

Graph-oriented modelling of multiscaled dynamical systems with a dynamical structure:

Challenges to the "Relational Growth Grammar" approach

Winfried Kurth

Brandenburg University of Technology at Cottbus, Chair for Practical Computer Science / Graphics Systems

Aspects of biological and/or chemical systems which are not yet well represented:

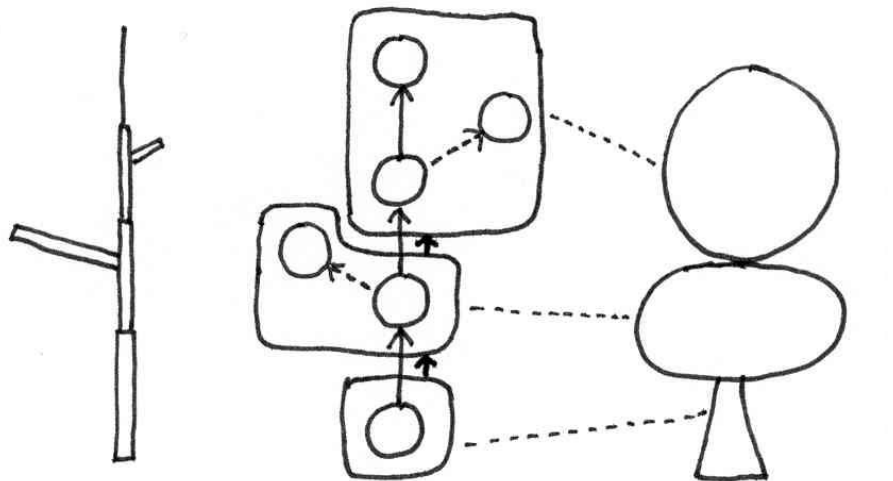
1. Multiscaled structures

(\cong different views on the same structure)

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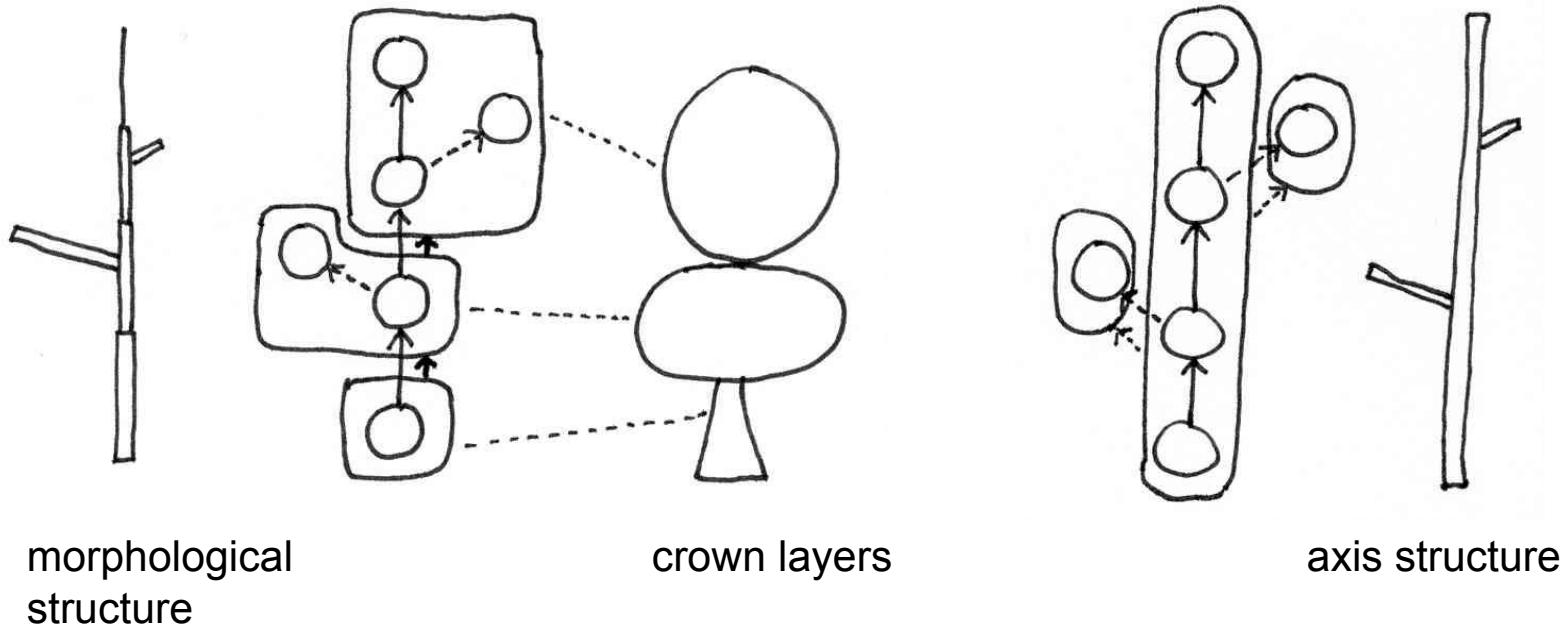
morphological
structure

