Toolbox 'BSBM'

This manual aims to show detailed guidelines on using our toolbox, including the usages of primary functions, our settings of initial values, and a control panel for reproducing our simulation results.

BSBM_ContPnl

An example of toolbox 'BSBM'

Description:

The Control Panel of Our Proposed Bayesian Method with a SBM-motivated Prior (BSBM)

This control panel serves as an example of analyzing data using our toolbox. It is written as a function which can be easily modified to use on computing clusters. Users can use this control panel to reproduce the results of our simulation study (Section 3 in our paper) and modify this file to analyze their data without difficulty.

Details:

This control panel shows a step-by-step procedure to analyze the simulated data we used in Section 3.1. The two simulated datasets we used in Section 3.2 can be analyzed by only changing the file names in the control panel. All the simulated data are provided under the subfolder Simulated_data:

- Simulation1.mat: The simulated data we used in Section 3.1;
- Simulation 2_1.mat: The simulated data we used in Section 3.2, with 2714 time points;
- Simulation2_2.mat: The simulated data we used in Section 3.2, with 1000 time points.

Each data contains two variables:

- Y: The simulated data, which is a real d×T matrix. d is the number of channels and T is the number of time points. For example, in Simulation1.mat, Y is a 50×1000 matrix;
- A_true: The true network structure, which is a binary d×d matrix. "A_true(i,j) = 1" indicates that there is a directional connection from channel j to channel i. For example, in Simulation1.mat, A_true is a 50×50 matrix.

The systems and the error structures we used to generate these simulated datasets are described in Section 3.1 and Section 3.2 of our paper in detail.

The control panel also includes the calculation of the clustering probabilities and the network edge probabilities. The analyses based on the clustering probabilities and the network edge probabilities, including generating figures, are described in Section 2.4, Section 3, and Section 4 of our paper in detail.

BSBM_EM

EM Algorithm for Setting Starting Values and Hyperparameter of the Partially Collapsed Gibbs Sampler

Description:

This is the primary function of the EM algorithm, which is used to set starting values and hyperparameter for the MCMC algorithm.

Usage:

```
[Gam_all, A_all, C_all, R_all, mu_all, m_all, B_all]
```

= BSBM_EM(Y, mite, 10, u0, e0, xi_sq, xi_sq2, A0, C0, mu0, R0, m0, B0, Gam0, b_sq)

Input:

Y The observed data \mathbf{Y} , a real $d \times T$ matrix,

where d is the number of channels and T is

the number of time points

mite The max number of iterations

10,u0 Hyperparamters in **B**'s prior

e0 Hyperparameter in R's prior

xi_sq Hyperparameter in **A**'s prior

xi_sq2 Hyperparameter in C's prior and μ 's prior

A0,C0,mu0,R0,m0,B0,Gam0 The initial values of \mathbf{A} , \mathbf{C} , $\boldsymbol{\mu}$, \mathbf{R} , \mathbf{m} , \mathbf{B} , $\mathbf{\Gamma}$

b_sq Fixed at 1 in current version of BSBM

Output:

Gam_all,A_all,C_all,R_all,mu_all,m_all,B_all record the updated values of Γ ,A,C,R, μ ,m,B at each iteration.

Details:

The outputs of BSBM_EM contain the values of parameters at every iteration. We use the updated values at the final iteration to set the starting values and hyperparameter of the MCMC algorithm.

Other functions under the subfolder *EM* (BSBM_Kalman, BSBM_EM_update_A_Gam, BSBM_EM_update_B, BSBM_EM_update_C, BSBM_EM_update_m, BSBM_EM_update_mu, BSBM_EM_update_R) are the updating steps in the EM algorithm.

BSBM_MCMC MCMC Simulation Algorithm for Posterior Inference

Description:

This is the primary function of the MCMC algorithm, which is used for posterior inference of our proposed Bayesian hierarchical model.

Usage:

[A_all, C_all, X, V, Q, R_all, mu_all, P_all, m_all, B_all, Gam_all]
= BSBM_MCMC(Y, mcmc_mite, l0, u0, e0, xi_sq, xi_sq2, X0,
A0, C0, mu0, Q0,R0, V0, P0, m0, B0, Gam0)

Input:

Y The observed data Y, a real $d \times T$ matrix,

where d is the number of channels and T is

the number of time points

mcmc_mite The max number of iterations

l0,u0 Hyperparamters in **B**'s prior

e0 Hyperparameter in **R**'s prior

xi_sq Hyperparameter in **A**'s prior

xi_sq2 Hyperparameter in C's prior and μ 's prior

X0 The initial values of the underlying neuronal

states X

A0,C0,mu0,R0,P0,m0,B0,Gam0 The initial values of A,C,μ,R,P,m,B,Γ

Q0,V0 Fixed at the identity matrices in current ver-

sion of BSBM

Output:

Gam_all,A_all,C_all,R_all,mu_all,P_all,m_all,B_all record the simulated values of Γ ,A,C,R, μ ,P,m,B at each iteration. X record the simulated values of X at the last iteration. V,Q are fixed at the identity matrices in the current version of BSBM.

Details:

The outputs of BSBM_MCMC contain the simulated markov chains of all parameters. We use the simulated values of \mathbf{m} and $\mathbf{\Gamma}$ to calculate the clustering probabilities and the network edge probabilities, which we defined in Secion 2.4 in our paper.

Other functions under the subfolder *MCMC* (BSBM_MCMC_Update_A, BSBM_MCMC_update_B, BSBM_MCMC_update_C, BSBM_MCMC_update_Gam,

BSBM_MCMC_update_m, BSBM_MCMC_update_mu, BSBM_MCMC_update_P, BSBM_MCMC_update_Q, BSBM_MCMC_update_R, BSBM_MCMC_update_X) are the simulation steps in the MCMC simulation algorithm.

Reference:

Huazhang Li, Yaotian Wang, Guofen Yan, Yinge Sun, Seiji Tanabe, Chang-Chia Liu, Mark S. Quigg, Tingting Zhang (2020). A Bayesian State-Space Approach to Mapping Directional Brain Networks, *Journal of the American Statistical Association*. Under Review.