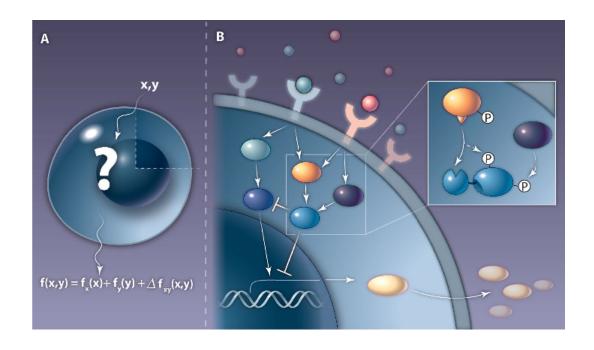
Atomizing BioModels

Transforming reaction-network models into rule-based models

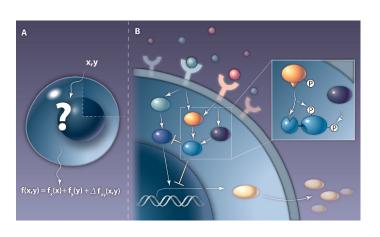
Jose-Juan Tapia James R. Faeder

Rule-based models are ...

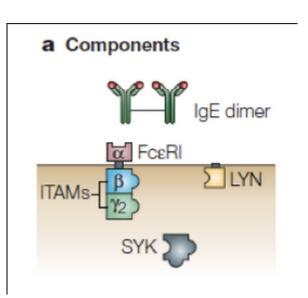


W. S. Hlavacek and J. R. Faeder, Sci. Signal. 2, pe46 (2009).

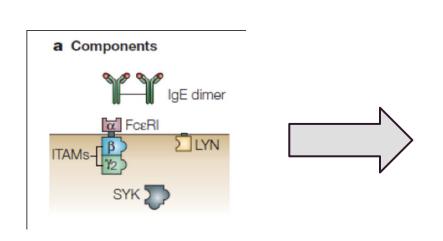
Mental model

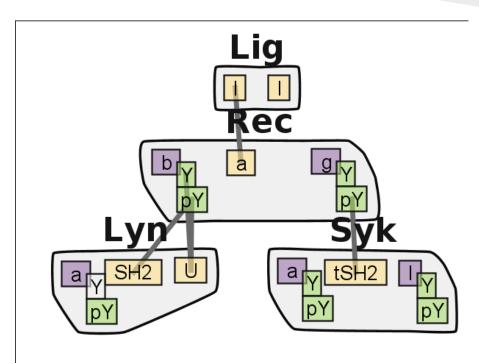






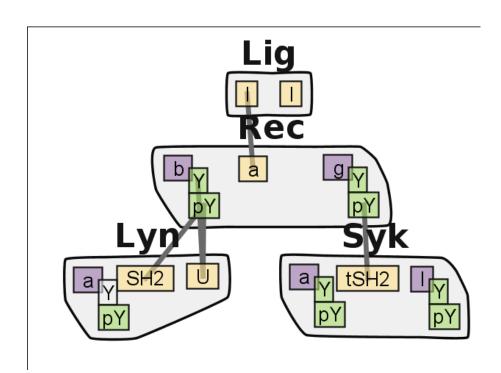
Rule Based Modeling (RBM)





Contact map

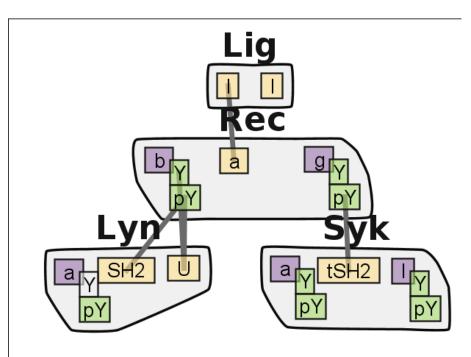
Terminology



Contact map

- Molecules: Indivisible entities that associate with other indivisible entities
- Components:
 Molecule's functional attributes
- Species: Unique configuration of one or more molecules

Terminology (2)



- Reaction:
 Transformation(s)
 applied to one or more species
- Rule: Compressed representation of a set of reactions. Uses patterns.