crown layers

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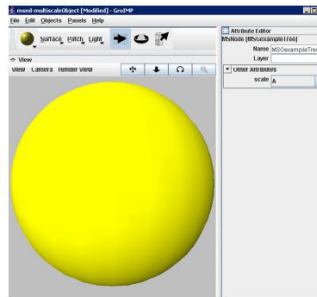
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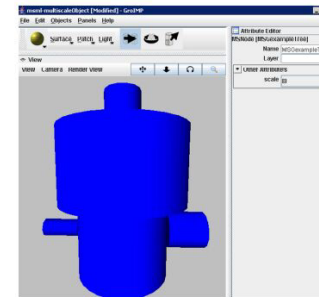
"Multiscaled Tree Graphs" (MTG):

Theoretical basis given by Godin & Caraglio 1998.

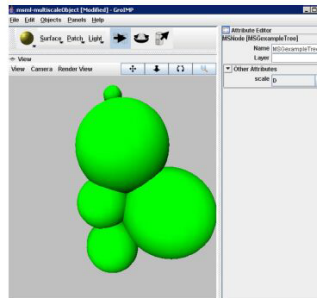
first attempt to integrate this in GroIMP: Diploma thesis by Sören Schneider (2006)



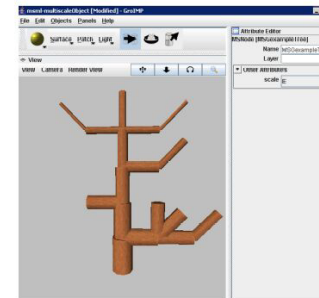
(a) Skale A



(b) Skale B



(c) Skale D



(d) Skale E

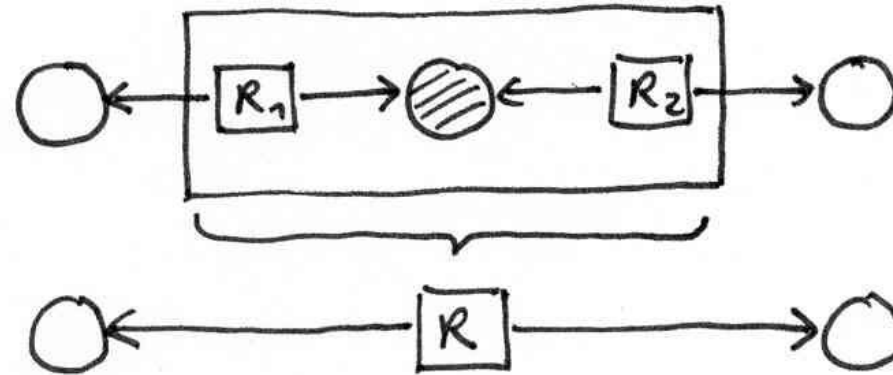
- including new XML-based data format (MSML)

