

A Path Formulation for the Generalized Pooling Problem

Mohammed Alfaki

3rd Nordic Optimization Symposium, Stockholm, KTH

March 13, 2009



Outline

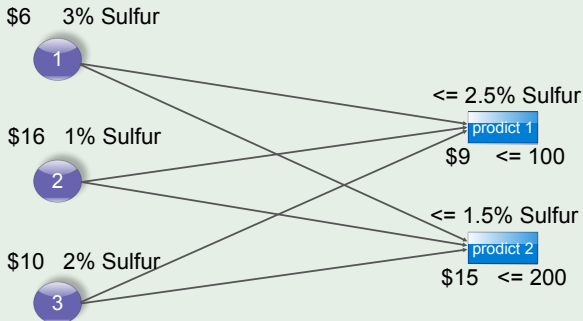
- 1 Introduction to the pooling problem
 - What is the pooling problem?
 - State of the art in global methods
 - Generalization and important applications
- 2 Formulations for general networks
 - Arc and path-oriented formulations
- 3 Branch-And-Bound
 - Linear relaxation
- 4 Column Generation
 - Column Generation-General idea
 - How it works here?
- 5 Conclusion



What is the pooling problem?

Example

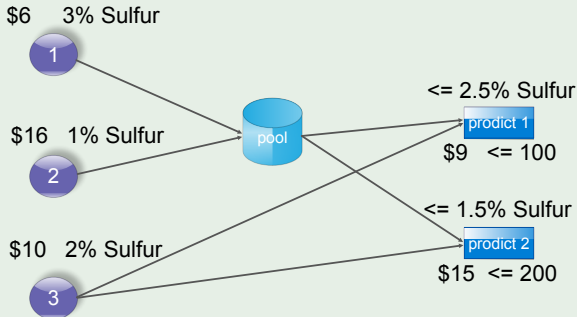
Blending problem:



What is the pooling problem?

Example

Pooling problem:



Haverly problem -1978



NP-hardness

Definitions

- Mixing raw materials in pools
- Two stage blending problem
- The model is non-linear
- Many local optima
- Hard to find global optimum

Theorem

The pooling problem is NP-hard even in the case of single-layer of pools.

Proof: *A poly reduction from the 3-dimensional matching problem to the single-layered pooling problem.*



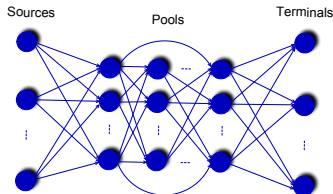
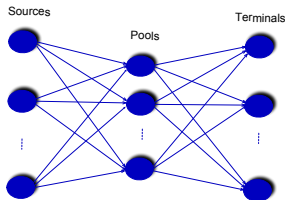
State of the art in global methods

- The mathematical models for the pooling problem based on non-linear formulation
- Nonlinearity appears in two type of constraints
 - 1 Quality balance at pools
 - 2 Quality bound at the terminals (products)
- Two formulations exist:
 - 1 p-formulation (**Haverly 1978**): consists of flow and quality variables
 - 2 q-formulation (**Ben-Tal et al. 1994**): replacing the quality variables with variable representing flow proportions.
- q-formulation is tighter than p-formulation
- **Tawarmalani & Sahinidis 2005** gave the pq-formulation, which is even tighter



Generalization and important applications

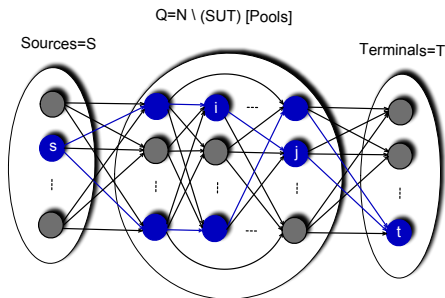
- The pq-formulation is not applicable when we have multi-layered of pools
- Important application do exist fo the multi-layered pooling problem:
 - 1 Pipeline transportation of dry gas
 - 2 Multi-period inventory model



Arc-oriented formulation

- $G = (N, A)$ is **DAG**
- Parameters:
 - 1 $q_i, i \in SUT, k \in K$
 - 2 u_i node capacity of $i \in N$
 - 3 c_{ij} unit cost of $(i, j) \in A$
- Variables:
 - 1 w_i S quality leaving $i \in Q$
 - 2 v_{ij} the flow along $(i, j) \in A$
- Minimize the cost
- Sulfur Quality balance

$$w_i = \frac{\sum_{j \in N_i^-} w_j v_{ji}}{\sum_{j \in N_i^-} v_{ji}}, i \in Q$$



