

Geometric Sciences of Information: Random musings

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f -divergences between location-scale families

- ▶ Multivariate location-scale family with standard density $p(x)$:
 $p_{\mu, \Sigma}(x) := |\Sigma|^{-1/2} p(\Sigma^{-1/2}(x - \mu))$, $x \in \mathbb{R}^d$, $p(x) = \tilde{p}(\|x^2\|)$ (include multivariate normal distributions, multivariate Cauchy distributions)
- ▶ f -divergence between $P, Q \ll \mu$ for strictly convex generator $f \in C^2$ with $f(1) = 0$: $I_f(P : Q) := \int_{\mathcal{X}} p f\left(\frac{q}{p}\right) d\mu$ (include KL divergence)
- ▶ $I_f(p_{\mu_1, \Sigma} : p_{\mu_2, \Sigma}) = h_f(\Delta_{\Sigma}^2(\mu_1, \mu_2))$ where
 $\Delta_{\Sigma}^2(\mu_1, \mu_2) := (\mu_2 - \mu_1)^{\top} \Sigma^{-1}(\mu_2 - \mu_1)$ for strictly increasing h_f
 (Mahalanobis distance Δ_{Σ})

Preserve relative comparisons:

$$I_f(p_{\mu_1, \Sigma} : p_{\mu_2, \Sigma}) < I_f(p_{\mu'_1, \Sigma} : p_{\mu'_2, \Sigma}) \Leftrightarrow \Delta_{\Sigma}^2(\mu_1, \mu_2) < \Delta_{\Sigma}^2(\mu'_1, \mu'_2)$$

\Rightarrow Voronoi diagrams, minimum enclosing ball, k -center clustering **independent** of f

- ▶ Since $\Delta_{\Sigma}^2(\mu_1, \mu_2) = \Delta_1(0, \Delta_{\Sigma}^2(\mu_1, \mu_2))$, multivariate f -div. amounts to univariate f -div (\rightarrow fast Monte Carlo estimations: eg., Jensen-Shannon div.): $I_f[p_{\mu_1, \Sigma}, p_{\mu_2, \Sigma}] = I_f[p_{0,1}, p_{\Delta_{\Sigma}(\mu_1, \mu_2), 1}]$
- ▶ Spectral matrix f -divergences for multivariate scale families:

$I_f[p_{\mu, \Sigma_1} : p_{\mu, \Sigma_2}] = E_f(|1 - \lambda_1|, \dots, |1 - \lambda_d|)$, where $E_f(\cdot)$ is a d -variate totally symmetric function and $\lambda_i \in \text{sp}(\Sigma_2 \Sigma_1^{-1})$.

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