

# Deriving the Gaussian from a Maximum Entropy Principle

Normal / Gaussian  $X \sim \mathcal{N}(X_0, \mu, \sigma^2) = \left( \frac{1}{\sigma \sqrt{2\pi}} \right) \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2\right)$

Normalization Constant  $\left( \frac{1}{\sigma \sqrt{2\pi}} \right)$

Where does this come from?

→ Gaussian / Normal is maximizing the entropy under prescribed mean & variance

→ Differential Entropy

Entropy is a Functional

$$H(p_0) = - \int_{-\infty}^{\infty} p_0(x) \ln(p_0(x)) dx$$

$\left( \frac{1}{\sigma \sqrt{2\pi}} \right) = \underset{p(x)}{\text{argmax}} H(p)$

$p^*(x) = \underset{p(x)}{\text{argmax}} H(p)$

s.t.:  $\int_{-\infty}^{\infty} p(x) dx = 1$  (normalization)

$\int_{-\infty}^{\infty} p(x) \cdot x dx = \mu$  (prescribed mean first order moment)

$\int_{-\infty}^{\infty} p(x) \cdot (x-\mu)^2 dx = \sigma^2$  (prescribed variance second order central moment)

Build a Lagrangian

$$\mathcal{L}(p, \lambda_0, \lambda_1, \lambda_2) = \int_{-\infty}^{\infty} p \ln p dx + \lambda_0 \cdot \left( \int_{-\infty}^{\infty} p(x) dx - 1 \right) + \lambda_1 \cdot \left( \int_{-\infty}^{\infty} p(x) \cdot x dx - \mu \right) + \lambda_2 \cdot \left( \int_{-\infty}^{\infty} p(x) \cdot (x-\mu)^2 dx - \sigma^2 \right)$$

$= \int_{-\infty}^{\infty} (p \ln p + \lambda_0 p + \lambda_1 p \cdot x + \lambda_2 p (x-\mu)^2) dx$

Another functional

Some scalars  $-\lambda_0 - \lambda_1 \mu - \lambda_2 \sigma^2$

→ Take derivative and set to 0

$\frac{\partial \mathcal{L}}{\partial p} \stackrel{!}{=} 0 \quad \text{I} \quad \Rightarrow \quad \frac{\partial \mathcal{L}}{\partial p} = \ln p + \frac{p}{p} + \lambda_0 + \lambda_1 x + \lambda_2 (x-\mu)^2$

$\frac{\partial \mathcal{L}}{\partial \lambda_0} = \int_{-\infty}^{\infty} p dx - 1 \stackrel{!}{=} 0 \quad \text{II}$

$\frac{\partial \mathcal{L}}{\partial \lambda_1} = \int_{-\infty}^{\infty} p \cdot x dx - \mu \stackrel{!}{=} 0 \quad \text{III}$

$\frac{\partial \mathcal{L}}{\partial \lambda_2} = \int_{-\infty}^{\infty} p \cdot (x-\mu)^2 dx - \sigma^2 \stackrel{!}{=} 0 \quad \text{IV}$

I rearrange for p

$\ln p(x) = -1 - \lambda_0 - \lambda_1 x - \lambda_2 (x-\mu)^2$

$= -1 - \lambda_0 - \lambda_1 x - \lambda_2 (x^2 - 2x\mu + \mu^2)$

$= -1 - \lambda_0 - \lambda_1 x - \lambda_2 x^2 + \lambda_2 2x\mu - \lambda_2 \mu^2$

$= \underbrace{-\lambda_2 x^2}_a + \underbrace{(2\lambda_2 \mu - \lambda_1)x}_b - \underbrace{1 - \lambda_0 - \lambda_2 \mu^2}_c$

Completing the square

$a = -\lambda_2 \quad b = 2\lambda_2 \mu - \lambda_1 \quad c = -1 - \lambda_0 - \lambda_2 \mu^2$

$= ax^2 + bx + c$

$= a(x-d)^2 + e$

$a = a \quad d = -\frac{b}{2a} \quad e = c - \frac{b^2}{4a}$

$a = -\lambda_2 \quad d = -\frac{(2\lambda_2 \mu - \lambda_1)}{2 \cdot (-\lambda_2)} = -1 - \lambda_0 - \lambda_2 \mu^2 - \frac{(2\lambda_2 \mu - \lambda_1)^2}{4(-\lambda_2)}$

