

On the f -divergences between hyperboloid and Poincaré distributions

Frank Nielsen (Sony CSL), Kazuki Okamura (Shizuoka Univ.)

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Introduction

Embedding: from **discrete** graph to **continuous** space

e.g. Sarker (2012): embedding of trees in **hyperbolic** plane with low distortion (not Euclidian plane)

→ probability distribution on hyperbolic space.

Review:

- hyperboloid distributions on the Minkowski space by Jensen (1981)
— analogy to the von-Mises Fisher distributions on the sphere
- Souriau-Gibbs distributions on by Barbaresco (2019)
— in the Poincaré disk with its Fisher information metric = the Poincaré hyperbolic Riemannian metric

Poincaré distributions

Tojo and Yoshino (2020); hyperboloid distribution realized on the upper half-plane \mathbb{H}

- **Parameter space:**

$$\Theta := \{(a, b, c) \in \mathbb{R}^3 : a > 0, c > 0, ac - b^2 > 0\} \simeq \text{Sym}^+(2, \mathbb{R}) \text{ by}$$
$$(a, b, c) \simeq \begin{bmatrix} a & b \\ b & c \end{bmatrix}.$$

- $|\theta| := ac - b^2 > 0$ and $\text{tr}(\theta) := a + c$ for $\theta = (a, b, c)$.

- **pdf:**

$$p_{\theta}(x, y) := \frac{\sqrt{|\theta|} \exp(2\sqrt{|\theta|})}{\pi} \exp\left(-\frac{a(x^2 + y^2) + 2bx + c}{y}\right) \frac{1}{y^2}, (x, y) \in \mathbb{H}$$

\exists **q -deformed** Poincaré distributions. (Tojo and Yoshino)

f -divergence

The f -divergence induced by a convex generator $f : (0, \infty) \rightarrow \mathbb{R}$ between two pdfs $p(x, y)$ and $q(x, y)$ on \mathbb{H} :

$$D_f[p : q] := \int_{\mathbb{H}} p(x, y) f\left(\frac{q(x, y)}{p(x, y)}\right) dx dy.$$

This measures dissimilarity between two distributions.

Theorem 1

Every f -divergence between two Poincaré distributions p_θ and $p_{\theta'}$ is a function of $(|\theta|, |\theta'|, \text{tr}(\theta'\theta^{-1}))$

Proof components:

(1) $D_f[p_\theta : p_{\theta'}]$ is invariant wrt $\text{SL}(2, \mathbb{R})$ -action:

$$D_f[p_\theta : p_{\theta'}] = D_f[p_{g^{-1}\theta} : p_{g^{-1}\theta'}], \quad g \in \text{SL}(2, \mathbb{R})$$

(2) Every action-invariant function $g(\theta, \theta')$ on \mathbb{H}^2 is a function of $(|\theta|, |\theta'|, \text{tr}(\theta'\theta^{-1}))$ — **maximal invariant** of the action.

Importance of the concept of maximal invariant

- Assume that one has a problem for which a function f which is **invariant** wrt some group action $f(gx) = f(x)$ but difficult to solve explicitly $f()$ from scratch
- For the group action, one finds a **maximal invariant** $m()$: It is an invariant and maximal, i.e.

$$m(x) = m(y) \implies \exists g \text{ s.t. } y = gx$$

- Then, $\exists h$ s.t. $f(x) = h(m(x))$. Solving/finding $h()$ *may be simpler* than solving/finding the original $f()$

See the book by Eaton(1989)

Proposition 1 (explicit formulae)

(i) (Kullback-Leibler) Let $f(u) = -\log u$. Then,

$$D_f [p_\theta : p_{\theta'}] = \frac{1}{2} \log \frac{|\theta|}{|\theta'|} + 2 \left(\sqrt{|\theta|} - \sqrt{|\theta'|} \right) + \left(\frac{1}{2} + \sqrt{|\theta|} \right) (\text{tr}(\theta' \theta^{-1}) - 2).$$

(ii) (squared Hellinger) Let $f(u) = (\sqrt{u} - 1)^2/2$. Then,

$$D_f [p_\theta : p_{\theta'}] = 1 - \frac{2|\theta|^{1/4}|\theta'|^{1/4} \exp \left(|\theta|^{1/2} + |\theta'|^{1/2} \right)}{|\theta + \theta'|^{1/2} \exp \left(|\theta + \theta'|^{1/2} \right)}.$$

(iii) (Neyman χ^2) Let $f(u) := (u - 1)^2$. Assume that $2\theta' - \theta \in \Theta$. Then,

$$D_f [p_\theta : p_{\theta'}] = \frac{|\theta'| \exp(4|\theta'|^{1/2})}{|\theta|^{1/2} |2\theta' - \theta|^{1/2} \exp(2(|\theta|^{1/2} + |2\theta' - \theta|^{1/2}))} - 1.$$

$|\theta + \theta'|$ and $|2\theta' - \theta|$ can be expressed by $|\theta|$, $|\theta'|$, and $\text{tr}(\theta' \theta^{-1})$.

