

iNEAT: Incomplete Network Alignment

Presented by Si Zhang (ASU)



Si Zhang



Hanghang Tong



Jie Tang



Jiejun Xu

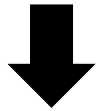


Wei Fan

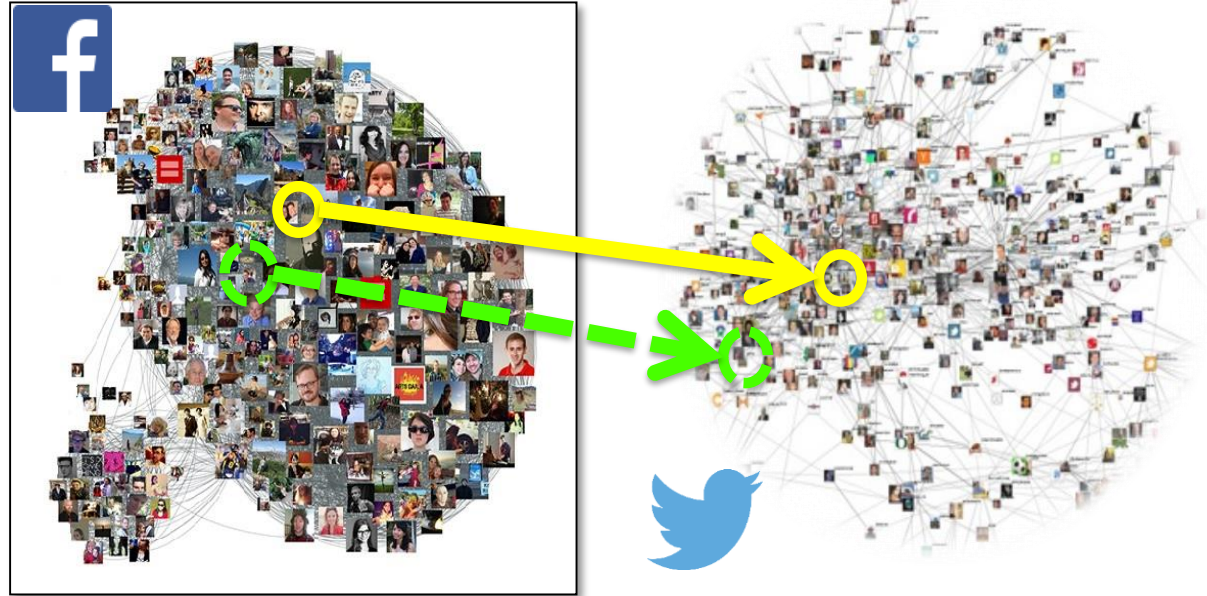
Why Incomplete Network Alignment?

- Networks are multi-sourced (**Variety**)

Network Alignment



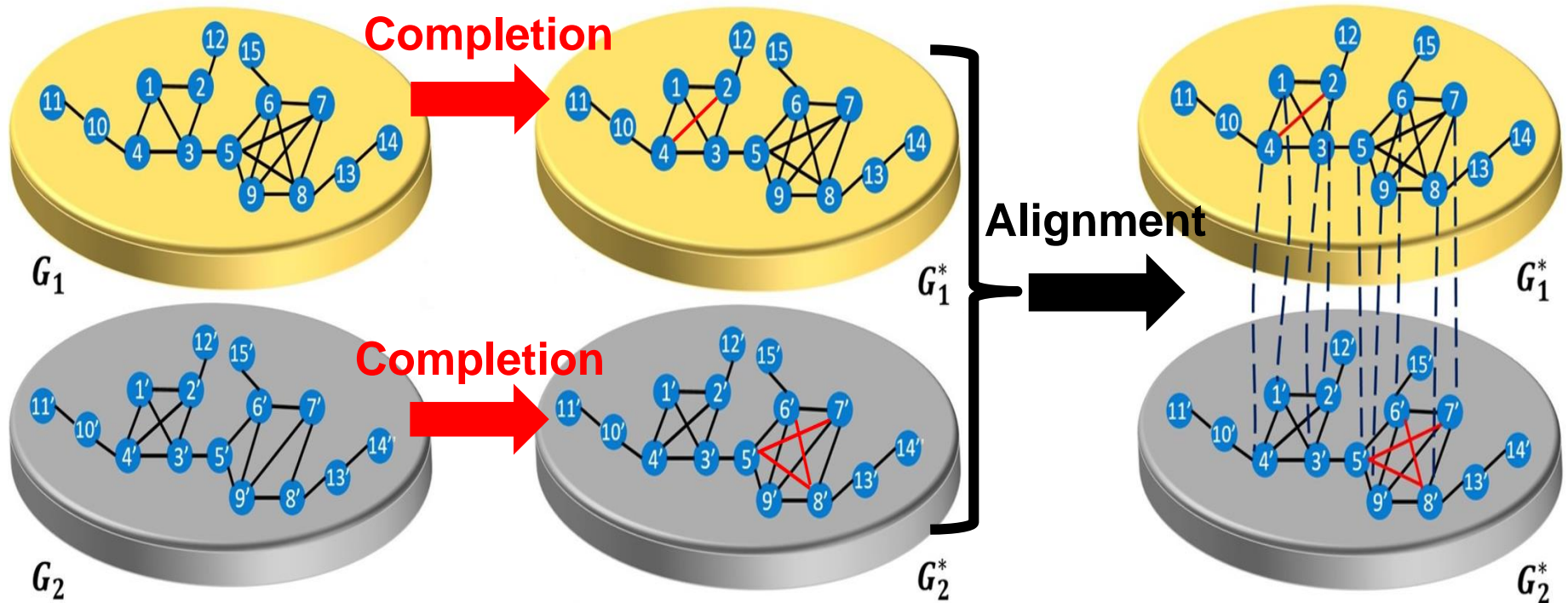
Find Someone
Like You



- Networks are often incomplete (**Veracity**)
 - Missing edges in multiple networks
- **Q:** how to align networks with missing edges?

How about Completion-Then-Alignment?

- A natural choice



- Completion ideally provides higher-quality networks

Completion-Then-Alignment: Limitations

- Limitation 1: alignment efficiency
 - Many alignment methods have $O(mn)$ complexity
of edges
 - Densified networks after completion intensify computation
 - **Q1:** how to make alignment and completion more efficient?
- Limitation 2: alignment accuracy
 - Potential introduced noisy links could affect alignment
 - **Q2:** how to complete networks to benefit alignment?

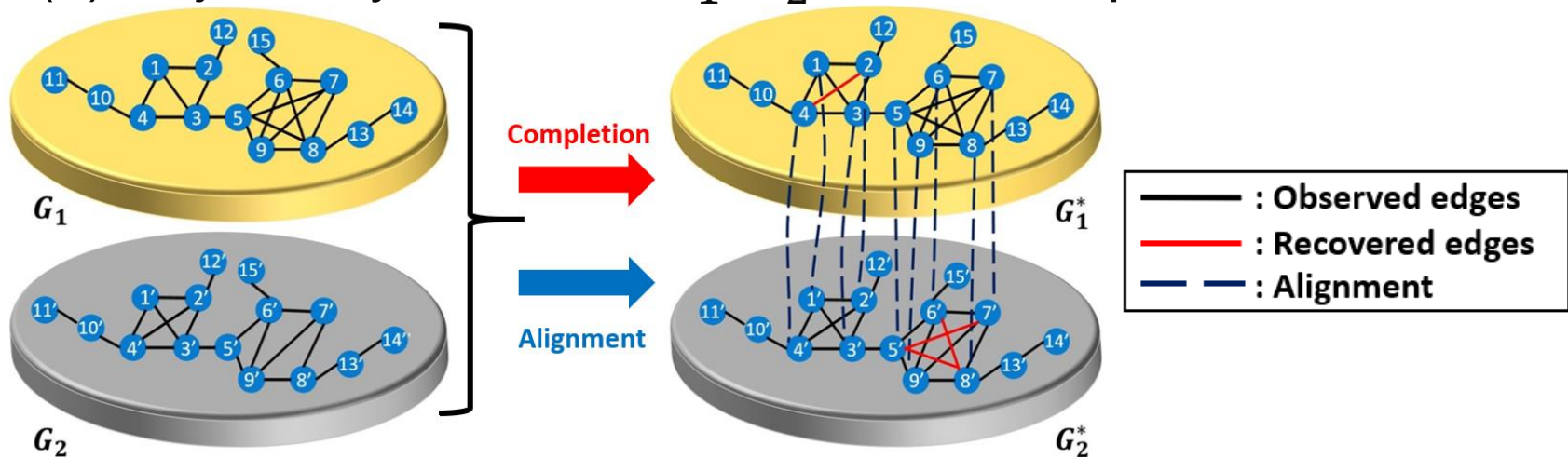
Prob. Def: Incomplete Network Alignment

■ Given:

- (1) Adjacency matrices A_1, A_2 of two incomplete networks
- (2) a prior alignment preference matrix H

■ Output:

- (1) Alignment matrix S
- (2) Adjacency matrices A_1^*, A_2^* of the complete networks



Preliminaries

- Network Alignment (FINAL, IsoRank)

$$\min_{\mathbf{s}} \alpha \mathbf{s}^T (\mathbf{D}_1 \otimes \mathbf{D}_2 - \mathbf{A}_1 \otimes \mathbf{A}_2) \mathbf{s} + (1 - \alpha) \|\mathbf{s} - \mathbf{h}\|_2^2$$

- Intuition: Alignment consistency
- Complexity: $O(mn)$ or $O(n^2)$ with approximation

- Network Completion (Factorization-Based)

$$\min_{\mathbf{U}_1, \mathbf{V}_1, \mathbf{U}_2, \mathbf{V}_2} J_1 = \frac{1}{2} \|P_{\Omega_1}(\mathbf{A}_1 - \mathbf{U}_1 \mathbf{V}_1^T)\|_F^2 + \frac{\lambda}{2} (\|\mathbf{U}_1\|_F^2 + \|\mathbf{V}_1\|_F^2) \\ + \frac{1}{2} \|P_{\Omega_2}(\mathbf{A}_2 - \mathbf{U}_2 \mathbf{V}_2^T)\|_F^2 + \frac{\lambda}{2} (\|\mathbf{U}_2\|_F^2 + \|\mathbf{V}_2\|_F^2)$$

within-network completion

- Intuition: low-rank characteristics of real networks

Zhang, Si, and Hanghang Tong. "FINAL: Fast Attributed Network Alignment." *KDD*. 2016.

Singh, Rohit, Jinbo Xu, and Bonnie Berger. "Global alignment of multiple protein interaction networks with application to functional orthology detection." *National Academy of Sciences*, (2008).

Rennie, Jasson DM, and Nathan Srebro. "Fast maximum margin matrix factorization for collaborative prediction." *ICML*. ACM, 2005.

Roadmap

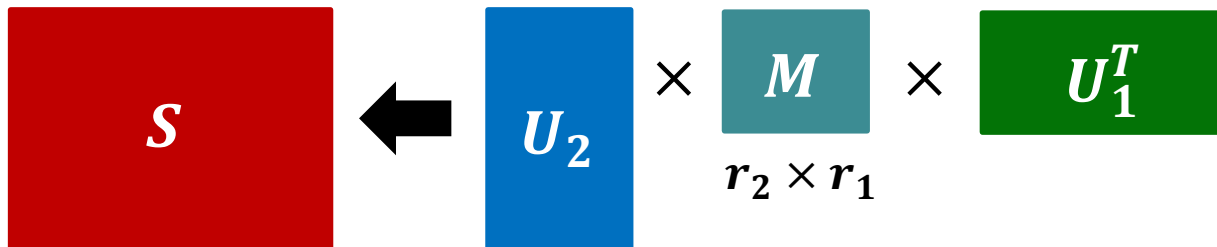
- Motivation and Background ✓
- Q1: Network Completion Helps Alignment
- Q2: Network Alignment Helps Completion
- iNEAT: Optimization Algorithm
- Experiments
- Conclusions

Network Completion Helps Alignment

- Alignment across complete networks

$$\min_s J_2 = \alpha \mathbf{s}^T (\mathbf{D} - \mathbf{A}_1^* \otimes \mathbf{A}_2^*) \mathbf{s} + (1 - \alpha) \|\mathbf{D}\mathbf{s} - \mathbf{h}\|_2^2$$

- $\mathbf{A}_1^* = \mathbf{U}_1 \mathbf{V}_1^T$, $\mathbf{A}_2^* = \mathbf{U}_2 \mathbf{V}_2^T$ are complete adjacency matrices
- Benefit: higher-quality of input networks
- Low-rank structure of alignment matrix
 - Low rank of networks \longrightarrow low rank of alignment matrix
 - $\mathbf{S} = \alpha \mathbf{U}_2 \mathbf{M} \mathbf{U}_1^T + (1 - \alpha) \mathbf{H}$ (proof in paper)
- Benefit: an alignment algorithm with a *linear* complexity



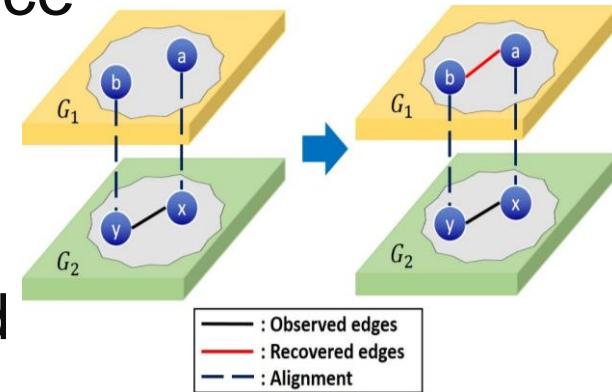
Roadmap

- Motivation and Background ✓
- Q1: Network Completion Helps Alignment ✓
- Q2: Network Alignment Helps Completion
- iNEAT: Optimization Algorithm
- Experiments
- Conclusions

Network Alignment Helps Completion

- Auxiliary confidence of edge existence

- (1) node-a and node-x are aligned
- (2) node-b and node-y are aligned
- (3) node-x and node-y are connected
- Infer: a potential edge (a, b)



$$A_1^*(a, b) \approx \sum_{x,y}^{n_2} S(a, b)S(b, y)A_2(x, y) = (S^T A_2 S)(a, b)$$

- Mathematically, we have

$$\begin{aligned}
 \min_{U_1, V_1, U_2, V_2, M} J_3 = & \frac{\beta}{2} \left\| P_{\bar{\Omega}_1} (U_1 V_1^T - U_1 M^T U_2^T A_2 U_2 M U_1^T) \right\|_F^2 \\
 & + \frac{\beta}{2} \left\| P_{\bar{\Omega}_2} (U_2 V_2^T - U_2 M U_1^T A_1 U_1 M^T U_2^T) \right\|_F^2
 \end{aligned}$$

cross-network completion

Roadmap

- Motivation and Background ✓
- Q1: Network Completion Helps Alignment ✓
- Q2: Network Alignment Helps Completion ✓
- **iNEAT: Optimization Algorithm**
- Experiments
- Conclusions

iNEAT: Optimization Algorithm

- Overall joint optimization problem

$$\min_{\mathbf{U}_1, \mathbf{V}_1, \mathbf{U}_2, \mathbf{V}_2, \mathbf{M}} J = J_1 + J_2 + J_3$$

$$s. t. \quad \mathbf{U}_1, \mathbf{V}_1, \mathbf{U}_2, \mathbf{V}_2, \mathbf{M} \geq \mathbf{0}$$

$$J_1 = \frac{1}{2} \|P_{\Omega_1}(\mathbf{A}_1 - \mathbf{U}_1 \mathbf{V}_1^T)\|_F^2 + \frac{\lambda}{2} (\|\mathbf{U}_1\|_F^2 + \|\mathbf{V}_1\|_F^2) + \frac{1}{2} \|P_{\Omega_2}(\mathbf{A}_2 - \mathbf{U}_2 \mathbf{V}_2^T)\|_F^2 + \frac{\lambda}{2} (\|\mathbf{U}_2\|_F^2 + \|\mathbf{V}_2\|_F^2)$$

within-network completion

$$J_2 = \alpha \mathbf{s}^T (\mathbf{D} - \mathbf{A}_1^* \otimes \mathbf{A}_2^*) \mathbf{s} + (1 - \alpha) \|\mathbf{D} \mathbf{s} - \mathbf{h}\|_2^2$$

Network alignment

$$J_3 = \frac{\beta}{2} \|P_{\bar{\Omega}_1}(\mathbf{U}_1 \mathbf{V}_1^T - \mathbf{U}_1 \mathbf{M}^T \mathbf{U}_2^T \mathbf{A}_2 \mathbf{U}_2 \mathbf{M} \mathbf{U}_1^T)\|_F^2 + \frac{\beta}{2} \|P_{\bar{\Omega}_2}(\mathbf{U}_2 \mathbf{V}_2^T - \mathbf{U}_2 \mathbf{M} \mathbf{U}_1^T \mathbf{A}_1 \mathbf{U}_1 \mathbf{M}^T \mathbf{U}_2^T)\|_F^2$$

cross-network completion

- Optimization Algorithm

- Block coordinate descent + Multiplicative update
- Mathematical details in paper
- Complexity: **linear** for both time and space complexity

- Output:

- (1) Alignment matrix $\mathbf{S} = \alpha \mathbf{U}_2 \mathbf{M} \mathbf{U}_1^T + (1 - \alpha) \mathbf{H}$;
- (2) Complete adjacency matrix $\mathbf{A}_1^* = \mathbf{U}_1 \mathbf{V}_1^T, \mathbf{A}_2^* = \mathbf{U}_2 \mathbf{V}_2^T$

Roadmap

- Motivation and Background ✓
- Q1: Network Completion Helps Alignment ✓
- Q2: Network Alignment Helps Completion ✓
- iNEAT: Optimization Algorithm ✓
- Experiments
- Conclusions

Experiments

■ Datasets

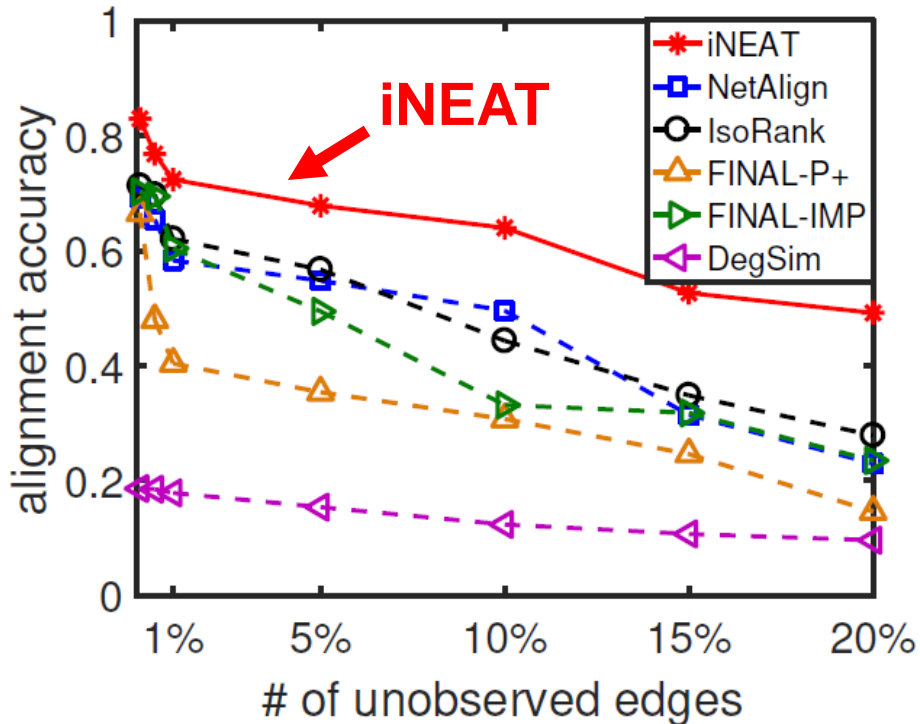
Category	Network	# of Nodes	# of Edges
Collaboration	GrQc	5,241	14,484
Infrastructure	Oregon	7,352	15,665
Social	Google+	23,628	39,194
Social	Youtube	1,134,890	2,987,624

■ Evaluation objectives

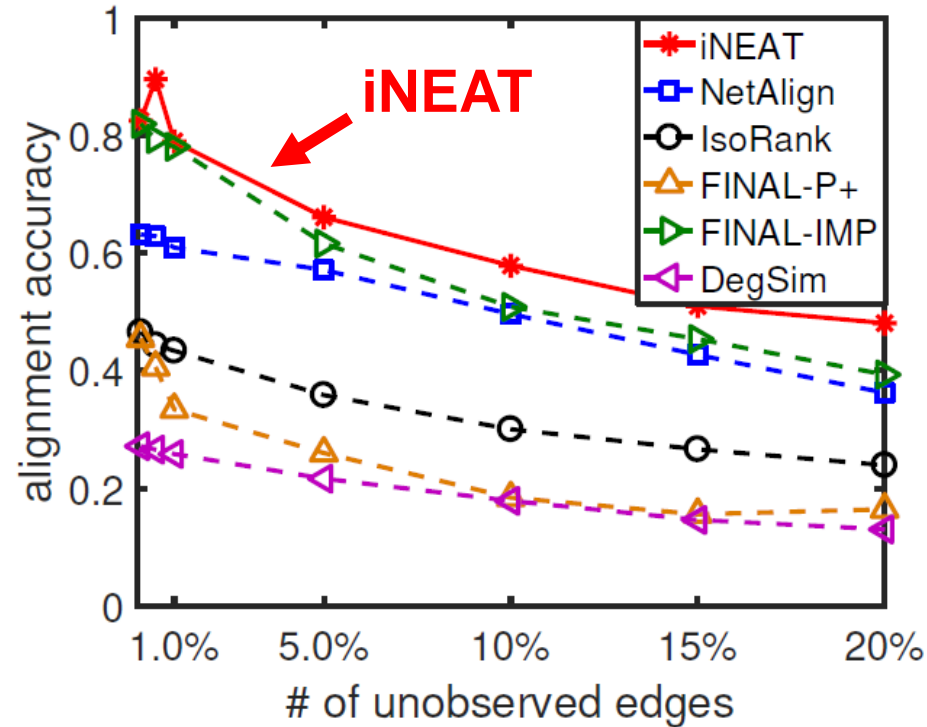
- Effectiveness of aligning the incomplete networks
- Effectiveness of multiple network completion
- Efficiency and scalability

