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', 'Case_Based', 'Genetic_Algorithms', 'Theory', 'Probabilistic_Method
        s', 'Rule_Learning', 'Neural_Networks'}
        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)

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

Graph info: Name:
   Type: Graph
   Number of nodes: 2708
   Number of edges: 5278
   Average degree: 3.8981
```

Building and Training Graph Convolutional Networks

```
In [ ]: # Parameters
        channels = 16
                               # Number of channels in the first layer
        dropout = 0.5  # Dropout rate for the features
12_reg = 5e-4  # L2 regularization rate
        es_patience = 200 # Patience for a
        learning_rate = 1e-2  # Learning rate
                             # Patience for early stopping
        # Preprocessing operations
        A = GraphConv.preprocess(A).astype('f4')
        # 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 #
        input_1 (InputLayer)
                                       [(None, 1433)]
        dropout (Dropout)
                                       (None, 1433)
                                                           0
                                                                       input_1[0][0]
                                       [(None, 2708)]
        input_2 (InputLayer)
                                                           0
                                                           22928
        graph_conv (GraphConv)
                                       (None, 16)
                                                                       dropout[0][0]
                                                                       input_2[0][0]
        dropout_1 (Dropout)
                                       (None, 16)
                                                                       graph_conv[0][0]
        graph_conv_1 (GraphConv)
                                       (None, 7)
                                                                       dropout_1[0][0]
                                                           112
                                                                       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,
callbacks=[
    EarlyStopping(patience=es_patience, restore_best_weights=True),
    tbCallBack_GCN
])
```

```
Epoch 1/400
1/1 [============] - 0s 443ms/step - loss: 0.1167 - acc: 0.1143 - val_loss: 0.3666 - val_ac
c: 0.2420
Epoch 2/400
n_batch_end) is slow compared to the batch update (0.216931). Check your callbacks.
c: 0.4760
Epoch 3/400
c: 0.5740
Epoch 4/400
1/1 [============ ] - 0s 271ms/step - loss: 0.0977 - acc: 0.7000 - val_loss: 0.3402 - val_ac
c: 0.5620
Epoch 5/400
1/1 [============ ] - 0s 280ms/step - loss: 0.0919 - acc: 0.6857 - val_loss: 0.3322 - val_ac
c: 0.5620
Epoch 6/400
1/1 [=========== ] - 0s 251ms/step - loss: 0.0869 - acc: 0.7500 - val loss: 0.3238 - val ac
c: 0.5900
Epoch 7/400
1/1 [==============] - 0s 275ms/step - loss: 0.0832 - acc: 0.7571 - val_loss: 0.3144 - val_ac
c: 0.6200
Fnoch 8/400
c: 0.6580
Epoch 9/400
c: 0.6880
Epoch 10/400
c: 0.7160
Epoch 11/400
c: 0.7280
Epoch 12/400
1/1 [==========] - 0s 297ms/step - loss: 0.0721 - acc: 0.8786 - val_loss: 0.2703 - val_ac
c: 0.7340
Epoch 13/400
c: 0.7460
Epoch 14/400
1/1 [============= ] - 0s 250ms/step - loss: 0.0648 - acc: 0.9143 - val_loss: 0.2570 - val_ac
c: 0.7580
Epoch 15/400
1/1 [============= ] - 0s 232ms/step - loss: 0.0658 - acc: 0.8857 - val_loss: 0.2513 - val_ac
c: 0.7560
Epoch 16/400
c: 0.7620
Epoch 17/400
1/1 [=========== ] - 0s 225ms/step - loss: 0.0613 - acc: 0.9071 - val loss: 0.2408 - val ac
c: 0.7600
Epoch 18/400
c: 0.7640
Epoch 19/400
1/1 [=========== ] - 0s 304ms/step - loss: 0.0603 - acc: 0.9000 - val loss: 0.2306 - val ac
c: 0.7600
Epoch 20/400
c: 0.7660
Epoch 21/400
1/1 [==========] - 0s 254ms/step - loss: 0.0580 - acc: 0.9000 - val_loss: 0.2203 - val_ac
c: 0.7680
Epoch 22/400
1/1 [============= ] - 0s 252ms/step - loss: 0.0553 - acc: 0.9357 - val_loss: 0.2159 - val_ac
c: 0.7680
Epoch 23/400
1/1 [===========] - 0s 246ms/step - loss: 0.0528 - acc: 0.9357 - val_loss: 0.2120 - val_ac
c: 0.7740
```

```
Epoch 24/400
1/1 [============= ] - 0s 273ms/step - loss: 0.0547 - acc: 0.8929 - val_loss: 0.2085 - val_ac
c: 0.7720
Epoch 25/400
c: 0.7720
Epoch 26/400
1/1 [============= ] - 0s 312ms/step - loss: 0.0512 - acc: 0.9714 - val_loss: 0.2021 - val_ac
c: 0.7760
Epoch 27/400
c: 0.7740
Epoch 28/400
1/1 [=========== ] - 0s 248ms/step - loss: 0.0503 - acc: 0.9214 - val loss: 0.1968 - val ac
c: 0.7700
Epoch 29/400
c: 0.7700
Epoch 30/400
1/1 [============= ] - 0s 222ms/step - loss: 0.0509 - acc: 0.9286 - val_loss: 0.1934 - val_ac
c: 0.7720
Epoch 31/400
c: 0.7700
Epoch 32/400
c: 0.7700
Epoch 33/400
c: 0.7700
Epoch 34/400
1/1 [============= ] - 0s 199ms/step - loss: 0.0462 - acc: 0.9643 - val_loss: 0.1864 - val_ac
c: 0.7700
Epoch 35/400
c: 0.7760
Epoch 36/400
c: 0.7800
Epoch 37/400
c: 0.7820
Epoch 38/400
c: 0.7760
Epoch 39/400
1/1 [=========== ] - 0s 218ms/step - loss: 0.0454 - acc: 0.9214 - val loss: 0.1794 - val ac
c: 0.7760
Epoch 40/400
1/1 [==========] - 0s 193ms/step - loss: 0.0408 - acc: 0.9643 - val_loss: 0.1787 - val_ac
c: 0.7720
Epoch 41/400
1/1 [===========] - 0s 198ms/step - loss: 0.0406 - acc: 0.9714 - val_loss: 0.1785 - val_ac
c: 0.7660
Epoch 42/400
c: 0.7700
Epoch 43/400
c: 0.7720
Epoch 44/400
c: 0.7740
Epoch 45/400
c: 0.7800
Epoch 46/400
c: 0.7840
Epoch 47/400
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c: 0.7880
Epoch 48/400
c: 0.7880
Epoch 49/400
1/1 [============= ] - 0s 173ms/step - loss: 0.0412 - acc: 0.9214 - val_loss: 0.1665 - val_ac
c: 0.7900
Epoch 50/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0382 - acc: 0.9571 - val_loss: 0.1652 - val_ac
c: 0.7860
Epoch 51/400
c: 0.7900
Epoch 52/400
1/1 [============= ] - 0s 238ms/step - loss: 0.0404 - acc: 0.9357 - val_loss: 0.1656 - val_ac
c: 0.7860
Epoch 53/400
c: 0.7840
Epoch 54/400
c: 0.7820
Epoch 55/400
1/1 [=========== ] - 0s 176ms/step - loss: 0.0368 - acc: 0.9571 - val loss: 0.1708 - val ac
c: 0.7740
Epoch 56/400
c: 0.7720
Epoch 57/400
c: 0.7760
Epoch 58/400
1/1 [=============== ] - 0s 202ms/step - loss: 0.0380 - acc: 0.9500 - val_loss: 0.1704 - val_ac
c: 0.7740
Epoch 59/400
c: 0.7780
Epoch 60/400
c: 0.7820
Epoch 61/400
c: 0.7800
Epoch 62/400
1/1 [===========] - 0s 184ms/step - loss: 0.0350 - acc: 0.9786 - val_loss: 0.1632 - val_ac
c: 0.7820
Epoch 63/400
1/1 [============= ] - 0s 184ms/step - loss: 0.0363 - acc: 0.9500 - val_loss: 0.1606 - val_ac
c: 0.7840
Epoch 64/400
c: 0.7800
Epoch 65/400
c: 0.7800
Epoch 66/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0361 - acc: 0.9286 - val_loss: 0.1585 - val_ac
c: 0.7820
Epoch 67/400
c: 0.7820
Epoch 68/400
c: 0.7760
Epoch 69/400
c: 0.7800
Epoch 70/400
1/1 [============= ] - 0s 185ms/step - loss: 0.0343 - acc: 0.9500 - val_loss: 0.1627 - val_ac
c: 0.7800
Epoch 71/400
```

```
c: 0.7800
Epoch 72/400
1/1 [============= ] - 0s 182ms/step - loss: 0.0356 - acc: 0.9643 - val_loss: 0.1616 - val_ac
c: 0.7760
Epoch 73/400
c: 0.7800
Epoch 74/400
c: 0.7720
Epoch 75/400
1/1 [============= ] - 0s 210ms/step - loss: 0.0314 - acc: 0.9643 - val_loss: 0.1570 - val_ac
c: 0.7780
Epoch 76/400
1/1 [============= ] - 0s 191ms/step - loss: 0.0318 - acc: 0.9857 - val_loss: 0.1551 - val_ac
c: 0.7800
Epoch 77/400
1/1 [============ ] - 0s 183ms/step - loss: 0.0327 - acc: 0.9500 - val loss: 0.1533 - val ac
c: 0.7820
Epoch 78/400
c: 0.7840
Fnoch 79/400
c: 0.7820
Epoch 80/400
c: 0.7820
Epoch 81/400
c: 0.7800
Epoch 82/400
c: 0.7800
Epoch 83/400
1/1 [============= ] - 0s 202ms/step - loss: 0.0341 - acc: 0.9643 - val_loss: 0.1561 - val_ac
c: 0.7740
Epoch 84/400
c: 0.7700
Epoch 85/400
1/1 [============= ] - 0s 184ms/step - loss: 0.0336 - acc: 0.9571 - val_loss: 0.1580 - val_ac
c: 0.7720
Epoch 86/400
c: 0.7680
Epoch 87/400
c: 0.7700
Epoch 88/400
1/1 [=========== ] - 0s 196ms/step - loss: 0.0318 - acc: 0.9429 - val loss: 0.1549 - val ac
c: 0.7740
Epoch 89/400
c: 0.7840
Epoch 90/400
1/1 [=========== ] - 0s 171ms/step - loss: 0.0293 - acc: 0.9643 - val loss: 0.1515 - val ac
c: 0.7840
Epoch 91/400
c: 0.7880
Epoch 92/400
1/1 [============= ] - 0s 178ms/step - loss: 0.0308 - acc: 0.9571 - val_loss: 0.1485 - val_ac
c: 0.7900
Epoch 93/400
1/1 [============= ] - 0s 183ms/step - loss: 0.0303 - acc: 0.9857 - val_loss: 0.1481 - val_ac
c: 0.7900
Epoch 94/400
c: 0.7780
```

```
Epoch 95/400
1/1 [============= ] - 0s 162ms/step - loss: 0.0299 - acc: 0.9500 - val_loss: 0.1517 - val_ac
c: 0.7760
Epoch 96/400
c: 0.7680
Epoch 97/400
1/1 [============= ] - 0s 182ms/step - loss: 0.0304 - acc: 0.9643 - val_loss: 0.1544 - val_ac
c: 0.7680
Epoch 98/400
1/1 [============= ] - 0s 173ms/step - loss: 0.0301 - acc: 0.9571 - val_loss: 0.1536 - val_ac
c: 0.7740
Epoch 99/400
1/1 [=========== ] - 0s 168ms/step - loss: 0.0285 - acc: 0.9786 - val loss: 0.1527 - val ac
c: 0.7780
Epoch 100/400
c: 0.7760
Epoch 101/400
1/1 [============= ] - 0s 166ms/step - loss: 0.0316 - acc: 0.9571 - val_loss: 0.1534 - val_ac
c: 0.7760
Epoch 102/400
c: 0.7760
Epoch 103/400
c: 0.7740
Epoch 104/400
c: 0.7740
Epoch 105/400
1/1 [============= ] - 0s 193ms/step - loss: 0.0311 - acc: 0.9643 - val_loss: 0.1517 - val_ac
c: 0.7680
Epoch 106/400
c: 0.7740
Epoch 107/400
c: 0.7700
Epoch 108/400
c: 0.7740
Epoch 109/400
c: 0.7660
Epoch 110/400
1/1 [=========== ] - 0s 168ms/step - loss: 0.0277 - acc: 0.9786 - val loss: 0.1538 - val ac
c: 0.7620
Epoch 111/400
c: 0.7620
Epoch 112/400
1/1 [============= ] - 0s 174ms/step - loss: 0.0283 - acc: 0.9571 - val_loss: 0.1567 - val_ac
c: 0.7660
Epoch 113/400
c: 0.7680
Epoch 114/400
c: 0.7640
Epoch 115/400
c: 0.7680
Epoch 116/400
c: 0.7600
Epoch 117/400
c: 0.7680
Epoch 118/400
```

```
c: 0.7820
Epoch 119/400
c: 0.7880
Epoch 120/400
1/1 [===========] - 0s 167ms/step - loss: 0.0269 - acc: 0.9714 - val_loss: 0.1425 - val_ac
c: 0.7960
Epoch 121/400
1/1 [==========] - 0s 160ms/step - loss: 0.0263 - acc: 0.9929 - val_loss: 0.1444 - val_ac
c: 0.7860
Epoch 122/400
c: 0.7820
Epoch 123/400
1/1 [============= ] - 0s 169ms/step - loss: 0.0286 - acc: 0.9857 - val_loss: 0.1514 - val_ac
c: 0.7760
Epoch 124/400
c: 0.7760
Epoch 125/400
c: 0.7700
Epoch 126/400
c: 0.7780
Epoch 127/400
c: 0.7740
Epoch 128/400
c: 0.7820
Epoch 129/400
c: 0.7820
Epoch 130/400
c: 0.7780
Epoch 131/400
c: 0.7740
Epoch 132/400
c: 0.7700
Epoch 133/400
c: 0.7820
Epoch 134/400
1/1 [===========] - 0s 224ms/step - loss: 0.0285 - acc: 0.9429 - val_loss: 0.1422 - val_ac
c: 0.7920
Epoch 135/400
c: 0.8000
Epoch 136/400
c: 0.7980
Epoch 137/400
1/1 [============= ] - 0s 212ms/step - loss: 0.0274 - acc: 0.9571 - val_loss: 0.1423 - val_ac
c: 0.7900
Epoch 138/400
c: 0.7900
Epoch 139/400
1/1 [============ ] - 0s 161ms/step - loss: 0.0275 - acc: 0.9571 - val loss: 0.1471 - val ac
c: 0.7880
Epoch 140/400
c: 0.7740
Epoch 141/400
1/1 [============= ] - 0s 223ms/step - loss: 0.0258 - acc: 0.9714 - val_loss: 0.1575 - val_ac
c: 0.7620
Epoch 142/400
```

```
c: 0.7540
Epoch 143/400
1/1 [============= ] - 0s 158ms/step - loss: 0.0254 - acc: 0.9786 - val_loss: 0.1590 - val_ac
c: 0.7440
Epoch 144/400
c: 0.7500
Epoch 145/400
c: 0.7580
Epoch 146/400
1/1 [============= ] - 0s 166ms/step - loss: 0.0263 - acc: 1.0000 - val_loss: 0.1432 - val_ac
c: 0.7740
Epoch 147/400
1/1 [============ ] - 0s 167ms/step - loss: 0.0280 - acc: 0.9429 - val_loss: 0.1404 - val_ac
c: 0.7880
Epoch 148/400
1/1 [=========== ] - 0s 161ms/step - loss: 0.0266 - acc: 0.9714 - val loss: 0.1402 - val ac
c: 0.7900
Epoch 149/400
c: 0.7820
Fnoch 150/400
c: 0.7800
Epoch 151/400
c: 0.7740
Epoch 152/400
c: 0.7640
Epoch 153/400
c: 0.7600
Epoch 154/400
1/1 [============ ] - 0s 163ms/step - loss: 0.0277 - acc: 0.9571 - val loss: 0.1513 - val ac
c: 0.7640
Epoch 155/400
c: 0.7660
Epoch 156/400
1/1 [============= ] - 0s 165ms/step - loss: 0.0256 - acc: 0.9500 - val_loss: 0.1492 - val_ac
c: 0.7640
Epoch 157/400
c: 0.7680
Epoch 158/400
c: 0.7740
Epoch 159/400
1/1 [=========== ] - 0s 164ms/step - loss: 0.0250 - acc: 0.9857 - val loss: 0.1436 - val ac
c: 0.7840
Epoch 160/400
c: 0.7800
Epoch 161/400
1/1 [=========== ] - 0s 162ms/step - loss: 0.0235 - acc: 0.9786 - val loss: 0.1425 - val ac
c: 0.7800
Epoch 162/400
c: 0.7800
Epoch 163/400
1/1 [===========] - 0s 163ms/step - loss: 0.0248 - acc: 0.9643 - val_loss: 0.1434 - val_ac
c: 0.7820
Epoch 164/400
1/1 [============= ] - 0s 165ms/step - loss: 0.0257 - acc: 0.9714 - val_loss: 0.1435 - val_ac
c: 0.7760
Epoch 165/400
1/1 [============] - 0s 169ms/step - loss: 0.0251 - acc: 0.9857 - val_loss: 0.1455 - val_ac
c: 0.7760
```

