

Genetics Networks (revisited)

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MIS 214/CS 274

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What is a genetic network?

Individual genes have a *function* (e.g. transforming a substance or binding to a substance)

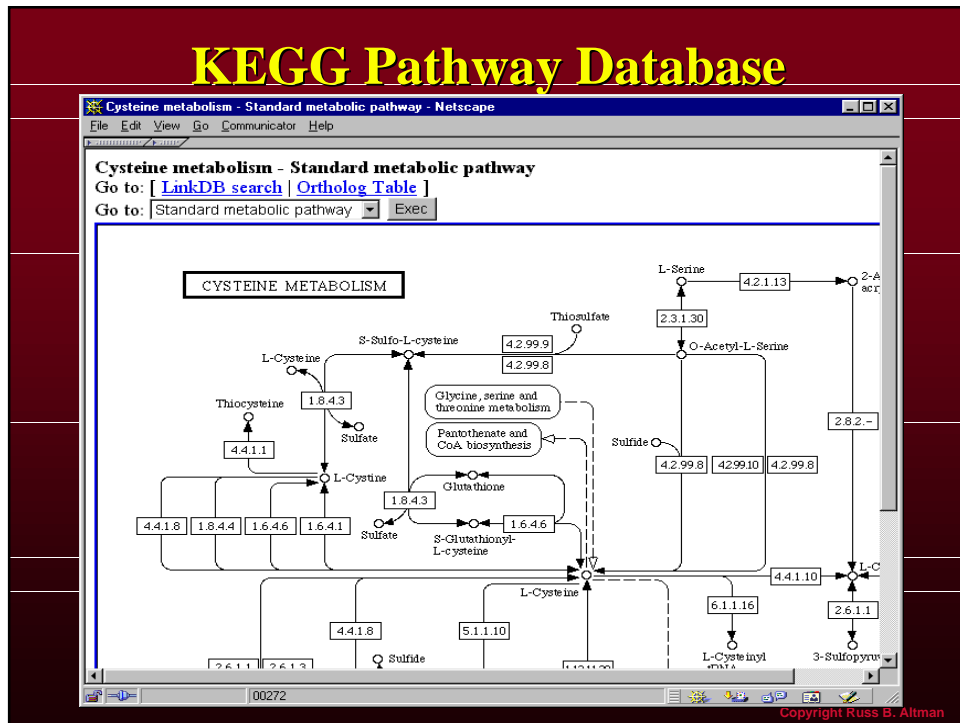
Sets of functions when sequenced can produce *pathways* (e.g. output of one transformation is the input to another)

Sets of pathways, as they interact with other pathways, create a *genetic network* of interactions.

The emergent properties of these networks constitute the “observables” when we study cells.

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KEGG Pathway Database



KEGG Pathway Database

DBGET Result: ENZYME 4.2.99.9 - Netscape

[[LinkDB](#)]

ENTRY	EC 4.2.99.9
NAME	O-Succinylhomoserine (thiol)-lyase Cystathionine gamma-synthase
CLASS	Lyases Carbon-oxygen lyases Other carbon-oxygen lyases
SYSNAME	O-Succinyl-L-homoserine succinate-lyase (adding cyste
REACTION	O-Succinyl-L-homoserine + L-Cysteine = Cystathionine
SUBSTRATE	O-Succinyl-L-homoserine L-Cysteine Hydrogen sulfide Methanethiol
PRODUCT	Cystathionine Succinate L-Homocysteine Methionine 2-Oxobutanoate NH3
COFACTOR	Pyridoxal phosphate
COMMENT	A pyridoxal-phosphate protein. Also reacts with hydro and methanethiol as replacing agents, producing homoc methionine respectively. In the absence of thiol, car catalyse beta,gamma-elimination to form 2-oxobutanoate

Document: Done

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EcoCYC Metaboli Database

<http://ecocyc.pangeasystems.com:1555/server.html>

E. coli Pathway: cysteine biosynthesis

More Detail Less Detail

Synonyms: cyssyn

Superclasses: [Individual amino acids](#)

Net reaction equation: L-serine + acetyl CoA + sulfide = L-cysteine + CoA + acetate

Comment: The pathway to sulfide synthesis feeds in to this pathway for cysteine biosynthesis.

Pathway interactions: Cysteine is used in the synthesis of methionine, both sulfur containing amino acids.

Superpathways: [sulfate assimilation and cysteine biosynthesis](#)

Locations of Mapped Genes:

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Gene Functions: X, Y, Z
 Pathways: X-Y, X-Z-Y, X-Z-X
 Networks: XYZ

Level in Cell

Time

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Studying Gene Networks

Until recently, difficult because

- (1) Data about interactions and timing of expression was very difficult to collect.**
- (2) No good methods to analyze these networks**

But now:

- (1) Methods for measuring the “network” are available.**
- (2) Progress made in thinking about computing on these network.**

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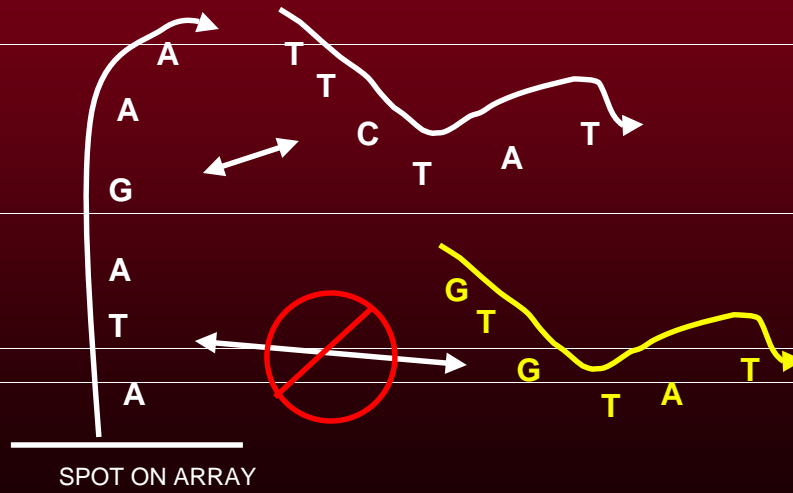
Availability of Relevant Data

