

Local Search Algorithms for the Maximum Carpool Matching Problem

Gilad Kutiel¹ Dror Rawitz²

¹Technion

²Bar Ilan University

ESA 2017

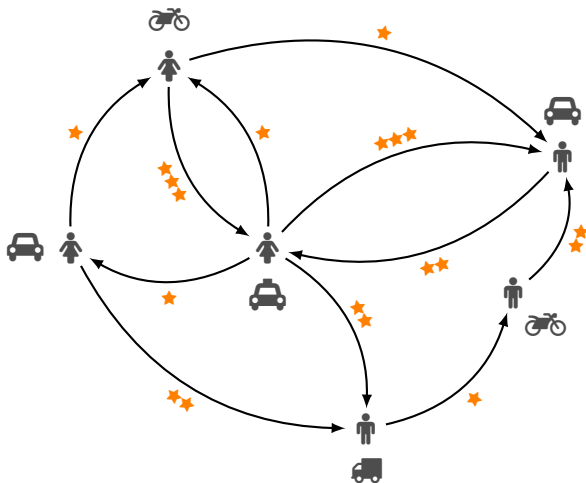
Maximum Carpool Matching (MCM)



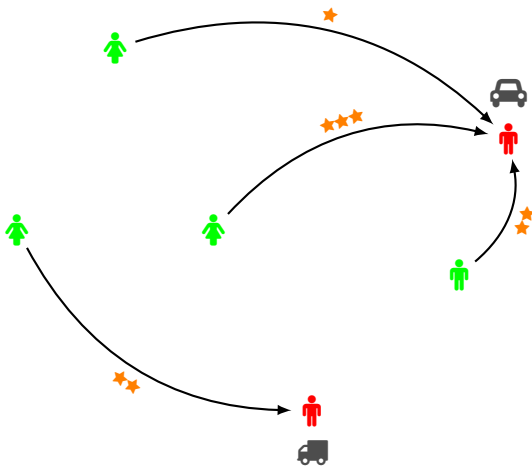
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Maximum Carpool Matching (MCM) cont.

Max 

s.t.



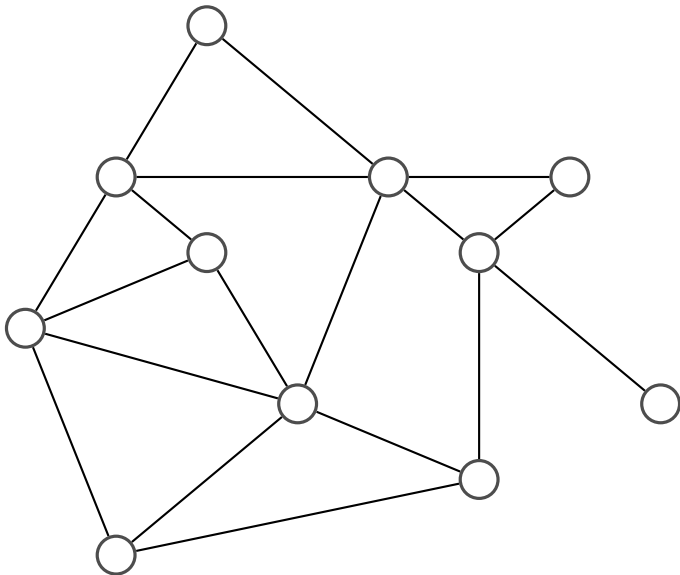
Maximum Carpool Matching (MCM) cont.

