

Vaccinating bridge nodes reduces cumulative deaths!

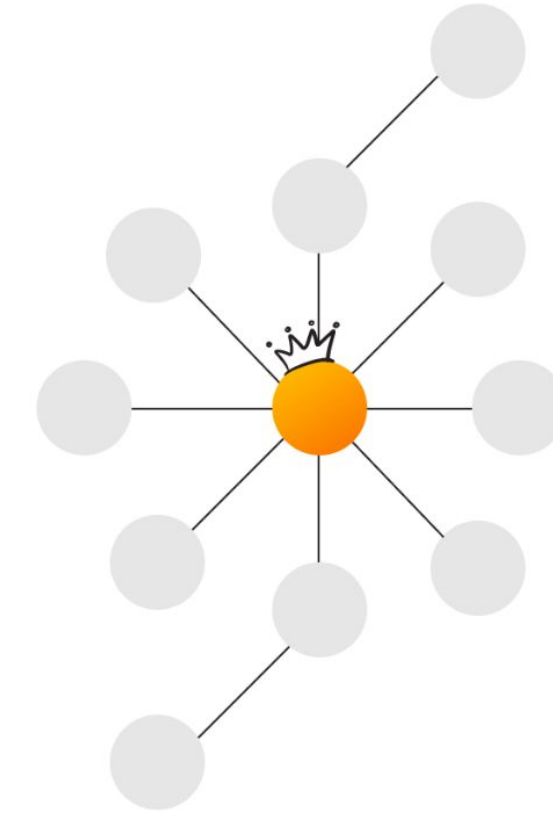
Evaluating Bridge-Node based Methods as Viable Alternatives to Degree-based Methods for Vaccine Allocation in Low Resource Settings



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Motivation

- **Vaccine scarcity** demands **efficient allocation**.
- Traditional **degree-based targeting** vaccinates **high-contact individuals** but **ignores network community structure**.
- **Real contact networks have communities** where **epidemics spread rapidly** within but **rely on bridge nodes** to propagate between.
- This is **rarely modeled**; **age-based** and **degree-based** methods favored.



Problem

Problem Statement: Given network $G=(V,E)$ with communities, limited vaccine supply ϕ , find target set T^* minimizing infections and deaths:

$$T^* = \operatorname{argmin} E[I(G,T)]$$

Hypothesis: Bridge-node targeting outperforms degree-based in low-resource settings.

Our Experimental Benchmark

2 datasets,
Demography vs
synthetic

7 node
selection
methods

3 infection
rates, 3 vaccine
availabilities

2
Simulation
Models

Random vs
Community-
Based Seeds

250+ experiments, 1 benchmark

Evaluation Datasets

LocationType	Nodes	Edges	AvgDegree
BasicsShop	1457	12659	17.37
Home	1648	1864	2.26
School	363	7412	40.83
SocialEvent	1420	19631	27.64
Work	1225	9952	16.24

Table 1: Network statistics for different social contact networks in the Zenodo population-based dataset

LocationType	Nodes	Edges	AvgDegree
Home	3997	5811	2.90
School	895	9805	21.91
SocialEvent	4000	50886	25.44
Work	2516	16688	13.26

Table 2: Network statistics for different social contact networks in the Synthetic population-based dataset

Bridge Node Detection Methods

Bridge-Hub Detector (BHD)

Random walks identify community boundaries; landing nodes flagged as bridges. Local-only.

Community-Based Bridge (CB)

Louvain partitioning + cross-community edge ratio weighted by distinct communities connected.

