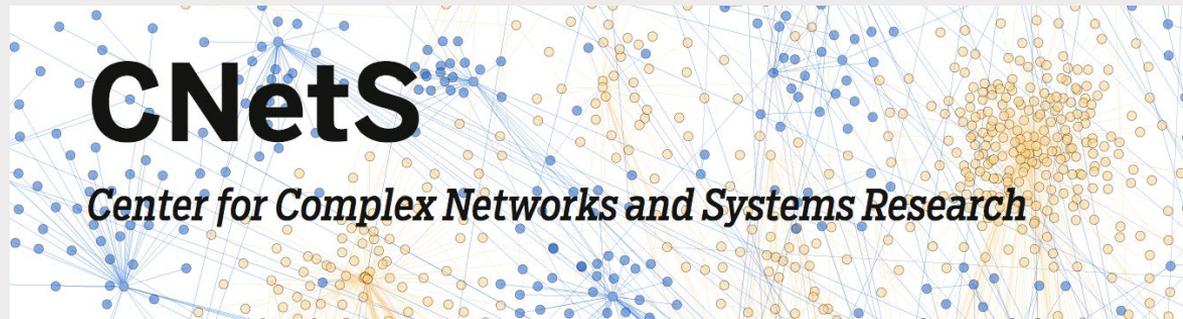


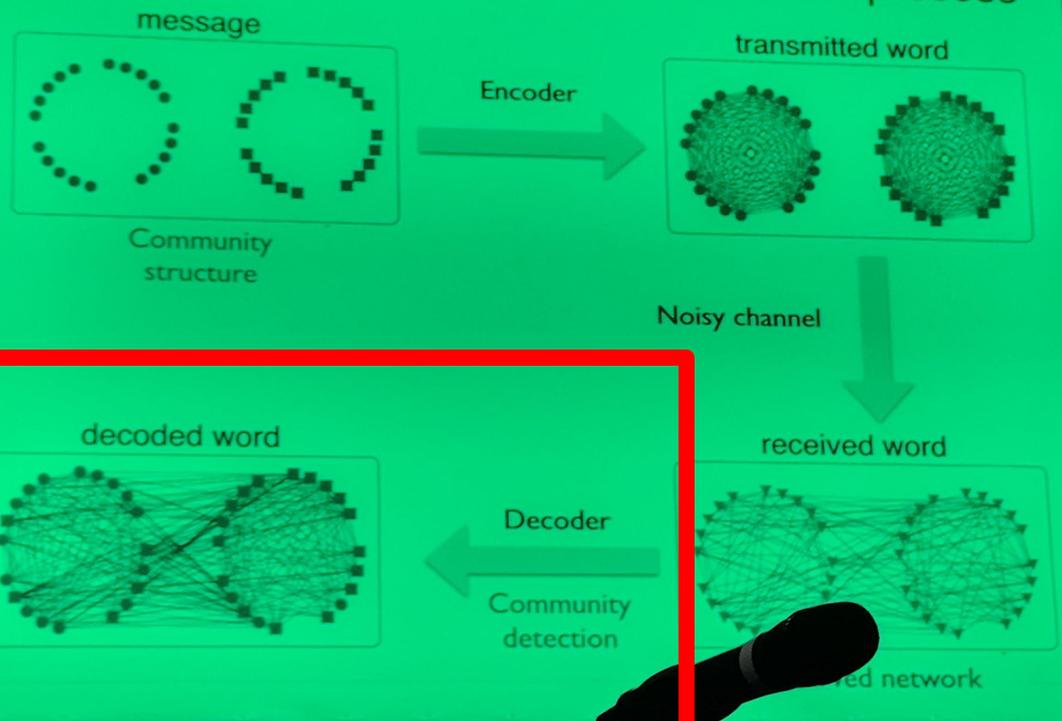
Error-Correcting Decoders for Communities in Networks

