



Design and analysis of multiple feedback loops using natural and synthetic genetic constructs

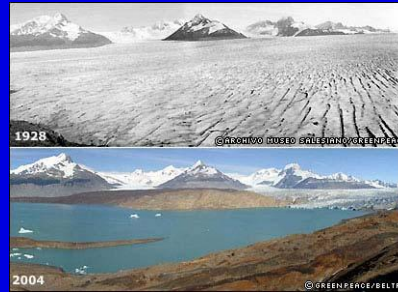
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Genotype to Phenotype: An Integrated Approach

Upsala Glacier, BBC website



Environment

Randomness

- Genome
- Transcriptome
- Proteome
- Metabalome



Phenotype

(physiological consequence)



Genotype to Phenotype: Modeling and Analysis

Modeling

Logical networks

ODE/PDE

Delayed ODE

Stochastic approach

Multiscale

Analysis

Data mining

Estimation theory

Nonlinear systems theory

Feedback control theory

Sensitivity

Key: Identify design principles

Simplicity in Biology



Alon, *Nature* 446:497 (2007)

- Diversity in genes NOT in motifs of regulatory networks
- These networks are robust (yet fragile)
- Combination of motifs yield new dynamical properties
- Network motifs conserved across organisms (animal, plant)

Analog Motifs in Natural Systems



— Freeman, *Nature*, **408**, 2000

- PI control (plasma calcium homeostasis in mammals)
- Negative feedback (autorepression)
- Positive feedback (growth in cell development)
- Negative/positive feedback (prolong weak signals)
- Feedforward (heat shock response)
- Cascades (insulin signaling pathways)



Engineered Versus Natural System

ENGINEERED SYSTEM

Design

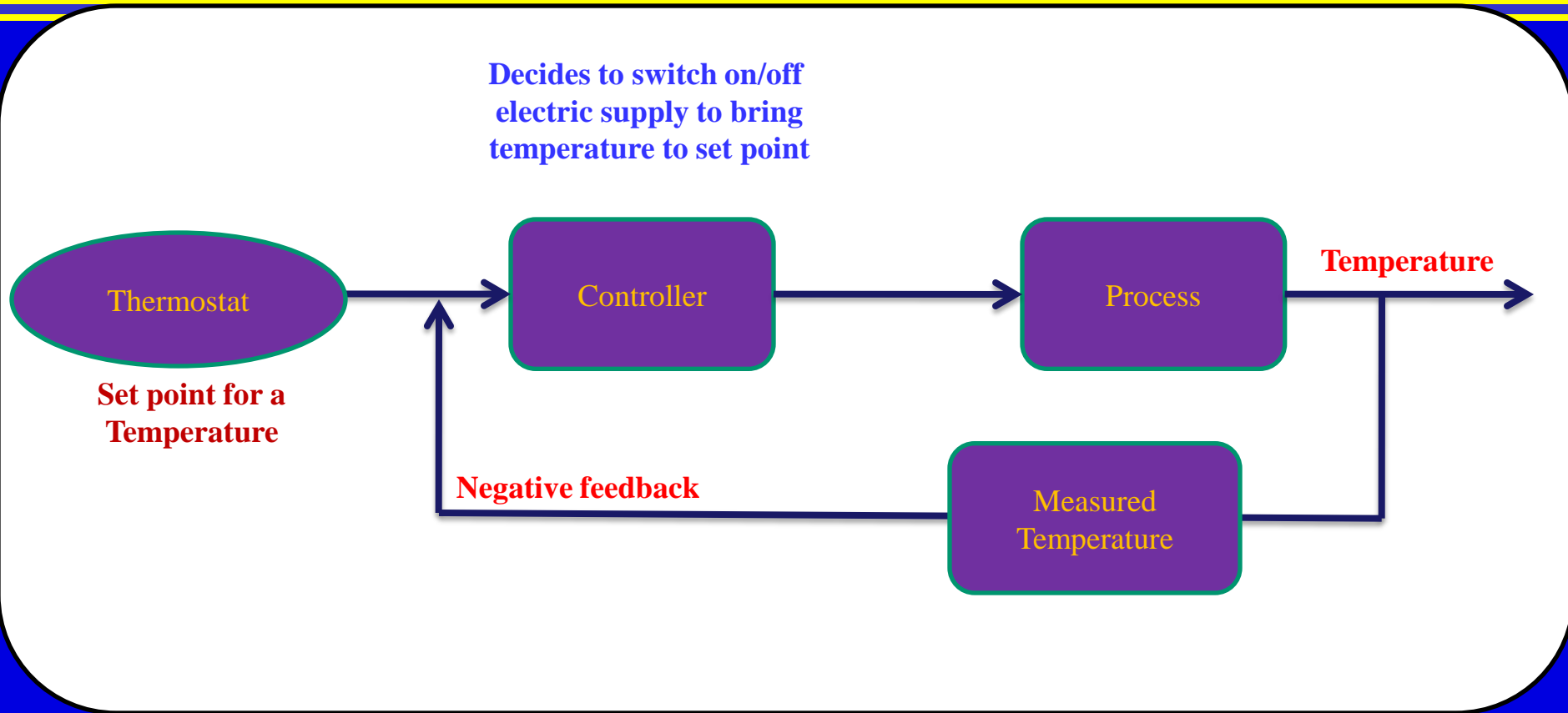
Operation

Optimization

Control

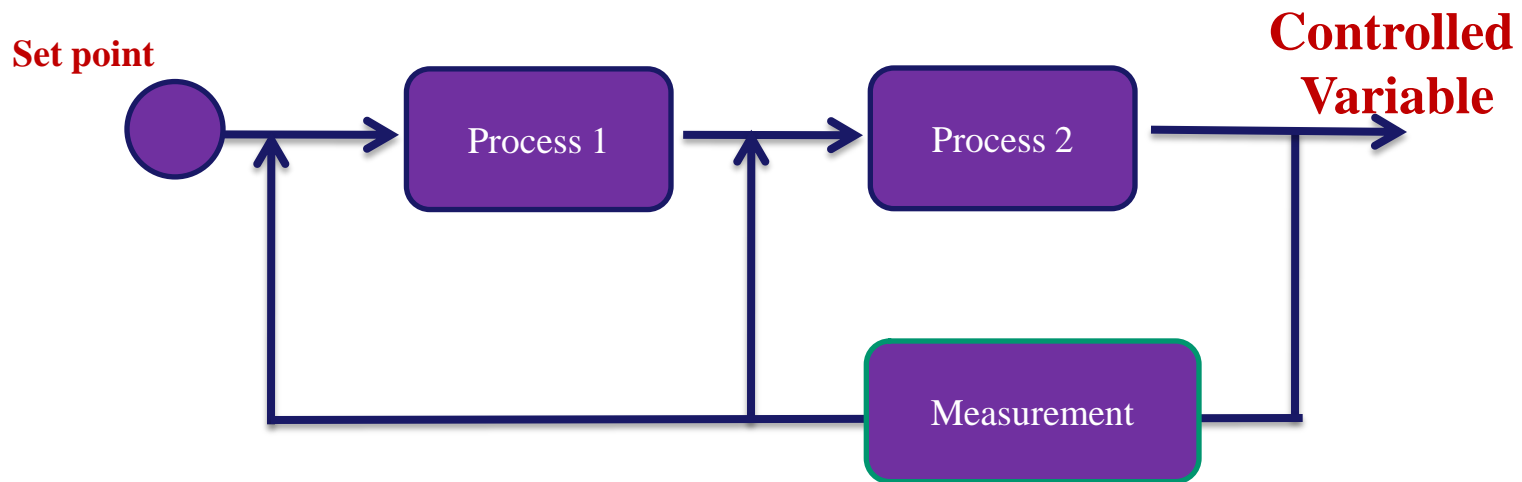
Engineered system: **bottom-up design** with known functionality of components
Natural system: **top down design** with unknown inherent property of various motifs

Engineered Systems : Room Heater



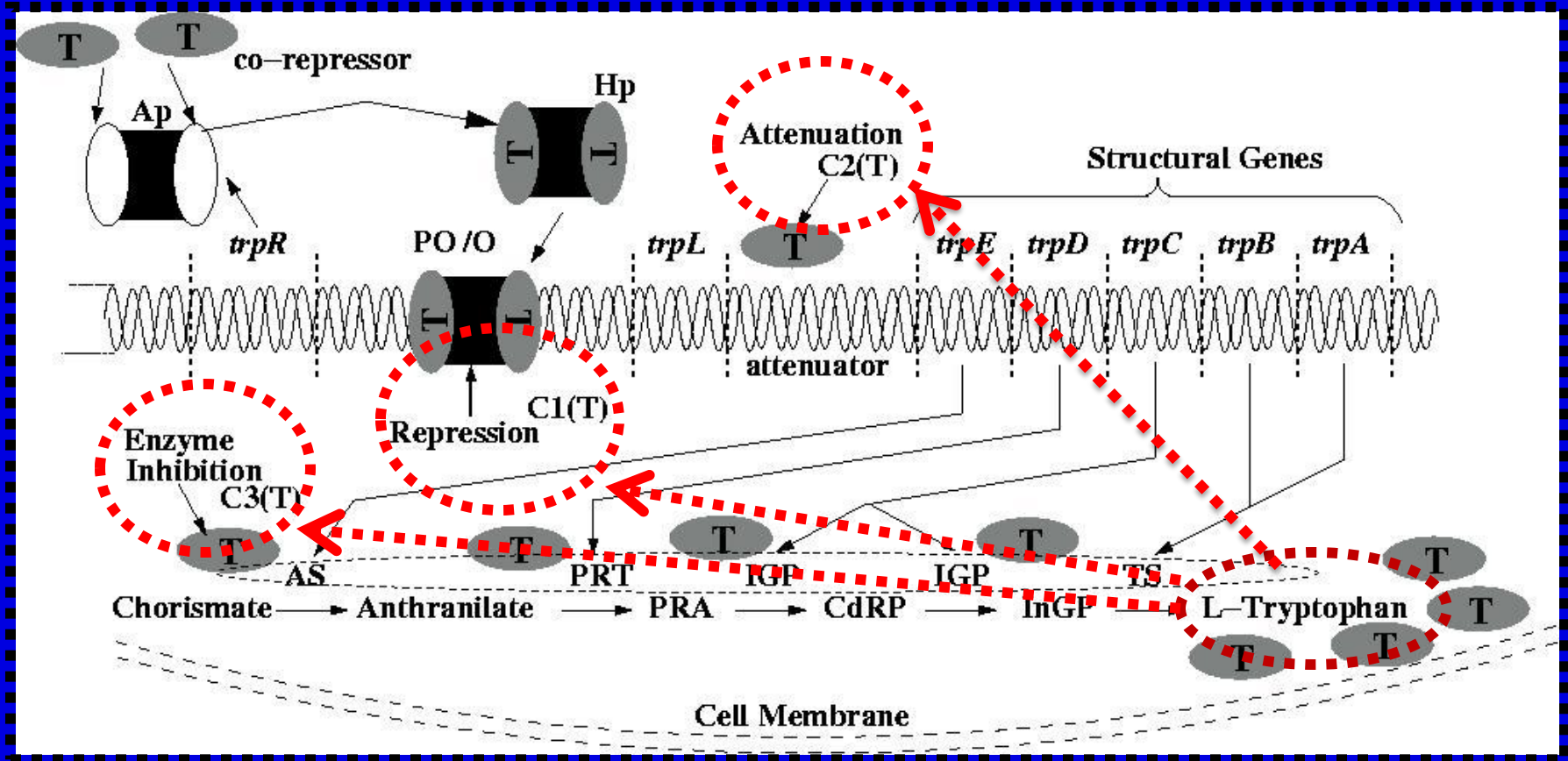
SINGLE INPUT SINGLE OUTPUT (SISO)

Multiple Input Multiple Output: a motif observed in Biological System

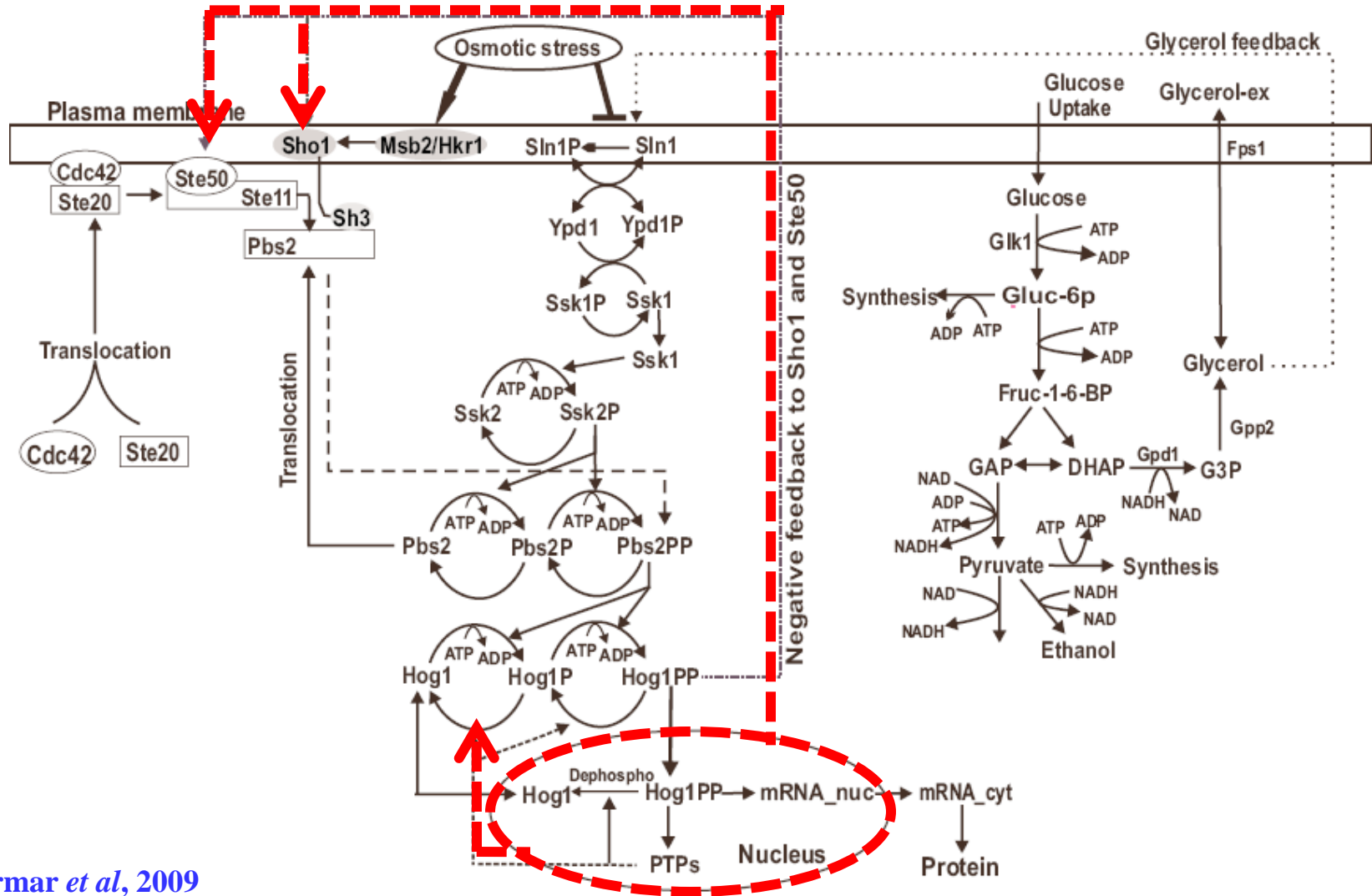


Single output is regulating the multiple upstream processes

Tryptophan in *E. coli* (bacteria)

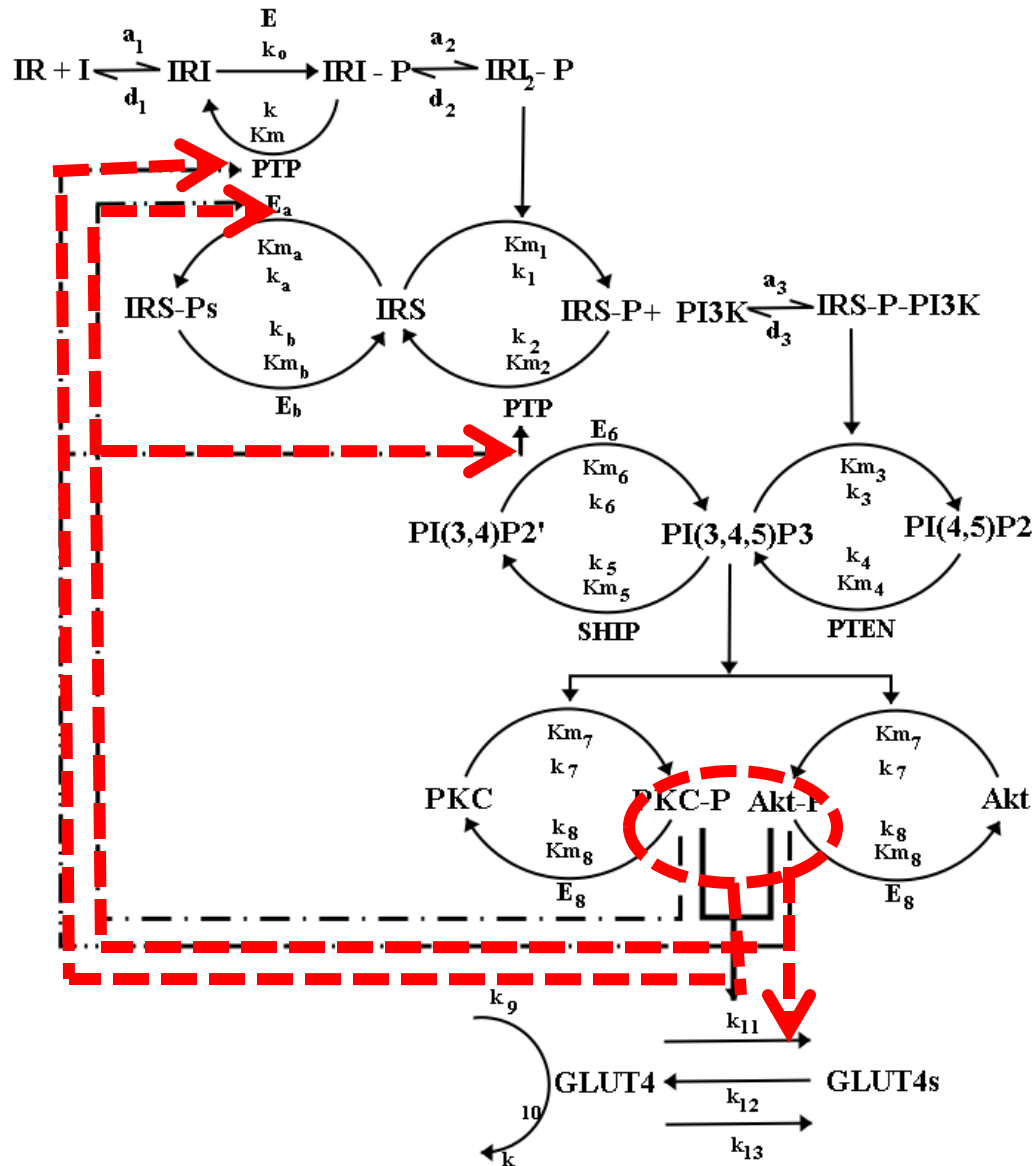


Osmotic Stress Pathway in Yeast



Ref. Parmar *et al*, 2009

Insulin Signaling Pathway in Mammals



Are multiple feedbacks in biological systems redundant or overkill?





