```
In [1]: | import dgl
        import dgl.function as fn
        import torch as th
        import torch.nn as nn
        import torch.nn.functional as F
        from dgl import DGLGraph
        gcn msg = fn.copy src(src='h', out='m')
        gcn_reduce = fn.sum(msg='m', out='h')
In [2]: class NodeApplyModule(nn.Module):
            def __init__(self, in_feats, out_feats, activation):
                 super(NodeApplyModule, self).__init__()
                 self.linear = nn.Linear(in_feats, out_feats)
                 self.activation = activation
            def forward(self, node):
                h = self.linear(node.data['h'])
                if self.activation is not None:
                    h = self.activation(h)
                return {'h' : h}
In [3]: class GCN(nn.Module):
            def init (self, in feats, out feats, activation):
                 super(GCN, self).__init__()
                 self.apply mod = NodeApplyModule(in feats, out feats, activation)
            def forward(self, g, feature):
                g.ndata['h'] = feature
                 g.update_all(gcn_msg, gcn_reduce)
                 g.apply nodes(func=self.apply mod)
                 return g.ndata.pop('h')
```

```
In [4]: class Net(nn.Module):
            def __init__(self):
                 super(Net, self).__init__()
                 self.gcn1 = GCN(1433, 16, F.relu)
                 self.gcn2 = GCN(16, 7, None)
            def forward(self, g, features):
                x = self.gcn1(g, features)
                 x = self.gcn2(g, x)
                 return x
        net = Net()
        print(net)
        Net(
           (gcn1): GCN(
            (apply_mod): NodeApplyModule(
               (linear): Linear(in_features=1433, out_features=16, bias=True)
            )
           (gcn2): GCN(
            (apply_mod): NodeApplyModule(
               (linear): Linear(in_features=16, out_features=7, bias=True)
            )
          )
        )
        from dgl.data import citation_graph as citegrh
In [5]:
        import networkx as nx
        def load cora data():
            data = citegrh.load cora()
            features = th.FloatTensor(data.features)
            labels = th.LongTensor(data.labels)
            train mask = th.BoolTensor(data.train mask)
            test mask = th.BoolTensor(data.test mask)
            g = data.graph
            # add self loop
            g.remove_edges_from(nx.selfloop_edges(g))
            g = DGLGraph(g)
            g.add_edges(g.nodes(), g.nodes())
            return g, features, labels, train mask, test mask
In [6]: | def evaluate(model, g, features, labels, mask):
            model.eval()
            with th.no_grad():
                 logits = model(g, features)
                 logits = logits[mask]
                 labels = labels[mask]
                 _, indices = th.max(logits, dim=1)
                 correct = th.sum(indices == labels)
                 return correct.item() * 1.0 / len(labels)
```

```
In [7]: import time
        import numpy as np
        g, features, labels, train_mask, test_mask = load_cora_data()
        optimizer = th.optim.Adam(net.parameters(), lr=1e-3)
        dur = []
        for epoch in range(50):
            if epoch >=3:
                t0 = time.time()
            net.train()
            logits = net(g, features)
            logp = F.log_softmax(logits, 1)
            loss = F.nll_loss(logp[train_mask], labels[train_mask])
            optimizer.zero_grad()
            loss.backward()
            optimizer.step()
            if epoch >=3:
                 dur.append(time.time() - t0)
            acc = evaluate(net, g, features, labels, test_mask)
            print("Epoch {:05d} | Loss {:.4f} | Test Acc {:.4f} | Time(s) {:.4f}".form
        at(
                     epoch, loss.item(), acc, np.mean(dur)))
```

C:\Users\qiqishaoshuai\Miniconda3\lib\site-packages\numpy\core\fromnumeric.p
y:3335: RuntimeWarning: Mean of empty slice.
 out=out, **kwargs)

