

AN INVOLUTION INEQUALITY FOR THE KULLBACK—LEIBLER DIVERGENCE

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Abstract. Let $R_t := tP + (1-t)Pt$, where $t \in (0, 1)$, P is a probability measure, and Pt is the push-forward image of P under a measurable involution ι . An inequality involving the Kullback–Leibler divergence of R_t from P is given. It is shown that the role of the involution is essential.

Let \mathcal{P} denote the set of all probability measures on a measurable space (X, Σ) . Take any P and Q in \mathcal{P} such that P absolutely continuous with respect to Q , with the Radon–Nikodym derivative $\frac{dP}{dQ}$ of P with respect to Q . Then the Kullback–Leibler (KL) divergence of Q from P can be defined by the formula

$$D(P||Q) := D_{\text{KL}}(P||Q) := \int_X dP \ln \frac{dP}{dQ}.$$

The KL divergence is well defined, in that the displayed integral exists and its value is in the interval $[0, \infty]$ [2]. In the original paper by Kullback and Leibler, the KL divergence was referred to as “the mean information for discrimination”, whereas the term “divergence” was used there for the symmetrized version $D(P||Q) + D(Q||P)$ of $D(P||Q)$; see pages 80 and 81 concerning formulas (2.4) and (2.6) in [2].

Suppose that $\iota: X \rightarrow X$ is an involution of X , that is, ι is a bijection and the inverse transformation ι^{-1} equals ι . For example, if X is an additive group, then the transformation $X \ni x \mapsto -x$ is an involution.

Suppose further that the involution ι is Σ -measurable. For any $P \in \mathcal{P}$, let $Pt = P\iota^{-1}$ denote the push-forward (or, in this case, equivalently, the “push-back”) probability measure, defined by the condition that $(Pt)(A) := P(\iota(A)) = P(\iota^{-1}(A))$ for all $A \in \Sigma$. In view of the example mentioned in the previous paragraph, Pt may be considered a generalized reflection of P .

THEOREM 1. *Take any $P \in \mathcal{P}$. For $t \in (0, 1)$, consider the mixture*

$$R_t := tP + (1-t)Pt \tag{1}$$

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of the probability measures P and P_t , so that P is absolutely continuous with respect to R_t . Then

$$\frac{D(P||R_{1-t})}{\ln(1-t)} \geq \frac{D(P||R_t)}{\ln t} \tag{2}$$

for all $t \in (0, 1/2)$.

Proof. In this proof, let t be an arbitrary number in the interval $(0, 1)$, unless further specified. Let $\mu := (P + P_t)/2$, so that $\mu t = \mu$. Let $p := \frac{dP}{d\mu}$. Then $\frac{dP_t}{d\mu} = p_t := p \circ t$ and hence $\frac{dR_t}{d\mu} = r_t := t p + (1-t) p_t$. It follows that

$$D(P||R_t) = \int_X d\mu p \ln \frac{p}{r_t} = \int_{[p>0]} d\mu p \ln \frac{p}{r_t} = J_t + \alpha \ln \frac{1}{t},$$

where

$$J_t := \int_{[p>0, p_t>0]} d\mu p \ln \frac{p}{r_t} = \int_{[p>0, p_t>0]} d(\mu t) p_t \ln \frac{p_t}{r_t t} = \int_{[p>0, p_t>0]} d\mu p_t \ln \frac{p_t}{r_{1-t}},$$

$$\alpha := \int_{[p>0, p_t=0]} d\mu p = \int_{[p_t=0]} d\mu p = P([p_t = 0]) \in [0, 1];$$

here and in what follows, $[p > 0] := \{x \in X : p(x) > 0\}$, and the sets $[p > 0, p_t > 0]$, $[p > 0, p_t = 0]$, $[p_t = 0]$ are defined similarly. Hence,

