

## Why "neural" community detection?

- Graph representation learning models allow exploiting a graph's (*i*) topology, (*ii*) temporal dynamics, and (*iii*) attribute features to obtain node, edge, or graph-level embeddings.



A function *f* fits (*learns*) a graph *G* and maps nodes to embeddings **H**, which are then used to obtain a set *C* of communities (*clusters*) [1].

- This joint exploration potentially improves on the detectability thresholds [2] of the graph's communities, while the obtained functions (*models*) may be used to predict unseen data.
- Real-world graphs for model evaluation are the norm in AI research  $\rightarrow$  but a *flawed* one!

<sup>[1]</sup> Passos et al., ACM CoNEXT/GNNet Workshop, 2024.

<sup>[2]</sup> Nadakuditi & Newman, Phys. Rev. Letters, 2012.



## How to evaluate those models?

- Real-world temporal graph data is scarce, of limited scope, and ground truths are dubious [3].
- To overcome it, we introduce the TADC-SBM generator, a <u>T</u>ime-varying, <u>A</u>ttributed, Degree <u>-C</u>orrected <u>S</u>tochastic <u>B</u>lock <u>M</u>odel [4] based on [5, 6] for benchmarks in *controlled* scenarios.



In addition to a block matrix **B**, we employ a transition matrix **T** to control the probability of nodes transitioning communities over time.

- We then focused on the "special" case where  $\boldsymbol{\tau} \coloneqq \eta \mathbf{I} + (1 - \eta) \frac{\mathbf{J} - \mathbf{I}}{k - 1}$  for our experiments.

[5] Ghasemian et al., Phys. Rev. X, 2016.[6] Tsitsulin et al., ACM Web Conference/GLB Workshop, 2021.

<sup>[3]</sup> Peel et al., Science Advances, 2017.[4] Passos et al., IEEE ISCC, 2025 (Accepted).



## **Benchmarking results**

- Model performance overall severely degraded as  $\eta \rightarrow 0$ .
- **Exception:** the TGC [7] model displayed good resilience.
- Traditional (*algorithmic*) approaches performed better or as good as SOTA neural models in most scenarios [4].
- The same in a previous study [1] with real-world graphs.
- Main takeaway: lots of room for further improvements, but "neural" community detection is not all we need.

## Limitations and future work

- Extending the model to support mixed memberships.
- Generating dynamic (node and edge-level) features.





 $\leftarrow$  Preprint and code available

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<sup>[1]</sup> Passos et al., ACM CoNEXT/GNNet Workshop, 2024. [4] Passos et al., IEEE ISCC, 2025 (Accepted). [7] Liu et al., ICLR, 2024.