: 0.4085 - lr: 9.0000e-04 Epoch 2/50 1472/1472 [=====>.] - ETA: 0s - loss: 0.0039 - accuracy: 0.4069

micro\_F1\_train: 0.43833661751715697

micro\_F1\_val: 0.42057704502219406

Epoch 00002: val\_accuracy improved from 0.40853 to 0.42058, saving model to model\_5\_save/weights-e65+02-0.4206.hdf5 1472/1472 [=====>.] - 11s 8ms/step - loss: 0.0039 - accuracy: 0.4069 - val\_loss: 1.3481 - val\_accuracy: 0.4206 - lr: 9.0000e-04 Epoch 3/50 1469/1472 [=====>.] - ETA: 0s - loss: 0.0043 - accuracy: 0.3920

micro\_F1\_train: 0.41496228850988653

micro\_F1\_val: 0.4053582752060875

Epoch 00003: val\_accuracy did not improve from 0.42058 1472/1472 [=====>.] - 11s 8ms/step - loss: 0.0043 - accuracy: 0.3921 - val\_loss: 1.3961 - val\_accuracy: 0.4054 - lr: 9.0000e-04 Epoch 4/50 1470/1472 [=====>.] - ETA: 0s - loss: 0.0041 - accuracy: 0.3982

micro\_F1\_train: 0.404566147992118

micro\_F1\_val: 0.39267596702599866

Epoch 00004: val\_accuracy did not improve from 0.42058 1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0041 - accuracy: 0.3981 - val\_loss: 1.4385 - val\_accuracy: 0.3927 - lr: 9.0000e-04 Epoch 5/50 1472/1472 [=====>.] - ETA: 0s - loss: 0.0045 - accuracy: 0.3993

micro\_F1\_train: 0.4146225453557111

micro\_F1\_val: 0.4036144578313253

Epoch 00005: val\_accuracy did not improve from 0.42058 1472/1472 [=====>.] - 12s 8ms/step - loss: 0.0045 - accuracy: 0.3993 - val\_loss: 1.4889 - val\_accuracy: 0.4036 - lr: 8.1000e-04 Epoch 6/50 1464/1472 [=====>.] - ETA: 0s - loss: 0.0040 - accuracy: 0.4125

micro\_F1\_train: 0.40612896650132496

micro\_F1\_val: 0.3979074191502854

Epoch 00006: val\_accuracy did not improve from 0.42058 1472/1472 [=====>.] - 11s 8ms/step - loss: 0.0040 - accuracy: 0.4127 - val\_loss: 1.4660 - val\_accuracy: 0.3979 - lr: 8.1000e-04 Epoch 7/50 1463/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.4053

micro\_F1\_train: 0.44900455255826593

micro\_F1\_val: 0.4437222574508561

Epoch 00007: val\_accuracy improved from 0.42058 to 0.44372, saving model to model\_5\_save/weights-e65+07-0.4437.hdf5  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.4051 - val\_loss: 1.3228 -  
val\_accuracy: 0.4437 - lr: 7.2900e-04 Epoch 8/50 1464/1472 [=====] - ETA: 0s - loss: 0.0041 -  
accuracy: 0.4098

micro\_F1\_train: 0.46041992253856084

micro\_F1\_val: 0.45640456563094484

Epoch 00008: val\_accuracy improved from 0.44372 to 0.45640, saving model to model\_5\_save/weights-e65+08-0.4564.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0041 - accuracy: 0.4097 - val\_loss: 1.2835 -  
val\_accuracy: 0.4564 - lr: 7.2900e-04 Epoch 9/50 1468/1472 [=====] - ETA: 0s - loss: 0.0040 -  
accuracy: 0.4121

micro\_F1\_train: 0.4449955833389957

micro\_F1\_val: 0.4376981610653139

Epoch 00009: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 12s 8ms/step -  
loss: 0.0040 - accuracy: 0.4120 - val\_loss: 1.3431 - val\_accuracy: 0.4377 - lr: 7.2900e-04 Epoch 10/50 1466/1472  
[=====] - ETA: 0s - loss: 0.0038 - accuracy: 0.4199

micro\_F1\_train: 0.4498878847591221

micro\_F1\_val: 0.43864933417882057

Epoch 00010: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 10s 6ms/step -  
loss: 0.0038 - accuracy: 0.4199 - val\_loss: 1.3248 - val\_accuracy: 0.4386 - lr: 7.2900e-04 Epoch 11/50 1469/1472  
[=====] - ETA: 0s - loss: 0.0036 - accuracy: 0.4151

micro\_F1\_train: 0.43677379900795

micro\_F1\_val: 0.4231135066582118

Epoch 00011: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 11s 8ms/step -  
loss: 0.0036 - accuracy: 0.4154 - val\_loss: 1.3184 - val\_accuracy: 0.4231 - lr: 6.5610e-04 Epoch 12/50 1464/1472  
[=====] - ETA: 0s - loss: 0.0039 - accuracy: 0.4049

micro\_F1\_train: 0.4566147992117959

micro\_F1\_val: 0.4495878249841471

Epoch 00012: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 11s 7ms/step -  
loss: 0.0039 - accuracy: 0.4050 - val\_loss: 1.2985 - val\_accuracy: 0.4496 - lr: 6.5610e-04 Epoch 13/50 1463/1472  
[=====] - ETA: 0s - loss: 0.0043 - accuracy: 0.4118

micro\_F1\_train: 0.4570224909968064

micro\_F1\_val: 0.44403931515535827

Epoch 00013: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 11s 7ms/step -  
loss: 0.0043 - accuracy: 0.4118 - val\_loss: 1.3001 - val\_accuracy: 0.4440 - lr: 5.9049e-04 Epoch 14/50 1467/1472  
[=====] - ETA: 0s - loss: 0.0036 - accuracy: 0.4145

micro\_F1\_train: 0.4394237956105185

micro\_F1\_val: 0.43024730500951175

Epoch 00014: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0036 - accuracy: 0.4146 - val\_loss: 1.3110 - val\_accuracy: 0.4302 - lr: 5.9049e-04 Epoch 15/50 1463/1472  
[=====] - ETA: 0s - loss: 0.0045 - accuracy: 0.4161

micro\_F1\_train: 0.4284161174152341

micro\_F1\_val: 0.40805326569435635

Epoch 00015: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0045 - accuracy: 0.4160 - val\_loss: 1.3578 - val\_accuracy: 0.4081 - lr: 5.3144e-04 Epoch 16/50 1462/1472  
[=====] - ETA: 0s - loss: 0.0034 - accuracy: 0.4142

micro\_F1\_train: 0.43677379900795



micro\_F1\_val: 0.42422320862396956

Epoch 00016: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 10s 7ms/step - loss: 0.0034 - accuracy: 0.4144 - val\_loss: 1.3229 - val\_accuracy: 0.4242 - lr: 5.3144e-04 Epoch 17/50 1465/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4107

micro\_F1\_train: 0.44526737786233606

micro\_F1\_val: 0.4275523145212429

Epoch 00017: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 10s 7ms/step - loss: 0.0036 - accuracy: 0.4112 - val\_loss: 1.2856 - val\_accuracy: 0.4276 - lr: 4.7830e-04 Epoch 18/50 1471/1472 [=====>.] - ETA: 0s - loss: 0.0038 - accuracy: 0.4223

micro\_F1\_train: 0.4536930080858871

micro\_F1\_val: 0.4380152187698161

Epoch 00018: val\_accuracy did not improve from 0.45640 1472/1472 [=====] - 11s 7ms/step - loss: 0.0038 - accuracy: 0.4222 - val\_loss: 1.2941 - val\_accuracy: 0.4380 - lr: 4.7830e-04 Epoch 19/50 1470/1472 [=====>.] - ETA: 0s - loss: 0.0042 - accuracy: 0.4132

micro\_F1\_train: 0.4732622137663926

micro\_F1\_val: 0.462428662016487

Epoch 00019: val\_accuracy improved from 0.45640 to 0.46243, saving model to model\_5\_save/weights-e65+19-0.4624.hdf5 1472/1472 [=====] - 10s 7ms/step - loss: 0.0042 - accuracy: 0.4130 - val\_loss: 1.2805 - val\_accuracy: 0.4624 - lr: 4.3047e-04 Epoch 20/50 1464/1472 [=====>.] - ETA: 0s - loss: 0.0035 - accuracy: 0.4249

micro\_F1\_train: 0.44003533328803424

micro\_F1\_val: 0.4275523145212429

Epoch 00020: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 11s 7ms/step - loss: 0.0035 - accuracy: 0.4252 - val\_loss: 1.3003 - val\_accuracy: 0.4276 - lr: 4.3047e-04 Epoch 21/50 1463/1472 [=====>.] - ETA: 0s - loss: 0.0037 - accuracy: 0.4263

micro\_F1\_train: 0.4439084052456343

micro\_F1\_val: 0.4329422954977806

Epoch 00021: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 11s 8ms/step - loss: 0.0037 - accuracy: 0.4262 - val\_loss: 1.3088 - val\_accuracy: 0.4329 - lr: 4.3047e-04 Epoch 22/50 1461/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4255

micro\_F1\_train: 0.46334171366446963

micro\_F1\_val: 0.44705136334812934

Epoch 00022: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 10s 7ms/step - loss: 0.0033 - accuracy: 0.4258 - val\_loss: 1.2816 - val\_accuracy: 0.4471 - lr: 3.8742e-04 Epoch 23/50 1465/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4363

micro\_F1\_train: 0.43459944282122714

micro\_F1\_val: 0.42232086239695626

Epoch 00023: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 11s 7ms/step - loss: 0.0033 - accuracy: 0.4366 - val\_loss: 1.3256 - val\_accuracy: 0.4223 - lr: 3.8742e-04 Epoch 24/50 1464/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4471

micro\_F1\_train: 0.4411904600122308

micro\_F1\_val: 0.4272352568167406

Epoch 00024: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 11s 7ms/step - loss: 0.0032 - accuracy: 0.4473 - val\_loss: 1.2703 - val\_accuracy: 0.4272 - lr: 3.4868e-04 Epoch 25/50 1472/1472 [=====] - ETA: 0s - loss: 0.0034 - accuracy: 0.4360

