Rule-Based Kinetic Modeling of Signal Transduction Networks

Part I. Motivation

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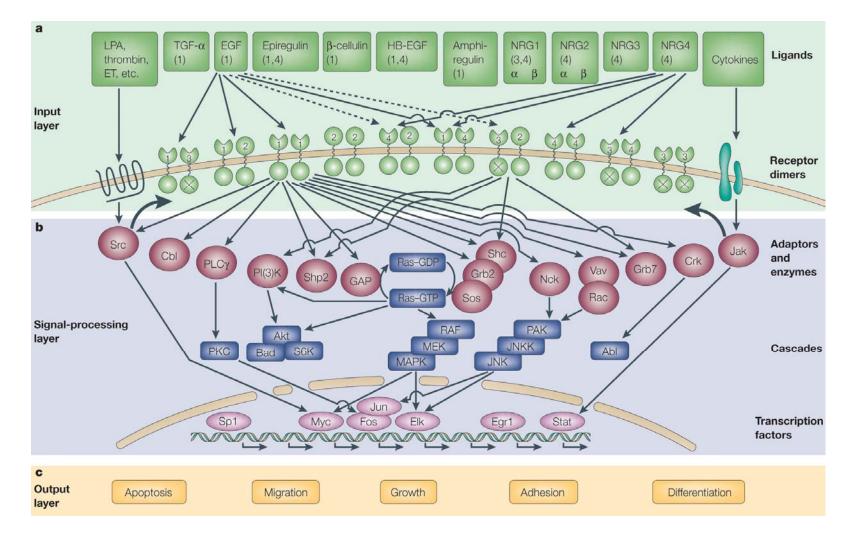
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Ambarish Nag Michael Monine



Example 1: Early events in signaling by Epidermal Growth factor Receptor

Large networks of proteins and other molecules are involved in signaling

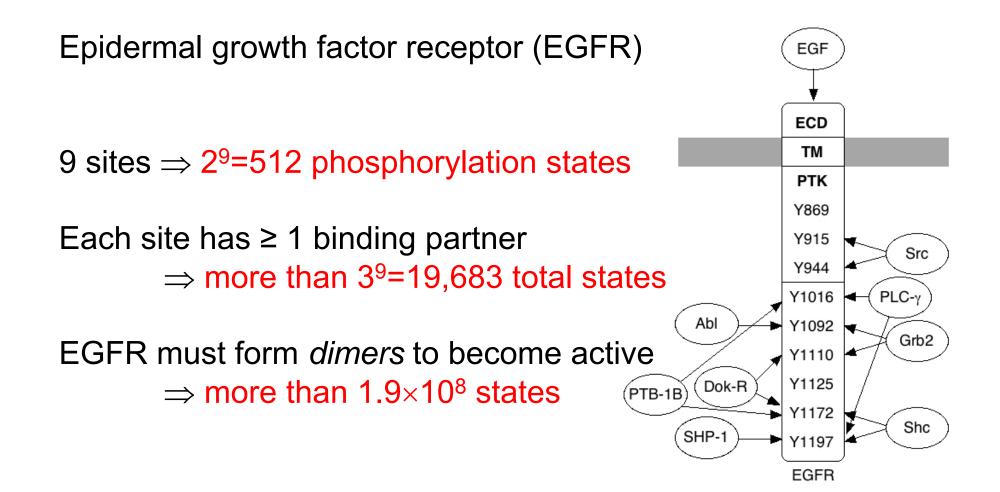


Yarden & Sliwkowski, Nature Rev. Mol. Cell Biol. 02: 127-137 (2001).

Phenomenological vs. Mechanistic Modeling

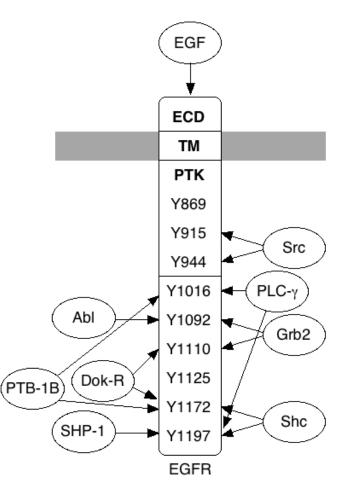
- Type of model depends on the questions one wants to ask (and answer).
- *Phenomenological models* are good for establishing correlations among the measured variables.
- Mechanistic models attempt to put known information into a model that can describe data and make predictions about how manipulating the components affects the outcome.

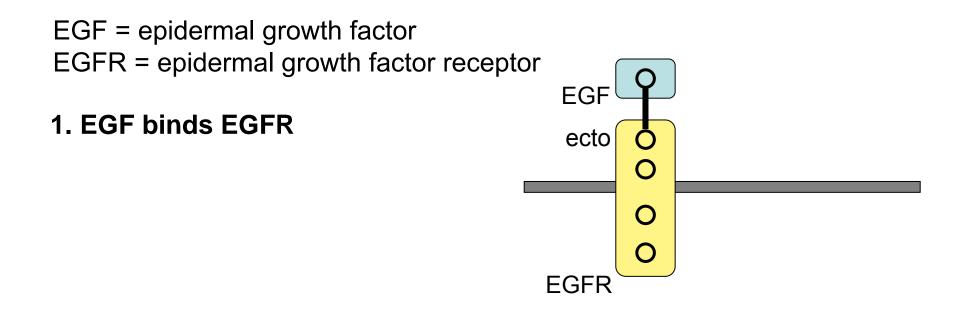
Multiplicity of sites and binding partners gives rise to combinatorial complexity



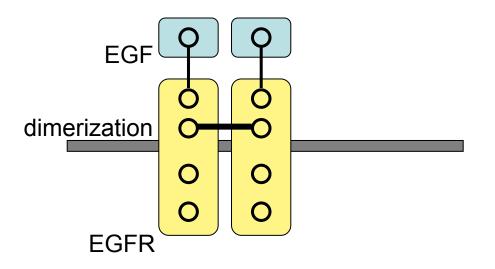
Multiplicity of sites and binding partners gives rise to combinatorial complexity

...but the number of interactions is relatively small.

