

# PreCog Task 3: ATLAS

Sudhanva Joshi - 2023102022  
ECE, IIIT Hyderabad

**Abstract**—This report details the analysis of a directed ALTAS Graph on various parameters in an attempt to correlate them to real-life strategies. The pipeline to test this model has been expounded upon below.

**Note:** All corresponding files are included in their respective folders.

## I. GRAPH PROPERTIES

Establishing a crucial *Win-Condition* can be defined as follows:

**A player wins if they are able to reach the node which only has neighbouring nodes with 0 outdegree.**

### A. Degree

As mentioned prior, Node *Degrees* is key in developing and reaching a winning strategy for the game.

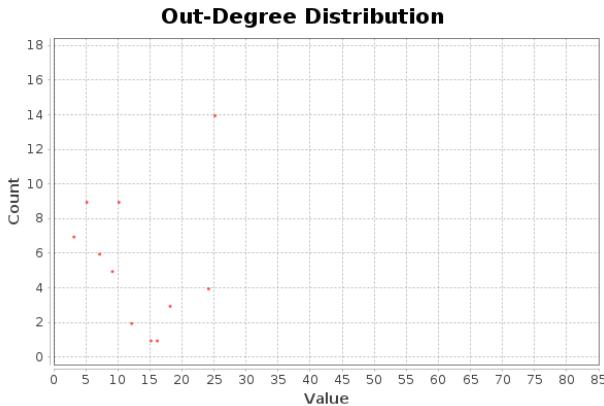


Fig. 1: Country Only

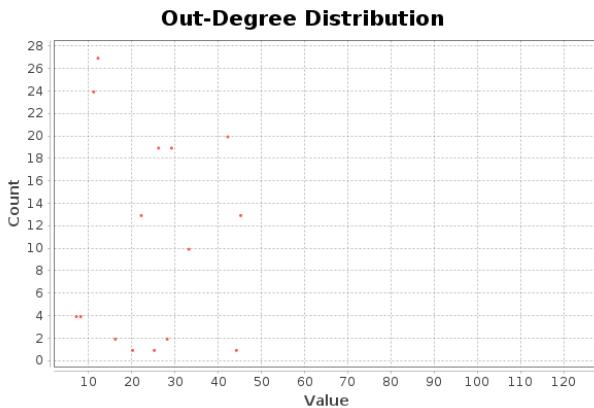


Fig. 2: City Only

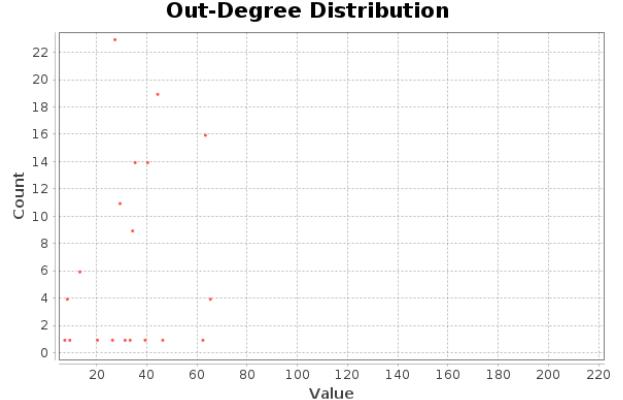


Fig. 3: Country+City

It is evident that as the sample size increases, the distribution of the outdegree of the nodes tends to follow a power-law distribution, which indicates that the game path may 'fan out' more as the game progresses, making it difficult to corner the opponent.

### B. Centrality

We can lock the opponent out of a cluster of nodes by observing the *Centrality* of nodes. H

- **Betweenness** - The number of shortest paths that pass through a node (*Bottleneck*).
- **Closeness** - The average distance of a node to all other nodes (*Options to Move yourself*).
- **Eccentricity** - The maximum distance of a node to all other nodes (*Distance from Objective-Low Outdegree neighbours*).

### C. Diameter

The obtained values for the *Diameter* of all 3 graphs are as follows:

$$d_{country} = 7, d_{city} = 6, d_{country+city} = 5$$

The diameter of the combined graph is the lowest, which indicates that the game may be more 'compact' and the player may have to make more strategic moves to win.

