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– **1. Exchangeability with known model parameters**

- (a) $p(y_1 = \mathbf{b}, y_2 = \mathbf{w}) = p(y_1 = \mathbf{w}, y_2 = \mathbf{b}) = 1/4$. Observations y_1 and y_2 are exchangeable. We put the balls back, ergo y_1 and y_2 are independent.
- (b) y_1 and y_2 are exchangeable, but not independent.
- (c) y_1 and y_2 are exchangeable, but not independent. However, there are 1 million balls for each color, and we only take twice. Hence, we can act as if two observations are independent.

– **5. Mixtures of independent distributions**

W.L.O.G, let's compute the covariance of θ_i and θ_j :

$$\begin{aligned}
 \text{cov}(\theta_i, \theta_j) &= \int_{\theta_i} \int_{\theta_j} p(\theta_i, \theta_j) (\theta_i - \mu_i) (\theta_j - \mu_j) d\theta_i d\theta_j \\
 &= \int_{\alpha} \int_{\theta_i} \int_{\theta_j} p(\theta_i, \theta_j | \alpha) p(\alpha) (\theta_i - \mu_i) (\theta_j - \mu_j) d\alpha d\theta_i d\theta_j \\
 &= \int_{\alpha} \int_{\theta_i} \int_{\theta_j} p(\theta_i | \alpha) (\theta_i - \mu_i) d\theta_i p(\theta_j | \alpha) (\theta_j - \mu_j) d\theta_j p(\alpha) d\alpha \\
 &= \int_{\alpha} \left(\int_{\theta} p(\theta | \alpha) (\theta - \mu) d\theta \right)^2 p(\alpha) d\alpha \\
 &\geq 0
 \end{aligned}$$

– **Continuous mixture models**

(a) We have

$$\begin{aligned}
 y | \theta &\sim \frac{1}{y!} \theta^y e^{-\theta} \\
 \theta | \alpha, \beta &\sim \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\beta\theta}
 \end{aligned}$$

Therefore, the prior predictive distribution of y :

$$\begin{aligned}
 p(y | \alpha, \beta) &= \int_{\theta} p(y, \theta | \alpha, \beta) d\theta \\
 &= \int_{\theta} p(y | \theta) p(\theta | \alpha, \beta) d\theta \\
 &= \frac{\beta^\alpha}{y! \Gamma(\alpha)} \int \theta^{\alpha+y-1} e^{-(\beta+1)\theta} d\theta \\
 &= \frac{\Gamma(\alpha+y)}{y! \Gamma(\alpha)} \frac{\beta^\alpha}{(\beta+1)^{\alpha+y}} \int \frac{(\beta+1)^{\alpha+y}}{\Gamma(\alpha+y)} \theta^{\alpha+y-1} e^{-(\beta+1)\theta} d\theta \\
 &= \frac{\Gamma(\alpha+y)}{y! \Gamma(\alpha)} \left(\frac{\beta}{\beta+1} \right)^\alpha \left(\frac{1}{\beta+1} \right)^y
 \end{aligned}$$

Hence, the negative binomial.

To calculate the mean and variance of $y|\alpha, \beta$, we have

$$\mathbb{E}(y) = \mathbb{E}(\mathbb{E}(y|\theta)) = \mathbb{E}(\theta) = \frac{\alpha}{\beta}$$

and

$$\mathbb{V}(y) = \mathbb{E}(\mathbb{V}(y|\theta)) + \mathbb{V}(\mathbb{E}(\theta|y)) = \mathbb{E}(\theta) + \mathbb{V}(\theta) = \frac{\alpha}{\beta^2}(\beta + 1)$$