2D hyperboloid distributions

Barndorff-Nielsen (1978); Jensen (1981)

Lobachevskii space:

$$\mathbb{L}^2 := \left\{ (x_0, x_1, x_2) \in \mathbb{R}^3 : x_0 = \sqrt{1 + x_1^2 + x_2^2} \right\} \simeq \mathbb{R}^2$$

Minkowski inner product:

$$[(x_0, x_1, x_2), (y_0, y_1, y_2)] := x_0 y_0 - x_1 y_1 - x_2 y_2.$$

- **Parameter space:**

$$\Theta_{\mathbb{L}^2} := \left\{ (\theta_0, \theta_1, \theta_2) \in \mathbb{R}^3 : \theta_0 > \sqrt{\theta_1^2 + \theta_2^2} \right\}.$$

- **pdf:** For $\theta \in \Theta_{\mathbb{L}^2}$,

$$p_{\theta}(x_1, x_2) := \frac{|\theta| \exp(|\theta|)}{2(2\pi)^{1/2}} \frac{\exp(-[\theta, \tilde{x}])}{\sqrt{1 + x_1^2 + x_2^2}}, \quad (x_1, x_2) \in \mathbb{R}^2$$

where we let $\tilde{x} := \left(\sqrt{1 + x_1^2 + x_2^2}, x_1, x_2 \right) \in \mathbb{L}^2$ and $|\theta| := [\theta, \theta]^{1/2}$.

Theorem 2

Every f -divergence between p_θ and $p_{\theta'}$ is a function of the triplet $([\theta, \theta], [\theta', \theta'], [\theta, \theta'])$, i.e., the pairwise Minkowski inner products of θ and θ' .

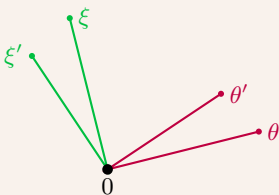
Geometric interpretation: $D_f [p_\theta : p_{\theta'}] \longleftrightarrow \triangle 0\theta\theta'$

Thm 2 \longleftrightarrow **side-angle-side theorem** in Euclidean and *hyperbolic* geometry

$$([\theta, \theta], [\theta', \theta'], [\theta, \theta']) = ([\xi, \xi], [\xi', \xi'], [\xi, \xi'])$$

$$\implies \triangle 0\theta\theta' \equiv \triangle 0\xi\xi' \implies D_f [p_\theta : p_{\theta'}] = D_f [p_\xi : p_{\xi'}]$$

Proof strategy is similar to the one of Theorem 1.



Proposition 2 (explicit formulae)

(i) (Kullback-Leibler) Let $f(u) = -\log u$. Then,

$$D_f[p_\theta : p_{\theta'}] = \log \left(\frac{|\theta|}{|\theta'|} \right) - |\theta'| + \frac{[\theta, \theta']}{[\theta, \theta]} + \frac{[\theta, \theta']}{|\theta|} - 1.$$

(ii) (squared Hellinger) Let $f(u) = (\sqrt{u} - 1)^2/2$. Then,

$$D_f[p_\theta : p_{\theta'}] = 1 - \frac{2|\theta|^{1/2}|\theta'|^{1/2} \exp(|\theta|/2 + |\theta'|/2)}{|\theta + \theta'| \exp(|\theta + \theta'|/2)}.$$

(iii) (Neyman χ^2) Let $f(u) := (u - 1)^2$. Assume that $2\theta' - \theta \in \Theta_{\mathbb{L}^2}$. Then,

$$D_f[p_\theta : p_{\theta'}] = \frac{|\theta'|^2 \exp(2|\theta'|)}{|\theta||2\theta' - \theta| \exp(|\theta| + |2\theta' - \theta|)} - 1.$$

This corresponds to Proposition 1 for Poincaré distributions.

Correspondence

Proposition 3 (Correspondence between the parameter spaces)

A bijection:

$$\Theta \longrightarrow \Theta_{\mathbb{L}}$$

$$\theta := (a, b, c) \mapsto \theta_{\mathbb{L}} := (a + c, a - c, 2b)$$

By this map,

(i) For $\theta, \theta' \in \Theta_{\mathbb{H}}$,

$$|\theta_{\mathbb{L}}|^2 = [\theta_{\mathbb{L}}, \theta_{\mathbb{L}}] = 4|\theta|, \quad |\theta'_{\mathbb{L}}|^2 = [\theta'_{\mathbb{L}}, \theta'_{\mathbb{L}}] = 4|\theta'|, \quad [\theta_{\mathbb{L}}, \theta'_{\mathbb{L}}] = 2|\theta| \operatorname{tr}(\theta' \theta^{-1}).$$

(ii) For every f and $\theta, \theta' \in \mathbb{H}$,

$$D_f^{\mathbb{L}} [p_{\theta_{\mathbb{L}}} : p_{\theta'_{\mathbb{L}}}] = D_f^{\mathbb{H}} [p_{\theta} : p_{\theta'}].$$

There is also a correspondence between the sample spaces, which is compatible with the above one.

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