

Generalized Kantorovich-Rubinstein Duality beyond Hausdorff and Kantorovich

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Kantorovich - Rubinstein Duality

discrete prob. distr.
↓
on X

Metric on X $\longrightarrow$ Metric on $\mathcal{D}X$

Method 1: Transport plans $\mathcal{S} \in \mathcal{D}(X \times X)$ s.t. $\sum_y \mathcal{S}(x, y) = \mu(x)$
 $\sum_x \mathcal{S}(x, y) = \nu(y)$

0.1	0	0	0.1
0.1	0.2	0	0.3
0.1	0.2	0.3	0.6
0.3	0.4	0.3	1

$\left. \begin{array}{c} \text{ } \\ \text{ } \\ \text{ } \end{array} \right\} \nu$

$\underbrace{\hspace{10em}}_{\mu}$

Cost of the plan $\mathcal{S}$:

$$\sum_{x,y} \mathcal{S}(x,y) \cdot d(x,y) = \mathbb{E}_{\mathcal{S}}(d)$$

Distance: $\mathcal{J}^d(\mu, \nu) = \inf \{ \mathbb{E}_{\mathcal{S}}(d) \mid \mathcal{S} \text{ is a transport plan} \}$

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Method 2: Price functions $f(x)$ = price at which a logistics firm buys
 $g(y)$ = price at which it sells back

s.t. $g(y) - f(x) \leq d(x, y)$: not cheaper to transport yourself

Distance: $\delta^K(d)(\mu, \nu) = \sup \{ \mathbb{E}_\nu(g) - \mathbb{E}_\mu(f) \mid (f, g) \text{ valid pair of price functions} \}$

Theorem The distances coincide: $\delta^K(d)(\mu, \nu) = \delta^W(d)(\mu, \nu)$

Predicate Liftings & Couplings

$$\begin{array}{ccc} \text{Expectation } \mathbb{E}: & (X \rightarrow [0,1]) & \longrightarrow (\mathcal{D}X \rightarrow [0,1]) \\ & f & \longmapsto (\mu \mapsto \mathbb{E}_\mu(f)) \end{array}$$


$$\text{Predicate lifting } \lambda: (X \rightarrow [0,1]) \longrightarrow (FX \rightarrow [0,1])$$

$$\begin{array}{l} \text{Transport plan: } g \in \mathcal{D}(X \times Y) \text{ s.t. } \mathcal{D}\pi_1(g) = \mu, \mathcal{D}\pi_2(g) = \nu \\ \text{Coupling: } t \in F(X \times Y) \text{ s.t. } F\pi_1(t) = t_1, F\pi_2(t) = t_2 \end{array}$$


Generalized Constructions (Fuzzy Relations)

$r: X \times Y \rightarrow [0,1]$ $t_1 \in FX$ $t_2 \in FY$ λ pred. lifting for F

Wasserstein lax extension:

$$W_\lambda(r)(t_1, t_2) = \inf \{ \lambda_{X \times Y}(r)(t) \mid t \text{ is coupling of } t_1, t_2 \}$$

Kantorovich lax extension:

truncated subtraction: $y \ominus x = \max(y - x, 0)$

$$K_\lambda^{\text{rel}}(r)(t_1, t_2) = \sup \{ \lambda_Y(g)(t_2) \ominus \lambda_X(f)(t_1) \mid (f, g) \text{ } r\text{-nonexp.} \}$$

$$(f: X \rightarrow [0,1], g: Y \rightarrow [0,1]) \text{ } r\text{-nonexp.} \Leftrightarrow \forall x, y. g(y) - f(x) \leq r(x, y)$$

Generalized Constructions (Pseudometrics)

d pseudometric on X $t_1, t_2 \in FX$ λ pred. lifting for F

Wasserstein lifting:

$$W_\lambda(d)(t_1, t_2) = \inf \{ \lambda_{X \times X}(d)(t) \mid t \text{ is coupling of } t_1, t_2 \}$$

Kantorovich lifting:

$$K_\lambda(d)(t_1, t_2) = \sup \{ | \lambda_X(f)(t_2) - \lambda_X(f)(t_1) | \mid f: (X, d) \rightarrow (\mathbb{R}, d_e) \}$$

nonexpansive
↓
↑
Euclidean distance

Generalized Duality

For which F and λ do we have ...?

