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MAX PLANCK INSTITUTE FOR
TERRESTRIAL MICROBIOLOGY



Information processing by bacterial quorum sensing systems

Ilka B. Bischofs

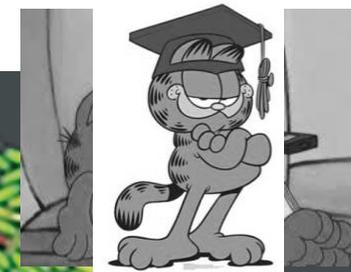
μ CATs-Lab @ BioQuant, Heidelberg



Understand



Control



Engineer

Adaptation

„The only constant in life is change.“



Information about the niche.

Adaptation

„The only constant in life is change.“

**Information
about environmental
conditions.**



Environmental conditions change.



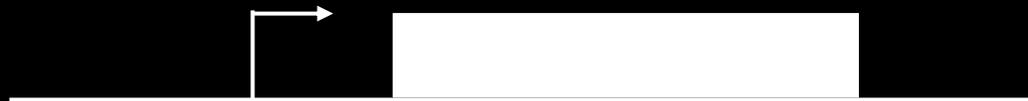
Adaptation

Environment

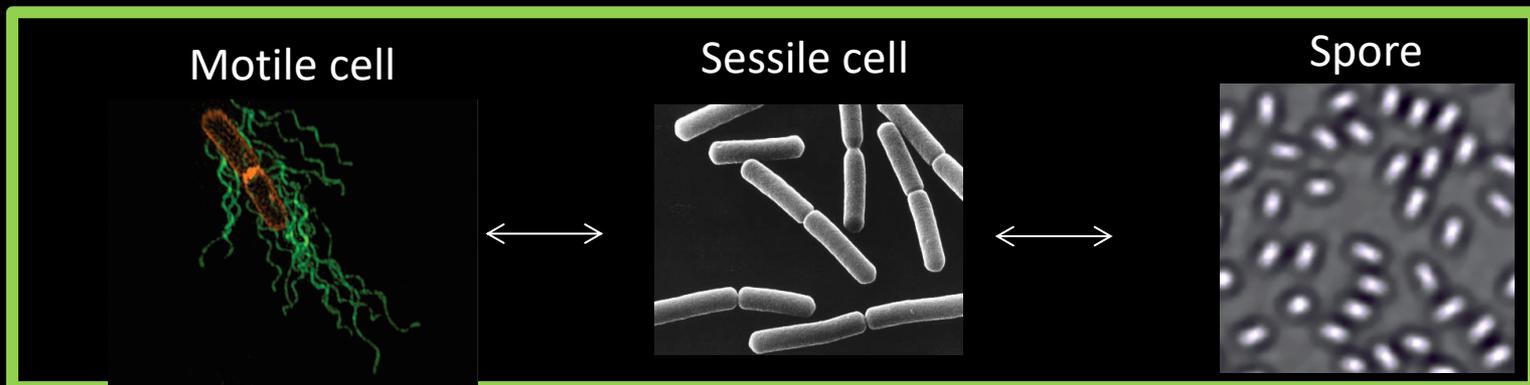


Cellular state

Gene expression

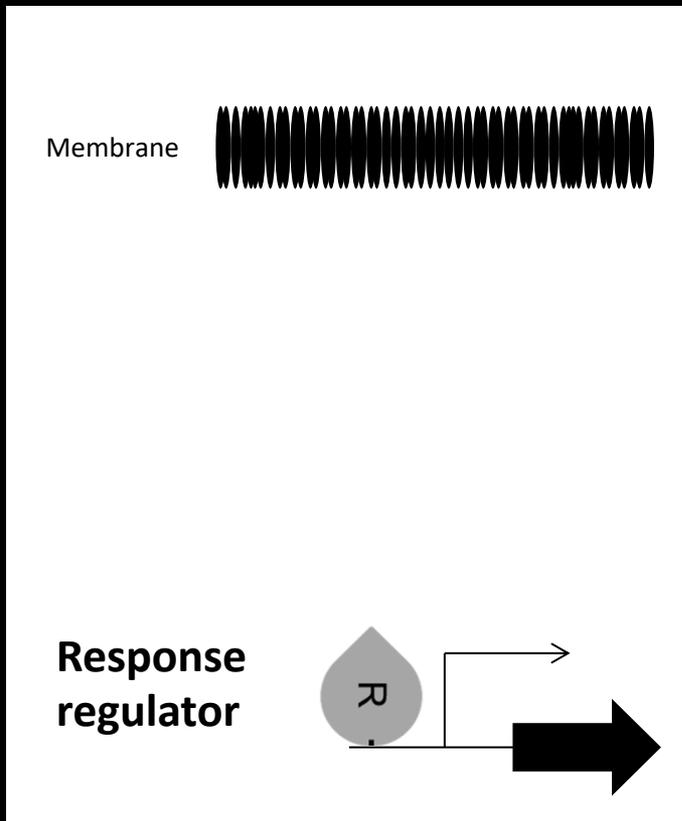


Cell differentiation



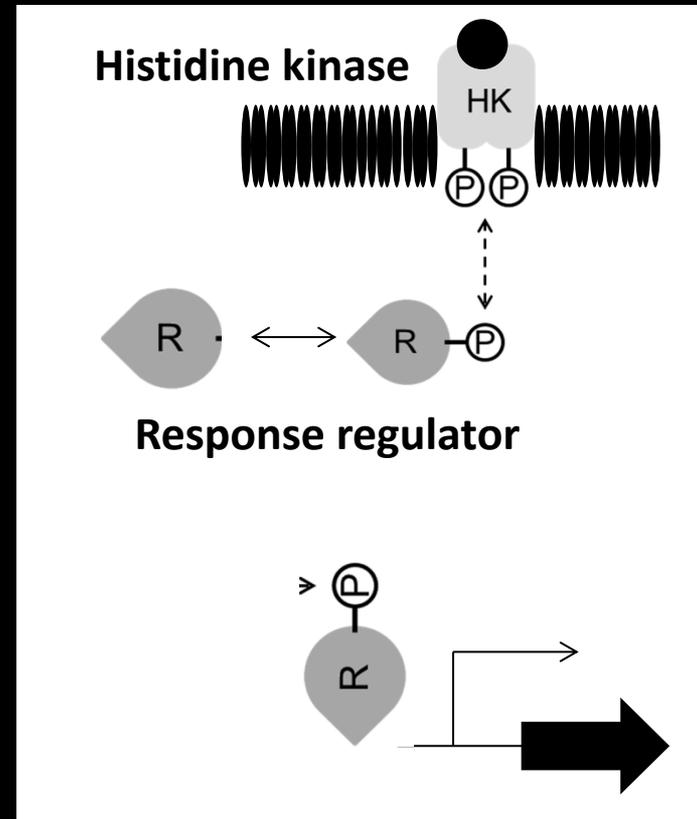
Signal Transduction in Bacteria

One component systems



e.g. *Lac repressor*

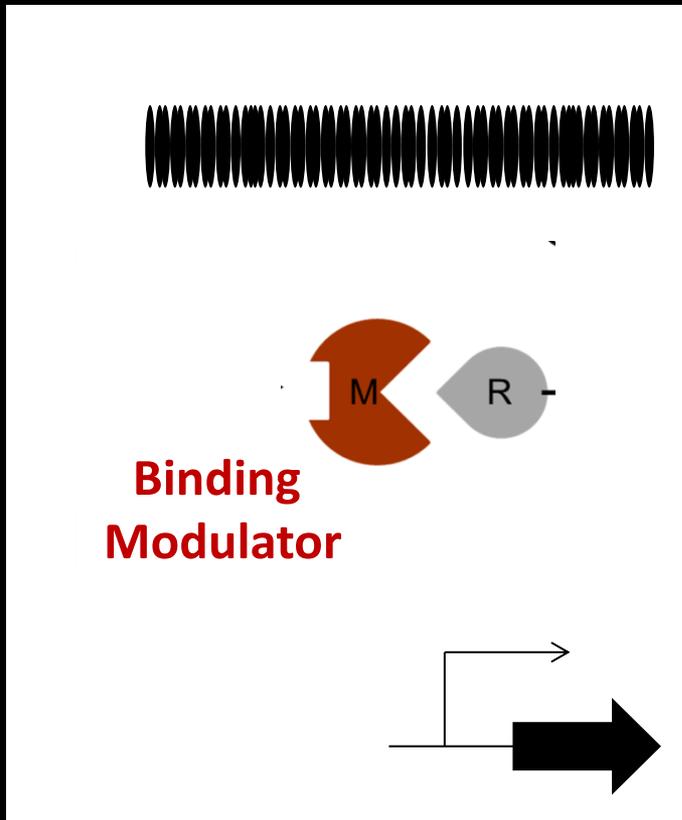
Two component systems



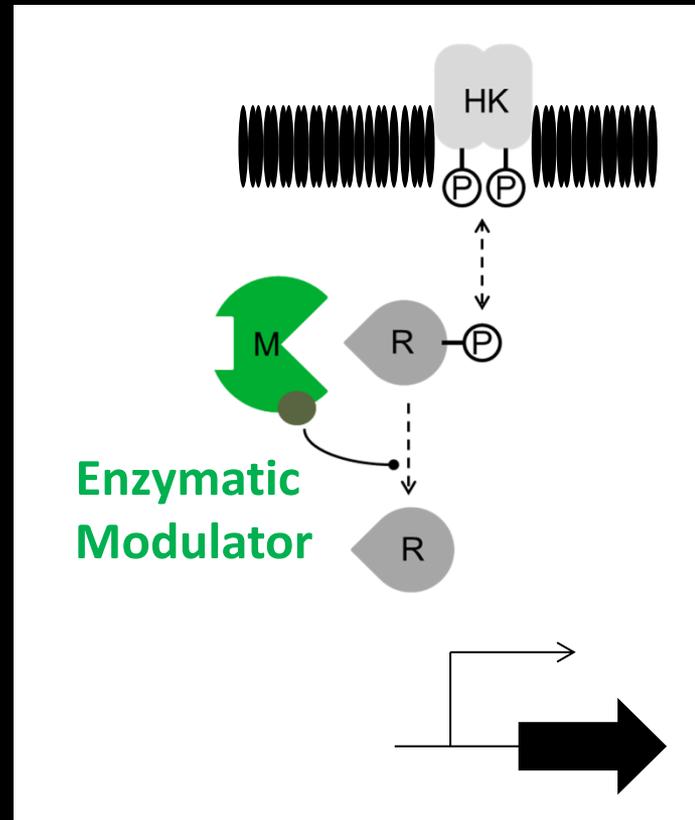
e.g. *chemotaxis CheA-CheY, EnvZ/OmpR*

Signal Transduction in Bacteria

One component systems



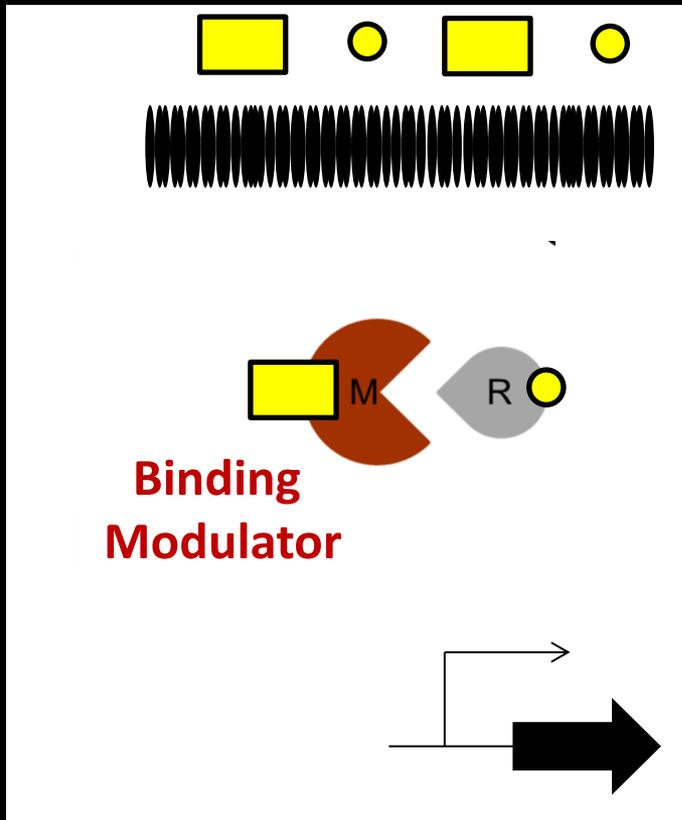
Two component systems



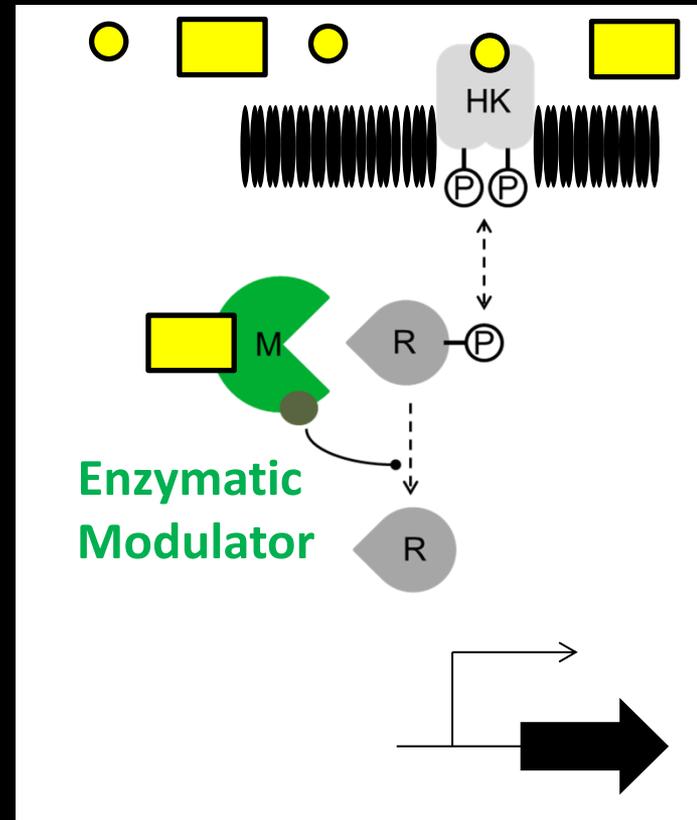
Auto-inducer signaling (Quorum sensing)