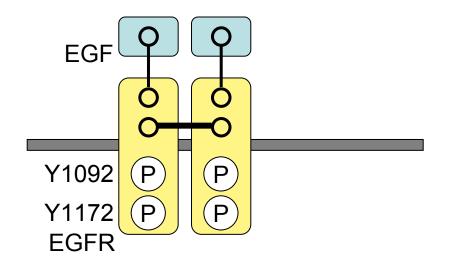




- 1. EGF binds EGFR
- 2. EGFR dimerizes

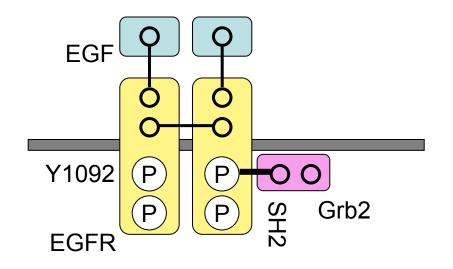


- 1. EGF binds EGFR
- 2. EGFR dimerizes
- 3. EGFR transphosphorylates itself



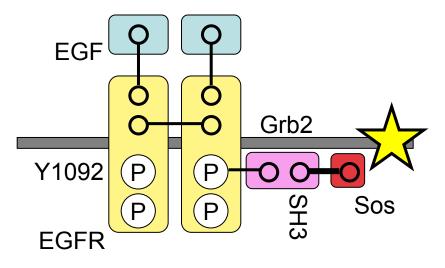
Grb2 pathway

- 1. EGF binds EGFR
- 2. EGFR dimerizes
- 3. EGFR transphosphorylates itself
- 4. Grb2 binds phospho-EGFR

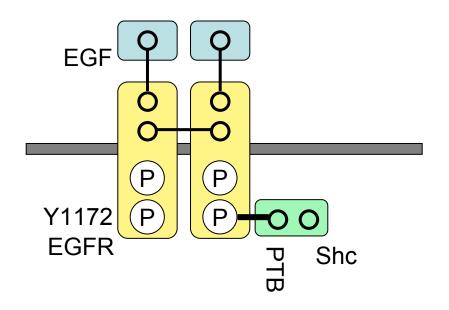


Grb2 pathway

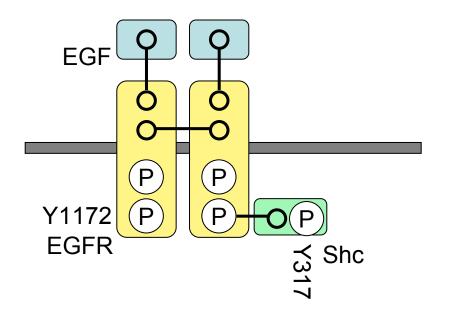
- 1. EGF binds EGFR
- 2. EGFR dimerizes
- 3. EGFR transphosphorylates itself
- 4. Grb2 binds phospho-EGFR
- 5. Sos binds Grb2 (Activation Path 1)



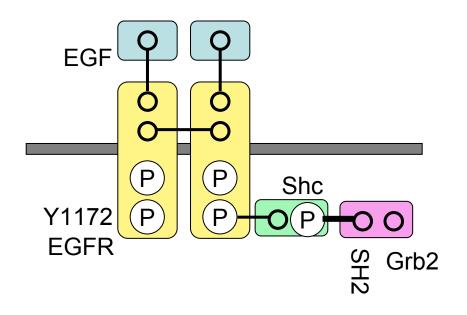
- 1. EGF binds EGFR
- 2. EGFR dimerizes
- 3. EGFR transphosphorylates itself
- 4. Shc binds phospho-EGFR



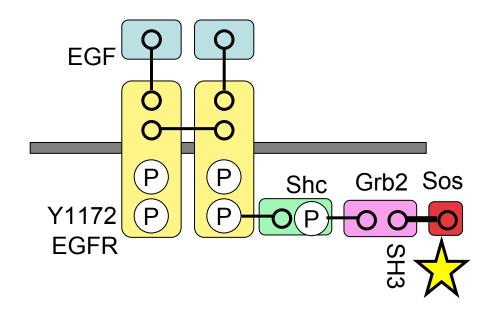
- 1. EGF binds EGFR
- 2. EGFR dimerizes
- 3. EGFR transphosphorylates itself
- 4. Shc binds phospho-EGFR
- **5. EGFR transphosphorylates Shc**



