

Fisher Information

$$\theta \mapsto q(\theta) \text{ smooth curve} \quad \dot{q}^*(\theta) = \frac{d}{dt} \log q(\theta) \quad \mathbb{E}_{q(\theta)}(\dot{q}^*(\theta)) = 0$$

Fisher information $I(\theta) = \text{Var}(\dot{q}^*(\theta)) = \mathbb{E}_{q(\theta)}(\dot{q}^*(\theta)^2)$

Gradient of $\phi: \mathcal{S}(X) \rightarrow \mathbb{R}$ $\frac{d}{d\theta} \phi(q(\theta)) = \langle \text{Grad } \phi(q(\theta)), \dot{q}^*(\theta) \rangle_{q(\theta)}$

Cramer-Rao inequality

$$\left(\frac{d}{d\theta} \phi(q(\theta)) \right)^2 = \left(\langle \text{Grad } \phi(q(\theta)), \dot{q}^*(\theta) \rangle_{q(\theta)} \right)^2$$

$$\leq \mathbb{E}_{q(\theta)}(\text{Grad } \phi(q(\theta))^2) \mathbb{E}_{q(\theta)}((\dot{q}^*(\theta))^2)$$

$$\boxed{\mathbb{E}_{q(\theta)}(\text{Grad } \phi(q(\theta))^2) \geq \left(\frac{d}{d\theta} \phi(q(\theta)) \right)^2 I(\theta)^{-1}}$$

Ex. 1 $\phi(q) = E_q(t)$ "t is an unbiased estimate of $\phi(q)$ " 2.

$$\text{Error } \phi(q) = t - E_q(t) = t - \phi(q) \quad \text{"error"}$$

$$E_q((\text{Error } \phi(q))^2) = \text{Var}_q(t) \quad \text{"error variance"}$$

$$\text{C-R:} \quad \text{Var}_{q(\theta)}(t) \geq \left(\frac{d}{d\theta} \phi(q(\theta)) \right)^2 I(\theta)^{-1}$$

Ex. 2 $\phi(q) = -E_q(\log q)$ 'Entropy'

$$\text{Error } \phi(q) = -\log q - (-E_q(\log q)) \Rightarrow \text{Statistical Physics}$$

$$\text{C-R:} \quad \text{Var}_{q(\theta)}(\log q(\theta)) \geq \left(\frac{d}{d\theta} \phi(q(\theta)) \right)^2 I(\theta)^{-1}$$