

$$1) \quad f(\mathbf{g}|\mathbf{h}, r) = \frac{\sqrt{r}}{\sqrt{2\pi}} \exp\left(-r\frac{(\mathbf{g}-\mathbf{h})^2}{2}\right) \Rightarrow f(\mathbf{y}_1, \dots, \mathbf{y}_n | \mathbf{g}, r) = \left(\frac{r}{2\pi}\right)^{\frac{n}{2}} \exp\left(-r\frac{\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2}{2}\right)$$

$$f(\theta|r) = \frac{\sqrt{r}}{\sqrt{2\pi}} \exp\left(-\frac{\lambda r(\theta - \mathcal{M})^2}{2}\right)$$

$$f(r) = \left(\frac{\lambda}{2}\right)^{\frac{n}{2}} r^{\frac{n}{2}-1} e^{-\frac{\lambda}{2} r}$$

$$\Rightarrow \quad P(\theta | r, \mathbf{g}) = \frac{f(\theta, r, \mathbf{g})}{f(r, \mathbf{g})} \cdot \frac{f(\theta|r)}{f(\theta|r)}$$

$$= \frac{f(\theta, r, \mathbf{g})}{f(\theta, r)} \cdot \frac{f(\theta|r)}{f(r, \mathbf{g})}$$

$$= f(\mathbf{g}|\theta, r) \cdot f(r) \cdot \frac{f(\theta|r)}{f(r, \mathbf{g})}$$

$$P(\theta | r, \mathbf{g}) = f(\mathbf{g}|\theta, r) \cdot f(r) \cdot \frac{f(\theta|r)}{f(r, \mathbf{g})}$$

$$\propto f(\theta, r, \mathbf{g}) = f(\mathbf{g}|\theta, r) f(r) f(\theta|r)$$

$$= \left(\frac{r}{2\pi}\right)^{\frac{n}{2}} \exp\left(-r\frac{\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2}{2}\right) \cdot \left(\frac{\lambda}{2}\right)^{\frac{n}{2}} r^{\frac{n}{2}-1} e^{-\frac{\lambda}{2} r} \cdot \frac{\sqrt{r}}{\sqrt{2\pi}} \exp\left(-\frac{\lambda r(\theta - \mathcal{M})^2}{2}\right)$$

$$\propto \exp\left(-r\frac{n\theta^2 - 2\theta\sum \mathbf{g} + \sum \mathbf{g}^2}{2}\right) \exp\left(-\frac{\lambda r(\theta - \mathcal{M})^2}{2}\right)$$

$$= \exp\left(-\frac{r}{2} \frac{n(\theta^2 - 2\theta\sum \mathbf{g} + \sum \mathbf{g}^2) - \lambda(\theta^2 - 2\theta\mathcal{M} + \mathcal{M}^2)}{2}\right)$$

$$= \exp\left\{-\frac{r\theta^2 + 2r\theta\sum \mathbf{g} - r\sum \mathbf{g}^2 - 2\lambda r\theta\mathcal{M} - \lambda\mathcal{M}^2}{2}\right\}$$

$$\propto \exp\left\{-\frac{r\theta^2 + 2r\theta\sum \mathbf{g} - r\sum \mathbf{g}^2 - 2\lambda r\theta\mathcal{M}}{2}\right\}$$

$$= \exp\left\{-\frac{\theta^2 r(n+\lambda) - 2\theta r(\sum \mathbf{g} + \lambda\mathcal{M})}{2}\right\}$$

$$\propto \exp\left\{-\frac{\left(\theta r(n+\lambda) - \sum r(\sum \mathbf{g} + \lambda\mathcal{M})\right)^2}{2(n+\lambda)}\right\}$$

$$= \exp\left\{-\frac{r(n+\lambda)\left(\theta - \frac{\sum \mathbf{g} + \lambda\mathcal{M}}{n+\lambda}\right)^2}{2}\right\} \sim N\left(\frac{\sum \mathbf{g} + \lambda\mathcal{M}}{n+\lambda}, \frac{1}{r(n+\lambda)}\right) \quad \therefore \quad \theta | r, \mathbf{g} \sim N\left(\frac{\sum \mathbf{g} + \lambda\mathcal{M}}{n+\lambda}, \frac{1}{r(n+\lambda)}\right)$$

$$\Rightarrow \quad P(r|\mathbf{g}) = \frac{f(r, \mathbf{g})}{f(\mathbf{g})} \quad P(\theta | r, \mathbf{g}) = \frac{f(r(n+\lambda))}{\sqrt{2\pi}} \exp\left[-\frac{r(n+\lambda)}{2}\left(\theta - \frac{\sum \mathbf{g} + \lambda\mathcal{M}}{n+\lambda}\right)^2\right]$$