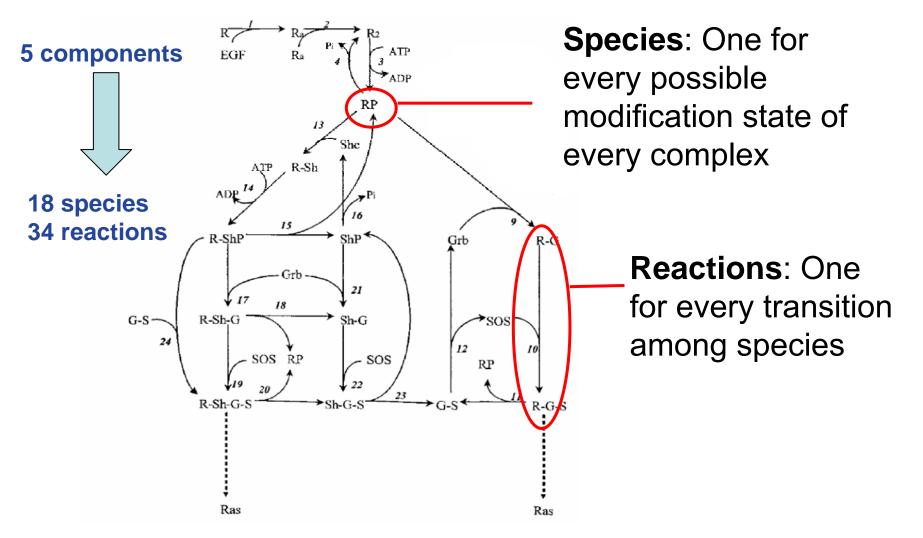
- 1. EGF binds EGFR
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- 5. EGFR transphosphorylates Shc
- 6. Grb2 binds phospho-Shc



- 1. EGF binds EGFR
- 2. EGFR dimerizes
- 3. EGFR transphosphorylates itself
- 4. Shc binds phospho-EGFR
- 5. EGFR transphosphorylates Shc
- 6. Grb2 binds phospho-Shc
- 7. Sos binds Grb2 (Activation Path 2)



A conventional model for EGFR signaling

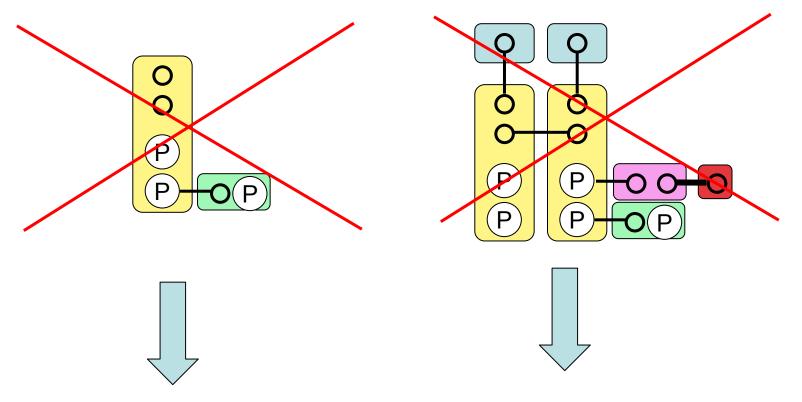


Kholodenko et al., J. Biol. Chem. 274, 30169 (1999)

Excluded from the scheme

No modified monomers

No complexes



1. Phosphorylation inhibits dimer breakup

2. Adaptor binding is competitive

Summary of the conventional approach

- Combinatorial complexity gives rise to a multitude of species and reactions.
- Modelers assume (often implicitly) only some of these combinations are important.
- Assumptions are based on convenience rather than physical knowledge.
- Assumptions may be valid under some conditions, but not others.
- These assumptions cannot be tested without addressing combinatorial complexity.

Experiments probe

affinities for multiple the kinetics of multiple **binding partners** phosphorylation sites EGFR DAPP1 ABL1 ABL2 FGR SRC SLA2 SLA2 SLA2 PIK3R2 PIK3R3 PLCG2 PIK3R PLCG1 BMX GRB2 GRB7 GRB10 FES YES1 MATK 1.2 GRB14 EB-1 GRAP2 EGFR CCM2 CRK Relative Phosphorylation 1.0 CRKL TLN1 pY1172 EPS8L2 NCK1 GULP1 NCK2 0.8 APS APPL ANKS1 SH3BP2 RABGAP1L LNK 0.6 RABGAP1 SH2B FRS3 O > PTPN6 FRS2 O 0.4 869 915 998 1016 1069 1092 1110 1125 1138 1172 1197 IRS4 🗢 PTPN11 IRS2 O E18941 0.2 IRS1 DOK5L MIST E185634 DOK5 0.0 DOK4 HSH2D DOK2 CTEN 0 5 10 15 20 25 30 35 DOK1 E109111 E136111 E138606 E129946 EGF Stimulation Time (min) E169291 RGS12 E105251 **SHB** NUMB NUMBL TNS MAPK8P2 TENC1 MAPK8P1 TENS1 DAB2 O DAB1 O SH2D1A NPPL1 0 min 5min 10min 30min АРВАЗ EAT2 APBA2 BRDG1 LCP2 APBA1 BLNK RIN1 RIN2 RIN3 SH2D3A CHN1 CHN1 APBB2 APBB1 RASA APBB3

Zhang et al., *Mol. Cell. Proteomics* **4**, 1240 (2005).

Richard B. Jones et al., *Nature* 439, 168-174 (2006).