micro\_F1\_train: 0.42821227152272884

micro\_F1\_val: 0.4128091312618896

Epoch 00025: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 11s 7ms/step - loss: 0.0034 - accuracy: 0.4360 - val\_loss: 1.3155 - val\_accuracy: 0.4128 - lr: 3.4868e-04 Epoch 26/50 1461/1472 [=====>.] - ETA: 0s - loss: 0.0034 - accuracy: 0.4376

micro\_F1\_train: 0.46103146021607666

micro\_F1\_val: 0.4497463538363982

Epoch 00026: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 10s 7ms/step - loss: 0.0034 - accuracy: 0.4370 - val\_loss: 1.2471 - val\_accuracy: 0.4497 - lr: 3.4868e-04 Epoch 27/50 1464/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4445

micro\_F1\_train: 0.4375891825779711

micro\_F1\_val: 0.4237476220672162

Epoch 00027: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 11s 7ms/step - loss: 0.0033 - accuracy: 0.4442 - val\_loss: 1.3127 - val\_accuracy: 0.4237 - lr: 3.4868e-04 Epoch 28/50 1469/1472 [=====>.] - ETA: 0s - loss: 0.0034 - accuracy: 0.4379

micro\_F1\_train: 0.46069171706190126

micro\_F1\_val: 0.4448319594166138

Epoch 00028: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 12s 8ms/step - loss: 0.0034 - accuracy: 0.4377 - val\_loss: 1.2606 - val\_accuracy: 0.4448 - lr: 3.4868e-04 Epoch 29/50 1469/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4372

micro\_F1\_train: 0.44710199089488345

micro\_F1\_val: 0.43389346861128725

Epoch 00029: val\_accuracy did not improve from 0.46243 1472/1472 [=====] - 10s 7ms/step - loss: 0.0037 - accuracy: 0.4372 - val\_loss: 1.2936 - val\_accuracy: 0.4339 - lr: 3.1381e-04 Epoch 30/50 1463/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4435

micro\_F1\_train: 0.4980634640212

micro\_F1\_val: 0.48224476854787574

Epoch 00030: val\_accuracy improved from 0.46243 to 0.48224, saving model to model\_5\_save/weights-e65+30-0.4822.hdf5 1472/1472 [=====] - 10s 7ms/step - loss: 0.0033 - accuracy: 0.4431 - val\_loss: 1.2207 - val\_accuracy: 0.4822 - lr: 3.1381e-04 Epoch 31/50 1466/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4563

micro\_F1\_train: 0.464360943126996

micro\_F1\_val: 0.4497463538363982

Epoch 00031: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 8ms/step - loss: 0.0033 - accuracy: 0.4562 - val\_loss: 1.2944 - val\_accuracy: 0.4497 - lr: 3.1381e-04 Epoch 32/50 1471/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4468

micro\_F1\_train: 0.46334171366446963

micro\_F1\_val: 0.45212428662016485

Epoch 00032: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 7ms/step - loss: 0.0032 - accuracy: 0.4468 - val\_loss: 1.2606 - val\_accuracy: 0.4521 - lr: 3.1381e-04 Epoch 33/50 1472/1472 [=====] - ETA: 0s - loss: 0.0031 - accuracy: 0.4446

micro\_F1\_train: 0.4678942719304206

micro\_F1\_val: 0.4573557387444515

Epoch 00033: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 10s 7ms/step - loss: 0.0031 - accuracy: 0.4446 - val\_loss: 1.2785 - val\_accuracy: 0.4574 - lr: 2.8243e-04 Epoch 34/50 1462/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4520

micro\_F1\_train: 0.4846775837466875

micro\_F1\_val: 0.47622067216233355

micro\_F1\_val: 0.4560875079264426

Epoch 00034: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 7ms/step - loss: 0.0032 - accuracy: 0.4519 - val\_loss: 1.2260 - val\_accuracy: 0.4762 - lr: 2.8243e-04 Epoch 35/50 1471/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4525

micro\_F1\_train: 0.4687776041312768

micro\_F1\_val: 0.4560875079264426

Epoch 00035: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 7ms/step - loss: 0.0032 - accuracy: 0.4526 - val\_loss: 1.2522 - val\_accuracy: 0.4561 - lr: 2.5419e-04 Epoch 36/50 1472/1472 [=====] - ETA: 0s - loss: 0.0033 - accuracy: 0.4529

micro\_F1\_train: 0.4751647754297751

micro\_F1\_val: 0.4622701331642359

Epoch 00036: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 8ms/step - loss: 0.0033 - accuracy: 0.4529 - val\_loss: 1.2560 - val\_accuracy: 0.4623 - lr: 2.5419e-04 Epoch 37/50 1469/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4481

micro\_F1\_train: 0.46809811782292593

micro\_F1\_val: 0.4530754597336715

Epoch 00037: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 7ms/step - loss: 0.0033 - accuracy: 0.4482 - val\_loss: 1.2611 - val\_accuracy: 0.4531 - lr: 2.2877e-04 Epoch 38/50 1466/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4502

micro\_F1\_train: 0.47468913501392945

micro\_F1\_val: 0.4619530754597337

Epoch 00038: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 7ms/step - loss: 0.0033 - accuracy: 0.4502 - val\_loss: 1.2585 - val\_accuracy: 0.4620 - lr: 2.2877e-04 Epoch 39/50 1467/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4575

micro\_F1\_train: 0.4618468437860977

micro\_F1\_val: 0.4497463538363982

Epoch 00039: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 15s 10ms/step - loss: 0.0031 - accuracy: 0.4575 - val\_loss: 1.2697 - val\_accuracy: 0.4497 - lr: 2.0589e-04 Epoch 40/50 1467/1472 [=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4549

micro\_F1\_train: 0.45790582319766254

micro\_F1\_val: 0.44419784400760937

Epoch 00040: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 18s 12ms/step - loss: 0.0031 - accuracy: 0.4551 - val\_loss: 1.2853 - val\_accuracy: 0.4442 - lr: 2.0589e-04 Epoch 41/50 1466/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4499

micro\_F1\_train: 0.46096351158524157

micro\_F1\_val: 0.4499048826886493

Epoch 00041: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 16s 11ms/step - loss: 0.0032 - accuracy: 0.4500 - val\_loss: 1.2664 - val\_accuracy: 0.4499 - lr: 1.8530e-04 Epoch 42/50 1469/1472 [=====>.] - ETA: 0s - loss: 0.0036 - accuracy: 0.4497

micro\_F1\_train: 0.44377250798396417

micro\_F1\_val: 0.42882054533925174

Epoch 00042: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 10s 7ms/step - loss: 0.0035 - accuracy: 0.4500 - val\_loss: 1.3148 - val\_accuracy: 0.4288 - lr: 1.8530e-04 Epoch 43/50 1468/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4501

micro\_F1\_train: 0.48515322416253315

micro\_F1\_val: 0.4722574508560558

Epoch 00043: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 8ms/step -

Epoch 00043: val\_accuracy did not improve from 0.46224 1472/1472 [=====] - 11s 8ms/step - loss: 0.0030 - accuracy: 0.4506 - val\_loss: 1.2288 - val\_accuracy: 0.4723 - lr: 1.6677e-04 Epoch 44/50 1463/1472 [=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4596

micro\_F1\_train: 0.4733301623972277

micro\_F1\_val: 0.45719720989220036

Epoch 00044: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 8ms/step - loss: 0.0031 - accuracy: 0.4598 - val\_loss: 1.2609 - val\_accuracy: 0.4572 - lr: 1.6677e-04 Epoch 45/50 1468/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4686

micro\_F1\_train: 0.47727118298566285

micro\_F1\_val: 0.46322130627774255

Epoch 00045: val\_accuracy did not improve from 0.48224 1472/1472 [=====] - 11s 8ms/step - loss: 0.0030 - accuracy: 0.4688 - val\_loss: 1.2430 - val\_accuracy: 0.4632 - lr: 1.5009e-04 Epoch 46/50 1465/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4684

micro\_F1\_train: 0.4992865393762316

micro\_F1\_val: 0.4833544705136335

Epoch 00046: val\_accuracy improved from 0.48224 to 0.48335, saving model to model\_5\_save/weights-e65+46-0.4834.hdf5 1472/1472 [=====] - 11s 7ms/step - loss: 0.0030 - accuracy: 0.4683 - val\_loss: 1.2139 - val\_accuracy: 0.4834 - lr: 1.5009e-04 Epoch 47/50 1464/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4742

micro\_F1\_train: 0.47557246721478563

micro\_F1\_val: 0.4613189600507292

Epoch 00047: val\_accuracy did not improve from 0.48335 1472/1472 [=====] - 11s 7ms/step - loss: 0.0031 - accuracy: 0.4743 - val\_loss: 1.2461 - val\_accuracy: 0.4613 - lr: 1.5009e-04 Epoch 48/50 1466/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4730

micro\_F1\_train: 0.4842019433308419

micro\_F1\_val: 0.47019657577679136

Epoch 00048: val\_accuracy did not improve from 0.48335 1472/1472 [=====] - 12s 8ms/step - loss: 0.0030 - accuracy: 0.4729 - val\_loss: 1.2445 - val\_accuracy: 0.4702 - lr: 1.5009e-04 Epoch 49/50 1463/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4715

micro\_F1\_train: 0.4725147788272066

micro\_F1\_val: 0.46163601775523144

Epoch 00049: val\_accuracy did not improve from 0.48335 1472/1472 [=====] - 11s 8ms/step - loss: 0.0032 - accuracy: 0.4716 - val\_loss: 1.2702 - val\_accuracy: 0.4616 - lr: 1.3509e-04 Epoch 50/50 1466/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4704

micro\_F1\_train: 0.5028198681796562

micro\_F1\_val: 0.4889029803424223

Epoch 00050: val\_accuracy improved from 0.48335 to 0.48890, saving model to model\_5\_save/weights-e65+50-0.4889.hdf5 1472/1472 [=====] - 10s 7ms/step - loss: 0.0032 - accuracy: 0.4705 - val\_loss: 1.2075 - val\_accuracy: 0.4889 - lr: 1.3509e-04 Time Taken for training (sec): 556.2567737102509

## Retrain 2

In [ ]:

```
NeuNet_2.load_weights("/content/model_5_save/weights-e65+50-0.4889.hdf5")
```

