



Absolute Robustness in Deterministic and Stochastic Chemical Reaction Networks

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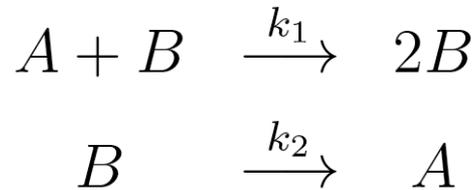
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Absolute robustness: Toy model



ODE:

$$A' = -k_1 AB + k_2 B$$

$$B' = k_1 AB - k_2 B$$

Mass conservation:

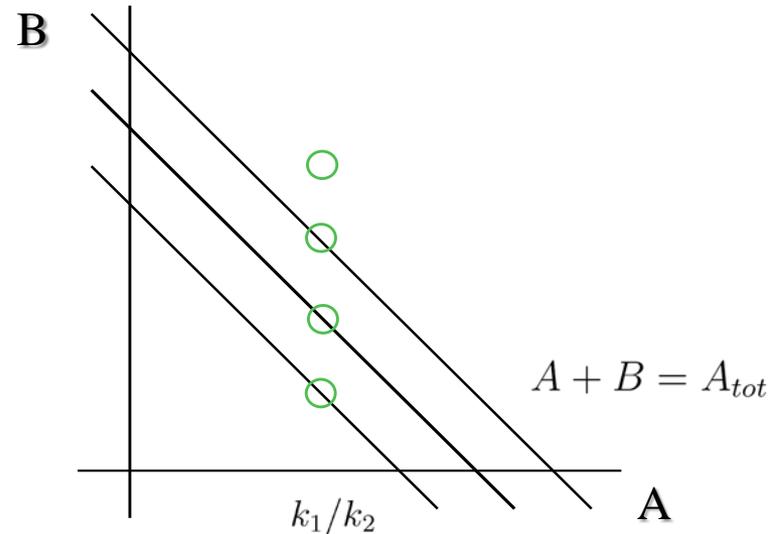
$$A + B = A_{tot}$$

At steady state:

$$k_1 AB = k_2 B$$

$$A = k_1/k_2$$

$$(B \neq 0)$$



- The steady state concentration of A is *absolutely robust* to changes in the total protein concentration A_{tot}
- If A is the active version of the protein, then this system is robust to protein fluctuations at steady state



Absolute robustness: General result

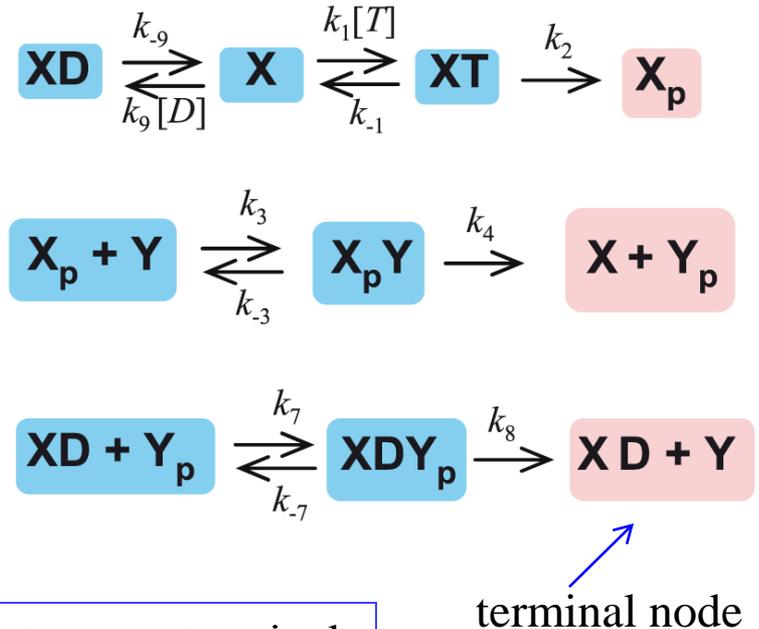
Deficiency:

$$\delta := \mathcal{C} - \ell - \text{rank } \Gamma$$

nodes (aka complexes)

connected components

nonterminal node

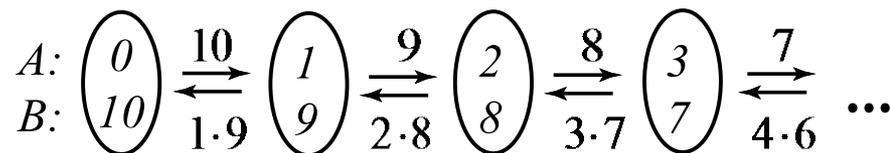
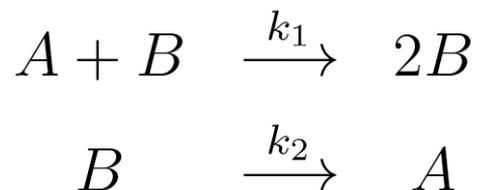


