Multi-player minimum cost flow problems with nonconvex costs and integer flows

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Outline

Introduction

Algorithm construction

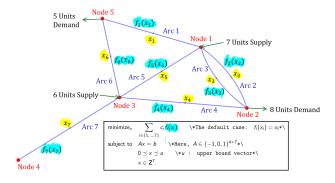
Algorithm description

Numerical example

Network Flow Problems

- Network flow problems: optimization problems associated with underlying directed network.
- ▶ They arise in numerous application settings and in many forms.
- Some common application areas: communication networks, transportation system, social network, power system, electro-mechanical systems etc.
- ► The minimum cost flow problem is the most fundamental among network flow problems.

Minimum Cost Flow Problem



Minimum Cost Flow Problem

minimize_x
$$\sum_{i \in \{1,\dots,n\}} c_i f_i(x_i) \qquad \text{`*Default Case:} \quad f_i(x_i) = x_i * \text{`}$$
 subject to
$$Ax = b$$

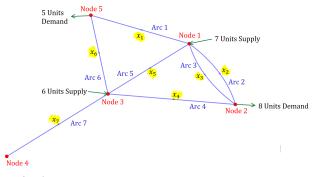
$$0 \leq x \leq u$$

$$x \in \mathbf{Z}^n.$$
 (1)

- ▶ Here, the network has n arcs, m+1 nodes.
- ightharpoonup A: reduced node-arc incidence matrix, dimension $m \times n$, full row rank.
- We re-index the m linearly independent columns of A as the first m columns.
- b : represents supplies to the nodes at different points.

Extension of Min. Cost Flow to a Multi-player Setup

- ▶ With each arc of the network graph we associate one player.
- ► Each of the players is trying to minimize its nonconvex cost function, subject to the network flow constraints.
- Our goal is to seek an efficient solution concept in this multi-player problem.



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Multi-player Extension

▶ The goal of the *i*th player for $i \in [n] = \{1, ..., n\}$, given other players' strategies $x_{-i} \in \mathbf{Z}^{n-1}$, is to solve:

```
minimize<sub>x_i</sub> f_i(x_i) \*proper: practically can be anything*\ subject to A(x_i,x_{-i})=b \*The constraints 0 \leq (x_i,x_{-i}) \leq u couple the players*\ x \in \mathbf{Z}^n.
```

(2)

Solution Concepts in Consideration

- A *vector optimal solution* that minimizes all the objectives simultaneously is unlikely to exist.
- ► The celebrated *Nash equilibrium* is also not very efficient in our setup because:
 - the constraint set of the problem has equality constraints, thus making any feasible point a Nash equilibrium, and
 - posteriori some of the players may decide to deviate from the Nash equilibrium in order to reduce their costs even more at the expense of the rest of the players.
- ▶ A more effective solution concept is the *Pareto optimal point*.

Pareto Optimal Point

A *Pareto optimal point* is a solution concept where none of the objective functions can be improved without worsening some of the other objective values.

Definition

(Pareto Optimal Point) In problem 2, a point $x^* \in P$ is Pareto optimal if it satisfies the following: there *does not* exist another point $\tilde{x} \in P$ such that

$$(\forall i \in [n])$$
 $f_i(\tilde{x}_i) \leq f_i(x_i^*),$

with at least one index $j \in [n]$ satisfying $f_j(\tilde{x}_j) < f_j(x_i^*)$.

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Related Work

- Our problem setup does not seem to be investigated in existing literature.
- ► Nonconvex network flow problems for very specific cost functions (Magnanti1984, Graves1985, Daskin2011, He2015).
- Integer multi-commodity flow problems (Brunetta2000, Ozdaglar2004).

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Decoupling the Last n-m **Players**

Theorem

(Existence of a decoupling variable) Denote

 $B = [A_1|A_2|\cdots|A_m]$. The equality constraint set

 $Q = \{x \in \mathbf{Z}^n \mid Ax = b\}$ is nonempty and for any vector $x, x \in Q$ is equivalent to saying that there exists a $z \in \mathbf{Z}^{n-m}$ such that

$$x = (d_1 - h_1^T z, \dots, d_m - h_m^T z, z_1, \dots, z_{n-m}),$$
 (3)

where d_i is the ith component of $d = B^{-1}b$, and $h_i^T \in \mathbf{Z}^{n-m}$ is the ith row of $B^{-1}A_{[1:m,m+1:n]}$.

