Community Detection on the Reddit Hyperlink Dataset

Alexander Rundle Department of Computer Science University of Exeter, Exeter, UK

Internal Supervision:

Internal Supervision:

University of Exeter, Exeter, UK

University of Exeter, Exeter, UK

Abstract—WRITE ABSTRACT AT END OF WRITING RE-**PORT**

I. INTRODUCTION: CONTEXTUALIZATION AND MOTIVATION

A. The Reddit Hyperlink Dataset

For this mini-project I have chosen to use the Reddit Hyperlink Dataset [1], hosted on the Stanford Large Network Dataset Collection [2]. The Reddit Hyperlink Dataset was created by Kumar et al. [1] to study cases of 'intercommunity conflict', where members of one Reddit community (called a 'Subreddit') collectively mobilise to participate in or attack another community [1].

The dataset itself consists of a Tab-Separated Values (TSV) file containing 40 months of Reddit comments and posts extracted between January 2014 and April 2017, in which a hyperlink to another subreddit was present. As such, each entry in the dataset represents a directed edge between two subreddit nodes and allows one to study communities that may exist between various subreddits. The dataset also provides the Post ID, the timestamp of the post, a label indicating explicit sentiment towards the target subreddit and a vector representing the text properties of the source post [1].

II. METHODS: DESCRIPTION AND IMPLEMENTATION

Before any community detection algorithms can be applied to the Reddit Hyperlink Dataset, some pre-processing must be carried out to ensure that it works with the NetworkX package [3]. First, the original TSV file is downloaded from the Stanford Large Network Dataset Collection and loaded into a temporary Pandas DataFrame [4], [5]. The temporary DataFrame is then saved as a Comma-Separated Values (CSV) file to allow for easier manipulation; following this, the CSV file is loaded into a new DataFrame in which only the SOURCE_SUBREDDIT and TARGET_SUBREDDIT fields are extracted.

As each row in the CSV file represents any post or comment 10 labels_louvain = community.best_partition(weighted) containing a hyperlink from the source subreddit to the target subreddit, rows can repeat source and target subreddits. To convert this DataFrame into an edgelist, the occurrences of each row are counted and then placed into a WEIGHT field for each edge. For example, if a row containing the source and target subreddits 'destinythegame' and 'gaming' respectively, appears 5 times within the CSV file then the weight for the

edge from 'destinythegame' to 'gaming' would be 5. The weighted edgelist is subsequently saved as a separate CSV file, which is provided within the submission.

A. Louvain Algorithm

Introduced in 2008 by Blondel et al. [6], the Louvain algorithm finds partitions of high modularity in large networks by unfolding the network into a complete hierarchical structure [6]. The algorithm itself is divided into two phases, which are repeated iteratively; initially in phase one, each node is assigned its own community so that there are as many communities as are nodes in the network. For each node i and its neighbours j of i, the gain in modularity that might occur if i was moved from its own community to the community of jis evaluated; node i is subsequently placed into the community for which the modularity gain is a positive maximum [6]. This is repeated for every node, repeatedly until a local maxima of modularity is achieved. In the second phase of the algorithm, a new network is built whose nodes are now the communities found during phase one. The weights on edges between the new nodes are found as the sum of the weights between nodes in the corresponding two communities [7]. Once this new network is built, the first phase of the algorithm is then reapplied and iterated; in this case, a 'pass' is denoted as a combination of the two phases [6]. The passes are then iterated until no changes are made and a global maxima of modularity

To apply this algorithm to the Reddit Hyperlink Dataset [1], I use the python-louvain package [8] for Python 3.9 [9].

```
import networkx as nx
import community
weighted = make_weighted(reddit_df)
# Make Spring Layout of Weighted Network
pos = nx.spring_layout(weighted, weight='WEIGHT')
# Get list of Louvain partitions
```

Listing 1. Louvain Algorithm applied to Reddit Hyperlink Dataset

B. Girvan-Newman Algorithm

[10]

C. Fluid Communities Algorithm

[11]

III. EXPERIMENTS AND RESULTS

IV. CONCLUSIONS

A. Future Work

REFERENCES

- S. Kumar, W. L. Hamilton, J. Leskovec, and D. Jurafsky, "Community interaction and conflict on the web," in *Proceedings of the 2018 World Wide Web Conference on World Wide Web*. International World Wide Web Conferences Steering Committee, 2018, pp. 933–943.
- [2] J. Leskovec and A. Krevl, "SNAP Datasets: Stanford large network dataset collection," http://snap.stanford.edu/data, Jun. 2014.
- [3] A. A. Hagberg, D. A. Schult, and P. J. Swart, "Exploring network structure, dynamics, and function using networkx," in *Proceedings of* the 7th Python in Science Conference, G. Varoquaux, T. Vaught, and J. Millman, Eds., Pasadena, CA USA, 2008, pp. 11 – 15.
- [4] The Pandas Development Team, "pandas-dev/pandas: Pandas," Feb. 2020. [Online]. Available: https://doi.org/10.5281/zenodo.3509134
- [5] Wes McKinney, "Data Structures for Statistical Computing in Python," in Proceedings of the 9th Python in Science Conference, Stéfan van der Walt and Jarrod Millman, Eds., 2010, pp. 56 – 61.
- [6] V. D. Blondel, J.-L. Guillaume, R. Lambiotte, and E. Lefebvre, "Fast unfolding of communities in large networks," *Journal of statistical* mechanics: theory and experiment, vol. 2008, no. 10, p. P10008, 2008.
- [7] A. Arenas, J. Duch, A. Fernández, and S. Gómez, "Size reduction of complex networks preserving modularity," *New Journal of Physics*, vol. 9, no. 6, p. 176, 2007.
- [8] T. Aynaud, "python-louvain x.y: Louvain algorithm for community detection," https://github.com/taynaud/python-louvain, 2020
- [9] G. Van Rossum and F. L. Drake, *Python 3 Reference Manual*. Scotts Valley, CA: CreateSpace, 2009.
- [10] M. Girvan and M. E. Newman, "Community structure in social and biological networks," *Proceedings of the national academy of sciences*, vol. 99, no. 12, pp. 7821–7826, 2002.
- [11] F. Parés, D. G. Gasulla, A. Vilalta, J. Moreno, E. Ayguadé, J. Labarta, U. Cortés, and T. Suzumura, "Fluid communities: A competitive, scalable and diverse community detection algorithm," in Complex Networks & Their Applications VI: Proceedings of Complex Networks 2017 (The Sixth International Conference on Complex Networks and Their Applications). Springer, 2018, pp. 229–240.