Reaction Network Model(RNM): reactions Rule-Based Model(RBM): rules

Rule-based modeling

Pros:

- Rich syntax that allows the modeler to encode structural and contextual information
- Very scalable

Cons:

- Syntax may be overkill for smaller models or phenomenological models.

Reaction Network Modeling

$$\begin{split} \mathrm{d} \mathrm{S}/\mathrm{d} t &= \sigma \mathrm{R}(t) - ac \mathrm{S}(t) \mathrm{I}(t) - \mu \mathrm{S}(t) \\ \mathrm{d} \mathrm{I}/\mathrm{d} t &= ac \mathrm{S}(t) \mathrm{I}(t) - \rho \mathrm{I}(t) - \delta \mathrm{I}(t) \\ \mathrm{d} \mathrm{R}/\mathrm{d} t &= \rho \mathrm{I}(t) - \sigma \mathrm{R}(t) - \mu \mathrm{R}(t) \\ \mathrm{d} \mathrm{D}/\mathrm{d} t &= \mu \mathrm{S}(t) + \delta \mathrm{I}(t) + \mu \mathrm{R}(t) \end{split}$$

RNM representation

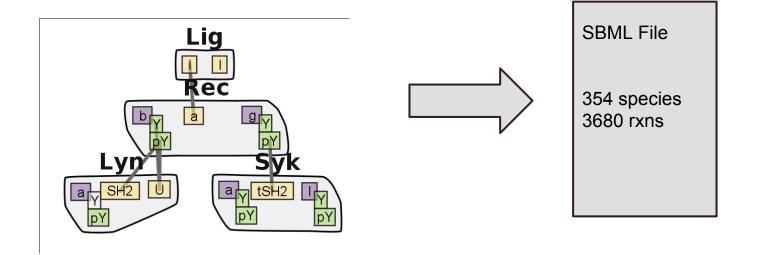
SBML

BNGL

$$A(b) + B(a) -> A(b!1).B(a!1)$$

RNM is adequate for small models.

Network expansion



19 rules 4 molecules

This phenomenon is known as combinatorial complexity

Reaction Network Modeling (2)

Pros:

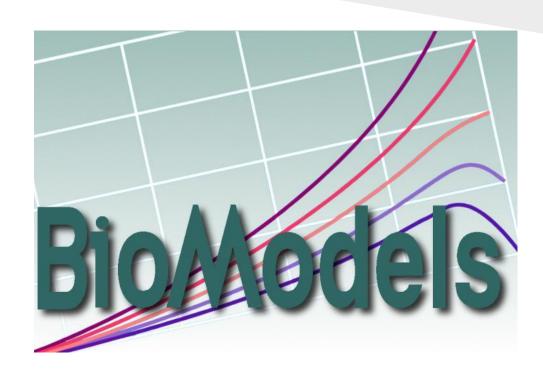
- Well understood theory of differential equations based chemical kinetics
- Simple representation suitable for smaller models
- Suitable for smaller models

Cons:

- Biological structural and contextual information is lost
- Non scalable

BioModels

There exist a large body of knowledge already encoded in SBML.



Our goals are

We could deeply understand and build upon years of RNM modeling knowledge in a semi-automated way.

Find a way to recover structural and contextual information that is no longer explicit in RNM models.

Make the benefits of RBM accessible to a larger public.

Presenting...

RBM



Atomizer

How does it work?

