

# Statistical Models in Computational Biology

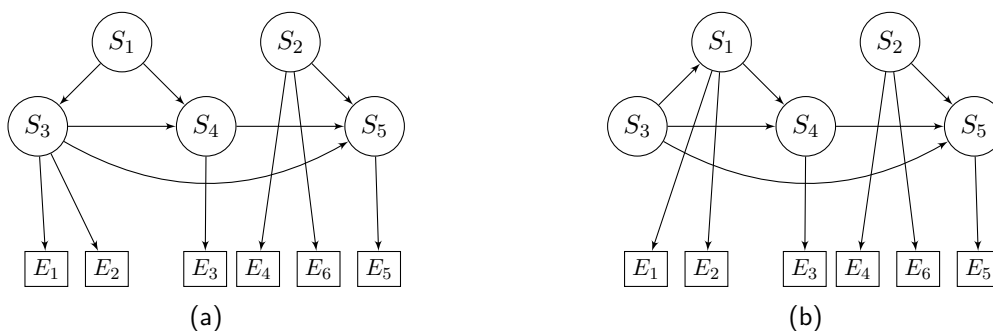
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Due 5th of May 2022

Please submit your project with the filename Lastname(s)\_Project8.pdf.

## Problem 20: Classical NEMs

(3 points)

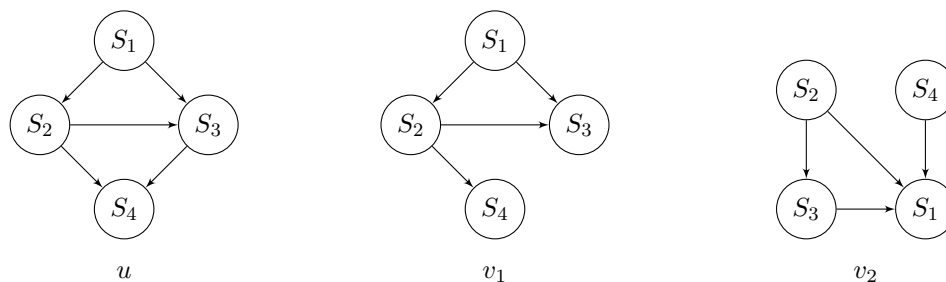


Solve this problem in R.

- For each model, construct the transitive closure (by adding edges) and define the corresponding adjacency matrices  $\Phi$  and  $\Theta$ , which represent the signalling pathways and the E-gene attachments. Determine the corresponding expected effect patterns ( $F$ ). (1.5 point)
- Assuming **no noise**, determine the discrete data  $D_1$  and  $D_2$  from both models. Given only the data, can you tell apart the two models? (0.5 point)
- Use the `mnem`<sup>1</sup> package for this question: Take  $D_1$  and  $D_2$  from the previous question. For each model, calculate the marginal log-likelihood ratio (network score) given the data by setting the false positive rate to be 5% and the false negative rate to be 1%. (1 point)

## Problem 21: Hidden Markov NEMs

(3 points)



<sup>1</sup>Please install this version: <https://www.bioconductor.org/packages/devel/bioc/html/mnem.html>

Solve this problem in R.

(Hint: The *mnem* package does not have an inbuilt function to compute the transition probabilities directly. From the lecture slides implement the different steps and use the functions in the *mnem* package wherever necessary. Please use `mnem:::enumerate.models` to enumerate networks.)

1. Using the definitions for HM-NEMs from the lecture, compute the transition probabilities from  $G_t = u$  to  $G_{t+1} \in \{v_1, v_2\}$  for different smoothness parameter  $\lambda \in \{0.1, \dots, 0.9\}$ . (2 points)
2. Plot the transition probabilities for  $v_1$  and  $v_2$  as a function of  $\lambda$ . Describe the transition probabilities as a function of  $\lambda$ . (1 point)

## Problem 22: Mixture NEMs

(4 points)



Given are two NEMs  $F_1$  and  $F_2$  with two S-genes  $\{S_1, S_2\}$  and two E-genes  $\{E_1, E_2\}$ . The data contains four cells  $\{C_1, C_2, C_3, C_4\}$ .  $\{C_1, C_3\}$  are perturbed by a knock-down of  $S_1$ , and  $\{C_2, C_3, C_4\}$  are perturbed by a knock-down of  $S_2$ . (Hint: You can choose to solve in R or by hand.)

1. Determine the cellular perturbation map  $\rho$ , where  $\rho_{ic} = 1$  if cell  $c$  is perturbed by a knock-down of S-gene  $i$ . (0.5 points)
2. Assume that  $\{C_1, C_2\}$  are generated from  $F_1$  and  $\{C_3, C_4\}$  are generated from  $F_2$ , compute the **noiseless** log odds matrix  $R$ , where  $R_{jc} > 0$  means that the perturbation on cell  $c$  has an effect on E-gene  $j$ :
  - (a) For each component  $k$ , compute the expected effect pattern  $(\rho^T \phi_k \theta_k)^T$ . Replace all non-zeros by 1. (1 points)
  - (b) Based on the component assignment for each cell, extract the corresponding column from the expected effect patterns computed above and put it into  $R$ . Replace all zeros by  $-1$ . (0.5 points)
3. Take  $R$  from the previous question. Given the vector of mixture weights  $\pi = (0.44, 0.56)$ , calculate the responsibilities  $\Gamma$ . Then, update the mixture weights. (2 points)