

# Modeling and Analysis of

## Biological Systems

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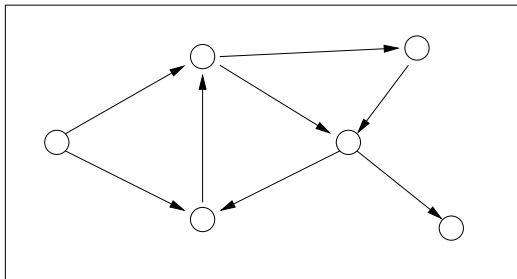
Part of the work described here is in collaboration with Alessandro Abate (Berkeley), Yu Bai (Stanford), Pat Lincoln, Merrill Knapp, Keith Laderoute, Carolyn Talcott (SRI)

# Modeling

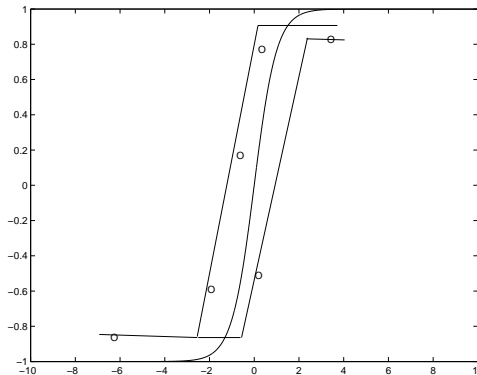
From experimental data to a formal model in some modeling formalism