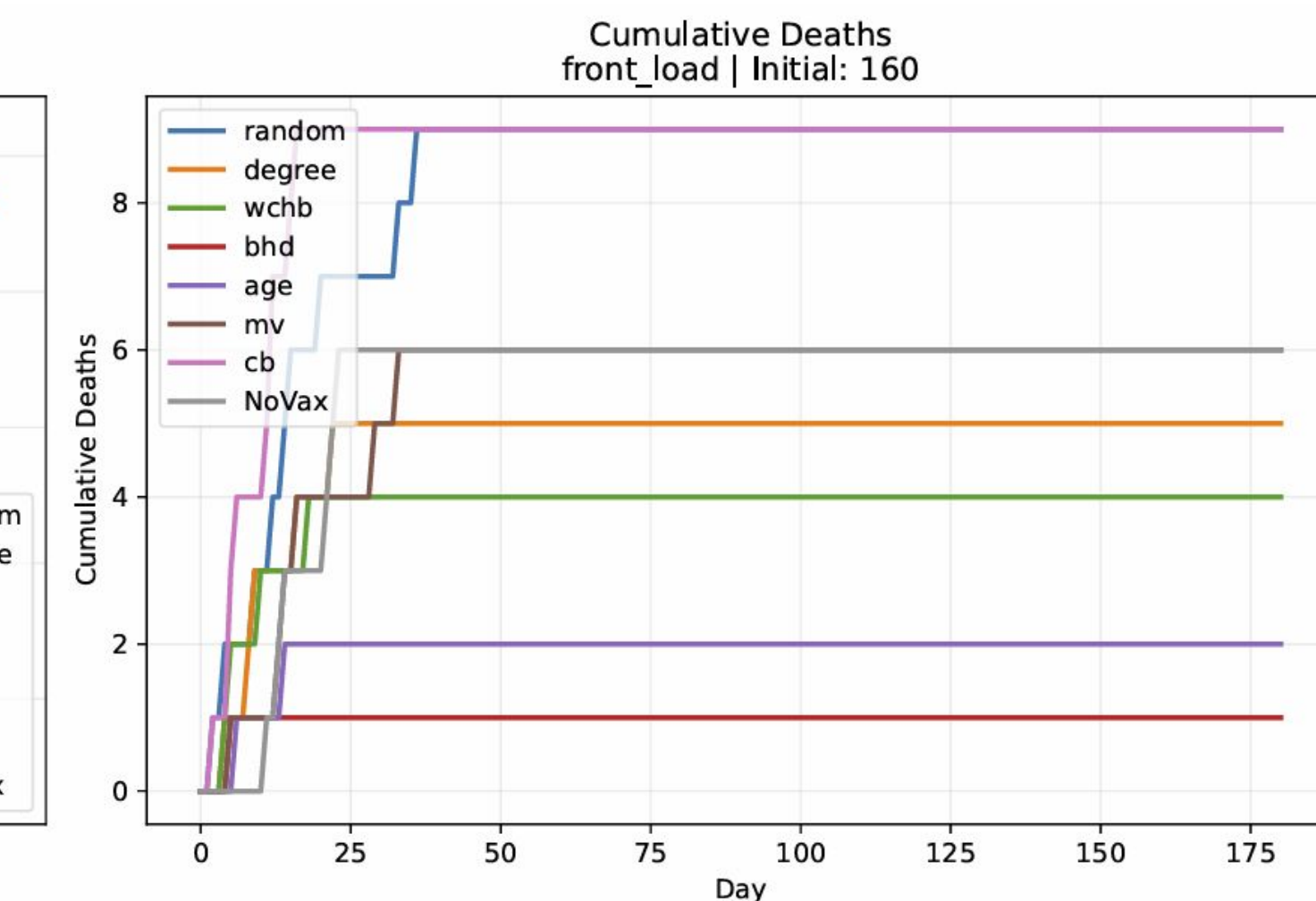