Duality:

- $K_\lambda(d) = W_\lambda(d)$ for all pseudometrics d
- $K_\lambda^{rel}(r) = W_\lambda(r)$ for all fuzzy relations r

Correspondence: Replace LHS with $K_\Lambda^{rel} = \sup_{\lambda' \in \Lambda} K_{\lambda'}^{rel}$

Generic Results

Theorem We always have $K_\lambda \leq W_\lambda$ and $K_\lambda^{\text{rel}} \leq W_\lambda$.

[Balden, Bonchi, Kerstan, König 2018]

Theorem We always have correspondences
 $K_{\underline{\Lambda}} = W_\lambda$ and $K_{\underline{\Lambda}}^{\text{rel}} = W_\lambda$.

[Wild, Schröder 2021]

$\underline{\Lambda}$ is based on a factor presentation of F and may thus be very large...

Overview of Results

✓ duality
 ~ correspondence
 ✗ correspondence, but no small Δ known

F	λ	New?	d	r	
D	$\mathbb{E}$	no	✓	✓	classical KR duality
P	sup	no	✓ ₁	~ ₂	duality for Hausdorff distance
Poly(D, P)	poly($\mathbb{E}$, sup)	no	~ ₃	✗	
D	$\mathbb{E}_p$	yes	✗	✗	p-Wasserstein distance
D	"generally"	yes	✓	✓	Lévy-Prokhorov distance
C	sup $\circ \mathbb{E}$	yes	✓	✗	Convex Powerset functor

duality known to fail in general

1 Balduen, Banchi, Kerstan, König 2018

2 Wild, Schröder 2021

3 Humeau, Patriscan, Rot 2025

p-Wasserstein distance

$p \geq 1$

$$\delta^{w,p}(d)(\mu, \nu) = \inf \left\{ \left(\mathbb{E}_g(d^p) \right)^{1/p} \mid g \text{ coupling of } \mu, \nu \right\}$$

pointwise

$\leadsto$ predicate lifting $\mathbb{E}_p(f)(\mu) = \left(\mathbb{E}_\mu(f^p) \right)^{1/p}$

Proposition. For $p=2$, $K_{\mathbb{E}_p} \neq W_{\mathbb{E}_p}$.

Lévy-Prokhorov Distance

$$\delta^{LP}(d)(\mu, \nu) = \inf \left\{ \varepsilon \mid \forall A \subseteq X. \mu(A) \leq \nu(A_\varepsilon^d) + \varepsilon \right\}$$

$$A_\varepsilon^d = \{x \mid d(A, x) \leq \varepsilon\}$$


Ky Fan representation: $\delta^{LP}(d) = W_\lambda(d)$, where

$$\lambda_\lambda(f)(\mu) = \inf_{\varepsilon \geq 0} \max(\varepsilon, \mu(\{x \mid f(x) > \varepsilon\}))$$

" $f(x)$ is high with high probability"

" f is generally true"

Lévy-Prokhorov Distance

Theorem We have $K_\lambda = W_\lambda$ and $K_\lambda^{\text{rel}} = W_\lambda$.

Proof sketch

$$\bullet W_\lambda(d)(\mu, \nu) = \inf_{\varepsilon \geq 0} \max(\varepsilon, W_{\mathbb{E}}(r^\varepsilon)(\mu, \nu))$$

$$\uparrow \\ r^\varepsilon(x, y) = \begin{cases} 0, & \text{if } r(x, y) < \varepsilon \\ 1, & \text{if } r(x, y) \geq \varepsilon \end{cases}$$

- Use classical duality. Costs are in $\{0, 1\}$, so prices are too.
- Transform price functions for $K_{\mathbb{E}}(r^\varepsilon)$ into ones for $K_\lambda(r)$.