Systems Analysis of the Tryptophan System in *Escherichia coli*

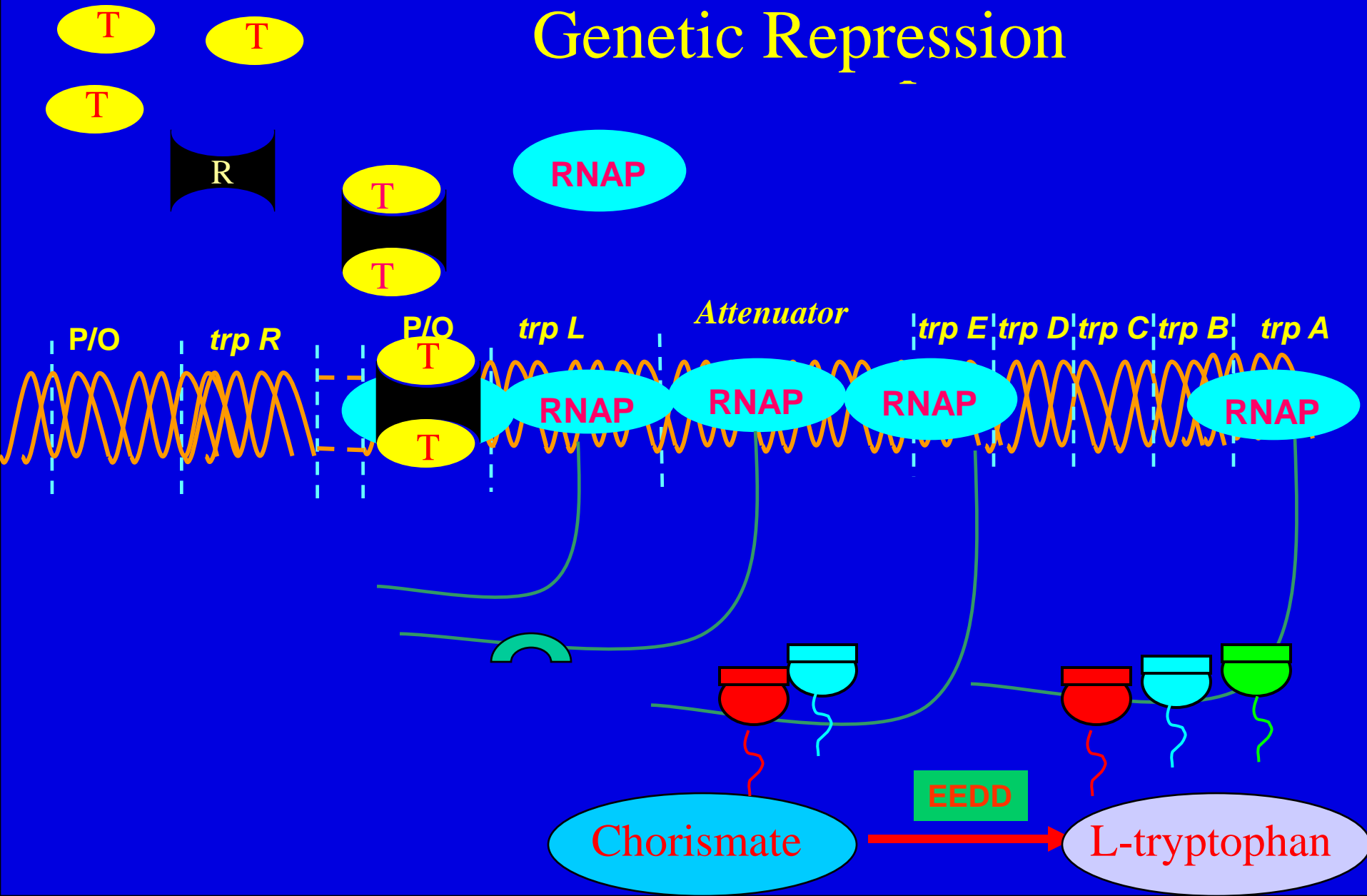
Modeling and Analysis of Tryptophan System in *Escherichia coli*



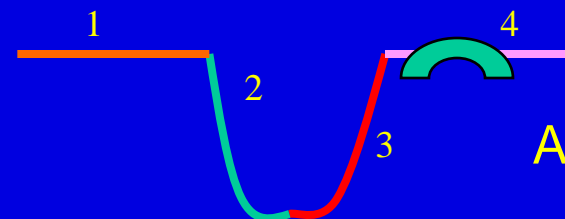
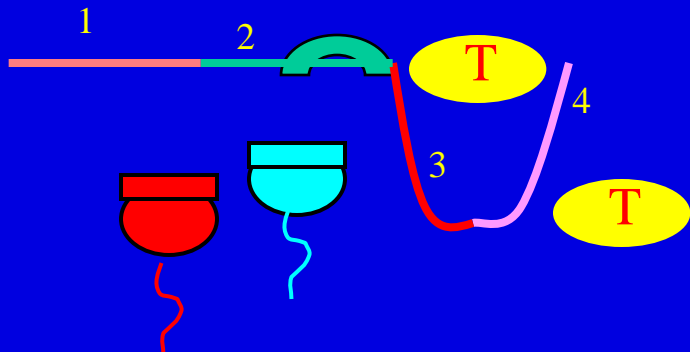
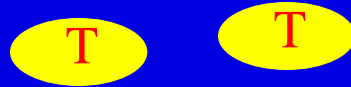
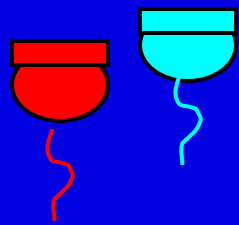
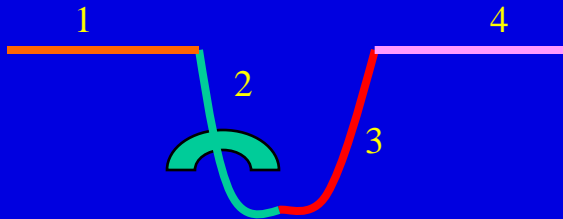
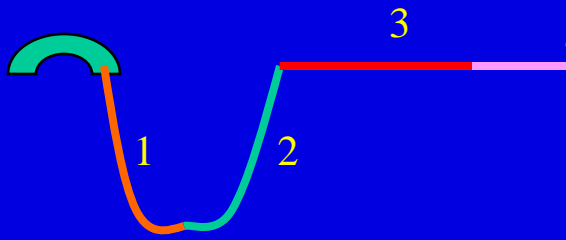
— Yanofsky and co-workers, 1972, 1984, 1987, 2000; Xie et al., 2003

- Goal: make tryptophan if none available in medium
- stop making tryptophan if available from medium
- Multiple feedback loop motif for autoregulation
- Widely occurring motif in biological systems (parallel cascade)
 - HOG pathway activation during osmotic shock (Hohmann, 2002)
 - Insulin signalling pathway (Sedaghat et al., 2002)
 - p53 regulation in cell cycle and apoptosis (Kohn, 1999)
 - circadian rhythms

F/B Mechanism I: Genetic Repression



F/B Mechanism II: Attenuation



Structural Enzymes

Anthranilate synthase

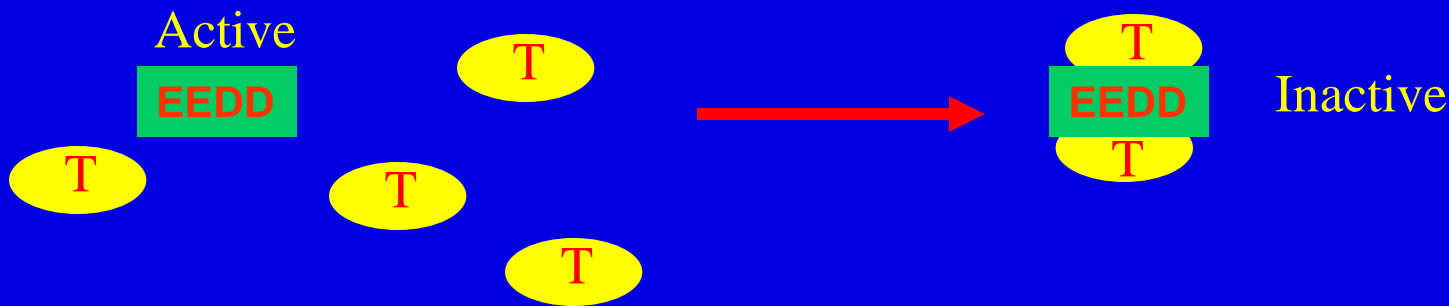
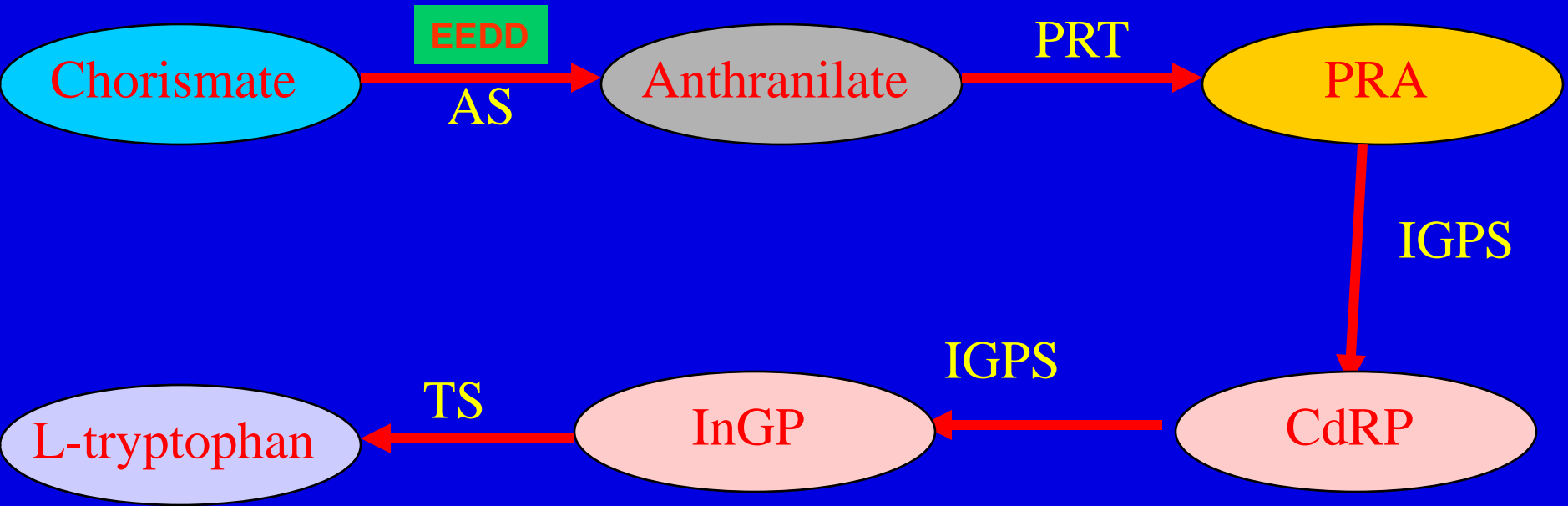
Phosphoribosyl
anthranilate
transferase

Indole glycerol
phosphate synthase

Tryptophan synthase



F/B Mechanism III: Enzyme Inhibition



Models



Bliss et al. (1982): repression and inhibition, time delays

Sinha (1988): detailed repression, tryptophan consumption constant

Sen and Liw (1990): non-constant tryptophan consumption

Santillan Mackey (2001): attenuation is modeled

Xiu et al. (2002): repressor autoregulation dynamics

Bhartiya et al. (2003): model simplifications, attenuation not modeled

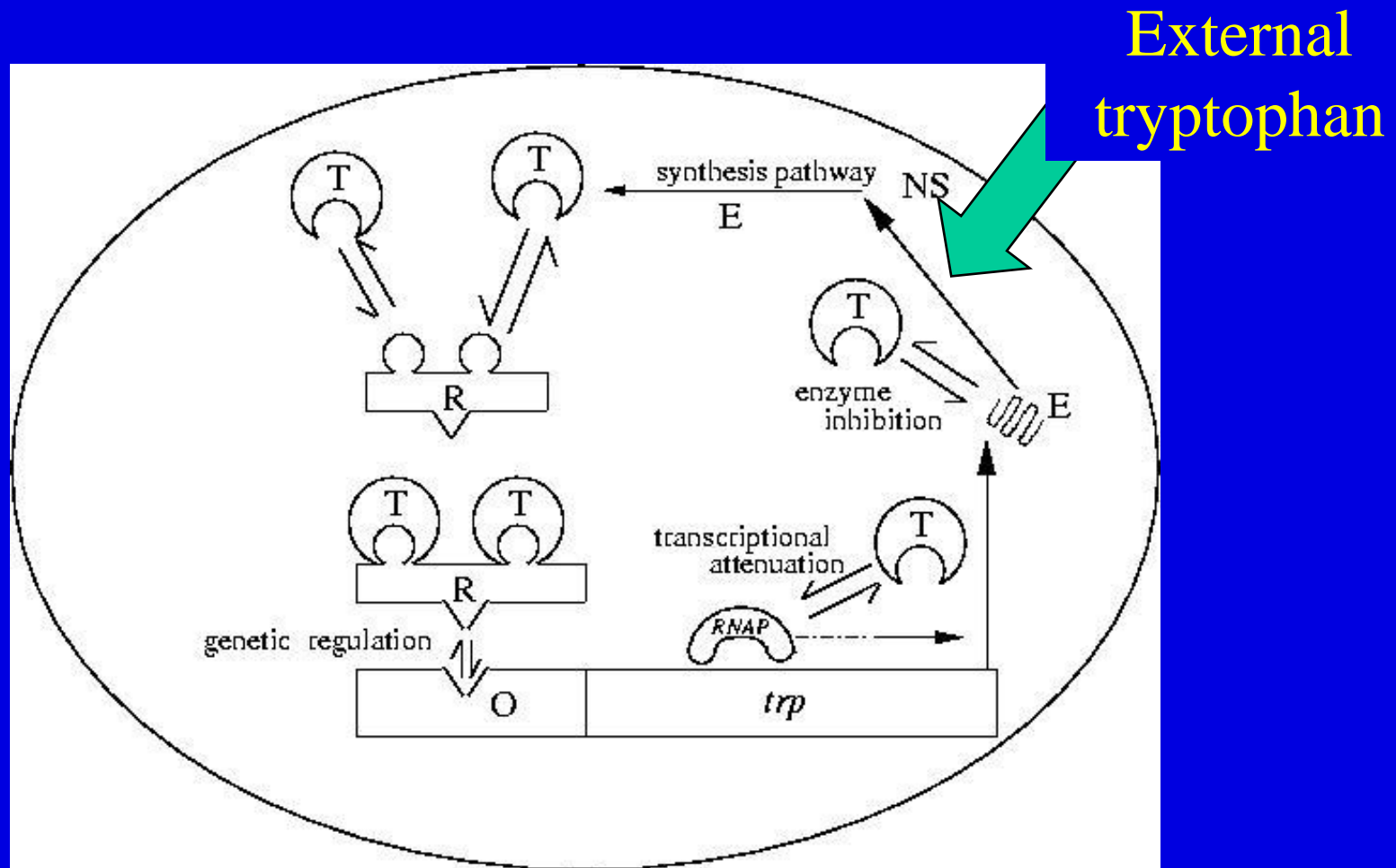
Ruhela et al. (2004): attenuation modeled

A Systems-Relevant Model for Tryptophan System



— Bhartiya, Rawool, Venkatesh, Eur. J. Biochem, **270**, 2003

— Ruhela, Bhartiya and Venkatesh, FEBS Letters, **563**, 2004



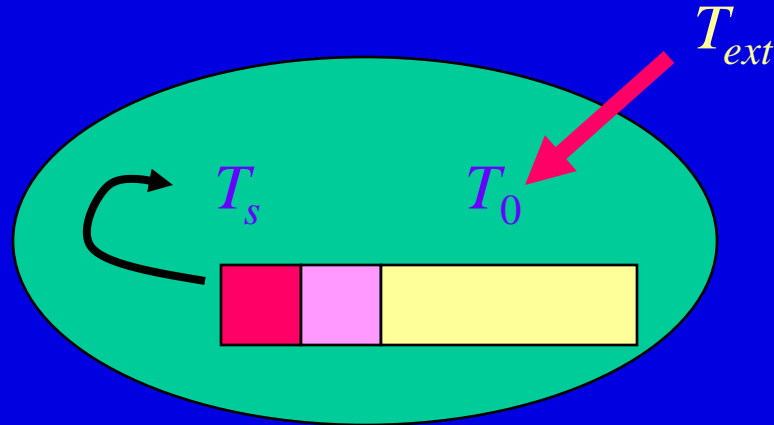
Tryptophan System Model

$$T_0 = T_{0,max} \frac{T_{ext}}{T_{ext} \left(1 + \frac{T_t}{f} \right) + e}$$

\nearrow
 100 μ M

Total tryptophan

$$T_t = T_s + T_0$$

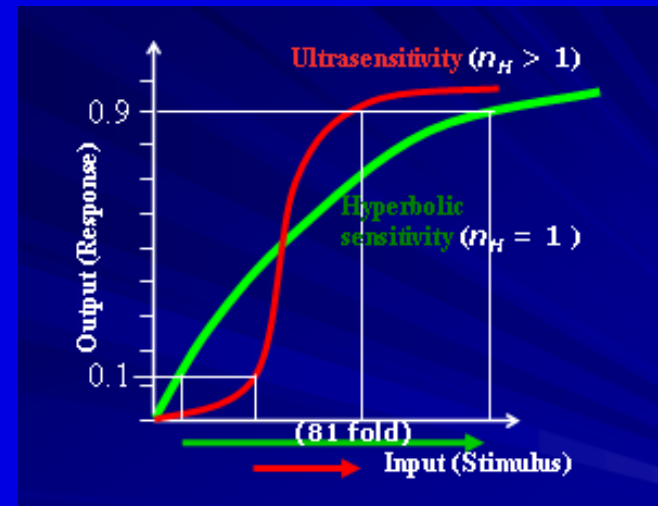


Operon Activation

$$\frac{d}{dt} O_F = k_1 O_t C_1(T_t) - k_{d,1} O_F - \mu O_F$$

Feedback Mechanism I: Genetic Repression

$$C_1(T_t) = \frac{K_1^{1.92}}{K_1^{1.92} + T_t^{1.92}}$$





Tryptophan System Model

Transcription

$$\frac{d}{dt} mRNA = k_2 O_R C_2(T_t) - k_{d,2} mRNA - \mu mRNA$$

Feedback Mechanism II: Attenuation

$$C_2(T_t) = \frac{K_1^{1.72}}{K_1^{1.72} + T_t^{1.72}}$$



Tryptophan System Model

Translation

$$\frac{d}{dt} E = k_3 mRNA - \mu E$$

Synthesis

$$\frac{d}{dt} T_s = k_4 C_3(T_t) E - g \frac{T_s}{k_g + T_s} - \mu T_s$$

Feedback Mechanism
III: Enzyme Inhibition

$$C_3(T_t) = \frac{K_1^{1.2}}{K_1^{1.2} + T_t^{1.2}}$$

- Enables delineation of process and regulator
- Does not use delay differential equations



