# Node Classification using Graph Convolutional Networks

300

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: {'Theory', 'Neural_Networks', 'Rule_Learning', 'Genetic_Algorithms', 'Probabilistic_Methods', 'C
        ase_Based', 'Reinforcement_Learning'}
        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
        12_{reg} = 5e-4
        es_patience = 150

# Number of training epochs
# 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 1/300
1/1 [============] - 0s 345ms/step - loss: 0.1168 - acc: 0.1286 - val_loss: 0.3655 - val_ac
c: 0.1780
Epoch 2/300
1/1 [=============] - ETA: 0s - loss: 0.1097 - acc: 0.2214WARNING:tensorflow:Method (on_trai
n_batch_end) is slow compared to the batch update (0.169520). Check your callbacks.
c: 0.2380
Epoch 3/300
c: 0.2640
Epoch 4/300
1/1 [============ ] - 0s 173ms/step - loss: 0.0969 - acc: 0.4214 - val_loss: 0.3370 - val_ac
c: 0.2860
Epoch 5/300
1/1 [===========] - 0s 173ms/step - loss: 0.0929 - acc: 0.4071 - val_loss: 0.3274 - val_ac
c: 0.3380
Epoch 6/300
1/1 [=========== ] - 0s 180ms/step - loss: 0.0877 - acc: 0.5000 - val loss: 0.3177 - val ac
c: 0.4080
Epoch 7/300
1/1 [==============] - 0s 170ms/step - loss: 0.0837 - acc: 0.5857 - val_loss: 0.3084 - val_ac
c: 0.4960
Fnoch 8/300
c: 0.5780
Epoch 9/300
c: 0.6360
Epoch 10/300
c: 0.6920
Epoch 11/300
c: 0.7280
Epoch 12/300
1/1 [============= ] - 0s 307ms/step - loss: 0.0727 - acc: 0.8714 - val_loss: 0.2664 - val_ac
c: 0.7480
Epoch 13/300
c: 0.7540
Epoch 14/300
1/1 [============= ] - 0s 355ms/step - loss: 0.0690 - acc: 0.8714 - val_loss: 0.2528 - val_ac
c: 0.7600
Epoch 15/300
1/1 [============ ] - 0s 291ms/step - loss: 0.0661 - acc: 0.8929 - val_loss: 0.2465 - val_ac
c: 0.7660
Epoch 16/300
c: 0.7640
Epoch 17/300
1/1 [=========== ] - 0s 248ms/step - loss: 0.0645 - acc: 0.9071 - val loss: 0.2356 - val ac
c: 0.7660
Epoch 18/300
c: 0.7700
Epoch 19/300
1/1 [=========== ] - 0s 436ms/step - loss: 0.0614 - acc: 0.9143 - val loss: 0.2266 - val ac
c: 0.7740
Epoch 20/300
c: 0.7740
Epoch 21/300
1/1 [============= ] - 0s 285ms/step - loss: 0.0578 - acc: 0.9143 - val_loss: 0.2185 - val_ac
c: 0.7760
Epoch 22/300
c: 0.7720
Epoch 23/300
1/1 [===========] - 0s 223ms/step - loss: 0.0560 - acc: 0.9286 - val_loss: 0.2121 - val_ac
c: 0.7760
```

```
Epoch 24/300
1/1 [============= ] - 0s 237ms/step - loss: 0.0542 - acc: 0.9286 - val_loss: 0.2094 - val_ac
c: 0.7740
Epoch 25/300
c: 0.7800
Epoch 26/300
1/1 [===========] - 0s 390ms/step - loss: 0.0546 - acc: 0.9214 - val_loss: 0.2042 - val_ac
c: 0.7800
Epoch 27/300
c: 0.7800
Epoch 28/300
1/1 [=========== ] - 0s 252ms/step - loss: 0.0506 - acc: 0.9357 - val loss: 0.1996 - val ac
c: 0.7800
Epoch 29/300
c: 0.7800
Epoch 30/300
1/1 [============= ] - 0s 214ms/step - loss: 0.0492 - acc: 0.9214 - val_loss: 0.1949 - val_ac
c: 0.7800
Epoch 31/300
c: 0.7820
Epoch 32/300
c: 0.7880
Epoch 33/300
c: 0.7860
Epoch 34/300
1/1 [============= ] - 0s 202ms/step - loss: 0.0497 - acc: 0.9286 - val_loss: 0.1864 - val_ac
c: 0.7840
Epoch 35/300
c: 0.7860
Epoch 36/300
c: 0.7860
Epoch 37/300
c: 0.7840
Epoch 38/300
c: 0.7780
Epoch 39/300
1/1 [============ ] - 0s 204ms/step - loss: 0.0412 - acc: 0.9857 - val loss: 0.1821 - val ac
c: 0.7800
Epoch 40/300
c: 0.7840
Epoch 41/300
1/1 [============= ] - 0s 174ms/step - loss: 0.0413 - acc: 0.9357 - val_loss: 0.1794 - val_ac
c: 0.7840
Epoch 42/300
c: 0.7820
Epoch 43/300
c: 0.7720
Epoch 44/300
c: 0.7760
Epoch 45/300
c: 0.7740
Epoch 46/300
c: 0.7740
Epoch 47/300
```

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```
c: 0.7740
Epoch 72/300
1/1 [============= ] - 0s 195ms/step - loss: 0.0349 - acc: 0.9929 - val_loss: 0.1608 - val_ac