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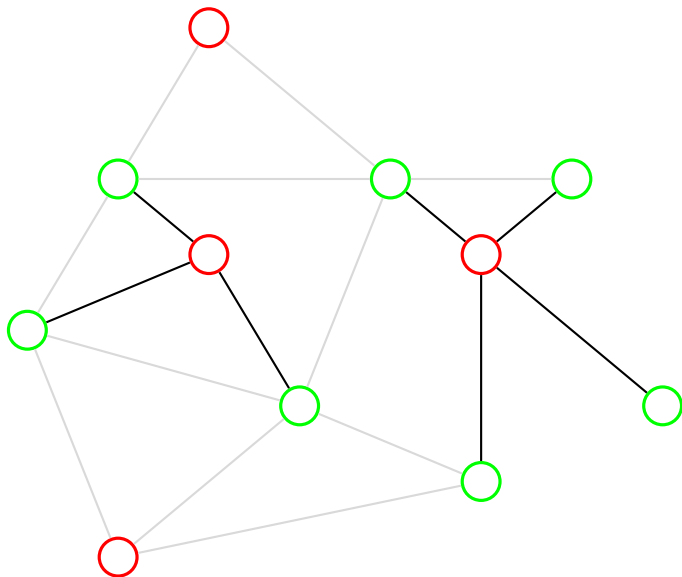
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Maximum Spanning Star Forest (MSSF)



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Hardness

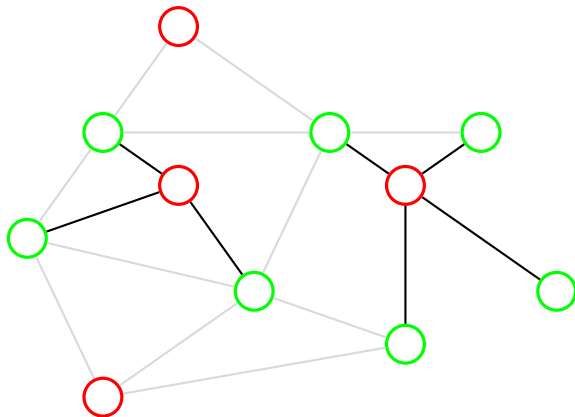
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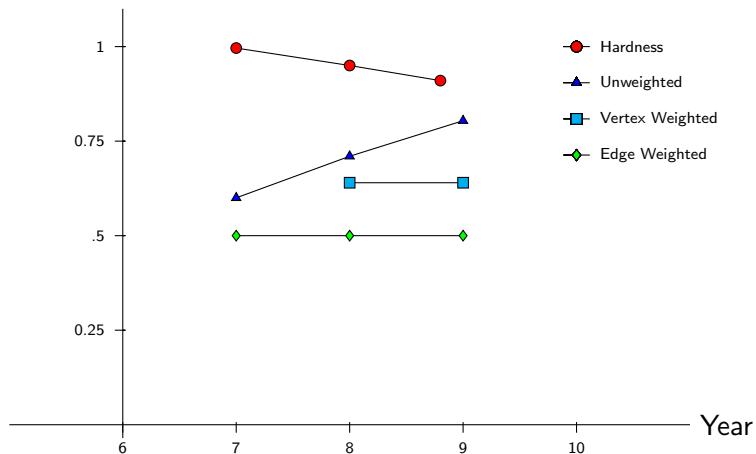
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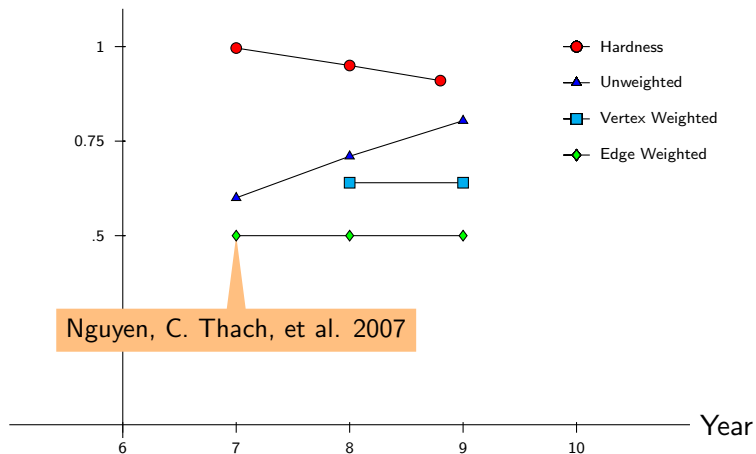
Previous Work (MSSF)

