

# Strong formulations for the pooling problem

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# Outline

- 1 Introduction
- 2  $\mathcal{NP}$ -hardness
- 3 PQ-formulation
- 4 New formulations
- 5 Computational results
- 6 Summary



What is pooling problem, and why it is hard to solve?

# Network topology

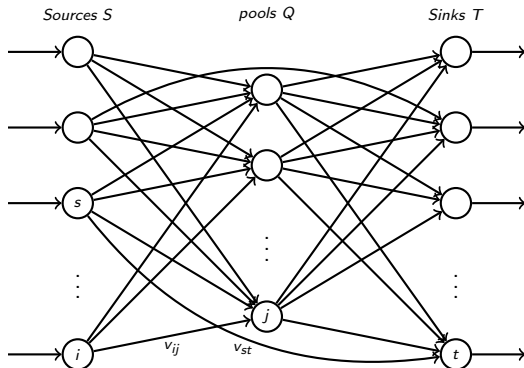
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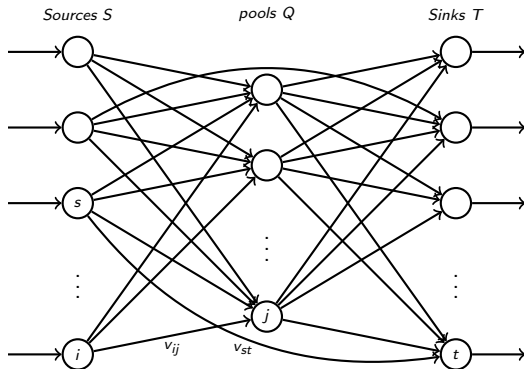
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$v_{ij}$  is flow along the arc  $(i, j) \in A$



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$\mathcal{F}(G, S, T, u)$  is the polytope

$$\sum_{j:(s,j) \in A} v_{sj} \leq u_s, s \in S$$

$$\sum_{j:(j,i) \in A} v_{ji} \leq u_i, i \in N \setminus S$$

$$\sum_{j:(i,j) \in A} v_{ij} = \sum_{j:(i,j) \in A} v_{ji}, i \in Q$$



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$$\text{MCFP} = \min \{ c^T v : v \in \mathcal{F}(G, S, T, u), 0 \leq v_{ij} \leq \max(u_i, u_j) \}$$



What is pooling problem, and why it is hard to solve?

# The pooling problem

For any  $v \in \mathcal{F}(G, S, T, u)$  associate a **quality matrix**  $w \in \mathbb{R}^{N \times K}$ :

$$w_i^k = \begin{cases} q_i^k & \text{if } i \in S \\ \frac{\sum_{j:(j,i) \in A} w_j^k v_{ji}}{\sum_{j:(j,i) \in A} v_{ji}} & \text{if } i \in N \setminus S \end{cases}$$



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The definition resembles the P-formulation



# Maximum 3-dimensional matching

## Definition

Given 3 disjoint finite sets  $I_1$ ,  $I_2$  &  $I_3$ , and a set  $E \subseteq I_1 \times I_2 \times I_3$ , find  $M \subseteq E$  of maximum cardinality s.t.  $\forall i \in I_1 \cup I_2 \cup I_3$  there is at most one  $(i_1, i_2, i_3) \in M$  s.t.  $i \in \{i_1, i_2, i_3\}$





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Reducing max 3D matching to the pooling problem

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- $A = \{(i_1, i_2) : \exists (i_1, i_2, i_3) \in E\} \cup \{(i_2, (i_1, i_3)) : (i_1, i_2, i_3) \in E\}$



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- $u_i = 1, \forall i \in N$  and  $c_{ij} = -1, \forall (i, j) \in A$
- $\forall t = (i_1, i_3), \text{ let } q_t^k = \begin{cases} 0 & k = k_{i_1} \\ 1 & k \neq k_{i_1} \end{cases}, q_{i_1}^k = \begin{cases} 0 & k = k_{i_1} \\ 1 & k \neq k_{i_1} \end{cases}$



# Consequence

Through out the proof we assume that  $|K|$  is changing with the problem size





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## Open question

Is the pooling problem  $\mathcal{NP}$ -hard for fixed  $|K|$ ?