Krishna C. Bathina and Filippo Radicchi

Indiana University



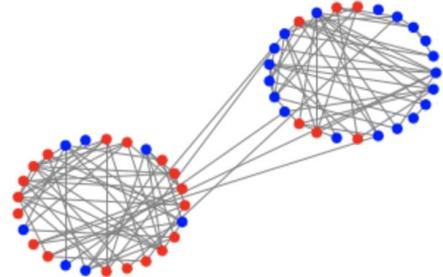
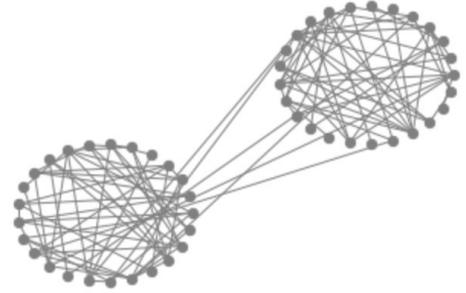
Community detection as a communication process



- Detectability threshold
- Noisy channel capacity
- Capacity achieving codes

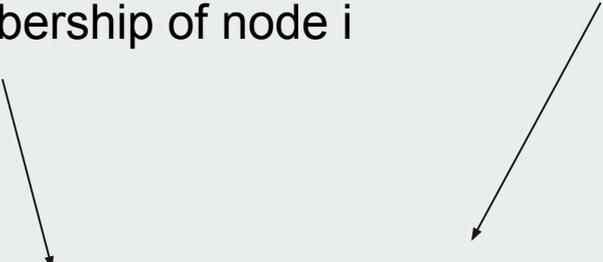
Message becomes Decoded

A distorted message is passed through a decoding algorithm and the original message is returned - at least partially



0 if community of i = community of j
1 if community of i \neq community of j

Community membership of node i


$$\sigma_i + \sigma_j + \theta_{ij} \pmod{2} = 0$$

If the equation is solved for all n^2 pairs of nodes, then the distorted message has been perfectly decoded

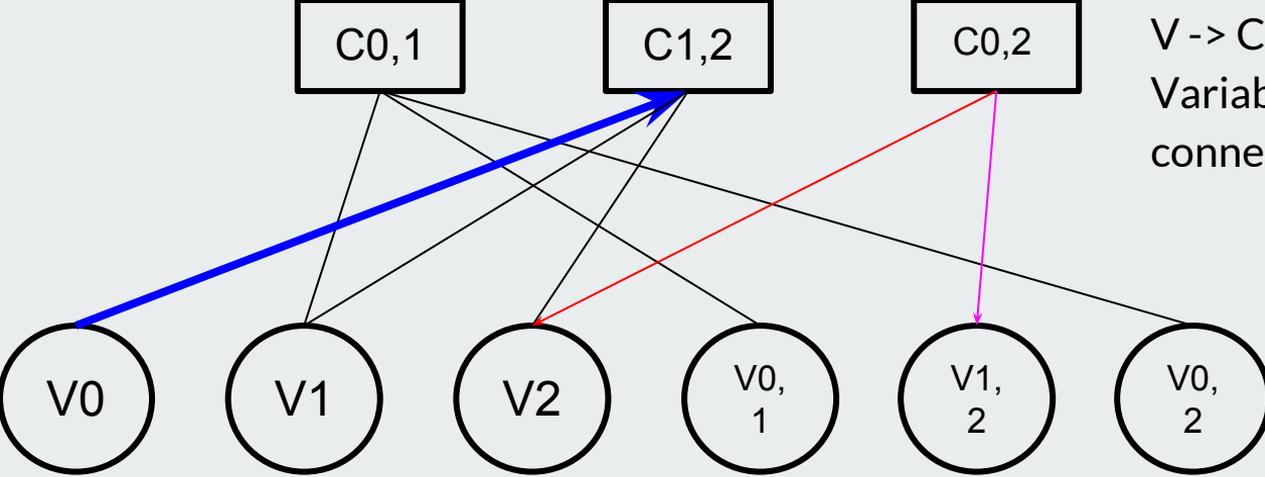
Gallager Codes (LDPC codes)