Tryptophan System in *Escherichia coli*: Regulator and Process

— Venkatesh, Bhartiya and Ruhela,

FEBS Letters, 563, 2004

Activation/Transcription

Translation

Tryptophan
synthesis

$$\frac{d}{dt}(O_R) = k_1 O_t C_1(T_t) - k_{d1} O_R - \mu O_R$$

O_R

$$\frac{d}{dt}(mRNA) = k_2 O_R C_2(T_t) - k_{d2} mRNA - \mu mRNA$$

$mRNA$

$$\frac{d}{dt}E = k_3 mRNA - \mu E$$

E

$$\frac{d}{dt}T_s = k_4 C_3(T_t) E - g \frac{T_s}{T_s + K_g} - \mu T_s$$

T_s

$$C_1(T_t) = \frac{K_{i,1}^{1.92}}{K_{i,1}^{1.92} + T_t^{1.92}}$$

Repression

$$C_2(T_t) = \frac{K_{i,2}^{1.72}}{K_{i,2}^{1.72} + T_t^{1.72}}$$

Attenuation

$$C_3(T_t) = \frac{K_{i,3}^{1.2}}{K_{i,3}^{1.2} + T_t^{1.2}}$$

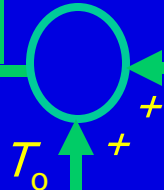
Inhibition

T_{ext}

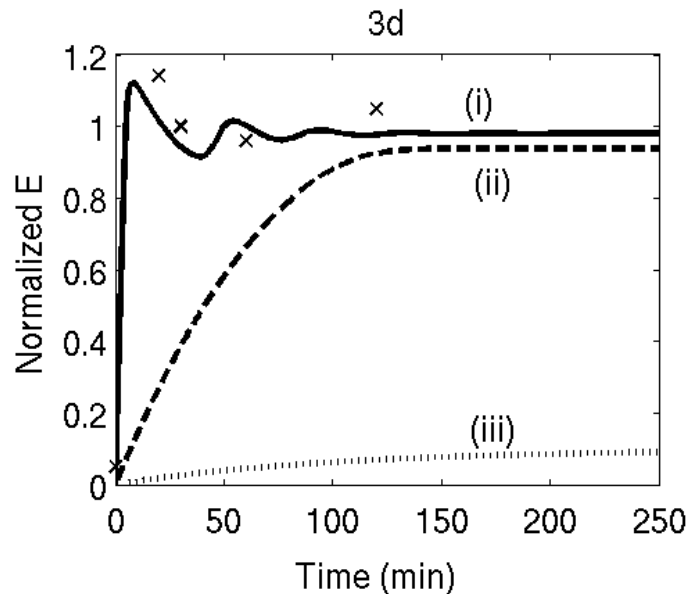
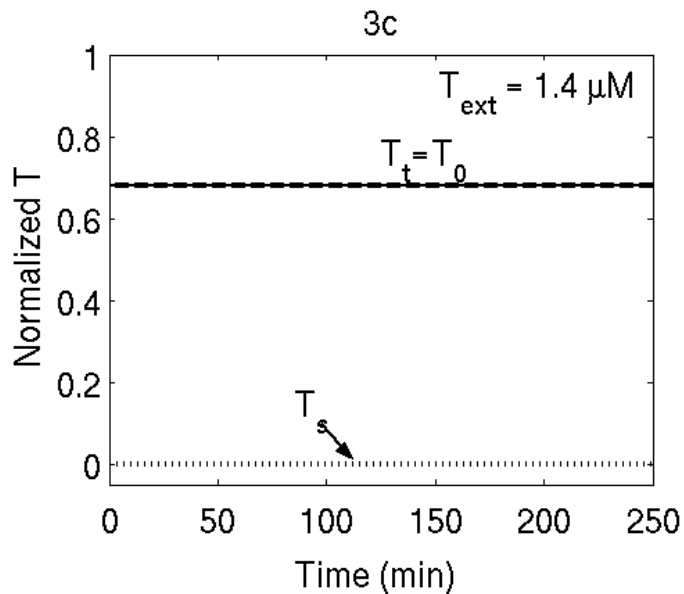
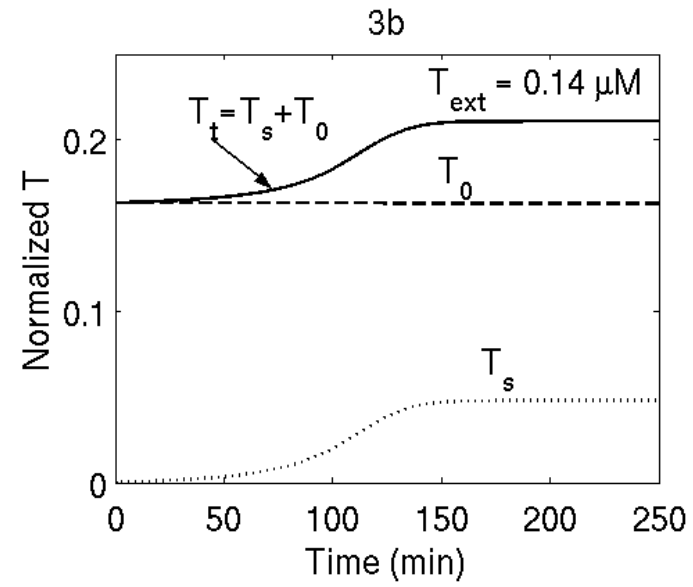
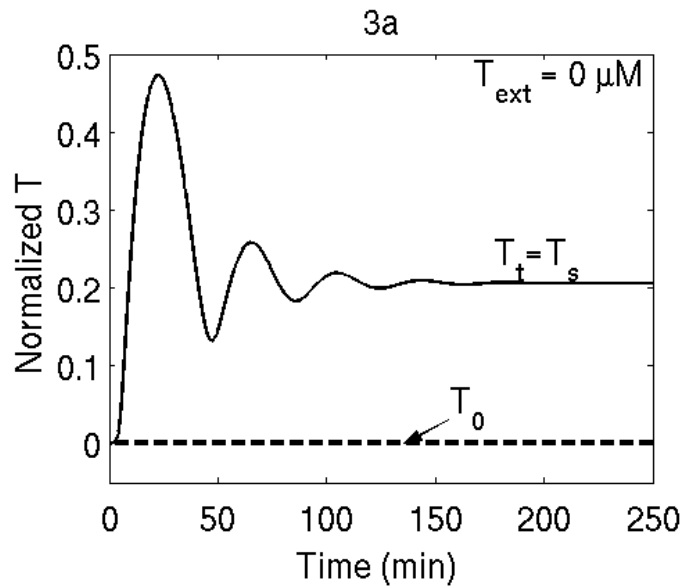
T_t

T_o

$$T_o = f(T_{ext}, T_t)$$



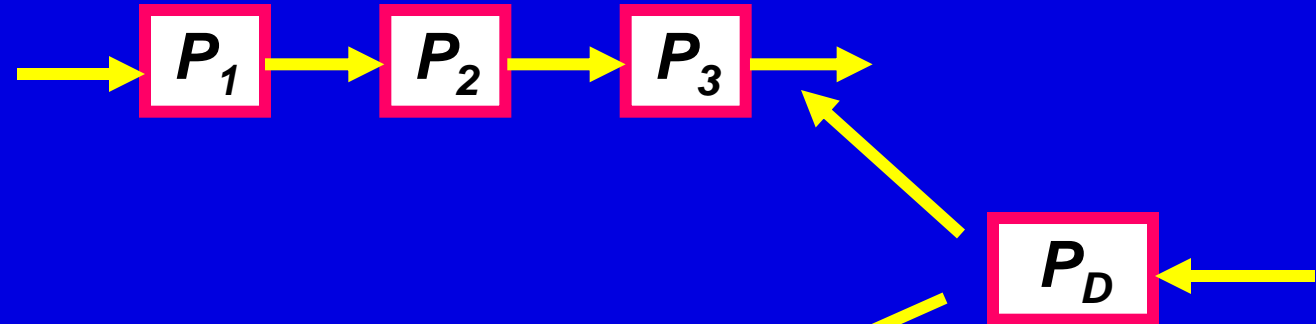
Model Simulation and Validation



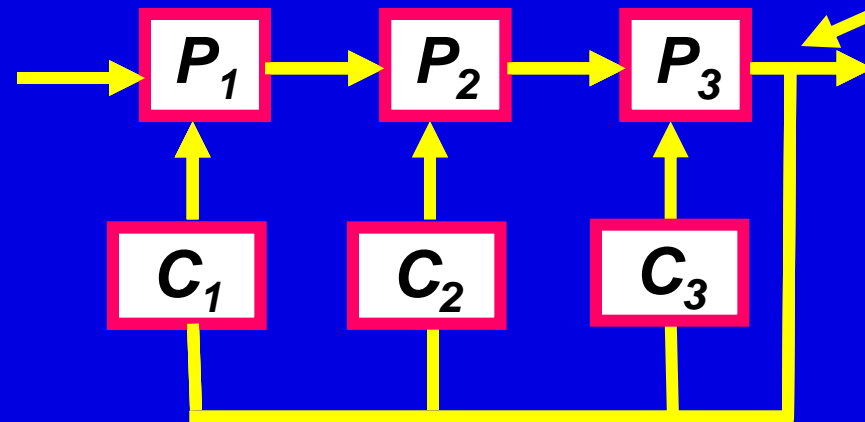


Network Structure: Multiple feedback loops for regulation of Processes-in-series

Open Loop



Closed Loop



$C_1C_2C_3$ active : Triple feedback loop



Results

Nominal performance

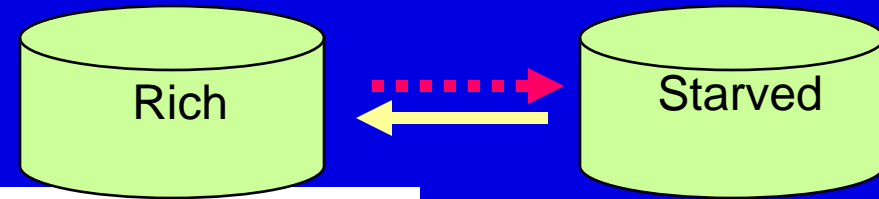
Robust performance

Frequency Response Analysis

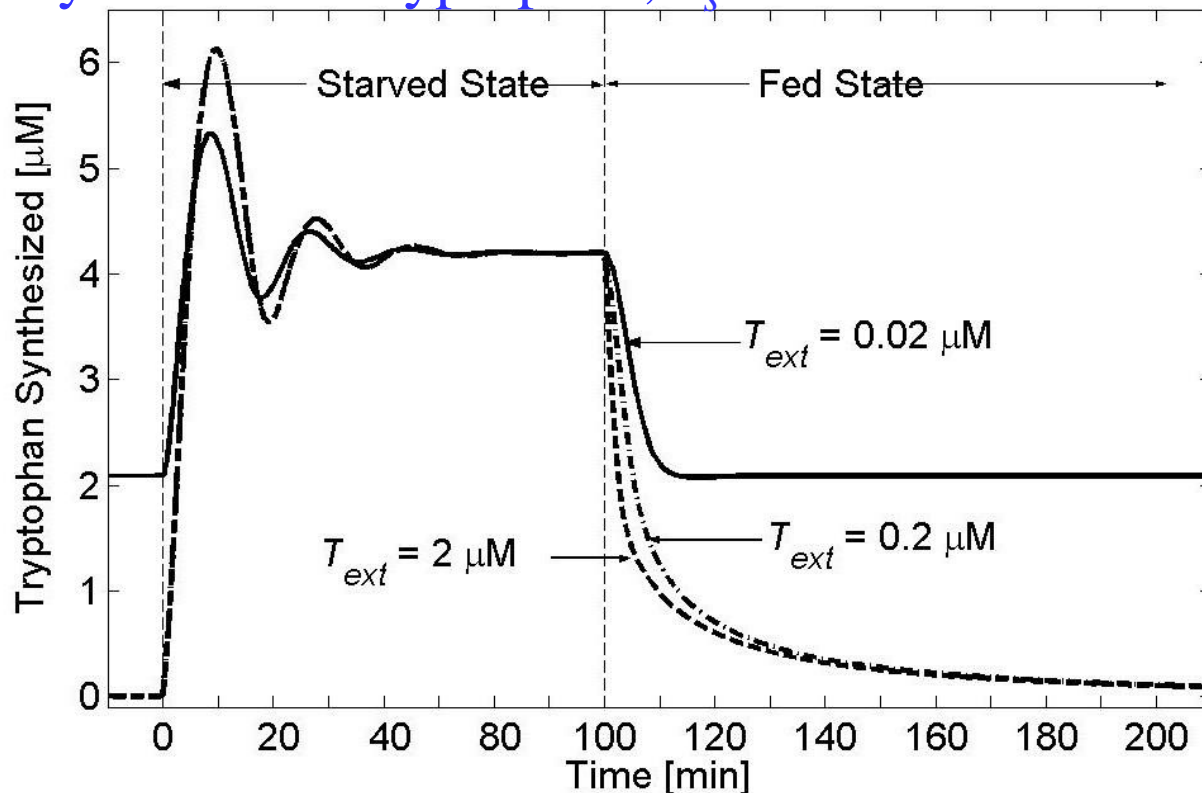
System Response Under Different Nutritional Levels

-Chaudhary, Bhartiya and Venkatesh, IET Systems Biology, 1, 2007

Studies with Nonlinear Model



Synthesized Tryptophan, T_s



- Homeostasis

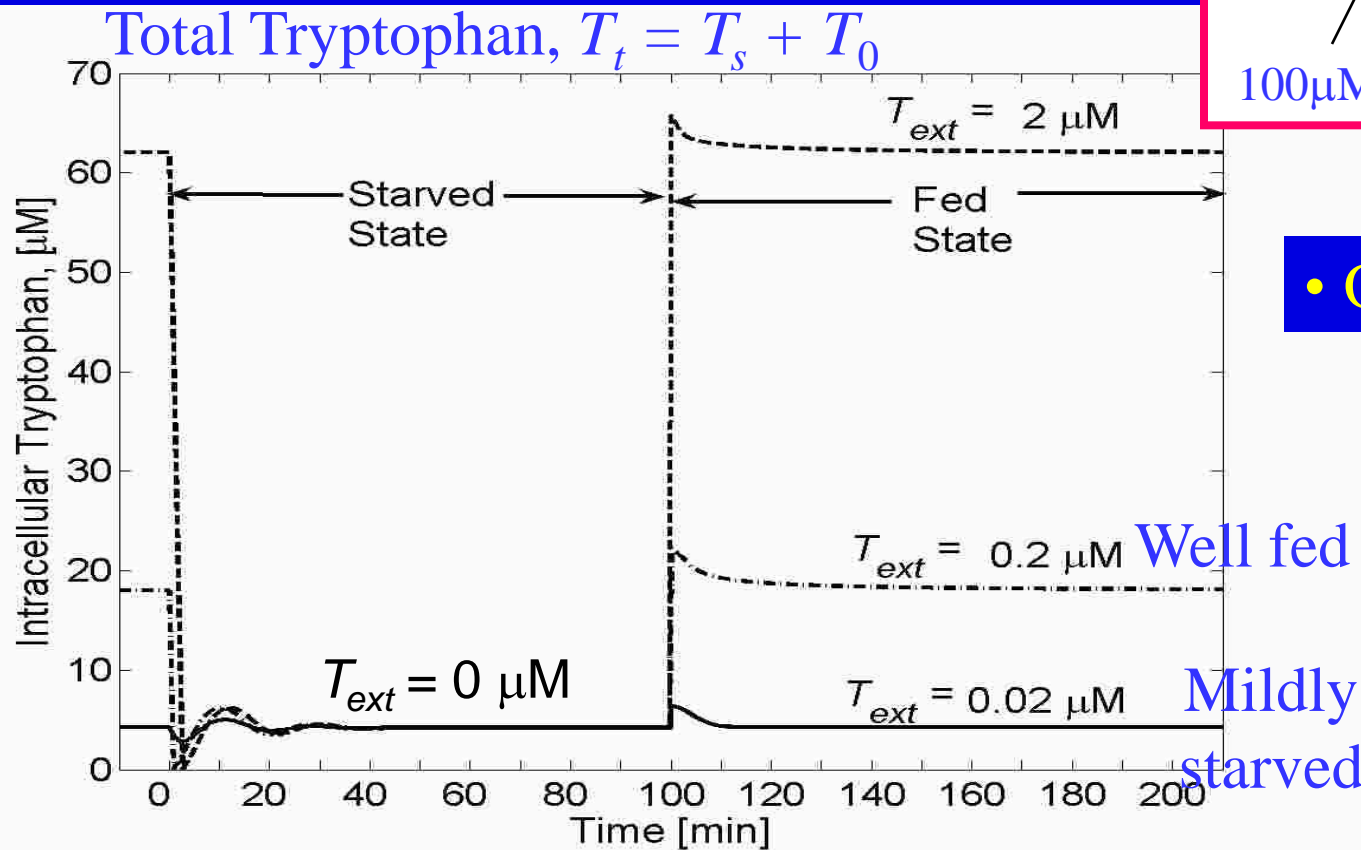
System Response Under Different Nutritional Levels

