

Introduction to rule-based modeling with BioNetGen and RuleBender

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Presenters:

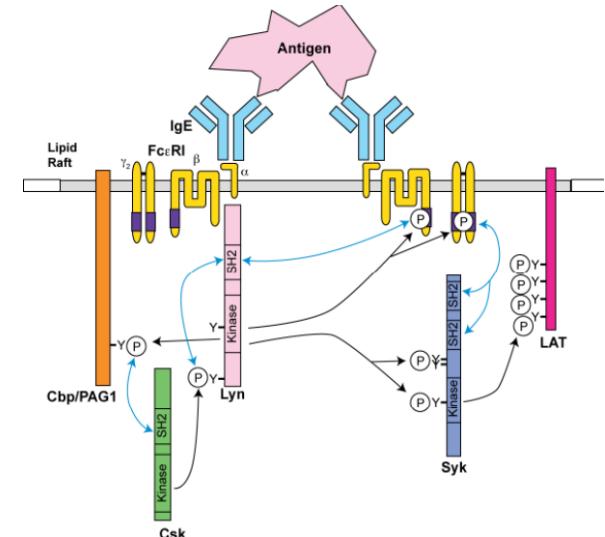
Cihan Kaya

Jose Juan Tapia

John Sekar

Sanjana Gupta

For additional information and references see
<http://bionetgen.org/index.php/Tutorials>



Agenda for Day 2

8:30-9:30 - Intro to Rule-Based Modeling (**Jim**)

9:30-9:45 - Break

9:45-11:00 - Basic models in RuleBender (**Jim**):

Lotka-Volterra, Polarization, Michaelis-Menten, CaMKII

11:00-11:30 - Modeling DA neurons (**Cihan**)

11:30-12:00 - Atomizer demo and models (**Jose**)

12:00-1:00 - Lunch

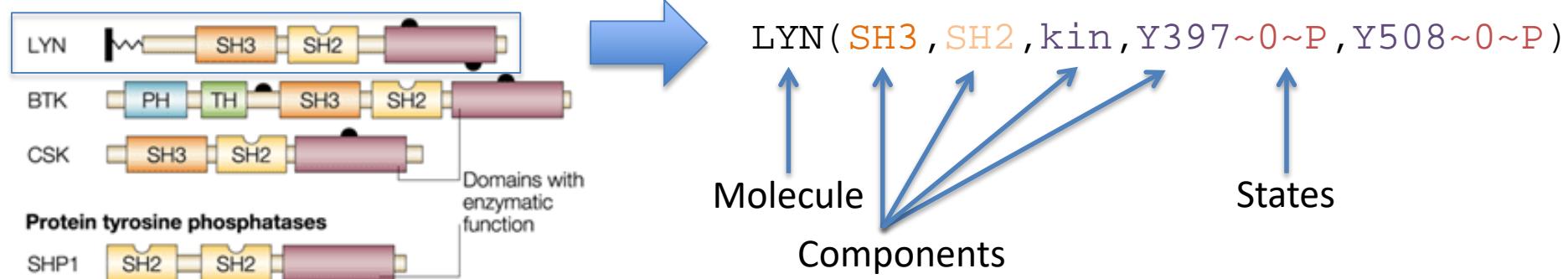
1:00-1:30 - Visualization demo (**John**)

1:30-2:30 - Clustering and confinement models (**Sanjana**)

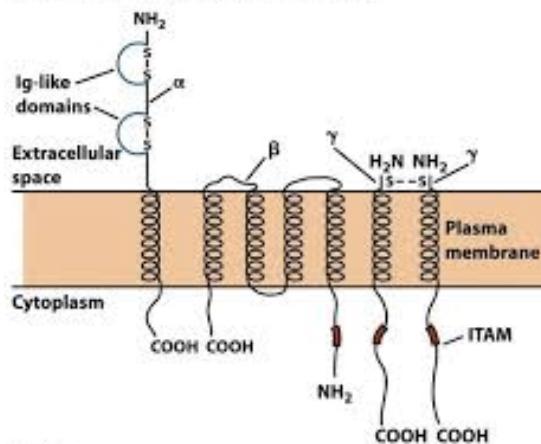
2:30-5:00 - Projects and open demos and discussions

What is Rule-based Modeling (RBM)?

Molecules are modeled as *structured objects*



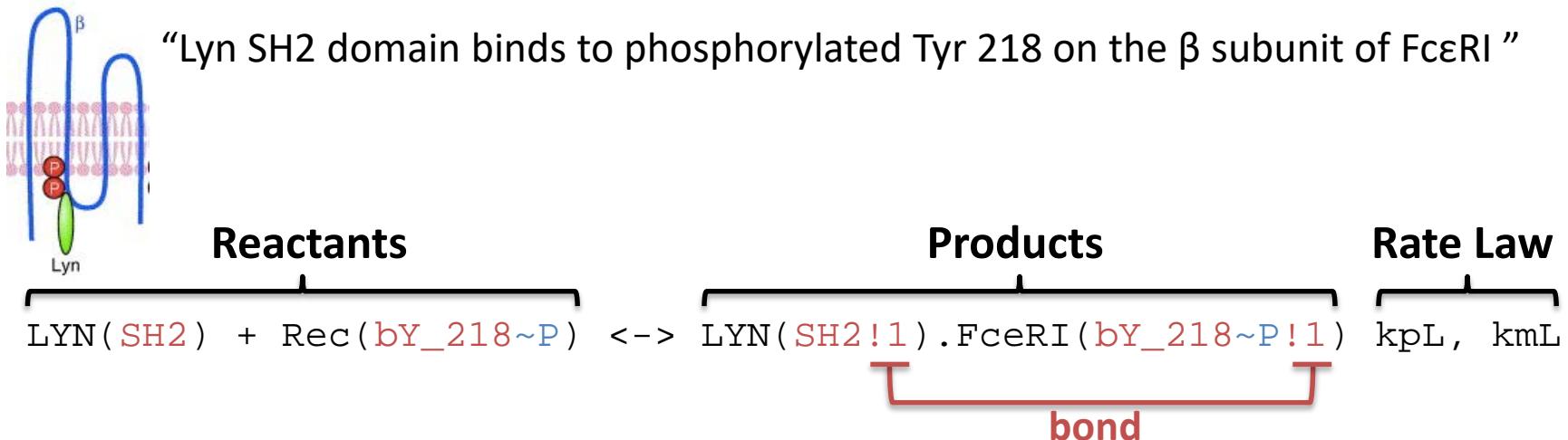
Fc ϵ RI: High-affinity IgE receptor



Fc ϵ RI(a_Ig, b_Y218~0~P, g_ITAM~0~P)

What is Rule-based Modeling (RBM)?

Rules define the interactions of molecules



Center – elements modified by the action of the rule

Context – elements required for reaction to occur but not modified

“Don’t write don’t care” – elements not mentioned may be in any state

→ One rule can generate reactions involving many different species

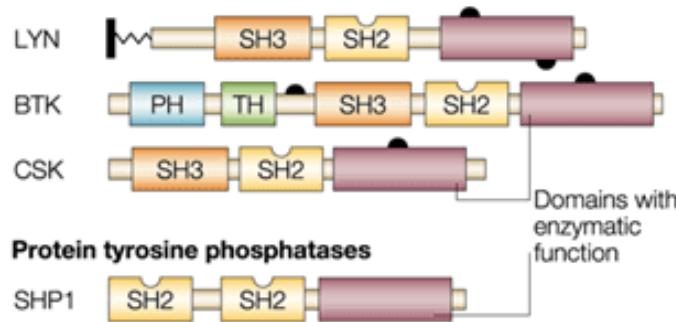
Reaction rate determined by **Mass Action kinetics**

$$\text{rate forward} = k_p L * [Lyn(SH2)] * [Rec(by_{218}~P)]$$

↑
rate per reactant set (instance rate)

Why was RBM developed?

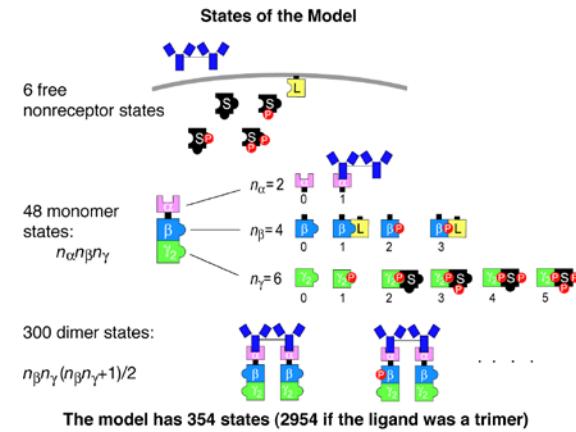
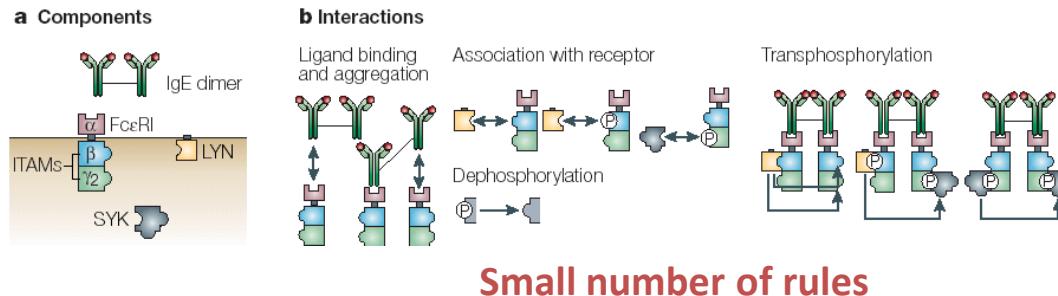
- Proteins are multi-functional



multiple sites of binding

multiple sites of posttranslational modification

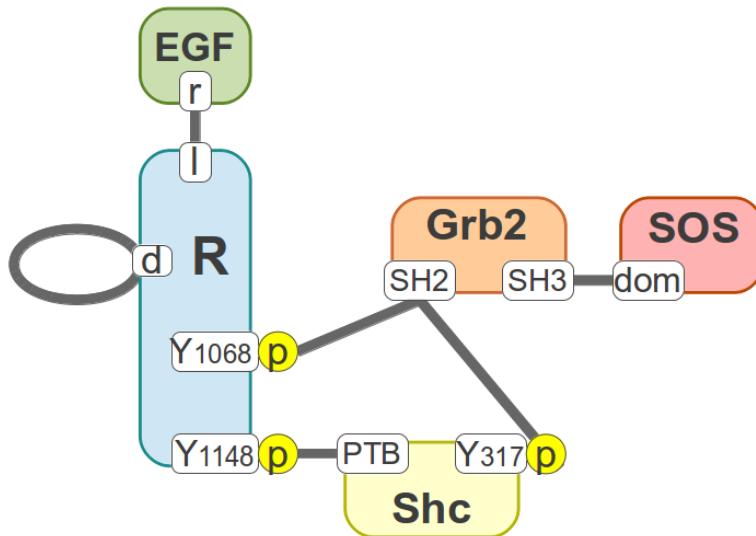
- Representing their known interactions requires handling of *combinatorial complexity*



