# What functions do gene expression levels represent?

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What functions do gene expression levels represent?

#### Previous talk

- Evolvability of sparse linear functions
- Based on the observation that gene expression levels are regulated by transcription factors networks that have low depth and fan-in



Identify a framework describing:

- More realistic models of gene expression
- Parameters corresponding to knobs tuned by evolution

- Review some gene regulation mechanisms
- Summarize a known framework (Markov chains, dynamical systems)
- Study the functional form of solutions (Kirchoff's Matrix Tree Theorem)
- Consider generalizations and implications for evolvability

## Gene expression basics

- DNA is transcribed into mRNA, which is translated into protein
- Gene expression level is proportional to the fraction of promoters bound by RNA polymerase (RNAp)



 Physical state of the promoter can make RNAp binding (un)favorable, equivalent to changing the "effective" amount of RNAp

# Simple 2-state system



- Markov chain with 2 states: promoter is empty or bound by RNAp
- a and b are rate constants
- Let x = probability RNAp is not bound
   y = probability RNAp is bound ⇒ Gene expression level ∝ y
- What do the rate constants a, b depend on?
- What kind of functions are x, y in terms of a, b?

L. Bintu, *et al.* Transcriptional Regulation by the Numbers 1: Models, Current Opinion in Genetics & Development 15(2):116-124 (2005)

### Simple 2-state system



- P = number of RNAp molecules
- N = number of non-specific (NS) sites where RNAp can bind to DNA

probability of state S  $\propto \exp(-E_S/k_BT)$ 



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### Simple 2-state system



- *P* = number of RNAp molecules
- N = number of non-specific (NS) sites where RNAp can bind to DNA

$$Z = \frac{N!}{P!(N-P)!} e^{-PE_{NS}/k_BT} + \frac{N!}{(P-1)!(N-P-1)!} e^{-(P-1)E_{NS}/k_BT} e^{-E_P/k_BT}$$

$$E_{NS}$$

$$E_{NS}$$

$$E_P$$

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### Simple 2-state system: Equilibrium probabilities



N = number of non-specific (NS) sites where RNAp can bind to DNA P = number of RNAp molecules

$$Z = \underbrace{\frac{N!}{P!(N-P)!} e^{-PE_{NS}/k_{B}T}}_{X} + \underbrace{\frac{N!}{(P-1)!(N-P-1)!} e^{-(P-1)E_{NS}/k_{B}T} e^{-E_{P}/k_{B}T}}_{Y}$$
Probability RNAp bound to promoter 
$$= \frac{Y}{X+Y} = \frac{1}{1+\frac{X}{Y}}$$

$$\approx \frac{1}{1+\frac{N}{P} e^{\Delta E/k_{B}T}} = \frac{1}{1+\frac{1}{[RNAP]} \frac{K_{S}}{K_{NS}}}$$

L. Bintu, *et al.* Transcriptional Regulation by the Numbers 1: Models, Current Opinion in Genetics & Development 15(2):116-124 (2005)

## (Part of) a real promoter system ...



H.D. Kim and E.K. O'Shea, A quantitative model of transcription factor activated gene expression, Nature Structural Molecular Biology, 15(11), 1192-1198 (2008).

#### Promoters can have many states





- $2 \times 3 \times 2 = 12$  states
- Example: promoter composed of 3 physical regions
- The regions can be empty or occupied
- RNAp can bind only when the 3rd region is (6/12 states)

## Transitions change occupancy at 1 promoter region



- This transition "adds" or "removes" 😔 at the 3rd promoter region
- These rates are physical constants

# Some transitions are due to binding / unbinding



- Rate of binding  $\propto$  [free  $^{12}$ ]
- Assumption: [free  ${}^{oldsymbol{ heta}}$ ] pprox constant w.r.t. promoter system dynamics

# (Part of) a real promoter system ...



- This graph is sparse and strongly connected
- Some transitions are irreversible

# Representation: Graph G(V, E)



- Vertices correspond to states
- Edges are directed and labeled by non-negative transition rates
- No self-edges

I. Mirzaev, J. Gunawardena, "Laplacian dynamics on general graphs", Bull Math Biol, 75:2118-49 2013.

# Continuous-time Markov chain



- $x_i$  = probability of state *i*
- $\frac{d\vec{x}}{dt}$  describes the rates at which the state probabilities change

$$\frac{dx_1}{dt} = ax_2 - (b+c)x_1$$
$$\frac{dx_2}{dt} = bx_1 + dx_4 - ax_2$$
$$\frac{dx_3}{dt} = cx_1 - ex_3$$
$$\frac{dx_2}{dt} = ex_3 - dx_4$$