-Chaudhary, Bhartiya and Venkatesh, IET Systems Biology, 1, 2007

Studies with Nonlinear Model

$$T_0 = T_{0,max} \frac{T_{ext}}{T_{ext} \left(1 + \frac{T_t}{f} \right) + e}$$

100μM

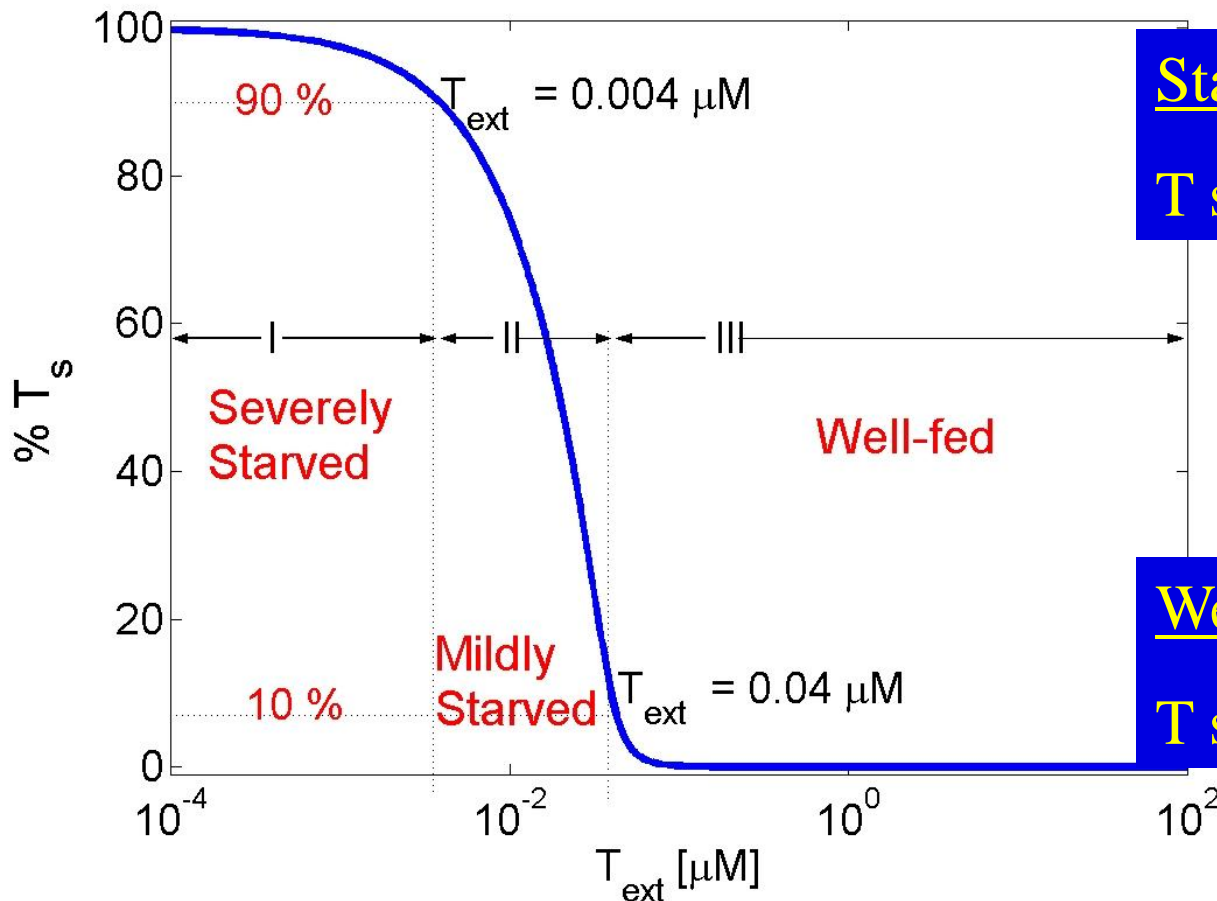


- Output disturbance



Characterization of Nutritional Status

Studies with Nonlinear Model



Starvation: $T_{ext} < 0.004 \mu\text{M}$

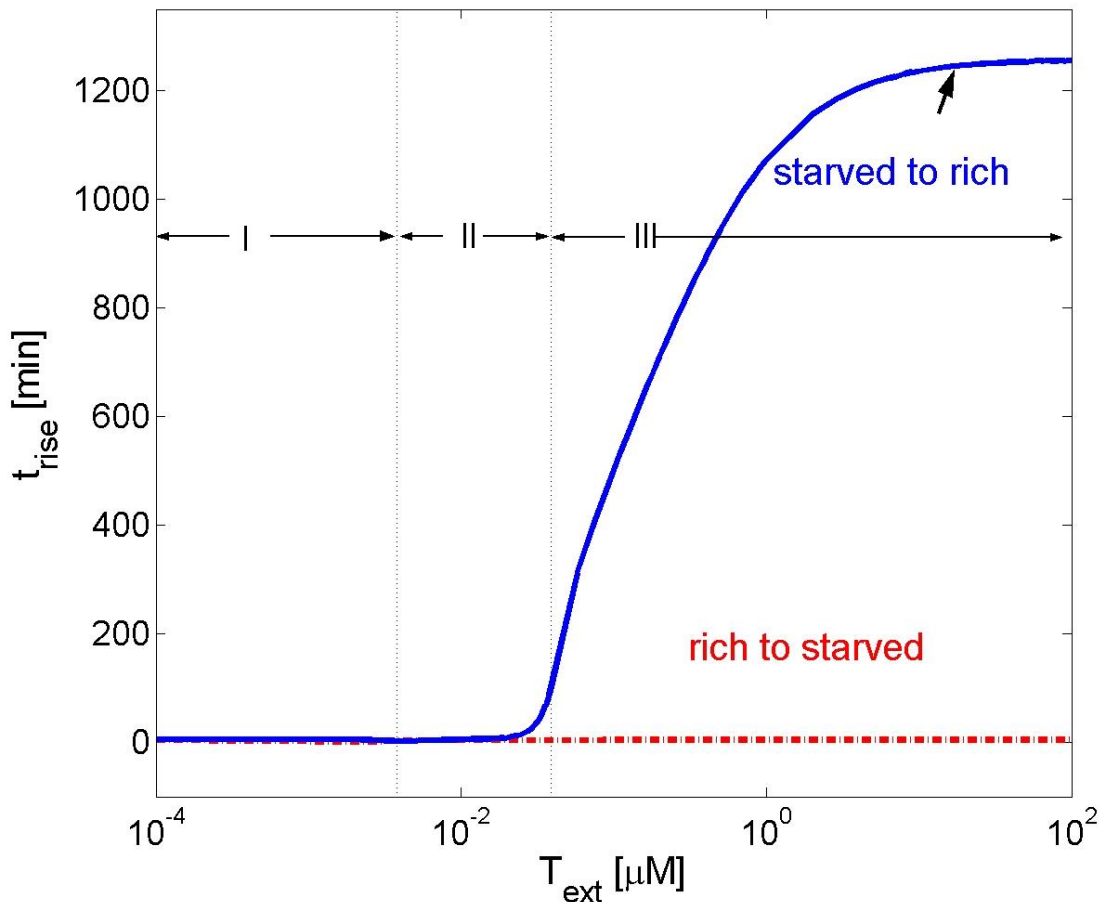
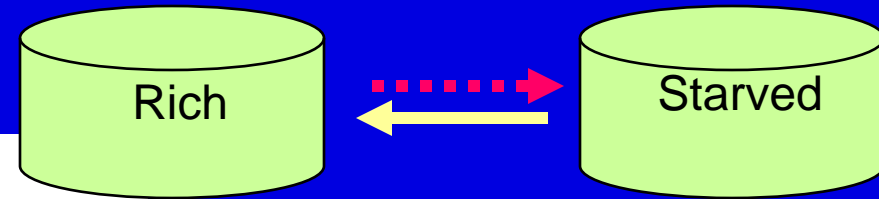
T synthesized $> 90\%$ T total

Well-fed: $T_{ext} > 0.04 \mu\text{M}$

T synthesized $< 10\%$ T total

Network Goal: Robust rise time necessary for survival during starvation

Studies with Nonlinear Model



Nominal Performance



- Rapid tryptophan synthesis in severely to mildly starved conditions
- Under starvation, rise time of 5 minutes regardless of initial state
- Under well-fed conditions, sluggish shut-off of synthesis
- Identified three regions of nutrition



Results

Nominal performance

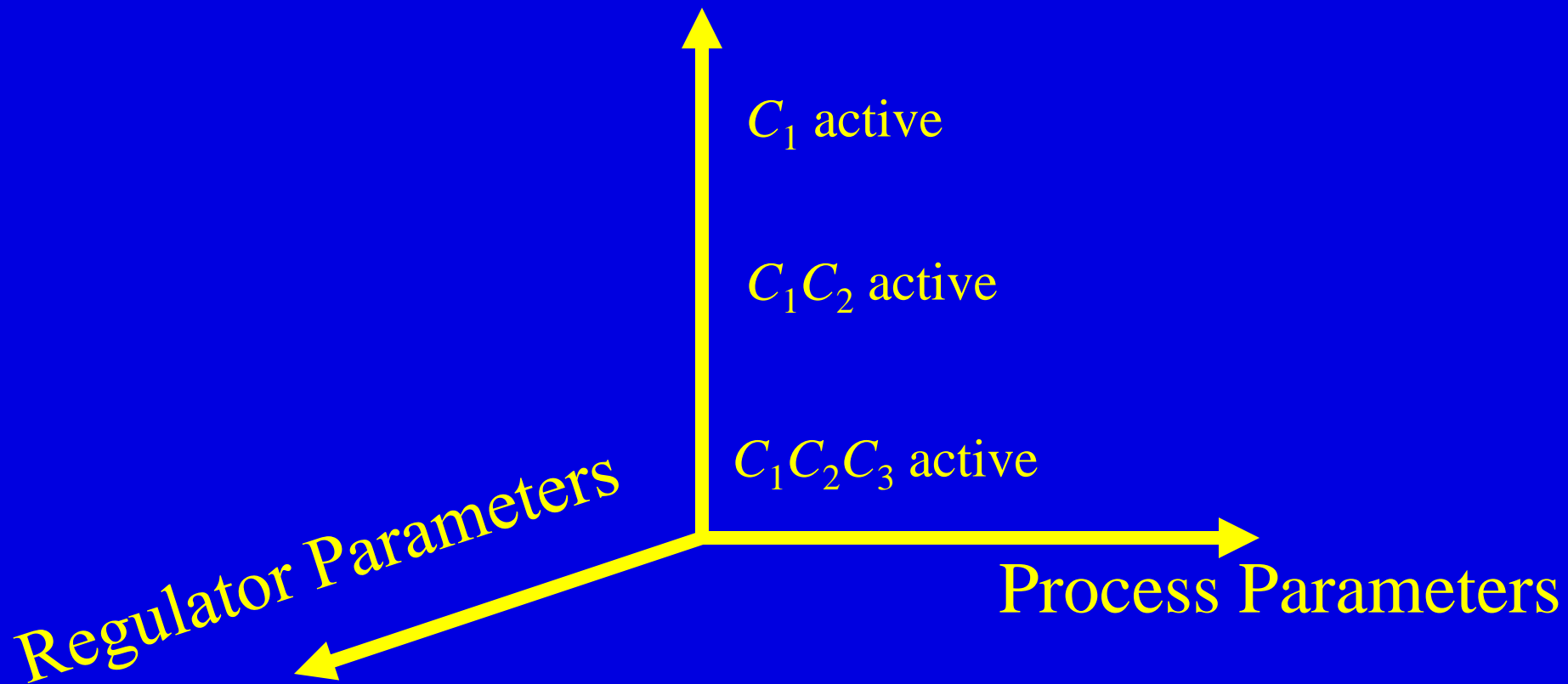
Robust performance

Frequency Response Analysis



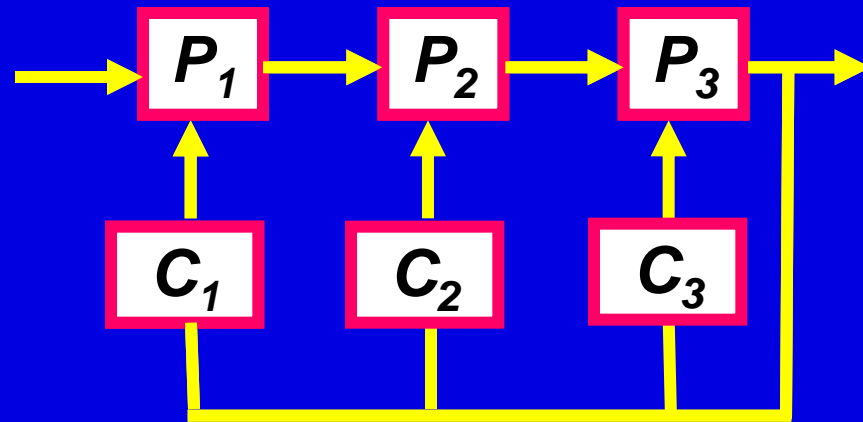
Perturbations

Regulatory Strategies





$C_1C_2C_3$ active v/s C_1 active Design

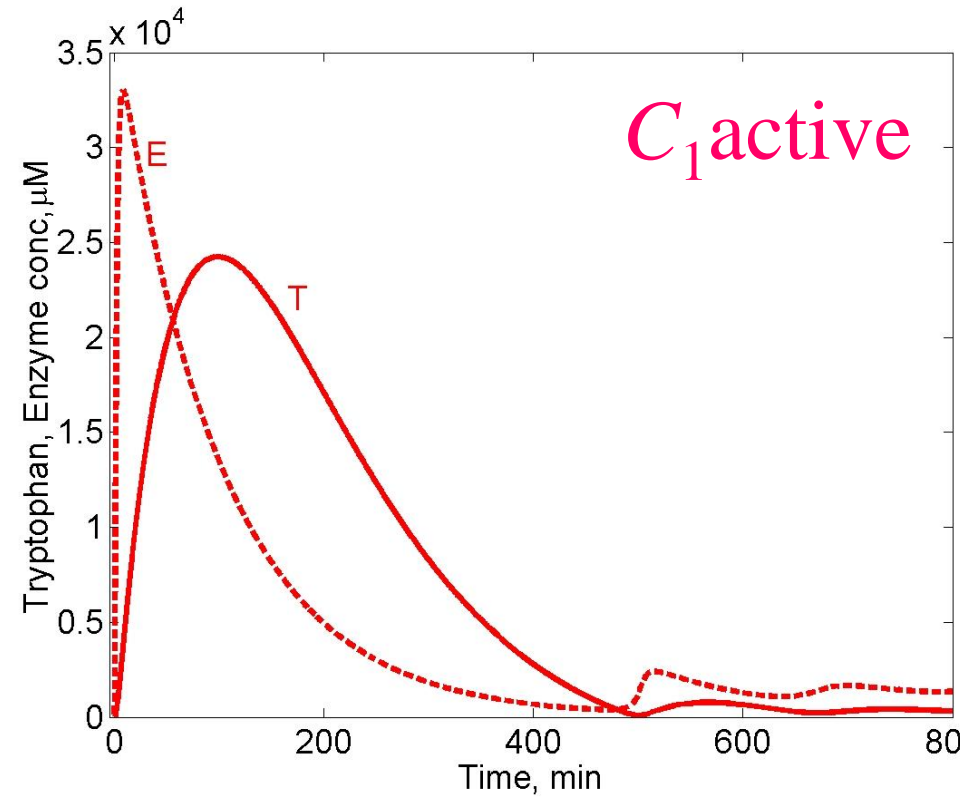
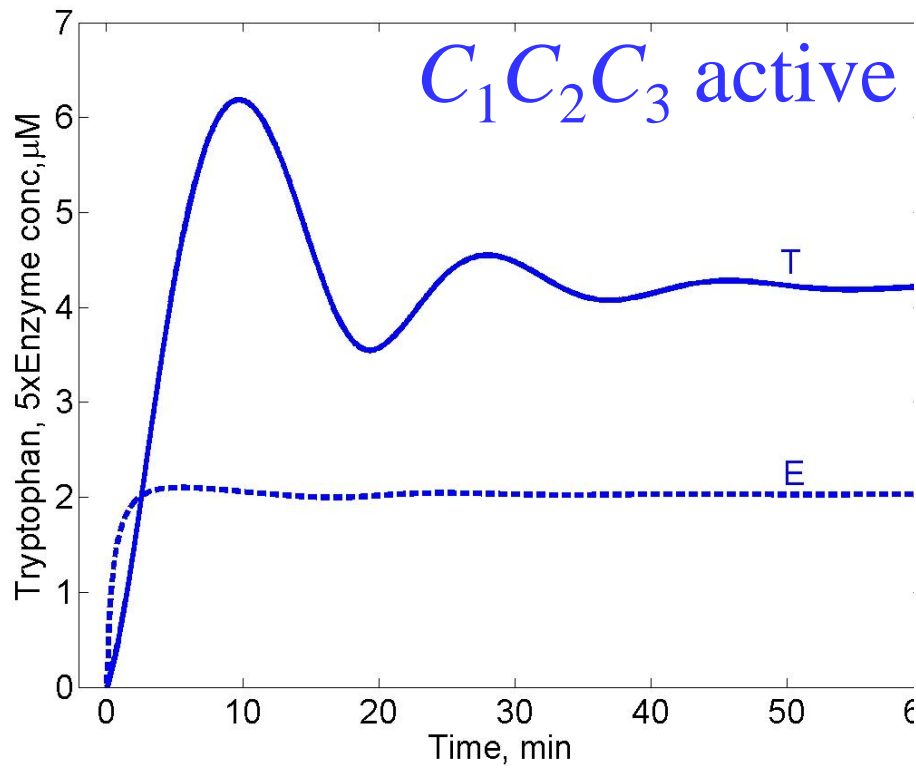


$C_1C_2C_3$ active : Triple feedback loop

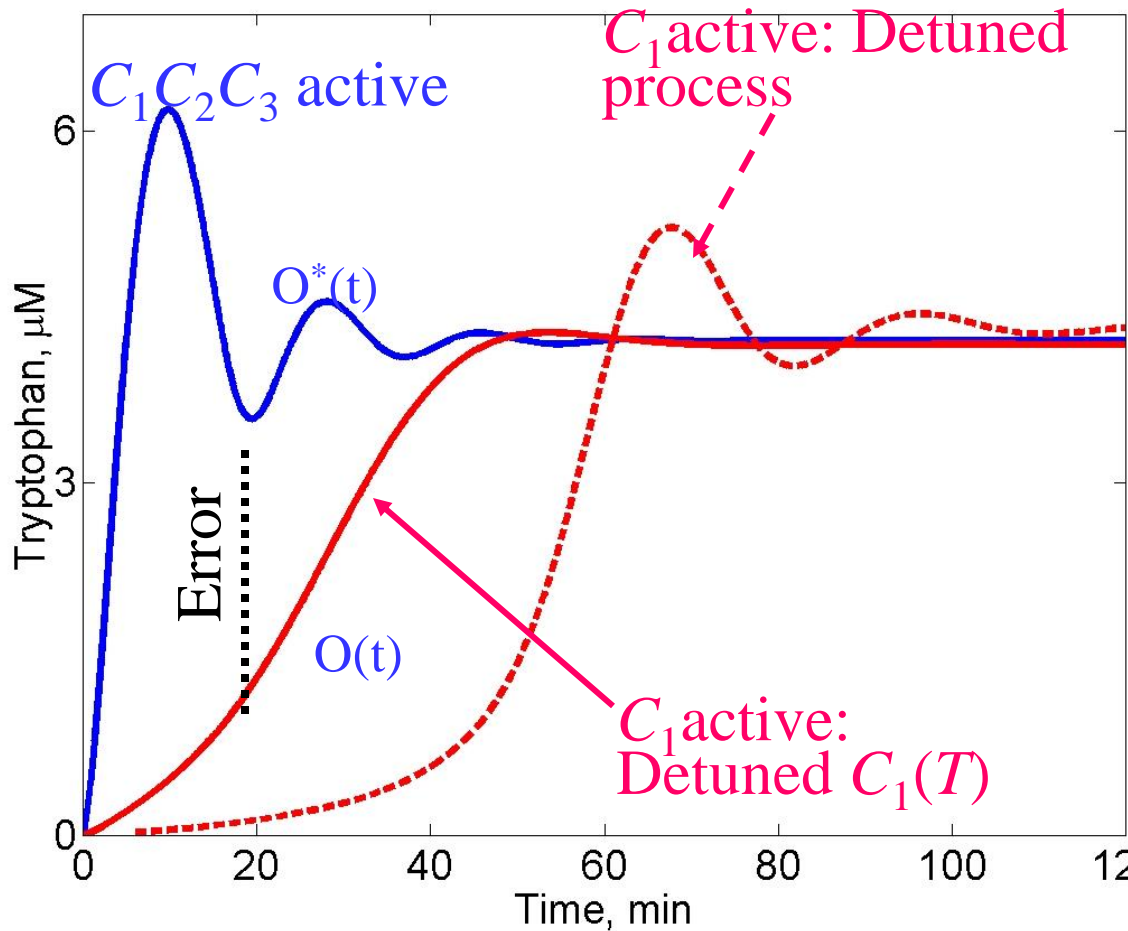
Are Multiple feedbacks loops a regulatory overkill?
(Freeman, Nature, 2003).