- but not yet satisfactorily finished and not yet well supported

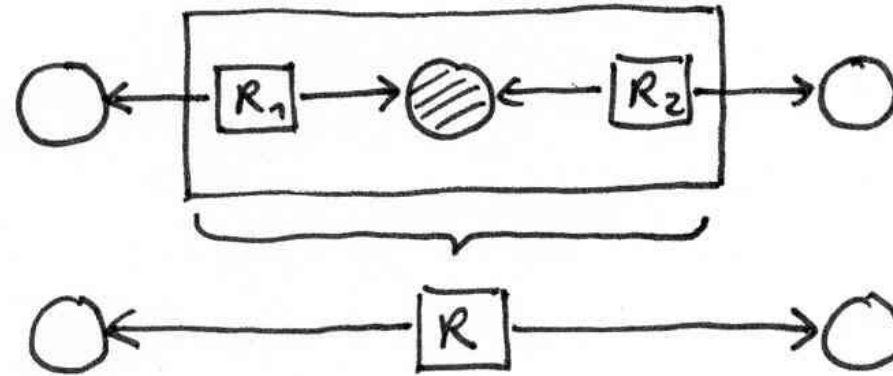
To be done:

- Extension of existing representation to all sorts of modules,
 - particularly: 3D representation of structures from MTG files
 - operations on aggregate modules (*e.g., complete deletion in case of a collision*)
 - local adaptation of resolution (zoom-in, zoom-out)
-
- solution of the runtime problems of the current MTG filter of GroIMP (XSLT is too slow)

multiscaled structures are common in biochemical networks:
different resolutions of reaction chains



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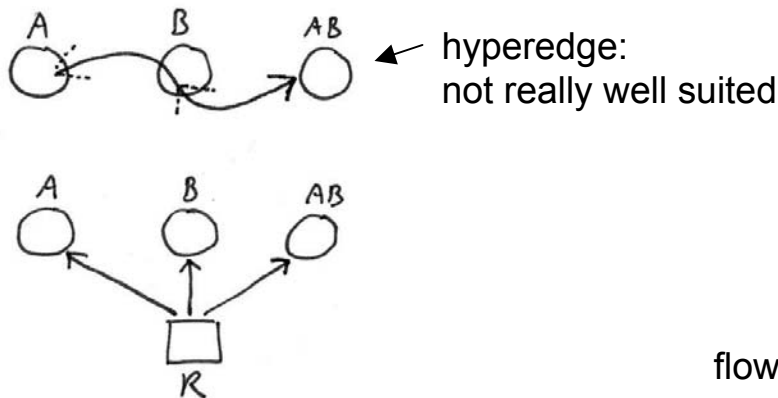
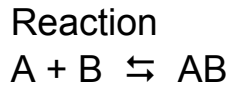


Remark:

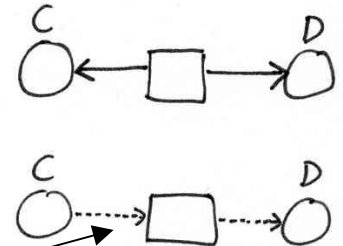
Representation of metabolic networks favorably with Petri nets.

Nodes: Substrate nodes and reaction nodes

Example:



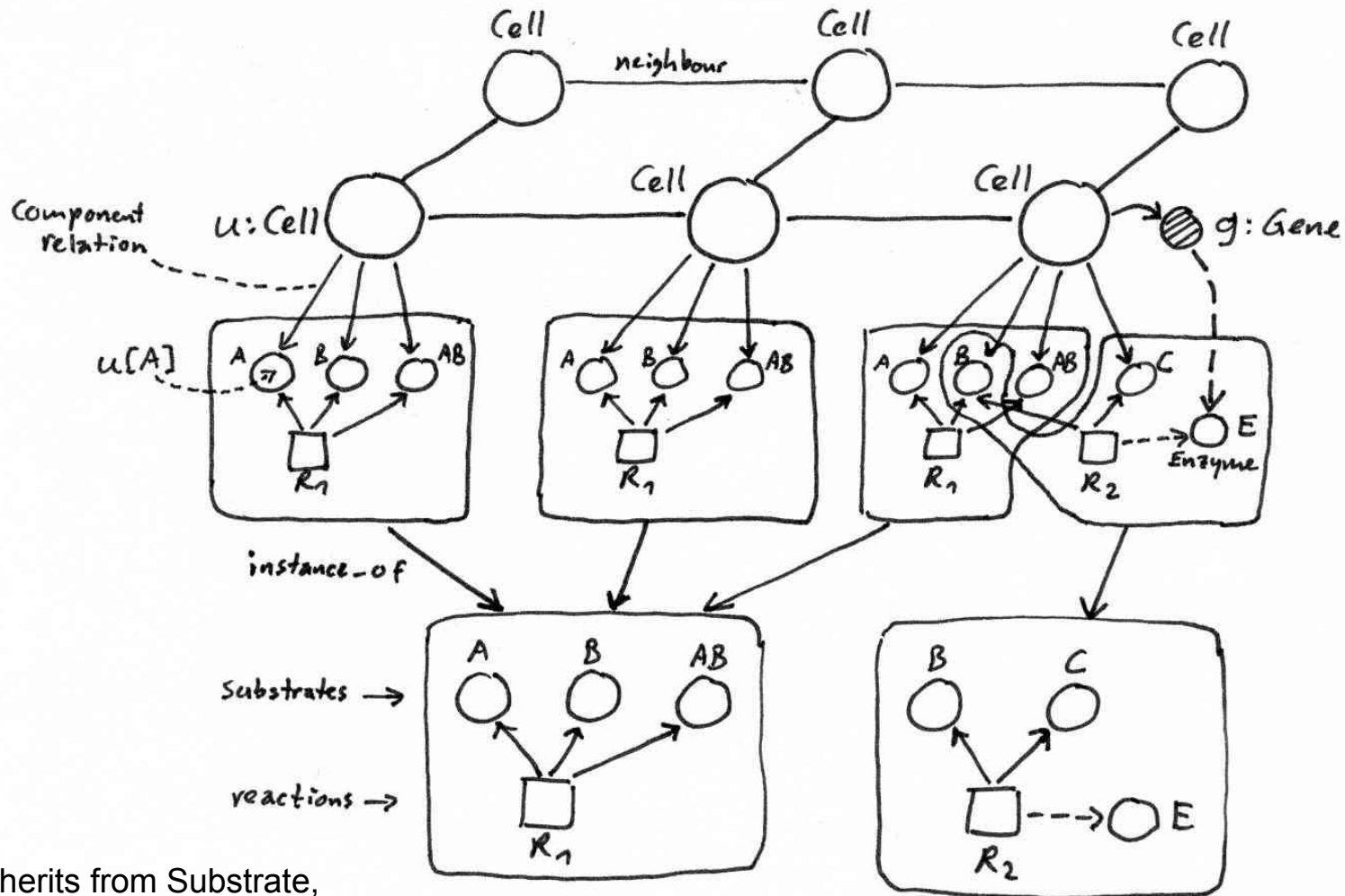
$C \rightarrow D$:



flow of matter

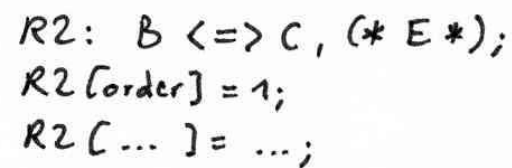
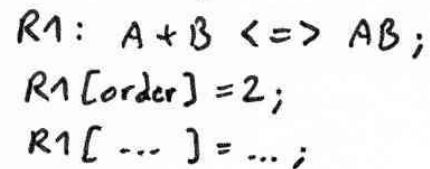


2. Instantiations of metabolic networks



A inherits from Substrate,

R_1 inherits from Reaction



3. Easy specification of sensitivity

e.g., collision detection:

A ==> F [RU(45) F B] F A;

this rule shall be applied only when none of the 3 Fs hits an existing object in space

suggestion:

A ==> b: (F [RU(45) F B] F A); /* trial */

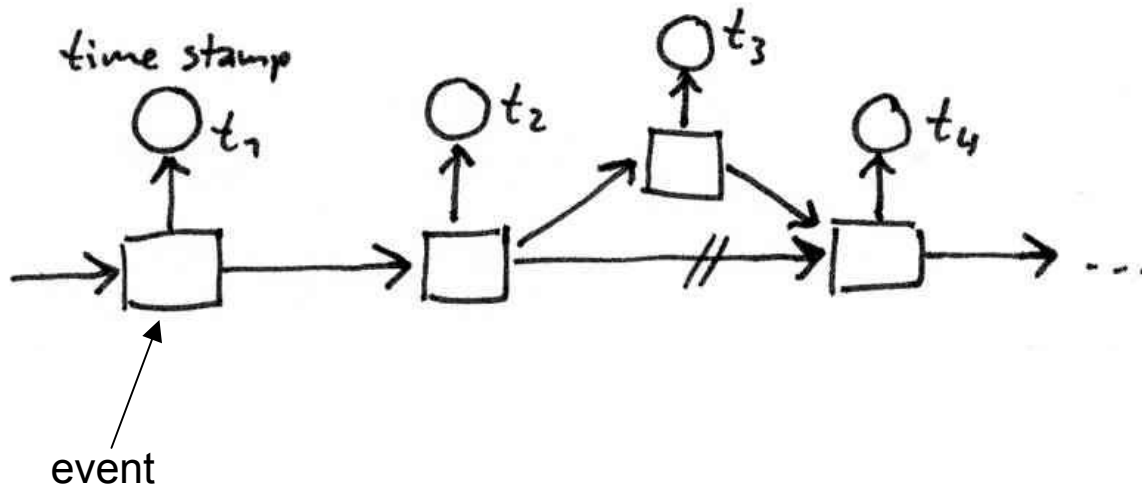
b:Compound (b.collission) ==> ; /* deletion */

4. Event queues

- or, more generally:
a more flexible time management

RGGs use discrete time steps of fixed length

to be preferred for realistic models: continuous time (faked),
event handling



- timed L-systems (Prusinkiewicz & Lindenmayer)
- differential L-systems (DOL-systems) (Prusinkiewicz et al.)
- scheduling mechanisms in SIMULA and other languages

to be developed:

a concept for general representation of **dynamics** in XL

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particularly:

5. Dynamics governed by differential equations

(chemical kinetics is just a special case)

Example: Xylem sap flow model HYDRA
(based on differential eq.).
Structure has impact on function

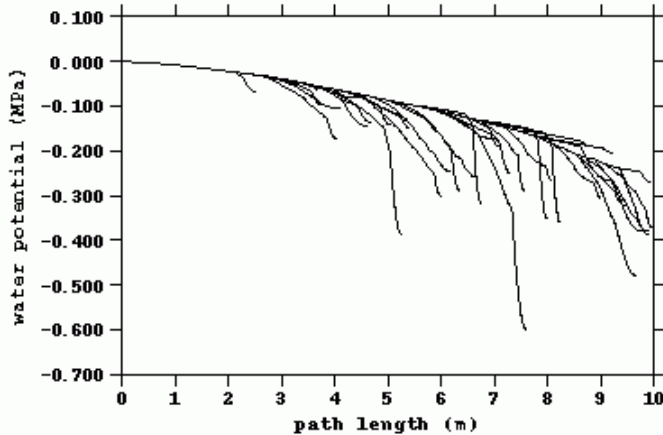
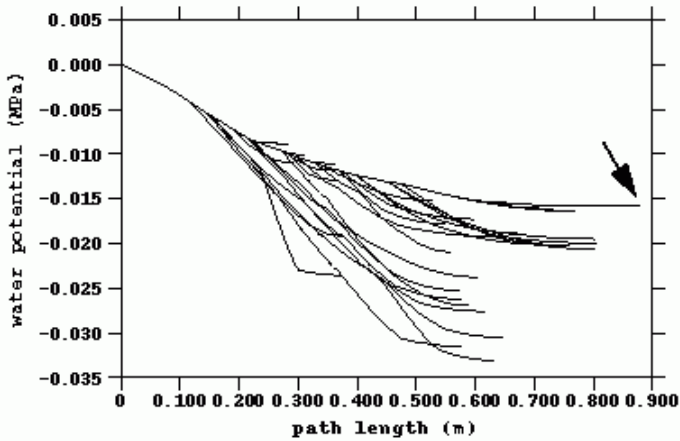
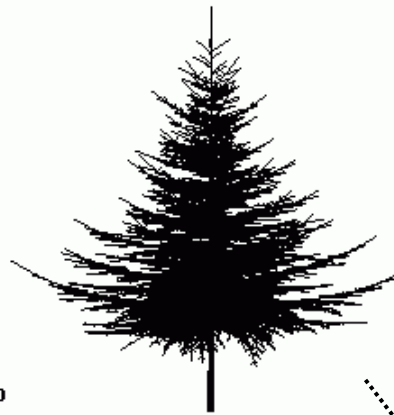
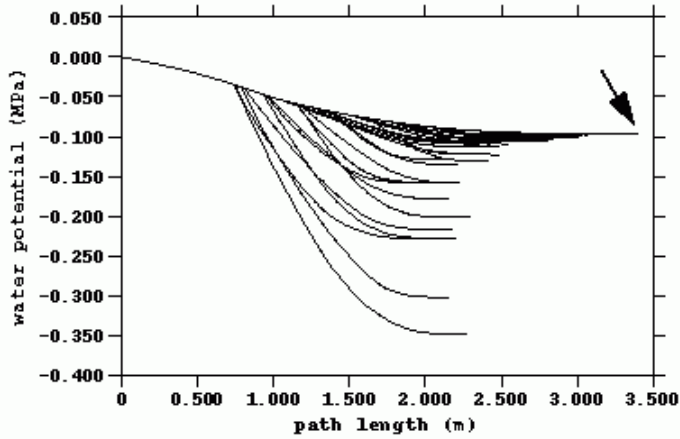
(Früh & K. 1999)