Effectiveness of network alignment

Oregon infrastructure network



Google+ social network

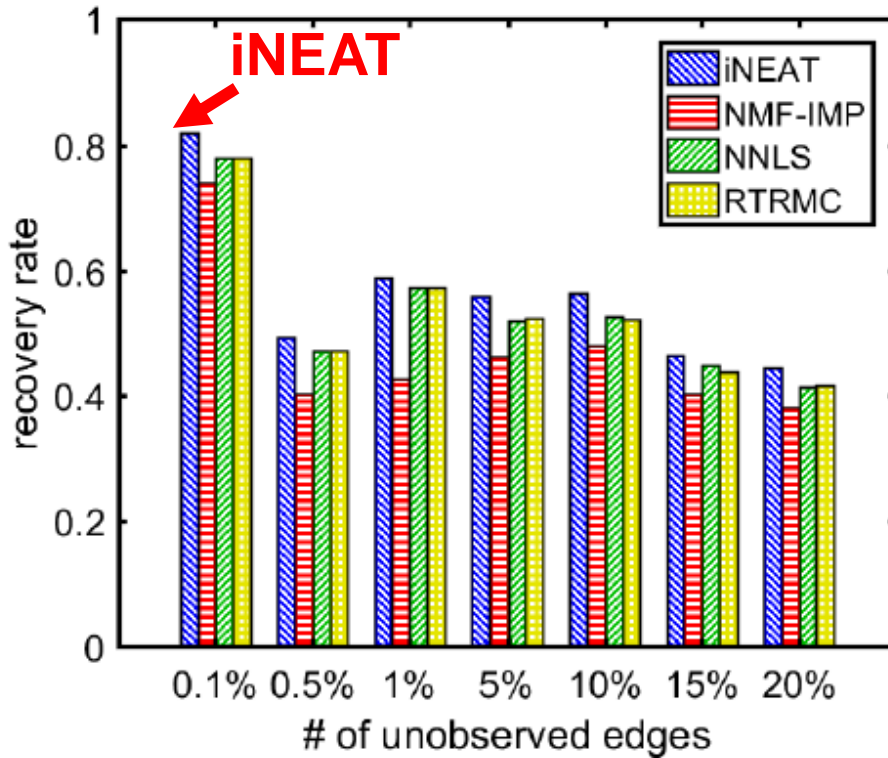


■ Observations:

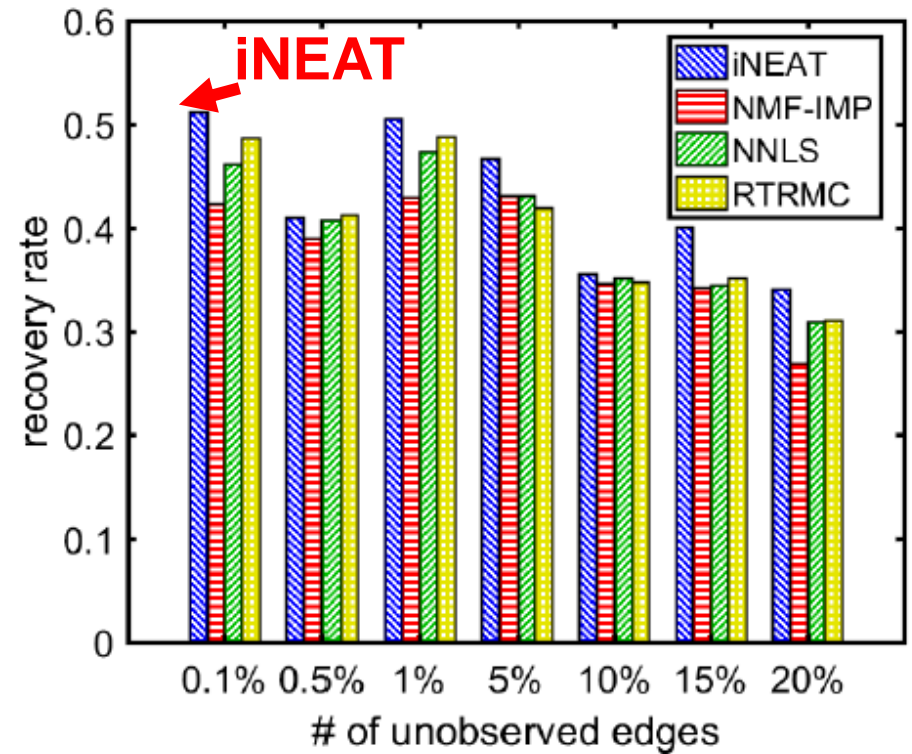
- (1) iNEAT achieves a better alignment accuracy
- (2) completion-then-alignment (FINAL-IMP) might be worse

