

Conditional Approach Gibbs Sampler

$$\begin{aligned} \underline{Z} &= \underline{D} \underline{Y} + \underline{\varepsilon} & \underline{\varepsilon} &\sim N(0, \Sigma_{\varepsilon}(\rho_{\varepsilon})) \\ \underline{Y} &\sim N(0, \underline{\Sigma}) & \underline{\Sigma} &:= \underline{\Sigma}(A, \sigma_{\varepsilon}^2, \sigma_{\eta}^2, K) \\ & & &= \underline{\Sigma}(B(A, r, D, \sigma_{\varepsilon}^2), \\ & & &\quad \underline{\Sigma}_{\eta}(\sigma_{\eta}^2, K_{\eta}), \\ & & &\quad \underline{\Sigma}_2(\sigma_{\varepsilon}^2, K_{\varepsilon})) \end{aligned}$$

$$\begin{aligned} \therefore p(\underline{Y} | \underline{Z}, \underline{\theta}) &\propto p(\underline{Z} | \underline{Y}, \underline{\theta}) p(\underline{Y} | \underline{\theta}) \\ &\propto \exp\left(-\frac{1}{2} (\underline{Z} - \underline{D} \underline{Y})^T \underline{\Sigma}_{\varepsilon}^{-1}(\underline{\theta}) (\underline{Z} - \underline{D} \underline{Y}) \right. \\ &\quad \left. - \frac{1}{2} \underline{Y}^T \underline{\Sigma}(\underline{\theta})^{-1} \underline{Y}\right) \\ &\propto \exp\left(-\frac{1}{2} \underline{Y}^T (\underline{\Sigma}(\underline{\theta})^{-1} + \underline{D}^T \underline{\Sigma}_{\varepsilon}^{-1}(\underline{\theta}) \underline{D}) \underline{Y} \right. \\ &\quad \left. + \underline{Y}^T (\underline{D}^T \underline{\Sigma}_{\varepsilon}^{-1}(\underline{\theta}) \underline{Z})\right) \\ \underline{\Sigma}_Y^* &:= (\underline{\Sigma}(\underline{\theta})^{-1} + \underline{D}^T \underline{\Sigma}_{\varepsilon}^{-1}(\underline{\theta}) \underline{D})^{-1} \\ \underline{\mu}_Y^* &:= \underline{\Sigma}_Y^* \underline{D}^T \underline{\Sigma}_{\varepsilon}^{-1}(\underline{\theta}) \underline{Z} \end{aligned}$$

$$\begin{aligned} p(\sigma_{\varepsilon}^{-2} | \underline{Y}, \underline{Z}, \underline{\theta}) &\propto p(\underline{Z} | \underline{Y}, \sigma_{\varepsilon}^{-2}, \underline{\theta}) p(\sigma_{\varepsilon}^{-2}) \\ \underline{\theta}^- &:= \underline{\theta} / \sigma_{\varepsilon}^2 \\ &\propto \exp\left(-\frac{\sigma_{\varepsilon}^{-2}}{2} (\underline{Z} - \underline{D} \underline{Y})^T \underline{\Sigma}_{\varepsilon}^{-1}(\underline{\theta}^-) (\underline{Z} - \underline{D} \underline{Y}) \right. \\ &\quad \left. + (\alpha_{\varepsilon} - 1) \ln \sigma_{\varepsilon}^{-2} - \beta_{\varepsilon} \sigma_{\varepsilon}^{-2}\right) \\ &\quad \exp\left(+\frac{n}{2} \ln \sigma_{\varepsilon}^{-2}\right) \end{aligned}$$

$$\Rightarrow \alpha^* = \alpha_{\varepsilon} - 1 + \frac{n}{2}$$

$$\beta^* = \beta_{\varepsilon} + (\underline{Z} - \underline{D} \underline{Y})^T \underline{\Sigma}_{\varepsilon}^{-1}(\underline{\theta}^-) (\underline{Z} - \underline{D} \underline{Y})$$

$$p(\tilde{\rho}_\varepsilon | \underline{y}, \underline{z}, \underline{\theta}) \propto p(\underline{z} | \underline{y}, \underline{\theta}) p(\tilde{\rho}_\varepsilon)$$

$$\propto \exp\left(-\frac{1}{2}(\underline{z} - \underline{D}\underline{y})^T \underline{\Sigma}_\varepsilon(\underline{\theta})^{-1}(\underline{z} - \underline{D}\underline{y})\right) \\ \exp\left(-\frac{1}{2} \ln |\underline{\Sigma}_\varepsilon(\underline{\theta})|\right)$$

$$\exp\left(-\frac{1}{2\sigma_{\rho_\varepsilon}^2}(\tilde{\rho}_\varepsilon - \mu_{\rho_\varepsilon})^2\right) \quad \rho_\varepsilon = 2\Phi(\tilde{\rho}_\varepsilon) - 1$$

$$p(\sigma_{11}^2 | \underline{y}, \underline{z}, \underline{\theta}^-) \propto p(\underline{z} | \underline{y}, \underline{\theta}^-) p(\underline{y} | \underline{\theta}^-) p(\underline{\theta}^-) \\ \underline{\theta}^- = \underline{\theta} \setminus \sigma_{11}^2$$

$$\propto p(\underline{y}_2 | \underline{y}_1, \underline{\theta}^-) p(\underline{y}_1 | \underline{\theta}^-) p(\underline{\theta}^-)$$

