Bayesian Learning and Inference in Recurrent Switching Linear Dynamical Systems

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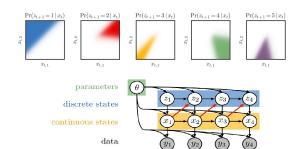
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Model

SLDS and rSLDS

- Observation $y_t = Cx_t + d + w_t$, $w_t \stackrel{\text{iid}}{\sim} \mathcal{N}(0, S)$
- Continous latent state $x_{t+1} = A_{z_{t+1}} x_t + b_{z_{t+1}} + \nu_t, \ \nu_t \overset{\mathrm{iid}}{\sim} \mathcal{N}(0, Q_{z_{t+1}})$
- Discrete latent state $z_t \in \{1, \dots, K\}$
 - SLDS $z_{t+1}|z_t \sim \pi_{z_t}$
 - rSLDS $z_{t+1}|z_t, x_t \sim \pi_{SB}(\nu_{t+1}), \ \nu_{t+1} = R_{z_t}x_t + r_{z_t}$





Stick Breaking Logitstic Regression

- $p(z|x) \sim \pi_{SB}(\nu), \ \nu = Rx + r$
- Link function: $\pi_{SB}(\nu) = \left(\pi_{SB}^{(1)}(\nu), \dots, \pi_{SB}^{(K)}(\nu)\right)$ (with $\sigma(x) = \frac{e^x}{1+e^x}$)

$$\pi_{SB}^{(k)}(\nu) = \begin{cases} \sigma(\nu_k) \prod_{j < k} \sigma(-\nu_j), & \text{if } k = 1, \dots, K - 1 \\ \prod_{1}^{K} \sigma(-\nu_j), & \text{if } k = K \end{cases}$$

- $p(z|x) \sim \prod_{k=1}^K \sigma(\nu_k)^{\mathbb{I}[z=k]} \sigma(-\nu_k)^{\mathbb{I}[z>k]}$ (Likelihood)
- With a Gaussian Prior p(x), the posterior $p(x|z) \propto p(x) * p(z|x)$ is non-Gaussian
- Bayesian updating is not efficient
 - Gibbs Sampler: sampling $x^{(i+1)} \sim p(x|z=z^{(i)})$, sampling $z^{(i+1)} \sim p(z|x=x^{(i+1)})$
 - Message Passing $m_{t\to(t+1)}(x_{t+1}) = \int \psi(x_t, y_t) \psi(x_t, z_{t+1}) \psi(x_t, x_{t+1}, z_{t+1}) m_{(t-1)\to t}(x_t) dx_t$



Polya-gamma augmentation

$$\frac{(e^{\nu})^{a}}{(1+e^{\nu})^{b}} = 2^{-b} \int_{0}^{\infty} e^{\kappa \nu} e^{-\omega \nu^{2}/2} p_{PG} \left(\omega \mid b, 0\right) d\omega \left(\kappa = a - \frac{b}{2}\right)$$
(2.1)

$$p(x_t|z_{t+1}) \propto \prod_{k=1}^{K-1} \frac{\left(e^{\nu_{t+1,k}}\right)^{\mathbb{I}[z_{t+1}=k]}}{\left(1+e^{\nu_{t+1,k}}\right)^{\mathbb{I}[z_{t+1}\geq k]}}$$
(2.2)

$$p(x_t|z_{t+1}) = \int p(x_t, \omega_t|z_{t+1}) d\omega_t = \int p(x_t|z_{t+1}, \omega_t) \frac{p(\omega_t)}{\rho(\omega_t)} d\omega_t$$
 (2.3)

$$\omega_{t,k} \mid x_t, z_{t+1} \sim \mathsf{PG}\left(\mathbb{I}\left[z_{t+1} \ge k\right], \nu_{t+1,k}\right)$$
 (2.4)

- $p(x_t|z_{t+1},\omega_t)$ is Gaussian
- Message Passing $\psi(x_t, z_{t+1}, \omega_t) \propto \mathcal{N}(\nu_{t+1} | \Omega_t^{-1} \kappa_{t+1}, \Omega_t^{-1})$
- Thus instantiating these auxiliary variables in a Gibbs sampler enables efficient block updates



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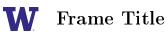
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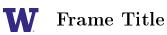
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