■ Auto-inducer ●

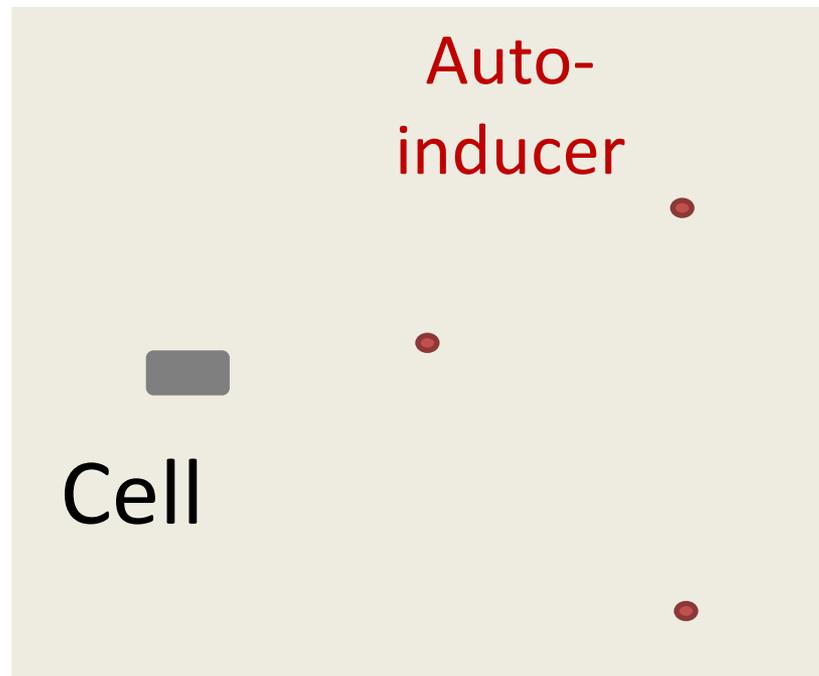
One component systems



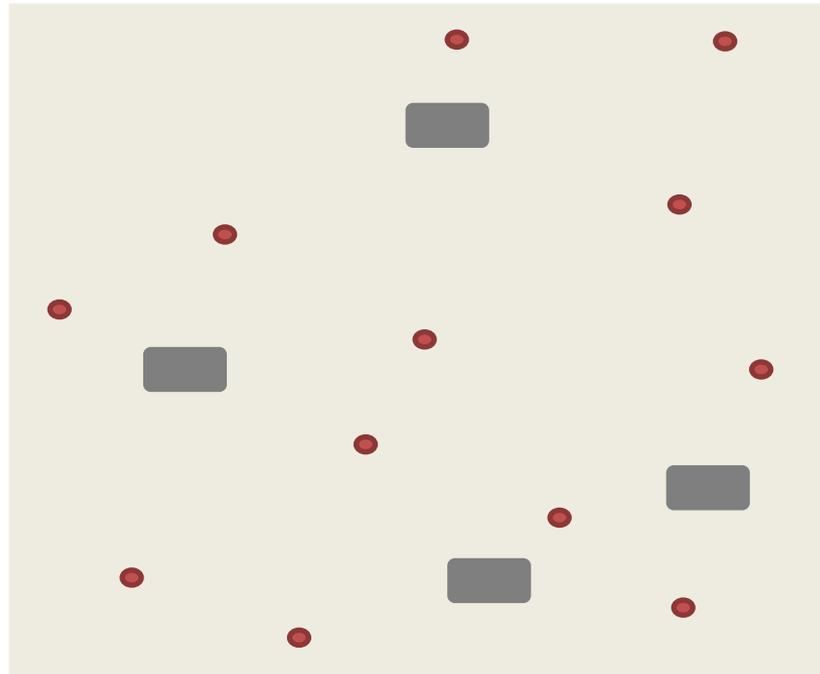
Two component systems



Quorum Sensing

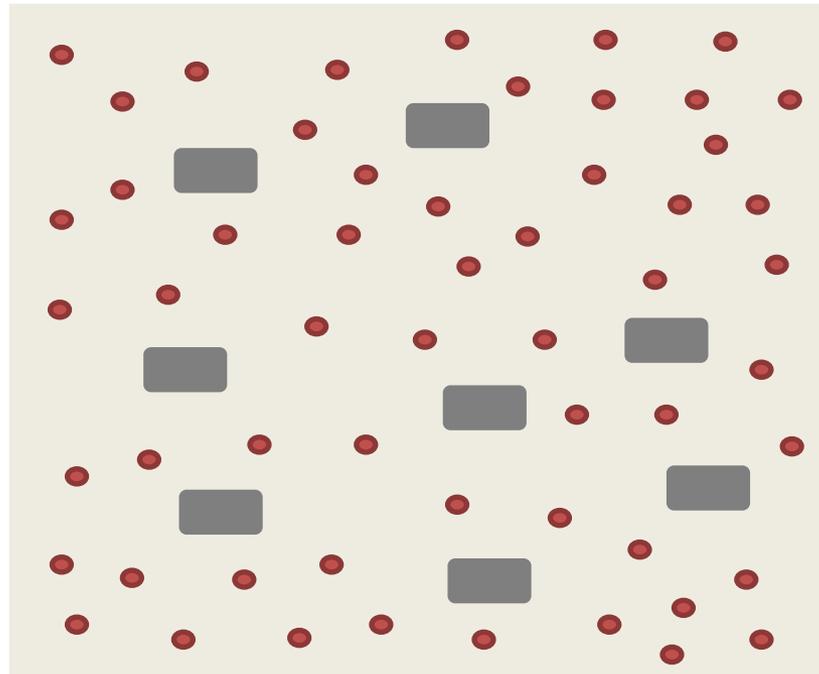


Quorum Sensing



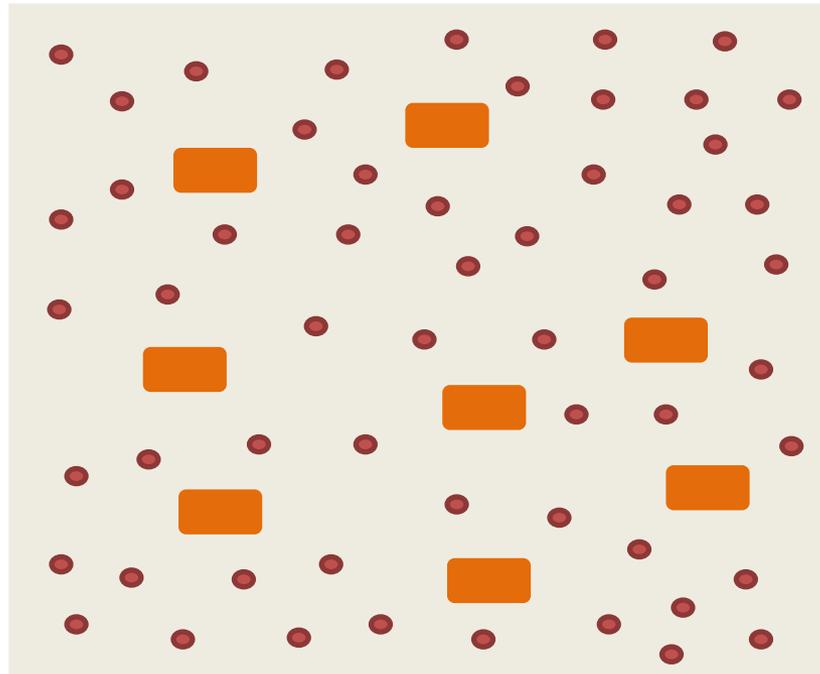
more cells – more signal

Quorum Sensing



*„The quorum“ –
a minimal behavioral unit*

Quorum Sensing



„The quorum response“

(Almost) 50 Years of Quorum Sensing

First observations

1964 Tomasz & Hotchkins (G+)

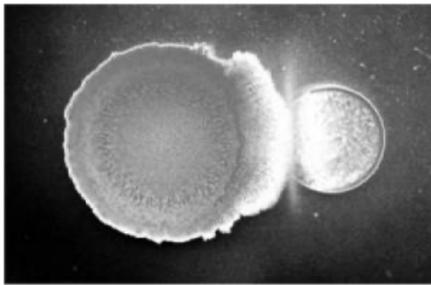
1970 Neaslon et al. (G-)



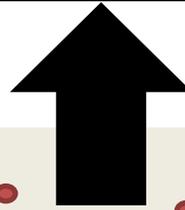
**Bioluminescence is
regulated by cell density.**

The Quorum Response Spectrum

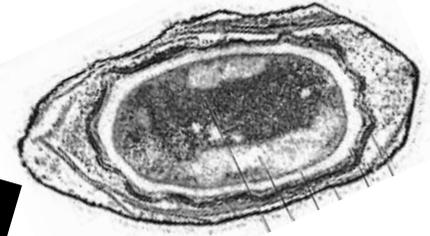
Production of exofactors



Horizontal gene transfer



Cell differentiation



- Cost of production
 - Benefit dependent on enzyme accumulation
- => Density-dependent fitness function
See work by Lingchong You

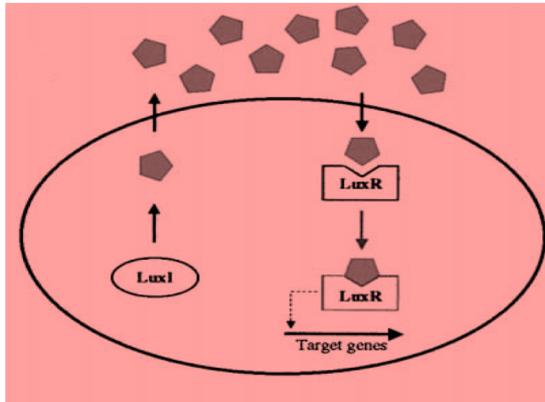
Host-microbe interactions

Biofilms



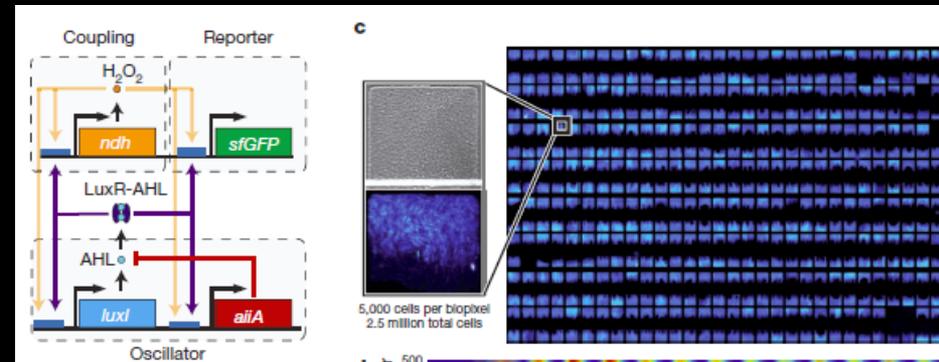
Quorum Sensing in Synthetic Biology Applications

LuxIR-type of QS Circuitry



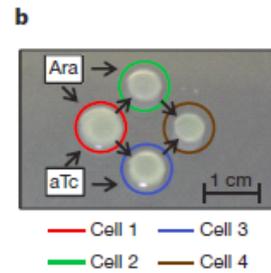
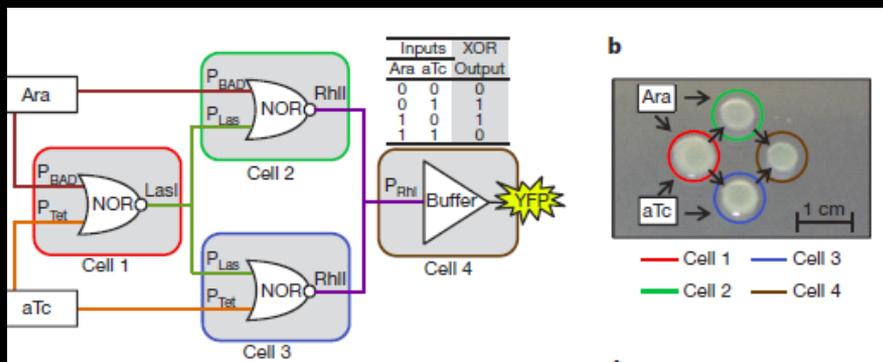
V. fischeri (Bassler, 2002)

