

Tropical Mechanism Design

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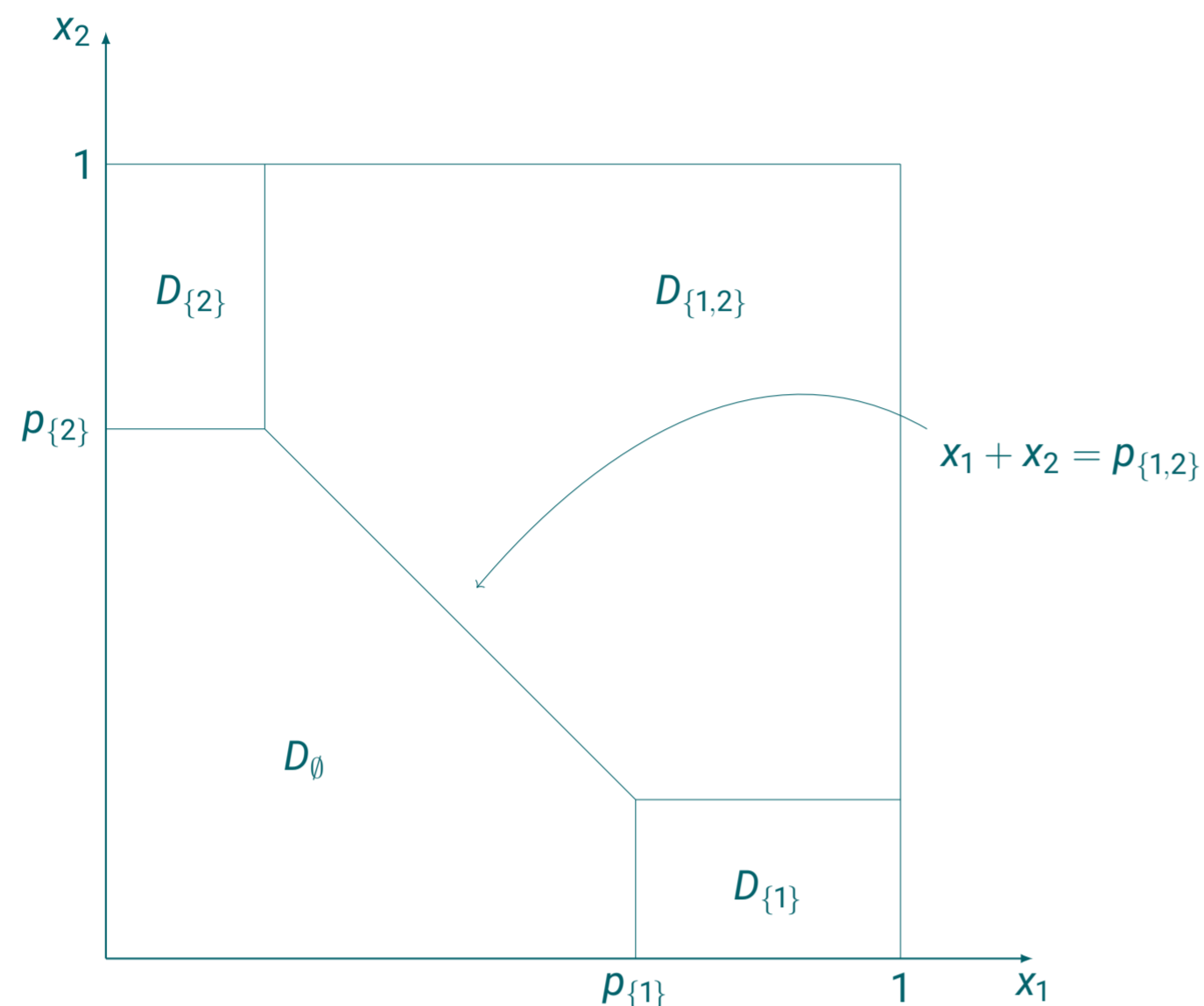
How to sell m items?