THEOREM: Suppose that $\delta=1$ and that there exist two nonterminal nodes that differ by a single variable A. Then A is absolutely robust.

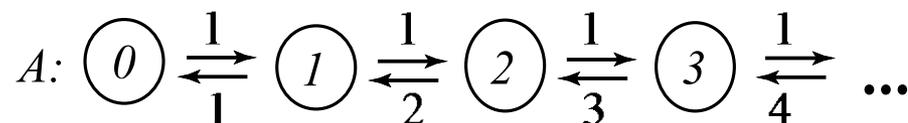
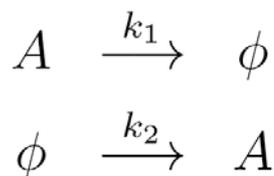
- In the example above, the variable \mathbf{Y}_p is absolutely robust, i.e. its concentration at steady state is independent of total protein concentrations \mathbf{X}_{tot} and \mathbf{Y}_{tot}
- Absolute robustness is a consequence of the structure of the network, not of specific parameter values



Stochasticity and absolute robustness: toy model



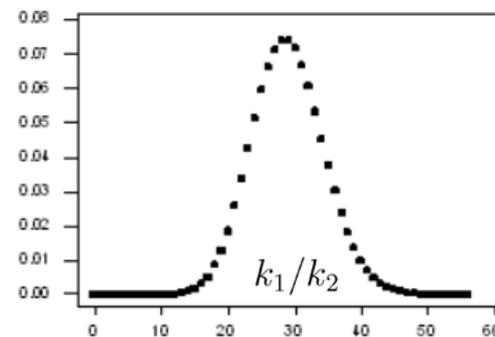
Notice the similarity with



which has Poisson distribution

$$P(A = k) = \frac{\lambda^k}{k!} e^{-\lambda}$$

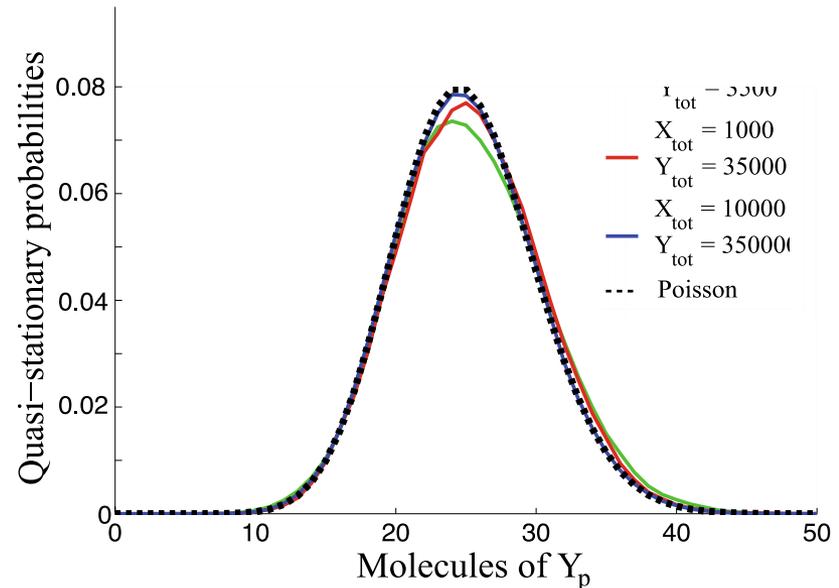
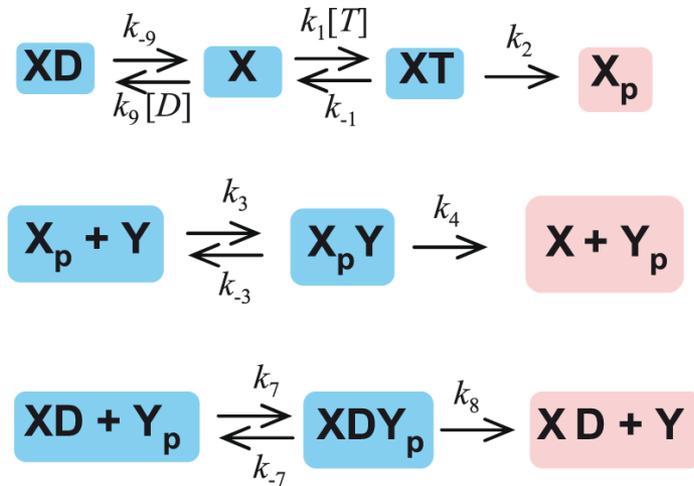
$$\begin{aligned}
 \lambda &= k_2/k_1 \\
 E(A) &= \lambda \\
 \sigma^2(A) &= \lambda
 \end{aligned}$$