Spruce (L-system model)

Spruce (3D measurement)

Thuja (3D measurement)

HYDRA until now separated
from GROGRA / GroIMP

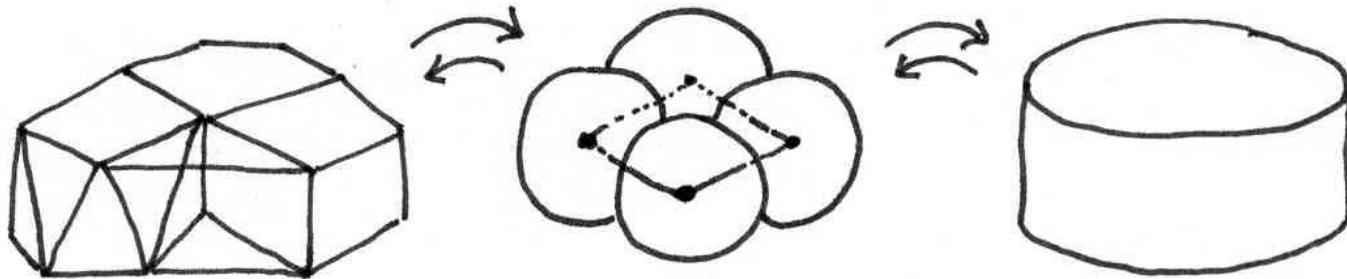


6. Structure and dynamics of 3D cell assemblies

RGGs are (until now) most often used to produce *tree-like structures*

more general 3D networks and structures should be possible even with the current release of XL

a question: how important is the exact geometry of 2D and 1D boundaries?



7. Rules transforming rules

why this?

Applications:

- Possibility to evolve rules by mutations, again described by rules
- metabolic and regulatory networks define rules for substrate concentrations and/or gene activation
- machine learning
(to learn means: to change rules, according to meta-rules for learning...)

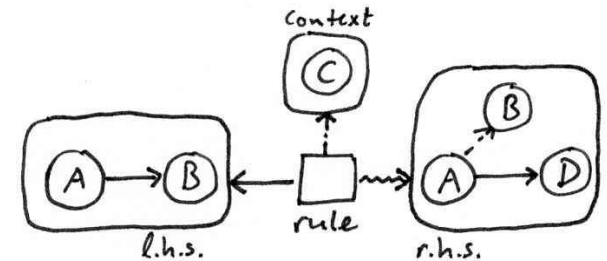
probably necessary for this purpose:

various unifications

particularly for the concepts of "edge" or "relation":

- graph edges $-0->$
- deduced relations $(>)*$
- component relation $\mathbf{a[x]}$
- rule arrows and context delimiters

e.g., rule $A \rightarrow B, (* C *) \implies A [B] D$:



- term-building relations

e.g. term $(x > 2*y)$:

