Node Classification using Graph Convolutional Networks

400

This node classification task uses CORA dataset from https://lings.soe.ucsc.edu/data

The dataset consists of 2708 nodes which correspond to scientific publications.

The nodes are classified into **7** categories indicating the topics of each document.

The edges indicate whether a document is cited by the other or vice versa.

Each node has 1433 features which is described by a 0/1-valued vector, indicating the bag-of-words from the dictionary.

This is an undirected graph problem

```
In [ ]: #importing dependencies
        import numpy as np
        import os
        import networkx as nx
        from keras.utils import to_categorical
        from sklearn.preprocessing import LabelEncoder
        from sklearn.utils import shuffle
        from sklearn.metrics import classification_report
        from spektral.layers import GraphConv
        from tensorflow.keras.models import Model
        from tensorflow.keras.layers import Input, Dropout, Dense
        from tensorflow.keras import Sequential
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.callbacks import TensorBoard, EarlyStopping
        import tensorflow as tf
        from tensorflow.keras.regularizers import 12
        from collections import Counter
        from sklearn.manifold import TSNE
        import matplotlib.pyplot as plt
```

Data Loading and Preprocessing

We are going to use the edges connecting the (from file cora.cites).

The nodes are loaded from file cora.content.

In cora.content file:

The **first** element indicates the **node name**

The **second** until the last second elements indicate the **node features**

The last element indicates the label of that particular node

In cora.cites file:

Each line indicates the tuple of connected nodes

Parsing the data

```
In [ ]: #parse the data
        labels = []
        nodes = []
        X = []
        for i,data in enumerate(all_data):
            elements = data.split('\t')
            labels.append(elements[-1])
            X.append(elements[1:-1])
            nodes.append(elements[0])
        X = np.array(X,dtype=int)
        N = X.shape[0] #the number of nodes
        F = X.shape[1] #the size of node features
        print('X shape: ', X.shape)
        #parse the edge
        edge_list=[]
        for edge in all_edges:
            e = edge.split('\t')
            edge_list.append((e[0],e[1]))
        print('\nNumber of nodes (N): ', N)
        print('\nNumber of features (F) of each node: ', F)
        print('\nCategories: ', set(labels))
        num_classes = len(set(labels))
        print('\nNumber of classes: ', num_classes)
        X shape: (2708, 1433)
        Number of nodes (N): 2708
        Number of features (F) of each node: 1433
        Categories: {'Reinforcement_Learning', 'Neural_Networks', 'Rule_Learning', 'Probabilistic_Methods', 'Genetic
        _Algorithms', 'Case_Based', 'Theory'}
        Number of classes: 7
```

Select examples for training, validation, and test then set the mask

```
#get the indices that do not go to traning data
    rest_idx = [x for x in range(len(labels)) if x not in train_idx]
    #get the first val_num
    val_idx = rest_idx[:val_num]
    test_idx = rest_idx[val_num:(val_num+test_num)]
    return train_idx, val_idx,test_idx

train_idx,val_idx,test_idx = limit_data(labels)

In []: #set the mask
    train_mask = np.zeros((N,),dtype=bool)
    train_mask[train_idx] = True

    val_mask = np.zeros((N,),dtype=bool)
    val_mask[val_idx] = True

test_mask = np.zeros((N,),dtype=bool)
    test_mask[test_idx] = True
```

Show Data Distribution

```
In []: print("All Data Distribution: \n{}".format(Counter(labels)))

All Data Distribution:
    Counter({'Neural_Networks': 818, 'Probabilistic_Methods': 426, 'Genetic_Algorithms': 418, 'Theory': 351, 'Case_Based': 298, 'Reinforcement_Learning': 217, 'Rule_Learning': 180})

In []: print("Training Data Distribution: \n{}".format(Counter([labels[i] for i in train_idx])))

    Training Data Distribution:
    Counter({'Reinforcement_Learning': 20, 'Probabilistic_Methods': 20, 'Neural_Networks': 20, 'Case_Based': 20, 'Theory': 20, 'Genetic_Algorithms': 20, 'Rule_Learning': 20})

In []: print("Validation Data Distribution: \n{}".format(Counter([labels[i] for i in val_idx])))

Validation Data Distribution:
    Counter({'Neural_Networks': 172, 'Genetic_Algorithms': 78, 'Probabilistic_Methods': 72, 'Theory': 63, 'Case_B ased': 58, 'Reinforcement_Learning': 35, 'Rule_Learning': 22})
```

Convert the labels to one hot encoding

```
In [ ]: def encode_label(labels):
    label_encoder = LabelEncoder()
    labels = label_encoder.fit_transform(labels)
    labels = to_categorical(labels)
    return labels, label_encoder.classes_
labels_encoded, classes = encode_label(labels)
```

Build a graph on NetworkX using the obtained nodes and edges list

```
In []: #build the graph
    G = nx.Graph()
    G.add_nodes_from(nodes)
    G.add_edges_from(edge_list)

#obtain the adjacency matrix (A)
    A = nx.adjacency_matrix(G)
    print('Graph info: ', nx.info(G))
    print('Adjacency matrix:')
    print(A)
```

```
Graph info: Name:
Type: Graph
Number of nodes: 2708
Number of edges: 5278
Average degree: 3.8981
Adjacency matrix:
  (0, 79)
              1
  (0, 1537)
               1
  (0, 2149)
               1
  (0, 2160)
  (0, 2675)
  (0, 2692)
               1
  (1, 108)
  (1, 486)
  (1, 1987)
  (1, 2126)
  (1, 2133)
               1
  (2, 2526)
               1
  (2, 2677)
               1
  (3, 202)
               1
  (3, 352)
  (3, 1522)
  (3, 2487)
               1
  (4, 285)
  (4, 1806)
  (5, 34)
  (5, 317)
  (5, 394)
  (5, 714)
               1
  (5, 2679)
               1
  (6, 13)
  (2698, 715) 1
  (2699, 1258) 1
  (2699, 1656) 1
  (2699, 2632) 1
  (2700, 419)
  (2700, 1078) 1
  (2700, 2526) 1
  (2701, 714)
              1
  (2701, 2539) 1
  (2701, 2634) 1
  (2701, 2679) 1
  (2702, 644)
  (2702, 937)
               1
  (2702, 1932) 1
  (2703, 1692) 1
  (2703, 1811) 1
  (2704, 294) 1
  (2705, 1839) 1
  (2706, 643) 1
  (2706, 1421) 1
  (2706, 1630) 1
  (2707, 1348) 1
  (2707, 1431) 1
  (2707, 1939) 1
  (2707, 2380) 1
```

