# 7. Transcriptional regulation from microarray data

Warning: Statistical physics. It only works on average.

http://regan.med.harvard.edu/CVBR-course.php

### How about the entire genetic regulatory network?



#### If we know enough interactions:



#### An overview of approaches

#### Co-expression networks

#### Boolean networks

### ODEs and regression

Bayesian networks

- Known literature
- Transcription factors
- Binding site information
- Perturbation experimental design



#### Sounds nice! How does clustering go?



## Coherent groups of genes & experiments









#### CLR – Context Likelihood of Relatedness







#### Boolean network approaches

- State of the system, or gene activity profile: (0,1,1,0,0,1,1,1,0,...,0)
- State changes in time: trajectory in nD space
- Structure of state space determines all possible dynamics

\* Ordered systems

S(t.)

C high-dimensional system as "attractor landscape (schematic)



- Only a small fraction of all cell states are stable
- The system does not visit all possible states
- Attractors:

All genes <- all possible k inputs need to be tested

#### Finding expression dependencies

- Bayesian networks
  - Directed Acyclic Graph
  - Its structure describes the conditional probabilities that best fit the data
  - Discrete or continuous
  - Extremely time-consuming on large networks

1: Maxímum k ínputs / gene

2: Find relevant part of gene space

**Regression based methods** 

 $\left[X_i^{(k)} = \sum h_{j \to i} \cdot X_j^{(k)} + p_i^{(k)}\right]$ 

 $N_{exp} \ll N_{gene}$ 

#### NIR – Network identification by Multiple regression

$$\begin{split} X_i^{(k)} &= \sum_j h_{j \to i} \cdot X_j^{(k)} + p_i^{(k)} \\ \sum \mathbf{Z}_j^{(k)} \cdot \mathbf{A}'_{ji} = - \mathbf{P}_i^{(k)} \end{split}$$

- o Iterative approximation
  - o Oth approximation:
  - o calculate  $P_i^{(k)}$
  - for each gene
    use n ∈ (Q, M)
    equations where pi < pc</li>
    solve



#### Summer break!

Next lecture:

Sep 14 12 PM