

1) a independent observations  $X_1, \dots, X_N$   $X_i \sim \text{Bernoulli}(w)$

$$P(X_i | w) = w^{x_i} (1-w)^{(1-x_i)}$$

$x \in \{0, 1\}$  sample space  
 $w \in [0, 1]$

$$p(w) = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} w^{a-1} (1-w)^{b-1} \quad w \sim \text{Beta}(a, b)$$

Bayes Rule  $p(w | x_1, \dots, x_N) = \frac{p(x_1, \dots, x_N | w) p(w | a, b)}{p(x_1, \dots, x_N)} = \prod_{i=1}^N [w^{x_i} (1-w)^{(1-x_i)}] \cdot w^{a-1} (1-w)^{b-1} \cdot \underbrace{\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} p(x_1, \dots, x_N)}_k$

$$\propto w^{\sum_{i=1}^N x_i + a - 1} (1-w)^{\sum_{i=1}^N (1-x_i) + b - 1}$$

$$\propto w | x_1, \dots, x_N, a, b \sim \text{Beta} \left( \sum_{i=1}^N x_i + a, \sum_{i=1}^N (1-x_i) + b \right)$$

1) b  $f(y_1, y_2) = f_{x_1, x_2}(g_1^{-1}(y_1, y_2), g_2^{-1}(y_1, y_2)) | J(y_1, y_2) |$

$$y_1 = x_1 / (x_1 + x_2) \rightarrow \begin{cases} x_1 = y_1 y_2 = g_1^{-1}(y_1, y_2) \\ x_2 = y_2 (1-y_1) = g_2^{-1}(y_1, y_2) \end{cases}$$

$$y_2 = x_1 + x_2$$

$$J(y_1, y_2) = \begin{vmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{vmatrix} = \begin{vmatrix} y_2 & y_1 \\ -y_2 & 1-y_1 \end{vmatrix} = |y_2|$$

Assume that  $x_1, x_2$  are independent

$$f_{x_1, x_2} = f_{x_1} f_{x_2}$$

$$= \frac{1}{\Gamma(a)} x_1^{a-1} \exp(-bx_1) \cdot \frac{1}{\Gamma(a_2)} x_2^{a_2-1} \exp(-bx_2)$$

$$f_{x_1, x_2}(y_1, y_2) = \frac{1}{\Gamma(a)\Gamma(a_2)} (y_1, y_2)^{a-1} \exp(-y_1 y_2) (y_2 (1-y_1))^{a_2-1} \exp(-y_2 (1-y_1)) \cdot |y_2|$$

$$= \frac{1}{\Gamma(a)\Gamma(a_2)} y_1^{a-1} (1-y_1)^{a_2-1} y_2^{(a_1+a_2-1)-1+1} \exp(-y_2)$$

$$\propto \underbrace{y_1^{a-1} (1-y_1)^{a_2-1}}_{\text{Beta}(a, a_2)} \underbrace{y_2^{a_1+a_2-1} \exp(-y_2)}_{\text{Gamma}(a_1+a_2, 1)}$$

1) c  $x_i \sim N(\theta, \sigma^2)$   $\sigma^2$  is known

$$\theta \sim N(m, v)$$

Find  $p(w | x_1, \dots, x_N)$

$$p(\theta | x_1, \dots, x_N) = p(x_1, \dots, x_N | \theta) p(\theta)$$

$$= p(x_1 | \theta) \dots p(x_N | \theta) p(\theta) \propto \exp\left(-\frac{(x_1 - \theta)^2}{2\sigma^2}\right) \dots \exp\left(-\frac{(x_N - \theta)^2}{2\sigma^2}\right) \exp\left(-\frac{(\theta - m)^2}{2v}\right)$$

$$= \exp\left(-\frac{\sum_{i=1}^N (x_i - \theta)^2}{2\sigma^2} - \frac{(\theta - m)^2}{2v}\right) \propto \exp\left(\theta^2\left(-\frac{N}{2\sigma^2} + \frac{1}{2v}\right) + \theta\left(\frac{\sum_{i=1}^N x_i}{\sigma^2} + \frac{m}{v}\right)\right)$$

$$\propto \exp\left(-\frac{1}{2} \left(\theta - \underbrace{\left(\frac{\sum_{i=1}^N x_i + m}{\frac{N}{\sigma^2} + \frac{1}{v}}\right)}_{\left(\frac{N}{\sigma^2} + \frac{1}{v}\right)}\right)^2\right)$$