Building and Training Graph Convolutional Networks

```
# Model definition
       X_in = Input(shape=(F, ))
       fltr_in = Input((N, ), sparse=True)
       dropout_1 = Dropout(dropout)(X_in)
        graph_conv_1 = GraphConv(channels,
                               activation='relu',
                               kernel_regularizer=12(12_reg),
                               use_bias=False)([dropout_1, fltr_in])
       dropout_2 = Dropout(dropout)(graph_conv_1)
        graph_conv_2 = GraphConv(num_classes,
                              activation='softmax',
                              use_bias=False)([dropout_2, fltr_in])
        # Build model
       model = Model(inputs=[X_in, fltr_in], outputs=graph_conv_2)
       optimizer = Adam(lr=learning_rate)
        model.compile(optimizer=optimizer,
                     loss='categorical_crossentropy',
                     weighted_metrics=['acc'])
        model.summary()
       tbCallBack_GCN = tf.keras.callbacks.TensorBoard(
           log_dir='./Tensorboard_GCN_cora',
        callback_GCN = [tbCallBack_GCN]
       Model: "model"
       Layer (type)
                                     Output Shape
                                                        Param #
                                                                   Connected to
        ______
        input_1 (InputLayer)
                                     [(None, 1433)]
                                                        0
       dropout (Dropout)
                                     (None, 1433)
                                                        0
                                                                    input_1[0][0]
       input_2 (InputLayer)
                                     [(None, 2708)]
                                                        0
       graph_conv (GraphConv)
                                     (None, 16)
                                                        22928
                                                                    dropout[0][0]
                                                                    input_2[0][0]
       dropout_1 (Dropout)
                                     (None, 16)
                                                        0
                                                                    graph_conv[0][0]
                                     (None, 7)
       graph_conv_1 (GraphConv)
                                                        112
                                                                    dropout_1[0][0]
                                                                   input 2[0][0]
        _____
       Total params: 23,040
       Trainable params: 23,040
       Non-trainable params: 0
In [ ]: # Train model
       validation_data = ([X, A], labels_encoded, val_mask)
       model.fit([X, A],
                 labels_encoded,
                 sample_weight=train_mask,
                 epochs=epochs,
                 batch_size=N,
                 validation_data=validation_data,
                 shuffle=False,
                     EarlyStopping(patience=es_patience, restore_best_weights=True),
                     tbCallBack_GCN
                 ])
```

```
Epoch 1/400
1/1 [=============] - 0s 317ms/step - loss: 0.1165 - acc: 0.1857 - val_loss: 0.3666 - val_ac
c: 0.3600
Epoch 2/400
n_batch_end) is slow compared to the batch update (0.153413). Check your callbacks.
c: 0.4420
Epoch 3/400
c: 0.4700
Epoch 4/400
1/1 [============= ] - 0s 155ms/step - loss: 0.0956 - acc: 0.6000 - val_loss: 0.3327 - val_ac
c: 0.4720
Epoch 5/400
1/1 [============= ] - 0s 156ms/step - loss: 0.0906 - acc: 0.6071 - val_loss: 0.3228 - val_ac
c: 0.4820
Epoch 6/400
1/1 [=========== ] - 0s 164ms/step - loss: 0.0862 - acc: 0.6357 - val loss: 0.3129 - val ac
c: 0.4960
Epoch 7/400
c: 0.5100
Fnoch 8/400
c: 0.5480
Epoch 9/400
c: 0.6080
Epoch 10/400
c: 0.6380
Epoch 11/400
c: 0.6700
Epoch 12/400
1/1 [==========] - 0s 158ms/step - loss: 0.0692 - acc: 0.8429 - val_loss: 0.2649 - val_ac
c: 0.6960
Epoch 13/400
c: 0.7160
Epoch 14/400
1/1 [============= ] - 0s 150ms/step - loss: 0.0645 - acc: 0.8714 - val_loss: 0.2515 - val_ac
c: 0.7320
Epoch 15/400
c: 0.7420
Epoch 16/400
c: 0.7480
Epoch 17/400
1/1 [=========== ] - 0s 155ms/step - loss: 0.0623 - acc: 0.8714 - val loss: 0.2319 - val ac
c: 0.7540
Epoch 18/400
c: 0.7640
Epoch 19/400
1/1 [=========== ] - 0s 151ms/step - loss: 0.0631 - acc: 0.8857 - val loss: 0.2220 - val ac
c: 0.7720
Epoch 20/400
c: 0.7800
Epoch 21/400
1/1 [============= ] - 0s 157ms/step - loss: 0.0594 - acc: 0.8714 - val_loss: 0.2147 - val_ac
c: 0.7740
Epoch 22/400
c: 0.7700
Epoch 23/400
1/1 [============] - 0s 149ms/step - loss: 0.0568 - acc: 0.8857 - val_loss: 0.2086 - val_ac
c: 0.7740
```

```
Epoch 24/400
1/1 [============= ] - 0s 156ms/step - loss: 0.0583 - acc: 0.8643 - val_loss: 0.2060 - val_ac
c: 0.7800
Epoch 25/400
c: 0.7880
Epoch 26/400
1/1 [==========] - 0s 154ms/step - loss: 0.0509 - acc: 0.9429 - val_loss: 0.2005 - val_ac
c: 0.7900
Epoch 27/400
c: 0.7900
Epoch 28/400
1/1 [=========== ] - 0s 152ms/step - loss: 0.0517 - acc: 0.9286 - val loss: 0.1949 - val ac
c: 0.7900
Epoch 29/400
c: 0.7880
Epoch 30/400
1/1 [============= ] - 0s 152ms/step - loss: 0.0481 - acc: 0.9357 - val_loss: 0.1908 - val_ac
c: 0.7900
Epoch 31/400
c: 0.7900
Epoch 32/400
c: 0.7820
Epoch 33/400
c: 0.7800
Epoch 34/400
1/1 [============= ] - 0s 147ms/step - loss: 0.0492 - acc: 0.9143 - val_loss: 0.1849 - val_ac
c: 0.7800
Epoch 35/400
c: 0.7840
Epoch 36/400
c: 0.7880
Epoch 37/400
c: 0.7860
Epoch 38/400
c: 0.7900
Epoch 39/400
1/1 [=========== ] - 0s 147ms/step - loss: 0.0436 - acc: 0.9500 - val loss: 0.1754 - val ac
c: 0.7860
Epoch 40/400
c: 0.7860
Epoch 41/400
1/1 [===========] - 0s 159ms/step - loss: 0.0430 - acc: 0.9500 - val_loss: 0.1739 - val_ac
c: 0.7840
Epoch 42/400
c: 0.7820
Epoch 43/400
c: 0.7760
Epoch 44/400
c: 0.7780
Epoch 45/400
c: 0.7760
Epoch 46/400
c: 0.7760
Epoch 47/400
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```
c: 0.7820
Epoch 48/400
c: 0.7820
Epoch 49/400
1/1 [============= ] - 0s 148ms/step - loss: 0.0389 - acc: 0.9571 - val_loss: 0.1680 - val_ac
c: 0.7840
Epoch 50/400
1/1 [============= ] - 0s 149ms/step - loss: 0.0381 - acc: 0.9571 - val_loss: 0.1667 - val_ac
c: 0.7880
Epoch 51/400
c: 0.7900
Epoch 52/400
1/1 [===========] - 0s 212ms/step - loss: 0.0384 - acc: 0.9714 - val_loss: 0.1643 - val_ac
c: 0.7840
Epoch 53/400
c: 0.7760
Epoch 54/400
c: 0.7860
Epoch 55/400
1/1 [============ ] - 0s 166ms/step - loss: 0.0376 - acc: 0.9643 - val loss: 0.1601 - val ac
c: 0.7820
Epoch 56/400
c: 0.7840
Epoch 57/400
c: 0.7860
Epoch 58/400
c: 0.7880
Epoch 59/400
c: 0.7920
Epoch 60/400
c: 0.7960
Epoch 61/400
c: 0.7960
Epoch 62/400
c: 0.7960
Epoch 63/400
1/1 [============= ] - 0s 149ms/step - loss: 0.0351 - acc: 0.9714 - val_loss: 0.1583 - val_ac
c: 0.7940
Epoch 64/400
c: 0.7980
Epoch 65/400
c: 0.7940
Epoch 66/400
1/1 [============= ] - 0s 169ms/step - loss: 0.0352 - acc: 0.9643 - val_loss: 0.1589 - val_ac
c: 0.7880
Epoch 67/400
c: 0.7880
Epoch 68/400
c: 0.7940
Epoch 69/400
c: 0.7900
Epoch 70/400
1/1 [============= ] - 0s 165ms/step - loss: 0.0338 - acc: 0.9571 - val_loss: 0.1558 - val_ac
c: 0.7920
Epoch 71/400
```

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```
c: 0.7880
Epoch 72/400
1/1 [============= ] - 0s 147ms/step - loss: 0.0335 - acc: 0.9714 - val_loss: 0.1554 - val_ac
c: 0.7920
Epoch 73/400
c: 0.7920
Epoch 74/400
c: 0.7880
Epoch 75/400
1/1 [============= ] - 0s 151ms/step - loss: 0.0339 - acc: 0.9643 - val_loss: 0.1521 - val_ac
c: 0.7980
Epoch 76/400
1/1 [============= ] - 0s 164ms/step - loss: 0.0324 - acc: 0.9571 - val_loss: 0.1516 - val_ac
c: 0.8020
Epoch 77/400
1/1 [=========== ] - 0s 149ms/step - loss: 0.0321 - acc: 0.9714 - val loss: 0.1515 - val ac
c: 0.8040
Epoch 78/400
c: 0.8040
Fnoch 79/400
c: 0.7960
Epoch 80/400
c: 0.7900
Epoch 81/400
c: 0.7880
Epoch 82/400
c: 0.7880
Epoch 83/400
1/1 [============= ] - 0s 147ms/step - loss: 0.0339 - acc: 0.9429 - val_loss: 0.1505 - val_ac
c: 0.7940
Epoch 84/400
c: 0.7960
Epoch 85/400
1/1 [============= ] - 0s 148ms/step - loss: 0.0310 - acc: 0.9714 - val_loss: 0.1486 - val_ac
c: 0.7940
Epoch 86/400
c: 0.7920
Epoch 87/400
c: 0.7880
Epoch 88/400
1/1 [=========== ] - 0s 149ms/step - loss: 0.0307 - acc: 0.9643 - val loss: 0.1501 - val ac
c: 0.7800
Epoch 89/400
c: 0.7760
Epoch 90/400
1/1 [=========== ] - 0s 151ms/step - loss: 0.0310 - acc: 0.9857 - val loss: 0.1517 - val ac
c: 0.7760
Epoch 91/400
c: 0.7760
Epoch 92/400
1/1 [============= ] - 0s 152ms/step - loss: 0.0312 - acc: 0.9571 - val_loss: 0.1518 - val_ac
c: 0.7760
Epoch 93/400
1/1 [============= ] - 0s 146ms/step - loss: 0.0303 - acc: 0.9643 - val_loss: 0.1513 - val_ac
c: 0.7740
Epoch 94/400
c: 0.7820
```

```
Epoch 95/400
1/1 [============= ] - 0s 151ms/step - loss: 0.0295 - acc: 0.9857 - val_loss: 0.1506 - val_ac
c: 0.7820
Epoch 96/400
c: 0.7800
Epoch 97/400
1/1 [============= ] - 0s 154ms/step - loss: 0.0298 - acc: 0.9714 - val_loss: 0.1499 - val_ac
c: 0.7800
Epoch 98/400
c: 0.7780
Epoch 99/400
1/1 [=========== ] - 0s 150ms/step - loss: 0.0305 - acc: 0.9500 - val loss: 0.1487 - val ac
c: 0.7780
Epoch 100/400
c: 0.7800
Epoch 101/400
1/1 [==========] - 0s 150ms/step - loss: 0.0307 - acc: 0.9643 - val_loss: 0.1484 - val_ac
c: 0.7760
Epoch 102/400
c: 0.7820
Epoch 103/400
c: 0.7760
Epoch 104/400
c: 0.7700
Epoch 105/400
1/1 [==========] - 0s 146ms/step - loss: 0.0309 - acc: 0.9500 - val_loss: 0.1493 - val_ac
c: 0.7760
Epoch 106/400
c: 0.7780
Epoch 107/400
c: 0.7760
Epoch 108/400
c: 0.7800
Epoch 109/400
c: 0.7840
Epoch 110/400
1/1 [=========== ] - 0s 175ms/step - loss: 0.0304 - acc: 0.9357 - val loss: 0.1442 - val ac
c: 0.7860
Epoch 111/400
c: 0.7920
Epoch 112/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0287 - acc: 0.9357 - val_loss: 0.1434 - val_ac
c: 0.7920
Epoch 113/400
c: 0.7840
Epoch 114/400
c: 0.7760
Epoch 115/400
c: 0.7680
Epoch 116/400
c: 0.7700
Epoch 117/400
c: 0.7640
Epoch 118/400
1/1 [==================] - 0s 159ms/step - loss: 0.0285 - acc: 0.9714 - val_loss: 0.1545 - val_ac
```

