

Stochastic block models for a collection of networks

Applications in ecology and sociology

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Collaborators

Joint work with



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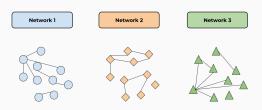


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Collection of networks : consensus in the structure

Objectives

Looking for commun patterns in networks involving non-common sets of nodes

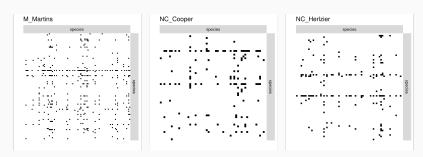


Applications

- Compare the structure of ecological networks
- Compare sociological networks : advices between lawyers, researchers or priests

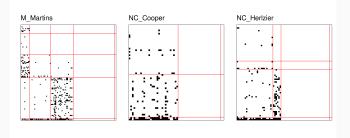
Three foodwebs

- Pine-firest stream food webs issued from Maine, North-Caroline and Nez-Zealand [Thompson and Townsend, 2003]
- Involve respectively 105, 58 and 71 species.
- $Y_{ij} = 1$ if i is eaten by j. Directed relation



Look for similarities and differences between network structures.

Separate SBMs

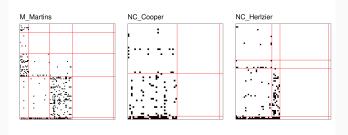


- Fitted SBM on each separately
- Reordered the matrices following the blocks
- Label the blocks following the average out-degrees order

Interpretation:

- In row : is eaten by...
- In col : eats...

Separate SBMs



- Two bottom groups in each matrix are basal species: eaten by many species and not eating anybody.
- Martins: has a separation into 5 blocks, the third one is a medium trophic level, which preys on basal species and is highly preyed by species of the 1st block.
 - Cooper. Higher trophic levels grouped together in the same block (lack of statistical power).
 - Herlzier: higher trophic level is separated into 2 blocks determined on how much they prey on the less preyed basal block.

Towards a joint modeling of the networks

- Need to model jointly the networks
- Identify the groups playing the same role through out the networks, with an unsupervised strategy.
- Let $(\mathbf{Y}^m)_{m=1,...,M}$ denote the collection of networks each involving n_m nodes.
- (Y^m) independent.

ì

$$\mathbf{Y}^m \sim \mathsf{SBM}_{n_m}(Q^m, \boldsymbol{\pi}^m, \boldsymbol{\alpha}^m)$$

• Conditions on the parameters $(\pi^m)_{m=1,...,M}$ and $(\alpha^m)_{m=1,...,M}$

First naive model

iid-colSBM

$$\mathbf{Y}^m \sim \mathsf{SBM}_{n_m}(Q, \pi, \alpha)$$

with $\pi_q > 0 \ \forall q \in \{1,\dots,Q\}$ and $\sum_{q=1}^Q \pi_q = 1$.

- $(Q-1)+Q^2$ unknown parameters, M clustering
- Too strict to be applied to the Thomson's dataset

A first relaxed model : π -colSBM

Same structure of connection α , specific proportions of blocks in each network

$\pi\text{-colSBM}$

$$\mathbf{Y}^m \sim \mathsf{SBM}_{n_m}(Q, \boldsymbol{\pi}^m, \boldsymbol{lpha})$$

On the block proportions

- $\pi_a^m \ge 0$
- If $\pi_q^m = 0$ then block q is not represented in network m

π -colSBM : different proportions

M=2 networks

$$\alpha = \begin{pmatrix} \alpha_{11} & \alpha_{12} & \alpha_{13} \\ \alpha_{12} & \alpha_{22} & \alpha_{23} \\ \alpha_{13} & \alpha_{23} & \alpha_{33} \end{pmatrix} \qquad \begin{array}{l} \pi^1 = [.25, .25, .50] \\ \pi^2 = [.20, .50, .30] \end{array}.$$

- Same connection structure between blocks
- Different block proportions
- $2 \times (3-1) + 3^2 = 15$ parameters.

