

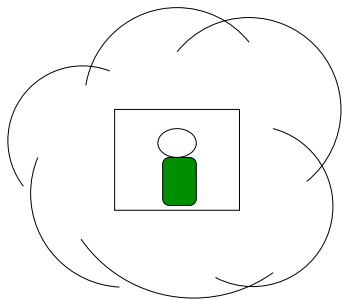
Capacitated Metric Labeling

Ankur Moitra, MIT

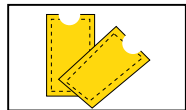
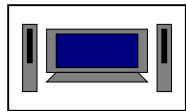
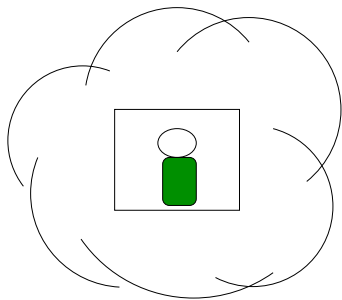
joint work with Matthew Andrews, MohammadTaghi Hajiaghayi and Howard Karloff

January 24, 2011

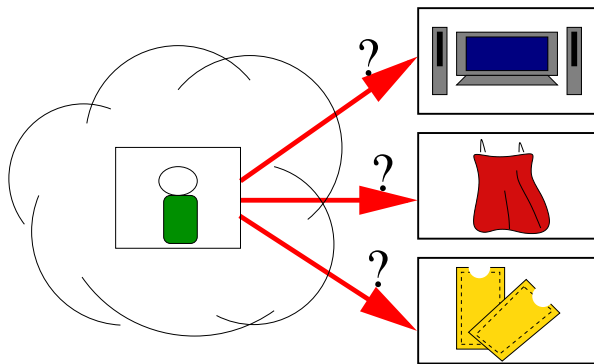
Classification (on a Social Network)



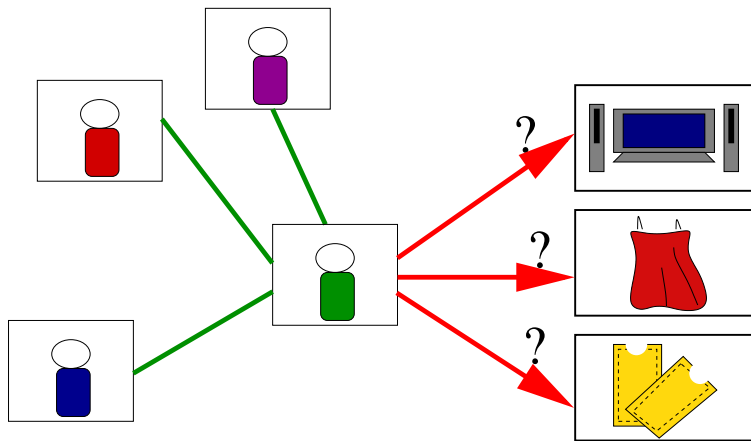
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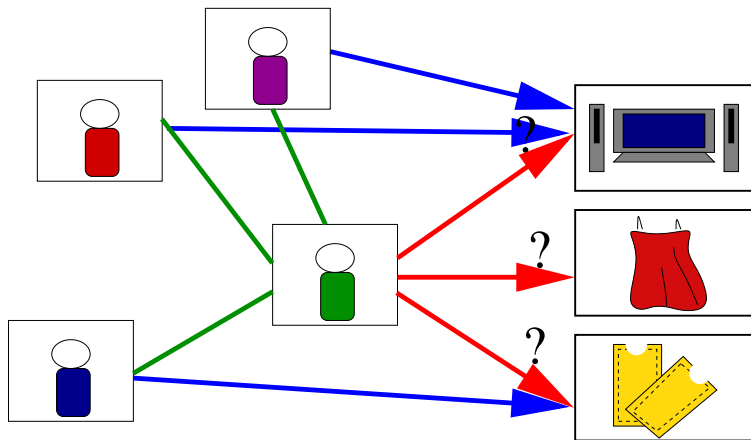
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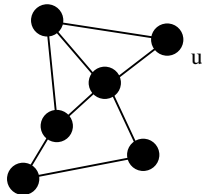


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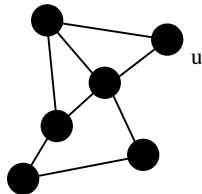
Metric Labeling

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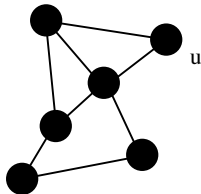


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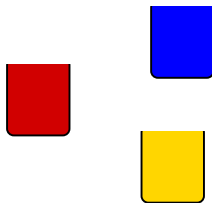


