

High-Dimensional Measures and Geometry

Lecture Notes from Feb 16, 2010

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We inspect the general result in the product space with our examples:

Examples: Fair and unfair coins: If $X = \{0, 1\}$, $\mu(\{0\}) = 1/2 = \mu(\{1\})$, $f(x) = x - 1/2$ for $x \in X$, then

$$L_f(\lambda) = \ln \left(\frac{1}{2} e^{\lambda/2} + \frac{1}{2} e^{-\lambda/2} \right)$$

To compute Legendre transform, maximize $t\lambda - L_f(\lambda)$, maximum is assumed at $\lambda^* = \ln \left(\frac{1+2t}{1-2t} \right)$, $-1/2 < t \leq 1/2$. [By taking derivative w. r. t. λ and equalizing to zero to solve for λ]. Thus, we have

$$L_f^*(\lambda) = - \left(\frac{1}{2} + t \right) \ln \left(\frac{1}{1+2t} \right) - \left(\frac{1}{2} - t \right) \ln \left(\frac{1}{1-2t} \right) = H \left(\frac{1}{2} - t \right) + \ln 2.$$

So, we obtain precisely what we had previously

$$\mu_n(\{x \in X : h(x) \leq nt\}) \leq 2^{-n} e^{nH(\frac{1}{2}-t)}.$$

Now, we let $\mu(\{0\}) = 1 - p$, $\mu(\{1\}) = p$ and take $f(x) = x - p$. Then,

$$L_f(\lambda) = \ln (pe^{\lambda(1-p)} + (1-p)e^{-p\lambda}) = \int e^{\lambda f} d\mu$$

and $t\lambda - L_f(\lambda)$ is maximal at

$$\lambda^* = \ln \left(\frac{(1-p)(t+p)}{p(1-p-t)} \right), \quad -p \leq t < 1-p.$$

Thus, we have

$$\begin{aligned} L_f^*(\lambda) &= \ln \left(pe^{(1-p)\ln\left(\frac{(1-p)(t+p)}{p(1-p-t)}\right)} + (1-p)e^{-p\ln\left(\frac{(1-p)(t+p)}{p(1-p-t)}\right)} \right) \\ &= -p \ln \left(\frac{(1-p)(t+p)}{p(1-p-t)} \right) + \ln \left(e^{\ln p + \ln\left(\frac{(1-p)(t+p)}{p(1-p-t)}\right)} + e^{\ln(1-p)} \right) \\ &= -p \ln \left(\frac{(1-p)(t+p)}{p(1-p-t)} \right) + \ln \left(e^{\ln p + \ln\left(\frac{(t+p)}{p(1-p-t)}\right)} \right) + \ln(1-p) \\ &= -p \ln \left(\frac{(1-p)(t+p)}{p(1-p-t)} \right) + \ln \left(\frac{(t+p)}{(1-p-t)} + 1 \right) + \ln(1-p) \\ &= -p \ln \left(\frac{(1-p)(t+p)}{p(1-p-t)} \right) + \ln \left(\frac{1}{(1-p-t)} \right) + \ln(1-p). \end{aligned}$$

So, we have

$$\begin{aligned}
L_f^*(\lambda) &= (t+p) \ln \left(\frac{(1-p)(t+p)}{p(1-p-t)} \right) - \ln \left(\frac{1-p}{1-p-t} \right) \\
&= (t+p) \ln \left(\frac{(t+p)}{p} \right) + (t+p) \ln \left(\frac{1-p}{1-p-t} \right) - \ln \left(\frac{1-p}{1-p-t} \right) \\
&= (t+p) \ln \left(\frac{(t+p)}{p} \right) + (t+p-1) \ln \left(\frac{(1-p)}{1-p-t} \right)
\end{aligned}$$

Using Varadhan's Lemma, we have, for $h = x_1 + \dots + x_n - np, h : X_n \rightarrow \mathbb{R}$ that

$$\begin{aligned}
\mu_n(\{x \in X_n : h(x) \leq nt\}) &\leq e^{-nL_f^*(t)} \\
&= \left[\left(\frac{p}{t+p} \right)^{t+p} \left(\frac{1-p}{1-p-t} \right)^{1-p-t} \right]^n \quad (*)
\end{aligned}$$

Setting $t = \frac{\alpha\beta}{n}, p = \frac{\beta^2}{n}$, we see that the R. H. S. of (*) becomes

$$\left(1 - \frac{\frac{\alpha\beta}{n}}{\frac{\alpha\beta}{n} + \frac{\beta^2}{n}} \right)^{\alpha\beta + \beta^2} \left(1 + \frac{\frac{\alpha\beta}{n}}{\frac{1-\alpha\beta}{n} - \frac{\beta^2}{n}} \right)^n \left(1 + \frac{\frac{\alpha\beta}{n}}{\frac{1-\alpha\beta}{n} - \frac{\beta^2}{n}} \right)^{-\alpha\beta - \beta^2}$$

which approaches to $\left[\left(1 - \frac{\alpha\beta}{\alpha\beta + \beta^2} \right)^{\alpha\beta + \beta^2} e^{\alpha\beta} \right] < 1$ as $n \rightarrow \infty$. So, this is a non-trivial asymptotic estimate.