c: 0.7720
Epoch 73/300
c: 0.7740
Epoch 74/300
c: 0.7800
Epoch 75/300
1/1 [============ ] - 0s 188ms/step - loss: 0.0335 - acc: 0.9714 - val_loss: 0.1589 - val_ac
c: 0.7800
Epoch 76/300
1/1 [============ ] - 0s 190ms/step - loss: 0.0335 - acc: 0.9357 - val_loss: 0.1589 - val_ac
c: 0.7880
Epoch 77/300
1/1 [=========== ] - 0s 179ms/step - loss: 0.0290 - acc: 1.0000 - val loss: 0.1585 - val ac
c: 0.7900
Epoch 78/300
1/1 [==============] - 0s 176ms/step - loss: 0.0365 - acc: 0.9714 - val_loss: 0.1591 - val_ac
c: 0.7940
Fnoch 79/300
c: 0.7880
Epoch 80/300
c: 0.7860
Epoch 81/300
c: 0.7840
Epoch 82/300
c: 0.7840
Epoch 83/300
1/1 [============= ] - 0s 168ms/step - loss: 0.0289 - acc: 0.9857 - val_loss: 0.1585 - val_ac
c: 0.7820
Epoch 84/300
c: 0.7820
Epoch 85/300
1/1 [============= ] - 0s 168ms/step - loss: 0.0333 - acc: 0.9357 - val_loss: 0.1565 - val_ac
c: 0.7860
Epoch 86/300
c: 0.7860
Epoch 87/300
c: 0.7860
Epoch 88/300
1/1 [=========== ] - 0s 187ms/step - loss: 0.0298 - acc: 0.9714 - val loss: 0.1544 - val ac
c: 0.7860
Epoch 89/300
c: 0.7860
Epoch 90/300
1/1 [=========== ] - 0s 185ms/step - loss: 0.0321 - acc: 0.9429 - val loss: 0.1571 - val ac
c: 0.7820
Epoch 91/300
c: 0.7840
Epoch 92/300
1/1 [===========] - 0s 173ms/step - loss: 0.0300 - acc: 0.9643 - val_loss: 0.1583 - val_ac
c: 0.7880
Epoch 93/300
1/1 [============= ] - 0s 202ms/step - loss: 0.0311 - acc: 0.9643 - val_loss: 0.1568 - val_ac
c: 0.7860
Epoch 94/300
c: 0.7840
```

```
Epoch 95/300
1/1 [============== ] - 0s 169ms/step - loss: 0.0279 - acc: 0.9857 - val_loss: 0.1526 - val_ac
c: 0.7880
Epoch 96/300
c: 0.7900
Epoch 97/300
1/1 [===========] - 0s 190ms/step - loss: 0.0297 - acc: 0.9500 - val_loss: 0.1472 - val_ac
c: 0.7920
Epoch 98/300
1/1 [============= ] - 0s 184ms/step - loss: 0.0292 - acc: 0.9714 - val_loss: 0.1465 - val_ac
c: 0.7900
Epoch 99/300
1/1 [=========== ] - 0s 169ms/step - loss: 0.0306 - acc: 0.9714 - val loss: 0.1469 - val ac
c: 0.7820
Epoch 100/300
c: 0.7720
Epoch 101/300
1/1 [============= ] - 0s 180ms/step - loss: 0.0325 - acc: 0.9643 - val_loss: 0.1493 - val_ac
c: 0.7700
Epoch 102/300
c: 0.7700
Epoch 103/300
c: 0.7740
Epoch 104/300
c: 0.7620
Epoch 105/300
1/1 [============= ] - 0s 171ms/step - loss: 0.0277 - acc: 0.9857 - val_loss: 0.1527 - val_ac
c: 0.7600
Epoch 106/300
c: 0.7660
Epoch 107/300
c: 0.7800
Epoch 108/300
c: 0.7760
Epoch 109/300
c: 0.7680
Epoch 110/300
1/1 [============ ] - 0s 170ms/step - loss: 0.0306 - acc: 0.9571 - val loss: 0.1550 - val ac
c: 0.7700
Epoch 111/300
1/1 [============= ] - 0s 170ms/step - loss: 0.0270 - acc: 0.9714 - val_loss: 0.1544 - val_ac
c: 0.7640
Epoch 112/300
1/1 [============= ] - 0s 211ms/step - loss: 0.0280 - acc: 0.9714 - val_loss: 0.1531 - val_ac
c: 0.7660
Epoch 113/300
c: 0.7760
Epoch 114/300
c: 0.7820
Epoch 115/300
c: 0.7800
Epoch 116/300
c: 0.7840
Epoch 117/300
c: 0.7800
Epoch 118/300
1/1 [=================] - 0s 162ms/step - loss: 0.0275 - acc: 0.9643 - val_loss: 0.1486 - val_ac
```

```
c: 0.7820
Epoch 119/300
c: 0.7900
Epoch 120/300
1/1 [==========] - 0s 188ms/step - loss: 0.0291 - acc: 0.9429 - val_loss: 0.1465 - val_ac
c: 0.7900
Epoch 121/300
1/1 [===========] - 0s 184ms/step - loss: 0.0285 - acc: 0.9786 - val_loss: 0.1444 - val_ac
c: 0.7880
Epoch 122/300
c: 0.7960
Epoch 123/300
1/1 [============= ] - 0s 201ms/step - loss: 0.0275 - acc: 0.9714 - val_loss: 0.1435 - val_ac
c: 0.7980
Epoch 124/300
c: 0.7980
Epoch 125/300
c: 0.7920
Epoch 126/300
1/1 [=========== ] - 0s 184ms/step - loss: 0.0288 - acc: 0.9714 - val loss: 0.1461 - val ac
c: 0.7880
Epoch 127/300
c: 0.7820
Epoch 128/300
c: 0.7760
Epoch 129/300
c: 0.7680
Epoch 130/300
c: 0.7680
Epoch 131/300
c: 0.7740
