

# Optimal transport between unequal dimensions

Brendan Pass (joint work R. McCann and partially with P.-A. Chiappori)

University of Alberta

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- Setting: probability measures  $d\mu = f dx$  on  $X \subset \mathbb{R}^m$  and  $d\nu = g dy$  on  $Y \subseteq \mathbb{R}^n$  and a surplus  $s(x, y)$ .
- $\Pi(\mu, \nu)$  is the set of probability measures on  $X \times Y$  which *project* to  $\mu$  and  $\nu$ ; that is  $\gamma(B \times Y) = \mu(B)$ ,  $\gamma(X \times A) = \nu(A)$  for all  $B \subset X, A \subset Y$ .
- **Monge-Kantorovich problem:** maximize the linear functional

$$\int_{X \times Y} s(x, y) d\gamma(x, y)$$

over the convex set  $\gamma \in \Pi(\mu, \nu)$ .

- Usually, take  $m = n$ .

# Motivation for unequal dimensional problem

- Matching in economics:  $X$  and  $Y$  might parameterize:
  - Buyers and sellers.
  - Firms and employees in a labour market
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- These may well be different.
- Other applications?

- The problem is **dual** to minimizing

$$\int_X u(x) d\mu(x) + \int_Y v(y) d\nu(y)$$

among  $u \in L^1(\mu)$ ,  $v \in L^1(\nu)$  such that  $u(x) + v(y) \geq s(x, y)$ .

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- We have  $u(x) + v(y) - s(x, y) = 0$   $\gamma$  a.e. and so

$$\begin{aligned} Du(x) &= D_x s(x, y), & Dv(y) &= D_y s(x, y) \\ D^2 u(x) &\geq D_{xx}^2 s(x, y), & D^2 v(y) &\geq D_{yy}^2 s(x, y) \end{aligned}$$

# Optimal transport preliminaries

- We will assume:

*Twist*:  $y \mapsto D_x s(x, y)$  is injective (for each fixed  $x$ ); therefore  $m \geq n$ .

*Non-degeneracy*:  $D_{xy}^2 s(x, y)$  has full rank at each  $(x, y)$ ; that is, the rank is  $n$ .

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- Caffarelli '92: For  $s(x, y) = x \cdot y$ , regularity holds for convex target  $Y$  and good  $f, g$ .
- Ma-Trudinger-Wang '05, Loeper '10...: Same holds for other  $s(x, y)$  satisfying a **deep structural condition**. If not, one can find densities for which the optimal map is discontinuous.

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- Can they be characterized by a PDE?
- Do we expect smoothness?



# Smoothness fails

## Theorem (P '12)

*There exists smooth densities  $f, g$ , bounded above and below, for which the optimal map is discontinuous, UNLESS  $s$  is of index form; that is,*

$$s(x, y) = b(I(x), y) + \alpha(x)$$

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- Example: If  $Y$  is an  $n$ -dimensional submanifold of  $\mathbb{R}^m$ , and  $s(x, y) = x \cdot y$ , regularity theory requires  $Y$  to be convex, which can only happen if  $Y$  is affine. In this case,  $x \cdot y = x_Y \cdot y$ , where  $x_Y \in \mathbb{R}^n$  is the projection of  $x$  onto  $Y$ .

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- Consequence: any regularity result will have to depend on the **interaction** between  $s$  and  $\mu, \nu$ .

# A partial differential equation for unequal dimensions

- When  $m > n$ , we guess:

$$g(y) = \int_{T^{-1}(y)} \frac{f(x)}{\sqrt{\det[DT(x)DT(x)^T]}} d\mathcal{H}^{m-n}(x)$$

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- Except for the **domain of integration**,  $T$  is **eliminated**, (almost) leaving a PDE for  $v$ .

# Partial differential equation cont.

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for  $p = Dv(y)$ ,  $Q = D^2v(y)$ .

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$$\begin{aligned} T^{-1}(y) &\subseteq \partial_s v(y) := \{x : v^s(x) + v(y) = s(x, y)\} \\ &\subseteq X_2(y, p, Q) := \{x \in X_1(y, p), D_{yy}^2 s(x, y) \leq Q\} \\ &\subseteq X_1(y, p) := \{x : D_y s(x, y) = p\} \end{aligned}$$

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- Replacing  $T^{-1}(y)$  with either  $X_i$  results in a **local** PDE.



# Characterizing solutions via a non-local PDE

## Theorem (McCann -P. '17)

If  $(u, v) = (v^s, u^s)$  are  $s$ -conjugate functions, then they maximize the dual problem if and only if  $v$  solves the **non-local** equation

$$g(y) = \int_{\partial_s v(y)} \frac{f(x) \det[D^2 v(y) - D_{yy}^2 s(x, y)]}{\sqrt{\det[D_{yx}^2 s(x, y) D_{yx}^2 s(x, y)^T]}} d\mathcal{H}^{m-n}(x)$$

$\mathcal{H}^n$  almost everywhere.

# Solving via local PDE

Local equation:  $g(y) = G_i(y, Dv(y), D^2v(y))$ , where

$$G_i(y, p, Q) = \int_{X_i(y,p,Q)} \frac{f(x) \det[Q - D_{yy}^2 s(x, y)]}{\sqrt{\det[D_{yx}^2 s(x, y) D_{yx}^2 s(x, y)^T]}} d\mathcal{H}^{m-n}(x)$$

for  $i = 1, 2$ .