In [ ]:

```
opt = optimizers.Adam(learning_rate=0.0001)
NeuNet_2.compile(loss='categorical_crossentropy', optimizer = opt, metrics=['accuracy'])
```

logs/20220211-031907 WARNING:tensorflow:write\_grads will be ignored in TensorFlow 2.0 for the TensorBoard Callback.  
Epoch 1/50 2/1472 [.....] - ETA: 5:17 - loss: 0.0054 - accuracy: 0.5000 WARNING:tensorflow:Callback method  
on\_train\_batch\_begin is slow compared to the batch time (batch time: 0.0040s vs on\_train\_batch\_begin time: 0.0199s).  
Check your callbacks. WARNING:tensorflow:Callback method on\_train\_batch\_end is slow compared to the batch time (batch  
time: 0.0040s vs on\_train\_batch\_end time: 0.0176s). Check your callbacks. 1465/1472 [=====>] -  
ETA: 0s - loss: 0.0033 - accuracy: 0.4795  
  
micro\_F1\_train: 0.4875314262417612  
  
micro\_F1\_val: 0.4771718452758402  
  
Epoch 00001: val\_accuracy improved from -inf to 0.47717, saving model to model\_5\_save2/weights-e65+50+01-0.4772.hdf5  
1472/1472 [=====>] - 13s 8ms/step - loss: 0.0033 - accuracy: 0.4793 - val\_loss: 1.2379 -  
val\_accuracy: 0.4772 Epoch 2/50 1465/1472 [=====>] - ETA: 0s - loss: 0.0031 - accuracy: 0.4799  
  
micro\_F1\_train: 0.4989467962220561  
  
micro\_F1\_val: 0.4843056436271401  
  
Epoch 00002: val\_accuracy improved from 0.47717 to 0.48431, saving model to model\_5\_save2/weights-e65+50+02-0.4843.hdf5  
1472/1472 [=====>] - 10s 7ms/step - loss: 0.0031 - accuracy: 0.4801 - val\_loss: 1.2088 -  
val\_accuracy: 0.4843 Epoch 3/50 1470/1472 [=====>] - ETA: 0s - loss: 0.0029 - accuracy: 0.4754  
  
micro\_F1\_train: 0.49038526873683497  
  
micro\_F1\_val: 0.4771718452758402  
  
Epoch 00003: val\_accuracy did not improve from 0.48431 1472/1472 [=====>] - 10s 7ms/step -  
loss: 0.0029 - accuracy: 0.4755 - val\_loss: 1.2315 - val\_accuracy: 0.4772 Epoch 4/50 1464/1472  
[=====>] - ETA: 0s - loss: 0.0032 - accuracy: 0.4707  
  
micro\_F1\_train: 0.47733913161649794  
  
micro\_F1\_val: 0.45989220038046924  
  
Epoch 00004: val\_accuracy did not improve from 0.48431 1472/1472 [=====>] - 10s 7ms/step -  
loss: 0.0032 - accuracy: 0.4703 - val\_loss: 1.2454 - val\_accuracy: 0.4599 Epoch 5/50 1463/1472  
[=====>] - ETA: 0s - loss: 0.0029 - accuracy: 0.4696  
  
micro\_F1\_train: 0.4914044981993613  
  
micro\_F1\_val: 0.47479391249207353  
  
Epoch 00005: val\_accuracy did not improve from 0.48431 1472/1472 [=====>] - 11s 7ms/step -  
loss: 0.0029 - accuracy: 0.4701 - val\_loss: 1.2199 - val\_accuracy: 0.4748 Epoch 6/50 1469/1472  
[=====>] - ETA: 0s - loss: 0.0029 - accuracy: 0.4725  
  
micro\_F1\_train: 0.4824352789291296  
  
micro\_F1\_val: 0.46480659480025366  
  
Epoch 00006: val\_accuracy did not improve from 0.48431 1472/1472 [=====>] - 11s 7ms/step -  
loss: 0.0029 - accuracy: 0.4724 - val\_loss: 1.2432 - val\_accuracy: 0.4648 Epoch 7/50 1470/1472  
[=====>] - ETA: 0s - loss: 0.0029 - accuracy: 0.4752  
  
micro\_F1\_train: 0.48658014541007  
  
micro\_F1\_val: 0.47305009511731133  
  
Epoch 00007: val\_accuracy did not improve from 0.48431 1472/1472 [=====>] - 11s 7ms/step -  
loss: 0.0029 - accuracy: 0.4752 - val\_loss: 1.2313 - val\_accuracy: 0.4731 Epoch 8/50 1461/1472  
[=====>] - ETA: 0s - loss: 0.0031 - accuracy: 0.4645  
  
micro\_F1\_train: 0.4974519263436842  
  
micro\_F1\_val: 0.4835129993658846  
  
Epoch 00008: val\_accuracy did not improve from 0.48431 1472/1472 [=====>] - 10s 7ms/step -  
loss: 0.0031 - accuracy: 0.4650 - val\_loss: 1.2218 - val\_accuracy: 0.4835 Epoch 9/50 1461/1472  
[=====>] - ETA: 0s - loss: 0.0029 - accuracy: 0.4745

micro\_F1\_train: 0.49058911462934024

micro\_F1\_val: 0.47463538363982244

Epoch 00009: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0029 - accuracy: 0.4743 - val\_loss: 1.2405 - val\_accuracy: 0.4746 Epoch 10/50 1465/1472  
[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4755

micro\_F1\_train: 0.48977373105931915

micro\_F1\_val: 0.4751109701965758

Epoch 00010: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0028 - accuracy: 0.4753 - val\_loss: 1.2182 - val\_accuracy: 0.4751 Epoch 11/50 1472/1472  
[=====] - ETA: 0s - loss: 0.0030 - accuracy: 0.4618

micro\_F1\_train: 0.4813481008357682

micro\_F1\_val: 0.4670259987317692

Epoch 00011: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 11s 8ms/step -  
loss: 0.0030 - accuracy: 0.4618 - val\_loss: 1.2397 - val\_accuracy: 0.4670 Epoch 12/50 1470/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4756

micro\_F1\_train: 0.48719168308758576

micro\_F1\_val: 0.4714648065948003

Epoch 00012: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 11s 7ms/step -  
loss: 0.0029 - accuracy: 0.4758 - val\_loss: 1.2198 - val\_accuracy: 0.4715 Epoch 13/50 1463/1472  
[=====>.] - ETA: 0s - loss: 0.0035 - accuracy: 0.4766

micro\_F1\_train: 0.46932119317795745

micro\_F1\_val: 0.45339251743817377

Epoch 00013: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 11s 8ms/step -  
loss: 0.0035 - accuracy: 0.4767 - val\_loss: 1.2638 - val\_accuracy: 0.4534 Epoch 14/50 1467/1472  
[=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4678

micro\_F1\_train: 0.4769993884623225

micro\_F1\_val: 0.4643310082435003

Epoch 00014: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 11s 8ms/step -  
loss: 0.0030 - accuracy: 0.4676 - val\_loss: 1.2529 - val\_accuracy: 0.4643 Epoch 15/50 1464/1472  
[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4771

micro\_F1\_train: 0.4944621865869403

micro\_F1\_val: 0.47622067216233355

Epoch 00015: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 12s 8ms/step -  
loss: 0.0028 - accuracy: 0.4771 - val\_loss: 1.2269 - val\_accuracy: 0.4762 Epoch 16/50 1464/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4786

micro\_F1\_train: 0.49228783040021745

micro\_F1\_val: 0.4776474318325935

Epoch 00016: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0029 - accuracy: 0.4788 - val\_loss: 1.2216 - val\_accuracy: 0.4776 Epoch 17/50 1466/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4754

micro\_F1\_train: 0.4898416796901543

micro\_F1\_val: 0.4741597970830691

Epoch 00017: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 13s 9ms/step -  
loss: 0.0029 - accuracy: 0.4747 - val\_loss: 1.2253 - val\_accuracy: 0.4742 Epoch 18/50 1461/1472  
[=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4829

micro\_F1\_train: 0.4948698783719508

micro\_F1\_val: 0.47479391249207353

Epoch 00018: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 11s 7ms/step - loss: 0.0032 - accuracy: 0.4826 - val\_loss: 1.2115 - val\_accuracy: 0.4748 Epoch 19/50 1470/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4818

micro\_F1\_train: 0.4982673099137052

micro\_F1\_val: 0.478281547241598

Epoch 00019: val\_accuracy did not improve from 0.48431 1472/1472 [=====] - 11s 7ms/step - loss: 0.0029 - accuracy: 0.4815 - val\_loss: 1.2206 - val\_accuracy: 0.4783 Epoch 20/50 1472/1472 [=====] - ETA: 0s - loss: 0.0028 - accuracy: 0.4784

micro\_F1\_train: 0.5096826798940002

micro\_F1\_val: 0.48922003804692454

Epoch 00020: val\_accuracy improved from 0.48431 to 0.48922, saving model to model\_5\_save2/weights-e65+50+20-0.4892.hdf5 1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.4784 - val\_loss: 1.2064 - val\_accuracy: 0.4892 Epoch 21/50 1472/1472 [=====] - ETA: 0s - loss: 0.0028 - accuracy: 0.4839

micro\_F1\_train: 0.5008493578854386

micro\_F1\_val: 0.4803424223208624

Epoch 00021: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 10s 7ms/step - loss: 0.0028 - accuracy: 0.4839 - val\_loss: 1.2194 - val\_accuracy: 0.4803 Epoch 22/50 1463/1472 [=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4802

micro\_F1\_train: 0.49276347081606303

micro\_F1\_val: 0.4733671528218136

Epoch 00022: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 12s 8ms/step - loss: 0.0031 - accuracy: 0.4805 - val\_loss: 1.2210 - val\_accuracy: 0.4734 Epoch 23/50 1469/1472 [=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4734

micro\_F1\_train: 0.49235577903105254

micro\_F1\_val: 0.47400126823081806

Epoch 00023: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 7ms/step - loss: 0.0031 - accuracy: 0.4731 - val\_loss: 1.2292 - val\_accuracy: 0.4740 Epoch 24/50 1465/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4756

micro\_F1\_train: 0.49636474825032273

micro\_F1\_val: 0.4785986049461002

Epoch 00024: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 10s 7ms/step - loss: 0.0030 - accuracy: 0.4754 - val\_loss: 1.2136 - val\_accuracy: 0.4786 Epoch 25/50 1467/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4800

micro\_F1\_train: 0.48868655296595775

micro\_F1\_val: 0.4724159797083069

Epoch 00025: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 8ms/step - loss: 0.0029 - accuracy: 0.4801 - val\_loss: 1.2261 - val\_accuracy: 0.4724 Epoch 26/50 1464/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4671