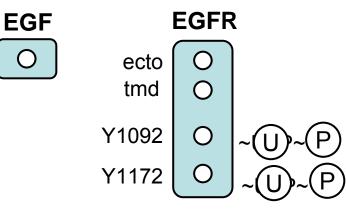
Rule-based modeling is a way to handle combinatorial complexity

Assumption of proteins modularity:

- Signalling molecules consist of functional domains
- Interactions depend on a limited set of features of signalling molecules, and are "local" with respect to these functional domains.
- The evolution of biological system is defined by <u>rules</u> describing activities, potential modifications and interactions of the domains of signaling molecules.
- Computer algorithm <u>automatically generates</u> all molecular species and reactions implied by rules.

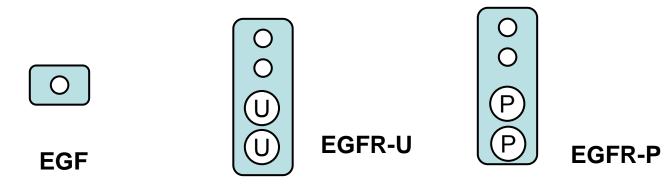
Instead of the list of species a user specifies

a) Biomolecules and their components



Components of proteins may have attributes, e.g. conformation or phosphorylation state.

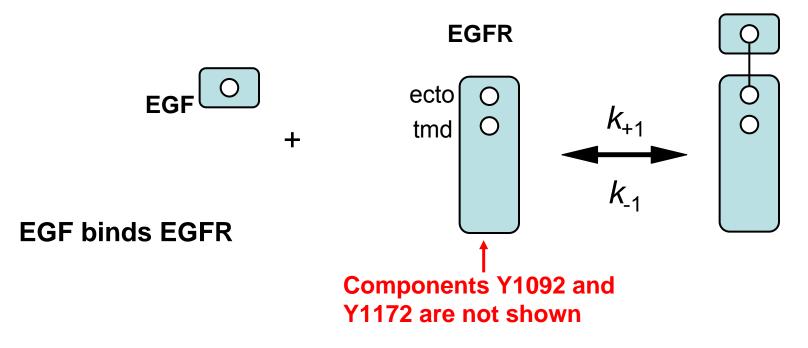
b) Species existing before simulation



Instead of the list of reactions a user specifies

c) Rules that generate reactions and species

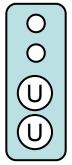
 User specifies a rule for each <u>experimentally-testable</u> feature of the system (*Example: kinetics of ligand-receptor binding is independent of receptor cytosolic tail modifications*).



Rules generate reactions and new chemical species

Initial set of species



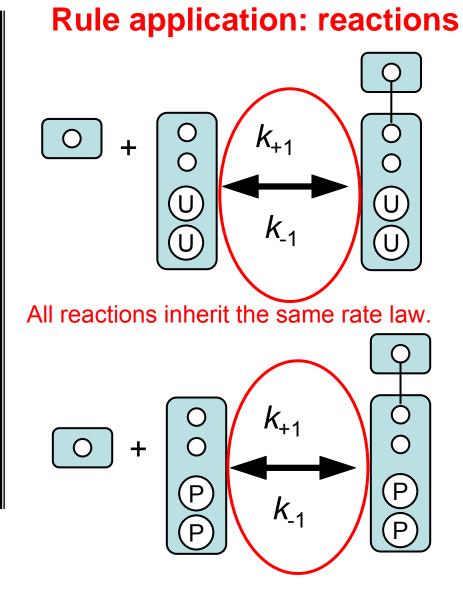


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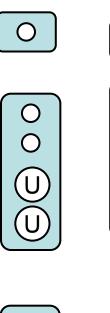
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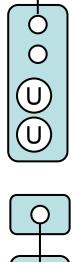
New set of species

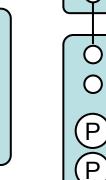


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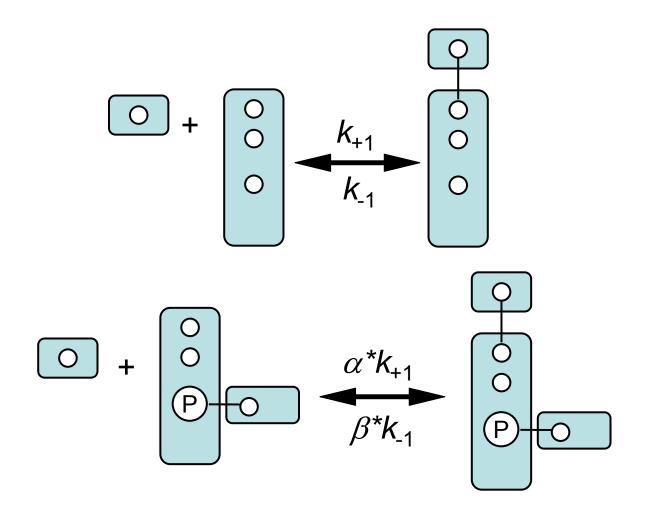
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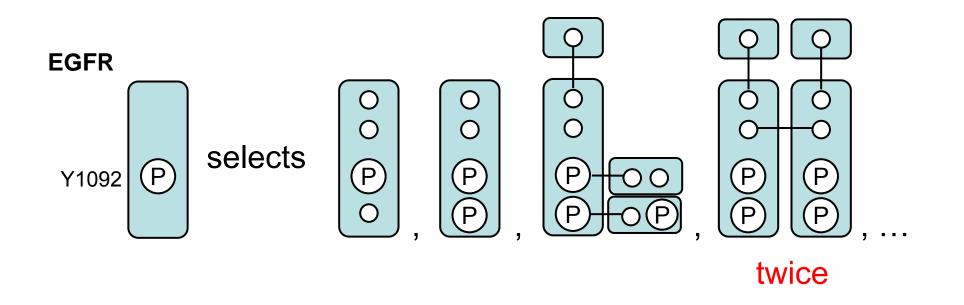