Path-oriented formulation

- Variables:

- 1 v_{ij} the flow along $(i, j) \in A$
- 2 v_{ij}^s flow originating from $s \in S$
- 3 x_p flow along path $p \in P$
- 4 y_i^s proportion of flow at pool i coming from $s \in S$

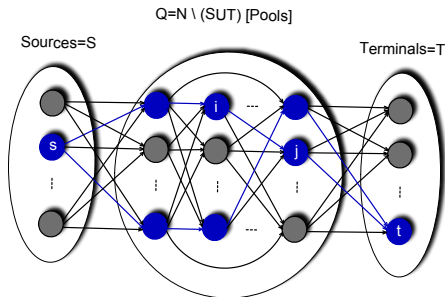
- Minimize the cost

- Proposition variables:

$$y_i^s = \frac{v_{ij}^s}{v_{ij}}, s \in S, i \in Q, (i, j) \in A$$

- legal flow:

$$v_{ij}^s - \sum_{p \in P_{ij}^s} x_p = 0, s \in S, (i, j) \in A$$

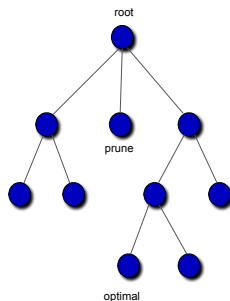


- P : set of directed paths from some source to some sink
- $P^s \subseteq P$: set of paths start with source $s \in S$
- $P_{ij} \subseteq P$: set of paths intersecting $(i, j) \in A$
- $P_{ij}^s = P^s \cap P_{ij}$



Linear relaxation

- The idea based on **B&B** algorithms for integer programming
- Subproblem: linear relaxation
- Solve the relaxed problem at each node to get upper or lower bounds
- Prune: by bound, infeasibility, optimality
- Stop when the tree is empty

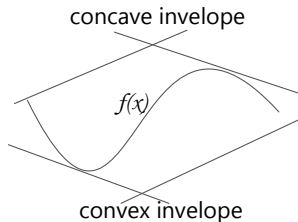


Convex and concave envelopes

- Putting $\zeta = yv$ the convex and concave envelopes on rectangle $\mathcal{C} = [\underline{y}, \bar{y}] \times [\underline{v}, \bar{v}]$ are:

$$\max\{\underline{y}\underline{v} + \underline{y}\bar{v} - \underline{y}\underline{v}, \bar{y}\underline{v} + \bar{y}\bar{v} - \bar{y}\bar{v}\}$$

$$\min\{\underline{y}\underline{v} + \bar{y}\bar{v} - \bar{y}\underline{v}, \bar{y}\underline{v} + \underline{y}\bar{v} - \underline{y}\bar{v}\}$$
- Replacing the bilinear constraints with $\zeta_{ij}^s - v_{ij}^s = 0$
- Adding the above bounds to give linear relaxation problem R



Column Generation—General idea

- Column generation helpful in problems with many variables/columns but relatively few constraints
- Exclude some of the variables
- The **master problem**: original problem with only a subset of variables
- The **subproblem**: to identify whether there is excluded column that reduces the objective function value



How it works here?

- The master problem R' will be the problem R with only $P' \subseteq P$
- The idea is to solve R' , and whenever we detect that the cost can be reduced by sending flow along any path $p \in P \setminus P'$, we augment P' by p
- We are looking for a path that reduces the cost
- The subproblem will be the **Shortest-Path** problem, which is solvable in polynomial time, since G is **DAG**



Algorithm for solving R

At each node in the **B&B** tree we will solve the following:

repeat

 solve the master problem R'

for $(\forall s \in S)$ **do**

 define arc lengths

$p \leftarrow$ shortest path from s to some $t \in T$ in G

if $(length(p) < 0)$ **then**

$P' \leftarrow P' \cup \{p\}$

end

end

until $(no\ path\ of\ length < 0\ found)$



Conclusion

- This is work in progress.
- The pooling problem is NP-hard
- The path-oriented formulation is equivalent to the pq-formulation, when we have one layer of pools
- Generalization of the strongest known formulation for one-layer instances

- Outlook
 - Experimental evaluation