Starvation:

$C_1C_2C_3$ active v/s C_1 active design



Starvation: Improve C_1 active mutant performance by retuning



- Retuning of single loop not sufficient to yield performance as in multiple loop design
- Multiple feedback architecture is key to meet physiological needs
- Rise time = 5 min
- Settling time = 30 min

Robust Performance Metrics

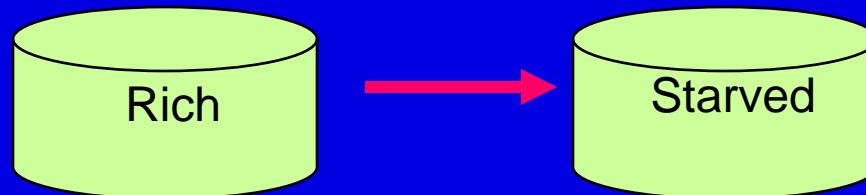


Network Goals

- Rise time (time needed to first attain 5% of final value)
- Root mean square error (error relative to nominal performance)

$$I(p, s) = \frac{1}{\sqrt{t_f}} \sqrt{\int_0^{t_f} [O(t, p, s) - O^*(t, p^*, s^*)]^2 dt}$$

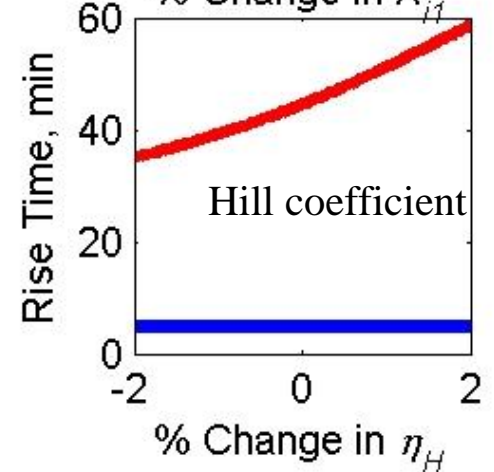
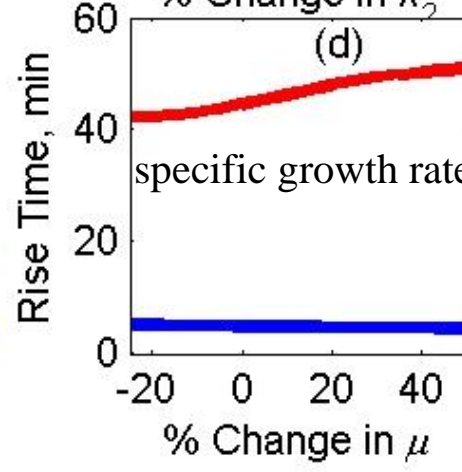
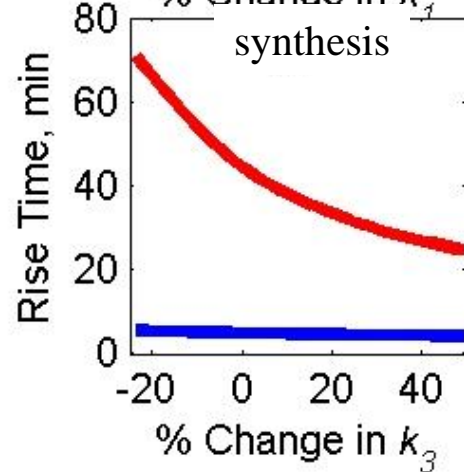
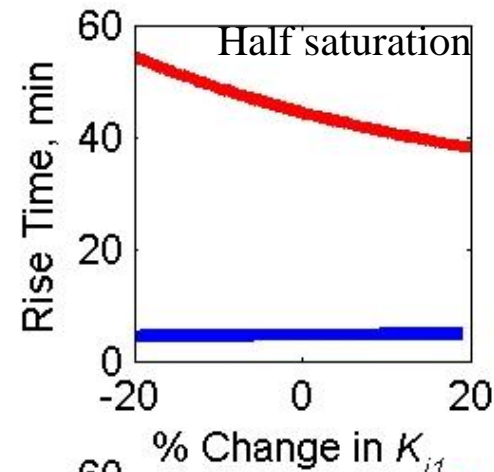
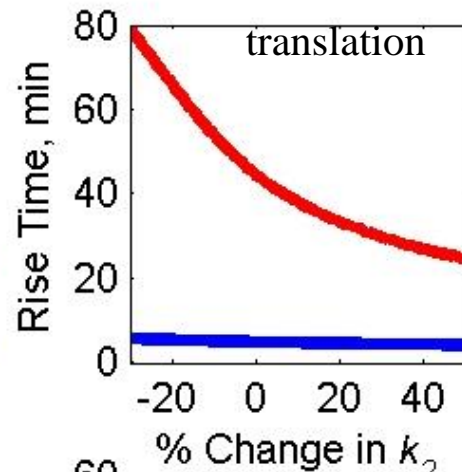
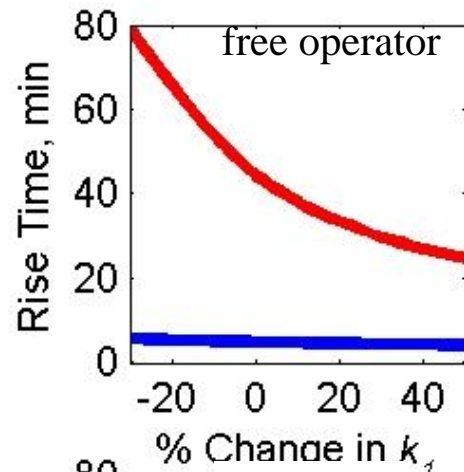
- Perturb one parameter at a time (co-ordinate directions only)



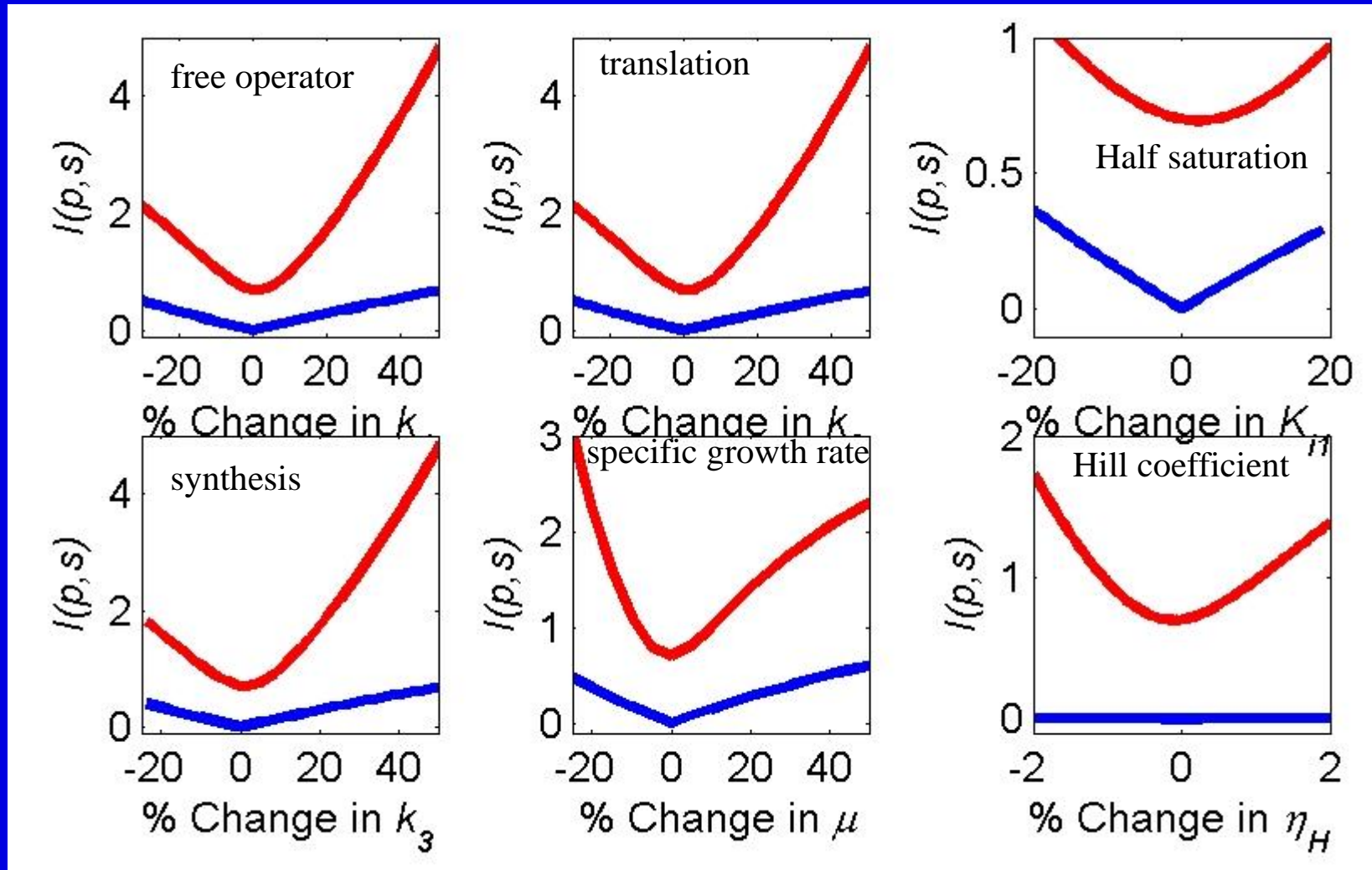
Robust Performance: $C_1C_2C_3$ active v/s C_1 active design; Metric: Rise time



Bhartiya, Chaudhary, Venkatesh and Doyle, Royal Society Interface, 2006



Robust Performance: $C_1C_2C_3$ active v/s C_1 active design; Metric: $I(p,s)$



Robust Performance: $C_1C_2C_3$ active v/s C_1 active design



- Multiple loop design yields superior dynamic performance
- Multiple loop design is robust thus making parameter values irrelevant (non-model based)
- Multiple loop design advantage for both *trp* physiological system as well as linearized system
- Robust to parameters yet fragile to structural mutations (HOT, RYF)



Results

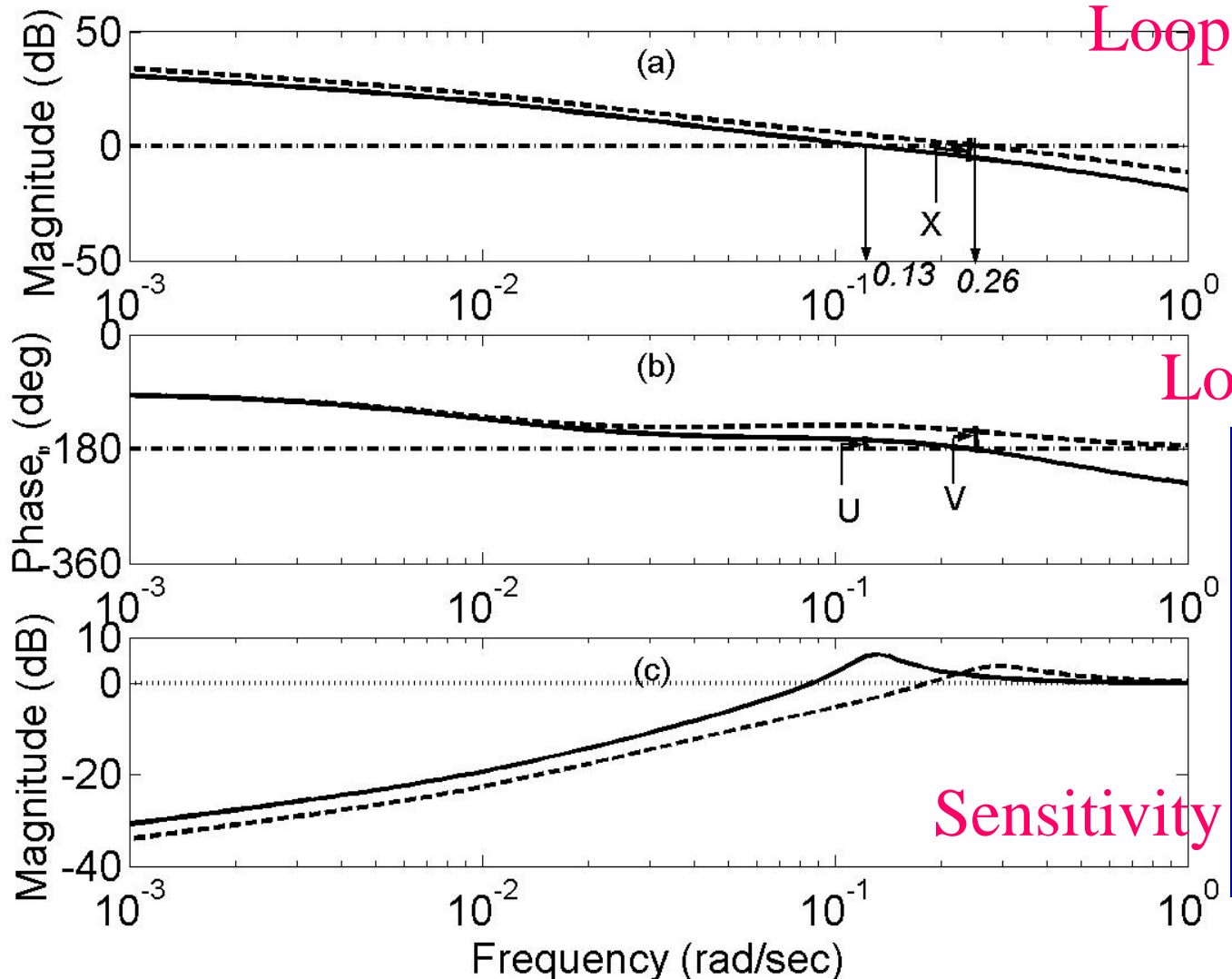
Nominal performance

Robust performance

Frequency Response Analysis

Frequency Response of Linearized *Trp* system

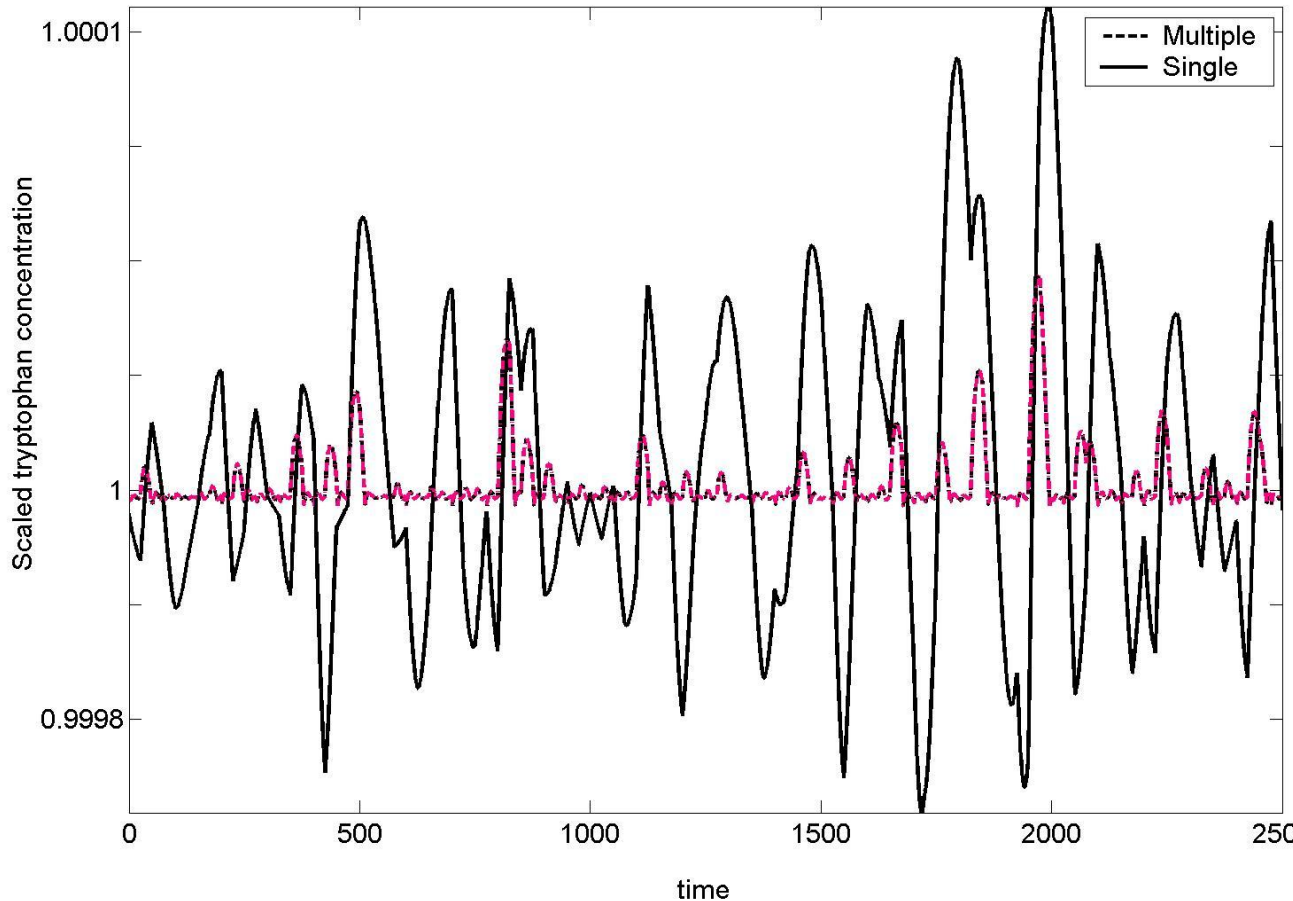
$C_1C_2C_3$ active versus C_1 active mutant



--- multiple
— single

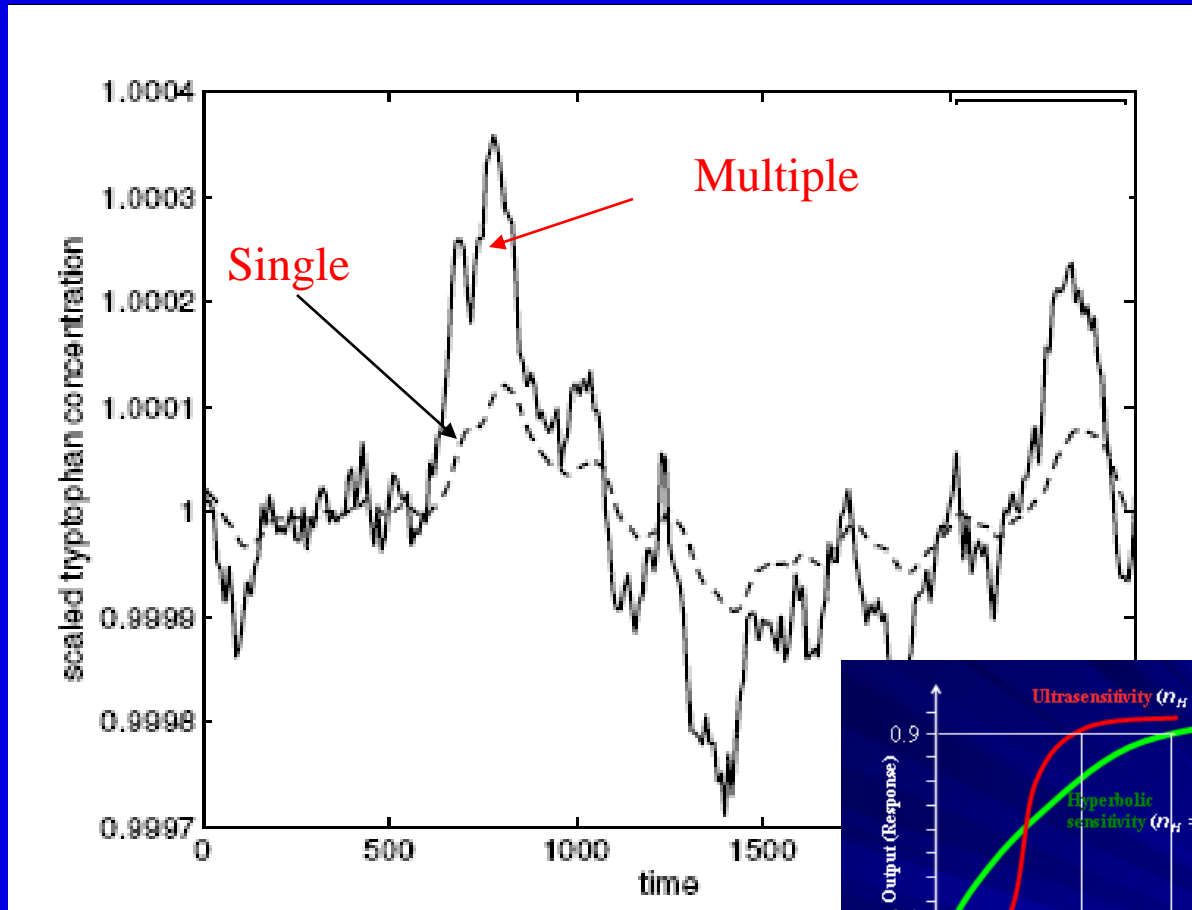
- Increased bandwidth, GM, PM, lower sensitivity peak
- Susceptibility to noise?