Epoch 132/300
c: 0.7760
Epoch 133/300
c: 0.7820
Epoch 134/300
1/1 [============= ] - 0s 217ms/step - loss: 0.0292 - acc: 0.9500 - val_loss: 0.1410 - val_ac
c: 0.7880
Epoch 135/300
c: 0.7920
Epoch 136/300
c: 0.7880
Epoch 137/300
1/1 [============= ] - 0s 247ms/step - loss: 0.0271 - acc: 0.9786 - val_loss: 0.1394 - val_ac
c: 0.7880
Epoch 138/300
c: 0.7820
Epoch 139/300
1/1 [=========== ] - 0s 253ms/step - loss: 0.0257 - acc: 0.9857 - val loss: 0.1445 - val ac
c: 0.7820
Epoch 140/300
c: 0.7820
Epoch 141/300
1/1 [============= ] - 0s 227ms/step - loss: 0.0237 - acc: 0.9786 - val_loss: 0.1490 - val_ac
c: 0.7840
Epoch 142/300
```

```
c: 0.7780
Epoch 143/300
1/1 [============= ] - 0s 253ms/step - loss: 0.0258 - acc: 0.9714 - val_loss: 0.1489 - val_ac
c: 0.7740
Epoch 144/300
c: 0.7740
Epoch 145/300
c: 0.7760
Epoch 146/300
1/1 [============= ] - 0s 232ms/step - loss: 0.0270 - acc: 0.9714 - val_loss: 0.1436 - val_ac
c: 0.7840
Epoch 147/300
1/1 [============= ] - 0s 247ms/step - loss: 0.0258 - acc: 0.9786 - val_loss: 0.1425 - val_ac
c: 0.7820
Epoch 148/300
1/1 [=========== ] - 0s 308ms/step - loss: 0.0252 - acc: 0.9571 - val loss: 0.1411 - val ac
c: 0.7740
Epoch 149/300
c: 0.7740
Fnoch 150/300
c: 0.7840
Epoch 151/300
c: 0.7840
Epoch 152/300
c: 0.7800
Epoch 153/300
c: 0.7740
Epoch 154/300
c: 0.7760
Epoch 155/300
c: 0.7760
Epoch 156/300
1/1 [===========] - 0s 193ms/step - loss: 0.0253 - acc: 0.9643 - val_loss: 0.1457 - val_ac
c: 0.7740
Epoch 157/300
1/1 [============= ] - 0s 215ms/step - loss: 0.0240 - acc: 0.9714 - val_loss: 0.1469 - val_ac
c: 0.7820
Epoch 158/300
c: 0.7780
Epoch 159/300
1/1 [=========== ] - 0s 232ms/step - loss: 0.0261 - acc: 0.9286 - val loss: 0.1454 - val ac
c: 0.7720
Epoch 160/300
c: 0.7780
Epoch 161/300
1/1 [=========== ] - 0s 219ms/step - loss: 0.0274 - acc: 0.9286 - val loss: 0.1423 - val ac
c: 0.7780
Epoch 162/300
c: 0.7800
Epoch 163/300
1/1 [============= ] - 0s 238ms/step - loss: 0.0242 - acc: 0.9714 - val_loss: 0.1457 - val_ac
c: 0.7740
Epoch 164/300
c: 0.7760
Epoch 165/300
1/1 [============] - 0s 230ms/step - loss: 0.0255 - acc: 0.9571 - val_loss: 0.1472 - val_ac
c: 0.7860
```

```
Epoch 166/300
1/1 [============ ] - 0s 193ms/step - loss: 0.0237 - acc: 0.9786 - val_loss: 0.1463 - val_ac
c: 0.7780
Epoch 167/300
1/1 [=============] - 0s 181ms/step - loss: 0.0273 - acc: 0.9571 - val_loss: 0.1458 - val_ac
c: 0.7660
Epoch 168/300
1/1 [===========] - 0s 204ms/step - loss: 0.0224 - acc: 0.9857 - val_loss: 0.1455 - val_ac
c: 0.7660
Epoch 169/300
1/1 [============= ] - 0s 289ms/step - loss: 0.0252 - acc: 0.9714 - val_loss: 0.1461 - val_ac
c: 0.7640
Epoch 170/300
1/1 [============ ] - 0s 236ms/step - loss: 0.0258 - acc: 0.9786 - val loss: 0.1458 - val ac
c: 0.7640
Epoch 171/300
c: 0.7660
Epoch 172/300
1/1 [===========] - 0s 201ms/step - loss: 0.0244 - acc: 0.9786 - val_loss: 0.1449 - val_ac
c: 0.7660
Epoch 173/300
c: 0.7720
Epoch 174/300
c: 0.7780
Epoch 175/300
c: 0.7820
Epoch 176/300
1/1 [============= ] - 0s 254ms/step - loss: 0.0232 - acc: 0.9786 - val_loss: 0.1460 - val_ac
c: 0.7800
Epoch 177/300
c: 0.7620
Epoch 178/300
c: 0.7560
Epoch 179/300
c: 0.7500
Epoch 180/300
c: 0.7460
Epoch 181/300
1/1 [=========== ] - 0s 171ms/step - loss: 0.0232 - acc: 0.9857 - val loss: 0.1456 - val ac
c: 0.7540
Epoch 182/300
c: 0.7620
Epoch 183/300
1/1 [============= ] - 0s 166ms/step - loss: 0.0239 - acc: 0.9714 - val_loss: 0.1447 - val_ac
c: 0.7660
Epoch 184/300
c: 0.7720
Epoch 185/300
c: 0.7740
Epoch 186/300
c: 0.7740
Epoch 187/300
c: 0.7760
Epoch 188/300
c: 0.7800
Epoch 189/300
1/1 [================] - 0s 203ms/step - loss: 0.0238 - acc: 0.9643 - val_loss: 0.1484 - val_ac
```

```
