

Interval analysis and Optimal Transport

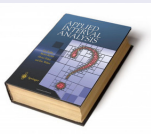
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LARIS - Université d'Angers - France

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<http://www.math.uu.se/swim2014/>

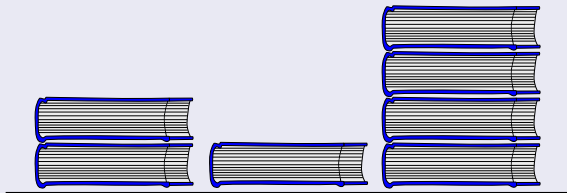
Outline

- 1 Introduction to Optimal Transport
 - Transportation
 - Optimal Transport
 - Some known results
- 2 A lower bound of the optimal value
 - Finite dimensional relaxation
- 3 An upper bound of the optimal value
 - Duality
 - Finite dimensional relaxation
- 4 Conclusion - Future work

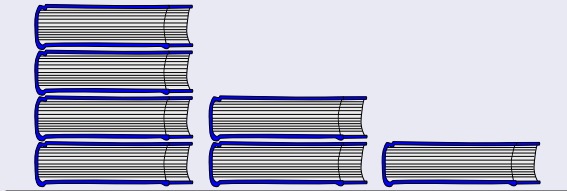
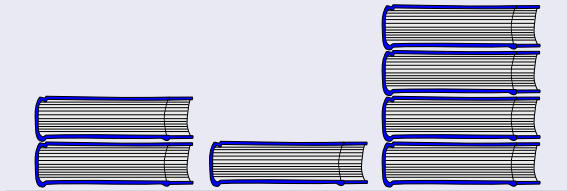
Example with books



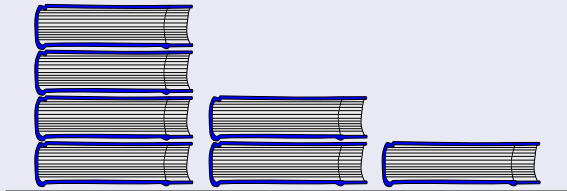
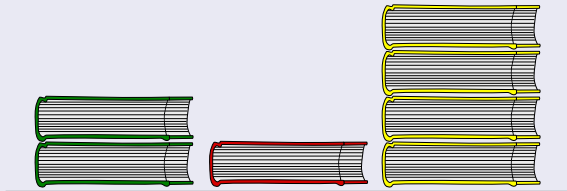
Example in the discrete case



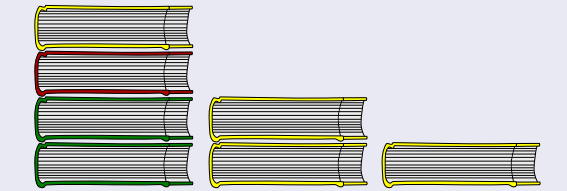
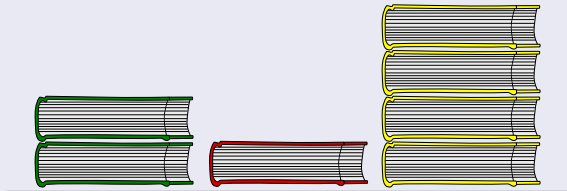
Example in the discrete case



Example in the discrete case



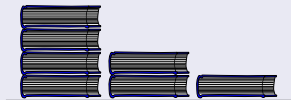
Example in the discrete case



Example in the discrete case



$$\mu = (2, 1, 4)$$



$$\nu = (4, 2, 1)$$

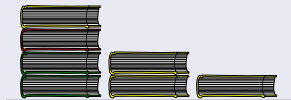
Transportation

	4	2	1
2			
1			
4			

Example in the discrete case



$$\mu = (2, 1, 4)$$



$$\nu = (4, 2, 1)$$

Transportation

	4	2	1
2			
1			
4			

Example in the discrete case



$$\mu = (2, 1, 4)$$



$$\nu = (4, 2, 1)$$

A plan transference π

	4	2	1	
2	2	0	0	
1	1	0	0	,
4	1	2	1	

$$\pi = \begin{pmatrix} 2 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 2 & 1 \end{pmatrix}$$

Plan transference problem

	4	2	1
2	.	.	.
1	.	.	.
4	.	.	.