Approximation Ratio

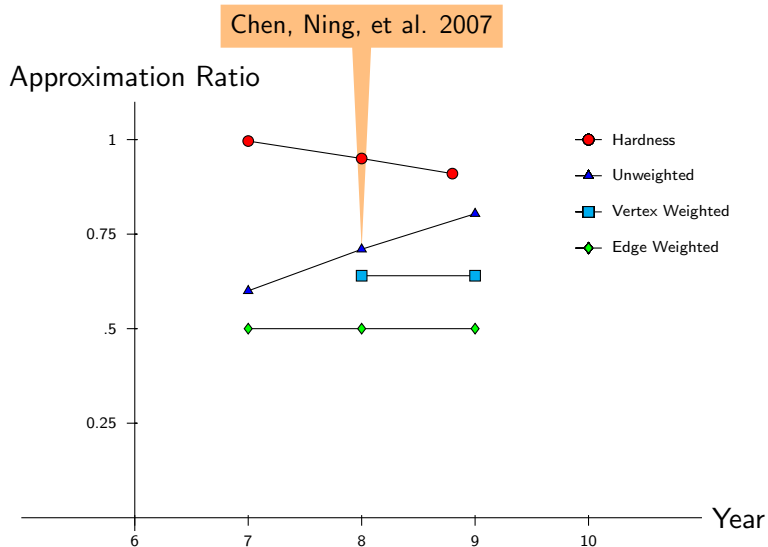


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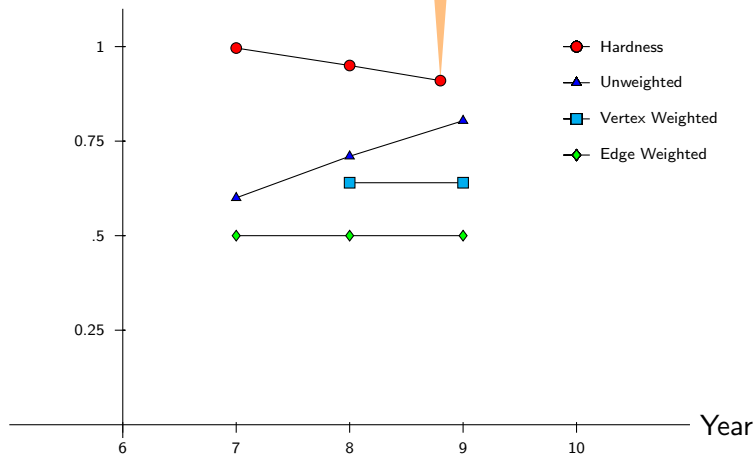
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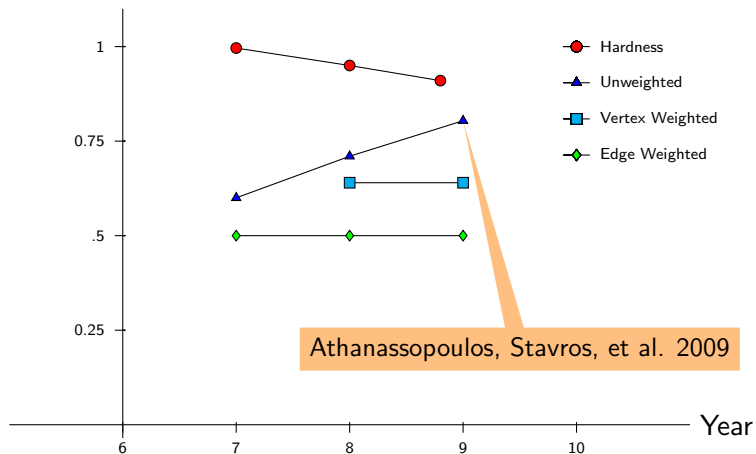
Deeparnab Chakrabarty et al. 2008

