### BIOINF - MATH 540: Mathematics of Biological Networks

Assignment date: November 13, 2019

Due date: December 3, 2019

Name:

## Problem Set 7: Evaluating Network Controllability from Data

Discrete-time-invariant linear control systems of the form

$$x[k+1] = Ax[k] + B_T u[k], (1)$$

can be used to analyze control over the cell state, where x[k] is the state at time k  $(N \times 1)$ , A is the state transition matrix  $(N \times N)$ ,  $B_T$  is the input matrix  $(N \times D_T)$ , and u[k] is the input function  $(N \times 1)$ .

In this assignment, the cell state refers to the type of cell (e.g. skin, muscle, etc) and we wish to reprogram the cell into a new type. We define each element in the state variable  $x_i[k]$  to be the gene expression level of a Topologically Associating Domain (TAD) i at time k, which is defined as the sum of the expression levels of all genes contained within the TAD (units: fragments per kilobase of transcript per million, FPKM). TADs are inherent structural units of the chromosome: contiguous segments of the 1-D genome for which empirical physical interactions can be observed [1]. The initial cell type here will be fibroblast, and the set of initial snapshots  $x_{\text{FIB}}^m$ , m = 1, ..., M represent the observed expression at the M = 8 different time points mentioned above [2]. A single arbitrary initial snapshot will be denoted  $x_I$ . A is derived from Dynamic Mode Decomposition (DMD) on the time series data (See: Additional information).

The input matrix  $B_T$  is defined from the regulatory set of T as follows:  $D_T$  = the number of TADs in the regulatory set of T,  $\{i_1 < i_2 < ... < i_{D_T}\}$  = the indices of TADs in the regulatory set of T, column j of  $B_T$  have a 1-entry in row  $i_j$ , and zeros elsewhere.

The following measures are used to evaluate control:

- 1. Minimal possible distance to target  $(\mu_1)$
- 2. Energy  $(\mu_2)$
- 3. Trace of controllability gramian  $(tr(W_c))$

Distance is defined as

$$d(u) = ||x[u, K, x_I] - x_F||_2$$
(2)

with the standard Euclidean norm. The term  $x[u, K, x_I]$  denotes the state of Equation 1 after K time steps, starting from state  $x_I$  and using input u.  $x_F$  is the target state. For each  $B_T$  and corresponding Equation 1, we can compute the optimal control  $u^*(B_T)$  that minimizes the distance d to  $x_F$ . The energy of an input signal u is defined as

$$e(u) = \sum_{k=0}^{K-1} u[k]^T u[k].$$
(3)

Both  $d(u^*)$  and  $e(u^*)$  quantify some capacity of Equation 1 to be steered from  $x_I$  to  $x_F$ . We define reprogramming measures 1 and 2 as

$$\mu_1(B) := d(u^*(B_T))$$
 (4)

$$\mu_2(B) := e(u^*(B_T)),$$
 (5)

where smaller distances and energies imply higher capacity for reprogramming.

The controllability gramian  $W_c$  is defined as

$$W_c = \sum_{k=0}^{K-1} A^k B B^T (A^k)^T. (6)$$

The trace of  $W_c$  is a measure of average controllability [3], where larger values imply greater control.

## Problem 1

From the data provided, compute the following measures for each individual TF. For this problem, let K = 3,  $x_I = x_{\text{FIB}}^8$ , and  $x_F = x_{\text{ESC}}$ 

- (a) Minimal possible distance to target  $(\mu_1)$
- (b) Energy of input that minimizes distance to target  $(\mu_2)$
- (c) Trace of controllability gramian  $(tr(W_c))$

### Problem 2

Where do the following TFs rank in each of the following measures:  $tr(W_c)$  and  $\mu_1$  (out of the 332 TFs)?

