

# Optimal Mechanism for Selling Two Goods with Uniformly Distributed Valuations

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1 Aug 2015

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- ▶ Based on the value of  $\hat{z}$ , the auctioneer decides the allocation probability  $q : D \rightarrow [0, 1]^2$ , and the payment from the buyer  $t : D \rightarrow \mathbb{R}$ .
- ▶ Consider a quasi-linear mechanism, where the utility function of the buyer is given by  $u(z, \hat{z}) = z \cdot q(\hat{z}) - t(\hat{z})$ .

## Optimal Auctions

- ▶ **Auctioneer's objective:** Maximize  $\mathbb{E}_z t(z)$ , subject to IC and IR constraints.
- ▶ **IC:**  $u(z) \geq u(z, \hat{z}) \forall z, \hat{z} \in D$ .
- ▶ **IR:**  $u(z) \geq 0 \forall z \in D$ .

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- ▶ Define  $\phi(z) = z - \frac{1-F(z)}{f(z)}$ . Assume that  $\phi$  is increasing. Then the optimal allocation is given by

$$q(z) = \begin{cases} 1 & \text{if } \phi(z) \geq 0 \\ 0 & \text{if } \phi(z) \leq 0. \end{cases}$$

- ▶ Myerson has also provided the solution for the single-item  $n$ -buyer setting.

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- ▶ The solution is known for cases when the distributions  $f_1$  and  $f_2$  give rise to a so-called *well-formed* partition of the support set  $D$ .
- ▶ Cases such as  $f = \text{Unif}[0, b_1] \times [0, b_2]$ ,  $f = \exp(\lambda_1) \times \exp(\lambda_2)$ , and  $f = \text{Beta} \times \text{Beta}$  belong to this category.

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- ▶ The optimal solution for all these cases was given by Daskalakis et al.<sup>2</sup>
- ▶ In this talk, we shall derive the formulation of the optimization problem as done by Daskalakis et al., discuss their solutions when  $f = Unif[0, b_1] \times [0, b_2]$ , and then the work that we have done to find the optimal solution for the case when  $f = Unif[c_1, c_1 + b_1] \times [c_2, c_2 + b_2]$ .

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## Primal problem

- ▶ Recall that the auctioneer's objective was to maximize  $\mathbb{E}_z t(z)$  w.r.t. IC and IR constraints.
- ▶ Rochet's theorem provides a necessary and sufficient condition for a mechanism to be IC and IR.

### Theorem

A quasi-linear mechanism satisfies IC and IR, iff  $u(\cdot)$  is convex,  $\nabla u(z) = q(z)$  a.e.  $z \in D$ , and  $u(z) \geq 0 \forall z \in D$ .<sup>3</sup>

- ▶ This theorem helps us formulate the optimization problem completely in terms of  $u$ .

$$\begin{aligned} & \max_u \int_D (z \cdot \nabla u(z) - u(z)) f(z) dz \\ & \text{s.t. } \{u \text{ convex, } \nabla u(z) \in [0, 1]^2 \text{ a.e. } z, \text{ and } u(z) \geq 0 \forall z.\} \end{aligned}$$

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<sup>3</sup>J.C.Rochet, "The Taxation Principle and the Multi-time Hamilton Jacobi Equations", Journal of Mathematical Economics 14, 2 (April 1985), 113-128.

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 $c(x, y) = (x_1 - y_1)_+ + (x_2 - y_2)_+$ .



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- ▶  $u(z) \geq 0 \Leftrightarrow u(0, 0) \geq 0$ , since  $u(0, 0) \geq 0$  combined with  $\nabla u \geq 0$  implies  $u(z) \geq 0$ .
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- ▶ We further consider  $u(0, 0) = 0$ , since fixing  $u(0, 0) > 0$  only reduces the objective function.
- ▶ The optimization problem now becomes

$$\max_u \int_D (z \cdot \nabla u(z) - u(z)) f(z) dz$$

$$\text{s.t. } \{u \text{ convex, } u(x) - u(y) \leq c(x, y) \forall (x, y), \text{ and } u(0, 0) = 0.\}$$

## Primal problem (contd...)

- ▶ Using integration by parts, the objective function can be written as,

$$\int_D (z \cdot \nabla u(z) - u(z)) f(z) dz = \int_D u d(\mu + \mu_s)$$

- ▶  $\mu(z) := -z \cdot \nabla f(z) - 3f(z)$ ,  $\mu_s(z) := f(z)(z \cdot n)\mathbf{1}(z \in \partial D)$ .

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- ▶  $\mu$  is the density of a measure absolutely continuous w.r.t.  $\mathcal{L}_2$ .  $\mu_s$ , w.r.t.  $\mathcal{L}_1$ ,  $n$  is the normal to the surface  $\partial D$ .