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*Assume  $(u, v) = (v^s, u^s)$  with  $v \in C^2(Y)$ . If  $v$  solves the local equation with  $i = 2$ , then  $(u, v)$  solves the dual problem.*

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*Conversely, assume also that  $u \in C^2(X)$ , and that each  $X_i(y, Dv(y), D^2v(y))$  is connected. Then, if  $(u, v)$  solves the dual problem,  $v$  solves the local equation a.e.*

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Remark:  $G_2$  is elliptic, so if  $v \in C^{2,\alpha}$  one can bootstrap and get that  $v$  is as smooth as  $G_2$  and  $g$ .

# One dimensional targets (joint with McCann and Chiappori)

- Take  $m > n = 1$ . We (try to) construct the optimal map explicitly one level set at a time, assuming  $T^{-1}(y) = X_1(y, p)$  for some  $p$ .

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- For fixed  $y$ , let  $k(y)$  be the (unique)  $k \in \mathbb{R}$  which *splits the population proportionately at  $y$* ; that is,

$$\mu(X_{\leq}(y, p)) = \nu((-\infty, y))$$

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- We say  $(s, \mu, \nu)$  is **nested** if, whenever  $y_0 < y_1$ ,

$$X_{\leq}(y_0, p(y_0)) \subseteq X_{<}(y_1, p(y_1)).$$



# Solution Theorem (roughly stated)

## Theorem (Chiappori-McCann-P ('16))

*If  $(s, \mu, \nu)$  is nested, then the map that assigns each  $x \in X_1(y, p(y))$  to  $y = T(x)$  is the unique optimizer.*

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- In this case, the  $i = 1$  version of the local equation holds:

$$g(y) = \int_{X_1(y, p(y))} \frac{p'(y) - s_{yy}(x, y)}{|D_x s_y|} f(x) d\mathcal{H}^{m-1}(x)$$

(Note that  $p(y) = v'(y)$ .)

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- The **velocity** of  $X_1(y, p(y))$  is given by  $\frac{p' - s_{yy}}{|D_x s_y|^2} D_x s_y$ . If the speed  $(p'(y) - s_{yy}(y, x))/|D_x s_y|$  is **positive** for each  $x \in X_1(y, p(y))$  and each  $y$ , the model is nested. Conversely, if nestedness holds, the speed is always **nonnegative**, and **positive for at least one**  $x \in X_1(y, p(y))$  for each  $y$ .

# Regularity for nested data when $n = 1$

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- To ensure smoothness of  $h$ , we need smoothness and bounds on  $f, g, s, \partial X$ , as well as **nondegeneracy**, and **transversality** of the intersection of  $\partial X$  and the  $X_1(y, p(y))$  (or at least  $H^{m-1}(\partial X \cap X_1(y, p(y))) = 0$  locally). We can differentiate these using a generalized divergence theorem and the coarea formula.

Example:

$$\begin{aligned}h_p(y, p(y)) &= \int_{X_1(y,p)} f(x) \frac{d\mathcal{H}^{m-1}(x)}{|D_x s_y(x, y)|} \\ &= \int_{X_{\leq}(y,p)} \nabla \cdot \left( f(x) \frac{D_x s_y(x, y)}{|D_x s_y(x, y)|^2} \right) d\mathcal{H}^m(x) \\ &\quad - \int_{\partial X \cup \bar{X}_{\leq}(y,p)} f(x) \frac{D_x s_y(x, y) \cdot \hat{n}_X}{|D_x s_y(x, y)|^2} d\mathcal{H}^{m-1}(x)\end{aligned}$$

$$h_y(y, p(y)) = \int_{X_1(y,p)} f(x) \frac{-s_{yy}(x, y) d\mathcal{H}^{m-1}(x)}{|D_x s_y(x, y)|} - g(y).$$

## Theorem (Chiappori-McCann-P ('16))

Fix an integer  $r \geq 1$ . Suppose there is an interval  $Y' = (y_0, y_1) \subset Y$  such that  $X' \cap \partial X \in C^1$  intersects  $\overline{X(y, k(y))}$  transversally for all  $y \in \overline{Y'}$ , where  $X' = \bigcup_{y \in Y'} \overline{X_1(y, p(y))}$ . Then  $\|p\|_{C^{r,1}(Y')}$  is controlled by the following quantities, all assumed positive and finite:  $\|\log f\|_{C^{r-1,1}(X')}$ ,  $\|\log g\|_{C^{r-1,1}(Y')}$ ,  $\|s_y\|_{C^{r,1}(X' \times Y')}$ ,  $\|\hat{n}_X\|_{(C^{r-2,1} \cap W^{1,1})(X' \cap \partial X)}$ ,  $\mathcal{H}^{m-1}[\partial^* X]$ ,

$$\inf_{y \in Y'} \mathcal{H}^{m-1}[X(y, k(y))] \quad (\text{proximity to ends of } Y) \quad (1)$$

$$\inf_{x \in X', y \in Y'} |D_x s_y(x, y)| \quad (\text{non-degeneracy}), \quad (2)$$

$$\inf_{x \in X' \cap \partial X, y \in Y'} 1 - (\hat{n}_X \cdot \hat{n}_{X_-})^2 \quad (\text{transversality}) \quad (3)$$

where  $\hat{n}_{X_-}(x, y) = D_x s_y / |D_x s_y|$ , and  $\mathcal{H}^{m-2}[\overline{X(y_0, k(y_0))} \cap \partial X]$ .



- The optimal map  $T$  satisfies  $\frac{\partial s}{\partial y}(x, T(x)) = p(T(x)) = v'(T(x))$ . With a *speed limit condition*,  $p'(y) - s_{yy}(x, y) > 0$ , this is as smooth as  $\frac{\partial s}{\partial y}$  and  $v' = p$  via the implicit function theorem.

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- The potential  $u$  is then one derivative smoother than  $T$  via  $Du(x) = D_x s(x, T(x))$ .

# Thanks

- Thank you!