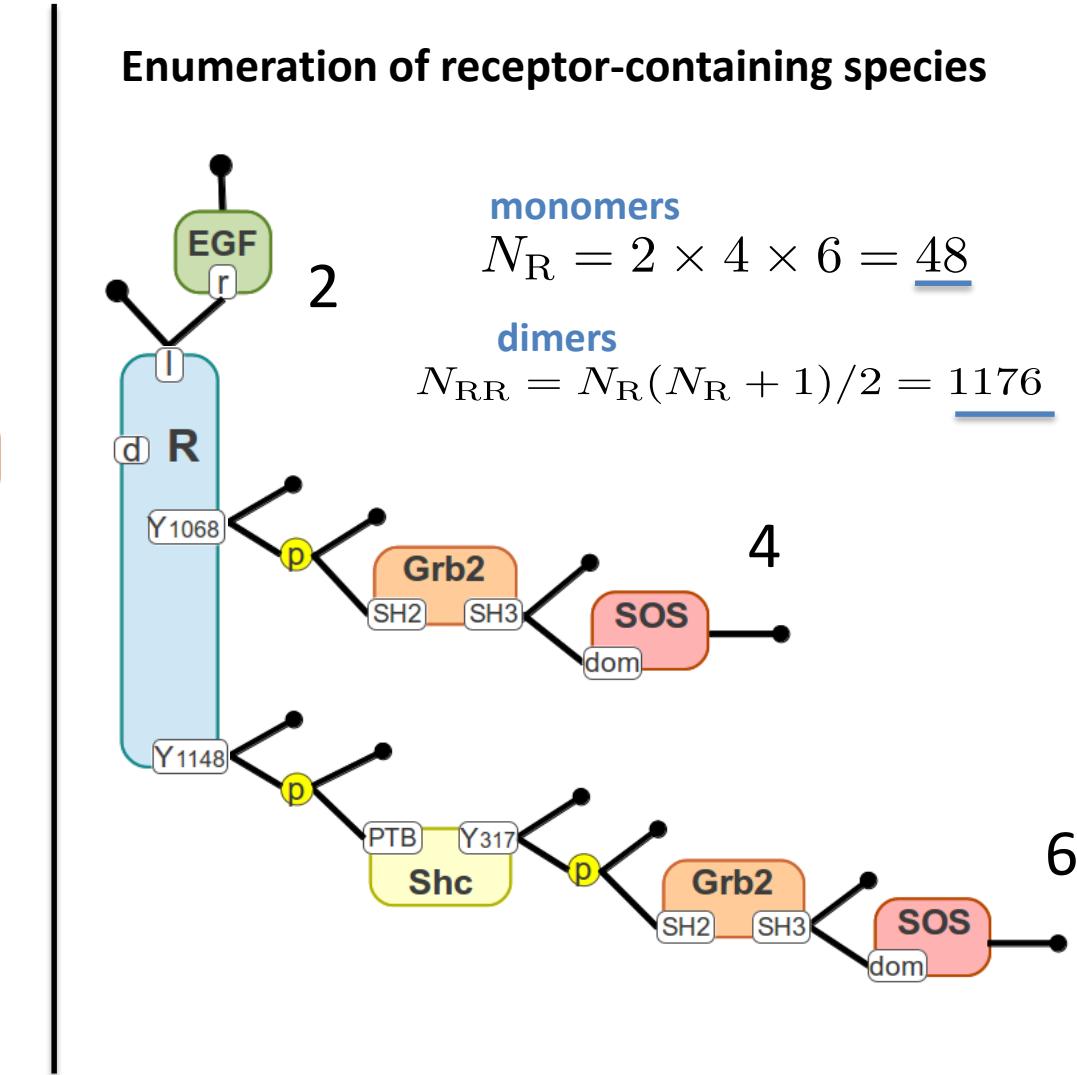
Small number of components and interactions → huge number of possible species and reactions

Combinatorial Complexity in Biochemical Interactions

Contact Map for molecules involved in EGFR signaling

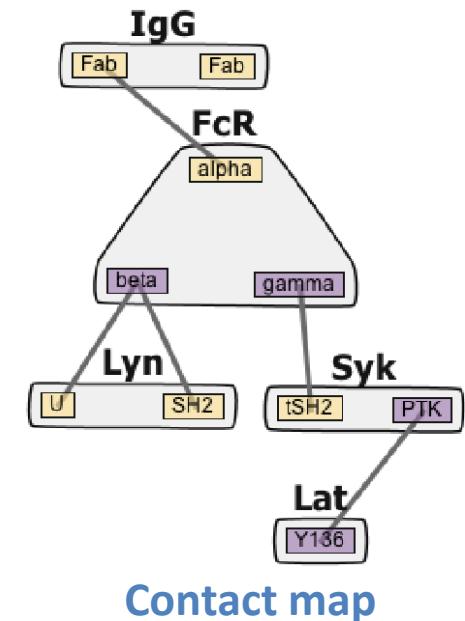


Enumeration of receptor-containing species



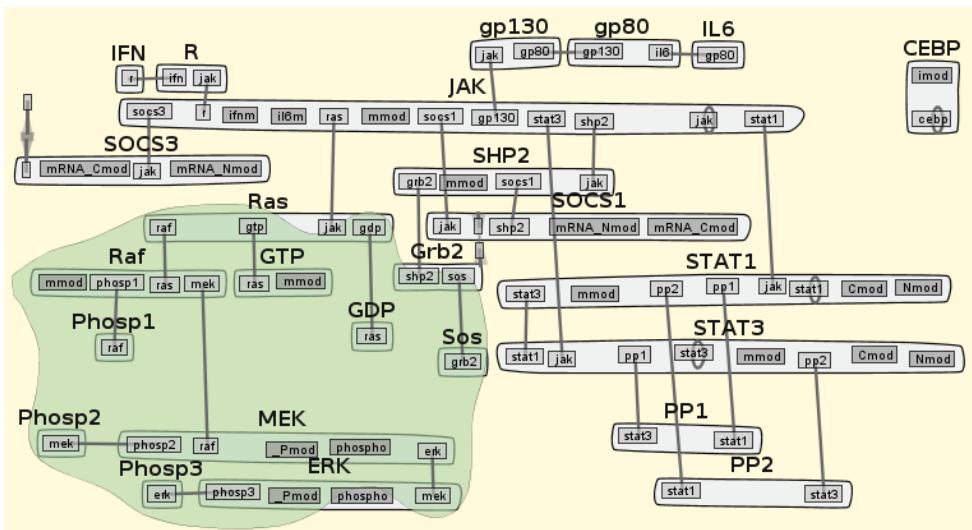
Why use RBM?

- *Concise and precise* representation of biochemical knowledge
 - rules are simple (less context) when interactions are modular
- *Flexible* with respect to simulation method
 - Deterministic / stochastic
 - Well-mixed / compartmental / spatial
- Structures and rules are *reusable*
 - Rule libraries
- Compact visual representation
 - Contact map and beyond

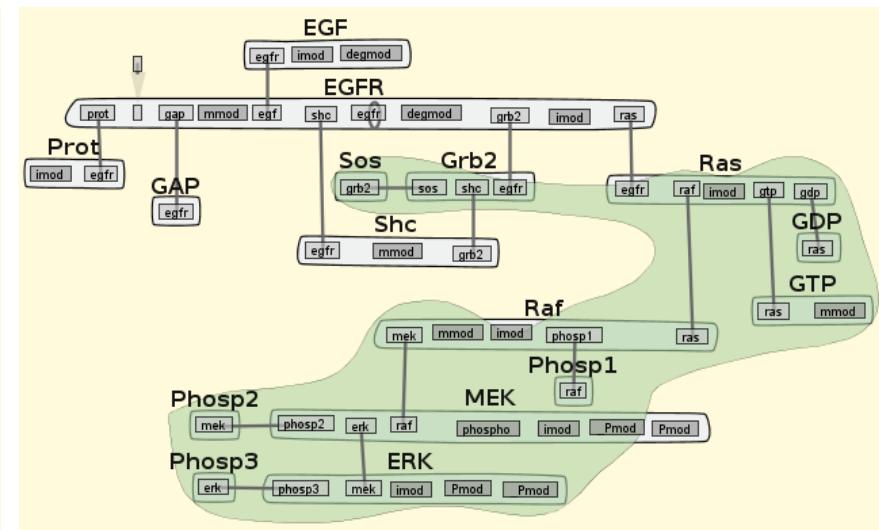


Model Comparison

Generated by MOSBIE, a model exploration system within RuleBender



BioModels 543



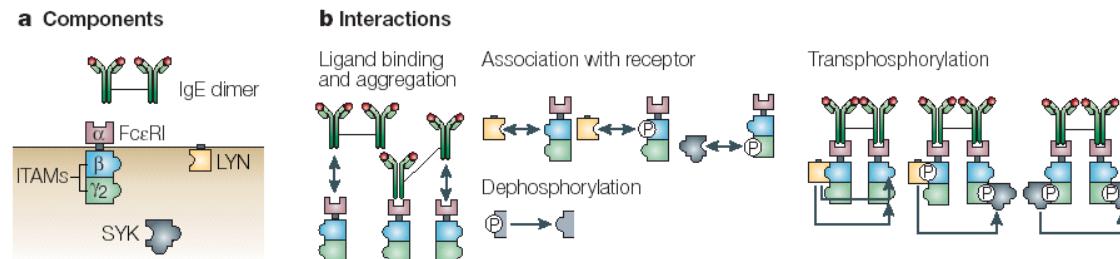
BioModels 19

Some Rule-Based Modeling Tools

- BioNetGen & relatives
 - RuleBender
 - NFsim
 - BioNetGen@VirtualCell
- Kappa & relatives
- ML-rules
- Simmune
- pySB
- SBML Level 3 Multi

Rule-Based Modeling protocol

1. Identify components and interactions



2. Translate into objects (molecules) and rules



3. Determine concentrations and rate constants

4. Simulate and analyze the model

ODE's

SSA

PDE's

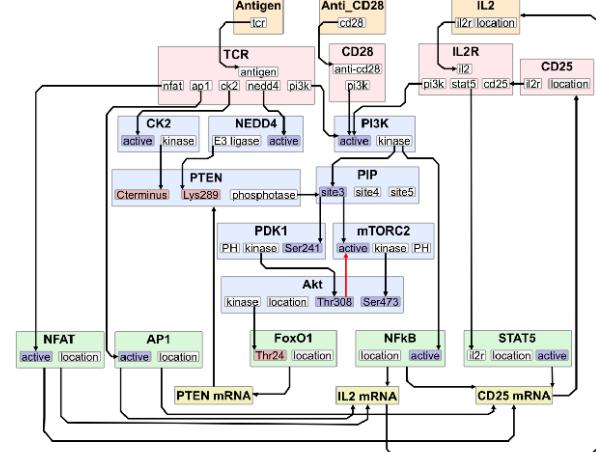
BD

[IgE dimer] 10 nM [Fc ϵ RI] 4×10^5 per cell
[Lyn] 3×10^4 per cell [SYK] 4×10^5 per cell