Solutions

$$\pi = \begin{array}{c|ccc} & 4 & 2 & 1 \\ \hline 2 & 2 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 4 & 1 & 2 & 1 \end{array}, \quad \tilde{\pi} = \begin{array}{c|ccc} & 4 & 2 & 1 \\ \hline 2 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 \\ 4 & 2 & 2 & 0 \end{array}.$$

Definition - Transference plan

A transference plan (or a transportation) π is a measure on the product space $X \times Y$ such that

$$\begin{cases} \pi(A \times Y) = \mu(A), \\ \pi(X \times B) = \nu(B). \end{cases}$$

all measurable subsets A of X and B of Y .

In the discrete case

$$\begin{cases} \forall i, \sum_j \pi_{ij} = \mu_i, \\ \forall j, \sum_i \pi_{ij} = \nu_j. \end{cases}$$

Comparing two plan transferences

Transportation π 

$$I(\pi) = 0 + 0 + 1 + 2 + 1 + 1 + 0 = 5$$

Transportation $\tilde{\pi}$ 

$$I(\tilde{\pi}) = 0 + 2 + 1 + 2 + 2 + 1 + 1 = 9$$

In the discrete case

$$\begin{aligned} \min_{\pi \in \mathbb{R}^n \otimes \mathbb{R}^m} \quad & \sum_{i,j} c_{ij} \pi_{ij} \\ \text{subject to} \quad & \forall i, \sum_j \pi_{ij} = \mu_i, \\ & \forall j, \sum_i \pi_{ij} = \nu_j. \end{aligned} \tag{1}$$

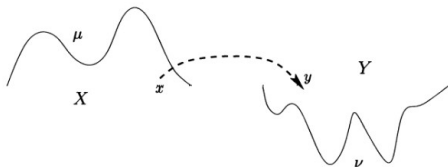
where c_{ij} are non negative real numbers which tells how much it costs to transport one unit of mass from location i to location j .

Kantorovich formulation

The *optimal transportation cost* between μ and ν is the value :

$$\begin{aligned} \mathcal{T}_c(\mu, \nu) = \inf_{\pi \in \mathcal{B}(X \times Y)} & \int_{X \times Y} c(x, y) d\pi(x, y) \\ \text{subject to} & \pi_X = \mu, \\ & \pi_Y = \nu \end{aligned} \quad (2)$$

The optimal π 's, i.e. those such that $I(\pi) = \mathcal{T}_c(\mu, \nu)$, if they exist, will be called *optimal transference plans*.

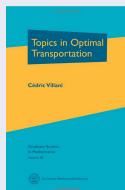


Remark

The *optimal transportation problem* is an infinite dimensional linear programming problem.

i.e. I is a linear cost function, and constraints are linear.

- 1 $c = \|x - y\|^p$, $p > 1$, the strict convexity of c guarantees that, if μ, ν are absolutely continuous with respect to Lebesgue measure, then there is a unique solution to the Kantorovich problem.
- 2 $c = \|x - y\|^2$, optimal transference plans are the (restrictions of) gradients of convex functions.
- 3 many others in



Topics in Optimal Transportation, Cédric Villani, AMS (2003)

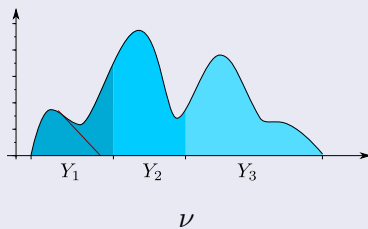
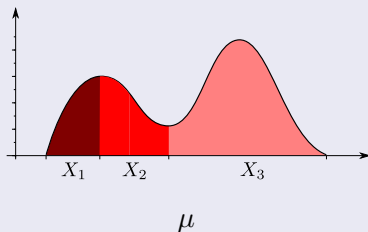
Proposition - Relaxation

Let μ and ν (with support X and Y) be absolutely continuous measures with respect to Lebesgue measure. If $\{X_i\}_i$ and $\{Y_j\}_j$ be finite pavings of X and Y . Suppose that $\mu(X_i) \in [\underline{\mu}_i, \bar{\mu}_i]$, $\nu(Y_j) \in [\underline{\nu}_j, \bar{\nu}_j]$, and $\forall x, y \in X_i \times Y_j, \underline{c}_{ij} \leq c(x, y)$,