Gene Arrays

- (1) Fix pieces of DNA of known sequence on a 2-dimensional array (e.g. all 6000 yeast genes)**
- (2) Gather samples from cells under two conditions (e.g. starved vs. not starved)**
- (3) Label the mRNA of the two samples with fluorescent dyes of different colors (green/red)**
- (4) Mix the samples and pour on the 2D array. mRNA that is complementary (in Watson-Crick sense) will anneal to pieces of DNA.**
- (5) Measure fluorescence to see if levels of mRNA in the one sample is same, more, less than the other.**

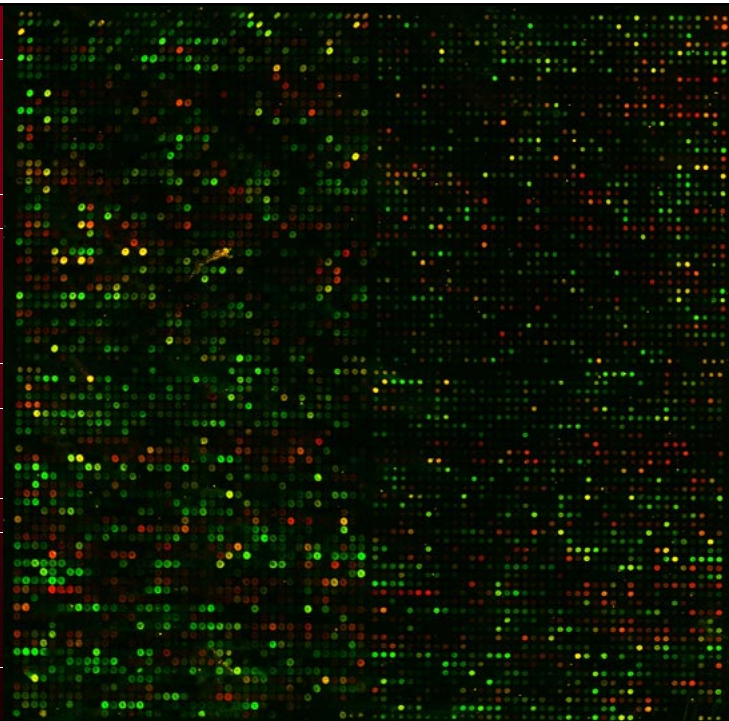
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DNA &/or RNA Annealing



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Typical
DNA
array
for
Yeast



What does this tell you?

mRNA is a proxy for how much protein is available in the cell.

DNA array gives a snapshot of the level of mRNA expression (and thus protein) in the cell at a particular time.

Can take multiple snapshots to watch the evolution of availability of mRNA over time, or in response to stimuli.

For example: in response to starvation, evolution of cancer, normal vs. disease, etc...

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What to do with array data?

Some obvious choices...

- 1. Cluster genes based on similar gene expression = a new metric for “similarity” separate from sequence measures.**
- 2. Try to infer genetic interactions based on timing and co-regulation of genes.**

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Cluster genes based on patterns of expression.

Cluster analysis = unsupervised learning

1. Define a distance metric between two gene expression patterns (e.g. Euclidean distance = $\text{SQRT}((x_1-x_2)^2 + (y_1-y_2)^2 \dots)$)

EITHER:

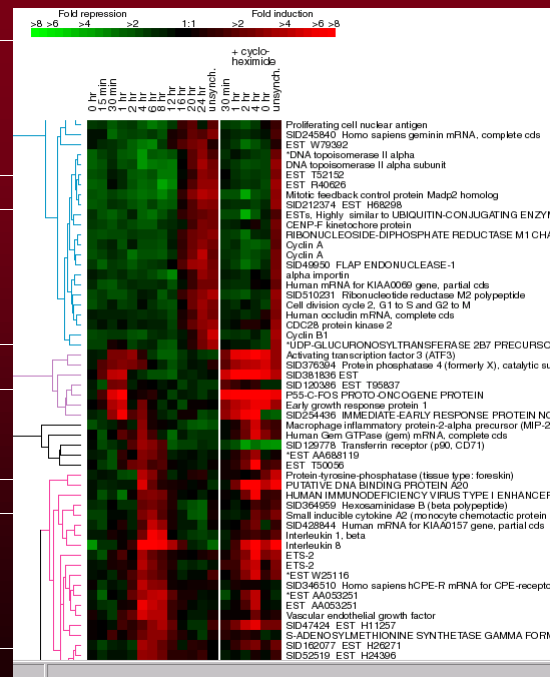
2. Create a distance matrix of all pairwise distances, and then find closest pairs, and combine them (like phylogenetic trees). $O(N^2)$

OR

2. Randomly generate “seed” points of the same dimensionality of the data, and map data points to closest seed. $O(N)$.

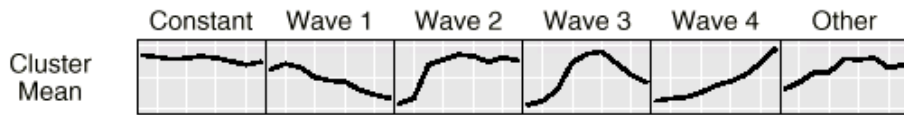
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Can build trees from cluster analysis, groups genes by common patterns of expression.



