

# Dynamical Allocation of Cellular Resources as an Optimal Control Problem: Novel Insights into Microbial Growth Strategies

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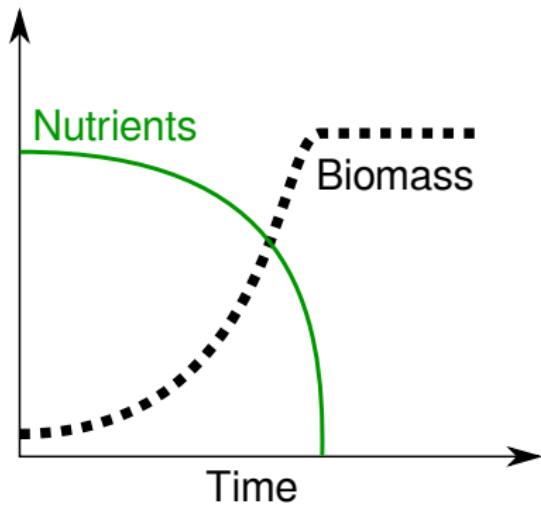
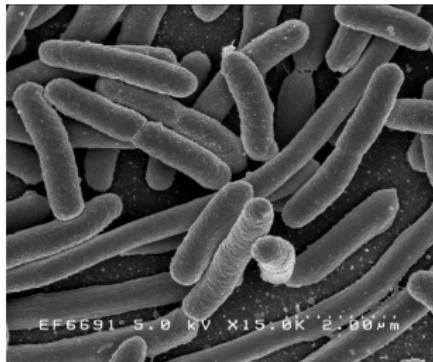
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October, 21th 2015

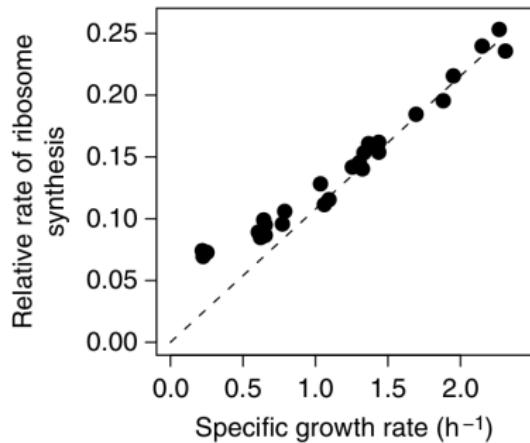
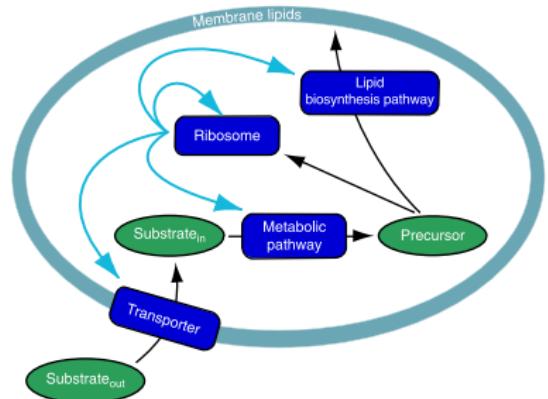
# MICROBIAL GROWTH



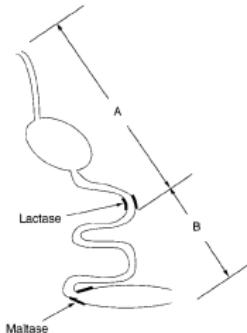
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Source picture: NIAID

# MOLECULAR COMPOSITION OF A MICROORGANISM



# DO MICROORGANISMS LIVE IN CONSTANT ENVIRONMENTS?



Mostly, not.

