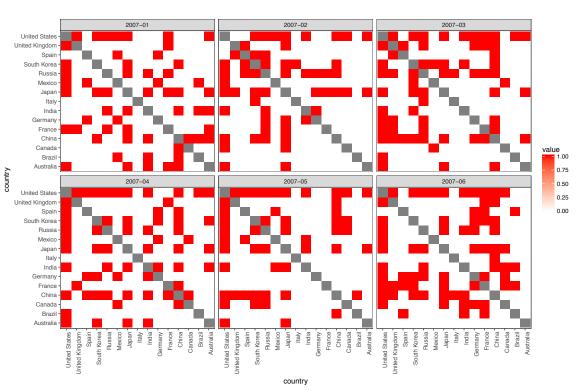
## Modeling dynamic international relationship



- Challenges:
  - Binary data with 21.7% non-zero entries
  - Each  $y_{ij,t}$  is a time series
  - Network structure
- Goal: Capturing the pattern

Latent space model

$$y_{ij,t}|\pi_{ij,t} \sim Bernoulli\{\pi_{ij,t}\}$$

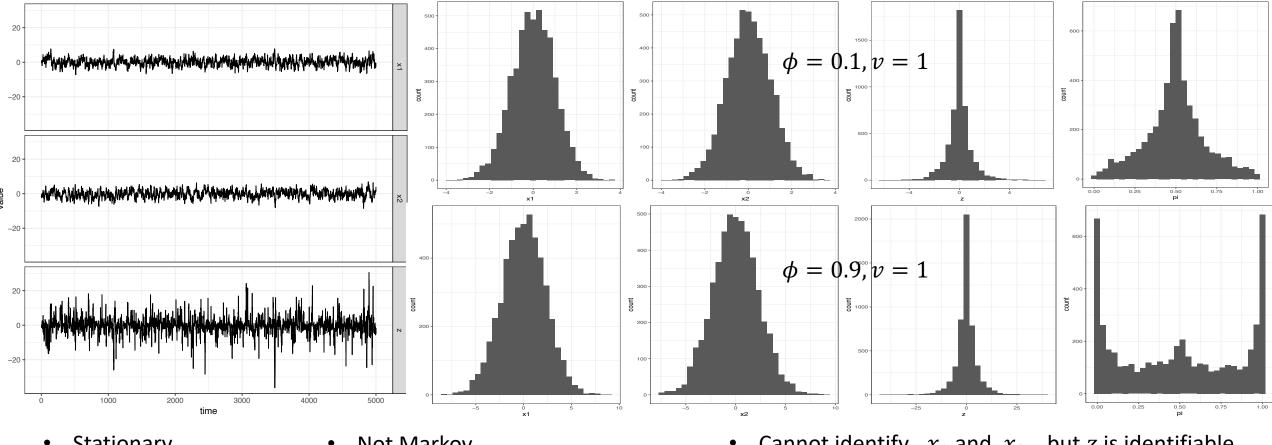
$$\pi_{ij,t} = \{1 + e^{-S_{ij,t}}\}^{-1}$$

$$S_{ij,t} = \mu_t + x_{i,t}^T x_{j,t}$$

- Latent factor  $x_{i,t} = \{x_{i1,t}, ..., x_{iH,t}\}^T$
- Stationary AR( $\overline{1}$ ) process  $\mu_t$   $x_{ih,t}$
- product of two AR(1) process

## Properties of the product of two AR(1) process

 $z_t = x_{1t} x_{2t}$ 



Stationary

Not Markov

 $x_{2,t-1} \longrightarrow x_{2,t} \longrightarrow x_{2,t+1} \longrightarrow x_{2,t+2} \longrightarrow$  $z_{t+1}$  $x_{1,t-1} \longrightarrow x_{1,t} \longrightarrow x_{1,t+1} \longrightarrow x_{1,t+2} \longrightarrow$ 

Cannot identify  $x_1$  and  $x_2$  , but z is identifiable

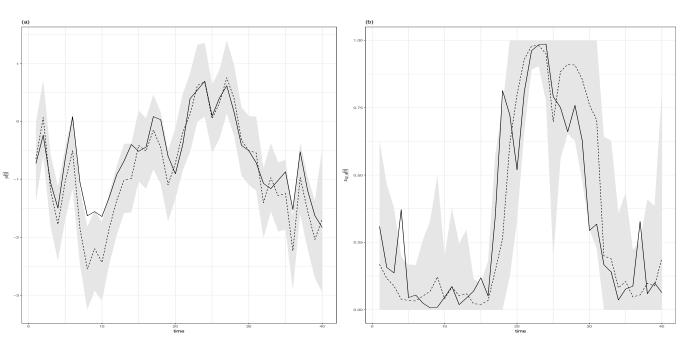
## Sampling scheme and applications

Gibbs Sampling with Polya-Gamma data augmentation

$$logit\pi_{ij} = \mu + x_i^T x_j$$

From the joint distribution of AR(1) process

$$\mu \sim MVN(0, s\Phi_{\mu}), x_{ih} \sim MVN(0, s\Phi_{x})$$



• Choice of  $H, \phi_{\mu}, v_{\mu}, \phi_{x}, v_{x}$ 

*H*: large for better model fitting; add shrinkage prior

s: reasonable small range

 $\phi$ : relative small for model fitting