micro\_F1\_train: 0.5055378134130597

micro\_F1\_val: 0.48652504755865567

Epoch 00026: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 12s 8ms/step - loss: 0.0029 - accuracy: 0.4671 - val\_loss: 1.1996 - val\_accuracy: 0.4865 Epoch 27/50 1463/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4819

micro\_F1\_train: 0.49949038526873685

micro\_F1\_val: 0.48161065313887125

Epoch 00027: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 7ms/step -  
loss: 0.0028 - accuracy: 0.4819 - val\_loss: 1.2103 - val\_accuracy: 0.4816 Epoch 28/50 1465/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4777

micro\_F1\_train: 0.4991506421145614

micro\_F1\_val: 0.4784400760938491

Epoch 00028: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 8ms/step -  
loss: 0.0029 - accuracy: 0.4781 - val\_loss: 1.2118 - val\_accuracy: 0.4784 Epoch 29/50 1466/1472  
[=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4833

micro\_F1\_train: 0.507983964123123

micro\_F1\_val: 0.4881103360811668

Epoch 00029: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 12s 8ms/step -  
loss: 0.0030 - accuracy: 0.4837 - val\_loss: 1.2035 - val\_accuracy: 0.4881 Epoch 30/50 1471/1472  
[=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4811

micro\_F1\_train: 0.5022762791329755

micro\_F1\_val: 0.48398858592263794

Epoch 00030: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 7ms/step -  
loss: 0.0030 - accuracy: 0.4810 - val\_loss: 1.2091 - val\_accuracy: 0.4840 Epoch 31/50 1472/1472  
[=====] - ETA: 0s - loss: 0.0028 - accuracy: 0.4966

micro\_F1\_train: 0.5013249983012842

micro\_F1\_val: 0.48113506658211797

Epoch 00031: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 8ms/step -  
loss: 0.0028 - accuracy: 0.4966 - val\_loss: 1.2122 - val\_accuracy: 0.4811 Epoch 32/50 1471/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4772

micro\_F1\_train: 0.5030916627029965

micro\_F1\_val: 0.4808180088776157

Epoch 00032: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0029 - accuracy: 0.4773 - val\_loss: 1.1952 - val\_accuracy: 0.4808 Epoch 33/50 1467/1472  
[=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4766

micro\_F1\_train: 0.48630835088672963

micro\_F1\_val: 0.4678186429930247

Epoch 00033: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0031 - accuracy: 0.4771 - val\_loss: 1.2345 - val\_accuracy: 0.4678 Epoch 34/50 1464/1472  
[=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4780

micro\_F1\_train: 0.4917442413535367

micro\_F1\_val: 0.4749524413443247

Epoch 00034: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0031 - accuracy: 0.4778 - val\_loss: 1.2235 - val\_accuracy: 0.4750 Epoch 35/50 1465/1472  
[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4810

micro\_F1\_train: 0.500373717469593

micro\_F1\_val: 0.4819277108433735

Epoch 00035: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 7ms/step -  
loss: 0.0028 - accuracy: 0.4812 - val\_loss: 1.2113 - val\_accuracy: 0.4819 Epoch 36/50 1465/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4898

micro\_F1\_train: 0.5001019229462527

micro\_F1\_val: 0.4835129993658846

Epoch 00036: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 8ms/step -



loss: 0.0029 - accuracy: 0.4898 - val\_loss: 1.2060 - val\_accuracy: 0.4835 Epoch 37/50 1470/1472

[=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4814

micro\_F1\_train: 0.48963783379764897

micro\_F1\_val: 0.47130627774254913

Epoch 00037: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 10s 7ms/step -

loss: 0.0032 - accuracy: 0.4813 - val\_loss: 1.2346 - val\_accuracy: 0.4713 Epoch 38/50 1468/1472

[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4813

micro\_F1\_train: 0.4980634640212

micro\_F1\_val: 0.478281547241598

Epoch 00038: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 7ms/step -

loss: 0.0028 - accuracy: 0.4818 - val\_loss: 1.2195 - val\_accuracy: 0.4783 Epoch 39/50 1466/1472

[=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4797

micro\_F1\_train: 0.4828429707141401

micro\_F1\_val: 0.46591629676601143

Epoch 00039: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 12s 8ms/step -

loss: 0.0031 - accuracy: 0.4794 - val\_loss: 1.2459 - val\_accuracy: 0.4659 Epoch 40/50 1465/1472

[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4752

micro\_F1\_train: 0.491200652306856

micro\_F1\_val: 0.47463538363982244

Epoch 00040: val\_accuracy did not improve from 0.48922 1472/1472 [=====] - 11s 7ms/step -

loss: 0.0028 - accuracy: 0.4752 - val\_loss: 1.2134 - val\_accuracy: 0.4746 Epoch 41/50 1472/1472

[=====] - ETA: 0s - loss: 0.0030 - accuracy: 0.4786

micro\_F1\_train: 0.511109601141537

micro\_F1\_val: 0.4930247305009512

Epoch 00041: val\_accuracy improved from 0.48922 to 0.49302, saving model to model\_5\_save2/weights-e65+50+41-0.4930.hdf5

1472/1472 [=====] - 12s 8ms/step - loss: 0.0030 - accuracy: 0.4786 - val\_loss: 1.1793 -

val\_accuracy: 0.4930 Epoch 42/50 1466/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4854

micro\_F1\_train: 0.4992185907453965

micro\_F1\_val: 0.4755865567533291

Epoch 00042: val\_accuracy did not improve from 0.49302 1472/1472 [=====] - 10s 7ms/step -

loss: 0.0029 - accuracy: 0.4853 - val\_loss: 1.2086 - val\_accuracy: 0.4756 Epoch 43/50 1463/1472

[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4815

micro\_F1\_train: 0.5030237140721614

micro\_F1\_val: 0.4833544705136335

Epoch 00043: val\_accuracy did not improve from 0.49302 1472/1472 [=====] - 10s 7ms/step -

loss: 0.0028 - accuracy: 0.4816 - val\_loss: 1.1886 - val\_accuracy: 0.4834 Epoch 44/50 1461/1472

[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4875

micro\_F1\_train: 0.4956173133111368

micro\_F1\_val: 0.47622067216233355

Epoch 00044: val\_accuracy did not improve from 0.49302 1472/1472 [=====] - 10s 7ms/step -

loss: 0.0028 - accuracy: 0.4871 - val\_loss: 1.2170 - val\_accuracy: 0.4762 Epoch 45/50 1471/1472

[=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4838

micro\_F1\_train: 0.5017326900862947

micro\_F1\_val: 0.4862079898541535

Epoch 00045: val\_accuracy did not improve from 0.49302 1472/1472 [=====] - 10s 7ms/step -

loss: 0.0030 - accuracy: 0.4839 - val\_loss: 1.1929 - val\_accuracy: 0.4862 Epoch 46/50 1471/1472

[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4954

```
[ ]: Epoch 46: loss: 0.0029 - accuracy: 0.4816
```

```
micro_F1_train: 0.4928993680777332
```

```
micro_F1_val: 0.4766962587190869
```

```
Epoch 00046: val_accuracy did not improve from 0.49302 1472/1472 [=====] - 11s 8ms/step -  
loss: 0.0028 - accuracy: 0.4953 - val_loss: 1.2260 - val_accuracy: 0.4767 Epoch 47/50 1470/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4816
```

```
micro_F1_train: 0.4875993748725963
```

```
micro_F1_val: 0.4724159797083069
```

```
Epoch 00047: val_accuracy did not improve from 0.49302 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0029 - accuracy: 0.4816 - val_loss: 1.2171 - val_accuracy: 0.4724 Epoch 48/50 1472/1472  
[=====] - ETA: 0s - loss: 0.0029 - accuracy: 0.4822
```

```
micro_F1_train: 0.500373717469593
```

```
micro_F1_val: 0.4825618262523779
```

```
Epoch 00048: val_accuracy did not improve from 0.49302 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0029 - accuracy: 0.4822 - val_loss: 1.1981 - val_accuracy: 0.4826 Epoch 49/50 1470/1472  
[=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4899
```

```
micro_F1_train: 0.5107019093565265
```

```
micro_F1_val: 0.49286620164870004
```

```
Epoch 00049: val_accuracy did not improve from 0.49302 1472/1472 [=====] - 10s 7ms/step -  
loss: 0.0029 - accuracy: 0.4897 - val_loss: 1.1890 - val_accuracy: 0.4929 Epoch 50/50 1467/1472  
[=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4813
```

```
micro_F1_train: 0.5160698511924985
```

```
micro_F1_val: 0.49603677869372226
```

```
Epoch 00050: val_accuracy improved from 0.49302 to 0.49604, saving model to model_5_save2/weights-e65+50+50-0.4960.hdf5  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.4816 - val_loss: 1.1668 -  
val_accuracy: 0.4960 Time Taken for training (sec): 562.9610755443573
```

### Retrain 3

```
In [ ]:
```

```
NeuNet_2.load_weights("/content/m2_weights-e65+50+50-0.4960.hdf5")
```

```
In [ ]:
```

```
opt = optimizers.Adam(learning_rate=0.0001)  
NeuNet_2.compile(loss='categorical_crossentropy', optimizer = opt, metrics=['accuracy'])
```

```
In [ ]:
```

```
logdir = os.path.join("logs", datetime.now().strftime("%Y%m%d-%H%M%S"))  
print(logdir)  
file_writer = tf.summary.create_file_writer(logdir + "/metrics")  
# tensorboard = TensorBoard(log_dir=logdir)  
tensorboard = TensorBoard(log_dir=logdir, histogram_freq=1, write_graph=True, write_grads=True)  
  
metric_calc_5 = metric_calc()  
  
# Saving model at every epoch if validation accuracy is improved from previous epoch  
filepath_m5="model_5_save3/weights-e65+50+50+{epoch:02d}-{val_accuracy:.4f}.hdf5"  
checkpoint_m5 = ModelCheckpoint(filepath=filepath_m5, monitor='val_accuracy', verbose=1,  
                                save_best_only=True,  
                                mode='auto')  
  
# earlystop_m5 = EarlyStopping(monitor='val_accuracy', min_delta=0.01, patience=20, verbose=1)  
# reduce_lr_m5 = ReduceLROnPlateau(monitor='val_loss', factor=0.9, patience=2, min_lr=0.0001)  
  
callback_list_5 = [metric_calc_5
```

```

callback_list_5 = [metric_calc_5,
                  checkpoint_m5,
                  #earlystop_m5,
                  #reduce_lr_m5,
                  tensorboard]

# fit network
verbose_5, epochs_5, batch_size_5 = 1, 50, 10
start = time.time()
history_5 = NeuNet_2.fit(X_train_std, y_ctg_train,
                        class_weight = class_wts,
                        validation_data=(X_test_std, y_ctg_test),
                        epochs=epochs_5,
                        batch_size=batch_size_5,
                        verbose=verbose_5,
                        callbacks=callback_list_5)

# Passing the F1 score data of each epoch to tensor board
# https://stackoverflow.com/questions/58102016/how-visualize-in-tensorboard-a-metric-callback
for i in range(len(metric_calc_5.metrics['micro_F1_train'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_train', metric_calc_5.metrics['micro_F1_train'][i])
file_writer.flush()

for i in range(len(metric_calc_5.metrics['micro_F1_val'])):
    with file_writer.as_default(step=i+1):
        tf.summary.scalar('micro_F1_val', metric_calc_5.metrics['micro_F1_val'][i])
file_writer.flush()

stop = time.time()
print('Time Taken for training (sec): ', stop-start)

```

logs/20220211-113500

WARNING:tensorflow:write\_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.