c: 0.7700
Epoch 190/300
c: 0.7660
Epoch 191/300
1/1 [============= ] - 0s 183ms/step - loss: 0.0256 - acc: 0.9571 - val_loss: 0.1499 - val_ac
c: 0.7620
Epoch 192/300
1/1 [===========] - 0s 171ms/step - loss: 0.0244 - acc: 0.9786 - val_loss: 0.1468 - val_ac
c: 0.7720
Epoch 193/300
c: 0.7620
Epoch 194/300
1/1 [===========] - 0s 159ms/step - loss: 0.0250 - acc: 0.9714 - val_loss: 0.1422 - val_ac
c: 0.7780
Epoch 195/300
c: 0.7780
Epoch 196/300
c: 0.7760
Epoch 197/300
c: 0.7820
Epoch 198/300
c: 0.7760
Epoch 199/300
c: 0.7840
Epoch 200/300
c: 0.7800
Epoch 201/300
c: 0.7860
Epoch 202/300
c: 0.7860
Epoch 203/300
c: 0.7840
Epoch 204/300
c: 0.7740
Epoch 205/300
1/1 [============= ] - 0s 180ms/step - loss: 0.0225 - acc: 0.9857 - val_loss: 0.1409 - val_ac
c: 0.7820
Epoch 206/300
c: 0.7840
Epoch 207/300
c: 0.7860
Epoch 208/300
c: 0.7840
Epoch 209/300
c: 0.7780
Epoch 210/300
1/1 [=========== ] - 0s 181ms/step - loss: 0.0233 - acc: 0.9714 - val loss: 0.1442 - val ac
c: 0.7780
Epoch 211/300
c: 0.7760
Epoch 212/300
1/1 [============= ] - 0s 191ms/step - loss: 0.0221 - acc: 0.9643 - val_loss: 0.1444 - val_ac
c: 0.7700
Epoch 213/300
```

```
c: 0.7740
Epoch 214/300
1/1 [============= ] - 0s 168ms/step - loss: 0.0229 - acc: 0.9786 - val_loss: 0.1402 - val_ac
c: 0.7780
Epoch 215/300
c: 0.7820
Epoch 216/300
c: 0.7860
Epoch 217/300
1/1 [============= ] - 0s 184ms/step - loss: 0.0217 - acc: 0.9714 - val_loss: 0.1370 - val_ac
c: 0.7880
Epoch 218/300
1/1 [============ ] - 0s 170ms/step - loss: 0.0224 - acc: 0.9643 - val_loss: 0.1405 - val_ac
c: 0.7880
Epoch 219/300
1/1 [=========== ] - 0s 173ms/step - loss: 0.0217 - acc: 0.9571 - val loss: 0.1449 - val ac
c: 0.7800
Epoch 220/300
c: 0.7800
Fnoch 221/300
c: 0.7740
Epoch 222/300
c: 0.7740
Epoch 223/300
c: 0.7760
Epoch 224/300
c: 0.7800
Epoch 225/300
c: 0.7940
Epoch 226/300
c: 0.7860
Epoch 227/300
1/1 [============= ] - 0s 177ms/step - loss: 0.0210 - acc: 0.9857 - val_loss: 0.1353 - val_ac
c: 0.7820
Epoch 228/300
c: 0.7840
Epoch 229/300
c: 0.7840
Epoch 230/300
1/1 [=========== ] - 0s 173ms/step - loss: 0.0229 - acc: 0.9714 - val loss: 0.1430 - val ac
c: 0.7860
Epoch 231/300
c: 0.7920
Epoch 232/300
1/1 [=========== ] - 0s 187ms/step - loss: 0.0219 - acc: 0.9786 - val loss: 0.1423 - val ac
c: 0.7900
Epoch 233/300
c: 0.7900
Epoch 234/300
1/1 [===========] - 0s 182ms/step - loss: 0.0196 - acc: 0.9786 - val_loss: 0.1372 - val_ac
c: 0.7920
Epoch 235/300
1/1 [============= ] - 0s 186ms/step - loss: 0.0221 - acc: 0.9714 - val_loss: 0.1362 - val_ac
c: 0.7780
Epoch 236/300
1/1 [============] - 0s 205ms/step - loss: 0.0221 - acc: 0.9714 - val_loss: 0.1371 - val_ac
c: 0.7760
```

```
Epoch 237/300
1/1 [============= ] - 0s 170ms/step - loss: 0.0223 - acc: 0.9500 - val_loss: 0.1414 - val_ac
c: 0.7700
Epoch 238/300
c: 0.7620
Epoch 239/300
1/1 [===========] - 0s 196ms/step - loss: 0.0205 - acc: 0.9643 - val_loss: 0.1513 - val_ac
c: 0.7620
Epoch 240/300
1/1 [============ ] - 0s 198ms/step - loss: 0.0198 - acc: 0.9857 - val_loss: 0.1525 - val_ac
c: 0.7660
Epoch 241/300
1/1 [============ ] - 0s 223ms/step - loss: 0.0214 - acc: 0.9857 - val loss: 0.1507 - val ac
c: 0.7660
Epoch 242/300
c: 0.7700
Epoch 243/300
1/1 [============= ] - 0s 195ms/step - loss: 0.0214 - acc: 0.9857 - val_loss: 0.1443 - val_ac
c: 0.7740
Epoch 244/300
c: 0.7740
Epoch 245/300
c: 0.7660
Epoch 246/300
c: 0.7720
Epoch 247/300
1/1 [============= ] - 0s 216ms/step - loss: 0.0233 - acc: 0.9643 - val_loss: 0.1383 - val_ac
c: 0.7740
Epoch 248/300
c: 0.7700
Epoch 249/300
c: 0.7720
Epoch 250/300
c: 0.7760
Epoch 251/300
c: 0.7840
Epoch 252/300
1/1 [=========== ] - 0s 171ms/step - loss: 0.0199 - acc: 1.0000 - val loss: 0.1385 - val ac