- Simulations show that the toy model has transient steady state distributions that are very similar to a Poisson distribution and centered around the deterministic mean



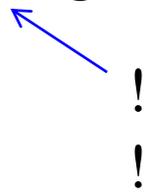
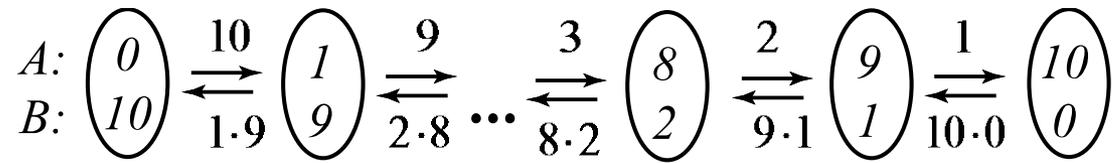
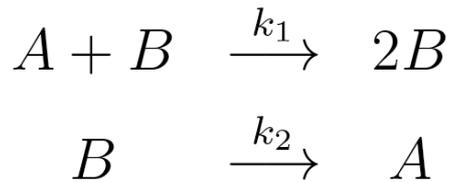
Stochasticity and absolute robustness: General simulations



- Gillespie simulations were carried out with more complex systems such as the EnvZ/OmpR osmoregulation model (above)
- After simulating absolutely robust systems for some time, they *appear* to converge towards the same Poisson distribution regardless of total concentrations
- Mean of the distribution corresponds to the steady state of deterministic system



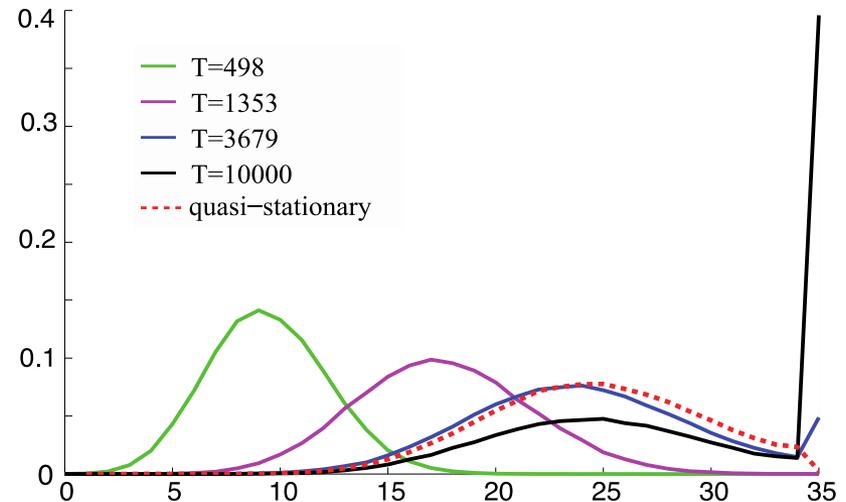
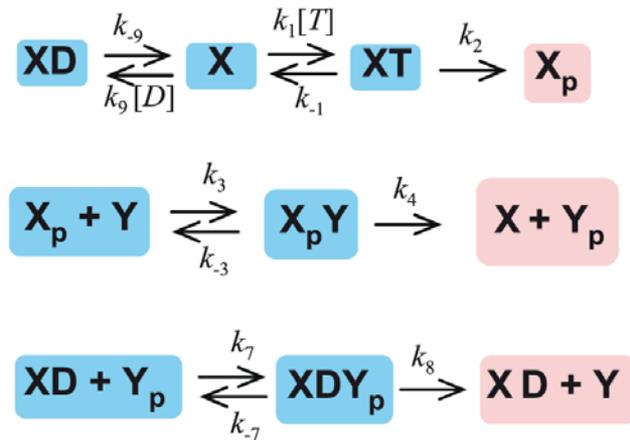
Stochastic extinction: Toy model



