

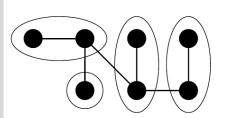


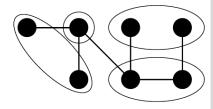
Bounds for the Number of Partitions of a Graph

ECDA 2019 - Bayreuth

Fabian Ball and Andreas Geyer-Schulz | March 18, 2019

INFORMATION SERVICES AND ELECTRONIC MARKETS





Outline



- 1. Motivation
- 2. Preliminaries
- 3. Lower Bound
- 4. Small Graphs
- 5. Estimated Number of k-Partitions
- 6. Conclusion



- We all know the following combinatorial results
 - 1. The number of *k*-partitions of an *n*-set is the Stirling number of the 2nd kind S(n,k) (or $\binom{n}{k}$)
 - 2. The number of partitions of an n-set is the Bell number, which is the sum over all $1 \le k \le n$ Stirling numbers (2nd kind)
- Both numbers grow rapidly with increasing n (and "intermediate" k)
- But what about graph clustering?
- Are all partitions actually part of the solution space?
- Independent of the optimization goal, we normally only want connected nodes within clusters
- Donath and Hoffman (1973) present lower bounds for a special application (cluster sizes are restricted to be smaller than a threshold



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Partition of a Set

- Set M of some entities (e.g. data points)
- Partition $P(M) = \{M_1, ..., M_k\}$ into subsets/cells M_i
 - Complete: $\bigcup_i M_i = V$
 - Disjoint: $\forall i \neq j : M_i \cap M_i = \emptyset$
 - No empty subsets: $\forall i : M_i \neq \emptyset$



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Number of Partitions

- "How many possibilities exist to divide n elements into a partition of $k \le n$ subsets?"
- Equivalent to the question: "How large is the search space of kmeans?" (for given k)
- Answer: S(n, k) (the Stirling numbers of the 2nd kind) (e.g. Rennie and Dobson, 1969)



- Defined recursively (non-recursive form exists, but not of importance here)
- $S(n,k) = k \cdot S(n-1,k) + S(n-1,k-1)$
- S(n,1) = S(n,n) = 1
- Explanation:
 - Shift perspective, add a new element (the n + 1th)
 - $S(n+1,k) = k \cdot S(n,k) + S(n,k-1)$
 - Either put the new element into one of the k existing cells; for each choice there exist S(n, k) possibilities
 - Or put the element into a new cell itself (the kth one), all other n elements must then be partitioned into k-1 cells
- The sum over all 1 < k < n is the Bell number B(n) (Bell, 1934)



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Simple Graph

- G = (V, E)
- Node set $V = \{1, 2, ..., n\}$ $(n < \infty)$
- Edge set $E \subseteq \{\{u,v\} \mid u,v \in V, u \neq v\}$
- ⇒ Undirected, finite, no loops, no weights, no multiple edges + connected

Graph Partition

- Defined as partitions above
- The set to be partitioned is the node set *V*
- Assumption: The subgraph induced by a cell is connected (i. e. we never have nodes in the same cell if they are not connected by a path within the cell)



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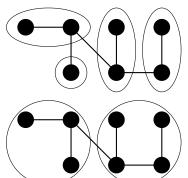
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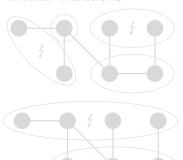
2 Preliminaries – Graph Partition Assumption Examples



"Valid" Partitions



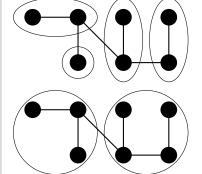
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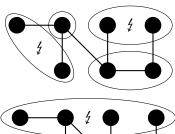
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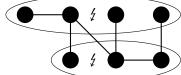


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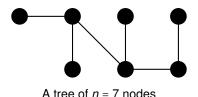
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Fact: A tree of n nodes and m = n - 1 edges is the sparsest possible connected graph

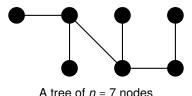


Theorem (Number of Partitions of a Tree

The number of possible partitions of a tree is $2^m = 2^{n-1}$



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A tree of H = 7 modes

Theorem (Number of Partitions of a Tree)

The number of possible partitions of a tree is $2^m = 2^{n-1}$.



