

Don't choose between workload balance and makespan minimization

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Joint work with Sébastien Deschamps

A modeling exercise

Input.

- Set U of tasks, $u \in U$ requires d_u minutes to be completed
- Set W of workers
- Bipartite graph $G = (U \cup W, E)$: edge uw = worker w has the skill for task u
- Parallel work on a task allowed

Output. Time spent by each worker on each task so as to minimize the maximum workload

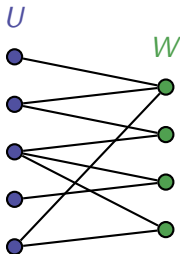
Exercise: Model this as a linear program

workload of a worker = total time worked by him

If there is no break, maximum workload = makespan
makespan = time required to complete all tasks

A modeling exercise, cont'd

$$\begin{aligned} \text{minimize} \quad & \max_{w \in W} \sum_{e \in \delta(w)} x_e \\ \text{s.t.} \quad & \sum_{e \in \delta(u)} x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R}_+ \quad \forall e \in E \end{aligned}$$



A student: "If workload balance is optimized, then maximal workload is minimized."

A modeling exercise, cont'd

$$\begin{array}{ll} \text{minimize} & \max_w \sum_{e \in \delta(w)} x_e \\ \text{s.t.} & \sum_{e \in \delta(u)} x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R}_+ \quad \forall e \in E \end{array}$$

$$\begin{array}{ll} \text{minimize} & \max_{w, w'} \sum_{e \in \delta(w)} x_e - \sum_{e \in \delta(w')} x_e \\ \text{s.t.} & \sum_{e \in \delta(u)} x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R}_+ \quad \forall e \in E \end{array}$$

A student: “Any optimal solution of the program on the right is an optimal solution of the problem on the left.”

Polymatroids

The student is right.

Theorem Fujishige 1980

Every polymatroid admits a unique **lexicographically minimal** point of its base polyhedron. Moreover, this point is also **lexicographically maximal**.

Base polyhedron of a polymatroid on a ground set $W =$

$$\{z \in \mathbb{R}_+^W : z(S) \leq f(S) \forall S \subseteq W, z(W) = f(W)\}$$

where f is **submodular** ($f(S) + f(T) \geq f(S \cap T) + f(S \cup T)$)

In our case: the map $S \subseteq W \mapsto d(N(S)) \in \mathbb{R}_+$ is submodular.

Beyond polymatroid

$$\begin{array}{ll} \text{minimize} & \max_w \sum_{e \in \delta(w)} b_e x_e \\ \text{s.t.} & \sum_{e \in \delta(u)} a_e x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R}_+ \quad \forall e \in E \end{array}$$

$$\begin{array}{ll} \text{minimize} & \max_{w, w'} \sum_{e \in \delta(w)} b_e x_e - \sum_{e \in \delta(w')} b_e x_e \\ \text{s.t.} & \sum_{e \in \delta(u)} a_e x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R}_+ \quad \forall e \in E \end{array}$$

where the a_e and b_e are positive real numbers.

Theorem Deschamps–M. 2024+

Any optimal solution of the program on the right is an optimal solution of the problem on the left (provided that the program on the right has a nonzero optimal value).

Beyond polymatroid

$$\begin{array}{ll} \text{minimize} & \max_w \sum_{e \in \delta(w)} b_e x_e \\ \text{s.t.} & \sum_{e \in \delta(u)} a_e x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R}_+ \quad \forall e \in E \end{array}$$

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where the a_e and b_e are positive real numbers.

Interpretation:

- a_{uw} = efficiency of worker w on task u
- b_{uw} = disutility of worker w on task u
- “If workload cannot be perfectly balanced, then optimizing workload balance minimizes maximal workload.”

Comments

Implies the result of the beginning (special case $a_e = b_e = 1$ for all $e \in E$)

Proof:

- elementary and short (less than one page)
- work by contradiction
- improve an optimal solution of one problem using an optimal solution of the other (“transfers”)

Realm of linear programming (in particular, polynomial algorithms)

Extensions?

♣ x_e in \mathbb{Z}_+ ?