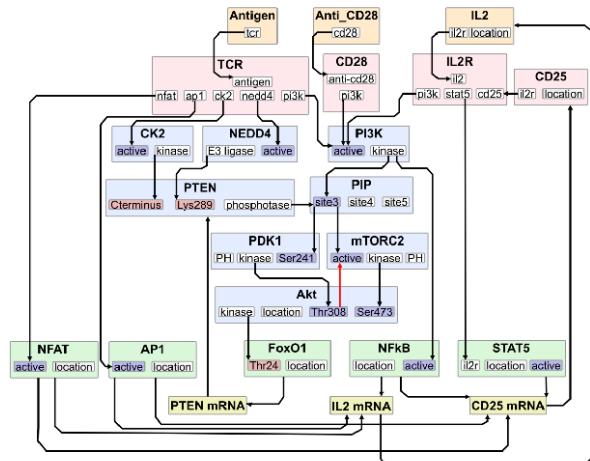
Systems Modeling Workflow

1. Construct model based on **known** and **hypothesized** interactions.

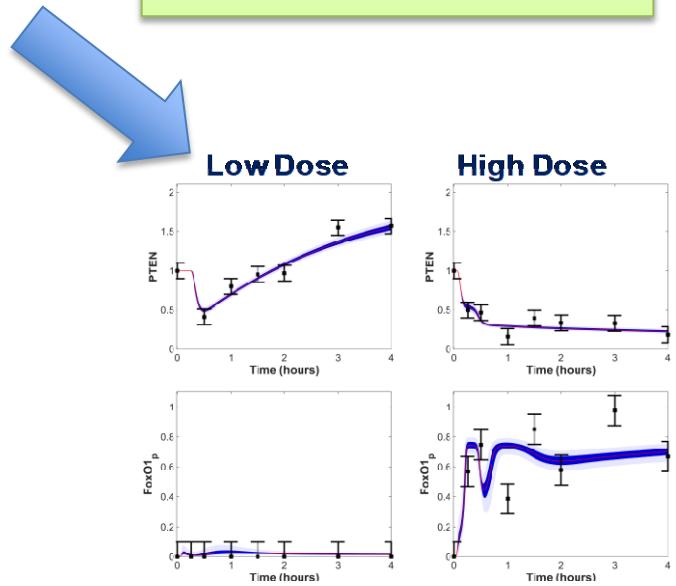
RuleBender
BioNetGen



Systems Modeling Workflow

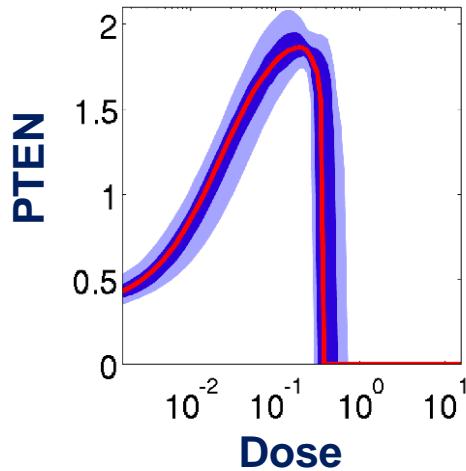
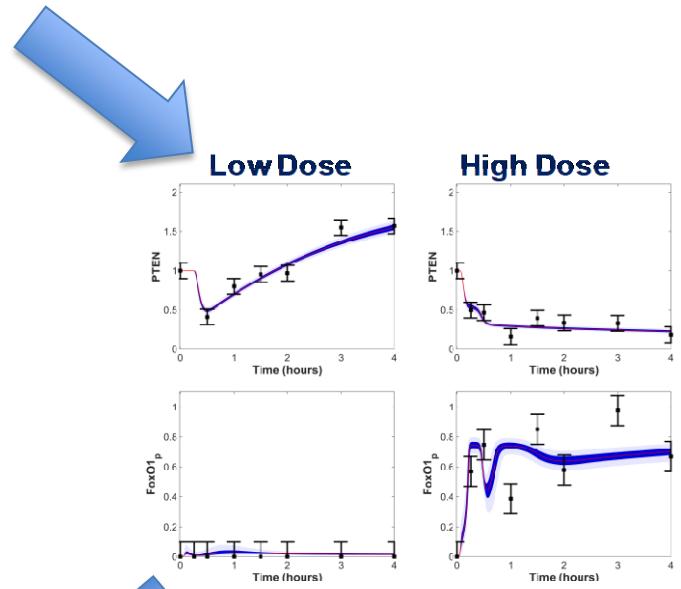
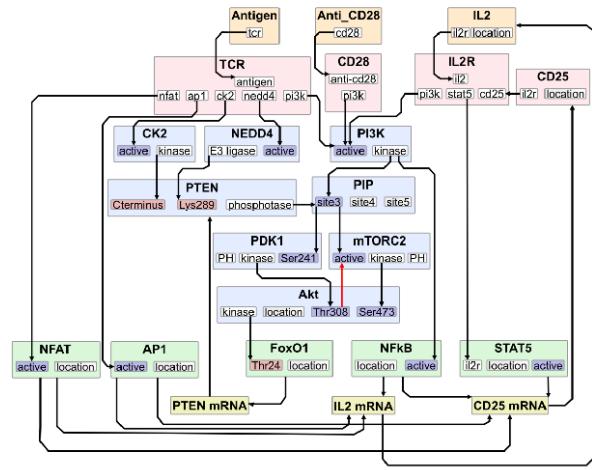


2. Calibrate model parameters to available data.



ptempest
BioNetFit

Systems Modeling Workflow

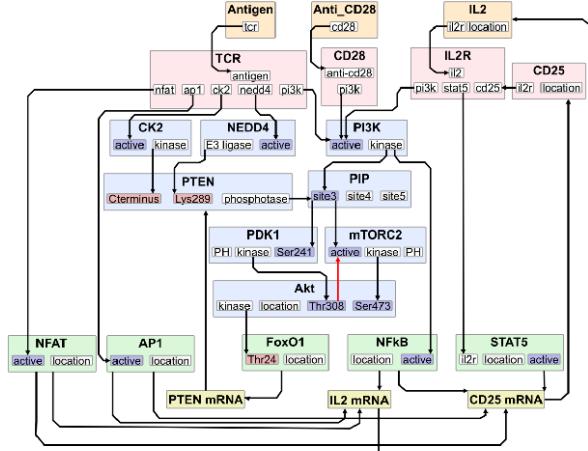
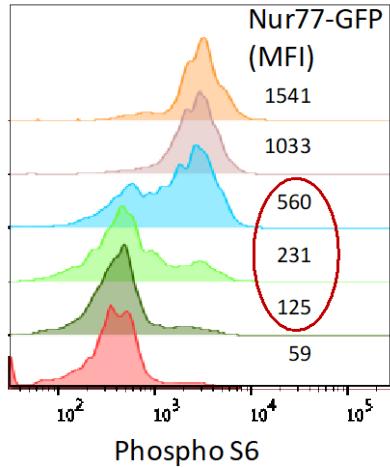


ptempest
RuleBender

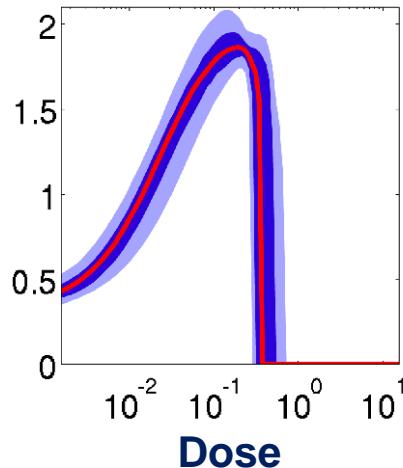
3. Analyze model to find novel behaviors.

Systems Modeling Workflow

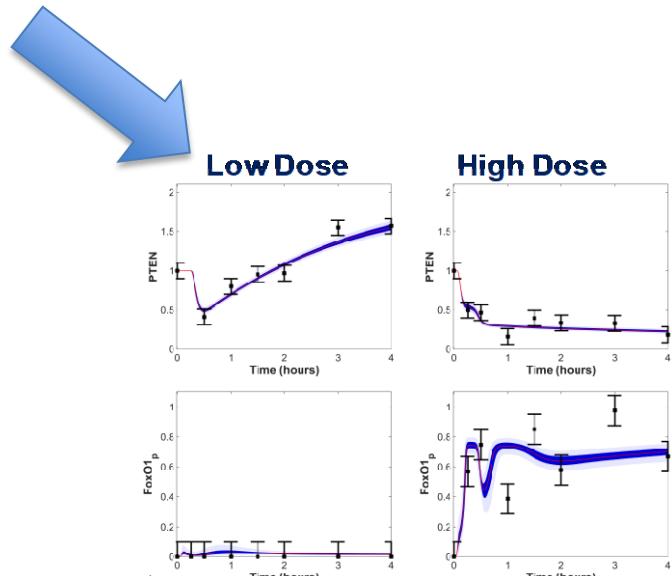
Increasing activation level ↑



PTEN

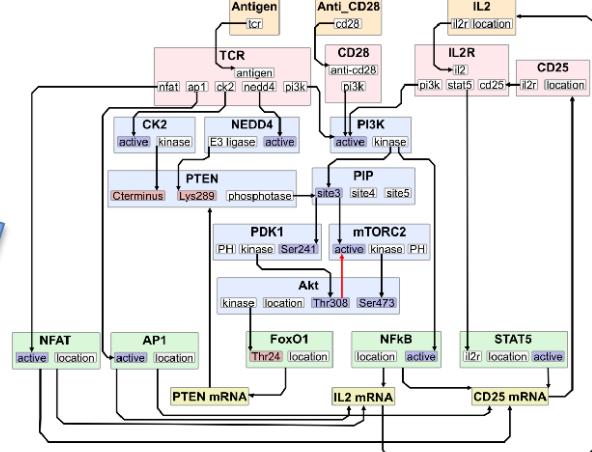


4. Perform experiments to test predictions.

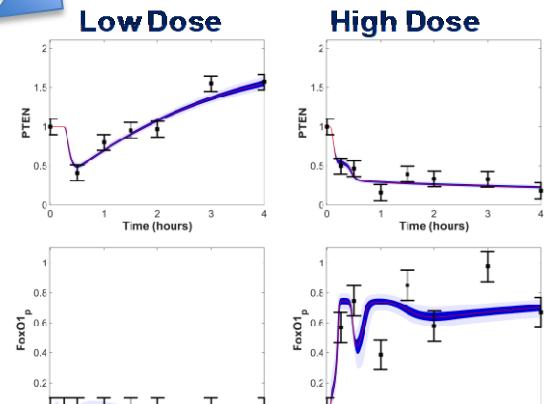


Systems Modeling Workflow

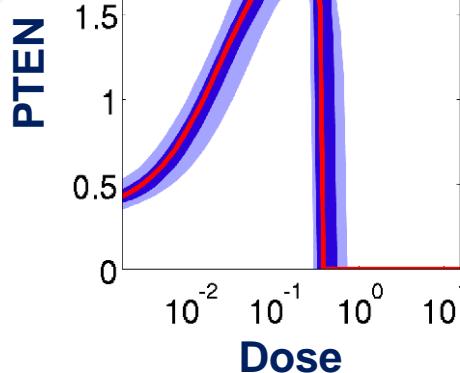
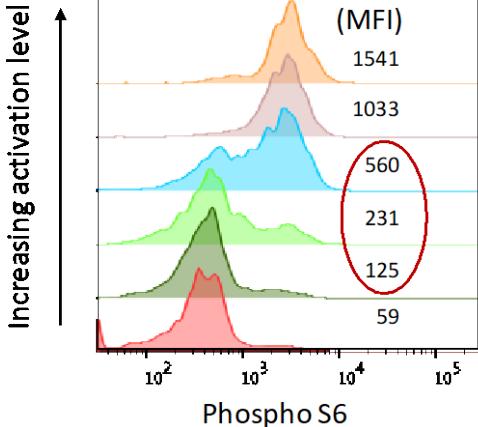
1. Construct model based on **known** and **hypothesized** interactions.



2. Calibrate model parameters to available data.



Repeat as needed



4. Perform experiments to test predictions.

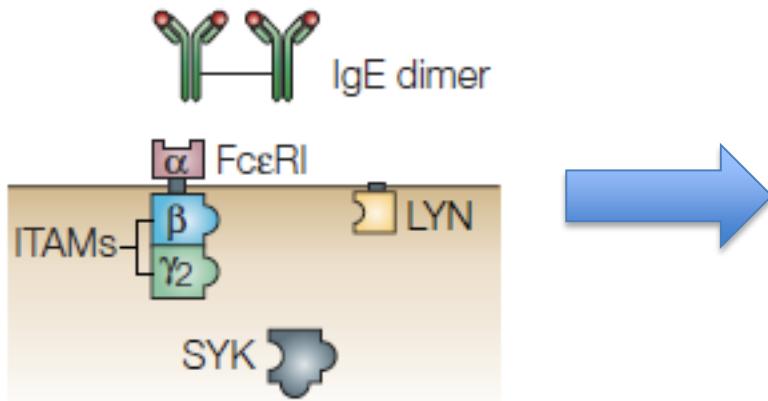
3. Analyze model to find novel behaviors.