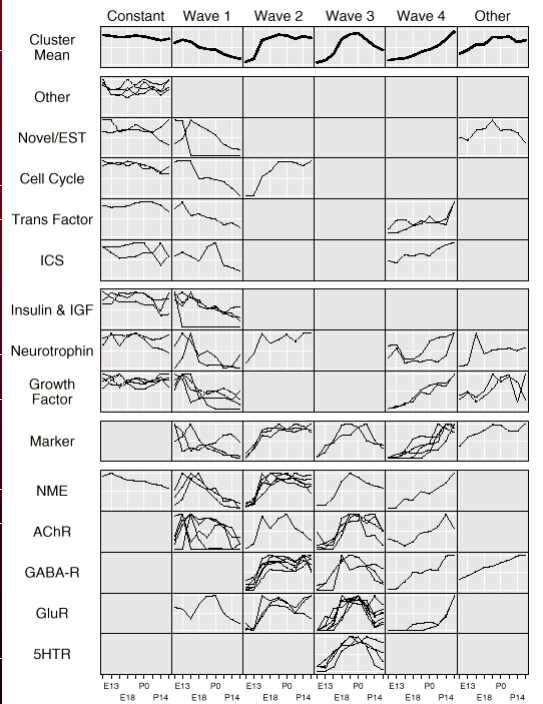
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Average of clustered wave forms



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**Typical
“wave
forms”
observed
(note: not
lots of
bumps)**



Reconstructing Genetic Network

Hard problem.

Given N genes, there are an exponential number of connections between the genes.

Relationships are not generally +/- but are continuous valued (e.g. concentration of molecule varies smoothly).

Must use knowledge about expected function and membership in pathways to prune the list of possible network interactions.

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Simplification: Boolean Network

1. All genes are either in **on** state or **off** state.
2. State of a gene at time T determined by the logical combination of states of regulatory genes at time $T-1$.
3. Can propagate the states over time to simulate the evolution of the network.

Very simplified, but still useful for seeing the kinds of properties that can emerge from these networks.

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Sample network

$$A' = B$$

$$B' = A \text{ or } C$$

$$C' = (A \text{ and } B) \text{ or } (B \text{ and } C) \text{ or } (A \text{ and } C)$$

equivalent to:

INPUT			OUTPUT		
<u>A</u>	<u>B</u>	<u>C</u>	<u>A'</u>	<u>B'</u>	<u>C'</u>
0	0	0	0	0	0
0	0	1	0	1	0
0	1	0	1	0	0
0	1	1	1	1	1
1	0	0	0	1	0
1	0	1	0	1	1
1	1	0	1	1	1
1	1	1	1	1	1

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Some sample state transitions

100-010-100-010-100...

101-011-111-111-111...

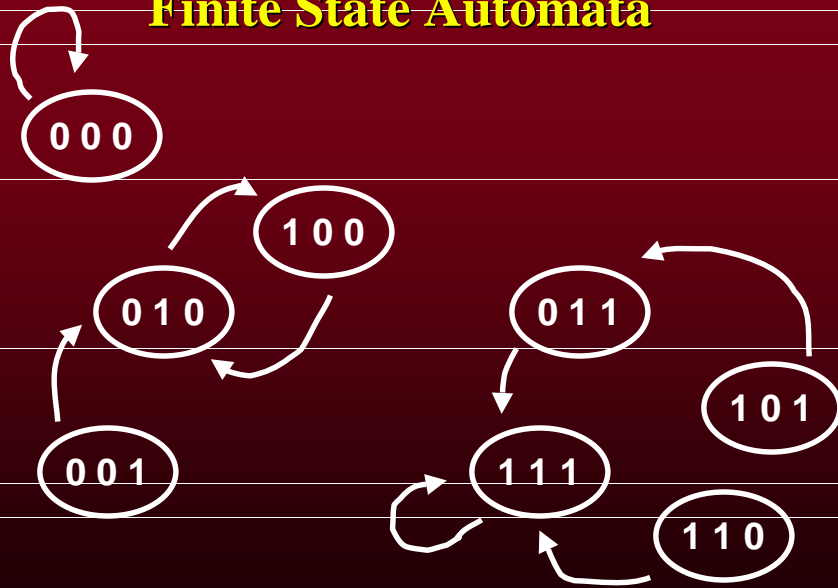
000-000-000...

001-010-100-010-100...

010-100-010-100-010...

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Finite State Automata



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Things to notice

Attractors can be *static* (only one node) or *dynamic* (sequence of nodes that repeat).

Different equilibria are possible, even for very simple network. These can correspond to:

- disease state
- resting state
- perturbed state

Transitions from one attractor to another require external events (starvation, food supply, heat, etc...) to create mutation in state, and move to different attractor.

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Correspondences

Genetic Network	Boolean Network
Genotype/DNA	Wiring and rules
Gene	Element of state
Expression pattern	State
Development	Trajectory
Mature cell	Attractor

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Old Premise (from Somogyi et al)

Gene for every function, function for every gene.

Complete reduction of organism into genes

Determination of protein structures/activities

Mapping molecular gene product interactions

Assembly of database of molecular mechanisms

Synthesis is sum-of-parts computer model.

