

(iii) uniform prior $[-1, 1]$ for ρ ...

Full conditional.

$$f(\rho | \mu, \sigma^2, y) \propto \prod_{t=2}^N N(y_t | (y_{t-1} - \mu)\rho + \mu, \sigma^2) \cdot \pi(\rho)$$

$$z_t = y_t - \mu \Rightarrow (z_{t-1} = y_{t-1} - \mu)$$

$$f(\rho | \mu, \sigma^2, y) \propto N(z_1 | z_{T-1}, \rho, \sigma^2) \cdot (-1 < \rho < 1)$$

$$\propto \exp\left[-\frac{1}{2\sigma^2} \sum_{t=2}^N (z_t - \rho z_{t-1})^2\right] \cdot (-1 < \rho < 1)$$

$$\propto \exp\left[-\frac{1}{2\sigma^2} \left(\rho^2 \sum_{t=2}^N z_t^2 - 2\rho \sum_{t=2}^N z_t z_{t-1} + \sum_{t=2}^N z_t^2\right)\right]$$

$$a = \sum_{t=2}^N z_t^2$$

$$b = \sum_{t=2}^N z_t z_{t-1}$$

$$\propto \exp\left[-\frac{1}{2\sigma^2} (a\rho^2 - 2b\rho)\right] = \exp\left[-\frac{1}{2\sigma^2} a \left(\rho^2 - \frac{2b\rho}{a} + \frac{b^2}{a^2}\right) + \frac{b^2}{a^2} \cdot \frac{1}{2\sigma^2}\right]$$

$$\propto \exp\left[\frac{1}{2\sigma^2} a \left(\rho - \frac{b}{a}\right)^2\right] = \exp\left[-\frac{1}{2} \left(\frac{a}{\sigma^2}\right) \left(\rho - \frac{b}{a}\right)^2\right] = \exp\left[-\frac{1}{2} \left(\frac{\rho - \frac{b}{a}}{\sqrt{\frac{1}{a \sigma^2}}}\right)^2\right]$$

$$\sim N\left(\frac{b}{a}, \frac{a}{\sigma^2}\right) \Rightarrow N\left(\frac{\sum_{t=2}^N z_t z_{t-1}}{\sum_{t=2}^N z_t^2}, \frac{\sigma^2}{\sum_{t=2}^N z_t^2}\right)$$

$$2.) p(x_i | \theta) = h(x_i) \exp\{\theta T(x_i) - c(\theta)\}$$

(a) Find expression for Jeffreys Prior

$$\pi(\theta) \propto \sqrt{I(\theta)} \Rightarrow \sqrt{I(\theta)} = -E \sqrt{\frac{\partial^2}{\partial \theta^2} [\log(L(\theta | x_1 \dots x_n))]}$$

$$I(\theta) = -n E \left[\frac{\partial^2}{\partial \theta^2} \log(L(\theta | x)) \right]$$

$$I(\theta) = -n E \left[\frac{\partial^2}{\partial \theta^2} n [\log(h(x)) + \theta T(x) - c(\theta)] \right]$$

$$I(\theta) = -n E \left[\frac{\partial}{\partial \theta} T(x) - c(\theta)' \right]$$

$$I(\theta) = -n E [-c(\theta)']$$

$$\Rightarrow \boxed{\pi(\theta) \propto \sqrt{nc(\theta)'}}$$

$$b.) \text{ iid Poisson} \Rightarrow \text{likelihood} = \frac{\theta^x e^{-\theta}}{x_i!} \Rightarrow \frac{1}{x_i!} \exp[x \log(\theta) - \theta]$$

$$c(\theta) = \exp(\theta)$$

$$\hookrightarrow c'(\theta) = \exp(\theta) d\theta$$

$$c'(\theta) = \frac{1}{\lambda} \exp(\theta)$$

$$c''(\theta) = \frac{1}{\lambda} d\theta \exp(\theta)$$

$$= \frac{1}{\lambda^2} \exp(\theta)$$

$$= \frac{\lambda}{\lambda^2} = \frac{1}{\lambda}$$

$$I(\lambda) = -n E [-c(\theta)'] = n E \left[\frac{1}{\lambda} \right] \Rightarrow \sqrt{\frac{n}{\lambda}}$$

$$\pi(\lambda) \propto \sqrt{I(\lambda)} = \boxed{\frac{1}{\sqrt{\lambda}}}$$

1.) y_1, \dots, y_n generated from...