$$\propto \exp\left(-\frac{\sigma_{11}^{-2}}{2} \underline{y}_1^T \underline{\Sigma}_{11}(\underline{\theta}^-)^{-1} \underline{y}_1 \right. \\ \left. + \frac{n_1}{2} \ln \sigma_{11}^{-2} + (\alpha_{11} - 1) \ln \sigma_{11}^{-2} - \beta_{11} \sigma_{11}^{-2}\right)$$

$$\Rightarrow \alpha_{11}^* = \alpha_{11} - 1 + n_1/2$$

$$\beta_{11}^* = \beta_{11} + \frac{1}{2} \underline{y}_1^T \underline{\Sigma}_{11}(\underline{\theta}^-)^{-1} \underline{y}_1$$

$$p(\sigma_{211}^2 | \underline{y}, \underline{z}, \underline{\theta}^-) \propto p(\underline{y}_2 | \underline{y}_{11}, \underline{\theta}^-) p(\underline{\theta}^-)$$

$$\underline{\theta}^- = \underline{\theta} \setminus \sigma_{211}^2$$

$$\propto \exp\left(-\frac{\sigma_{211}^{-2}}{2} (\underline{y}_2 - \underline{B}\underline{y}_1)^T \underline{\Sigma}_{211}(\underline{\theta}^-) (\underline{y}_2 - \underline{B}\underline{y}_1) \right. \\ \left. + \frac{n_2}{2} \ln \sigma_{211}^{-2} + (\alpha_{211} - 1) \ln \sigma_{211}^{-2} - \beta_{211} \sigma_{211}^{-2}\right)$$

$$\Rightarrow \alpha_{211}^* = \alpha_{211} - 1 + n_2/2$$

$$\beta_{211}^* = \beta_{211} + \frac{1}{2} (\underline{y}_2 - \underline{B}\underline{y}_1)^T \underline{\Sigma}_{211}(\underline{\theta}^-) (\underline{y}_2 - \underline{B}\underline{y}_1)$$

$$p(A | \underline{y}, \underline{z}, \underline{\theta}^-) \propto p(\underline{z} | \underline{y}, \underline{\theta}^-) p(\underline{y}_1 | \underline{y}, \underline{\theta}^-) p(A)$$

$$\underline{\theta}^- \equiv \underline{\theta}^{1A}$$

$$\propto \exp\left(-\frac{1}{2} (\underline{y}_2 - A \underline{B}(\underline{\theta}) \underline{y}_1)^T \underline{\Sigma}_{21}^{-1}(\underline{\theta}) (\underline{y}_2 - A \underline{B}(\underline{\theta}) \underline{y}_1) - \frac{\sigma_A^2}{2} (A - \mu_A)^2\right)$$

$$\propto \exp\left(-\frac{1}{2} A^2 \left(\sigma_A^2 + \underline{y}_1^T \underline{B}^{-T} \underline{\Sigma}_{21}^{-1}(\underline{\theta}) \underline{B}^{-1} \underline{y}_1\right) + A \left(\underline{y}_1^T \underline{B}^{-T} \underline{\Sigma}_{21}^{-1}(\underline{\theta}) \underline{y}_2 + \sigma_A^{-2} \mu_A\right)\right)$$

$$\Rightarrow \sigma_A^{2*} = \sigma_A^2 + \underline{y}_1^T \underline{B}^{-T} \underline{\Sigma}_{21}^{-1}(\underline{\theta}) \underline{B}^{-1} \underline{y}_1$$

$$\mu_A^* = \sigma_A^{2*} \left(\underline{y}_1^T \underline{B}^{-T} \underline{\Sigma}_{21}^{-1}(\underline{\theta}) \underline{y}_2 + \sigma_A^{-2} \mu_A \right)$$

$$p(\Delta, r | \underline{y}, \underline{z}, \underline{\theta}^-) \propto \exp\left(-\frac{1}{2} (\underline{y}_2 - A \underline{B}(\underline{\theta}))^T \underline{\Sigma}_{21}^{-1}(\underline{\theta}) (\underline{y}_2 - A \underline{B}(\underline{\theta}))\right)$$

$$\underline{\theta}^- \equiv \underline{\theta}^{1A, \tilde{r}}$$

$$-\frac{1}{2\sigma_r^2} (\tilde{r} - \mu_r) - \frac{1}{2D_1} (\Delta_1 - \mu_{\Delta_1})$$

$$-\frac{1}{2D_2} (\Delta_2 - \mu_{\Delta_2}) \quad ; r = e^{\tilde{r}}$$

$$p(K_{21} | \underline{y}, \underline{z}, \underline{\theta}^-) \propto \exp\left(-\frac{1}{2} (\underline{y}_2 - A \underline{B})^T \underline{\Sigma}_{21}^{-1}(\underline{\theta}) (\underline{y}_2 - A \underline{B}) - \frac{1}{2\sigma_{K_{21}}^2} (\tilde{K}_{21} - \mu_{K_{21}})^2 \quad ; K_{21} = e^{\tilde{K}_{21}}\right)$$

$$\underline{\theta}^- \equiv \underline{\theta}^{1A, \tilde{K}_{21}}$$

$$p(K_n | \underline{y}, \underline{z}, \underline{\theta}^-) \propto \exp\left(-\frac{1}{2} \underline{y}_1^T \underline{\Sigma}_{11}^{-1}(\underline{\theta}) \underline{y}_1 - \frac{1}{2\sigma_{K_n}^2} (\tilde{K}_n - \mu_{K_n})^2\right)$$

$$\underline{\theta}^- \equiv \underline{\theta}^{1A, \tilde{K}_n}$$

$$K_n = e^{\tilde{K}_n}$$