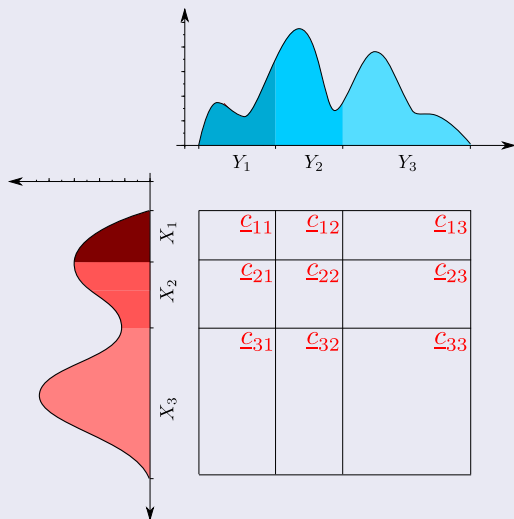
$$\begin{aligned} \mathcal{I} &= \min_{\pi_{ij} \in \mathbb{R}^n \otimes \mathbb{R}^m} \sum_{i,j} \underline{c}_{ij} \pi_{ij} \\ \text{subject to } \forall i, \underline{\mu}_i &\leq \sum_j \pi_{ij} \leq \bar{\mu}_i, \\ \forall j, \underline{\nu}_j &\leq \sum_i \pi_{ij} \leq \bar{\nu}_j, \\ \forall i, \forall j, \pi_{ij} &\geq 0. \end{aligned}$$

then $\mathcal{I} \leq \mathcal{T}_c(\mu, \nu)$.

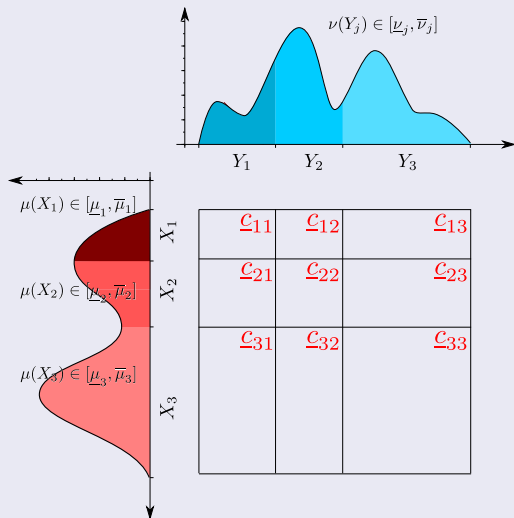
Spatial discretization



Spatial discretization



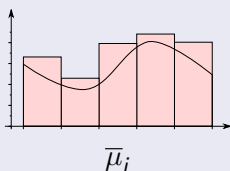
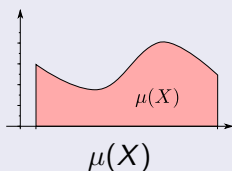
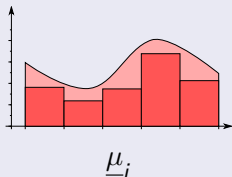
Spatial discretization



Enclosing

If $\mu = f(x)dx$, and $[f]$ an inclusion function for f then

$$\int_X f(x)dx \in \sum_i [f](X_i)\lambda(X_i)$$



$$\sum_i \underline{f}(X_i)\lambda(X_i) \leq$$

$$\int_X f(x)dx$$

$$\leq \sum_i \bar{f}(X_i)\lambda(X_i)$$

Proof

Let $\{X_i\}, \{Y_j\}$ be a pavings, let $\pi_{ij} = \pi(X_i \times Y_j)$ then
 $\forall \pi, \exists \xi_{ij} \in X_i \times Y_j,$

$$\sum_{i,j} c(\xi_{ij})\pi_{ij} = \int_{X \times Y} c(x, y) d\pi(x, y) \quad (3)$$

Since $\underline{c}_{ij} \leq c(\xi_{ij})$ and $\pi_{ij} \geq 0$, then
 $\forall \pi,$

$$\sum_{i,j} \underline{c}_{ij}\pi_{ij} \leq \int_{X \times Y} c(x, y) d\pi(x, y) \quad (4)$$

Proof

Let μ and ν (with support X and Y) be absolutely continuous measures with respect to Lebesgue measure. If $\{X_i\}_i$ and $\{Y_j\}_j$ be finite pavings of X and Y . Suppose that $\mu(X_i) \in [\underline{\mu}_i, \bar{\mu}_i]$, $\nu(Y_j) \in [\underline{\nu}_j, \bar{\nu}_j]$, and $\forall x, y \in X_i \times Y_j, \underline{c}_{ij} \leq c(x, y)$,

$$\begin{aligned} \mathcal{K} &= \min_{\pi_{ij} \in \mathbb{R}^n \otimes \mathbb{R}^m} \sum_{i,j} \underline{c}_{ij} \pi_{ij} \\ \text{subject to } \forall i, \mu_i &= \sum_j \pi_{ij} = \mu_i, \\ \forall j, \nu_j &= \sum_i \pi_{ij} = \nu_j, \\ \forall i, \forall j, \pi_{ij} &\geq 0. \end{aligned}$$

then $\mathcal{K} \leq \mathcal{T}_c(\mu, \nu)$.