c: 0.7920
Epoch 253/300
c: 0.7860
Epoch 254/300
1/1 [============= ] - 0s 173ms/step - loss: 0.0216 - acc: 0.9786 - val_loss: 0.1371 - val_ac
c: 0.7820
Epoch 255/300
c: 0.7780
Epoch 256/300
c: 0.7760
Epoch 257/300
c: 0.7680
Epoch 258/300
c: 0.7660
Epoch 259/300
c: 0.7680
Epoch 260/300
1/1 [==================] - 0s 158ms/step - loss: 0.0198 - acc: 1.0000 - val_loss: 0.1424 - val_ac
```

7/12/2022

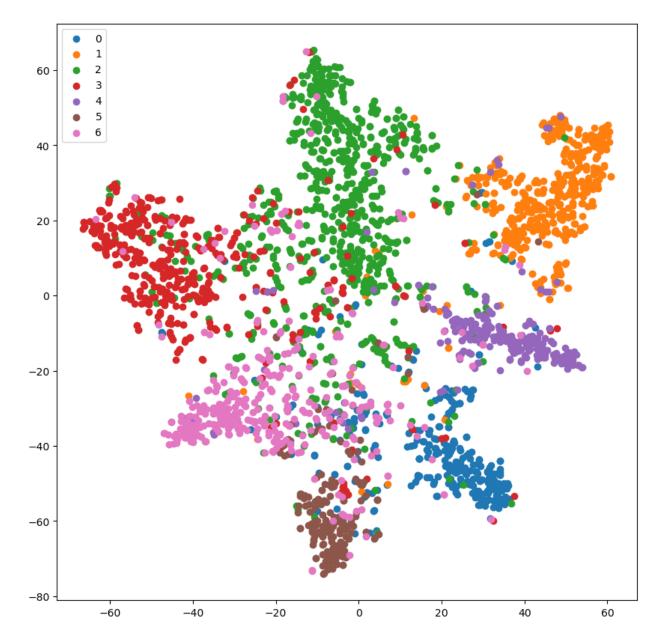
```
c: 0.7640
Epoch 261/300
c: 0.7600
Epoch 262/300
1/1 [==========] - 0s 160ms/step - loss: 0.0195 - acc: 1.0000 - val_loss: 0.1443 - val_ac
c: 0.7620
Epoch 263/300
1/1 [===========] - 0s 159ms/step - loss: 0.0208 - acc: 0.9786 - val_loss: 0.1440 - val_ac
c: 0.7660
Epoch 264/300
c: 0.7660
Epoch 265/300
1/1 [==========] - 0s 160ms/step - loss: 0.0209 - acc: 0.9714 - val_loss: 0.1424 - val_ac
c: 0.7720
Epoch 266/300
c: 0.7780
Epoch 267/300
c: 0.7860
Epoch 268/300
c: 0.7900
Epoch 269/300
c: 0.7860
Epoch 270/300
c: 0.7820
Epoch 271/300
c: 0.7760
Epoch 272/300
c: 0.7640
Epoch 273/300
c: 0.7740
Epoch 274/300
c: 0.7780
Epoch 275/300
c: 0.7820
Epoch 276/300
1/1 [============= ] - 0s 158ms/step - loss: 0.0202 - acc: 0.9857 - val_loss: 0.1364 - val_ac
c: 0.7880
Epoch 277/300
c: 0.7920
Epoch 278/300
c: 0.7920
Epoch 279/300
1/1 [============= ] - 0s 156ms/step - loss: 0.0220 - acc: 0.9643 - val_loss: 0.1394 - val_ac
c: 0.7880
Epoch 280/300
c: 0.7840
Epoch 281/300
1/1 [=========== ] - 0s 158ms/step - loss: 0.0196 - acc: 0.9786 - val loss: 0.1480 - val ac
c: 0.7700
Epoch 282/300
c: 0.7620
Epoch 283/300
1/1 [============= ] - 0s 166ms/step - loss: 0.0215 - acc: 0.9786 - val_loss: 0.1511 - val_ac
c: 0.7660
Epoch 284/300
```

```
c: 0.7640
    Epoch 285/300
    c: 0.7680
    Epoch 286/300
    c: 0.7720
    Epoch 287/300
    c: 0.7760
    Epoch 288/300
    1/1 [============= ] - 0s 162ms/step - loss: 0.0248 - acc: 0.9500 - val_loss: 0.1358 - val_ac
    c: 0.7800
    Epoch 289/300
    1/1 [============= ] - 0s 155ms/step - loss: 0.0191 - acc: 0.9857 - val_loss: 0.1395 - val_ac
    c: 0.7780
    Epoch 290/300
    1/1 [=========== ] - 0s 161ms/step - loss: 0.0204 - acc: 0.9643 - val loss: 0.1463 - val ac
    c: 0.7680
    Epoch 291/300
    c: 0.7460
    Fnoch 292/300
    1/1 [============= ] - 0s 160ms/step - loss: 0.0195 - acc: 0.9857 - val_loss: 0.1583 - val_ac
    c: 0.7420
    Epoch 293/300
    c: 0.7440
    Epoch 294/300
    1/1 [=========== ] - 0s 170ms/step - loss: 0.0205 - acc: 0.9714 - val loss: 0.1533 - val ac
    c: 0.7580
    Epoch 295/300
    c: 0.7660
    Epoch 296/300
    1/1 [============= ] - 0s 173ms/step - loss: 0.0210 - acc: 1.0000 - val_loss: 0.1410 - val_ac
    c: 0.7760
    Epoch 297/300
    c: 0.7820
    Epoch 298/300
    1/1 [============= ] - 0s 158ms/step - loss: 0.0219 - acc: 0.9714 - val_loss: 0.1345 - val_ac
    c: 0.7780
    Epoch 299/300
    c: 0.7720
    Epoch 300/300
    c: 0.7640
Out[]: <tensorflow.python.keras.callbacks.History at 0x20585267308>
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: recall f1-score precision support Case\_Based 0.80 0.69 0.74 114 Genetic\_Algorithms 0.86 0.87 0.86 156 Neural\_Networks 0.79 0.70 0.74 290 Probabilistic\_Methods 0.76 0.71 0.73 172 Reinforcement\_Learning 0.63 0.80 0.70 85 Rule\_Learning 0.52 0.88 0.66 60 Theory 0.57 0.54 0.56 123 0.73 1000 accuracy 0.74 0.71 1000 0.70 macro avg weighted avg 0.74 0.73 0.73 1000

### 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 1/300
1/1 [=============] - 0s 244ms/step - loss: 0.2190 - acc: 0.1143 - val_loss: 0.4341 - val_ac
c: 0.2740
Epoch 2/300
n_batch_end) is slow compared to the batch update (0.156036). Check your callbacks.
