

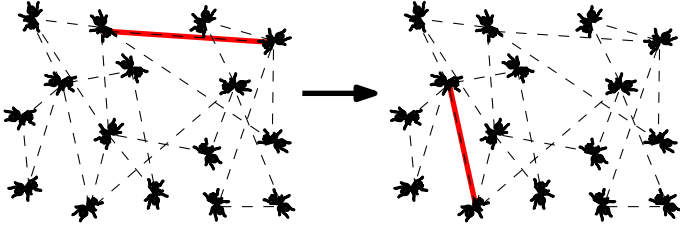
Friend or Foe?

Population Protocols can perform Community Detection

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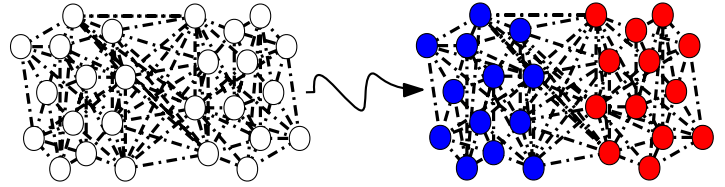
Population protocols

At each round a random edge is chosen and the two corresponding agents interact.



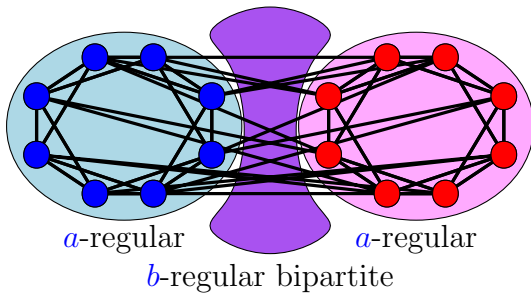
Reconstruction problem

Given graph generated by Regular Stochastic Block Model, find original partition.



Regular Stochastic Block Model

A graph $G = (V_1 \cup V_2, E)$ s.t. $|V_1| = |V_2|$, $G|_{V_1}$, $G|_{V_2} \sim$ random a -regular graphs, $G|_{E(V_1, V_2)} \sim$ random b -regular bipartite graph.



Theorem

$G = (V_1 \cup V_2, E)$ Regular Stochastic Block Model s.t. $d\epsilon^4 \gg b \log^2 n$, then w.h.p. $CSL(m, T)$ with $m = \Theta(\epsilon^{-1} \log n)$ and $T = \Theta(\log n)$ labels all nodes but a set U with size $|U| \leq \sqrt{\epsilon n}$, in such a way that

- nodes' labels in the same community agree on at least $5/6$ of entries, and
- nodes' labels in different communities differ in more than $1/6$ of entries.

A Taste of Spectral Analysis

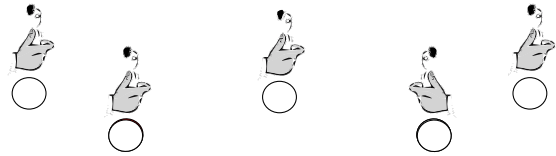
CSL is a **linear** dynamics

$$\mathbf{x}^{(t)} = \begin{pmatrix} \circ \\ \circ \\ \bullet \\ \bullet \\ \bullet \end{pmatrix} \quad \boxed{\mathbf{x}^{(t)} = W^{(t)} \cdot \mathbf{x}^{(t-1)} = (W^{(t)} \dots W^{(1)}) \cdot \mathbf{x}^{(0)}}$$

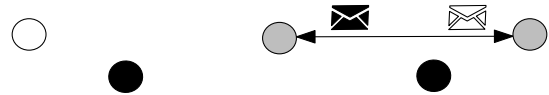
$E[W] = P$ transition matrix of r.w.
eigenvec. $\mathbf{v}_1, \dots, \mathbf{v}_n$, eigenval. $\lambda_1, \dots, \lambda_n$.

$$\underset{\text{community}}{C} \quad \overset{\text{sensitive}}{S} \quad \underset{\text{labeling}}{L} \quad (m, T)$$

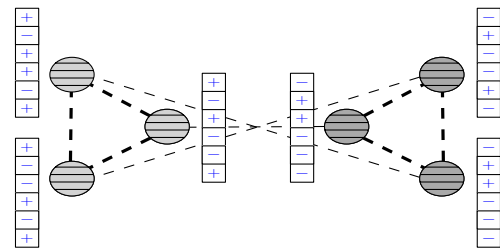
- At the outset $\mathbf{x}_u^{(0)} \sim \text{Unif}(\{-1, +1\}^m)$.



- In each round, the endpoints of the random edge choose a random index $j \in [m]$ and set $\mathbf{x}_u(j) = \mathbf{x}_v(j) = \frac{\mathbf{x}_u(j) + \mathbf{x}_v(j)}{2}$;



- At the T -th update of j -th component, u sets $\mathbf{h}_u(j) = \text{sgn}(\mathbf{x}_u(j))$.



$$\frac{1}{n} \sum_{u \in V} \mathbf{x}^{(0)}(u) = \text{average vector}$$

$$\frac{1}{n} \sum_{u \in V_1} \mathbf{x}^{(0)}(u) - \frac{1}{n} \sum_{u \in V_2} \mathbf{x}^{(0)}(u) = \text{difference vector}$$

$$\boxed{\mathbf{x}^{(t)} = \frac{1}{n} (\mathbf{1}^T \mathbf{x}^{(0)}) \mathbf{1} + \left(\frac{a-b}{a+b} \right)^t \frac{1}{n} (\chi^T \mathbf{x}^{(0)}) \chi + \mathbf{e}^{(t)}}$$