- + Define **prices p_J** for each bundle of items $J \subseteq \{1, \dots, m\}$
- + Buyer has valuation $x_j \in [0, 1]$ according to a probability function f_j for each item $j \in \{1, \dots, m\}$
- + Buyer takes **bundle which maximizes her utility**:

$$u(\mathbf{x}) = \max_{J \subseteq \{1, \dots, m\}} \left(\sum_{j \in J} x_j - p_J \right)$$

The set of all \mathbf{x} for which the buyer takes bundle J is called D_J
 $\rightarrow u$ is a *tropical polynomial* and the regions D_J are separated by its vanishing locus

Geometric representation for $m = 2$:



Our Goal: Choose optimal prices to **maximize the expected revenue**:

$$\text{Rev}(\mathbf{p}) = \sum_{J \subseteq \{1, \dots, m\}} \int_{D_J} p_J \, d\mathbf{f}(\mathbf{x}) = \int_{[0,1]^m} (\nabla u(\mathbf{x}) \cdot \mathbf{x} - u(\mathbf{x})) \, d\mathbf{f}(\mathbf{x})$$

Duality

- + Use the utility u instead of \mathbf{p} as optimization variable and define the **optimization program**:

$$\begin{aligned} \sup_u \int_{[0,1]^m} (\nabla u(\mathbf{x}) \cdot \mathbf{x} - u(\mathbf{x})) \, d\mathbf{f}(\mathbf{x}) \\ \text{s.t.: } 0 \leq \frac{\partial u(\mathbf{x})}{\partial x_j} \leq 1, \quad u: [0,1]^m \rightarrow \mathbb{R} \text{ convex} \end{aligned}$$

- + **Remark:** This generalizes the framework to lotteries - we sell tickets for the objects and the buyer gets object j with a probability of $\frac{\partial u(\mathbf{x})}{\partial x_j}$ (see [3])

- + The **dual optimization problem** (by [1]):

$$\begin{aligned} \inf_{z_1, \dots, z_m} \int_{[0,1]^m} \sum_{j=1}^m z_j(\mathbf{x}) \, d\mathbf{x} \\ \text{s.t.: } \sum_{j=1}^m \frac{\partial z_j(\mathbf{x})}{\partial x_j} \leq (m+1)f(\mathbf{x}) + \mathbf{x} \cdot \nabla f(\mathbf{x}) \\ z_j(0, \mathbf{x}_{-j}) = 0, \quad z_j(1, \mathbf{x}_{-j}) \geq 1, \quad z_j(\mathbf{x}) \geq 0, \quad j \in \{1, \dots, m\} \end{aligned}$$

where $z_j(\lambda, \mathbf{x}_{-j}) = z_j((x_1, \dots, x_{j-1}, \lambda, x_{j+1}, \dots, x_m))$