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New Premise (from Somogyi et al)

Gene function distributed across parallel network

Identify genes and genetic network elements

Determine states of network (expression patterns)

Map out alternative trajectories/attractors

Reverse engineer the network

parallel trajectories suggest shared input

temporal links determined by wave shapes

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RNA Structure Computations

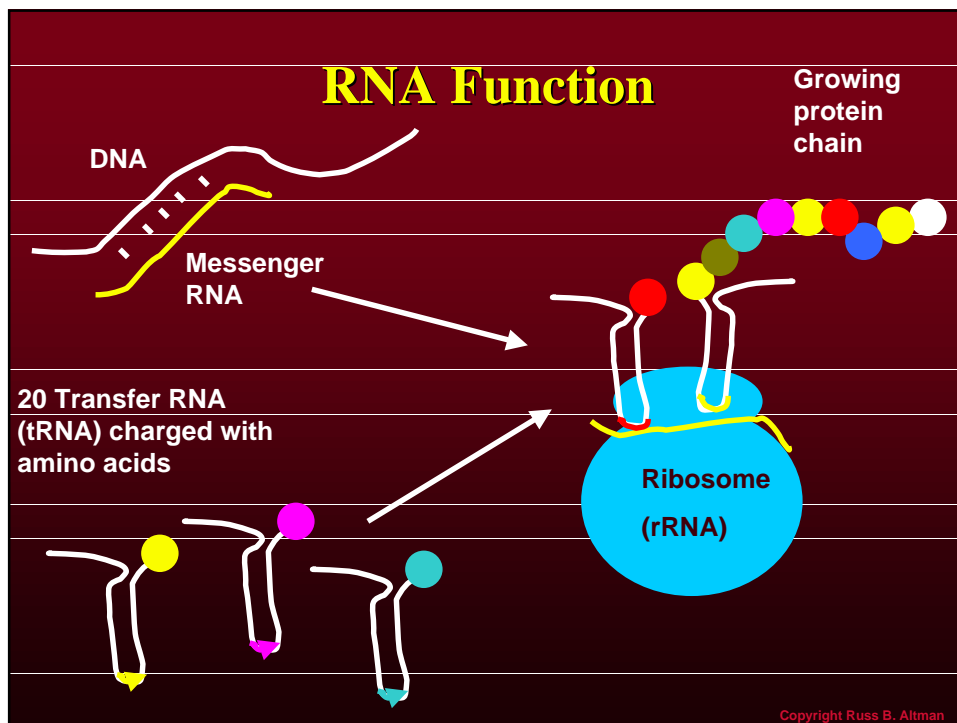
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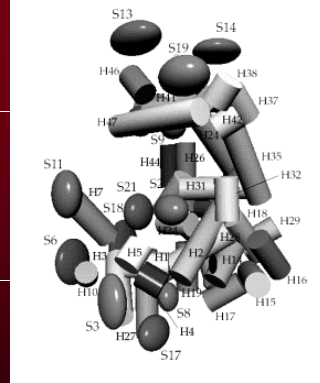
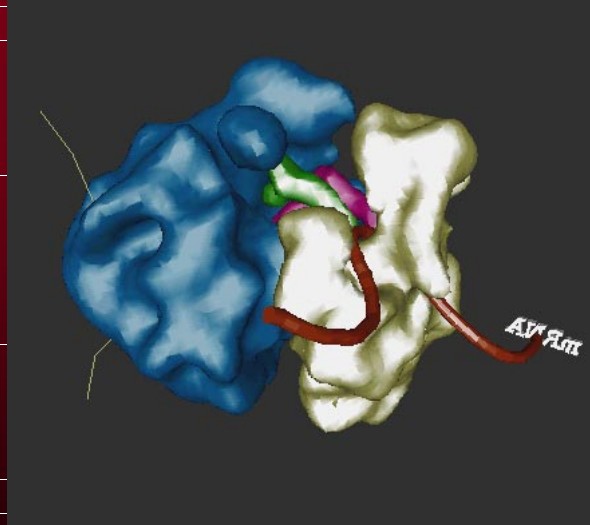
Biological roles RNA

0. Like DNA, RNA contains 4 subunits (AUGC). It is less stable than DNA, so is not a storage media.
1. the DNA code a gene is copied into messenger RNA (mRNA)
2. mRNA is the version of the genetic code translated at the ribosome.
3. the ribosome is made up RNA (ribosomal RNA or rRNA)
4. The individual amino acids are brought to the ribosome, as it reads the mRNA by molecules called transfer RNAs (tRNA)

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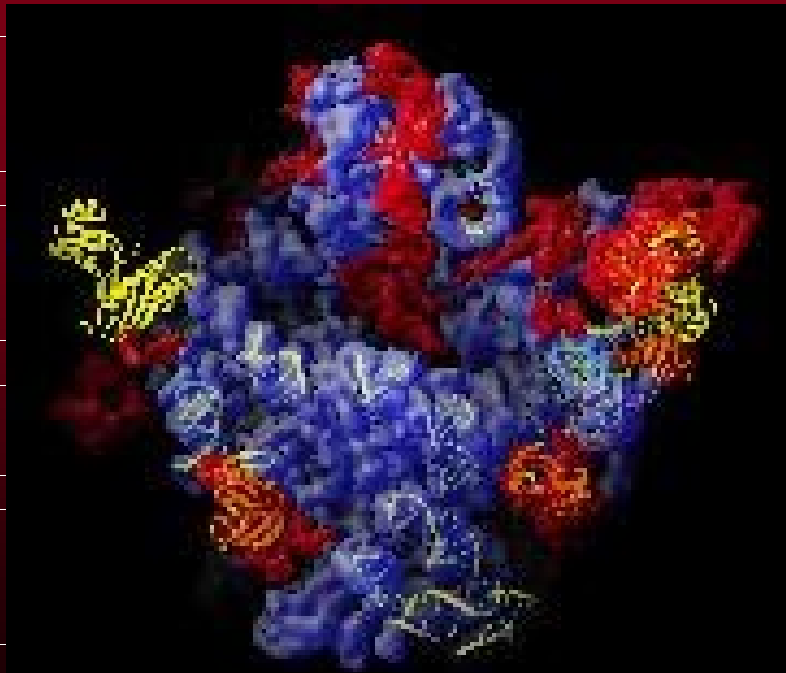


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<http://www.wadsworth.org/BMS/>

<http://www-smi.stanford.edu/projects/helix/ribo.html>
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Insert tRNA pix here.

