



Machine Learning with Graphs (MLG)

# HW1: Learning to Identify High Betweenness Nodes

**Deadline: 2020.03.30 (Mon.) 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://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

# HW1: Learning to Identify High BC Nodes

- 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 30 (Monday), 2020**
  - Submit your code: **.py** or **.ipynb** (preferred)
  - Submit a report (PDF):  $\geq 10$  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