Describes system at some level of abstractions

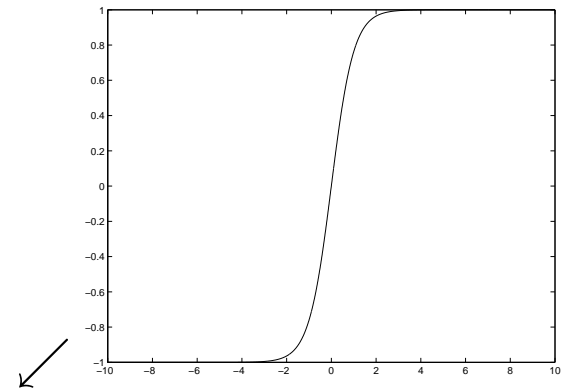
Discrete/Logical



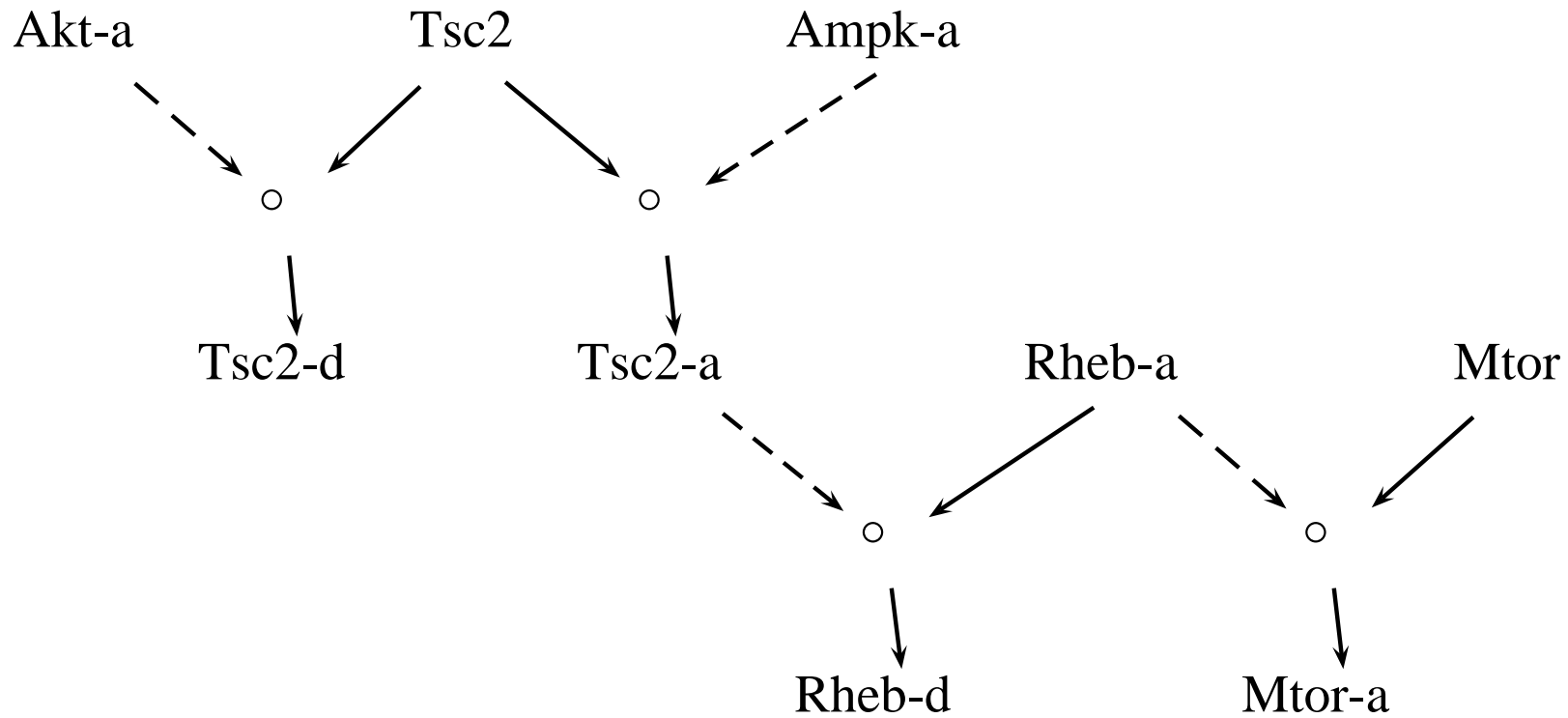
Hybrid



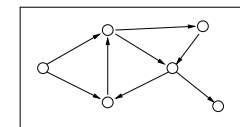
Continuous



# Logical Models: Example

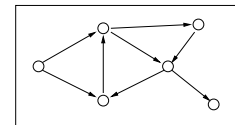


What does the diagram **mean**?

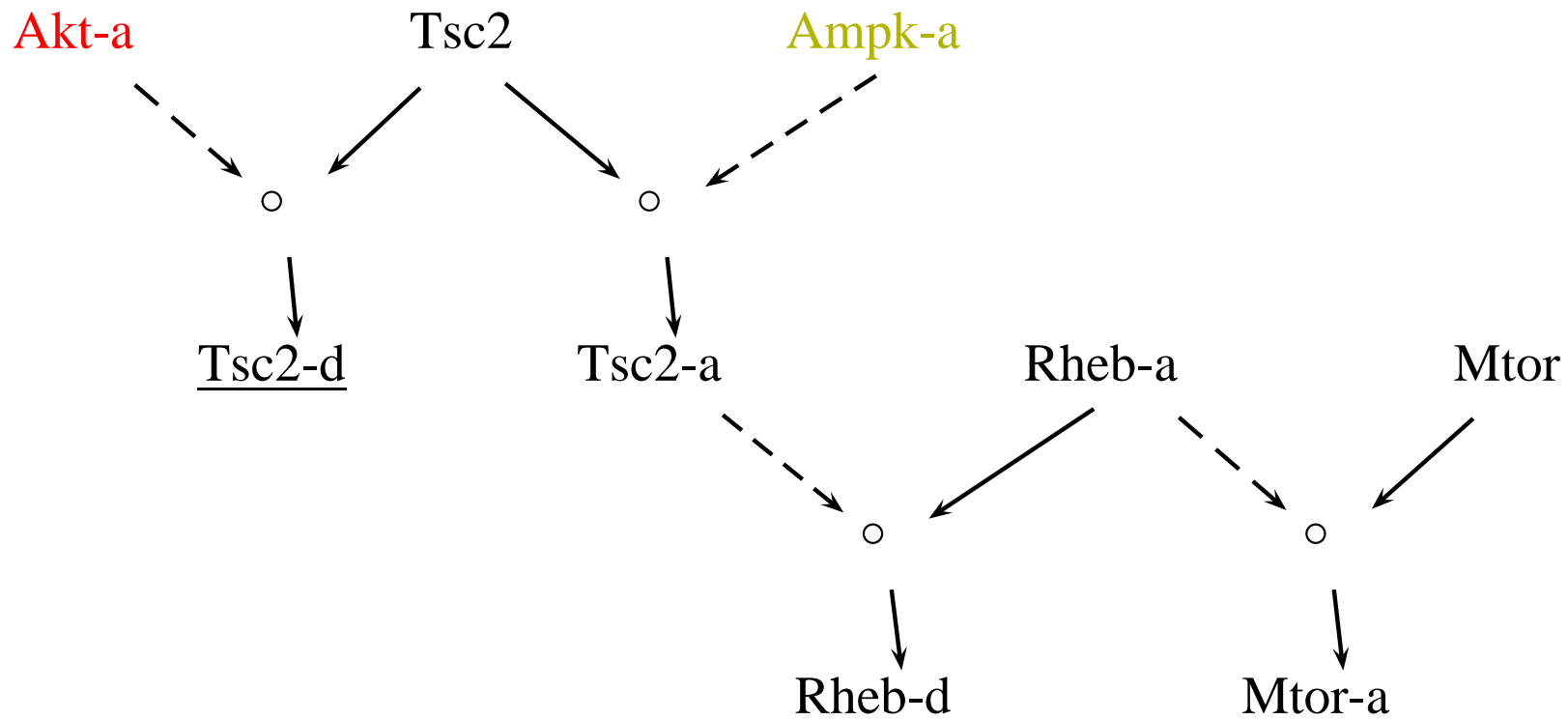


## Logical Models

- Describes the process at a high level of abstraction.
- Only qualitative behavior of the system is modeled
- Interpreted as a **petri-net**, lack of rate information means that it is interpreted as a **1-safe petri-net**
- **Pathway Logic** tool displays and analyzes these models
- Has been used to build models of **signaling pathways** in mammalian cells

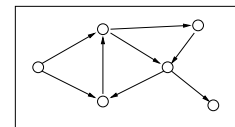


## Example



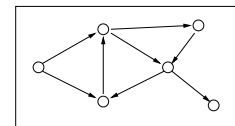
If **Akt-act** present, but **Ampk-act** is absent, then Tsc2 is deactivated and

