Emergence of gene regulatory networks under functional constraints

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in collaboration with

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EUROPEAN UNION EUROPEAN REGIONAL DEVELOPMENT FUND



Number of protein-coding genes

E. coli	\sim	4000
Yeast		6000
Fruit fly		14000
Human		23000



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			a			

Interactions between genes

Number of regulatory genes grows faster than linearly with the total number of genes.



Genetic information

Genome sequencing does not bring all essential information.



Source: http://xbase.ac.uk

Genetic information

Gene regulation plays an important role.



Source: http://xbase.ac.uk

Gene regulation

Cell differentiation

- different gene expression patterns correspond to different tissues
- many regulatory mechanisms
- transcription factors (TFs) promote/block other genes



Cell-division cycle

 gene expression patterns correspond to different phases

Gene regulation

Cell differentiation

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- transcription factors (TFs) promote/block other genes



...00011...

... 0 1 1 0 0 ...

Cell-division cycle

 gene expression patterns correspond to different phases



Baker's yeast cell-cycle

Regulatory network and gene expression pattern



Source: F. Li, T. Long, Y. Lu, Q. Ouyang and C. Tang, PNAS 101, 4781 (2004).

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Emergence of gene regulatory networks

0 0 0

Baker's yeast cell-cycle

Regulatory network and gene expression pattern



Dynamical models of gene networks

Focusing on transcriptional dynamics

- Boolean networks (genes on-off) (e.g. Stuart Kauffman)
- Threshold networks (*e.g.* inspired by neural dynamics, Andreas Wagner)
- Differential equations with rates (*e.g.* Albert Goldbeter, John Tyson)
- Piece-wise linear input-output relations (*e.g.* Hidde de Jong)

• ...

Phenotype

$$\mathbf{S}(t) = \\ (S_1(t), S_2(t), \dots, S_N(t))$$



$\mathbf{S}(t+1) = G(\mathbf{S}(t), \mathbf{W})$ $\mathbf{S}(0) \xrightarrow{\mathbf{W}} \mathbf{S}(1) \xrightarrow{\mathbf{W}} \dots \xrightarrow{\mathbf{W}} \mathbf{S}(t)$

Farget pattern

$$S^{target}(0) = 1 1 0 0 1$$

$$S^{target}(1) = 0 1 1 0 0$$

$$\vdots$$

$$S^{target}(T) = 0 0 0 1 1$$

Fitness

$$F(\mathbf{S}) = \exp(-f\sum_{t=0}^{l} |\mathbf{S}(t) - \mathbf{S}^{target}(t)|)$$

Source: Z. Burda, A. Krzywicki, O.C. Martin, M. Zagorski, PNAS 108, 17263 (2011).

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Phenotype



Transcriptional dynamics $\mathbf{S}(t+1) = G(\mathbf{S}(t), \mathbf{W})$ $\mathbf{S}(0) \xrightarrow{\mathbf{W}} \mathbf{S}(1) \xrightarrow{\mathbf{W}} \dots \xrightarrow{\mathbf{W}} \mathbf{S}(t)$

$\begin{array}{l} \text{rget pattern} \\ \mathbf{S}^{target}(0) &= 1\ 1\ 0\ 0\ 1 \\ \mathbf{S}^{target}(1) &= 0\ 1\ 1\ 0\ 0 \\ \vdots \\ \mathbf{S}^{target}(T) &= 0\ 0\ 0\ 1\ 1 \end{array}$

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Starget pattern $S^{target}(0) = 11001$ $S^{target}(1) = 01100$ \vdots $S^{target}(T) = 00011$

Source: Z. Burda, A. Krzywicki, O.C. Martin, M. Zagorski, PNAS 108, 17263 (2011).

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Phenotype



Transcriptional dynamics

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Emergence of gene regulatory networks

Biophysical modelling of interactions



Biophysical modelling of interactions



The binding energy between two molecules is additive between facing elements with each mismatch contributing a penalty ε .

$$P_{ij} = rac{n_j W_{ij}}{1 + n_j W_{ij}} ext{ where } n_j = n S_j(t)$$

Source: U. Gerland, J. Moroz and T. Hwa, PNAS 99, 12015 (2002).

Expression dynamics and MCMC sampling

$$S_i(t+1) = \underbrace{\left[1 - \prod_j (1 - P_{ij}(t))
ight]}_{ ext{Activators}} \underbrace{\prod_{j'} (1 - P_{ij'}(t))}_{ ext{Repressors}}$$
 $P_{ij}(t) = rac{1}{1 + \exp\left(arepsilon d_{ij} - \log nS_j(t)
ight)}$

Metropolis sampling of space of viable genotypes.



Network of essential interactions

Example of matrix \mathbf{W} with corresponding regulatory network for yeast cell-cycle expression pattern.



Most frequent network in the ensemble





Network made of most frequent interactions

Edge usage

Activatory interactions tend to form a long feed forward cascade

	Ch.	ABF	4 8 5	Qul's	Ch5.6	Clb1,2	Menu	Cae 20	Suris	,;; ;;;	Contraction of the second seco
Cln3	0.04	0	0	0	0	0	0	0	0	0	0
MBF	1.	0.5	0.54	0.01	0	0	0	0	0	0	0
SBF	1.	0.51	0.52	0.01	0	0	0	0	0	0	0
Cln1,2	0	0.59	0.62	0	0	0	0	0	0	0	0
Clb5,6	0	0.39	0.4	0.38	0.06	0	0	0	0	0	0
Clb1,2	0	0	0	0	0.99	0.01	0	0	0	0	0
Mcm1	0	0	0	0	0.87	0.98	0	0	0	0	0
Cdc20	0	0	0	0	0.02	0.1	1.	0	0	0	0
Swi5	0.01	0	0	0.01	0.01	0.05	0.09	1.	0	0.01	0.01
Sic1	1.	0	0	0	0	0.01	0.01	0	1.	0	0
Cdh1	0	0	0	0	0	0	0	0	0.01	1.	0

Regulatory network

Interactions present in >60% of networks



Results

- much of the cell-cycle network is reproduced under functional constraints
- motifs found are compatible with the imposed functional capability
- GRNs are as *sparse* as possible to maintain the imposed expression pattern

Similar outcome for two other cell-cycles:

- fission yeast, Davidich and Borhholdt (PloS One, 2008)
- mammals, Faure et al. (Bioinformatics, 2006)

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- Z. Burda, A. Krzywicki, O.C. Martin, M. Zagorski, Motifs emerge from function in model gene regulatory networks, PNAS 108, 17263-17268 (2011).
- Z. Burda, A. Krzywicki, O.C. Martin, M. Zagorski, Distribution of essential interactions in model gene regulatory networks under mutation-selection balance, Phys. Rev. E 82, 011908 (2010).

Thank you for your attention