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- ▶ 1/2-approximation (unweighted)

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- ▶ $\frac{1}{2} + \frac{1}{2|G|} - \varepsilon$ (unweighted)

Local Search

Unweighted, $|\text{truck}| = O(1)$

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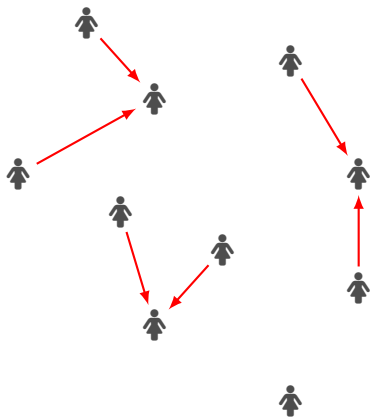
Local Search - Unweighted, $|E| = O(1)$

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 - ▶ Remove $k \leq K$ arcs
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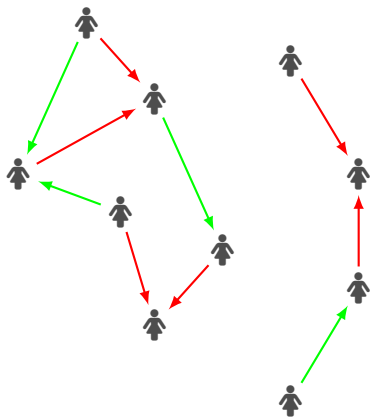
Local Search - Analysis (Star Graph)