$$= \frac{f(r, \mathbf{g})}{f(\theta, r, \mathbf{g})} \cdot \frac{f(\theta, r, \mathbf{g})}{f(\theta, r)} \cdot \frac{f(\theta, r)}{f(\mathbf{g})}$$

$$= \frac{1}{f(\theta | r, \mathbf{g})} \cdot f(\mathbf{g} | \theta, r) \cdot \frac{f(\theta, r)}{f(\mathbf{g})}$$

$$\propto \frac{1}{f(\theta | r, \mathbf{g})} \cdot f(\mathbf{g} | \theta, r) \cdot f(\theta, r) \quad , \quad P(\theta, r) = P(\theta|r) P(r) = \frac{\sqrt{r}}{\sqrt{2\pi}} \exp\left(-\frac{\lambda r(\theta - \mathcal{M})^2}{2}\right) \left(\frac{\lambda}{2}\right)^{\frac{n}{2}} r^{\frac{n}{2}-1} e^{-\frac{\lambda}{2} r}$$

$$= \frac{\sqrt{2\pi}}{\sqrt{r(n+\lambda)}} \exp\left[\frac{r(n+\lambda)}{2}\left(\theta - \frac{\sum \mathbf{g} + \lambda\mathcal{M}}{n+\lambda}\right)^2\right] \left(\frac{r}{2\pi}\right)^{\frac{n}{2}} \exp\left(-r\frac{\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2}{2}\right) \frac{\sqrt{r}}{\sqrt{2\pi}} \exp\left(-\frac{\lambda r(\theta - \mathcal{M})^2}{2}\right) \left(\frac{\lambda}{2}\right)^{\frac{n}{2}} r^{\frac{n}{2}-1} e^{-\frac{\lambda}{2} r}$$

$$\propto \frac{1}{r^{\frac{n+1}{2}}} \exp\left[\frac{r(n+\lambda)}{2}\left(\theta - \frac{\sum \mathbf{g} + \lambda\mathcal{M}}{n+\lambda}\right)^2\right] (r)^{\frac{n}{2}} \exp\left(-r\frac{\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2}{2}\right) \sqrt{r} \exp\left(-\frac{\lambda r(\theta - \mathcal{M})^2}{2}\right) r^{\frac{n}{2}-1} e^{-\frac{\lambda}{2} r}$$

$$= r^{\frac{n+1}{2}-1} \exp\left[\frac{r(n+\lambda)}{2}\left(\theta - \frac{\sum \mathbf{g} + \lambda\mathcal{M}}{n+\lambda}\right)^2\right] \exp\left(-r\frac{\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2}{2}\right) \exp\left(-\frac{\lambda r(\theta - \mathcal{M})^2}{2}\right) \exp\left(-r\frac{\lambda}{2}\right)$$

$$= r^{\frac{n+1}{2}-1} \exp\left\{\frac{r(n+\lambda)}{2}\left(\frac{\theta(n+\lambda) - (n\sum \mathbf{g} + \lambda\mathcal{M})}{n+\lambda}\right)^2 - r\frac{\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2}{2} - \frac{\lambda r(\theta - \mathcal{M})^2}{2} - r\frac{\lambda}{2}\right\}$$

$$= r^{\frac{n+1}{2}-1} \exp\left\{-\frac{r}{2}\left[n+\lambda\left(\frac{(n\sum \mathbf{g} + \lambda\mathcal{M}) - \theta(n+\lambda)}{n+\lambda}\right)^2 + \sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + \lambda(\theta - \mathcal{M})^2 + b\right]\right\}$$

$$= r^{\frac{n+1}{2}-1} \exp\left\{-\frac{r}{2}\left[n+\lambda\left(\frac{(n\sum \mathbf{g} + \lambda\mathcal{M}) - \theta(n+\lambda)}{n+\lambda}\right)^2 + \sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + \lambda(\theta - \mathcal{M})^2 + b\right]\right\}$$

$$= r^{\frac{n+1}{2}-1} \exp\left\{-\frac{r}{2}\left[\underbrace{\left((n\sum \mathbf{g} + \lambda\mathcal{M}) - \theta(n+\lambda)\right)^2}_{n+\lambda} + (n+\lambda)\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + (n+\lambda)\left(\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + n(\sum \mathbf{g} - \mathbf{g})^2 + n(\sum \mathbf{g} - \mathbf{g})^2\right) + (\lambda n + \lambda^2)\left(\theta^2 - 2\theta\mathcal{M} + \mathcal{M}^2\right) + (n+\lambda)b\right]\right\}$$