π -colSBM : nested structures

$$\pi_q^m \geq 0$$

$$\alpha = \begin{pmatrix} \alpha_{11} & \alpha_{12} & \alpha_{13} \\ \alpha_{12} & \alpha_{22} & \alpha_{23} \\ \alpha_{13} & \alpha_{23} & \alpha_{33} \end{pmatrix} \qquad \pi^1 = [.25, .25, .50] \\ \pi^2 = [.40, 0, .60].$$

- Blocks 1 and 3 are represented in the two networks while block 2 only exists in network 1.
- $3-1+3-2+3^2=14$ parameters

π -colSBM : partially nested structures

$$\alpha = \begin{pmatrix} \alpha_{11} & \alpha_{12} & \alpha_{13} \\ \alpha_{21} & \alpha_{22} & \cdot \\ \alpha_{31} & \cdot & \alpha_{33} \end{pmatrix} \qquad \pi^{1} = [.25, .75, 0] \\ \pi^{2} = [.40, 0, .60].$$

- The two networks share block 1 (for instance super predators or basal species)
- The remaining nodes of each network not equivalent in terms of connectivity.
- Blocks 2 and 3 never interact because their elements do not belong to the same network and so α_{23} and α_{32} are not required to define the model.
- (2-1)+(2-1)+7=11 parameters.

Number of parameters

Let S be the support $M \times Q$ matrix such that

$$S_{mq} = egin{cases} 1 & ext{if } \pi_q^m > 0 \ 0 & ext{otherwise} \ . \end{cases}$$

Then,

$$Nb(\pi\text{-}colSBM) = \sum_{m=1}^{M} \left(\sum_{q=1}^{Q} S_{qm} - 1\right) + \sum_{q,r=1}^{Q} \mathbf{1}_{(S'S)_{qr} > 0}$$

Varying density model : δ -colSBM

δ -colSBM

$$\mathbf{Y}^m \sim \mathsf{SBM}_{n_m}(Q, \boldsymbol{\pi}, \delta^m \boldsymbol{\alpha})$$

with $\pi_q > 0$,

- M networks exhibit similar intra- and inter blocks connectivity patterns but with proper densities.
- δ^m be a density parameter, specific to each network. $\delta^1 = 1$.
- Mimics differences of effort sampling or abundances
- $(Q-1) + Q^2 + (M-1)$ parameters.

Varying density and block proportion model

$\delta\pi ext{-colSBM}$

$$\mathbf{Y}^m \sim \mathsf{SBM}_{n_m}(Q, \boldsymbol{\pi}^m, \delta^m \boldsymbol{\alpha})$$

with $\pi_a^m \geq 0$

- Most flexible model
- $Nb(\pi$ -colSBM) + (M-1) parameters.

Summary

M independent networks.

$$\mathbf{Y}^m \sim \mathsf{SBM}(\mathit{Q}^m, \pi^m, lpha^m)$$

Model name	Block prop.	Connexion param.	Nb of param.
iid-colSBM	$\pi_q^m = \pi_q, \ \pi_q > 0$	$\alpha_{qr}^m = \alpha_{qr}$	$(Q-1) + Q^2$
π -colSBM	$\pi_q^m, \pi_q^m \geq 0$	$\alpha_{qr}^m = \alpha_{qr}$	$\leq M(Q-1)+Q^2$
δ -colSBM	$\pi_q^m = \pi_q, \ \pi_q > 0$	$\alpha_{qr}^m = \delta^m \alpha_{qr}$	$(Q-1)+Q^2+(M-1)$
$\delta\pi$ -colSBM	$\pi_q^m, \pi_q^m \geq 0$	$\alpha_{qr}^m = \delta^m \alpha_{qr}$	$\leq M(Q-1)+Q^2+M-1$
sep-SBM	$\pi_q^m, \pi_q^m > 0$	α_{qr}^{m}	$\sum_{m=1}^{M} (Q_m - 1) + Q_m^2$

Identifiability

Demonstrated for the most complex SBM, upto label switching of the blocks and permutation of the networks, under light conditions.

For π -coISBM, let us define $\mathcal{Q}_m = \{q \in \{1, \dots, Q\} | \pi_q^m > 0\}$.

