FCModeler: Dynamic Graph Display and Fuzzy Modeling of Regulatory and Metabolic Maps

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Gamble.

FCModeler Goals

Capture the intuitions of biologists and provide a modeling framework for assessing large amounts of information

Test effects of hypotheses

Use graph theoretic approaches to analyze network structure and behavior and search for critical paths in the network

Develop simulation tools based on fuzzy methods that model changes in the network

Key Components of FCModeler

Dynamic graph visualization package written in Java

Network analysis using graph theoretic methods

Network modeling using fuzzy cognitive maps

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Software Structure

METNET Database Stores node and interaction information MySQL

FCModeler Dynamic Graph Visualization Graph Theoretic Structure Analysis Java-based XML Graph Files

Network Modeling Software FCM implementations Currently in Matlab, will become Javabased

Current Efforts

FCModeler Development

- Incorporation of different levels of information such as functional links
- Finding critical points in interacting pathways
- Network analysis and parallel paths between nodes
- Visualization of graphs in 3D virtual reality systems

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For more information see: www.eng.iastate.edu/~julied/research

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Dynamic Graph Visualization

- Display graph data using different layouts
- Select subgraphs for further analysis
 Visualize simulation results

Graph Layout, Dot

dot from *Graphviz* package -AT&T research labs (http://www.research.att.c om/sw/tools/graphviz/) Developed for noncyclic graphs. Good for hierarchical data, handles large graphs well



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 Elements can separated into different levels based on cellular location, type, etc
 More intuitive

Slower layout, doesn't always converge



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GEM Layout

 Spring Embedder based layout,
 Places attractive gravitational force at the barycenter of the layout.

Nodes are attracted to adjacent nodes to it and repelled by other node



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Edge and Node Mappings

- User settable mappings for edges and nodes
 - Color
 - Width
 - Arrowhead type
 - Node shape
- User can move nodes and edges and store the coordinates for future reference



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Subgraph Selection

 Selectable subsets for further analysis based on vertices, cycles, etc

😤 Subgraph Algorithm Selection

-algorithms

- all edges connecting selected vertices
- Closed p-th neighborhood of selected vertices
- O all vertices incident with a selected edge
- 🔘 all selected vertices, & all selected edges with both end-vertices selected

ok 📔 cancel

🗛 FCModeler - reducedfasncycles.xml file view graph properties ڬ 🖽 🔍 <> ncoal rxr nr3c1 mapk8 rar jun vip-g vip vipr1 adcv prka

mapkl

X

mapk3

rela

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Dynamic Graph Visualization

- ◆ Java [™] interface that reads and displays data from a mySQL (www.mysql.com) database of links and nodes.
- Based on DIVA package for node/edge features www-cad.eecs.berkeley.edu/diva/index.html
- Uses Java 1.4 or higher

Graph Theoretic Analysis

Strongly Connected Components Cycle Analysis and Self Organizing Maps

Alternate Path Analysis

(SCC) Strongly Connected Components



 Every node is reachable from every other node in the same SCC.
 In metabolic pathway graphs, the SCC's show which components have bi-directional paths between them.

 Only nodes in an SCC can be in cycles

Cycle Search

- Elementary circuits, or cycles, are a series of nodes that start and end in the same place.
- Preliminary results have found existing pathways in the map as well as new relationships and oversights in the models.
- Cycles in a graph show how one node can affect another



Cycle Similarity Metrics

Many overlapping cycles, need to be grouped Sample Metrics Common Elements $d(A,B) = \frac{card(A \cap B)}{card(A \cup B)}$ • Overlap (subsethood) $S(A,B) = \frac{card(A \cap B)}{card(B)}; \quad S(B,A) = \frac{card(A \cap B)}{card(A)}$ Edit Distance



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Cycle Similarity Analysis

👹 cycles

file filter

8-cycle: [prka, e41, nr3c1, e93, mapk8, e424, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka] 9-cycle: [mapk3, e24, nr3c1, e93, mapk8, e424, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka, e111, mapk3] 10-cycle: [mapk3, e98, ncoa1, e37, nr3c1, e93, mapk8, e424, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka, e111, mapk3] 10-cycle: [rar, e440, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka, e111, mapk3, e98, ncoa1, e103, nxr, e334, rar] 10-cycle: [rar, e440, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka, e109, mapk1, e97, ncoa1, e103, nxr, e334, rar] 10-cycle: [rar, e440, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka, e109, mapk1, e97, ncoa1, e103, nxr, e334, rar] 10-cycle: [mapk1, e97, ncoa1, e37, nr3c1, e93, mapk8, e424, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka, e109, mapk1] 9-cycle: [rela, e43, nr3c1, e93, mapk8, e424, jun, e409, vip-g, e278, vip, e279, vipr1, e180, adcy, e76, prka, e112, rela]