```
c: 0.7660
Epoch 119/400
c: 0.7680
Epoch 120/400
1/1 [============= ] - 0s 176ms/step - loss: 0.0278 - acc: 0.9857 - val_loss: 0.1520 - val_ac
c: 0.7660
Epoch 121/400
1/1 [============ ] - 0s 153ms/step - loss: 0.0284 - acc: 0.9786 - val_loss: 0.1487 - val_ac
c: 0.7720
Epoch 122/400
c: 0.7720
Epoch 123/400
1/1 [============= ] - 0s 161ms/step - loss: 0.0245 - acc: 1.0000 - val_loss: 0.1426 - val_ac
c: 0.7740
Epoch 124/400
c: 0.7860
Epoch 125/400
c: 0.7840
Epoch 126/400
1/1 [=========== ] - 0s 160ms/step - loss: 0.0282 - acc: 0.9571 - val loss: 0.1442 - val ac
c: 0.7740
Epoch 127/400
c: 0.7760
Epoch 128/400
c: 0.7740
Epoch 129/400
c: 0.7740
Epoch 130/400
c: 0.7760
Epoch 131/400
c: 0.7760
Epoch 132/400
c: 0.7820
Epoch 133/400
c: 0.7860
Epoch 134/400
1/1 [============= ] - 0s 155ms/step - loss: 0.0251 - acc: 0.9714 - val_loss: 0.1392 - val_ac
c: 0.7900
Epoch 135/400
c: 0.7920
Epoch 136/400
c: 0.8000
Epoch 137/400
1/1 [============= ] - 0s 152ms/step - loss: 0.0261 - acc: 0.9571 - val_loss: 0.1388 - val_ac
c: 0.7920
Epoch 138/400
c: 0.7880
Epoch 139/400
1/1 [=========== ] - 0s 152ms/step - loss: 0.0263 - acc: 0.9429 - val loss: 0.1420 - val ac
c: 0.7700
Epoch 140/400
c: 0.7720
Epoch 141/400
1/1 [============= ] - 0s 149ms/step - loss: 0.0277 - acc: 0.9357 - val_loss: 0.1467 - val_ac
c: 0.7700
Epoch 142/400
```