- Linear code based on a *Low-Density Parity Check* matrix H
- Check nodes - pairs of nodes in graph (3)
- Variable nodes - number of bits in codeword (6)

$$N + \frac{N(N-1)}{2}$$

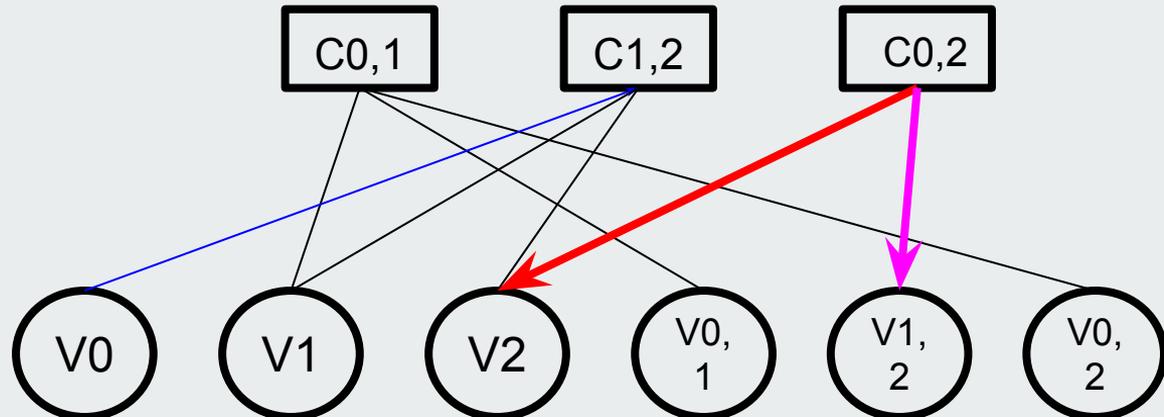
$$\mathcal{H} = \begin{bmatrix} 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \end{bmatrix}$$

$$\frac{N(N-1)}{2}$$



V -> C Message = probability of Variable's bit according the connected Check nodes

C -> V Message = probability of Variable's bit that would satisfy the other connected Variable nodes



a priori log likelihood ratio (LLR)

- Prior belief about the message given the information received
- Determines which steady state value the algorithm will converge to

Variable nodes i - Logarithm of the ratio of the community memberships given the received information bit

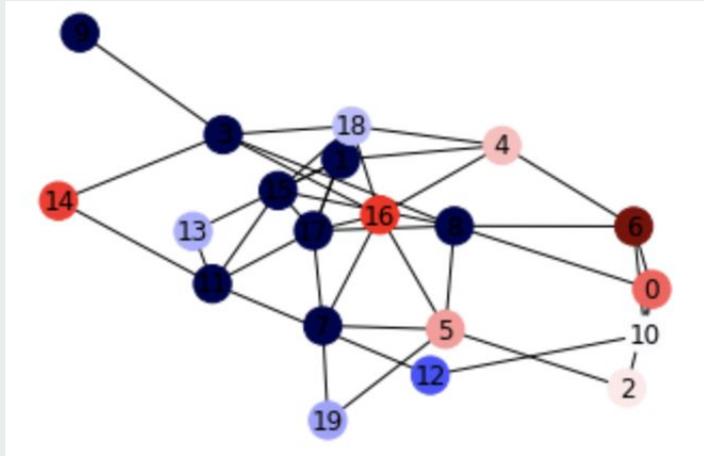
$$\ell_i = \log \frac{P(\sigma_i = 0 | s_i)}{P(\sigma_i = 1 | s_i)}$$

Variable nodes ij - Logarithm of the ratio of the parity bits given the existence of an edge

$$\ell_{ij} = \log \frac{P(\theta_{ij} = 0 | A_{ij})}{P(\theta_{ij} = 1 | A_{ij})}$$