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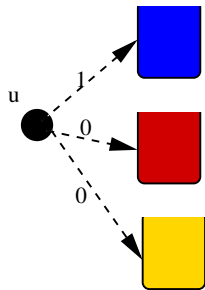
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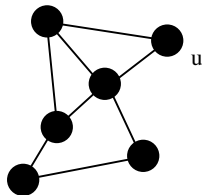


$\phi: V \times L \rightarrow \mathbb{R}^+$

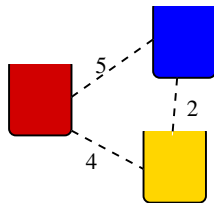


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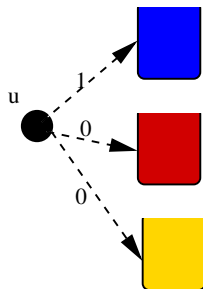
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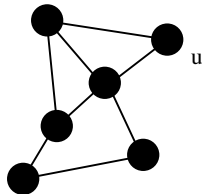


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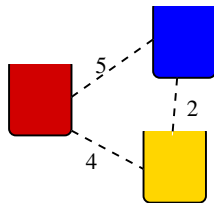


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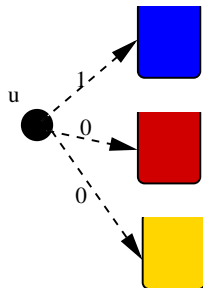
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assignment cost + $\underbrace{\text{pair-wise cost}}_{\text{capacity} \times \text{distance}}$

Metric Labeling Problem: (introduced by Kleinberg and Tardos)

$$\min_{f:V \rightarrow L} \sum_{u \in V} \phi(u, f(u)) + \sum_{(u,v) \in E} w(u,v) d_L(f(u), f(v))$$

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Applications to classification problems in statistical physics, biometry, machine vision, ...

Also encodes MAP estimation problem for Markov Random Fields

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There is a polynomial time $O(\log k)$ -approximation algorithm for METRIC LABELING

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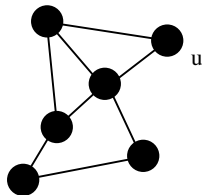
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Problem

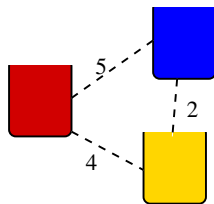
What if an approximation algorithm returns a highly imbalanced (balanced) solution, and our goal is a balanced (imbalanced) solution?

Capacitated Metric Labeling

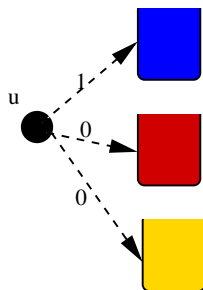
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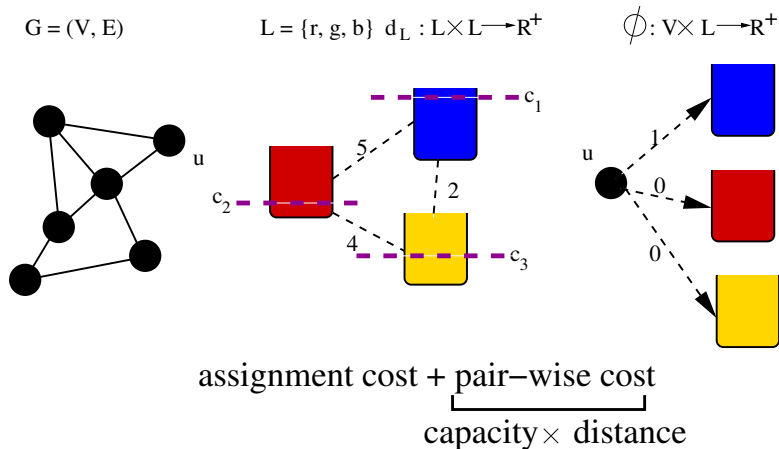


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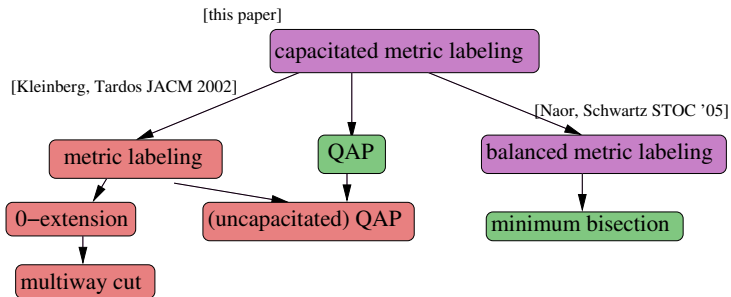