Noise Simulation with in vivo regulators: Injected at end transcription- A Langevin Approach

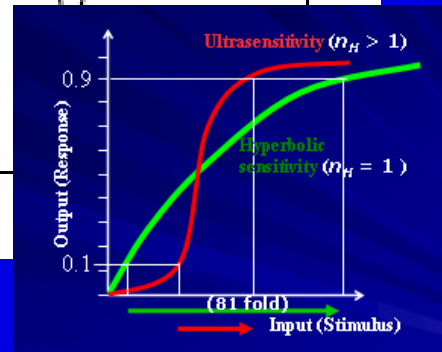


- Fixed step Euler
 - Single realization
 - Multiple feedback mitigates noise better! (contrary to intuition)
- ↓
- What is the role of nonlinear regulators?

Simulation with Sub-Sensitive Regulation



- η_H for C_1 , C_2 and C_3 is 0.5
 - Multiple feedback is more noisy
- ↓
- Observation: Is ultrasensitivity is responsible for noise mitigation?





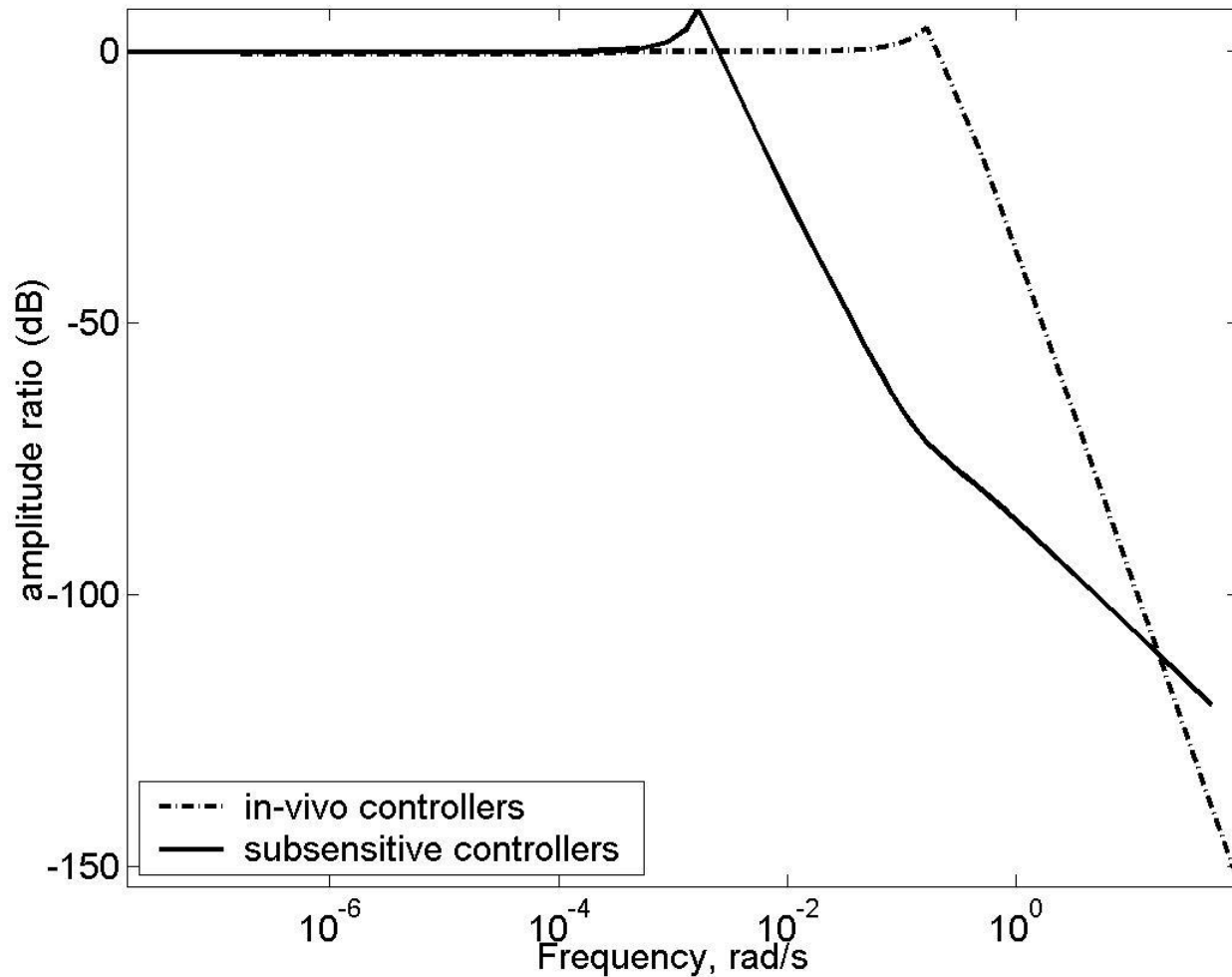
Frequency Response of *Trp* System Using Describing Functions

- **Quasi-linearisation:** Approximation of a nonlinear system by a linear one, which depend on some properties of the input
- **Describing Functions:** quasi-linear approximating functions—describe the transfer characteristics of a nonlinearity
- The graphical method described by Gelb and Vander Velde (1968) used to plot frequency response
- The tryptophan system is quasi-linearised around steady state concentration $T = 4.21 \mu\text{M}$
- Since the Hill equation represents an asymmetric nonlinearity, we divide it into two regions



Role Of Ultrasensitivity in Multiple Loop

Design Bavdekar, Venkatesh and Bhartiya, AIChE Annual Meeting, Indianapolis, 2005



•Subsensitive Design: η_H for C_1 , C_2 and C_3 is 0.5

•Observation: ultrasensitivity results in higher roll-off as well as retain higher bandwidth



Conclusions

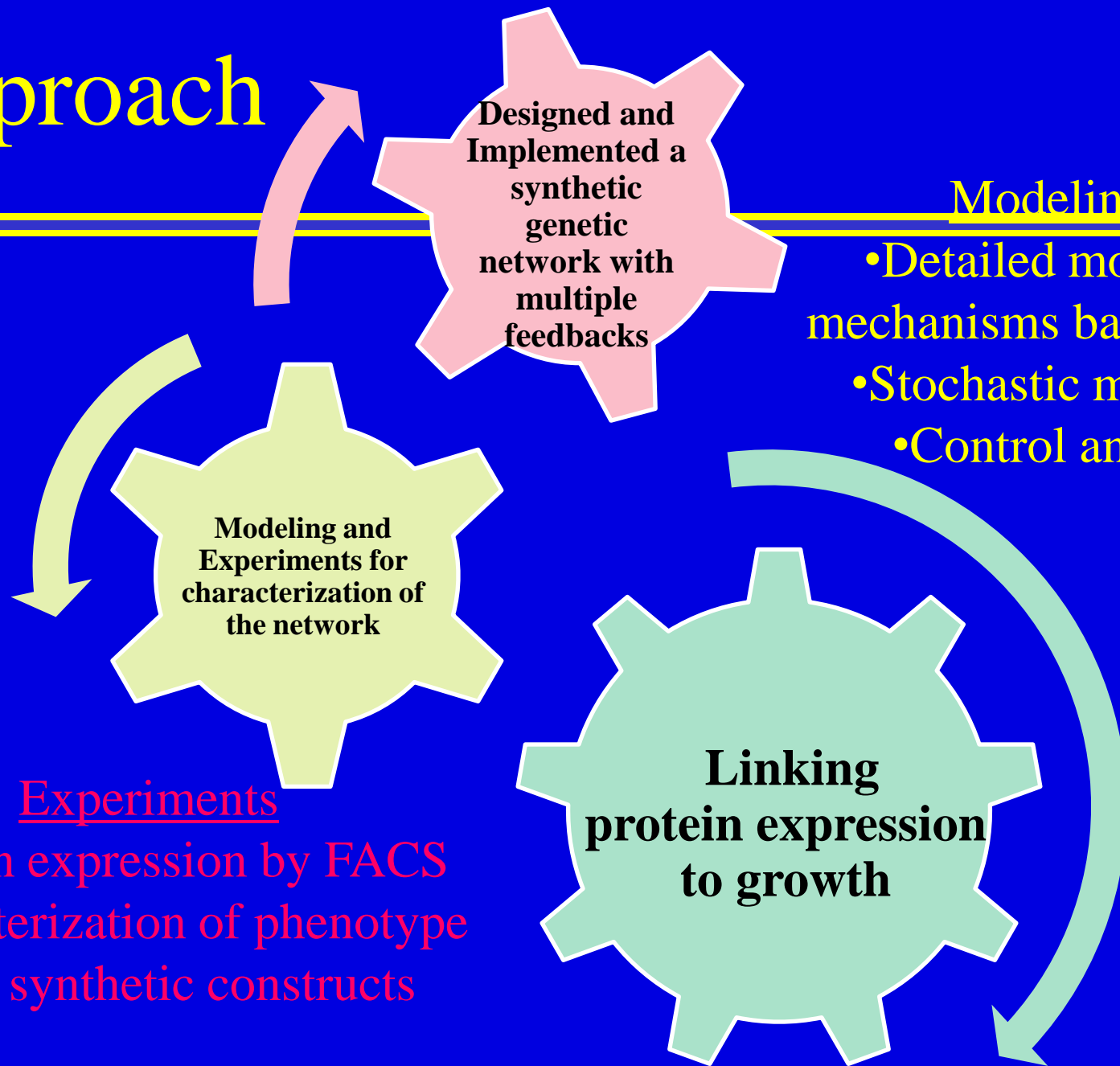
- Tryptophan System
 - Multiple feedback loops give bacterium a niche for survival during severe starvation
 - Nonlinear regulators counter the effect of fluctuations in nutritional environment
 - A prototype for analysis of naturally evolved systems



Implementation of multiple feed back loop strategy in a synthetic network



Approach



Designed and Implemented a synthetic genetic network with multiple feedbacks

Modeling –

- Detailed molecular mechanisms based model
- Stochastic modeling
- Control analysis

Modeling and Experiments for characterization of the network

Experiments

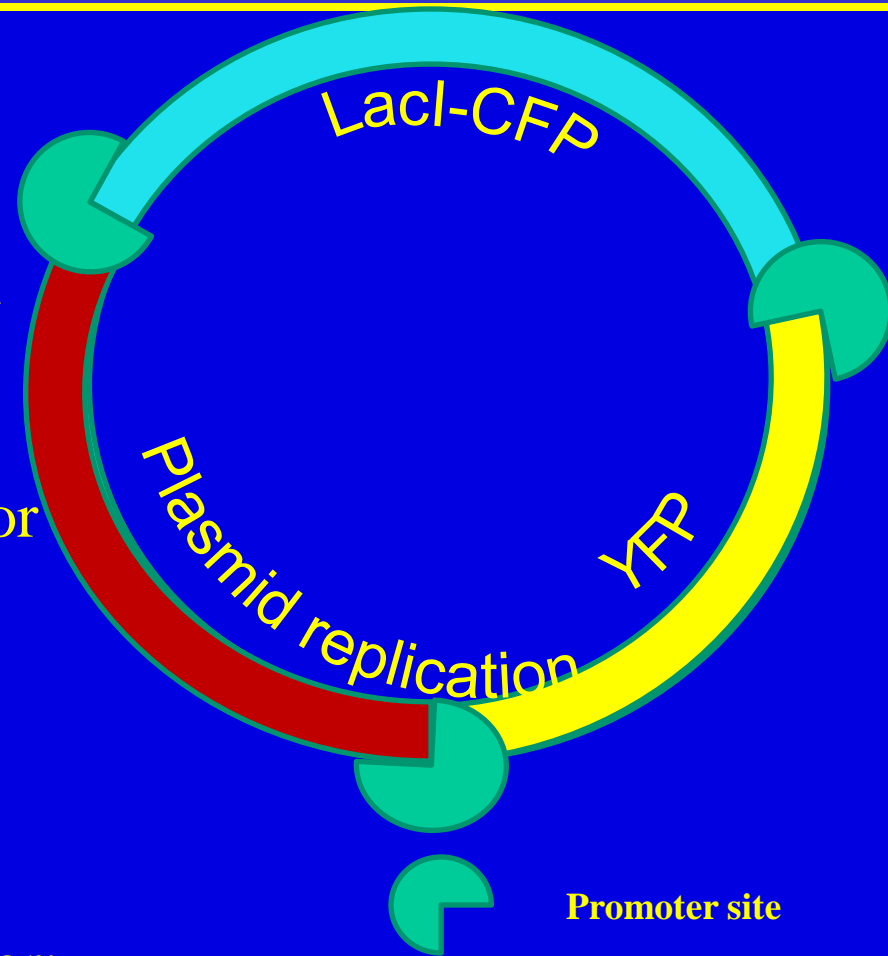
- Protein expression by FACS
- Characterization of phenotype in the synthetic constructs

Linking protein expression to growth

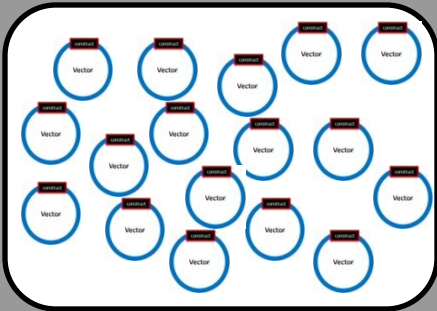
Components of Synthetic Constructs



- Use of existing bio-bricks
 - Four promoter sites used for the constructs: **pTet**, **pLac**, **pMB1** and **pLacOP**.
 - **pMB1** and **pLacOP** : promoters for plasmid replication.
 - To characterize amount of LacI:
LacI-CFP fusion protein.
 - To characterize plasmid copy number:
YFP expression.

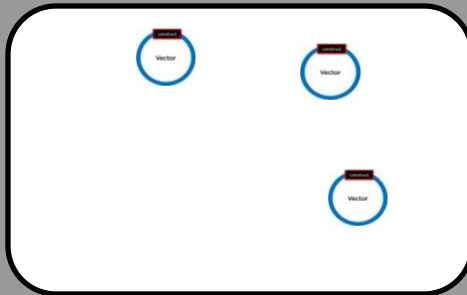
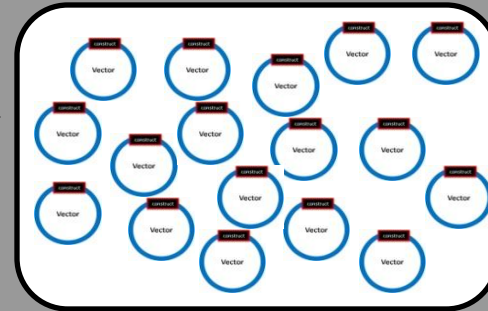