$$= -n^2\sum \mathbf{g}^2 - 2n\sum \mathbf{g}\lambda\mathcal{M} - \lambda^2\mathcal{M}^2 + 2\theta n\sum \mathbf{g} + 2\theta\lambda\sum \mathbf{g} + 2\theta\lambda^2\mathcal{M} - \theta^2 n^2 - 2\theta n\lambda - \theta^2\lambda^2 + (n+\lambda)\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + n^2\sum \mathbf{g}^2 + \lambda n\sum \mathbf{g}^2 - 2n\sum \mathbf{g}\theta - 2\lambda\sum \mathbf{g}\theta + \theta^2\theta^2 + \lambda n\theta^2 + \lambda n\theta^2 - 2\lambda n\theta\mathcal{M} + \mathcal{M}^2\lambda n + \lambda^2\theta^2 - 2\lambda^2\theta\mathcal{M} + \lambda^2\mathcal{M}^2 + (n+\lambda)b$$

$$= (n+\lambda)\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + n\lambda\left(\sum \mathbf{g}^2 - 2\sum \mathbf{g}\mathcal{M} + \mathcal{M}^2\right) + (n+\lambda)b = (n+\lambda)\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + n\lambda\left(\sum \mathbf{g} - \mathcal{M}\right)^2 + (n+\lambda)b$$

$$= r^{\frac{n+1}{2}-1} \exp\left\{-\frac{r}{2}\left(\sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + \frac{n\lambda}{n+\lambda}\left(\sum \mathbf{g} - \mathcal{M}\right)^2 + b\right)\right\}$$

$$\sim \text{Gamma}\left(\frac{n+\alpha}{2}, \sum_{i=1}^n (\mathbf{y}_i - \mathbf{g})^2 + \frac{n\lambda}{n+\lambda}\left(\sum \mathbf{g} - \mathcal{M}\right)^2 + b\right)$$

2)

$$p(y|\theta) = \frac{e^{-\theta} \theta^y}{y!}$$

$$L(\theta|\mathbf{y}) = \frac{e^{-n\theta} \theta^{\sum_{i=1}^n y_i}}{\prod_{i=1}^n (y_i!)} = \frac{e^{-n\theta} \theta^{\sum_{i=1}^n y_i}}{\prod_{i=1}^n (y_i!)}$$

$$L(\theta|\mathbf{y}) = -n\theta + (\sum_{i=1}^n y_i) \ln(\theta) - \sum_{i=1}^n \ln(y_i!)$$

$$\frac{\partial}{\partial \theta} L(\theta|\mathbf{y}) = -n + \frac{1}{\theta} \sum_{i=1}^n y_i$$

$$\frac{\partial^2}{\partial \theta^2} L(\theta|\mathbf{y}) = -\frac{1}{\theta^2} \sum_{i=1}^n y_i \quad \therefore E\left[-\frac{\partial^2}{\partial \theta^2} L(\theta|\mathbf{y})\right] = \frac{n}{\theta} = I(\theta)$$

$\therefore p(\theta) \propto I^{-1/2}(\theta) = \sqrt{\frac{n}{\theta}} \propto \theta^{-1/2} = \theta^{\frac{1}{2}-1} \exp(-0 \cdot \theta)$, it appears to follow a Gamma distribution with $\alpha = \frac{1}{2}$, but it does not have a corresponding β . Hence, it is an improper prior.

3)

a) $f(x|\theta) = \theta^\tau (1-\theta)^{1-\tau}$, $\tau = \frac{\theta}{1-\theta} \Leftrightarrow \theta = \frac{\tau}{1+\tau}$

$$f(y|\theta) = \binom{n}{y} \theta^y (1-\theta)^{n-y} = \binom{n}{y} \left(\frac{\theta}{1-\theta}\right)^y (1-\theta)^n$$

$$\Rightarrow f(y|\tau) = \binom{n}{y} \tau^y (1+\tau)^{-n}$$

$$\propto \tau^y (1+\tau)^{-n}$$

b) $L(\theta|y) = \ln \binom{n}{y} + y \ln \theta + (n-y) \ln(1-\theta)$

$$\frac{\partial}{\partial \theta} L(\theta|y) = \frac{y}{\theta} - \frac{n-y}{1-\theta}$$

$$\frac{\partial^2}{\partial \theta^2} L(\theta|y) = -\frac{y}{\theta^2} - \frac{n-y}{(1-\theta)^2}$$