SPECIFYING A RULE-BASED MODEL

Defining Molecules

Molecules are the basic objects in a
BNG model

BioNETGEN Language



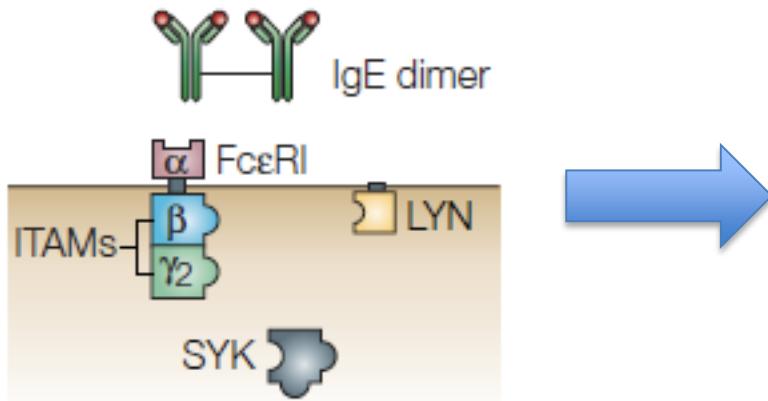
IgE(a , a)
FcεRI(a , b~U~P , g2~U~P)
Lyn(U , SH2)
Syk(tSH2 , lY~U~P , aY~U~P)

- Components** represent molecule elements
- Domains
 - Motifs
 - Properties

Defining Molecules

Molecules are the basic objects in a BNG model

BioNETGEN Language



IgE(a , a)

Fc ϵ RI(a , b~U~P , g2~U~P)

Lyn(U , SH2)

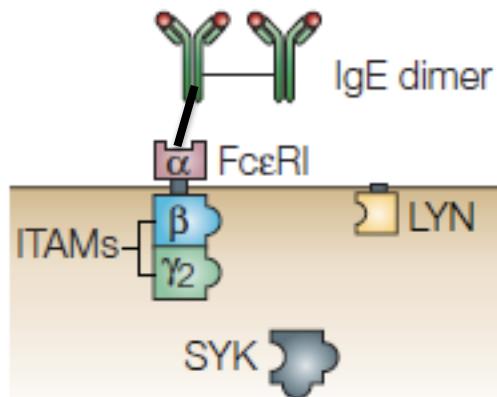
Syk(tSH2 , lY~U~P , aY~U~P)

- Components** may have different **states** representing
- posttranslational modifications
 - conformational state
 - ...

Binding

Molecules bind other molecules through components

BioNETGEN Language



$\text{IgE}(a, a!1). \text{Fc}\epsilon\text{RI}(a!1, b\sim U, g2\sim U)$

Bonds are formed by linking two components. The ‘.’ indicates a set of molecules forming a complex.

$\text{Fc}\epsilon\text{RI}(a, b\sim U!1, g2\sim U). \text{Lyn}(U!1)$

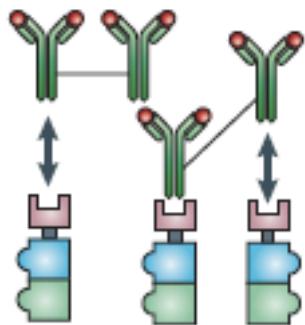
Components may have both states and bonds.

$\text{Lyn}(\text{SH2}!1, \text{Cterm}\sim P!1)$

Bonds may occur within a molecule.

Defining Interaction Rules

Ligand binding
and aggregation



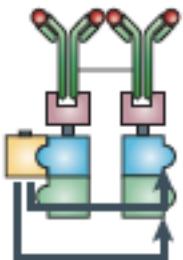
BioNETGEN Language

$\text{IgE}(a, \underline{a}) + \text{Fc}\epsilon\text{RI}(\underline{a}) \rightleftharpoons \text{IgE}(a, \underline{a!1}).\text{Fc}\epsilon\text{RI}(\underline{a!1})$

...

binding and dissociation

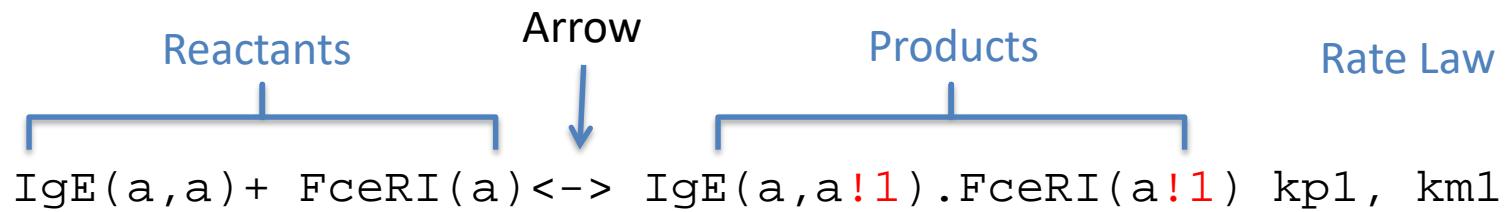
Transphosphorylation



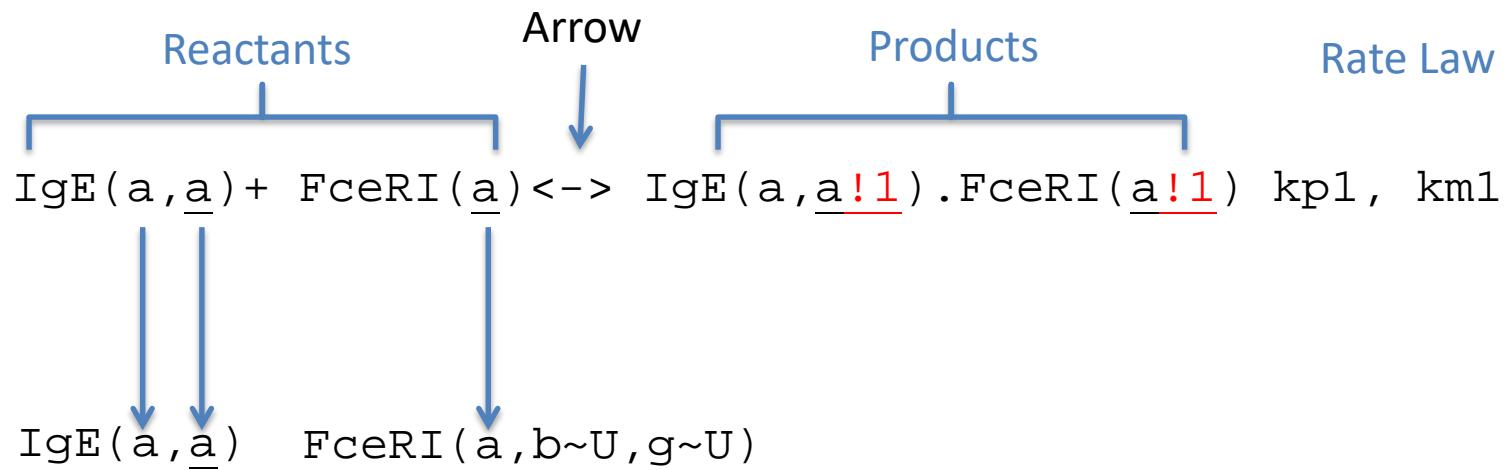
$\text{Lyn}(U!1).\text{Fc}\epsilon\text{RI}(b!1).\text{Fc}\epsilon\text{RI}(b\sim\underline{\text{U}}) \rightarrow \backslash$
 $\text{Lyn}(U!1).\text{Fc}\epsilon\text{RI}(b!1).\text{Fc}\epsilon\text{RI}(b\sim\underline{\text{P}})$