Random

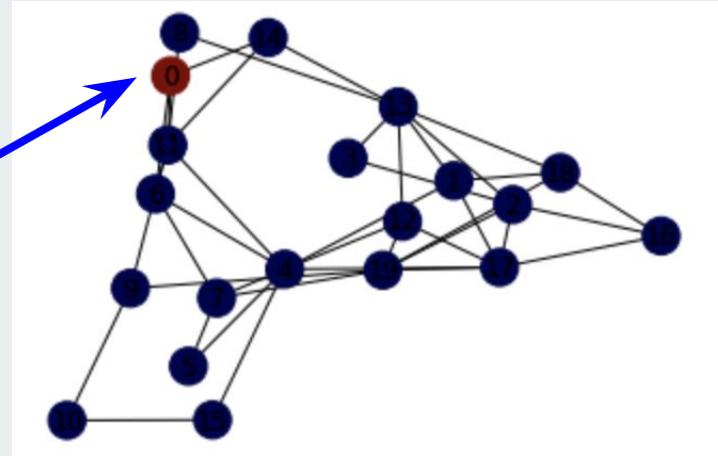
$$l_N \sim \mathcal{U}(-1, 1)$$



Regular

$$l_i = 1 \quad l_{N \setminus i} = 0$$

Very confident
about
community
membership



$$\ell_{ij} = \log \frac{P(\theta_{ij} = 0 | A_{ij})}{P(\theta_{ij} = 1 | A_{ij})}$$

$$= \begin{cases} \log(p_{in}) - \log(p_{out}) & \text{if } A_{ij} = 1 \\ \log(1 - p_{in}) - \log(1 - p_{out}) & \text{if } A_{ij} = 0 \end{cases}$$

$A_{i,j}$	$\theta_{i,j}$	$P(A_{i,j} \theta_{i,j})$	$P(\theta_{i,j} A_{i,j})$
1	0	P_{in}	$\frac{P_{in}}{P_{in}+P_{out}}$
1	1	P_{out}	$\frac{P_{out}}{P_{in}+P_{out}}$
0	0	$1 - P_{in}$	$\frac{1-P_{in}}{2-(P_{in}+P_{out})}$
0	1	$1 - P_{out}$	$\frac{1-P_{out}}{2-(P_{in}+P_{out})}$