Extendibility and refinement of rules



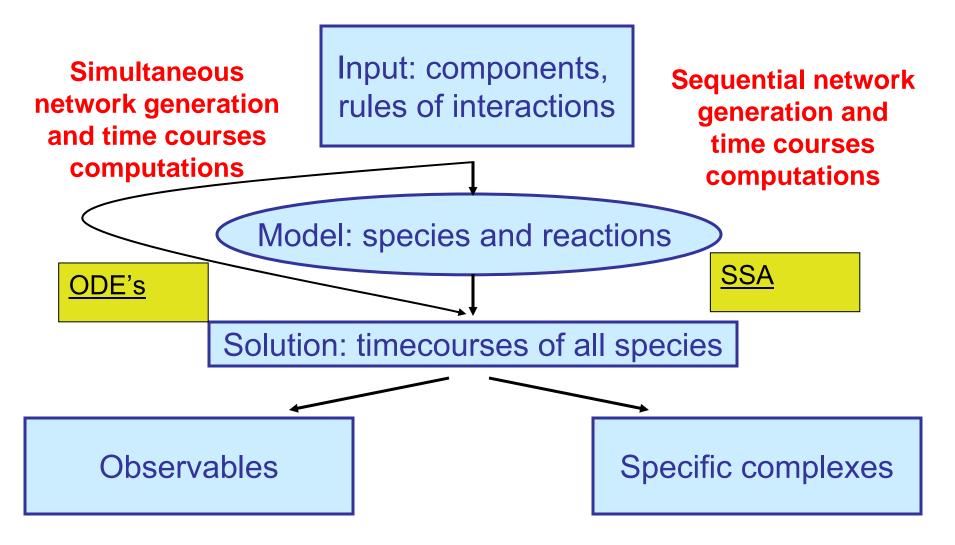
Revise rules to account for context (steric clashes, cooperativity).

Predictions are reported as "observables", corresponding to groups of species with the same properties



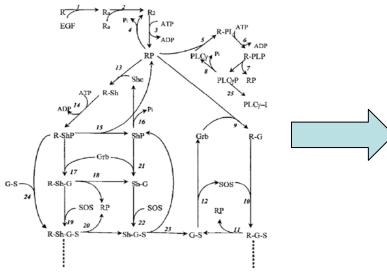
Pattern that selects EGFR phosphorylated at Y1092.

BioNetGen modeling



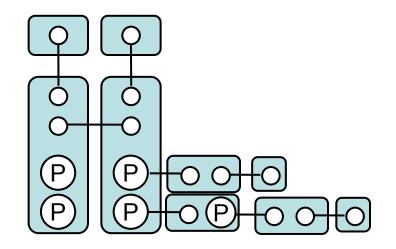
Rule-based version of a reaction scheme

- **18** species
- 34 reactions
- 37 parameters



Kholodenko et al. JBC (1999).

356 species3749 reactions

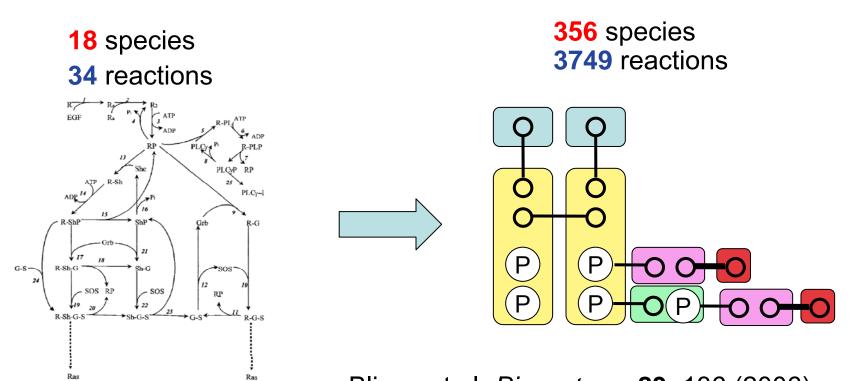


Blinov et al. Biosystems (2006).