Mtor-act is high

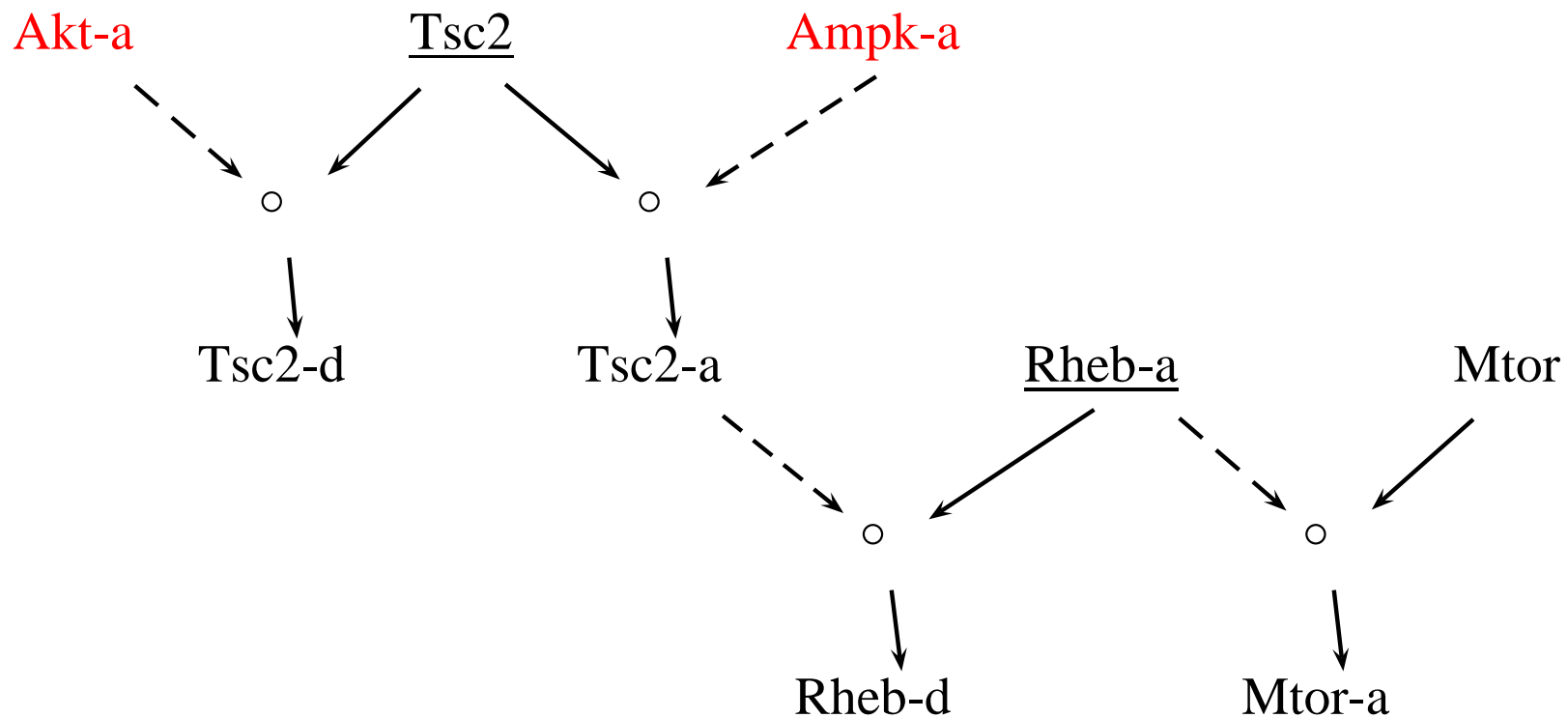


## Extended Logical Models

- **2-valued** interpretation gives limited information,
- Explored extensions of the 2-valued semantics by considering **3-valued qualitative interpretation** of the rules in the model
- Built a tool that analyzes these networks on 3 values
- Enables analysis of networks where a **resource** is be **shared** between two competing pathways

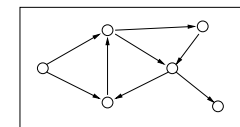


## Example



Difference in Mtor-act in the cases:

- (1) Akt-a and Ampk-a are both present,
- (2) Akt-a is present, but Ampk-a is not.

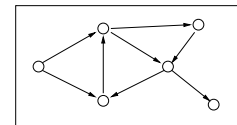


# Stochastic Extensions of Discrete Logical Models

Limited rate information can be added in the discrete logical models in the form of **transition probabilities**

The resulting model can be simulated using a **simplified** variant of Gillespie's stochastic simulation algorithm (with tau-leaping)

We obtain **time abstract** stochastic simulations of the stochastic pathway logic models in this way