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- ▶ We have  $\int_D d(\mu + \mu_s) = -1$ .
- ▶ To make this 0, we add a point measure  $\mu_p$  of 1 at  $(0, 0)$ .
- ▶ Defining  $\bar{\mu} = \mu + \mu_s + \mu_p$ , we have the objective function to be  $\int_D u d\bar{\mu}$ . Observe that defining  $\mu_p$  causes no harm to the objective function since  $u(0, 0) = 0$ .

# Primal and the dual

**The Primal problem:**

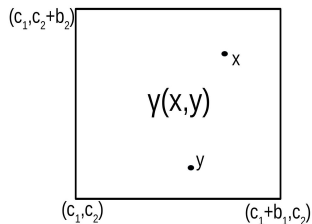
$$\begin{aligned} & \max_u \int_D u d\bar{\mu} \\ \text{s.t. } & \{u \text{ convex}, u(x) - u(y) \leq c(x, y) \forall (x, y), u(0, 0) = 0.\} \end{aligned}$$

**The Dual problem:**

$$\begin{aligned} & \inf_{\gamma} \int_{D \times D} c(x, y) d\gamma(x, y) \\ \text{s.t. } & \{\gamma(\cdot, D) = \gamma_1(\cdot), \gamma(D, \cdot) = \gamma_2(\cdot), \text{ and } \gamma_1 - \gamma_2 \succeq_{\text{cvx}} \bar{\mu}.\} \end{aligned}$$

where we say the measure  $\alpha$  *convex dominates* measure  $\beta$  ( $\alpha \succeq_{\text{cvx}} \beta$ ) if for every convex and increasing function, we have  $\int_D f d\alpha \geq \int_D f d\beta$ .

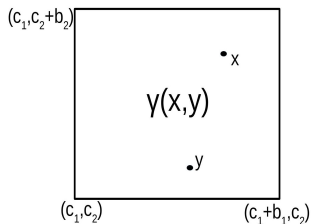
# The Optimal Transport problem



- ▶ The dual problem is a version of *optimal transport* problem.

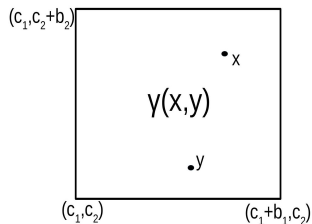


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- ▶  $c(x, y) \rightarrow$  Cost of transporting unit mass from  $x$  to  $y$ .
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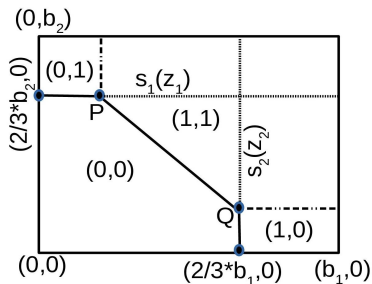
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- ▶  $\gamma(x, y) \rightarrow$  The differential mass transported from  $x$  to  $y$ .
- ▶ We need to find a way to minimize the cost of transportation subject to the constraint that  $\gamma_1 - \gamma_2 \succeq_{c \vee x} \bar{\mu}$ .

## The solution by Daskalakis et al.

- ▶ The problem of optimal transport was solved by Daskalakis et al. for  $f_1$  and  $f_2$  that give rise to a *well-formed* canonical partition of the support set  $D$ .

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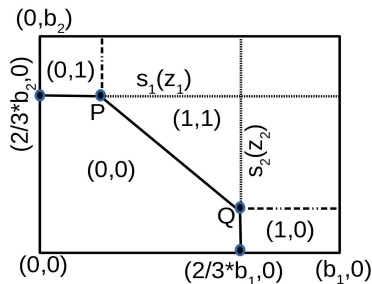
- ▶ The problem of optimal transport was solved by Daskalakis et al. for  $f_1$  and  $f_2$  that give rise to a *well-formed* canonical partition of the support set  $D$ .
- ▶ The solution when  $f = \text{Unif}[0, b_1] \times [0, b_2]$  is:



where the line joining  $P$  and  $Q$  is  $z_1 + z_2 = \frac{2b_1 + 2b_2 - \sqrt{2b_1b_2}}{3}$ .

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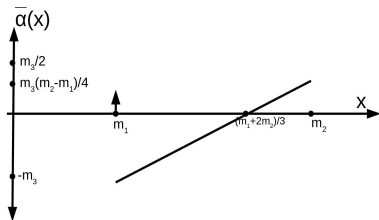
- ▶ The optimal  $\gamma$  that they provide is such that  $\gamma_1 - \gamma_2 = \bar{\mu}$ .