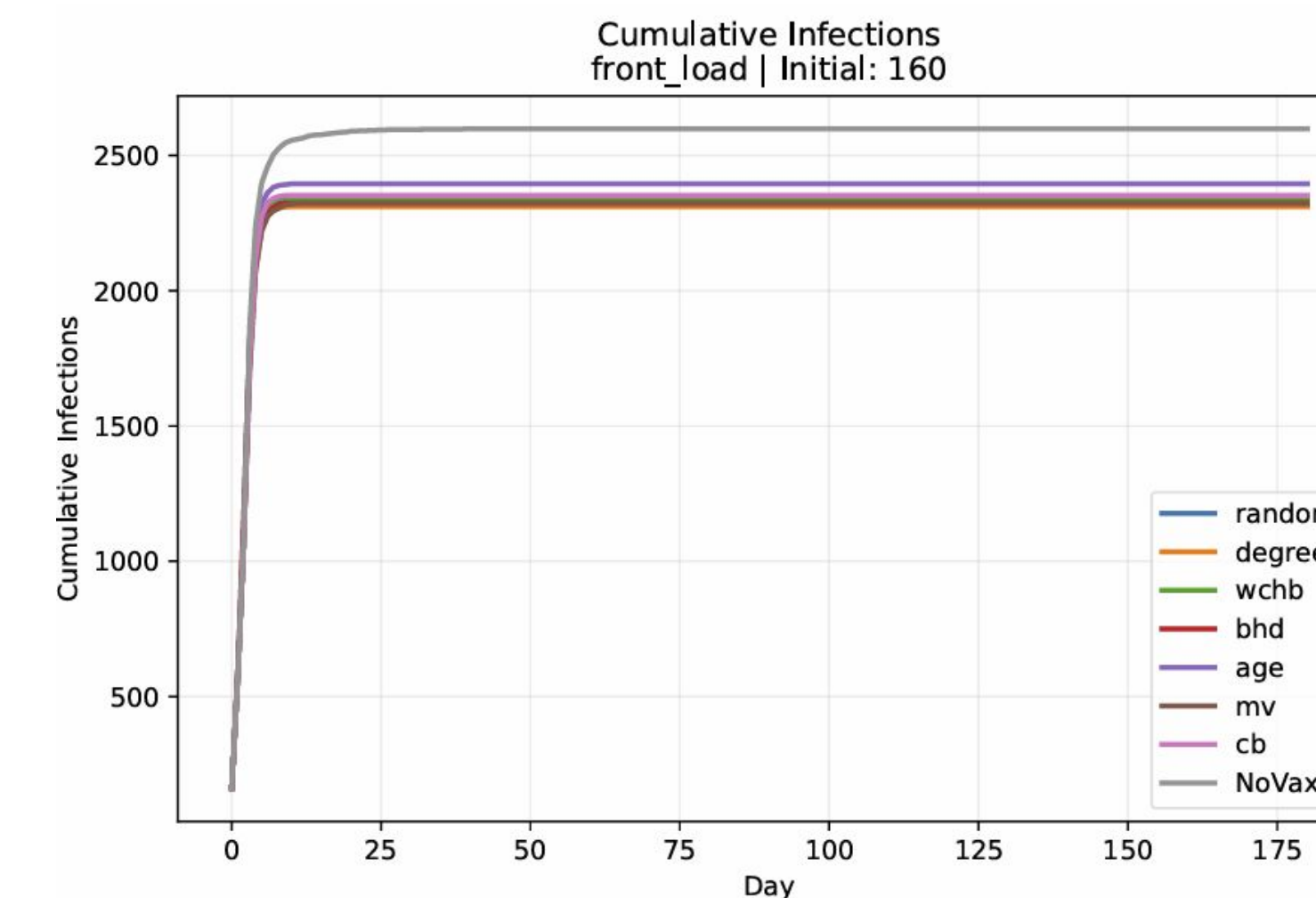
Weighted Community Hub-Bridge

