

From Petri Nets to Differential Equations

An Integrative Approach for Biochemical Network Analysis

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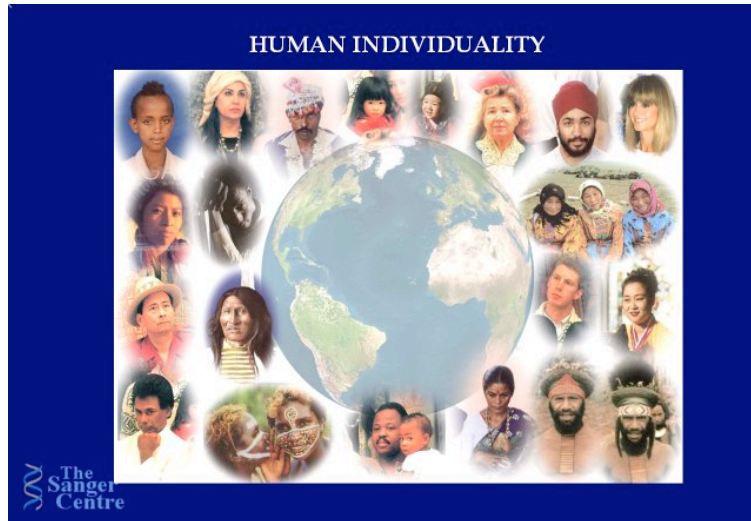
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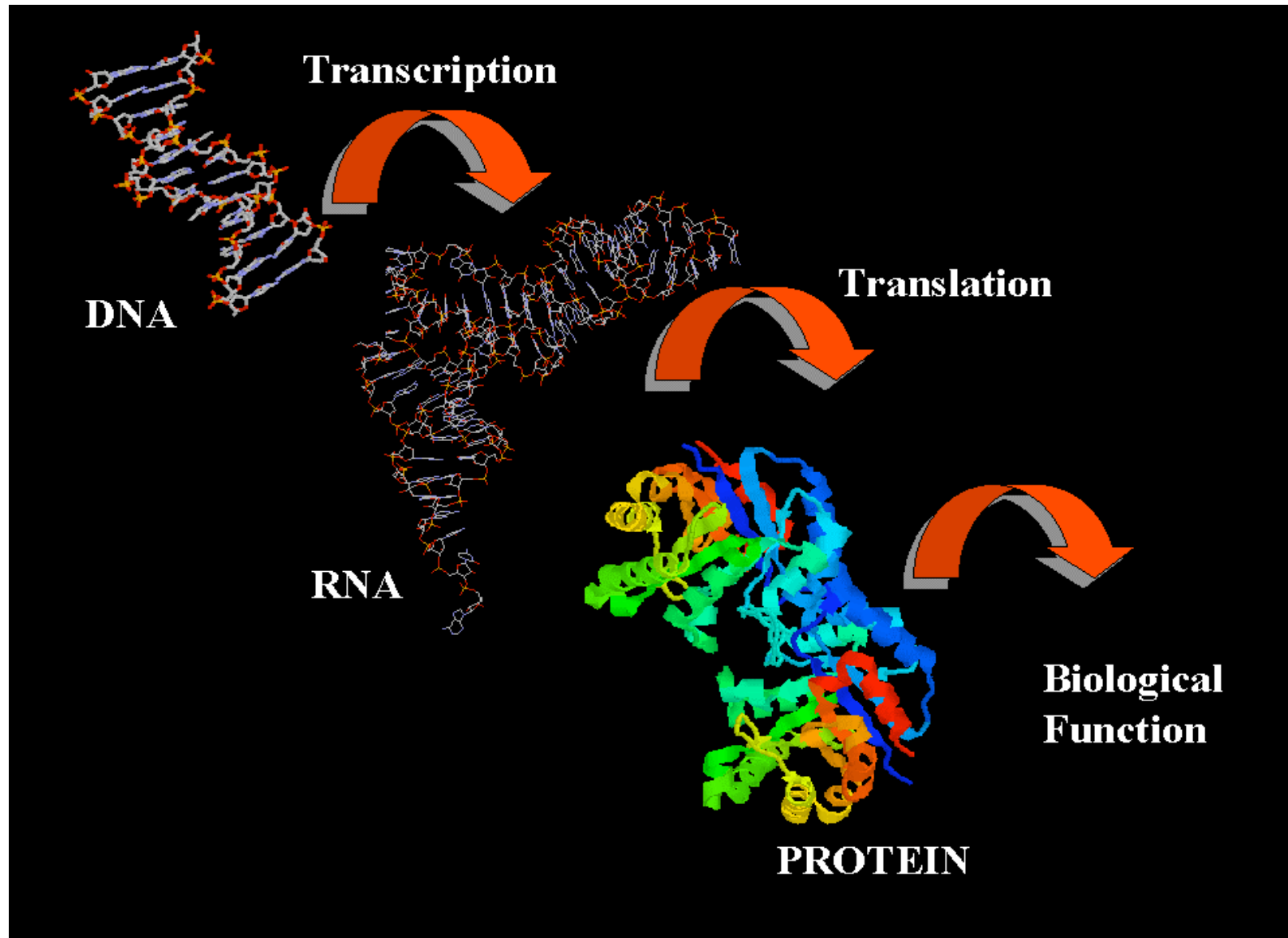
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Long-term vision

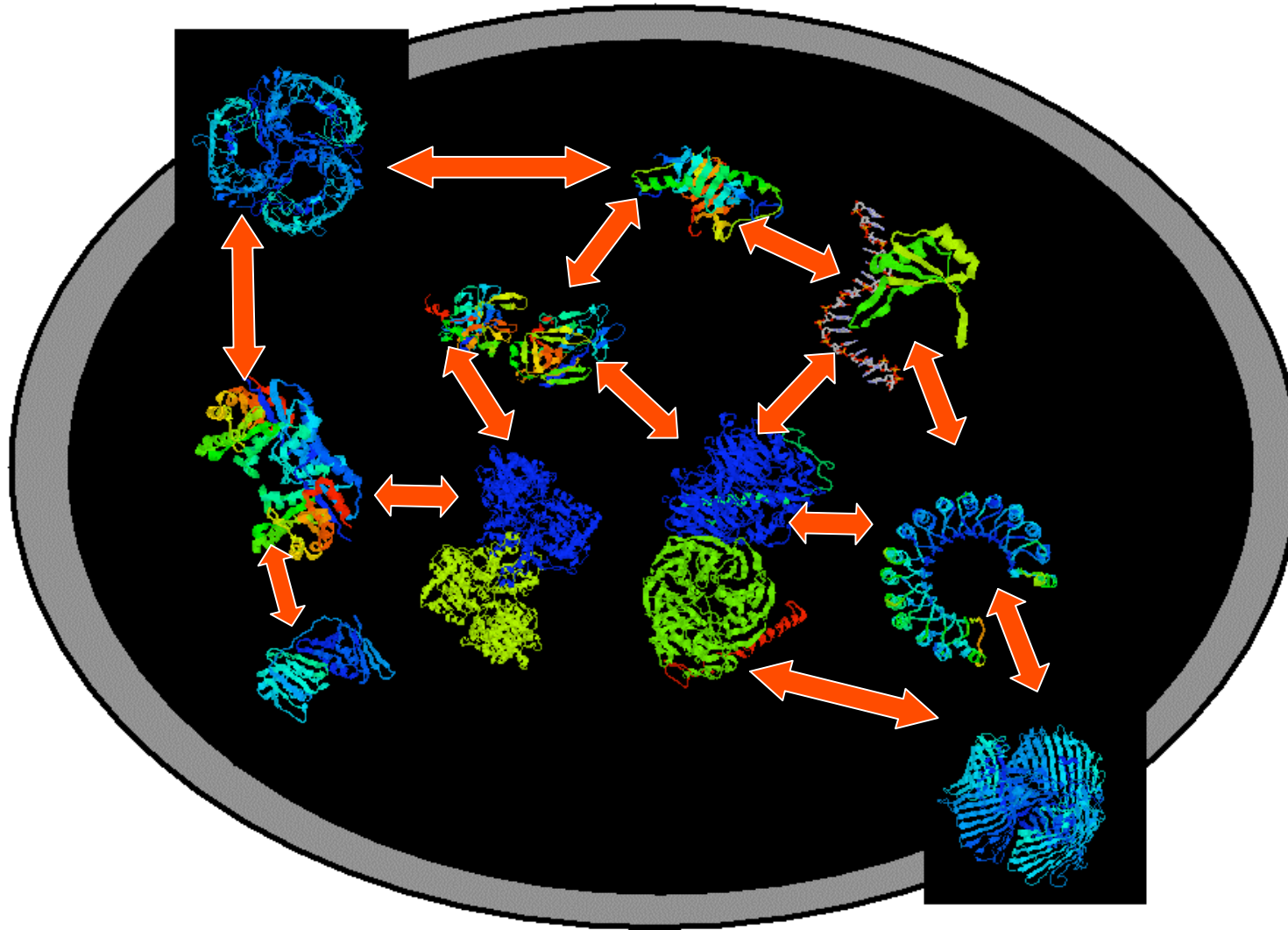


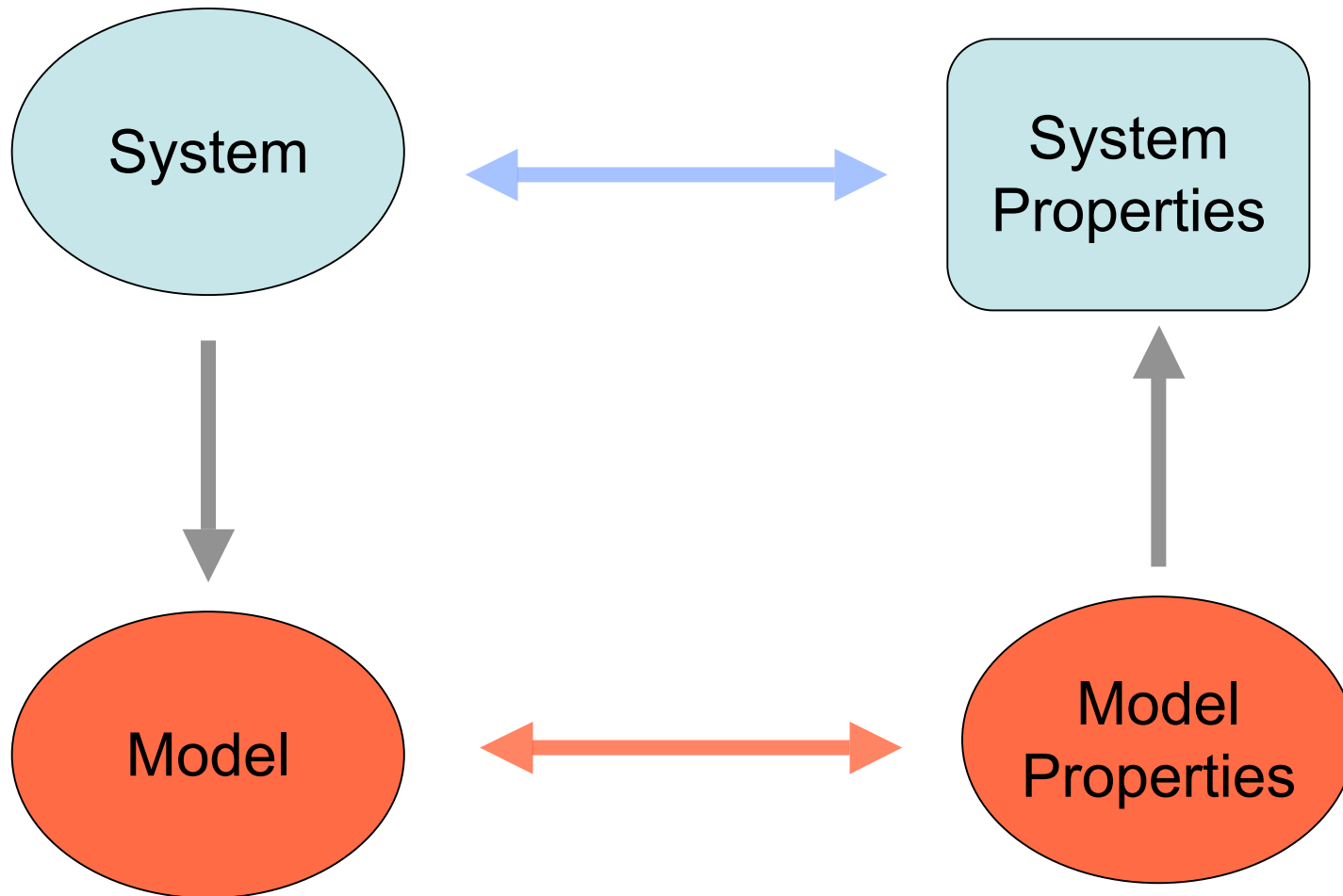
- Current trial-and-error drug prescription procedure
 - "adjust" an individual to a given drug by testing reaction over several weeks
- Vision: model-based drug prescription as part of medical patient treatment by a physician:
 - what and how much is necessary to substitute a given underfunction/malfunction
- Drug discovery process:
 - *in-silico* **quantitative** modelling before major development of drug