Advanced Biosensors



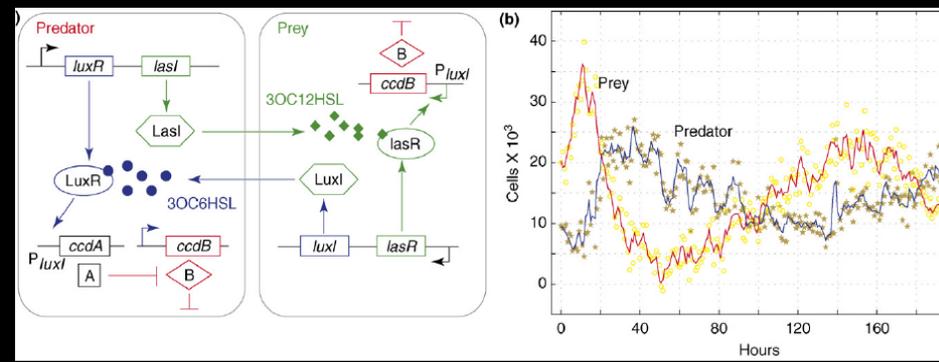
Prindle et al., Nature 2012

Multi-cellular Computing



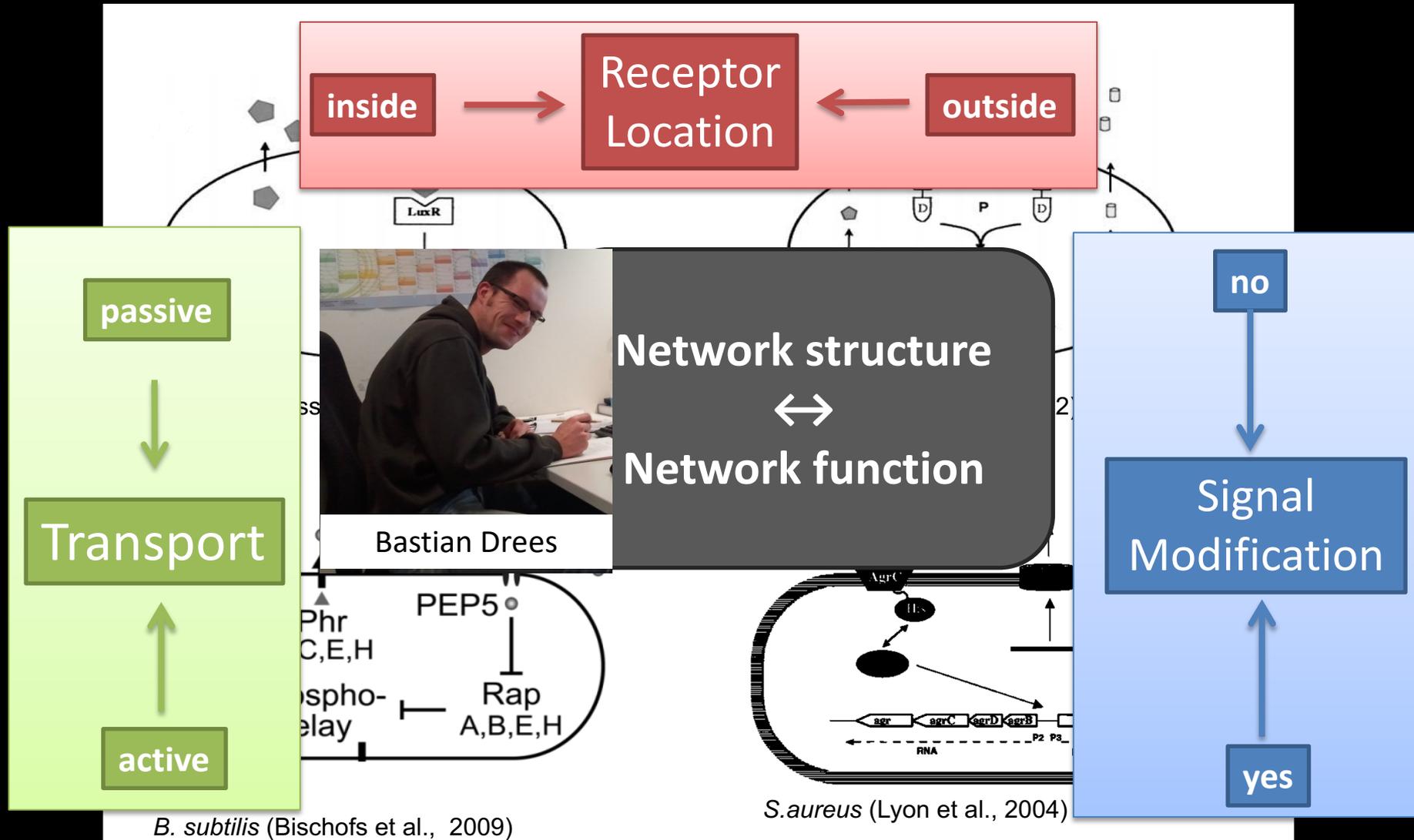
Tamsir et al., Nature 2011

Synthetic Ecology

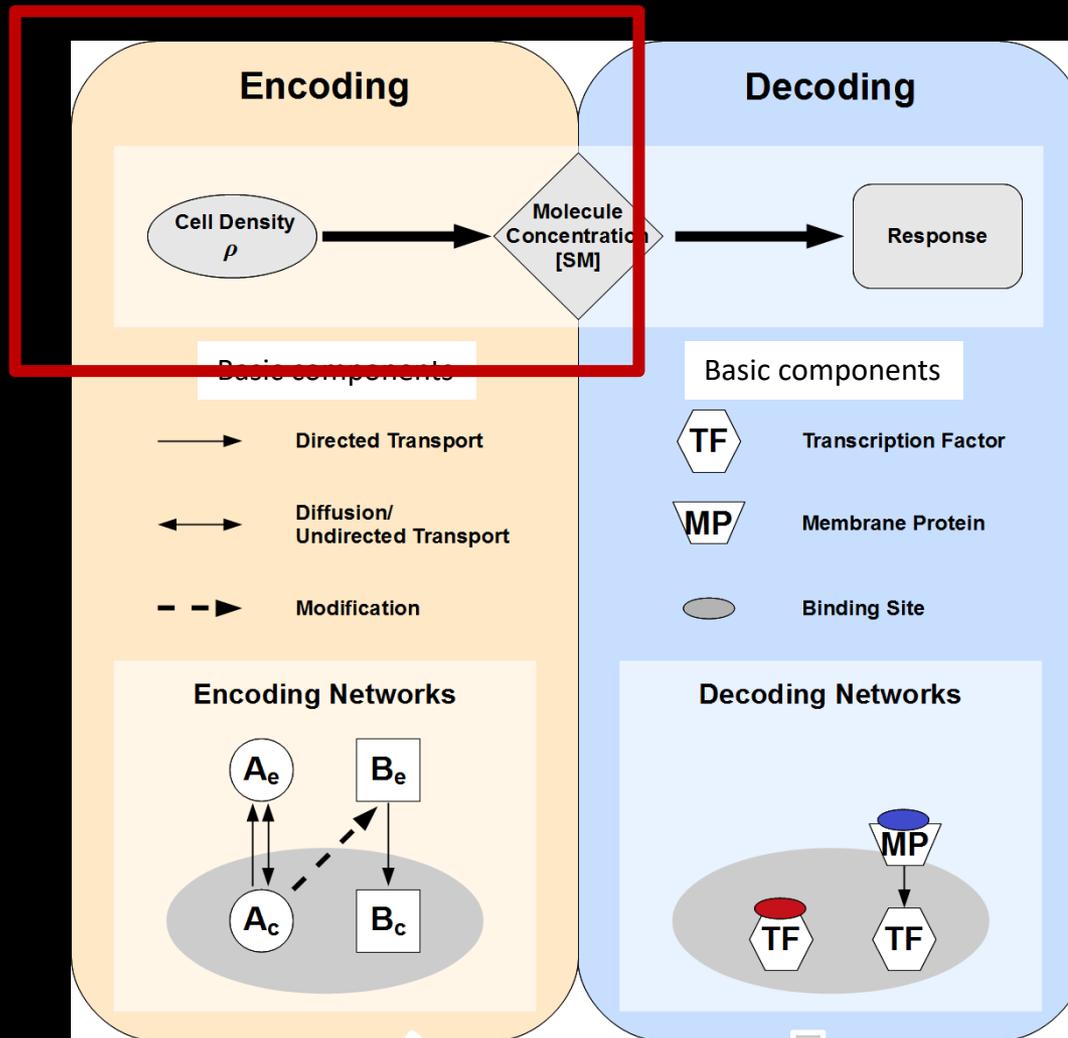


Brenner et al., Trends in Biotechnology, 2008¹⁴

Diversity of Quorum Sensing Architectures

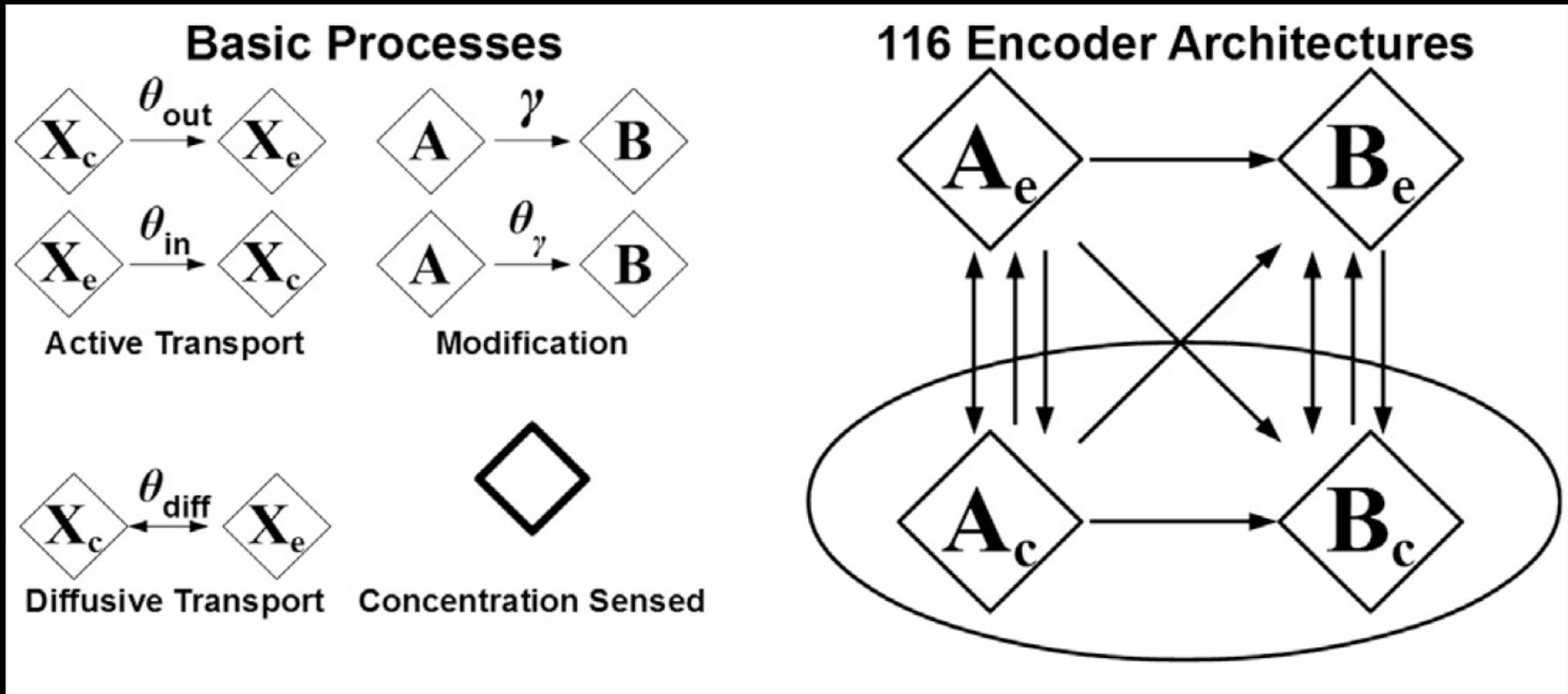


Adopting a modular view on QSS



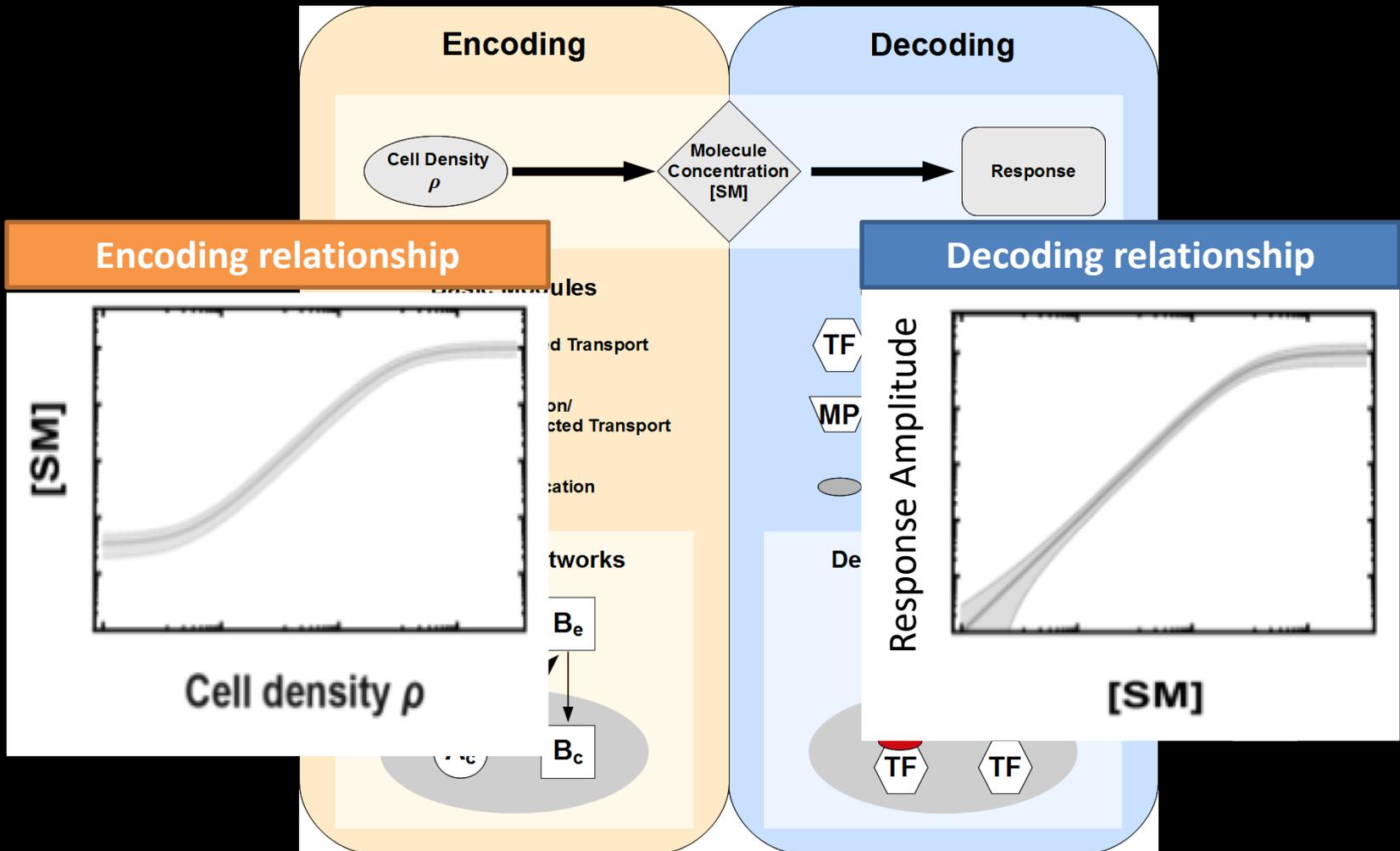
Feedback (often but not all cases)

Diversity of Encoding Architectures

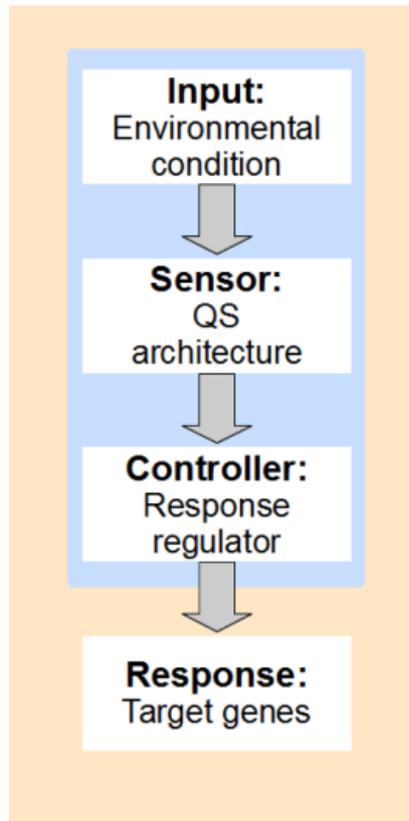


99 networks (85%) are capable of encoding information about cell density into SM concentration.