Composite: $S = \alpha \cdot \text{Bridge} + (1-\alpha) \cdot \text{Hub}$. Balances inter/intra-community connectivity.

Modularity Vitality (MV)

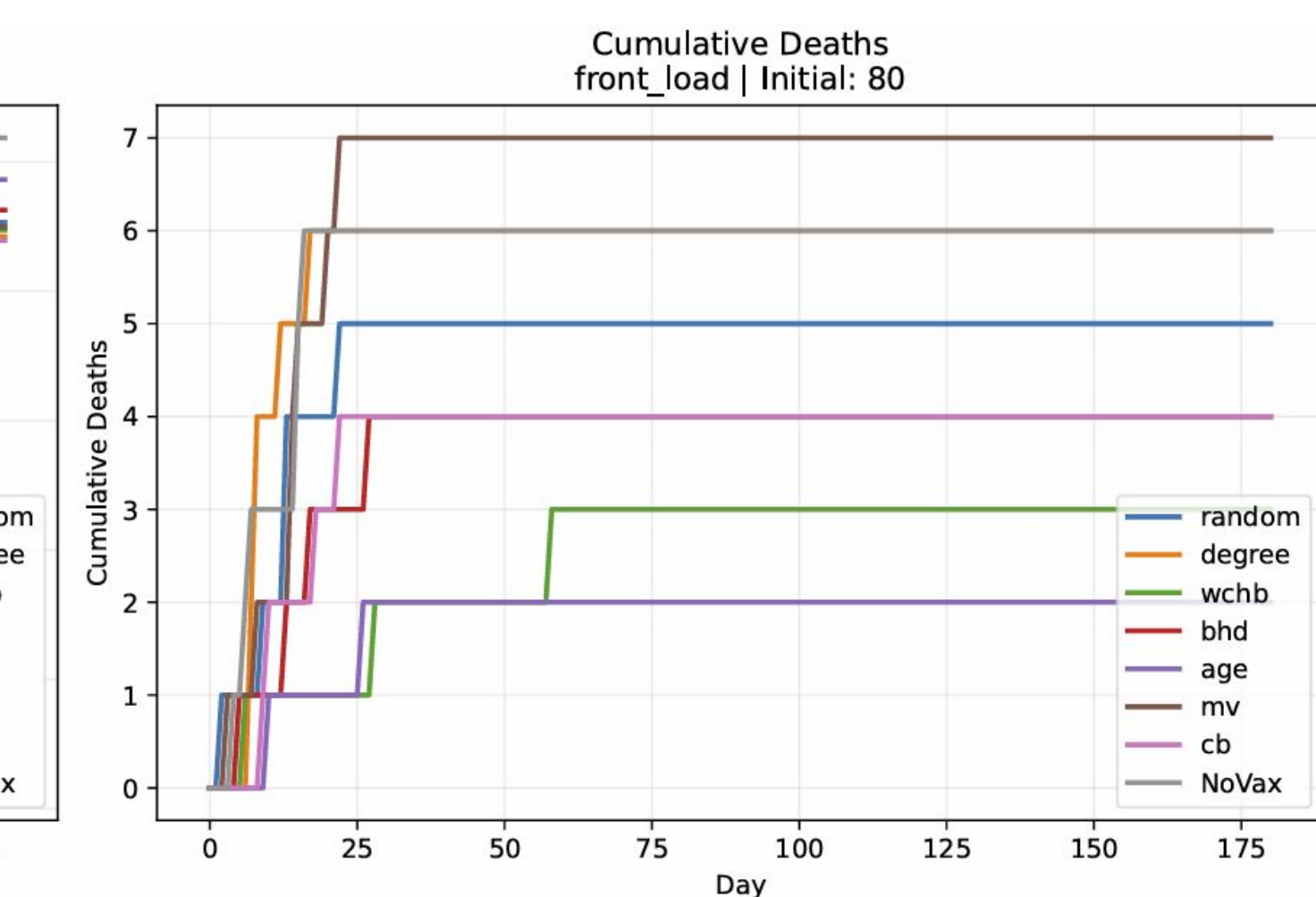
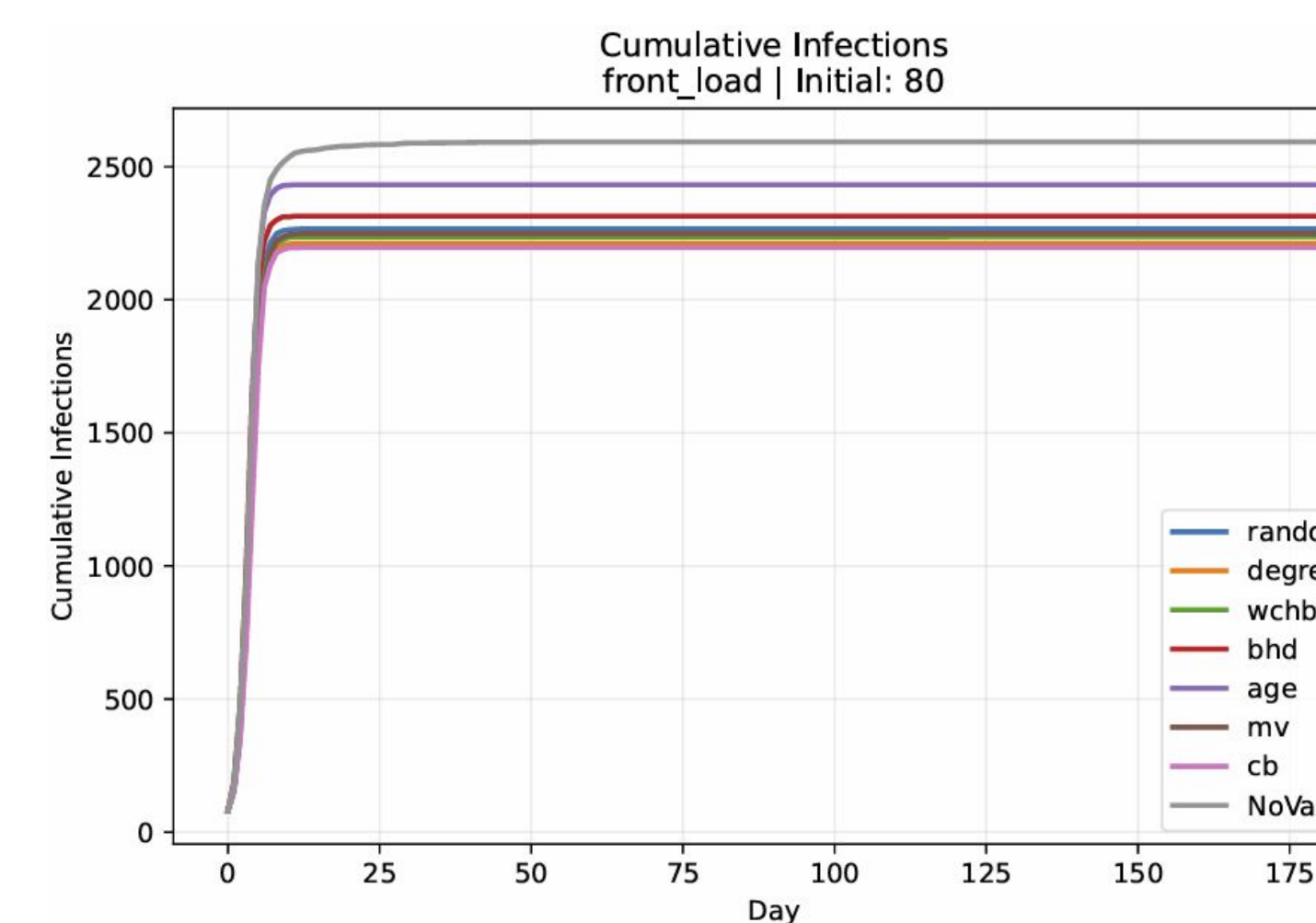
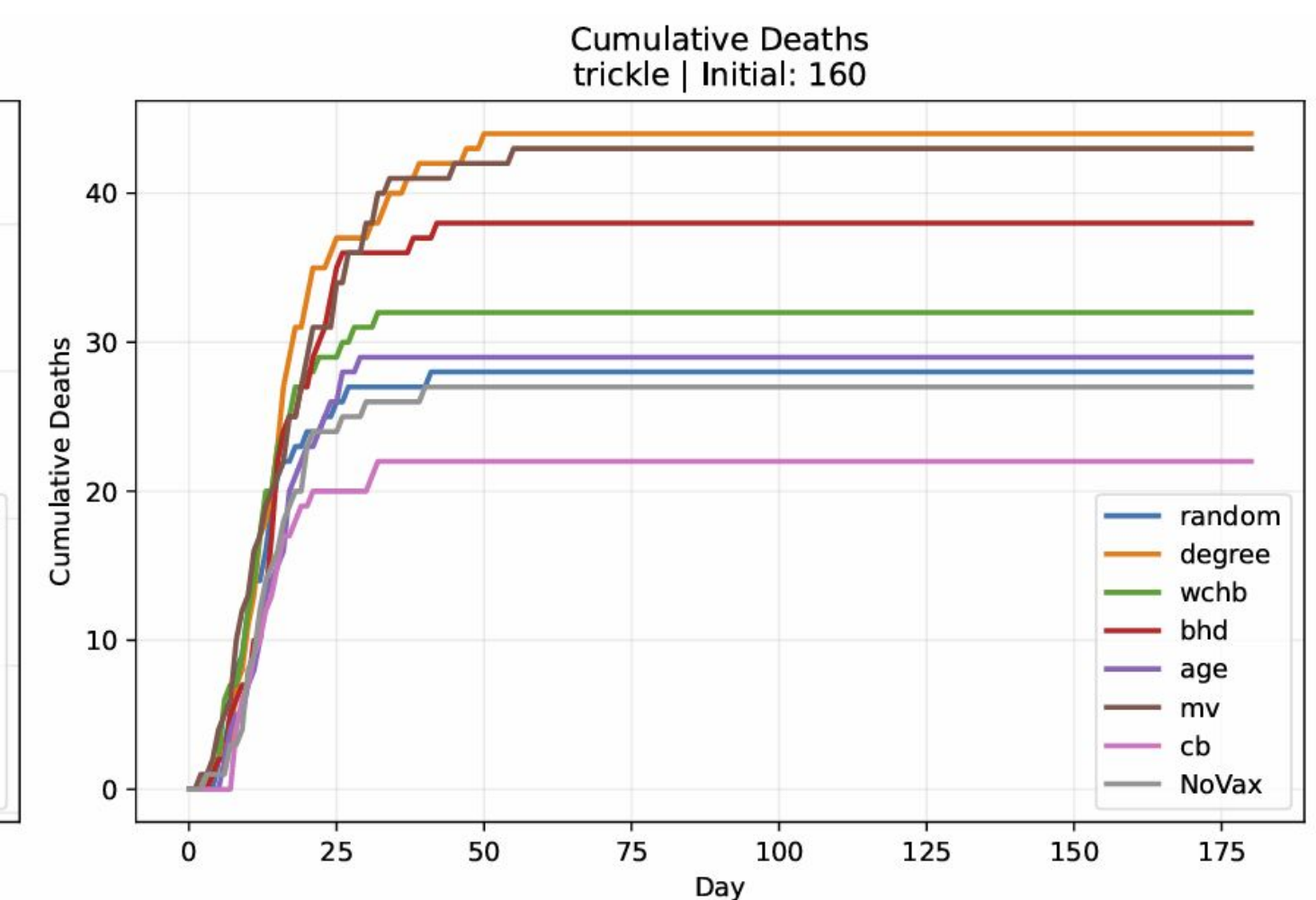
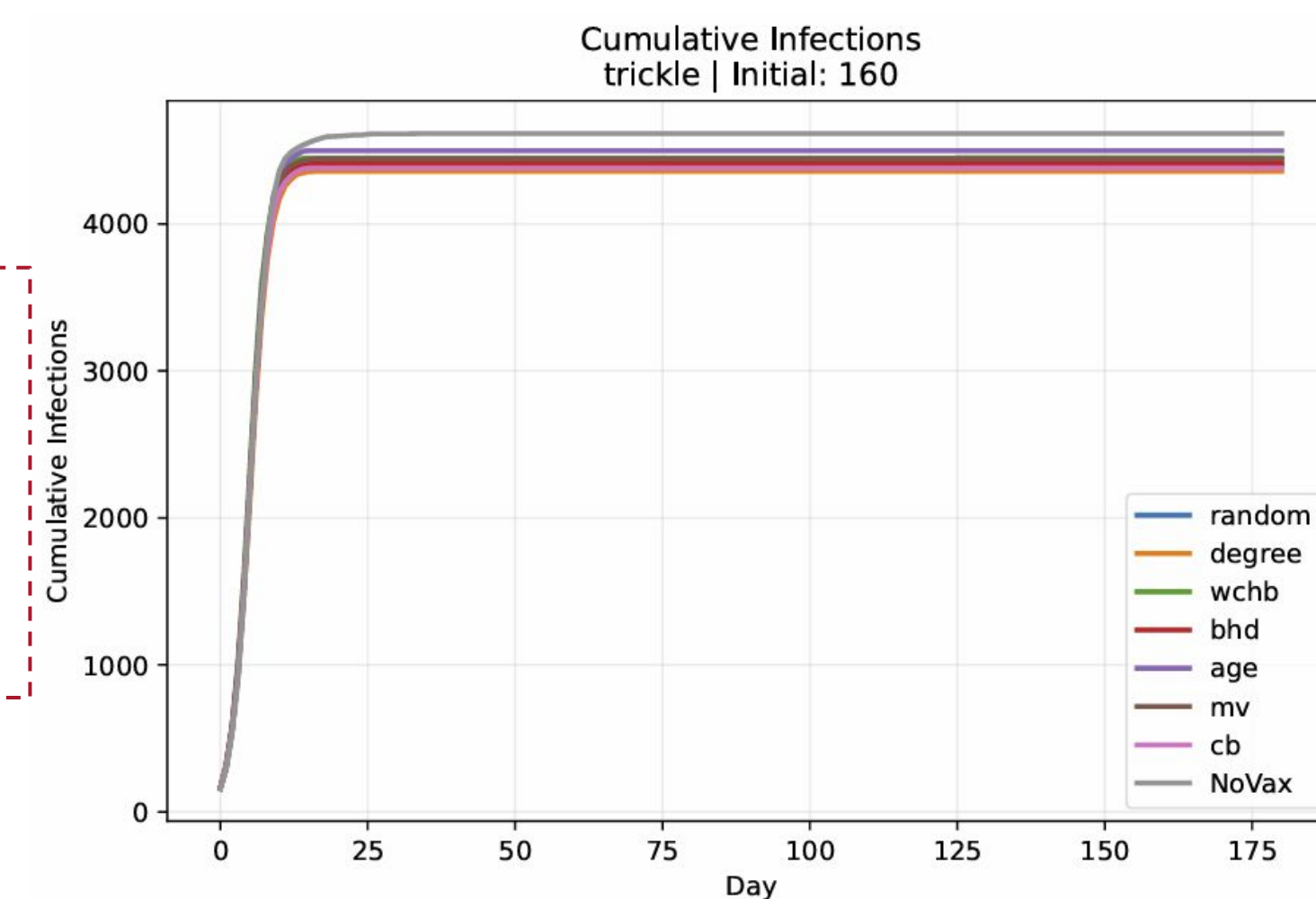
$\Delta Q = Q(G) - Q(G \setminus v)$. Nodes whose removal increases modularity act as structural glue.

Key Experimental Results



Random initialisations,
trickle vaccination,
initial infected nodes:
160, Zenodo dataset

Random initialisations,
trickle vaccination,
initial infected nodes:
160, Synthetic dataset



Random initialisations,
front-load vaccination,
initial infected nodes:
80, Zenodo dataset

Insights from Large-scale Experiments

Overall Performance Ranking: BHD > Age \approx WCHB > MV > Degree > CB > Random > NoVax

- ✓ **Bridge > Degree for Deaths:** Bridge nodes outperform degree, upto 50% lower deaths.
 - ✓ **Front-load Amplifies Gap:** Early deployment improves bridge methods' performance.
 - ✓ **Clustered Seeding Favors Bridges:** Upto 70% lower cumulative deaths than NoVax.
 - ✓ **Local Info Suffices:** BHD matches global information-based methods.
 - ✓ **Network demographic affects methods:** Age-based performs well when average population is older (Synthetic dataset), but underperforms bridge nodes in other scenarios.
- What's Next?** Eval on larger datasets, city-scale population, additional demographic data.