

# Distributional Convergence of Empirical Entropic Optimal Transport and Applications

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## Abstract

### *Keywords*

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## 1 Introduction

The goal of this manuscript is showing the asymptotic behavior of the kernel distance when the parameter of the family is estimated from the data (data-driven parameter). Recall first some notation:  $\{k_\lambda : \lambda \in \Lambda\}$  is a family of kernels where  $\Lambda$  is a parameter space to be specified later. For each of the kernels  $k_\lambda$  we will denote by  $\mathcal{H}_{k,\lambda}$  its

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associated RKHS and the unit ball of such space as  $\mathcal{F}_{k,\lambda}$ . For a given Borel's measure  $S$ , the mean embedding is defined as

$$\mu_S(\cdot) = \int_{\mathcal{X}} k_{\lambda}(\cdot, y) \, dS(y), \quad (1)$$

where the integral is understood in the Pettis' sense. The interest of mean embedding lays in the following definition property: for every  $f \in \mathcal{H}_{k,\lambda}$ , we have that  $S(f) = \langle f, \mu_S \rangle_{\mathcal{H}_{k,\lambda}}$ , where  $\langle \cdot, \cdot \rangle_{\mathcal{H}_{k,\lambda}}$  denotes the inner product. In terms of the Riesz's representation theorem for Hilbert's spaces, the mean embedding is the dual element of the integral functional induced by  $S$  in  $\mathcal{H}_{k,\lambda}$  (provided integrability assumptions).

In Cárcamo et al. (2024) we used the following set of assumptions.

**(Reg)** *Regularity assumption.*  $\mathcal{X}$  is a separable metric space and each kernel is continuous as a real function of one variable (with the other kept fixed).

**(Dom)** *Dominance assumption.* There exists a constant  $c > 0$  such that  $k_{\lambda} \ll c k$ , for all  $\lambda \in \Lambda$ . Further,  $k$  is bounded on the diagonal, that is,  $\sup_{x \in \mathcal{X}} (k(x, x)) < \infty$

(CAMBIAR LA CONDICIÓN DE DOMINANCIA. ES EXCESIVA).

**(Ide)** *Identifiability assumption.* If  $P \neq Q$ , there exists  $\lambda \in \Lambda$  such that  $\mu_P^{\lambda} \neq \mu_Q^{\lambda}$ .

**(Par)** *Continuous parametrization.*  $\Lambda$  is a compact subset of  $\mathbb{R}^k$  (with  $k \in \mathbb{N}$ ) and, for a fixed  $(x, y) \in \mathcal{X} \times \mathcal{X}$ , the function  $\lambda \mapsto k_{\lambda}(x, y)$  is continuous from  $\Lambda$  to  $\mathbb{R}$  MODIFICAR LA REGULARIDAD PARA INCLUIR LO QUE NECESITAMOS SOBRE DIFERENCIABILIDAD.

**(Sam)** *Sampling scheme.* The sampling scheme is balanced, that is,  $\frac{n}{(n+m)} \rightarrow \theta$ , with  $\theta \in [0, 1]$ , as  $n, m \rightarrow \infty$ .

Under a small variation of them, we will exploit the following Prof. Cárcamo's idea: if

$$\begin{aligned} \psi(\lambda, P - Q) &= \sup_{f \in \mathcal{F}_{k,\lambda}} ((P - Q)(f)) = \|\mu_P^{\lambda} - \mu_Q^{\lambda}\|_{\mathcal{H}_{k,\lambda}} \\ &= \left( \int_{\mathcal{X}} \int_{\mathcal{X}} k_{\lambda}(x, y) \, d(P - Q)(y) \, d(P - Q)(x) \right)^{1/2}, \end{aligned} \quad (2)$$

we can use the integral expression to compute the derivative explicitly.

Some questions around (2):

1. What is the appropriate domain for the new functional in order to compute the Hadamard directional derivative? As we can see, in (2), the argument of  $\sigma$  has been extended. Additionally, the integral expression is valid for every element of  $\ell^{\infty}(\mathcal{F}_{k,\Lambda})$ .

2. What is the new process? Obviously the empirical process is involved in the second argument. But for the first we should have to add assumptions on the parameter estimation (M-estimators, etc).
3. Empirical results, code (C++) and so: having the asymptotic distribution under the alternative, we can detect or explore examples where Gretton's heuristics is not working (interaction between the two terms of the limit, see below).

**Extension of mean embedding to the space  $\mathcal{C}_{\text{bpl}}(\mathcal{F}_{k,\Lambda}, \rho)$**

## 2 Main results

### 2.1 Differentiability results

**Lemma 1.** *Resultado sobre la propiedad Lipschitz marginal (extensión del lema de Shapiro).*

*Proof.* □

**Theorem 2.** *Let us assume that the family of kernels  $\{k_\lambda : \lambda \in \Lambda\}$  satisfies (Dom), (Ide) and (Par).*

*If  $P, Q \in \mathcal{MB}_p(\mathcal{X})$  such that  $P \neq Q$ , then the mapping  $\psi$  in (2) is Hadamard directionally differentiable at  $P - Q$*

*tangentially to  $\mathcal{C}_{\text{bpl}}(\mathcal{F}_{k,\Lambda}, \rho)$ , the subset of  $\ell^\infty(\mathcal{F}_{k,\Lambda})$  constituted by bounded, pre-linear and continuous functionals with respect to the distance  $\rho$  in (??). In such a case, the (directional) derivative of  $\psi$  at  $P - Q$  is given by*

$$\begin{aligned} \psi'_{(\lambda, P-Q)}(\zeta, g) &= g(h^{+, \lambda}) \\ &+ \frac{1}{2 \|\mu_{P-Q}^\lambda\|_{\mathcal{H}_{k,\lambda}}} \int_{\mathcal{X}} \int_{\mathcal{X}} \partial_\lambda k_\lambda(x, y)(\zeta) \, d(P-Q)(y) \, d(P-Q)(x), \end{aligned} \quad (3)$$

*with  $g \in \mathcal{C}(\mathcal{F}_{k,\Lambda}, \rho)$  where the functions  $h^{+, \lambda}$  are defined in (??).*

*Proof.* By definition of Hadamard directional differentiability and Lemma 1, given  $g \in \mathcal{C}_{\text{bpl}}(\mathcal{F}_{k,\Lambda}, \rho)$  and  $\zeta \in \Lambda$  and sequences  $(\zeta_j)_{j \in \mathbb{N}} \in \Lambda^\mathbb{N}$  and  $(t_j)_{j \in \mathbb{N}} \in \mathbb{R}^\mathbb{N}$  such that  $d_\Lambda(\zeta_j, \lambda) \rightarrow 0$  and  $t_j \searrow 0$  when  $j \rightarrow \infty$  we have to show that

$$\lim_{j \rightarrow \infty} \frac{\psi(\lambda + t_j \zeta_j, P - Q + t_j g) - \psi(\lambda, P - Q) - t_j \psi'_{(\lambda, P-Q)}(\zeta, g)}{t_j} = 0. \quad (4)$$

To begin with, note that

$$\frac{\psi(\lambda + t_j \zeta_j, P - Q + t_j g) - \psi(\lambda, P - Q) - t_j \psi'_{(\lambda, P-Q)}(\zeta, g)}{t_j} = L_1 + L_2 + L_3, \quad (5)$$

where

$$\begin{aligned}
L_1 &= \frac{\sup_{f \in \mathcal{F}_{k,\lambda}} ((P - Q + t_j g)(f)) - \sup_{f \in \mathcal{F}_{k,\lambda}} ((P - Q)(f))}{t_j}, \\
L_2 &= \frac{1}{L_4} \int_{\mathcal{X}} \int_{\mathcal{X}} \frac{k_{\lambda+t_j \zeta_j}(x, y) - k_{\lambda}(x, y)}{t_j} d(P - Q)(y) d(P - Q)(x), \\
L_3 &= \frac{1}{L_4} \left( g \left( \int_{\mathcal{X}} (k_{\lambda+t_j \zeta_j}(\cdot, y) - k_{\lambda}(\cdot, y)) d(P - Q)(y) \right) \right. \\
&\quad + \int_{\mathcal{X}} g (k_{\lambda+t_j \zeta_j}(\cdot, y) - k_{\lambda}(\cdot, y)) d(P - Q)(y) \\
&\quad \left. + t_j \left( \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})}^2 - \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda})}^2 \right) \right), \\
L_4 &= \left( \left\| \mu_{P-Q+t_j g}^{\lambda+t_j \zeta_j} \right\|_{\mathcal{H}_{k,\lambda+t_j \zeta_j}} + \left\| \mu_{P-Q+t_j g}^{\lambda} \right\|_{\mathcal{H}_{k,\lambda}} \right)^{1/2}.
\end{aligned} \tag{6}$$

The convergence of  $L_1$  was proved in Cárcamo et al. (2024, Lemma 4) and the limit is  $g(h^{+,\lambda})$ , where  $h^{+,\lambda}$  was defined in [INSERTAR ECUACIÓN](#). Provided the convergence of  $L_4$ , by (Par) converges to

$$\frac{1}{2 \left\| \mu_{P-Q}^{\lambda} \right\|_{\mathcal{H}_{k,\lambda}}} \int_{\mathcal{X}} \int_{\mathcal{X}} \partial_{\lambda} k_{\lambda}(x, y)(\zeta) d(P - Q)(y) d(P - Q)(x). \tag{7}$$

Now we continue with  $L_4$ . By definition

$$\begin{aligned}
\left\| \mu_{P-Q+t_j g}^{\lambda+t_j \zeta_j} \right\|_{\mathcal{H}_{k,\lambda+t_j \zeta_j}} &= \left( \int_{\mathcal{X}} \int_{\mathcal{X}} k_{\lambda+t_j \zeta_j}(x, y) d(P - Q)(y) d(P - Q)(x) \right. \\
&\quad + t_j g \left( \mu_{P-Q}^{\lambda+t_j \zeta_j} \right) \\
&\quad + t_j \int_{\mathcal{X}} g (k_{\lambda+t_j \zeta_j}(\cdot, y)) d(P - Q)(y) \\
&\quad \left. + t_j^2 \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})}^2 \right)^{1/2}.
\end{aligned} \tag{8}$$

From top to bottom in (8):

1.  $\int_{\mathcal{X}} \int_{\mathcal{X}} k_{\lambda+t_j \zeta_j}(x, y) d(P - Q)(y) d(P - Q)(x)$  converges to  $\int_{\mathcal{X}} \int_{\mathcal{X}} k_{\lambda}(x, y) d(P - Q)(y) d(P - Q)(x) = \left\| \mu_{P-Q}^{\lambda} \right\|_{\mathcal{H}_{k,\lambda}}^2$  by (Dom) and (Par) by Cauchy-Schwarz's inequality and dominated convergence theorem.

2. By  $g \in \mathcal{C}_{\text{bpl}}(\mathcal{F}_{k,\Lambda}, \rho)$

$$\left| g \left( \mu_{\mathbf{P}-\mathbf{Q}}^{\lambda+t_j \zeta_j} \right) \right| \leq \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})} \left\| \mu_{\mathbf{P}-\mathbf{Q}}^\lambda \right\|_{\mathcal{H}_{k,\lambda}}, \quad (9)$$

so by (Dom),

$$\left| g \left( \mu_{\mathbf{P}-\mathbf{Q}}^{\lambda+t_j \zeta_j} \right) \right| \leq \|g\|_{\ell^\infty(\mathcal{F}_{k,\Lambda})} \int_{\mathcal{X}} \sqrt{k(x, x)} \, d(\mathbf{P} + \mathbf{Q})(y), \quad (10)$$

and  $t_j g \left( \mu_{\mathbf{P}-\mathbf{Q}}^{\lambda+t_j \zeta_j} \right)$  goes to 0.

3. Analogously,  $\left| g \left( k_{\lambda+t_j \zeta_j} \right) \right| \leq \|g\|_{\ell^\infty(\mathcal{F}_{k,\Lambda})} \sqrt{k(y, y)}$ , so we can conclude that  $t_j \int_{\mathcal{X}} g \left( k_{\lambda+t_j \zeta_j}(\cdot, y) \right) d(\mathbf{P} - \mathbf{Q})(y)$  is also tending to 0.

4. Finally, since  $\|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})} \leq \|g\|_{\ell^\infty(\mathcal{F}_{k,\Lambda})}$ , then  $t_j^2 \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})}$  converges to 0.

In conclusion,  $L_4 \longrightarrow 2 \left\| \mu_{\mathbf{P}-\mathbf{Q}}^\lambda \right\|_{\mathcal{H}_{k,\lambda}}$  when  $j \longrightarrow \infty$ . Additionally, the bounds shown in the previous enumeration provides the domination condition to apply the dominated convergence theorem to the integrals in  $L_3$ . Now, let us proof the convergence of integrands:

- 1.
- 2.

□

## References

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### **3 Supplementary material**