

SIR Model on Random Graphs

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SIR Model

- Used to model the spread of certain diseases
- Population divided into three states: susceptible, infected, and recovered
- $s(t)$, $i(t)$, and $r(t)$ are the fractions of the population in each state

$$\frac{ds}{dt} = -\beta is \quad \frac{di}{dt} = \beta is - \gamma i \quad \frac{dr}{dt} = \gamma i$$

β = rate at which infected individuals contact other individuals.

γ = rate at which infected individuals recover.

Random Graphs

Erdős-Renyi Random Graphs: $G(n, p)$, where n = number of nodes, p = probability that a possible edge exists.

Watts-Strogatz Random Graphs: $G(n, k, p)$, where n = number of nodes, k = number of nearest neighbors to which each node starts connected, p = probability of rewiring an edge

- Have the small-world property

Barabási-Albert Model: $G(n, m)$, where n = number of nodes, m = number of edges attached from a new node to existing nodes

- Generates scale-free graphs using preferential attachment

Stochastic Block Model (SBM):

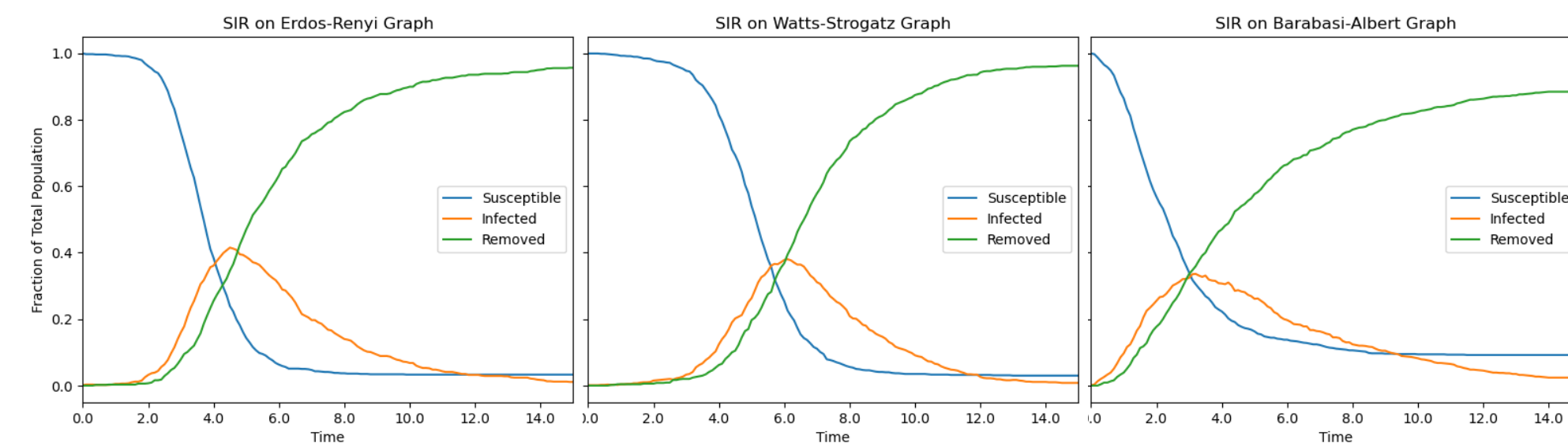
- Generates graphs that contain communities
- Degree of a node: number of nodes to which it is connected
- Degree distribution: probability distribution for the degrees across the whole network

Goal

Analyze SIR on various random graphs and explore the outbreak sizes.

SIR on Random Graphs

- Each node represents an individual
- An edge indicates that two individuals are in contact
- The disease can only spread between nodes in contact
- For each time step dt , a susceptible individual in contact with an infected individual becomes infected with probability βdt , an infected individual recovers with probability γdt
- Probability a susceptible node in contact with an infected node becomes infected: $T = \frac{\beta}{\beta + \gamma}$



Simulated SIR for 1000 time steps on Erdős-Renyi, Watts-Strogatz, and Barabási-Albert random graphs with mean degree $z = 10$, $dt = .1$, $\beta = .3$, $\gamma = .05$

Generating Functions [1]

$G_0(x) = \sum_{k=0}^{\infty} p_k x^k$ generates the degree distribution for a randomly chosen vertex in the network.