(a) POU5F1 (b) SOX2 (c) KLF4 (d) MYC (e) NANOG

## Answer for Problem 2

- (a) POU5F1  $\mu_1$ : 332; tr( $W_c$ ): 1
- (b) SOX2  $\mu_1$ : 1; tr( $W_c$ ): 2
- (c) KLF4  $\mu_1$ : 2; tr( $W_c$ ): 3
- (d) MYC  $\mu_1$ : 3; tr( $W_c$ ): 4
- (e) NANOG  $\mu_1$ : 4;  $tr(W_c)$ : 5

#### Additional information

## MATLAB data variable explanation

- A: State transition matrix.
- X\_fib: TAD-level RNA-seq expression time series data on fibroblasts  $(x_{\text{FIB}})$
- X\_esc: TAD-level RNA-seq expression for embryonic stem cells  $(x_{ESC})$
- B: Control matrix
- TF\_names: Names of transcription factors. These correspond to the columns of B

### **DMD** computation

Recall from the main text we want to identify an A matrix such that Y = AX, where X and Y are the first and last M-1 columns of the initial data matrix  $x_{\rm FIB}^m$ , respectively. A straight-forward identification of the matrix A is then computed by  $A \approx \bar{A} := YX^{\dagger}$ , where  $^{\dagger}$  represents the Moore-Penrose pseudoinverse. DMD computes the eigendecomposition  $\bar{A}\Phi = \Phi\Lambda$  of the linear operator  $\bar{A}$ 

## Computation of control evaluations

The so-called controllability gramian  $W_c$  is an important characterizing matrix for any linear system 1, and is defined as

$$W_c = \sum_{k=0}^{K-1} A^k B B^T (A^k)^T, (7)$$

The trace of  $W_c$  is a measure of the systems average controllability [3], where larger values imply greater

control. For comparison with our targeted control measures, we also compute  $tr(W_c)$ :

$$\operatorname{tr}(W_c) = \operatorname{tr}\left(\sum_{k=0}^{K-1} A^k B B^T (A^k)^T\right)$$
$$= \operatorname{tr}\left(\sum_{k=0}^{K-1} B B^T (A^k)^T A^k\right)$$
$$= \operatorname{tr}\left(B B^T \sum_{k=0}^{K-1} (A^k)^T A^k\right)$$

That is,  $\sum_{k=0}^{K-1} (A^k)^{\mathrm{T}} A^k$  can be computed once.

Computation of  $\mu_1$  and  $\mu_4$  is almost strictly linear algebra. For any (A, B), if A is  $N \times N$  and B is  $N \times D$ , define the  $N \times KD$  matrix C as

$$C = (B, AB, \dots, A^{K-2}B, A^{K-1}B),$$
(8)

Finally, let

$$z := x_F - A^K x_I. (9)$$

The minimal-distance control  $u_*$  is given by

$$u_* := C^{\dagger} z, \tag{10}$$

and  $\mu_1(B)$  is the corresponding minimum distance given by

$$\mu_1(B) = \|Cu_* - z\| = \|\left(CC^{\dagger} - I_N\right)z\|_2. \tag{11}$$

 $\mu_2(B)$  is the energy of  $u_*$ , i.e.

$$\mu_2(B) = e(u_*) = \sum_{k=0}^{K-1} u_*^{\mathrm{T}}[k] u_*[k]$$
(12)

# References

- [1] Jie Chen, Alfred O Hero III, and Indika Rajapakse. Spectral identification of topological domains. Bioinformatics, 32(14):2151–2158, 2016.
- [2] Scott Ronquist, Geoff Patterson, Lindsey A Muir, Stephen Lindsly, Haiming Chen, Markus Brown, Max S Wicha, Anthony Bloch, Roger Brockett, and Indika Rajapakse. Algorithm for cellular reprogramming. *Proceedings of the National Academy of Sciences*, 114(45):11832–11837, 2017.
- [3] R.W. Brockett. Finite Dimensional Linear Systems. John Wiley & Sons, Inc., New York, USA, 1970.