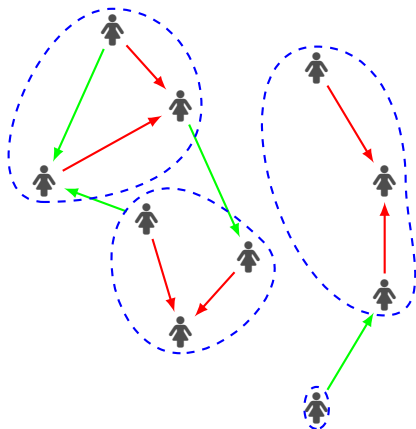
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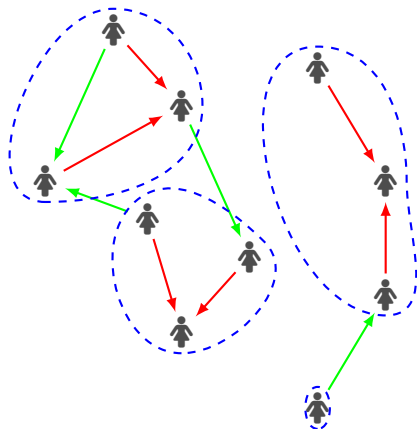
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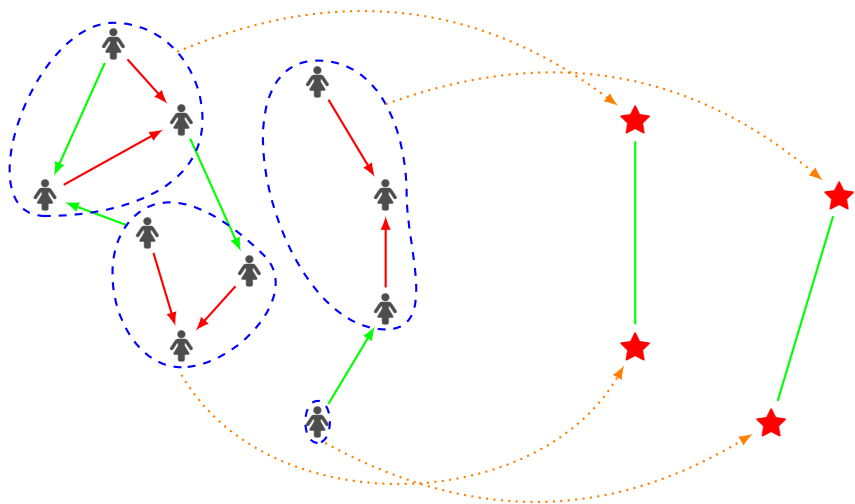
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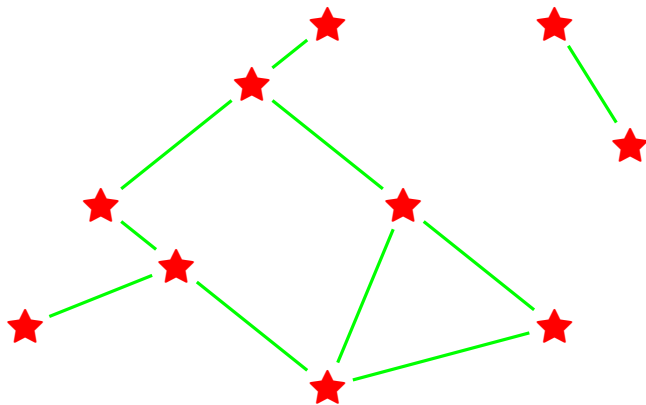
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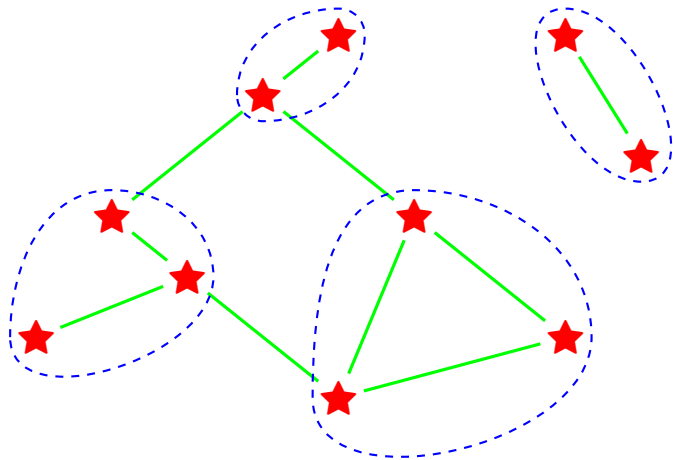
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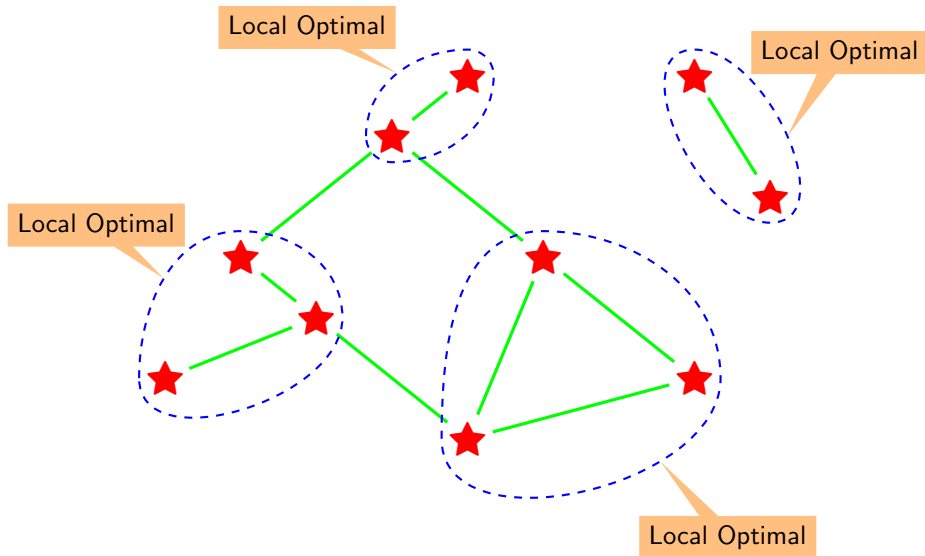
Local Search - Analysis cont.