Uniform  $[c_1, c_1 + b_1] \times [c_2, c_2 + b_2]$

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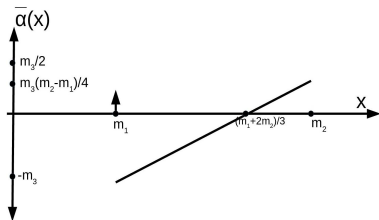
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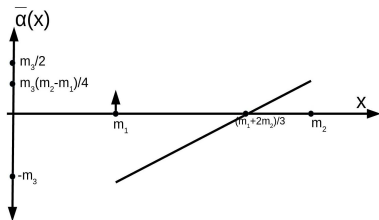


- ▶ One can prove that  $\bar{\alpha} \succeq_{\text{cvx}} 0$  for any  $m_2 \geq m_1 \geq 0$ ,  $m_3 \geq 0$ .



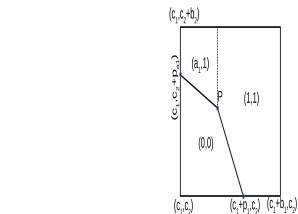
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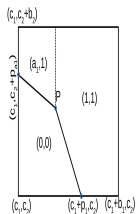


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- ▶ We thus find a solution which does not relax the convexity constraint.

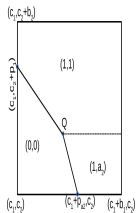
# The general structure



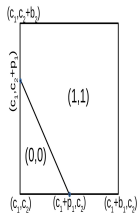
(a)



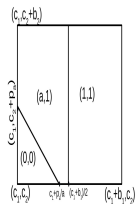
(b)



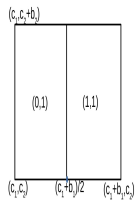
(f)



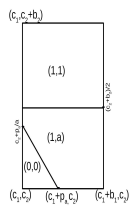
(c)



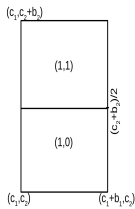
(d)



(e)



(g)



(h)

# Conjecture

## Conjecture

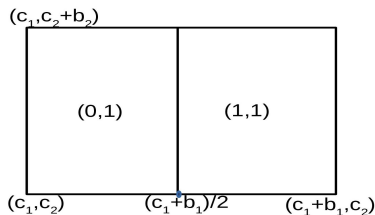
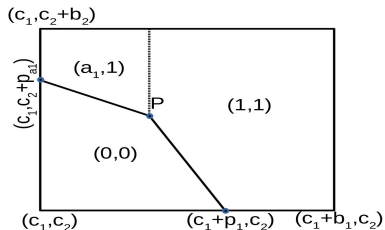
*Consider  $z \sim \text{Unif}[c_1, c_1 + b_1] \times [c_2, c_2 + b_2]$ . The structure of the optimal solution for any  $(c_1, c_2, b_1, b_2) \in \mathbb{R}_+^4$  is one among the eight structures (a)-(h).*

In other words, defining  $E_x$  to be the set of all  $(c_1, c_2, b_1, b_2)$  such that the optimal solution has the structure “x”, “x” taking any alphabet from (a) to (h), we conjecture that  $\bigcup_{x=a}^h E_x = \mathbb{R}_+^4$ .

# Structures when $c_1 = 0$

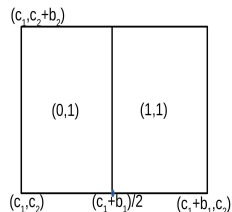
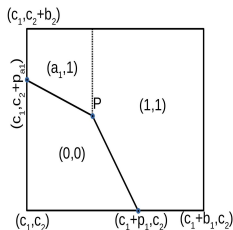
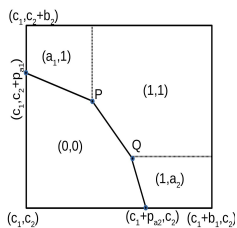
## Theorem

Consider  $z \sim \text{Unif}[0, b_1] \times [c_2, c_2 + b_2]$ . Then the optimal solution has one of the following structures when  $\frac{b_1}{b_2} \geq 2$ :



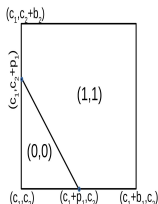
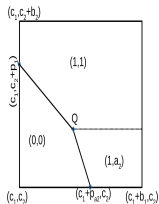
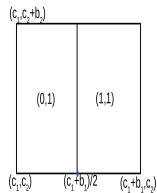
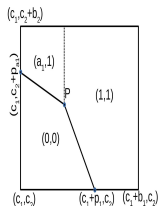
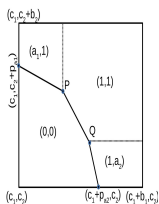
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