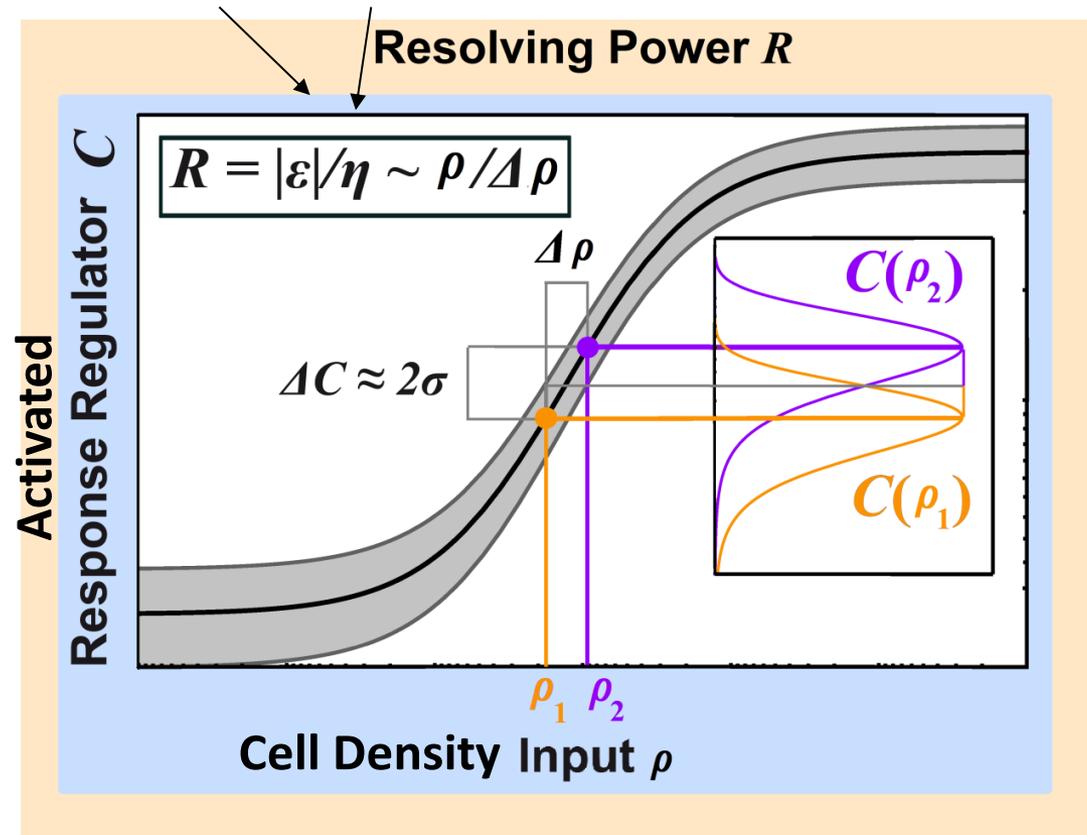
Adopting a Modular View on QSS



Defining a Performance Criterion

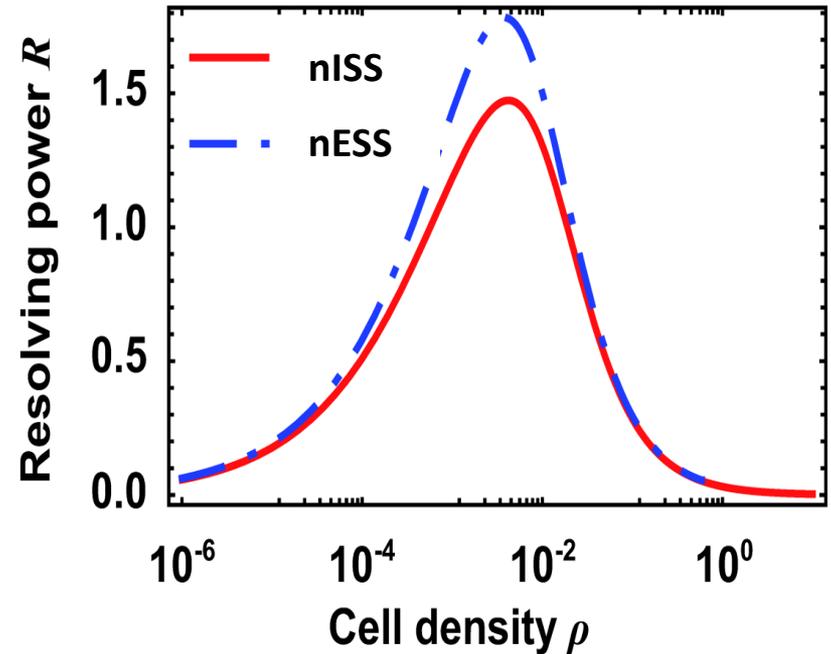
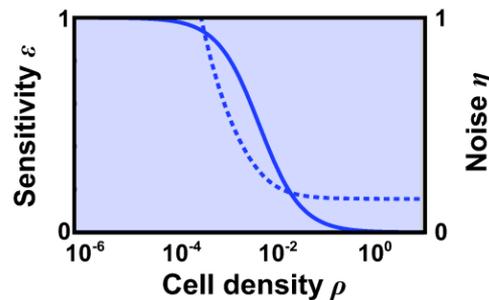
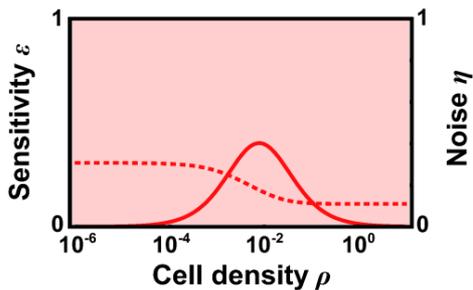
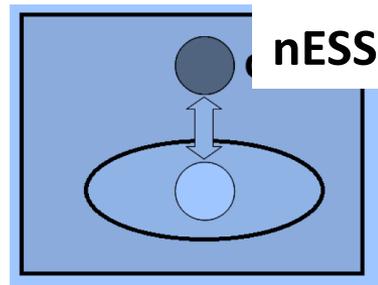
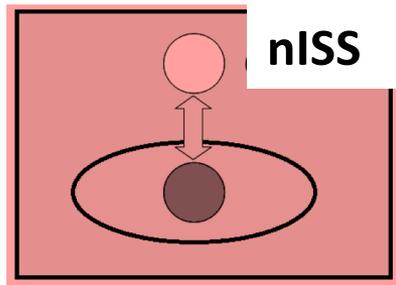


Elasticity coefficient Relative noise (CV)



Relation to mutual information:
$$M_{\max} = \log_2 \left(\frac{1}{\sqrt{2\pi e}} \int \frac{R(I)}{I} dI \right)$$

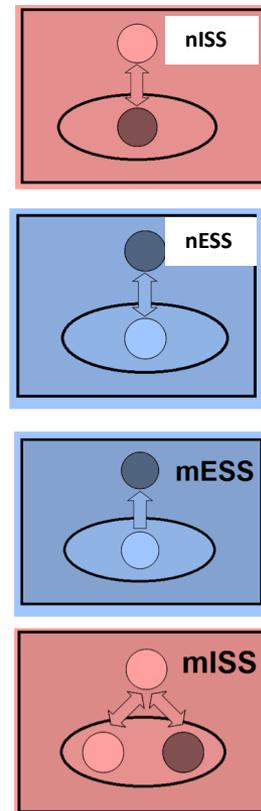
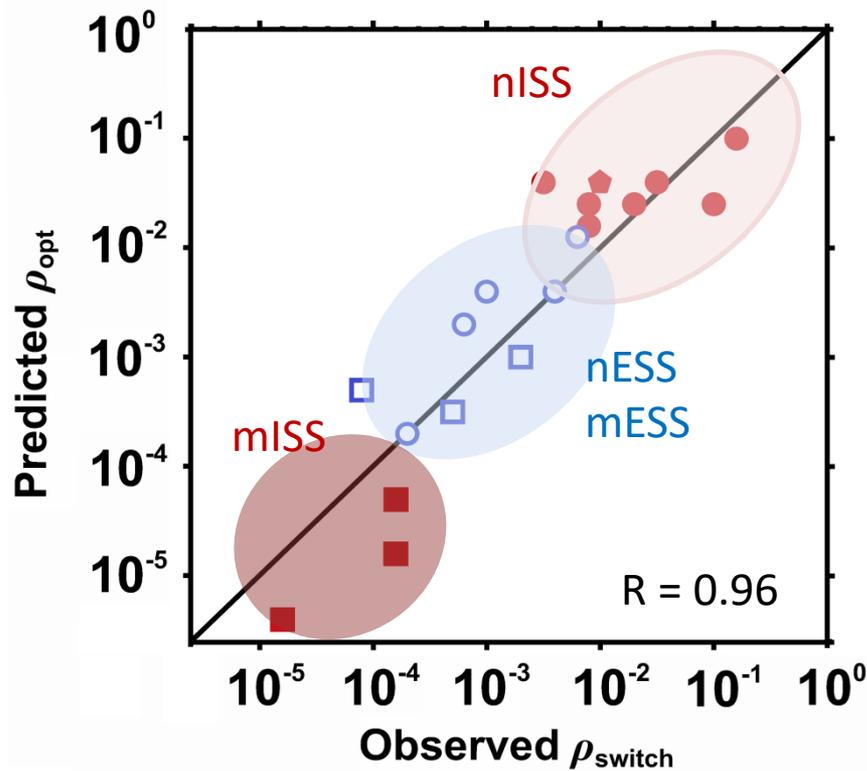
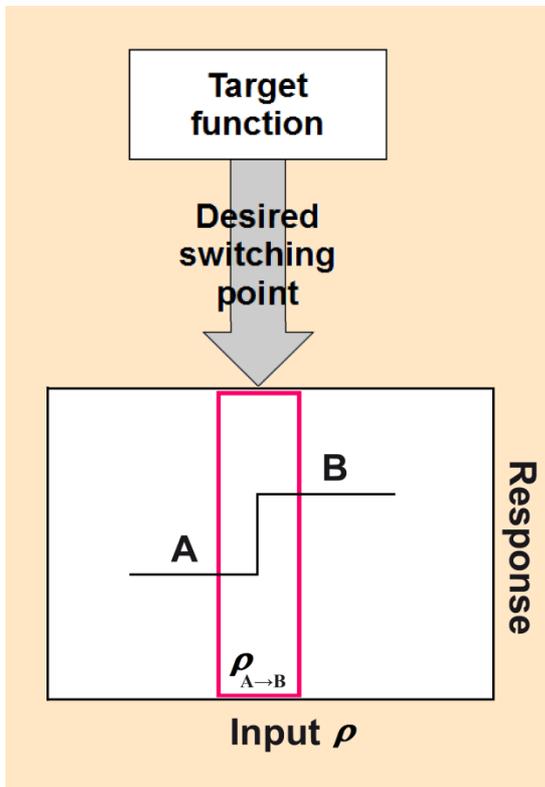
Quantifying Performance



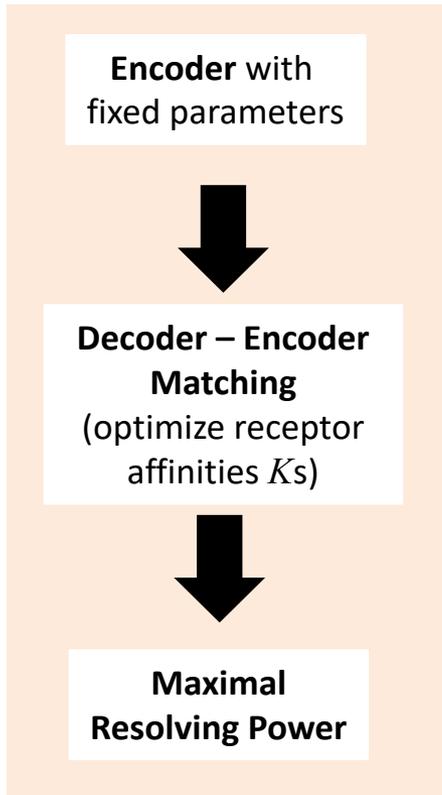
QS architectures have distinct sensitivity and noise characteristics.

QSS achieve optimal performance at a certain cell density input.

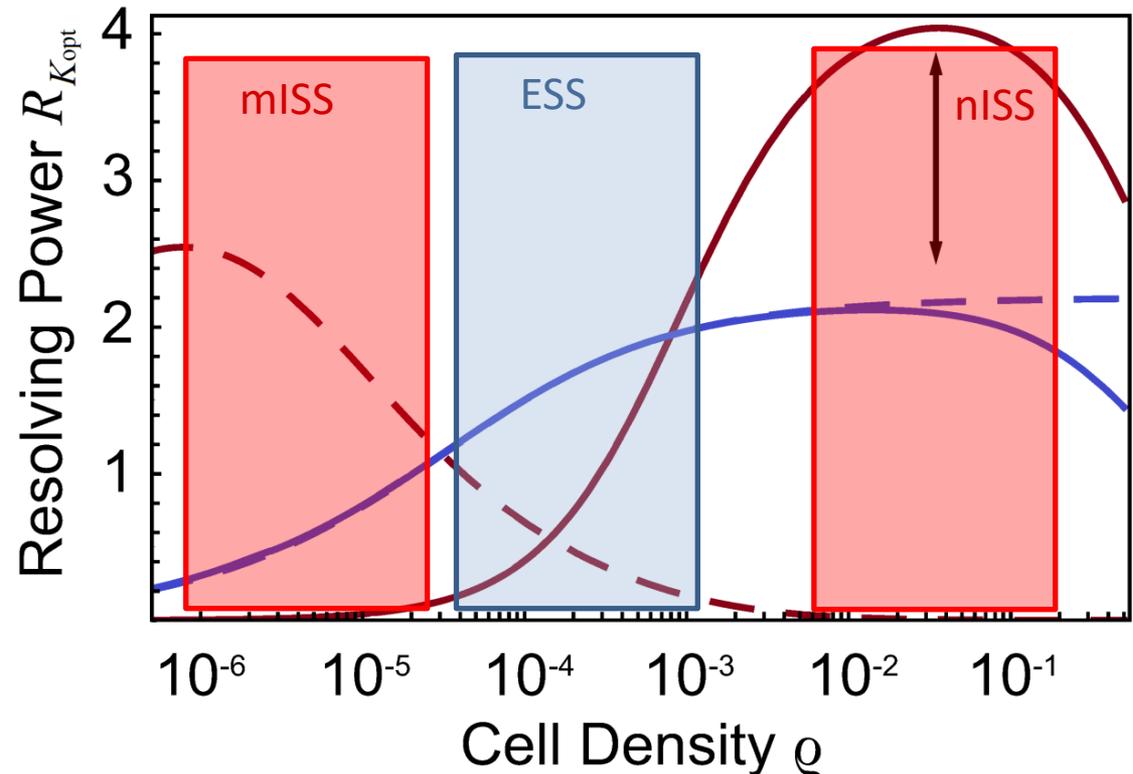
The optimal sensory performance is matched to the „quorum“



Performance Trade-offs



Resolving power with a „matched“ decoder.



Trade-off



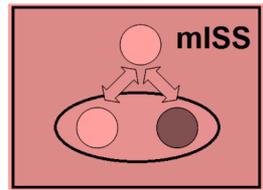
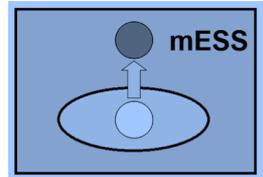
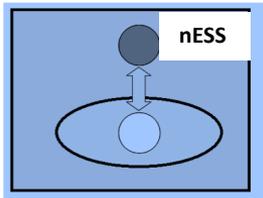
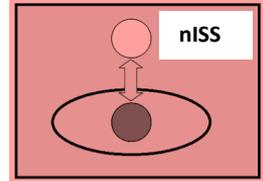
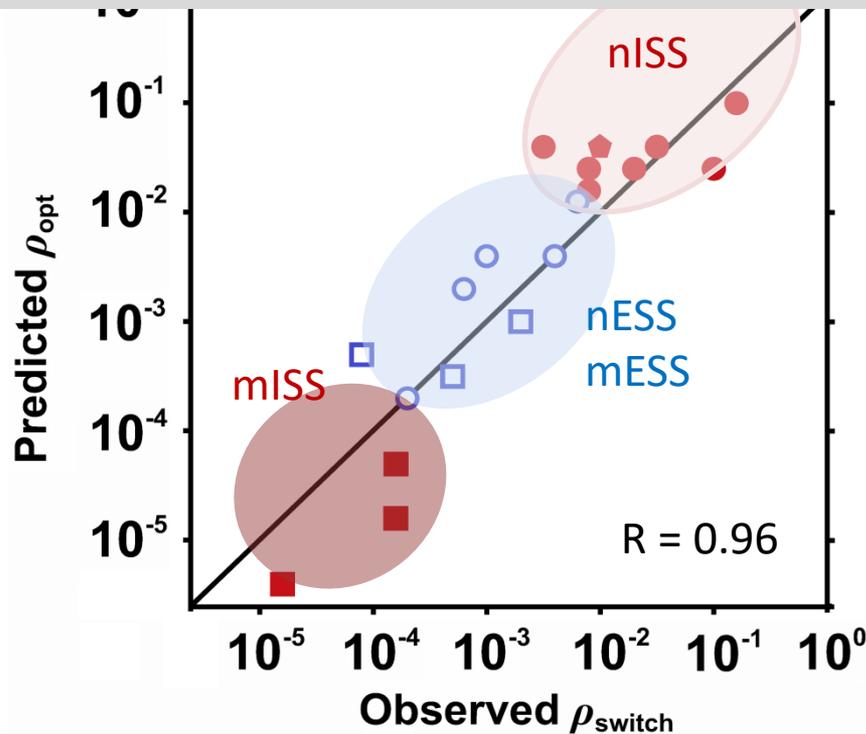
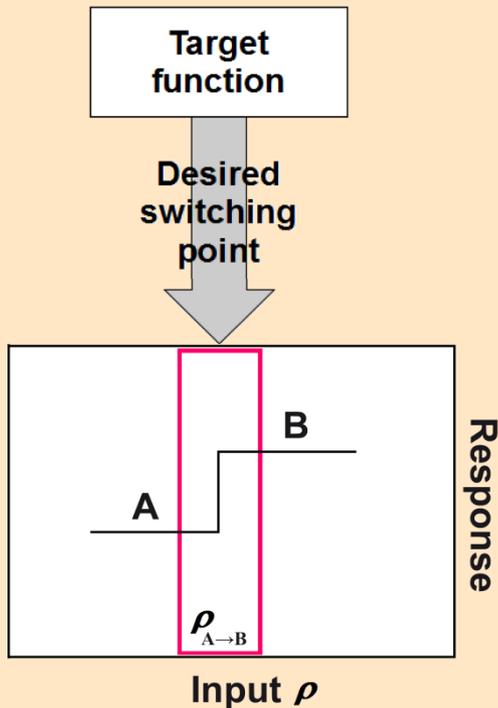
Architectural Complexity
(better sensitivity characteristics)
nESS

Architectural Simplicity
(better noise characteristics)
nISS

Conclusions

Design Principle for Synthetic Biology

Low cell density sensing may require other QSS architectures than the simple LuxIR-type system.