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RNA has 3D structure

AUUCGGCGACGAU

**Primary
Structure**

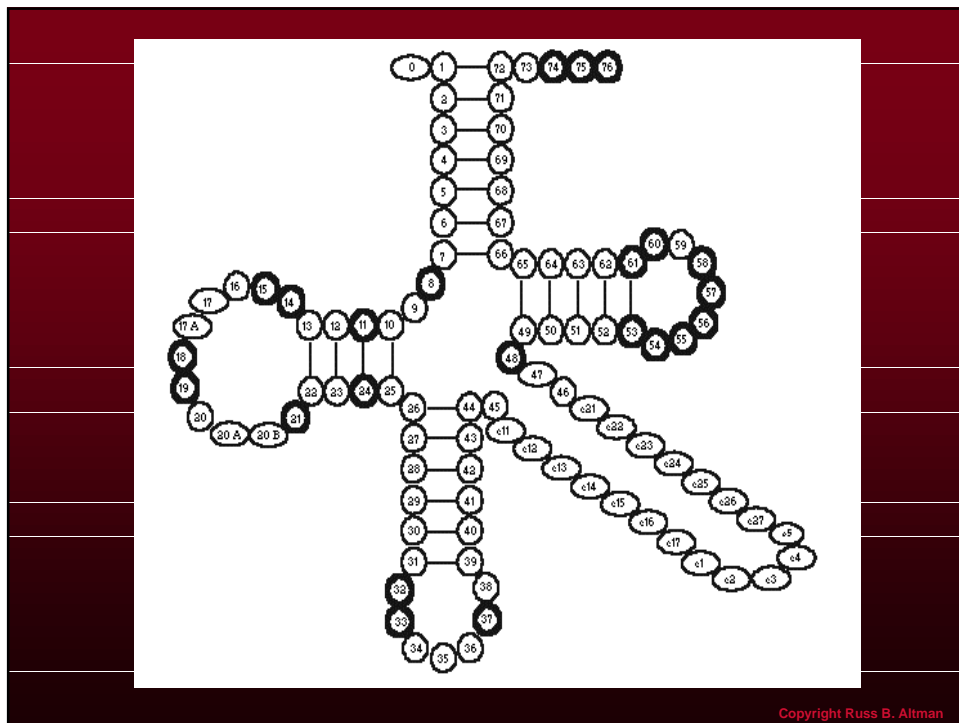
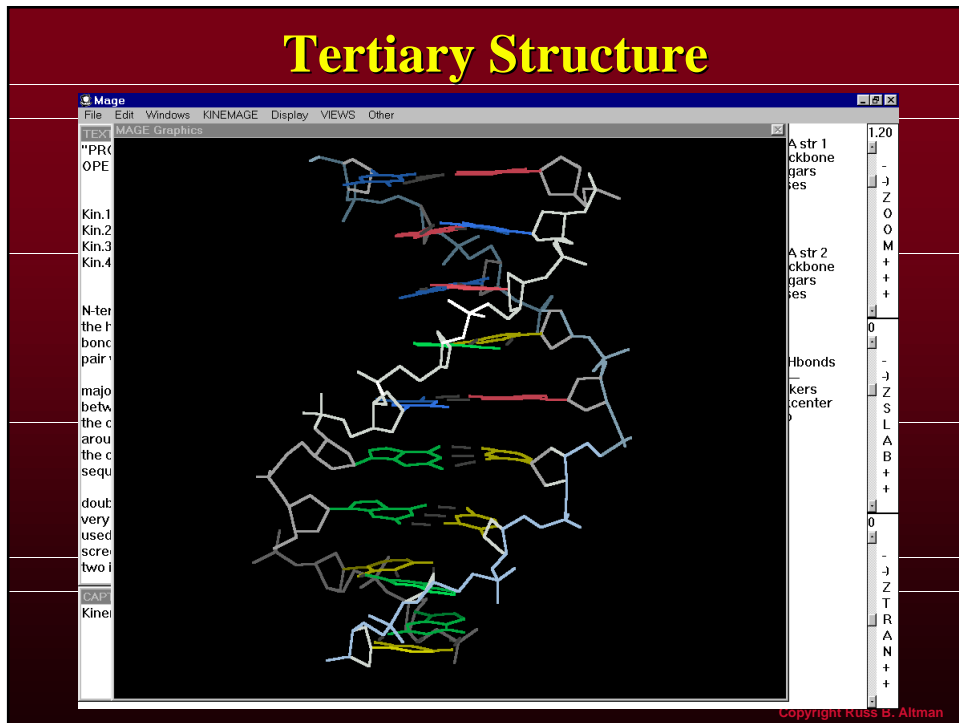