- Although this system transiently seems to converge to a Poisson distribution, when the system reaches the state (10,0) it gets stuck at the boundary (!)
- This extinction event cannot happen in the deterministic system, but in the stochastic case it happens with probability 1 given enough time
- Simple example of a significant qualitative difference between a continuous chemical reaction model and its corresponding stochastic system



Stochastic extinction: General result



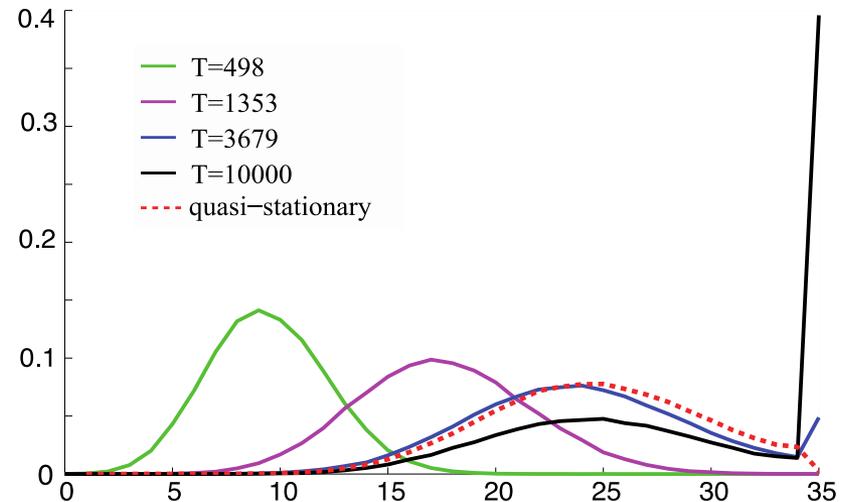
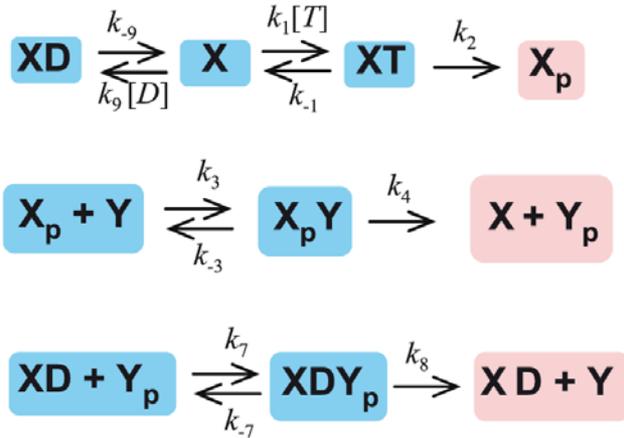
Definition: A chemical reaction network is *conservative* if every species is included in some mass conservation relation.

Definition: Given a state c , we say that a node in the network graph is *off* if at least one of the species in the complex is equal to zero.

Theorem: Suppose that $\delta=1$, there exist two nonterminal nodes that differ by a single variable A , and the system is conservative. Then for every solution of the stochastic system, every nonterminal node eventually turns (and stays) off.