Evolution of QSS target genes:

Does the QSS architecture constrain the evolution of QSS target genes?

It's really hard to find parameters to model QSS ...

Calculations based on guestimates 23...

Traditional Quorum Sensing Research



The QS Paradigm



**Homogeneous
population**

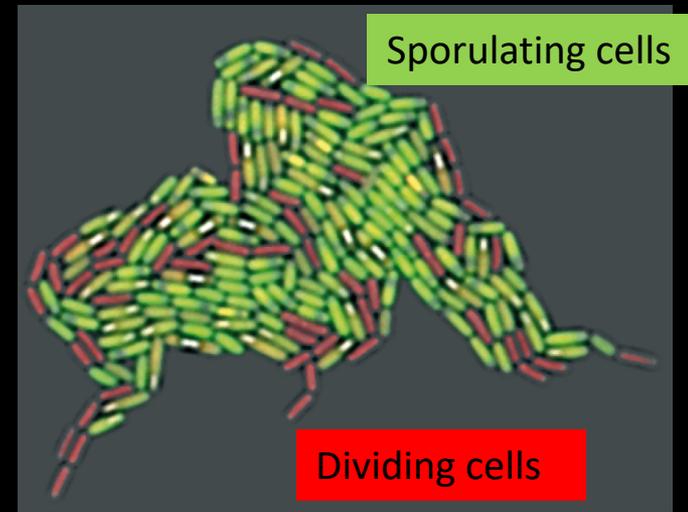
**Cell density
sensing**

**Synchronized
response**

Complex adaptive traits



Heterogeneous population



Distribution of different phenotypes

Division of labor

J. van Gestel et al.,
Plos Biology, 2015

Bet hedging

Reviewed in: Veening et al.,
Annu. Rev. Microbiol. 2008

Population-based quantitative trait

Our mission



**Molecular
Determinants &
Mechanisms**

**Complex
Adaptive Traits
(CATs)**

understand

Organizing Principles

**Molecular
Level**

**Population
Level**

engineer

control

Fluorescence timelapse microscopy



Stephanie Trauth



Marika Ziesack



Charlotte Kaspar

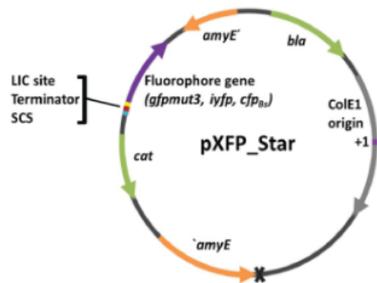


Svenja Schwindt



Group of K. Rohr

Genetic engineering



Fluorescence reporter

Trauth & Bischofs, 2014

Timelapse microscopy

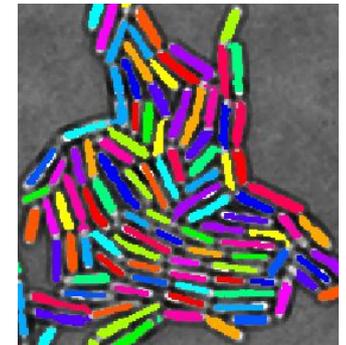
Hydrogels



Microfluidics

Group of D. Kohlheyer

Image Processing



Lilushvili V. et al., 2015

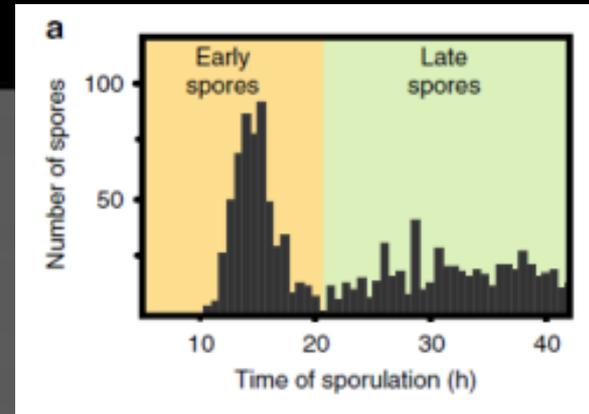
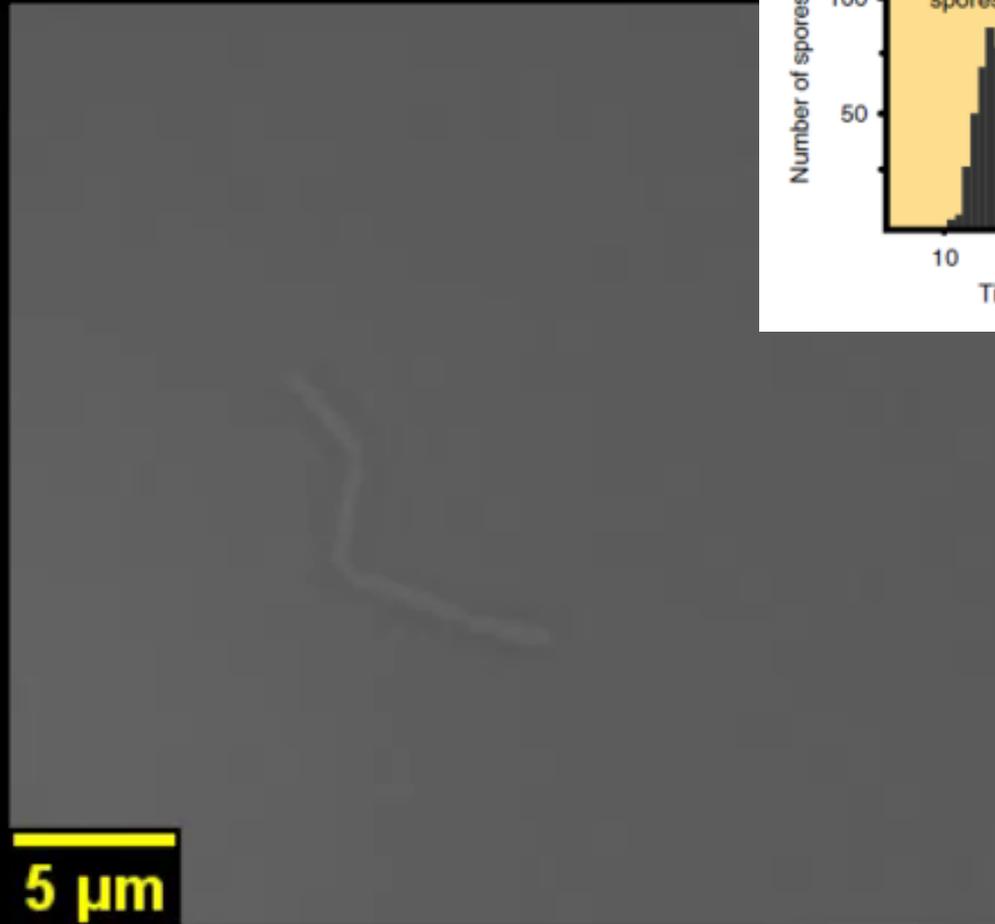
Heterochronic population response



04:00:00 h



Alper Mutlu



Going beyond the QS paradigm



**Homogeneous
population**

**Cell density
sensing**

**Synchronized
response**

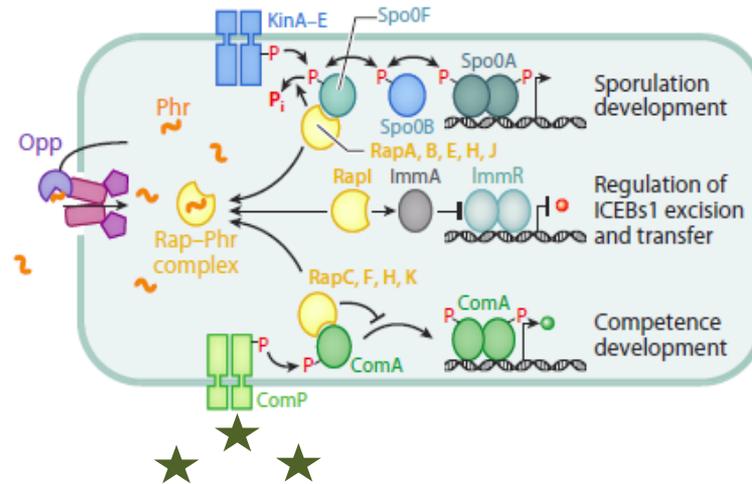


**Heterogeneous
population**

?

**Heterochronic
response**

Communication in *B. subtilis*



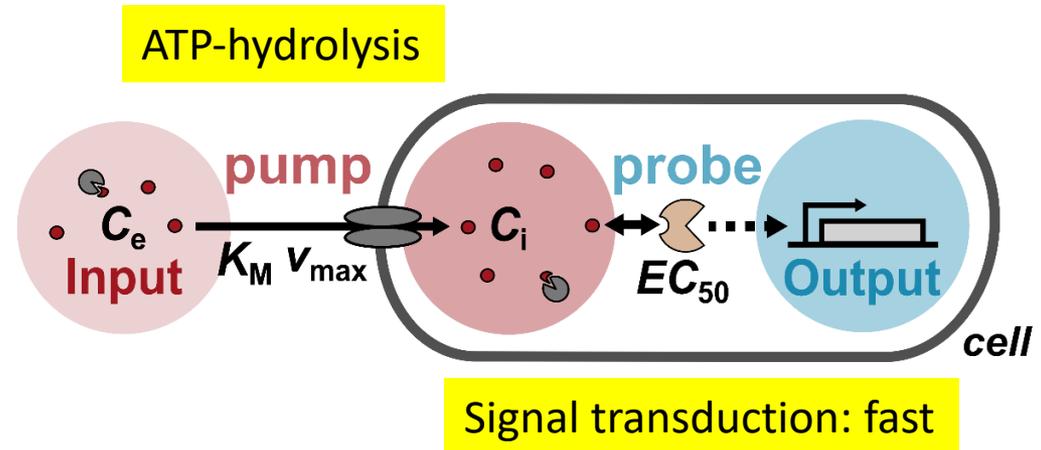
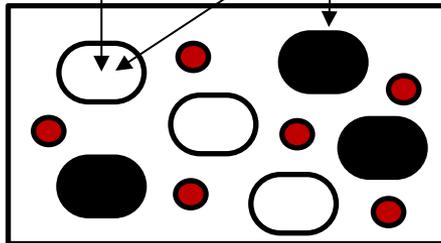
*Neiditch et al.,
Ann. Rev. Genetics (2017)*

Rap-Phr-Systems are commonly referred as **quorum sensing** systems.
However, there is little experimental evidence for a cell-density dependent type of regulation.

The Pump-Probe Model

Phenomenological (ODE-type) model

Signal producer Signal uptake



Assumptions

Population model

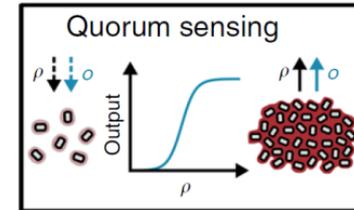
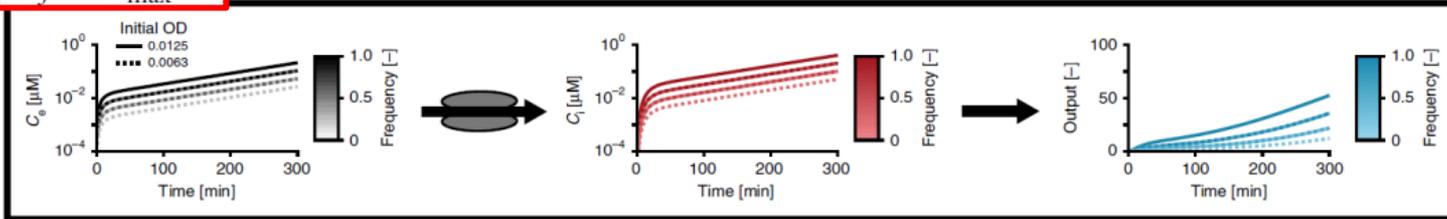
- Heterogeneous population: Signal produced by a subpopulation of cells present at frequency f
- Signal uptake by all cells.
- Exponentially growing population (same growth rate & well-mixed)
- Different starting cell densities (OD)

Signal transduction model

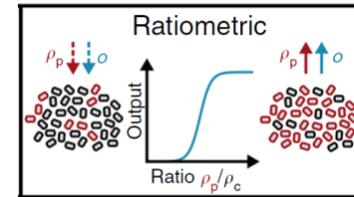
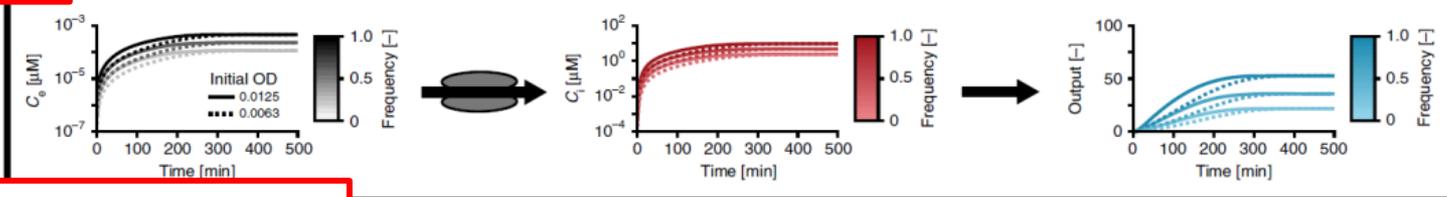
- Signal uptake with Michaelis-Menten pump kinetics
- Signal degradation (and dilution by cell growth)
- Output is a function of the intracellular signal concentration using a Hill function