Atomizing in a nutshell Identify what every reaction does

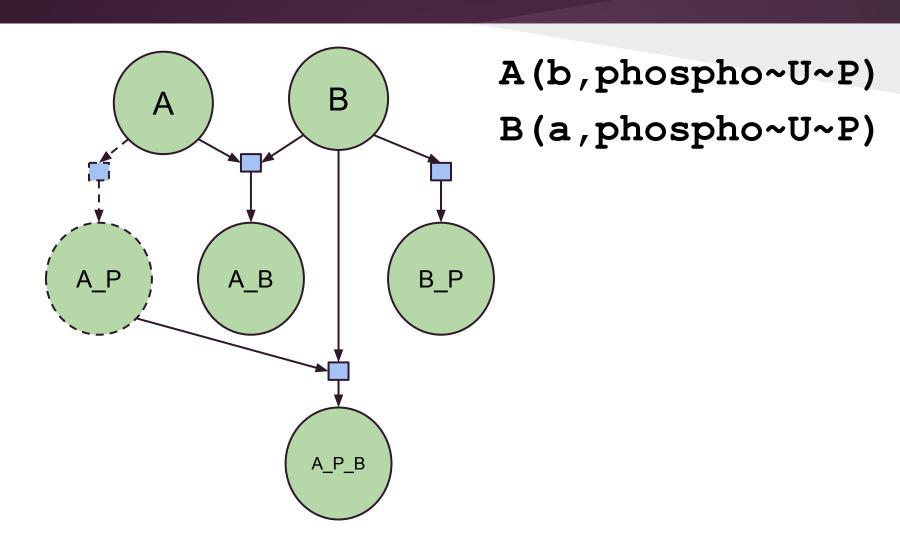
Reaction stoichiometry information

Naming conventions

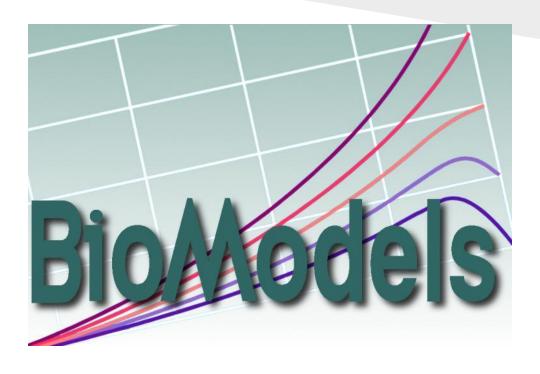
Lexical analysis

Complexation and phspaiyuiyiuyiu

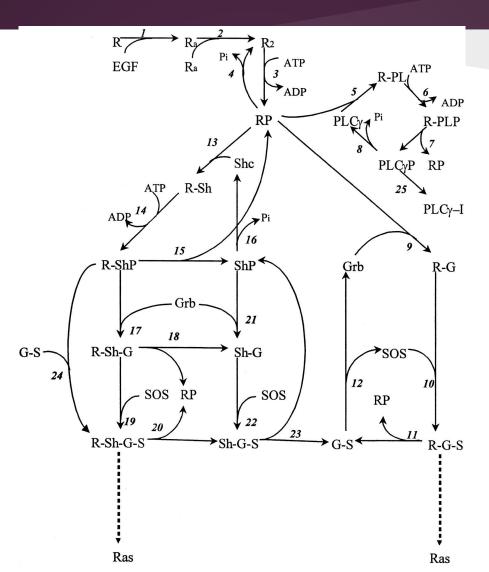
Composition Graph



Atomize BioModels

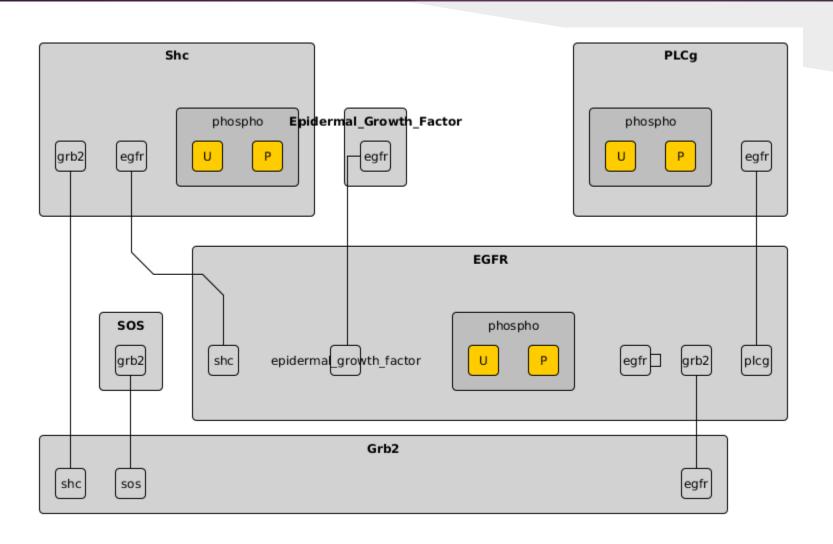


Epidermal Growth Factor

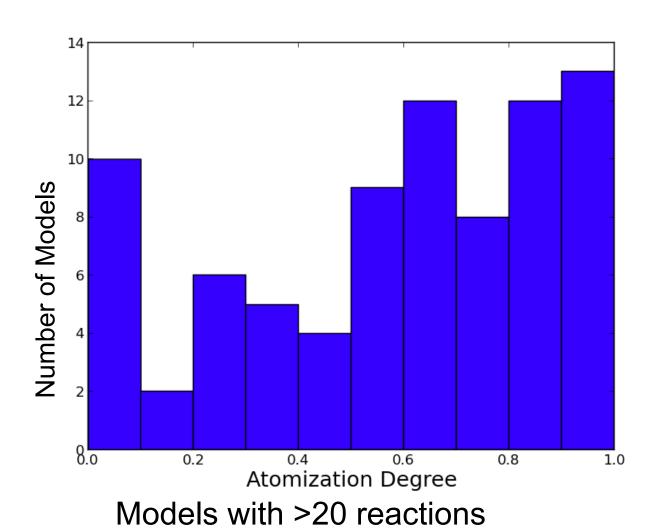


Kholodenko 1999

BioModels 48 atomized



Atomization degree = structured / total molecules



Model fusion

