

Task 2.

$$z_t = A z_{t-1} + \varepsilon$$

matrix

$$\varepsilon \sim \mathcal{N}(0, \Sigma)$$

$$x_t = B z_t + \eta$$

matrix

$$\eta \sim \mathcal{N}(0, \Gamma)$$

$$x_t \sim \mathcal{N}(B z_t, \Gamma), \quad z_t \sim \mathcal{N}(A z_{t-1}, \Sigma)$$

from lecture:

Expectation-Maximization (EM) algorithm

provide initial estimate. $\theta^* = \theta^{(0)}$

E-step: $q^* = \underset{\text{given } \theta^*}{\operatorname{argmax}} \operatorname{ELBO}[q, \theta^*]$

$$= \underset{q}{\operatorname{argmin}} \operatorname{KL}[q \parallel p_{\theta^*}(z|x)]$$

approximate posterior

M-step: $\theta^* = \underset{\theta}{\operatorname{argmax}} \operatorname{ELBO}[q^*, \theta]$

$$= \underset{\theta}{\operatorname{argmax}} E_{q^*} [\log p_{\theta}(x, z)]$$

until $|\Delta \operatorname{ELBO}| < \varepsilon$

M-step: $\max E_q [\log p_\theta(x, z)]$ w.r.t $\{B, \Gamma, A, \Sigma, \mu_0\}$

$$p_\theta(x, z) = p(x|z) p(z) \quad \text{obs. model} \quad \text{latent model}$$

$$= p(x_1|z_1) p(z_1) \prod_{t=1}^T p(x_t|z_t) p(z_t|z_{t-1})$$

$$\Rightarrow E_q [\log p_\theta(x, z)] = E_q [\log p(z_1)]$$

$$+ E_q \left[\sum_{t=2}^T \log p(z_t|z_{t-1}) \right]$$

$$+ E_q \left[\sum_{t=1}^T \log p(x_t|z_t) \right]$$

gaussian
in this case.

M-step \rightarrow Expr. $\theta^* = \arg \max_{\theta} ELBO[q^*, \theta]$, w.r.t $\{B, \Sigma, \Gamma\}$

$$= E_q \left[-\frac{M}{2} \log(2\pi) - \frac{1}{2} \log |\Sigma| - \frac{1}{2} (z_1 - \mu_0)^T \Sigma^{-1} (z_1 - \mu_0) \right]$$

$$+ E_q \left[-\frac{M(T-1)}{2} \log(2\pi) - \frac{T-1}{2} \log |\Sigma| - \frac{1}{2} \sum_{t=2}^T (z_t - A z_{t-1})^T \Sigma^{-1} (z_t - A z_{t-1}) \right]$$

$$+ E_q \left[-\frac{NT}{2} \log(2\pi) - \frac{T}{2} \log |\Gamma| - \frac{1}{2} \sum_{t=1}^T (x_t - B z_t)^T \Gamma^{-1} (x_t - B z_t) \right]$$

only last part dep on B

$$\frac{\partial}{\partial B} E_q \left[-\frac{NT}{2} \log(2\pi) - \frac{T}{2} \log |\Gamma| - \frac{1}{2} \sum_{t=1}^T (x_t - B z_t)^T \Gamma^{-1} (x_t - B z_t) \right]$$

$$\frac{\partial}{\partial B} E_q \left[-\frac{1}{2} \sum_{t=1}^T \left(x_t^T \Gamma^{-1} x_t - x_t^T \Gamma^{-1} B z_t - z_t^T B^T \Gamma^{-1} x_t + z_t^T B^T \Gamma^{-1} B z_t \right) \right]$$

$$= E_q \left[\frac{\partial}{\partial B} \left(-\frac{1}{2} \sum_{t=1}^T \left(x_t^T \cancel{\Gamma^{-1} x_t} - x_t^T \Gamma^{-1} B z_t - z_t^T B^T \Gamma^{-1} x_t + z_t^T B^T \Gamma^{-1} B z_t \right) \right) \right]$$