Decoupled Problems for the n-m **Players**

▶ In z we can decouple the optimization problems for players m+1, m+2, ..., n as follows

minimize_{$$z_i$$} $f_i(z_i)$
subject to $0 \le z_i \le u_i$ $z_i \in \mathbf{Z}$. (4)

Set of different optimal solutions for player m+i for $i \in [n-m]$ is

$$D_i = \{z_{i,1}, z_{i,2}, \ldots, z_{i,p_i}\}.$$

▶ Define, $D = \times_{i=1}^{n-m} D_i \neq \emptyset$

- ▶ We provide each player $i \in [m]$ with its own local copy of z, denoted by $z^{(i)} \in \mathbf{Z}^{n-m}$, which acts as its decision variable.
- ► For any $i \in [m]$, $x_i = d_i h_i^T z^{(i)}$.
- ► The copy z⁽ⁱ⁾ has to be in consensus with the rest of the first m players:

$$z^{(i)} = z^{(j)} \quad \forall j \in [m] \setminus \{i\}.$$

- ► The copy $z^{(i)}$ has to satisfy the *flow bound constraints*, *i.e.*, $0 \le d_i h_i^T z^{(i)} \le u_i$ for all $i \in [m]$.
- ▶ For the last n-m players $z_i \in D_i$, so:

$$z^{(i)} \in D$$

For all $i \in [m]$, the *i*th player's optimization problem in variable $z^{(i)}$ can be written as:

minimize_{$$z^{(i)}$$} $\bar{f}_i\left(z^{(i)}\right) = f_i(d_i - h_i^T z^{(i)})$
subject to $z^{(i)} = z^{(j)}, \quad j \in [m] \setminus \{i\}$
 $0 \le d_i - h_i^T z^{(i)} \le u_i$
 $z^{(i)} \in D.$ (5)

$$q_i(z^{(i)}) = (d_i - h_i^T z^{(i)})(d_i - h_i^T z^{(i)} - 1) \cdots (d_i - h_i^T z^{(i)} - u_i) = 0,$$
 (6)

$$z_j(z^{(i)}) = (z_j^{(i)} - z_{j,1})(z_j^{(i)} - z_{j,2})\dots(z_j^{(i)} - z_{j,p_i}) = 0, \quad j \in [n-m].$$
 (7)

$$\mathscr{F} = \{z \in \mathbf{Z}^{n-m} \mid (\forall k \in [m]) \ q_k(z) = 0, (\forall j \in [n-m]) \ r_i(z) = 0\}.$$

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$$\mathscr{F} = \{z \in \mathbf{Z}^{n-m} \mid (\forall k \in [m]) \ q_k(z) = 0, (\forall j \in [n-m]) \ r_i(z) = 0\}.$$

For $i \in [m]$, *i.e.*, each of these players are optimizing over a common constraint set \mathscr{F} :

- ▶ When ℱ is nonempty? How to find the points in it?
- ► A little algebraic geometry will take us a long way...

A Little Algebraic Geometry

▶ The *ideal* generated by polynomials $f_1, f_2, ..., f_m \in \mathbf{C}[x]$ is the set

ideal
$$\{f_1,\ldots,f_m\} = \left\{\sum_{i=1}^m h_i f_i \mid (\forall i \in [m]) \ h_i \in \mathbf{C}[x]\right\}.$$

- * analogous to span of vectors *\
- A Groebner basis G_≻ is particular kind of generating set of an ideal I over a field C[x] * analogous to basis set of a span*\
- ▶ Reduced Groebner basis G_{reduced,≻} is the most compact Groebner basis for an ideal I. * analogous to orthonormal basis of a span *\
- ► There are many computer algebra packages to compute reduced Groebner basis such as Macaulry2, SINGULAR, FGb, Maple, Mathematica etc.

When is \mathscr{F} Nonempty?

Theorem

The set \mathscr{F} is nonempty if and only

$$G_{reduced,\succ} \neq \{1\},$$

where $G_{reduced,\succ}$ is the reduced Groebner basis of ideal $\{q_1,\ldots,q_m,r_1,\ldots,r_{n-m}\}$ with respect to any ordering.

Computing Points in ${\mathscr F}$

For
$$i = [n - m]$$
 calculate

$$G_{n-m-i} = G_{reduced,\succ_{lex}} \cap \mathbf{C}[z_{n-m-i+1},z_{n-m-i+2},\ldots,z_{n-m}].$$

Theorem

$$G_0 = \mathscr{F}$$
.

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Finding the Pareto optimal points from \mathscr{F}

```
for i \in [m]
  X_i := d_i - h_i^T \mathscr{F}
                    \* The inverse operator is denoted X_i^{-1} *\
end for
```

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for i \in [m]
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  X_i := d_i - h_i^T \mathscr{F}
end for
  Sort the elements of the \{X_i\}_{i=1}^ms with respect to cardinality of the elements in a
descending order.
  Denote the index set of the sorted set by \{s_1,\ldots,s_m\} such that |X|_{s_1}\geq\cdots\geq |X|_{s_m}.
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Finding the Pareto optimal points from \mathscr{F}