```
c: 0.7640
Epoch 143/400
1/1 [============= ] - 0s 148ms/step - loss: 0.0252 - acc: 0.9643 - val_loss: 0.1485 - val_ac
c: 0.7700
Epoch 144/400
c: 0.7700
Epoch 145/400
c: 0.7740
Epoch 146/400
1/1 [============= ] - 0s 162ms/step - loss: 0.0258 - acc: 0.9643 - val_loss: 0.1458 - val_ac
c: 0.7800
Epoch 147/400
1/1 [============= ] - 0s 149ms/step - loss: 0.0262 - acc: 0.9857 - val_loss: 0.1448 - val_ac
c: 0.7800
Epoch 148/400
1/1 [=========== ] - 0s 152ms/step - loss: 0.0283 - acc: 0.9643 - val loss: 0.1435 - val ac
c: 0.7900
Epoch 149/400
c: 0.7980
Fnoch 150/400
c: 0.7920
Epoch 151/400
c: 0.7860
Epoch 152/400
c: 0.7800
Epoch 153/400
c: 0.7700
Epoch 154/400
c: 0.7700
Epoch 155/400
c: 0.7680
Epoch 156/400
1/1 [============= ] - 0s 158ms/step - loss: 0.0240 - acc: 0.9857 - val_loss: 0.1460 - val_ac
c: 0.7640
Epoch 157/400
1/1 [============= ] - 0s 169ms/step - loss: 0.0256 - acc: 0.9643 - val_loss: 0.1438 - val_ac
c: 0.7660
Epoch 158/400
c: 0.7740
Epoch 159/400
1/1 [============ ] - 0s 151ms/step - loss: 0.0264 - acc: 0.9786 - val loss: 0.1394 - val ac
c: 0.7760
Epoch 160/400
c: 0.7920
Epoch 161/400
1/1 [=========== ] - 0s 157ms/step - loss: 0.0238 - acc: 0.9643 - val loss: 0.1378 - val ac
c: 0.8000
Epoch 162/400
c: 0.8020
Epoch 163/400
1/1 [===========] - 0s 154ms/step - loss: 0.0243 - acc: 0.9643 - val_loss: 0.1400 - val_ac
c: 0.7940
Epoch 164/400
1/1 [============= ] - 0s 159ms/step - loss: 0.0278 - acc: 0.9643 - val_loss: 0.1413 - val_ac
c: 0.7900
Epoch 165/400
1/1 [============] - 0s 150ms/step - loss: 0.0247 - acc: 0.9929 - val_loss: 0.1424 - val_ac
c: 0.7860
```

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Epoch 166/400
1/1 [============= ] - 0s 167ms/step - loss: 0.0252 - acc: 0.9929 - val_loss: 0.1421 - val_ac
c: 0.7920
Epoch 167/400
c: 0.7960
Epoch 168/400
1/1 [============= ] - 0s 150ms/step - loss: 0.0235 - acc: 0.9857 - val_loss: 0.1426 - val_ac
c: 0.7900
Epoch 169/400
1/1 [============= ] - 0s 151ms/step - loss: 0.0271 - acc: 0.9786 - val_loss: 0.1424 - val_ac
c: 0.7840
Epoch 170/400
1/1 [=========== ] - 0s 150ms/step - loss: 0.0266 - acc: 0.9714 - val loss: 0.1419 - val ac
c: 0.7820
Epoch 171/400
c: 0.7820
Epoch 172/400
1/1 [============= ] - 0s 158ms/step - loss: 0.0250 - acc: 0.9643 - val_loss: 0.1399 - val_ac
c: 0.7860
Epoch 173/400
c: 0.7860
Epoch 174/400
c: 0.7880
Epoch 175/400
c: 0.7800
Epoch 176/400
1/1 [============= ] - 0s 168ms/step - loss: 0.0221 - acc: 0.9857 - val_loss: 0.1465 - val_ac
c: 0.7740
Epoch 177/400
c: 0.7820
Epoch 178/400
c: 0.7780
Epoch 179/400
c: 0.7820
Epoch 180/400
c: 0.7840
Epoch 181/400
1/1 [=========== ] - 0s 173ms/step - loss: 0.0246 - acc: 0.9714 - val loss: 0.1376 - val ac
c: 0.7900
Epoch 182/400
c: 0.7880
Epoch 183/400
1/1 [============= ] - 0s 161ms/step - loss: 0.0219 - acc: 0.9929 - val_loss: 0.1360 - val_ac
c: 0.7960
Epoch 184/400
c: 0.7980
Epoch 185/400
c: 0.7960
Epoch 186/400
c: 0.7900
Epoch 187/400
c: 0.7920
Epoch 188/400
c: 0.7800
Epoch 189/400
1/1 [=================] - 0s 154ms/step - loss: 0.0235 - acc: 0.9857 - val_loss: 0.1439 - val_ac
```

```
c: 0.7760
Epoch 190/400
c: 0.7840
Epoch 191/400
1/1 [===========] - 0s 188ms/step - loss: 0.0251 - acc: 0.9643 - val_loss: 0.1422 - val_ac
c: 0.7920
Epoch 192/400
1/1 [============= ] - 0s 213ms/step - loss: 0.0234 - acc: 0.9571 - val_loss: 0.1414 - val_ac
c: 0.7900
Epoch 193/400
c: 0.7840
Epoch 194/400
1/1 [============= ] - 0s 247ms/step - loss: 0.0219 - acc: 0.9786 - val_loss: 0.1423 - val_ac
c: 0.7860
Epoch 195/400
c: 0.7840
Epoch 196/400
c: 0.7800
Epoch 197/400
c: 0.7840
Epoch 198/400
c: 0.7880
Epoch 199/400
c: 0.7920
Epoch 200/400
c: 0.7880
Epoch 201/400
c: 0.7920
Epoch 202/400
c: 0.7940
Epoch 203/400
c: 0.8000
Epoch 204/400
c: 0.8000
Epoch 205/400
1/1 [============= ] - 0s 199ms/step - loss: 0.0218 - acc: 0.9714 - val_loss: 0.1344 - val_ac
c: 0.7900
Epoch 206/400
c: 0.7880
Epoch 207/400
c: 0.7880
Epoch 208/400
1/1 [============= ] - 0s 274ms/step - loss: 0.0225 - acc: 0.9929 - val_loss: 0.1367 - val_ac
c: 0.7860
Epoch 209/400
c: 0.7820
Epoch 210/400
1/1 [=========== ] - 0s 220ms/step - loss: 0.0215 - acc: 0.9929 - val loss: 0.1387 - val ac
c: 0.7840
Epoch 211/400
c: 0.7840
Epoch 212/400
1/1 [============= ] - 0s 193ms/step - loss: 0.0219 - acc: 0.9786 - val_loss: 0.1410 - val_ac
c: 0.7860
Epoch 213/400
```

```
c: 0.7820
Epoch 214/400
1/1 [============ ] - 0s 201ms/step - loss: 0.0219 - acc: 0.9643 - val_loss: 0.1416 - val_ac
c: 0.7800
Epoch 215/400
c: 0.7760
Epoch 216/400
c: 0.7720
Epoch 217/400
1/1 [============= ] - 0s 196ms/step - loss: 0.0254 - acc: 0.9500 - val_loss: 0.1405 - val_ac
c: 0.7700
Epoch 218/400
1/1 [============= ] - 0s 173ms/step - loss: 0.0219 - acc: 0.9786 - val_loss: 0.1400 - val_ac
c: 0.7720
Epoch 219/400
1/1 [=========== ] - 0s 166ms/step - loss: 0.0203 - acc: 1.0000 - val loss: 0.1413 - val ac
c: 0.7760
Epoch 220/400
c: 0.7640
Fnoch 221/400
c: 0.7600
Epoch 222/400
c: 0.7620
Epoch 223/400
c: 0.7640
Epoch 224/400
c: 0.7640
Epoch 225/400
c: 0.7720
Epoch 226/400
c: 0.7740
Epoch 227/400
1/1 [============== ] - 0s 155ms/step - loss: 0.0210 - acc: 0.9857 - val_loss: 0.1347 - val_ac
c: 0.7740
Epoch 228/400
c: 0.7720
Epoch 229/400
c: 0.7760
Epoch 230/400
1/1 [=========== ] - 0s 166ms/step - loss: 0.0217 - acc: 0.9929 - val loss: 0.1340 - val ac
c: 0.7760
Epoch 231/400
c: 0.7760
Epoch 232/400
1/1 [=========== ] - 0s 186ms/step - loss: 0.0217 - acc: 0.9786 - val loss: 0.1374 - val ac
c: 0.7800
Epoch 233/400
c: 0.7720
Epoch 234/400
1/1 [==========] - 0s 180ms/step - loss: 0.0203 - acc: 0.9786 - val_loss: 0.1424 - val_ac
c: 0.7760
Epoch 235/400
c: 0.7780
Epoch 236/400
1/1 [============] - 0s 169ms/step - loss: 0.0208 - acc: 0.9857 - val_loss: 0.1426 - val_ac
c: 0.7780
```

```
Epoch 237/400
1/1 [============= ] - 0s 177ms/step - loss: 0.0200 - acc: 0.9929 - val_loss: 0.1410 - val_ac
c: 0.7760
Epoch 238/400
c: 0.7760
Epoch 239/400
1/1 [============ ] - 0s 164ms/step - loss: 0.0209 - acc: 0.9786 - val_loss: 0.1381 - val_ac
c: 0.7740
Epoch 240/400
1/1 [============= ] - 0s 153ms/step - loss: 0.0212 - acc: 0.9857 - val_loss: 0.1376 - val_ac
c: 0.7760
Epoch 241/400
1/1 [============ ] - 0s 192ms/step - loss: 0.0212 - acc: 0.9571 - val loss: 0.1357 - val ac
c: 0.7720
Epoch 242/400
c: 0.7740
Epoch 243/400
1/1 [============= ] - 0s 162ms/step - loss: 0.0211 - acc: 0.9929 - val_loss: 0.1331 - val_ac
c: 0.7700
Epoch 244/400
c: 0.7720
Epoch 245/400
c: 0.7760
Epoch 246/400
c: 0.7800
Epoch 247/400
1/1 [============= ] - 0s 161ms/step - loss: 0.0207 - acc: 0.9929 - val_loss: 0.1389 - val_ac
c: 0.7760
Epoch 248/400
c: 0.7780
Epoch 249/400
c: 0.7820
Epoch 250/400
c: 0.7860
Epoch 251/400
c: 0.7800
Epoch 252/400
1/1 [=========== ] - 0s 159ms/step - loss: 0.0241 - acc: 0.9357 - val loss: 0.1360 - val ac
c: 0.7740
Epoch 253/400
c: 0.7640
Epoch 254/400
1/1 [============= ] - 0s 152ms/step - loss: 0.0194 - acc: 0.9857 - val_loss: 0.1409 - val_ac
c: 0.7660
Epoch 255/400
c: 0.7660
Epoch 256/400
c: 0.7700
Epoch 257/400
c: 0.7700
Epoch 258/400
c: 0.7740
Epoch 259/400
c: 0.7700
Epoch 260/400
1/1 [=================] - 0s 156ms/step - loss: 0.0206 - acc: 0.9643 - val_loss: 0.1428 - val_ac
```

```
c: 0.7700
Epoch 261/400
c: 0.7700
Epoch 262/400
1/1 [============= ] - 0s 159ms/step - loss: 0.0218 - acc: 0.9714 - val_loss: 0.1413 - val_ac
c: 0.7720
Epoch 263/400
1/1 [============= ] - 0s 158ms/step - loss: 0.0212 - acc: 0.9857 - val_loss: 0.1415 - val_ac
c: 0.7700
Epoch 264/400
c: 0.7640
Epoch 265/400
1/1 [============ ] - 0s 150ms/step - loss: 0.0227 - acc: 0.9643 - val_loss: 0.1414 - val_ac
c: 0.7640
Epoch 266/400
c: 0.7560
Epoch 267/400
c: 0.7460
Epoch 268/400
1/1 [=========== ] - 0s 177ms/step - loss: 0.0227 - acc: 0.9571 - val loss: 0.1404 - val ac
c: 0.7540
Epoch 269/400
c: 0.7640
Epoch 270/400
1/1 [===========] - 0s 171ms/step - loss: 0.0220 - acc: 0.9786 - val_loss: 0.1364 - val_ac
c: 0.7640
Epoch 271/400
c: 0.7780
Epoch 272/400
1/1 [============= ] - 0s 157ms/step - loss: 0.0191 - acc: 0.9786 - val_loss: 0.1343 - val_ac
c: 0.7880
Epoch 273/400
c: 0.7940
Epoch 274/400
c: 0.7940
Epoch 275/400
c: 0.7840
Epoch 276/400
1/1 [============= ] - 0s 160ms/step - loss: 0.0210 - acc: 0.9714 - val_loss: 0.1366 - val_ac
c: 0.7680
Epoch 277/400
c: 0.7640
Epoch 278/400
1/1 [============== ] - 0s 170ms/step - loss: 0.0229 - acc: 0.9500 - val_loss: 0.1410 - val_ac
c: 0.7700
Epoch 279/400
1/1 [============= ] - 0s 158ms/step - loss: 0.0213 - acc: 0.9857 - val_loss: 0.1404 - val_ac
c: 0.7740
Epoch 280/400
c: 0.7780
Epoch 281/400
1/1 [=========== ] - 0s 152ms/step - loss: 0.0199 - acc: 1.0000 - val loss: 0.1388 - val ac
c: 0.7840
Epoch 282/400
c: 0.7840
Epoch 283/400
1/1 [============= ] - 0s 174ms/step - loss: 0.0221 - acc: 0.9571 - val_loss: 0.1373 - val_ac
c: 0.7880
Epoch 284/400
```

```
c: 0.7820
Epoch 285/400
c: 0.7740
Epoch 286/400
c: 0.7680
Epoch 287/400
c: 0.7640
Epoch 288/400
1/1 [============= ] - 0s 153ms/step - loss: 0.0207 - acc: 0.9786 - val_loss: 0.1374 - val_ac
c: 0.7740
Epoch 289/400
1/1 [============= ] - 0s 156ms/step - loss: 0.0211 - acc: 0.9786 - val_loss: 0.1384 - val_ac
c: 0.7740
Epoch 290/400
1/1 [=========== ] - 0s 173ms/step - loss: 0.0211 - acc: 0.9643 - val loss: 0.1405 - val ac
c: 0.7720
Epoch 291/400
c: 0.7720
Fnoch 292/400
c: 0.7700
Epoch 293/400
c: 0.7700
Epoch 294/400
c: 0.7740
Epoch 295/400
c: 0.7720
Epoch 296/400
c: 0.7720
Epoch 297/400
c: 0.7680
Epoch 298/400
1/1 [============= ] - 0s 151ms/step - loss: 0.0191 - acc: 0.9786 - val_loss: 0.1380 - val_ac
c: 0.7700
Epoch 299/400
c: 0.7680
Epoch 300/400
c: 0.7740
Epoch 301/400
1/1 [=========== ] - 0s 174ms/step - loss: 0.0210 - acc: 0.9857 - val loss: 0.1397 - val ac
c: 0.7640
Epoch 302/400
c: 0.7660
Epoch 303/400
1/1 [=========== ] - 0s 182ms/step - loss: 0.0181 - acc: 0.9857 - val loss: 0.1475 - val ac
c: 0.7720
Epoch 304/400
c: 0.7640
Epoch 305/400
1/1 [============= ] - 0s 188ms/step - loss: 0.0185 - acc: 0.9857 - val_loss: 0.1430 - val_ac
c: 0.7700
Epoch 306/400
1/1 [============= ] - 0s 179ms/step - loss: 0.0197 - acc: 1.0000 - val_loss: 0.1388 - val_ac
c: 0.7820
Epoch 307/400
1/1 [===========] - 0s 161ms/step - loss: 0.0204 - acc: 0.9714 - val_loss: 0.1342 - val_ac
c: 0.7800
```

```
Epoch 308/400
1/1 [============= ] - 0s 183ms/step - loss: 0.0193 - acc: 0.9929 - val_loss: 0.1317 - val_ac
c: 0.7780
Epoch 309/400
c: 0.7800
Epoch 310/400
1/1 [===========] - 0s 205ms/step - loss: 0.0186 - acc: 1.0000 - val_loss: 0.1301 - val_ac
c: 0.7760
Epoch 311/400
1/1 [============= ] - 0s 190ms/step - loss: 0.0211 - acc: 0.9714 - val_loss: 0.1312 - val_ac
c: 0.7720
Epoch 312/400
1/1 [=========== ] - 0s 192ms/step - loss: 0.0203 - acc: 0.9857 - val loss: 0.1333 - val ac
c: 0.7680
Epoch 313/400
c: 0.7660
Epoch 314/400
1/1 [===========] - 0s 164ms/step - loss: 0.0206 - acc: 0.9786 - val_loss: 0.1388 - val_ac
c: 0.7780
Epoch 315/400
c: 0.7680
Epoch 316/400
c: 0.7680
Epoch 317/400
c: 0.7760
Epoch 318/400
1/1 [============= ] - 0s 168ms/step - loss: 0.0204 - acc: 0.9786 - val_loss: 0.1394 - val_ac
c: 0.7820
Epoch 319/400
c: 0.7860
Epoch 320/400
c: 0.7900
Epoch 321/400
c: 0.7880
Epoch 322/400
c: 0.7900
Epoch 323/400
1/1 [============ ] - 0s 152ms/step - loss: 0.0201 - acc: 0.9643 - val loss: 0.1318 - val ac
c: 0.7920
Epoch 324/400
c: 0.7940
Epoch 325/400
1/1 [============= ] - 0s 170ms/step - loss: 0.0188 - acc: 0.9786 - val_loss: 0.1301 - val_ac
c: 0.7880
Epoch 326/400
c: 0.7820
Epoch 327/400
c: 0.7660
Epoch 328/400
c: 0.7600
Epoch 329/400
c: 0.7480
Epoch 330/400
c: 0.7560
Epoch 331/400
```

```
c: 0.7620
Epoch 332/400
c: 0.7700
Epoch 333/400
1/1 [===========] - 0s 159ms/step - loss: 0.0178 - acc: 1.0000 - val_loss: 0.1440 - val_ac
c: 0.7740
Epoch 334/400
1/1 [============= ] - 0s 171ms/step - loss: 0.0211 - acc: 0.9500 - val_loss: 0.1376 - val_ac
c: 0.7680
Epoch 335/400
c: 0.7820
Epoch 336/400
1/1 [===========] - 0s 160ms/step - loss: 0.0202 - acc: 0.9643 - val_loss: 0.1315 - val_ac
c: 0.7760
Epoch 337/400
c: 0.7800
Epoch 338/400
c: 0.7680
Epoch 339/400
c: 0.7640
Epoch 340/400
c: 0.7660
Epoch 341/400
c: 0.7700
Epoch 342/400
c: 0.7740
Epoch 343/400
c: 0.7780
Epoch 344/400
c: 0.7800
Epoch 345/400
c: 0.7820
Epoch 346/400
c: 0.7800
Epoch 347/400
1/1 [============= ] - 0s 156ms/step - loss: 0.0192 - acc: 0.9786 - val_loss: 0.1341 - val_ac
c: 0.7760
Epoch 348/400
c: 0.7720
Epoch 349/400
c: 0.7740
Epoch 350/400
1/1 [============= ] - 0s 183ms/step - loss: 0.0198 - acc: 0.9786 - val_loss: 0.1341 - val_ac
c: 0.7820
Epoch 351/400
c: 0.7760
Epoch 352/400
1/1 [=========== ] - 0s 178ms/step - loss: 0.0170 - acc: 0.9857 - val loss: 0.1376 - val ac
c: 0.7740
Epoch 353/400
c: 0.7800
Epoch 354/400
1/1 [============= ] - 0s 181ms/step - loss: 0.0202 - acc: 0.9500 - val_loss: 0.1394 - val_ac
c: 0.7740
Epoch 355/400
```