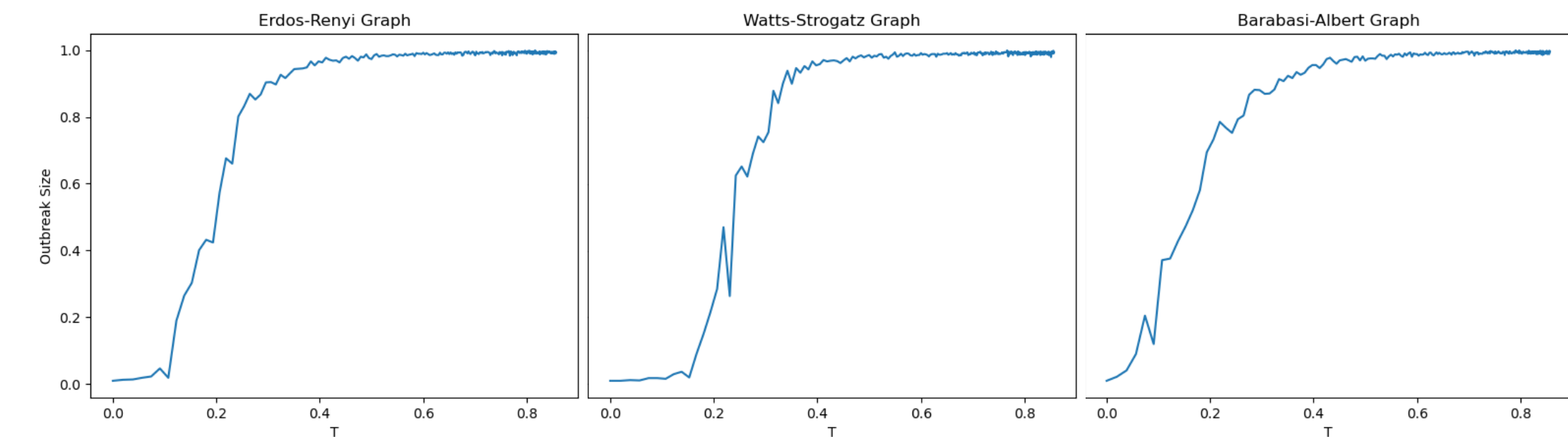
$G_1(x) = \frac{G'_0(x)}{G'_0(1)}$ generates the excess degree distribution for a vertex reached by following a randomly chosen edge