C:\Users\qiqishaoshuai\Miniconda3\lib\site-packages\numpy\core_methods.py:16

1: RuntimeWarning: invalid value encountered in double scalars

ret = ret.dtype.type(ret / rcount)

```
Epoch 00000 | Loss 1.9850 |
                           Test Acc 0.1510
                                              Time(s) nan
Epoch 00001
             Loss 1.9659
                            Test Acc 0.1550
                                              Time(s) nan
Epoch 00002
              Loss 1.9495
                            Test Acc 0.1640
                                              Time(s) nan
Epoch 00003
             Loss 1.9341
                            Test Acc 0.2250
                                              Time(s) 0.0628
Epoch 00004
              Loss 1.9185
                            Test Acc 0.2520
                                              Time(s) 0.0643
Epoch 00005
              Loss 1.9020
                            Test Acc 0.2700
                                              Time(s) 0.0645
Epoch 00006
              Loss 1.8846
                            Test Acc 0.2910
                                              Time(s) 0.0666
Epoch 00007
              Loss 1.8664
                            Test Acc 0.3070
                                              Time(s) 0.0686
Epoch 00008
              Loss 1.8475
                            Test Acc 0.3180
                                              Time(s) 0.0705
Epoch 00009
              Loss 1.8284
                            Test Acc 0.3260
                                              Time(s) 0.0704
Epoch 00010
              Loss 1.8090
                            Test Acc 0.3350
                                              Time(s) 0.0702
Epoch 00011
             Loss 1.7891
                            Test Acc 0.3500
                                              Time(s) 0.0708
             Loss 1.7691
Epoch 00012
                            Test Acc 0.3610
                                              Time(s) 0.0714
Epoch 00013
              Loss 1.7488
                            Test Acc 0.3590
                                              Time(s) 0.0725
Epoch 00014
              Loss 1.7287
                            Test Acc 0.3650
                                              Time(s) 0.0722
Epoch 00015
              Loss 1.7088
                            Test Acc 0.3760
                                              Time(s) 0.0730
Epoch 00016
             Loss 1.6895
                            Test Acc 0.3790
                                              Time(s) 0.0728
Epoch 00017
             Loss 1.6702
                            Test Acc 0.3900
                                              Time(s) 0.0729
Epoch 00018
              Loss 1.6506
                            Test Acc 0.3950
                                              Time(s) 0.0725
Epoch 00019
              Loss 1.6309
                            Test Acc 0.4000
                                              Time(s) 0.0740
Epoch 00020
              Loss 1.6115
                            Test Acc 0.4080
                                              Time(s) 0.0737
                            Test Acc 0.4130
Epoch 00021
              Loss 1.5924
                                              Time(s) 0.0740
Epoch 00022
             Loss 1.5741
                            Test Acc 0.4220
                                              Time(s) 0.0736
Epoch 00023
              Loss 1.5559
                            Test Acc 0.4240
                                              Time(s) 0.0734
Epoch 00024
              Loss 1.5378
                            Test Acc 0.4310
                                              Time(s) 0.0731
Epoch 00025
              Loss 1.5199
                            Test Acc 0.4370
                                              Time(s) 0.0728
Epoch 00026
              Loss 1.5021
                            Test Acc 0.4470
                                              Time(s) 0.0725
Epoch 00027
              Loss 1.4844
                            Test Acc 0.4560
                                              Time(s) 0.0725
Epoch 00028
              Loss 1.4670
                            Test Acc 0.4750
                                              Time(s) 0.0724
Epoch 00029
              Loss 1.4496
                            Test Acc 0.4930
                                              Time(s) 0.0723
Epoch 00030
              Loss 1.4324
                            Test Acc 0.5040
                                              Time(s) 0.0722
              Loss 1.4153
Epoch 00031
                            Test Acc 0.5150
                                              Time(s) 0.0720
Epoch 00032
              Loss 1.3982
                            Test Acc 0.5310
                                              Time(s) 0.0718
                            Test Acc 0.5470
Epoch 00033
              Loss 1.3808
                                              Time(s) 0.0718
Epoch 00034
              Loss 1.3635
                            Test Acc 0.5590
                                              Time(s) 0.0718
Epoch 00035
                            Test Acc 0.5610
                                              Time(s) 0.0716
              Loss 1.3465
Epoch 00036
              Loss 1.3297
                            Test Acc 0.5710
                                              Time(s) 0.0716
Epoch 00037
              Loss 1.3132
                            Test Acc 0.5790
                                              Time(s) 0.0720
Epoch 00038
              Loss 1.2968
                            Test Acc 0.5890
                                              Time(s) 0.0721
Epoch 00039
              Loss 1.2806
                            Test Acc 0.5960
                                              Time(s) 0.0721
Epoch 00040
                            Test Acc 0.6080
                                              Time(s) 0.0725
              Loss 1.2644
              Loss 1.2483
Epoch 00041
                            Test Acc 0.6190
                                              Time(s) 0.0725
Epoch 00042
              Loss 1.2324
                            Test Acc 0.6320
                                              Time(s) 0.0726
Epoch 00043
              Loss 1.2167
                            Test Acc 0.6450
                                              Time(s) 0.0726
Epoch 00044
              Loss 1.2011
                            Test Acc 0.6610
                                              Time(s) 0.0727
Epoch 00045
                            Test Acc 0.6730
              Loss 1.1858
                                              Time(s) 0.0726
Epoch 00046
             Loss 1.1706
                            Test Acc 0.6770
                                              Time(s) 0.0725
Epoch 00047
              Loss 1.1556
                            Test Acc 0.6830
                                              Time(s) 0.0725
Epoch 00048
              Loss 1.1406
                            Test Acc 0.6870
                                              Time(s) 0.0724
Epoch 00049 | Loss 1.1258 | Test Acc 0.6940 | Time(s) 0.0723
```

In []: