

# **Analytic Theory of Power Law Graphs**

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# Outline



- B<sup>⊗K</sup> Graphs
- (B+I)<sup>⊗K</sup> Graphs

Introduction

• Summary

- Kronecker Graphs
- Graphs as Matrices
- Algorithm Comparison



### Power Law Modeling of Kronecker Graphs



- Real world data (internet, social networks, ...) has connections on all scales (i.e power law)
- Can be modeled with Kronecker Graphs:  $G^{\otimes k} = G^{\otimes k-1} \otimes G$ 
  - Where "⊗"denotes the Kronecker product of two matrices



### **Graphs as Matrices**



- Graphs can be represented as a sparse matrices
  - Multiply by adjacency matrix  $\rightarrow$  step to neighbor vertices
  - Work-efficient implementation from sparse data structures
- Most algorithms reduce to products on semi-rings: C = A "+"."x" B
  - "x" : associative, distributes over "+"
  - − □"+" : associative, commutative
  - Examples: +.\* min.+ or.and



# **Algorithm Comparison**

Algorithm (Problem)	Canonical Complexity	Array-Based Complexity	Critical Path (for array)
Bellman-Ford (SSSP)	<i>©</i> (mn)	<i>©</i> (mn)	<i>Θ</i> ( <i>n</i> )
Generalized B-F (APSP)	NA	<i>©</i> ( <i>n</i> <sup>3</sup> log <i>n</i> )	<i>Θ</i> (log <i>n</i> )
Floyd-Warshall (APSP)	<i>©</i> ( <i>n</i> <sup>3</sup> )	<i>©</i> ( <i>n</i> <sup>3</sup> )	<i>Θ</i> ( <i>n</i> )
Prim (MST)	<i>©</i> ( <i>m</i> + <i>n</i> log <i>n</i> )	<i>Θ</i> ( <i>n</i> <sup>2</sup> )	<i>Θ</i> ( <i>n</i> )
Borůvka (MST)	$\Theta(m \log n)$	<i>Θ</i> ( <i>m</i> log <i>n</i> )	<i>Θ</i> (log <sup>2</sup> <i>n</i> )
Edmonds-Karp (Max Flow)	<i>©</i> ( <i>m</i> <sup>2</sup> <i>n</i> )	<i>Θ</i> ( <i>m</i> <sup>2</sup> <i>n</i> )	<i>®</i> (mn)
Push-Relabel (Max Flow)	⊖(mn²)	<i>O</i> ( <i>mn</i> <sup>2</sup> )	?
	(or <i>©</i> ( <i>n</i> <sup>3</sup> ))		
Greedy MIS (MIS)	$\Theta(m+n \log n)$	$\Theta(mn+n^2)$	$\Theta(n)$
Luby (MIS)	<i>®</i> ( <i>m</i> + <i>n</i> log <i>n</i> )	$\Theta(m \log n)$	$\Theta(\log n)$

Majority of selected algorithms can be represented with array-based constructs with equivalent complexity.

$$n = |V|$$
 and  $m = |E|$ .)

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# Outline

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- Definitions
- Bipartite Graphs
- Degree Distribution



Kronecker Product

- Let B be a N<sub>B</sub>xN<sub>B</sub> matrix
- Let C be a N<sub>C</sub>xN<sub>C</sub> matrix
- Then the Kronecker product of B and C will produce a  $N_B N_C x N_B N_C$  matrix A:

$$A = B \otimes C = \begin{pmatrix} b_{1,1}C & b_{1,2}C & \dots & b_{1,M_B}C \\ b_{2,1}C & b_{2,2}C & \dots & b_{2,M_B}C \\ \vdots & \vdots & & \vdots \\ b_{N_B,1}C & b_{N_B,2}C & \dots & b_{N_B,M_B}C \end{pmatrix}$$

Kronecker Graph (Leskovec 2005 & Chakrabati 2004)

- Let G be a NxN adjacency matrix
- Kronecker exponent to the power k is:

$$G^{\otimes k} = G^{\otimes k-1} \otimes G$$



# **Types of Kronecker Graphs**

#### **Explicit**

• G only 1 and 0s

#### **Stochastic**

 G contains probabilities

#### **Instance**

 A set of M points (edges) drawn from a stochastic





# **Kronecker Product of a Bipartite Graph**



- Fundamental result [Weischel 1962] is that the Kronecker product of two complete bipartite graphs is two complete bipartite graphs
- More generally

$$B(n_1, m_1) \otimes B(n_2, m_2) \stackrel{P}{=} B(n_1 n_2, m_1 m_2) \cup B(n_2 m_1, n_1 m_2)$$



### Degree Distribution of Bipartite Kronecker Graphs

 Kronecker exponent of a bipartite graph produces many independent bipartite graphs

$$B(n,m)^{\otimes k} \stackrel{P}{=} \bigcup_{r=0}^{k-1} \bigcup_{r=0}^{\binom{k-1}{r}} B(n^{k-r}m^r, n^r m^{k-r})$$