# Continuous-time Markov chain



$$\frac{d\vec{x}}{dt} = \begin{pmatrix} -(b+c) & a & 0 & 0 \\ b & -a & 0 & d \\ c & 0 & -e & 0 \\ 0 & 0 & e & -d \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} \equiv \mathcal{L}\vec{x}$$

# Continuous-time Markov chain



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- $\mathcal{L}(G) = \text{Laplacian of graph } G$
- Columns of *L* sum to zero
- Conservation of mass:  $x_1(t) + \cdots + x_n(t) = c$

### Focus on strongly connected graphs



- Physically meaningful setting
- Results can be generalized to graphs that are not strongly connected
- Strongly connected:  $\geq$  1 directed path between any 2 vertices  $\geq$  1 spanning tree rooted at each vertex

### Spanning trees rooted at vertex 1



- A tree is a graph with no cycles
- A spanning tree has all *n* vertices  $\Rightarrow n 1$  edges
- Rooted at 1: All paths lead to 1

# Spanning trees rooted at vertex 2



# All spanning trees



### Solve for steady state when G is strongly connected

- Steady state: x(t) converges to a stationary distribution x\*
- Want to solve for  $\vec{x}^*$  s.t.  $\frac{d\vec{x}}{dt} = \mathcal{L}\vec{x}^* = 0$
- Strongly connected  $\Rightarrow$  rank of  $\mathcal{L}$  is exactly n-1
- Recall: columns of  $\mathcal{L}$  sum to zero

 $\Rightarrow 1^T \mathcal{L} = 0 \Rightarrow 1$  is a left eigenvector with eigenvalue 0

- $\Rightarrow \exists$  corresponding right eigenvector  $\vec{\rho}$  s.t.  $\mathcal{L}\vec{\rho} = \mathbf{0}$
- $\Rightarrow$  Solution  $\vec{x}^* \propto \vec{\rho}$

- $\mathcal{L}$  is not full-rank  $\Rightarrow det(\mathcal{L}) = 0$
- Determinant expansion ⇒ ∀j, det(L) = ∑<sub>k</sub>(-1)<sup>j+k</sup>ℓ<sub>jk</sub> det(L<sub>j,k</sub>) = 0 det(L<sub>i,k</sub>) = (j, k)-minor of L = determinant of submatrix made by deleting the j-th row and k-th col

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- Fact for Laplacians:  $det(\mathcal{L}_{j,k}) = (-1)^{j+k} det(\mathcal{L}_{k,k})$
- Then:  $\forall j, \sum_k \ell_{jk} \det(\mathcal{L}_{k,k}) = 0 \Rightarrow \left| \mathcal{L}\vec{\rho} = 0 \text{ for } \rho_k = \det(\mathcal{L}_{k,k}) \right|$

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- Kirchhoff's Matrix Tree Theorem ⇒

$$\det(\mathcal{L}_{k,k}) = (-1)^{n-1} \sum_{T \in \Theta_k} \prod_{e \in T} a_e$$

where n = number of vertices

- $\Theta_k = \{\text{spanning trees rooted at vertex } k\}$
- $a_e$  = weight of edge e

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- Thus, the entries of  $\vec{\rho}$  are homogeneous polynomials in the rates (degree = |T| = n 1)

### Steady state solution: Rational polynomial in the rates



## Steady state solution: Rational polynomial in the rates



- $\rho_i > 0$  since  $\geq 1$  spanning tree at each vertex
- Normalize for steady state:  $x_i^* = \rho_j / (\rho_1 + \rho_2 + \rho_3 + \rho_4)$

#### Gene expression level at steady state

- Gene expression level  $\propto$  probability promoter bound by RNAp
- For example, suppose RNAp can bind to states 2 and 4

Gene expression level  $\propto x_2^* + x_4^* = \frac{\rho_2 + \rho_4}{\rho_1 + \rho_2 + \rho_3 + \rho_4} = \frac{1}{1 + \frac{\rho_1 + \rho_3}{\rho_2 + \rho_4}}$ 



- Gene expression level = rational homogeneous polynomial
- Each term is degree n 1 and positive (product of rates)

# Gene expression level: Rational polynomial of rates



T. Ahsendorf, F. Wong, R. Eils, J. Gunawardena. A framework for modelling gene regulation that accommodates non-equilibrium mechanisms, submitted (2013)

# What is the functional form of the rates?

- Example real system: 4/5 rates are constants
- More complicated dependencies are possible, e.g., dimerization

## What could be under selection in this framework?

Sequence mutations to DNA can occur at:

- Promoter regions that bind transcription factors and other molecules
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Mutations can perturb the model by:

- Adding new states (new site where TF can bind; new TF that can bind)
- Changing rates (modify binding properties of DNA or protein)
- Removing states (*e.g.*, send a rate to 0)

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