

$$L(\omega) = \prod p_{\omega}(x_i, y_i)$$

$$= \prod p_{\omega}(y_i | x_i) p(x_i)$$

$$y_i = f_{\omega}(x_i) + \varepsilon_i \quad \varepsilon_i \sim N(0, 1)$$

$$y_i | x_i \stackrel{D}{=} N(f_{\omega}(x_i), 1).$$

$$\propto \prod_{i=1}^n \left[\exp\left(-\frac{(y_i - f_{\omega}(x_i))^2}{2}\right) p(x_i) \right]$$

$$Y_i = f_{\omega}(x_i) + \varepsilon_i$$

$$Y_i = \mathbb{1} \{ f_{\omega}(x_i) + \varepsilon_i \geq 0 \}$$

$$\begin{aligned} L(\omega) &= \prod_{i=1}^n p_{\omega}(x_i, y_i) \\ &= \prod_{i=1}^n p_{\omega}(y_i | x_i) p(x_i) \end{aligned}$$

$$Y_i | x_i = \text{Ber}(q_i(x_i))$$

$$p_{\omega}(y_i | x_i) = \mathbb{P}(Y_i = y_i | x_i = x_i)$$

$$\begin{aligned} q(x_i) &= \mathbb{P}(Y_i = 1 | x_i = x_i) \\ &= \mathbb{P}(f_{\omega}(x_i) + \varepsilon_i \geq 0) \end{aligned}$$

$$\begin{aligned} &= \int_{\mathbb{R}} \mathbb{1} \{ f_{\omega}(x_i) + g \geq 0 \} \\ &\quad \times p(g) dg. \end{aligned}$$

where p is the density of $\omega(x_i, \cdot)$.

$$L(w) = \prod_{i=1}^N p_w(x_i, y_i) \\ = \prod_{i=1}^N p_w(y_i | x_i) p(x_i)$$

$$q(x_i) = P(Y_i = 1 | X_i = x_i)$$

$$z_i \stackrel{\text{iid}}{\sim} \mathcal{B}(\theta) \quad \theta \in (0, 1).$$

$$L^z(\theta | (z_i)) = \prod \theta^{z_i} (1 - \theta)^{1 - z_i}$$

Applying the analogy with
 $\theta \leftrightarrow p(x_i)$ and $z_i \leftarrow y_i$.

we get

$$L(w) = \prod_{i=1}^N \left[q(x_i)^{y_i} (1 - q(x_i))^{1 - y_i} \times p(x_i) \right]$$

$$\prod \theta^{z_i} (1 - \theta)^{1 - z_i} = \theta^{\sum z_i} (1 - \theta)^{N - \sum z_i}$$

$$L(\omega) = \prod_{i=1}^n p_{\omega}(x_i, y_i) \\ = \prod_{i=1}^n p_{\omega}(y_i | x_i) p(x_i)$$

$$q(x_i) = P(Y_i = 1 | X_i = x_i)$$

$$p_{\omega}(y_i | x_i) = q(x_i)^{y_i} (1 - q(x_i))^{1 - y_i}$$

Bayes formula.

- $p(x, y)$ a complete likelihood
- $p(x|y) p(y)$

$$p(y|x) = \frac{p(x|y) p(y)}{p(x)}.$$

Binomial PG model

$$Y_i \sim \text{Bin}(q) \quad q \in (0, 1)$$

$$L(q | (y_i)) = \prod q^{y_i} (1-q)^{1-y_i}.$$

Beta distribution on $(0, 1)$
with density

$$t \mapsto \frac{t^{\alpha-1} (1-t)^{\beta-1}}{B(\alpha, \beta)} \quad \alpha, \beta > 0.$$

$t \in (0, 1)$

Then posterior

$$p(q | (y_i)) \propto L(q | (y_i)) p_{\text{prior}}(q)$$

$$\propto \left[\prod q^{y_i} (1-q)^{1-y_i} \right] q^{\alpha-1} (1-q)^{\beta-1}$$

$$\propto q^{\sum y_i + \alpha - 1} (1-q)^{N - \sum y_i + \beta - 1}.$$

$$= \text{Beta}(\sum y_i + \alpha, N - \sum y_i + \beta).$$

$$y_i = f_{\omega}(x_i) + \sigma \varepsilon_i \quad \varepsilon_i \stackrel{iid}{\sim} N(0,1)$$

$$(y_i)_{i=1}^N = N_n \left(\begin{pmatrix} f_{\omega}(x_1) \\ \vdots \\ f_{\omega}(x_n) \end{pmatrix}, \sigma^2 I_n \right)$$

$$\Sigma^{-1} = \sigma^{-2} I_n$$

$$y_i = f_w(x_i) + \sigma \varepsilon_i \quad \varepsilon_i \stackrel{\text{iid}}{\sim} \mathcal{N}(0, 1)$$

$$w \mapsto p_{\text{prior}}(w)$$

$$p(w | \mathcal{D}) = ?$$

$$L(w; \mathcal{D}) = \prod \frac{1}{(\sigma \sqrt{2\pi})^{1/2}} \exp\left(-\frac{(y_i - f_w(x_i))^2}{2\sigma^2}\right)$$

$$p(w | \mathcal{D}) \propto L(w; \mathcal{D}) p_{\text{prior}}(w).$$

$$\text{in the case } p_{\text{prior}}(w) = \mathcal{N}(w_0, S_0)$$

$$p(w | \mathcal{D}) \propto \exp\left(-\frac{\|y - \phi_X w\|_B^2}{2} - \frac{1}{2} \langle S_0^{-1} (w - w_0), w - w_0 \rangle\right)$$

We want to show that

$p(w | \mathcal{D})$ is Gaussian.

We show

$$p(\omega | D) \propto \exp(-\omega^T A \omega + b^T \omega)$$

$$\propto \exp\left(-\frac{1}{2} \langle \Sigma_N^{-1} (\omega - m_N), \omega - m_N \rangle\right)$$

where

$$\Sigma_N^{-1} = 2A$$

$$\Sigma_N^{-1} m_N = b$$

$$p(\omega | D) \propto \exp\left(-\frac{\|y - \Phi_x \omega\|_B^2}{2} - \frac{1}{2} \langle \Sigma_0^{-1} (\omega - m_0), \omega - m_0 \rangle\right)$$

$$\propto \exp\left(-\frac{1}{2} \langle (B \Phi_x^T \Phi_x + \Sigma_0^{-1}) \omega, \omega \rangle\right)$$

$$+ \langle B \Phi_x^T y + \Sigma_0^{-1} m_0, \omega \rangle.$$

$$\Sigma_z^{-1} = \beta \Phi_x^T \Phi_x + \Sigma_0^{-1}$$

$$m_z = \Sigma_z (\beta \Phi_x^T y + \Sigma_0^{-1} m_0).$$