- Same number of parameters as in reaction scheme
- Physical basis for rate parameters (e.g. binding constants)

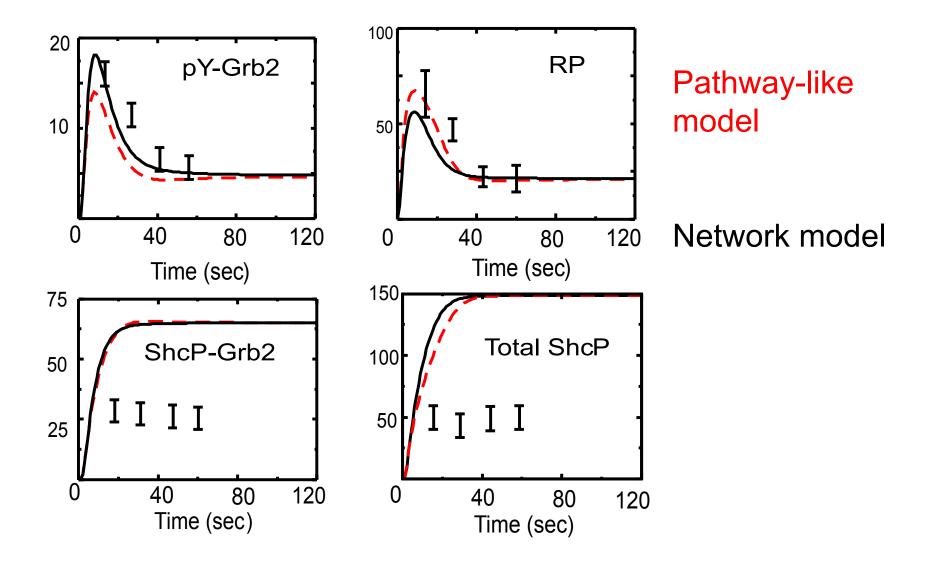
Rule-based version of the Kholodenko model

- 5 molecule types
- 23 reaction rules
- No new rate parameters (!)

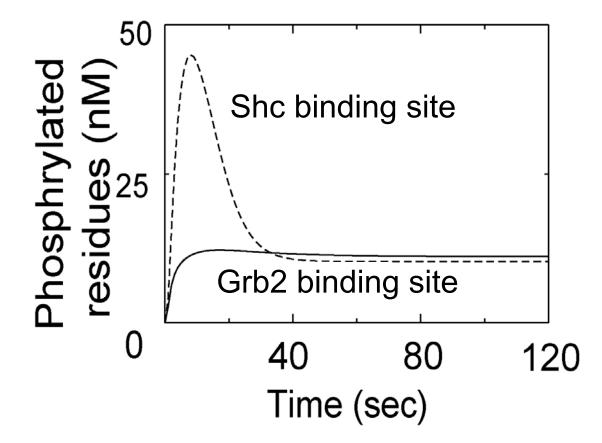


Blinov et al. *Biosystems* 83, 136 (2006).

Fit to experimental measurements



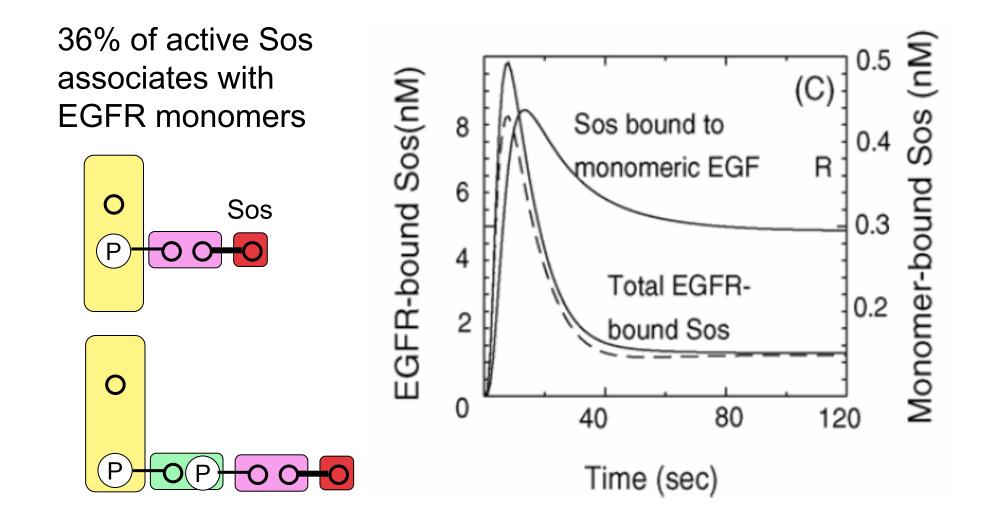
Rule-based model predicts distinct kinetics for two phosphorylation sites



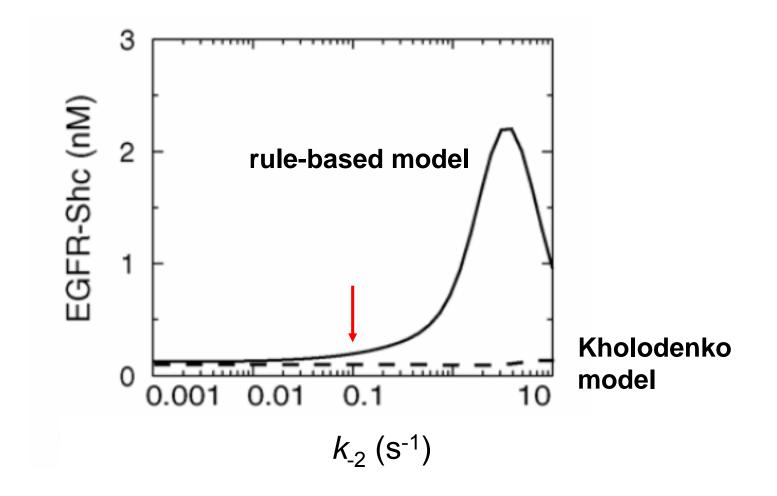
Recent experiments suggest that binding partner can affect phoshporylation kinetics.