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Capacitated Metric Labeling



Optimization



Our Results

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For $k = O(1)$, there is a polynomial time $O(\log n)$ -approximation algorithm for CAPACITATED METRIC LABELING

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Definition

The congestion an instance I of CAPACITATED METRIC LABELING is the minimum value of C so that scaling the label capacities up by a factor of C has a zero cost solution

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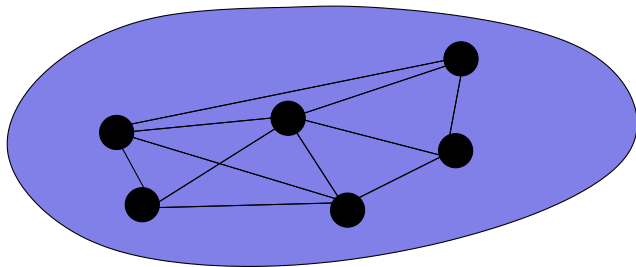
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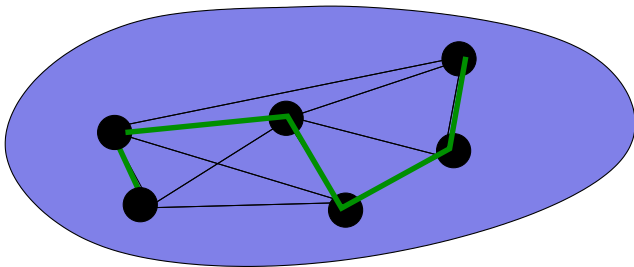
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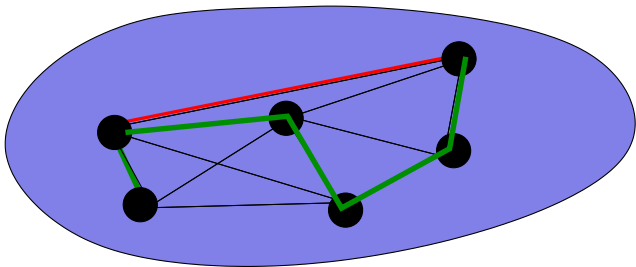
Hierarchical Decompositions



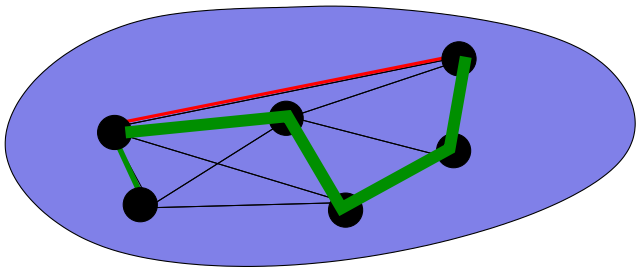
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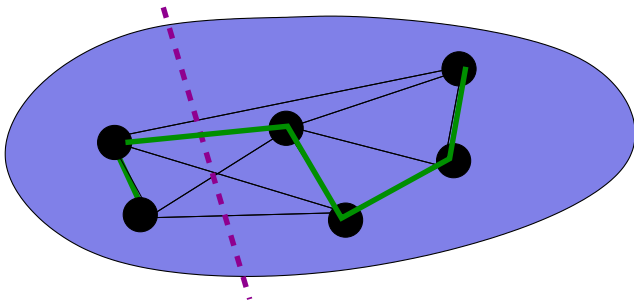
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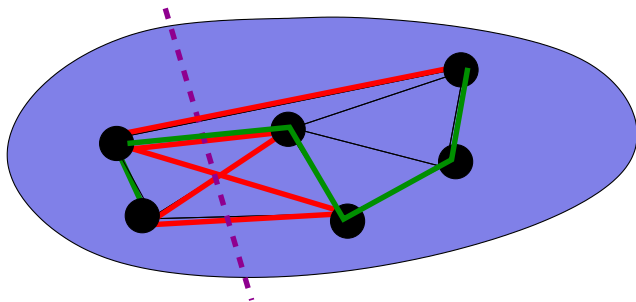
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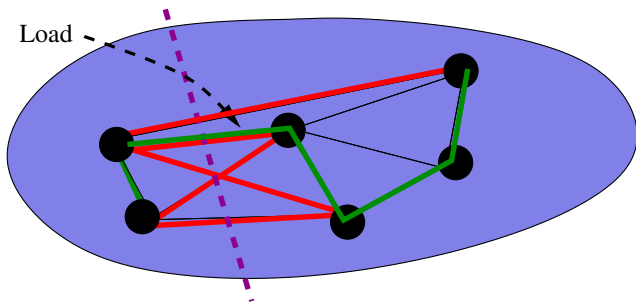
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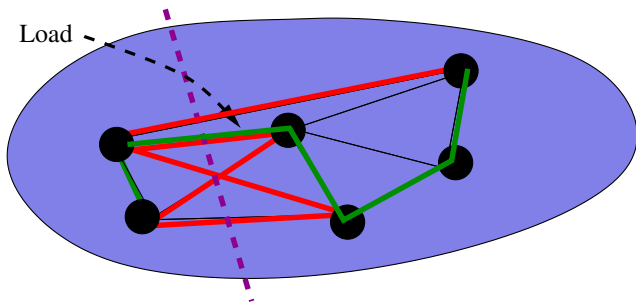