$$y_T = \mu_T + \rho(y_{T-1} - \mu) + \varepsilon_T \quad \varepsilon_T \sim N(0, \sigma^2)$$

$$y_T = \log(r_T) \quad y_T = \mu_T + \rho y_{T-1} - \rho \mu \Rightarrow y_T - \rho y_{T-1} = \mu(1-\rho) \\ \Rightarrow \mu = \frac{y_T - \rho y_{T-1}}{1-\rho}$$

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Midterm 2

(a) Full conditional for precision...

(i) $f(\sigma^2 | \mu^T, \rho^T, y)$ - Gamma prior w/ parameters a and b

$$z_T = y_T - \mu \Rightarrow z_{T-1} = y_{T-1} - \mu$$

$$f(\sigma^2 | \mu^T, \rho^T, y) \propto \prod_{t=2}^N (z_t | z_{t-1}, \rho, \sigma^2) \cdot \text{IG}(\sigma^2 | a^2, b^2)$$

$$\propto (\sigma^2)^{-\frac{n-1}{2}} \exp\left\{-\frac{1}{2\sigma^2} \sum_{t=2}^N (z_t - \rho z_{t-1})^2\right\} (\sigma^2)^{-(a-1)} \exp\left\{-\frac{b}{\sigma^2}\right\}$$

→ Full conditional for precision follows Inverse Gamma

$$\sim \text{IG}\left(a + \frac{n-1}{2}, b + \frac{1}{2} \sum_{t=2}^N (z_t - \rho z_{t-1})^2\right)$$

(ii) Long term Average μ - Gaussian prior mean ξ and variance τ^2

$$p(\mu | \sigma^2, y) \propto p(y | \mu, \sigma^2) \cdot \pi(\mu)$$

$$p(y_i | y_{i-1}, \mu, \sigma^2, \rho) \propto \prod_{i=1}^N \exp\left[-\frac{(y_i - \mu)^2}{2\sigma^2}\right] \cdot \pi(\mu)$$

$$\propto \exp\left[-\frac{1}{2\sigma^2} \sum_{i=1}^N (y_i - \mu)^2\right] \cdot \pi(\mu)$$

$$\propto \exp\left[-\frac{1}{2\sigma^2} \left(n\mu^2 - 2\sum_{i=1}^N y_i \mu + \sum_{i=1}^N y_i^2\right)\right]$$

$$\propto \exp\left[-\frac{1}{2\sigma^2} \left(\mu - \sum_{i=1}^N y_i / n\right)^2 + A\right] \quad \rightarrow \bar{y} = \frac{\sum_{i=1}^N y_i}{n}$$

$$p(\mu | \sigma^2, \rho, y) \sim N(\mu | \bar{y}, \frac{\sigma^2}{n})$$

$$f(\mu | \sigma^2, \rho, y) \propto \prod_{t=2}^N f(y_t | \mu(1-\rho) + \rho y_{t-1}, \sigma^2) \cdot \pi(\mu) \propto \prod_{t=2}^N f\left(\frac{y_t - \rho y_{t-1}}{1-\rho} | \mu \left(\frac{\sigma}{1-\rho}\right)^2\right) \cdot \pi(\mu)$$

$$y_T^* = \frac{y_T - \rho y_{T-1}}{1-\rho} \quad \bar{y}^* = \frac{\sum_{t=2}^N y_t^*}{n-1} \quad \sigma^* = \frac{\sigma}{1-\rho}$$

$$f(\mu | \sigma^2, \rho, y) \propto N(\bar{y}^* | \mu, \frac{\sigma^{*2}}{n-1}) \cdot N(\xi, \tau^2)$$

$$\Rightarrow \sim N\left[\frac{\frac{\bar{y}^*}{\frac{\sigma^{*2}}{n-1}} + \frac{\xi}{\tau^2}}{\frac{1}{\sigma^{*2}} + \frac{1}{\tau^2}}, \left(\frac{n-1}{\sigma^{*2}} + \frac{1}{\tau^2}\right)^{-1}\right]$$