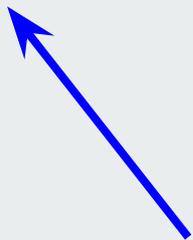
Stochastic block model

P_{in} is large enough to convey information about the community assignments



$$P_{in} - P_{out} \geq \frac{2\sqrt{k}}{N}$$

$$P_{in} + P_{out} = \frac{2k}{N}$$



Tunable parameter



$$P_{in} = \alpha \frac{k + \sqrt{k}}{N}$$

$$P_{out} = \max\left(0, \frac{\alpha k}{N} - P_{in}\right)$$

2 communities

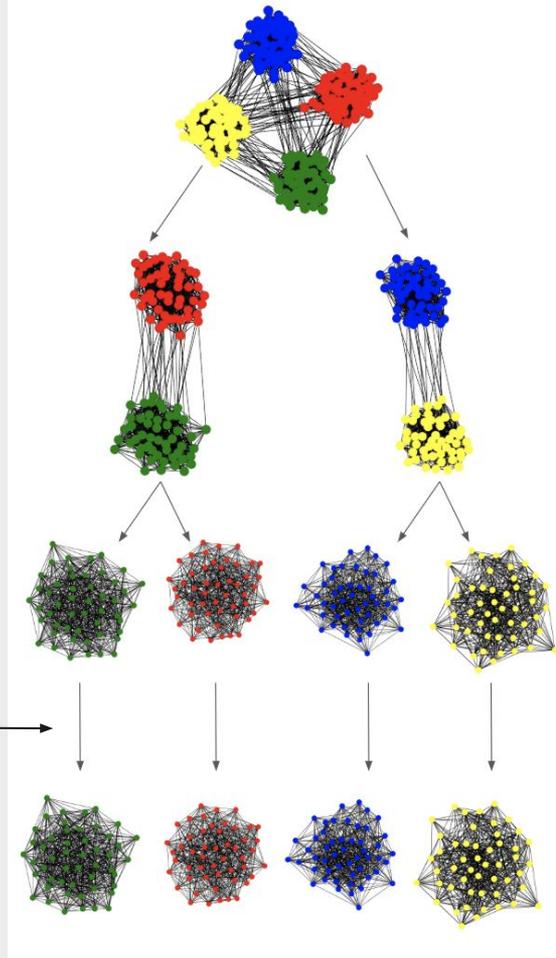
Stochastic Block Model

Decelle, Aurelien, et al. "Asymptotic analysis of the stochastic block model for modular networks and its algorithmic applications." *Physical Review E* 84.6 (2011): 066106.

Algorithm

1. Choose starting condition
2. Run on network
 - a. Iterative decoding - new P_{in} and P_{out} for each iteration
 - b. Return 2 sub-networks
3. Repeat on each subnetwork until no new splits are formed

Split 2



Original Algorithm (Reformulated Gallagher)

$$F(a, x) = \log \frac{1 + a \tanh \frac{x}{2}}{1 - a \tanh \frac{x}{2}}$$

Best *a priori* initial estimate of message

$$\zeta_{i \rightarrow j}^{t=0} = l_i$$

Update to LLR based on all other nodes

$$\zeta_{i \rightarrow j}^t = \zeta_{i \rightarrow j}^{t=0} + \sum_{k \in N \setminus i, j} F\left[\tanh \frac{l_{i,j}}{2}, \zeta_{k \rightarrow i}^{t-1}\right]$$

$$l_i^f = l_i + \sum_{k \in N \setminus i} F\left[\tanh \frac{l_{i,k}}{2}, \zeta_{k \rightarrow i}^{f-1}\right]$$

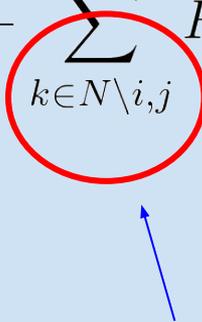
$$l_{i,j}^f = l_{i,j} + F\left[\tanh \frac{l_{i,j}}{2}, \zeta_{j \rightarrow i}^{f-1}\right]$$

Hard Decision

$$\left\{ \begin{array}{l} \sigma_i = 0 \text{ if } l_i^f > 0 \\ \theta_{i,j} = 0 \text{ if } l_{i,j}^f > 0 \end{array} \right.$$

Reduced Algorithm

- Original algorithm - messages are updated by all pairs of nodes
 - Even if $A_{ij} = 0$
- Assume messages passed between unconnected nodes are constant

$$\zeta_{i \rightarrow j}^t = \zeta_{i \rightarrow j}^{t=0} + \sum_{k \in N \setminus \{i, j\}} F\left[\tanh \frac{\ell_{i, j}}{2}, \zeta_{k \rightarrow i}^{t-1}\right]$$


LLR iterates on all node pairs, even if not connected

Constants

$A_{i,j}$	$\theta_{i,j}$	$P(A_{i,j} \theta_{i,j})$	$P(\theta_{i,j} A_{i,j})$
1	0	P_{in}	$\frac{P_{in}}{P_{in}+P_{out}}$
1	1	P_{out}	$\frac{P_{out}}{P_{in}+P_{out}}$
0	0	$1 - P_{in}$	$\frac{1-P_{in}}{2-(P_{in}+P_{out})}$
0	1	$1 - P_{out}$	$\frac{1-P_{out}}{2-(P_{in}+P_{out})}$

$$\ell_{con} = P(\theta = 0|A = 1) - P(\theta = 1|A = 1) = \frac{P_{in} - P_{out}}{P_{in} + P_{out}}$$

$$\ell_{non} = P(\theta = 0|A = 0) - P(\theta = 1|A = 0) = \frac{P_{out} - P_{in}}{2 - P_{in} - P_{out}}$$

