

Consider the measurement model

$$y_j = a_j x_j, \text{ where } \log a_j \sim \mathcal{N}(\log a_0, \sigma^2), \quad a_i + a_j \text{ for } i \neq j, \quad a_j \perp x_j \text{ and } x_j > 0.$$

We can write this as

$$y = a \cdot x, \text{ where } a \cdot x := [a_1 x_1, \dots, a_n x_n]^T \text{ (elementwise product).}$$

Then we obtain the likelihood by marginalizing over the nuisance parameter  $a$ :

$$\begin{aligned} P(y|x) &= \frac{P(y, x)}{P(x)} = \int_{\mathbb{R}^n} \frac{P(y, x, a)}{P(x)} da = \int_{\mathbb{R}^n} \frac{P(y, x, a)}{P(x)P(a)} P(a) da \quad \left( \begin{array}{l} x + a \\ \Rightarrow P(x, a) = P(x)P(a) \end{array} \right) \\ &= \int_{\mathbb{R}^n} \frac{P(y, x, a)}{P(x, a)} P(a) da = \int_{\mathbb{R}^n} P(y|x, a) P(a) da. \end{aligned}$$

What is  $P(y|x, a)$ ? Since we now condition on both  $x$  and  $a$ , we obtain from the measurement model that

$$y = a \cdot x \quad \Rightarrow \quad P(y|a, x) = \delta(y - a \cdot x),$$

$\underbrace{\hspace{1.5cm}}$   
 these are  
 now fixed

where  $\delta$  is (formally) the Dirac delta distribution.

Therefore

$$P(y|x) = \int_{\mathbb{R}^n} \delta(y - a \cdot x) P(a) da.$$

Change of variables:  $v_i = a_i x_i \Rightarrow dv_i = x_i da_i$

$$\begin{aligned} \Rightarrow P(y|x) &= \frac{1}{x_1 \cdots x_n} \int_{\mathbb{R}^n} \delta(y - v) P\left(\frac{v}{x}\right) dv \quad \left| \quad \frac{v}{x} := \left[ \frac{v_1}{x_1}, \dots, \frac{v_n}{x_n} \right]^T \right. \\ &= \frac{1}{x_1 \cdots x_n} P\left(\frac{y}{x}\right). \end{aligned} \quad (*)$$

By the independence of  $a_i$  and  $a_j$  for  $i \neq j$ , we obtain

$$P\left(\frac{y}{x}\right) = \prod_{j=1}^n P_{a_j}\left(\frac{y_j}{x_j}\right).$$



Since  $a_j \sim \log N(\log a_0, \sigma^2)$ , it follows by change of variables that

$$p_{a_j}(a) = p_{\log a_j}(\log a) \frac{d \log a}{da} = \frac{1}{\sqrt{2\pi\sigma^2}} \frac{1}{a} \exp\left(-\frac{1}{2\sigma^2} (\log a - \log a_0)^2\right).$$

Hence

$$P\left(\frac{y_i}{x}\right) = \left(\frac{1}{\sqrt{2\pi\sigma^2}}\right)^n \frac{x_1 \cdots x_n}{y_1 \cdots y_n} \exp\left(-\frac{1}{2\sigma^2} \sum_{j=1}^n \left(\log \frac{y_j}{a_0 x_j}\right)^2\right).$$

Plugging this into (x) yields

$$P(y|x) = \left(\frac{1}{\sqrt{2\pi\sigma^2}}\right)^n \frac{1}{y_1 \cdots y_n} \exp\left(-\frac{1}{2\sigma^2} \sum_{j=1}^n \left(\log \frac{y_j}{a_0 x_j}\right)^2\right).$$