**AUUCG^GG
| | | |
UAAGC^AA**

**Secondary
Structure**

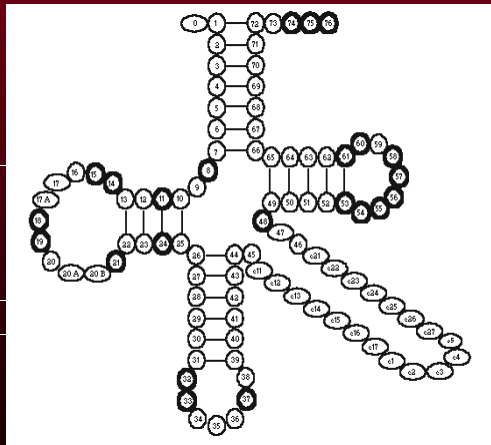
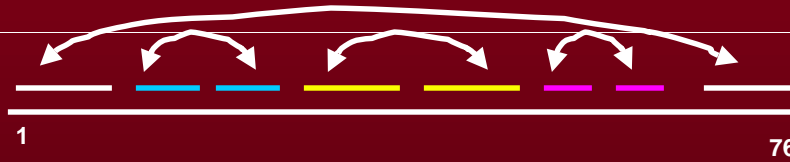
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Tertiary Structure



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RNA folding is usually nested



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RNA can have pseudoknots

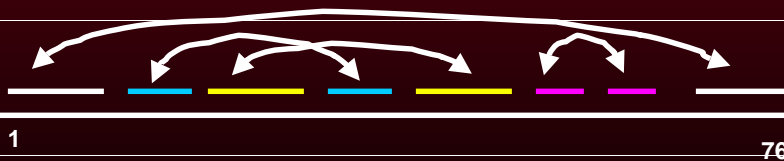
Normal base pairing (for base pairs i - j and i' - j'):

$$i < j < i' < j' \text{ or } i < i' < j' < j$$

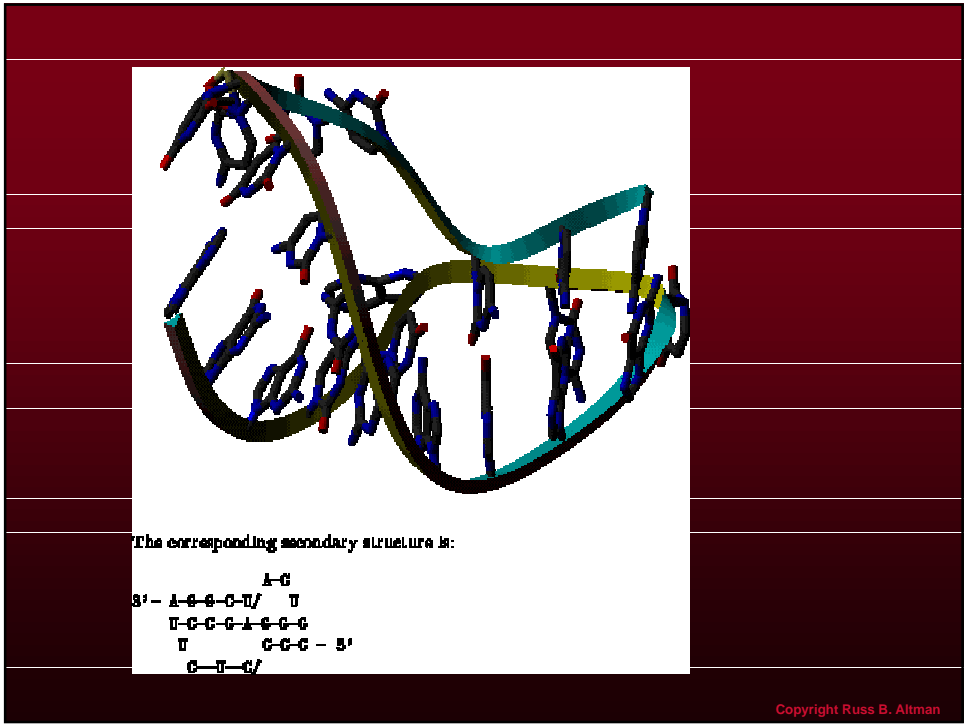
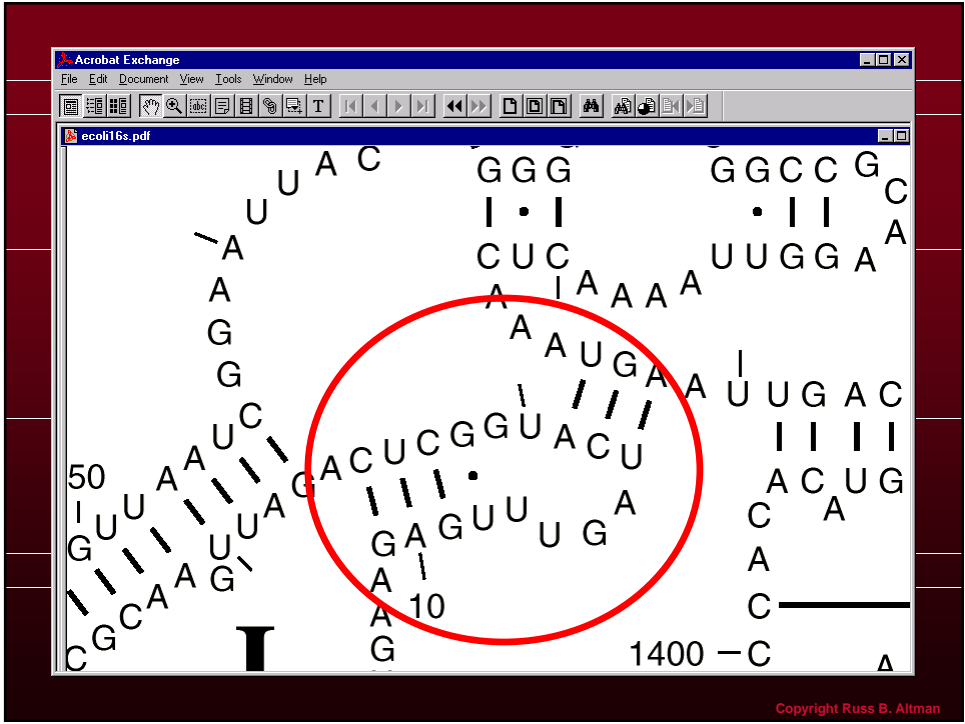
Pseudoknots

$$i < i' < j < j'$$

Yellow and blue helices form a pseudoknot.

