Popularity Adjusted Block Models are Generalized Random Dot Product Graphs

JSM Speed Presentation

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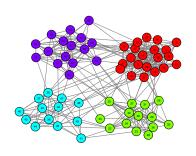


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Community Detection for Networks

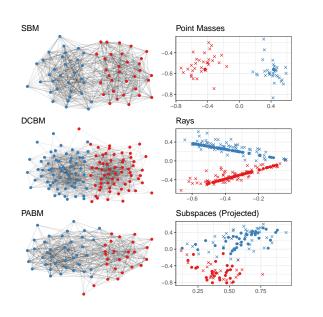


Def Popularity Adjusted Block Model (Sengupta and Chen, 2017):

Let each vertex $i \in [n]$ have K popularity parameters $\lambda_{i1},...,\lambda_{iK} \in [0,1].$

Then $A \sim \mathsf{BernoulliGraph}(P)$ is a PABM if each $P_{ij} = \lambda_{iz_j}\lambda_{jz_i}$

Connecting Block Models to the GRDPG



- K-means clustering
- Gaussian mixture models
- K-means with cosine similarity
- GMM on angles

• ???

Orthogonal Spectral Clustering

Theorem (KTT): If $P = V\Lambda V^{\top}$ and $B = nVV^{\top}$, then $B_{ij} = 0$ if $z_i \neq z_j$.

Algorithm: Orthogonal Spectral Clustering:

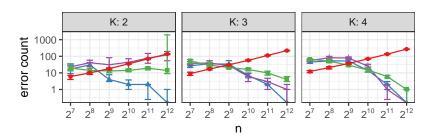
- 1. Let V be the eigenvectors of A corresponding to the K(K+1)/2 most positive and K(K-1)/2 most negative eigenvalues.
- 2. Compute $B = |nVV^{\top}|$ applying $|\cdot|$ entry-wise.
- 3. Construct graph G using B as its similarity matrix.
- 4. Partition G into K disconnected subgraphs.

Theorem (KTT): \forall pairs (i,j) belonging to different communities, $\max_{i,j} B_{ij} = O_P \Big(\frac{(\log n)^c}{\sqrt{n\rho_n}} \Big).$

Corollary: OSC results in zero clustering error as $n \to \infty$, with probability 1.

Simulation Study

- Modularity Maximization
- Orthogonal Spectral Clustering
- Sparse Subspace Clustering on Adj. Matrix
- Sparse Subspace Clustering on ASE



Thank you

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arXiv preprint: https://arxiv.org/abs/2109.04010

GitHub repository: https://github.com/johneverettkoo/pabm-grdpg

R package: https://github.com/johneverettkoo/osc