component state change

Parts of a rule

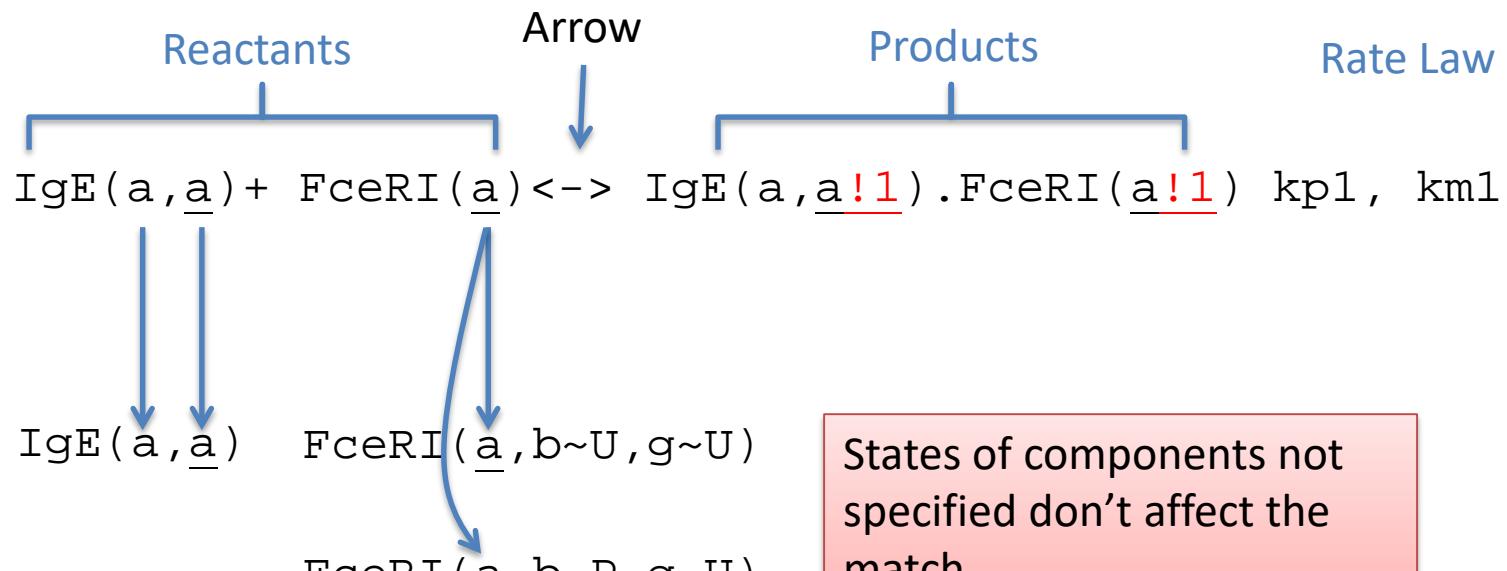


Parts of a rule



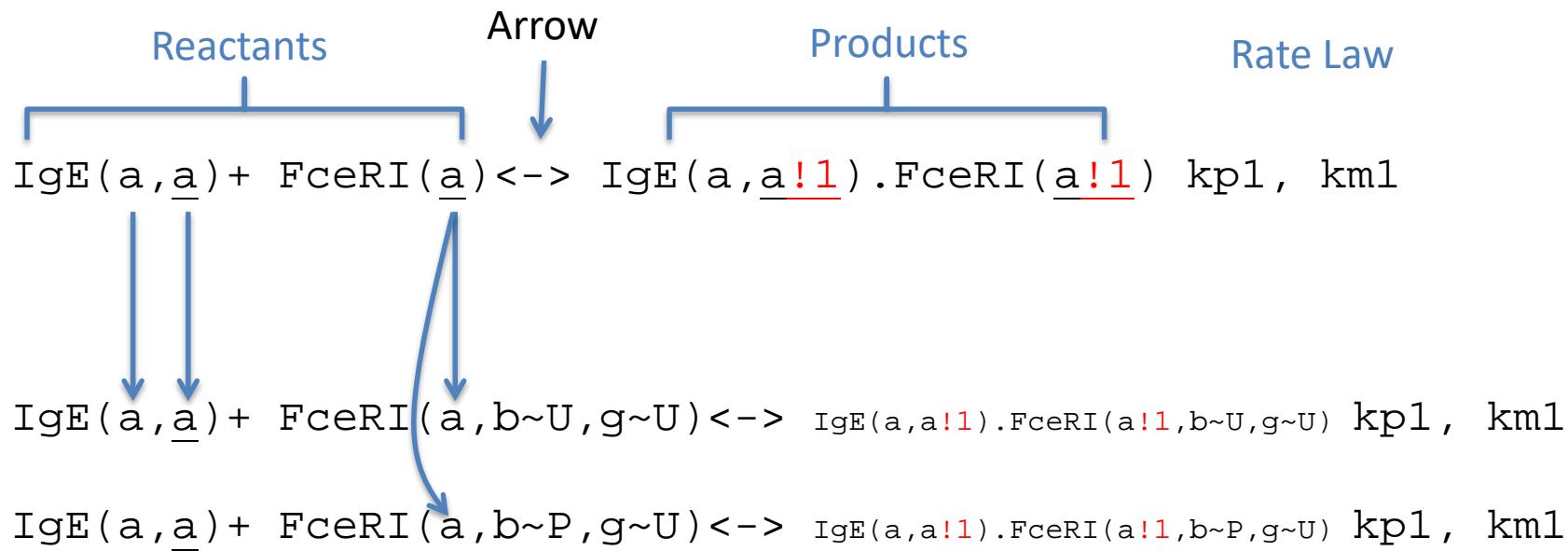
Reactant patterns
select properties of
each reactant
molecule.

Parts of a rule



Reactant patterns
select properties of each reactant molecule.

Parts of a rule

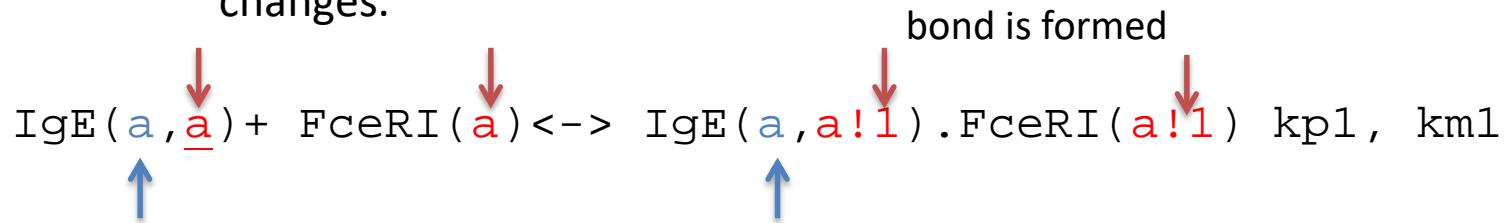


Reactant patterns
select properties of
each reactant
molecule.

Because patterns can match
many different species,
each rule can generate
many reactions.

Center and context

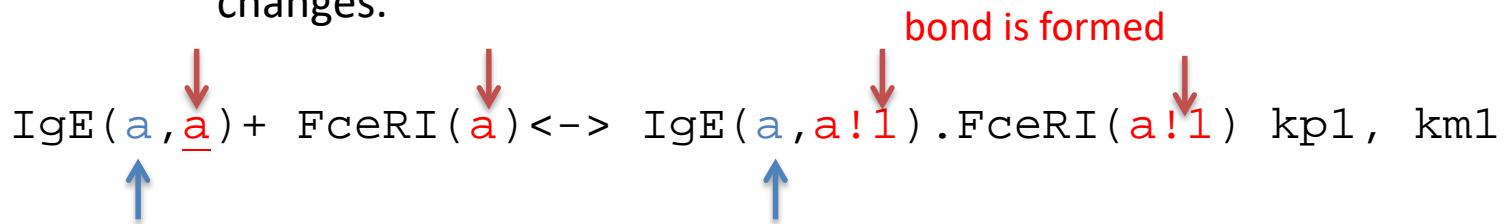
The **center** of a rule is the part that the rule changes.



The **context** is the part that is necessary for the rule to happen but is unchanged.

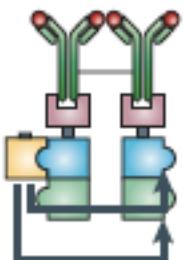
Center and context

The **center** of a rule is the part that the rule changes.



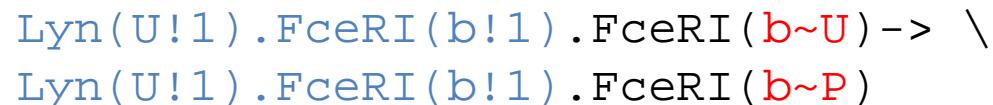
The **context** is the part that is necessary for the rule to happen but is unchanged.

Transphosphorylation



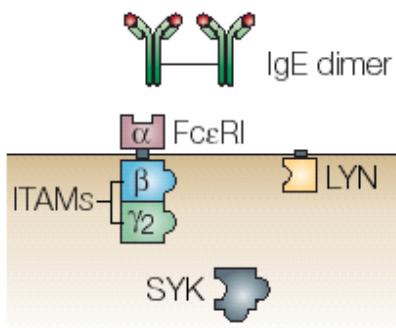
Context can represent complex biochemistry.

component state is changed

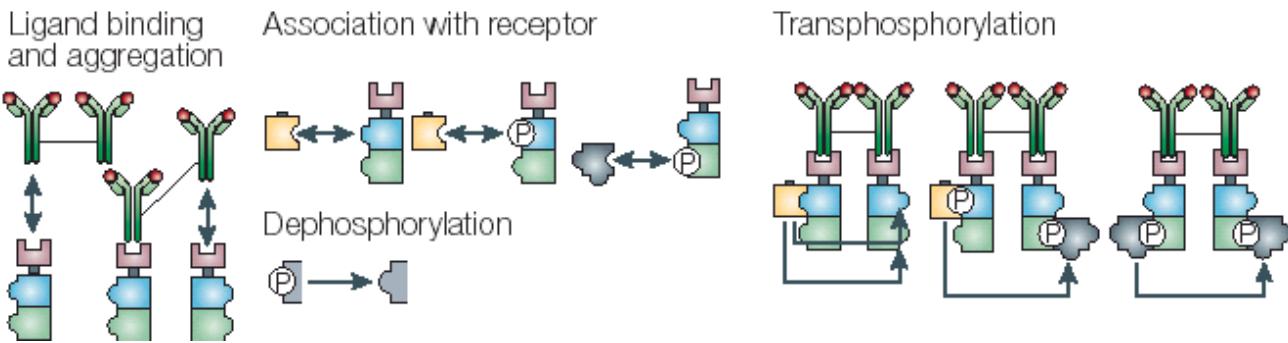


Composition of a Rule-Based Model

a Components



b Interactions



Molecules

```
begin molecules
Lig(1,1)
Lyn(U,SH2)
Syk(tSH2,1~U~P,a~U~P)
Rec(a,b~U~P,g~U~P)
end molecules
```

Reaction Rules

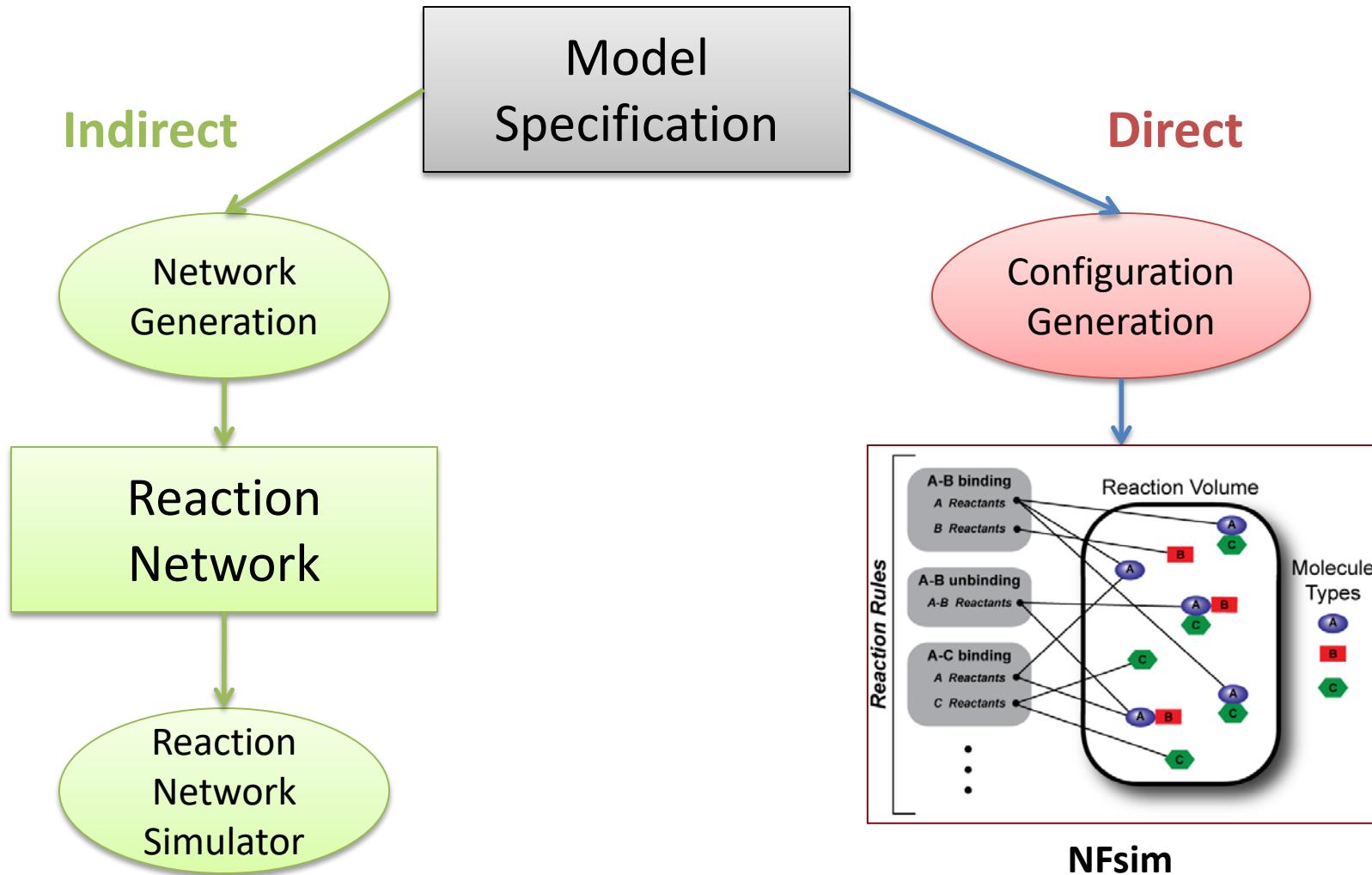
```
begin reaction_rules
# Ligand-receptor binding
1 Rec(a) + Lig(1,1) <-> Rec(a!1).Lig(1!1,1) kp1, km1
   Rec(a) + Lig(1,1) <-> Rec(a!1).Lig(1!1,1) kp1, km1

# Receptor-aggregation
2 Rec(a) + Lig(1,1!1) <-> Rec(a!2).Lig(1!2,1!1) kp2, km2

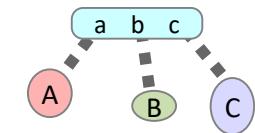
# Constitutive Lyn-receptor binding
3 Rec(b~Y) + Lyn(U,SH2) <-> Rec(b~Y!1).Lyn(U!1,SH2) kpL, kmL
...
```