```
Epoch 166/400
1/1 [============= ] - 0s 165ms/step - loss: 0.0249 - acc: 0.9643 - val_loss: 0.1478 - val_ac
c: 0.7620
Epoch 167/400
c: 0.7660
Epoch 168/400
1/1 [============= ] - 0s 163ms/step - loss: 0.0241 - acc: 0.9643 - val_loss: 0.1491 - val_ac
c: 0.7660
Epoch 169/400
1/1 [============= ] - 0s 181ms/step - loss: 0.0263 - acc: 0.9357 - val_loss: 0.1453 - val_ac
c: 0.7780
Epoch 170/400
1/1 [=========== ] - 0s 168ms/step - loss: 0.0268 - acc: 0.9500 - val loss: 0.1435 - val ac
c: 0.7800
Epoch 171/400
c: 0.7840
Epoch 172/400
1/1 [===========] - 0s 168ms/step - loss: 0.0242 - acc: 0.9643 - val_loss: 0.1413 - val_ac
c: 0.7860
Epoch 173/400
c: 0.7840
Epoch 174/400
c: 0.7800
Epoch 175/400
c: 0.7840
Epoch 176/400
1/1 [============= ] - 0s 164ms/step - loss: 0.0259 - acc: 0.9643 - val_loss: 0.1415 - val_ac
c: 0.7840
Epoch 177/400
c: 0.7820
Epoch 178/400
c: 0.7820
Epoch 179/400
c: 0.7820
Epoch 180/400
c: 0.7760
Epoch 181/400
1/1 [=========== ] - 0s 159ms/step - loss: 0.0252 - acc: 0.9714 - val loss: 0.1407 - val ac
c: 0.7760
Epoch 182/400
c: 0.7720
Epoch 183/400
1/1 [============= ] - 0s 165ms/step - loss: 0.0242 - acc: 0.9643 - val_loss: 0.1462 - val_ac
c: 0.7740
Epoch 184/400
c: 0.7680
Epoch 185/400
c: 0.7760
Epoch 186/400
c: 0.7820
Epoch 187/400
c: 0.7800
Epoch 188/400
c: 0.7800
Epoch 189/400
1/1 [==================] - 0s 163ms/step - loss: 0.0219 - acc: 0.9714 - val_loss: 0.1332 - val_ac
```

```
c: 0.7700
Epoch 190/400
c: 0.7680
Epoch 191/400
1/1 [============= ] - 0s 189ms/step - loss: 0.0233 - acc: 0.9786 - val_loss: 0.1350 - val_ac
c: 0.7660
Epoch 192/400
1/1 [============= ] - 0s 162ms/step - loss: 0.0246 - acc: 0.9857 - val_loss: 0.1377 - val_ac
c: 0.7660
Epoch 193/400
c: 0.7660
Epoch 194/400
1/1 [============ ] - 0s 160ms/step - loss: 0.0216 - acc: 0.9857 - val_loss: 0.1446 - val_ac
c: 0.7600
Epoch 195/400
c: 0.7600
Epoch 196/400
c: 0.7620
Epoch 197/400
c: 0.7640
Epoch 198/400
c: 0.7640
Epoch 199/400
c: 0.7560
Epoch 200/400
c: 0.7600
Epoch 201/400
c: 0.7720
Epoch 202/400
c: 0.7820
Epoch 203/400
c: 0.7860
Epoch 204/400
c: 0.7920
Epoch 205/400
1/1 [============= ] - 0s 161ms/step - loss: 0.0238 - acc: 0.9786 - val_loss: 0.1398 - val_ac
c: 0.7840
Epoch 206/400
c: 0.7820
Epoch 207/400
c: 0.7840
Epoch 208/400
1/1 [============= ] - 0s 156ms/step - loss: 0.0231 - acc: 0.9786 - val_loss: 0.1438 - val_ac
c: 0.7780
Epoch 209/400
c: 0.7680
Epoch 210/400
1/1 [============ ] - 0s 164ms/step - loss: 0.0224 - acc: 0.9714 - val loss: 0.1482 - val ac
c: 0.7640
Epoch 211/400
c: 0.7640
Epoch 212/400
1/1 [============= ] - 0s 165ms/step - loss: 0.0239 - acc: 0.9857 - val_loss: 0.1442 - val_ac
c: 0.7740
Epoch 213/400
```

```
c: 0.7720
Epoch 214/400
1/1 [============= ] - 0s 161ms/step - loss: 0.0229 - acc: 0.9857 - val_loss: 0.1401 - val_ac
c: 0.7720
Epoch 215/400
c: 0.7720
Epoch 216/400
c: 0.7700
Epoch 217/400
1/1 [============= ] - 0s 161ms/step - loss: 0.0239 - acc: 0.9714 - val_loss: 0.1388 - val_ac
c: 0.7700
Epoch 218/400
1/1 [============= ] - 0s 167ms/step - loss: 0.0233 - acc: 0.9571 - val_loss: 0.1397 - val_ac
c: 0.7760
Epoch 219/400
1/1 [=========== ] - 0s 167ms/step - loss: 0.0221 - acc: 0.9643 - val loss: 0.1433 - val ac
c: 0.7680
Epoch 220/400
c: 0.7680
Fnoch 221/400
c: 0.7620
Epoch 222/400
c: 0.7600
Epoch 223/400
c: 0.7720
Epoch 224/400
c: 0.7780
Epoch 225/400
c: 0.7760
Epoch 226/400
c: 0.7740
Epoch 227/400
1/1 [============= ] - 0s 170ms/step - loss: 0.0207 - acc: 0.9714 - val_loss: 0.1417 - val_ac
c: 0.7700
Epoch 228/400
1/1 [============= ] - 0s 173ms/step - loss: 0.0210 - acc: 0.9929 - val_loss: 0.1430 - val_ac
c: 0.7760
Epoch 229/400
c: 0.7720
Epoch 230/400
1/1 [=========== ] - 0s 171ms/step - loss: 0.0206 - acc: 0.9857 - val loss: 0.1477 - val ac
c: 0.7600
Epoch 231/400
c: 0.7540
Epoch 232/400
1/1 [===========] - 0s 169ms/step - loss: 0.0225 - acc: 0.9643 - val loss: 0.1498 - val ac
c: 0.7520
Epoch 233/400
c: 0.7600
Epoch 234/400
1/1 [============= ] - 0s 174ms/step - loss: 0.0205 - acc: 0.9714 - val_loss: 0.1385 - val_ac
c: 0.7800
Epoch 235/400
1/1 [============= ] - 0s 173ms/step - loss: 0.0212 - acc: 0.9857 - val_loss: 0.1350 - val_ac
c: 0.7820
Epoch 236/400
1/1 [===========] - 0s 182ms/step - loss: 0.0224 - acc: 0.9786 - val_loss: 0.1327 - val_ac
c: 0.7940
```

