

Department of Computer Science and Technology

Machine Learning

Homework 2

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2020280401

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- 2 (GCN)
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3 CogDL

3.1 Testing CogDL Models

In this assignment, we are asked to run two models on two different datasets by running the provided scripts. The first script is as follows: python scripts/train.py –task unsupervised_node_classification – dataset wikipedia –model deepwalk; by running this script, we would be training and testing the DeepWalk model on the Wikipedia dataset for the task of node classification. The results of this process is provided in the figure below (Figure 1).

```
python scripts/train.py --task unsupervised node classification --dataset wikipedia --model deepwalk
(cogdl) → cogdl git:(
ailed to import Deep Graph Library (DGL)
Namespace(cpu=False, dataset=['wikipedia'], device_id=[0], enhance=None, hidden_size=128, iteration=10, lr=0.01, max_epoch=500, model=
 'deepwalk'], num_shuffle=5, patience=100, save_dir='.', seed=[1], task='unsupervised_node_classification', walk_length=80, walk_num=4
0, weight_decay=0.0005, window_size=5, worker=10)
Downloading http://snap.stanford.edu/node2vec/POS.mat
Processing...
Done!
node number: 4777
generating random walks..
100%
                                               | 40/40 [00:59<00:00, 1.48s/it]
training word2vec...
 Variant
                             Micro-F1 0.1
                                            | Micro-F1 0.3
                                                             | Micro-F1 0.5
                                                                               | Micro-F1 0.7
                                                                                                | Micro-F1 0.9
 ('wikipedia', 'deepwalk') | 0.4253±0.0000
                                            0.4720±0.0000
                                                             0.4951±0.0000
                                                                               | 0.5056±0.0000
                                                                                                 0.5214±0.0000
```

Figure 1: Result of running the first script

Accordingly, the second script (python scripts/train.py -task node_classification -dataset citeseer - model gcn) analyzes the same task but for training and testing the GCN model on the citeseer dataset. The obtained results are demonstrated in the below figure (Figure 2).

```
ailed to import Deep Graph Library (DGL)
lamespace(cpu=False, dataset=['citéseer'], device_id=[0], dropout=0.5, enhance=None, hidden_size=64, lr=0.01, max_epoch=500, missing_:
ate=-1, model=['gcn'], num_classes=None, num_features=None, patience=100, save_dir='.', seed=[1], task='node_classification', weight_d
cav=0.0005)
 ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.x
 ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.tx
 ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.allx
 ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.y
 ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.ty
ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.ally
ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.graph
ownloading https://github.com/kimiyoung/planetoid/raw/master/data/ind.citeseer.test.index
rocessing..
one!
poch: 440, Train: 1.0000, Val: 0.7000: 88%|
                                                                                                             | 440/500 [00:05<00:00, 84.81it/s]
Valid accurracy = 0.73
est accuracy = 0.72
 Variant
  ('citeseer', 'gcn') | 0.7200<u>±</u>0.0000
```

Figure 2: Result of running the second script

3.2 API Design

Similar to what is demonstrated in the two scripts of the previous section, the main entry point of the model training is train.py located in the scripts folder.

3.3 Implementation

In this assignment, I implemented the Hierarchical Graph Pooling with Structure Learning model [1]. The pull request id for this implementation is .

3.4 Suggestions

There are a number of suggestions that I believe would improve the CogDL experience, both as a user and as a contributor:

1. hi

3.5 Contributions

References

[1] Zhen Zhang et al. "Hierarchical graph pooling with structure learning". In:arXiv(2019) arXiv:1911.05954