

## Second order structure of a finite sample space exponential manifold

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On a finite sample space  $\Omega$ , all strictly positive densities have the exponential form  $q = \exp(u - K_p(u)) = e_p(u)$ , where  $p$  is a strictly positive reference density,  $u$  is a random variable such that  $E_p(u) = 0$ ,  $K_p(u)$  is a normalising constant. As  $u = \log(q/p) - E_p(\log(q/p))$ , all densities are parameterized by the chart  $s_p: q \mapsto u$ . Given a subspace  $V_p \subset L_0^2(p)$  of  $p$ -centered random variables, the exponential family  $\mathcal{E}$  is the family of densities  $e_p(v)$ ,  $v \in V$ . In the atlas of the charts  $s_q$ ,  $q \in \mathcal{E}_p(V)$ , the transition mappings from the reference  $p$  to the reference  $q$  are  $V \ni u \mapsto \exp(u - K_p(u)).p \mapsto u - K_p(u) - \log(q/p) = u - E_q(u) - (E_q(u) + K_p(u))$ . Hence, we have an *affine atlas*, with *exponential transport*  $U_p^q: V_p \ni u \mapsto u - E_q(u) \in V_q$ . The velocity of a curve  $t \mapsto p(t)$  in the chart centered at  $p$  is  $(d/dt)(\log(p(t)/p) - E_p(\log(p(t)/p))) = \dot{p}(t)/p(t) - E_p(\dot{p}(t)/p(t)) = \dot{u}(t) - E_{p(t)}(\dot{u}(t)) = U_p^{p(t)} \dot{u}(t)$ . The velocity at  $t = 0$  in the chart  $s_{p(0)}$  is  $Dp(0) = \dot{u}(0)$  so that the expression of the *tangent space* at  $p$  is  $T_p \mathcal{E} = V_p$ . The expression of the tangent bundle is the *statistical bundle*  $S\mathcal{E}$  consisting of couples  $(q, v)$ ,  $q \in \mathcal{E}$  and  $v \in V_q$ . The *statistical gradient* of a mapping  $\phi: \mathcal{E} \rightarrow \mathbb{R}$  is the section  $\text{grad } \phi$  such that  $(d/dt)\phi[p(t)] = E_{p(t)}(\text{grad } \phi(p(t))(D/dt)p(t))$ . There is a Riemannian metric  $V_p \times V_p \ni (u, v) \mapsto E_p(uv)$  and the dual of the exponential transport is the *mixture transport*. The Levi-Civita connection and the two affine transports define three different geometries on the statistical bundle. The different covariant derivatives and their Hessians result. We consider applications of this set-up and shortly discuss the relation of these geometries with the Gini dissimilarity index.

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[3] L Malagó, G. Pistone, Entropy **17(6)**, 4215 (2015).