```
Epoch 237/400
1/1 [============= ] - 0s 159ms/step - loss: 0.0210 - acc: 0.9786 - val_loss: 0.1312 - val_ac
c: 0.7920
Epoch 238/400
c: 0.7900
Epoch 239/400
1/1 [============= ] - 0s 156ms/step - loss: 0.0216 - acc: 0.9643 - val_loss: 0.1325 - val_ac
c: 0.7840
Epoch 240/400
1/1 [============= ] - 0s 153ms/step - loss: 0.0203 - acc: 0.9786 - val_loss: 0.1368 - val_ac
c: 0.7760
Epoch 241/400
1/1 [=========== ] - 0s 162ms/step - loss: 0.0212 - acc: 0.9643 - val loss: 0.1416 - val ac
c: 0.7660
Epoch 242/400
c: 0.7540
Epoch 243/400
1/1 [============= ] - 0s 162ms/step - loss: 0.0231 - acc: 0.9714 - val_loss: 0.1478 - val_ac
c: 0.7500
Epoch 244/400
c: 0.7560
Epoch 245/400
c: 0.7600
Epoch 246/400
c: 0.7720
Epoch 247/400
1/1 [============= ] - 0s 183ms/step - loss: 0.0199 - acc: 0.9786 - val_loss: 0.1395 - val_ac
c: 0.7820
Epoch 248/400
c: 0.7820
Epoch 249/400
c: 0.7800
Epoch 250/400
c: 0.7760
Epoch 251/400
c: 0.7760
Epoch 252/400
1/1 [=========== ] - 0s 167ms/step - loss: 0.0216 - acc: 0.9929 - val loss: 0.1418 - val ac
c: 0.7700
Epoch 253/400
1/1 [==========] - 0s 158ms/step - loss: 0.0196 - acc: 0.9786 - val_loss: 0.1475 - val_ac
c: 0.7620
Epoch 254/400
1/1 [============= ] - 0s 214ms/step - loss: 0.0222 - acc: 0.9714 - val_loss: 0.1504 - val_ac
c: 0.7580
Epoch 255/400
c: 0.7700
Epoch 256/400
c: 0.7700
Epoch 257/400
c: 0.7600
Epoch 258/400
c: 0.7680
Epoch 259/400
c: 0.7760
Epoch 260/400
1/1 [==================] - 0s 161ms/step - loss: 0.0205 - acc: 0.9929 - val_loss: 0.1387 - val_ac
```

```
c: 0.7760
Epoch 261/400
c: 0.7640
Epoch 262/400
1/1 [============= ] - 0s 153ms/step - loss: 0.0227 - acc: 0.9571 - val_loss: 0.1386 - val_ac
c: 0.7600
Epoch 263/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0207 - acc: 0.9643 - val_loss: 0.1385 - val_ac
c: 0.7580
Epoch 264/400
c: 0.7620
Epoch 265/400
1/1 [============= ] - 0s 178ms/step - loss: 0.0196 - acc: 0.9857 - val_loss: 0.1355 - val_ac
c: 0.7620
Epoch 266/400
c: 0.7660
Epoch 267/400
c: 0.7660
Epoch 268/400
c: 0.7740
Epoch 269/400
c: 0.7760
Epoch 270/400
c: 0.7740
Epoch 271/400
c: 0.7780
Epoch 272/400
c: 0.7680
Epoch 273/400
c: 0.7620
Epoch 274/400
c: 0.7620
Epoch 275/400
c: 0.7600
Epoch 276/400
1/1 [============ ] - 0s 210ms/step - loss: 0.0198 - acc: 0.9929 - val_loss: 0.1418 - val_ac
c: 0.7600
Epoch 277/400
c: 0.7640
Epoch 278/400
c: 0.7720
Epoch 279/400
1/1 [===========] - 0s 426ms/step - loss: 0.0199 - acc: 0.9714 - val_loss: 0.1477 - val_ac
c: 0.7740
Epoch 280/400
c: 0.7760
Epoch 281/400
1/1 [=========== ] - 0s 275ms/step - loss: 0.0205 - acc: 0.9857 - val loss: 0.1437 - val ac
c: 0.7780
Epoch 282/400
c: 0.7760
Epoch 283/400
1/1 [============= ] - 0s 190ms/step - loss: 0.0214 - acc: 0.9857 - val_loss: 0.1387 - val_ac
c: 0.7800
Epoch 284/400
```

```
c: 0.7780
Epoch 285/400
1/1 [============= ] - 0s 225ms/step - loss: 0.0194 - acc: 0.9857 - val_loss: 0.1348 - val_ac
c: 0.7860
Epoch 286/400
c: 0.7880
Epoch 287/400
c: 0.7800
Epoch 288/400
1/1 [============= ] - 0s 223ms/step - loss: 0.0225 - acc: 0.9500 - val_loss: 0.1392 - val_ac
c: 0.7840
Epoch 289/400
1/1 [============ ] - 0s 201ms/step - loss: 0.0187 - acc: 0.9714 - val_loss: 0.1401 - val_ac
c: 0.7800
Epoch 290/400
1/1 [===========] - 0s 461ms/step - loss: 0.0205 - acc: 0.9714 - val loss: 0.1406 - val ac
c: 0.7900
Epoch 291/400
c: 0.7940
Fnoch 292/400
c: 0.7880
Epoch 293/400
c: 0.7880
Epoch 294/400
c: 0.7800
Epoch 295/400
c: 0.7720
Epoch 296/400
1/1 [===========] - 0s 201ms/step - loss: 0.0203 - acc: 0.9643 - val_loss: 0.1382 - val_ac
c: 0.7680
Epoch 297/400
c: 0.7660
Epoch 298/400
1/1 [============= ] - 0s 244ms/step - loss: 0.0204 - acc: 0.9929 - val_loss: 0.1387 - val_ac
c: 0.7620
Epoch 299/400
c: 0.7800
Epoch 300/400
c: 0.7740
Epoch 301/400
1/1 [=========== ] - 0s 228ms/step - loss: 0.0215 - acc: 0.9643 - val loss: 0.1343 - val ac
c: 0.7780
Epoch 302/400
c: 0.7760
Epoch 303/400
1/1 [=========== ] - 0s 204ms/step - loss: 0.0205 - acc: 0.9857 - val loss: 0.1390 - val ac
c: 0.7760
Epoch 304/400
1/1 [============== ] - 0s 188ms/step - loss: 0.0186 - acc: 1.0000 - val_loss: 0.1422 - val_ac
c: 0.7720
Epoch 305/400
1/1 [===========] - 0s 185ms/step - loss: 0.0189 - acc: 0.9786 - val_loss: 0.1472 - val_ac
c: 0.7620
Epoch 306/400
1/1 [============= ] - 0s 203ms/step - loss: 0.0203 - acc: 0.9571 - val_loss: 0.1531 - val_ac
c: 0.7540
Epoch 307/400
1/1 [============] - 0s 221ms/step - loss: 0.0218 - acc: 0.9714 - val_loss: 0.1530 - val_ac
c: 0.7520
```

```
Epoch 308/400
1/1 [============= ] - 0s 186ms/step - loss: 0.0204 - acc: 0.9714 - val_loss: 0.1477 - val_ac
c: 0.7640
Epoch 309/400
c: 0.7880
Epoch 310/400
1/1 [============= ] - 0s 197ms/step - loss: 0.0199 - acc: 0.9857 - val_loss: 0.1340 - val_ac
c: 0.7900
Epoch 311/400
c: 0.7840
Epoch 312/400
1/1 [=========== ] - 0s 244ms/step - loss: 0.0218 - acc: 0.9500 - val loss: 0.1323 - val ac
c: 0.7780
Epoch 313/400
c: 0.7820
Epoch 314/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0184 - acc: 0.9714 - val_loss: 0.1373 - val_ac
c: 0.7760
Epoch 315/400
c: 0.7580
Epoch 316/400
c: 0.7560
Epoch 317/400
c: 0.7480
Epoch 318/400
1/1 [============ ] - 0s 174ms/step - loss: 0.0194 - acc: 0.9929 - val_loss: 0.1533 - val_ac
c: 0.7480
Epoch 319/400
c: 0.7460
Epoch 320/400
c: 0.7500
Epoch 321/400
c: 0.7600
Epoch 322/400
c: 0.7740
Epoch 323/400
1/1 [=========== ] - 0s 198ms/step - loss: 0.0193 - acc: 0.9857 - val loss: 0.1359 - val ac
c: 0.7840
Epoch 324/400
c: 0.7840
Epoch 325/400
1/1 [===========] - 0s 168ms/step - loss: 0.0198 - acc: 0.9786 - val_loss: 0.1344 - val_ac
c: 0.7820
Epoch 326/400
c: 0.7860
Epoch 327/400
c: 0.7800
Epoch 328/400
c: 0.7580
Epoch 329/400
c: 0.7420
Epoch 330/400
c: 0.7340
Epoch 331/400
1/1 [==================] - 0s 172ms/step - loss: 0.0197 - acc: 0.9786 - val_loss: 0.1529 - val_ac
```