$= -1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \cdot \left( \mu - \frac{\lambda_1}{2\lambda_2} \right)^2$

$= \mu - \frac{\lambda_1}{2\lambda_2} \quad e = -1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \left( \mu - \frac{\lambda_1}{2\lambda_2} \right)^2$

$\ln(p) = a \cdot (x-d)^2 + e$

$= -\lambda_2 \cdot \left( x - \left( \mu - \frac{\lambda_1}{2\lambda_2} \right) \right)^2 - 1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \left( \mu - \frac{\lambda_1}{2\lambda_2} \right)^2$

$= -\lambda_2 \cdot \left( x - \mu + \frac{\lambda_1}{2\lambda_2} \right)^2 - 1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \left( \mu - \frac{\lambda_1}{2\lambda_2} \right)^2$

$y := x - \mu + \frac{\lambda_1}{2\lambda_2} \quad \Leftrightarrow x = y + \mu - \frac{\lambda_1}{2\lambda_2}$

$= -\lambda_2 y^2 - 1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \left( \mu - \frac{\lambda_1}{2\lambda_2} \right)^2 \quad | \exp(\dots)$

$p(y) = \exp(-\lambda_2 y^2 - 1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \left( \mu - \frac{\lambda_1}{2\lambda_2} \right)^2)$

plug into III

$\int_{-\infty}^{\infty} p(x) \cdot x dx = \mu$

$\int_{-\infty}^{\infty} p(y) \cdot \left( y + \mu - \frac{\lambda_1}{2\lambda_2} \right) dy = \mu$

$\int_{-\infty}^{\infty} p(y) \cdot y dy + \int_{-\infty}^{\infty} p(y) \cdot \left( \mu - \frac{\lambda_1}{2\lambda_2} \right) dy = \mu$

Due to symmetry of the Gaussian  $\Rightarrow$  vanishes

$0 + \int_{-\infty}^{\infty} p(y) \cdot \left( \mu - \frac{\lambda_1}{2\lambda_2} \right) dy = \mu$

$\left( \mu - \frac{\lambda_1}{2\lambda_2} \right) \cdot \int_{-\infty}^{\infty} p(y) dy = \mu$

$= 1 \quad \text{because of II}$

$\mu - \frac{\lambda_1}{2\lambda_2} = \mu$

$-\frac{\lambda_1}{2\lambda_2} = 0$

$\boxed{\lambda_1 = 0}$

plug  $\lambda_1 = 0$  into  $p(y)$

$p(y) = \exp(-\lambda_2 y^2 - 1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \cdot \left( \mu - \frac{\lambda_1}{2\lambda_2} \right)^2)$

$p(y) = \exp(-\lambda_2 y^2 - 1 - \lambda_0 - \lambda_2 \mu^2 + \lambda_2 \mu^2)$

$y = x - \mu - \frac{\lambda_1}{2\lambda_2}$

$p(x) = \exp(-\lambda_2 \cdot (x-\mu)^2 - 1 - \lambda_0)$

Subst:  $z := x - \mu \quad \Leftrightarrow x = z + \mu$

$p(z) = \exp(-\lambda_2 \cdot z^2 - 1 - \lambda_0)$

plug into II

$\int_{-\infty}^{\infty} p(x) dx = 1$

$= \int_{-\infty}^{\infty} p(z) dz = 1$

$\int_{-\infty}^{\infty} \exp(-\lambda_2 z^2 - 1 - \lambda_0) dz = 1$

$\exp(-1 - \lambda_0) \int_{-\infty}^{\infty} \exp(-\lambda_2 z^2) dz = 1$

Similar to normalization derivation

$\sqrt{\frac{\pi}{\lambda_2}}$

$e^{-1-\lambda_0} \cdot \sqrt{\frac{\pi}{\lambda_2}} = 1 \quad (*)$

plug  $p(z)$  into IV

$\int_{-\infty}^{\infty} p(x) \cdot (x-\mu)^2 dx = \sigma^2$

$\int_{-\infty}^{\infty} p(z) \cdot (z+\mu-\mu)^2 dz = \sigma^2$

$\int_{-\infty}^{\infty} p(z) z^2 dz = \sigma^2$

$\int_{-\infty}^{\infty} \exp(-\lambda_2 z^2 - 1 - \lambda_0) \cdot z^2 dz = \sigma^2$