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result holds when $a_e = b_e = 1$ for all $e \in E$ (Frank–Murota, 2022);
counterexamples otherwise.

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♣ a_e and b_e in \mathbb{R} , or even other dependencies?

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not yet investigated

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♣ x_e in \mathbb{R} ?

next slides

Negative numbers allowed

$$\begin{array}{ll} \text{minimize} & \max_w \sum_{e \in \delta(w)} b_e x_e \\ \text{s.t.} & \sum_{e \in \delta(u)} a_e x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R} \quad \forall e \in E \end{array}$$

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where the a_e and b_e are positive real numbers.

Proposition Deschamps–M. 2024+

Any optimal solution of the program on the **left** is an optimal solution of the problem on the **right**.

... but the program on the left has not always an optimal solution...

Negative numbers allowed

$$\begin{array}{ll} \text{minimize} & \max_w \sum_{e \in \delta(w)} b_e x_e \\ \text{s.t.} & \sum_{e \in \delta(u)} a_e x_e = d_u \quad \forall u \in U \\ & x_e \in \mathbb{R} \quad \forall e \in E \end{array}$$

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where the a_e and b_e are positive real numbers.

Proposition Deschamps–M. 2024+

Given any feasible solution, there is a feasible solution with an objective value on the **left** that is at least as good, and with an objective value equal to **0** on the **right**.

Comments

Concrete implication: not clear

Quite easy to establish: realm of linear algebra

Other “dependencies”

$$\text{minimize} \quad \max_w f_w(x_{\delta(w)})$$

$$\text{s.t.} \quad f_u(x_{\delta(u)}) = d_u \quad \forall u \in U$$

$$x_e \in \mathbb{R} \quad \forall e \in E$$

$$\text{minimize} \quad \max_{w, w'} f_w(x_{\delta(w)}) - f_{w'}(x_{\delta(w')})$$

$$\text{s.t.} \quad f_u(x_{\delta(u)}) = d_u \quad \forall u \in U$$

$$x_e \in \mathbb{R} \quad \forall e \in E$$

where the maps f_v are continuous and componentwise increasing self-bijections of \mathbb{R} .

Theorem Deschamps–M. 2024+

Given any feasible solution, there is a feasible solution with an objective value on the left that is at least as good, and with an objective value equal to 0 on the right.

Ingredients of the proof

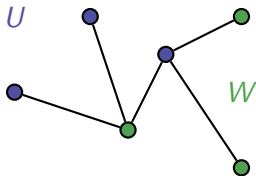
Lemma (“Implicit function theorem”) Deschamps–M. 2024+

Let $f: \mathbb{R}^k \rightarrow \mathbb{R}$ be a componentwise increasing self-bijection of \mathbb{R} . Given $y \in \mathbb{R}^{k-2}$, there exists a continuous decreasing self-map φ of \mathbb{R} such that $\varphi(t)$ is the unique number such that $f(y, \varphi(t), t) = 0$ for $t \in \mathbb{R}$.

- Natural and not difficult to prove
- Reminds the implicit function theorem from calculus
- Should have been already noticed (?)

Ingredients of the proof, cont'd

$$X := \{x \in \mathbb{R}^E : f_u(x_{\delta(u)}) = d_u, \forall u \in U\}$$



Lemma Deschamps–M. 2024+

Suppose that G is a tree. Then there is a unique x^* minimizing $\max_w f_w(x_{\delta(w)})$ over X . Moreover, all $f_w(x_{\delta(w)}^*)$ are equal.

- Natural but less immediate to prove
- Relies on the previous “implicit function lemma”

Comments

Proof provides an (approximation) algorithm

- assuming oracle to “invert” the f_v
- binary-search over an arbitrary spanning tree

Proof reminds of the Poincaré–Miranda theorem

Theorem Miranda (1940)

Consider continuous functions $f_1, \dots, f_n: [-1, 1]^n \rightarrow \mathbb{R}$ such that f_i is nonpositive when $x_i = -1$ and nonnegative when $x_i = 1$. Then there is a point where all functions are 0 simultaneously.

THANK YOU