```
c: 0.7380
Epoch 332/400
c: 0.7480
Epoch 333/400
1/1 [==========] - 0s 200ms/step - loss: 0.0200 - acc: 0.9714 - val_loss: 0.1456 - val_ac
c: 0.7540
Epoch 334/400
1/1 [============= ] - 0s 185ms/step - loss: 0.0184 - acc: 0.9714 - val_loss: 0.1412 - val_ac
c: 0.7620
Epoch 335/400
c: 0.7680
Epoch 336/400
1/1 [============= ] - 0s 184ms/step - loss: 0.0215 - acc: 0.9500 - val_loss: 0.1345 - val_ac
c: 0.7820
Epoch 337/400
c: 0.7840
Epoch 338/400
c: 0.7800
Epoch 339/400
1/1 [=========== ] - 0s 168ms/step - loss: 0.0187 - acc: 0.9714 - val loss: 0.1378 - val ac
c: 0.7780
Epoch 340/400
c: 0.7780
Epoch 341/400
1/1 [============= ] - 0s 159ms/step - loss: 0.0200 - acc: 0.9857 - val_loss: 0.1416 - val_ac
c: 0.7680
Epoch 342/400
c: 0.7680
Epoch 343/400
c: 0.7660
Epoch 344/400
c: 0.7640
Epoch 345/400
c: 0.7620
Epoch 346/400
c: 0.7660
Epoch 347/400
1/1 [============ ] - 0s 160ms/step - loss: 0.0199 - acc: 0.9571 - val_loss: 0.1356 - val_ac
c: 0.7600
Epoch 348/400
c: 0.7640
Epoch 349/400
c: 0.7620
Epoch 350/400
1/1 [============= ] - 0s 228ms/step - loss: 0.0190 - acc: 0.9857 - val_loss: 0.1418 - val_ac
c: 0.7640
Epoch 351/400
c: 0.7660
Epoch 352/400
1/1 [=========== ] - 0s 187ms/step - loss: 0.0183 - acc: 0.9929 - val loss: 0.1451 - val ac
c: 0.7740
Epoch 353/400
c: 0.7760
Epoch 354/400
1/1 [============ ] - 0s 162ms/step - loss: 0.0184 - acc: 0.9857 - val_loss: 0.1477 - val_ac
c: 0.7780
Epoch 355/400
```

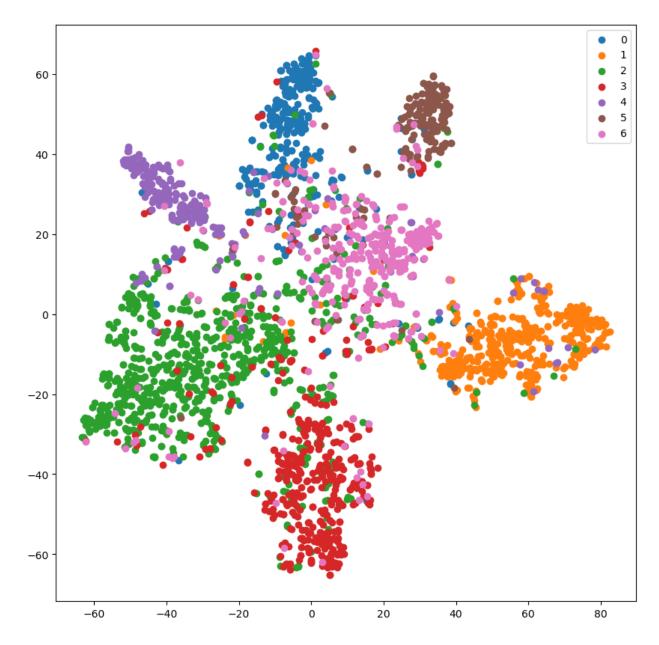
```
c: 0.7740
Epoch 356/400
1/1 [============= ] - 0s 162ms/step - loss: 0.0184 - acc: 0.9857 - val_loss: 0.1428 - val_ac
c: 0.7780
Epoch 357/400
c: 0.7740
Epoch 358/400
c: 0.7760
Epoch 359/400
1/1 [============= ] - 0s 183ms/step - loss: 0.0192 - acc: 0.9786 - val_loss: 0.1417 - val_ac
c: 0.7720
Epoch 360/400
1/1 [============= ] - 0s 187ms/step - loss: 0.0208 - acc: 0.9786 - val_loss: 0.1434 - val_ac
c: 0.7640
Epoch 361/400
1/1 [=========== ] - 0s 198ms/step - loss: 0.0193 - acc: 0.9714 - val loss: 0.1434 - val ac
c: 0.7620
Epoch 362/400
c: 0.7700
Fnoch 363/400
c: 0.7740
Epoch 364/400
c: 0.7660
Epoch 365/400
c: 0.7740
Epoch 366/400
c: 0.7700
Epoch 367/400
1/1 [==========] - 0s 157ms/step - loss: 0.0180 - acc: 0.9929 - val_loss: 0.1394 - val_ac
c: 0.7700
Epoch 368/400
c: 0.7720
Epoch 369/400
1/1 [============= ] - 0s 166ms/step - loss: 0.0186 - acc: 0.9857 - val_loss: 0.1345 - val_ac
c: 0.7720
Epoch 370/400
c: 0.7800
Epoch 371/400
c: 0.7780
Epoch 372/400
1/1 [=========== ] - 0s 174ms/step - loss: 0.0198 - acc: 0.9857 - val loss: 0.1342 - val ac
c: 0.7800
Epoch 373/400
c: 0.7780
Epoch 374/400
1/1 [=========== ] - 0s 162ms/step - loss: 0.0176 - acc: 0.9857 - val loss: 0.1369 - val ac
c: 0.7700
Epoch 375/400
c: 0.7640
Epoch 376/400
1/1 [============ ] - 0s 169ms/step - loss: 0.0179 - acc: 0.9786 - val_loss: 0.1399 - val_ac
c: 0.7680
Epoch 377/400
1/1 [============= ] - 0s 165ms/step - loss: 0.0191 - acc: 0.9643 - val_loss: 0.1377 - val_ac
c: 0.7700
Epoch 378/400
1/1 [===========] - 0s 171ms/step - loss: 0.0186 - acc: 0.9714 - val_loss: 0.1347 - val_ac
c: 0.7740
```