# PQ-formulation

- Q-formulation (Ben-Tal et. al 94): introduce the proportion

$$y_i^s = \frac{V_{si}}{\sum_{t:(i,t) \in A} V_{it}}, \quad s \in S, \quad i \in Q$$



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$$\sum_{s \in S} y_i^s v_{it} = v_{it}, \quad i \in Q, t \in T$$

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- PQ-formulation dominates both the P- and the Q-formulation



# TP-formulation

- analogous to the proportion  $y_i^s$ , define the terminal proportion

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  - they do not have equal strength
  - none of them dominates the other



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  - 2  $\text{Vex}(y_i^t v_{si})$ , and  $\text{Cav}(y_i^t v_{si})$





# Branching strategy

A procedure for choosing the branching pair in the STP-formulation

- 1 for each  $(s, i, t)$

$$\text{violation} = \min \{v^*y^* - \text{Vex}(vy), \text{Cav}(vy) - v^*y^*\}$$

- 2 find the maximum violation pair in  $\{y_i^s v_{it}\}$
- 3 find the maximum violation pair in  $\{y_i^t v_{si}\}$
- 4 choose the pair with minimum violation from (2) and (3)



# Comparing the relaxations

Red value  $\implies$  better lower bounds

Problem	Objective function value			Global Solution
	PQ-relaxation	TP-relaxation	STP-relaxation	
Adhya1	<b>-840.27</b>	-856.25	<b>-840.27</b>	-549.80
Adhya2	-574.78	-574.78	-574.78	-549.80
Adhya3	-574.78	-574.78	-574.78	-561.05
Adhya4	<b>-961.93</b>	-967.43	<b>-961.93</b>	-877.65
Bental4	-550.00	<b>-541.67</b>	<b>-541.67</b>	-450.00
Bental5	-3500.00	-3500.00	-3500.00	-3500.27
Foulds2	-1100.00	-1100.00	-1100.00	-1100.00
Foulds3	-8.00	-8.00	-8.00	-8.00
Foulds4	-8.00	-8.00	-8.00	-8.00
Foulds5	-8.00	-8.00	-8.00	-8.00
Haverly1	-500.00	-500.00	-500.00	-400.00
Haverly2	-1000.00	-1000.00	-1000.00	-600.00
Haverly3	<b>-800.00</b>	-875.00	<b>-800.00</b>	-750.00
Rt2	-6034.87	<b>-5528.00</b>	<b>-5528.25</b>	-4391.83



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- formulate the PQ-, TP-, and STP-formulation on GAMS



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- neglect the RLT cuts in the local search procedure
  - apply the `RELAXATION_ONLY_EQUATIONS` option in BARON



# Computational experiments with BARON

Problem	PQ-formulation		TP-formulation		STP-formulation	
	#nodes	time(sec)	#nodes	time(sec)	#nodes	time(sec)
Adhya1	21	0.20	39	0.61	7	0.37
Adhya2	33	0.19	15	0.30	1	0.27
Adhya3	31	0.38	37	0.61	19	0.82
Adhya4	1	0.18	17	0.27	1	0.30
Bental4	1	0.02	1	0.01	1	0.02
Bental5	0	0.03	0	0.10	0	0.17
Foulds2	0	0.02	0	0.01	0	0.02
Foulds3	0	0.84	0	0.36	0	8.95
Foulds4	0	1.18	0	0.50	0	1.57
Foulds5	0	0.30	0	0.42	0	1.64
Haverly1	1	0.02	1	0.02	1	0.02
Haverly2	5	0.02	1	0.01	1	0.03
Haverly3	1	0.02	1	0.01	1	0.02
Rt2	13	0.14	1	0.10	1	0.19

0 nodes means that the problem solved during the preprocessing



# Large problems

Randomly generated problems with

- specified  $|S|$ ,  $|Q|$ ,  $|T|$ , and  $|K|$





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- specified  $|S|$ ,  $|Q|$ ,  $|T|$ , and  $|K|$
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problem	problem sets size				probability
	$ S $	$ Q $	$ T $	$ K $	
problem1	25	15	20	10	0.60
problem2	40	25	35	15	0.52
problem3	60	35	50	20	0.30



# Computational results—large problems

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Instance	PQ-formulation				STP-formulation			
	BARON		Our Code		BARON		Our Code	
	# node	time(sec)	# node	time(sec)	# node	time(sec)	# node	time(sec)
problem1	0	739.20	298	273.47	—	> 2 days	151	830.28
problem2	—	> 2 days	532	4273.84	—	> 2 days	303	41100.15
problem3	—	> 2 days	714	41387.13	—	> 2 days	640	162526.54





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- a branching strategy suite with the STP-formulation



Thank you.



# proof illustration

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$$\sum_{j:(j,i_2) \in A} w_j^{k_{i_1}} v_{ji_2} = 0 \quad (u_{i_2} = 1) \implies v_{ji_2} = \begin{cases} 0 & \text{if } j \neq i_1 \\ 1 & \text{if } j = i_1 \end{cases}$$



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then the path  $(i_1, i_2, t)$  is the only one