Epoch 1/50

2/1472 [.....] - ETA: 5:01 - loss: 0.0036 - accuracy: 0.5000 WARNING:tensorflow:Callback method `on\_train\_batch\_begin` is slow compared to the batch time (batch time: 0.0051s vs `on\_train\_batch\_begin` time: 0.0180s). Check your callbacks.

WARNING:tensorflow:Callback method `on\_train\_batch\_end` is slow compared to the batch time (batch time: 0.0051s vs `on\_train\_batch\_end` time: 0.0182s). Check your callbacks.

1463/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4925

micro\_F1\_train: 0.49582115920364206

micro\_F1\_val: 0.4819277108433735

Epoch 00001: val\_accuracy improved from -inf to 0.48193, saving model to model\_5\_save3/weights-e65+50+50+01-0.4819.hdf5

1472/1472 [=====>.] - 12s 7ms/step - loss: 0.0030 - accuracy: 0.4922 - val\_loss: 1.2308 - val\_accuracy: 0.4819

Epoch 2/50

1467/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4859

micro\_F1\_train: 0.5053339675205545

micro\_F1\_val: 0.4903297400126823

Epoch 00002: val\_accuracy improved from 0.48193 to 0.49033, saving model to model\_5\_save3/weights-e65+50+50+02-0.4903.hdf5

1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0027 - accuracy: 0.4859 - val\_loss: 1.1882 - val\_accuracy: 0.4903

Epoch 3/50

1466/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4797

micro\_F1\_train: 0.4980634640212

micro\_F1\_val: 0.4809765377298668

Epoch 00003: val\_accuracy did not improve from 0.49033

1472/1472 [=====>.] - 10s 7ms/step - loss: 0.0030 - accuracy: 0.4796 - val\_loss: 1.2130 - val\_accuracy: 0.4810

Epoch 4/50

1466/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4870

micro\_F1\_train: 0.5104301148331861

micro\_F1\_val: 0.4906467977171845

Epoch 00004: val\_accuracy improved from 0.49033 to 0.49065, saving model to model\_5\_save3/weights-e65+50+50+04-0.4906.hdf5  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0027 - accuracy: 0.4871 - val\_loss: 1.1850 - val\_accuracy: 0.4906  
Epoch 5/50  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4832  
  
micro\_F1\_train: 0.5033634572263369  
  
micro\_F1\_val: 0.4836715282181357  
  
Epoch 00005: val\_accuracy did not improve from 0.49065  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0032 - accuracy: 0.4833 - val\_loss: 1.1906 - val\_accuracy: 0.4837  
Epoch 6/50  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4828  
  
micro\_F1\_train: 0.5048583271047088  
  
micro\_F1\_val: 0.48826886493341787  
  
Epoch 00006: val\_accuracy did not improve from 0.49065  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0030 - accuracy: 0.4837 - val\_loss: 1.1922 - val\_accuracy: 0.4883  
Epoch 7/50  
1471/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4876  
  
micro\_F1\_train: 0.5007814092546036  
  
micro\_F1\_val: 0.4809765377298668  
  
Epoch 00007: val\_accuracy did not improve from 0.49065  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0029 - accuracy: 0.4877 - val\_loss: 1.2134 - val\_accuracy: 0.4810  
Epoch 8/50  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4866  
  
micro\_F1\_train: 0.5077121695997826  
  
micro\_F1\_val: 0.4885859226379201  
  
Epoch 00008: val\_accuracy did not improve from 0.49065  
1472/1472 [=====] - 12s 8ms/step - loss: 0.0028 - accuracy: 0.4866 - val\_loss: 1.1988 - val\_accuracy: 0.4886  
Epoch 9/50  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4881  
  
micro\_F1\_train: 0.5051301216280492  
  
micro\_F1\_val: 0.4881103360811668  
  
Epoch 00009: val\_accuracy did not improve from 0.49065  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0029 - accuracy: 0.4881 - val\_loss: 1.1922 - val\_accuracy: 0.4881  
Epoch 10/50  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4882  
  
micro\_F1\_train: 0.48345450839165593  
  
micro\_F1\_val: 0.463855421686747  
  
Epoch 00010: val\_accuracy did not improve from 0.49065  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0029 - accuracy: 0.4884 - val\_loss: 1.2326 - val\_accuracy: 0.4639  
Epoch 11/50  
1472/1472 [=====] - ETA: 0s - loss: 0.0028 - accuracy: 0.4807  
  
micro\_F1\_train: 0.5090711422164843  
  
micro\_F1\_val: 0.4881103360811668  
  
Epoch 00011: val\_accuracy did not improve from 0.49065  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0028 - accuracy: 0.4807 - val\_loss: 1.1934 - val\_accuracy: 0.4881  
Epoch 12/50  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4854

```
micro_F1_train: 0.5015967928246245

micro_F1_val: 0.48319594166138236

Epoch 00012: val_accuracy did not improve from 0.49065
1472/1472 [=====] - 11s 8ms/step - loss: 0.0029 - accuracy: 0.4856 - val_loss: 1.2165 - val_accuracy: 0.4832
Epoch 13/50
1466/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4952

micro_F1_train: 0.506489094244751

micro_F1_val: 0.4868421052631579

Epoch 00013: val_accuracy did not improve from 0.49065
1472/1472 [=====] - 10s 7ms/step - loss: 0.0027 - accuracy: 0.4950 - val_loss: 1.2064 - val_accuracy: 0.4868
Epoch 14/50
1471/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4900

micro_F1_train: 0.5092749881089896

micro_F1_val: 0.49001268230818007

Epoch 00014: val_accuracy did not improve from 0.49065
1472/1472 [=====] - 10s 7ms/step - loss: 0.0029 - accuracy: 0.4901 - val_loss: 1.1918 - val_accuracy: 0.4900
Epoch 15/50
1465/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4898

micro_F1_train: 0.5045185839505334

micro_F1_val: 0.487000634115409

Epoch 00015: val_accuracy did not improve from 0.49065
1472/1472 [=====] - 11s 8ms/step - loss: 0.0029 - accuracy: 0.4900 - val_loss: 1.1977 - val_accuracy: 0.4870
Epoch 16/50
1466/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4915

micro_F1_train: 0.5028198681796562

micro_F1_val: 0.4850982878883957

Epoch 00016: val_accuracy did not improve from 0.49065
1472/1472 [=====] - 10s 7ms/step - loss: 0.0027 - accuracy: 0.4916 - val_loss: 1.2027 - val_accuracy: 0.4851
Epoch 17/50
1467/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4851

micro_F1_train: 0.5108378066181967

micro_F1_val: 0.4914394419784401

Epoch 00017: val_accuracy improved from 0.49065 to 0.49144, saving model to model_5_save3/weights-e65+50+50+17-0.4914.hdf5
1472/1472 [=====] - 12s 8ms/step - loss: 0.0029 - accuracy: 0.4852 - val_loss: 1.1875 - val_accuracy: 0.4914
Epoch 18/50
1472/1472 [=====] - ETA: 0s - loss: 0.0029 - accuracy: 0.4926

micro_F1_train: 0.5039070462730176

micro_F1_val: 0.48462270133164237

Epoch 00018: val_accuracy did not improve from 0.49144
1472/1472 [=====] - 10s 7ms/step - loss: 0.0029 - accuracy: 0.4926 - val_loss: 1.2140 - val_accuracy: 0.4846
Epoch 19/50
1471/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4846

micro_F1_train: 0.4954134674186315

micro_F1_val: 0.47685478757133803

Epoch 00019: val_accuracy did not improve from 0.49144
1472/1472 [=====] - 10s 7ms/step - loss: 0.0029 - accuracy: 0.4926 - val_loss: 1.2140 - val_accuracy: 0.4846
```

```
1472/1472 [=====] - 10s 7ms/step - loss: 0.0029 - accuracy: 0.4847 - val_loss: 1.2161 - val_accuracy: 0.4769
Epoch 20/50
1466/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4913

micro_F1_train: 0.5210980498742951

micro_F1_val: 0.49889029803424223

Epoch 00020: val_accuracy improved from 0.49144 to 0.49889, saving model to model_5_save3/weights-e65+50+50+20-0.4989.hdf5
1472/1472 [=====] - 11s 7ms/step - loss: 0.0029 - accuracy: 0.4917 - val_loss: 1.1676 - val_accuracy: 0.4989
Epoch 21/50
1463/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4946

micro_F1_train: 0.5189916423184073

micro_F1_val: 0.49873176918199114

Epoch 00021: val_accuracy did not improve from 0.49889
1472/1472 [=====] - 10s 7ms/step - loss: 0.0028 - accuracy: 0.4943 - val_loss: 1.1814 - val_accuracy: 0.4987
Epoch 22/50
1466/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4905

micro_F1_train: 0.5075083237072773

micro_F1_val: 0.48478123018389346

Epoch 00022: val_accuracy did not improve from 0.49889
1472/1472 [=====] - 10s 7ms/step - loss: 0.0028 - accuracy: 0.4908 - val_loss: 1.1993 - val_accuracy: 0.4848
Epoch 23/50
1463/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4935

micro_F1_train: 0.49955833389957194

micro_F1_val: 0.48050095117311353

Epoch 00023: val_accuracy did not improve from 0.49889
1472/1472 [=====] - 10s 7ms/step - loss: 0.0028 - accuracy: 0.4933 - val_loss: 1.2056 - val_accuracy: 0.4805
Epoch 24/50
1468/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4985

micro_F1_train: 0.5135557518516002

micro_F1_val: 0.4938173747622067

Epoch 00024: val_accuracy did not improve from 0.49889
1472/1472 [=====] - 10s 7ms/step - loss: 0.0028 - accuracy: 0.4985 - val_loss: 1.1761 - val_accuracy: 0.4938
Epoch 25/50
1471/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.5004

micro_F1_train: 0.5155942107766529

micro_F1_val: 0.49429296131896006

Epoch 00025: val_accuracy did not improve from 0.49889
1472/1472 [=====] - 11s 8ms/step - loss: 0.0028 - accuracy: 0.5006 - val_loss: 1.1810 - val_accuracy: 0.4943
Epoch 26/50
1463/1472 [=====>.] - ETA: 0s - loss: 0.0032 - accuracy: 0.4900

micro_F1_train: 0.5227967656451723

micro_F1_val: 0.49968294229549776

Epoch 00026: val_accuracy improved from 0.49889 to 0.49968, saving model to model_5_save3/weights-e65+50+50+26-0.4997.hdf5
1472/1472 [=====] - 11s 7ms/step - loss: 0.0032 - accuracy: 0.4895 - val_loss: 1.1879 - val_accuracy: 0.4997
Epoch 27/50
1468/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4939
```