$e^{-1-\lambda_0} \cdot \int_{-\infty}^{\infty} \exp(-\lambda_2 z^2) \cdot z^2 dz = \sigma^2$

$I(2) = \int_{-\infty}^{\infty} \exp(-\lambda_2 z^2) dz = \sqrt{\frac{\pi}{\lambda_2}}$

$-\frac{dI}{d\lambda_2} = \int_{-\infty}^{\infty} (-z^2) \exp(-\lambda_2 z^2) dz$

$e^{-1-\lambda_0} \cdot \left( -\frac{d}{d\lambda_2} \left( \sqrt{\frac{\pi}{\lambda_2}} \right) \right) = \sigma^2$

$\frac{d}{d\lambda_2} \left( \pi^{\frac{1}{2}} \cdot \lambda_2^{-\frac{1}{2}} \right) = \pi^{\frac{1}{2}} \cdot \left( -\frac{1}{2} \right) \cdot \lambda_2^{-\frac{3}{2}}$

$= -\frac{1}{2} \sqrt{\frac{\pi}{\lambda_2^3}}$

$e^{-1-\lambda_0} \cdot (-1) \cdot \left( -\frac{1}{2} \sqrt{\frac{\pi}{\lambda_2^3}} \right) = \sigma^2$

$e^{-1-\lambda_0} \cdot \frac{1}{2} \sqrt{\frac{\pi}{\lambda_2^3}} = \sigma^2 \quad (**)$

Solving a system of nonlinear equations

$(*) \quad e^{-1-\lambda_0} \cdot \sqrt{\frac{\pi}{\lambda_2}} = 1 \quad \rightarrow e^{-1-\lambda_0} = \sqrt{\frac{\lambda_2}{\pi}}$

$(**) \quad e^{-1-\lambda_0} \cdot \frac{1}{2} \sqrt{\frac{\pi}{\lambda_2^3}} = \sigma^2 \quad \rightarrow e^{-1-\lambda_0} = 2\sigma^2 \sqrt{\frac{\lambda_2^3}{\pi}}$

$\sqrt{\frac{\lambda_2}{\pi}} = 2\sigma^2 \sqrt{\frac{\lambda_2^3}{\pi}} \quad | ( )^2$

$\frac{\lambda_2}{\pi} = 4\sigma^4 \cdot \frac{\lambda_2^3}{\pi}$

$\lambda_2^2 = \frac{1}{4\sigma^4}$

$\lambda_2 = \pm \frac{1}{2\sigma^2} \quad (\text{we need "+"})$

$\boxed{\lambda_2 = \frac{1}{2\sigma^2}}$

plug into  $(*)$

$e^{-1-\lambda_0} \cdot \sqrt{\frac{\pi}{\frac{1}{2\sigma^2}}} = 1$

$e^{-1-\lambda_0} \cdot \sigma \sqrt{2\pi} = 1$

$e^{-1-\lambda_0} = \frac{1}{\sigma \sqrt{2\pi}} \quad | \ln(\dots)$

$-1 - \lambda_0 = -\ln(\sigma \sqrt{2\pi})$

$\boxed{\lambda_0 = \ln(\sigma \sqrt{2\pi}) - 1}$

plug into  $p(x)$

$p(x) = \exp(-\lambda_2 (x-\mu)^2 - 1 - \lambda_0)$

$p(x) = \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2 - 1 - (\ln(\sigma \sqrt{2\pi}) - 1)\right)$

$= \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2 - \ln(\sigma \sqrt{2\pi})\right)$

$= \exp(-\ln(\sigma \sqrt{2\pi})) \cdot \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2\right)$

$\boxed{p(x) = \frac{1}{\sigma \sqrt{2\pi}} \cdot \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2\right)}$

Gaussian / Normal Distribution