Proof.

By induction.

$$\mathbf{n} = \mathbf{1} \ 2^{1-1} = 1$$
, the trivial and the singleton partition are the same

 $n = 2 2^{2-1} = 2$, there is the trivial and the singleton partition

$$\mathbf{n} = \mathbf{3} \ 2^{3-1} = \mathbf{4}$$
, trivial + singleton + 2× connected pairs

$$\mathbf{n} + \mathbf{1} \ 2^{(n+1)-1} = 2^n = 2 \cdot 2^{n-1} = 2^{n-1} + 2^{n-1}$$

- (a) is added as a singleton cell or
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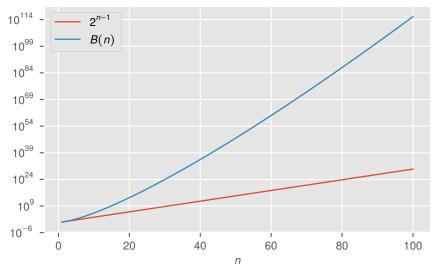
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3 Lower Bound vs. Upper Bound (Partitions)





Motivation

Preliminaries

Lower Bound

Small Graphs

Estimated Number of k-Partitions



Theorem (Number of *k*-Partitions of a Tree)

The number of possible partitions of a tree into k cells is $\binom{n-1}{k-1}$.

Proof.

- Two cells are formed by picking (removing) an edge that connects the two subtrees
- $1 \le k \le n$ cells are formed by picking k 1 unique edges
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Alternate Proof for the Number of Partitions

- Using the binomial theorem $(x + y)^n = \sum_{k=0}^n \binom{n}{k} x^{n-k} y^k$
- For x = y = 1:

$$2^{n} = \sum_{k=0}^{n} \binom{n}{k} = \sum_{k=1}^{n+1} \binom{n}{k-1}$$
 (1)

$$2^{n-1} = \sum_{k=1}^{n} \binom{n-1}{k-1} \tag{2}$$



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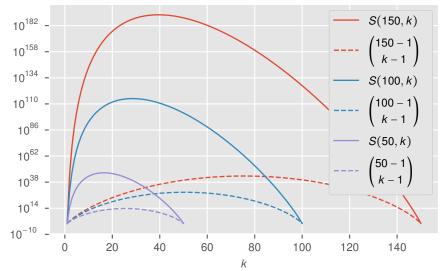
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Motivation Preliminaries Lower Bound Small Graphs Estimated Number of k-Partitions Conclusion

3 Lower Bound vs. Upper Bound (k-Partitions)





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3 Lower Bound - Conclusion



- The number of clustering partitions of a tree are the exact lower bound for the number of partitions
- \blacksquare Still exponential, but had to be expected (argument: graph clustering is $\mathcal{NP}\text{-hard})$

4 Small Graphs - Overview



- Small connected graphs with n = 2, ..., 13 nodes
- Downloaded from http://www.graphclasses.org/smallgraphs.html
- Includes *all* graphs with n = 2, ..., 5 nodes and some graphs with n = 6, ..., 13
- Total number of investigated graphs is 501
- We enumerated the full search space (given our assumption of connected nodes within clusters) for all these graphs
- Already (very) time and memory consuming for "larger" n (in the range of several minutes and more than 40 GB)
- Graph with the largest search space: $\overline{X_{196}}$ (n = 13; m = 39 # Partitions is 4,880,943, but B(13) = 27,644,437)

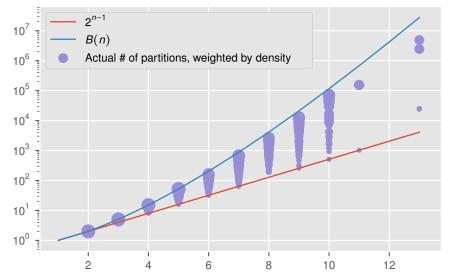
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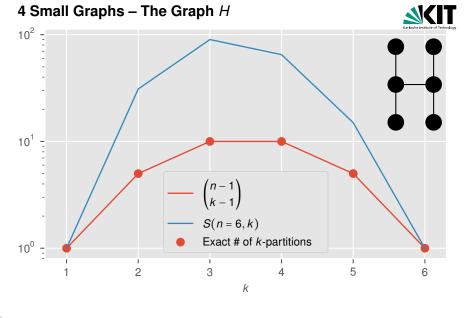
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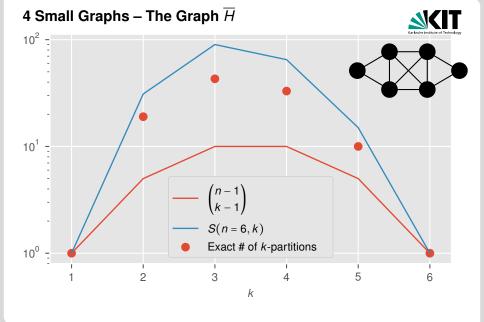


