

XDATA PI meeting

Spectral Clustering for Divide-and-Conquer Graph Matching

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Abstract

We present a parallelized bijective graph matching algorithm that leverages seeds and is designed to match very large graphs. Our algorithm combines spectral graph embedding with existing state-of-the-art seeded graph matching procedures. We justify our approach by proving that modestly correlated, large stochastic block model random graphs are correctly matched utilizing very few seeds through our divide-and-conquer procedure. We also demonstrate the effectiveness of our approach in matching very large graphs in simulated and real data examples.



V. Lyzinski, D.L. Sussman, D.E. Fishkind, H. Pao, L. Chen,
J.T. Vogelstein, Y. Park, C.E. Priebe,
“Spectral Clustering for Divide-and-Conquer Graph Matching,”
Parallel Computing, accepted for publication, 2015.



V. Lyzinski

Background

Given two graphs, $G_1 = (V_1, E_1)$ and $G_2 = (V_2, E_2)$, the Graph Matching Problem (GMP) seeks an alignment between the vertex sets V_1 and V_2 that best preserves structure across the graphs. In *bijective* graph matching, we further assume $|V_1| = |V_2| = n$, and the alignment sought by GMP is a bijection between V_1 and V_2 .

Graph Matching Problem

Find a bijection $\psi : V_1 \rightarrow V_2$ minimizing the quantity

$$|\{(i, j) \in V_1 \times V_1 \text{ s.t. } [i \sim_{G_1} j, \psi(i) \not\sim_{G_2} \psi(j)] \text{ or } [i \not\sim_{G_1} j, \psi(i) \sim_{G_2} \psi(j)]\}|, \quad (1)$$

i.e. the problem seeks to minimize the number of edge disagreements between G_2 and “ $\psi(G_1)$ ”. Equivalently stated, if A and B are the respective adjacency matrices of G_1 and G_2 , then this problem seeks to minimize $\|A - PBP^T\|_F^2$, over all permutation matrices $P \in \Pi(n) := \{n \times n \text{ permutation matrices}\}$, with $\|\cdot\|_F$ the matrix Frobenius norm.

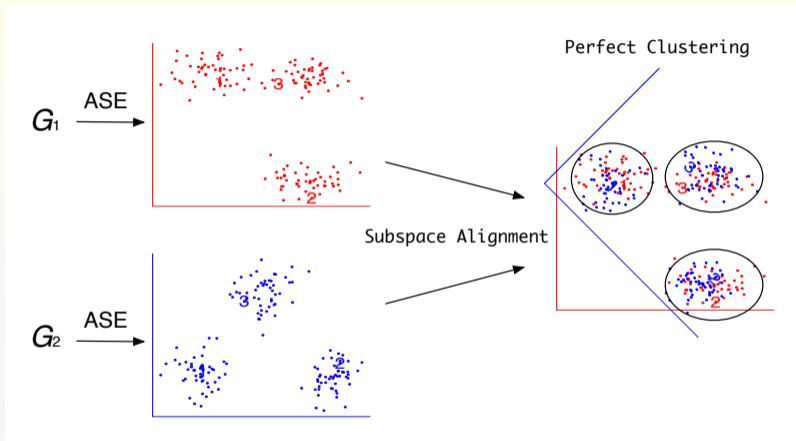
Background

In the seeded graph matching problem (SGMP), we further assume the presence of a latent alignment ϕ between the vertex sets of G_1 and G_2 . Our task is to then efficiently leverage the information in a partial observation of the latent alignment, i.e. a *seeding*, to estimate the remaining latent alignment.

Seeded Graph Matching Problem

Given subsets of the vertices $S_1 \subset V_1$ and $S_2 \subset V_2$ called *seeds* with $|S_1| = |S_2| = s$ and a bijective seeding function $\phi_S : S_1 \rightarrow S_2$, the task is to use ϕ_S to estimate ϕ by finding the bijection extending ϕ_S which minimizes (1).

Divide-and-Conquer Seeded Graph Matching



$$\Omega(C_{i,1}, G_1) \xleftrightarrow{SGM} \Omega(C_{i,2}, G_2) \Rightarrow \psi^{(i)}$$

$$\psi = \bigoplus_{i=1}^k \psi^{(i)}$$

Theorems

Theorem 1: Perfect Clustering




[EJS2014]

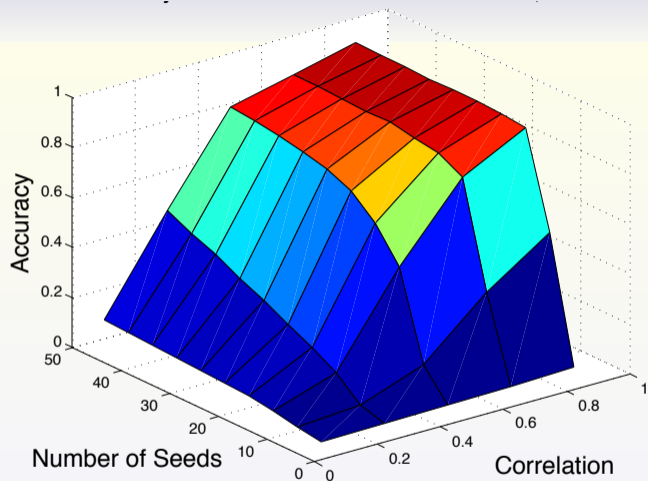
Theorem 2: Seeded Graph Matching

[JMLR2014]

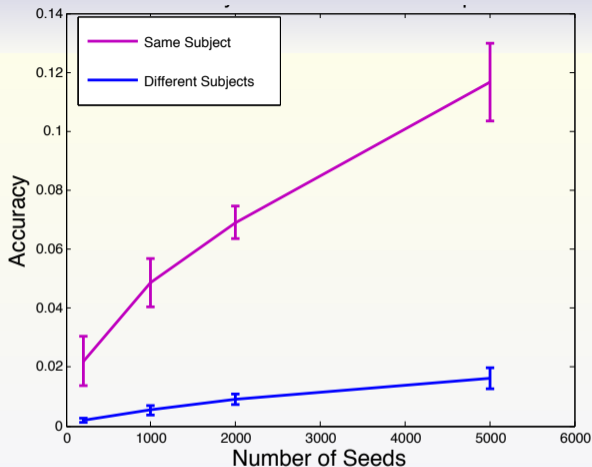
Theorem 3: Subspace Alignment

[PARCO2015]

-  Vince Lyzinski, Daniel Sussman, Minh Tang, Avanti Athreya, Carey E. Priebe, "Perfect Clustering for Stochastic Blockmodel Graphs via Adjacency Spectral Embedding," *Electronic Journal of Statistics*, accepted for publication, 2014.
-  Vince Lyzinski, Donniell E. Fishkind, and Carey E. Priebe, "Seeded graph matching for correlated Erdos-Renyi graphs," *Journal of Machine Learning Research*, vol. 15, no. Nov, pp. 3513-3540, 2014.
-  V. Lyzinski, D.L. Sussman, D.E. Fishkind, H. Pao, L. Chen, J.T. Vogelstein, Y. Park, C.E. Priebe, "Spectral Clustering for Divide-and-Conquer Graph Matching," *Parallel Computing*, accepted for publication, 2015.



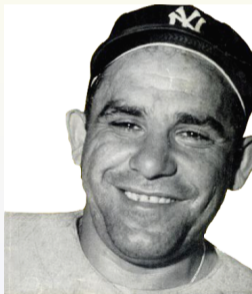
Fraction of unseeded vertices correctly matched across two $K = 900$ block, $\vec{n} = 30 \cdot \vec{1}$, $d = 10$ dimensional ρ -correlated SBM's with s seeds drawn uniformly at random from the 27000 vertices.



The fraction of the unseeded vertices correctly matched for graphs 8 and 29 (within-subject) and for graphs 1 and 8 (across-subject). For the 8–29 pair, $n = 20,541$, $d = 30$. For the 1–8 pair, $n = 18,694$, $d = 30$, we cluster using k -means, reclustering any clusters of size ≥ 800 . We plot the fraction of the vertices correctly matched in each of the two experiments for number of seeds $s = 200, 1000, 2000$, and 5000 .

Yogi Berra:

*"In theory there is no difference between theory and practice.
In practice, there is."*



Leopold Kronecker to Hermann von Helmholtz (1888):

*“The wealth of your practical experience
with sane and interesting problems
will give to mathematics
a new direction and a new impetus.”*



Kronecker



Helmholtz