1) d  $x_1, \dots, x_N \text{ iid } \sim N(\bar{\theta}, \sigma^2 = 1/w)$

$$w \sim \text{Gamma}(a, b) \propto w^{a-1} e^{-bw}$$

$$w \in (0, \infty)$$

$$p(w | x) \propto \underbrace{[w^{n/2} \exp(-\frac{1}{2} w \sum (x_i - \bar{\theta})^2)]}_{\text{Likelihood}} \underbrace{[w^{a-1} e^{-bw}]}_{\text{Prior}}$$

$$\propto w^{\frac{n}{2} + a - 1} e^{-w(b + \frac{1}{2} \sum (x_i - \bar{\theta})^2)}$$

$$\sim \text{Gamma}\left(\frac{n}{2} + a, b + \frac{1}{2} \sum (x_i - \bar{\theta})^2\right)$$

$$w = \frac{1}{\sigma^2} \rightarrow \frac{dw}{d\sigma^2} = -\frac{1}{(\sigma^2)^2}$$

$$p_{\sigma^2}(\sigma^2) = p_w(w) \det\left(-\frac{1}{(\sigma^2)^2}\right) = \sigma^{2(-a - \frac{n}{2} - 1)} \exp\left(\frac{-1}{\sigma^2} \left[b + \frac{1}{2} \sum (x_i - \bar{\theta})^2\right]\right)$$

$$\text{IG} \propto x^{-a-1} e^{-\frac{b}{x}}$$

$$\sigma^2 \sim \text{IG}\left(a + \frac{n}{2}, b + \frac{1}{2} \sum (x_i - \bar{\theta})^2\right)$$

$$1) e \quad x_1, \dots, x_N \stackrel{iid}{\sim} N(\bar{\theta}, \sigma^2 = 1/w)$$

$$w \sim \text{Gamma}(a, b) \sim w^{a-1} e^{-bw}$$

$$x_i \sim N(\theta, \sigma_i^2)$$

$$p(\theta | x_1, \dots, x_N) \propto p(\theta) p(x_1, \dots, x_N | \theta)$$

$$\propto p(\theta / \prod_{i=1}^N p(x_i | \theta))$$

$$\propto \exp\left(-\frac{1}{2v}(\theta - m)^2\right) \prod_{i=1}^N \exp\left(-\frac{1}{2\sigma_i^2}(x_i - \theta)^2\right)$$

$$\propto \exp\left(-\frac{1}{2v}(\theta^2 - 2m\theta)\right) \cdot \exp\left(\sum_{i=1}^N -\frac{1}{2\sigma_i^2}(\theta^2 - 2x_i\theta)\right)$$

$$\propto \exp\left(-\frac{1}{2}\left[\left(\frac{1}{v} + \sum_{i=1}^N \frac{1}{\sigma_i^2}\right)\theta^2 - 2\left(\frac{m}{v} + \sum_{i=1}^N \frac{x_i}{\sigma_i^2}\right)\theta\right]\right)$$

$$\propto \exp\left(-\frac{1}{2}\left(\frac{1}{v} + \sum_{i=1}^N \frac{1}{\sigma_i^2}\right)\left(\theta - \frac{\frac{1}{v}m + \sum_{i=1}^N \frac{1}{\sigma_i^2}x_i}{\frac{1}{v} + \sum_{i=1}^N \frac{1}{\sigma_i^2}}\right)^2\right)$$

$$\sim N\left(\frac{\frac{1}{v}m + \sum_{i=1}^N \frac{1}{\sigma_i^2}x_i}{\frac{1}{v} + \sum_{i=1}^N \frac{1}{\sigma_i^2}}, \sqrt{\frac{1}{\left(\frac{1}{v} + \sum_{i=1}^N \frac{1}{\sigma_i^2}\right)^2}}\right)$$

$$x_i \sim N(\theta, \sigma^2)$$

$$\theta \sim N(m, v)$$

$$\hat{\theta} = \frac{\frac{1}{v}m + \frac{n}{\sigma^2}\bar{x}}{\frac{1}{v} + \frac{n}{\sigma^2}}$$

$$= cm + (1-c)\bar{x}$$

$$c = \frac{\frac{1}{v}}{\frac{1}{v} + \frac{n}{\sigma^2}}$$

1)f

$$(x|\sigma^2) \sim N(0, \sigma^2)$$

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

$$\frac{1}{\sigma^2} \sim \text{Ga}(a, b) \text{ prior}$$

$$f(1/\sigma^2) = \frac{b^a}{\Gamma(a)} \left(\frac{1}{\sigma^2}\right)^{a-1} \exp(-b/\sigma^2)$$