$$\begin{aligned} \frac{D(P||R_t)}{\ln(1/t)} - \alpha &= \frac{1}{2\ln(1/t)} \int_{[p>0, p_t>0]} d\mu \left(p \ln \frac{p}{r_t} + p_t \ln \frac{p_t}{r_{1-t}} \right) \\ &= \frac{1}{2\ln(1/t)} \int_{[p>0, p_t>0]} d\mu \left(p \ln \frac{p}{t p + (1-t) p_t} + p_t \ln \frac{p_t}{t p_t + (1-t) p} \right) \\ &= \frac{1}{2} \int_{[p>0, p_t>0]} \mu(dx) p(x) F(t, \rho(x)), \end{aligned}$$

where $\rho(x) := \frac{p_t(x)}{p(x)}$ and

$$F(t, u) := \frac{1}{\ln(1/t)} \left(\ln \frac{1}{t + (1-t)u} + u \ln \frac{1}{t + (1-t)/u} \right).$$

So, it suffices to show that

$$\Delta(t, u) := F(t, u) - F(1-t, u) > 0 \tag{3}$$

for $t \in (0, 1/2)$ and $u \in (0, 1) \cup (1, \infty)$ (clearly, $\Delta(t, 1) = 0$).

It is indeed not very hard to show that (3) is true. Note that $\Delta(t, 1/u) = \Delta(t, u)/u$ for $u > 0$. So, without loss of generality $u > 1$. That is, it suffices to show that

$$h(u) := \Delta(t, u) \ln(1-t) \ln t$$

is positive for $t \in (0, 1/2)$ and $u > 1$. Let

$$h_2(u) := h''(u) \frac{(1-t+tu)^2(t+(1-t)u)^2u}{1+u}$$

$$= -(u^2+1)(1-t)^2t^2 \ln\left(\frac{1}{t}-1\right) - u((1-2t^2)(1-t)^2 \ln(1-t) + t^2(1-4t+2t^2) \ln t).$$

So, $h_2(u)$ is a quadratic polynomial in u , with the coefficient of u^2 equal to $-(1-t)^2t^2 \ln\left(\frac{1}{t}-1\right) < 0$ for $t \in (0, 1/2)$. So, h_2 is concave and $h_2(\infty-) = -\infty$.

Next, $h_2(1) = g(t) := t^2 \ln t - (1-t)^2 \ln(1-t)$, $g(0+) = g(1/2) = 0$ and $g''(t) = -2 \ln\left(\frac{1}{t}-1\right) < 0$, so that g is concave and hence $h_2(1) = g(t) > 0$ for $t \in (0, 1/2)$. Because h_2 is concave and $h_2(\infty-) = -\infty$, it follows that $h_2(u)$ and hence $h''(u)$ switch their sign exactly once, from $+$ to $-$, as u increases from 1 to ∞ . So, for some real $c = c(t) > 1$, the function h is convex on $[1, c]$ and concave on $[c, \infty)$. At that, $h(1) = h'(1) = 0$ and $h(\infty-) = \infty$. So, $h > 0$ on $(1, \infty)$. That is, indeed $\Delta(t, u) > 0$ for $t \in (0, 1/2)$ and $u > 1$. \square

REMARK 1. Following the lines of the above proof of Theorem 1, one can see that inequality (2) turns into the equality if and only if $p\iota = p$ μ -almost everywhere on the set $A^+ := [p > 0, p\iota > 0]$, that is, if and only if $P(A^+ \cap A) = (P\iota)(A^+ \cap A)$ for all $A \in \Sigma$.

REMARK 2. The role of the involution ι in Theorem 1 is essential. Theorem 1 will fail to hold in general if, in the definition (1) of R_t , the involution $P\iota$ of P is replaced by an arbitrary $Q \in \mathcal{P}$. E.g., take $X = \{-1, 0, 1\}$, $t = \frac{1}{10}$, $P = \frac{1}{100}(57\delta_{-1} + 42\delta_0 + \delta_1)$, $Q = \frac{1}{100}(41\delta_{-1} + 12\delta_0 + 47\delta_1)$, where δ_x denotes the Dirac probability measure at x . Then inequality (2) will be false if $P\iota$ is replaced by Q in the definition (1) of R_t .

Theorem 1 is a generalization of an inequality conjectured on the MathOverflow site [1] and proved there by the author of the present note.

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