- U X

Common Element Similarity Metric							
Cycle	1	2	3	4	5	6	7
Number							
1	0	0.89	0.80	0.50	0.50	0.80	0.89
2	0.89	0	0.90	0.58	0.46	0.73	0.80
3	0.80	0.90	0	0.67	0.54	0.82	0.73
4	0.50	0.58	0.67	0	0.82	0.54	0.46
5	0.50	0.46	0.54	0.82	0	0.67	0.46
6	0.80	0.73	0.82	0.54	0.67	0	0.73
7	0.89	0.80	0.73	0.46	0.46	0.73	0

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Clustering Cycles

Each graph shows the model of the corresponding map unit, which is the generalized median of the cycles assigned to that map unit.

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Results of cycle clustering

- Similar cycles with overlap are found
- Shows parallel paths in the data structure



Alternate Paths

- Often partial information is known (e.g. *a* inhibits *d*)
 Checks for other paths that go from *a* to *d* and checks overall effect
 - a->b->c->d has a net inhibitory effect
- Helps check for consistency (e.g. b activates d), alternate paths only inhibit d so information is missing



Metabolic Network Modeling

Metabolic networks model net accumulation of biomolecules in organisms.

Regulatory networks modulate the action of metabolic networks

Nodes represent specific biochemicals such as proteins, RNA, and small molecules, or stimuli, such as light, heat, or nutrients.

Links show interactions between nodes

Link Types

- Conversion link (black arrow), a node is converted into another node, and used up in the process.
 Regulatory link (green and red arrows), node
 - activates or deactivates another node, not used up.
- Catalytic link (blue arrows) an enzyme that enables a chemical conversion and not used up in the process.



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Regulatory Link Models

- Edge weights {-1/N,0,1/N}, where N is the number of nodes going into the link
- As more information becomes known about the underlying biology, the functional link models will be updated.
- Regulatory input nodes have selffeedback to keep the nodes on until they are inhibited.



RNA

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Conversion Link Models

For causal relationships between nodes the link is modeled the same as a regulatory link

- When information about the rate of change in a reaction is available:
 - Simple difference equations can model the gradually rising and falling levels of the nodes
 - Step size depends on the reaction rate and the relationship between the nodes

Catalyzing Link Models

- Catalyzed reactions add a dummy node that acts upon a conversion link. This allows one link to modify another link.
- Another method is an augmented edge matrix that operates on the edges between the nodes.



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Fuzzy Cognitive Maps

- Fuzzy Cognitive Maps show interactions between different variables
 - Fuzzy signed digraphs represent causal flow between objects or concepts
 - Constructed using expert knowledge or neural learning



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Gibberellin Example

 Create metabolic map of system using expert knowledge
 Model the activation and regulation of Gibberellin (GA)
 Metabolites are held constant until inhibited by Gibberellin

Assumptions

- RNA assumed active until inhibited
- Precursor small molecule IPP present
- Metabolites and other factors active unless inhibited



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FCM of Gibberellin Activation

Model shows the regulation and homeostatic control of Gibberellin(GA):

- Regulates its own production
- Controls the activity of other biomolecules

Oscillation of GA levels directs the generation of biomolecules that are implicated in the formation of new cellular proliferation centers, (meristems)

Many key features of this model, including timing, can be tested experimentally by globally monitoring temporal profiles of mRNA, protein, and metabolite

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Metabolic Networks

- Metabolic networks form the basis for the net accumulation of biomolecules in organisms.
- Regulatory networks modulate the action of metabolic networks, leading to physiological and morphological changes.
- Modeling tool represents the interactions within and between these networks
 - Nodes represent specific biochemicals such as proteins, RNA, and small molecules, or stimuli, such as light, heat, or nutrients.
 - Links show interactions between nodes

Simple Fuzzy Cognitive Maps

- Graph edges are {-1,0,1}. Used when the direction of causality is agreed on, but not its degree
- Concepts either occur or do not occur
- Can test out hypotheses
- Concepts are usually summed then thresholded to get the next state

$$\mathbf{C}(t_{n+1}) = S[\mathbf{C}(t_n)\mathbf{E}]$$



Nested FCMs

- If more information is known about the links between concepts, then more detailed functional links can be used to combine information
 - Differential equations
 - Fuzzy approximators



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