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Savageau (1998), Am. Natural., 122(6):732-44  
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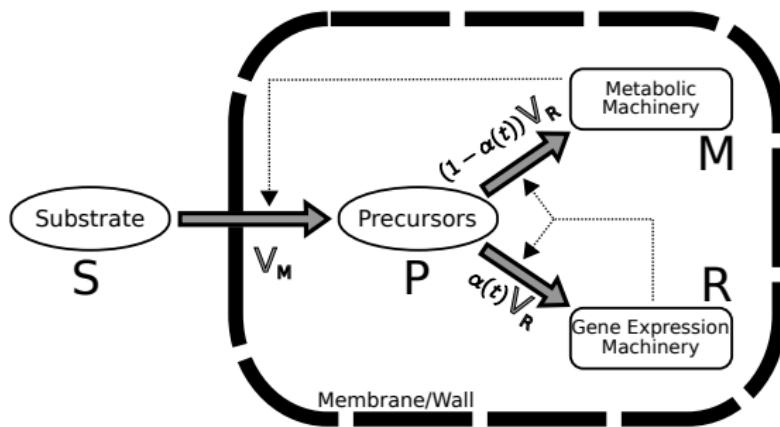
# OUR PROJECT: A DYNAMICAL PERSPECTIVE ON GROWTH CONTROL STRATEGIES

- ▶ Is considering balanced-growth a critical assumption to understand growth control strategies?
- ▶ Can we gain additional information by extending growth rate studies to dynamical environments?

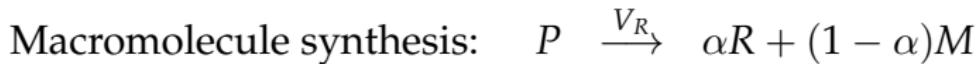
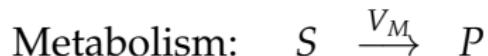
Tools:

- ▶ A simple model of resource allocation
- ▶ Optimal control theory
- ▶ Fluorescent reporters of gene expression (experiments)

# SELF-REPLICATOR MODEL OF RESOURCE ALLOCATION



Two biochemical (macro)reactions:



# TWO-DIMENSIONAL DYNAMICAL SYSTEM

Volume:  $V_{\text{ol}} = \beta(M + R) \Rightarrow$  Growth rate:  $\mu = \beta \frac{V_R}{V_{\text{ol}}} = \beta v_R$

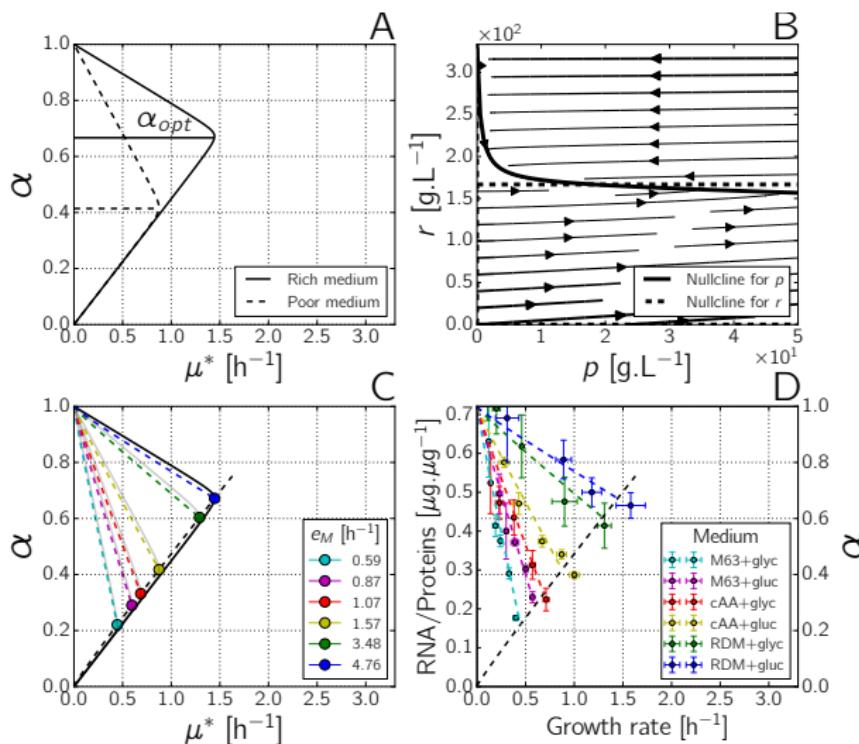
Michaelis-Menten kinetics  $\Rightarrow v_R = \frac{k_R \cdot p}{K_R + p} \cdot r$