$$= E_q \left[-\frac{1}{2} \sum_{t=1}^T \left(-\Gamma^{-1} x_t z_t^T - \Gamma^{-1} z_t x_t^T + (\Gamma^{-1})^T B z_t z_t^T + \Gamma^{-1} B z_t z_t^T \right) \right]$$

$$\Gamma \text{ sym.} \Rightarrow \Gamma^{-1} \text{ symmetric} \quad \Gamma^{-1} = (\Gamma^{-1})^T$$

$$\Rightarrow = E_q \left[\Gamma^{-1} x_t z_t^T - \Gamma^{-1} B z_t z_t^T \right] \stackrel{!}{=} 0$$

using linearity of exp val and B indep of q can be pulled out

$$\Rightarrow E_q \left[\sum_{t=1}^T \Gamma^{-1} x_t z_t^T \right] - B E_q \left[\sum_{t=1}^T \Gamma^{-1} z_t z_t^T \right] \stackrel{!}{=} 0$$

$$B = \underbrace{\sum_{t=1}^T x_t E_q [z_t^T]}_{\text{}} \left(\sum_{t=1}^T E_q [z_t z_t^T] \right)^{-1}$$

two first part dep on Σ

$$= \frac{\partial}{\partial \Sigma} \left(E_q \left[-\frac{M}{2} \log(2\pi) - \frac{1}{2} \log |\Sigma| - \frac{1}{2} (z_1 - \mu_0)^T \Sigma^{-1} (z_1 - \mu_0) \right] \right. \\ \left. + E_q \left[-\frac{M(T-1)}{2} \log(2\pi) - \frac{T-1}{2} \log |\Sigma| - \frac{1}{2} \sum_{t=2}^T (z_t - A z_{t-1})^T \Sigma^{-1} (z_t - A z_{t-1}) \right] \right)$$

cookbook

$$-\frac{1}{2} \frac{\partial}{\partial \Sigma} \log |\Sigma| = -\frac{1}{2} \cancel{(\Sigma^{-1})^T} \cdot \cancel{(\Sigma^{-1})} (\Sigma^{-1})^T = (\Sigma^{-1})^T \xrightarrow{\text{sym}} \frac{1}{2} \Sigma^{-1}$$

$$-\frac{T-1}{2} \frac{\partial}{\partial \Sigma} \log |\Sigma| = -\frac{T-1}{2} \Sigma^{-1} \quad \Sigma \text{ is sym}$$

$$\frac{\partial}{\partial \Sigma} \left(-\frac{1}{2} \left(z_1^T \Sigma^{-1} z_1 - z_1^T \Sigma^{-1} \mu_0 - \mu_0^T \Sigma^{-1} z_1 + \mu_0^T \Sigma^{-1} \mu_0 \right) \right)$$

$$= -\frac{1}{2} \left(-\Sigma^{-1} z_1 z_1^T \Sigma^{-1} + \Sigma^{-1} z_1 \mu_0^T \Sigma^{-1} + \Sigma^{-1} \mu_0 z_1^T \Sigma^{-1} - \Sigma^{-1} \mu_0 \mu_0^T \Sigma^{-1} \right)$$

$$\frac{\partial}{\partial \Sigma} \left(-\frac{1}{2} (z_t - A z_{t-1})^T \Sigma^{-1} (z_t - A z_{t-1}) \right) = -\frac{1}{2} \frac{\partial}{\partial \Sigma} \left(z_t^T \Sigma^{-1} z_t - z_t^T \Sigma^{-1} A z_{t-1} - z_{t-1}^T A^T \Sigma^{-1} z_t \right. \\ \left. + z_{t-1}^T A^T \Sigma^{-1} A z_{t-1} \right)$$

$$= -\frac{1}{2} \left(-\Sigma^{-1} z_t z_t^T \Sigma^{-1} + \Sigma^{-1} z_t z_{t-1}^T A^T \Sigma^{-1} + \Sigma^{-1} A z_{t-1} z_t^T \Sigma^{-1} - \Sigma^{-1} A z_{t-1} z_{t-1}^T A^T \Sigma^{-1} \right)$$