micro\_F1\_train: 0.5266698376027723

micro\_F1\_val: 0.5031705770450222

Epoch 00027: val\_accuracy improved from 0.49968 to 0.50317, saving model to model\_5\_save3/weights-e65+50+50+27-0.5032.hdf5

1472/1472 [=====] - 11s 8ms/step - loss: 0.0028 - accuracy: 0.4942 - val\_loss: 1.1790 - val\_accuracy: 0.5032

Epoch 28/50

1469/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4990

micro\_F1\_train: 0.5113813956648774

micro\_F1\_val: 0.4879518072289157

Epoch 00028: val\_accuracy did not improve from 0.50317

1472/1472 [=====] - 11s 7ms/step - loss: 0.0027 - accuracy: 0.4987 - val\_loss: 1.1930 - val\_accuracy: 0.4880

Epoch 29/50

1463/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4925

micro\_F1\_train: 0.5030916627029965

micro\_F1\_val: 0.4825618262523779

Epoch 00029: val\_accuracy did not improve from 0.50317

1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.4924 - val\_loss: 1.2150 - val\_accuracy: 0.4826

Epoch 30/50

1467/1472 [=====>.] - ETA: 0s - loss: 0.0033 - accuracy: 0.4907

micro\_F1\_train: 0.5048583271047088

micro\_F1\_val: 0.48319594166138236

Epoch 00030: val\_accuracy did not improve from 0.50317

1472/1472 [=====] - 10s 7ms/step - loss: 0.0033 - accuracy: 0.4906 - val\_loss: 1.2151 - val\_accuracy: 0.4832

Epoch 31/50

1462/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4948

micro\_F1\_train: 0.5087993476931439

micro\_F1\_val: 0.48589093214965123

Epoch 00031: val\_accuracy did not improve from 0.50317

1472/1472 [=====] - 11s 7ms/step - loss: 0.0029 - accuracy: 0.4946 - val\_loss: 1.2138 - val\_accuracy: 0.4859

Epoch 32/50

1465/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4976

micro\_F1\_train: 0.5185839505333968

micro\_F1\_val: 0.4950856055802156

Epoch 00032: val\_accuracy did not improve from 0.50317

1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.4979 - val\_loss: 1.1941 - val\_accuracy: 0.4951

Epoch 33/50

1465/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4986

micro\_F1\_train: 0.5168172861316844

micro\_F1\_val: 0.4931832593532023

Epoch 00033: val\_accuracy did not improve from 0.50317

1472/1472 [=====] - 11s 7ms/step - loss: 0.0027 - accuracy: 0.4983 - val\_loss: 1.2055 - val\_accuracy: 0.4932

Epoch 34/50

1471/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4892

micro\_F1\_train: 0.5060134538289054

micro\_F1\_val: 0.48652504755865567

Epoch 00034: val\_accuracy did not improve from 0.50317

1472/1472 [=====] - 11s 8ms/step - loss: 0.0029 - accuracy: 0.4894 - val\_loss:

1.2000 - val\_accuracy: 0.4865  
Epoch 35/50  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4941

micro\_F1\_train: 0.5085275531698036

micro\_F1\_val: 0.4879518072289157

Epoch 00035: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0028 - accuracy: 0.4943 - val\_loss:  
1.2029 - val\_accuracy: 0.4880  
Epoch 36/50  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0030 - accuracy: 0.4952

micro\_F1\_train: 0.49887884759122103

micro\_F1\_val: 0.4784400760938491

Epoch 00036: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0030 - accuracy: 0.4951 - val\_loss:  
1.2321 - val\_accuracy: 0.4784  
Epoch 37/50  
1471/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4917

micro\_F1\_train: 0.5004416661004281

micro\_F1\_val: 0.4800253646163602

Epoch 00037: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0029 - accuracy: 0.4918 - val\_loss:  
1.2241 - val\_accuracy: 0.4800  
Epoch 38/50  
1471/1472 [=====>.] - ETA: 0s - loss: 0.0031 - accuracy: 0.4918

micro\_F1\_train: 0.5081198613847931

micro\_F1\_val: 0.48668357641090676

Epoch 00038: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 10s 7ms/step - loss: 0.0031 - accuracy: 0.4919 - val\_loss:  
1.2062 - val\_accuracy: 0.4867  
Epoch 39/50  
1462/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4960

micro\_F1\_train: 0.5205544608276144

micro\_F1\_val: 0.4982561826252378

Epoch 00039: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0027 - accuracy: 0.4962 - val\_loss:  
1.1810 - val\_accuracy: 0.4983  
Epoch 40/50  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4989

micro\_F1\_train: 0.5156621594074879

micro\_F1\_val: 0.48922003804692454

Epoch 00040: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 12s 8ms/step - loss: 0.0028 - accuracy: 0.4988 - val\_loss:  
1.1953 - val\_accuracy: 0.4892  
Epoch 41/50  
1466/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4937

micro\_F1\_train: 0.5168852347625196

micro\_F1\_val: 0.49286620164870004

Epoch 00041: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.4941 - val\_loss:  
1.1772 - val\_accuracy: 0.4929  
Epoch 42/50  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4948

micro\_F1\_train: 0.5077121695997826

micro\_F1\_val: 0.4825618262523779



Epoch 00042: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0027 - accuracy: 0.4945 - val\_loss:  
1.2122 - val\_accuracy: 0.4826  
Epoch 43/50  
1467/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4918  
  
micro\_F1\_train: 0.5109057552490317  
  
micro\_F1\_val: 0.48478123018389346  
  
Epoch 00043: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 12s 8ms/step - loss: 0.0028 - accuracy: 0.4919 - val\_loss:  
1.1918 - val\_accuracy: 0.4848  
Epoch 44/50  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4978  
  
micro\_F1\_train: 0.5163416457158388  
  
micro\_F1\_val: 0.4936588459099556  
  
Epoch 00044: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0027 - accuracy: 0.4976 - val\_loss:  
1.1886 - val\_accuracy: 0.4937  
Epoch 45/50  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4975  
  
micro\_F1\_train: 0.514914724468302  
  
micro\_F1\_val: 0.4879518072289157  
  
Epoch 00045: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0028 - accuracy: 0.4977 - val\_loss:  
1.1950 - val\_accuracy: 0.4880  
Epoch 46/50  
1463/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4928  
  
micro\_F1\_train: 0.5125365223890739  
  
micro\_F1\_val: 0.48922003804692454  
  
Epoch 00046: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.4930 - val\_loss:  
1.1919 - val\_accuracy: 0.4892  
Epoch 47/50  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.5016  
  
micro\_F1\_train: 0.5163416457158388  
  
micro\_F1\_val: 0.4895370957514268  
  
Epoch 00047: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.5016 - val\_loss:  
1.1825 - val\_accuracy: 0.4895  
Epoch 48/50  
1468/1472 [=====>.] - ETA: 0s - loss: 0.0029 - accuracy: 0.4980  
  
micro\_F1\_train: 0.5115852415573826  
  
micro\_F1\_val: 0.48779327837666453  
  
Epoch 00048: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0029 - accuracy: 0.4979 - val\_loss:  
1.1995 - val\_accuracy: 0.4878  
Epoch 49/50  
1470/1472 [=====>.] - ETA: 0s - loss: 0.0027 - accuracy: 0.4980  
  
micro\_F1\_train: 0.5165454916083441  
  
micro\_F1\_val: 0.4930247305009512  
  
Epoch 00049: val\_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 8ms/step - loss: 0.0027 - accuracy: 0.4981 - val\_loss:  
1.1881 - val\_accuracy: 0.4930  
Epoch 50/50  
1469/1472 [=====>.] - ETA: 0s - loss: 0.0028 - accuracy: 0.4948

```
micro_F1_train: 0.5219134334443161
```

```
micro_F1_val: 0.49698795180722893
```

```
Epoch 00050: val_accuracy did not improve from 0.50317  
1472/1472 [=====] - 11s 7ms/step - loss: 0.0028 - accuracy: 0.4951 - val_loss:  
1.1838 - val_accuracy: 0.4970  
Time Taken for training (sec): 563.025274515152
```

```
In [ ]:
```

```
# http://localhost:6006/  
%load_ext tensorboard  
%tensorboard --logdir logs --host localhost
```

## Model 2 Retrain Predictions

```
In [ ]:
```

```
NeuNet_2.load_weights("/content/m2_weights-e65+50+50+27-0.5032.hdf5")
```

```
In [ ]:
```

```
px_data_std = pd.DataFrame(Scaler.transform(px_data))  
px_data_std = px_data_std.drop(feature_correlated,axis=1)
```

```
In [ ]:
```

```
# %%timeit  
y_pred_5 = NeuNet_2.predict(px_data_std)
```

```
In [ ]:
```

```
np.argmax(y_pred_5,axis=1)
```

```
Out[ ]:
```

```
array([ 3,  3, 12, ..., 14, 14, 14])
```

```
In [ ]:
```

```
y_pred_pd_5 = pd.DataFrame(np.argmax(y_pred_5,axis=1),index=px_data.index)  
# y_pred_pd[0] = y_pred_pd[0]+1
```

```
In [ ]:
```

```
%%timeit  
t = y_pred_pd_5.to_numpy().reshape((145,145))
```

The slowest run took 18.32 times longer than the fastest. This could mean that an intermediate result is being cached.

