

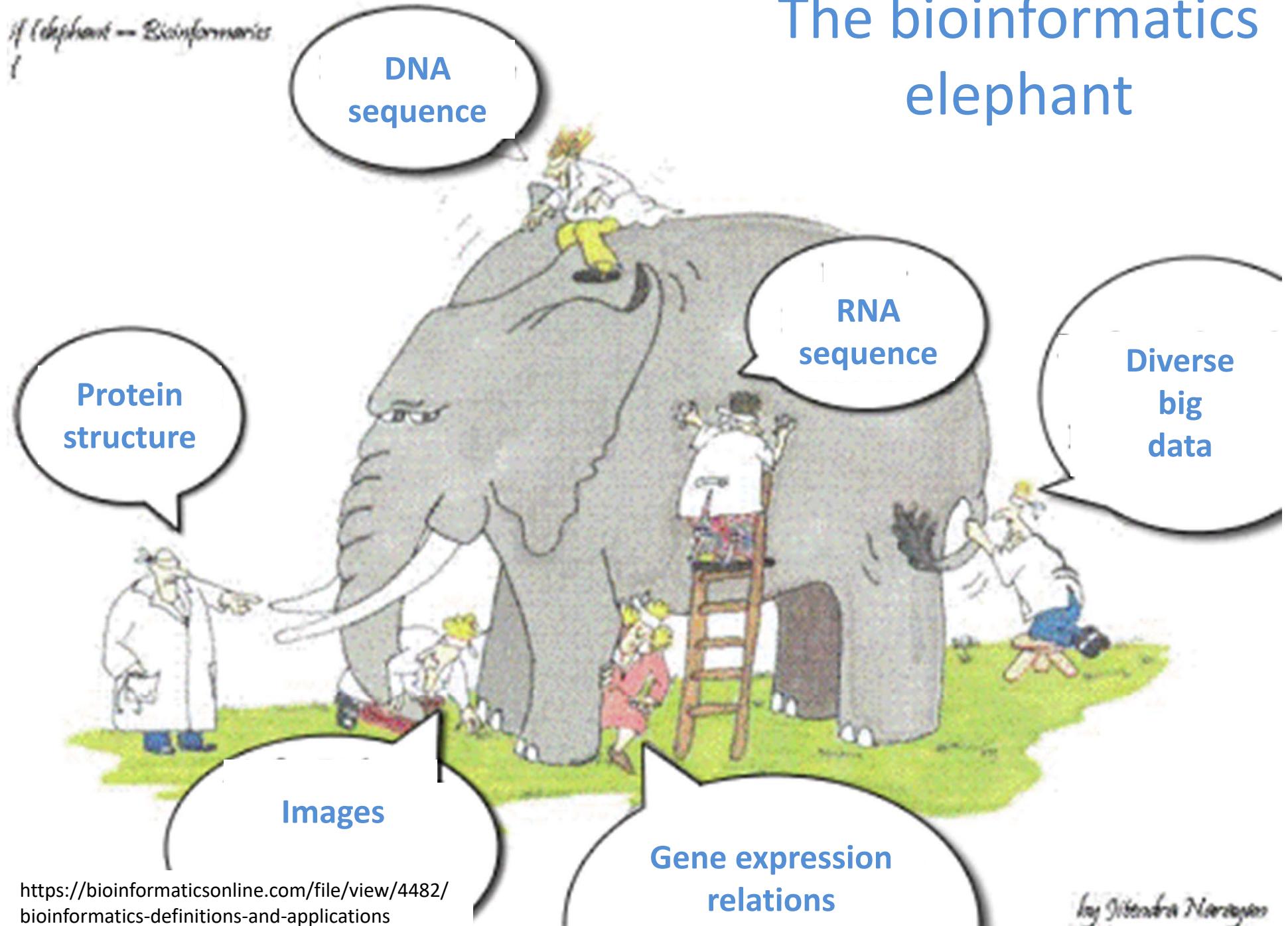
Bipartite graphs for computational modeling in systems biology: from KEGG to Petri nets

Ina Koch
Molecular Bioinformatics
Institute of Computer Science
Goethe University Frankfurt am Main

ina.koch@bioinformatik.uni-frankfurt.de
www.bioinformatik.uni-frankfurt.de

November, 14th 2023
3rd Edition of the Workshop "Metabolism and mathematical models: Two for a tango"

The bioinformatics elephant

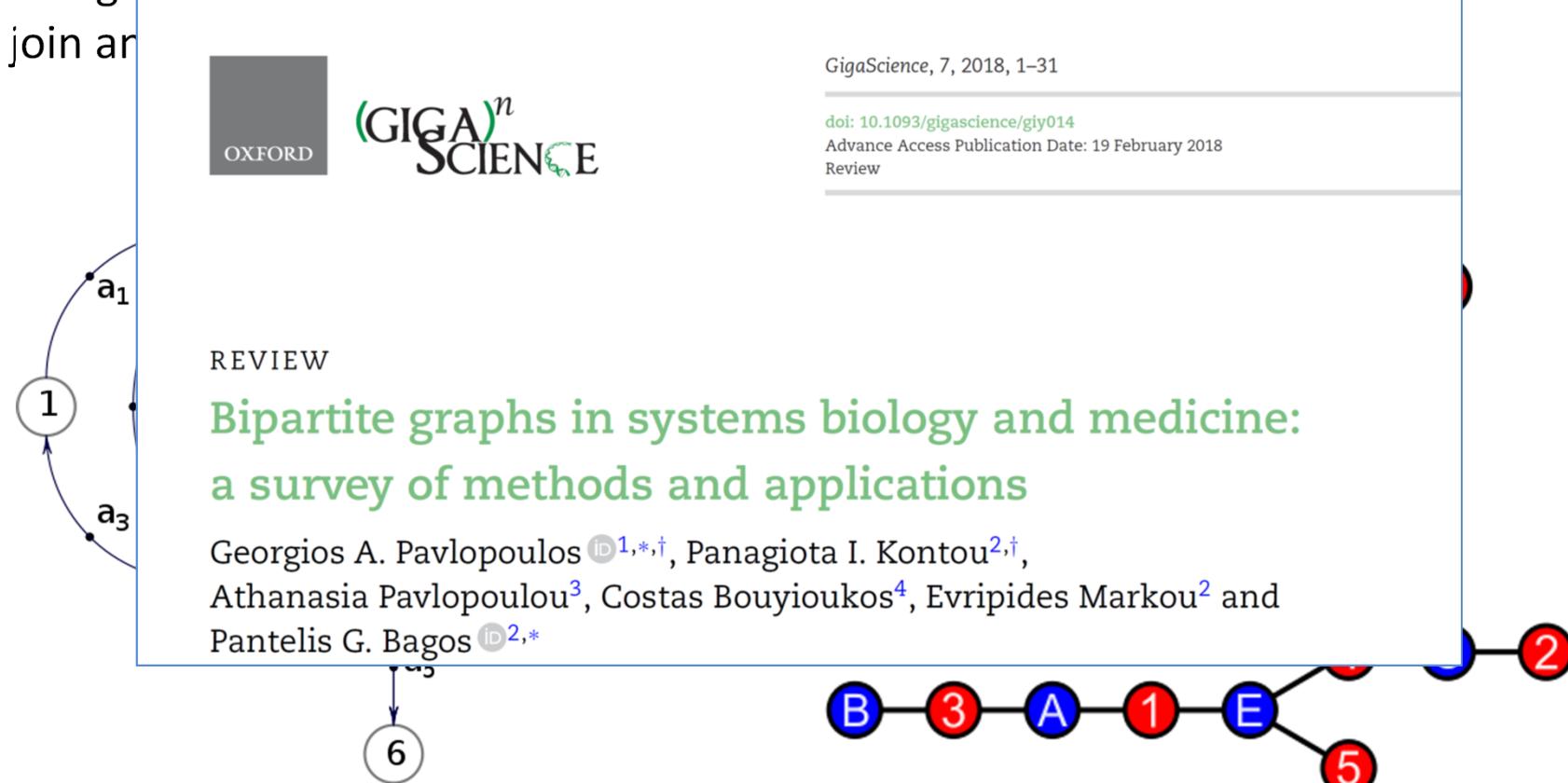


Challenges of data integration

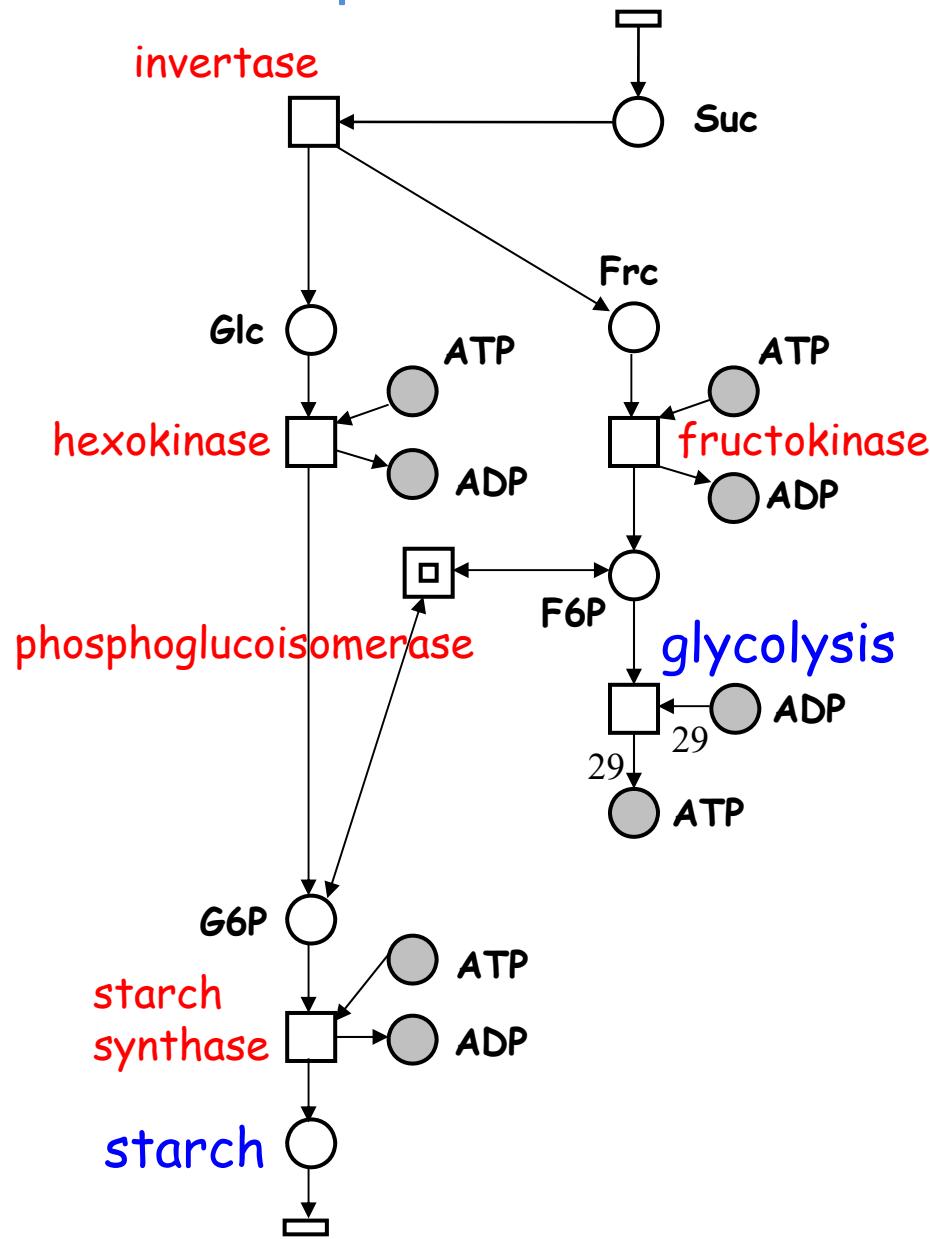
- ❖ Incomplete data
- ❖ Different time points and different locations in the cell
- ❖ Incomplete data
The quality and quantity of the data determine the modeling approach
- ❖ Different experiments under varying experimental conditions
- ❖ Different scales: genomics, transcriptomics, proteomics, metabolomics, interactomics, imaging, ...

Hypergraphs and bipartite graphs

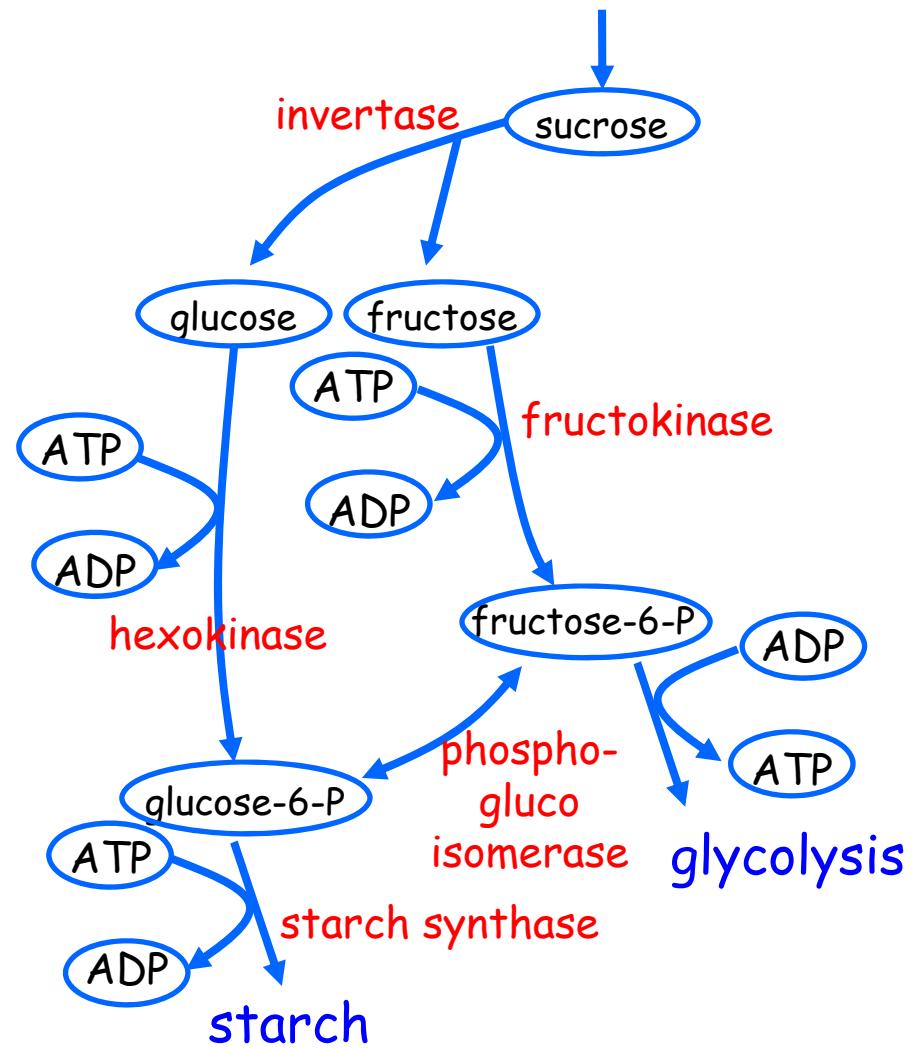
A **hypergraph** is a generalization
of a graph in which edges can join an
arbitrary number of vertices.



Bipartite graph (Petri net) representation

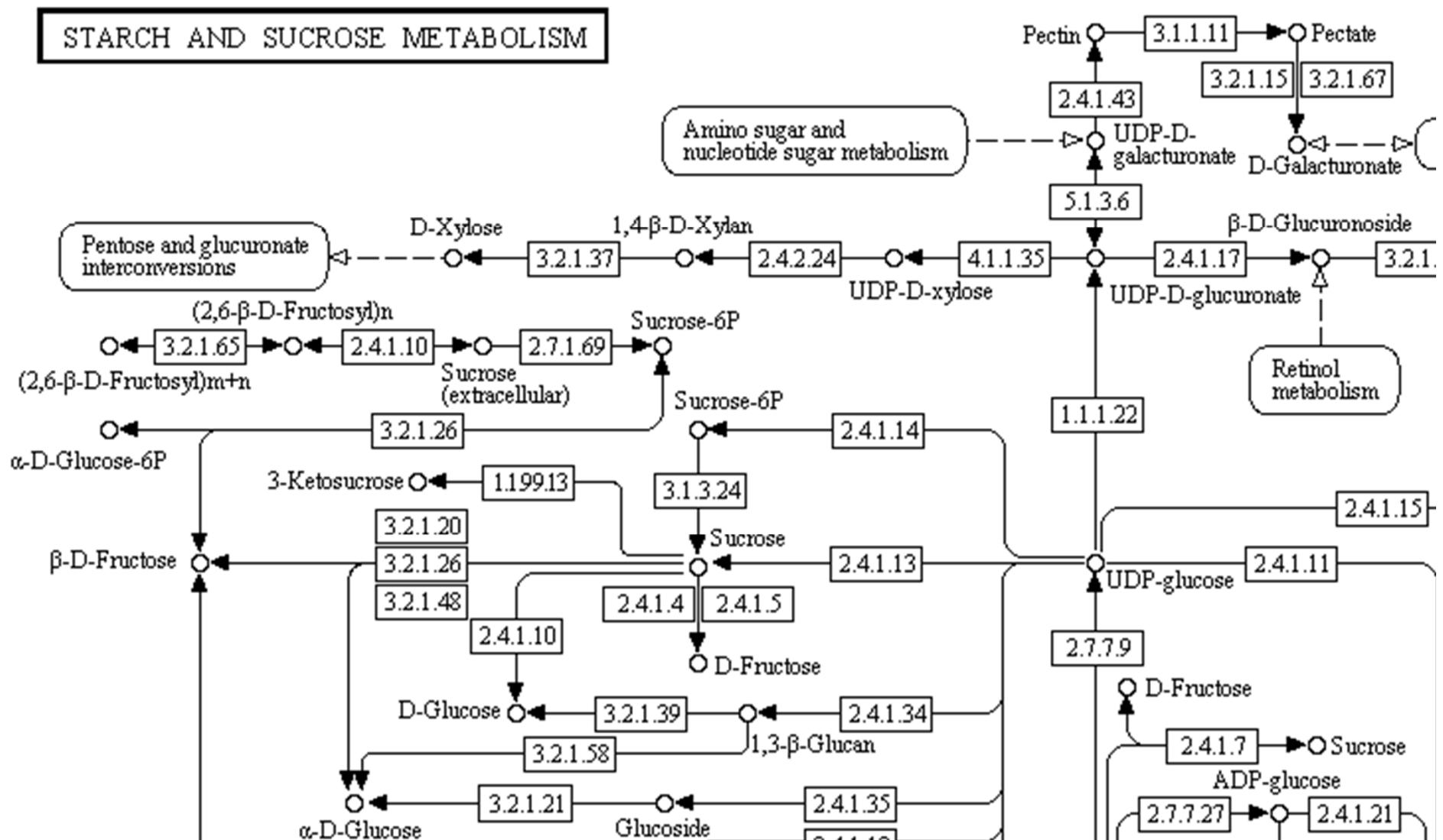


Hypergraph representation



Bipartite graph (KEGG) representation

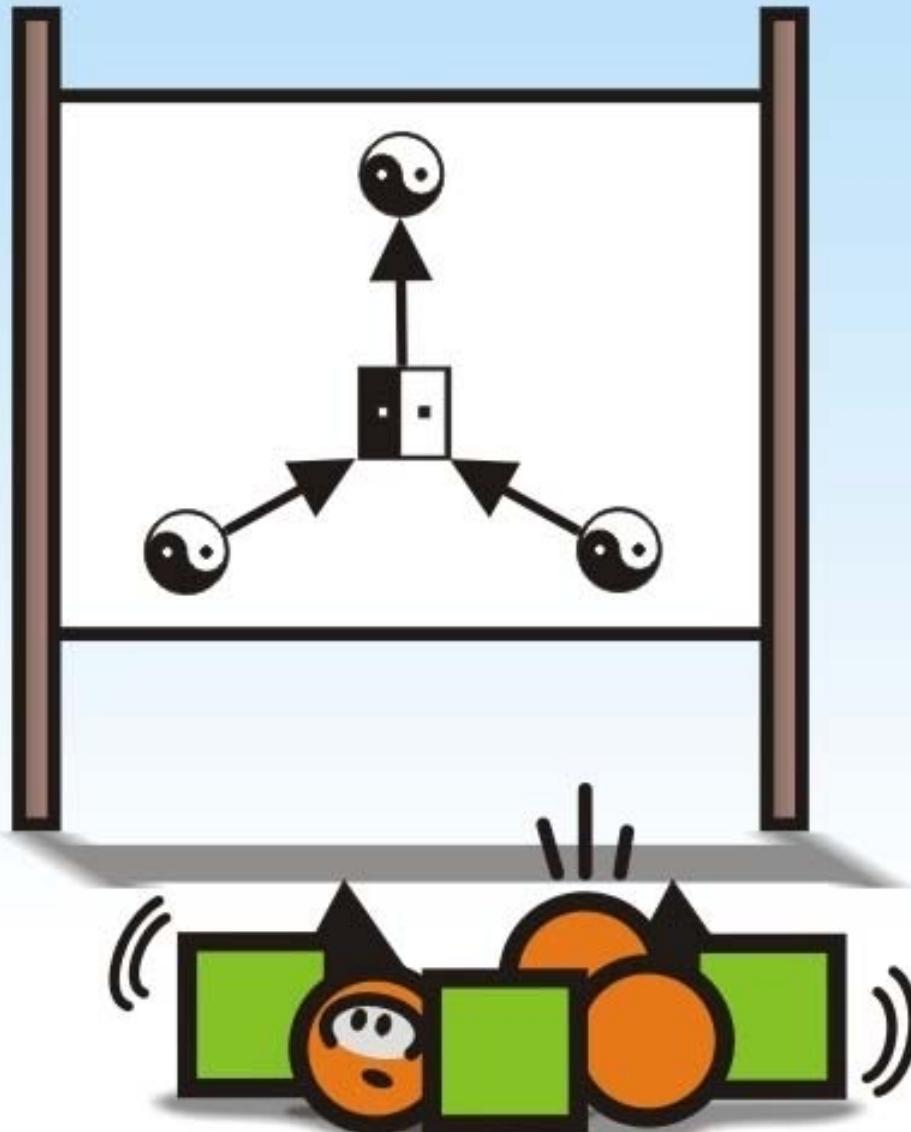
<http://www.genome.jp/kegg/pathway/map/map00500.html>



How and to what extent
can we analyze and predict
the system's behavior
without
knowing kinetic parameters?

Petri nets

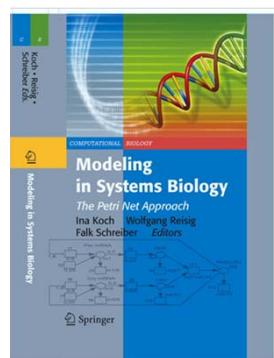
Petri nets have nothing to do with Petri dishes



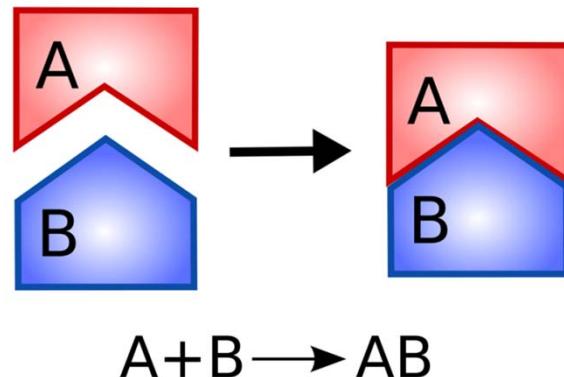
Carl Adam Petri
(1926 – 2010)

Petri net definitions

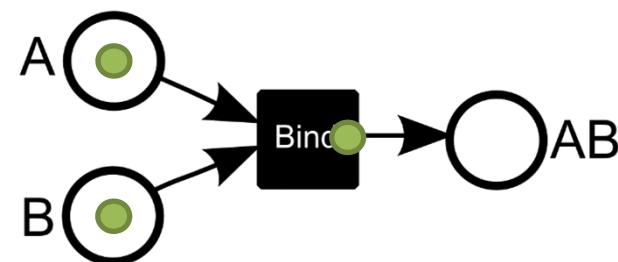
Elements	Symbols	Interpretation
Transitions	□	Active elements: processes, e.g., reactions
Places	○	Passive elements: biological species, e.g., compounds, define pre-/post-conditions
Tokens	● 1	Movable objects: molecules, proteins, enzymes, cofactors, bacteria, or cells, marking as system state
Edges or arcs	→	Relations, weighted to quantify pre-/post-conditions



Binding process



Petri net



Firing rule: untimed (P/T net),
timed-discrete, stochastic, continuous

Koch, Reisig, Schreiber (2011) Modeling in Systems Biology – The Petri Net Approach, Springer

Outline

Part I - Metabolism

- ❖ Modeling of the central carbon metabolism in potato tubers



Stéphanie Boue

Part II – Signal transduction

- ❖ Modeling of TNFR1-induced signaling pathway as Petri net, including the NF-κB pathway



Leonie Amstein

Outline

Part I - Metabolism

- ❖ Modeling of the central carbon metabolism in potato tubers



Stéphanie Boue

Part II – Signal transduction

- ❖ Modeling of TNFR1-induced signaling pathway as Petri net, including the NF-κB pathway



Leonie Amstein

Sucrose-to-starch-pathway in potato tuber

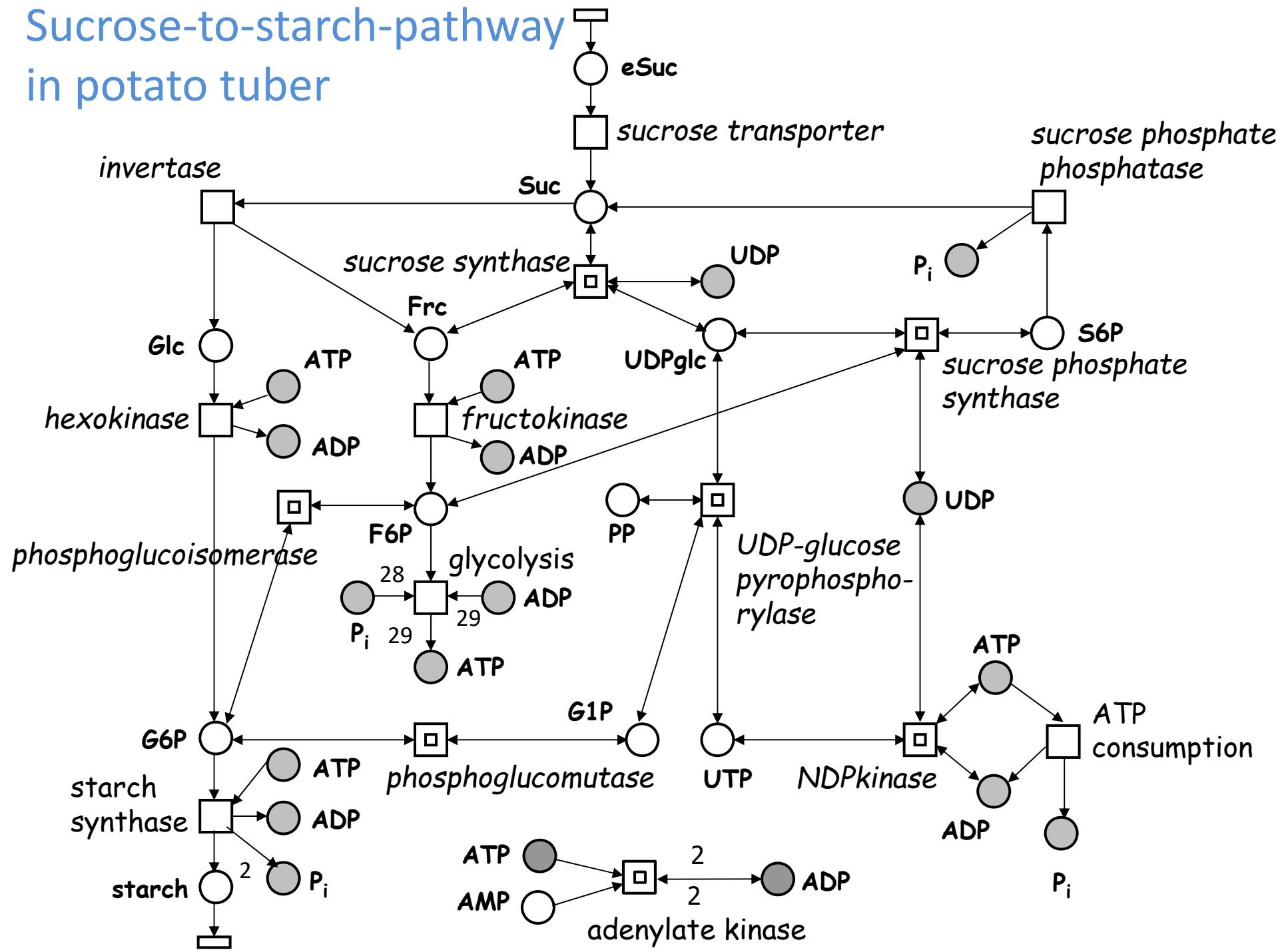


Björn Junker

sucrose synthase:	$\text{Suc} + \text{UDP} \leftrightarrow \text{UDPGlc} + \text{Frc}$
UDP-glucose pyrophosphorylase:	$\text{UDPGlc} + \text{PP} \leftrightarrow \text{G1P} + \text{UTP}$
phosphoglucomutase:	$\text{G6P} \leftrightarrow \text{G1P}$
fructokinase:	$\text{Frc} + \text{ATP} \rightarrow \text{F6P} + \text{ADP}$
phosphoglucoisomerase:	$\text{G6P} \leftrightarrow \text{F6P}$
hexokinase:	$\text{Glc} + \text{ATP} \rightarrow \text{G6P} + \text{ADP}$
invertase:	$\text{Suc} \rightarrow \text{Glc} + \text{Frc}$
sucrose phosphate synthase:	$\text{F6P} + \text{UDPGlc} \leftrightarrow \text{S6P} + \text{UDP}$
sucrose phosphate phosphatase:	$\text{S6P} \rightarrow \text{Suc} + \text{P}_i$
glycolysis (b):	$\text{F6P} + 29 \text{ ADP} + 28 \text{ P}_i \rightarrow 29 \text{ ATP}$
NDPkinase:	$\text{UDP} + \text{ATP} \leftrightarrow \text{UTP} + \text{ADP}$
sucrose transporter:	$\text{eSuc} \rightarrow \text{Suc}$
ATP consumption (b):	$\text{ATP} \rightarrow \text{ADP} + \text{P}_i$
starch synthesis:	$\text{G6P} + \text{ATP} \rightarrow 2\text{P}_i + \text{ADP} + \text{starch}$
adenylate kinase:	$\text{ATP} + \text{AMP} \leftrightarrow 2\text{ADP}$
pyrophosphatase:	$\text{PP} \rightarrow 2\text{P}_i$
adenylate kinase:	$\text{AdK} \leftrightarrow \text{AMP} + \text{ADP}$

Sucrose-to-starch-pathway

in potato tuber

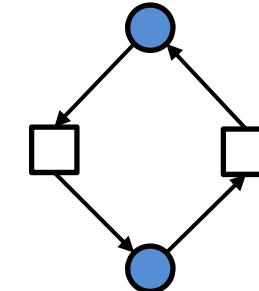


Invariant analysis at steady state

Substance conservations

Place invariants

- ❖ Sets of places whose weighted sum of tokens remains constant
- ❖ Substance conservations



Place invariant: $C^T x = 0$

Transition invariant: $C y = 0$

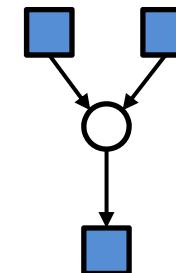
Solutions: **minimal, nonnegative, nontrivial, integer**

Minimal: $\exists z: \text{supp}(z) \subseteq \text{supp}(u)$ and the largest common divisor of all non-zero entries of u is 1

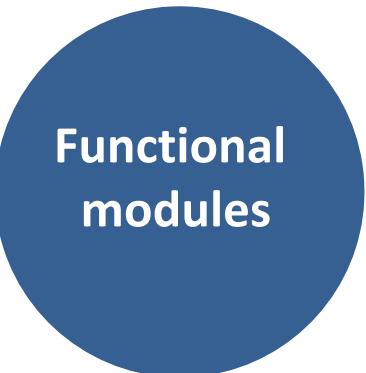
Functional modules

Transition invariants

- ❖ Multi-sets of transitions whose firing reproduces an arbitrary initial marking
Lautenbach (1973) *GMD Report No. 82*
- ❖ Basic functional modules
(elementary modes)
Schuster *et al.* (1993) *Second Gauss Symposium*



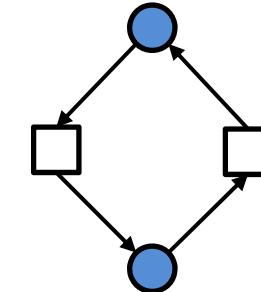
Invariant analysis for the sucrose-to-starch-pathway



Place invariants

The net is not **covered by P-invariants**.

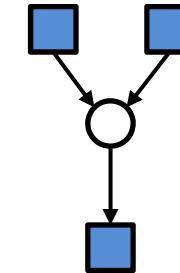
1. UDPglc, UTP, UDP
2. ATP, AMP, ADP
3. G6P, F6P, G1P, UTP, ATP(2), ADP, S6P, P_i, PP(2)



Transition invariants

The net is not **covered by transition invariants!**

- Trivial:
1. SPS, [SPS_rev](#)
 2. UGPase, UGPASE_rev
 3. SuSy_SuSy_rev
 4. PGM, PGM_rev
 5. NDPkin, NDPkin_rev
 6. [AdK](#), [AdK_rev](#)
 7. PGI, PGI_rev



Removing [SPS_rev](#), [AdK](#), and [AdK_rev](#), the Petri net becomes **covered by transition invariants**

Transition invariant analysis

Invariant number	ATP used for cycling							
	sucrose cleavage		hexoses go into		ATP cons	Inv	Inv	SuSy
	SuSy Inv	Glyc	StaSy	SuSy_rev		SPS, SPP	SPS, SPP	
8	x		x	x				x
9	x		x	x	x			
10	x		x	x				
11		x	x	x		x		
12		x	x	x			x	
13		x	x	x				x
14	x	x	x	x				
15	x	x	x					
16	x	x			x			
17	x	x				x		
18	x	x					x	
19	x	x		x				

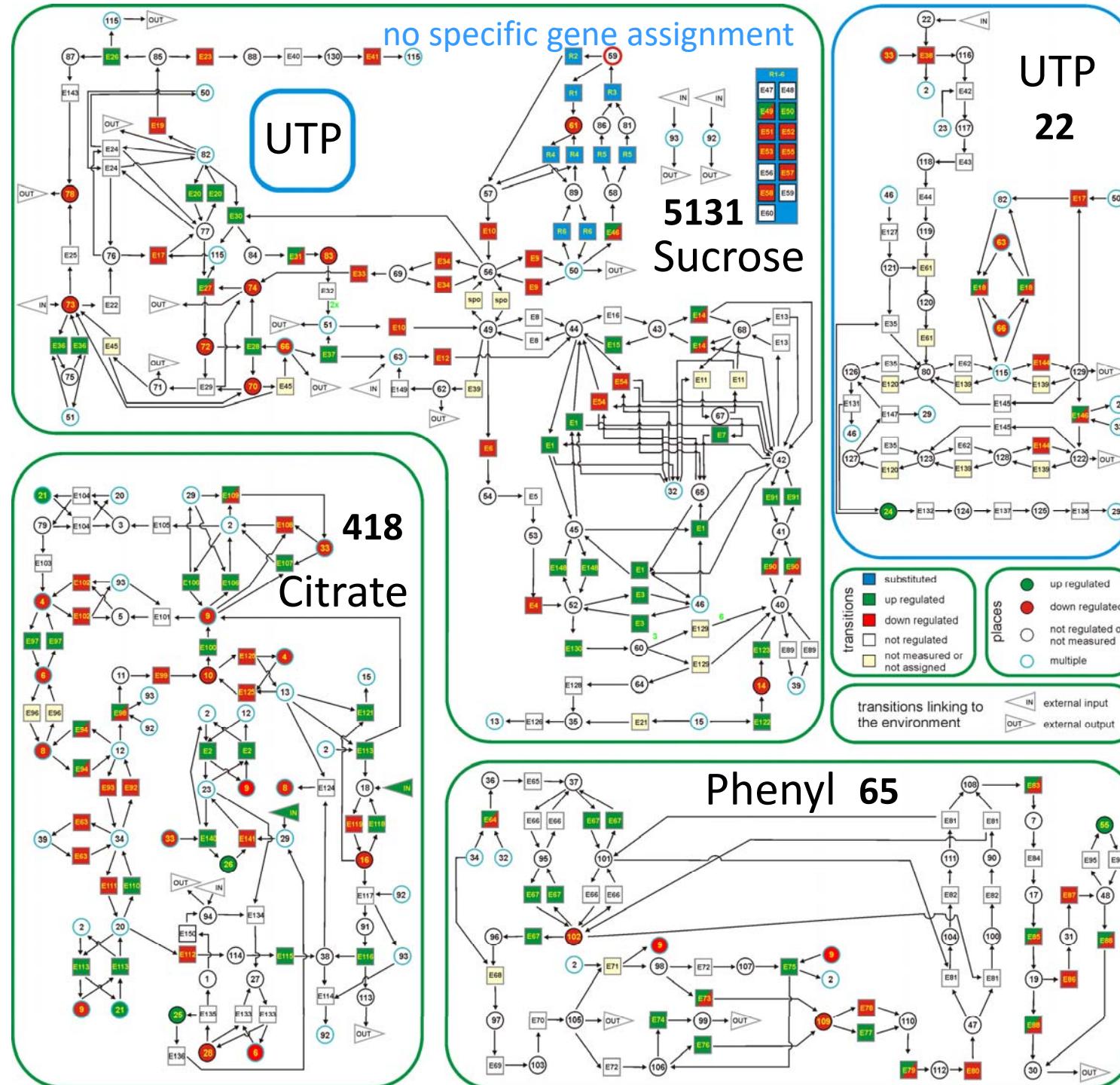
Arabidopsis thaliana

Complete network:
Sucrose-,
UTP-,
Citrate-,
and
Phenyl-
Pathway

transition
invariants of
each subnet

Koch et al. (2017)
Frontiers in genetics

MODEL1801090001



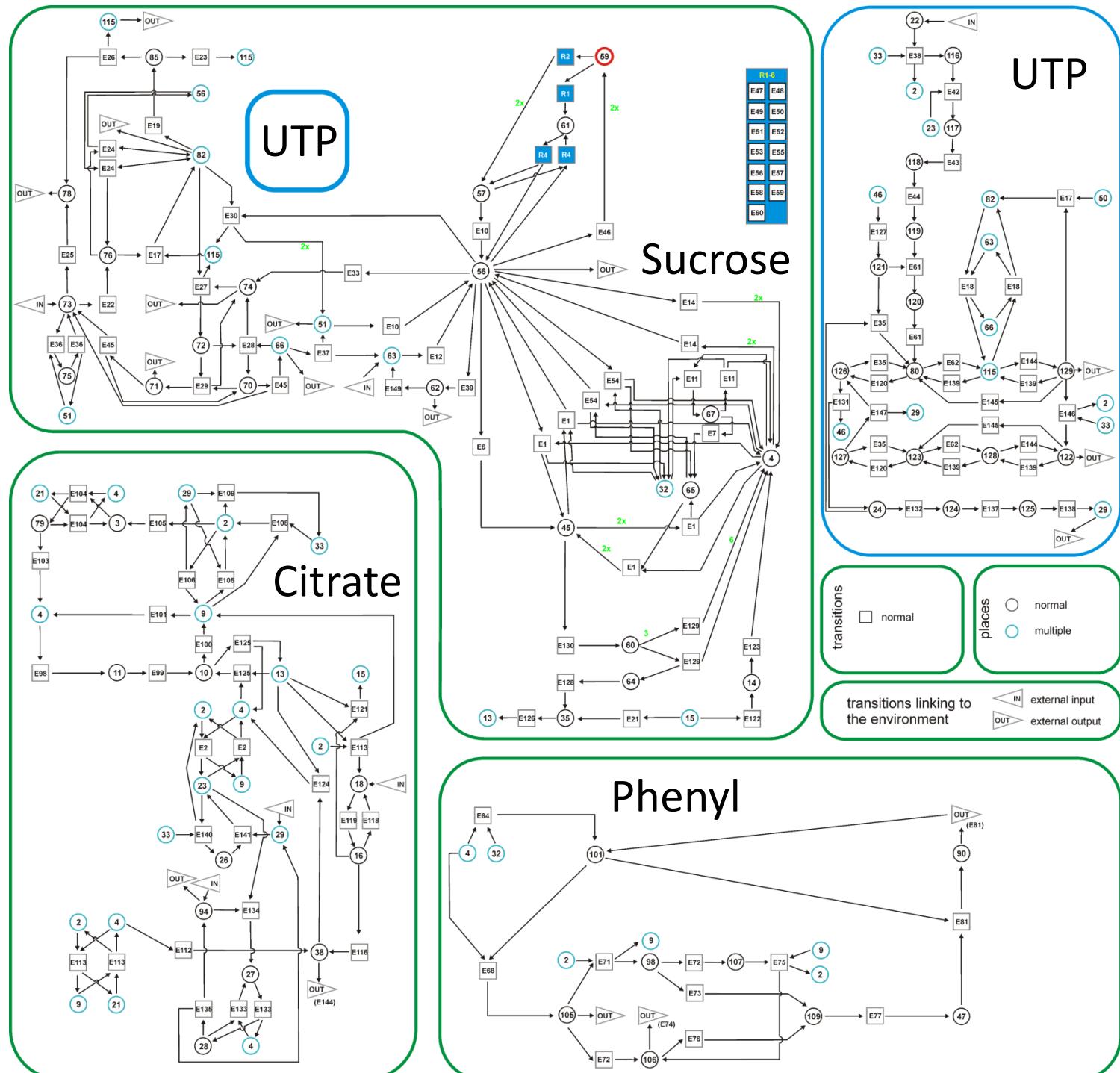
Reduced network

	Net	Reduced
#pl.	130	61
#tr.	232	123
#ed.	539	309

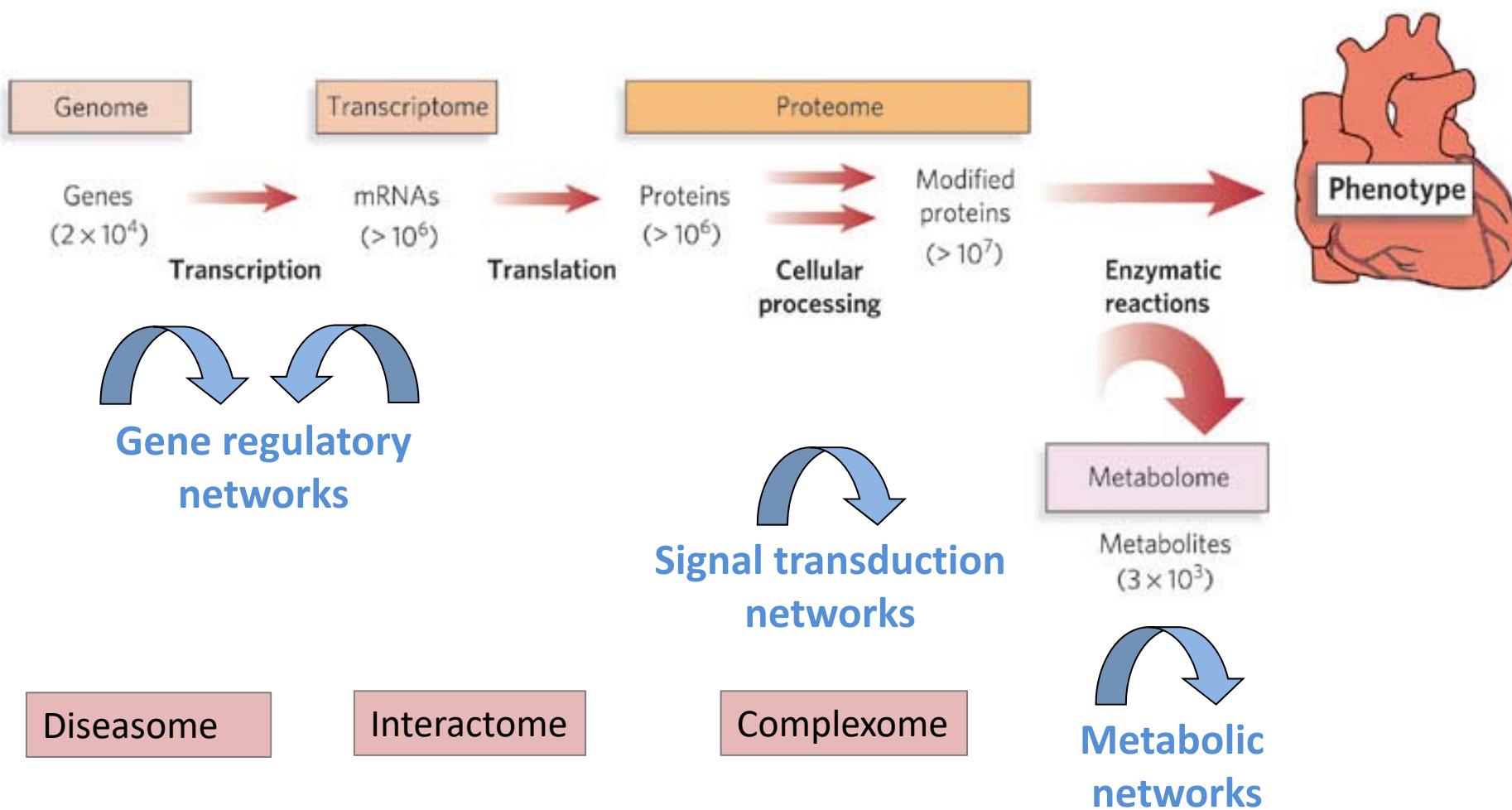
#transition
invariants
9775

Koch et al. (2017)
Frontiers in genetics

MODEL1801090001



Data of different scales require different modeling methods



Outline

Part I - Metabolism

- ❖ Modeling of the central carbon metabolism in potato tubers



Stéphanie Boue

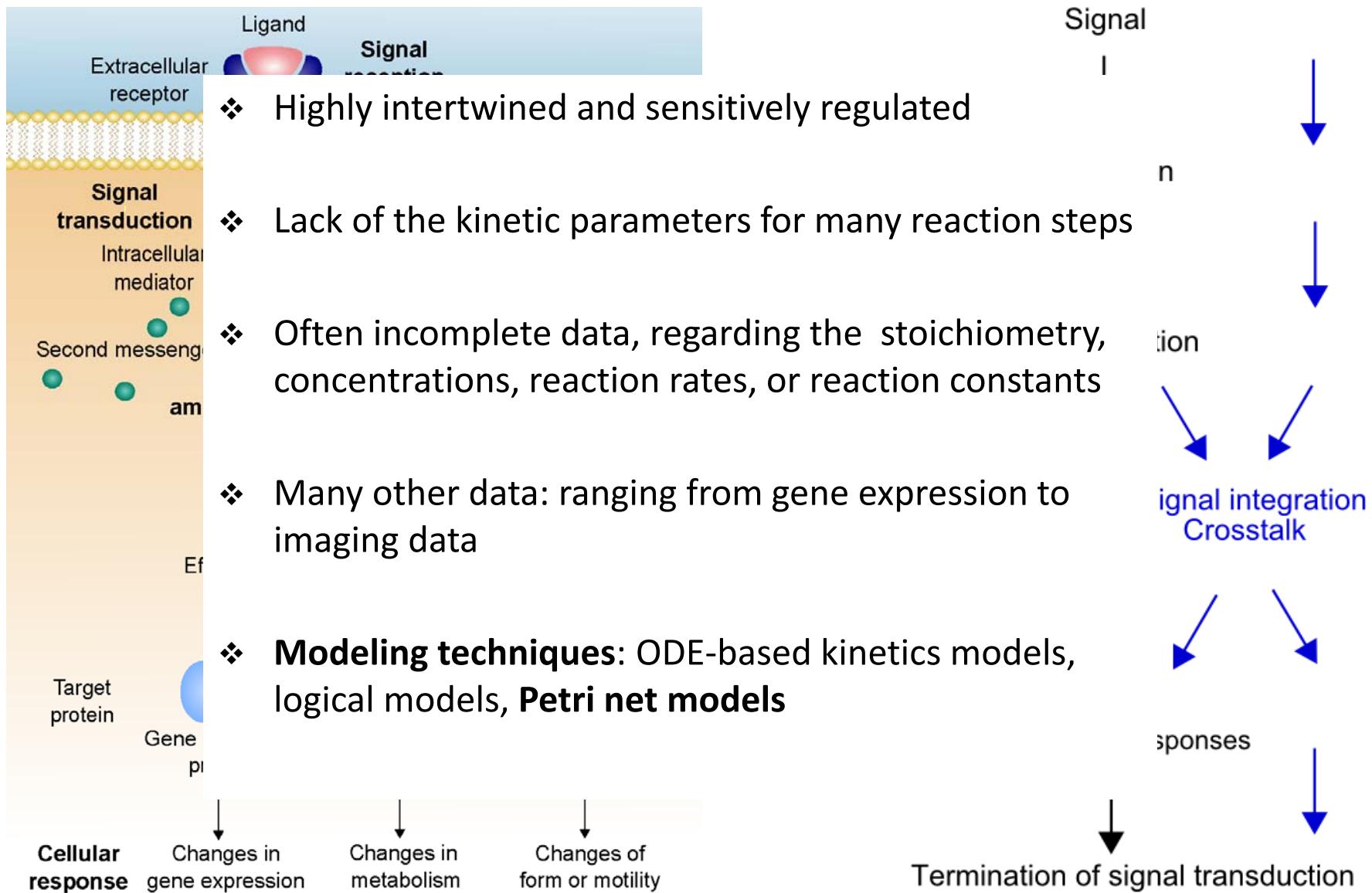
Part II – Signal transduction

- ❖ Modeling of TNFR1-induced signaling pathway as Petri net, including the NF-κB pathway

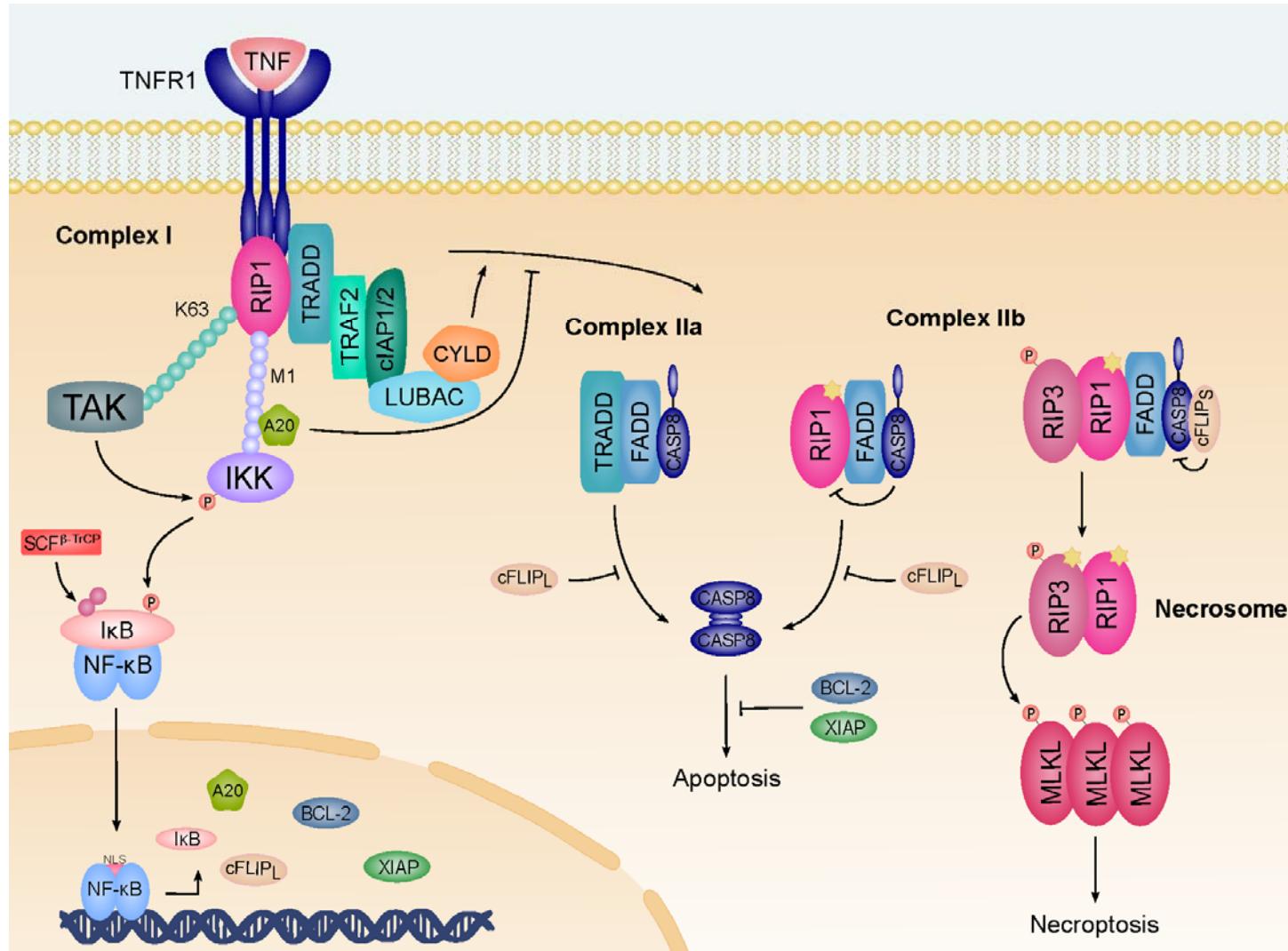


Leonie Amstein

Functionality in signaling pathways

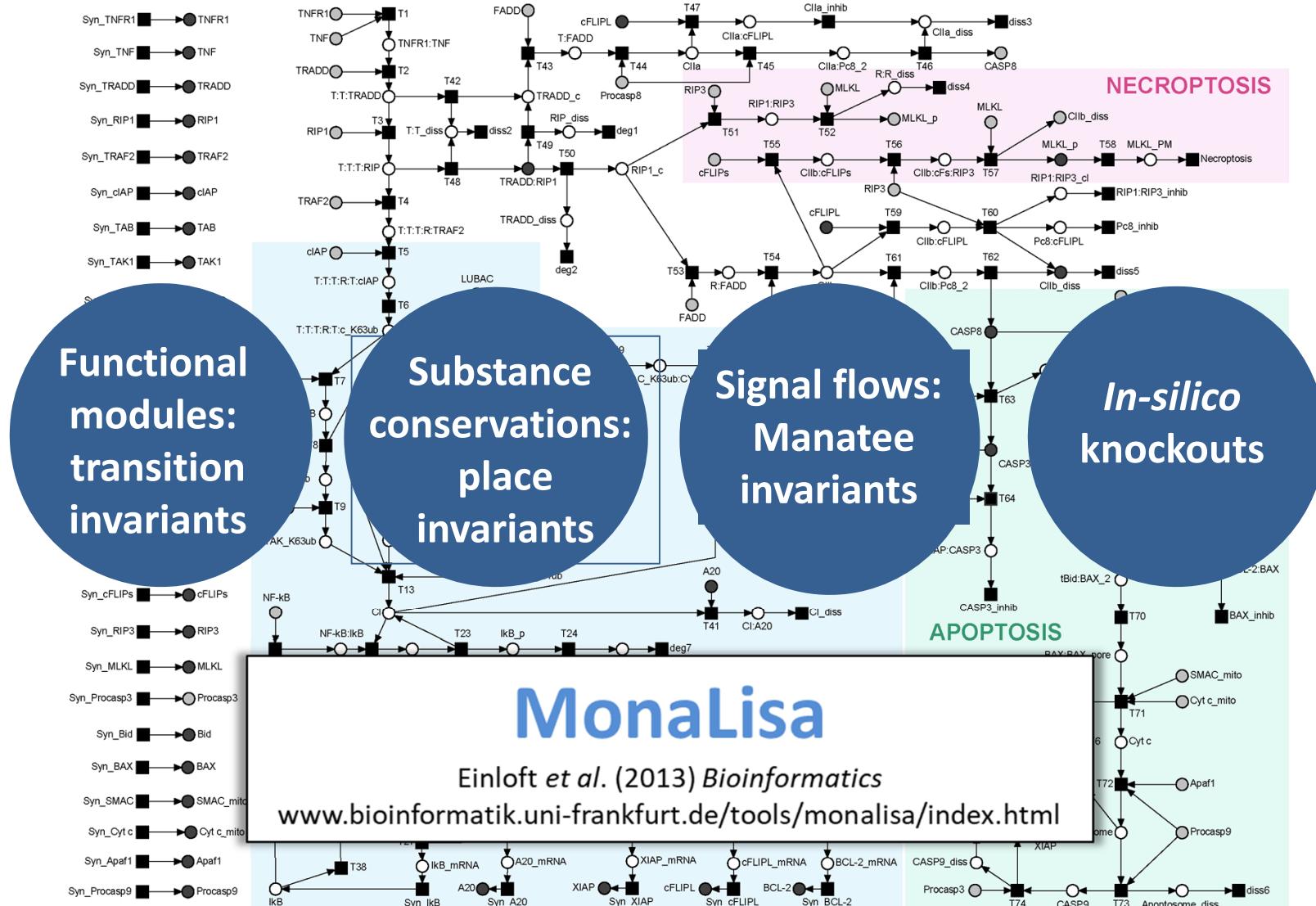


The TNFR1 signal transduction pathway



Amstein et al. (2017) BMC Systems Biol ; Amstein et al. (2022) PLoS Comp Biol

The Petri net analysis

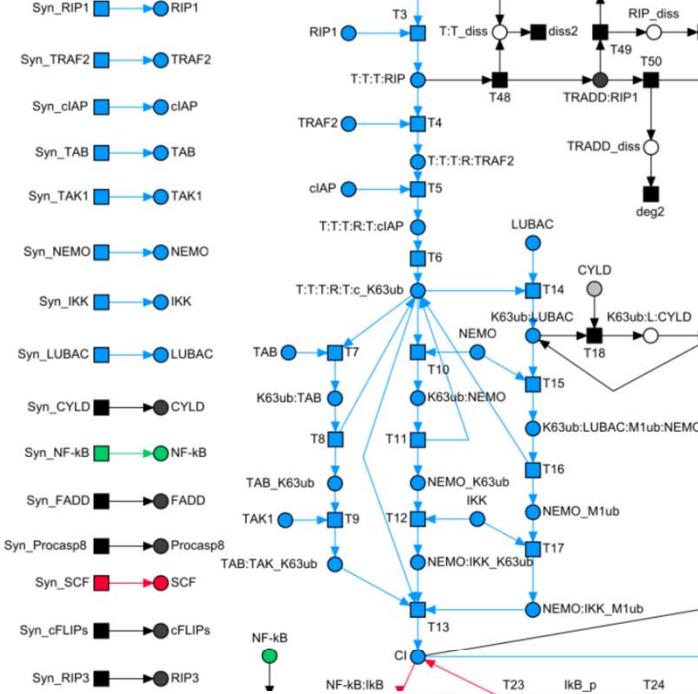


Malik-Sheriff et al. (2020) Nucl Acids Res

BioModels ID: MODEL2210170001

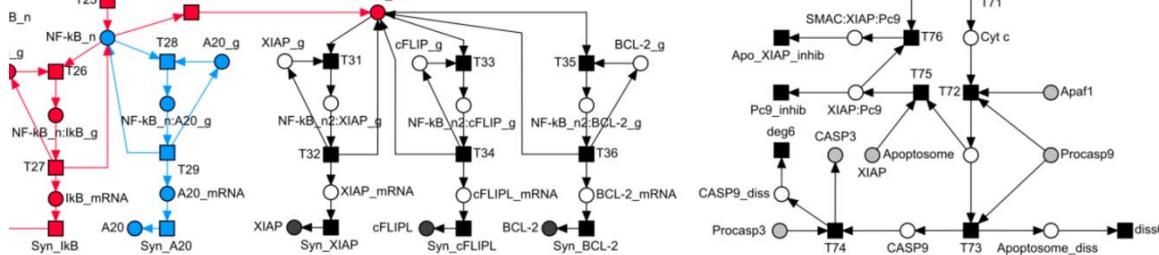
Manatee invariants predict signaling pathways

Formation of complex I and dissociation via A20



**Activation of NF-κB,
degradation of IκB
and gene
expression of IκB,
formation of the
inhibitory complex**

Turnover of NF-κB

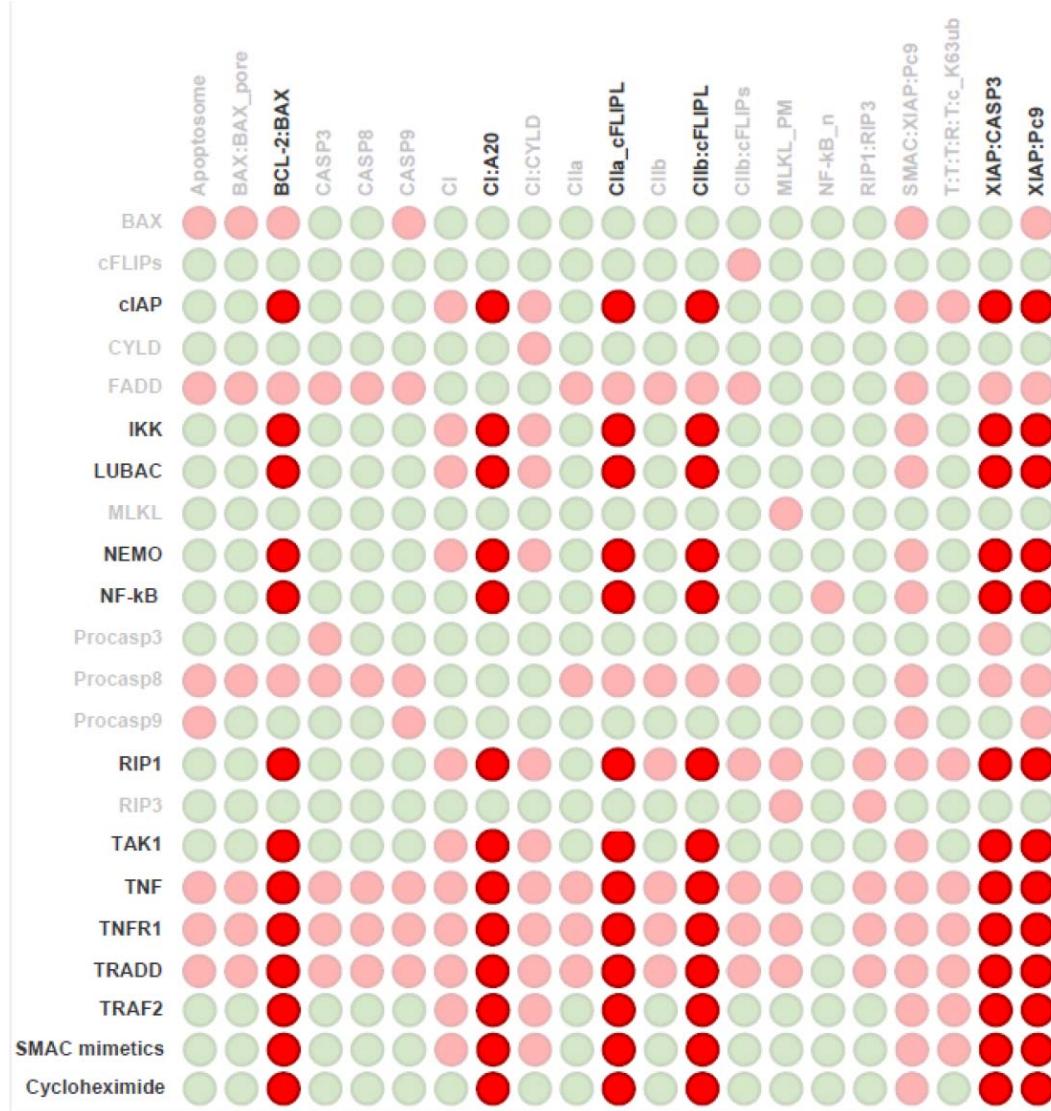


279 complete signal flows: Manatee invariants



Amstein *et al.*
(2017)
BMC Systems Biol

In-silico knockouts of the TNFR1 Petri net



Hannig et al. (2018) Bioinformatics

- ❖ Knockout matrix based on Manatee invariants
- ❖ 20 *in-silico* knockouts, 2 therapies (SMAC mimetic, cycloheximide), 21 complexes
- ❖ Identification of knockouts that overcome the robust *survival* response
- ❖ Identification of functional



isiKnock
Jennifer Hannig
(former Scheidel)

Christoph Welsch

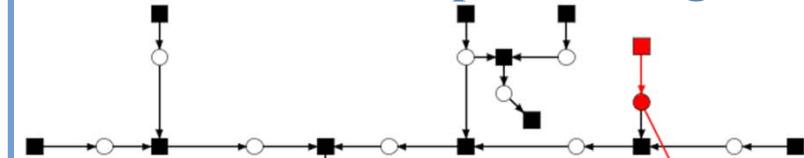


Heiko Mühl



Ongoing projects

IL-22 receptor binding



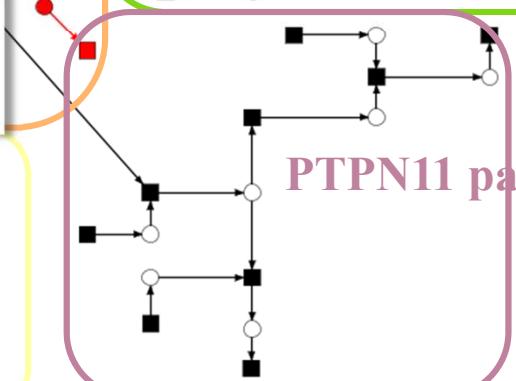
Classical Petri net and development
of a mass-action reaction model
showing the interactions between
IL-6 and IL-22 pathways
unpublished



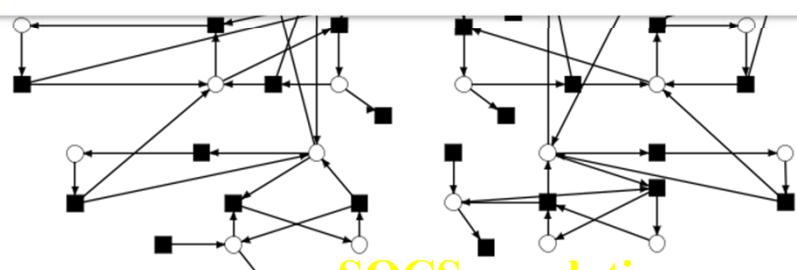
STAT binding

Koch & Büttner (2023) *Am J Physiol: Cell Physiol*
Computational modeling of signal transduction networks
without kinetic parameters: Petri net approaches

IL-6 receptor binding



PTPN11 pathway



SOCS regulation



Marcus Keßler

Ongoing projects

Influence of the gut microbiota on the translocation of bacteria and fungi in patients with ACLF

Agent-based model on the interaction between antibiotics and microbiota

- ❖ Bacterial movement through gut motility (peristalsis, laminar flow, water absorption)
- ❖ **Bacterial metabolism** (production of SCFA)
- ❖ Bacterial growth in dependence on the pH value



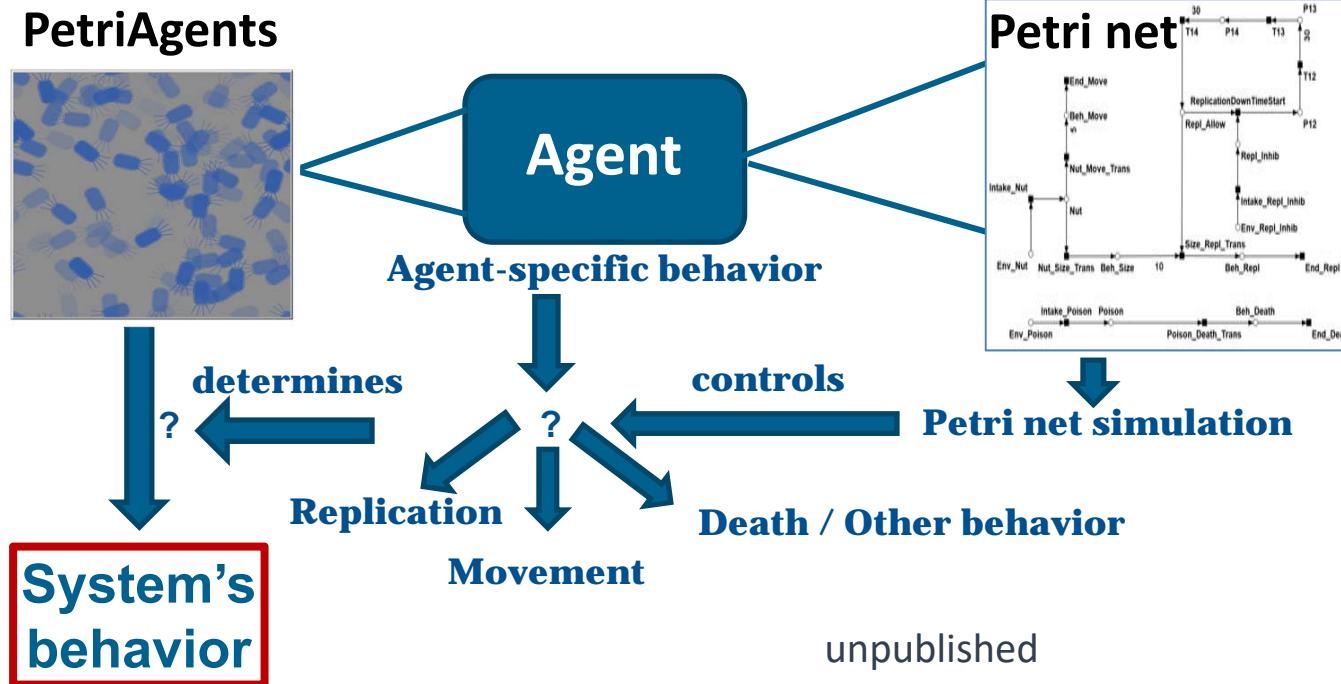
Kristiyana Tsenova



Jörg Ackermann

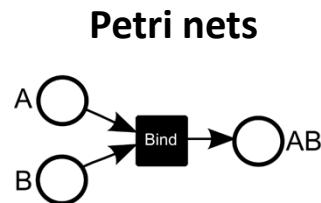
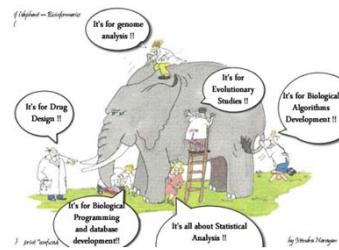


Marius Kirchner

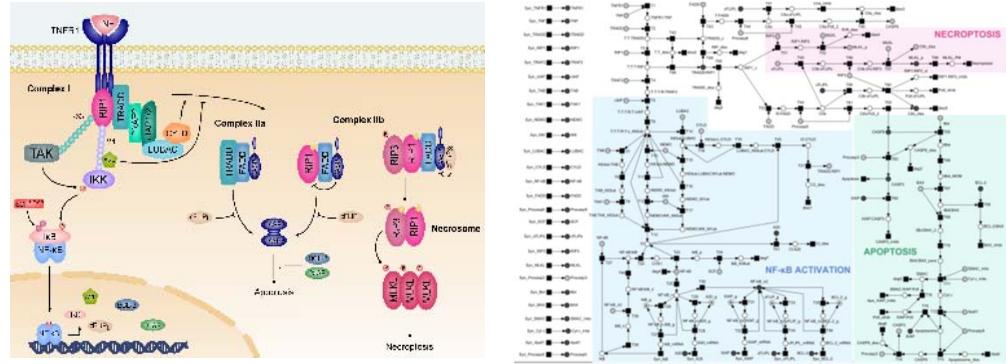


Summary

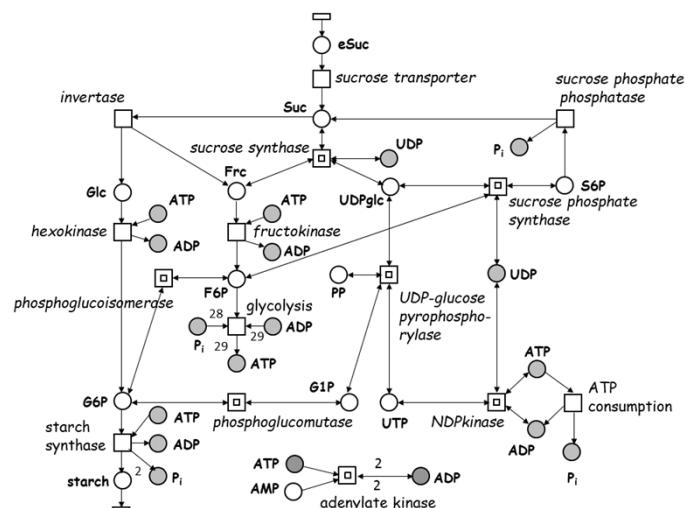
Challenges of data integration



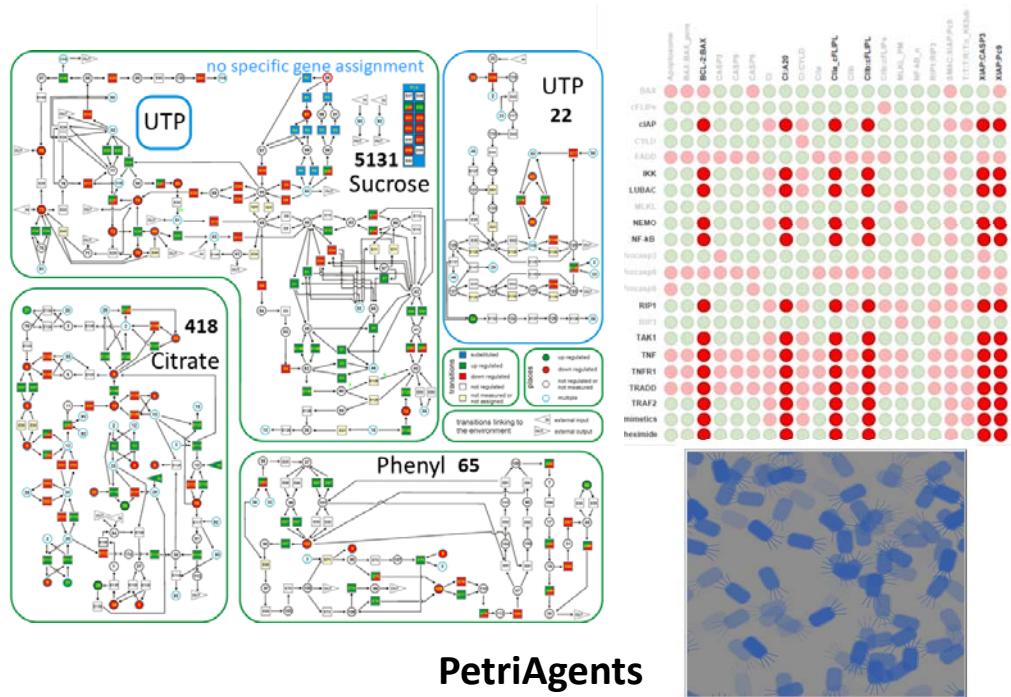
Petri net model of the TNFR1-mediated signaling pathways



Petri net model of the carbon metabolism in potato tubers and in *Arabidopsis thaliana*



Knockout analyses



Take-home messages

- ❖ The quality and quantity of the data determine the choice of the modeling method -> **Check your data** carefully
- ❖ The system's behavior can be predicted **without knowing the kinetic parameters**, using, e.g., Petri nets -> invariant and knockout analyses
- ❖ **Place invariants** describe **substance conservations**
- ❖ **Transition invariants** represent **functional modules**
- ❖ **Manatee invariants** are **linear combinations** of transition invariants -> **complete signaling pathways**
- ❖ **The bottleneck** is the availability of data of sufficient quality and quantity, especially for machine learning applications
- ❖ The exploration of **all possible system states** is still a challenge

There are many databases related to metabolism,
and there are for a given organism,
many metabolic networks that have been
reconstructed that provide different
and not always compatible information.

How the community could address these issues?

- ❖ Standardization
- ❖ Computational verification methods
- ❖ Database improvement by automatic curation,
 checking for contradictions
- ❖ **Validation, validation, validation!**

Acknowledgments

University Jena

Prof. Dr. Stefan Schuster

University Koblenz

Prof. Dr. Falk Schreiber

Goethe University Frankfurt

Prof. Dr. Ivan Đikić

Prof. Dr. Martin-Leo Hansmann

Prof. Dr. Christoph Welsch

Prof. Dr. Heiko Mühl

Prof. Dr. Maria Vehreschild

Ecole Normale Supérieure Paris

Prof. Dr. Denis Thieffry

University of Evry

Val d'Essonne, Évry

Prof. Dr. Anna Niarakis



Ein Clusterprojekt des Landes Hessen



Federal Ministry
of Education
and Research



Exzellente Forschung für
Hessens Zukunft

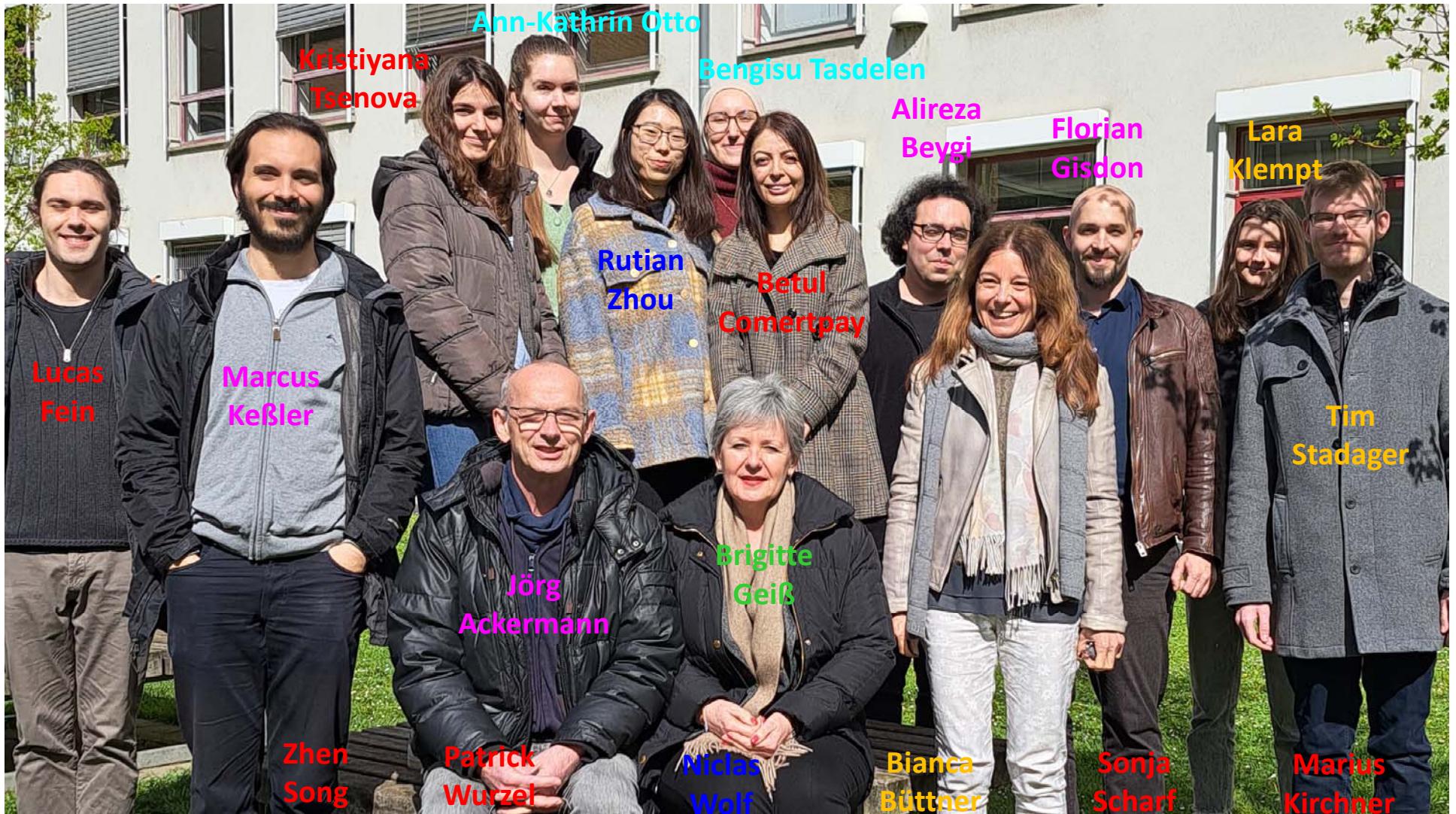


HESSEN



April 2023

Mol BI



Administration

Bachelor student

Master student

Ph.D. student

Postdoc, Alumni



*Thank you
for
your attention!*

System's invariants

Transition \ Place	r_1	r_2	r_{3f}	r_{3b}
C	-2	-1	-1	+1
O_2	-1	-1	0	0
CO	+2	0	+2	-2
CO_2	0	+1	-1	+1
init	0	0	0	0

place (P-) invariant: $C^T x = 0$

transition (T-) invariant: $C y = 0$

0: steady-state constraint

Search for **minimal nonnegative, nontrivial integer** solutions

Minimal: $\exists z: \text{supp}(z) \subseteq \text{supp}(u)$ and the largest common divisor of all non-zero entries of u is 1

P-invariants

$$\begin{aligned} -2x_1 - 1x_2 + 2x_3 &= 0 \\ -1x_1 - 1x_2 + 1x_4 &= 0 \\ -1x_1 + 2x_3 - 1x_4 &= 0 \\ +1x_1 - 2x_3 + 1x_4 &= 0 \\ +3x_1 + 2x_2 - 1x_5 &= 0 \\ -2x_3 - 1x_4 + 1x_5 &= 0 \end{aligned}$$

T-invariants

$$\begin{aligned} -2y_1 - 1y_2 - 1y_3 + 1y_4 + 3y_5 &= 0 \\ -1y_1 - 1y_2 + 2y_5 &= 0 \\ +2y_1 + 2y_3 - 2y_4 - 2y_6 &= 0 \\ +1y_2 - 1y_3 + 1y_4 - 1y_6 &= 0 \\ -1y_5 + 1y_6 &= 0 \end{aligned}$$

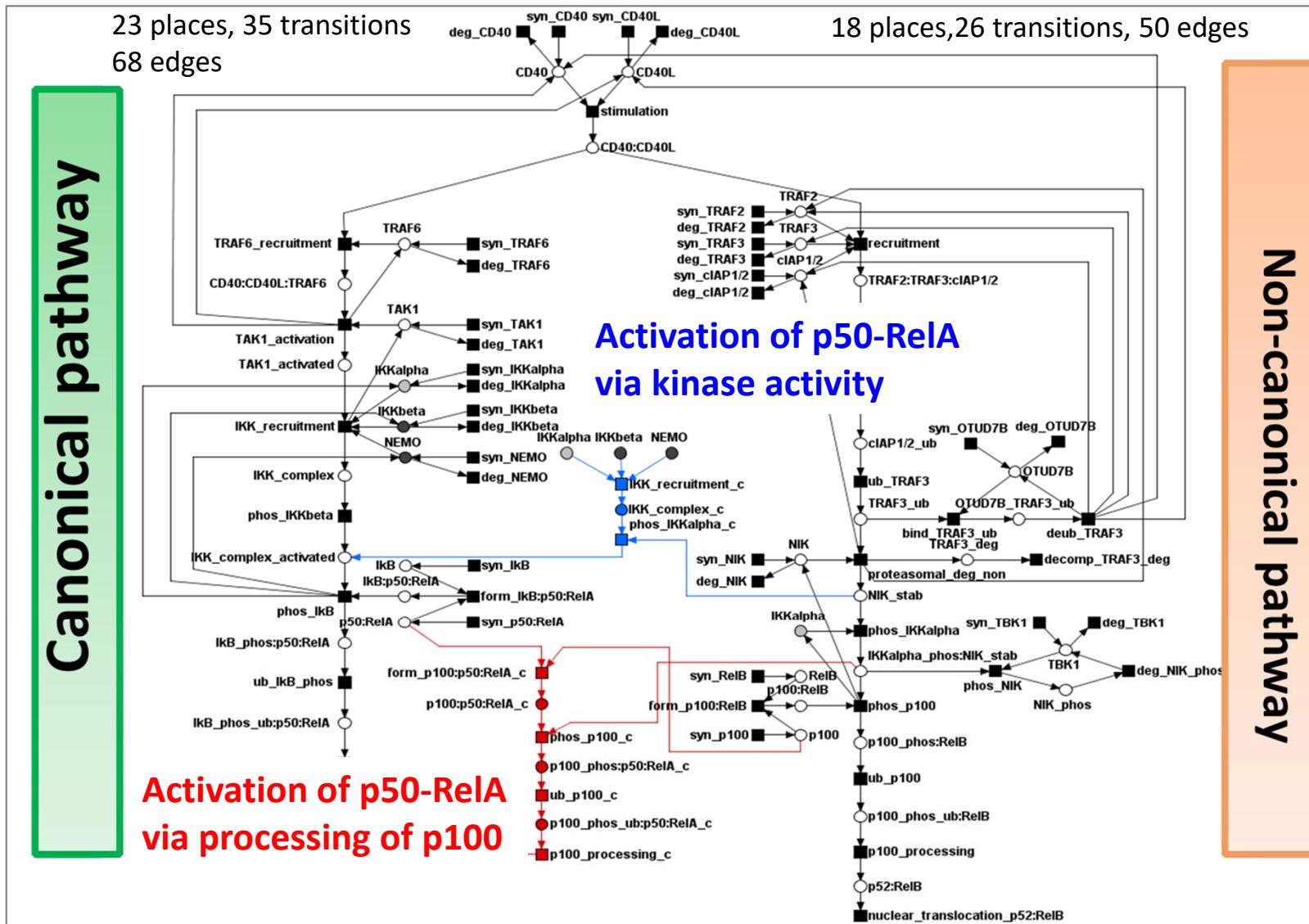
Parikh vector: vector of firing frequencies

The canonical and non-canonical NF-κB pathways and their crosstalk: a comparative study based on Petri nets

Trares et al. (2022) Biosystems; BioModels ID: MODEL2207210001/2/3



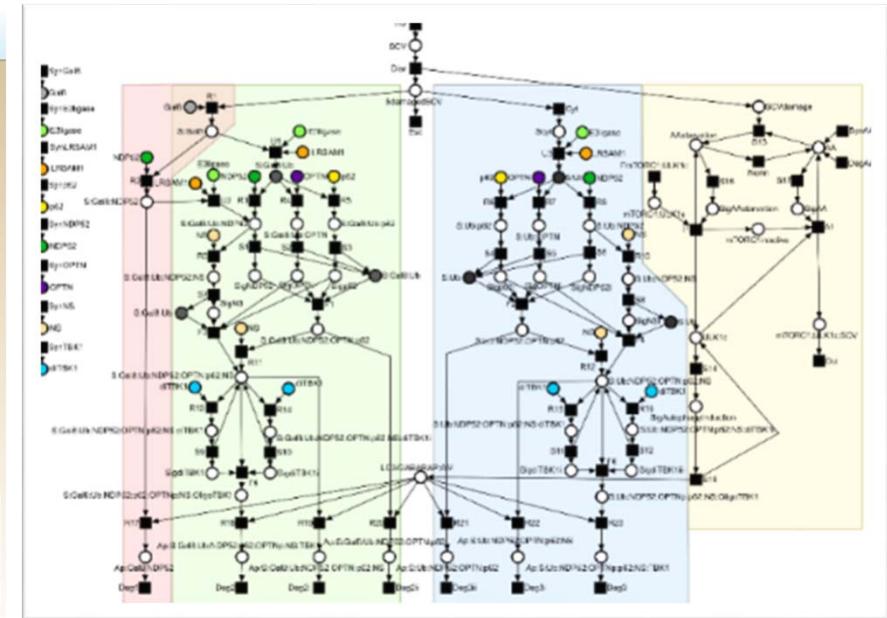
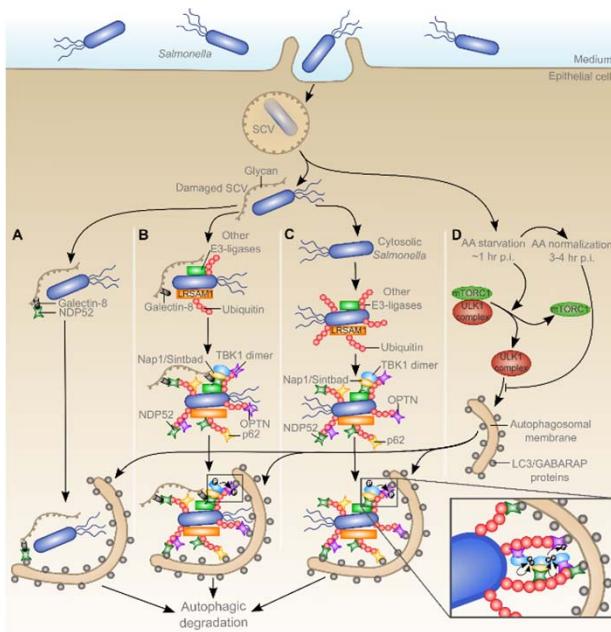
Kira Trares



Xenophagy in epithelial HeLa cells after *Salmonella Typhimurium* infection



Jennifer Hannig



- ❖ Classical Petri net
- ❖ Stochastic Petri net



Ivan Dikic

Hannig *et al.* (2018) *Bioinformatics*
Scheidel *et al.* (2016) *PLoS Computational Biology*

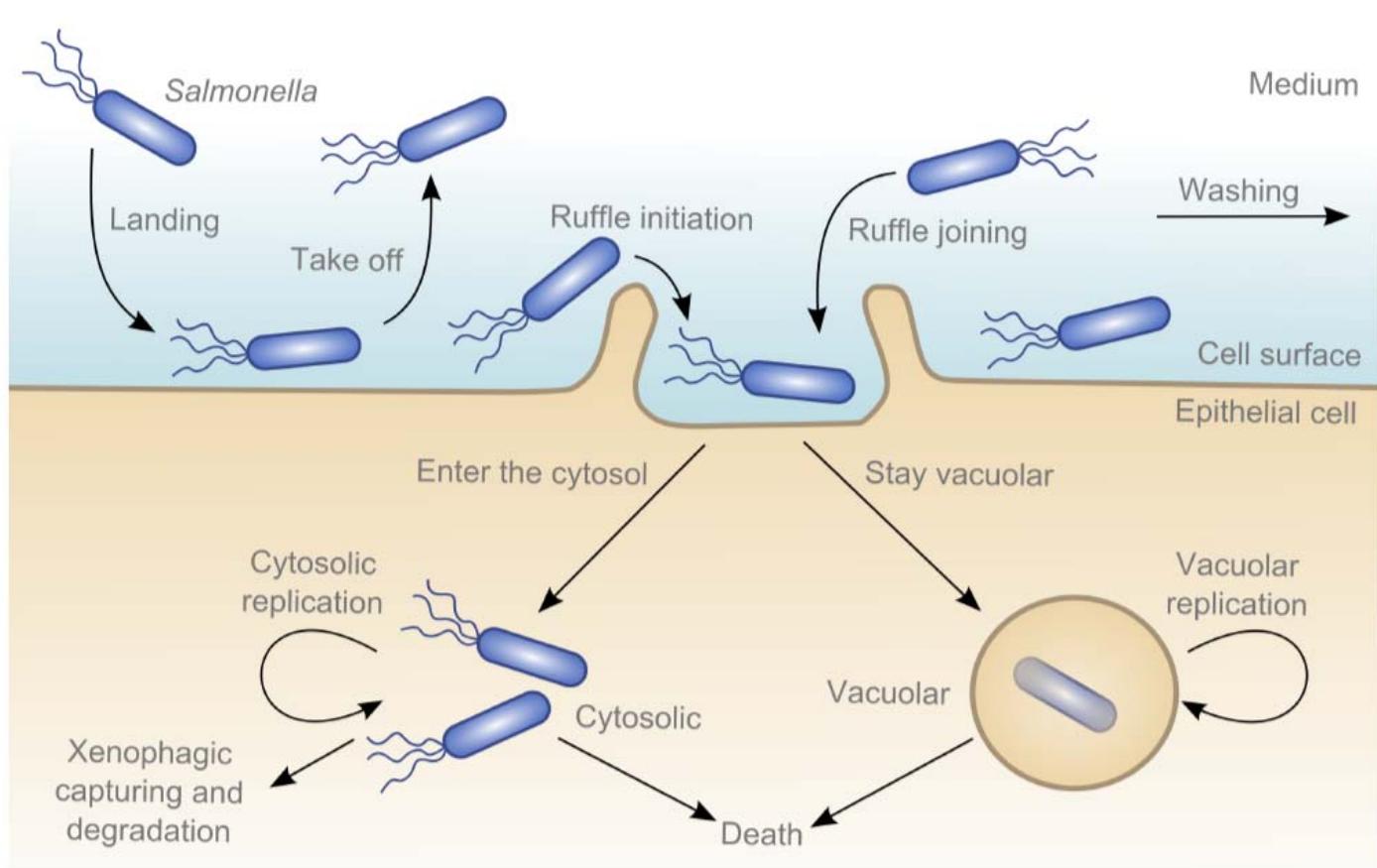
Agent-based model of *Salmonella* movement on the cell surface



Nasrin Alikhani
Chamgordani



Jennifer Hannig

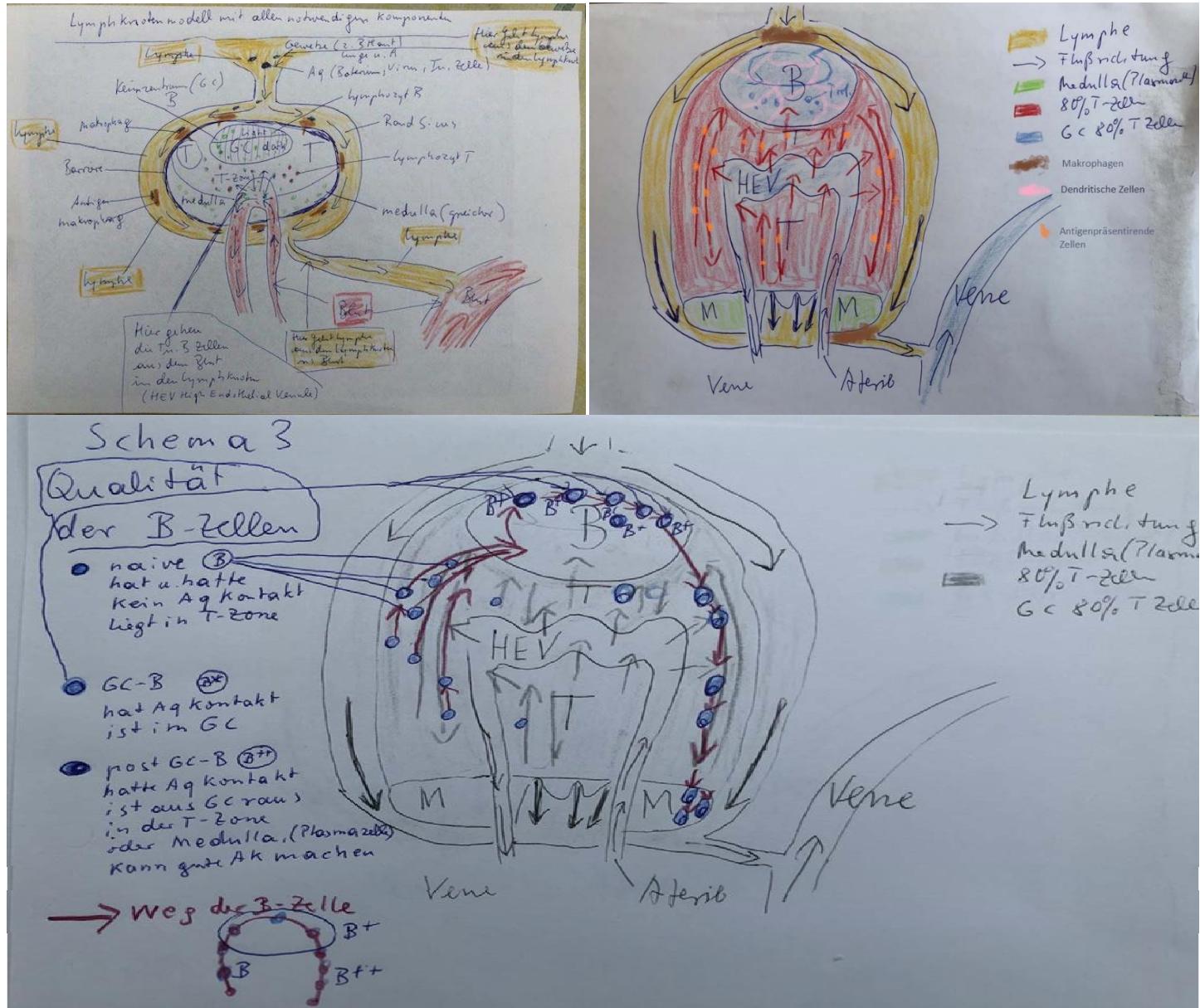


unpublished

A Petri net model of the human lymph node



Martin-Leo Hansmann



unpublished