

Demostrar que para $X \sim \mathcal{N}(\mu, \sigma^2)$, la entropía de X es

$$S(X) = \frac{1}{2} \log_2(2\pi e \sigma^2).$$

Tomando

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad (1)$$

y

$$S(X) = - \int_{\mathbb{R}} f(x) \log_2 f(x) dx \quad (2)$$

para el caso continuo y sabiendo que

$$\log_b x = \frac{\log_a x}{\log_a b} \quad (3)$$

Entonces:

$$\begin{aligned} S(X) &= (1)(2) = - \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \log_2 \left(\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right) dx \\ &= - \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \cdot \frac{1}{\log 2} \log \left(\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right) dx \\ &= - \frac{1}{\log 2} \left(\int_{\mathbb{R}} f(x) \log \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right) dx + \int_{\mathbb{R}} f(x) \cdot \left(-\frac{(x-\mu)^2}{2\sigma^2} \right) dx \right) \\ &= - \frac{1}{\log 2} \left(\log \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right) \underbrace{\int_{\mathbb{R}} f(x) dx}_{=1} - \frac{1}{2\sigma^2} \underbrace{\int_{\mathbb{R}} f(x)(x-\mu)^2 dx}_{=\mathbb{E}[(X-\mu)^2]} \right) \\ &= \frac{-\log \left(\frac{1}{\sqrt{2\pi\sigma}} \right)}{\log 2} + \underbrace{\frac{\sigma^2}{2\sigma^2 \log 2}}_{\frac{1}{2 \log 2}} \\ &= \frac{-\log 1 + \log(\sqrt{2\pi\sigma})}{\log 2} + \frac{1}{2 \log 2} \\ &= \frac{\log(\sqrt{2\pi\sigma})}{\log 2} + \frac{1}{2 \log 2} \\ &= \log_2(\sqrt{2\pi\sigma}) + \frac{\log e}{2 \log 2} \\ &= \log_2(\sqrt{2\pi\sigma}) + \frac{1}{2} \log_2 e \\ &= \frac{1}{2} \left(\log_2 ((\sqrt{2\pi\sigma})^2) + \log_2 e \right) \\ &= \frac{1}{2} \log_2(2\pi\sigma^2 e) \end{aligned}$$

De esta manera se puede observar que la entropía no depende de la media de la gaussiana sino de la varianza.