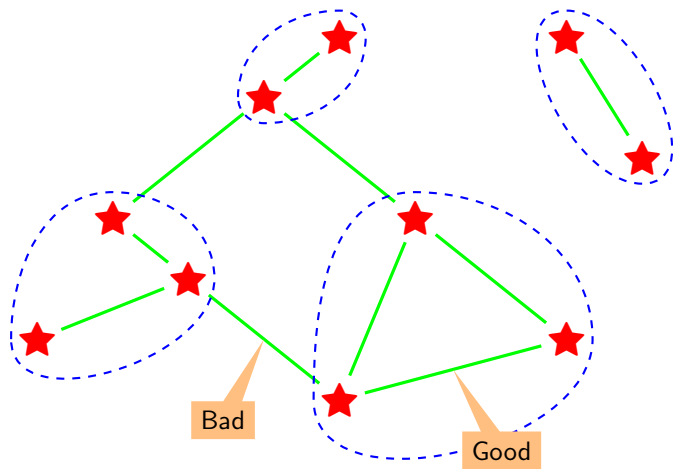
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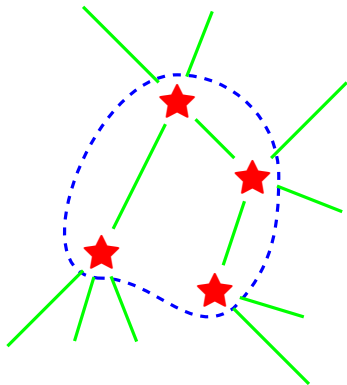


Local Search - Analysis cont.