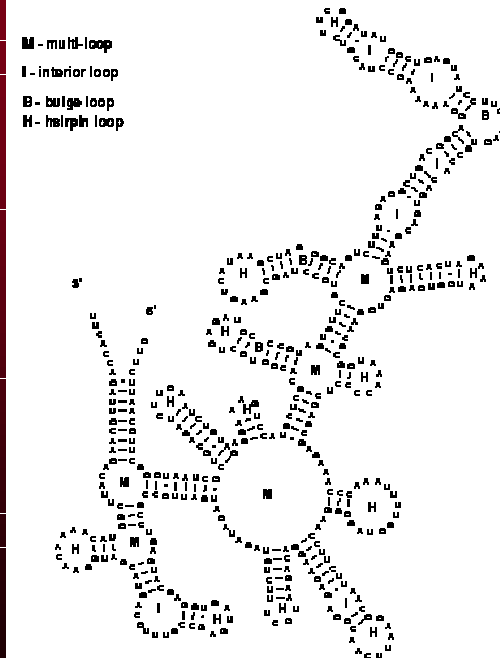


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Bacillus subtilis RNase P RNA

- M - multi-loop
- I - interior loop
- B - bulge loop
- H - hairpin loop



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Show sample RNA folds

Show sample RNA structures

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Predicting RNA Secondary Structure

1. Based on phylogenetic comparisons

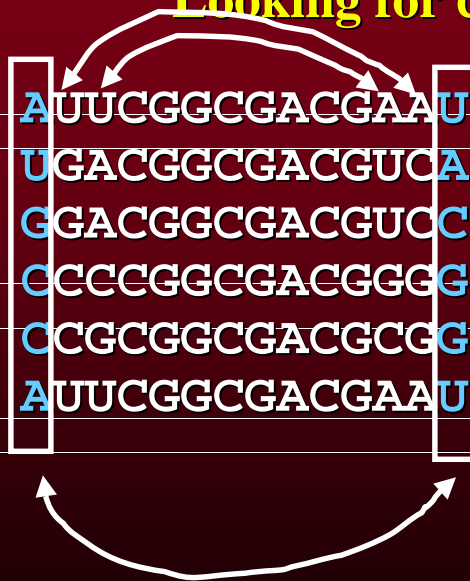
- look at columns of a multiple alignment
- find WC or nonCanonical relationships that are preserved despite mutations.

2. Based on energetics

- individual contributions to energy are additive
- assumption that optimal secondary structure is that which has lowest energy.

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Looking for covariation



The consistent covariation of the two columns in a Watson-Crick manner indicates that there is some sort of relationship between those two positions in the secondary structure.

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Covariation is taken as a gold standard for other prediction methods.

AUUCGGCGACGAAU

Primary Structure



**AUUCG G G
| | | | C
UAAGC A**

Secondary Structure

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RNA Folding energetics

Based on the observation that the stability of an RNA fold can be decomposed into the contributions of individual energies.

Favorable contributions from:

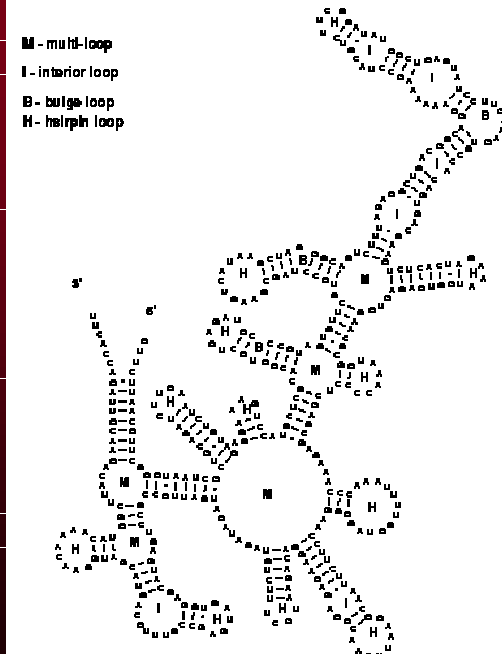


1. Hydrogen bonds of base-pairs
2. Favorable “stacking” interactions of bases (parallel planes of rings)
3. Some Ad hoc basepairs created in irregular structures (such as loops of 4 = tetraloops)

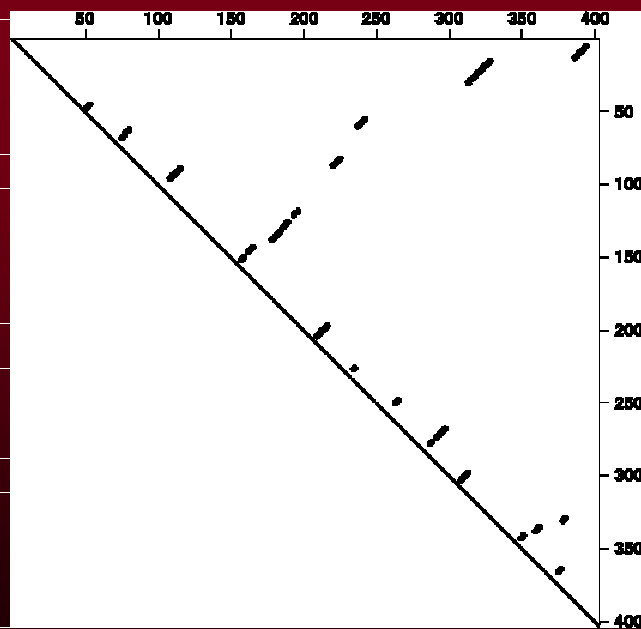
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Bacillus subtilis RNase P RNA