Proof

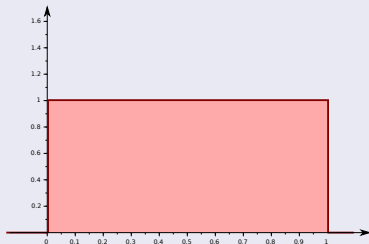
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$$\begin{aligned} \underline{\mathcal{T}} &= \min_{\pi_{ij} \in \mathbb{R}^n \otimes \mathbb{R}^m} \sum_{i,j} \underline{c}_{ij} \pi_{ij} \\ \text{subject to } \forall i, \underline{\mu}_i &\leq \sum_j \pi_{ij} \leq \bar{\mu}_i, \\ \forall j, \underline{\nu}_j &\leq \sum_i \pi_{ij} \leq \bar{\nu}_j, \\ \forall i, \forall j, \pi_{ij} &\geq 0. \end{aligned}$$

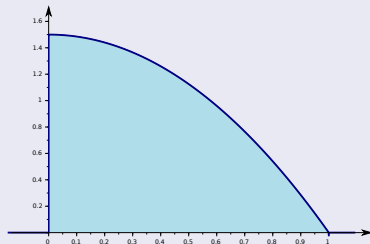
then $\underline{\mathcal{T}} \leq \mathcal{T}_c(\mu, \nu)$.

Example

$$X = Y = [0, 1]$$



$$\mu = 1 dx$$



$$\nu = 1.5(1 - y^2) dy$$

$$c(x, y) = \|x - y\|^2$$

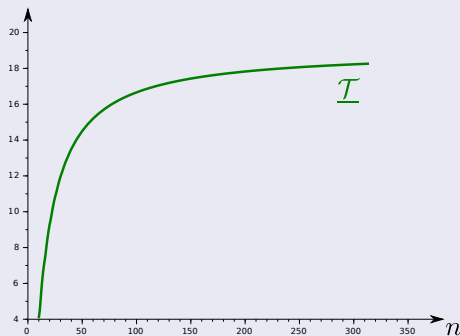


Figure : Guaranteed lower bounds of $\mathcal{T}_c(\mu, \nu)$ where $n = \text{Card}\{X_i\}_i$

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Linear programming - Duality

Primal problem

$$\begin{aligned} \min_{x \in \mathbb{R}^n} \quad & c^T x \\ \text{subject to} \quad & Ax = b, \\ & x \geq 0. \end{aligned}$$

Dual problem

$$\begin{aligned} \max_{y \in \mathbb{R}^m} \quad & b^T y \\ \text{subject to} \quad & y_i \in \mathbb{R}, \\ & A^T y \leq c. \end{aligned} \tag{5}$$

Duality

$$\begin{aligned} & \inf_{\pi \in \mathcal{B}(X \times Y)} \int_{X \times Y} c(x, y) d\pi(x, y) \\ & \text{subject to } \pi_X = \mu, \\ & \pi_Y = \nu \end{aligned}$$

$$\begin{aligned} & \sup_{\phi, \psi \in \mathcal{C}_b(X, Y)} \int_X \phi(x) d\mu(x) + \int_Y \psi(y) d\nu(y) \\ & \text{subject to } \phi(x) + \psi(y) \leq c(x, y). \end{aligned} \tag{6}$$

where $\mathcal{C}_b(X, Y)$ denotes the set of all pairs of bounded and continuous functions $\phi : X \rightarrow \mathbb{R}$ and $\psi : Y \rightarrow \mathbb{R}$.

If X is compact and Hausdorff, $\mathcal{C}_b(X)^* = \{\text{Radon measure}\}$

Kantorovich Duality

The minimum of the Kantorovich problem is equal to

$$\begin{aligned} \mathcal{T}_c(\mu, \nu) = & \sup_{\phi, \psi \in \mathcal{C}_b(X, Y)} \int_X \phi(x) d\mu(x) + \int_Y \psi(y) d\nu(y) \\ & \text{subject to } \phi(x) + \psi(y) \leq c(x, y). \end{aligned} \quad (7)$$