- 1. $\forall m : n_m \geq 2|Q_m|$
- 2. $(\alpha \cdot \pi^m)_q \neq (\alpha \cdot \pi^m)_r$ for all $(q \neq r) \in \mathcal{Q}_m^2$
- 3. $\forall q = 1, \ldots, Q, \quad \exists m : q \in \mathcal{Q}_m$
- 4. Each diagonal entry of α is unique

Inference

VEM algorithm

- \blacksquare Direct extension of VEM previously described for $\it iid\mbox{-}\rm colSBM$ and $\it \pi\mbox{-}\rm colSBM$
- Less obvious with $\delta_m \alpha$: M step not explicit.

Model selection

ICL can be directly extended for iid-coISBM and the δ -coISBM

$$ICL(Q) = \mathcal{I}(\hat{\tau}, \hat{\theta}) - \frac{Q-1}{2} \log \left(\sum_{m \in \mathcal{M}} n_m \right)$$
$$-\frac{1}{2} \left(\frac{Q(Q+1)}{2} + \nu(\delta) \right) \log \left(\sum_{m \in \mathcal{M}} \frac{n_m(n_m-1)}{2} \right), (1)$$

where $\nu(\delta) = M - 1$ for $\delta colSBM$ and 0 otherwise.

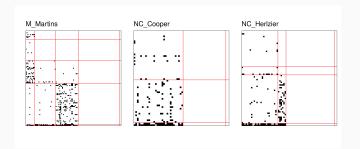
Model selection

- ullet For iid-colSBM and the δ -colSBM
- π_q^m possibly null. Asymptotic approximation do not hold
- Each couple (Q, S) defines a model.

$$ICL(Q,S) = \mathcal{I}(\hat{\tau}, \hat{\theta}) - \sum_{m=1}^{M} \frac{|Q_m| - 1}{2} \log(n_m) - \frac{1}{2} \left(\sum_{q,r=1}^{Q} \mathbf{1}_{(S'S)_{qr} > 0} + \nu(\delta) \right) \log \left(\sum_{m=1}^{M} \frac{n_m(n_m - 1)}{2} \right) (2)$$

Application on the foodwebs





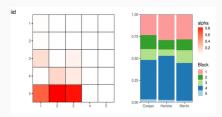
Separate sbm

Model choice

Model	ICL
sepSBM	-2080
iid-colSBM	-1966
$\pi ext{-coISBM}$	-1982
δ -colSBM	-1969
$\delta\pi ext{-coISBM}$	-1989

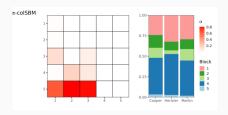
 $\,\blacksquare\,$ Reject sepSBM : commun structure in the networks

iid-colSBM: the prefered model



- Makes 5 blocks
- Block 3 (light green) is a small block of intermediate trophic level species with some within block predation.
- The higher trophic level is divided into 2 more blocks,
 - block 2 (dark green) only preys on the 2 basal blocks
 - block 1 (pink) preys on the intermediate block 3 level but only on the most connected basal species block.

π -colSBM



- Also 5 blocks.
- There are no empty blocks
- the block proportions are roughly corresponding to the ones of iid-colSBM .
- Flexibility of the π -colSBM of little use compared to the iid-colSBM on this collection.

Conclusion

- The three networks do share a commun structure.
- We can identify the species playing the same role across networks (ecosytems)
- Other results
 - Quality of prediction when missing data.
 - Application in sociology : advices between lawyers, researchers or priests
 - Clustering of networks. Application on a database of 80 networks.

Work in progress

- Develop a wide variety of models
- Very active research field in our group
- Various extensions in progress
 - Taking into account the incertitude of reconstruction of the networks (data from metagenomics)
 - Extension to large multilayer networks such as interactome
 - Looking for tools to compare networks: plant health submitted to combination of stress

References i



Thompson, R. M. and Townsend, C. R. (2003).

Impacts on stream food webs of native and exotic forest : An intercontinental comparison. Ecology, 84(1):145–161.



Vissault, S., Cazelles, K., Bergeron, G., Mercier, B., Violet, C., Gravel, D., and Poisot, T. (2020). rmangal: An R package to interact with Mangal database. R package version 2.0.2.