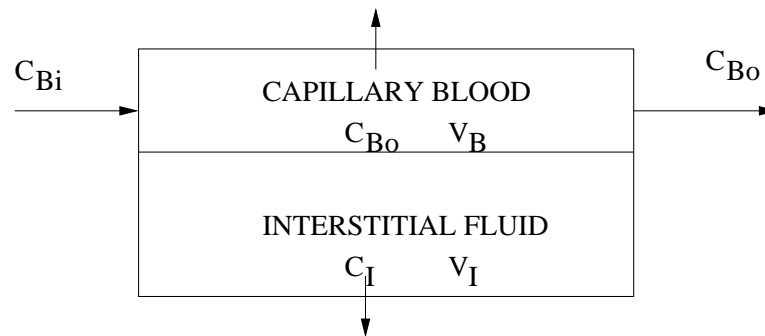




# Continuous Dynamical System Models

Continuous differential equations are used classically to build models

Example. A 6-compartment PD model of glucose metabolism in human body



The mass balances for a typical physiologic compartment:

$$V_B \dot{C}_{B_o} = Q_B(C_{B_i} - C_{B_o}) + PA(C_I - C_{B_o}) - r_{RBC}$$

$$V_I \dot{C}_I = PA(C_{B_o} - C_I) - r_T$$

$V$ : volume,  $C$ : concentration,  $Q$ : flow,  $r$ : rate

Hard to determine the rate constants, still uncertainty in predicated values



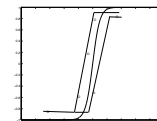
# Hybrid Systems

Combines **differential equations** with **discrete boolean logic**

$$\frac{d}{dt}[RapA] = \underline{DRapA} - LRapA * [RapA] - k_{12} * [Pep5i] * [RapA]$$

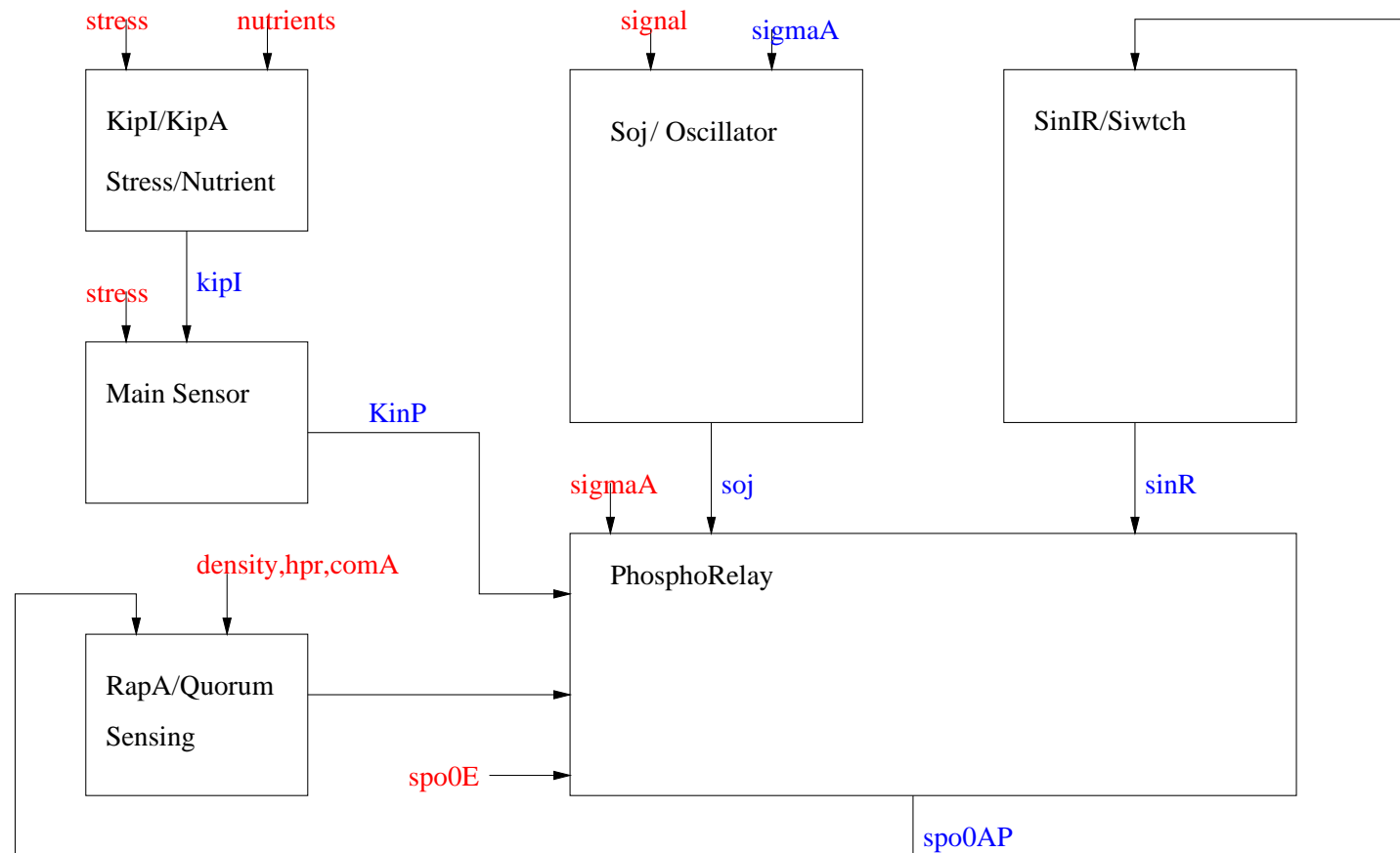
```
DRapA = IF ( comAP_high? ) THEN 1  
        ELSIF ( spo0AP2_high? ) THEN 0  
        ELSIF ( hpr_high? ) THEN 0  
        ELSE 1/2 ENDIF;
```

- **Richer language** for modeling at different levels of abstraction
- **Nondeterminism** allows for unknown parameter values
- **Compositional** development of complete model