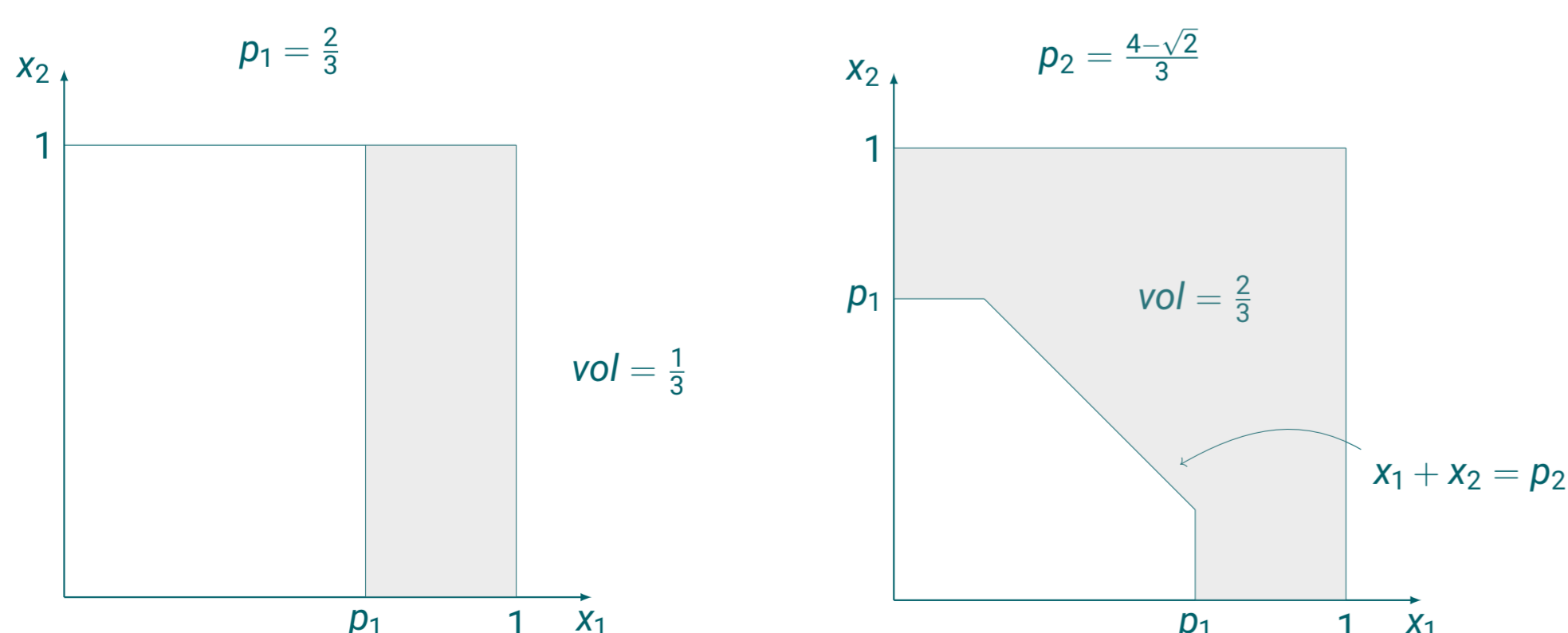
SJA-prices

- + Consider the case, where $f_j \equiv 1$ for all $j \in \{1, \dots, m\}$, so \mathbf{x} is **drawn uniformly**
- + Choose prices depending on the cardinality of the bundle, that is if $|J_1| = |J_2|$, then $p_{J_1} = p_{J_2} =: p_{|J_1|}$. If p_1, \dots, p_{r-1} are already defined, define p_r s.t.:

$$\text{vol} \left(\bigcup_{J \subseteq \{1, \dots, r\}} \left\{ \mathbf{x} \in [0,1]^m \mid \sum_{j \in J} x_j \geq p_{|J|} \right\} \right) = \frac{r}{m+1} \quad (1)$$

- + Proven to be **optimal for $m \leq 6$** (see [1])

SJA-prices for $m = 2$:



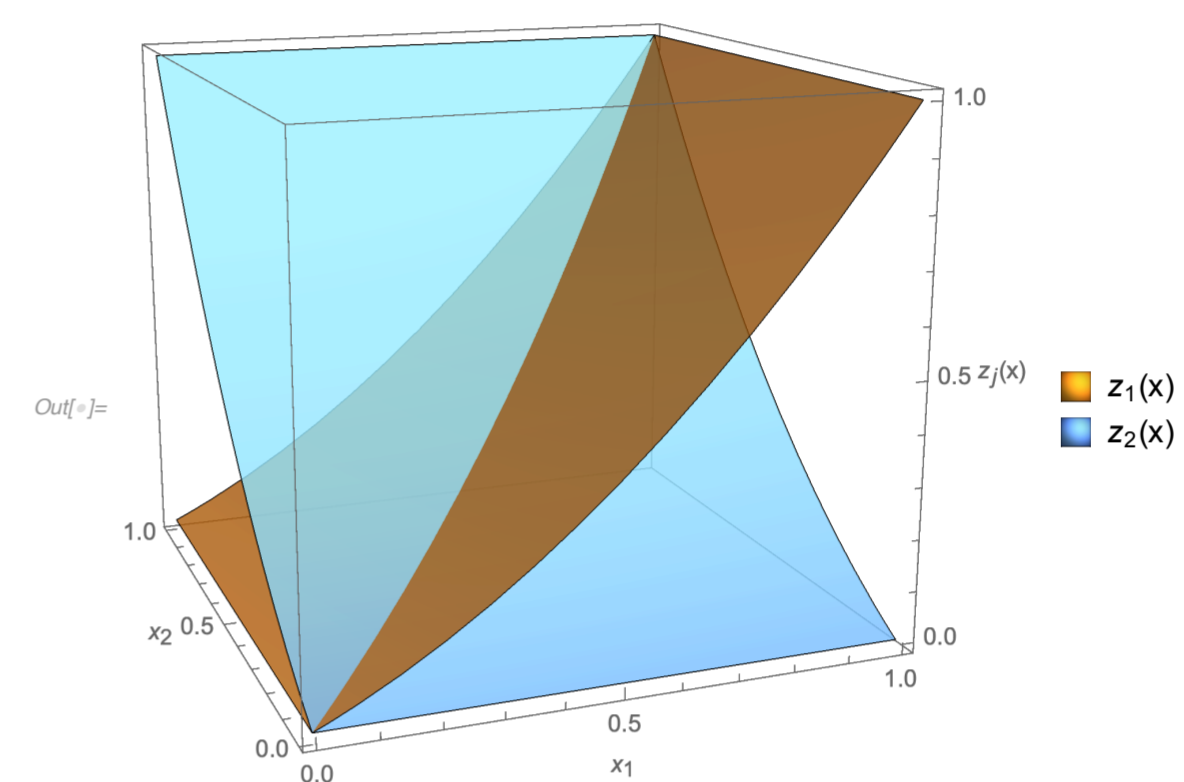
Duality for SJA

- + Consider the dual problem for uniform distributions of \mathbf{x} :

$$\begin{aligned} \inf_{z_1, \dots, z_m} \int_{[0,1]^m} \sum_{j=1}^m z_j(\mathbf{x}) \, d\mathbf{x} \\ \text{s.t.: } \sum_{j=1}^m \frac{\partial z_j(\mathbf{x})}{\partial x_j} \leq (m+1) \\ z_j(0, \mathbf{x}_{-j}) = 0, \quad z_j(1, \mathbf{x}_{-j}) \geq 1, \quad z_j(\mathbf{x}) \geq 0, \quad j \in \{1, \dots, m\} \end{aligned}$$

- + **Intuitive description:** We search functions z_1, \dots, z_m s.t. z_j starts with 0 at the "lower" facet (according to the j -th coordinate) of the $[0, 1]^m$ -dimensional hypercube and ends with at least 1 at the "upper" facet, while the sum of all z_j cannot grow too steeply (in their respective direction)

A feasible solution for $m = 2$:



Min-Cost-Flow

- + We want to **discretize** $\frac{\partial z_j(\mathbf{x})}{\partial x_j}$ on the m -dimensional $(N \times \dots \times N)$ -grid

- + **Construction of graph:**

Nodes: $V = U \cup B \cup \{s, t\}$, with
 $U = \{u_\alpha \mid \alpha \in \{1, \dots, N\}^m\}$ for the cells of the grid
 $B = \{b_{\alpha-j}^j \mid \alpha-j \in \{1, \dots, N\}^{m-1}, j \in \{1, \dots, m\}\}$ as supplementary nodes on the boundary.

Edges: $E = S \cup F \cup T$, with
 $S = \{(s, u) \mid u \in U\}$,
 $F = \{(u_\alpha, b_{\alpha-j}^j) \mid j \in \{1, \dots, m\}\}$,
 $T = \{(b, t) \mid b \in B\}$