□

Convex Powerset Functor

subfunctor of $\mathcal{PD}$

$$\mathcal{C}X = \{\emptyset \neq D \subseteq \mathcal{D}X \mid D \text{ is convex}\}$$

Predicate lifting: $\lambda = \text{sup} \circ \mathbb{E}$ $\lambda_X(f)(A) = \sup_{\mu \in A} \mathbb{E}_\mu(f)$

Known results:

Hausdorff distance

- $W_\lambda = W_{\text{sup}} \circ W_{\mathbb{E}} = \delta^H \circ \delta^K \leftarrow \text{Kantorovich distance}$
- Duality fails for $\mathcal{PD}$

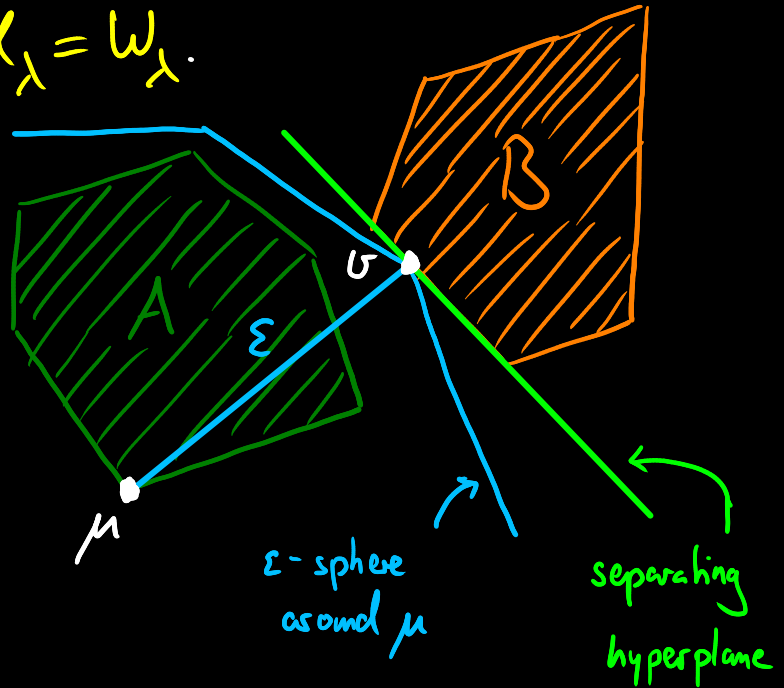
Convex Powerset Functor

Theorem We have $K_\lambda = W_\lambda$.

Proof sketch.

$$\begin{aligned} \delta^H(\delta^k(d))(A, B) \\ = \delta^k(d)(\mu, \nu) = \varepsilon \end{aligned}$$

Extract price function
from hyperplane



Computing Distances

Compute $W_\lambda(d)(\text{conv}\{\overset{\text{finitely supported}}{\mu_1, \dots, \mu_n}, \text{conv}\{\sigma_1, \dots, \sigma_n\})$

- Algorithm based on W_λ : need to solve exponentially many LPs
- Algorithm based on K_λ : need to solve polynomially many LPs

$$K_\lambda(d)(A, B) = \sup \left\{ \left| \sup_{\nu \in B} E_\nu(f) - \sup_{\mu \in A} E_\mu(f) \right| \mid f: (X, d) \rightarrow \mathbb{R} \text{ nonexp.} \right\}$$

Idea: These suprema are reached at some μ_i and σ_j

$\leadsto$ Two LP for each i and j

Beyond Sets

More general dualities:

metric spaces

- Hausdorff functor $H: \text{Met} \rightarrow \text{Met}$, $HX = \text{closed subsets of } X$
- Kantorovich functor $K: \text{Met} \rightarrow \text{Met}$, $KX = \text{Radon prob. measures on } X$

pseudometric spaces

Theorem Duality holds also for $C: \text{PMet} \rightarrow \text{PMet}$,

$CX = \text{non-empty convex subsets of } KX$

Open: What about Lévy-Prokhorov?

Future Work

- Duality for Lévy-Prokhorov beyond sets
- Correspondence for $\mathcal{C}$ over fuzzy relations

conjecture: $\max(K_{\text{supo}E}^{\text{rel}}, K_{\text{info}E}^{\text{rel}}) = W_{\text{supo}E}$

- Distinguishing formulae
- Generalized Lévy-Prokhorov  WIP
- Dualities / correspondences involving lax couplings
- Fully categorical Kantorovich-Rubinstein duality