One network architecture – different control functions

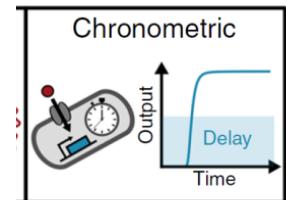
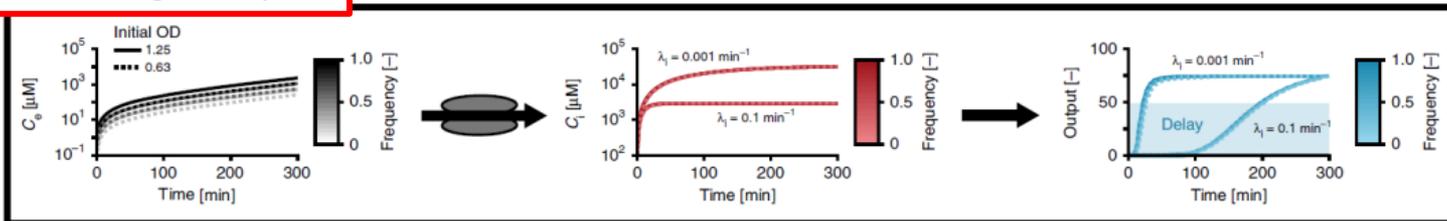
$$\pi_{\text{eff}} = f\pi - v_{\text{max}} > 0$$



$$\pi_{\text{eff}} < 0$$



Saturated signal import

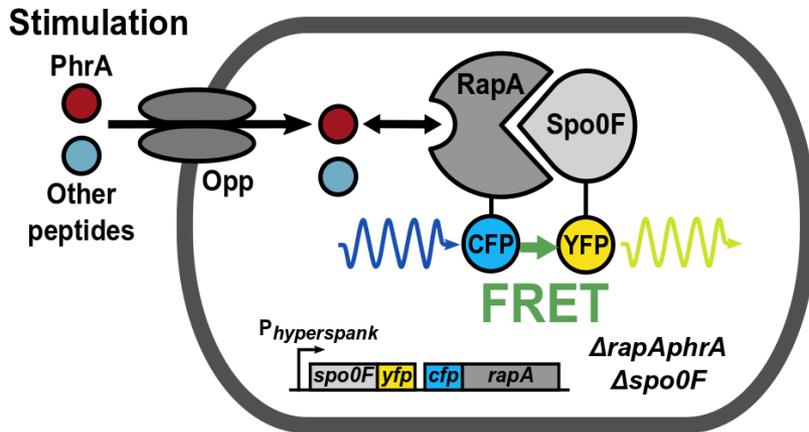


Control functions depend on network parameters and operating conditions

not (well) defined

Fluorescence (Förster) resonance energy transfer (FRET)

A new tool for quantitative studies of PhrA-signaling in bacterial cells



Tool for studying
protein-protein interactions
in vivo



(relatively) **short-ranged** interaction

Förster radius:
 $R_0 \sim \text{nm}$



Energy transfer rate:

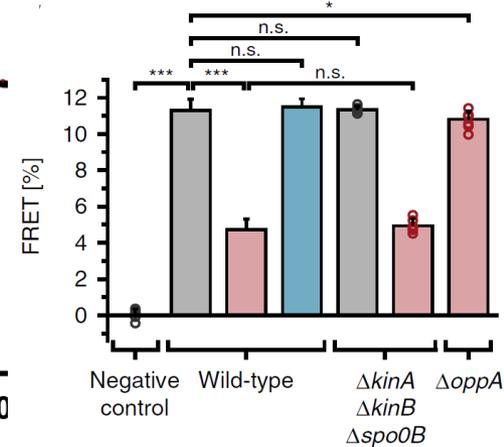
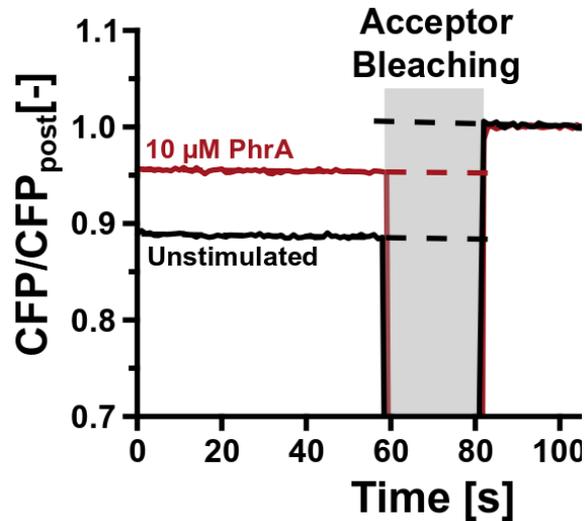
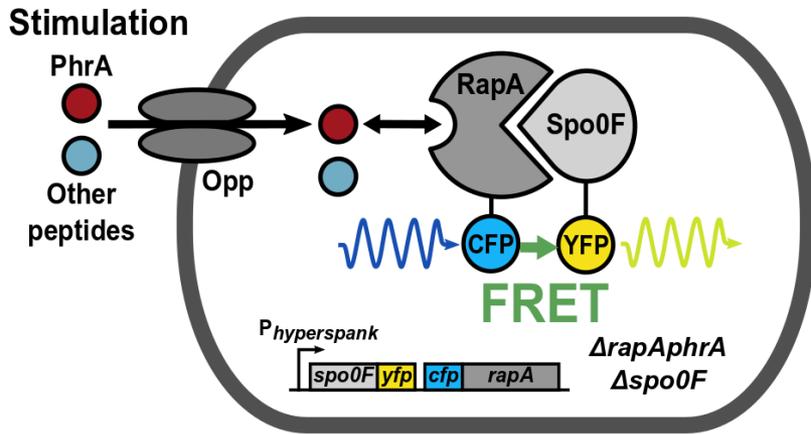
$$k_T = 1/\tau_D * \left(\frac{R_0}{R}\right)^6$$

τ_D : Fluorescence lifetime of the donor

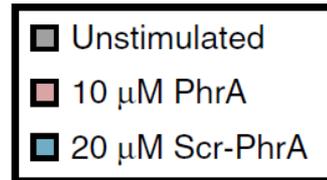
R: Distance between chromophores

Acceptor-photobleaching experiments

FRET efficiency:
$$\text{FRET} = \frac{\text{CFP}_{\text{post}} - \text{CFP}_{\text{pre}}}{\text{CFP}_{\text{post}}} \cdot 100\%$$



Population-average measurement



FRET specifically reports on PhrA

FRET controls



Molecular crowding controls



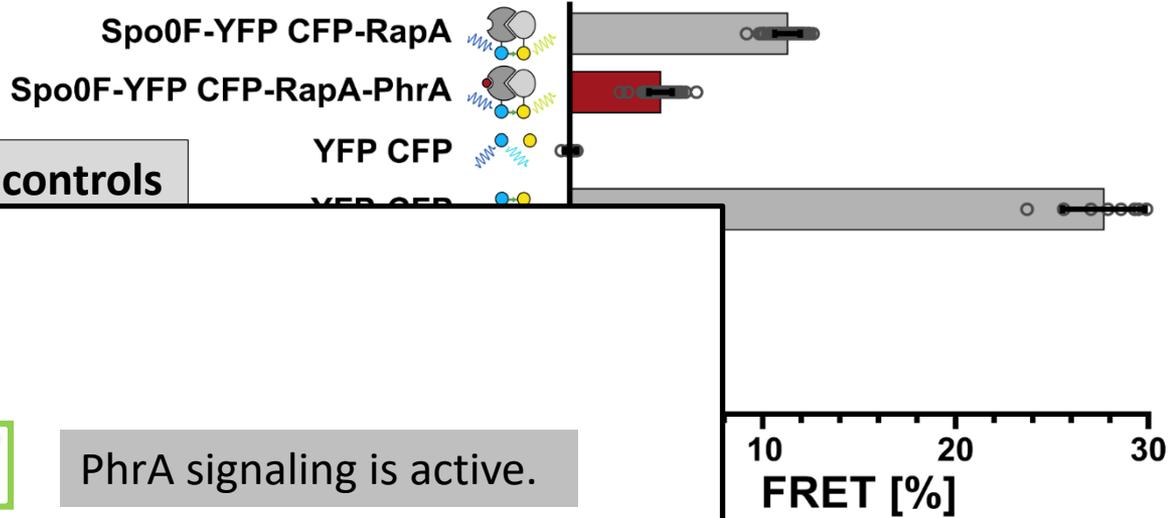
Bleed



Valida



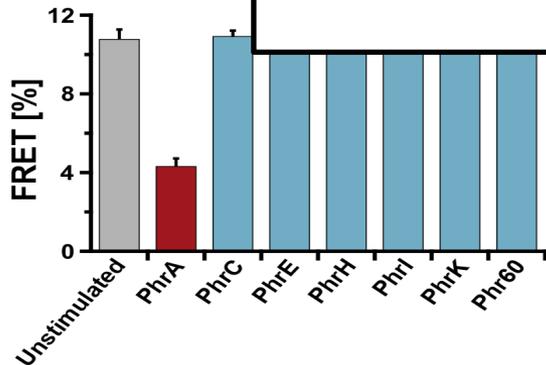
PhrA signaling is active.



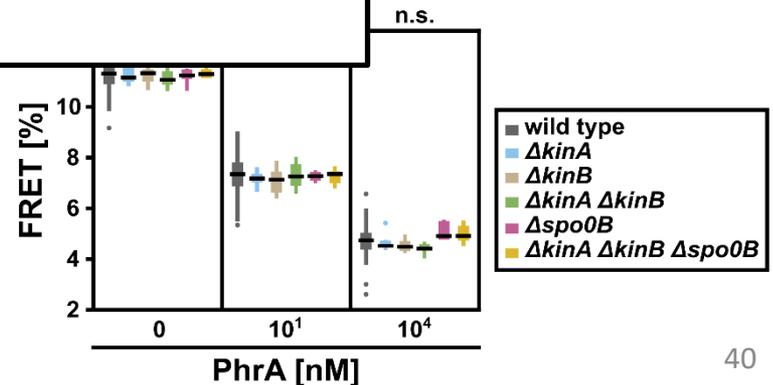
Reporter controls



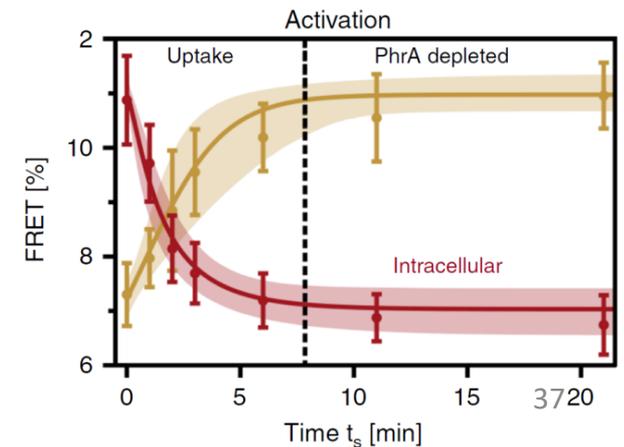
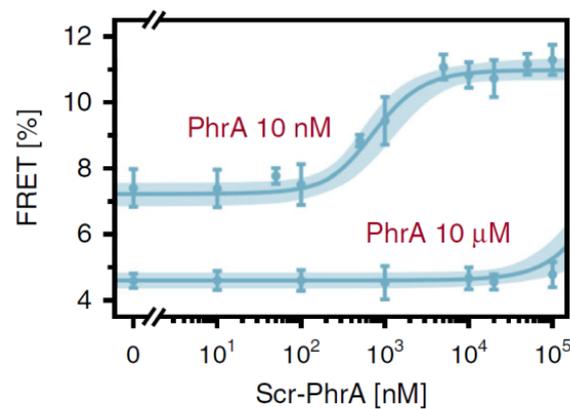
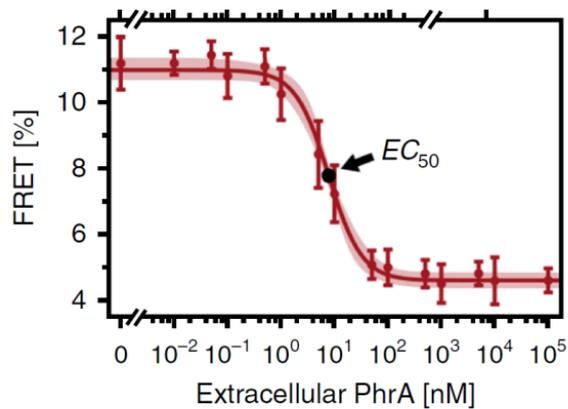
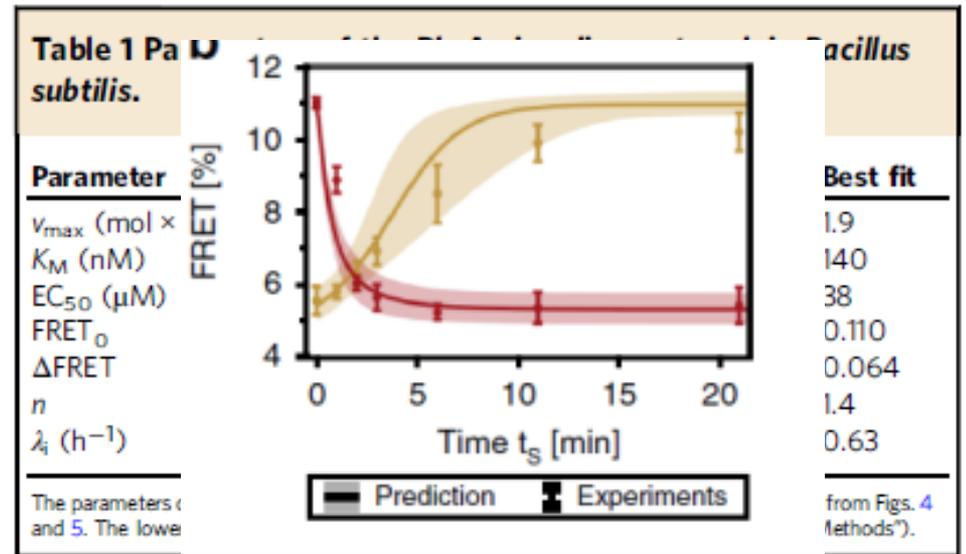
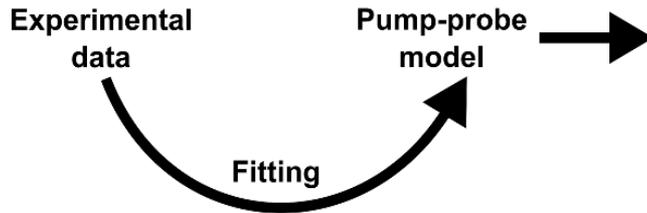
No cross-talk



phosphorelay signaling



PhrA processing is well described by the pump-probe model



Cells accumulate PhrA and trigger a response at μM -levels

PhrA is „integrated“ over a relatively long time.

$$\tau = 1/(\mu + \lambda_i) \sim 50 \text{ min}$$

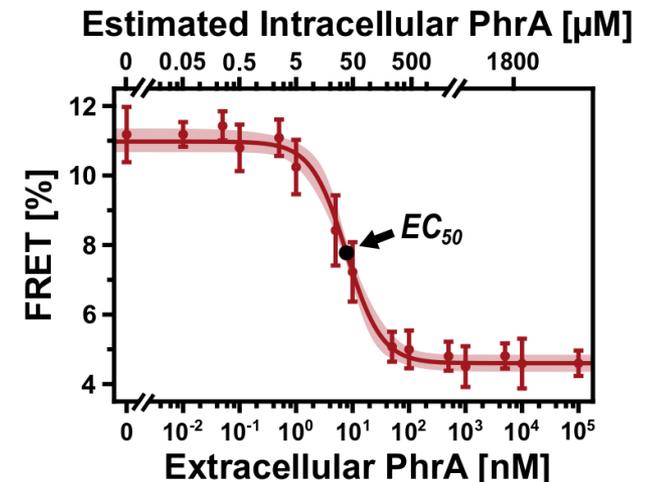
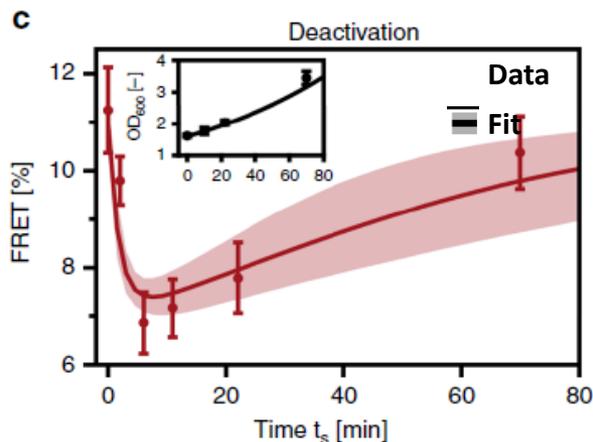
Model: Large signal amplification upon extra- to intracellular conversion.