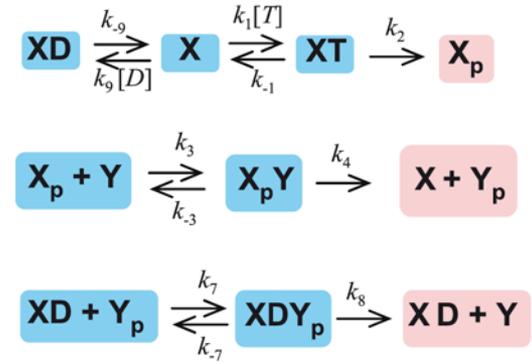
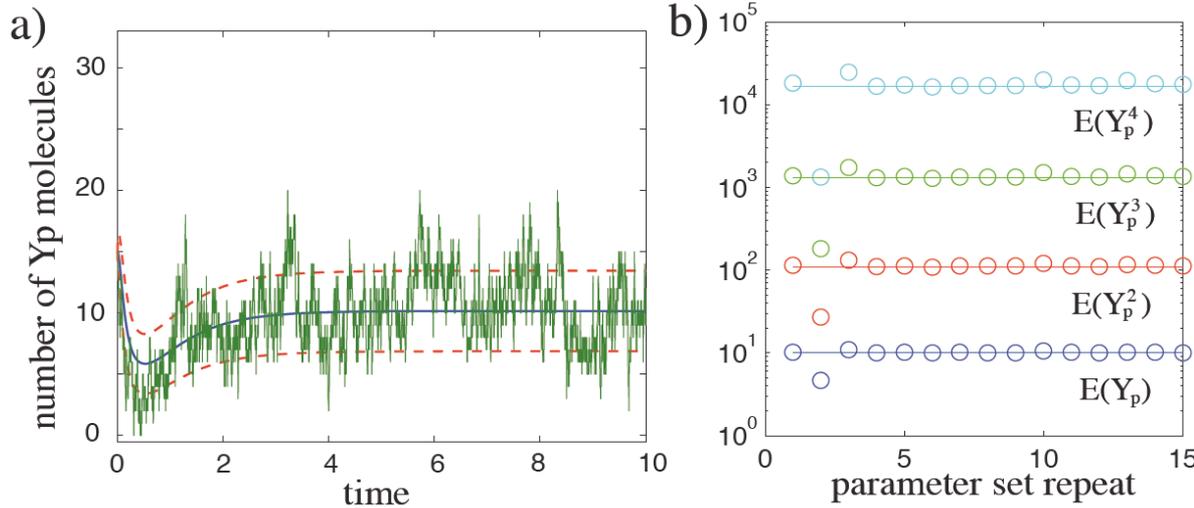
Stochastic extinction: General result



- In the particular example above, eventually the variables XD , X , XT , X_pY , XDY_p must vanish, since they are the only variables in their respective nodes. Therefore $\text{X}_p = \text{X}_{\text{tot}}$ in the long run, since all other forms of X vanish
- Since $\text{XD} = 0$ in the long run, the only reaction that can destroy Y_p can never take place. Also, since $\text{X}_p + \text{Y}$ is eventually off, it must eventually hold $\text{Y} = 0$. We conclude that $\text{Y}_p = \text{Y}_{\text{tot}}$ in this system after sufficiently long time.
- A similar result holds for conservative systems under the same assumptions as in Shinar et al.



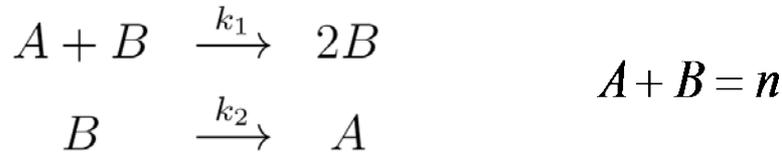
Transient Behavior: Computational Results



- Moment closure analysis to estimate mean, variance and higher moments of Yp in quasistationary distribution, using Joao Hepanha's StochDynTools software for Matlab
- Randomized each parameter value over six orders of magnitude, and compared $E(Y_p^k)$ with the moments of the Poisson distribution



Transient Behavior: Mathematical Analysis



Factorial moments:

$$y_1 = E(A), \quad y_2 = E(A(A-1)), \quad y_m = E(A(A-1)\dots(A-m+1)), \quad m \geq 0$$

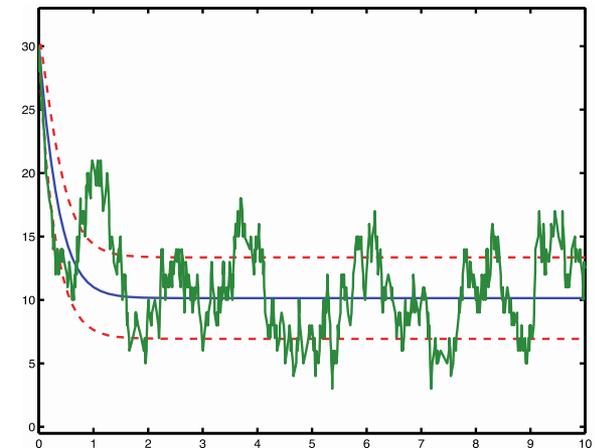
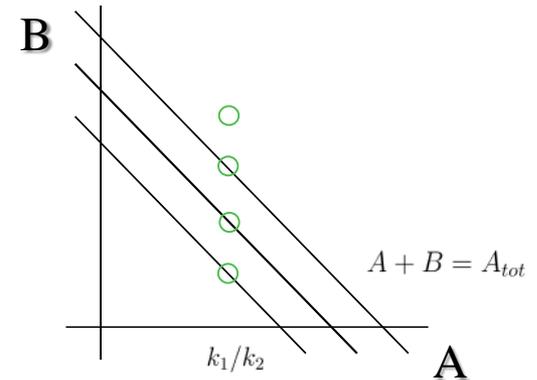
One can calculate for this system:

$$y'_m = mk_2(n-m+1)y_{m-1} - mk_2y_m - mk_1(n-m)y_m + mk_1y_{m+1}, \quad m \geq 1$$

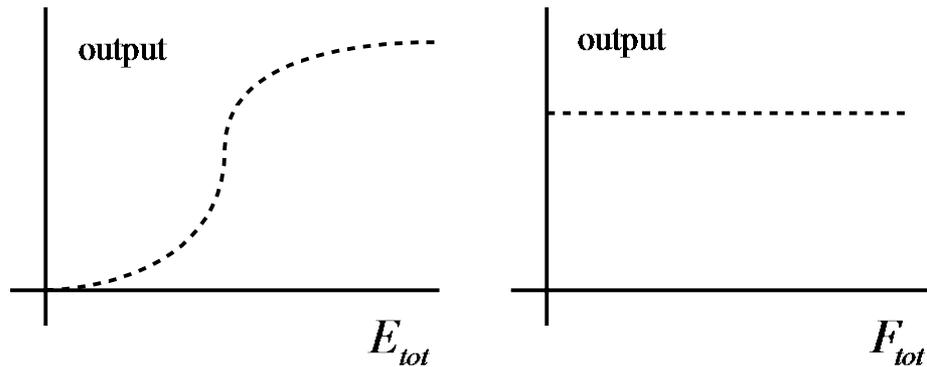
For large n , setting $y_m = 0$ and dividing by n :