Effectiveness of network completion

GrQc collaboration network



Google+ social network

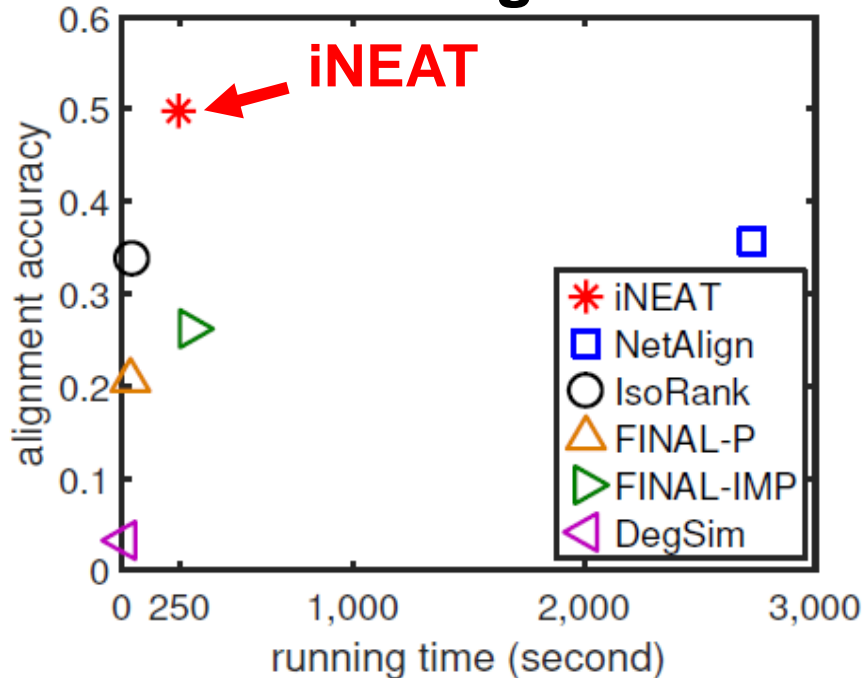


■ Observation:

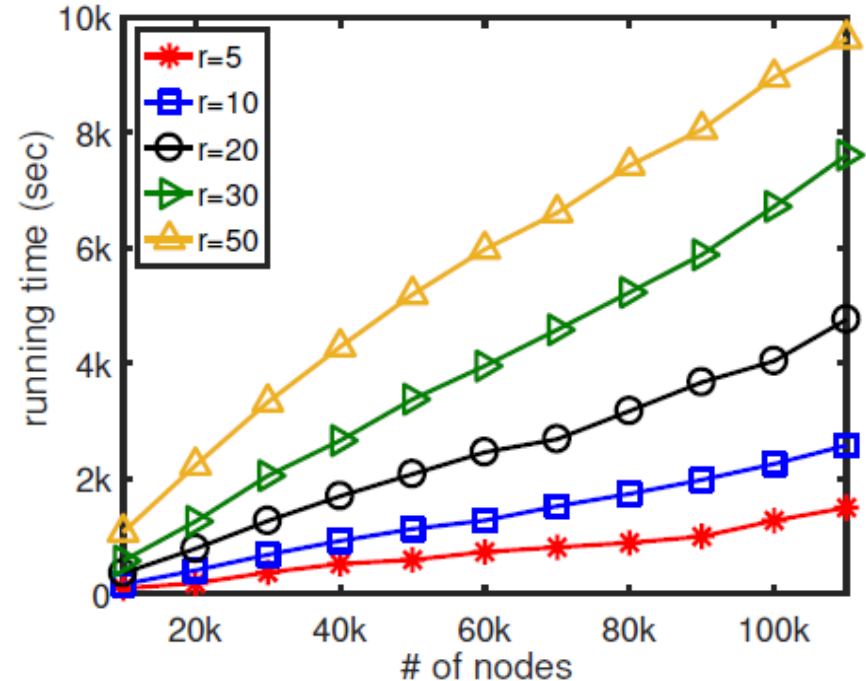
- Multiple networks completion is benefitted from alignment

Efficiency and Scalability

Balance between accuracy & running time



Scalability



Observations:

- (1) iNEAT has a better accuracy-time trade-off;
- (2) iNEAT has a **linear** complexity w.r.t # of nodes

Roadmap

- Motivation and Background ✓
- Q1: Network Completion Helps Alignment ✓
- Q2: Network Alignment Helps Completion ✓
- iNEAT: Optimization Algorithm ✓
- Experiments ✓
- **Conclusions**

Conclusions

- Incomplete Network Alignment
 - Q1: Alignment efficiency
 - A1: Network completion helps alignment
 - Q2: Alignment accuracy
 - A2: Network alignment helps completion
- Results
 - Better network alignment and completion performance
 - Fast, and *linear* complexity
- More in paper
 - Details of the optimization algorithm
 - Proof of low-rank structure of alignment matrix

Thank You!