



Machine Learning with Graphs (MLG)

HW1: Learning to Identify High Betweenness Nodes

Deadline: 2021.03.19 (Fri.) 23:59

Submission: Code (.py/.ipynb) and Report (PDF)

HW1: Learning to Identify High BC Nodes

- Implement the paper that uses NN-based learning to identify high BC nodes
 - “*Learning to Identify High Betweenness Centrality Nodes from Scratch: A Novel Graph Neural Network Approach*”
ACM CIKM 2019 <https://arxiv.org/abs/1905.10418>
 - Any package can be used, e.g., PyTorch, Tensorflow, Keras, NetworkX, DGL, and PyTorch Geometric
 - Highly recommend **PyTorch Geometric**
 - A geometric deep learning extension library for PyTorch
 - https://github.com/rusty1s/pytorch_geometric
 - <https://www.google.com/search?q=pytorch%20geometric>
- HW1 is for you to get familiar with NN implementations and graph/network analysis packages

HW1: Learning to Identify High BC Nodes

- **2 test datasets:** ~5K synthetic graph, ~1M YouTube graph
 - Ground-truth ranking, i.e., high BC nodes, is provided
- Evaluation metrics (defined in the paper)
 - Recommendation metric: **Top-N% Accuracy** (N=1, 5, 10)
 - Ranking metric: **Kendall tau distance**
 - **Wall-clock running time** (in seconds)
- **Reference code** (official implementation in Tensorflow)
 - <https://github.com/FFrankyy/DrBC>
 - Note: You can NOT copy & paste the code.
We will check the code plagiarism!
- You can follow all of the settings mentioned in the paper
 - E.g., Hyperparameter settings in Table 2

You implementation is not required to exactly follow the paper, i.e., you can do any you feel reasonable to improve the method

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- Compared methods
 - RK <https://github.com/ecrc/BeBeCA>
 - k-BC <https://github.com/ecrc/BeBeCA>
 - KADABRA <https://github.com/natema/kadabra>
- Required reproducibility
 - Table 3, 4, 5, 6, 7, 8, and 9 in the paper
- Bonus: use DrBC to find high-closeness and high-CC nodes
- HW1 Submission via Moodle
 - Deadline: **23:59, March 19 (Friday), 2021**
 - Submit your code: **.py** or **.ipynb** (preferred)
 - Submit a report (PDF): ≥ 10 pages (you cannot include code in report)
- Content in the report
 - Implementation details (describe how you implement)
 - Reproduced experimental results, along with analysis and insights
- Learning to learn by yourself. Good Luck!

Final Remark

- **What if I cannot successfully reproduce DrBC by following the detailed settings of the paper?**
 - Need not to totally follow the paper!!
(The paper is not always correct)
 - You can modify DrBC to make the model converged and generate accurate results
 - Write the details of what you modify in the report
 - Figure out what are the potential problems
 - Write the possible reasons in the report
- **What if I totally cannot understand DrBC algorithm?**
 - You should come up with a supervised learning method (not necessary NN models) to train and predict BC values
 - Follow the experimental settings of DrBC
 - Write the details/results in the report

But you may receive a lower score,
it depends on you method