Model aggregation: a building-block approach to creating large macromolecular regulatory networks

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"Process that combines two or more submodels into a single unified model that contains the combined information (without redundancies) across the original collection."

How do I map species?

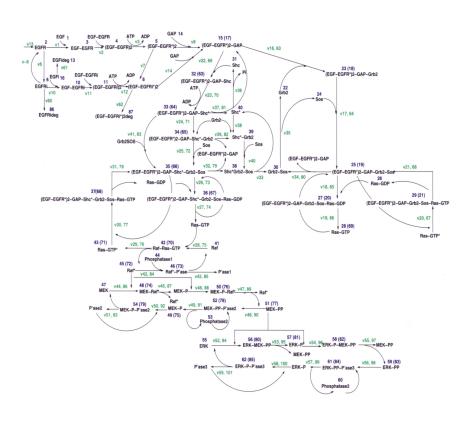


Image taken from Schoeberl et. al.

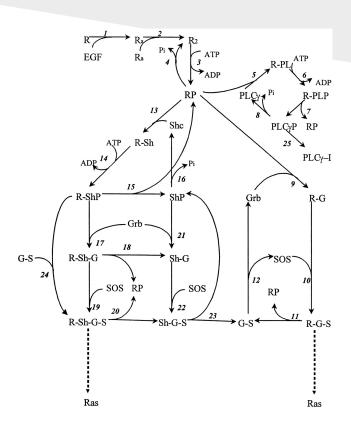
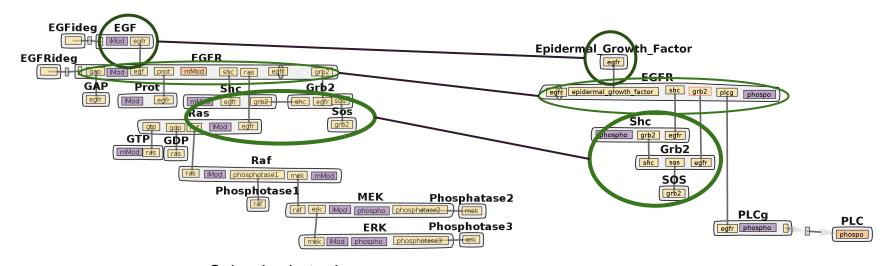


Image taken from Kholodenko et. al.

