Making a computer out of junk DNA

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Overview of development

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 - Cell signaling

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Development

How does an organism develop from a single cell?

How does a brain develop?





Cells receive signals from neighboring cells and their environment, causing them to differentiate.



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- It can tell it for example, to stop growing or form a synapse with another cell.



Proteins involved in the MAPK/Erk pathway Communicates surface receptors with DNA



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Activation of eukaryotic transcription



Gene networks



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- Some regulatory proteins enhance and others repress transcription, by binding to regulatory elements.



Analogous to digital logic



Relatively small number of inputs and outputs





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A pine tree has 7 times

Yet the bladder wort has only 3% noncoding DNA
Complication: "Junk DNA"

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Encyclopedia of DNA Elements

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- Is most of doing anything "useful"?
- There appear to be quite a few examples, but most of the function of it is still unknown.



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RNA is "sticky". e.g. It has a lot of secondary structure



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On physical grounds, it is doing something!

RNA Chemical equilibration



Relaxation

$$\tau_{\rho} \frac{d\rho_i}{dt} = -\rho_i + \frac{C_i}{1 + \sum_j \rho_j K_{ij}}$$

Equilibration time, of order 1s

RNA creation

RNA is both degraded and created. For species i $\tau_C \frac{dC_i}{dt} = -C_i + f(C_i, \{\rho_k\})$

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Creation rate, *f*, is regulated by the total concentration of *i* and the concentration of unbound molecules.

Assumption:

All this "junk" will evolve to do something useful

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i.e. Inputs to the genome will give sensible outputs





 J_{ij} 's



 J_{ij} 's chosen to give optimal outputs given inputs



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$$s_i = \tanh(\frac{\beta}{N}\sum_j J_{ij}s_j)$$


A cell





Neural net / RNA mappingspin $\longrightarrow s_i \iff \rho_i \xleftarrow$ Unbound
concentrationspin coupling $\longrightarrow J_{ij} \iff K_{ij} \longleftarrow$ Equilibrium
Constant

$$\rho_i = \delta \frac{1+s_i}{2} + b$$
$$K_{ij} = \epsilon \frac{1+J_{ij}}{2} + a$$

 $t \to \infty$: Want RNA's to give neural net behavior

$$f(C_i, \{\rho_k\}) = \frac{C_i}{\rho_i} S[\frac{4}{\epsilon} (\frac{C_i}{\rho_i} - 1) - 2(1 + 2\frac{a}{\epsilon}) \sum_{j=1}^N \rho_j \frac{2(\delta + 2b)}{\epsilon} \sum_j K_{ij} + (\frac{2a}{\epsilon} + 1)(\delta + 2b)N]$$

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Minimum spin states for Boltzmann machine/ Hopfield

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An ugly mess!

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$$f(C_i, \{\rho_k\}) = \frac{C_i}{\rho_i} S(\frac{4}{\epsilon} (\frac{C_i}{\rho_i} - 1) - A)$$

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An ugly mess!
$$f(C_i, \{\rho_k\}) = \frac{C_i}{\rho_i} S(\frac{4}{\epsilon} (\frac{C_i}{\rho_i} - 1) - A)$$
A is a constant

Creation function



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Equivalent to a neural network





Fraction of mistakes as a function of number of hidden units.

Hidden units are initially scrambled.

50 neurons, 3 patterns. 2 output units.

Is this plausible?



Use a known mechanism

Presence of RNA increases its own transcription

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Possible models for how this happens:



"Role of non-coding RNA transcription around gene regulatory elements in transcription factor recruitment" Naomichi Takemata and Kunihiro Ohta RNA BIOLOGY 2017



Making a computer out of junk



Equivalent to:

DEEP NEURAL NETWORK



neuralnetworksanddeeplearning.com - Michael Nielsen, Yoshua Bengio, Ian Goodfellow, and Aaron Courville, 2016.