$$mk_2y_{m-1} = mk_1y_m \qquad y_m = \frac{k_2}{k_1}y_{m-1} = \lambda y_{m-1}$$

These are the factorial moments of the Poisson distribution

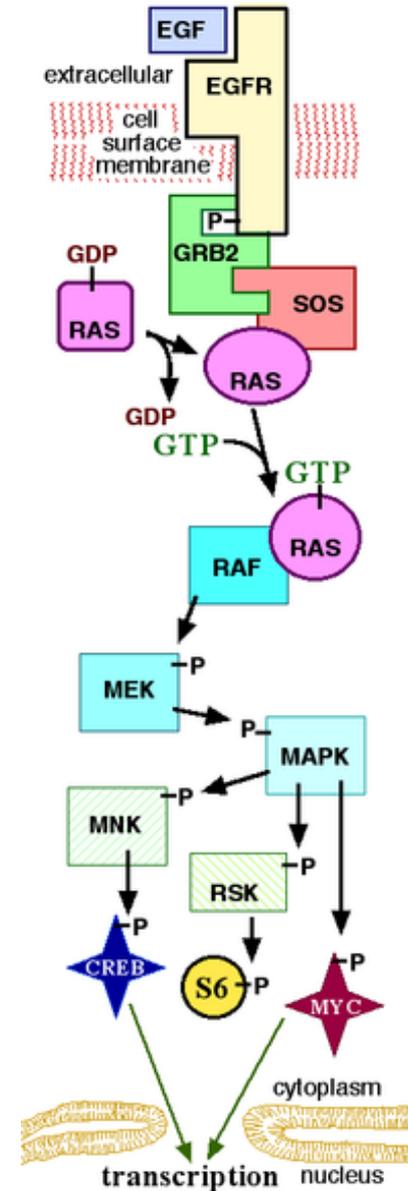


Absolutely robust dose responses



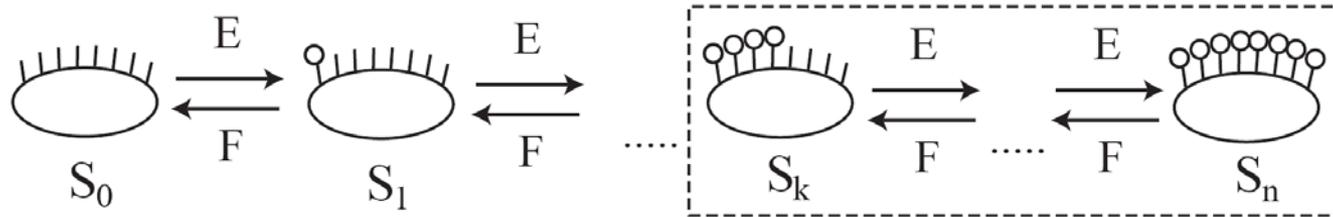
QUESTION: Can a system be absolutely robust to *some* total protein concentrations, but not *others*?

- This behavior can lead to robustness in signal transduction, where systems respond selectively to certain environmental signals while suppressing others
- Since such partial absolute robustness is a weaker statement than full absolute robustness, one can expect that weaker assumptions can lead to it





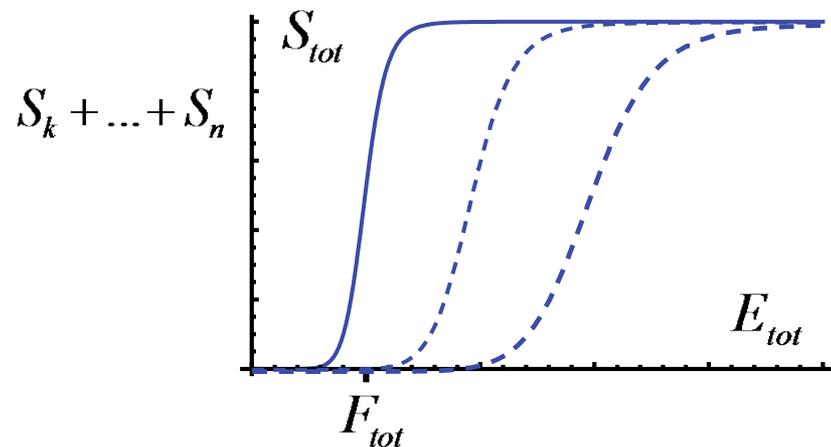
Example: Multisite phosphorylation



$$S_0 + S_1 + \dots + S_n = S_{tot}$$

$$E = E_{tot}$$

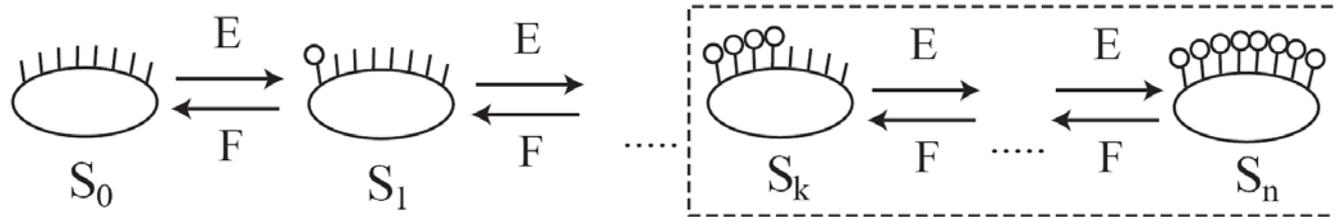
$$F = F_{tot}$$



- In this model the active substrate concentration is an ultrasensitive function of the kinase $E=E_{tot}$
- However, the system is highly dependent on small changes of the total phosphatase F_{tot} as well as S_{tot}