100000 loops, best of 5: 4.02  $\mu$ s per loop

```
In [ ]:
```

```
# y_pred_pd_5
```

```
In [ ]:
```

```
y_pred_pd_5.value_counts()
```

```
Out[ ]:
```

Out [ ]:

```
14    3394
15    2805
0     2582
11    2144
10    1654
2     1625
12    1409
6     1366
5     1082
3      966
8      644
4      621
13     258
16     218
9      146
1       56
7       55
dtype: int64
```

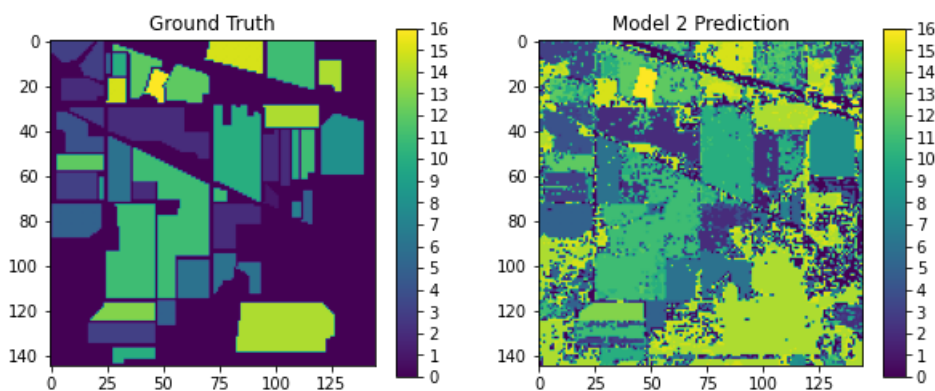
In [ ]:

```
# px_data.index
```

In [ ]:

```
figr,axis = plt.subplots(1,2,figsize=(10,10))
im0 = axis[0].imshow(mat_gt['indian_pines_gt'])#,cmap='jet')
axis[0].set_title('Ground Truth')
plt.colorbar(im0,ax=axis[0],shrink=0.4,aspect=16, ticks=range(0,17,1))

im1 = axis[1].imshow(y_pred_pd_5.to_numpy().reshape((145,145))#,cmap='jet')
axis[1].set_title('Model 2 Prediction')
plt.colorbar(im1,ax=axis[1],shrink=0.4,aspect=16, ticks=range(0,17,1))
# plt.savefig('NeuNet_4_e100.png')
plt.show()
```



In [ ]:

```
y_pred_5_test = NeuNet_2.predict(X_test_std)
y_pred_pd_5_test = pd.DataFrame(np.argmax(y_pred_5_test,axis=1),index=y_test.index)
```

In [ ]:

```
# plot_confusion_matrix(px_class,y_pred_pd_5)
plot_confusion_matrix(y_test,y_pred_pd_5_test)
```

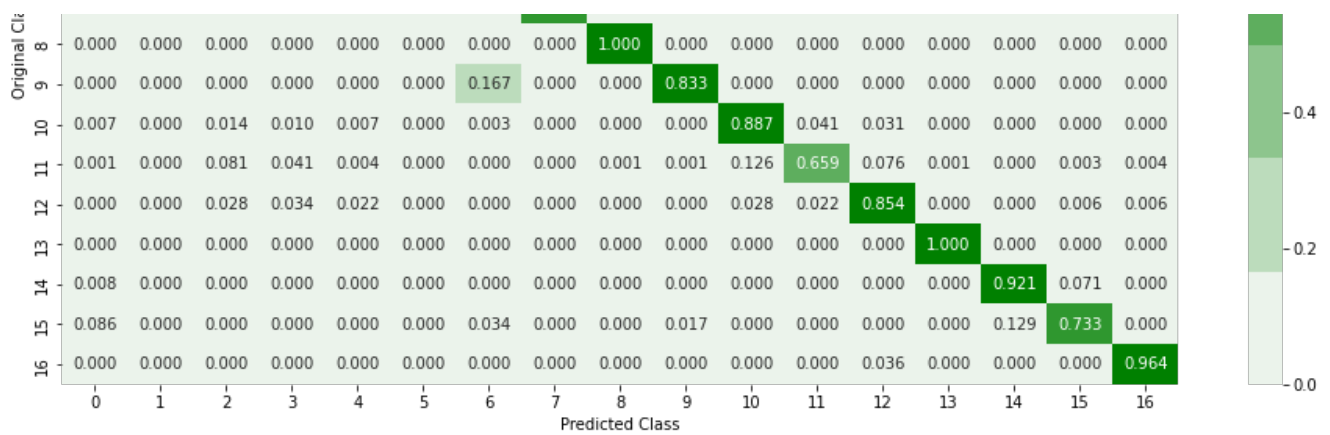
Number of misclassified points 49.68294229549778

----- Confusion matrix -----

-----

738.0	4.0	106.0	88.0	73.0	177.0	203.0	7.0	47.0	32.0	119.0	71.0	119.0	16.0	677.0	718.0	38.0
-------	-----	-------	------	------	-------	-------	-----	------	------	-------	------	-------	------	-------	-------	------





Sum of rows in recall matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

In [ ]:

```
kappa_score = cohen_kappa_score(y_test,y_pred_pd_5_test)
```

In [ ]:

```
print('kappa_score for model 2 (17 classes) = ', kappa_score)
```

kappa\_score for model 2 (17 classes) = 0.44586439564171576

### Confusion matrix without class 0

In [ ]:

```
# code reference: appliedaicourse.com case studies
def plot_confusion_matrix_1(test_y, predict_y):
    """
    This function generates the confusion matrix ignoring row and column associated with Class 0.
    Also evaluates the micro and macro F1 score for the above.
    """
    print('Confusion / Precision / Recall matrix without class 0')
    C = confusion_matrix(test_y, predict_y)
    # print("Number of misclassified points ", (len(test_y)-np.trace(C))/len(test_y)*100)
    print("Number of misclassified points ", (np.sum(C[1:,1:])-np.trace(C[1:,1:]))/np.sum(C[1:,1:])*100)
    # C = 17x17 matrix, each cell (i,j) represents number of points of class i are predicted class j

    #Precision matrix
    A = (C[1:,1:]/C[1:,1:].sum(axis=0))
    #divide each element of the confusion matrix with the sum of elements in that column

    #Recall matrix
    B = ((C[1:,1:].T)/(C[1:,1:].sum(axis=1))).T
    #divide each element of the confusion matrix with the sum of elements in that row

    labels = list(range(0,17,1))
    cmap=sb.light_palette("green")
    # representing C in heatmap format
    print("-"*50, "Confusion matrix", "-"*50)
    plt.figure(figsize=(16,8))
    sb.heatmap(C[1:,1:], annot=True, cmap=cmap, fmt=".1f", xticklabels=labels[1:17], yticklabels=labels[1:17])
    plt.xlabel('Predicted Class')
    plt.ylabel('Original Class')
    plt.show()

    # representing B in heatmap format
    print("-"*50, "Precision matrix", "-"*50)
    plt.figure(figsize=(16,8))
    sb.heatmap(A, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels[1:17], yticklabels=labels[1:17])
    plt.xlabel('Predicted Class')
    plt.ylabel('Original Class')
```

```

plt.show()
print("Sum of columns in precision matrix",A.sum(axis=0))

# representing A in heatmap format
print("-"*50, "Recall matrix" , "-"*50)
plt.figure(figsize=(16,8))
sb.heatmap(B, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels[1:17], yticklabels=labels[1:17])
plt.xlabel('Predicted Class')
plt.ylabel('Original Class')
plt.show()
print("Sum of rows in recall matrix",B.sum(axis=1))

#sum of all True positives
TP = np.trace(C[1:,1:])

#sum of all True positives and False Positives
TP_FP = np.sum(C[1:,1:].sum(axis=1))

#sum of all True positives and False Negatives
TP_NP = np.sum(C[1:,1:].sum(axis=0))

#micro F1 score evaluation
micro_Pr = TP / TP_FP
micro_Re = TP / TP_NP
micro_F1 = 2 * (micro_Pr * micro_Re)/(micro_Pr + micro_Re)

print('\n micro F1 score ignoring class 0 : ', micro_F1)

#macro F1 score evaluation
macro_Pr = np.trace(B)/16
macro_Re = np.trace(A)/16

macro_F1 = 2 * (macro_Pr * macro_Re)/(macro_Pr + macro_Re)

print('\n macro F1 score ignoring class 0 : ', macro_F1)

AA = np.trace(B)/16
print('\n Average Accuracy ignoring class 0 = ',AA)

```

In [ ]:

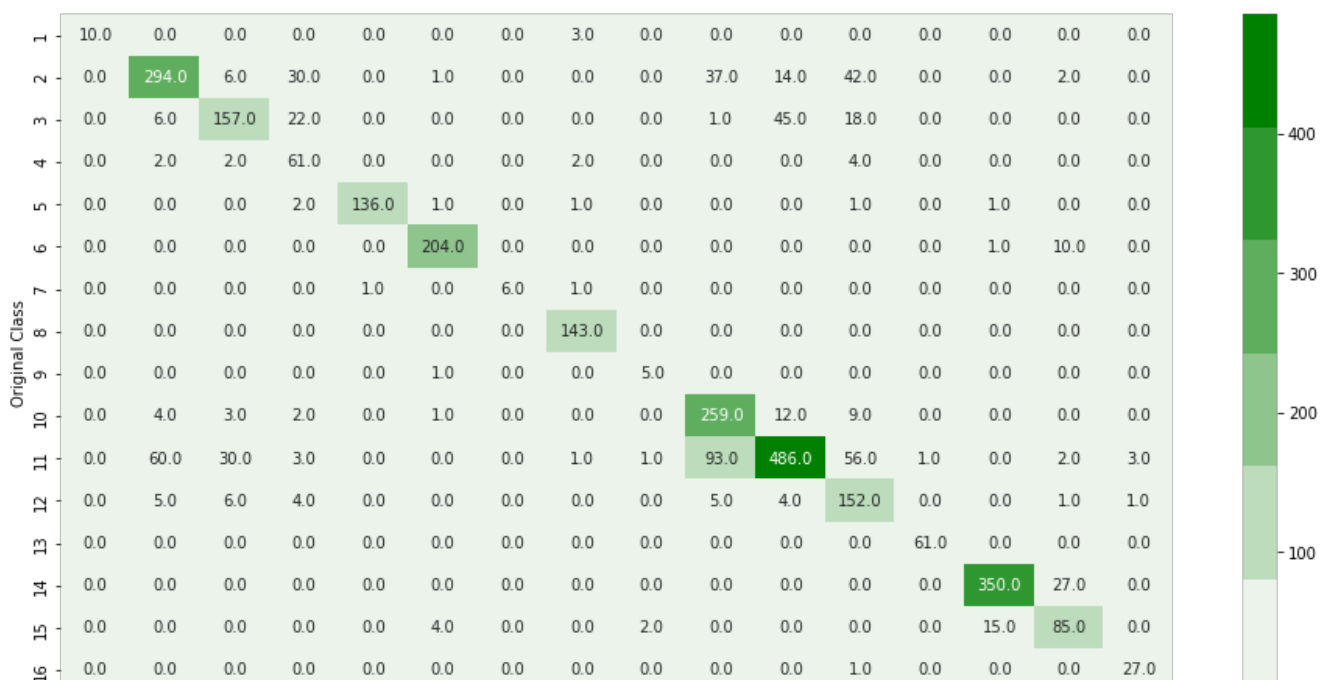
```

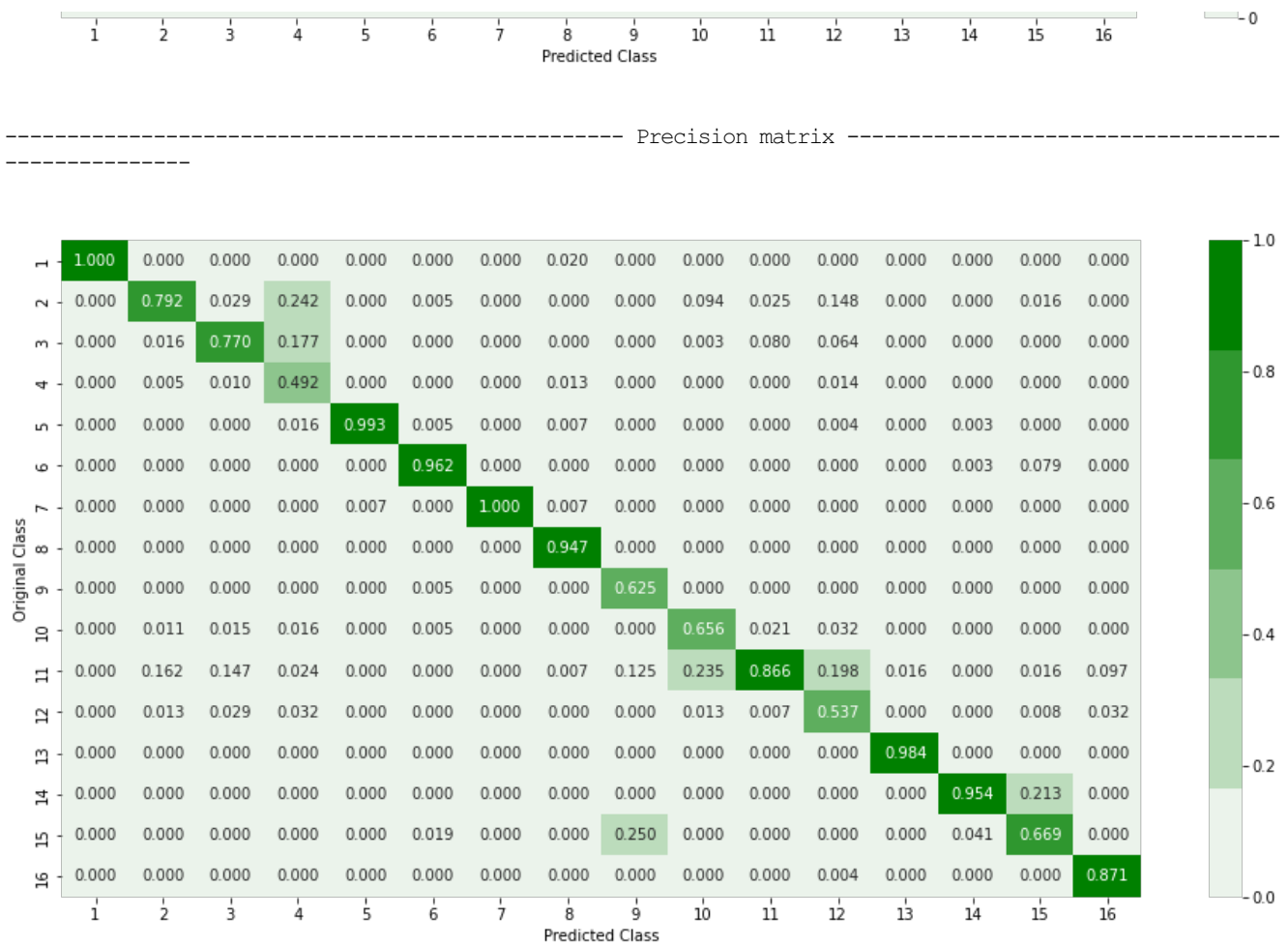
# plot_confusion_matrix_1(px_class,y_pred_pd_5)
plot_confusion_matrix_1(y_test,y_pred_pd_5_test)

```

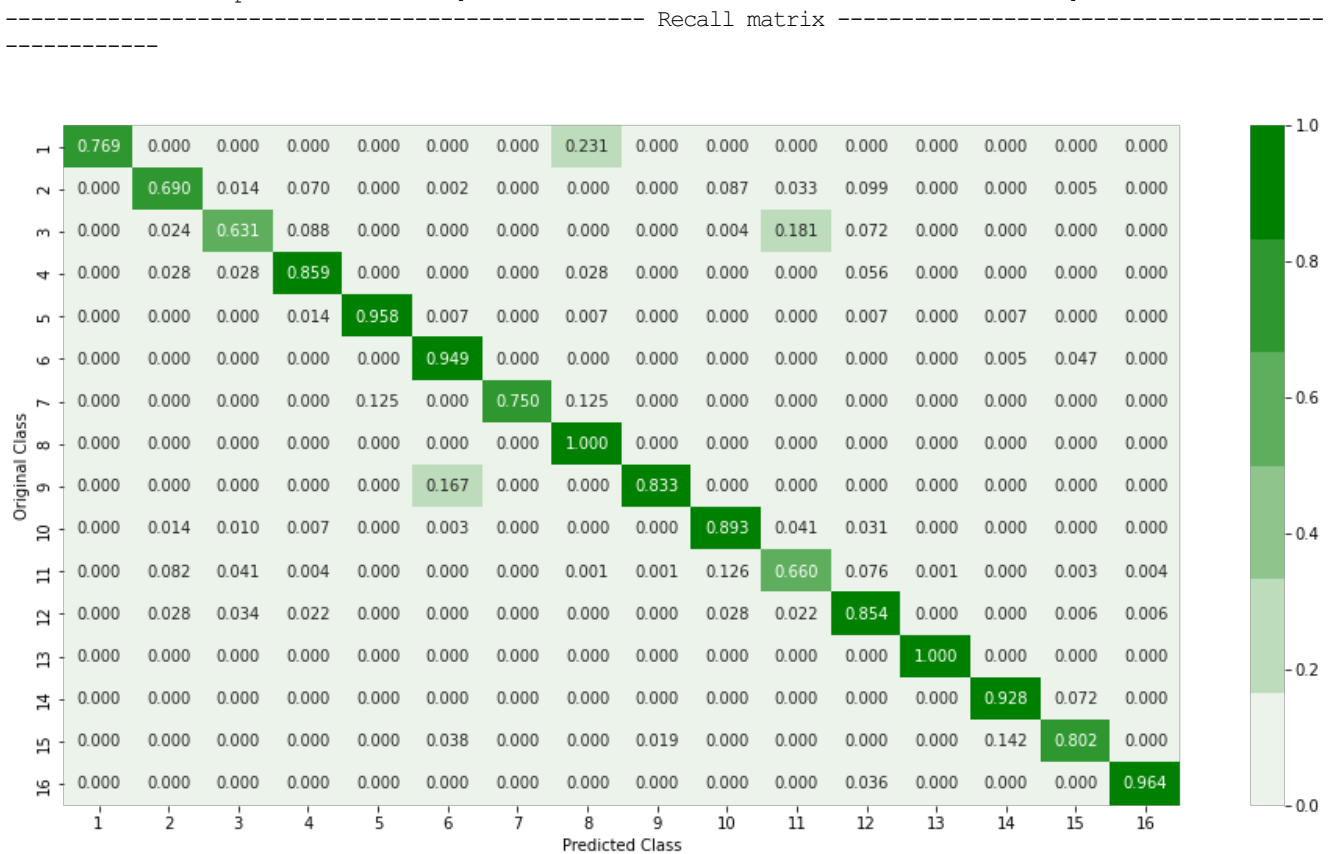
Confusion / Precision / Recall matrix without class 0  
Number of misclassified points 20.104952443424075

----- Confusion matrix -----  
-----





Sum of columns in precision matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]



Sum of rows in recall matrix [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

micro F1 score ignoring class 0 : 0.7989504755657593

macro F1 score ignoring class 0 : 0.8328771622914929

Average Accuracy ignoring class 0 = 0.8463051401778926

In [ ]:

```
y_test[y_test[0]!=0]
```

Out[ ]:

	0
546	11
10878	5
9176	2
5888	10
7486	10
...	...
6908	10
1394	15
8568	3
3658	15
12402	2

3075 rows × 1 columns

In [ ]:

```
kappa_score_16 = cohen_kappa_score(y_test[y_test[0]!=0],y_pred_pd_5_test[y_test[0]!=0])
```

In [ ]:

```
print('kappa_score for model 2 (16 classes ignoring class 0) = ', kappa_score_16)
```

kappa\_score for model 2 (16 classes ignoring class 0) = 0.7663891128653534