A'

$$E_q \left[-\frac{1}{2} \Sigma^{-1} - \frac{T-1}{2} \Sigma^{-1} - \frac{1}{2} \Sigma^{-1} \left(z_1 \mu_0^T + \mu_0 z_1^T - z_1 z_1^T - \mu_0 \mu_0^T \right) \Sigma^{-1} \right. \\ \left. - \frac{1}{2} \Sigma^{-1} \sum_{t=2}^T \left(z_t z_t^T A^T + A z_{t-1} z_t^T - z_t z_t^T - A z_{t-1} z_{t-1}^T A^T \right) \Sigma^{-1} \right]$$

B'

= 0 / .2

$$\Sigma^{-1}(-1 - T + 1) = \Sigma^{-1}A'\Sigma^{-1} - \Sigma^{-1}B'\Sigma^{-1} \stackrel{!}{=} 0 \quad | \cdot \Sigma$$

$$-T = \Sigma^{-1}A' + \Sigma^{-1}B' \quad | \cdot \Sigma$$

$$-\Sigma T = A' + B'$$

$$\Sigma = -\frac{1}{T} (A' + B')$$

$$\Rightarrow \Sigma = -\frac{1}{T} \left\{ -E_q(z, \mu_0^T) - E_q[\mu_0 z^T] + E_q[z_1 z_1^T] + E[\mu_0 \mu_0^T] \right. \\ \left. + \sum_{t=2}^T \left(-E_q[z_t z_{t-1}^T] A^T - A E_q[z_{t-1} z_t^T] + E_q[z_t z_t^T] + A E_q[z_{t-1} z_{t-1}^T] A^T \right) \right\}$$

only last part dep on Γ .

$$\frac{\partial}{\partial \Gamma} E_q \left[-\frac{NT}{2} \log(2\pi) - \frac{T}{2} \log |\Gamma| - \frac{1}{2} \sum_{t=1}^T (x_t - B z_t)^T \Gamma^{-1} (x_t - B z_t) \right]$$

$$-\frac{T}{2} \frac{\partial}{\partial \Gamma} \log |\Gamma| = -\frac{T}{2} \Gamma^{-1}$$

$$-\frac{1}{2} \sum_{t=1}^T (x_t - B z_t)^T \Gamma^{-1} (x_t - B z_t) = -\frac{1}{2} \text{tr} \left(\underbrace{\Gamma^{-1} (x_t - B z_t) (x_t - B z_t)^T}_{A'} \right)$$

$$\Rightarrow -\frac{1}{2} \frac{\partial}{\partial \Gamma} \text{tr} (\Gamma^{-1} A') = -\frac{1}{2} (-\Gamma^{-1} A' \Gamma^{-1})^T A'$$

$$\Gamma_{\text{sym}} \Rightarrow \frac{1}{2} \Gamma^{-1} A^T \Gamma^{-1}$$

$$\Rightarrow \frac{1}{2} \Gamma^{-1} (x_t - B z_t) (x_t - B z_t)^T \Gamma^{-1}$$

$$\frac{1}{2} \Gamma^{-1} (x_t x_t^T - x_t z_t^T B^T - B z_t x_t^T + B z_t z_t^T B^T) \Gamma^{-1}$$

$$\Rightarrow E_q \left[-\frac{1}{2} \Gamma^{-1} + \underbrace{\frac{1}{2} \sum_{t=1}^T \Gamma^{-1} (x_t x_t^T - x_t z_t^T B^T - B z_t x_t^T + B z_t z_t^T B^T) \Gamma^{-1}}_Y \right] \stackrel{!}{=} 0 \quad 1-2$$

$$\Rightarrow T \Gamma^{-1} = \sum_{t=1}^T \Gamma^{-1} E_q(Y) \Gamma^{-1} \quad | \cdot \Gamma$$

$$T = \sum_{t=1}^T \Gamma^{-1} E_q(Y) \quad | \cdot \Gamma$$

$$\Gamma = \frac{1}{T} \sum_{t=1}^T E_q(Y)$$

$$\Gamma = \frac{1}{T} \sum_{t=1}^T \left\{ x_t x_t^T - x_t E[z_t^T] B^T - B E_q[z_t] x_t^T + B E_q[z_t z_t^T] B^T \right\}$$