Update to LLR if $A_{i,s} = 1$

Update to LLR for all
unconnected nodes

$$\zeta_{i \rightarrow j}^{t=0} = l_i$$

Best *a priori* initial estimate

$$\zeta_{i \rightarrow j}^t = \zeta_{i \rightarrow j}^{t=0} + (N - k_i - 1)F\left[\frac{\ell_{non}}{2}, \mathcal{Z}^{t-1}\right] + \sum_{s \in \mathcal{N}_i \setminus j} F\left[\frac{\ell_{con}}{2}, \zeta_{s \rightarrow i}^{t-1}\right]$$

Number of unconnected nodes to node i excluding node j

$$\mathcal{Z}^{t=0} = \frac{\sum_{i=1}^N (N - k_i - 1) \ell_i}{N(N - 1) - 2M}$$

**Sum of all initial messages sent
to unconnected nodes**

Total number of non-edges

**Best *a priori*
initial
estimate**

Average update to LLR between connected nodes

$$\mathcal{Z} = \mathcal{Z}^{t=0} + F\left[\frac{\ell_{non}}{2}, \mathcal{Z}^{t-1}\right] + \frac{\sum_{i=1}^N \sum_{j \in \mathcal{N}_i} F\left[\frac{\ell_{con}}{2}, \zeta_{i \rightarrow j}^{t-1}\right]}{N(N - 1) - 2M}$$

Number of unconnected nodes to node i

$$l_i^f = l_i + \sum_{s \in \mathcal{N}_i} + F \left[\frac{l_{con}}{2}, \zeta_{s \rightarrow i}^{f-1} \right] + (N - k_i) F \left[\frac{l_{non}}{2}, \mathcal{Z}^{f-1} \right]$$

$$l_{i,j}^f = \log \frac{P_{in}}{P_{out}} + F \left[\tanh \frac{\zeta_{i \rightarrow j}^{f-1}}{2}, \zeta_{j \rightarrow i}^{f-1} \right]$$

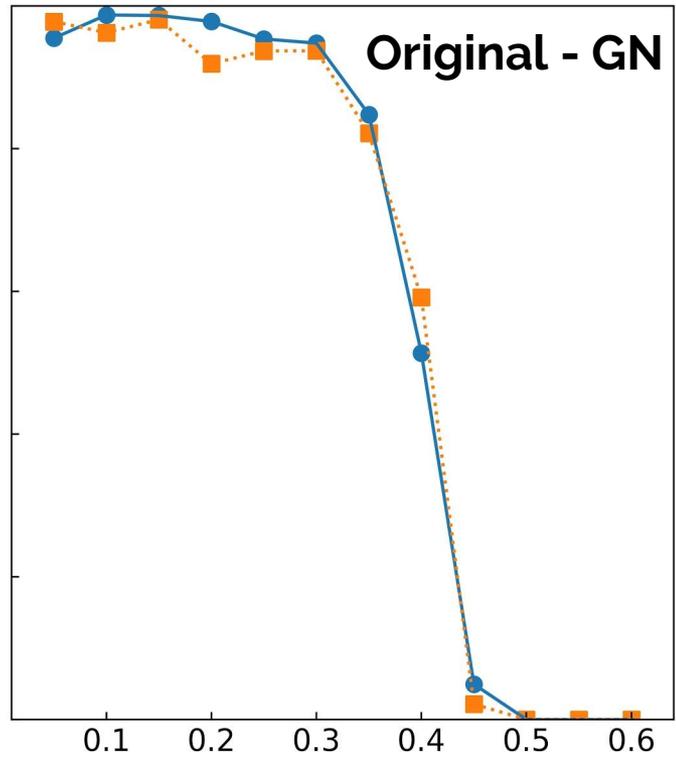
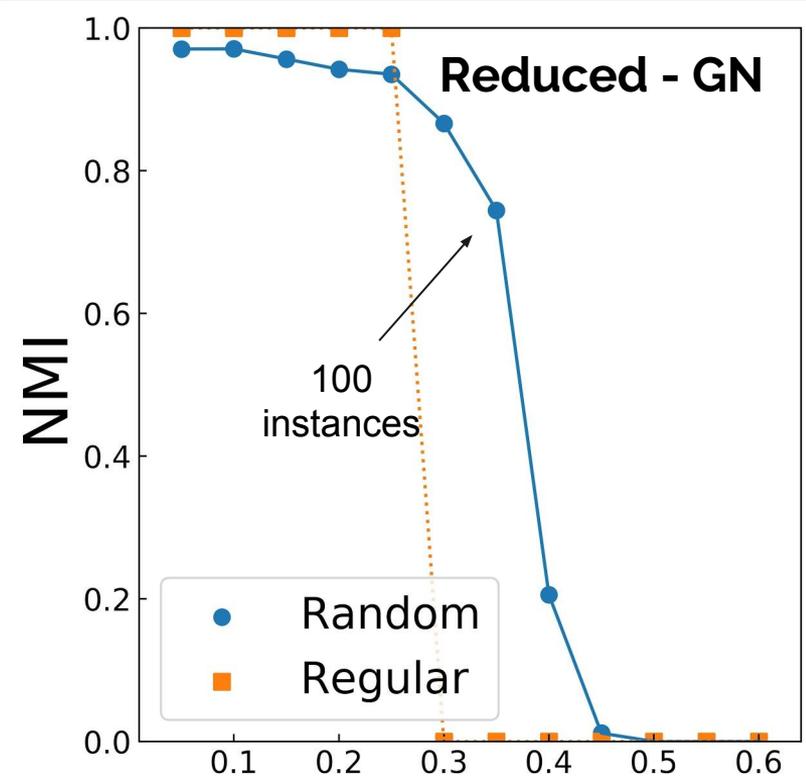
Hard Decision