c: 0.3960
Epoch 3/300
c: 0.4840
Epoch 4/300
1/1 [============= ] - 0s 166ms/step - loss: 0.1108 - acc: 0.6786 - val_loss: 0.3318 - val_ac
c: 0.5440
Epoch 5/300
1/1 [============= ] - 0s 167ms/step - loss: 0.0873 - acc: 0.7571 - val_loss: 0.3002 - val_ac
c: 0.5800
Epoch 6/300
1/1 [=========== ] - 0s 163ms/step - loss: 0.0665 - acc: 0.8286 - val loss: 0.2746 - val ac
c: 0.5720
Epoch 7/300
1/1 [=============] - 0s 162ms/step - loss: 0.0528 - acc: 0.9000 - val_loss: 0.2741 - val_ac
c: 0.5540
Fnoch 8/300
c: 0.5360
Epoch 9/300
c: 0.5520
Epoch 10/300
c: 0.5540
Epoch 11/300
c: 0.5580
Epoch 12/300
c: 0.5460
Epoch 13/300
c: 0.5560
Epoch 14/300
1/1 [============= ] - 0s 236ms/step - loss: 0.0388 - acc: 1.0000 - val_loss: 0.3613 - val_ac
c: 0.5520
Epoch 15/300
1/1 [============= ] - 0s 255ms/step - loss: 0.0366 - acc: 0.9929 - val_loss: 0.3517 - val_ac
c: 0.5840
Epoch 16/300
c: 0.5860
Epoch 17/300
1/1 [=========== ] - 0s 301ms/step - loss: 0.0318 - acc: 0.9857 - val loss: 0.3657 - val ac
c: 0.5740
Epoch 18/300
c: 0.5320
Epoch 19/300
1/1 [=========== ] - 0s 252ms/step - loss: 0.0259 - acc: 1.0000 - val loss: 0.4228 - val ac
c: 0.5200
Epoch 20/300
c: 0.5060
Epoch 21/300
1/1 [============ ] - 0s 373ms/step - loss: 0.0244 - acc: 0.9786 - val_loss: 0.4186 - val_ac
c: 0.5080
Epoch 22/300
c: 0.5340
Epoch 23/300
1/1 [============] - 0s 242ms/step - loss: 0.0220 - acc: 0.9714 - val_loss: 0.3840 - val_ac
c: 0.5500
```

```
Epoch 24/300
1/1 [============ ] - 0s 344ms/step - loss: 0.0216 - acc: 0.9786 - val_loss: 0.3680 - val_ac
c: 0.5540
Epoch 25/300
1/1 [============] - 0s 229ms/step - loss: 0.0239 - acc: 0.9571 - val_loss: 0.3533 - val_ac
c: 0.5380
Epoch 26/300
1/1 [============ ] - 0s 241ms/step - loss: 0.0189 - acc: 0.9929 - val_loss: 0.3533 - val_ac
c: 0.5280
Epoch 27/300
1/1 [============= ] - 0s 310ms/step - loss: 0.0214 - acc: 0.9714 - val_loss: 0.3637 - val_ac
c: 0.5020
Epoch 28/300
1/1 [=========== ] - 0s 262ms/step - loss: 0.0201 - acc: 0.9857 - val loss: 0.3750 - val ac
c: 0.4960
Epoch 29/300
c: 0.4980
Epoch 30/300
1/1 [============= ] - 0s 253ms/step - loss: 0.0211 - acc: 0.9786 - val_loss: 0.3725 - val_ac
c: 0.5060
Epoch 31/300
c: 0.5120
Epoch 32/300
c: 0.5180
Epoch 33/300
c: 0.5260
Epoch 34/300
1/1 [============= ] - 0s 256ms/step - loss: 0.0222 - acc: 0.9857 - val_loss: 0.3493 - val_ac
c: 0.5260
Epoch 35/300
c: 0.5300
Epoch 36/300
c: 0.5280
Epoch 37/300
c: 0.5560
Epoch 38/300
c: 0.5560
Epoch 39/300
1/1 [=========== ] - 0s 238ms/step - loss: 0.0219 - acc: 0.9857 - val loss: 0.3266 - val ac
c: 0.5500
Epoch 40/300
c: 0.5600
Epoch 41/300
1/1 [============ ] - 0s 264ms/step - loss: 0.0221 - acc: 0.9857 - val_loss: 0.3279 - val_ac
c: 0.5540
Epoch 42/300
c: 0.5460
Epoch 43/300
c: 0.5340
Epoch 44/300
c: 0.5300
Epoch 45/300
c: 0.5200
Epoch 46/300
c: 0.5180
Epoch 47/300
1/1 [================] - 0s 219ms/step - loss: 0.0211 - acc: 0.9857 - val_loss: 0.3594 - val_ac
```

```
c: 0.5140
Epoch 48/300
c: 0.5220
Epoch 49/300
1/1 [==========] - 0s 187ms/step - loss: 0.0189 - acc: 0.9929 - val_loss: 0.3621 - val_ac
c: 0.5200
Epoch 50/300
1/1 [============= ] - 0s 172ms/step - loss: 0.0194 - acc: 0.9929 - val_loss: 0.3711 - val_ac
c: 0.5160
Epoch 51/300
c: 0.4980
Epoch 52/300
1/1 [============ ] - 0s 161ms/step - loss: 0.0174 - acc: 0.9929 - val_loss: 0.3757 - val_ac
c: 0.4960
Epoch 53/300
c: 0.4940
Epoch 54/300
c: 0.4900
Epoch 55/300
1/1 [=========== ] - 0s 152ms/step - loss: 0.0180 - acc: 0.9857 - val loss: 0.3665 - val ac
c: 0.4940
Epoch 56/300
c: 0.5000
Epoch 57/300
c: 0.5240
Epoch 58/300
1/1 [=============== ] - 0s 190ms/step - loss: 0.0169 - acc: 1.0000 - val_loss: 0.3208 - val_ac
c: 0.5380
Epoch 59/300
c: 0.5460
Epoch 60/300
c: 0.5640
Epoch 61/300
c: 0.5660
Epoch 62/300
c: 0.5660
Epoch 63/300
1/1 [============= ] - 0s 174ms/step - loss: 0.0170 - acc: 0.9929 - val_loss: 0.3328 - val_ac
c: 0.5680
Epoch 64/300
c: 0.5680
Epoch 65/300
c: 0.5560
Epoch 66/300
c: 0.5360
Epoch 67/300
c: 0.5060
Epoch 68/300
c: 0.4860
Epoch 69/300
c: 0.4640
Epoch 70/300
1/1 [============= ] - 0s 178ms/step - loss: 0.0239 - acc: 0.9643 - val_loss: 0.4616 - val_ac
c: 0.4520
Epoch 71/300
```

```
c: 0.4600
Epoch 72/300
1/1 [============ ] - 0s 201ms/step - loss: 0.0196 - acc: 0.9857 - val_loss: 0.4800 - val_ac
c: 0.4520
Epoch 73/300
c: 0.4560
Epoch 74/300