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for i \in [m]
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   Sort the elements of the \{X_i\}_{i=1}^ms with respect to cardinality of the elements in a
descending order.
   Denote the index set of the sorted set by \{s_1,\ldots,s_m\} such that |X|_{s_1}\geq\cdots\geq |X|_{s_m}.
for i \in [m]
    X_{s_i}^* := \operatorname{argmin}_{X_{s_i} \in X_{s_i}} f_{s_i}(x_{s_i}) \* Univariate optimization problem *\
  \mathscr{F}^*_{s_i} := \bigcup_{\mathsf{X}_{\mathsf{S}_i} \in \mathsf{X}^*_{\mathsf{S}_i}} (\mathsf{X}^*_{s_i})^{-1}(\mathsf{X}_{\mathsf{S}_i}) \* Lemma. The set \mathscr{F}^*_{s_i} is nonempty *\
   if i \le m
      X_{s_{i+1}} := \left\{ d_{s_{i+1}} - h_{s_{i+1}}^T z \mid z \in \mathscr{F}_{s_i}^* \right\}.
   end if
end for
return \mathscr{F}^*_{s_m} \* Theorem. Any member of \mathscr{F}^*_{s_m} is a Pareto optimal point
*\
```

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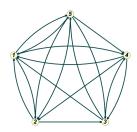
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Numerical Example: 16 Players

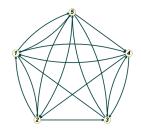


A multi-player transportation problem

- 5 nodes, 16 arcs
- · Nodes 2, 4 are retail centers
- Nodes 1, 3 are warehouses
- · Node 5 transshipment node
- Products need to be shipped from warehouses to retail centers
- Arcs represent different modes of shipment from one node to another and carried out by different organization
- Cost of a certain type of shipment depends on the number of products shipped

$$b = (9, -13, 15, -11), u = (5, 6, 6, 10, 10, 7, 11, 13, 16, 12, 4, 5, 6, 14, 13, 15)$$

Pareto Optimal Solutions



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The Pareto optimal points are

and

Concluding Remarks

- ▶ Methodology is numerically efficient when $m < \frac{n}{2}$.
- Calculating Groebner basis can be numerically challenging for very large system.
- ▶ Based on structure of the network it may happen that \mathscr{F} is empty \Rightarrow penalty based approach.

Thank You! Questions?

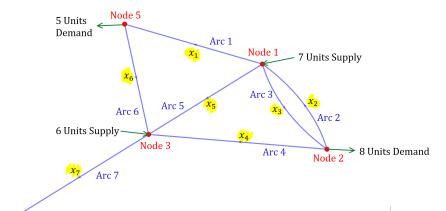
Cost Functions

Player	Cost function	
1	$-\frac{x_1^4}{30} - \frac{13x_1^3}{15} + \frac{259x_1^2}{30} - \frac{263x_1}{15} + 1$	
2	$\frac{77x_2^5}{120} - \frac{247x_2^4}{24} + \frac{471x_2^3}{8} - \frac{3365x_2^2}{24} + \frac{6779x_2}{60} + 1$	
3	$\frac{47x_3^4}{24} - \frac{133x_3^3}{4} + \frac{4897x_3^2}{24} - \frac{2123x_3}{4} + 485$	
4	$\frac{323x_{4}^{5}}{3360} - \frac{2179x_{4}^{4}}{1120} + \frac{47393x_{4}^{3}}{3360} - \frac{48709x_{4}^{2}}{1120} + \frac{7885x_{4}}{168} + 5$	
5	$(x_5-1)^2$	
6	$-\frac{x_6^4}{8} + \frac{25x_6^3}{12} - \frac{71x_6^2}{8} + \frac{95x_6}{12} + 10$	
7	x ₇ – 5	
8	$\frac{11x_8^7}{1260} - \frac{7x_8^6}{36} + \frac{119x_8^5}{72} - \frac{479x_8^4}{72} + \frac{4609x_8^3}{360} - \frac{803x_8^2}{72} + \frac{155x_8}{28} + 1$	
9	$-\frac{15}{16}x_9^3 + \frac{365x_9^2}{16} - \frac{2865x_9}{16} + \frac{7315}{16}$	
10	$(x_{10}-10)^2$	
11	$\frac{5x_{11}^4}{6} - \frac{35x_{11}^3}{3} + \frac{355x_{11}^2}{6} - \frac{370x_{11}}{3} + 90$	
12	$\frac{5x_{12}^4}{6} - \frac{25x_{12}^3}{3} + \frac{175x_{12}^2}{6} - \frac{110x_{12}}{3} + 15$	
13	$\frac{5x_{13}^4}{6} - 15x_{13}^3 + \frac{595x_{13}^2}{6} - 280x_{13} + 285$	
14	$\frac{5x_{14}^4}{6} - \frac{85x_{14}^3}{3} + \frac{2155x_{14}^2}{6} - \frac{6020x_{14}}{3} + 4165$	

Cost Functions

Player	Cost function	
15	x ₁₅ - 7	
16	$ \begin{pmatrix} x_{16} + 1, \\ 0, \\ (x_{16} + 1)^3, \\ -\frac{x_{16}^3}{2} + \frac{13x_{16}^2}{2} - \frac{244x_{16}}{3} + 330, \end{pmatrix} $	if $0 \le x_{16} \le 3$ if $4 \le x_{16} \le 6$ if $7 \le x_{16} \le 9$ else

A Generic Network



Node 4

The Minimum Cost Flow Problem

- ► There is a *directed connected graph* that represents the network.
- ► There is *flow* of some commodity along the arcs of the graphs.
- Each arc incurs a *cost* depending on the amount of flow.
- ► The flow is often taken to be *integral*.
- ► The goal is to *minimize the total cost* of all flows subject to the *network constraints*.