$$\left\{ \begin{array}{l} \sigma_i = 0 \text{ if } l_i^f > 0 \\ \theta_{i,j} = 0 \text{ if } l_{i,j}^f > 0 \end{array} \right.$$

Best Estimate of LLR

Girvan Newman

Mostly either completely perfect or completely imperfect recovery



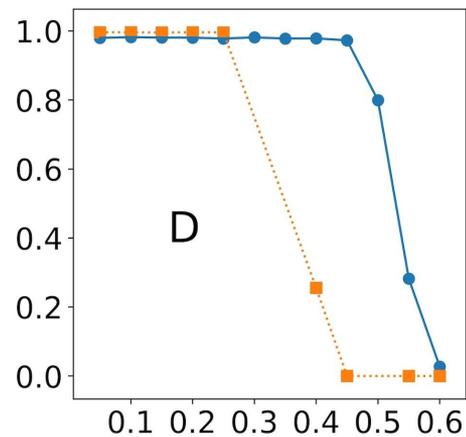
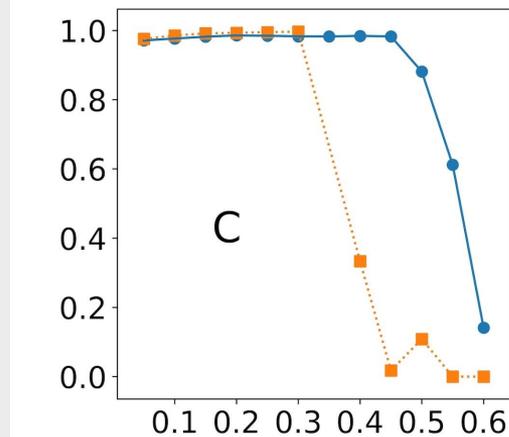
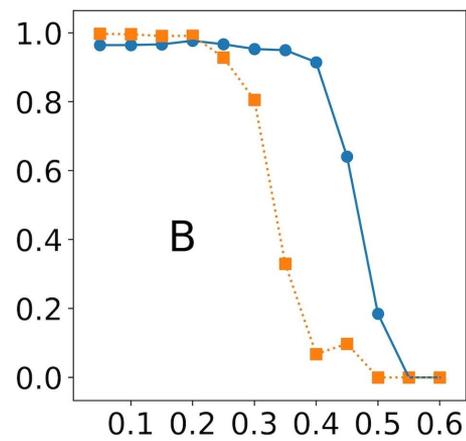
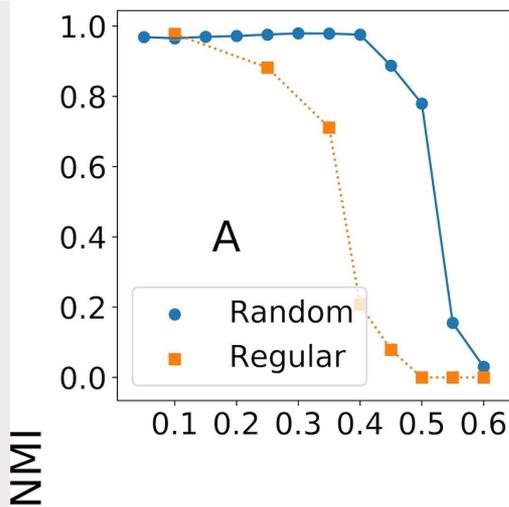
Mixing Parameter