Model with concentration variables (dimensionless):

Precursors:  $\frac{d\hat{p}}{dt} = E_M \cdot (1 - \hat{r}) - \frac{\hat{p}}{K + \hat{p}} \cdot \hat{r} \cdot (1 + \hat{p})$

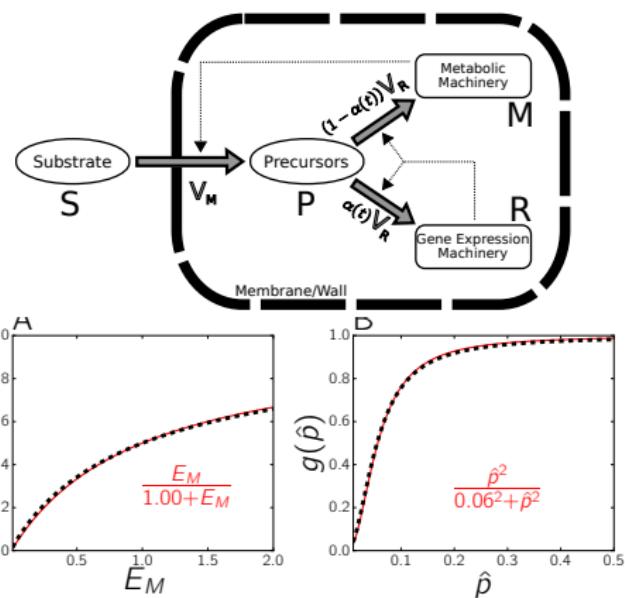
GEM:  $\frac{d\hat{r}}{dt} = \frac{\hat{p}}{K + \hat{p}} \cdot \hat{r} \cdot (\alpha - \hat{r})$

# MODEL PREDICTS THE STEADY-STATE GROWTH LAWS



Giordano *et al*, in preparation; from data in Scott *et al*, Science, 2010

# ALTERNATIVE CONTROL STRATEGIES FOR OPTIMAL RESOURCE ALLOCATION



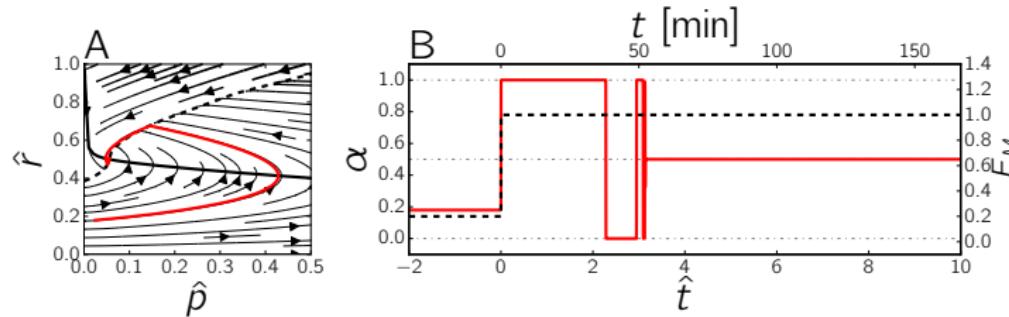
The two strategies are equivalent for steady-state growth!

# WHAT IF WE OPTIMIZE DURING A GROWTH TRANSITION?

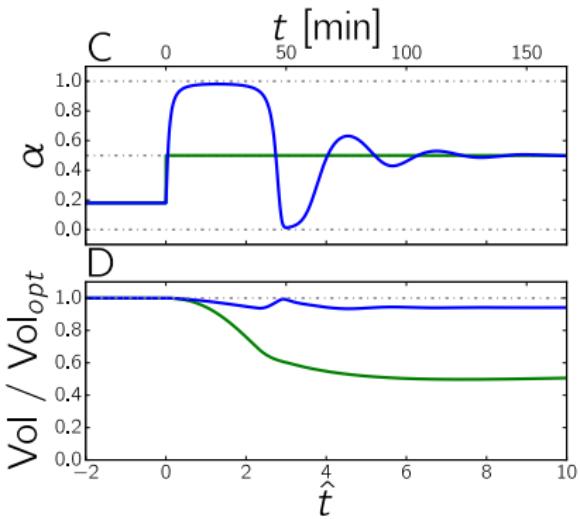
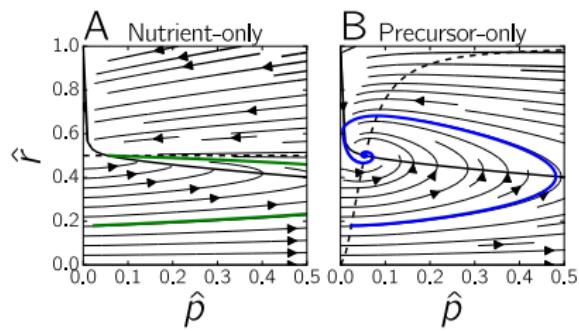
New objective: maximize biomass during a transition (upshift at  $t = 0$ )