```
Epoch 379/400
    1/1 [============= ] - 0s 178ms/step - loss: 0.0178 - acc: 0.9643 - val_loss: 0.1309 - val_ac
    c: 0.7760
    Epoch 380/400
    1/1 [============] - 0s 163ms/step - loss: 0.0173 - acc: 0.9929 - val_loss: 0.1286 - val_ac
    c: 0.7780
    Epoch 381/400
    1/1 [============= ] - 0s 163ms/step - loss: 0.0187 - acc: 0.9714 - val_loss: 0.1291 - val_ac
    c: 0.7860
    Epoch 382/400
    1/1 [============= ] - 0s 162ms/step - loss: 0.0199 - acc: 0.9857 - val_loss: 0.1321 - val_ac
    c: 0.7820
    Epoch 383/400
    1/1 [=========== ] - 0s 160ms/step - loss: 0.0179 - acc: 0.9786 - val loss: 0.1352 - val ac
    c: 0.7740
    Epoch 384/400
    c: 0.7740
    Epoch 385/400
    1/1 [============= ] - 0s 168ms/step - loss: 0.0190 - acc: 0.9857 - val_loss: 0.1388 - val_ac
    c: 0.7740
    Epoch 386/400
    c: 0.7760
    Epoch 387/400
    c: 0.7760
    Epoch 388/400
    c: 0.7760
    Epoch 389/400
    1/1 [============= ] - 0s 165ms/step - loss: 0.0182 - acc: 0.9786 - val_loss: 0.1380 - val_ac
    c: 0.7720
    Epoch 390/400
    c: 0.7680
    Epoch 391/400
    c: 0.7660
    Epoch 392/400
    c: 0.7700
    Epoch 393/400
    c: 0.7700
    Epoch 394/400
    1/1 [============ ] - 0s 212ms/step - loss: 0.0175 - acc: 0.9929 - val loss: 0.1378 - val ac
    c: 0.7740
    Epoch 395/400
    c: 0.7780
    Epoch 396/400
    1/1 [============= ] - 0s 228ms/step - loss: 0.0187 - acc: 0.9714 - val_loss: 0.1401 - val_ac
    c: 0.7680
    Epoch 397/400
    1/1 [=============== ] - 0s 207ms/step - loss: 0.0189 - acc: 0.9643 - val_loss: 0.1424 - val_ac
    c: 0.7640
    Epoch 398/400
    c: 0.7600
    Epoch 399/400
    c: 0.7640
    Epoch 400/400
    c: 0.7620
Out[]: <tensorflow.python.keras.callbacks.History at 0x1d54c319148>
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
                                    0.83
                                              0.72
           Case_Based
                           0.64
                                                         114
   Genetic_Algorithms
                           0.86
                                     0.88
                                              0.87
                                                         156
      Neural_Networks
                                     0.60
                                              0.70
                                                         290
                           0.83
 Probabilistic_Methods
                           0.79
                                     0.73
                                              0.76
                                                         172
Reinforcement_Learning
                           0.58
                                     0.74
                                              0.65
                                                         85
                                     0.70
        Rule_Learning
                           0.62
                                              0.66
                                                         60
               Theory
                           0.57
                                     0.67
                                                         123
                                              0.62
                                              0.72
                                                        1000
             accuracy
            macro avg
                           0.70
                                     0.74
                                              0.71
                                                        1000
                           0.74
                                     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
c: 0.2720
Epoch 2/400
n_batch_end) is slow compared to the batch update (0.170995). Check your callbacks.
c: 0.3240
Epoch 3/400
c: 0.4220
Epoch 4/400
1/1 [============= ] - 0s 195ms/step - loss: 0.1108 - acc: 0.7500 - val_loss: 0.3372 - val_ac
c: 0.5280
Epoch 5/400
1/1 [============= ] - 0s 175ms/step - loss: 0.0861 - acc: 0.8143 - val_loss: 0.3042 - val_ac
c: 0.5540
Epoch 6/400
1/1 [=========== ] - 0s 185ms/step - loss: 0.0652 - acc: 0.8571 - val loss: 0.2783 - val ac
c: 0.5660
Epoch 7/400
1/1 [=============] - 0s 194ms/step - loss: 0.0562 - acc: 0.8786 - val_loss: 0.2625 - val_ac
c: 0.5940
Fnoch 8/400
c: 0.5820
Epoch 9/400
c: 0.5780
Epoch 10/400
c: 0.5780
Epoch 11/400
c: 0.5820
Epoch 12/400
c: 0.5760
Epoch 13/400
c: 0.5780
Epoch 14/400
1/1 [============ ] - 0s 357ms/step - loss: 0.0383 - acc: 1.0000 - val_loss: 0.3708 - val_ac
c: 0.5740
Epoch 15/400
1/1 [============ ] - 0s 287ms/step - loss: 0.0369 - acc: 0.9929 - val_loss: 0.3751 - val_ac
c: 0.5800
Epoch 16/400
c: 0.5840
Epoch 17/400
1/1 [=========== ] - 0s 241ms/step - loss: 0.0319 - acc: 0.9857 - val loss: 0.3856 - val ac
c: 0.5780
Epoch 18/400
c: 0.5760
Epoch 19/400
1/1 [=========== ] - 0s 228ms/step - loss: 0.0270 - acc: 0.9857 - val loss: 0.3563 - val ac
c: 0.5980
Epoch 20/400
c: 0.6160
Epoch 21/400
1/1 [============= ] - 0s 233ms/step - loss: 0.0233 - acc: 0.9786 - val_loss: 0.3607 - val_ac
c: 0.6020
Epoch 22/400
c: 0.5660
Epoch 23/400
1/1 [===========] - 0s 454ms/step - loss: 0.0220 - acc: 0.9643 - val_loss: 0.4130 - val_ac
c: 0.5160
```

```
Epoch 24/400
1/1 [============= ] - 0s 305ms/step - loss: 0.0209 - acc: 0.9857 - val_loss: 0.4197 - val_ac
c: 0.4860
Epoch 25/400
1/1 [=============] - 0s 410ms/step - loss: 0.0225 - acc: 0.9714 - val_loss: 0.4139 - val_ac
c: 0.4680
Epoch 26/400
1/1 [============ ] - 0s 333ms/step - loss: 0.0239 - acc: 0.9643 - val_loss: 0.3842 - val_ac
c: 0.4860
Epoch 27/400
1/1 [============ ] - 0s 406ms/step - loss: 0.0199 - acc: 0.9857 - val_loss: 0.3616 - val_ac
c: 0.5100
Epoch 28/400
1/1 [=========== ] - 0s 310ms/step - loss: 0.0211 - acc: 0.9786 - val loss: 0.3461 - val ac
c: 0.5200
Epoch 29/400
c: 0.5060
Epoch 30/400
1/1 [============= ] - 0s 357ms/step - loss: 0.0234 - acc: 0.9714 - val_loss: 0.3607 - val_ac
c: 0.4900
Epoch 31/400
c: 0.5100
Epoch 32/400
c: 0.5280
Epoch 33/400
c: 0.5400
Epoch 34/400
1/1 [============= ] - 0s 295ms/step - loss: 0.0251 - acc: 0.9714 - val_loss: 0.3326 - val_ac
c: 0.5400
Epoch 35/400
c: 0.5440
Epoch 36/400
c: 0.5400
Epoch 37/400
c: 0.5480
Epoch 38/400
c: 0.5440
Epoch 39/400
1/1 [=========== ] - 0s 372ms/step - loss: 0.0233 - acc: 1.0000 - val loss: 0.3449 - val ac
c: 0.5420
Epoch 40/400
c: 0.5320
Epoch 41/400
1/1 [============ ] - 0s 220ms/step - loss: 0.0235 - acc: 0.9929 - val_loss: 0.3275 - val_ac
c: 0.5200
Epoch 42/400
c: 0.5240
Epoch 43/400
c: 0.5160
Epoch 44/400
c: 0.5200
Epoch 45/400
c: 0.5300
Epoch 46/400
c: 0.5580
Epoch 47/400
1/1 [================] - 0s 175ms/step - loss: 0.0217 - acc: 0.9857 - val_loss: 0.3086 - val_ac
```

```
c: 0.5620
Epoch 48/400
c: 0.5680
Epoch 49/400
1/1 [==========] - 0s 213ms/step - loss: 0.0206 - acc: 0.9786 - val_loss: 0.3140 - val_ac
c: 0.5760
Epoch 50/400
1/1 [============= ] - 0s 184ms/step - loss: 0.0208 - acc: 0.9857 - val_loss: 0.3169 - val_ac
c: 0.5720
Epoch 51/400
c: 0.5820
Epoch 52/400
1/1 [============ ] - 0s 190ms/step - loss: 0.0212 - acc: 0.9857 - val_loss: 0.3079 - val_ac
c: 0.5800
Epoch 53/400
c: 0.5700
Epoch 54/400
c: 0.5400
Epoch 55/400
1/1 [=========== ] - 0s 199ms/step - loss: 0.0191 - acc: 0.9857 - val loss: 0.3325 - val ac
c: 0.5140
Epoch 56/400
c: 0.5100
Epoch 57/400
c: 0.4920
Epoch 58/400
1/1 [============== ] - 0s 170ms/step - loss: 0.0214 - acc: 0.9500 - val_loss: 0.3475 - val_ac
c: 0.5020
Epoch 59/400
c: 0.5160
Epoch 60/400
c: 0.5160
Epoch 61/400
1/1 [=========== ] - 0s 207ms/step - loss: 0.0191 - acc: 0.9929 - val loss: 0.3801 - val ac
c: 0.5160
Epoch 62/400
c: 0.5080
Epoch 63/400
1/1 [============= ] - 0s 182ms/step - loss: 0.0215 - acc: 0.9786 - val_loss: 0.3763 - val_ac
c: 0.5060
Epoch 64/400
c: 0.4980
Epoch 65/400
c: 0.5040
Epoch 66/400
1/1 [============ ] - 0s 174ms/step - loss: 0.0221 - acc: 0.9786 - val_loss: 0.3760 - val_ac
c: 0.4980
Epoch 67/400
c: 0.4980
Epoch 68/400
c: 0.4920
Epoch 69/400
1/1 [============== ] - 0s 210ms/step - loss: 0.0213 - acc: 1.0000 - val_loss: 0.3643 - val_ac
c: 0.5080
Epoch 70/400
1/1 [============= ] - 0s 171ms/step - loss: 0.0243 - acc: 0.9714 - val_loss: 0.3603 - val_ac
c: 0.5080
Epoch 71/400
```