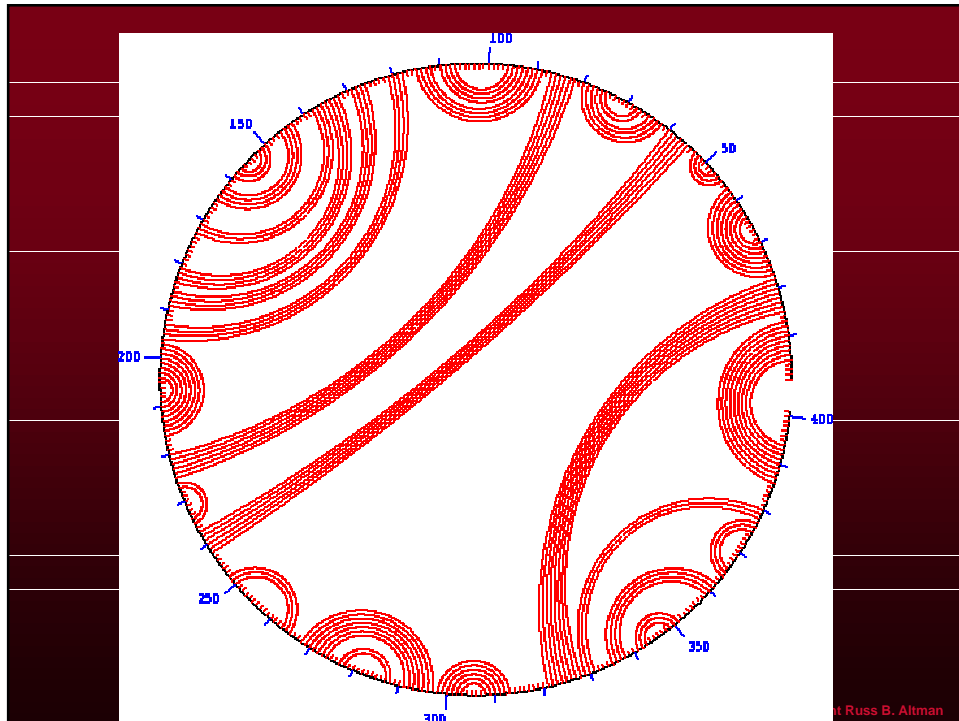
- M - multi-loop
- I - interior loop
- B - bulge loop
- H - hairpin loop



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Simple energies

G -- C = -3 (most favorable Watson-Crick)

A -- U = -2

G -- U = -1 (note: non-WC = non-canonical)

more elaborate schemes developed by Turner et al.

(<http://www.ibc.wustl.edu/~zucker/cgi-bin/efiles.cgi?T=37#STACK>)

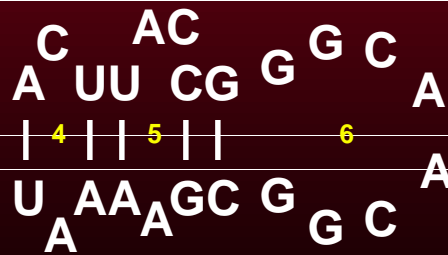
		A	C	G	U	
5' --> 3'		-----				
UX		A	0.4	0.4	0.4	-1.1
AY	→	C	0.4	0.4	-2.3	0.4
3' <-- 5'		G	0.4	-1.8	0.4	-0.8
		U	-0.9	0.4	-1.1	0.4

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RNA Folding Energies

Unfavorable contributions from:

4. Symmetric bulges in helices
5. Asymmetric bulges in helices
6. Increasing size of loop at end of helix
7. Multiple branches from a single loop (next slide)



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RNA Folding Algorithm

Use dynamic programming approach.

See Michael Zuker's home page for comprehensive description of state-of-the-art algorithms.

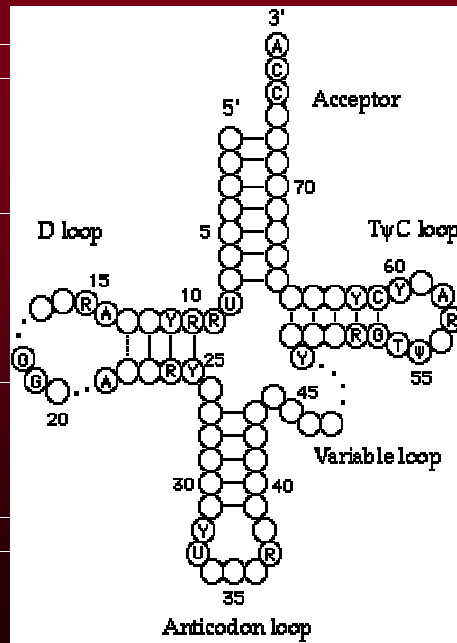
<http://alice.wustl.edu/~zucker/Bio-5495/RNAfold-html/> = OLD

<http://www.ibc.wustl.edu/~zucker/rna/>

Recursions available at:

<http://www.ibc.wustl.edu/~zucker/Bio-5495/RNAfold-html/node3.html>

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The MFOLD Program

<http://www.ibc.wustl.edu/~zucker/rna/form1.shtml>

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