

Formulation of the problem:

$$H(x) = - \int_{-\infty}^{\infty} p(x) \ln p(x) dx \leftarrow \text{Maximize this}$$

Constraints: $\int p(x) dx = 1$, $\int p(x) x dx = \mu$,
 $\int p(x) (x - \mu)^2 dx = \sigma^2$

$$L(p, \alpha_1, \alpha_2, \alpha_3) = \int p(x) \ln p(x) dx + \alpha_1 (1 - \int p(x) dx) + \alpha_2 (\mu - \int p(x) x dx) + \alpha_3 (\sigma^2 - \int p(x) (x - \mu)^2 dx)$$

$$\alpha_1, \alpha_2, \alpha_3 \geq 0$$

$$\min_p \max_{\alpha_1, \alpha_2, \alpha_3} L(p, \alpha_1, \alpha_2, \alpha_3) \leftarrow \text{objective function}$$

Why can we use dual formulation here?

$$\frac{\partial L}{\partial p} = p \cdot \frac{1}{p} + \ln p - \alpha_1 - \alpha_2 x - \alpha_3 (x - \mu)^2 = 0$$
$$= 1 + \ln p - \alpha_1 - \alpha_2 x - \alpha_3 (x^2 - 2\mu x + \mu^2) = 0$$

$$\ln p = -1 + \alpha_1 + \alpha_2 x + \alpha_3 (x - \mu)^2$$

$$p = \exp(-1 + \alpha_1 + \alpha_2 X + \alpha_3 (X - \mu)^2)$$

$$= \exp(-1 + \alpha_1 + \alpha_2 X + \alpha_3 (X^2 - 2\mu X + \mu^2))$$

$$= \exp[\alpha_3 X^2 + (\alpha_2 - 2\mu\alpha_3)X + (\alpha_3 \mu^2 + \alpha_1 - 1)]$$

$$\int p dx = 1$$

$$\int p x dx = \mu$$

$$\mu \int p dx = \mu$$

$$\begin{aligned} & \alpha_3 (X - \mu + k)^2 \\ &= \alpha_3 (X^2 + \mu^2 + k^2 - 2\mu X - 2k\mu + 2kX) \\ &= \alpha_3 X^2 + \alpha_3 \mu^2 + \alpha_3 k^2 - 2\mu\alpha_3 X - 2k\alpha_3 \mu + 2k\alpha_3 X \end{aligned}$$

$$= \alpha_3 X^2 + \alpha_3 \mu^2 + (2k\alpha_3 - 2\mu\alpha_3)X + \alpha_3 k^2 - 2k\alpha_3 \mu$$

$$\alpha_2 = 2k\alpha_3 \Rightarrow k = \frac{\alpha_2}{2\alpha_3}$$

$$p = \exp\left[\alpha_3 \left(X - \mu + \frac{\alpha_2}{2\alpha_3}\right)^2 - \frac{\alpha_2^2}{4\alpha_3} + \alpha_2 \mu + \alpha_1 - 1\right]$$

$$y = X - \mu + k \Rightarrow X = y + \mu - k = y + \mu - \frac{\alpha_2}{2\alpha_3} \quad p dy = dx$$

$$p = \exp\left[\alpha_3 y^2 - \frac{\alpha_2^2}{4\alpha_3} + \alpha_2 \mu + \alpha_1 - 1\right]$$

$$\int p x dx = \mu$$



$$\int_{-\infty}^{\infty} \exp \left[\alpha_3 y^2 - \frac{\alpha_2^2}{4\alpha_3} + \alpha_2 \mu + \alpha_1 - 1 \right] (y + \mu - k) dy = \mu$$

odd function

$\int p \mu dy = \mu$

$$\Rightarrow k=0 \Rightarrow \alpha_2=0$$

$$p = \exp [\alpha_3 x^2 - 2\mu \alpha_3 x + \alpha_3 \mu^2 + \alpha_1 - 1]$$

$$= \exp [\alpha_3 (\underline{x-\mu})^2 + (\alpha_1 - 1)] \quad \text{--- (*)}$$

$$\int p(x) dx = 1$$

$$\int \exp [\alpha_3 (x-\mu)^2 + (\alpha_1 - 1)] dx = 1$$

$$\int \exp(\alpha_3 (x-\mu)^2) \cdot \exp(\alpha_1 - 1) dx = 1$$

$$\int \exp(\alpha_3 (x-\mu)^2) dx = \exp(1 - \alpha_1)$$

According to Gaussian integral

$$\int_{-\infty}^{\infty} \exp(-a(x+b)^2) dx = \sqrt{\frac{\pi}{a}} \quad a > 0$$

$$\sqrt{\frac{\pi}{-\alpha_3}} = \exp(1 - \alpha_1) \quad \text{--- ①}$$

$$\int p(x)(x-\mu)^2 dx = \sigma^2$$

$$\int \exp[\alpha_3(x-\mu)^2 + (\alpha_1 - 1)] (x-\mu)^2 dx = \sigma^2$$

$$z = x - \mu, \quad dz = dx$$

$$\int \exp[\alpha_3 z^2 + (\alpha_1 - 1)] z^2 dz = \sigma^2$$

$$\int \exp[\alpha_3 z^2] z^2 dz = \sigma^2 \cdot \exp(1 - \alpha_1)$$

$$\frac{1}{2} \sqrt{\frac{\pi}{(-\alpha_3)^3}} = \sigma^2 \exp(1 - \alpha_1) \quad \text{--- (2)}$$

With ①② $\frac{1}{2} \sqrt{\frac{\pi}{(-\alpha_3)^3}} = \sigma^2 \sqrt{\frac{\pi}{-\alpha_3}}$

$$\sqrt{\frac{1}{(-\alpha_3)^3}} = 2\sigma^2 \sqrt{\frac{1}{-\alpha_3}}$$

$$1 = 2\sigma^2 \sqrt{\frac{1}{-\alpha_3}} \cdot \sqrt{(-\alpha_3)^3}$$

$$1 = 2\sigma^2 \sqrt{(-\alpha_3)^2} = 2\sigma^2 |-\alpha_3| = -2\sigma^2 \alpha_3$$

$$\Rightarrow \alpha_3 = \frac{1}{-2\sigma^2}$$

Hence, $\sqrt{\frac{\pi}{\frac{1}{2\sigma^2}}} = \exp(1 - \alpha_1) \Rightarrow \sqrt{2\sigma^2 \pi} = \exp(1 - \alpha_1)$

$$1 - \alpha_1 = \ln(\sqrt{2\sigma^2 \pi})$$

$$\alpha_1 = 1 - \ln(\sqrt{2\sigma^2\pi})$$

Plug all of these back into (*)

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

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

→ This is Gaussian distribution!