Reduced LFR

- Small - 10-50 nodes/community
- Big - 20-100 nodes/community

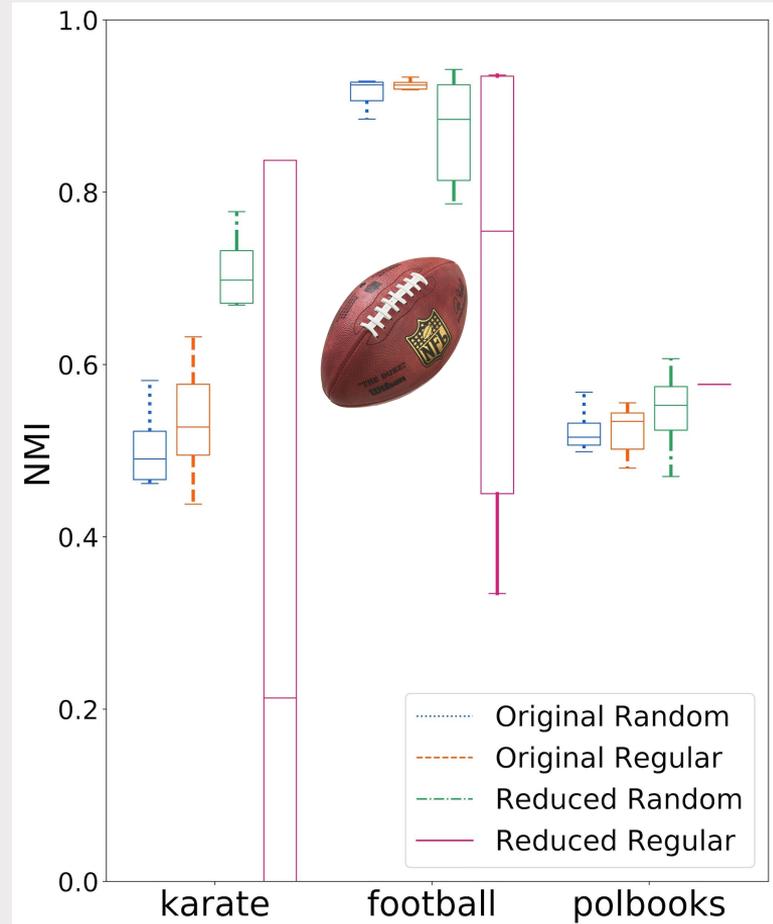
- A. 1000 Small
- B. 1000 Big
- C. 5000 Small
- D. 5000 Big

Mostly either completely perfect or completely imperfect recovery



Using Metadata

- Zachary Karate Club
- NCAA Football leagues
- US political books sold on Amazon during the 2004 election



Thank you!