Characteristics of promoters used for Plasmid Replication



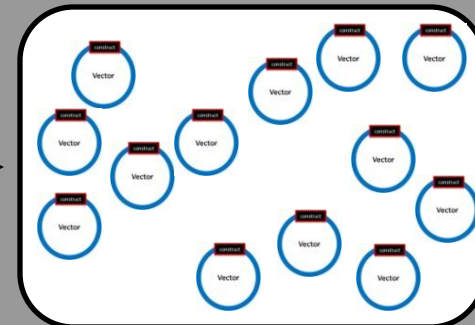
pMB1

On addition of IPTG
Plasmid copy number
does not change

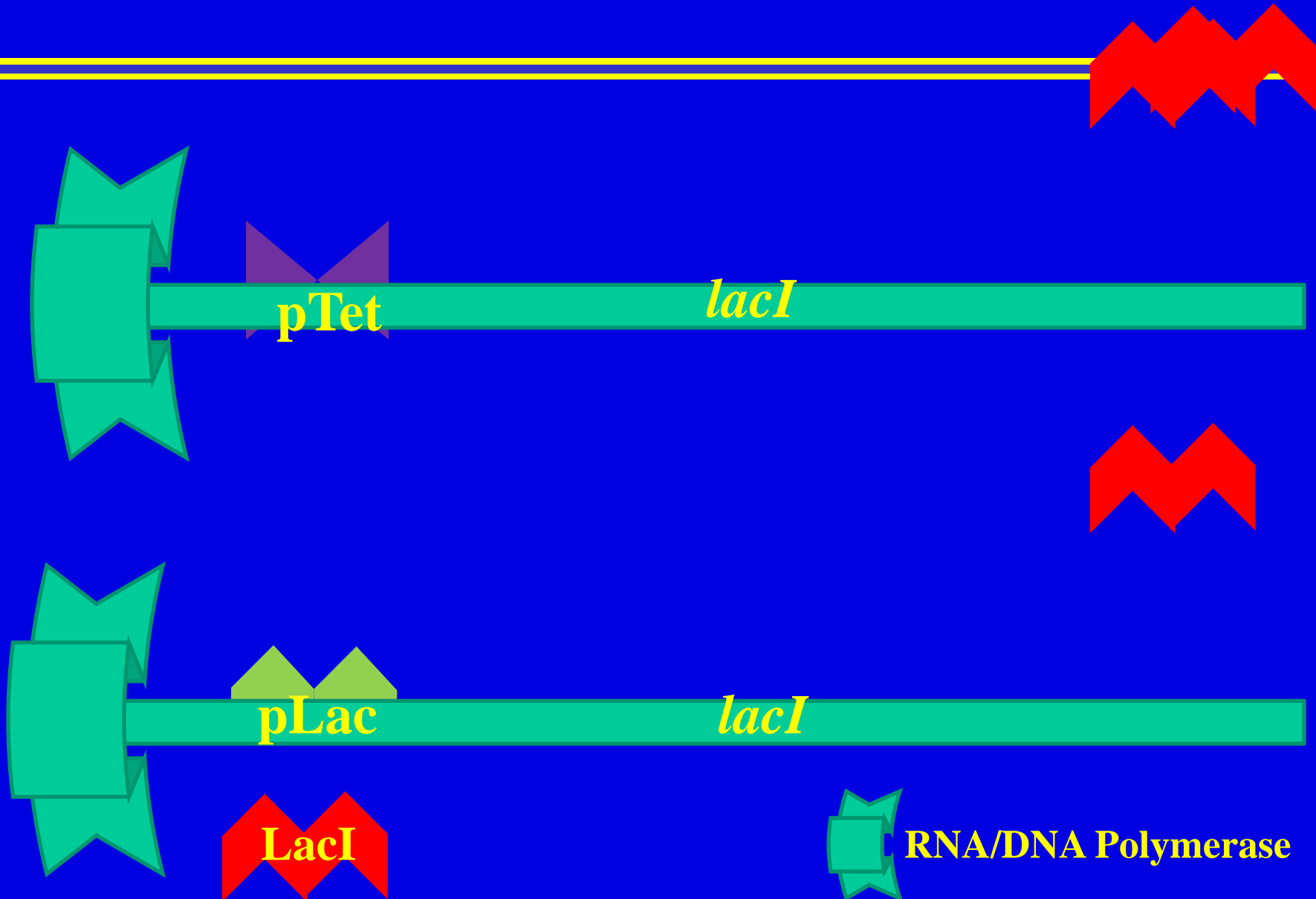


pLacOP

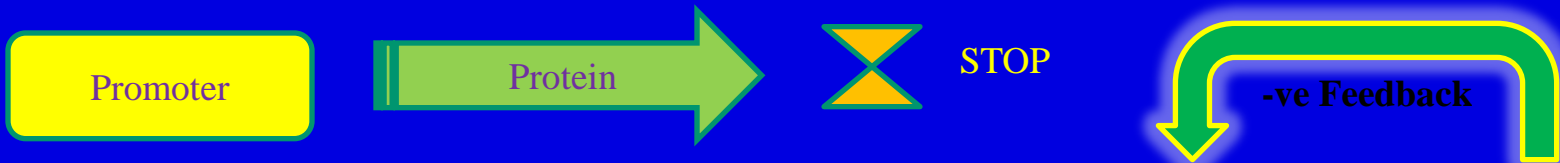
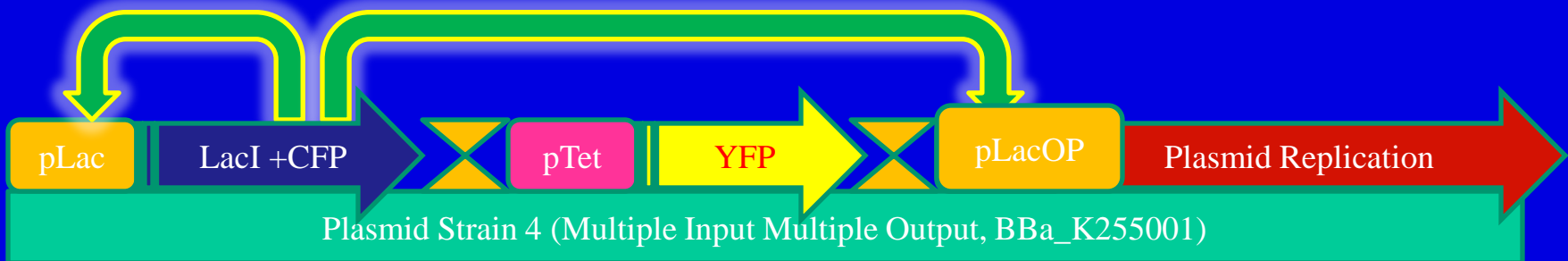
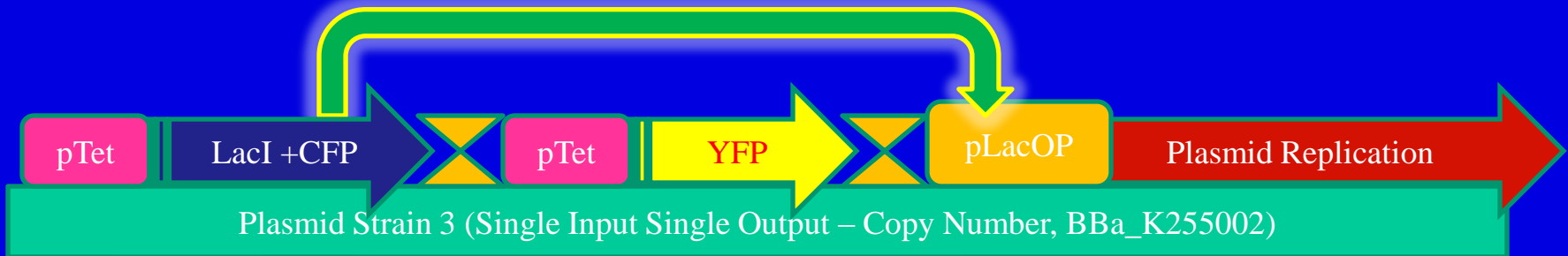
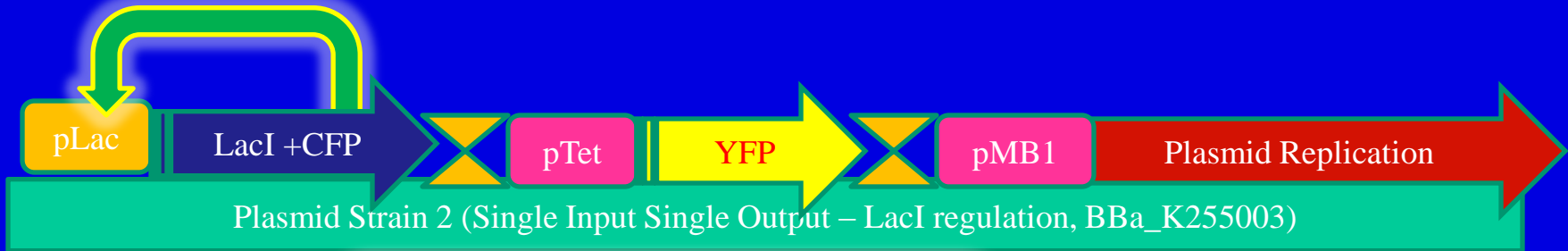
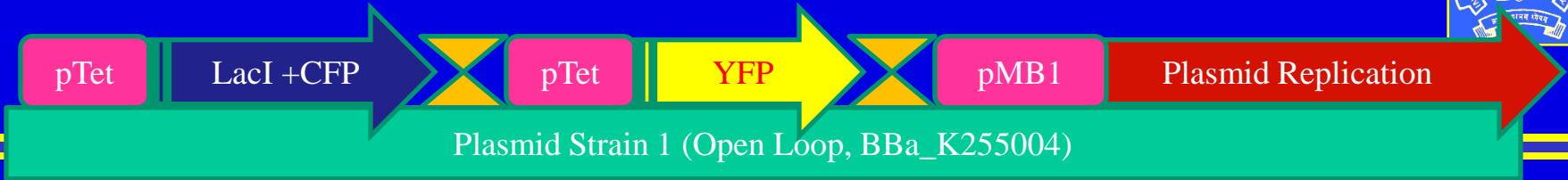
On addition of IPTG
Plasmid copy number
increases



LacI regulation in pTet and pLac



Constructs





SYNTHETIC CONSTRUCTS

NO
CONTROL

SINGLE INPUT
SINGLE OUTPUT

MULTIPLE INPUT
MULTIPLE
OUTPUT

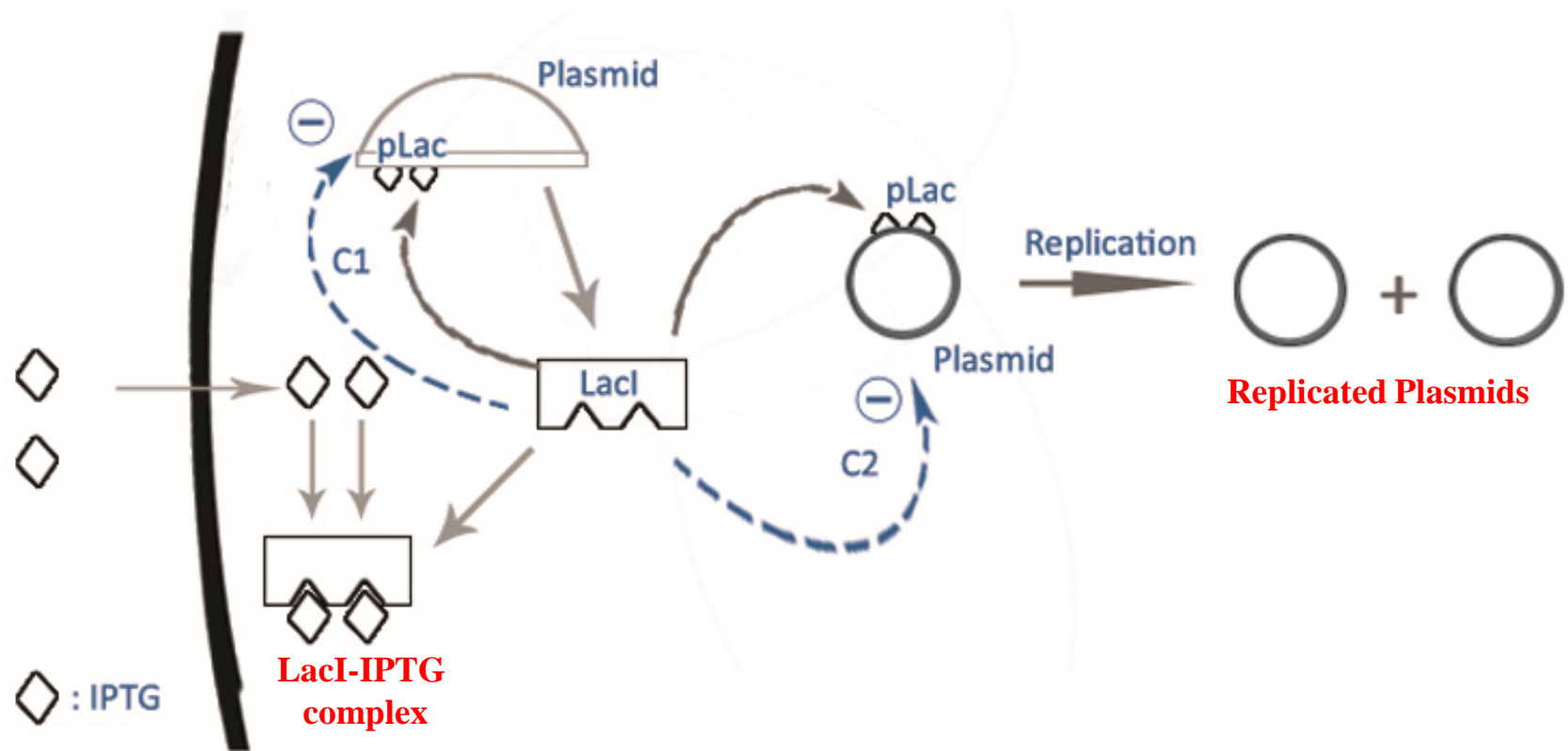
OPEN LOOP
(STRAIN 1)

SISO_LacI :
Regulation of
LacI (STRAIN
2)

SISO_CN :
Regulation of
Plasmid Copy
Number
(STRAIN 3)

MIMO:
Regulation of
Plasmid Copy
Number and LacI
(STRAIN 4)

Molecular Map of the Construct

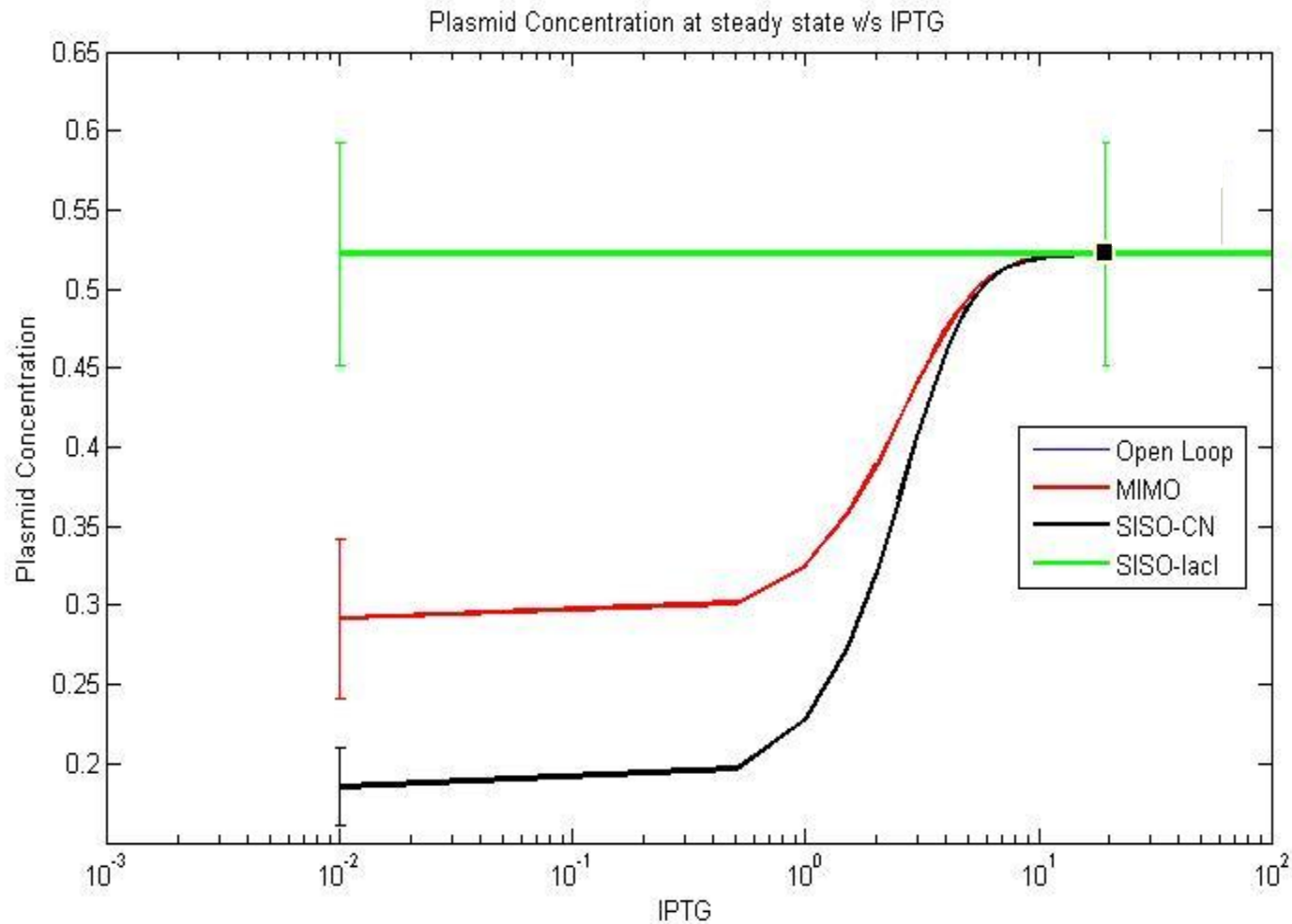




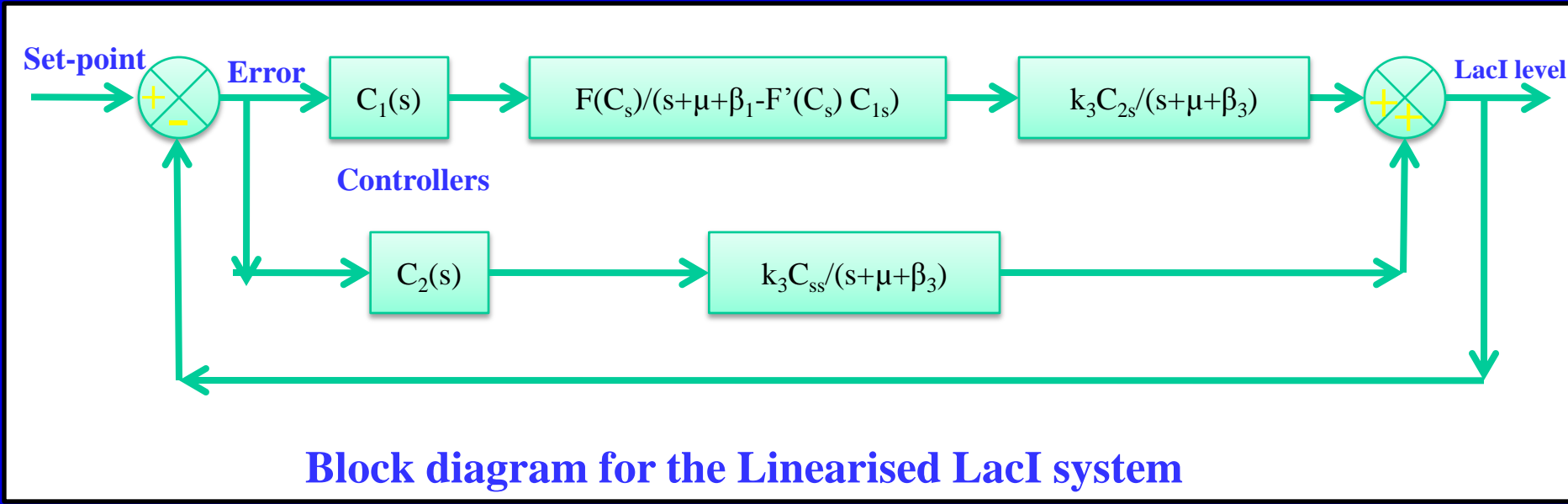
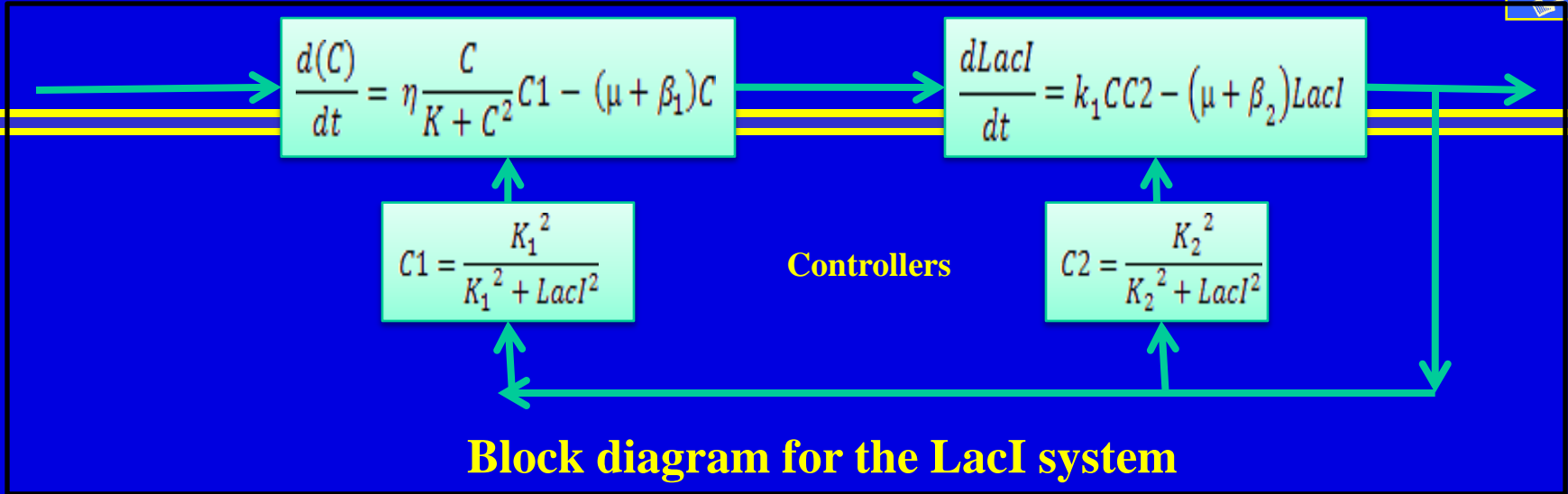
Modeling Methodologies

- Detailed Dynamic Modeling using all known molecular interactions
- Stochastic Analysis on a simplified model using Langevin approach
- Frequency response analysis on the linearised model