Hierarchical Decompositions

Theorem (Räcke)

There is a distribution μ on decomposition trees so that for all edges,

$$E_{T \leftarrow \mu}[\text{load}_T(e)] \leq O(\log n)w(e)$$



Rounding to a Tree

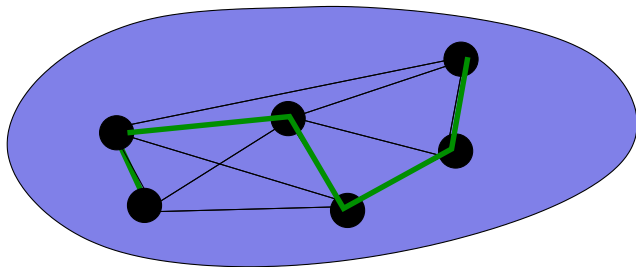
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$$COST(f, G) \leq COST(f, T)$$

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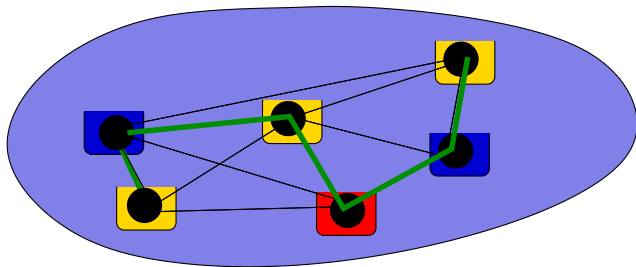
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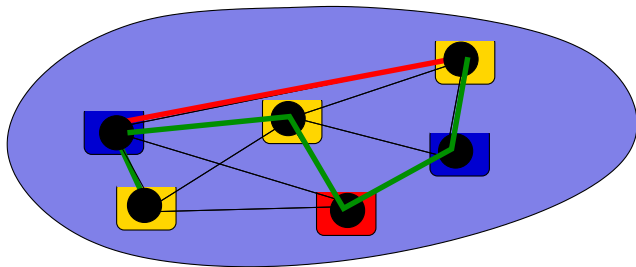
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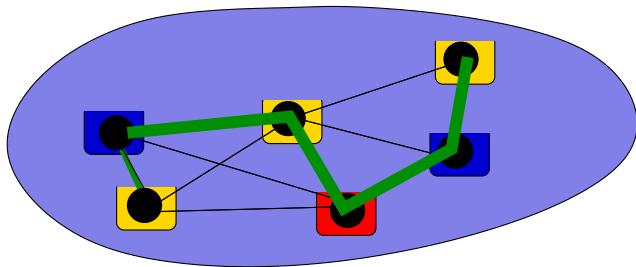
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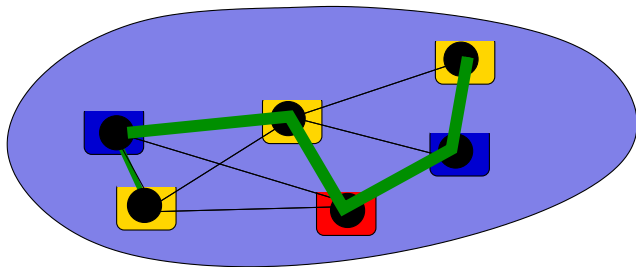
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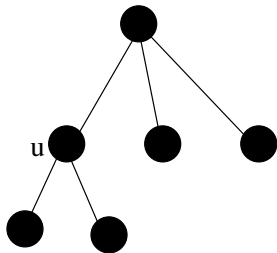
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$COST(f, G) \leq COST(f, T)$ and $E_{T \leftarrow \mu}[COST(f, T)] \leq O(\log n)COST(f, G)$



