



Data Science HW3

Department of Computer Science
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Due Date: 2023/05/12 (FRI) 23:59

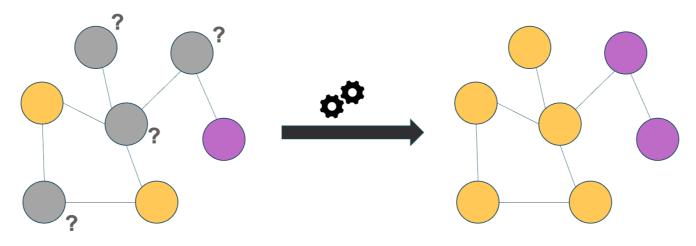
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HW3 Node classification



- Build a node classification model
- Given an unknown graph dataset
 - Train your model using the training nodes
 - Predict the labels of the testing nodes.





HW3 is hosted on Kaggle

- HW 3 Kaggle link
 - https://www.kaggle.com/t/d1097b49b16014572e81c3bafd7e3bb9
- Deadline: 2023/05/12 (FRI) 23:59
- Fill your Kaggle name in the google form
- We will use the result on Kaggle to score this homework
- You also need to hand in your code on eeclass.
 - The instructions for submitting files will be described in later pages.
 - Do not directly copy any source on the internet.





Dataset description

- A modified graph data
 - Each node has a predefined feature.
- Dataset file name description





Output format

- For each testing instance, there is a unique id
- Output your prediction to csv file with the following format and submit to Kaggle
- Remember to write the first line 'Id,Predict'
- Output csv file example:

ld	Predict
0	1
1	0
2	1
3	0
4	1
5	1





Evaluation

Using Accuracy

- There are two leaderboards on Kaggle
 - Public
 - Can be seen during competition
 - Private
 - Can be seen after competition



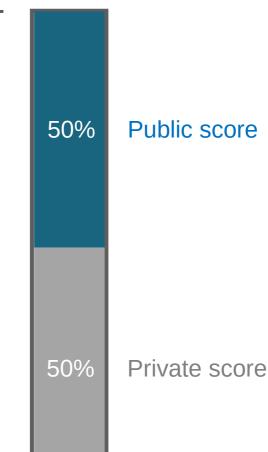


Public and Private leaderboard

- Public (Can be seen during competition)
 - 50% testing data
 - For reference

Testing data

- **Private** (Can be seen after competition)
 - the other 50%
 - Use this result for final scoring







Scoring

	Public	Private*	Your HW3 Score
Baseline 0	0.750	0.740	0
Baseline 60	0.814	0.786	60
Baseline 70	0.834	0.806	70
Baseline 80	0.838	0.822	80
Baseline 88	0.850	0.840	Top 20%: 100 20%~80%: 92 Others: 88

- You will get **0**, if your **private score** is between *baseline 0* and *baseline 60*
- You will get **60**, if your **private score** is between *baseline 60* and *baseline 70*
- You will get 70, if your private score is between baseline 70 and baseline 80
- And so on

^{*} The private score of each baseline may be adjusted based on the results of classmates.





Scoring

- Baseline scores
 - There are benchmarks on the leaderboard for reference

Public	Private						
--------	---------	--	--	--	--	--	--

This leaderboard is calculated with approximately 50% of the test data. The final results will be based on the other 50%, so the final standings may be different.

#	Team	Members	Score	Entries	Last
# T	baseline_88		0.85000		
# T	baseline_80		0.83800		
# T	baseline_70		0.83400		
1	baseline_60		0.81400		
1	baseline_0		0.75000		





Other rules

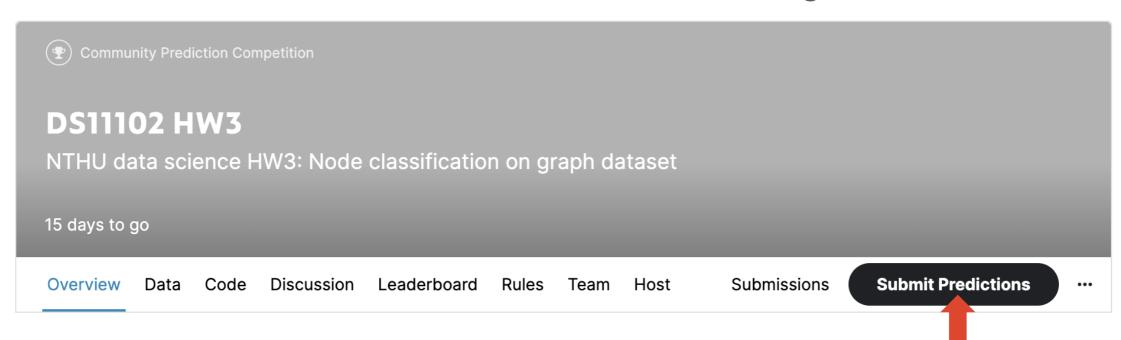
- You can submit 20 times per day
- You can choose 5 predictions for final scoring
 - Kaggle will use the best one to be your result
- We will publish private leaderboard one time on eeclass 3 days before the deadline of this homework.
 - This won't affect your final score, just for you to check your status.





How to submit

• Click 'Submit Predictions' button on the navigation bar







How to submit

Submit to Competition

File Upload Notebook



DS11102 HW3

You have 19 submissions remaining today. This resets in an hour.