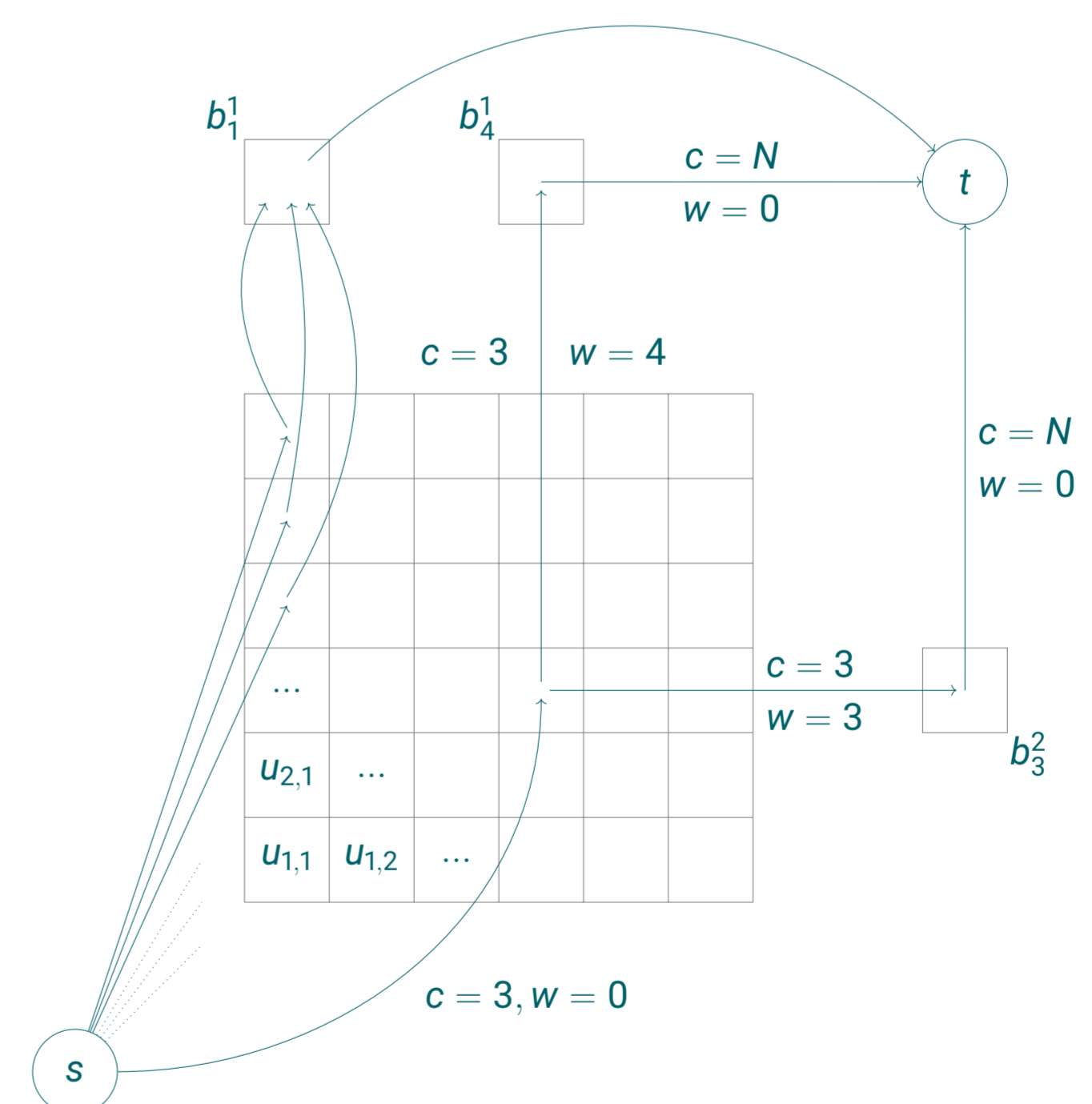
Capacities: $c(e) = \begin{cases} m+1, & e \in S \cup F \\ N, & e \in T \end{cases}$ Weights: $w(e) = \begin{cases} 0, & e \in S \cup T \\ N+1-\alpha_j, & e = (u_\alpha, b_{\alpha-j}^j) \end{cases}$

\rightarrow Edge-capacities simulate the constraints, weights simulate the objective value

- + Every $(s-t)$ -flow of capacity mN^m represents a feasible solution for the dual program

- + Get an **optimal solution** by searching a **min-cost-flow** for $N \rightarrow \infty$

Graph for $m = 2, N = 6$:



Advantages:

- + **Simpler proof** for the optimality of SJA-prices in the 2-item case than existing one
- + May be **adaptable for an arbitrary amount** of items (instead of just $m \leq 6$)
- + May help to find a proof for a greater class of distributions and thus fix a flaw in [2]

References

- [1] Y. Giannakopoulos and E. Koutsoupias. Duality and optimality of auctions for uniform distributions. *SIAM J. Comput.*, 47(1):121–165, 2018.
- [2] Y. Giannakopoulos and E. Koutsoupias. Selling two goods optimally. *Inf. Comput.*, 261(Part):432–445, 2018.
- [3] J. Rochet. The taxation principle and multi-time hamilton-jacobi equations. *Journal of Mathematical Economics*, 14(2):113 – 128, 1985.