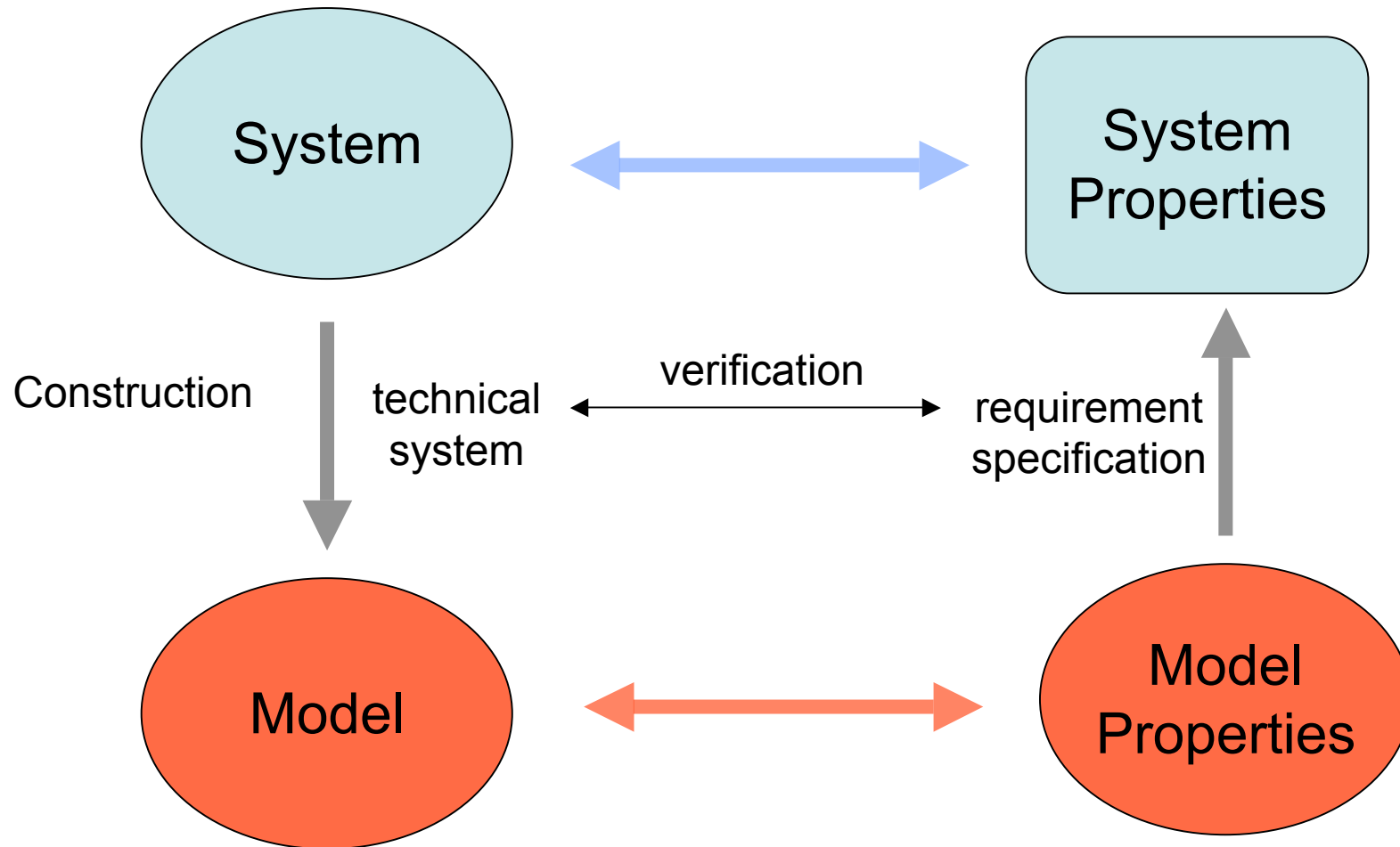
Biological function?

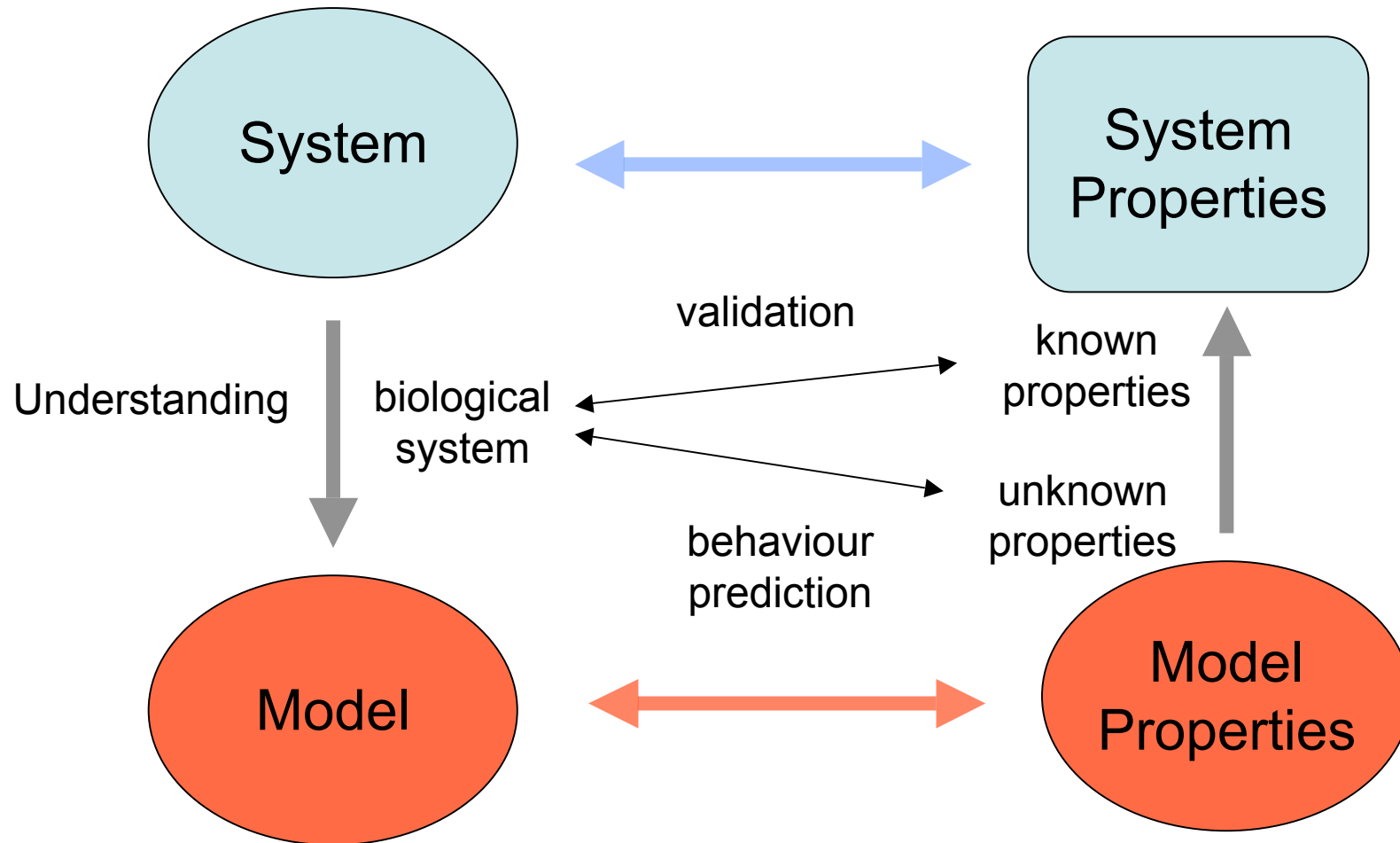


... by interaction in *networks*



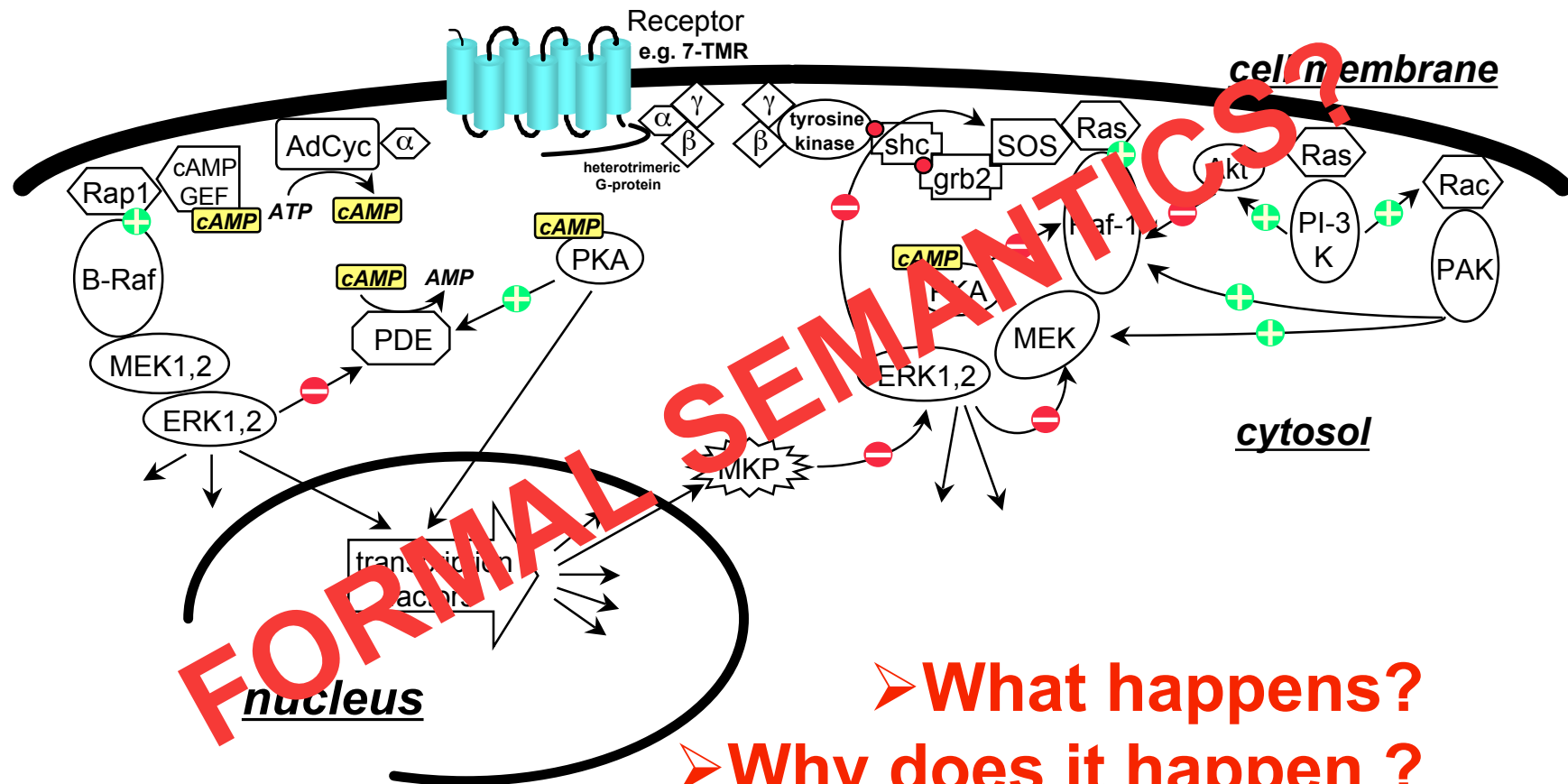






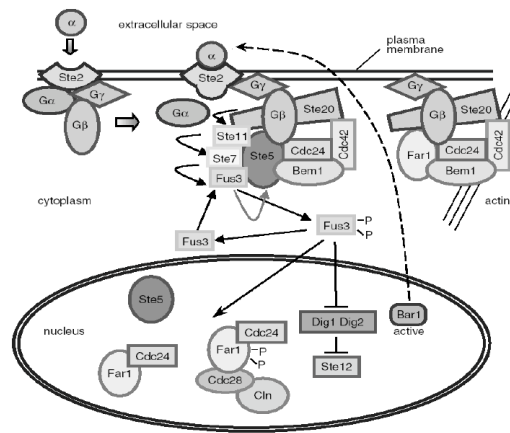
Biochemical networks

We can describe the general topology and single biochemical steps.
However, we do not understand the network function as a whole.



- What happens?
- Why does it happen ?
- How is specificity achieved?

$$\begin{aligned}
\frac{d\alpha}{dt} &= -v_1 \\
\frac{dSte2}{dt} &= -v_2 + v_3 - v_5 \\
\frac{dSte2_{active}}{dt} &= v_2 - v_3 - v_4 \\
\frac{dSst2_{active}}{dt} &= v_4 - v_5 \\
\frac{dG\alpha\beta\gamma}{dt} &= -v_6 + v_9 \\
\frac{dG\alpha GTP}{dt} &= v_6 - v_7 - v_8 \\
\frac{dG\alpha GDP}{dt} &= v_7 + v_8 - v_9 \\
\frac{dG\beta\gamma}{dt} &= v_6 - v_9 - v_{10} + v_{11} + v_{21} + v_{23} + v_{25} + v_{27} + v_{32} \\
&\quad - v_{42} + v_{43} \\
\frac{dSte5}{dt} &= -v_{12} + v_{13} + v_{17} + v_{21} + v_{23} + v_{25} + v_{27} + v_{32} \\
\frac{dSte11}{dt} &= -v_{12} + v_{13} + v_{17} + v_{21} + v_{23} + v_{25} + v_{27} + v_{32} \\
\frac{dSte7}{dt} &= -v_{14} + v_{15} + v_{17} + v_{21} + v_{23} + v_{25} + v_{27} + v_{32} \\
\frac{dFus3}{dt} &= -v_{14} + v_{15} + v_{17} + v_{21} + v_{23} + v_{25} + v_{27} - v_{29} \\
&\quad + v_{30} + v_{33} \\
\frac{dSte20}{dt} &= -v_{18} + v_{19} + v_{21} + v_{23} + v_{25} + v_{27} + v_{32}
\end{aligned}$$

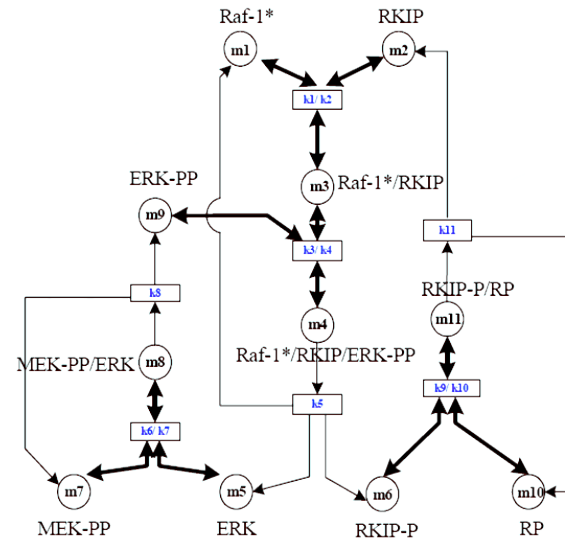


$$\begin{aligned}
v_1 &= \alpha[t] \cdot Bar1_{active}[t] \cdot k_1 \\
v_2 &= Ste2[t] \cdot \alpha[t] \cdot k_2 \\
v_3 &= Ste2_{active}[t] \cdot k_3 \\
v_4 &= Ste2_{active}[t] \cdot k_4 \\
v_5 &= Ste2[t] \cdot k_5 \\
v_6 &= Ste2_{active}[t] \cdot G\alpha\beta\gamma[t] \cdot k_6 \\
v_7 &= G\alpha GTP[t] \cdot k_7 \\
v_8 &= G\alpha GTP[t] \cdot Sst2_{active}[t] \cdot k_8 \\
v_9 &= G\alpha GDP[t] \cdot G\beta\gamma[t] \cdot k_9 \\
v_{10} &= G\beta\gamma[t] \cdot C[t] \cdot k_{10} \\
v_{11} &= D[t] \cdot k_{11} \\
v_{12} &= Ste5[t] \cdot Ste11[t] \cdot k_{12} \\
v_{13} &= A[t] \cdot k_{13} \\
v_{14} &= Ste7[t] \cdot Fus3[t] \cdot k_{14} \\
v_{15} &= B[t] \cdot k_{15} \\
v_{16} &= A[t] \cdot B[t] \cdot k_{16} \\
v_{17} &= C[t] \cdot k_{17} \\
v_{18} &= D[t] \cdot Ste20[t] \cdot k_{18}
\end{aligned}$$

READABILITY?

What is a biochemical network model?

1. Structure



graph
QUALITATIVE

2. Kinetics (if you can)

$$d[\text{Raf1}^*]/dt = k1 * m1 * m2 + k2 * m3 + k5 * m4$$

$$k1 = 0.53; k2 = 0.0072; k5 = 0.0315$$

reaction rates
QUANTITATIVE

3. Initial conditions

$$[\text{Raf1}^*]_{t=0} = 2 \mu\text{Molar}$$

marking , concentrations

Bionetworks: some problems

❑ knowledge

→ PROBLEM 1

- ⇒ uncertain
- ⇒ growing, changing
- ⇒ time-consuming wet-lab experiments
- ⇒ some data estimated
- ⇒ results distributed over independent data bases, papers, journals, . . .

❑ various, mostly ambiguous representations

→ PROBLEM 2

- ⇒ verbose descriptions
- ⇒ diverse graphical representations
- ⇒ contradictory and / or fuzzy statements

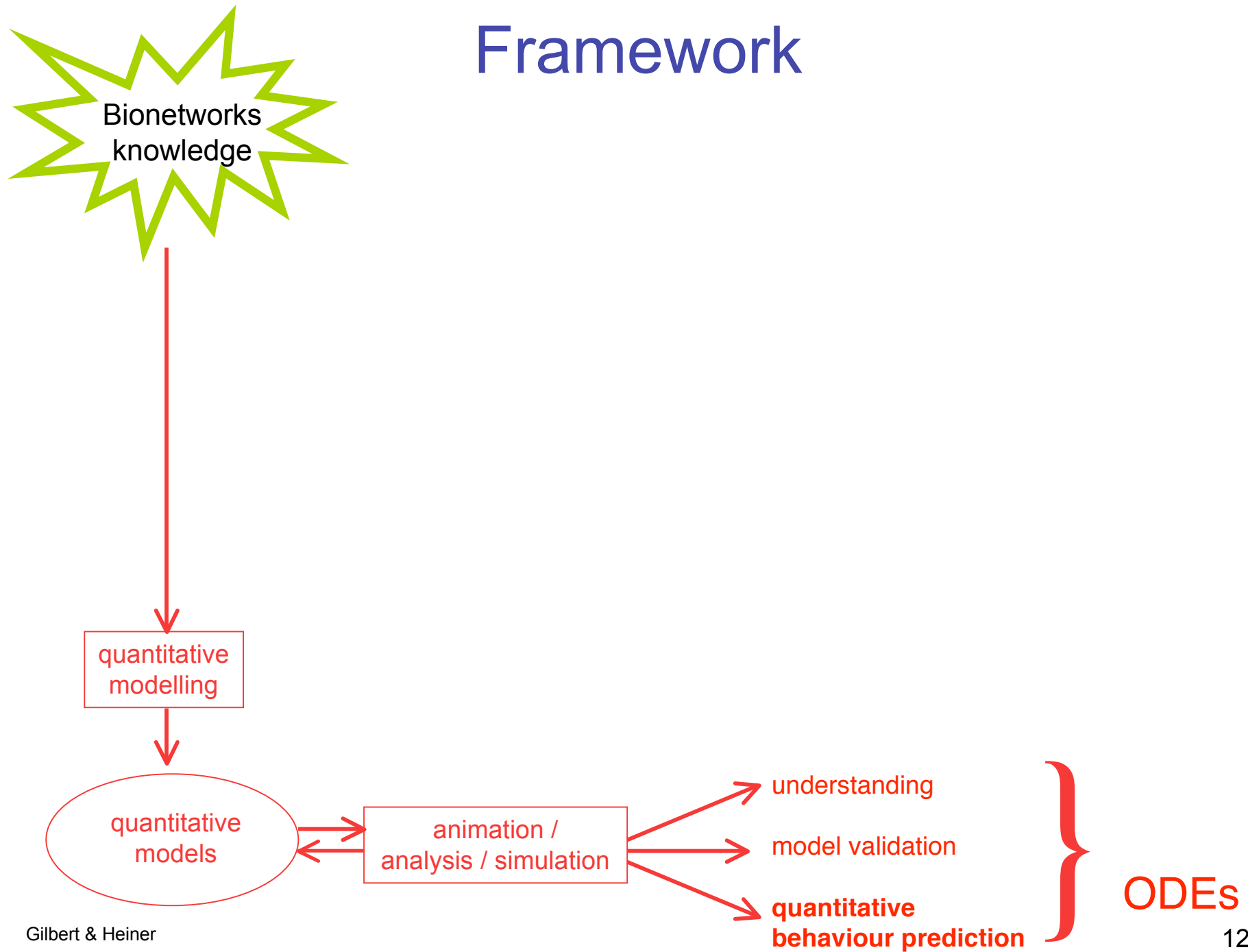
❑ network structure

→ PROBLEM 3

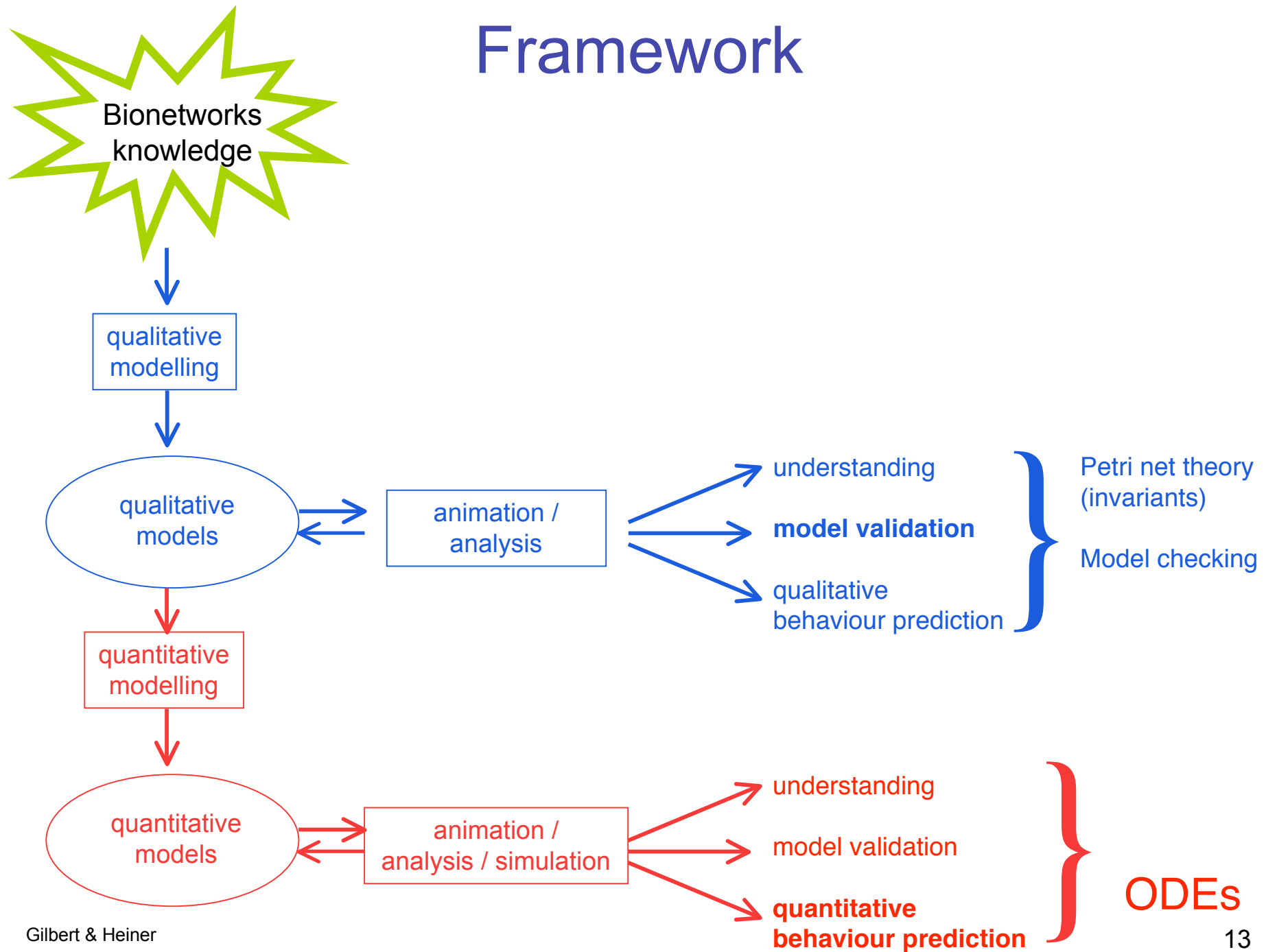
- ⇒ tend to grow fast
- ⇒ dense, apparently unstructured
- ⇒ hard to read

⇒ models are full of ASSUMPTIONS ⇐

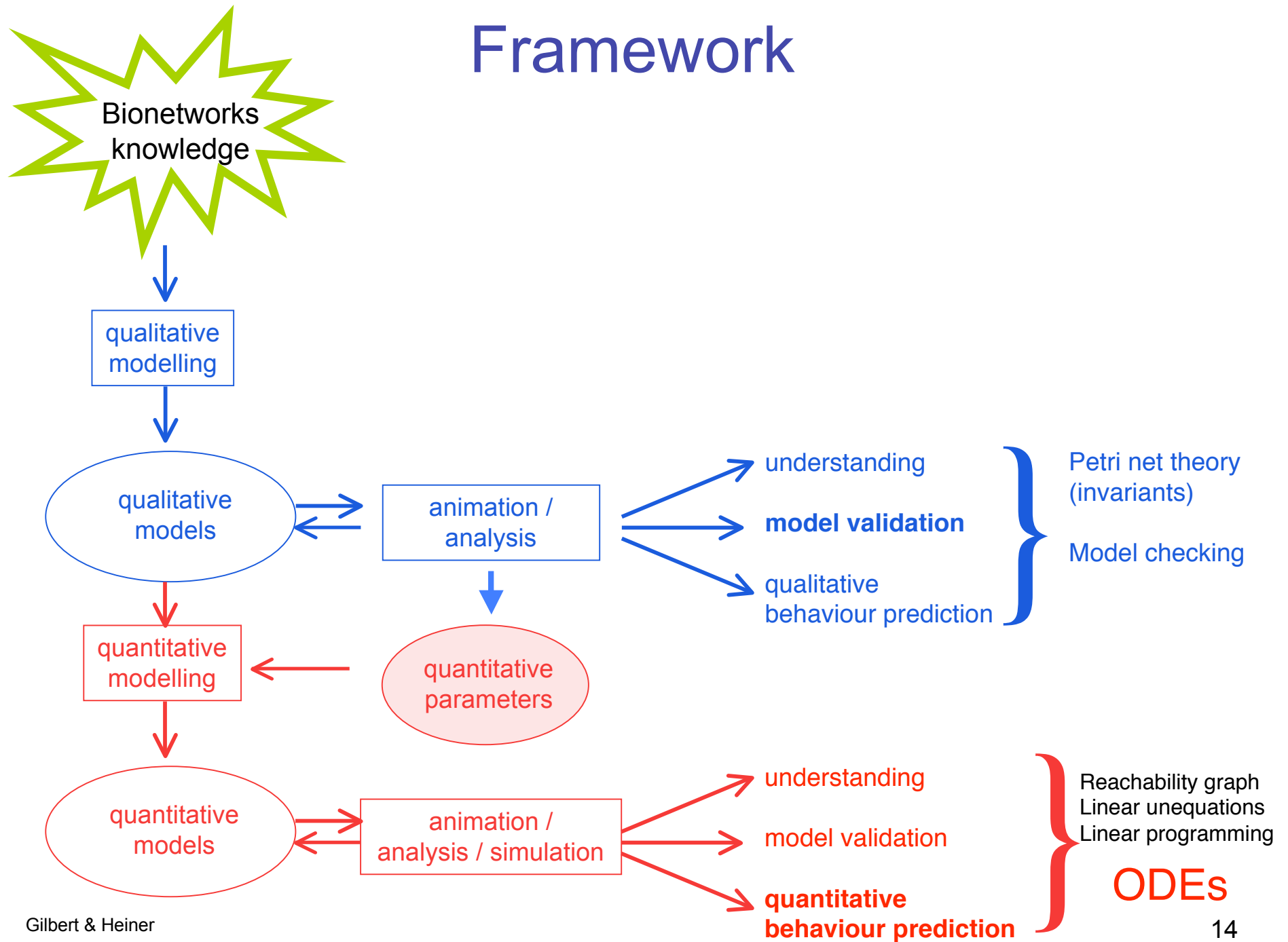
Framework



Framework

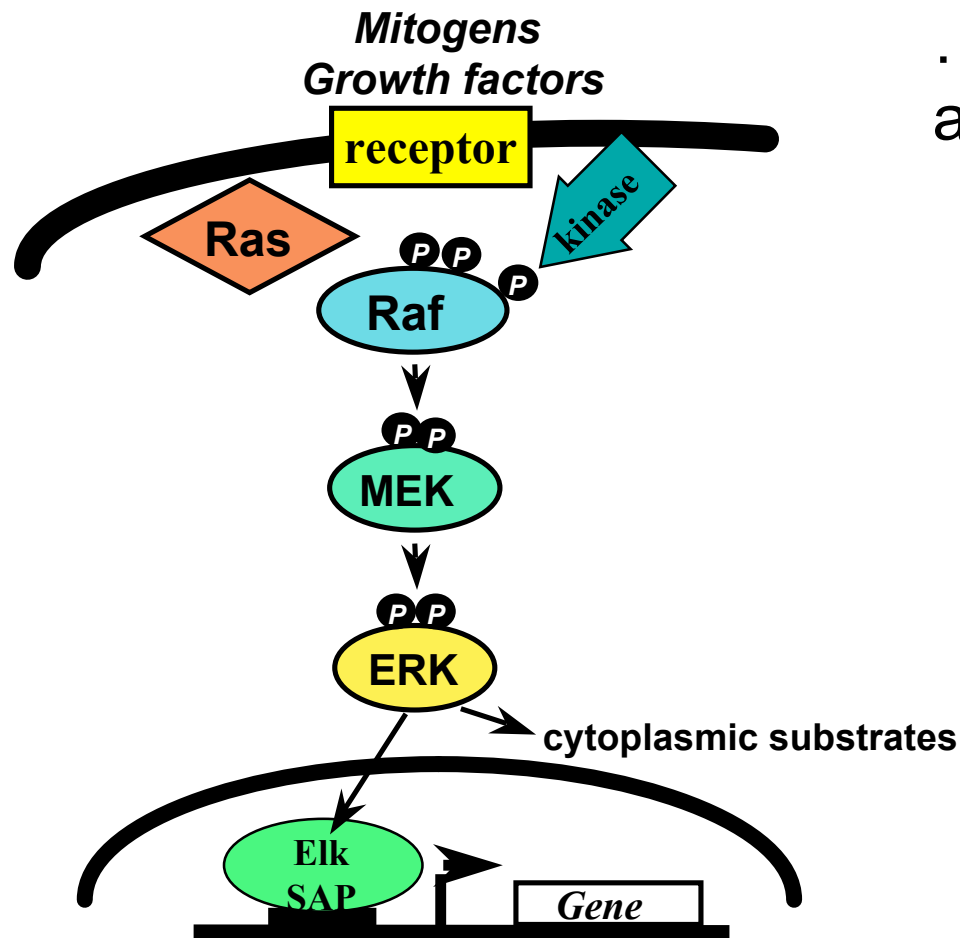


Framework



Ras-Raf-MEK-ERK signalling pathway

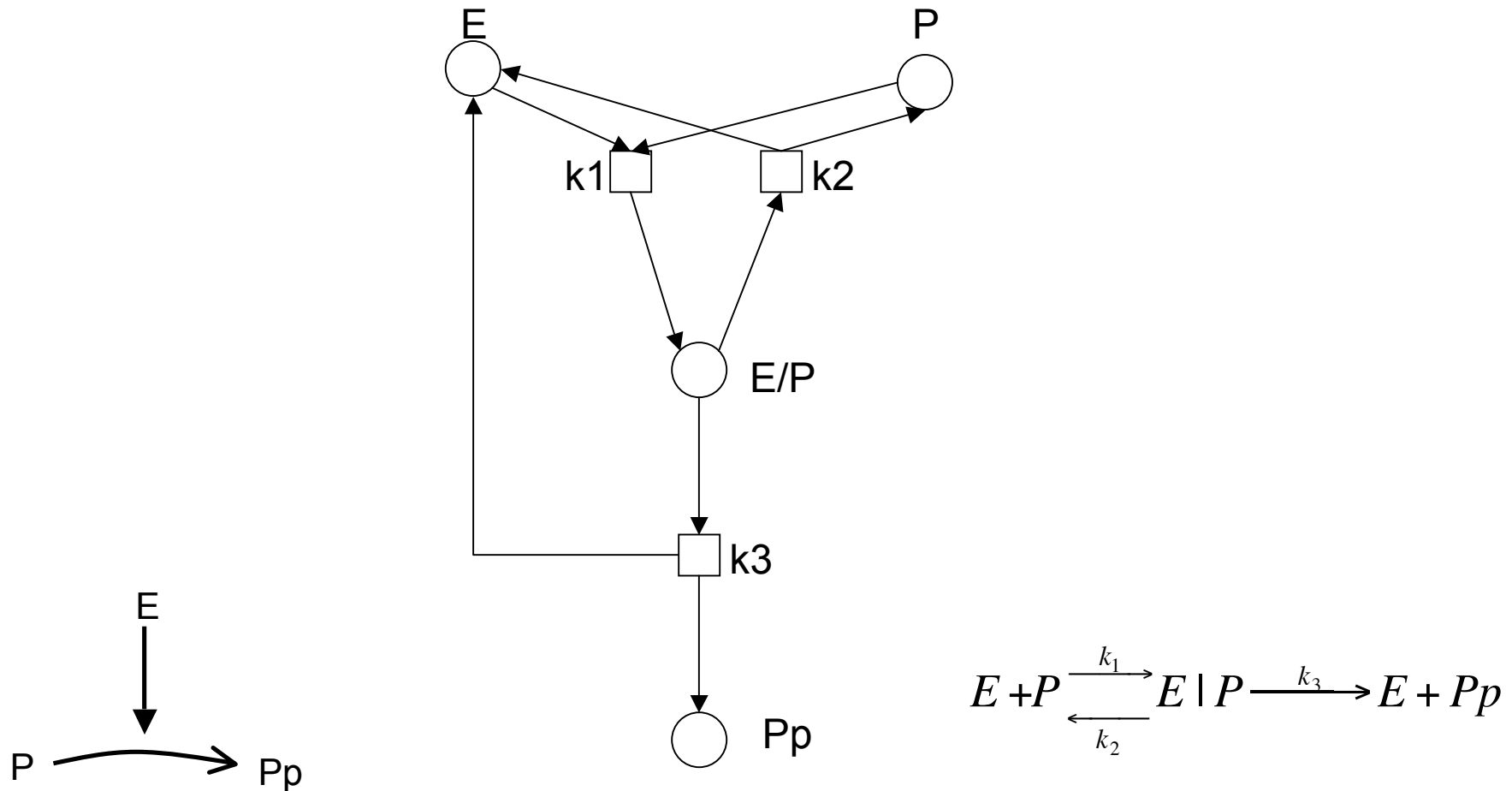
...one pathway...



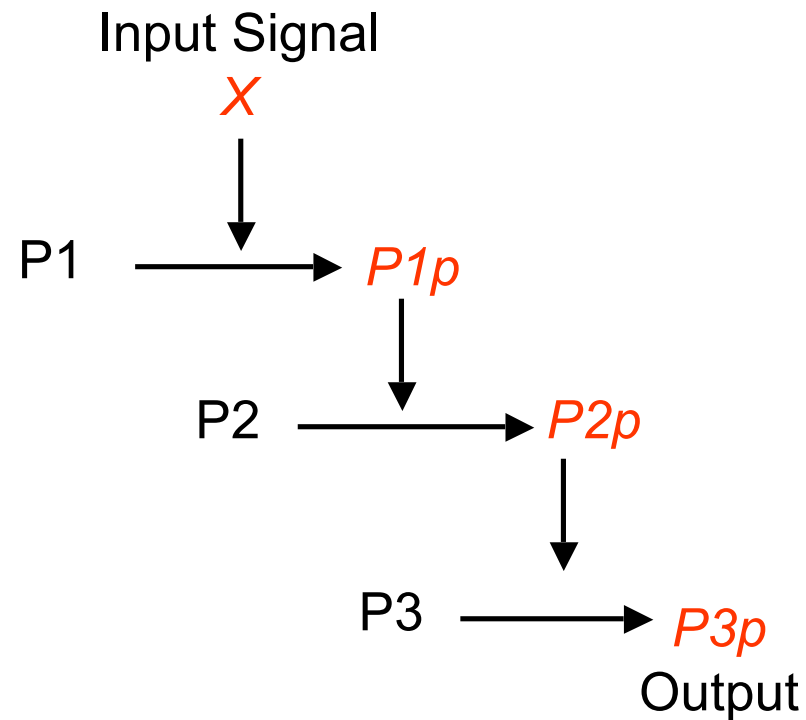
...mediates many functions, and hence diseases...

- Proliferation
- Cell cycle
- Differentiation
- Transformation
- Survival
- Cytoskeleton
- Adhesion
- Motility
- etc.

Network building block - enzymatic reaction



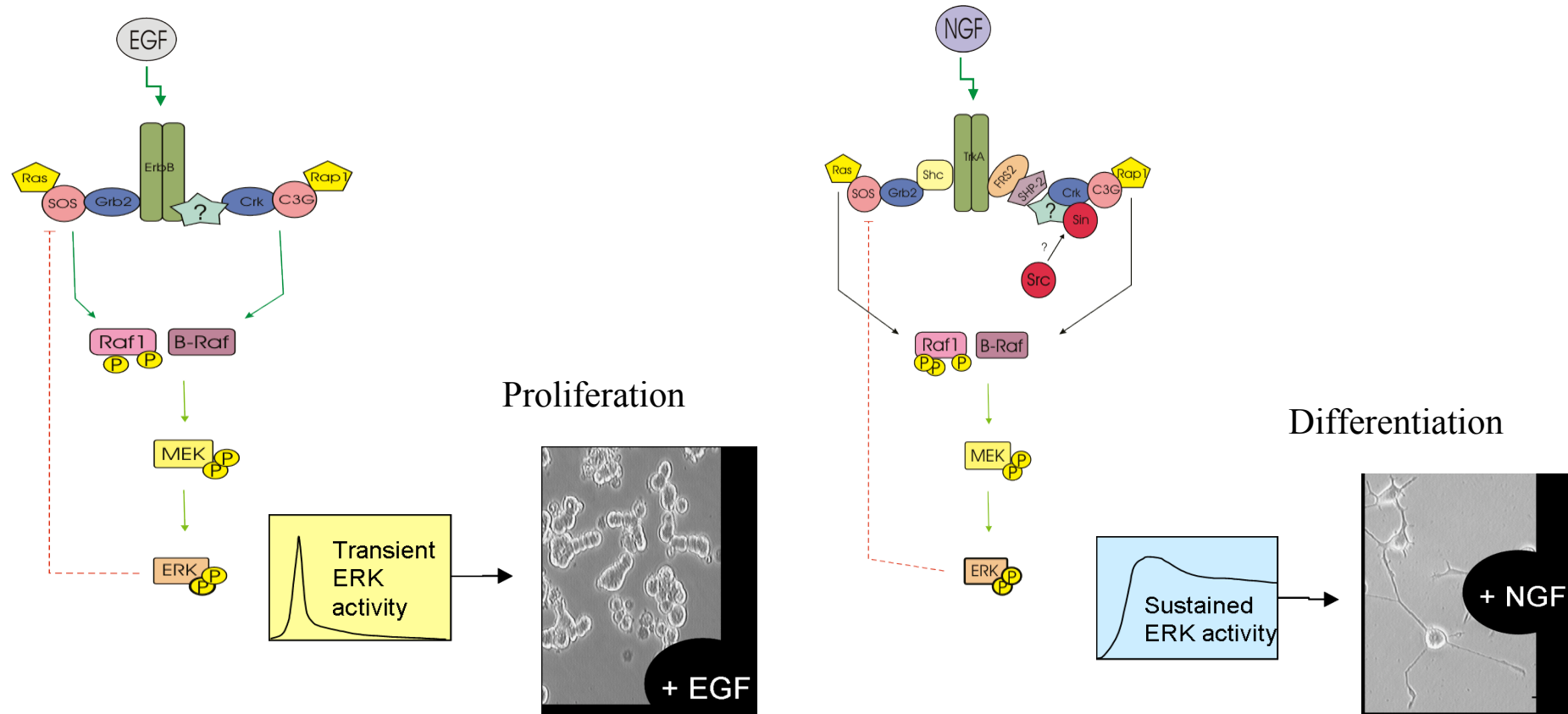
Signalling cascade of enzymatic reactions



Product become enzyme at next stage

Continuous system & model

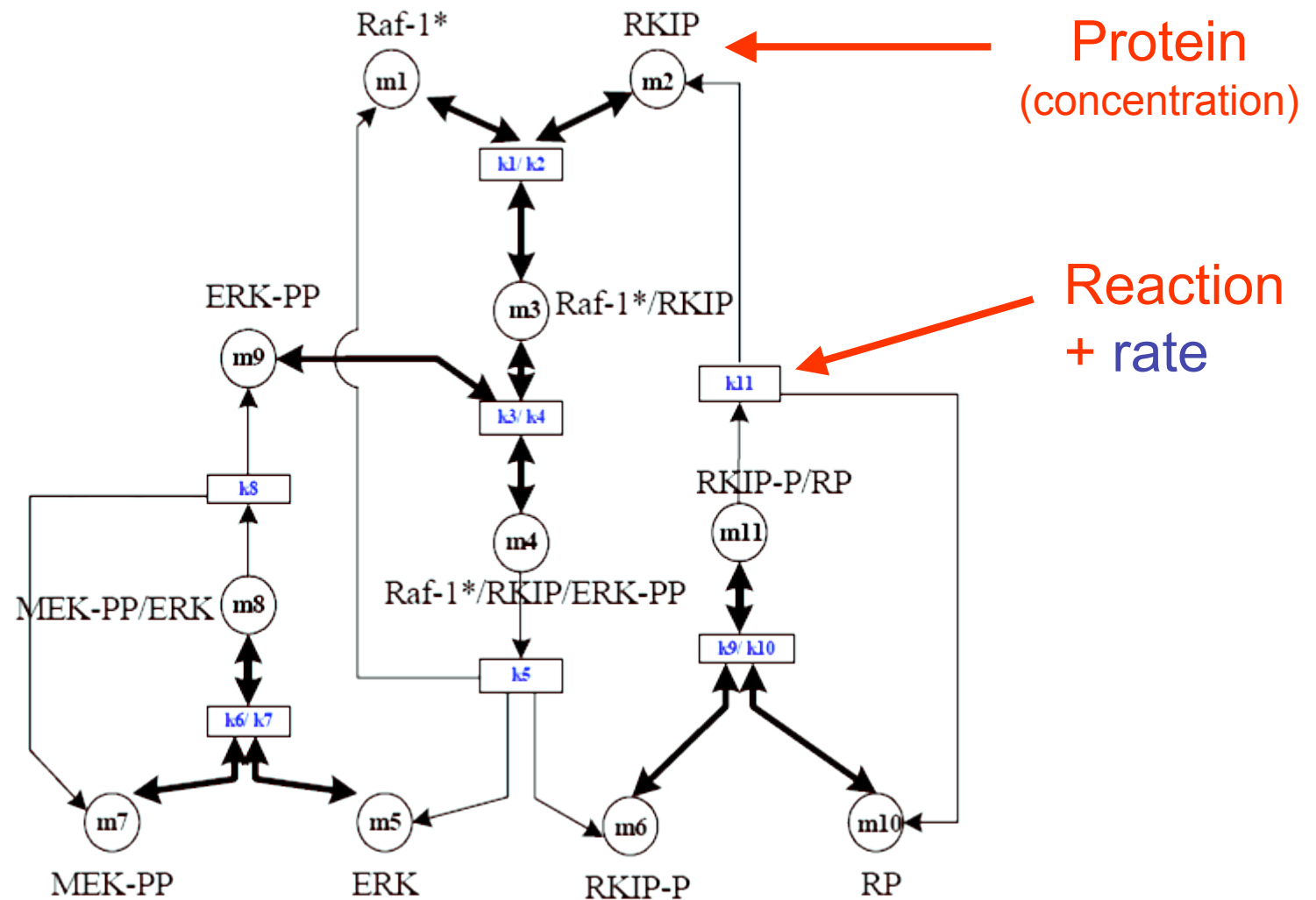
Effect of addition of EGF vs NGF

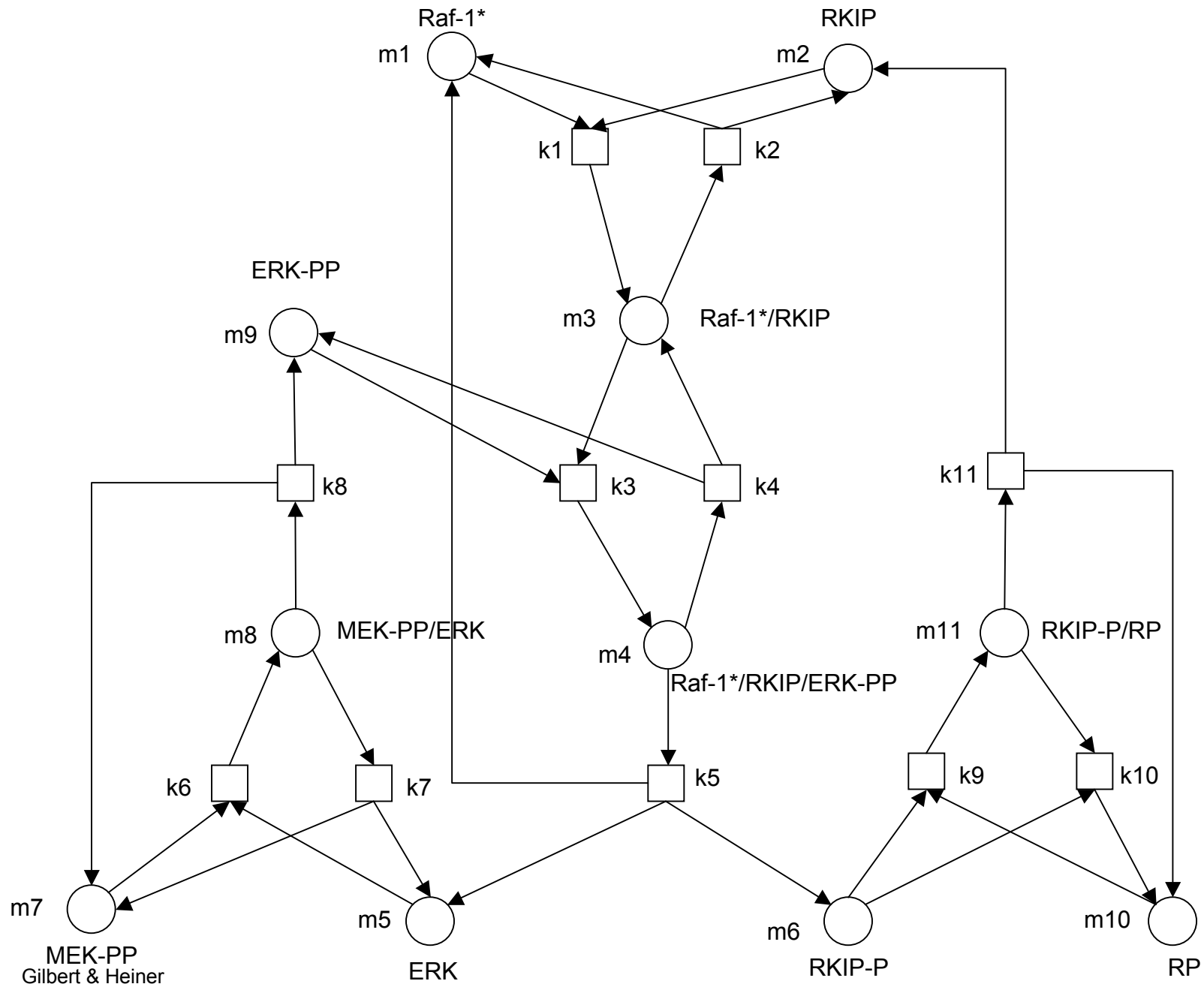


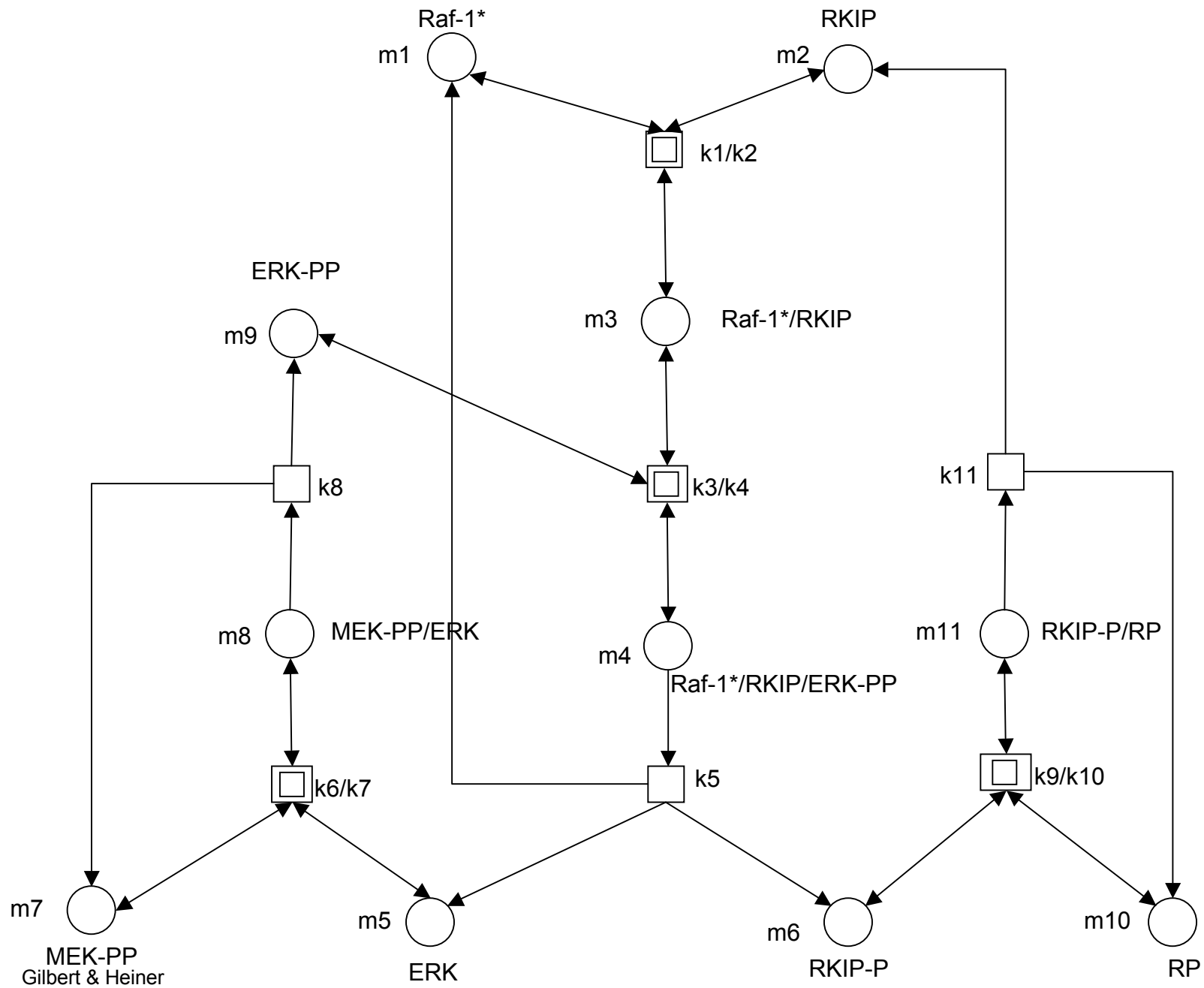
Case study: small model network

RKIP inhibited ERK pathway

Cho et al, CMSB03







Step 1 - Qualitative analysis

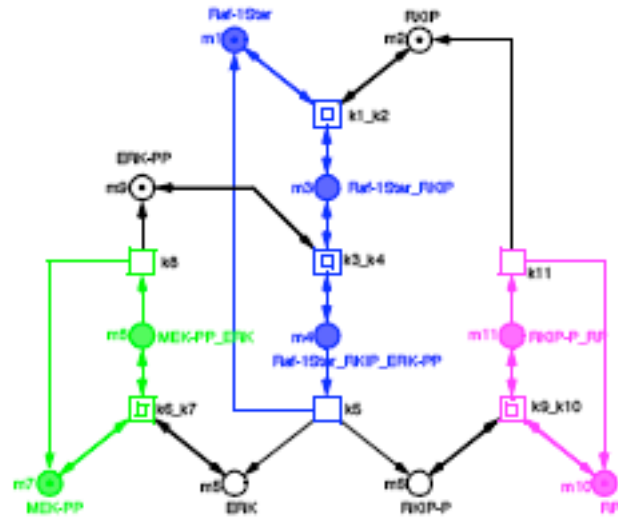
- Structural properties Tool supported
- General behavioural properties:
 - Boundedness
 - Liveness
 - Reversability
- Place & transition invariants:
 - CTI
 - CPI
- Model checking of temporal-logic properties:
 - special behavioural properties

Systematic construction of the minimal marking

- Each P-invariant gets at least one token
- In signal transduction:
 - exactly 1 token, corresponding to species conservation
 - token in least active state
- All (non-trivial) T-invariants become realizable
 - to make the net live
- Minimal marking
 - minimization of the state space

⇒ UNIQUE INITIAL MARKING ⇐

P-invariants



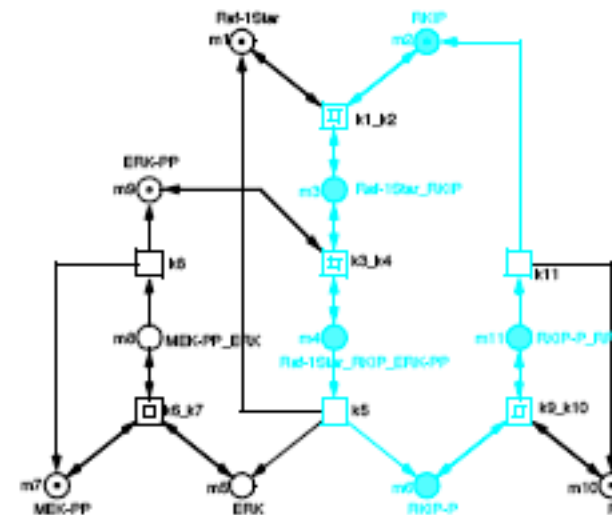
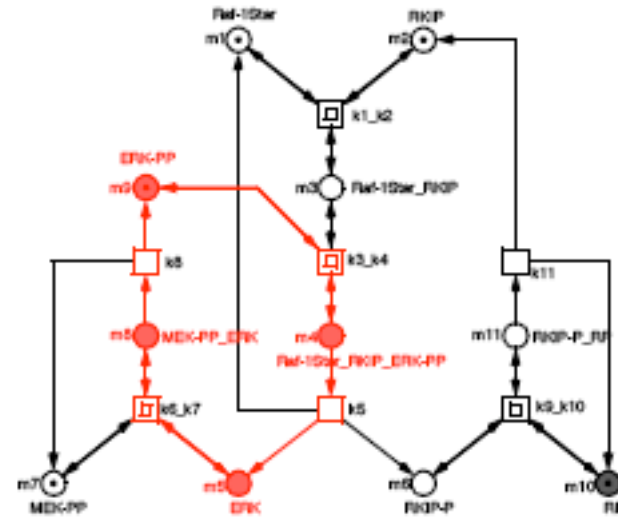
P-INV1: MEK

P-INV2: RAF-1STAR

P-INV3: RP

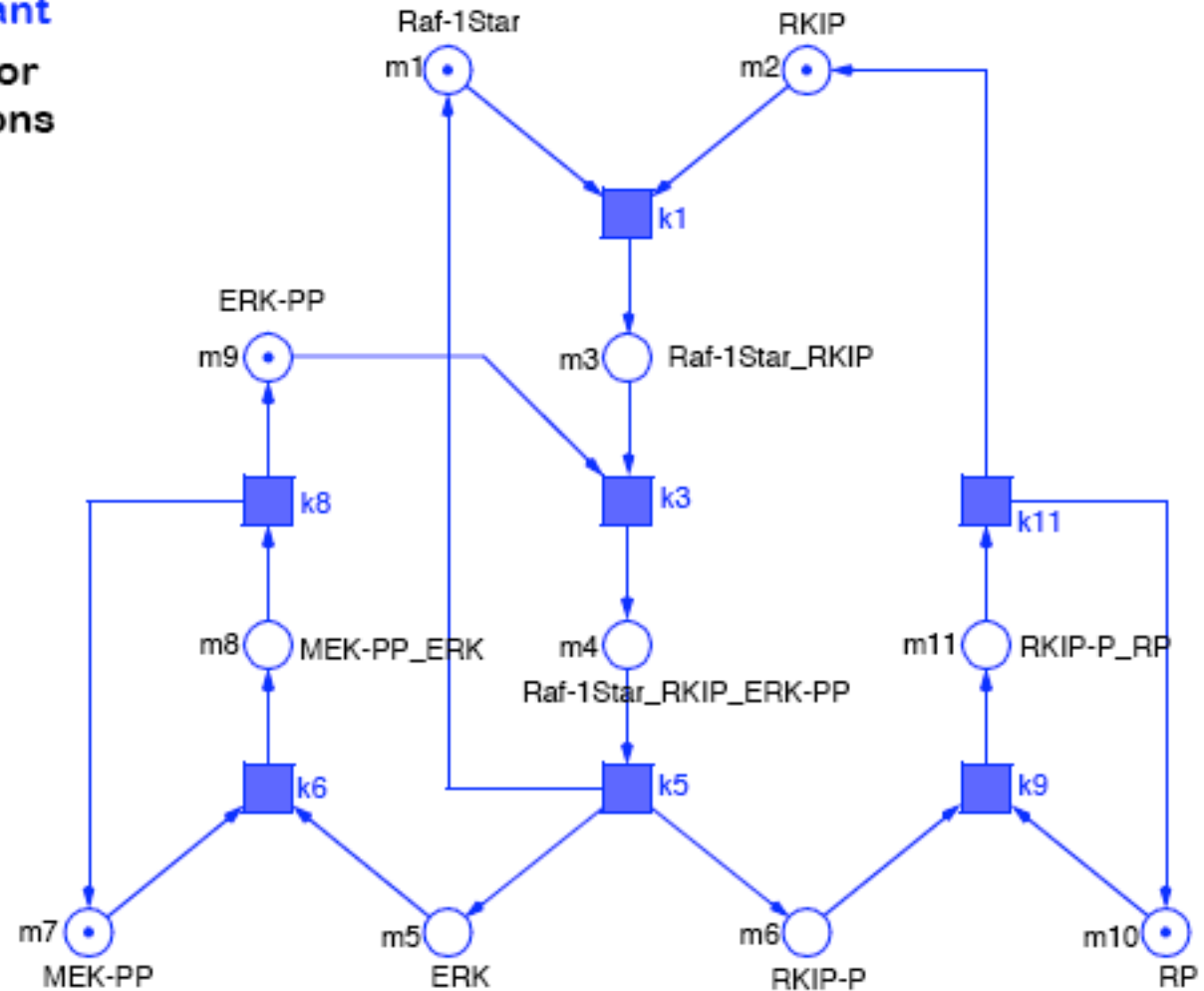
P-INV4: ERK

P-INV5: RKIP



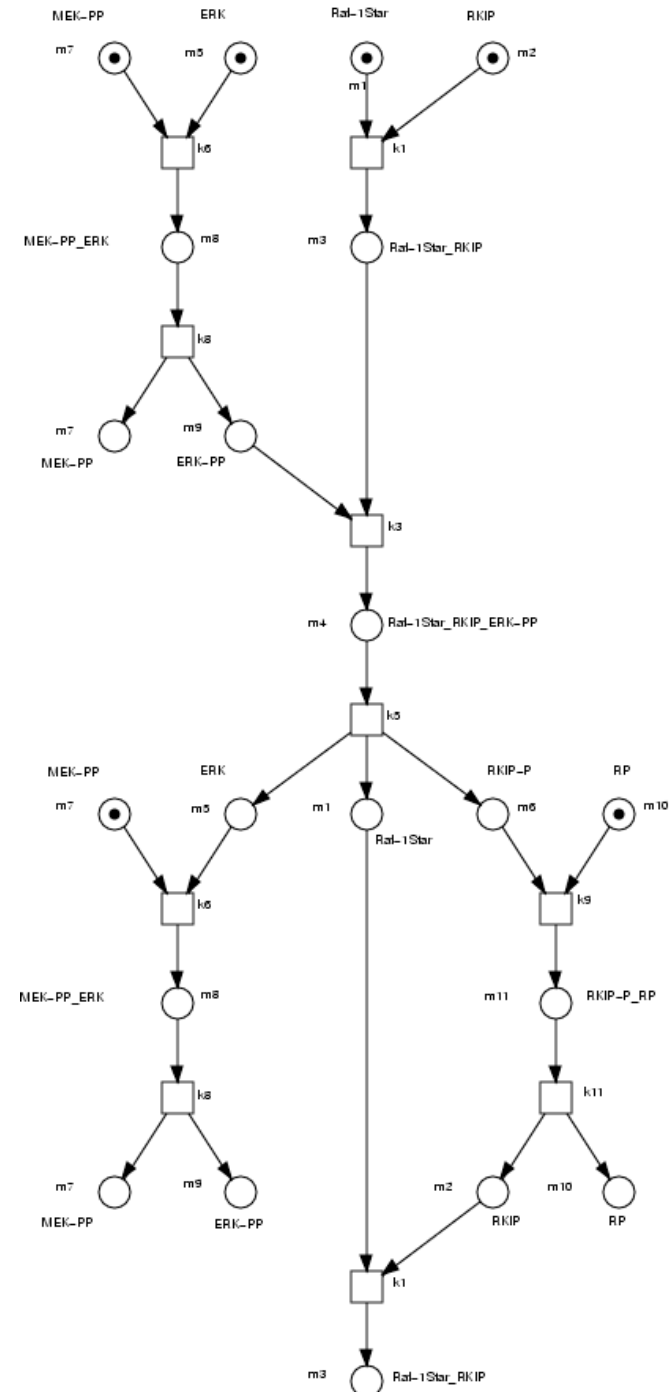
T-invariants

-> **non-trivial T-invariant**
+ **four trivial ones for reversible reactions**



Partial order run of the non-trivial T-invariant

- Partial order structure
- Illustrates *causal* & *concurrent* behaviour
- Labelled condition/event net
 - Events - transition occurrences
 - Conditions: input/output compounds



CTL queries

1. There are reachable states where ERK is phosphorylated and RKIP is not phosphorylated.

EF [(ERK-PP \vee RAF1*/RKIP/ERK-PP) \wedge RKIP]

2. The phosphorylation of ERK does not depend on the phosphorylation of RKIP.

EG [ERK \rightarrow E (\neg RKIP-P \vee RKIP-P/RP) U ERK-PP]

3. A cyclic behaviour w.r.t. RKIP is possible forever.

AG [(RKIP \rightarrow EF (\neg RKIP)) \wedge (\neg RKIP \rightarrow EF (RKIP))]

Qualitative analysis - summary

- Validation criterion 0
 - all expected structural properties hold
 - all expected general behavioural properties hold
- Validation criterion 1
 - CTI
 - no minimal T-invariant without biological interpretation
 - no known biological behaviour without corresponding T-invariant
- Validation criterion 2
 - CPI
 - no minimal P-invariant without biological interpretation
- Validation criterion 3
 - all expected special behavioural properties expressed by temporal logic formulae hold

Step 2: quantitative analysis

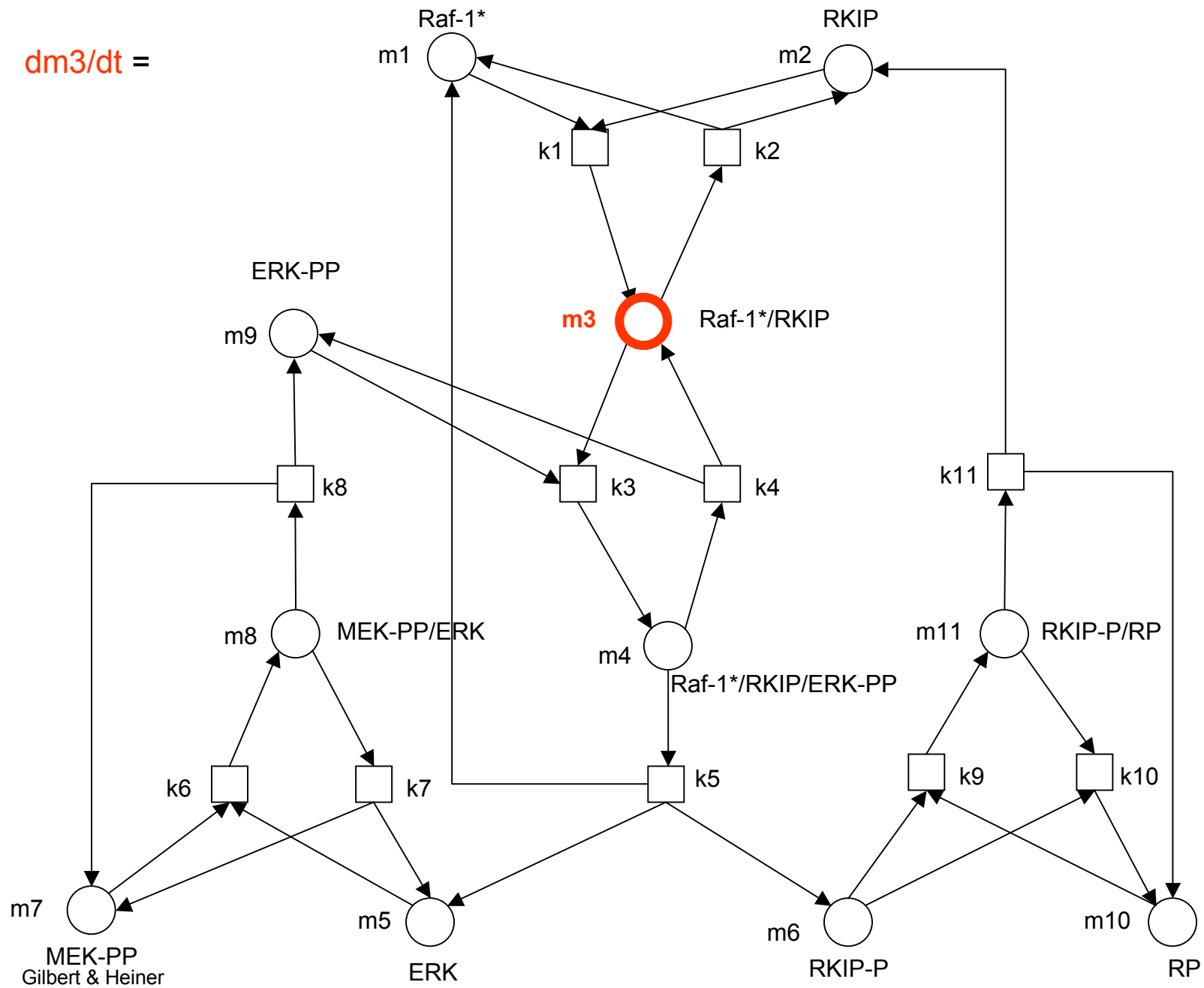
- Derive differential equations using graph structure & rate constants
- Add (13 alternative) initial conditions derived via qualitative analysis
- Compute steady-states to show “equivalence” of alternatives

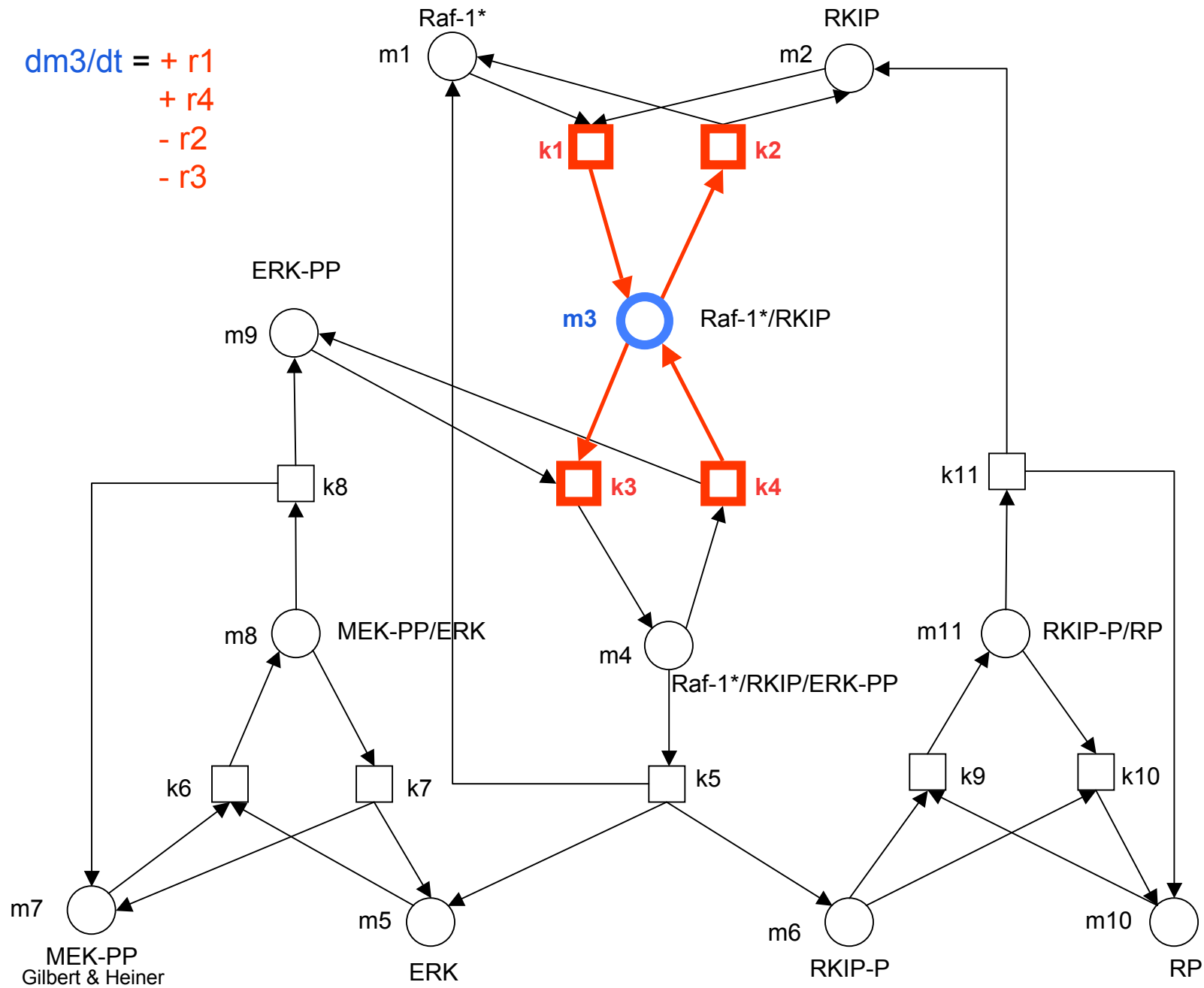
From (time-less) discrete to (timed) continuous Petri nets

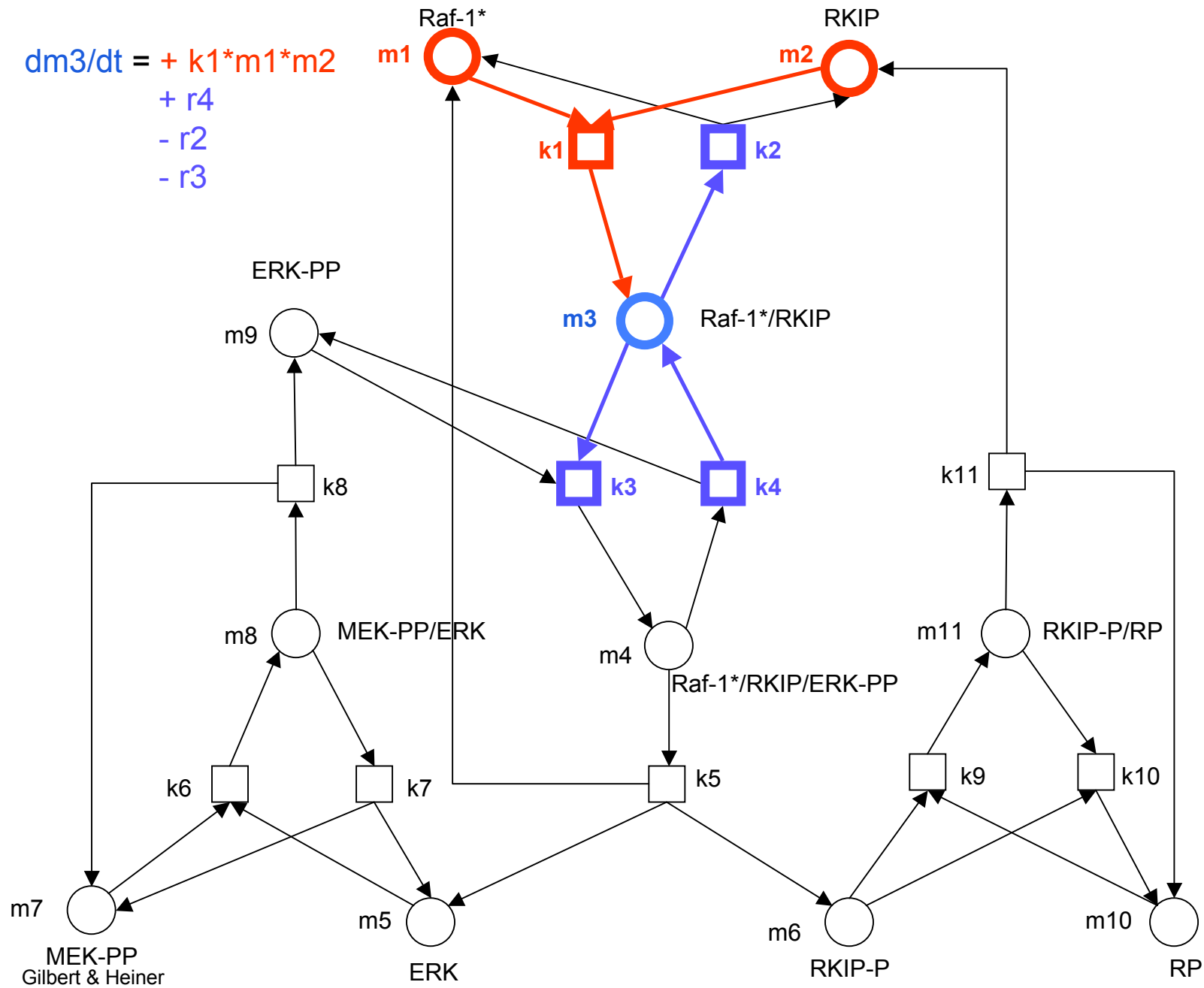
- Keep the structure
- Add rate function to each transition
 - Pre-places appear as parameters \rightarrow state dependent
 - Function defines the *continuous* flow
- Each place gets 1 token (real number)
 - Represents current *continuous* concentration
- Continuous state space

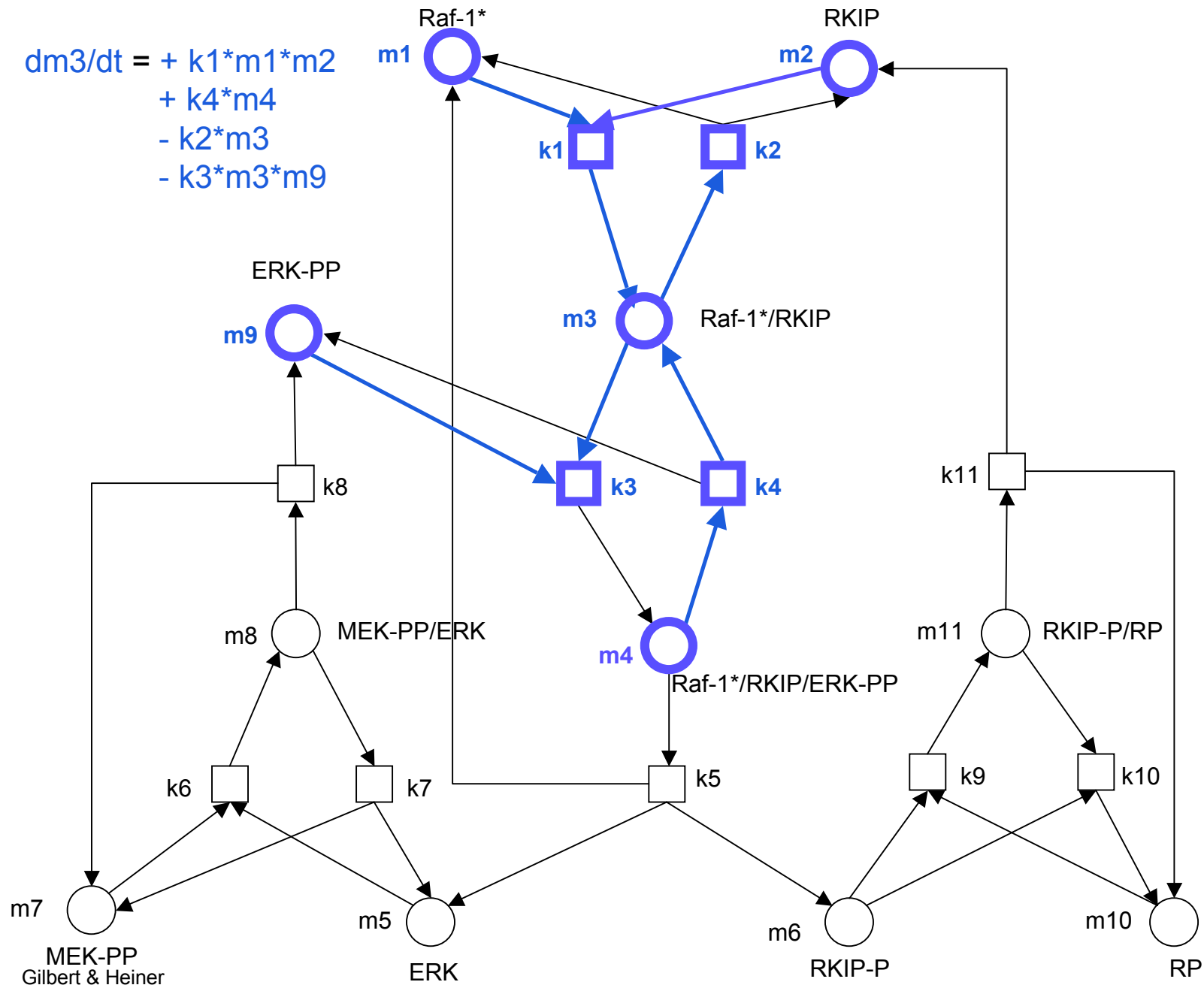
\Rightarrow Continuous Pn defines a system of ODEs \Leftarrow

$dm3/dt =$









Validation via ODE simulation

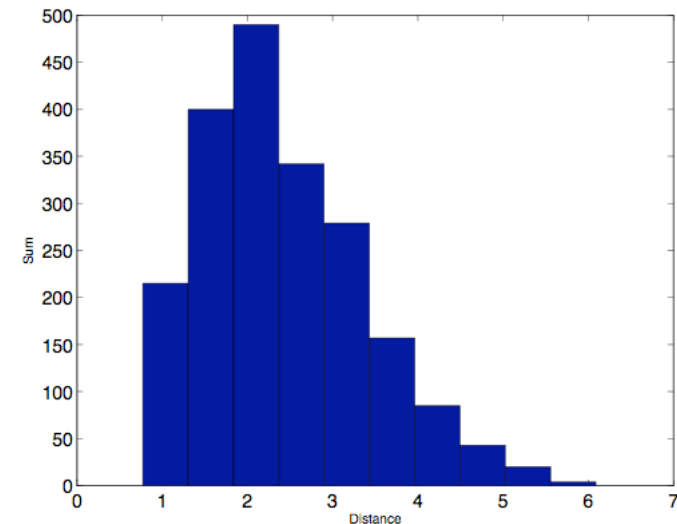
Species	S1	S2	S3	S4	S5	S6	S7	S8	S9	S10	S11	S12	S13
Raf-1*	1	0	0	1	1	1	1	1	0	0	1	1	1
RKIP	1	0	0	0	0	0	0	1	0	0	1	0	0
Raf-1*_RKIP	0	1	0	0	0	0	0	0	1	1	0	0	0
Raf-1*_RKIP_ERK-PP	0	0	1	0	0	0	0	0	0	0	0	0	0
ERK	0	0	0	1	0	0	1	1	1	0	0	0	0
RKIP-P	0	0	0	1	1	0	0	0	0	0	0	0	1
MEK-PP	1	1	1	1	0	0	1	1	1	0	0	1	1
MEK-PP_ERK	0	0	0	0	1	1	0	0	0	1	1	0	0
ERK-PP	1	1	0	0	0	0	0	0	0	0	0	1	1
RP	1	1	1	1	1	0	0	1	1	1	1	0	1
RKIP-P_RP	0	0	0	0	0	1	1	0	0	0	0	1	0

Cho et al

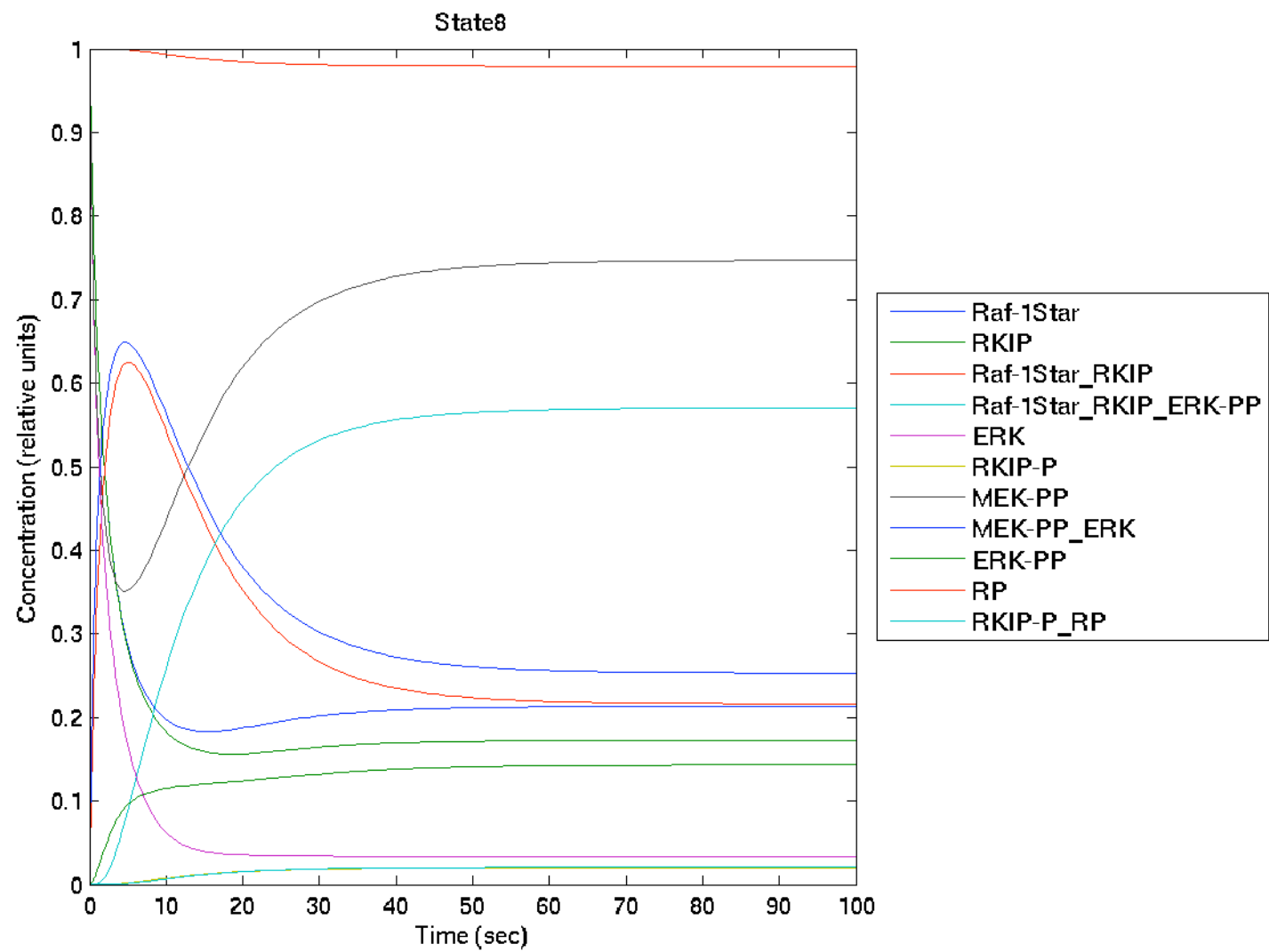
Biochemist

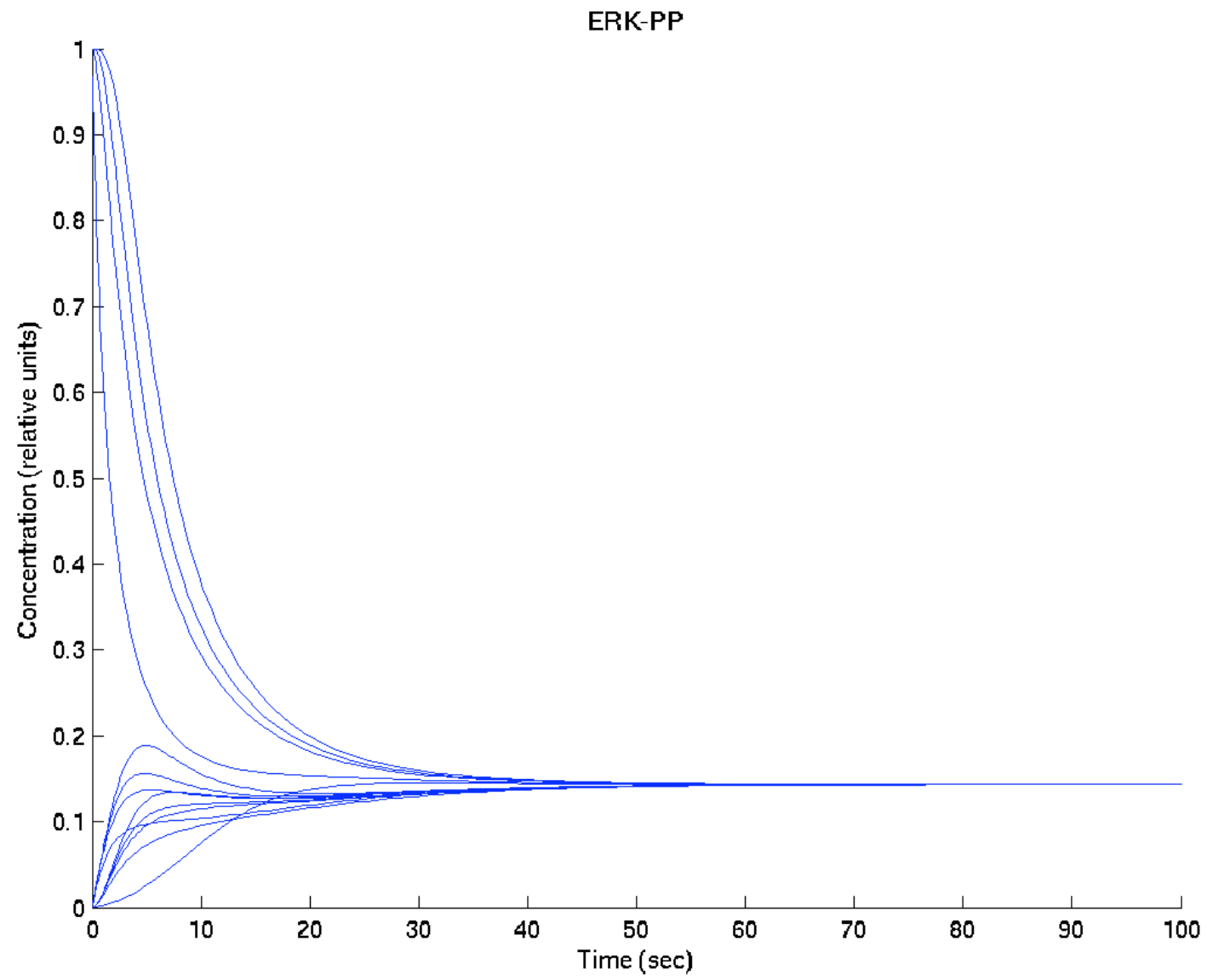
13 'Good' state configurations

Species	Mean steady state concentration	Standard Deviation
Raf-1*	0.2133	0.1225 * 1.0e-04
RKIP	0.1727	0.0854 * 1.0e-04
Raf-1*_RKIP	0.2163	0.5546 * 1.0e-04
Raf-1*_RKIP_ERK-PP	0.5704	0.4346 * 1.0e-04
ERK	0.0332	0.0135 * 1.0e-04
RKIP-P	0.0200	0.0169 * 1.0e-04
MEK-PP	0.7469	0.6020 * 1.0e-04
MEK-PP_ERK	0.2531	0.6020 * 1.0e-04
ERK-PP	0.1433	0.1846 * 1.0e-04
RP	0.9793	0.0471 * 1.0e-04
RKIP-P_RP	0.0207	0.0473 * 1.0e-04



Distribution of 'bad' steady states as euclidean distances from the 'good' final steady state





Quantitative analysis - summary

- Reasonable initial states of the quantitative model can be constructed by help of the qualitative model
- All “live” discrete states result in the same steady state
- No other (theoretically possible) initial states are “close” to this steady state
- Hold for this self-contained case study
 - Other case studies confirm these findings

Overall summary

- We use a 2-step approach;
 - qualitative model to validate the structure;
 - qualitative & quantitative model share the same structure
- Continuous pn are structured notations for ODEs
- Many open questions...

...Outlook

- Deeper insights into the relation between the discrete and the continuous world
- The whole truth about biomolecular systems
⇒ inherently stochastic behaviour!!!
- Continuous models are just an approximation of reality

Acknowledgements

- Rainer Breitling (Groningen)
- Alex Tovchigrechko (Cottbus)
- Simon Rogers (Glasgow)

Support from the UK DTI Bioscience
Beacons programme

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