$$E\left[-\frac{\partial^2}{\partial \theta^2} L(\theta|y)\right] = \frac{n}{\theta} + \frac{n}{1-\theta} = n \left(\frac{1}{\theta(1-\theta)}\right) \propto \frac{1}{\theta(1-\theta)} = I(\theta)$$

$$\Rightarrow p(\theta) \propto \theta^{\frac{1}{2}-1} (1-\theta)^{\frac{1}{2}-1}, \quad \theta = \frac{\tau}{1+\tau}, \quad d\theta = \frac{1}{(1+\tau)^2} d\tau$$

$$\Rightarrow I^{\frac{1}{2}}(\tau) = I^{\frac{1}{2}}\left(\frac{\tau}{1+\tau}\right) \left|\frac{d\theta}{d\tau}\right|$$

$$= \left(\frac{\tau}{1+\tau}\right)^{\frac{1}{2}-1} \left(1 - \frac{\tau}{1+\tau}\right)^{\frac{1}{2}-1} \frac{1}{(1+\tau)^2}$$

$$= \frac{1}{\Gamma(\frac{1}{2})^2} \frac{1}{(1+\tau)^2}$$

$$= \tau^{\frac{1}{2}-1} (1+\tau)^{-2} \propto p(\tau)$$

c)

$\theta = \frac{\tau}{1+\tau}$ is a one-to-one transformation, then

$$I(\theta) = E\left[\left(\frac{\partial \ln p(y|\theta)}{\partial \theta}\right)^2 \middle| \theta\right], \text{ where } \frac{\partial \ln p(y|\theta)}{\partial \theta} = \frac{\partial \ln p(y|\frac{\tau}{1+\tau})}{\partial \tau} \cdot \frac{\partial \tau}{\partial \theta}$$

$$= \left(\frac{\partial \tau}{\partial \theta}\right)^2 E\left[\left(\frac{\partial \ln p(y|\tau)}{\partial \tau}\right)^2 \middle| \tau\right]$$

$$\Rightarrow I^{\frac{1}{2}}(\theta) = I^{\frac{1}{2}}(\tau) \left|\frac{\partial \tau}{\partial \theta}\right|$$

$$p(\theta) = p(\tau) \left|\frac{\partial \tau}{\partial \theta}\right|$$

$$4) \ a) \ y \sim B(n, \theta) \quad , \quad \theta \sim \text{Beta}(\alpha, \beta)$$

$$\Rightarrow P(y|\theta) = \binom{n}{y} \theta^y (1-\theta)^{n-y}$$

$$P(\theta) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1}$$

$$\begin{aligned} \Rightarrow P(y) &= \int_0^1 P(y|\theta) P(\theta) d\theta \\ &= \binom{n}{y} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \int_0^1 \theta^{\alpha+y-1} (1-\theta)^{n+y-\beta-1} d\theta \\ &= \binom{n}{y} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \frac{\Gamma(\alpha+y)\Gamma(n+\beta-y)}{\Gamma(n+\alpha+\beta)} \int_0^1 \frac{\Gamma(n+\alpha+\beta)}{\Gamma(\alpha+y)\Gamma(n+\beta-y)} \theta^{\alpha+y-1} (1-\theta)^{n+y-\beta-1} d\theta \\ &= \binom{n}{y} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \frac{\Gamma(\alpha+y)\Gamma(n+\beta-y)}{\Gamma(n+\alpha+\beta)} \end{aligned}$$

$$b) \text{ For a fixed constant } C, \quad P(y) = C \text{ if and only if } \alpha = \beta = 1.$$

$$\Rightarrow \text{ If } \alpha, \beta \neq 0 \text{ for at least one, } P(y_i) \neq P(y_j) \quad \forall i \neq j$$

$$\begin{aligned} P(y|\alpha', \beta') &= \binom{n}{y} \frac{\Gamma(\alpha'+\beta')}{\Gamma(\alpha')\Gamma(\beta')} \frac{\Gamma(y+\alpha')\Gamma(n-y+\beta')}{\Gamma(n+\alpha'+\beta')} \\ &= \frac{n!}{y!(n-y)!} \frac{(\alpha'-1)!(\beta'-1)!}{(\alpha'-1)!(\beta'-1)!} \frac{(y+\alpha'-1)!(n-y+\beta'-1)!}{(n+\alpha'+\beta'-1)!} \\ &\propto \frac{(y+\alpha'-1)!(n-y+\beta'-1)!}{y!(n-y)!} = 1 \quad \text{iff } \alpha' = \beta' = 1 \end{aligned}$$

\therefore If the beta-binomial probability is constant in y , then the prior distribution has to have $\alpha = \beta = 1$.