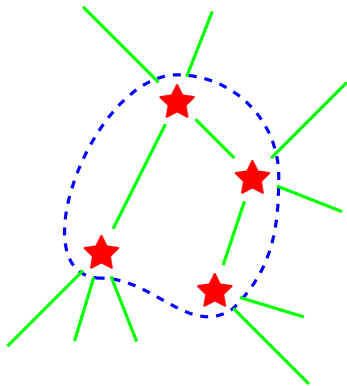
Local Search - Rough Analysis

► Let $r := \frac{\#\star}{\#\text{Good edges}}$



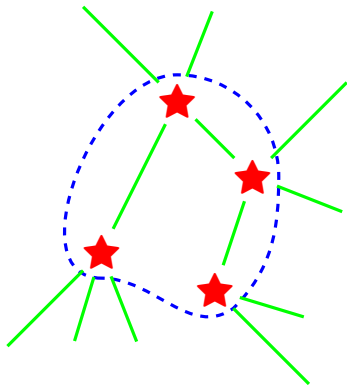
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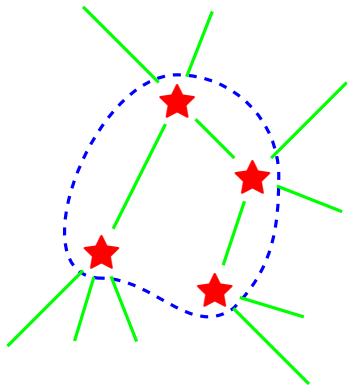
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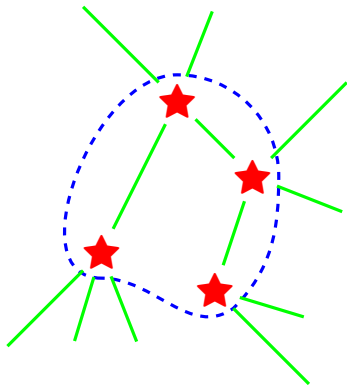
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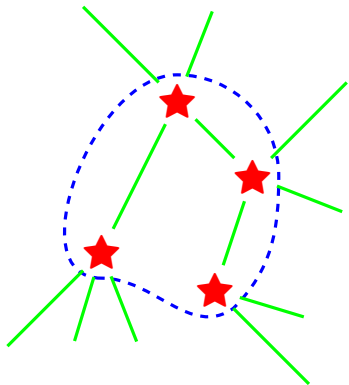
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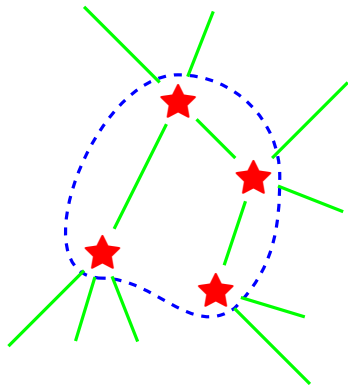
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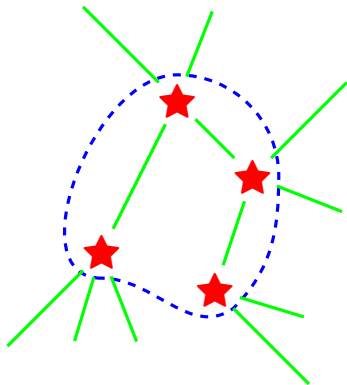
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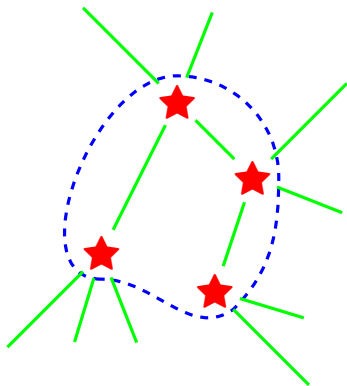
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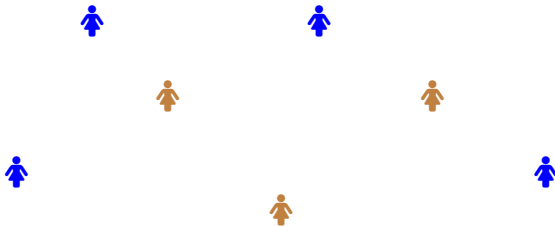
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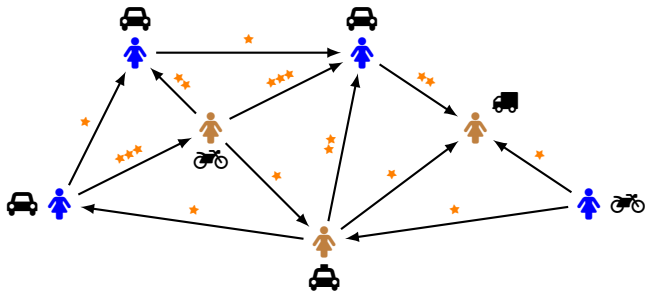
$$\frac{r + 1/2(|\star| - r)}{|\star|} \geq \frac{1 + 1/2(|\star| - 1)}{|\star|} = \frac{1}{2} + \frac{1}{2|\star|}$$

Submodularity

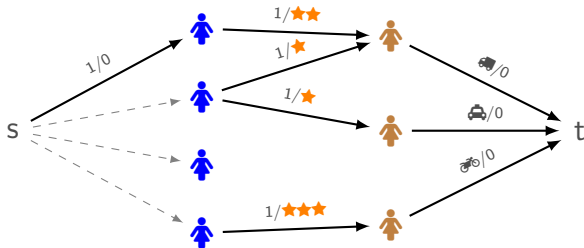
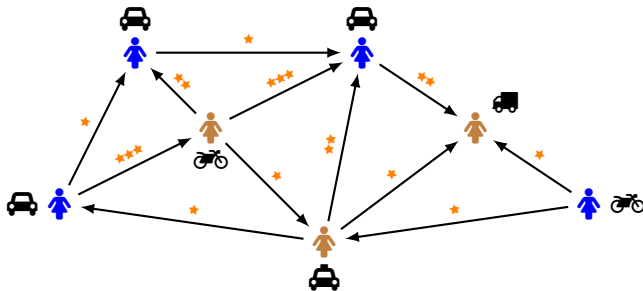
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Theorem