BioNetGen language

Methods for simulating RBMs

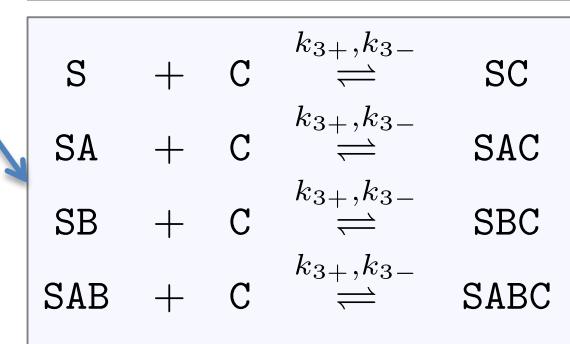
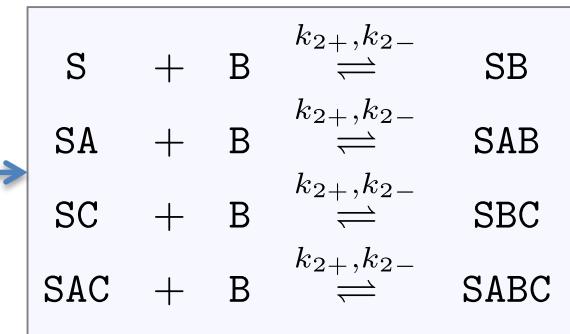
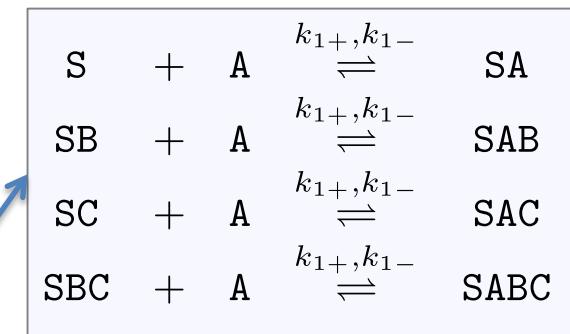
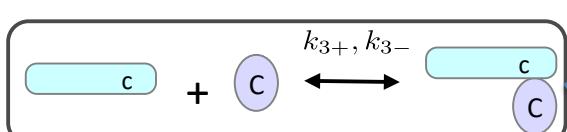
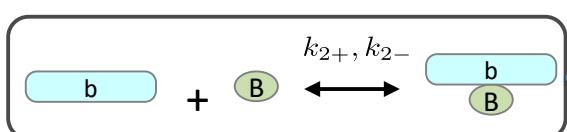
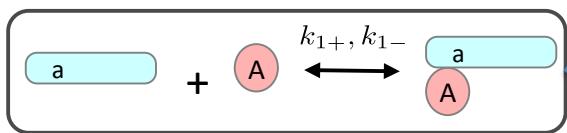


Indirect Methods – Network Generation



contact map

reaction rules

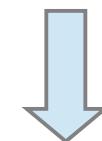


reactions

S, A, B, C,
SA, SB, SC,
SAB, SAC, SBC,
SABC

species

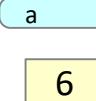
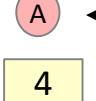
4 molecule types
3 rules



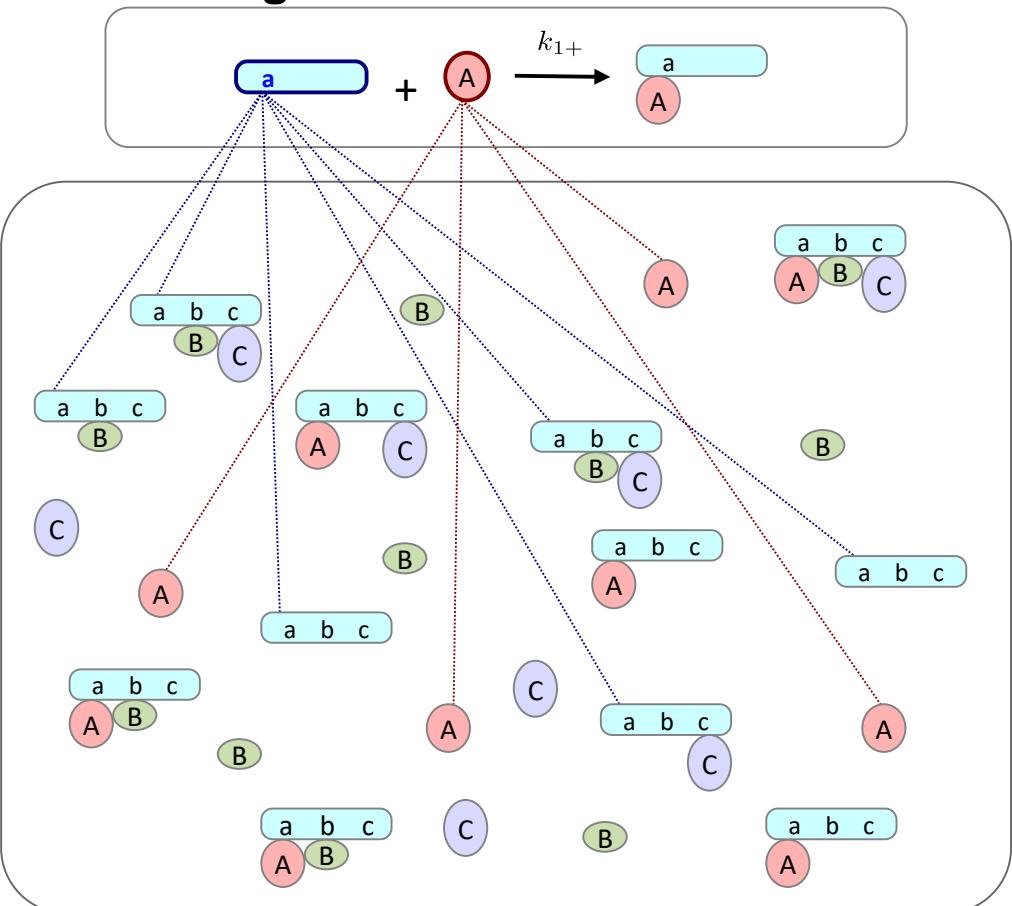
11 species
12 reactions

Direct Methods (NFsim)

reaction rules

		$a + A \xrightleftharpoons{k_{1+}, k_{1-}} aA$	$24k_{1+}$
		$b + B \xrightleftharpoons{k_{2+}, k_{2-}} bB$	$30k_{2+}$
		$c + C \xrightleftharpoons{k_{3+}, k_{3-}} cC$	$21k_{3+}$
			$6k_{1-}$
			$6k_{2-}$
			$5k_{3-}$

event generation



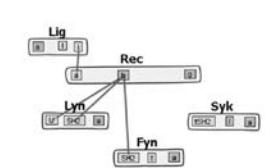
system configuration

total propensity

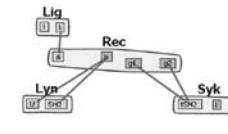
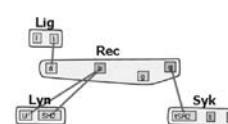
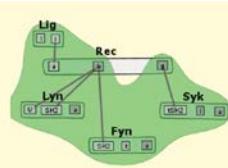
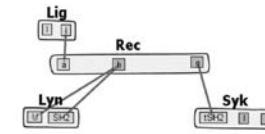
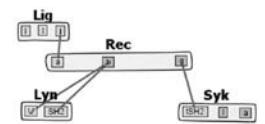
$$24k_{1+} + 6k_{1-} + 30k_{2+} + 6k_{2-} + 21k_{3+} + 5k_{3-}$$

Fc ϵ RI signaling models

a



b



NFSIM can simulate models of greatly increased complexity with manageable increase in cost.

RuleBender

An eclipse RCP application

rulebender.org

RuleBender

Model Simulation Results

egfr_net.bnlg

```
#Dephosphorylation
egfr(Y1068~pY) -> egfr(Y1068~Y) km3
egfr(Y1148~pY) -> egfr(Y1148~Y) km3

# Shc transphosph
egfr(r!2,Y1148~pY!1).Shc(PTB!1,Y317~Y) -> egfr(r!2,Y1148~pY!1).Shc(PTB!1,Y317~pY) km13
Shc(PTB!1,Y317~pY) -> Shc(PTB!1,Y317~Y) km14

# Y1068 activity
egfr(Y1068~pY) + Grb2(SH2,SH3) <-> egfr(Y1068~pY!1).Grb2(SH2!1,SH3)
egfr(Y1068~pY) + Grb2(SH2,SH3!2) <-> egfr(Y1068~pY!1).Grb2(SH2!1,SH3)
egfr(Y1068~pY!1).Grb2(SH2!1,SH3) + Sos(dom) <-> egfr(Y1068~pY!1).Grb2(SH2,SH3)

# Y1148 activity
egfr(Y1148~pY) + Shc(PTB,Y317~Y) <-> egfr(Y1148~pY!1).Shc(PTB!1,Y317~Y)
egfr(Y1148~pY) + Shc(PTB,Y317~pY) <-> egfr(Y1148~pY!1).Shc(PTB!1,Y317~pY)
egfr(Y1148~pY) + Shc(PTB,Y317~pY!1).Grb2(SH2!1,SH3) <-> \
```