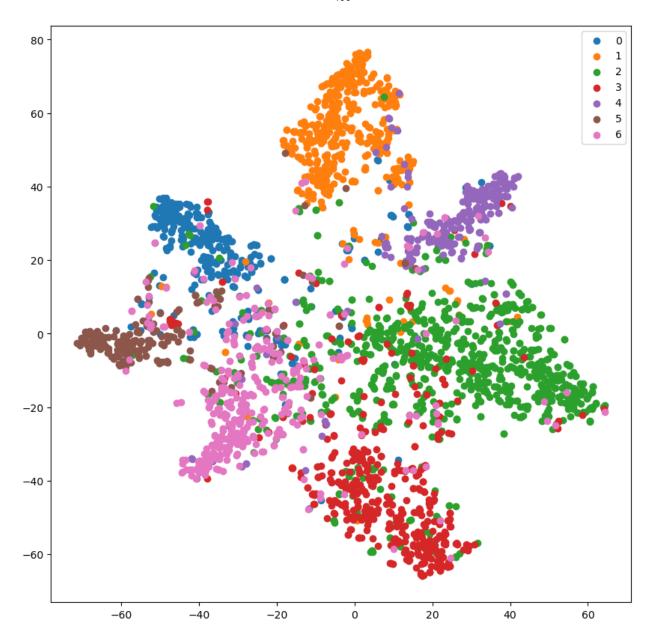
```
c: 0.7700
Epoch 356/400
1/1 [============ ] - 0s 168ms/step - loss: 0.0190 - acc: 0.9643 - val_loss: 0.1403 - val_ac
c: 0.7780
Epoch 357/400
c: 0.7680
Epoch 358/400
c: 0.7620
Epoch 359/400
1/1 [============ ] - 0s 174ms/step - loss: 0.0196 - acc: 0.9857 - val_loss: 0.1404 - val_ac
c: 0.7640
Epoch 360/400
1/1 [============= ] - 0s 173ms/step - loss: 0.0181 - acc: 0.9929 - val_loss: 0.1388 - val_ac
c: 0.7680
Epoch 361/400
1/1 [=========== ] - 0s 166ms/step - loss: 0.0191 - acc: 0.9857 - val loss: 0.1375 - val ac
c: 0.7720
Epoch 362/400
c: 0.7760
Fnoch 363/400
1/1 [============= ] - 0s 195ms/step - loss: 0.0182 - acc: 0.9857 - val_loss: 0.1390 - val_ac
c: 0.7800
Epoch 364/400
c: 0.7760
Epoch 365/400
c: 0.7800
Epoch 366/400
c: 0.7840
Epoch 367/400
c: 0.7860
Epoch 368/400
c: 0.7820
Epoch 369/400
1/1 [===========] - 0s 202ms/step - loss: 0.0186 - acc: 0.9786 - val_loss: 0.1312 - val_ac
c: 0.7840
Epoch 370/400
c: 0.7860
Epoch 371/400
c: 0.7840
Epoch 372/400
1/1 [=========== ] - 0s 183ms/step - loss: 0.0194 - acc: 0.9643 - val loss: 0.1339 - val ac
c: 0.7880
Epoch 373/400
c: 0.7840
Epoch 374/400
1/1 [=========== ] - 0s 168ms/step - loss: 0.0181 - acc: 0.9643 - val loss: 0.1376 - val ac
c: 0.7820
Epoch 375/400
c: 0.7760
Epoch 376/400
1/1 [===========] - 0s 168ms/step - loss: 0.0195 - acc: 0.9643 - val_loss: 0.1396 - val_ac
c: 0.7780
Epoch 377/400
1/1 [============= ] - 0s 162ms/step - loss: 0.0187 - acc: 0.9714 - val_loss: 0.1383 - val_ac
c: 0.7800
Epoch 378/400
1/1 [============] - 0s 173ms/step - loss: 0.0183 - acc: 0.9857 - val_loss: 0.1369 - val_ac
c: 0.7800
```

```
Epoch 379/400
    1/1 [============= ] - 0s 175ms/step - loss: 0.0174 - acc: 0.9857 - val_loss: 0.1366 - val_ac
    c: 0.7780
    Epoch 380/400
    1/1 [============] - 0s 173ms/step - loss: 0.0186 - acc: 0.9929 - val_loss: 0.1375 - val_ac
    c: 0.7740
    Epoch 381/400
    1/1 [============= ] - 0s 174ms/step - loss: 0.0176 - acc: 0.9929 - val_loss: 0.1372 - val_ac
    c: 0.7680
    Epoch 382/400
    1/1 [============= ] - 0s 177ms/step - loss: 0.0183 - acc: 0.9786 - val_loss: 0.1365 - val_ac
    c: 0.7680
    Epoch 383/400
    1/1 [=========== ] - 0s 160ms/step - loss: 0.0181 - acc: 0.9857 - val loss: 0.1349 - val ac
    c: 0.7720
    Epoch 384/400
    c: 0.7780
    Epoch 385/400
    1/1 [============= ] - 0s 167ms/step - loss: 0.0182 - acc: 0.9857 - val_loss: 0.1349 - val_ac
    c: 0.7720
    Epoch 386/400
    c: 0.7720
    Epoch 387/400
    c: 0.7660
    Epoch 388/400
    c: 0.7660
    Epoch 389/400
    1/1 [============= ] - 0s 167ms/step - loss: 0.0179 - acc: 0.9929 - val_loss: 0.1327 - val_ac
    c: 0.7660
    Epoch 390/400
    c: 0.7660
    Epoch 391/400
    c: 0.7680
    Epoch 392/400
    c: 0.7700
    Epoch 393/400
    c: 0.7720
    Epoch 394/400
    1/1 [=========== ] - 0s 173ms/step - loss: 0.0193 - acc: 0.9857 - val loss: 0.1353 - val ac
    c: 0.7660
    Epoch 395/400
    1/1 [============= ] - 0s 167ms/step - loss: 0.0175 - acc: 0.9786 - val_loss: 0.1392 - val_ac
    c: 0.7780
    Epoch 396/400
    1/1 [============ ] - 0s 170ms/step - loss: 0.0194 - acc: 0.9571 - val_loss: 0.1405 - val_ac
    c: 0.7780
    Epoch 397/400
    c: 0.7780
    Epoch 398/400
    c: 0.7660
    Epoch 399/400
    c: 0.7600
    Epoch 400/400
    c: 0.7700
Out[]: <tensorflow.python.keras.callbacks.History at 0x1cac77c8988>
In [ ]: # Evaluate model
    X_te = X[test_mask]
```

```
A_te = A[test_mask,:][:,test_mask]
y_te = labels_encoded[test_mask]
y_pred = model.predict([X_te, A_te], batch_size=N)
report = classification_report(np.argmax(y_te,axis=1), np.argmax(y_pred,axis=1), target_names=classes)
print('GCN Classification Report: \n {}'.format(report))
GCN Classification Report:
                       precision
                                   recall f1-score
                                                     support
                           0.74
                                    0.71
                                             0.72
           Case_Based
                                                        114
   Genetic_Algorithms
                           0.85
                                    0.88
                                              0.87
                                                        156
      Neural_Networks
                           0.75
                                    0.72
                                              0.73
                                                        290
 Probabilistic_Methods
                           0.82
                                    0.62
                                              0.71
                                                        172
Reinforcement_Learning
                          0.68
                                    0.78
                                             0.73
                                                         85
                           0.51
                                    0.82
        Rule_Learning
                                             0.62
                                                         60
                                    0.57
              Theory
                           0.56
                                             0.57
                                                        123
                                              0.72
                                                       1000
             accuracy
            macro avg
                           0.70
                                    0.73
                                              0.71
                                                       1000
                           0.73
                                    0.72
                                              0.72
                                                       1000
         weighted avg
```