c: 0.4580
Epoch 75/300
1/1 [============= ] - 0s 210ms/step - loss: 0.0216 - acc: 0.9857 - val_loss: 0.4468 - val_ac
c: 0.4800
Epoch 76/300
1/1 [============= ] - 0s 161ms/step - loss: 0.0271 - acc: 0.9500 - val_loss: 0.4306 - val_ac
c: 0.4760
Epoch 77/300
1/1 [=========== ] - 0s 161ms/step - loss: 0.0224 - acc: 0.9929 - val loss: 0.4241 - val ac
c: 0.5020
Epoch 78/300
c: 0.4800
Fnoch 79/300
c: 0.4720
Epoch 80/300
1/1 [=============== ] - 0s 210ms/step - loss: 0.0248 - acc: 0.9714 - val_loss: 0.4232 - val_ac
c: 0.4700
Epoch 81/300
c: 0.5140
Epoch 82/300
c: 0.5280
Epoch 83/300
1/1 [=========== ] - 0s 193ms/step - loss: 0.0287 - acc: 0.9714 - val loss: 0.3770 - val ac
c: 0.5220
Epoch 84/300
c: 0.5040
Epoch 85/300
1/1 [============= ] - 0s 209ms/step - loss: 0.0278 - acc: 0.9714 - val_loss: 0.4135 - val_ac
c: 0.5000
Epoch 86/300
c: 0.5060
Epoch 87/300
c: 0.5040
Epoch 88/300
1/1 [=========== ] - 0s 171ms/step - loss: 0.0316 - acc: 0.9643 - val loss: 0.4021 - val ac
c: 0.5180
Epoch 89/300
c: 0.5180
Epoch 90/300
1/1 [=========== ] - 0s 214ms/step - loss: 0.0274 - acc: 1.0000 - val loss: 0.4091 - val ac
c: 0.5040
Epoch 91/300
c: 0.5060
Epoch 92/300
1/1 [==========] - 0s 219ms/step - loss: 0.0299 - acc: 0.9643 - val_loss: 0.4170 - val_ac
c: 0.5180
Epoch 93/300
1/1 [============ ] - 0s 180ms/step - loss: 0.0287 - acc: 0.9929 - val_loss: 0.4149 - val_ac
c: 0.5240
Epoch 94/300
1/1 [===========] - 0s 206ms/step - loss: 0.0336 - acc: 0.9786 - val_loss: 0.3969 - val_ac
c: 0.5360
```

```
Epoch 95/300
1/1 [============= ] - 0s 210ms/step - loss: 0.0307 - acc: 0.9714 - val_loss: 0.3773 - val_ac
c: 0.5320
Epoch 96/300
c: 0.5340
Epoch 97/300
1/1 [============= ] - 0s 185ms/step - loss: 0.0292 - acc: 0.9857 - val_loss: 0.3790 - val_ac
c: 0.5360
Epoch 98/300
1/1 [============ ] - 0s 179ms/step - loss: 0.0287 - acc: 0.9929 - val_loss: 0.3855 - val_ac
c: 0.5380
Epoch 99/300
1/1 [=========== ] - 0s 181ms/step - loss: 0.0310 - acc: 0.9786 - val loss: 0.3907 - val ac
c: 0.5280
Epoch 100/300
c: 0.5220
Epoch 101/300
1/1 [===========] - 0s 198ms/step - loss: 0.0345 - acc: 0.9714 - val_loss: 0.4063 - val_ac
c: 0.5220
Epoch 102/300
c: 0.5160
Epoch 103/300
c: 0.5200
Epoch 104/300
c: 0.5160
Epoch 105/300
1/1 [============ ] - 0s 171ms/step - loss: 0.0340 - acc: 0.9643 - val_loss: 0.4099 - val_ac
c: 0.5060
Epoch 106/300
1/1 [=========== ] - 0s 180ms/step - loss: 0.0304 - acc: 0.9714 - val loss: 0.4108 - val ac
c: 0.5060
Epoch 107/300
c: 0.5060
Epoch 108/300
c: 0.5340
Epoch 109/300
c: 0.5520
Epoch 110/300
1/1 [=========== ] - 0s 179ms/step - loss: 0.0319 - acc: 0.9714 - val loss: 0.3799 - val ac
c: 0.5620
Epoch 111/300
c: 0.5620
Epoch 112/300
1/1 [============= ] - 0s 161ms/step - loss: 0.0319 - acc: 0.9714 - val_loss: 0.3892 - val_ac
c: 0.5440
Epoch 113/300
c: 0.5440
Epoch 114/300
c: 0.5420
Epoch 115/300
c: 0.5300
Epoch 116/300
c: 0.5180
Epoch 117/300
c: 0.5080
Epoch 118/300
1/1 [==================] - 0s 173ms/step - loss: 0.0321 - acc: 0.9929 - val_loss: 0.4104 - val_ac
```

```
c: 0.5020
Epoch 119/300
c: 0.5080
Epoch 120/300
1/1 [============= ] - 0s 205ms/step - loss: 0.0357 - acc: 0.9714 - val_loss: 0.3894 - val_ac
c: 0.5220
Epoch 121/300
1/1 [============= ] - 0s 188ms/step - loss: 0.0332 - acc: 0.9714 - val_loss: 0.3838 - val_ac
c: 0.5180
Epoch 122/300
c: 0.5160
Epoch 123/300
1/1 [============= ] - 0s 157ms/step - loss: 0.0373 - acc: 0.9500 - val_loss: 0.3771 - val_ac
c: 0.5240
Epoch 124/300
c: 0.5340
Epoch 125/300
c: 0.5280
Epoch 126/300
c: 0.5200
Epoch 127/300
c: 0.5220
Epoch 128/300
c: 0.5040
Epoch 129/300
c: 0.4940
Epoch 130/300
c: 0.4940
Epoch 131/300
c: 0.4900
Epoch 132/300
c: 0.4980
Epoch 133/300
c: 0.5000
Epoch 134/300
1/1 [============= ] - 0s 163ms/step - loss: 0.0328 - acc: 0.9857 - val_loss: 0.4117 - val_ac
c: 0.4940
Epoch 135/300
c: 0.5020
Epoch 136/300
c: 0.5040
Epoch 137/300
1/1 [============ ] - 0s 181ms/step - loss: 0.0355 - acc: 0.9857 - val_loss: 0.4059 - val_ac
c: 0.4960
Epoch 138/300
c: 0.4980
Epoch 139/300
c: 0.4960
Epoch 140/300
c: 0.4900
Epoch 141/300
1/1 [============ ] - 0s 157ms/step - loss: 0.0346 - acc: 0.9857 - val_loss: 0.4376 - val_ac
c: 0.4940
Epoch 142/300
```

```
c: 0.4860
     Epoch 143/300
     c: 0.4860
     Epoch 144/300
     1/1 [============== ] - 0s 378ms/step - loss: 0.0352 - acc: 0.9786 - val_loss: 0.4419 - val_ac