$$G_0(x; T) = G_0(1 + (x - 1)T)$$

$$G_1(x; T) = G_1(1 + (x - 1)T)$$

Outbreak Size

$S(T) = 1 - G_0(u; T)$, where u is the solution to $u = G_1(u; T)$



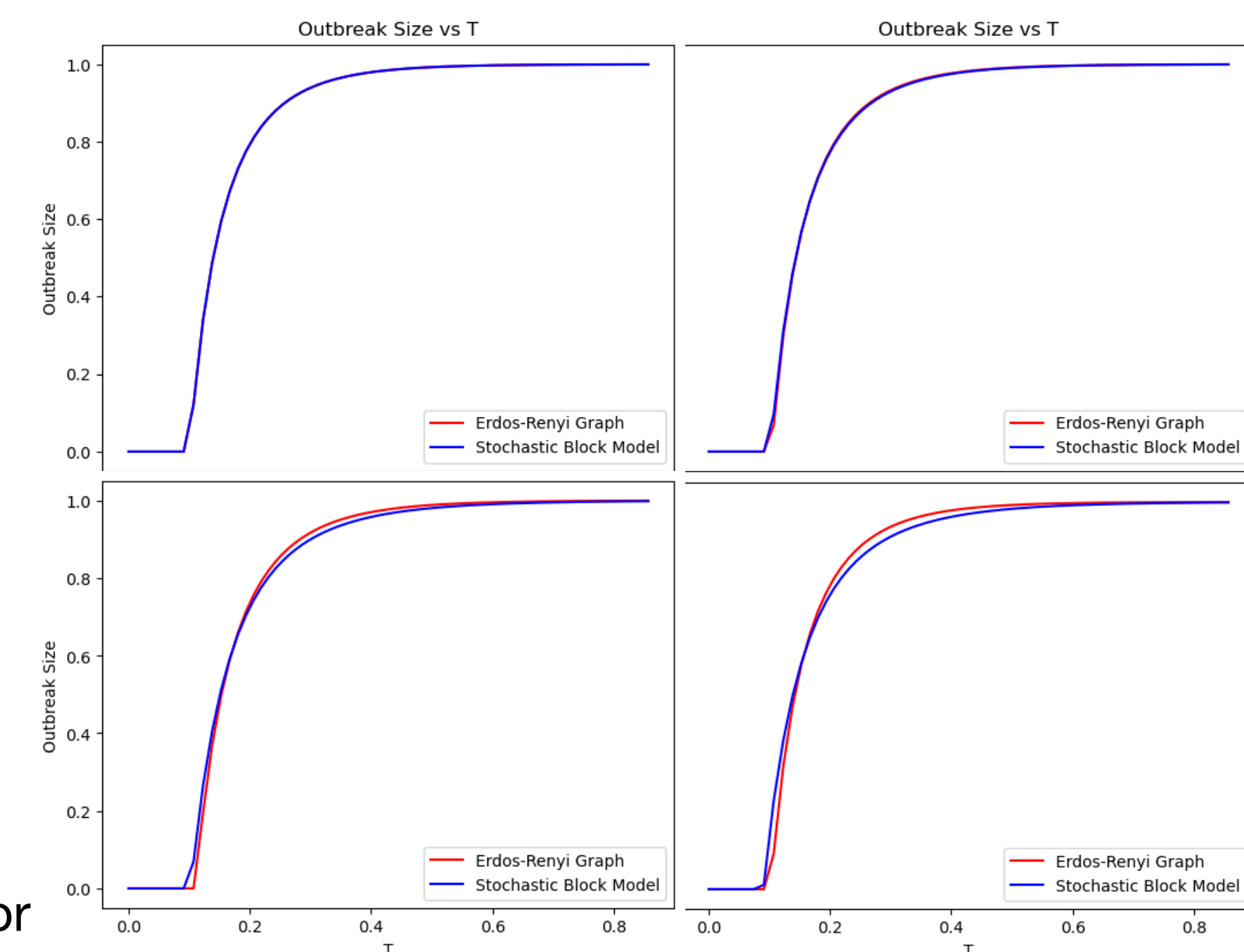
SIR with 1000 time steps with $z = 10$, $dt = .1$, final outbreak size as a function of T

- For an Erdős-Renyi graph: $G_0(x) = G_1(x) = e^{-z(1-x)}$ $S(T) = 1 - u$
- For a SBM with community sizes N_1, N_2, \dots, N_n and edge probability matrix P : $G_0^i(\mathbf{x}) = e^{-P_{i1}N_1(1-x_1)} e^{-P_{i2}N_2(1-x_2)} \dots e^{-P_{in}N_n(1-x_n)}$ generates the degree distribution by community for a random vertex in community i for large communities.

$$G_1^{i,j} = \frac{\frac{\partial G_0^i(\mathbf{x})}{\partial x_i}}{\frac{\partial G_0^i(\mathbf{x})}{\partial x_j}} = G_0^j(\mathbf{x})$$

generates the excess degree distribution by

community for a vertex reached following an edge from community i to community j . $u_i = G_0^i(\mathbf{u}; T)$ $S_i(T) = 1 - u_i$ is the outbreak size of community i $S(T)$ is the weighted average of each $S_i(T)$



$S(T)$ vs T for an Erdős-Renyi graph and a SBM with the same mean degree and number of nodes

TOP LEFT: SBM with equal community sizes, $P_{ii} = .05$, $P_{ij} = .01$, $i \neq j$
 TOP RIGHT: Equal community sizes, $P_{ii} = .05$, P_{ij} $i \neq j$ is a uniformly distributed random variable in $[0, .02]$
 BOTTOM LEFT: Equal community sizes, P_{ii} is a uniformly distributed random variable in $[0, 1)$, $P_{ij} = .01$, $i \neq j$
 BOTTOM RIGHT: Unequal community sizes, $P_{ii} = .05$, $P_{ij} = .01$, $i \neq j$

References

[1] Newman, M. E. (2002). Spread of epidemic disease on networks. *Physical review E*, 66(1), 016128.

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