Get hidden layer representation for GCN

```
In [ ]: layer_outputs = [layer.output for layer in model.layers]
        activation_model = Model(inputs=model.input, outputs=layer_outputs)
        activations = activation_model.predict([X,A],batch_size=N)
        #Get t-SNE Representation
        #get the hidden layer representation after the first GCN layer
        x_tsne = TSNE(n_components=2).fit_transform(activations[3])
In [ ]: def plot_tSNE(labels_encoded,x_tsne):
            color_map = np.argmax(labels_encoded, axis=1)
            plt.figure(figsize=(10,10))
            for cl in range(num classes):
                indices = np.where(color_map==cl)
                indices = indices[0]
                plt.scatter(x_tsne[indices,0], x_tsne[indices, 1], label=cl)
            plt.legend()
            plt.show()
        plot_tSNE(labels_encoded,x_tsne)
```



Comparison to Fully-Connected Neural Networks

Building and Training FNN

```
model_fnn.compile(optimizer=optimizer,
              loss='categorical_crossentropy',
              weighted_metrics=['acc'])
#define TensorBoard
tbCallBack_FNN = TensorBoard(
    log_dir='./Tensorboard_FNN_cora',
#Train model
validation_data_fnn = (X, labels_encoded, val_mask)
model_fnn.fit(
                X,labels_encoded,
                sample_weight=train_mask,
                epochs=epochs,
                batch_size=N,
                validation_data=validation_data_fnn,
                shuffle=False,
                callbacks=[
                  EarlyStopping(patience=es_patience, restore_best_weights=True),
                 tbCallBack_FNN
          ])
```

```
Epoch 1/400
1/1 [============] - 0s 295ms/step - loss: 0.2194 - acc: 0.1000 - val_loss: 0.4332 - val_ac
c: 0.3020
Epoch 2/400
n_batch_end) is slow compared to the batch update (0.149612). Check your callbacks.
c: 0.4300
Epoch 3/400
c: 0.5300
Epoch 4/400
1/1 [============= ] - 0s 171ms/step - loss: 0.1126 - acc: 0.6643 - val_loss: 0.3317 - val_ac
c: 0.5460
Epoch 5/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0875 - acc: 0.8214 - val_loss: 0.3026 - val_ac
c: 0.5500
Epoch 6/400
1/1 [=========== ] - 0s 174ms/step - loss: 0.0668 - acc: 0.9000 - val loss: 0.2765 - val ac
c: 0.5600
Epoch 7/400
1/1 [=============] - 0s 174ms/step - loss: 0.0503 - acc: 0.9429 - val_loss: 0.2610 - val_ac
c: 0.5620
Fnoch 8/400
c: 0.5820
Epoch 9/400
c: 0.5900
Epoch 10/400
c: 0.5800
Epoch 11/400
c: 0.5600
Epoch 12/400
c: 0.5660
Epoch 13/400
c: 0.5600
Epoch 14/400
1/1 [============= ] - 0s 172ms/step - loss: 0.0388 - acc: 0.9857 - val_loss: 0.3576 - val_ac
c: 0.5660
Epoch 15/400
1/1 [============ ] - 0s 172ms/step - loss: 0.0358 - acc: 0.9929 - val_loss: 0.3682 - val_ac
c: 0.5660
Epoch 16/400
c: 0.5800
Epoch 17/400
1/1 [=========== ] - 0s 181ms/step - loss: 0.0310 - acc: 0.9857 - val loss: 0.3705 - val ac
c: 0.5840
Epoch 18/400
c: 0.5760
Epoch 19/400
1/1 [=========== ] - 0s 178ms/step - loss: 0.0277 - acc: 0.9857 - val loss: 0.3936 - val ac
c: 0.5660
Epoch 20/400
c: 0.5440
Epoch 21/400
1/1 [============= ] - 0s 172ms/step - loss: 0.0256 - acc: 0.9643 - val_loss: 0.4114 - val_ac
c: 0.5380
Epoch 22/400
1/1 [============= ] - 0s 201ms/step - loss: 0.0242 - acc: 0.9714 - val_loss: 0.4263 - val_ac
c: 0.5180
Epoch 23/400
1/1 [===========] - 0s 189ms/step - loss: 0.0235 - acc: 0.9643 - val_loss: 0.4224 - val_ac
c: 0.5120
```

```
Epoch 24/400
c: 0.5120
Epoch 25/400
1/1 [===========] - 0s 188ms/step - loss: 0.0190 - acc: 0.9929 - val_loss: 0.3987 - val_ac
c: 0.4940
Epoch 26/400
1/1 [============= ] - 0s 187ms/step - loss: 0.0207 - acc: 0.9857 - val_loss: 0.3921 - val_ac
c: 0.4820
Epoch 27/400
1/1 [============= ] - 0s 180ms/step - loss: 0.0215 - acc: 0.9714 - val_loss: 0.3759 - val_ac
c: 0.4940
Epoch 28/400
1/1 [=========== ] - 0s 175ms/step - loss: 0.0205 - acc: 0.9786 - val loss: 0.3540 - val ac
c: 0.5060
Epoch 29/400
c: 0.5040
Epoch 30/400
1/1 [============= ] - 0s 183ms/step - loss: 0.0216 - acc: 0.9714 - val_loss: 0.3413 - val_ac
c: 0.5060
Epoch 31/400
c: 0.4980
Epoch 32/400
c: 0.4980
Epoch 33/400
c: 0.5080
Epoch 34/400
1/1 [============ ] - 0s 202ms/step - loss: 0.0210 - acc: 0.9929 - val_loss: 0.3237 - val_ac
c: 0.5160
Epoch 35/400
c: 0.5180
Epoch 36/400
c: 0.5240
Epoch 37/400
c: 0.5260
Epoch 38/400
c: 0.5240
Epoch 39/400
1/1 [=========== ] - 0s 171ms/step - loss: 0.0210 - acc: 1.0000 - val loss: 0.3181 - val ac
c: 0.5220
Epoch 40/400
c: 0.5200
Epoch 41/400
1/1 [============ ] - 0s 179ms/step - loss: 0.0203 - acc: 1.0000 - val_loss: 0.3248 - val_ac
c: 0.5180
Epoch 42/400
c: 0.5160
Epoch 43/400
c: 0.5180
Epoch 44/400
c: 0.5280
Epoch 45/400
c: 0.5440
Epoch 46/400
c: 0.5540
Epoch 47/400
```

```
c: 0.5720
Epoch 48/400
c: 0.5780
Epoch 49/400
1/1 [============= ] - 0s 176ms/step - loss: 0.0191 - acc: 0.9857 - val_loss: 0.3214 - val_ac
c: 0.5720
Epoch 50/400
1/1 [============ ] - 0s 178ms/step - loss: 0.0184 - acc: 0.9929 - val_loss: 0.3259 - val_ac
c: 0.5560
Epoch 51/400
c: 0.5420
Epoch 52/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0171 - acc: 0.9929 - val_loss: 0.3447 - val_ac
c: 0.5300
Epoch 53/400
c: 0.5160
Epoch 54/400
c: 0.5280
Epoch 55/400
c: 0.5320
Epoch 56/400
c: 0.5340
Epoch 57/400
c: 0.5260
Epoch 58/400
1/1 [============== ] - 0s 173ms/step - loss: 0.0194 - acc: 0.9643 - val_loss: 0.3555 - val_ac
c: 0.5200
Epoch 59/400
c: 0.5180
Epoch 60/400
c: 0.5100
Epoch 61/400
c: 0.5140
Epoch 62/400
c: 0.5040
Epoch 63/400
1/1 [============= ] - 0s 181ms/step - loss: 0.0174 - acc: 0.9786 - val_loss: 0.3682 - val_ac
c: 0.5060
Epoch 64/400
c: 0.4900
Epoch 65/400
1/1 [============== ] - 0s 180ms/step - loss: 0.0183 - acc: 0.9929 - val_loss: 0.4134 - val_ac
c: 0.4660
Epoch 66/400
1/1 [============ ] - 0s 201ms/step - loss: 0.0214 - acc: 0.9786 - val_loss: 0.4259 - val_ac
c: 0.4740
Epoch 67/400
c: 0.4920
Epoch 68/400
c: 0.5020
Epoch 69/400
c: 0.5020
Epoch 70/400
1/1 [============= ] - 0s 246ms/step - loss: 0.0217 - acc: 0.9929 - val_loss: 0.3641 - val_ac
c: 0.5180
Epoch 71/400
```

```
c: 0.5220
Epoch 72/400
1/1 [============ ] - 0s 163ms/step - loss: 0.0236 - acc: 0.9857 - val_loss: 0.3590 - val_ac
c: 0.5260
Epoch 73/400
c: 0.5340
Epoch 74/400
c: 0.5300
Epoch 75/400
1/1 [============= ] - 0s 167ms/step - loss: 0.0252 - acc: 0.9643 - val_loss: 0.3879 - val_ac
c: 0.5140
Epoch 76/400
1/1 [============ ] - 0s 165ms/step - loss: 0.0229 - acc: 1.0000 - val_loss: 0.4053 - val_ac
c: 0.5160
Epoch 77/400
1/1 [=========== ] - 0s 163ms/step - loss: 0.0245 - acc: 0.9786 - val loss: 0.4160 - val ac
c: 0.5060
Epoch 78/400
c: 0.5020
Fnoch 79/400
c: 0.5020
Epoch 80/400
c: 0.4980
Epoch 81/400
c: 0.4920
Epoch 82/400
c: 0.4880
Epoch 83/400
c: 0.4840
Epoch 84/400
c: 0.4960
Epoch 85/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0248 - acc: 0.9857 - val_loss: 0.3803 - val_ac
c: 0.5140
Epoch 86/400
c: 0.5240
Epoch 87/400
c: 0.5220
Epoch 88/400
1/1 [=========== ] - 0s 213ms/step - loss: 0.0266 - acc: 0.9714 - val loss: 0.3823 - val ac
c: 0.5200
Epoch 89/400
c: 0.5220
Epoch 90/400
1/1 [=========== ] - 0s 187ms/step - loss: 0.0251 - acc: 1.0000 - val loss: 0.4017 - val ac
c: 0.5320
Epoch 91/400
c: 0.5320
Epoch 92/400
1/1 [============ ] - 0s 204ms/step - loss: 0.0290 - acc: 0.9714 - val_loss: 0.3809 - val_ac
c: 0.5340
Epoch 93/400
1/1 [============= ] - 0s 181ms/step - loss: 0.0281 - acc: 0.9786 - val_loss: 0.3708 - val_ac
c: 0.5120
Epoch 94/400
1/1 [===========] - 0s 184ms/step - loss: 0.0280 - acc: 0.9786 - val_loss: 0.3732 - val_ac
c: 0.5200
```

```
Epoch 95/400
1/1 [============= ] - 0s 183ms/step - loss: 0.0280 - acc: 0.9786 - val_loss: 0.3835 - val_ac
c: 0.5020
Epoch 96/400
c: 0.4840
Epoch 97/400
1/1 [============= ] - 0s 185ms/step - loss: 0.0331 - acc: 0.9571 - val_loss: 0.3863 - val_ac
c: 0.4940
Epoch 98/400
c: 0.4940
Epoch 99/400
1/1 [=========== ] - 0s 188ms/step - loss: 0.0289 - acc: 0.9929 - val loss: 0.4000 - val ac
c: 0.5020
Epoch 100/400
c: 0.5060
Epoch 101/400
1/1 [============= ] - 0s 194ms/step - loss: 0.0281 - acc: 0.9857 - val_loss: 0.3857 - val_ac
c: 0.5140
Epoch 102/400
c: 0.5200
Epoch 103/400
c: 0.5220
Epoch 104/400
c: 0.5200
Epoch 105/400
1/1 [============ ] - 0s 180ms/step - loss: 0.0309 - acc: 0.9857 - val_loss: 0.4063 - val_ac
c: 0.5120
Epoch 106/400
1/1 [=========== ] - 0s 196ms/step - loss: 0.0288 - acc: 0.9929 - val loss: 0.4131 - val ac
c: 0.5080
Epoch 107/400
1/1 [=============== ] - 0s 184ms/step - loss: 0.0316 - acc: 0.9643 - val_loss: 0.4179 - val_ac
c: 0.4900
Epoch 108/400
c: 0.4820
Epoch 109/400
c: 0.4880
Epoch 110/400
1/1 [=========== ] - 0s 183ms/step - loss: 0.0333 - acc: 0.9643 - val loss: 0.4181 - val ac
c: 0.4920
Epoch 111/400
1/1 [============ ] - 0s 193ms/step - loss: 0.0320 - acc: 0.9714 - val_loss: 0.4136 - val_ac
c: 0.5040
Epoch 112/400
1/1 [============= ] - 0s 179ms/step - loss: 0.0319 - acc: 0.9929 - val_loss: 0.4083 - val_ac
c: 0.5000
Epoch 113/400
c: 0.5000
Epoch 114/400
c: 0.5080
Epoch 115/400
c: 0.5200
Epoch 116/400
c: 0.5300
Epoch 117/400
c: 0.5060
Epoch 118/400
1/1 [==================] - 0s 187ms/step - loss: 0.0306 - acc: 0.9786 - val_loss: 0.4230 - val_ac
```

```
c: 0.4940
Epoch 119/400
c: 0.4860
Epoch 120/400
1/1 [============= ] - 0s 174ms/step - loss: 0.0311 - acc: 0.9929 - val_loss: 0.4225 - val_ac
c: 0.4840
Epoch 121/400
1/1 [============= ] - 0s 179ms/step - loss: 0.0317 - acc: 0.9857 - val_loss: 0.4206 - val_ac
c: 0.4800
Epoch 122/400
c: 0.4860
Epoch 123/400
1/1 [============ ] - 0s 179ms/step - loss: 0.0327 - acc: 0.9786 - val_loss: 0.4155 - val_ac
c: 0.4800
Epoch 124/400
c: 0.5000
Epoch 125/400
c: 0.5080
Epoch 126/400
1/1 [=========== ] - 0s 194ms/step - loss: 0.0333 - acc: 0.9786 - val loss: 0.3811 - val ac
c: 0.5180
Epoch 127/400
c: 0.5360
Epoch 128/400
c: 0.5440
Epoch 129/400
1/1 [=============== ] - 0s 190ms/step - loss: 0.0318 - acc: 0.9857 - val_loss: 0.3764 - val_ac
c: 0.5460
Epoch 130/400
c: 0.5440
Epoch 131/400
c: 0.5440
Epoch 132/400
c: 0.5500
Epoch 133/400
c: 0.5620
Epoch 134/400
1/1 [============ ] - 0s 192ms/step - loss: 0.0333 - acc: 0.9643 - val_loss: 0.4050 - val_ac
c: 0.5540
Epoch 135/400
c: 0.5480
Epoch 136/400
c: 0.5400
Epoch 137/400
1/1 [============= ] - 0s 192ms/step - loss: 0.0329 - acc: 0.9857 - val_loss: 0.4317 - val_ac
c: 0.5340
Epoch 138/400
c: 0.5140
Epoch 139/400
1/1 [=========== ] - 0s 179ms/step - loss: 0.0331 - acc: 0.9786 - val loss: 0.4734 - val ac
c: 0.4980
Epoch 140/400
c: 0.4780
Epoch 141/400
1/1 [============= ] - 0s 176ms/step - loss: 0.0337 - acc: 0.9714 - val_loss: 0.5007 - val_ac
c: 0.4620
Epoch 142/400
```

```
c: 0.4560
Epoch 143/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0326 - acc: 0.9714 - val_loss: 0.4854 - val_ac
c: 0.4740
Epoch 144/400
c: 0.4860
Epoch 145/400
c: 0.5060
Epoch 146/400
1/1 [============= ] - 0s 167ms/step - loss: 0.0301 - acc: 0.9786 - val_loss: 0.4251 - val_ac
c: 0.5120
Epoch 147/400
1/1 [============ ] - 0s 179ms/step - loss: 0.0299 - acc: 0.9929 - val_loss: 0.4064 - val_ac
c: 0.5160
Epoch 148/400
1/1 [=========== ] - 0s 168ms/step - loss: 0.0300 - acc: 0.9857 - val loss: 0.3998 - val ac
c: 0.5180
Epoch 149/400
c: 0.5240
Fnoch 150/400
c: 0.5260
Epoch 151/400
1/1 [============== ] - 0s 177ms/step - loss: 0.0310 - acc: 0.9643 - val_loss: 0.3877 - val_ac
c: 0.5340
Epoch 152/400
c: 0.5380
Epoch 153/400
c: 0.5440
Epoch 154/400
c: 0.5540
Epoch 155/400
c: 0.5440
Epoch 156/400
1/1 [============= ] - 0s 186ms/step - loss: 0.0342 - acc: 0.9786 - val_loss: 0.3939 - val_ac
c: 0.5520
Epoch 157/400
c: 0.5340
Epoch 158/400
c: 0.5280
Epoch 159/400
1/1 [=========== ] - 0s 172ms/step - loss: 0.0382 - acc: 0.9500 - val loss: 0.4253 - val ac
c: 0.5200
Epoch 160/400
c: 0.5100
Epoch 161/400
1/1 [=========== ] - 0s 173ms/step - loss: 0.0333 - acc: 0.9786 - val loss: 0.4192 - val ac
c: 0.5080
Epoch 162/400
c: 0.5060
Epoch 163/400
1/1 [============= ] - 0s 170ms/step - loss: 0.0376 - acc: 0.9429 - val_loss: 0.4141 - val_ac
c: 0.4960
Epoch 164/400
c: 0.5160
Epoch 165/400
1/1 [===========] - 0s 187ms/step - loss: 0.0375 - acc: 0.9714 - val_loss: 0.4071 - val_ac
c: 0.5240
```

```
Epoch 166/400
1/1 [============= ] - 0s 180ms/step - loss: 0.0341 - acc: 0.9857 - val_loss: 0.4032 - val_ac
c: 0.5340
Epoch 167/400
1/1 [============] - 0s 190ms/step - loss: 0.0350 - acc: 0.9786 - val_loss: 0.4084 - val_ac
c: 0.5280
Epoch 168/400
1/1 [============= ] - 0s 182ms/step - loss: 0.0359 - acc: 0.9857 - val_loss: 0.4179 - val_ac
c: 0.5320
Epoch 169/400
1/1 [============ ] - 0s 179ms/step - loss: 0.0364 - acc: 0.9786 - val_loss: 0.4292 - val_ac
c: 0.5180
Epoch 170/400
1/1 [=========== ] - 0s 182ms/step - loss: 0.0361 - acc: 0.9714 - val loss: 0.4383 - val ac
c: 0.5060
Epoch 171/400
c: 0.5120
Epoch 172/400
1/1 [============= ] - 0s 178ms/step - loss: 0.0357 - acc: 0.9857 - val_loss: 0.4508 - val_ac
c: 0.5020
Epoch 173/400
c: 0.5060
Epoch 174/400
c: 0.5180
Epoch 175/400
c: 0.5160
Epoch 176/400
1/1 [============ ] - 0s 193ms/step - loss: 0.0377 - acc: 0.9857 - val_loss: 0.4260 - val_ac
c: 0.5300
Epoch 177/400
c: 0.5420
Epoch 178/400
c: 0.5500
Epoch 179/400
c: 0.5740
Epoch 180/400
1/1 [============== ] - 0s 204ms/step - loss: 0.0370 - acc: 0.9714 - val_loss: 0.3997 - val_ac
c: 0.5880
Epoch 181/400
1/1 [=========== ] - 0s 182ms/step - loss: 0.0368 - acc: 0.9643 - val loss: 0.3957 - val ac
c: 0.5840
Epoch 182/400
c: 0.5680
Epoch 183/400
1/1 [============ ] - 0s 177ms/step - loss: 0.0387 - acc: 0.9714 - val_loss: 0.3953 - val_ac
c: 0.5520
Epoch 184/400
c: 0.5360
Epoch 185/400
c: 0.5320
Epoch 186/400
c: 0.5220
Epoch 187/400
c: 0.5200
Epoch 188/400
c: 0.5160
Epoch 189/400
1/1 [==================] - 0s 205ms/step - loss: 0.0364 - acc: 0.9857 - val_loss: 0.4504 - val_ac
```

```
c: 0.4940
      Epoch 190/400
      1/1 [=========== ] - 0s 176ms/step - loss: 0.0375 - acc: 0.9786 - val loss: 0.4448 - val ac
      c: 0.4940
      Epoch 191/400
      1/1 [============= ] - 0s 192ms/step - loss: 0.0377 - acc: 0.9786 - val_loss: 0.4346 - val_ac
      c: 0.5100
      Epoch 192/400
      1/1 [============= ] - 0s 183ms/step - loss: 0.0382 - acc: 0.9714 - val_loss: 0.4156 - val_ac
      c: 0.5340
      Epoch 193/400
      1/1 [============== ] - 0s 178ms/step - loss: 0.0367 - acc: 0.9714 - val_loss: 0.4048 - val_ac
      c: 0.5420
      Epoch 194/400
      1/1 [============ ] - 0s 174ms/step - loss: 0.0398 - acc: 0.9643 - val_loss: 0.3999 - val_ac
      c: 0.5420
      Epoch 195/400
      1/1 [==========] - 0s 173ms/step - loss: 0.0409 - acc: 0.9786 - val_loss: 0.4003 - val_ac
      c: 0.5360
      Epoch 196/400
      c: 0.5320
      Epoch 197/400
      1/1 [=========== ] - 0s 165ms/step - loss: 0.0373 - acc: 0.9786 - val loss: 0.3948 - val ac
      c: 0.5460
      Epoch 198/400
      1/1 [==============] - 0s 180ms/step - loss: 0.0363 - acc: 0.9857 - val_loss: 0.3924 - val_ac
      c: 0.5480
      Epoch 199/400
      1/1 [============ ] - 0s 197ms/step - loss: 0.0393 - acc: 0.9714 - val_loss: 0.3879 - val_ac
      c: 0.5440
      Epoch 200/400
      1/1 [============== ] - 0s 177ms/step - loss: 0.0363 - acc: 0.9786 - val_loss: 0.3863 - val_ac
      c: 0.5480
      Epoch 201/400
      1/1 [============= ] - 0s 176ms/step - loss: 0.0359 - acc: 0.9786 - val_loss: 0.3932 - val_ac
      c: 0.5320
      Epoch 202/400
      c: 0.5200
      Epoch 203/400
      c: 0.5140
      Epoch 204/400
      c: 0.5200
      Epoch 205/400
      1/1 [============ ] - 0s 167ms/step - loss: 0.0406 - acc: 0.9714 - val_loss: 0.4034 - val_ac
      c: 0.5220
      Epoch 206/400
      1/1 [=============== ] - 0s 170ms/step - loss: 0.0357 - acc: 0.9929 - val_loss: 0.3997 - val_ac
      c: 0.5260
      Epoch 207/400
      c: 0.5260
      Epoch 208/400
      c: 0.5280
Out[]: <tensorflow.python.keras.callbacks.History at 0x1cad57705c8>
In [ ]: # Evaluate model
      y_pred = model_fnn.predict(X_te)
      report = classification_report(np.argmax(y_te,axis=1), np.argmax(y_pred,axis=1), target_names=classes)
      print('FCNN Classification Report: \n {}'.format(report))
```