Schulze et al., *Mol. Syst. Biol.* (2005)

Also predicts monomers make substantial contribution to steady state Sos activation

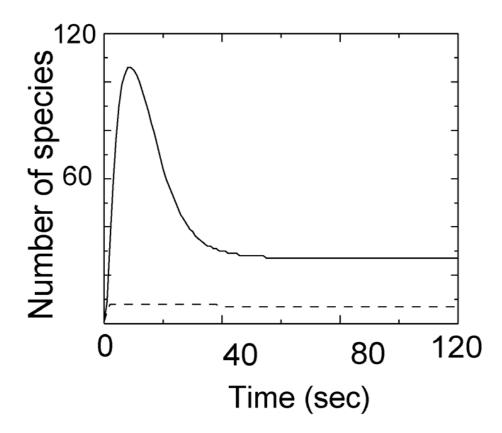


Strong differences when dimer dissociation rate is varied



Molecular diversity

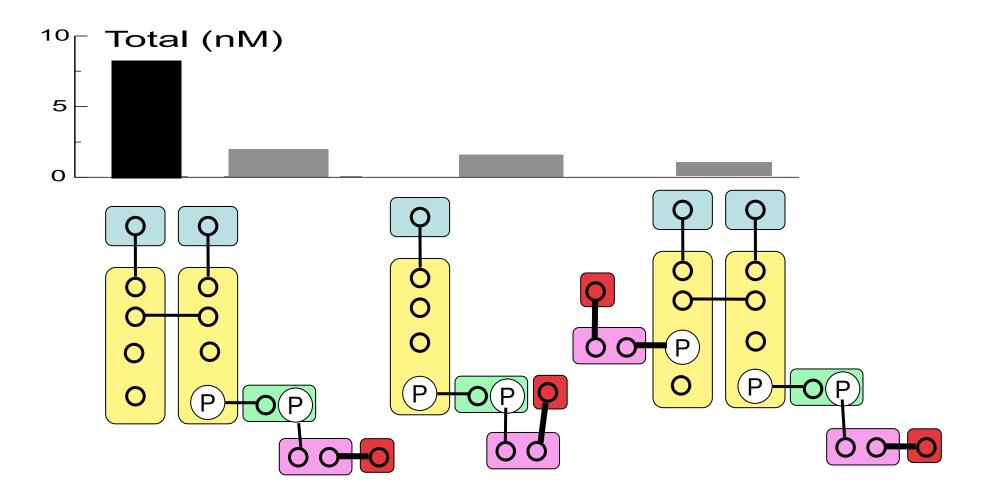
Much larger number of distinct chemical species is predicted to participate in signaling at short times than at steady state

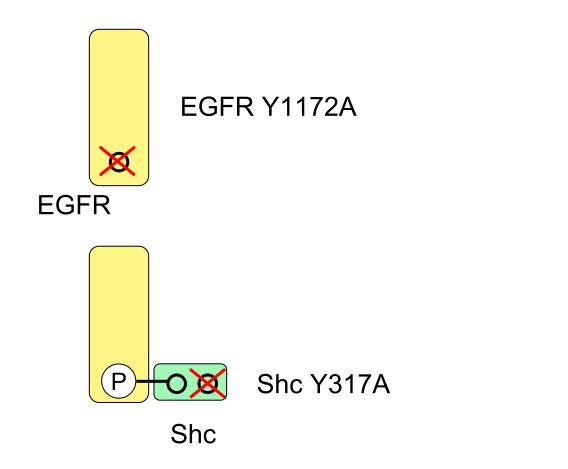


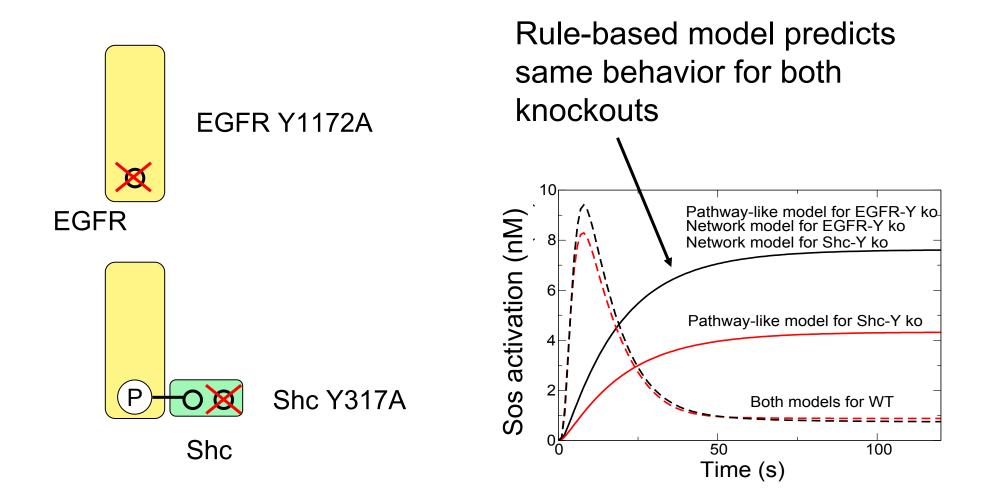
Solid line – network, dashed line – pathway-like