```
c: 0.5140
Epoch 72/400
1/1 [============= ] - 0s 194ms/step - loss: 0.0213 - acc: 0.9929 - val_loss: 0.3742 - val_ac
c: 0.5200
Epoch 73/400
c: 0.5140
Epoch 74/400
c: 0.5160
Epoch 75/400
1/1 [============ ] - 0s 191ms/step - loss: 0.0238 - acc: 0.9714 - val_loss: 0.3846 - val_ac
c: 0.5240
Epoch 76/400
1/1 [============= ] - 0s 179ms/step - loss: 0.0270 - acc: 0.9571 - val_loss: 0.3786 - val_ac
c: 0.5180
Epoch 77/400
1/1 [=========== ] - 0s 173ms/step - loss: 0.0229 - acc: 0.9786 - val loss: 0.3664 - val ac
c: 0.5280
Epoch 78/400
c: 0.5280
Fnoch 79/400
c: 0.5200
Epoch 80/400
c: 0.5240
Epoch 81/400
c: 0.5240
Epoch 82/400
c: 0.5220
Epoch 83/400
c: 0.5160
Epoch 84/400
c: 0.5060
Epoch 85/400
1/1 [===========] - 0s 179ms/step - loss: 0.0228 - acc: 1.0000 - val_loss: 0.3929 - val_ac
c: 0.5080
Epoch 86/400
c: 0.5060
Epoch 87/400
c: 0.5020
Epoch 88/400
1/1 [=========== ] - 0s 177ms/step - loss: 0.0251 - acc: 0.9786 - val loss: 0.4034 - val ac
c: 0.5240
Epoch 89/400
c: 0.4960
Epoch 90/400
1/1 [=========== ] - 0s 168ms/step - loss: 0.0279 - acc: 0.9500 - val loss: 0.4133 - val ac
c: 0.5060
Epoch 91/400
c: 0.5020
Epoch 92/400
1/1 [============= ] - 0s 163ms/step - loss: 0.0252 - acc: 0.9857 - val_loss: 0.4068 - val_ac
c: 0.5200
Epoch 93/400
1/1 [============ ] - 0s 163ms/step - loss: 0.0249 - acc: 0.9929 - val_loss: 0.4095 - val_ac
c: 0.5140
Epoch 94/400
1/1 [===========] - 0s 164ms/step - loss: 0.0244 - acc: 0.9857 - val_loss: 0.4104 - val_ac
c: 0.5280
```

```
Epoch 95/400
1/1 [============= ] - 0s 172ms/step - loss: 0.0289 - acc: 0.9786 - val_loss: 0.3983 - val_ac
c: 0.5400
Epoch 96/400
c: 0.5380
Epoch 97/400
1/1 [============= ] - 0s 177ms/step - loss: 0.0261 - acc: 0.9929 - val_loss: 0.3858 - val_ac
c: 0.5240
Epoch 98/400
1/1 [============ ] - 0s 184ms/step - loss: 0.0293 - acc: 0.9714 - val_loss: 0.3877 - val_ac
c: 0.5220
Epoch 99/400
1/1 [=========== ] - 0s 175ms/step - loss: 0.0269 - acc: 0.9786 - val loss: 0.3835 - val ac
c: 0.5260
Epoch 100/400
c: 0.5260
Epoch 101/400
1/1 [============= ] - 0s 196ms/step - loss: 0.0327 - acc: 0.9643 - val_loss: 0.3792 - val_ac
c: 0.5200
Epoch 102/400
c: 0.5280
Epoch 103/400
c: 0.5300
Epoch 104/400
c: 0.5480
Epoch 105/400
1/1 [============= ] - 0s 195ms/step - loss: 0.0299 - acc: 0.9571 - val_loss: 0.3728 - val_ac
c: 0.5480
Epoch 106/400
1/1 [=========== ] - 0s 203ms/step - loss: 0.0320 - acc: 0.9714 - val loss: 0.3766 - val ac
c: 0.5520
Epoch 107/400
c: 0.5360
Epoch 108/400
c: 0.5320
Epoch 109/400
c: 0.5280
Epoch 110/400
1/1 [=========== ] - 0s 176ms/step - loss: 0.0282 - acc: 0.9929 - val loss: 0.3975 - val ac
c: 0.5200
Epoch 111/400
1/1 [============= ] - 0s 185ms/step - loss: 0.0294 - acc: 0.9857 - val_loss: 0.4070 - val_ac
c: 0.5200
Epoch 112/400
1/1 [==========] - 0s 190ms/step - loss: 0.0351 - acc: 0.9643 - val_loss: 0.4114 - val_ac
c: 0.5160
Epoch 113/400
c: 0.5120
Epoch 114/400
c: 0.5160
Epoch 115/400
c: 0.5080
Epoch 116/400
c: 0.5140
Epoch 117/400
c: 0.5240
Epoch 118/400
```

```
c: 0.5220
Epoch 119/400
c: 0.5280
Epoch 120/400
1/1 [============= ] - 0s 192ms/step - loss: 0.0318 - acc: 0.9929 - val_loss: 0.4261 - val_ac
c: 0.5300
Epoch 121/400
1/1 [===========] - 0s 218ms/step - loss: 0.0353 - acc: 0.9786 - val_loss: 0.4202 - val_ac
c: 0.5280
Epoch 122/400
c: 0.5140
Epoch 123/400
1/1 [============ ] - 0s 163ms/step - loss: 0.0325 - acc: 0.9857 - val_loss: 0.4260 - val_ac
c: 0.4820
Epoch 124/400
c: 0.4700
Epoch 125/400
c: 0.4540
Epoch 126/400
c: 0.4640
Epoch 127/400
c: 0.4500
Epoch 128/400
c: 0.4560
Epoch 129/400
c: 0.4600
Epoch 130/400
c: 0.4760
Epoch 131/400
c: 0.4940
Epoch 132/400
c: 0.5060
Epoch 133/400
c: 0.5100
Epoch 134/400
1/1 [============= ] - 0s 164ms/step - loss: 0.0327 - acc: 0.9786 - val_loss: 0.3887 - val_ac
c: 0.5180
Epoch 135/400
c: 0.5160
Epoch 136/400
1/1 [============== ] - 0s 168ms/step - loss: 0.0312 - acc: 1.0000 - val_loss: 0.3800 - val_ac
c: 0.5320
Epoch 137/400
1/1 [============= ] - 0s 160ms/step - loss: 0.0318 - acc: 0.9929 - val_loss: 0.3813 - val_ac
c: 0.5280
Epoch 138/400
c: 0.5260
Epoch 139/400
1/1 [=========== ] - 0s 180ms/step - loss: 0.0332 - acc: 0.9786 - val loss: 0.4027 - val ac
c: 0.5200
Epoch 140/400
c: 0.5160
Epoch 141/400
1/1 [============= ] - 0s 173ms/step - loss: 0.0341 - acc: 0.9714 - val_loss: 0.4484 - val_ac
c: 0.5000
Epoch 142/400
```