To go further: partitionning a collection of networks

- If the networks in a collection do not have the same connectivity structure, we aim to partition them accordingly.
- Finding a partition $\mathcal{G} = (\mathcal{M}_g)_{g=1,...,G}$ of $\{1,\ldots,M\}$. such that

$$\forall g \in \{1, \dots, G\}, \quad \forall m \in \mathcal{M}_g, \quad \mathbf{Y}^m \sim \mathsf{SBM}(K^g, \pi^m, \alpha^g)$$

networks belonging to the subcollection \mathcal{M}_g share the same mesoscale structure given by $\pi\text{-colSBM}$.

Scoring a partition

ullet To any partition ${\mathcal G}$ we associate the following score :

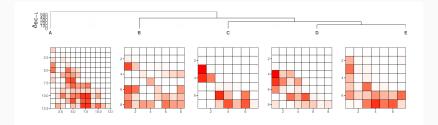
$$\mathsf{Sc}(\mathcal{G}) = \sum_{g=1}^{G} \mathsf{BIC\text{-}L}((\mathbf{Y}^m)_{m \in \mathcal{M}_g}, \widehat{K^g}).$$

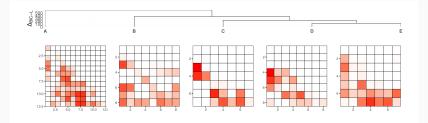
• Best partition $\mathcal G$ is chosen as follows :

$$\mathcal{G}^* = \operatorname*{arg\,max} \mathsf{Sc}(\mathcal{G}).$$

- 67 networks issued from the Mangal database belonging to 33 datasets. [Vissault et al., 2020]
- predation networks which are all directed networks with more than 30 species,
- number of species ranges from 31 to 106 (3395 in total) by network
- Density ranging from .01 to .32 (14934 total predation links).

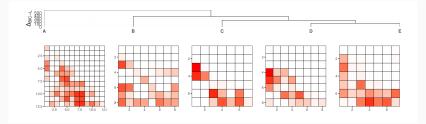
Aim use our model to propose partition of the networks into group of networks with common mesoscale structure.





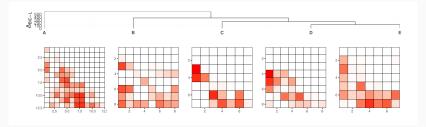
Groupe A

- 7 networks and 12 blocks are required to describe this group of networks
- 5 networks are issued from the same dataset (id : 80).
- These 5 networks populate the 12 blocks, while the other 2 networks only populate parts of them.
- Average density is about 0.18
- Blocks 1 to 3 represent the higher trophic levels, blocks 4 to 8 the intermediate ones and block 9 to 12 the lower ones.



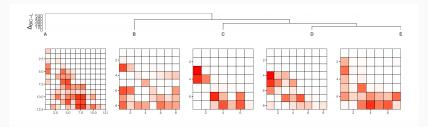
Group B: structure with 8 blocks

- 26 networks with heterogeneous size and density.
- Issued from various datasets
- Most networks populate only parts of the 8 blocks
- Block 4 is represented in only 5 networks where it is either an intermediate or a bottom trophic level.
- Species from top trophic levels prey on basal species.



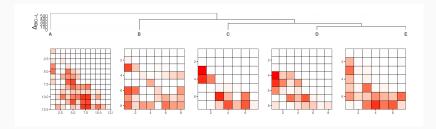
Group C: structure with 7 blocks

- 6 networks with density ranging from .06 to .11.
- All networks are represented in 5 or 6 of the 7 blocks, including the first three blocks.
- 3 of the 5 networks of dataset 48 (diff. collecting sites).
- Top trophic level divided into 2 blocks, species from those blocks preying only on intermediate trophic level species.



Group D: structure with 7 blocks

- 23 networks.
- The 10 networks from dataset 157 (stream food webs from New Zealand) are divided between groups B and D based on the type of ecosystem. The data from group B were collected in creeks, while the one from group D were collected on streams.



Group E: structure with 7 blocks