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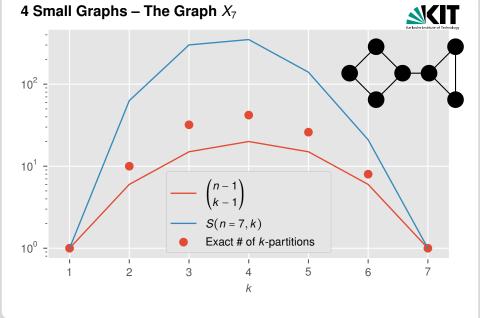


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Remark: Everything that follows is work in progress...

General Ideas

- 1. Modify the recursive formula for the Stirling numbers (2nd kind), so that it takes adjacency information into account
- 2. Weight the number of *k*-partitions by the graph density
- 3. Estimate the number of *k*-partitions as a mixture of lower and upper bound

Additional Observations:

- Number of ways to partition a graph into n-1 cells is $m \le \binom{n}{2}$
- If the graph is complete, then $m = \binom{n}{2} = S(n, n-1)$



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Modify the Formula

Reminder: $S(n, k) = k \cdot S(n-1, k) + S(n-1, k-1)$

- The idea stays the same, but the adjacency structure restricts *k*
- $S_G(n, k; m) = \overline{d}(n, k, m) \cdot S_G(n-1, k; m) + S_G(n-1, k-1; m)$
- $\overline{d}(n, k, m) \le k$ is the expected average number of neighbors (the degree) of a node
- Normally, especially in real-world graphs, the average number of neighbors is much smaller than k

Issues

- How can \overline{d} be estimated? (mean number of neighbors is $\frac{2m}{n}$)
- How does *m* change subject to *n*?
- The recursive idea of "adding" an additional element does no address where to add a node (and it depends!)

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Weight by Density

- We observed a decreased actual number of partitions if less edges exist
- The density $\rho = m/\binom{n}{2}$ is a measure of how many edges exist in contrast to a fully connected graph
- The actual number of graph k-partitions could be estimated as $\rho(G_{n_i=k}) \cdot S(n,k)$
- $G_{n_i=k}$ is a coarsened graph that results from the formation of clusters:

$$\rho = \frac{4}{6} \qquad \qquad \rho(G_{n_i=3}) = \frac{2}{3}$$

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As before: How to estimate the number of edges with decreasing n'

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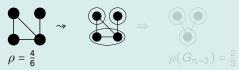
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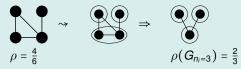
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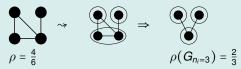
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Issues

As before: How to estimate the number of edges with decreasing n?



Estimate Mixture of Lower / Upper Bound

- The absolute lower (as presented earlier) and upper (2nd kind Stirling numbers) bounds are known
- The true number of *k*-partitions lies between these bounds
- This true number can be written as the linear combination $\# Partitions(n, k) = \lambda \cdot lb(n, k) + (1 \lambda) \cdot ub(n, k) = \lambda \cdot \binom{n-1}{k-1} + (1 \lambda) \cdot S(n, k)$ for some parameter $\lambda \in [0, 1]$
- The estimator for λ can possibly be based on several observable graph parameters (e. g. node degree distribution)

Issues

- How to estimate λ ?
- The obvious values n and m are not sufficient, as there can exist various graphs with the same n and m but a different number of partitions

Motivation

March 18, 2019



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- This true number can be written as the linear combination $\# \text{Partitions}(n, k) = \lambda \cdot \text{lb}(n, k) + (1 \lambda) \cdot \text{ub}(n, k) = \lambda \cdot \binom{n-1}{k-1} + (1 \lambda) \cdot S(n, k) \text{ for some parameter } \lambda \in [0, 1]$
- The estimator for λ can possibly be based on several observable graph parameters (e. g. node degree distribution)

Issues

- How to estimate λ ?
- The obvious values n and m are not sufficient, as there can exist various graphs with the same n and m but a different number of partitions

Motivation



Estimate Mixture of Lower / Upper Bound

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6 Conclusion I



What is the essence of this talk?