Problems

1 error, 0 warnings, 0 others

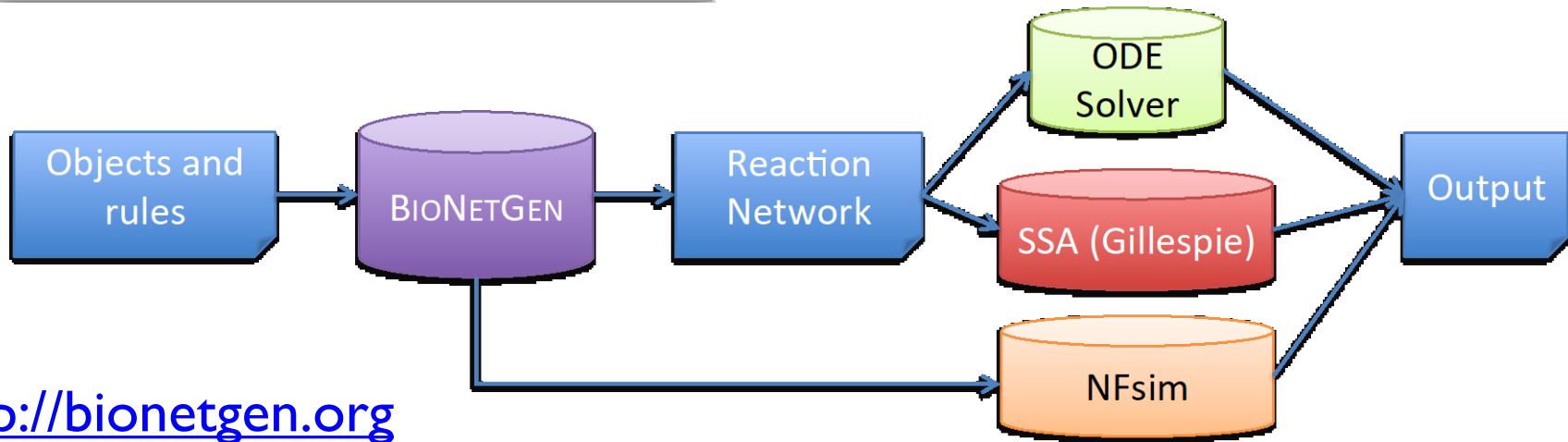
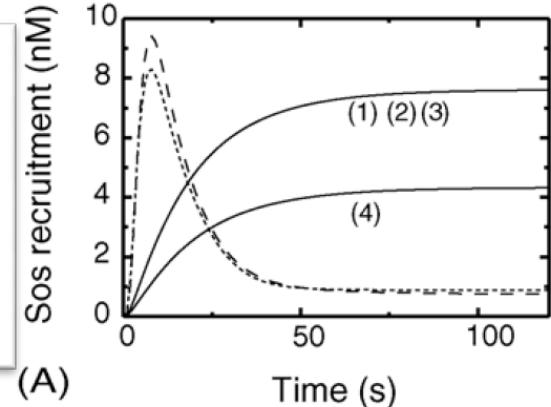
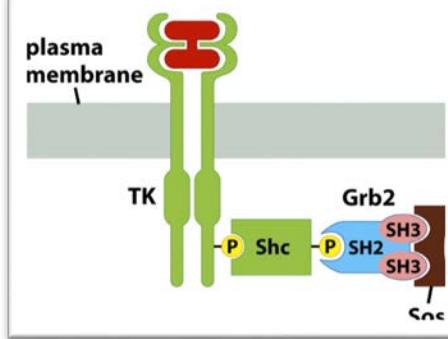
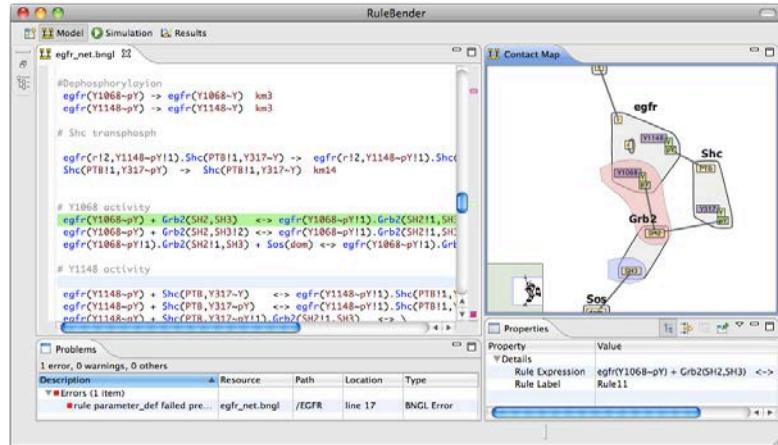
Description	Resource	Path	Location	Type
Errors (1 item)	egfr_net.bnlg	/EGFR	line 17	BNGL Error
rule parameter_def failed pre...				

Contact Map

Properties

Property	Value
Rule Expression	egfr(Y1068~pY) + Grb2(SH2,SH3) <->
Rule Label	Rule11

Basic RBM workflow with BioNetGen

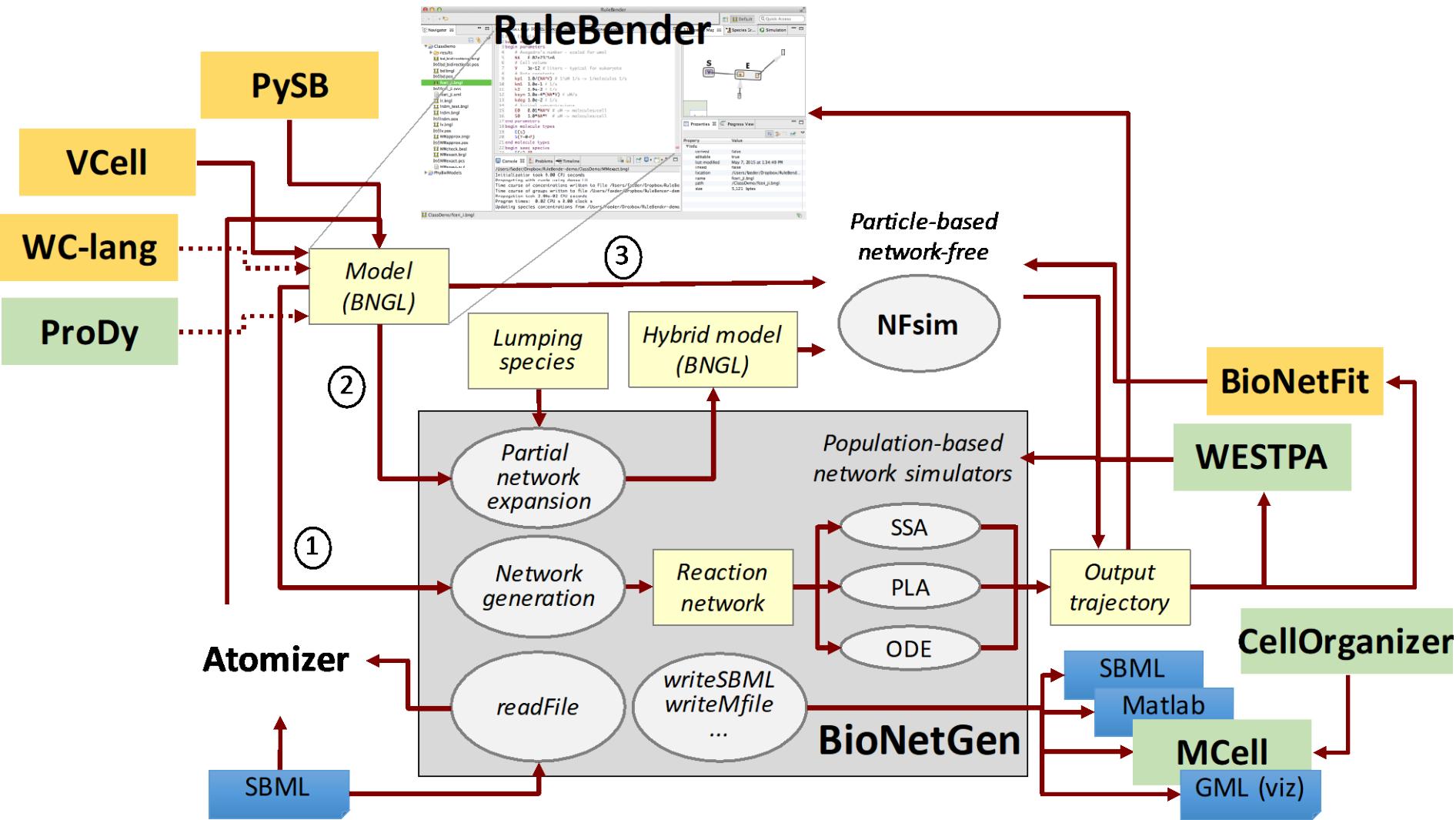


<http://bionetgen.org>

<http://rulebender.org>

<http://nfsim.org>

BioNetGen and Friends



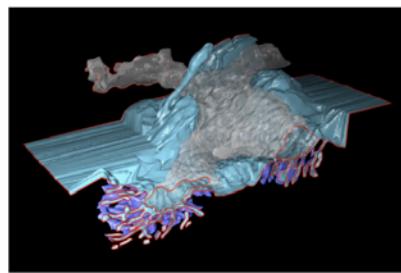
Development of network-free MCell

Beyond a well-mixed cell....

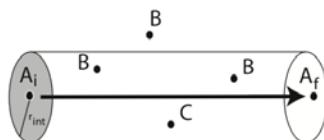
Jose Juan Tapia

MCell

1. Realistic 3D Geometry

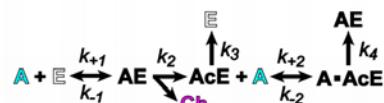
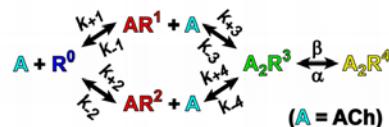