Drag and drop file to upload

(e.g., .csv, .zip, .gz, .7z)

Upload your answer csv file here

or

Browse Files

Your submission should be a CSV file with 1000 rows and a header. You can upload a zip/gz/7z archive.

DESCRIPTION

Enter a description

You can write some description about the answer csv file

0 / 500

> kaggle competitions submit -c ds11102-hw3 -f submission.csv -m ...





Cancel





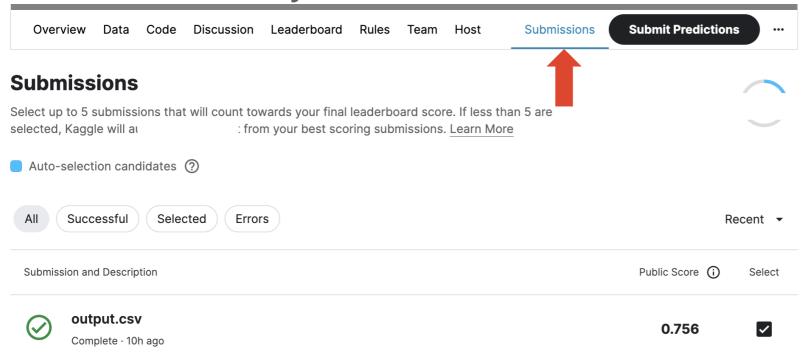
Click to submit





Choose predictions for final scoring

You can see all your submissions in 'Submissions'



Remember to choose 5 predictions before the deadline





Sample code description

Baseline 0 method





Sample code



- Load data based on provided dataset name.
- No need to modify this file.
- Define the details of model.

- Train and evaluate the model.
- Export the submission csv file.





data_loader.py





train.py

```
78
        # Load data
        features, graph, num_classes, \
79
         train_labels, val_labels, test_labels, \
80
81
         train_mask, val_mask, test_mask = load_data()
82
83
        # Initialize the model (Baseline Model: GCN)
84
         """TODO: build your own model in model.py and replace GCN() with your model"""
85
         in_size = features.shape[1]
86
        out_size = num_classes
        model = GCN(in_size, 16, out_size).to(device)
87
```





model.py

```
class GCN(nn.Module):
        Baseline Model:
        - A simple two-layer GCN model, similar to https://github.com/tkipf/pygcn
10

    Implement with DGL package

11
12
        def __init__(self, in_size, hid_size, out_size):
13
             super().__init__()
             self.layers = nn.ModuleList()
14
15
            # two-layer GCN
             self.layers.append(
17
                 GraphConv(in_size, hid_size, activation=F.relu)
             self.layers.append(GraphConv(hid_size, out_size))
20
             self.dropout = nn.Dropout(0.5)
21
22
        def forward(self, g, features):
23
             h = features
             for i, layer in enumerate(self.layers):
                if i != 0:
25
                    h = self.dropout(h)
                h = layer(g, h)
27
             return h
```





model.py





How to run

- Run:
- python3 train.py \
 --epochs {num of epochs} \
 --es_iters {num of iters to trigger early stopping}\
 --use-gpu

Example:

```
python3 train.py --es_iters 30 --epochs 300 --use_gpu
```





Once you successfully run the sample code...

```
Training...
Early stopping monitoring on
                            Accuracy 0.1980
Epoch 00000
             Loss 1.1011
Epoch 00001
             Loss 1.0950
                            Accuracy 0.2140
Epoch 00002
             Loss 1.0918
                            Accuracy 0.5120
             Loss 1.0873
Epoch 00003
                            Accuracy 0.5920
Epoch 00004
             Loss 1.0790
                            Accuracy 0.6120
Epoch 00005 | Loss 1.0724
                            Accuracy 0.6340
Epoch 00006
             Loss 1.0644
                            Accuracy 0.6500
Epoch 00007 |
             Loss 1.0541
                            Accuracy 0.6860
                            Accuracy 0.7100
Epoch 00008
             Loss 1.0476
Epoch 00009
             Loss 1.0333
                            Accuracy 0.7240
Epoch 00010
             Loss 1.0253
                            Accuracy 0.7180
```

.

```
Epoch 00155
              Loss 0.2243
                            Accuracy 0.7900
Epoch 00156
             Loss 0.2472
                            Accuracy 0.7880
Epoch 00157 |
             Loss 0.1898
                            Accuracy 0.7900
Epoch 00158 |
             Loss 0.1982
                            Accuracy 0.7880
Epoch 00159 |
             Loss 0.2302
                            Accuracy 0.7840
                            Accuracy 0.7860
Epoch 00160 |
             Loss 0.1814
Early stopping at epoch=161
Testing...
Export predictions as csv file.
```

* Please remember to upload your output csv file to Kaggle for scoring.





HW3 Conclusion

- Submit your code to eeclass before deadline
 - 1. All your python scripts file which is able to read dataset, train your model, export a prediction csv file
 - 2. a README.md file
 - describe how to run your code
 - The predictions must be derived from a machine learning or deep learning model.
- Evaluate your model performance on Kaggle.
 - Submit your model prediction file
 - We will use the result on Kaggle to score this homework.





For Your Reference

- DGL: https://docs.dgl.ai/index.htm
- GAT: https://arxiv.org/pdf/1710.10903v3.pdf
- SSP: https://arxiv.org/pdf/2008.09624v1.pdf
- GRACE: https://arxiv.org/abs/2006.04131v2
- State-of-the-art on Node Classification : https://paperswithcode.com/task/node-classification