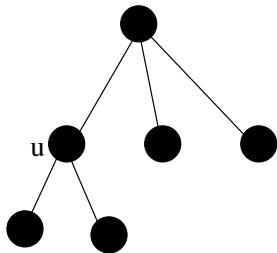
Dynamic Programming

T



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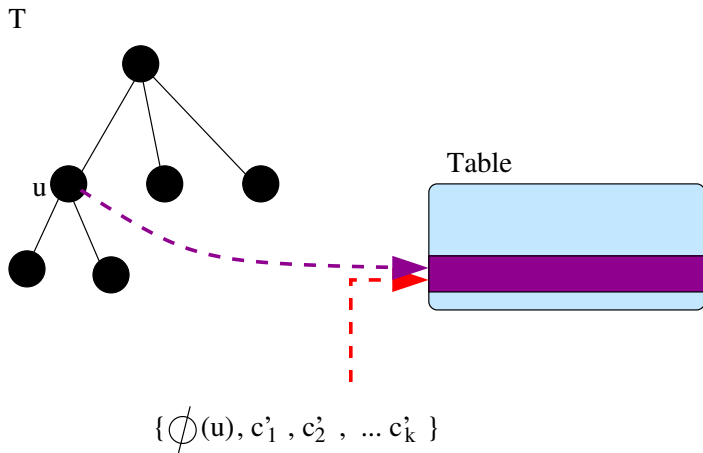
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Table



Dynamic Programming



Our Results

Theorem

For $k = O(1)$, there is a polynomial time $O(\log n)$ -approximation algorithm for CAPACITATED METRIC LABELING

... this is the regime of interest for many classification problems

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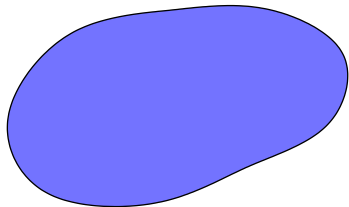
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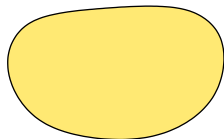
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Approximating Congestion

$G = (V, E)$

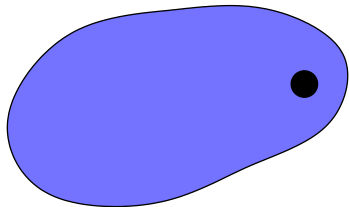


$L \quad d_L$

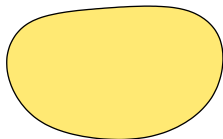


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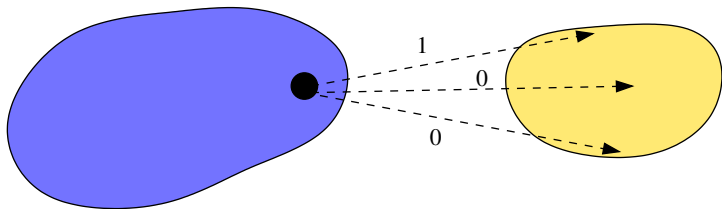
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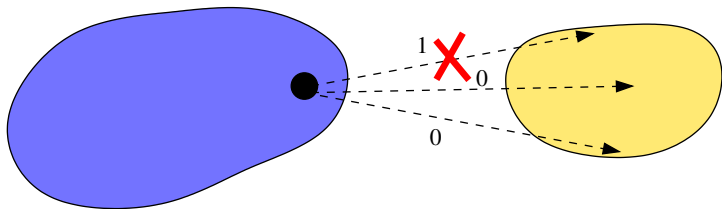
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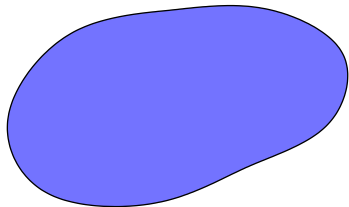
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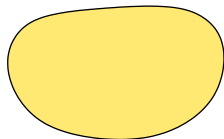


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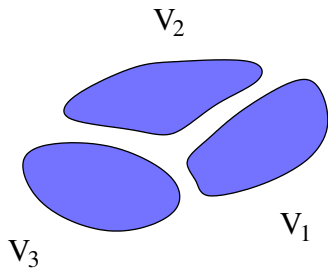


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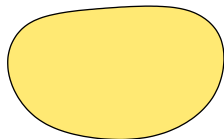


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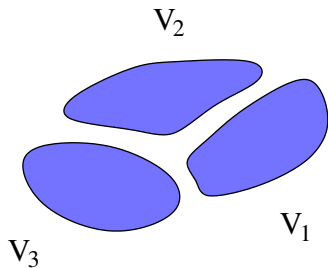


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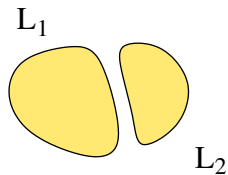