FCNN Classification Report: precision recall f1-score support Case_Based 0.54 0.54 0.54 114 Genetic_Algorithms 0.62 0.79 0.70 156 Neural_Networks 0.70 0.56 0.62 290 0.60 172 Probabilistic_Methods 0.71 0.52 Reinforcement_Learning 0.44 0.45 0.44 85 Rule_Learning 0.38 0.75 0.50 60 Theory 0.43 0.42 0.42 123 0.57 1000 accuracy 0.54 0.58 0.55 1000 macro avg

0.59

0.57

Get hidden layer representation for FNN

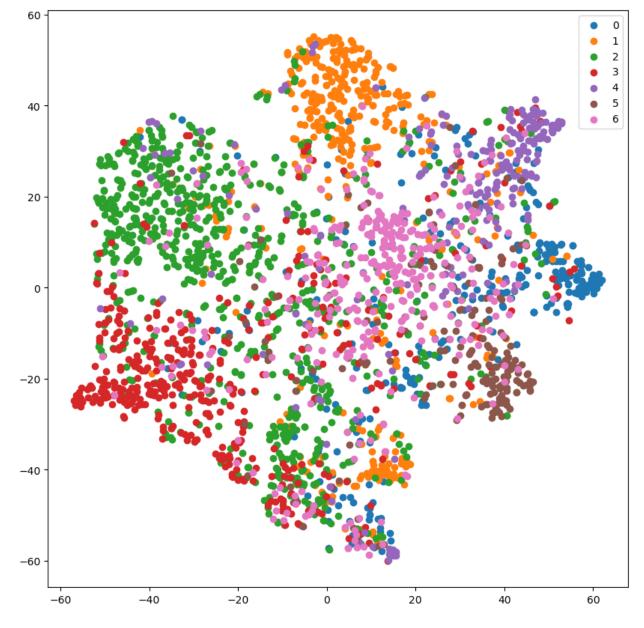
weighted avg

```
In [ ]: layer_outputs = [layer.output for layer in model_fnn.layers]
    activation_model = Model(inputs=model_fnn.input, outputs=layer_outputs)
    activations = activation_model.predict([X])
```

0.57

1000

In []: x_tsne = TSNE(n_components=2).fit_transform(activations[3])
plot_tSNE(labels_encoded,x_tsne)



In []: ### END OF NOTEBOOK ###