$$J(\alpha) = \int_0^{\tau} \mu(t, \hat{p}, \hat{r}, \alpha) dt$$

Optimal solution: bang-bang-singular regulatory strategy

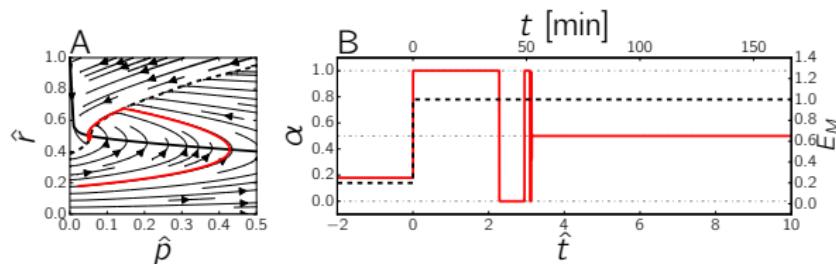


# PERFORMANCE OF CONTROL STRATEGIES DURING GROWTH TRANSITION

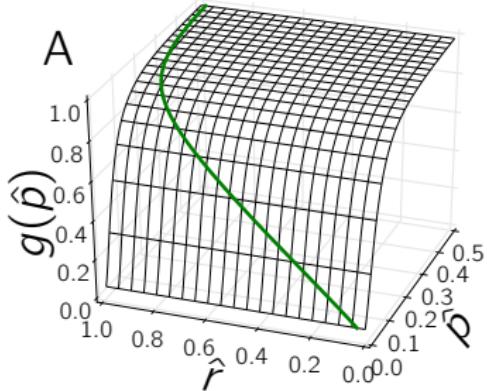


Control strategies are no longer equivalent in dynamic environment.

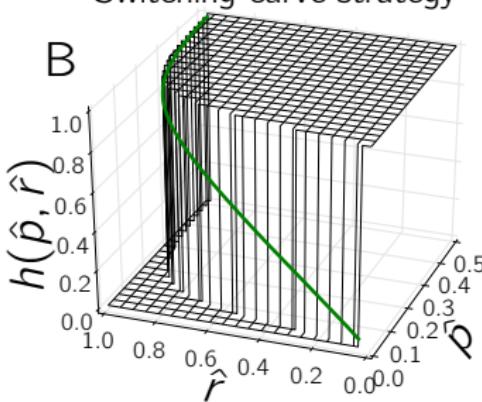
# AND IF WE CAN MEASURE SEVERAL VARIABLES?