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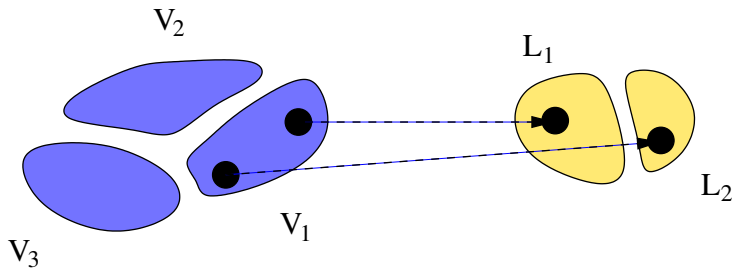
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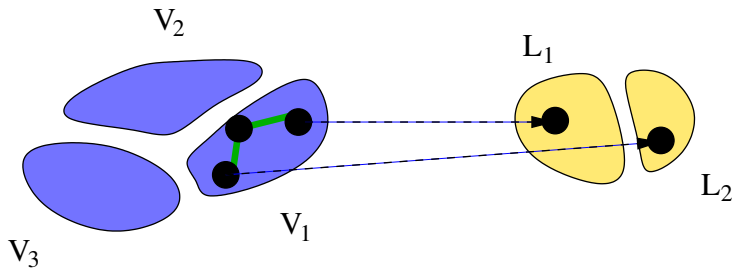
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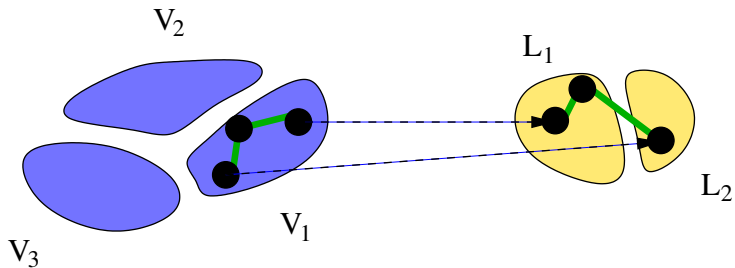
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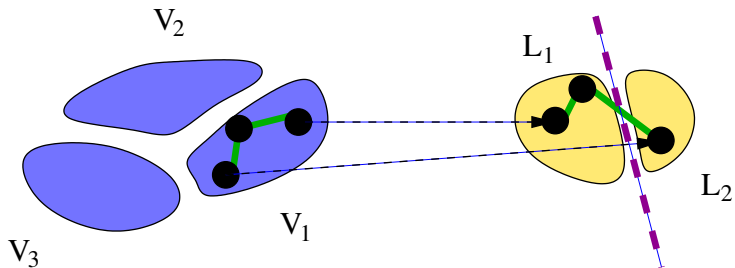
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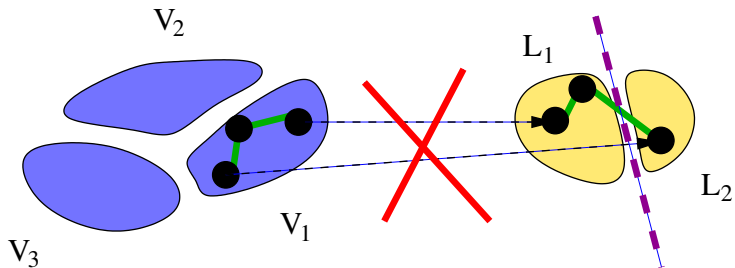
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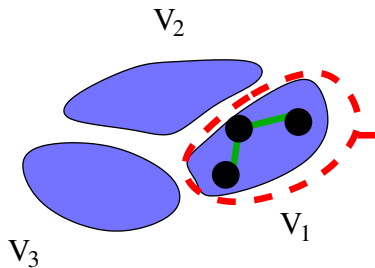
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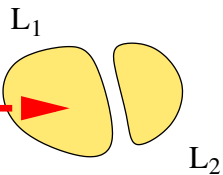


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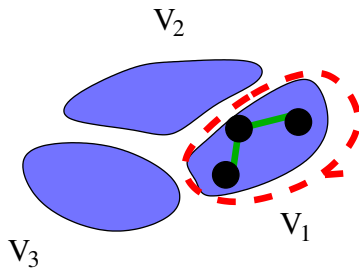


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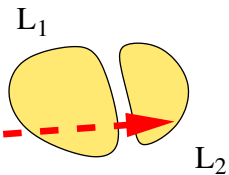