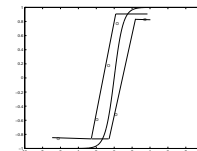
# A HybridSal Model

A **component based view** of the Sporulation Initiation Network in *B.Subtilis*



# The HybridSal Abstractor

- Creates a **conservative discrete approximation** of the **hybrid model**
- The discrete abstraction has all **behaviors** of the original nondeterministic (partially unspecified) model
- The abstractor works **compositionally** and abstracts the models by abstracting its **components** of the model
- It can **ignore** certain parts of the model and **focus** on other parts of **interest to the biologist**
- It can create **multiple abstract views** of the same base model
- Unknown rate constants can be **symbolically constrained**, such as  $(k_{12} > k_{21})$



## The Sal Model Checker

- The discrete abstract model is explored using a **symbolic model checker**

- Routinely search through state space of size  $2^{100}$  and beyond

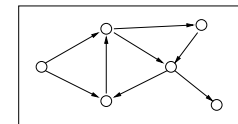
- Can extract **interesting behaviors** that the model exhibits:

*Under the given environment, can the cell go into a high SpooAP state?*

- Can also **provably verify** that certain things **never** happen

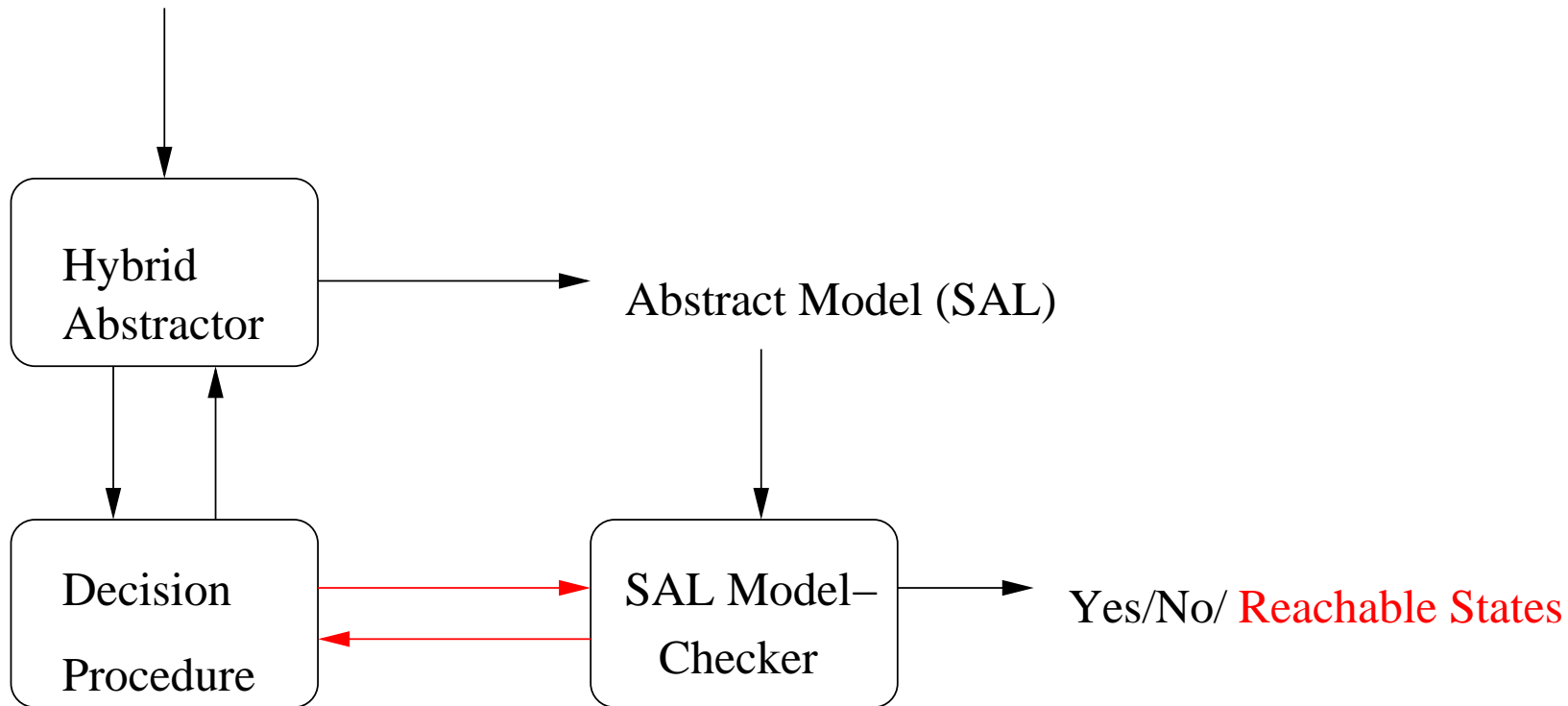
*It is impossible for the concentrations of proteins A and B to be high simultaneously.*

*If the cell enters a particular configuration, it does not get out of it unless the environmental signals change.*

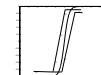


# The HybridSal Tool Architecture

HybridSAL Model



The **decision procedure** for the **quantifier-free theory of reals** powers the HybridSal tools



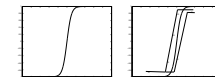
## Model Simplification

It is **computationally difficult** to analyze continuous and hybrid models with unknown parameters

Need to **simplify** the model without compromising much on the behaviors of the model

A **new** method for **model simplification** based on a **symbolic procedure** for reasoning about the real numbers:

- Not all terms in an ODE are equally important. Some reactions are **more influential** in determining the overall behavior
- The fact that certain terms **contribute little** to the **overall dynamics** can be formally stated and proved using a symbolic reasoning engine (that decides the theory of reals).



## Model Simplification: Example

Model of tetracycline resistance in bacteria (UPenn):

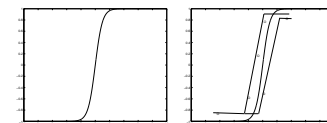
$$d[TetR]/dt = f_1 - k_d[TetR] - k_+[Tc][TetR] + k_-[TetRTc]$$

$$d[TetRTc]/dt = k_+[Tc][TetR] - k_-[TetRTc] - k_d[TetRTc]$$

$$d[Tc]/dt = k_i([Tc]^0 - [Tc]) - k_p[Tc][TetA] - k_+[Tc][TetR] + k_-[TetRTc] - k_d[Tc]$$

$$d[TetA]/dt = f_2 - k_d[TetA]$$

$$f_1, f_2 = \begin{cases} b_1 & \text{if } [TetR] > 1/c_1 \\ b_1 + k_1 & \text{otherwise} \end{cases}$$



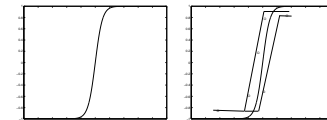


## Model Simplification: Example

We **simplify** the above system into the following system:

$$\begin{aligned}d[TetR]/dt &= f_1 - k_+[Tc][TetR] + k_-[TetRTc] \\d[TetRTc]/dt &= k_+[Tc][TetR] - k_-[TetRTc] - k_d[TetRTc] \\d[Tc]/dt &= k_i[Tc]^0 - k_p[Tc][TetA] \\d[TetA]/dt &= f_2 - k_d[TetA]\end{aligned}$$

This is a *sound* simplification: prove that the contribution of **eliminated terms** is much smaller than that of the few *dominating* retained terms



## Model Simplification: Technique

Let  $[TetR]_0, [TetRTc]_0, [Tc]_0, [TetA]_0$  denote the equilibrium concentrations. Hence, we get the formula  $C$ :

$$\begin{aligned}
 f_1 - k_d[TetR]_0 - k_+[Tc]_0[TetR]_0 + k_-[TetRTc]_0 &= 0 \wedge \\
 k_+[Tc]_0[TetR]_0 - (k_- + k_d)[TetRTc]_0 &= 0 \wedge \\
 k_i([Tc]^0 - [Tc]_0) - k_p[Tc]_0[TetA]_0 - k_+[Tc]_0[TetR]_0 + \\
 + k_-[TetRTc]_0 - k_d[Tc]_0 &= 0 \wedge \\
 f_2 - k_d[TetA]_0 &= 0
 \end{aligned}$$

We prove the following:

$$C \Rightarrow 10k_d[TetR]_0 < f_2$$

$$\begin{aligned}
 C \Rightarrow (10k_i[Tc]_0 < k_i[Tc]^0 \wedge 10k_+[Tc]_0[TetR]_0 < k_p[Tc]_0[TetA]_0 \wedge \\
 10k_-[TetRTc]_0 < k_i[Tc]^0 \wedge 10k_d[Tc]_0 < k_i[Tc]^0)
 \end{aligned}$$

# Model Simplification: Example

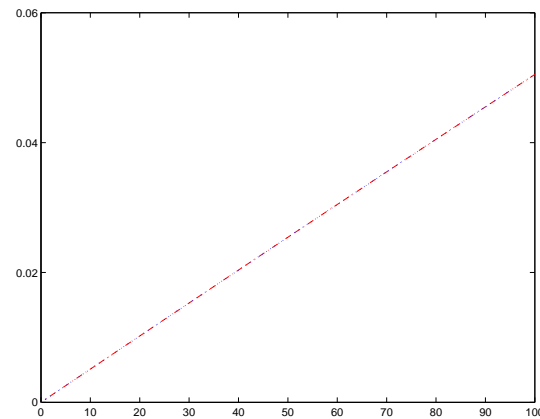
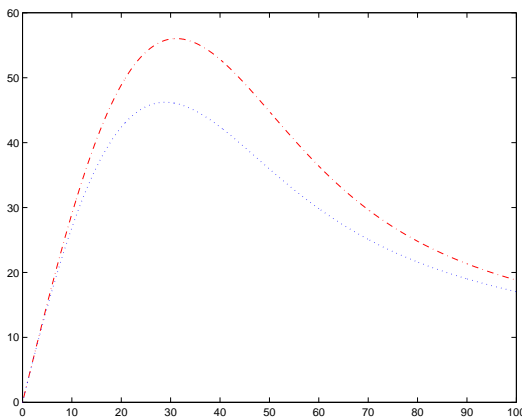
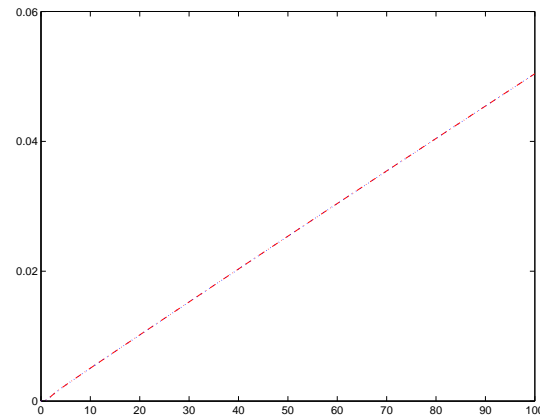
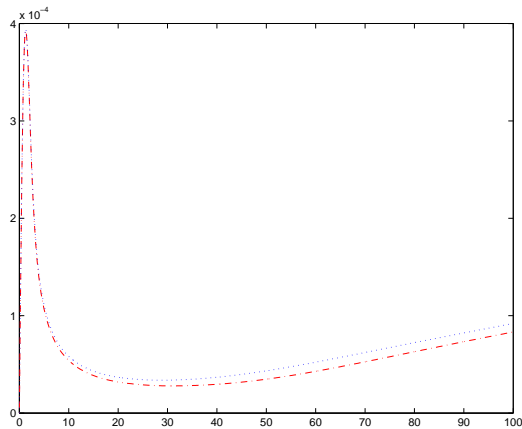


Table 1: Simulation plots for **100** time steps of the original model (red) and simplified model (blue) for each of the four species.

# Model Simplification: Example

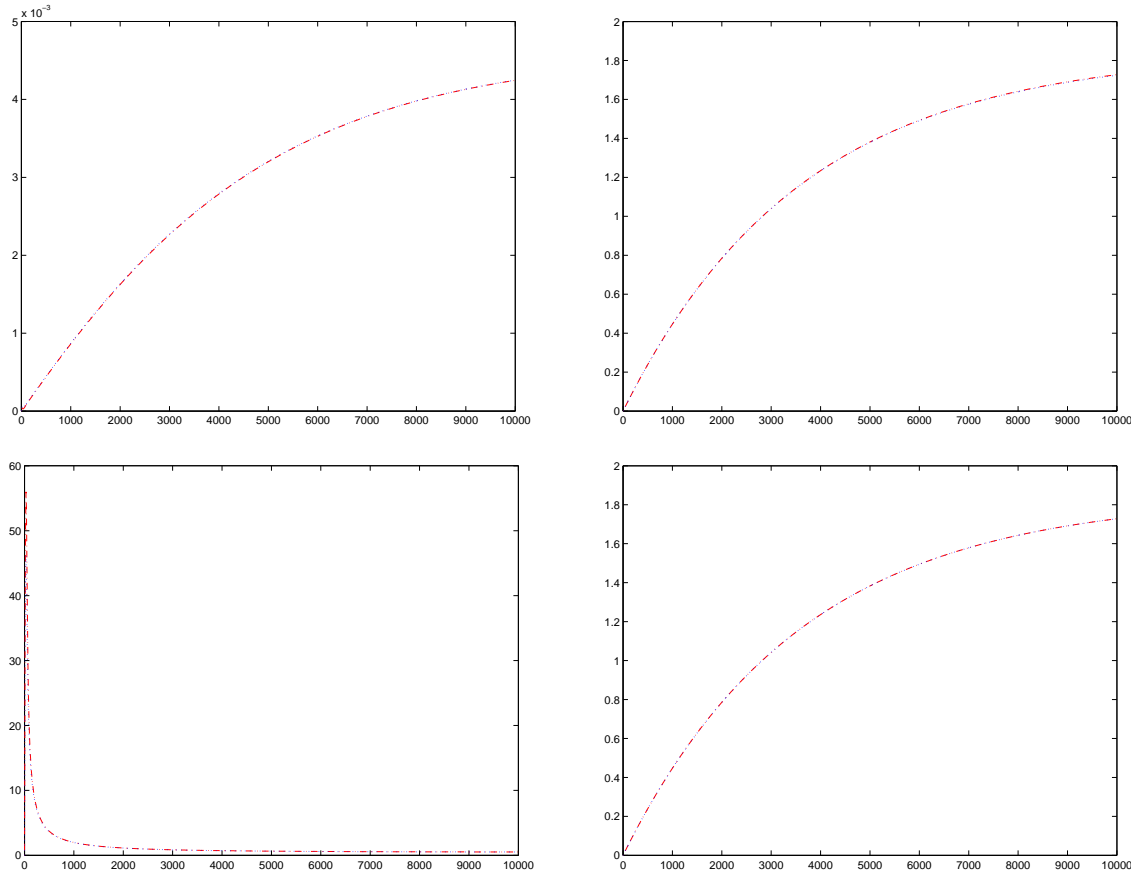


Table 2: Simulation plots for 10000 time steps of the original model (red) and simplified model (blue) for each of the four species.

## Symbolic Procedure for the Reals

The core technology we use in verifying model simplification steps and performing hybrid abstraction:

Given a set of nonlinear equations and inequalities:

$$p \approx 0, \quad p \in P$$

$$q > 0, \quad q \in Q$$

$$r \geq 0, \quad r \in R$$

where  $P, Q, R \subset \mathbb{Q}[\vec{x}]$  are sets of polynomials over  $\vec{x}$

Is the above set satisfiable over the reals?

## Generalized Simplex for Nonlinear Constraints

If all polynomials are **linear**, then **Simplex** LP solver can be used

- Introduce **slack variables** s.t. all inequality constraints are of the form  $v > 0$ , or  $w \geq 0$

$$\begin{array}{l} P = 0, \quad Q > 0, \quad R \geq 0 \quad \mapsto \\ \underline{P = 0}, \quad \underline{Q - \vec{v} = 0}, \quad \underline{R - \vec{w} = 0}, \quad \vec{v} > 0, \quad \vec{w} \geq 0 \end{array}$$

- **Search** for a polynomial  $p$  s.t.

$$\begin{array}{l} \underline{P = 0} \quad \Rightarrow \quad p \approx 0 \\ \vec{v} > 0, \quad \vec{w} \geq 0 \quad \Rightarrow \quad p > 0 \end{array}$$

- To search for  $p$ , compute the **Gröbner basis** for  $P$  using different possible orderings (pivot)

## Example: Nonlinear Constraint Solving

Consider  $E = \{x^3 = x, x > 2\}$ .

$$\begin{array}{r} x^3 - x \approx 0, \quad x - v - 2 \approx 0 \\ \hline (v + 2)^3 - (v + 2) \approx 0, \quad x - v - 2 \approx 0 \\ \hline (v + 2)(v + 1)(v + 3) \approx 0, \quad x - v - 2 \approx 0 \\ \hline \perp \end{array}$$

Computing GB and projecting it onto the slack variables discovers the witness  $p$  for unsatisfiability

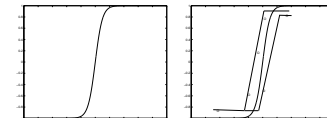
A. Tiwari, “An algebraic approach for the unsatisfiability of nonlinear constraints.” In *Computer Science Logic, CSL 2005*, Vol 3634 of LNCS, pp 248–262, Springer.

## A General Principle for Physical Systems?

The analysis of models of various biological systems shows that *whenever the system is **stable**, it is also **stable** in a sense **stronger than asymptotic stability** (in case of linear systems)*

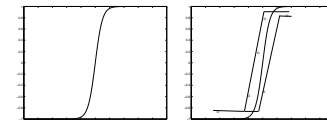
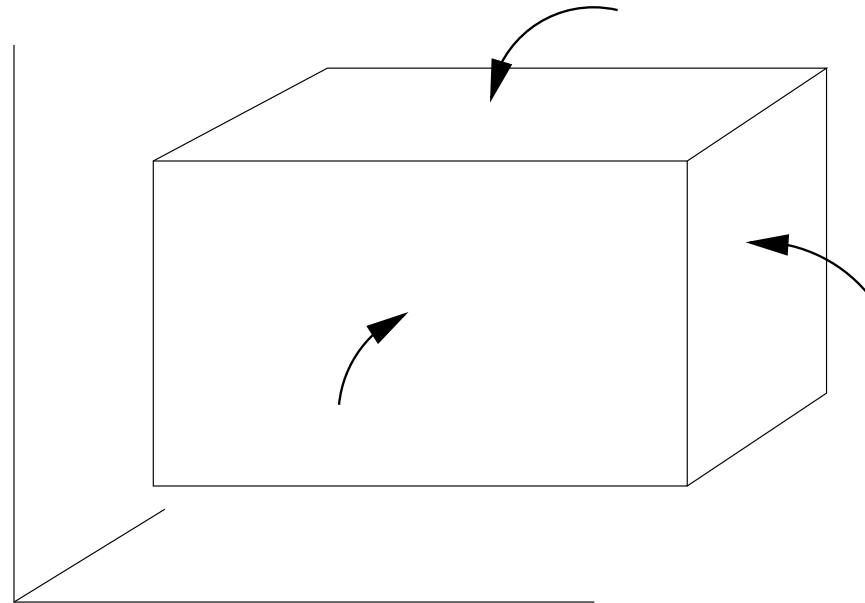
We have called this notion **box stability**

**Box stability.** A system is box stable around an equilibrium point  $\vec{x}_0$  if there is a rectangular box  $\vec{l} \leq \vec{x} \leq \vec{u}$  containing  $\vec{x}_0$  s.t. the vector field points inwards on all surfaces of the box.





## Box stability: Illustration



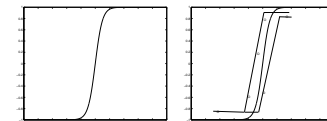
## Box stability: Examples

Examples of systems that were found to be **box stable**:

- Glucose and insulin metabolism in human body
- Tetracycline resistance in bacteria (UPenn)
- Regulation of induction in the *lac* operon in *E. Coli* (UPenn)
- *B. subtilis* sporulation initiation network (LBL)
- Delta-Notch intercellular signaling mechanism (Stanford)

Example that was **not box-stable**:

cardiovascular model of heart (oscillatory)

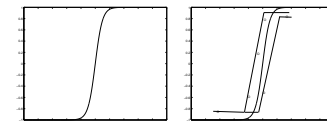


## Box stability: Why?

- Computationally more **tractable**
- Helps in **safety verification** of systems
- It has a very **natural** interpretation
- Implies asymptotic stability in the linear case

### Implications of the general observation

- Yields **new computational methods** to analyze linear and nonlinear systems
- As a principle, allows generation of **constraints** on **unknown parameters**
- Gives new techniques for defining **sensitivity** and analyzing it



# The BioSal Summary