Only k+1 different kinds of nodes in this graph, with degree distribution

$$Count[Deg = n^r m^{k-r}] = \binom{k}{r} n^{k-r} m^r$$



#### **Explicit Degree Distribution**





# **Instance Degree Distribution**



 An instance graph drawn from a stochastic bipartite graph is just the sum of Poisson distributions taken from the explicit bipartite graph



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- (B+I)<sup>⊗K</sup> Graphs
- Summary

- Bipartite + Identity Graphs
- Permutations and substructure
- Degree Distribution
- Iso Parametric Ratio



- Bipartite Kronecker graphs highlight the fundamental structures in a Kronecker graph, but
  - Are not connected (i.e. many independent bipartite graphs)
- Adding identity matrix creates connections on all scales
  - Resulting explicit graph has diameter = 2
  - Sub-structures in the graph are given by

$$(B+I)^{\otimes k} \stackrel{P}{=} \sum_{r=1}^{k} {}^{"} {\binom{k}{r}} {}^{"} {\underset{l}{\bigcup}} {}^{N^{k-1}} B^{\otimes k}$$

- Where "" indicates permutations are required to add the matrices
- Sub-structure can be revealed by applying permutation that "groups" vertices by their bipartite sub-graph



# **Bipartite Permutation**



- Left: unpermuted  $(B+I)^{\otimes 4}$  kronecker graph
- Right: permuted (B+I)<sup>⊗4</sup> kronecker graph



# **Identifying Substructure**



• Permuting specific terms shows their contributions to the graph



# **Quantifying Substructure**



 Connections between bipartite subgraphs are the Kronecker product of corresponding 2x2 matrices, e.g. B(1,1)<sup>84</sup>8I(2)



# **Substructure Degree Distribution**



- Only k+1 different kinds of nodes in this graph, with same degree distribution, only differing values of vertex degree
- $(B+I)^{\otimes k}$  is steeper than  $B^{\otimes k}$



### **Example Result: Iso-Parametric Ratio**



- Iso-parametric ratios measure the "surface" to "volume" of a sub-graph
- Can analytically compute for a Kronecker graph: (B+I)<sup>&k</sup>
- Shows large effect of including "half" or "all" of bipartite sub-graph



### Kronecker Graph Theory -Summary of Current Results-

Quantity	Graph: B(n,m) <sup>⊗k</sup>	Graph: (B+I) <sup>⊗k</sup>	
Degree Distribution	$Count[Deg = n^r m^{k-r}] = \binom{k}{r} n^{k-r} m^r$	$Count[Deg = (n+1)^r (m+1)^{k-r}] = \binom{k}{r} n^{k-r} m^r$	
Betweenness Centrality	$Count[C_b = (n/m)^{2r-k}(n^{k-r}m^r - 1)] = \binom{k}{r}n^k$	$r^{-r}m^{r}$	
Diameter	$Diam(B^{\otimes k}) = \infty$	$Diam((B+I)^{\otimes k}) = 2$	
Eigenvalues	$eig(B(n,m)^{\otimes k}) = \{\overbrace{(nm)^{k/2},, (nm)^{k/2}}^{2^{k-1}}, \overbrace{-(nm)^{k/2},, -(nm)^{k/2}}^{2^{k-1}}\}$		
	$eig((B+I)^{\otimes k}) = \{((nm)^{1/2}+1)^k, ((nm)^{1/2}+1)^{k-1}, ((nm)^{1/2}-1)^2((nm)^{1/2}+1)^{k-2}, \ldots\}$		
Iso-parametric Ratio "half"	$IsoPar(n_k(i)) = \infty$	$IsoPar(n_k(i)) = 2(n+1)^{k-r}(m+1)^r - 2$	
Iso-parametric Ratio "all"	$IsoPar(n_k(i) \cup m_k(i)) = 0$ $IsoPar(n_k(i) \cup m_k(i)) = 0$	$)) = 2\frac{n^{r}m^{k-r}(n+1)^{k-r}(m+1)^{r} + n^{k-r}m^{r}(n+1)^{r}(m+1)^{k-r}}{2n^{k}m^{k} + n^{r}m^{k-r} + n^{k-r}m^{r} + [\chi \text{ terms}]} - 2$	



# Reference

- Book: "Graph Algorithms in the Language of Linear Algebra"
- Editors: Kepner (MIT-LL) and Gilbert (UCSB)
- Contributors
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  - Chakrabart (CMU)
  - Dunlavy (Sandia)
  - Faloutsos (CMU)
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  - Kahn (MIT-LL & Brown)
  - Kegelmeyer (Sandia)
  - Kepner (MIT-LL)
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Graph Algorithms in the Language of Linear Algebra

Jeremy Kepner and John Gilbert (editors)

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