## Fisher information metric on the space of Gaussians

Consider the space of univariate Gaussian distributions parametrised by  $(\mu, \sigma) \in \mathbb{R} \times (0, \infty)$ . The Gaussian Kullback-Leibler divergence from  $(\mu, \sigma)$  to  $(\mu_1, \sigma_1)$  is:

$$K(\mu_1, \sigma_1) := \mathrm{KL}\Big(\mathcal{N}\left(\mu, \sigma^2\right) : \mathcal{N}\left(\mu_1, \sigma_1^2\right)\Big) = \log\frac{\sigma_1}{\sigma} + \frac{(\mu - \mu_1)^2 + \sigma^2 - \sigma_1^2}{2\sigma_1^2}$$

This has a global minimum when the points  $(\mu, \sigma)$  and  $(\mu_1, \sigma_1)$  coincide,  $K(\mu, \sigma) = 0$ . We can show this because, at this point, the gradient vanishes:

$$\nabla K(\mu, \sigma) = \begin{bmatrix} \frac{\partial K}{\partial \mu} \\ \frac{\partial K}{\partial \sigma} \end{bmatrix} \Big|_{(\sigma, \mu)} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

and the Hessian (matrix of second derivatives) is positive definite:

$$\nabla^2 K(\mu, \sigma) = \begin{bmatrix} \frac{\partial^2 K}{\partial \mu^2} & \frac{\partial^2 K}{\partial \mu \partial \sigma} \\ \frac{\partial^2 K}{\partial \sigma \partial \mu} & \frac{\partial^2 K}{\partial \sigma^2} \end{bmatrix} \bigg|_{(\mu, \sigma)} = \begin{bmatrix} \frac{1}{\sigma^2} & 0 \\ 0 & \frac{2}{\sigma^2} \end{bmatrix}$$

An equivalent way to write this is as a metric tensor

$$g = \frac{\mathrm{d}\mu^2 + 2\,\mathrm{d}\sigma^2}{\sigma^2}$$

so that  $g(\vec{u}, \vec{v}) = \vec{u}^T(\nabla^2 K)\vec{v}$  for any vectors  $\vec{u}, \vec{v} \in \mathbb{R}^2$ .

Under a change of coordinates  $\mu = \sqrt{2}x$ ,  $\sigma = y$ , the metric g is (twice) the metric of the Poincaré half-plane model of hyperbolic space.