```
c: 0.4760
Epoch 143/400
1/1 [============ ] - 0s 178ms/step - loss: 0.0344 - acc: 0.9714 - val_loss: 0.4529 - val_ac
c: 0.4720
Epoch 144/400
c: 0.4760
Epoch 145/400
c: 0.4860
Epoch 146/400
1/1 [============ ] - 0s 167ms/step - loss: 0.0334 - acc: 0.9714 - val_loss: 0.3973 - val_ac
c: 0.5140
Epoch 147/400
1/1 [============= ] - 0s 176ms/step - loss: 0.0321 - acc: 0.9786 - val_loss: 0.3882 - val_ac
c: 0.5120
Epoch 148/400
1/1 [=========== ] - 0s 166ms/step - loss: 0.0327 - acc: 0.9857 - val loss: 0.3842 - val ac
c: 0.5280
Epoch 149/400
c: 0.5360
Fnoch 150/400
c: 0.5360
Epoch 151/400
1/1 [============== ] - 0s 177ms/step - loss: 0.0326 - acc: 0.9786 - val_loss: 0.4046 - val_ac
c: 0.5300
Epoch 152/400
c: 0.5120
Epoch 153/400
c: 0.5080
Epoch 154/400
c: 0.5100
Epoch 155/400
c: 0.5060
Epoch 156/400
1/1 [============= ] - 0s 169ms/step - loss: 0.0331 - acc: 0.9643 - val_loss: 0.4190 - val_ac
c: 0.4960
Epoch 157/400
c: 0.4940
Epoch 158/400
c: 0.4760
Epoch 159/400
1/1 [=========== ] - 0s 176ms/step - loss: 0.0309 - acc: 0.9857 - val loss: 0.4404 - val ac
c: 0.4700
Epoch 160/400
c: 0.4800
Epoch 161/400
1/1 [=========== ] - 0s 179ms/step - loss: 0.0309 - acc: 0.9857 - val loss: 0.4320 - val ac
c: 0.4780
Epoch 162/400
c: 0.4940
Epoch 163/400
1/1 [============ ] - 0s 169ms/step - loss: 0.0301 - acc: 0.9929 - val_loss: 0.4250 - val_ac
c: 0.4960
Epoch 164/400
1/1 [============= ] - 0s 182ms/step - loss: 0.0297 - acc: 0.9857 - val_loss: 0.4280 - val_ac
c: 0.4980
Epoch 165/400
c: 0.4920
```

```
Epoch 166/400
1/1 [============= ] - 0s 170ms/step - loss: 0.0314 - acc: 0.9786 - val_loss: 0.4232 - val_ac
c: 0.5020
Epoch 167/400
c: 0.5080
Epoch 168/400
1/1 [============= ] - 0s 170ms/step - loss: 0.0341 - acc: 0.9571 - val_loss: 0.4097 - val_ac
c: 0.5100
Epoch 169/400
c: 0.5240
Epoch 170/400
1/1 [=========== ] - 0s 162ms/step - loss: 0.0297 - acc: 0.9857 - val loss: 0.3876 - val ac
c: 0.5320
Epoch 171/400
c: 0.5360
Epoch 172/400
1/1 [============= ] - 0s 176ms/step - loss: 0.0297 - acc: 0.9857 - val_loss: 0.3712 - val_ac
c: 0.5500
Epoch 173/400
c: 0.5480
Epoch 174/400
c: 0.5580
Epoch 175/400
c: 0.5440
Epoch 176/400
1/1 [============ ] - 0s 170ms/step - loss: 0.0349 - acc: 0.9643 - val_loss: 0.4174 - val_ac
c: 0.5280
Epoch 177/400
c: 0.5240
Epoch 178/400
1/1 [============== ] - 0s 179ms/step - loss: 0.0307 - acc: 0.9929 - val_loss: 0.4443 - val_ac
c: 0.5060
Epoch 179/400
c: 0.4920
Epoch 180/400
c: 0.4800
Epoch 181/400
1/1 [=========== ] - 0s 224ms/step - loss: 0.0370 - acc: 0.9643 - val loss: 0.4603 - val ac
c: 0.4960
Epoch 182/400
c: 0.5100
Epoch 183/400
1/1 [==========] - 0s 184ms/step - loss: 0.0391 - acc: 0.9500 - val_loss: 0.4436 - val_ac
c: 0.5020
Epoch 184/400
c: 0.5000
Epoch 185/400
c: 0.4940
Epoch 186/400
c: 0.4860
Epoch 187/400
c: 0.4900
Epoch 188/400
c: 0.4860
Epoch 189/400
1/1 [=================] - 0s 224ms/step - loss: 0.0360 - acc: 0.9643 - val_loss: 0.4333 - val_ac
```

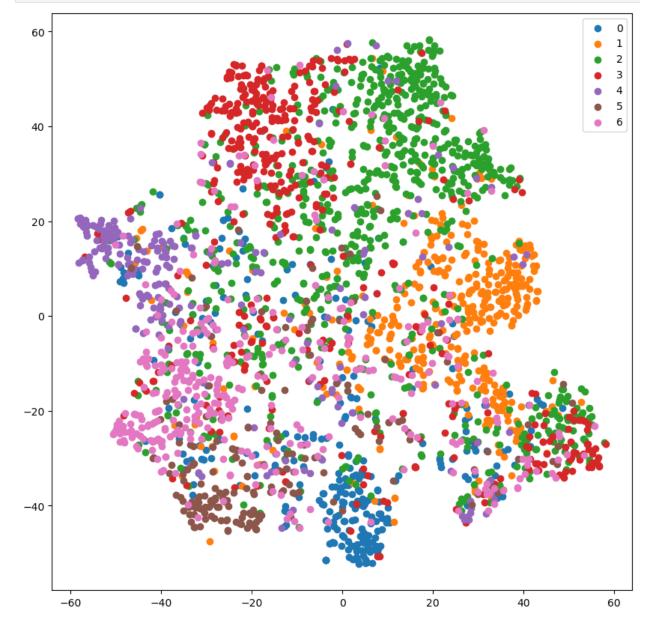
```
c: 0.4940
     Epoch 190/400
    1/1 [=========== ] - 0s 206ms/step - loss: 0.0343 - acc: 0.9929 - val loss: 0.4245 - val ac
     c: 0.5040
    1/1 [============= ] - 0s 182ms/step - loss: 0.0369 - acc: 0.9643 - val_loss: 0.4188 - val_ac
     c: 0.4980
    Epoch 192/400
    1/1 [============= ] - 0s 191ms/step - loss: 0.0374 - acc: 0.9786 - val_loss: 0.4203 - val_ac
    c: 0.4840
    Epoch 193/400
    c: 0.4760
    Epoch 194/400
    1/1 [============= ] - 0s 200ms/step - loss: 0.0369 - acc: 0.9714 - val_loss: 0.4386 - val_ac
     c: 0.4540
    Epoch 195/400
    c: 0.4440
    Epoch 196/400
    c: 0.4460
    Epoch 197/400
    1/1 [=========== ] - 0s 200ms/step - loss: 0.0417 - acc: 0.9714 - val loss: 0.4699 - val ac
    c: 0.4480
    Epoch 198/400
    c: 0.4440
    Epoch 199/400
    1/1 [============ ] - 0s 166ms/step - loss: 0.0356 - acc: 0.9929 - val_loss: 0.4679 - val_ac
    c: 0.4480
     Epoch 200/400
    1/1 [============== ] - 0s 170ms/step - loss: 0.0420 - acc: 0.9500 - val_loss: 0.4586 - val_ac
    c: 0.4580
    Epoch 201/400
    1/1 [============ ] - 0s 196ms/step - loss: 0.0367 - acc: 0.9929 - val_loss: 0.4475 - val_ac
    c: 0.4540
    Epoch 202/400
    c: 0.4640
    Epoch 203/400
    c: 0.4640
    Epoch 204/400
    c: 0.4620
    Epoch 205/400
    1/1 [============ ] - 0s 207ms/step - loss: 0.0422 - acc: 0.9714 - val_loss: 0.4335 - val_ac
    c: 0.4600
    Epoch 206/400
    c: 0.4400
     Epoch 207/400
     Out[]: <tensorflow.python.keras.callbacks.History at 0x1d55a884d08>
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 0.49 0.59 0.53 114 Genetic_Algorithms 0.67 0.81 0.74 156 Neural_Networks 0.73 0.52 0.61 290 0.54 0.60 172 Probabilistic_Methods 0.66 Reinforcement_Learning 0.43 0.52 0.47 85 Rule_Learning 0.43 0.73 0.54 60 Theory 0.43 0.43 0.43 123 0.58 1000 accuracy 0.55 0.59 0.56 1000 macro avg weighted avg 0.60 0.58 0.58 1000

Get hidden layer representation for FNN

```
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])
```





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