

Distributional Convergence of Empirical Entropic Optimal Transport and Applications

Javier Cárcamo ^{*} Antonio Cuevas [†]
Luis Alberto Rodríguez Ramírez [‡]

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Contents

1	Introduction	2
1.1	Preliminaries	3
1.2	State of the art: different approaches to parametric families of kernels	3
1.3	Our contribution	3
2	Main results	4
2.1	Differentiability results	4
2.2	Statistic results	7
2.3	Empirical results	7
3	Notas	7
4	Supplementary material	9

Abstract

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^{*}EHU-UPV

[†]UAM

[‡]Institute for Mathematical Stochastics, University of Göttingen, Goldschmidtstraße 7, 37077 Göttingen

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 Λ is a parameter space to be specified later. For each of the kernels k_λ we will denote by $\mathcal{H}_{k,\lambda}$ its 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 $k_\lambda \ll ck$, 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.* Λ 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 Λ 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.

$$h^{+, \lambda} = \frac{\mu_P^\lambda - \mu_Q^\lambda}{\|\mu_P^\lambda - \mu_Q^\lambda\|_{\mathcal{H}_{k, \lambda}}}, \quad (3)$$

1.1 Preliminaries

- Reproducing Kernel Hilbert Spaces.
- Mean embedding.
- Empirical process.
- U-statistics.

Generalized mean embedding

Las aplicaciones continuas y prelineales satisfacen la mayoría de propiedades que les pedimos a las medidas de Borel $\mathcal{MB}_p(\mathcal{X})$ para que haya mean embedding.

Lemma 1. *Given $g \in \mathcal{C}_{\text{bpl}}(\mathcal{F}_{k, \Lambda}, \rho)$ and $\lambda \in \Lambda$*

- *For every $\lambda \in \Lambda$, $\|g\|_{\ell^\infty(\mathcal{F}_{k, \lambda})} = \|g\|_{\mathcal{H}_{k, \lambda}^*}$.*
-

Proof. The expression $\|g\|_{\ell^\infty(\mathcal{F}_{k, \lambda})} = \|g\|_{\mathcal{H}_{k, \lambda}^*}$ is a direct consequence of Dudley (2014, Lemma 2.30, p. 88) \square

1.2 State of the art: different approaches to parametric families of kernels

- Median.
- Argmax.
- Rayleigh's quotient.

1.3 Our contribution

- Asymptotic result for data-driven kernel distance.
- Negative result on median heuristic for Gaussian kernel.
- Our proposal: data-driven corrected estimation (with the true asymptotic distribution).

2 Main results

2.1 Differentiability results

Previous lemma to make life easier later.

Lemma 2. *Resultado sobre la propiedad Lipschitz marginal (extensión del lema de Shapiro).*

Proof. □

Differentiability under the null. Continuity of ψ

Theorem 3. *Let us assume that the family of kernels $\{k_\theta : \theta \in \Lambda\}$ satisfies (Dom) and (Par). The mapping ψ in (2) is Hadamard directionally differentiable at $(\lambda, 0)$ tangentially to $\Lambda \times \mathcal{C}_{\text{bpl}}(\mathcal{F}_{k, \Lambda}, \rho)$. In such a case, the (directional) derivative of ψ at $(\lambda, 0)$ is given by*

$$\psi'_{(\lambda, 0)}(\zeta, g) = \|g\|_{\ell^\infty(\mathcal{F}_{k, \lambda})} = \|g\|_{\mathcal{H}_{k, \lambda}^*}, \quad (4)$$

with $g \in \mathcal{C}(\mathcal{F}_{k, \Lambda}, \rho)$.

Proof. By definition of Hadamard directional differentiability and Lemma 2, 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 $d_\Lambda(\zeta_j, \zeta) \rightarrow 0$ and $t_j \searrow 0$ when $j \rightarrow \infty$ we have to show that

$$\lim_{j \rightarrow \infty} \frac{\psi(\lambda + t_j \zeta_j, t_j g) - \psi(\lambda, 0) - t_j \psi'_{(\lambda, 0)}(\zeta, g)}{t_j} = 0. \quad (5)$$

Under the hypothesis of the theorem, note that

$$\frac{\psi(\lambda + t_j \zeta_j, t_j g) - \psi(\lambda, 0) - t_j \psi'_{(\lambda, 0)}(\zeta, g)}{t_j} = \|g\|_{\ell^\infty(\mathcal{F}_{k, \lambda + t_j \zeta_j})} - \|g\|_{\ell^\infty(\mathcal{F}_{k, \lambda})}. \quad (6)$$

Now observe that $\|g\|_{\ell^\infty(\mathcal{F}_{k, \lambda + t_j \zeta_j})} = \sqrt{g(k_{\lambda + t_j \zeta_j}(\cdot_1, \cdot_2))}$. By (Dom), (Par) and dominated convergence theorem, $k_{\lambda + t_j \zeta_j}(\cdot_1, y)$ converges to $k_\lambda(\cdot_1, y)$ on the metric ρ for every $y \in \mathcal{X}$. Hence, by continuity of g , $g(k_{\lambda + t_j \zeta_j}(\cdot_1, y))$ to $g(k_\lambda(\cdot_1, y))$ pointwise in $y \in \mathcal{X}$.

At this point, it is worth to mention that by (Dom)

$$|g(k_{\lambda + t_j \zeta_j}(\cdot_1, y)) - g(k_\lambda(\cdot_1, y))| \leq 2 \|g\|_{\ell^\infty(\mathcal{F}_{k, \Lambda})} \sqrt{k(y, y)}, \quad (7)$$

so, by dominated convergence theorem the convergence of $g(k_{\lambda + t_j \zeta_j}(\cdot_1, \cdot_2))$ to $g(k_\lambda(\cdot_1, \cdot_2))$ is also given in the metric ρ . Thanks to the continuity of g , the first part of the proof is ended. □

Differentiability under the alternative

Theorem 4. *Let us assume that the family of kernels $\{k_\theta : \theta \in \Lambda\}$ satisfies (Dom), (Ide) and (Par). If $P, Q \in \mathcal{MB}_p(\mathcal{X})$ such that $P \neq Q$, then the mapping ψ in (2) is Hadamard directionally differentiable at $(\lambda, P - Q)$ tangentially to $\Lambda \times \mathcal{C}_{\text{bpl}}(\mathcal{F}_{k,\Lambda}, \rho)$, the subset of $\ell^\infty(\mathcal{F}_{k,\Lambda})$ constituted by bounded, prelinear and continuous functionals with respect to the distance ρ in (??). In such a case, the (directional) derivative of ψ at $(\lambda, P - Q)$ is given by*

$$\begin{aligned} \psi'_{(\lambda, P-Q)}(\zeta, g) &= g(h^{+, \lambda}) \\ &+ \frac{1}{2 \|\mu_{P-Q}^\lambda\|_{\mathcal{H}_{k,\lambda}}} \int_{\mathcal{X}} \int_{\mathcal{X}} \partial_\lambda k_\lambda(x, y)(\zeta) \, d(P-Q)(y) \, d(P-Q)(x), \end{aligned} \quad (8)$$

with $g \in \mathcal{C}(\mathcal{F}_{k,\Lambda}, \rho)$ and $\zeta \in \Lambda$; where the functions $h^{+, \lambda}$ are defined in (3).

Proof. By definition of Hadamard directional differentiability and Lemma 2, 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 $d_\Lambda(\zeta_j, \zeta) \rightarrow 0$ and $t_j \searrow 0$ when $j \rightarrow \infty$ we have to show that

$$\lim_{j \rightarrow \infty} \frac{\psi(\lambda + t_j \zeta_j, P - Q + t_j g) - \psi(\lambda, P - Q) - t_j \psi'_{(\lambda, P-Q)}(\zeta, g)}{t_j} = 0. \quad (9)$$

To begin with, note that

$$\frac{\psi(\lambda + t_j \zeta_j, P - Q + t_j g) - \psi(\lambda, P - Q) - t_j \psi'_{(\lambda, P-Q)}(\zeta, g)}{t_j} = L_1 + L_2 + L_3, \quad (10)$$

where

$$\begin{aligned} L_1 &= \frac{\sup_{f \in \mathcal{F}_{k,\lambda}} ((P - Q + t_j g)(f)) - \sup_{f \in \mathcal{F}_{k,\lambda}} ((P - Q)(f))}{t_j}, \\ L_2 &= \frac{1}{L_4} \int_{\mathcal{X}} \int_{\mathcal{X}} \frac{k_{\lambda+t_j \zeta_j}(x, y) - k_\lambda(x, y)}{t_j} \, d(P-Q)(y) \, d(P-Q)(x), \\ L_3 &= \frac{1}{L_4} \left(g \left(\mu_{P-Q}^{\lambda+t_j \zeta_j} - \mu_{P-Q}^\lambda \right) + \int_{\mathcal{X}} g \left(k_{\lambda+t_j \zeta_j}(\cdot, y) - k_\lambda(\cdot, y) \right) \, d(P-Q)(y) \right. \\ &\quad \left. + t_j \left(\|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})}^2 - \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda})}^2 \right) \right), \\ L_4 &= \left(\left\| \mu_{P-Q+t_j g}^{\lambda+t_j \zeta_j} \right\|_{\mathcal{H}_{k,\lambda+t_j \zeta_j}} + \left\| \mu_{P-Q+t_j g}^\lambda \right\|_{\mathcal{H}_{k,\lambda}} \right)^{1/2}. \end{aligned} \quad (11)$$

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 (3). Provided the convergence of L_4 , by (Par), L_2 converges to

$$\frac{1}{2 \|\mu_{\mathbf{P}-\mathbf{Q}}^\lambda\|_{\mathcal{H}_{k,\lambda}}} \int_{\mathcal{X}} \int_{\mathcal{X}} \partial_\lambda k_\lambda(x, y)(\zeta) \, d(\mathbf{P}-\mathbf{Q})(y) \, d(\mathbf{P}-\mathbf{Q})(x). \quad (12)$$

Now we continue with L_4 . By definition

$$\begin{aligned} \left\| \mu_{\mathbf{P}-\mathbf{Q}+t_j g}^{\lambda+t_j \zeta_j} \right\|_{\mathcal{H}_{k,\lambda+t_j \zeta_j}} &= \left(\int_{\mathcal{X}} \int_{\mathcal{X}} k_{\lambda+t_j \zeta_j}(x, y) \, d(\mathbf{P}-\mathbf{Q})(y) \, d(\mathbf{P}-\mathbf{Q})(x) \right. \\ &\quad + t_j g \left(\mu_{\mathbf{P}-\mathbf{Q}}^{\lambda+t_j \zeta_j} \right) \\ &\quad + t_j \int_{\mathcal{X}} g(k_{\lambda+t_j \zeta_j}(\cdot, y)) \, d(\mathbf{P}-\mathbf{Q})(y) \\ &\quad \left. + t_j^2 \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})} \right)^{1/2}. \end{aligned} \quad (13)$$

From top to bottom in (13):

1. $\int_{\mathcal{X}} \int_{\mathcal{X}} k_{\lambda+t_j \zeta_j}(x, y) \, d(\mathbf{P}-\mathbf{Q})(y) \, d(\mathbf{P}-\mathbf{Q})(x)$ converges to $\int_{\mathcal{X}} \int_{\mathcal{X}} k_\lambda(x, y) \, d(\mathbf{P}-\mathbf{Q})(y) \, d(\mathbf{P}-\mathbf{Q})(x) = \|\mu_{\mathbf{P}-\mathbf{Q}}^\lambda\|_{\mathcal{H}_{k,\lambda}}^2$ by (Dom) and (Par) by Cauchy-Schwarz's inequality and dominated convergence theorem.
2. By $g \in \mathcal{C}_{\text{bpl}}(\mathcal{F}_{k,\Lambda}, \rho)$

$$\left| g \left(\mu_{\mathbf{P}-\mathbf{Q}}^{\lambda+t_j \zeta_j} \right) \right| \leq \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j \zeta_j})} \|\mu_{\mathbf{P}-\mathbf{Q}}^\lambda\|_{\mathcal{H}_{k,\lambda}}, \quad (14)$$

so by (Dom),

$$\left| g \left(\mu_{\mathbf{P}-\mathbf{Q}}^{\lambda+t_j \zeta_j} \right) \right| \leq \|g\|_{\ell^\infty(\mathcal{F}_{k,\Lambda})} \int_{\mathcal{X}} \sqrt{k(x, x)} \, d(\mathbf{P}+\mathbf{Q})(y), \quad (15)$$

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

3. Analogously, $|g(k_{\lambda+t_j \zeta_j})| \leq \|g\|_{\ell^\infty(\mathcal{F}_{k,\Lambda})} \sqrt{k(y, y)}$, so we can conclude that $t_j \int_{\mathcal{X}} g(k_{\lambda+t_j \zeta_j}(\cdot, y)) \, d(\mathbf{P}-\mathbf{Q})(y)$ 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, $L_4 \rightarrow 2 \|\mu_{P-Q}^\lambda\|_{\mathcal{H}_{k,\lambda}}$ when $j \rightarrow \infty$. To end up this proof, we see the convergence of L_3 . Firstly, we tackle $\int_{\mathcal{X}} g(k_{\lambda+t_j\zeta_j}(\cdot, y) - k_\lambda(\cdot, y)) d(P-Q)(y)$. Pointwise convergence of $k_{\lambda+t_j\zeta_j}(\cdot, y)$ to $k_\lambda(\cdot, y)$ is given by (Par) and by (Dom) and Folland (1999, Theorem 2.27), convergence in metric ρ is also given. By continuity of g , $g(k_{\lambda+t_j\zeta_j}(\cdot, y) - k_\lambda(\cdot, y))$ is converging to 0 pointwise when $j \rightarrow \infty$. Since this expression is also bounded by the third item of the previous enumeration, by dominated convergence theorem we have the desired limit.

Secondly, convergence $g(\mu_{P-Q}^{\lambda+t_j\zeta_j} - \mu_{P-Q}^\lambda)$ to 0 is to be proved. By (Dom), (Par), and definition of mean embedding, $\mu_{P-Q}^{\lambda+t_j\zeta_j}$ converges to μ_{P-Q}^λ pointwise. Now, recall that

$$\left| \mu_{P-Q}^{\lambda+t_j\zeta_j}(x) - \mu_{P-Q}^\lambda(x) \right| \leq 2 \sqrt{k(x, x)} \int_{\mathcal{X}} \sqrt{k(y, y)} d(P+Q)(y), \quad (16)$$

by (Dom). So, by virtue of dominated convergence theorem, the convergence is also given in the metric ρ . By continuity of g , this term is also done.

Finally, for the last term it is enough to observe that $\left| \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda+t_j\zeta_j})}^2 - \|g\|_{\ell^\infty(\mathcal{F}_{k,\lambda})}^2 \right| \leq 2 \|g\|_{\ell^\infty(\mathcal{F}_{k,\Lambda})}^2$. Hence, L_3 is converging to 0 and the proof is ended. \square

2.2 Statistic results

Our asymptotic results

Delta method y palante.

Negative result on the median heuristic for the Gaussian kernel

Con el Teorema 3 y la teoría de Gretton, sale que bajo la nula va a 0.

2.3 Empirical results

3 Notas

1. What happens when the estimated parameter goes to 0 or ∞ in the Gaussian kernel? The limit of the estimated parameter should belong to the parameter space (see Theorem 3). **Hablando en plata, bajo la nula y usando la heurística de la mediana, el parámetro va a infinito (la mediana va a 0). Se va a la mierda.**

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).

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4 Supplementary material