

Goal: $P(\alpha, \beta | Y) \propto \prod_j P(Y_j | \alpha, \beta) P(\alpha, \beta)$

$$P(Y_j | \alpha, \beta) = \int_0^1 P(Y_j, \theta_j | \alpha, \beta) d(\theta_j) = \int_0^1 P(Y_j | \theta_j) P(\theta_j | \alpha, \beta) d\theta_j$$

$$= \int_0^1 \binom{n_j}{y_j} \theta_j^{y_j} (1-\theta_j)^{n_j-y_j} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \theta_j^{\alpha-1} (1-\theta_j)^{\beta-1} d\theta_j$$

$$= \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \binom{n_j}{y_j} \int_0^1 \theta_j^{y_j+\alpha-1} (1-\theta_j)^{n_j-y_j+\beta-1} d\theta_j$$

$$= \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \binom{n_j}{y_j} \times \frac{\Gamma(y_j+\alpha) \Gamma(n_j-y_j+\beta)}{\Gamma(y_j+\alpha+n_j-y_j+\beta)}$$

$$\propto \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \frac{\Gamma(y_j+\alpha) \Gamma(n_j-y_j+\beta)}{\Gamma(y_j+\alpha+n_j-y_j+\beta)}$$

Beta-Binomial

$$\propto \frac{\Gamma(\alpha+y_j)}{\Gamma(\alpha)} \cdot \frac{\Gamma(\beta+n-y_j)}{\Gamma(\beta)}$$

$$\frac{\Gamma(\alpha+\beta+n_j)}{\Gamma(\alpha+\beta)}$$

Using this

$$P(\alpha, \beta | Y) \propto \prod_{j=1}^n \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) \Gamma(\beta)} \frac{\Gamma(y_j + \alpha) \Gamma(n_j - y_j + \beta)}{\Gamma(y_j + \alpha + n_j - y_j + \beta)} P(\alpha, \beta)$$

Rmk: $X \sim \text{Beta}(\alpha, \beta)$ $E[X] = \frac{\alpha}{\alpha + \beta}$ $\text{Var}(X) = \frac{n(1-n)}{\alpha + \beta - 1}$

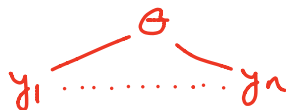
One idea: weakly informative priors $P(\frac{1}{\alpha + \beta}, \sqrt{\alpha + \beta}) \propto 1$

$$\Rightarrow P(\alpha, \beta) \propto (\alpha + \beta)^{-5/2}$$

They also use a reparameterization in $r = \text{logit}(\frac{\alpha}{\alpha + \beta})$

$$s = \log(\alpha + \beta) \text{ and use } P(r, s) \propto \alpha \beta (\alpha + \beta)^{-7/2}$$

Ex: $Y_1, \dots, Y_n | \theta \stackrel{iid}{\sim} N(\theta, \sigma^2)$
 $\theta \sim N(\theta_0, \tau^2)$



$$Y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} | \theta \sim N(\vec{\theta}, \sigma^2 I)$$

$$1_n \cdot \theta \sim N(\theta_0, \tau^2 11^T)$$

$$Y \sim N(\vec{\theta}_0, \sigma^2 I + \tau^2 11^T)$$

Remark: Correlated Data.