2. Random Walk Diffusion

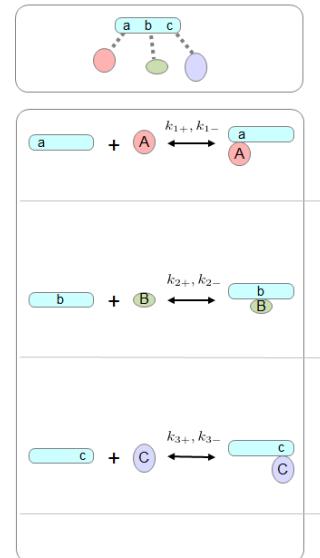


$A + B \rightarrow \text{products}$

3. Stochastic Biochemical State Transition

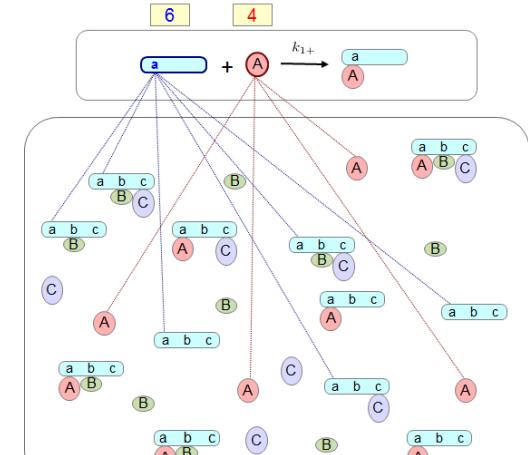


contact map



Network-free Simulation

Find reactant sites and compute propensity

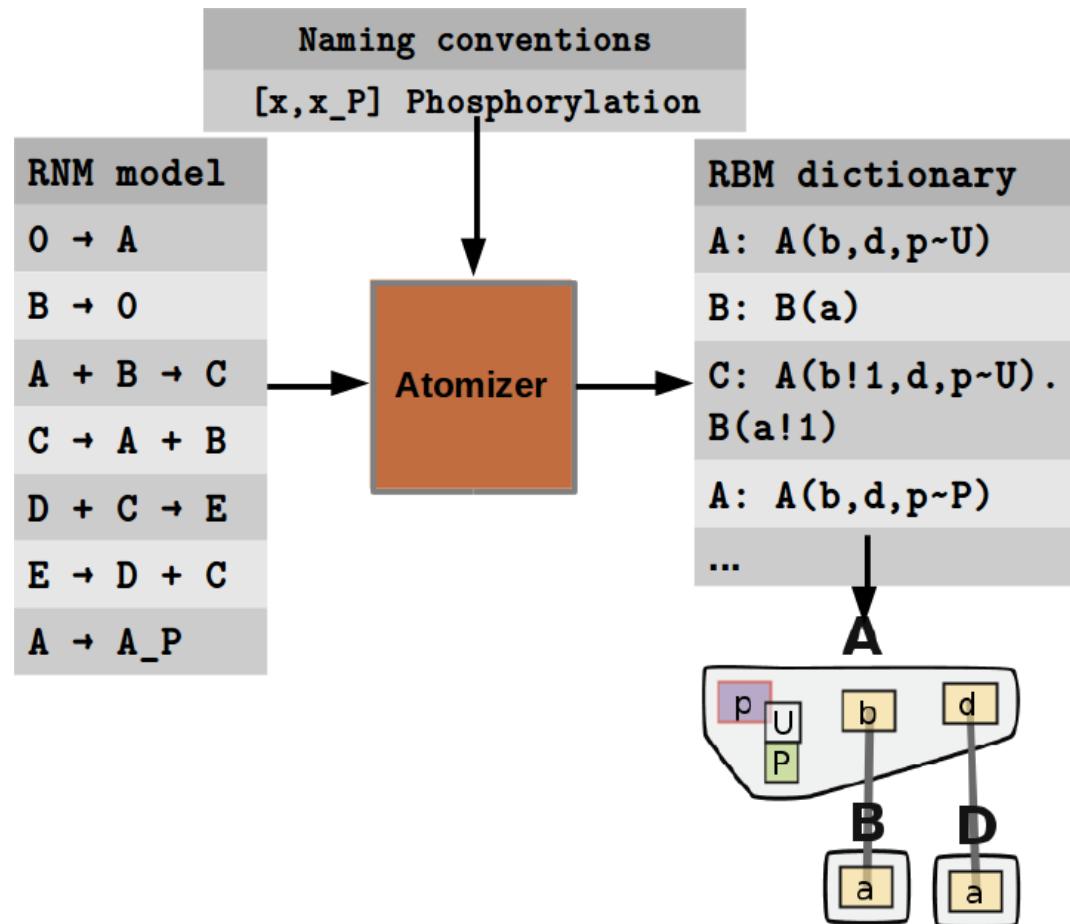


system

Particle-based

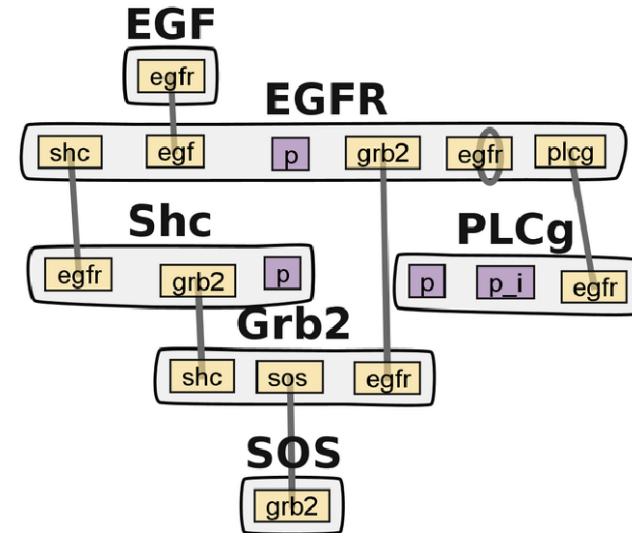
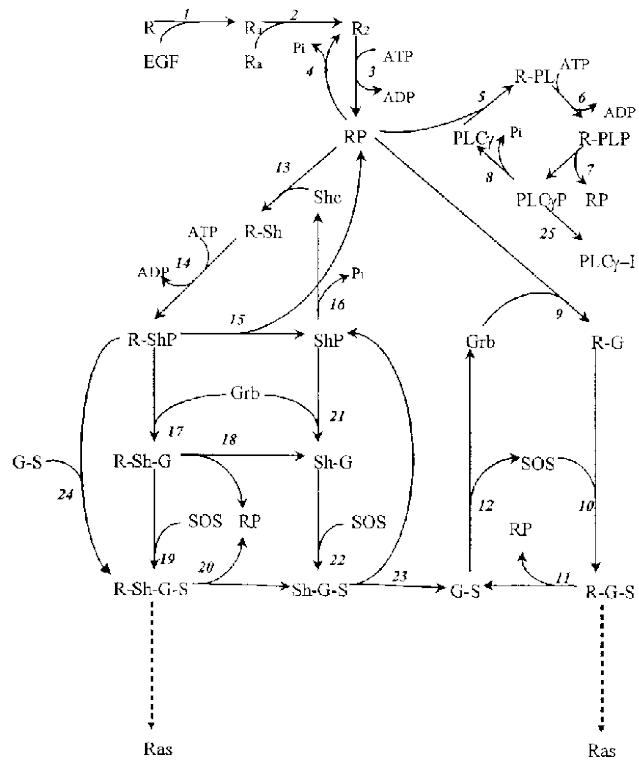
Atomizer can recover implicit molecular structure reaction networks

1. Reaction stoichiometry
2. Naming conventions
3. Annotation



Automating structure recovery

“Atomization”



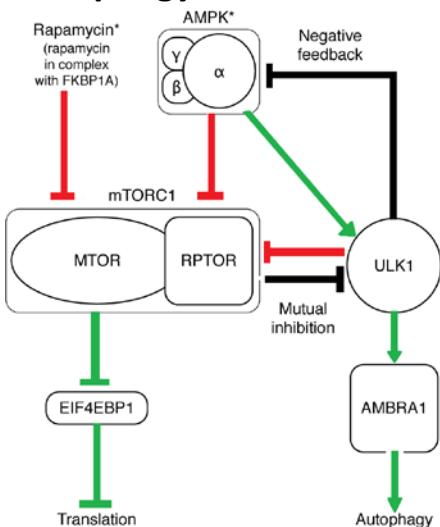
Motivation

- Find molecules and interactions used to build model
- Find out what has been modeled?
- Find the implicit assumptions
- Automate study of relaxing model assumptions

Visualizing rule-based models

John Sekar
Jose Juan Tapia

Autophagy network



Rule-based model

```
R1: rapa(mtor)+MTOR(HEAT,FRB)<->rapa(mtor!).MTOR(HEAT,FRB!) a1,d1
R2: MTOR(HEAT,FRB)+RPTOR(WD40)<->MTOR(HEAT!,FRB).RPTOR(WD40!) a2,d2
R3a: RPTOR(RNC,ulk1,S792~O,S855_S859~O)+ULK1(straptor) ->RPTOR(RNC,ulk1!,S792~O,S855_S859~O).ULK1(straptor!) a3
R3b: RPTOR(ulk1!)>RPTOR(ulk1)+ULK1(straptor) d3
R3c: RPTOR(ulk1!,S792~P,S855_S859~PP).ULK1(straptor!)>RPTOR(ulk1,S792~P,S855_S859~PP)+ULK1(straptor) d3max
R4a: RPTOR(RNC,ulk1,S792~O,S855_S859~O)+EIF4EBP1(RCR,S65_T70~O)->RPTOR(RNC!,ulk1,S792~O,S855_S859~O).EIF4EBP1
R4b: RPTOR(RNC!)EIF4EBP1(RCR!)>RPTOR(RNC
R5a: AMPK(ulk1,T172~P)+ULK1(stampk,S758~O)->AM
R5b: AMPK(ulk1!)ULK1(straptor!)>AMPK(ulk1!)+ULK1
R6: MTOR(HEAT!)RPTOR(WD40!,ulk1!)ULK1(strap
>MTOR(HEAT!)RPTOR(WD40!,ulk1!)ULK1(strapto
R7: MTOR(HEAT!)RPTOR(WD40!,RNC!)EIF4EBP1(R
>MTOR(HEAT!).RPTOR(WD40!,RNC!)EIF4EBP1(R
R8a: RPTOR(ulk1!,S792~O).ULK1(straptor!,S317~P,S
R8b: RPTOR(ulk1!,S855_S859~O).ULK1(straptor!,S3
R8c: RPTOR(ulk1!,S855_S859~P).ULK1(straptor!,S3
R9: ULK1(straptor!,S317~P,S778~P)+AMBRA1(ST~O)>
R10: ULK1(straptor!,S317~P,S778~P)+AMPK(ST~O)U
R11a: AMPK(ulk1!,T172~P,ST~O).ULK1(stampk!strap
R12: AMPK(T172~P)+RPTOR(S792~O)>AMPK(T172~P)
R13a: ULK1(S758~P)>ULK1(S758~O) u0
R13b: EIF4EBP1(S65_T70~P)>EIF4EBP1(S65_T70~O)
R13c: RPTOR(S792~P)>RPTOR(S792~O) u1
R13d: RPTOR(S855_S859~P)>RPTOR(S855_S859~O) u
R13e: RPTOR(S855_S859~PP)>RPTOR(S855_S859~P) u
R13f: AMBRA1(ST~P)>AMBRA1(ST~O) u0
R13g: AMPK(ST~P)>AMPK(ST~O) u2
R13h: ULK1(straptor!,S317~P)>ULK1(straptor!,S317~O) u0
R13i: ULK1(S778~P)>ULK1(S778~O) u0
R13j: RPTOR(S792~P)>RPTOR(S792~O) u0
```

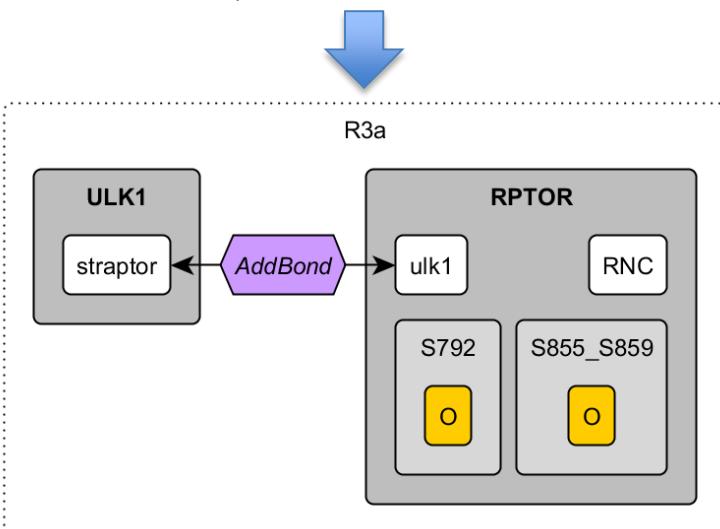
Rules are code!

Why rule visualization is essential

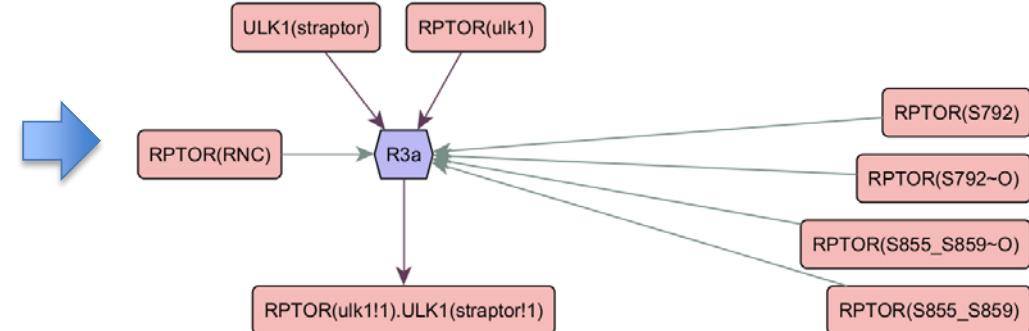
1. Debugging
2. Model reuse / sharing
3. Communicating with collaborators
4. Model comparison
5. Model integration

Automated rule visualization

R3a: RPTOR(RNC,ulk1,S792~O,S855_S859~O) + ULK1(straptor) -> \
RPTOR(RNC,ulk1!1,S792~O,S855_S859~O).ULK1(straptor!1) a3



Compact rule visualization



Regulatory graph

Faeder Lab Current and Past

John Sekar, PhD
Jose Juan Tapia
Robert Sheehan
Cihan Kaya (w/ I. Bahar)
Sanjana Gupta (w/ R. Lee)
Kyle Davidson
Rob Clark, PhD

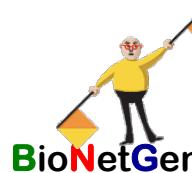


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Natasa Miskov-Zivanov, PhD (Pitt-ECE)
Dipak Barua, PhD (UMIST)

Jintao Liu, PhD (UCSD)
Justin Hogg, PhD (Apple)

Arshi Arora (MSKCC)
Qingyang Ding (Tsinghua)
Tyler McLaughlin (Rice)
Ilya Korsunksy (NYU)
Mate Nagy (Yale)
Niketh Nair (Oak Ridge)
Lori Stover (Macmillan)

Computational Collaborators



RuleBender
NFSim
Atomizer

MCell
WESTPA

Experimental Collaborators

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Michael Turner
William Hawse
Ashley Menk

Sarah Gaffen (PITT-Immunology)

Peter Setlow (UCHC)
Yongqing Li (ECU)
Shiwei Wang

Alexander Sorkin (PITT-Cell Biology)

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