General Insights: If X is a metric space with probability measure μ such that $A(t) = \{x : d(x, A) \leq t\}$ has measure $\mu(A(t)) \geq 1 - e^{-ct^2}$ for A with $\mu(A) \geq \frac{1}{2}$, then any 1-Lipschitz function $f : X \rightarrow \mathbb{R}$ concentrates about its median.

To see this, we consider

$$A_+ = \{x : f(x) \geq m_f\}$$

and

$$A_- = \{x : f(x) \leq m_f\},$$

both of which has measure at least 1/2. Then, we have

$$\mu(A_+(t)) \geq 1 - e^{-ct^2}, \mu(A_-(t)) \geq 1 - e^{-ct^2}$$

and

$$\mu(A_+(t) \cap A_-(t)) \geq 1 - 2e^{-ct^2}.$$

This implies that

$$\mu(\{s : |f(x) - m_f| \leq t\}) \geq 1 - 2e^{-ct^2}.$$

6 Measure Concentration on High Dimensional Spheres

6.1 Gaussians as Limits of Projected Spherical Measures

We recall that we induced normalized surface measure on S^{n-1} by map $x \rightarrow \frac{x}{\|x\|}$ from γ_n . Now, we will induce measure in reverse direction.

6.1.1 Lemma. *Non-normalized Riemannian surface measure of S^{n-1} is*

$$|S^{n-1}| = \frac{2\pi^{n/2}}{\Gamma(n/2)}$$

with

$$\Gamma(t) = \int_0^\infty x^{t-1} e^{-x} dx, \quad t > 0.$$

Proof. Consider $\rho_n(x) = \frac{1}{(2\pi)^{n/2}} e^{-\|x\|^2}$, the standard Gaussian density and let $S^{n-1}(r) = \{x \in \mathbb{R}^n : \|x\| = r\}$. Integration in polar co-ordinates gives us

$$\begin{aligned} 1 &= \int_{\mathbb{R}^n} d\rho_n = \frac{1}{(2\pi)^{n/2}} \int_0^\infty |S^{n-1}(r)| e^{-r^2/2} dr \\ &= \frac{1}{(2\pi)^{n/2}} |S^{n-1}(1)| \int_0^\infty r^{n-1} e^{-r^2/2} dr \end{aligned}$$

Substituting $t = r^2/2$, $r dr = dt$, $r^{n-2} = 2^{(n-2)/2} t^{(n-2)/2}$, we have

$$\begin{aligned} 1 &= \frac{1}{(2\pi)^{n/2}} |S^{n-1}(1)| \int_0^\infty 2^{(n-2)/2} t^{(n-2)/2} e^{-t} dt \\ &= \frac{1}{(2\pi)^{n/2}} |S^{n-1}(1)| 2^{(n-2)/2} \Gamma(n/2) \end{aligned}$$

Solving for $|S^{n-1}(1)|$, we obtain

$$|S^{n-1}(1)| = \frac{(2\pi)^{n/2} 2^{(n-2)/2}}{\Gamma(n/2)} = \frac{2\pi^{n/2}}{\Gamma(n/2)}.$$

□

6.1.2 Theorem. *Let $\tilde{S}^{n-1} \subset \mathbb{R}^n$ be the sphere*

$$\tilde{S}^{n-1} = \{x \in \mathbb{R}^n : \|x\| = \sqrt{n}\}$$

and $\tilde{\mu}_n$ be the rotation invariant probability measure on \tilde{S}^{n-1} . Consider $\Phi : \tilde{S}^{n-1} \rightarrow \mathbb{R}$ be given by $\Phi(x_1, \dots, x_n) = x_1$ and let ν_n be the induced measure on \mathbb{R} , i. e.,

$$\nu_n(A) = \tilde{\mu}_n(\Phi^{-1}(A)) \text{ for Borel sets } A \subset \mathbb{R}.$$

If γ_1 is the standard Gaussian measure with density $\rho_1(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$, then for any Borel set $A \subset \mathbb{R}$, $\lim_{n \rightarrow \infty} \nu_n(A) = \gamma_1(A)$ and the density of ν_n converges uniformly on compact sets to ρ_1 .