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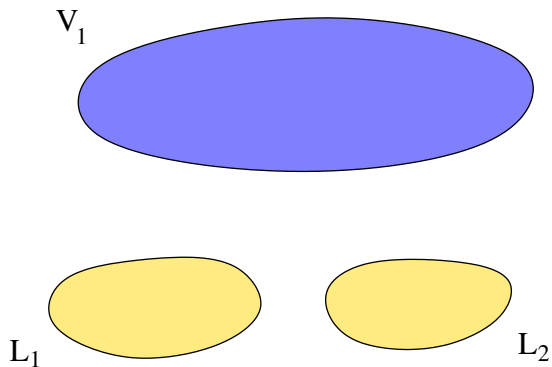
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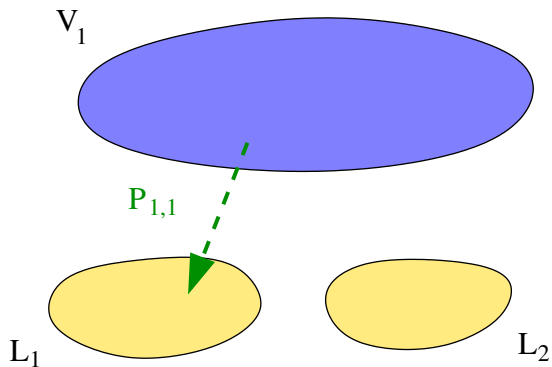
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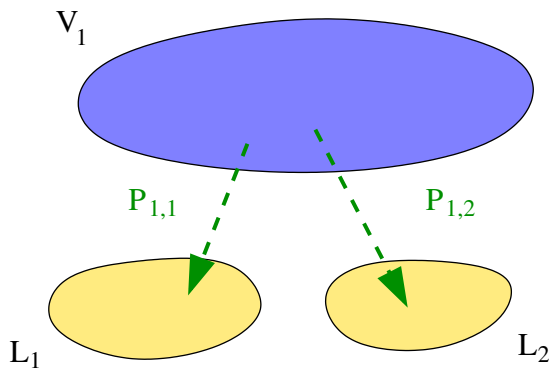
Conditional Probability



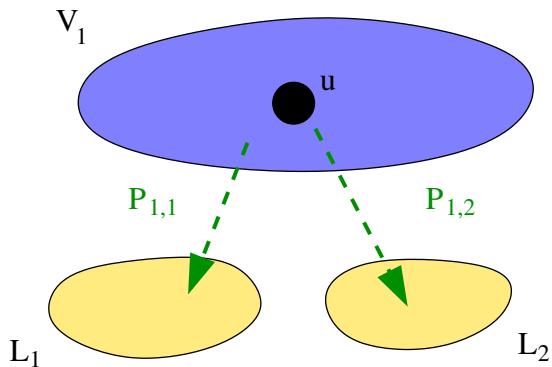
Conditional Probability



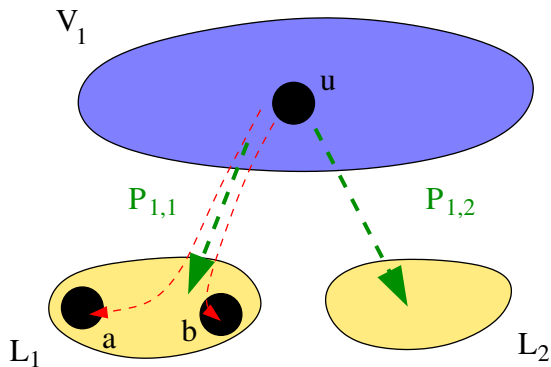
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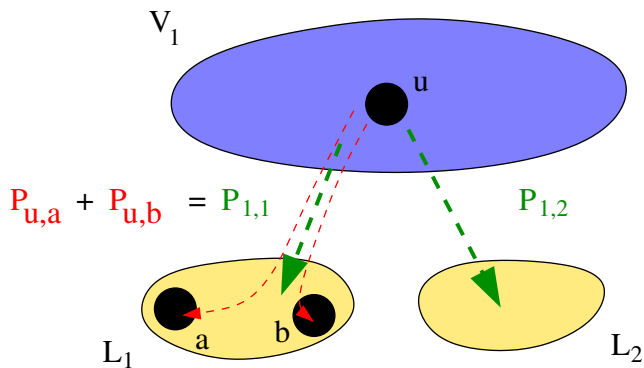
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Two Level Rounding

Procedure

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- 1 For each component V_i :

Choose $V_i \rightarrow L_j$ according to $P_{i,j}$

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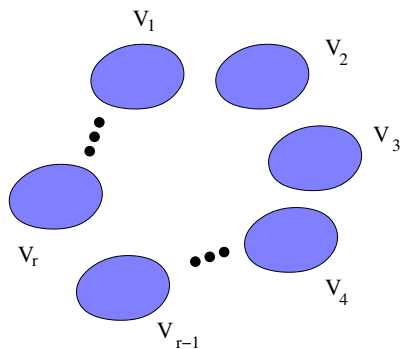
Choose $V_i \rightarrow L_j$ according to $P_{i,j}$