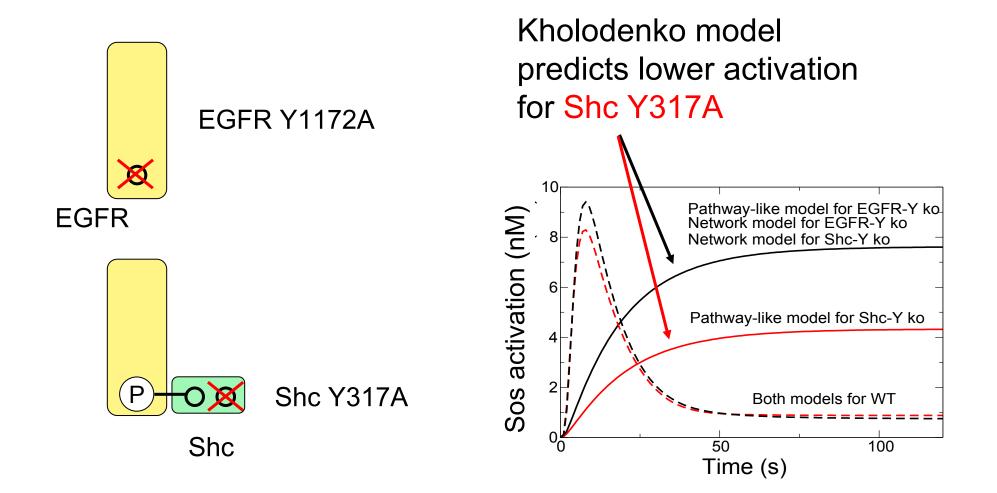
Dominant molecular complexes

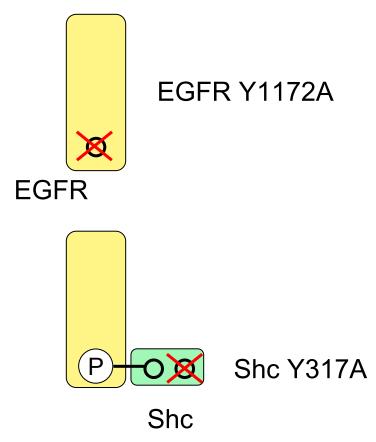
Few chemical species are predicted to account for almost all recruited Sos at steady state.











Kholodenko model predicts lower activation for Shc Y317A

... because mutant Shc blocks binding of Grb2 (competitive binding)

What do we gain

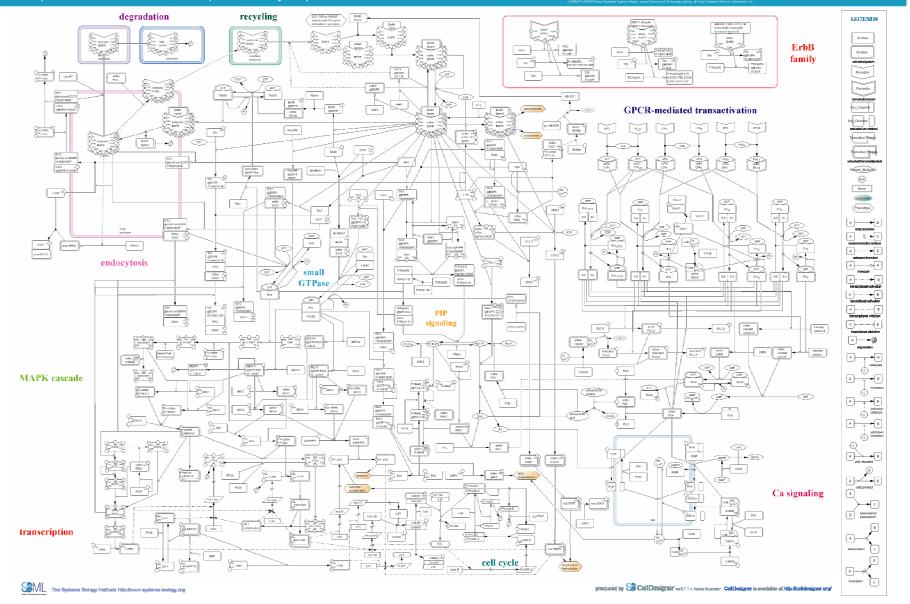
- <u>New quantitative predictions</u> about specific domains, complexes, and interactions, <u>in contact with kind of</u> <u>experiments biologists do</u> (monitoring levels, knocking out and over-expression of specific domains).
- <u>New qualitative predictions</u> (tracing reaction sequences, dominant molecular species).
- Testing <u>hypotheses about signalling mechanisms</u>, e.g. competitive versus non-competitive protein binding.
- Testing <u>effects of specific genetic manipulations</u>, e.g. effects of knock-outs.

BNGL as a collaborative framework

Grand challenge: create a comprehensive rule-based model

Epidermal Growth Factor Receptor Pathway Map

Kanae Oda (1.2), Yukiko Matsuoka (3), Hiroaki Kitano (1.2.4) (1) The System Bology Instate, (3) Department of Fundamental Science and Technology, Mills University,



Spatial modeling meets rule-based approach: BioNetGen at the Virtual Cell

- Rules can be used to generate spatial models accounting for combinatorial complexity:
 - Populate compartments with initial species.
 - Define "compartment-based" rules, with some rules generating species inside compartments and some rules defining trafficking between compartments.
 - Finally, define "spatial-based" rules, with each reactant and product species having a spatial location.