Example: Multisite phosphorylation

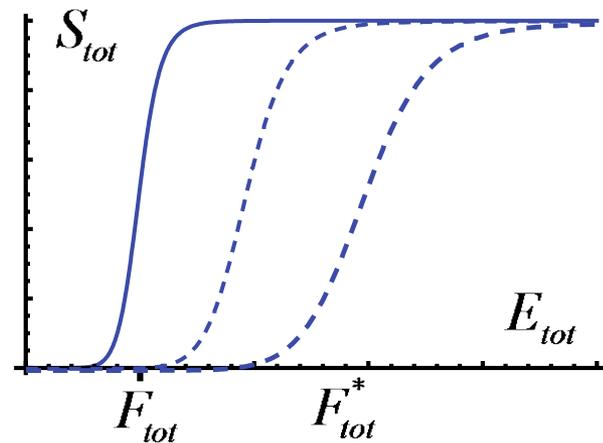


$$S_0 + S_1 + \dots + S_n = S_{tot}$$

$$E = E_{tot}$$

$$F = F_{tot}$$

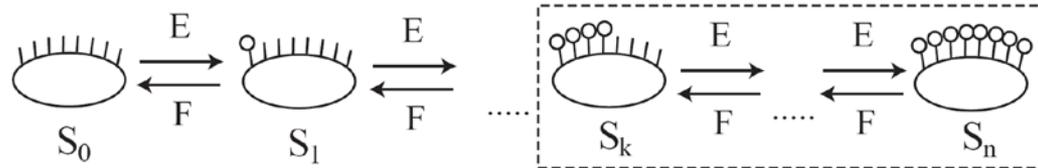
$$S_k + \dots + S_n$$



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Example: Multisite phosphorylation



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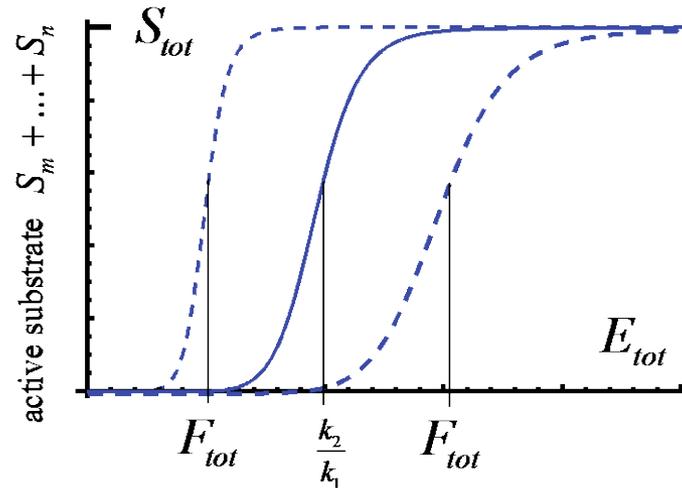
$$E = E_{tot}$$

$$F + B = F_{tot}$$

at steady state:

$$B' = k_1 F B - k_2 B = 0$$

$$\Rightarrow F = k_2 / k_1 \quad (B \neq 0)$$



- In this new system the dose response is not dependent on F_{tot} , but only on the rate parameters
- A similar change can be implemented to create (approximated) independence from S_{tot}



Conclusions

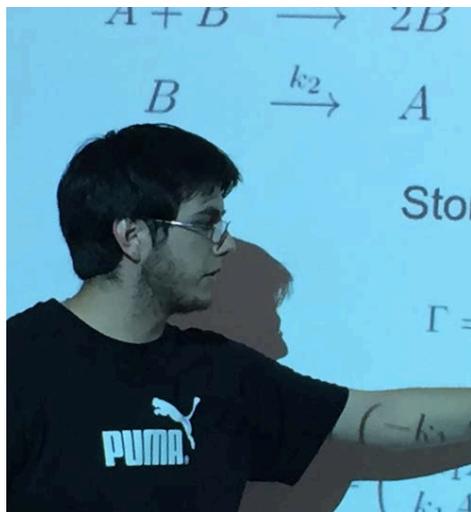
- Absolute robustness is a structural property that enables independence from protein variability for a steady state output
- After sufficient time, an extinction event leads to a significantly different dynamics from the deterministic case, and absolute robustness is lost
- Computational and analytical evidence shows that the absolutely robust variable has Poisson quasistationary distribution

Future work

- Investigate which chemical reactions can give rise to absolutely robust dose responses
- Prove that transient behavior of stochastic systems approximates a Poisson distribution
- Find experimental evidence for this mechanism in signal transduction in bifunctional proteins, such as in two component signaling, olfactory receptor selection, etc



Thanks!



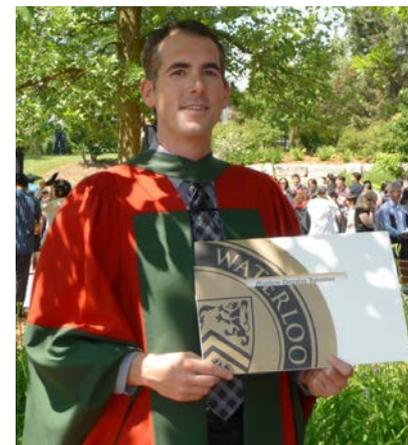
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Questions?

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