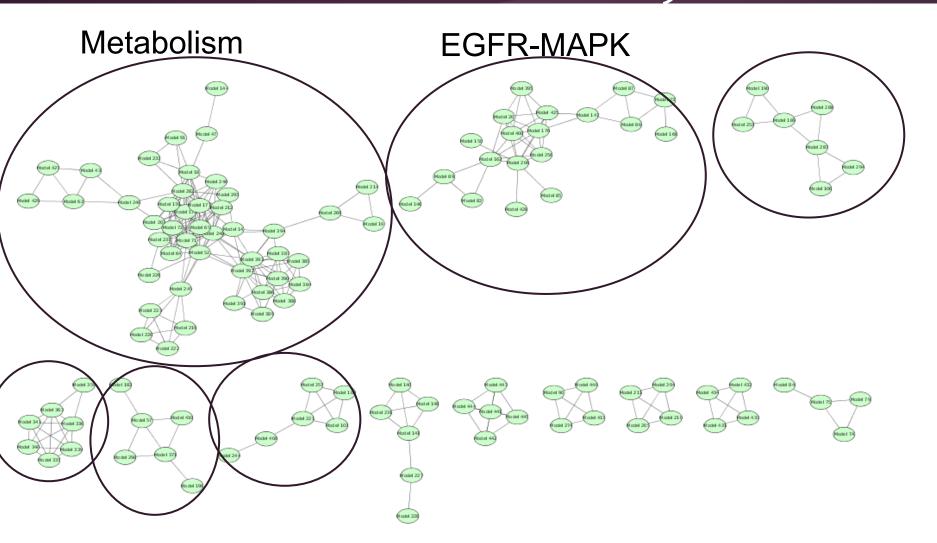
RBM Model fusion (2)



Schoeberl et. al.

Annotation information is used to reliably map species with different names across models

Model interaction(each node is a BioModels model)



Naming convention analysis

Patterns	Associated process	Example
[`+P',`+p',`+P']	Phosporylation	$x \to xP$
$[' + PP', ' + _PP']$	Double phosporylation	$xP \rightarrow xPP$
['+i']	Internalization	$\mathtt{x}->\mathtt{xi}$
$[`-_n',`+_c']$	Compartment transfer from nucleus to cytoplasm	x_n, x_c
$['+_ubiq']$	Ubiquitination	$\mathtt{x-} > \mathtt{x_ubiq}$
['K','KK']	Kinase, Kinase kinase	MAP, MAPK, MAPKK
['+H']	Adding a hydrogen-related modification	NAD, NADH
['+R']	Receptor	EGF, EGFR
['+c']	[Cyclic version, cytoplasm, casp3 substrate]	x, cx
['+2', '+3', '+4']	[Dimer, Trimer, Tetramer] - [Protein family]	x o x2

Naming convention analysis

+ P

27.4669509595

+ p

21.5778251599

- T+ D

9.0618336887

+ 2

7.4669509595

+ a

6.908315565

- D+ T

6.7356076759

- P+ M

5.5991471215

- n+ c

4.8614072495

Metric is the product of the number of times an annotation appears across the database multiplied by the percentage of models it appears in

Model verification: BioModels 109

Model annotation tells us that...

Cdk1Y10:

• Cyclin-dependent kinase

Cdk1Y11:

- Cyclin-dependent kinase
- Cyclin A

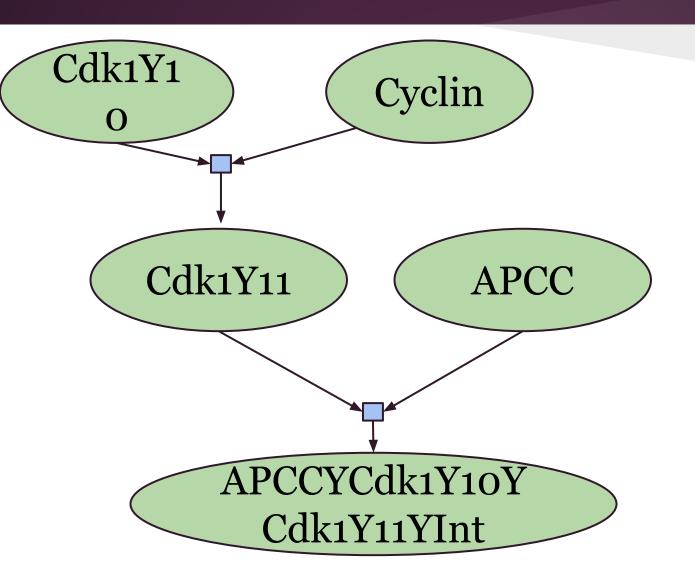
APCC:

ubiquitin ligase

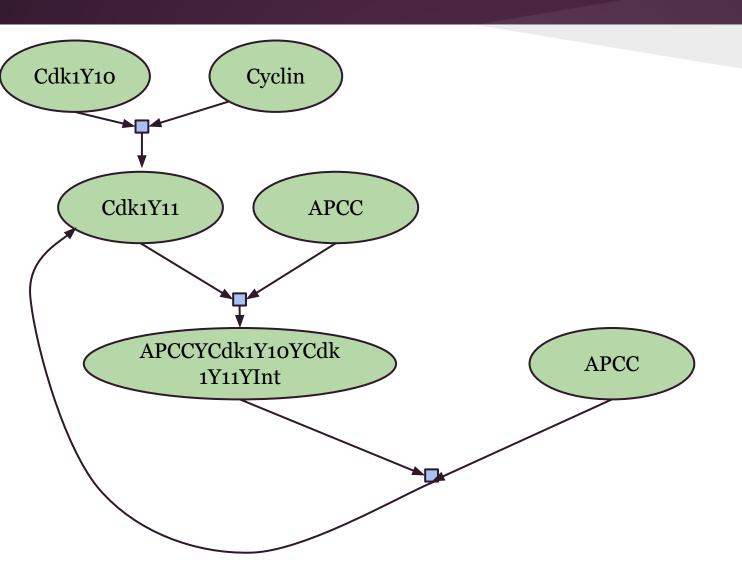
APCCYCdk1Y10YCdk1Y11YInt:

- Cyclin-dependent kinase
- ubiquitin ligase
- Cyclin A

Model composition (BM 109)

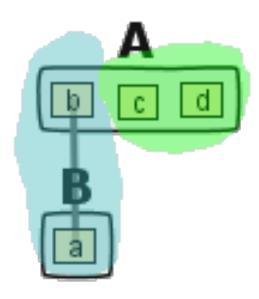


And then...??



Reaction center and context

$$A(b,c,d) + B(a) -> A(b!1,c,d).B(a!1)$$



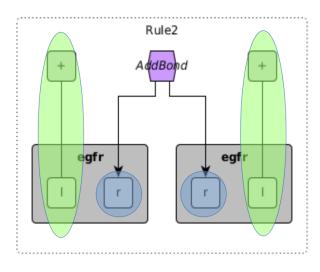
SBML Molecules contain minimal context information

$$A + B < -> C$$

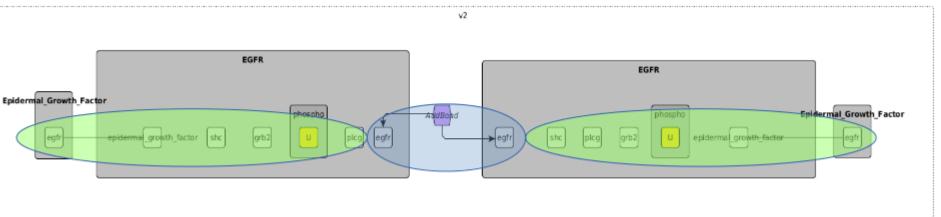
What is A? (Single molecule, umbrella name for a series of compound series of A molecules, etc).

Impossible to know without extensive annotation information. So we have to take them at face value.

Kholodenko's reaction context

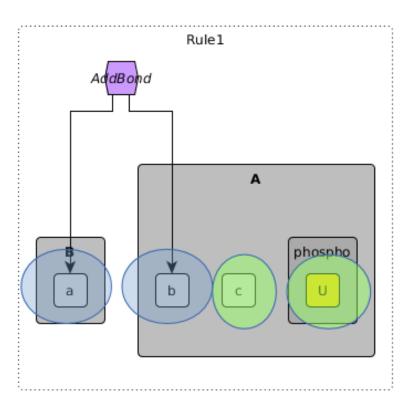


Manually constructed RBM version of Kholenko's model.

