

# Characterization of Simplicial Complexes by Counting Simplets Beyond Four Nodes: Online Appendix

## 1 CHARACTERIZATION OF NODES IN A SIMPLICIAL COMPLEX

Simplets and SC3 can also be used to characterize nodes, and the results can be used as input for node-level tasks. Specifically, we can characterize a node in an SC by the absolute count of each simplet in its *egonet*. For an SC  $\mathcal{K} = (V, E)$ , the *egonet* of a node  $v \in V$  is the subcomplex induced by  $v$  and its 1-hop neighbors, i.e.,  $\mathcal{K}[V_v]$  where  $V_v = \bigcup_{\sigma \in E \text{ s.t. } v \in \sigma} \sigma$ . As a result of characterization, each node is represented as a integer vector whose dimensionality is the same as the number of simplets (i.e.,  $s_k$ ).

## 2 APPLICATION: NODE CLASSIFICATION

We present an experiment on node classification.

**Data:** We used the 12 datasets from the following domains:

- **coauthorship:** cD, cMG, cMH
- **school:** chs, cps
- **email:** eEu, eEn
- **tags:** taau, taso
- **threads:** thau, thms, thso

For abbreviations, refer to Table 4 in the main paper. For each dataset, we use 100 nodes uniformly at random among those whose *egonet* has at least 10 nodes and at most  $10^4$  nodes. We use the domain of each dataset as the class of each node in the dataset.

As a result, we use 1,200 nodes and the following five classes: coauthorship, school, email, tags, and threads.

**Experimental Protocol:** We split the nodes randomly into a training set (80%) and a test set (20%). For each node, we repeated SC3 five times and averaged the counts of simplets in its *egonet*, which were used as input features, as described in Section 1. We used a gradient boosting method [1, 2] implemented in the Python scikit-learn library as a classifier.

**Competitors:** We compared SC3 with two simple baselines: (1) majority selection (i.e., yielding always a majority class as output), and (2) random guessing.

**Results:** The node-classification results are presented in Table 1. We use f1-micro scores as the accuracy measure. Our simplet-based approach significantly outperformed both baselines. Interestingly, its accuracy was slightly higher when  $k = 4$  than when  $k = 5$ .

Table 1: F1-micro Scores on Node Classification.

$k$	simplet based method	majority selection	random guessing
4	0.88	0.25	0.2
5	0.85	0.25	0.2

## REFERENCES

- [1] Leo Breiman. 1997. *Arcing the edge*. Technical Report. Citeseer.
- [2] Jerome H Friedman. 2001. Greedy function approximation: a gradient boosting machine. *Annals of statistics* (2001), 1189–1232.