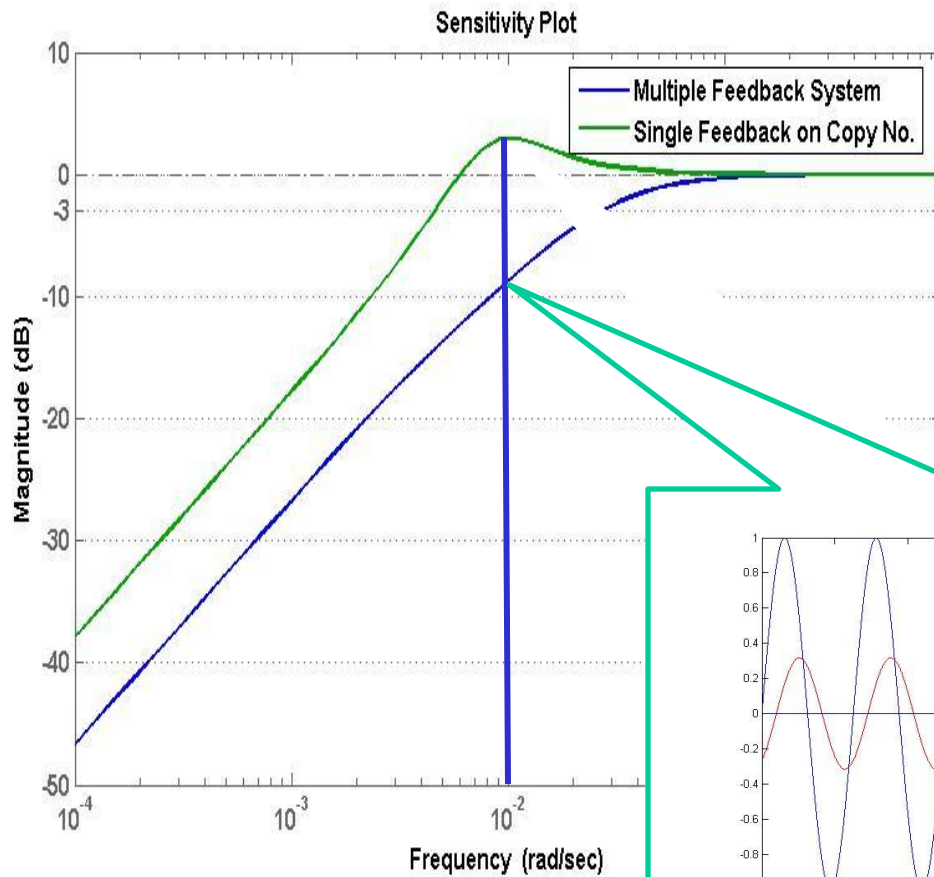
Prediction of Steady State Expression of YFP (Plasmid Copy Number)



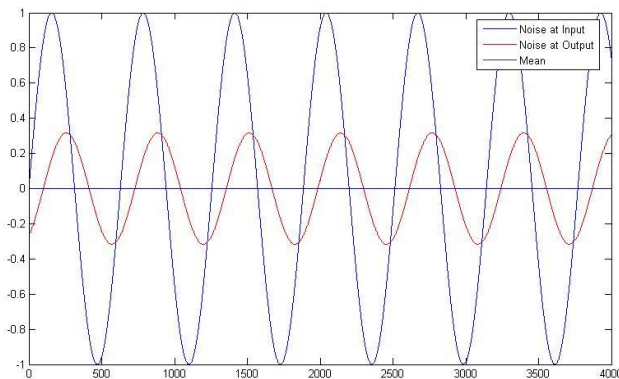
Control Analysis to Characterize System Behavior



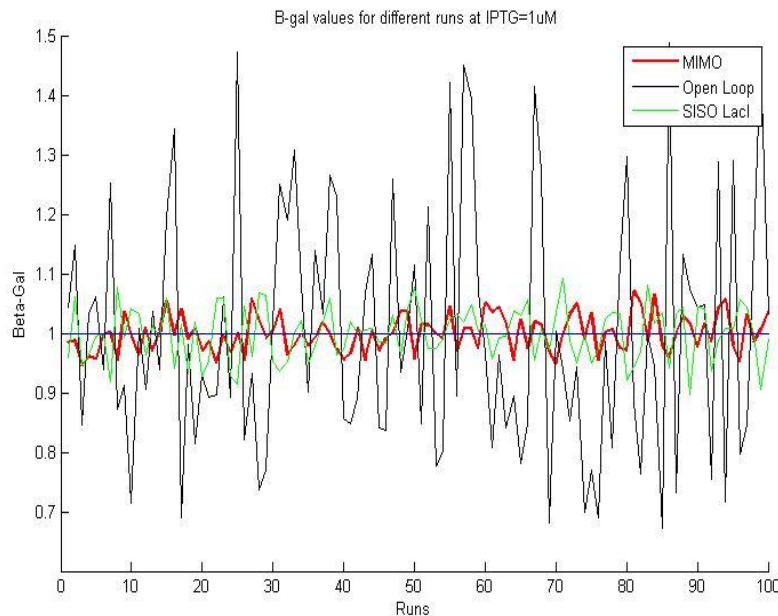
Frequency Response Analysis



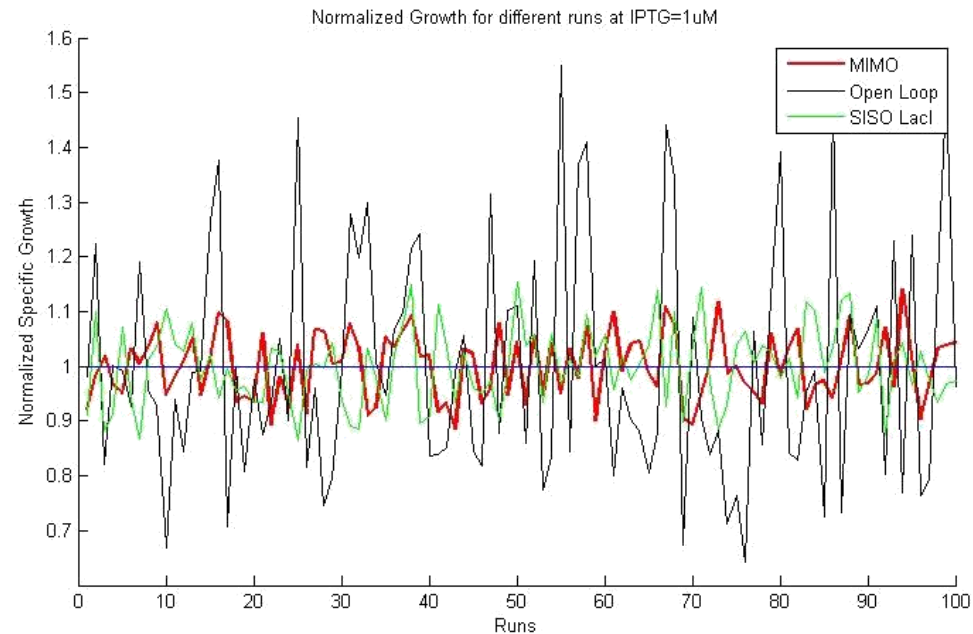
- **Higher bandwidth**
- **Higher Phase margin**
- **Noise Attenuation**



Stochastic Modeling on Growth Rate



**NORMALIZED β -gal
EXPRESSION**



**SPECIFIC GROWTH
RATE**

For perturbation of the kinetic parameters around the mean value, we see MIMO has the least variance compared to open loop or a single feedback system

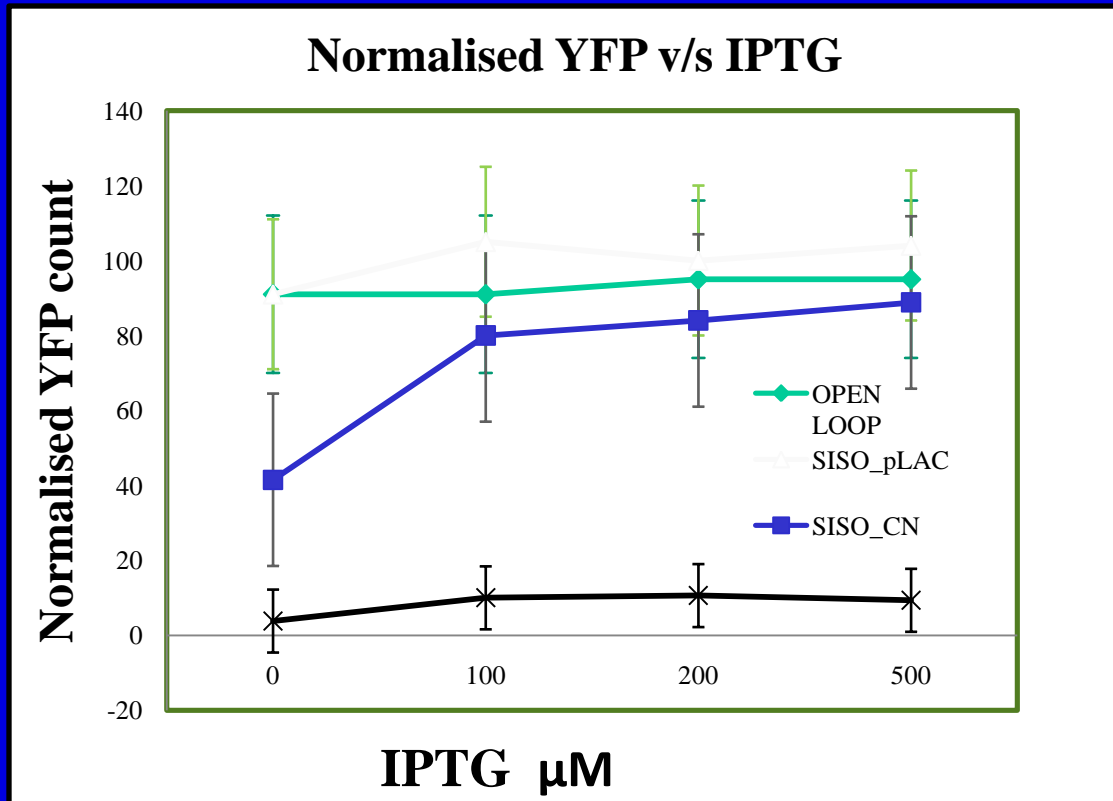


Experimental Validation

- Experiments with various IPTG concentrations were conducted.
- Protein expression measured as YFP using FACS to quantify plasmid copy number.
- Mean and Variance obtained from the distribution.



Experimental YFP expression (characterizing Plasmid Copy Number)



- **Open Loop and SISO_LacI:**
No increase in YFP with inducer
- **SISO_CN and MIMO:**
expression increase with inducer

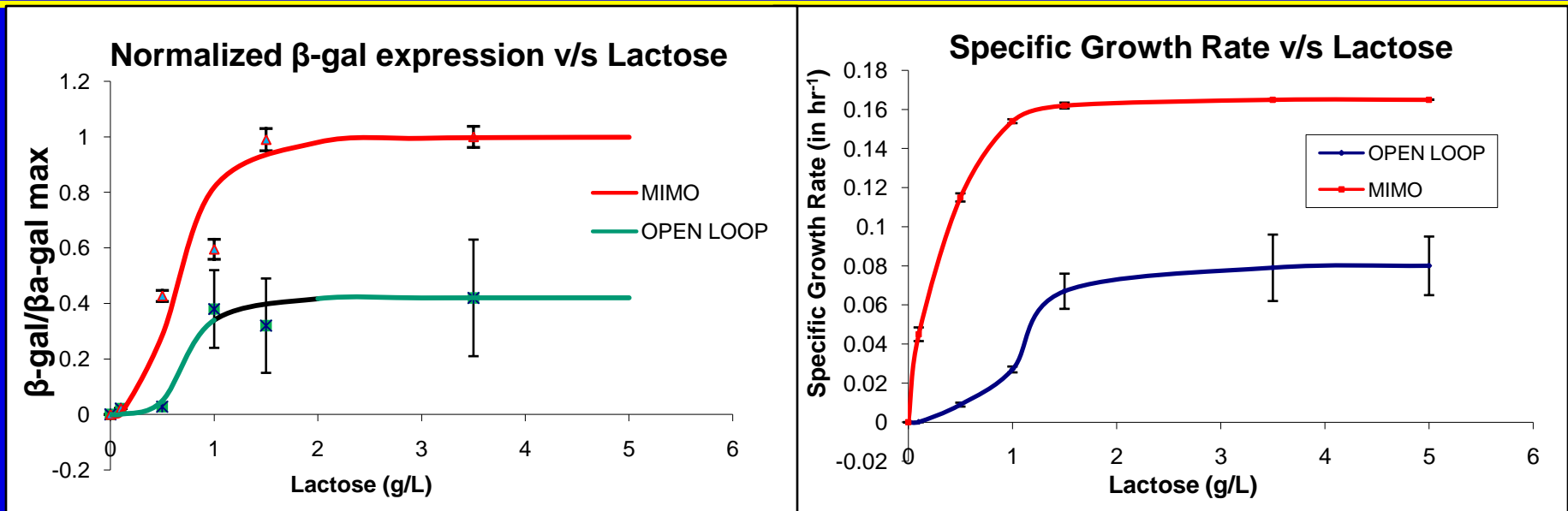
Higher variance
in open loop



Characterization of LacI expression

- An indirect measure of LacI was obtained by measuring β -galactosidase from the *lacZ* of the host.
- Further the growth rate of the four transformants were also enumerated.

Experimental Results



Noise in protein expression propagates to growth

The variance in specific growth rate is less compared to that observed in protein expression.

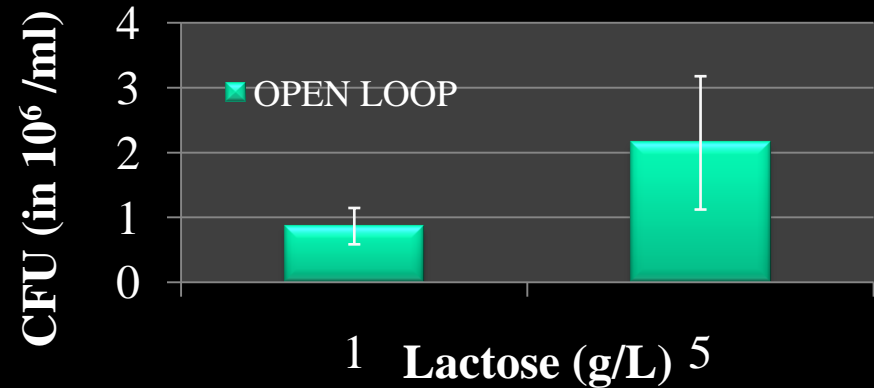
Agar Plate Experiments

Strains were grown on agar plate with different lactose concentrations.

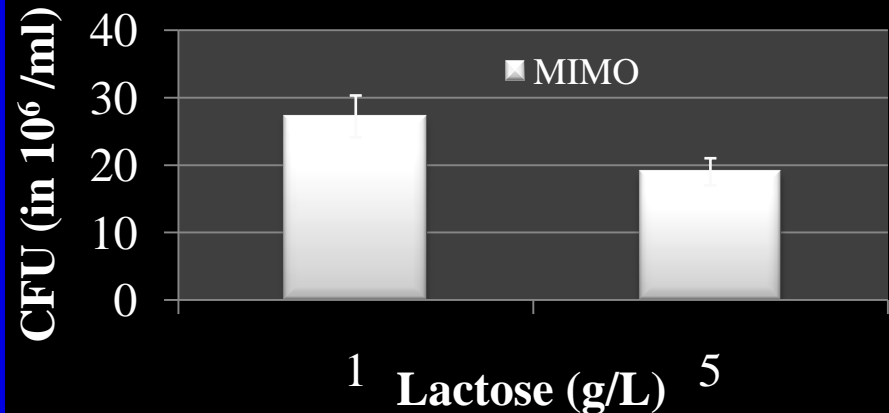
Colony Forming Units in the agar plates were counted.

Variance in Open Loop is 40 % and MIMO is 10%.

Agar Plate Experiment (without IPTG)



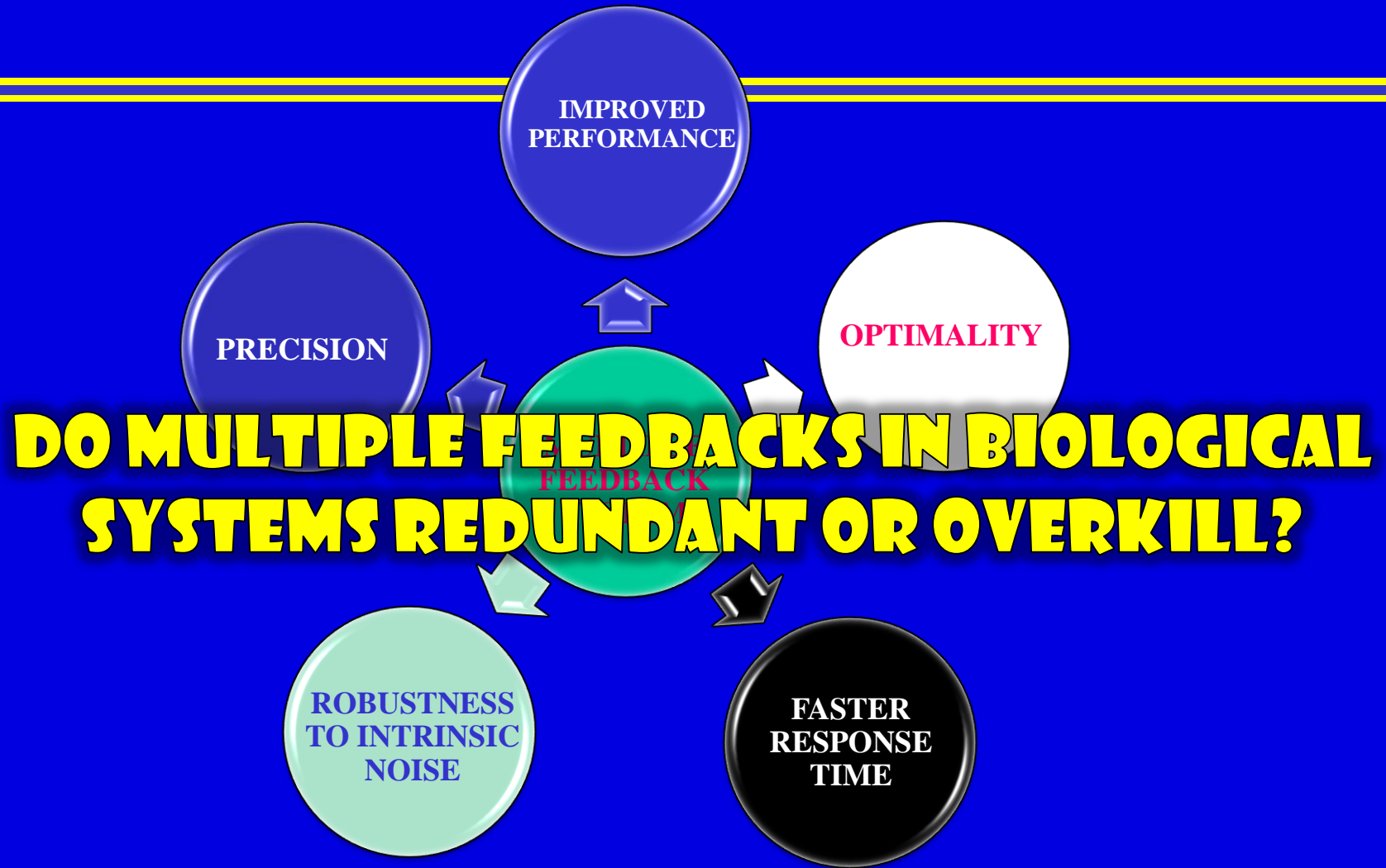
Agar Plate Experiments (without IPTG)





Recapitulating...

- Robustness in protein expression which leads to low variance in specific growth rate.
- The noise in protein expression is filtered leading to a decrease in the variance in growth rate. This may be due to metabolism and division process.
- The transformants with the synthetic network yields distinct phenotypic response.





Collaborators / Students

- Prof. Sharad Bhratiya (IITB)
- Dr. Mukund Thattai (NCBS)
- Nikhil Chaudhary (PhD, IITB)
- iGEM team 2009
- Pushkar Malakar (PhD, IITB)



Thank you!!!