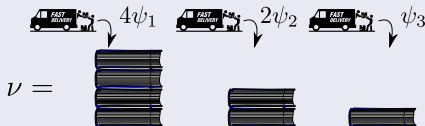
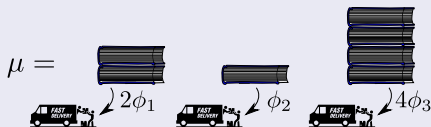
Interpretation in the discrete case

$$\begin{aligned} & \sup_{(\phi_i) \in \mathbb{R}^n, (\psi_j) \in \mathbb{R}^m} \sum_i \phi_i \mu_i + \sum_j \psi_j \nu_j \\ & \text{subject to} \quad \phi_i + \psi_j \leq c_{ij} \end{aligned} \tag{8}$$



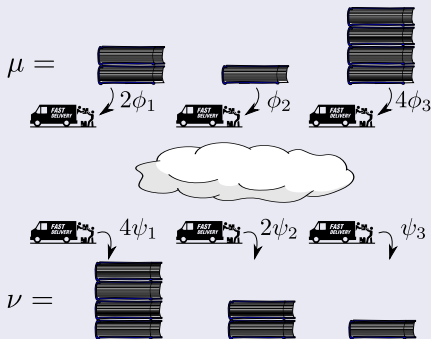
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Interpretation in the discrete case

$$\begin{aligned} & \sup_{(\phi_i) \in \mathbb{R}^n, (\psi_j) \in \mathbb{R}^m} \sum_i \phi_i \mu_i + \sum_j \psi_j \nu_j \\ & \text{subject to} \quad \phi_i + \psi_j \leq c_{ij} \end{aligned} \tag{10}$$



Proposition - Relaxation

Let μ and ν (with support X and Y) be absolutely continuous measures with respect to Lebesgue measure. If $\{X_i\}_i$ and $\{Y_j\}_j$ be finite pavings of X and Y . Suppose that $\mu(X_i) \in [\underline{\mu}_i, \bar{\mu}_i]$, $\nu(Y_j) \in [\underline{\nu}_j, \bar{\nu}_j]$, and $\forall x, y \in X_i \times Y_j, c(x, y) \leq \bar{c}_{ij}$,

$$\begin{aligned} \bar{\mathcal{T}} = & \sup_{(\phi_i) \in \mathbb{R}^n, (\psi_j) \in \mathbb{R}^m} \sum_i \phi_i \bar{\mu}_i + \sum_j \psi_j \bar{\nu}_j \\ & \text{subject to} \quad \phi_i + \psi_j \leq \bar{c}_{ij} \end{aligned} \quad (11)$$

then $\mathcal{T}_c(\mu, \nu) \leq \bar{\mathcal{T}}$.

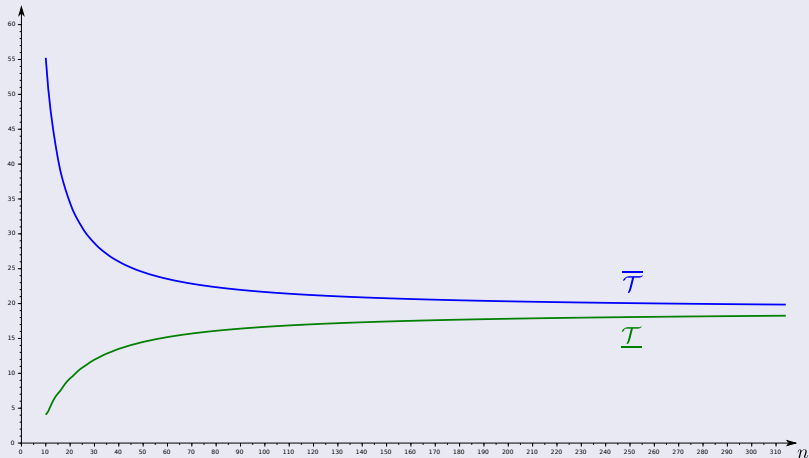


Figure : Guaranteed upper bounds of $\mathcal{T}_c(\mu, \nu)$ where $n = \text{Card}\{X_i\}_i$

Software

- filib - FI_LIB - A fast interval library,
<http://www2.math.uni-wuppertal.de/~xsc/software/filib.html>
- GLPK - GNU Linear Programming Kit (GLPK),
<http://www.gnu.org/software/glpk/>
- GMP - GNU Multiple Precision Arithmetic Library,
<https://gmplib.org/>
- Source code is available on my webpage.

Future work

- Compute guaranteed enclosures of the solution combining linear programming and constraint propagation.
- Generalize this methodology to other problems (D. Henrion & J.B. Lasserre):
 - Probability and Markov Chains
 - Optimal Control with occupation measures (ODE),
 - Others as in *Moments, Positive Polynomials and Their Applications*, J.B Lasserre, Imperial College Press Optimization Series (2009)

Tack för din uppmärksamhet !