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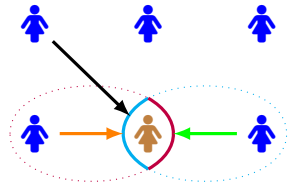
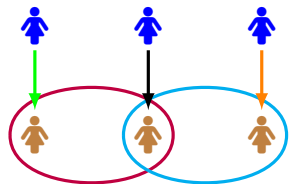
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Corollary

MCM is 1/2-approximable

Submodular cont.

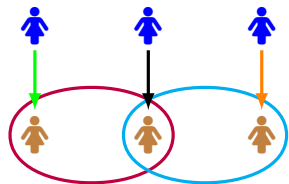
$F(A \cup B)$



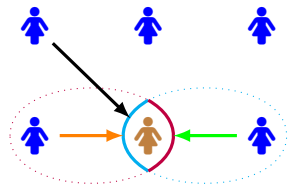
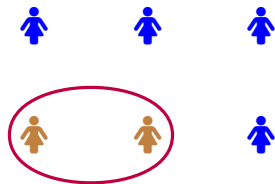
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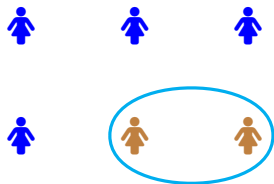


$F(A)$



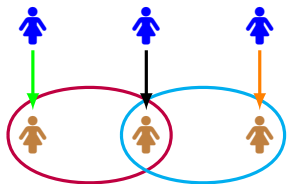
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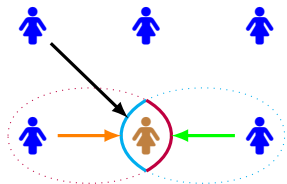
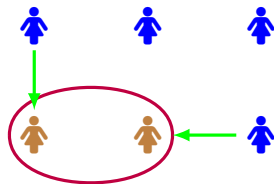


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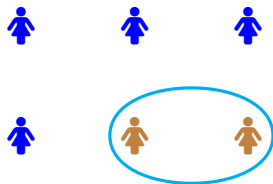
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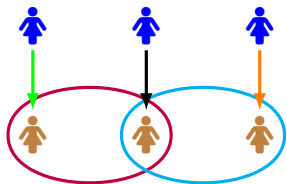
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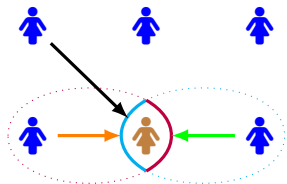
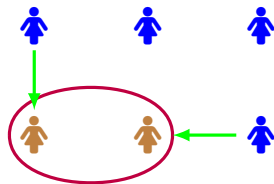
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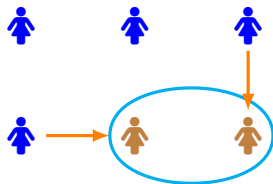


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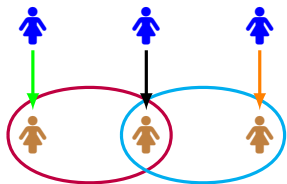
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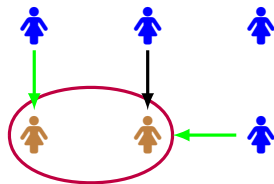


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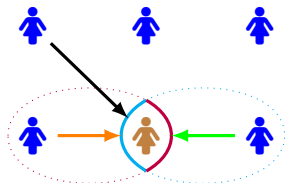
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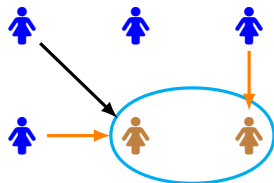
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- ▶ Is 1/2 the best we can hope for ?

Thank You.