     c: 0.4880
     Epoch 145/300
     c: 0.4840
     Epoch 146/300
     1/1 [============= ] - 0s 300ms/step - loss: 0.0361 - acc: 0.9786 - val_loss: 0.4413 - val_ac
     c: 0.4860
     Epoch 147/300
     c: 0.4840
     Epoch 148/300
     1/1 [=========== ] - 0s 185ms/step - loss: 0.0317 - acc: 0.9857 - val loss: 0.4329 - val ac
     c: 0.4720
     Epoch 149/300
     1/1 [==============] - 0s 168ms/step - loss: 0.0323 - acc: 0.9857 - val_loss: 0.4172 - val_ac
     c: 0.4840
     Fnoch 150/300
     1/1 [============ ] - 0s 239ms/step - loss: 0.0334 - acc: 0.9857 - val_loss: 0.4040 - val_ac
     c: 0.4980
     Epoch 151/300
     c: 0.5020
     Epoch 152/300
     c: 0.4920
     Epoch 153/300
     c: 0.4980
     Epoch 154/300
     1/1 [============ ] - 0s 145ms/step - loss: 0.0334 - acc: 0.9714 - val loss: 0.3843 - val ac
     c: 0.4920
     Epoch 155/300
     c: 0.4940
     Epoch 156/300
     1/1 [============= ] - 0s 145ms/step - loss: 0.0386 - acc: 0.9571 - val_loss: 0.3795 - val_ac
     c: 0.4980
     Epoch 157/300
     c: 0.4960
Out[]: <tensorflow.python.keras.callbacks.History at 0x205928e1508>
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
                      0.50
                            0.58
                                  0.54
                                         114
            Case Based
                      0.77
                            0.67
                                  0.72
       Genetic_Algorithms
                                         156
         Neural_Networks
                      0.73
                            0.49
                                  0.58
                                         290
                            0.55
     Probabilistic_Methods
                      0.65
                                  0.60
                                         172
     Reinforcement_Learning
                      0.43
                            0.51
                                  0.47
                                         85
                            0.77
                                  0.41
          Rule_Learning
                      0.28
                                          60
              Theory
                      0.44
                            0.46
                                  0.45
                                         123
                                  0.55
             accuracy
                                         1000
                      0.54
                            0.58
                                  0.54
                                         1000
             macro avg
           weighted avg
                      0.61
                            0.55
                                  0.57
                                         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 [ ]: x_tsne = TSNE(n_components=2).fit_transform(activations[3])
        plot_tSNE(labels_encoded,x_tsne)
                                                                                                                 1
                                                                                                                 2
                                                                                                                 3
                                                                                                                 4
          40
                                                                                                                 5
          20
           0
         -20
         -40
                                                                                   20
                    -60
                                   -40
                                                   -20
                                                                    0
                                                                                                  40
                                                                                                                  60
```

 $file: ///C:/Users/USER\ PC/Documents/HCMUT/221/Mathematical/Graph\_Convolutional\_Networks\_Node\_Classification/output/300.html$ 

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