- Clustering/Partitioning of a graph (w.r.t. the assumptions) did not get any easier
- However, the number of edges in comparison to the number of nodes (c.f. density) has a large impact on the actual number of possible partitions (see also Good et al. (2010), who argue on the performance of modularity clustering)
- The theoretical results and insights could help for the analysis of clustering strategies and algorithms

March 18, 2019

6 Conclusion II



What's next?

- The shown results are for graphs, which have explicit relations
- Could be applied for clustering in metric spaces as well if certain assumptions/restrictions are introduced (e.g. maximal allowed distance between the nearest data points in a cluster)
- Not mentioned here: (Graph) Symmetry restricts the search space as well (Ball, 2019)

Conjecture

As large real-world graphs are generally very sparse, the actual number o clustering partitions (the size of the search space) asymptotically approaches the lower bound.

6 Conclusion II



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March 18, 2019

Thank you for your attention!

Find the slides of this talk and additional material on Github: https://github.com/KIT-IISM-EM/ECDA2019



References I



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References II



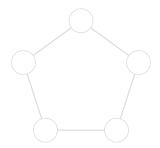


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8 2-Regular Graphs I



- We have proven an exact bound for trees
- Another quite restricted class are regular graphs:
 - The degree of a node is the number of neighbors
 - In a *k*-regular graph, every node has exactly *k* neighbors
 - Special case: 2-regular graphs are determined only by the number of nodes n ≥ 3



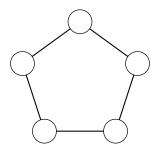
The 2-regular graph with n = 5 nodes

References 2-Regular Graphs

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The 2-regular graph with n = 5 nodes

References 2-Regular Graphs

8 2-Regular Graphs II



Theorem (Number of k-Partitions of a 2-regular graph)

The number of partitions of a 2-regular graph into k parts is $\binom{n}{n-k}$ for k > 1. For k = 1 it is 1.

Proof

- For k = 1 there clearly exists only the trivial partition that consists of one large cell containing all nodes.
- For $1 < k \le n$ choosing n k out of n edges can be thought of removing the edge and merging the two incident nodes:
 - Every merger means to merge two cells
 - The result is a coarsened 2-regular graph of k nodes
 - Each node represents a cell of the partition
 - How many ways exist to pick n k edges?
 - \Rightarrow It is just $\binom{n}{n-k}$

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References

8 2-Regular Graphs III



Hint

- The binomial coefficient has the property $\binom{n}{n-k} = \binom{n}{k}$
- Choosing $\binom{n}{k}$ edges means to separate the 2-regular graph into k cells (for k > 1; c.f. our argument for partitioning trees)
- To actually separate the graph into two disconnected subgraphs, we must remove two edges
- This is why picking $\binom{n}{n-1} = \binom{n}{1} = n$ is not the correct solution of the number of partitions into k = 1 cells

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8 2-Regular Graphs IV



Theorem (Number of Partitions of a 2-regular graph)

The number of partitions of a 2-regular graph is $2^n - n$.

Proof

- We use the binomial theorem again: $(x + y)^n = \sum_{k=0}^n \binom{n}{k} x^{n-k} y^k$ for x = y = 1
- $2^{n} = \sum_{k=0}^{n} {n \choose k} = \sum_{k=0}^{n} {n \choose n-k} = \underbrace{\binom{n}{n}}_{=1} + \underbrace{\binom{n}{n-1}}_{=n} + \underbrace{\sum_{k=2}^{n} \binom{n}{n-k}}_{\text{divide into } k=2,...,n} \text{ certains}$
- The first term represents the number of partitions into k = 1 cells
- The second term is "too much"

$$\Rightarrow 2^n - \binom{n}{n-1} = 2^n - n$$



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2-Regular Graphs