- 2 For each component V_i mapped to L_j , for each $u \in V_i$:

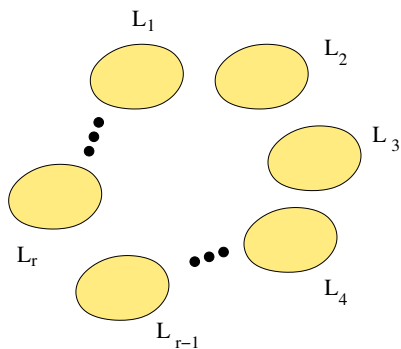
Choose $u \rightarrow a$ according to $\frac{P_{u,a}}{P_{i,j}}$

An Integrality Gap

$$|V| = r^2 \quad |V_i| = r$$

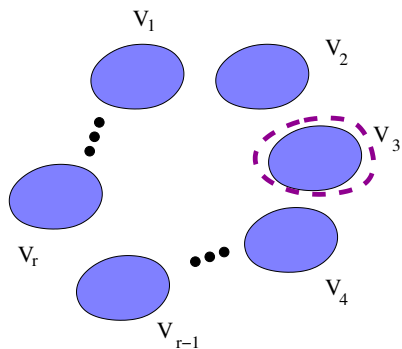


$$|L| = r^2 \quad |L_j| = r$$

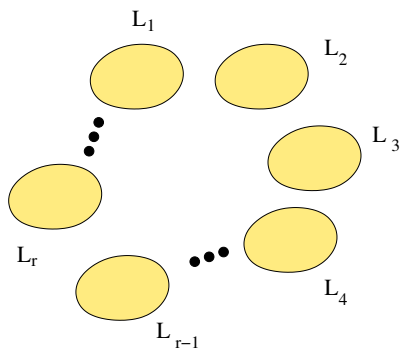


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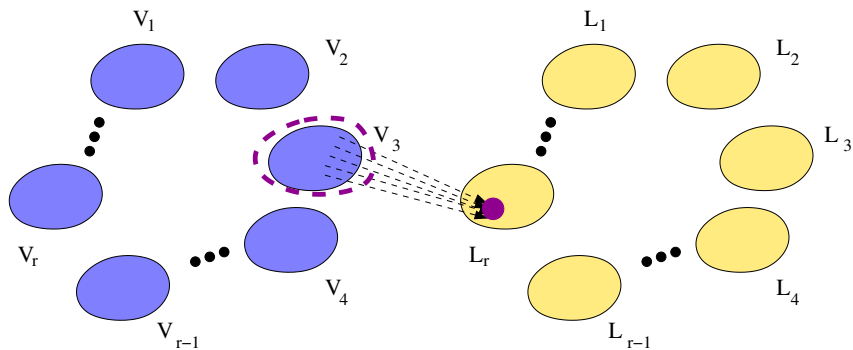
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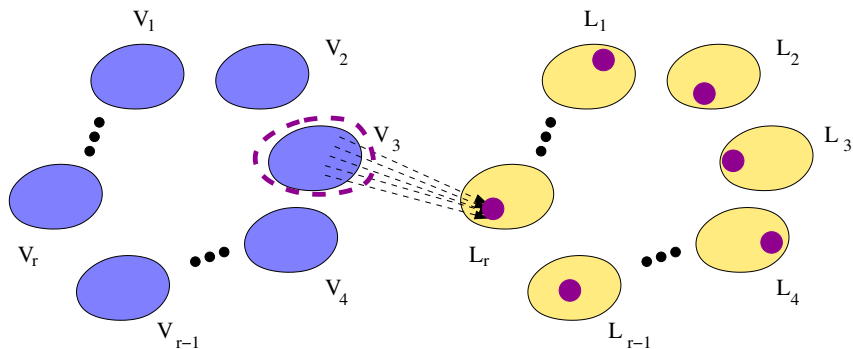
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Planning for Rounding

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Observation

In an integral solution the conditional expectation (of $a \in L_1$) is also bounded by the label capacity (of a)

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Let X_1, X_2, \dots, X_T be the expectation for a label a

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Theorem

There is a polynomial time $O(\log k)$ -approximation algorithm for the congestion of CAPACITATED METRIC LABELING

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Open Question

Can the notion of congestion be used to give bi-criteria hardness for other (graph partitioning) problems?

Questions?

Thanks!