Intracellular FRET response is triggered in the μM -regime.

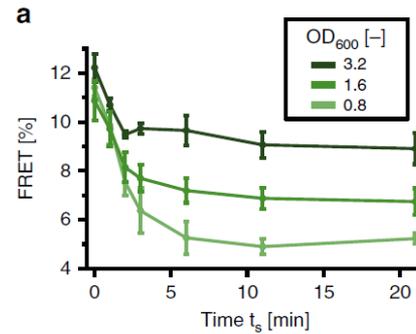
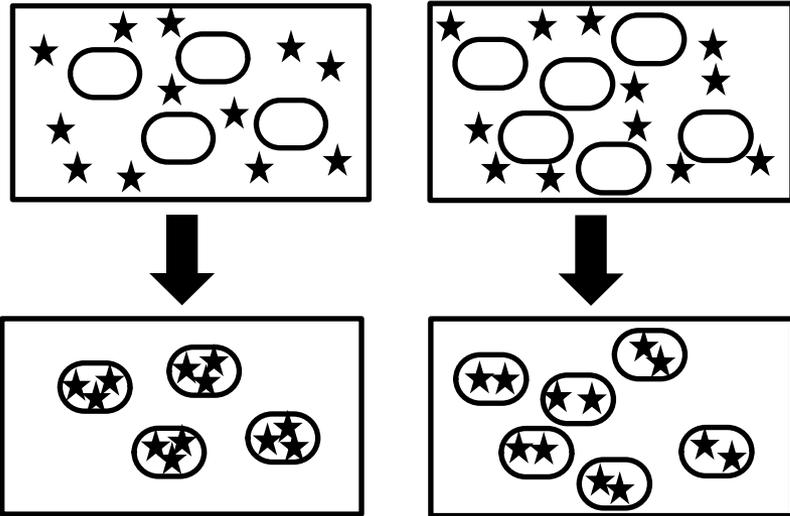
THE PROBE

Dynamics of recovery

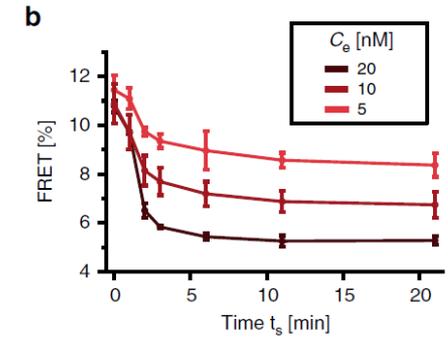
Intracellular FRET-response



Dose-dependent signal processing

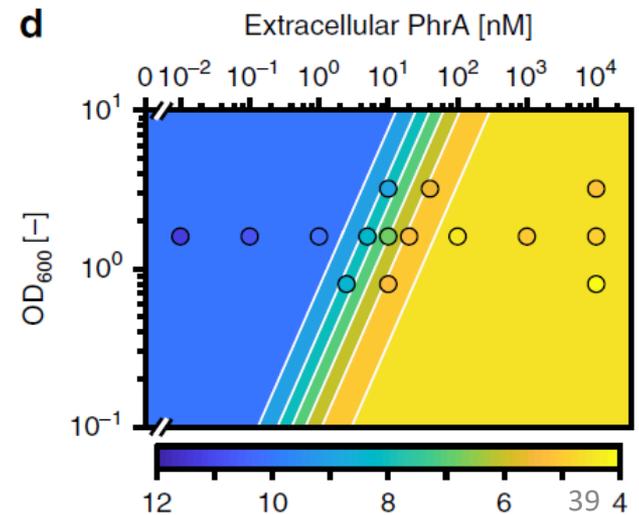
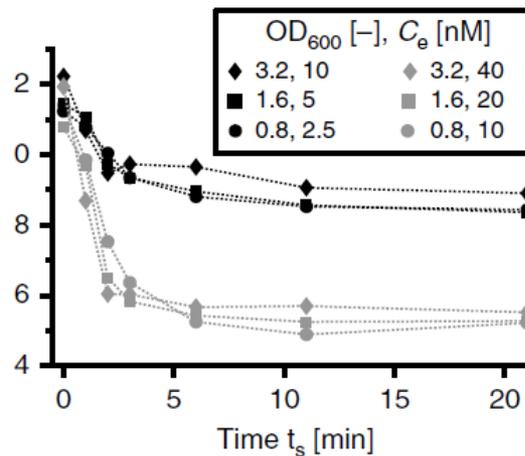


CELL DENSITY EFFECT

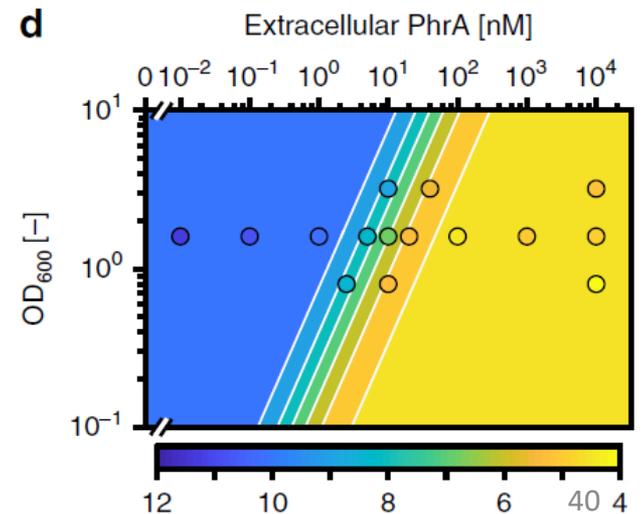
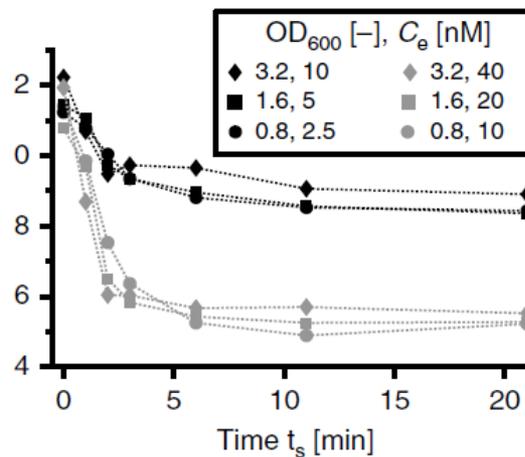
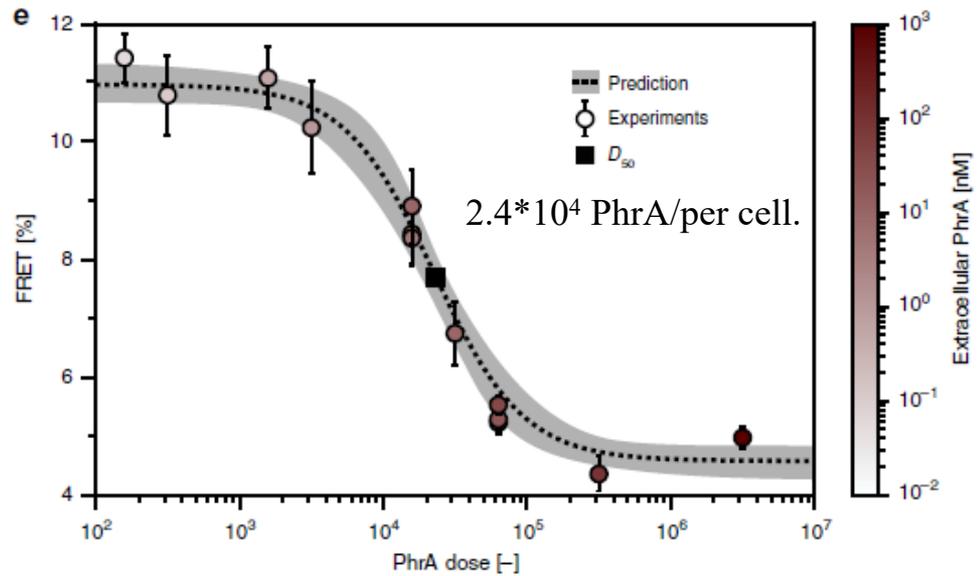
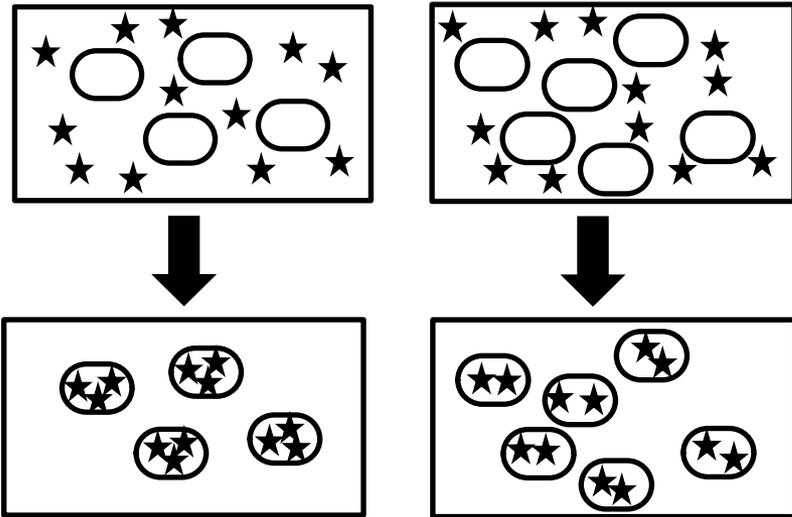


CONCENTRATION EFFECT

DOSE-DEPENDENT EFFECT

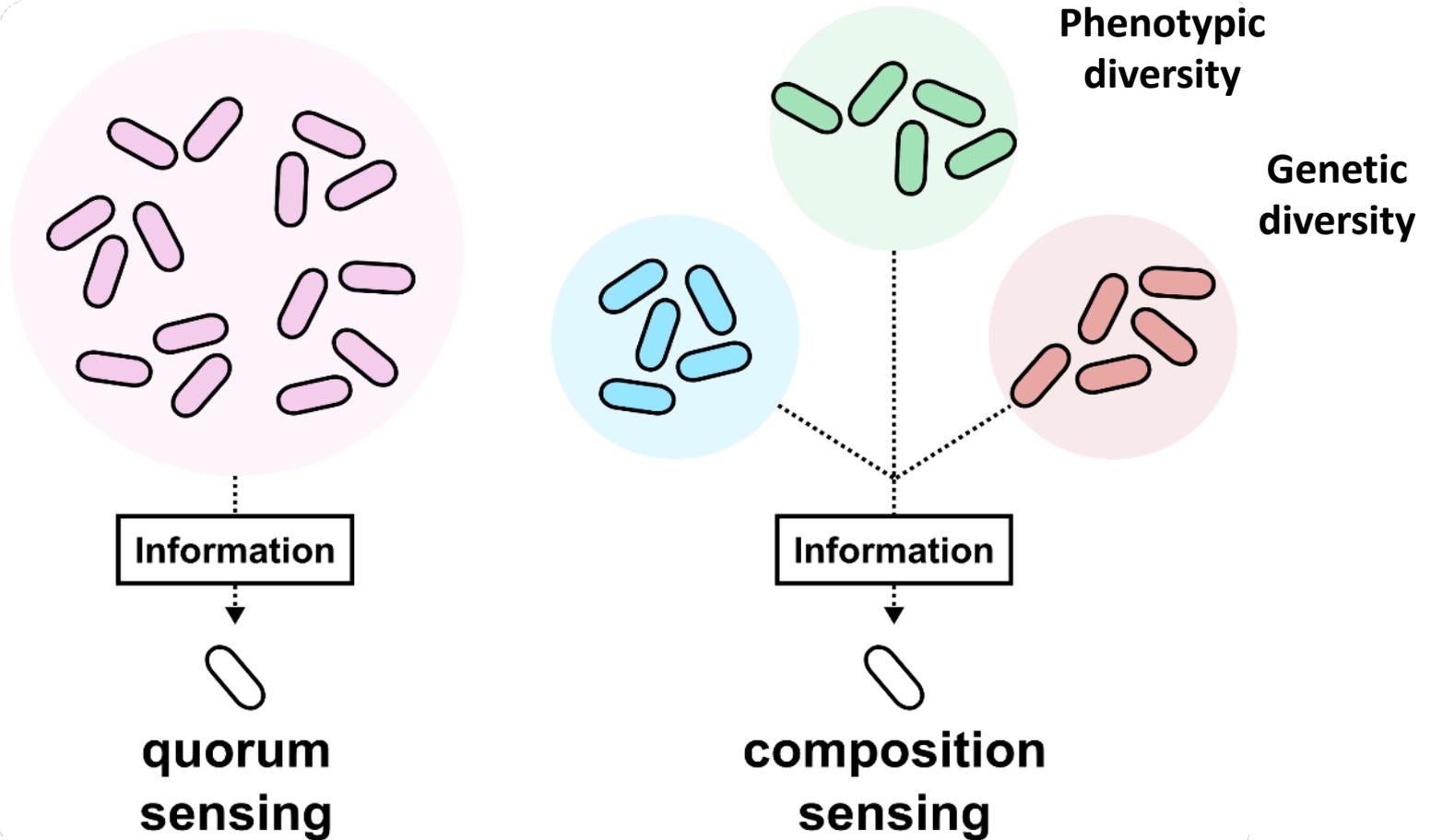


Dose-dependent signal processing

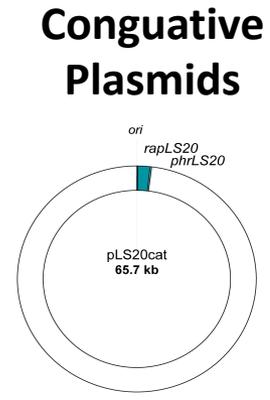
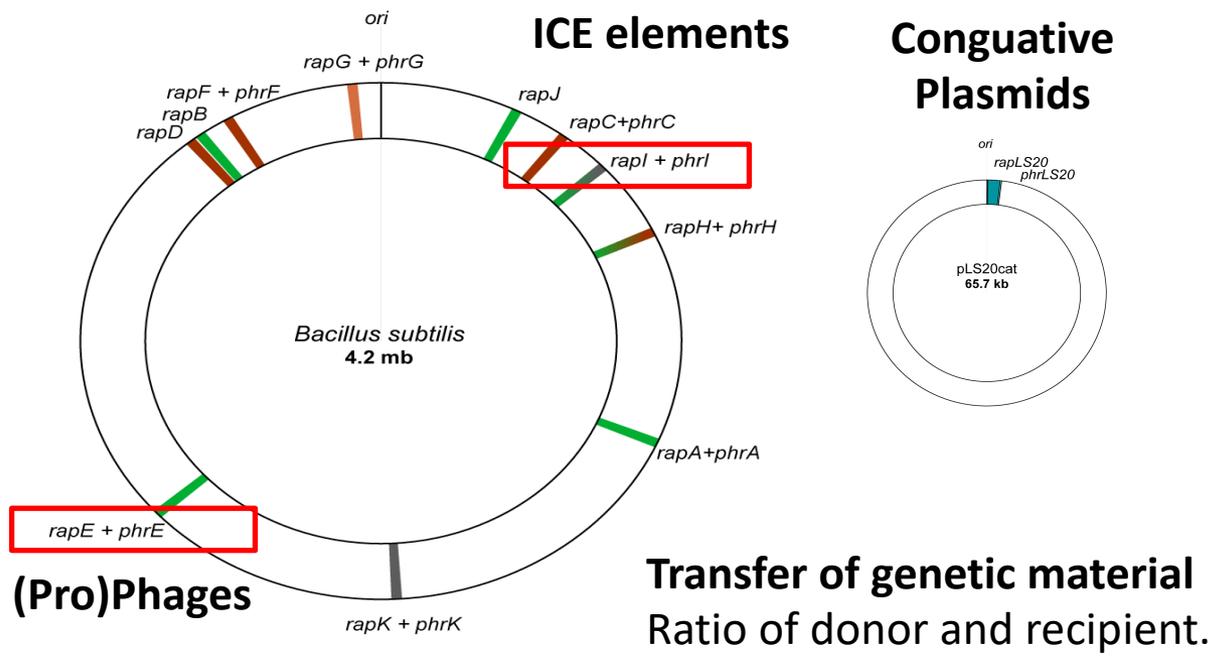


**DOSE-DEPENDENT
EFFECT**

Composition Sensing



PP Systems: Frequency-dependent regulation

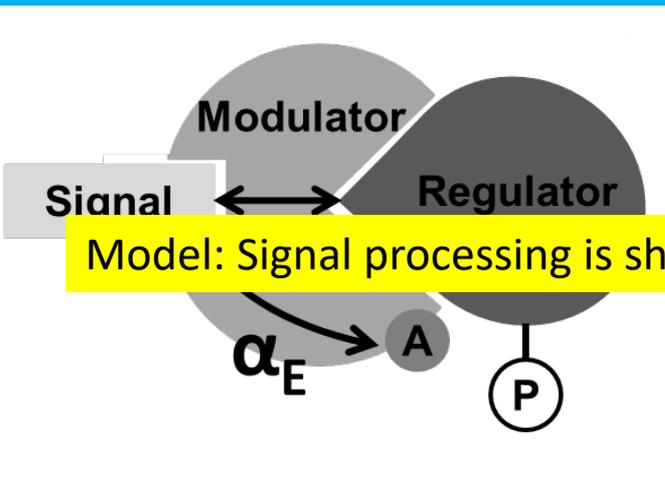


Frequency dependent investments into mating in yeast (Banderas et al., 2016)
Conjugation in *Enterococcus faecalis*, Banderas et al. , BioArXiv 2019.
See also upcoming work by Avigdor Eldar on phages (Tel Aviv University).

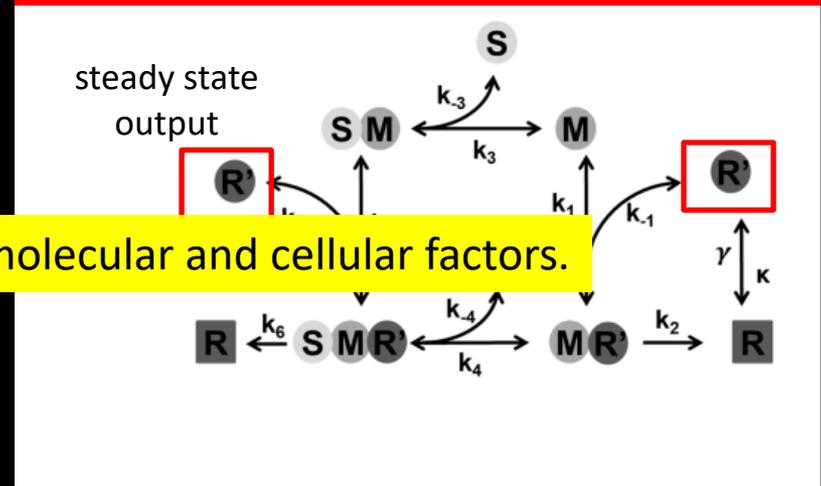
Future work

The Enzymatic Modulator Model

Molecular view



Network motif



Molecular Parameters

Binding Allostery

$$\alpha_B = \frac{K'_M}{K_M}$$

Enzymatic Allostery

$$\alpha_E = \frac{k_6}{k_2}$$

Relative affinity

$$\beta_E = \frac{K_3}{K_M} = \frac{K_4}{K'_M}$$

Cellular Parameters („Context“)

$$\mu = \frac{[M_T]}{[R_T]}$$

$$\kappa_r = \frac{\kappa}{\gamma + \kappa}$$

$$l = \frac{k_2}{\gamma + \kappa}$$

$$K_r = \frac{K_M}{[R_T]}$$

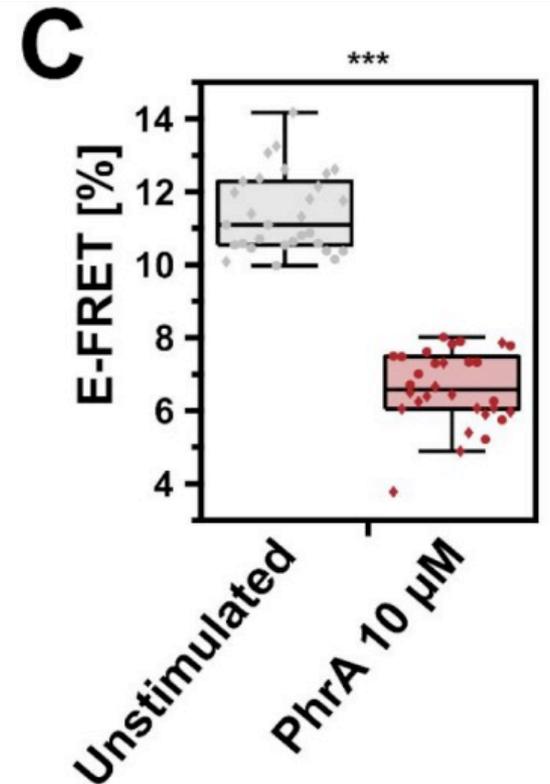
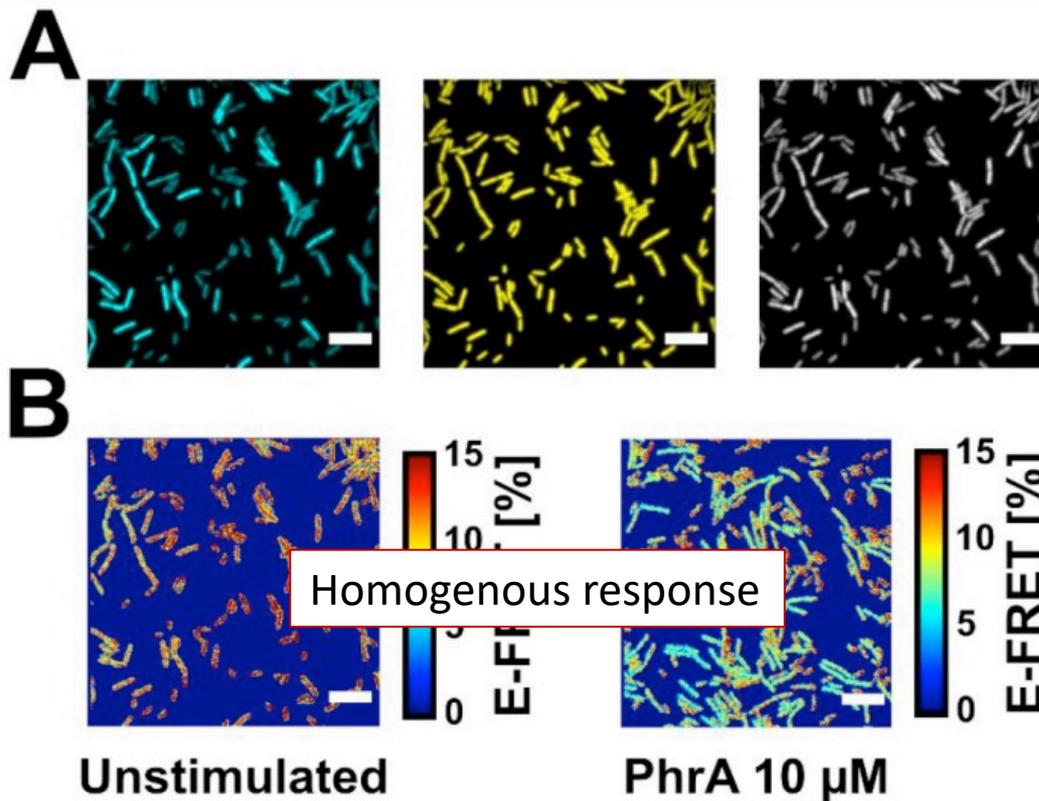
Signal processing by single cells

E-FRET Imaging (Zal & Gascoigne, 2004)

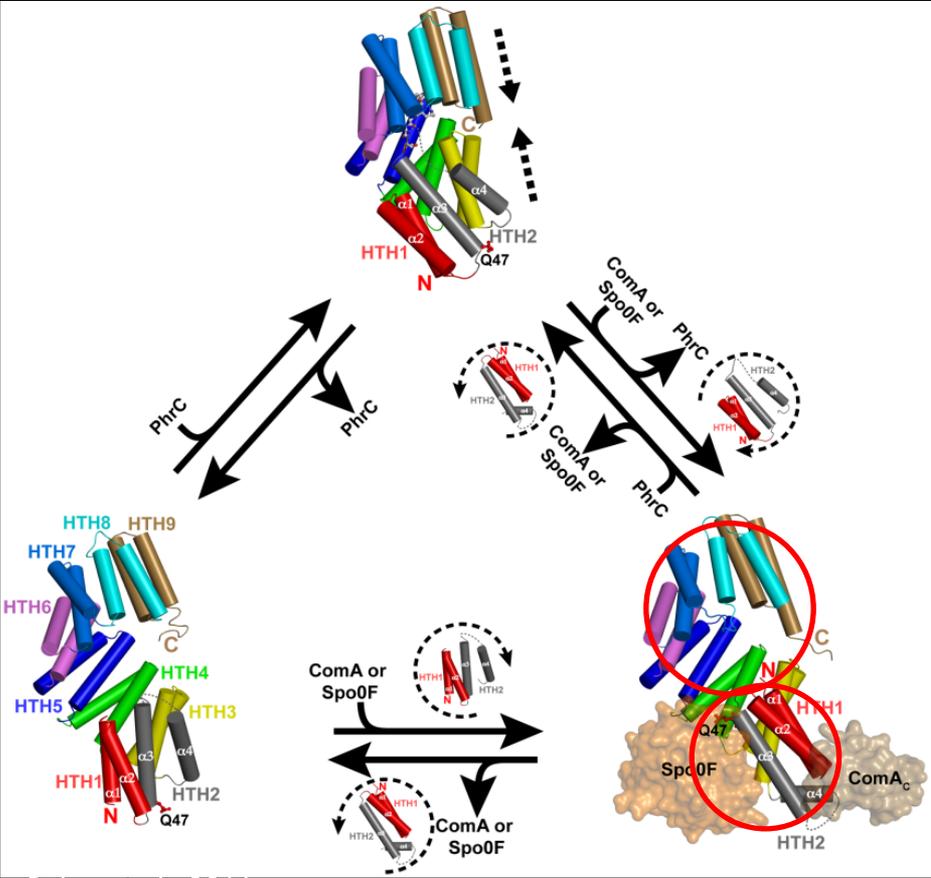
$$E_{\text{app}} = \frac{I_{\text{DA}} - aI_{\text{AA}} - dI_{\text{DD}}}{I_{\text{DA}} - aI_{\text{AA}} + (G - d)I_{\text{DD}}}$$

abcd: correction factors

G: calibration factor

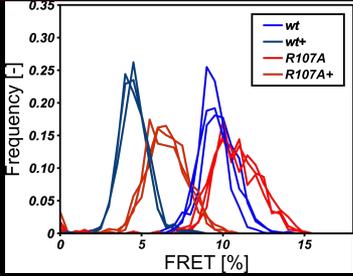


Signal processing by engineered receptors

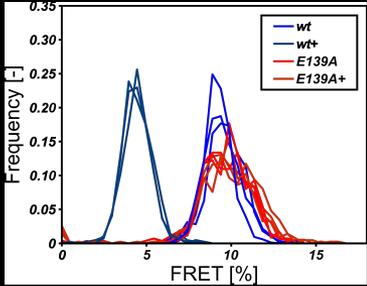


Baker et al., 2011
 Parashar et al., 2011
 Gallego et al., 2013
 Parashar et al., 2013

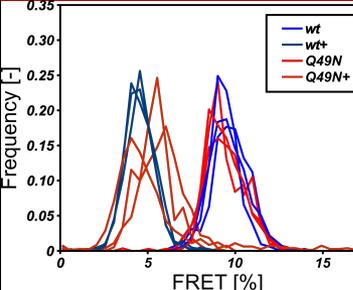
Allosteric residue



Allosteric residue



Catalytic residue



Summary

One sensory function – different architectures

The architecture of QSS could constrain the operating regime for cell density sensing.

One architecture – different sensory functions

Pump-probe networks could implement versatile control functions, including the ability for composition sensing and frequency-dependent regulation.

FRET is a powerful way to quantitatively interrogate signal processing in bacteria by monitoring protein-protein interactions in the cell.

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