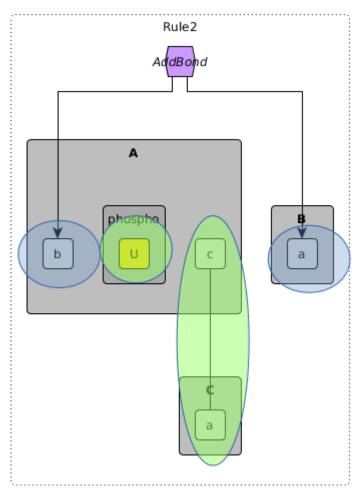


Automatically translated version of Kholodenko's model

Redundancy is the redundant way to redundantly go redundantly

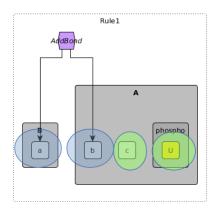


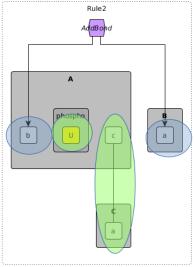
For 'A' and 'B' to bind all other states must be unoccupied.

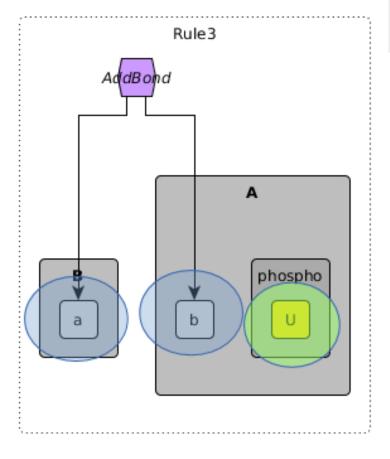


For 'A' and 'B' to bind 'A' must be bound to 'C' already

Context factorization

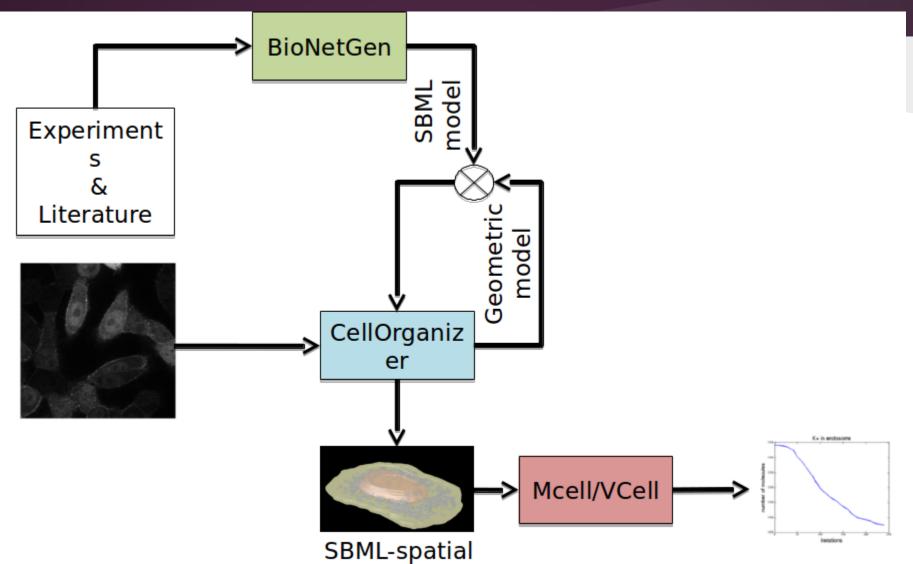






For A and B to bind it does not matter if C is part of the complex or not

Pipeline



Now you can atomize too!

```
http://ratomizer.appspot.
com/translate
```

BNG2.pl <sbmlfile.xml>

Thank you

Faeder Lab:

- Dr. James Faeder
- Dr. Justin Hogg
- Dr. Leonard Harris
- John Sekar

MCell Team

Devin Sullivan Aaron Wise



Q&A