**Note:** Simulated using *GEPHI*.

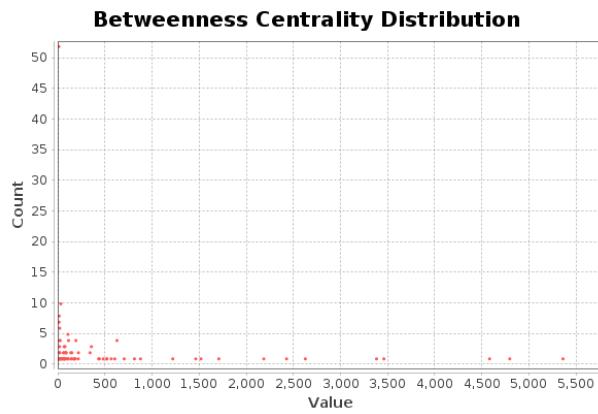


Fig. 4: Country Only

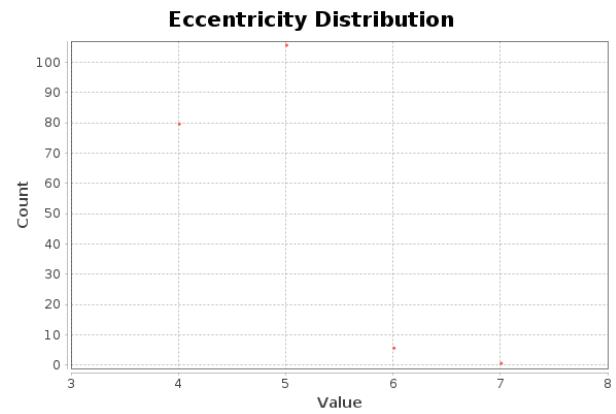


Fig. 7: Country Only

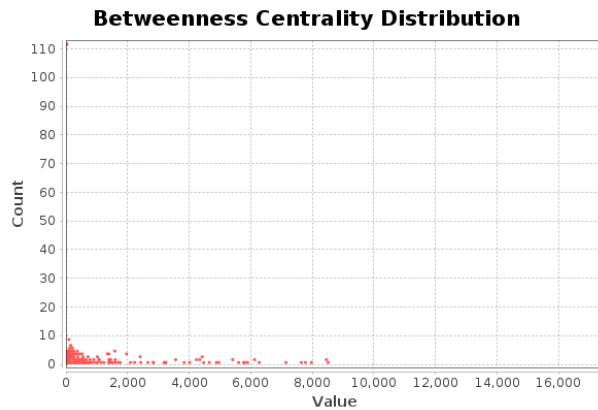


Fig. 5: City Only

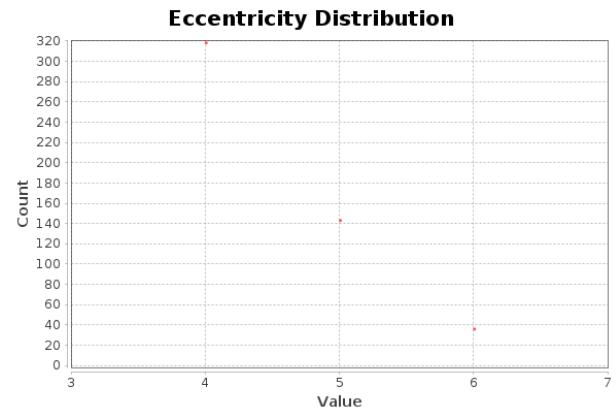


Fig. 8: City Only

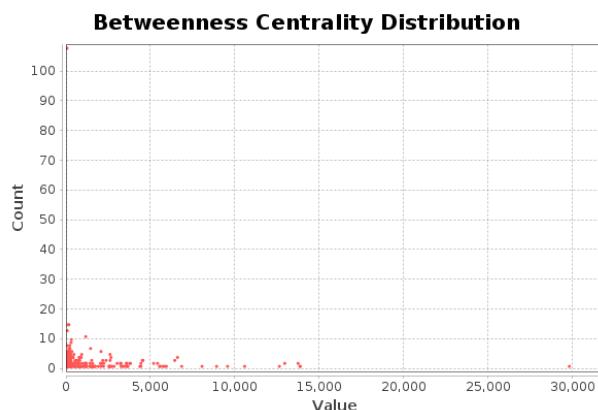


Fig. 6: Country+City

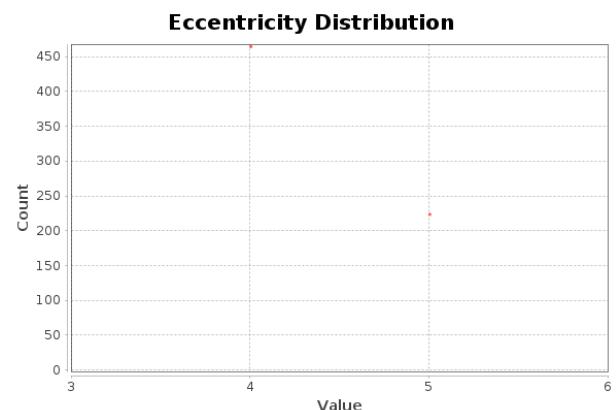


Fig. 9: Country+City

The addition of more chokepoints to defend or 'capture' (*traverse first*) makes it more difficult to establish control over the opponent.

Eccentricity lowering as the sample size increases implies that the game map is smaller; the player can reach (*or force their opponent*) towards fringe nodes faster.

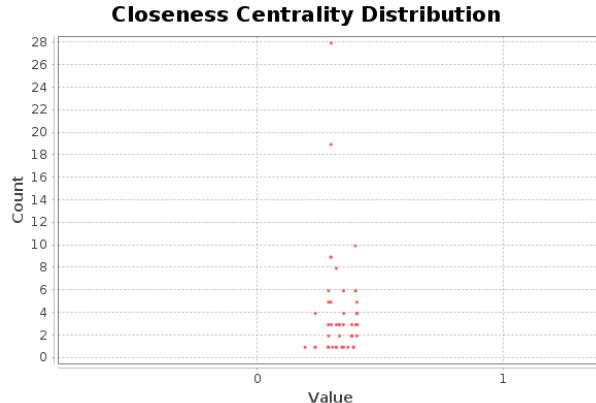


Fig. 10: Country Only

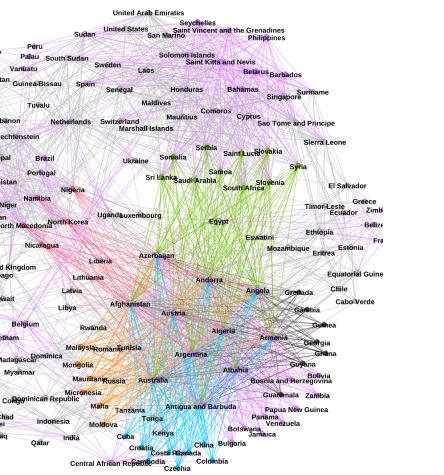


Fig. 13: Country Only

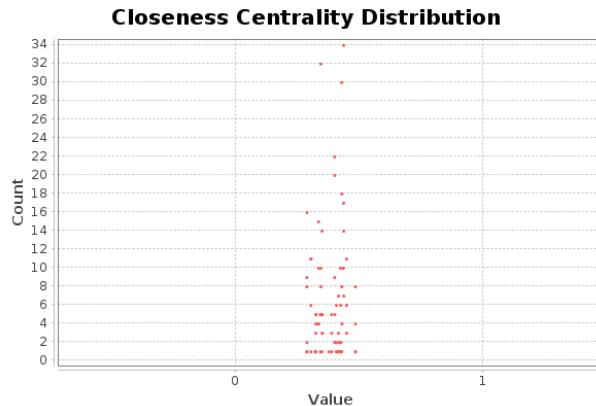


Fig. 11: City Only

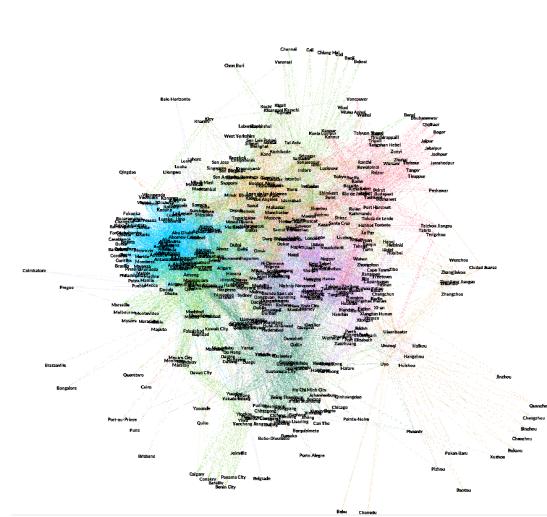


Fig. 14: City Only

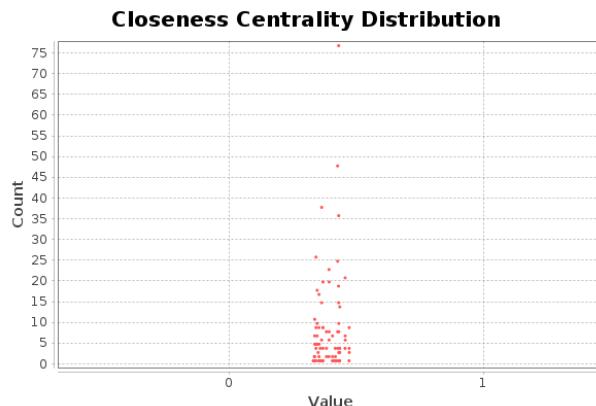


Fig. 12: Country+City

Tighter spread of scatterplot implies that the network is more densely connected, which makes it easier to maneuver yourself.

## II. CLUSTERING

Using **Modularity Classes** to identify and visualize clusters of *Densely Connected Components* in the ATLAS Graph:

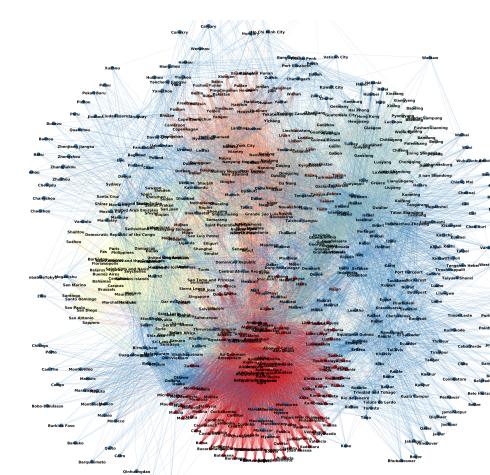


Fig. 15: Country+City

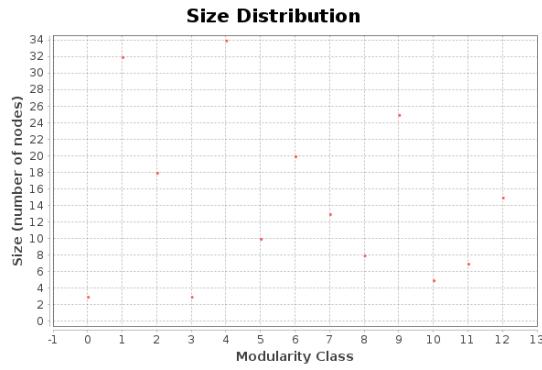


Fig. 16: Country Only

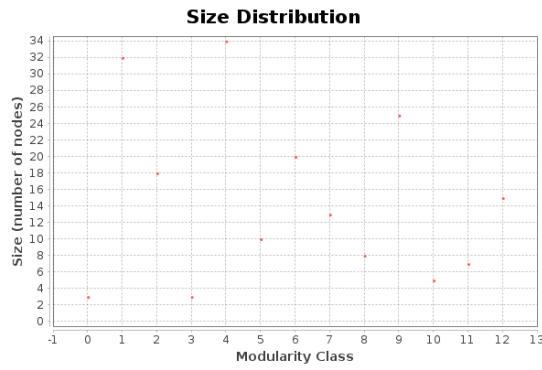


Fig. 17: City Only

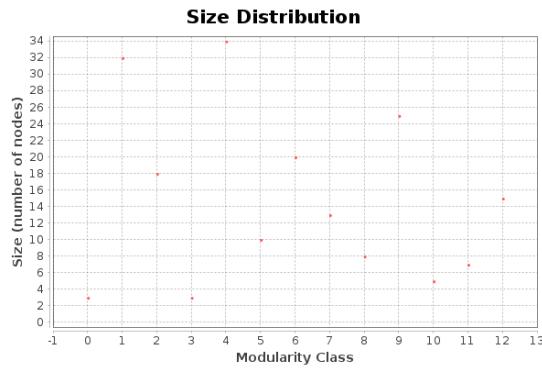


Fig. 18: Country+City

#### • Country & City Graph:

- High modularity; clusters = regions/cultures.
- Focus on intra-cluster chains; bridging between clusters is harder.
- Moderate modularity; clusters = geographic/economic ties.
- Exploit well-connected clusters; transitions require high-betweenness words.

#### • Country + City Graph:

- Lower modularity; more interconnected.
- Longer chains possible; increased competition for central nodes.

#### LINK PREDICTION

Given below is the comparative result for both the *Node2Vec* and *PyG* approaches to solving the *Edge Masking* problem on the ATLAS Graph. **Note:** 20% of Edges were Masked during testing.

```

Training unsupervised node2vec model.
Completed training probabilities: 100%
Generating walks (CPU: 3): 100%
Generating walks (CPU: 4): 100%
Generating walks (CPU: 2): 100%
Generating walks (CPU: 1): 100%
Node2Vec scores on masked edges - Mean: 0.1285, Std: 0.1208

Training unsupervised GAE model.
/home/sudhujoshi/.local/lib/python3.12/site-packages/torch_geometric/deprecation.py:26: UserWarning:
'train_test_split.edges' is deprecated, use 'transforms.RandomLinkSplit' instead
warnings.warn(out)

Epoch: 010, Loss: 0.3727, Test AUC: 0.6695, AP: 0.5937
Epoch: 010, Loss: 0.0990, Test AUC: 0.8892, AP: 0.8925
Epoch: 020, Loss: 0.0340, Test AUC: 0.9227, AP: 0.9208
Epoch: 030, Loss: 0.0969, Test AUC: 0.9367, AP: 0.9337
Epoch: 040, Loss: 0.0720, Test AUC: 0.9452, AP: 0.9385
Epoch: 050, Loss: 0.0540, Test AUC: 0.9518, AP: 0.9456
Epoch: 060, Loss: 0.0504, Test AUC: 0.9533, AP: 0.9456
Epoch: 070, Loss: 0.0454, Test AUC: 0.9566, AP: 0.9465
Epoch: 080, Loss: 0.0448, Test AUC: 0.9585, AP: 0.9500
Epoch: 090, Loss: 0.0438, Test AUC: 0.9577, AP: 0.9474

Final GAE Test Results:
AUC: 0.9912
Average Precision: 0.9500

```

Fig. 19: Result for Link Prediction, 693 Nodes

#### Parameter Results

##### • Mean and Std:

- **Mean:** Average performance score across runs.
- **Std:** Variability of results; low std = consistent performance.
- High mean + low std = good and consistent performance.

##### • AUC (Area Under the Curve):

- Measures ability to distinguish between existing and non-existent edges.
- Range: 0 to 1 (1 = perfect, 0.5 = random).
- High AUC = effective edge prediction.

##### • Average Precision:

- Summarizes precision-recall curve for edge prediction.
- Range: 0 to 1 (1 = perfect).
- High average precision = accurate predictions with few false positives.

It is observed that *Node2Vec* trades accuracy for speed due to the usage of **Random Walks**, while *PyG* is sluggish but more accurate due to the usage of **Tensors**.

#### REFERENCES

- Ulrik Brandes, A Faster Algorithm for Betweenness Centrality, in Journal of Mathematical Sociology 25(2):163-177, (2001)
- Vincent D Blondel, Jean-Loup Guillaume, Renaud Lambiotte, Etienne Lefebvre, Fast unfolding of communities in large networks, in Journal of Statistical Mechanics: Theory and Experiment 2008 (10), P1000
- R. Lambiotte, J.-C. Delvenne, M. Barahona Laplacian Dynamics and Multiscale Modular Structure in Networks 2009
- Robert Tarjan, Depth-First Search and Linear Graph Algorithms, in SIAM Journal on Computing 1 (2): 146–160 (1972)