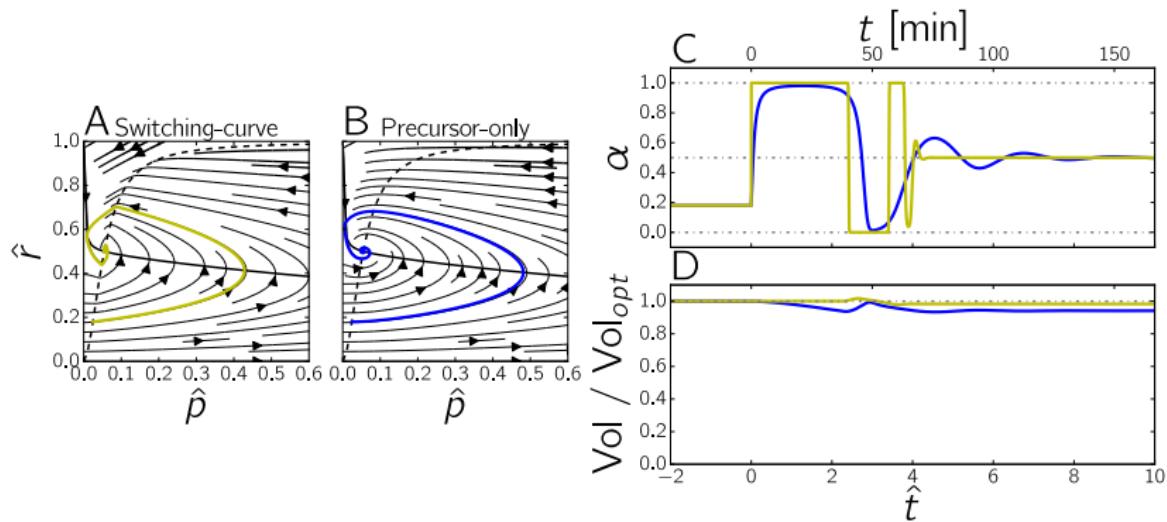
Precursor-only strategy



Switching-curve strategy

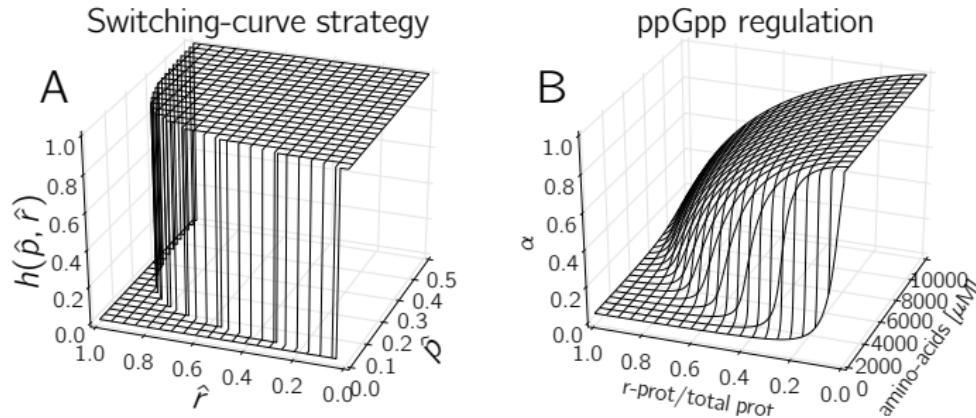


# IS A STRATEGY MEASURING TWO VARIABLES BETTER?



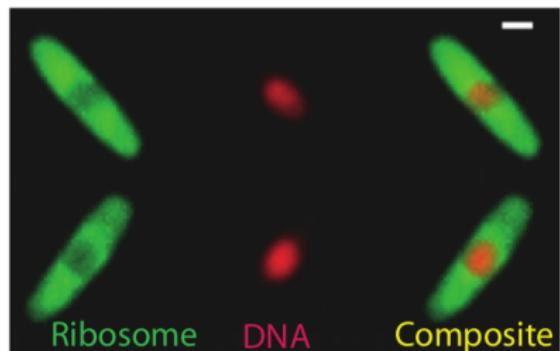
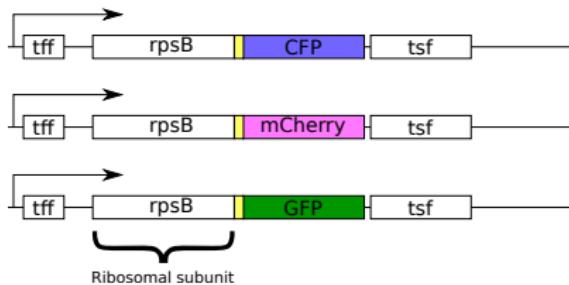
# DOES THE STRATEGY CORRESPOND TO ACTUAL REGULATORY MECHANISMS?

If we take a model of the ppGpp regulatory system in *E. coli* (Bosdriesz *et al*, 2015)...



... we obtain a likely candidate.

# EXPERIMENTAL VALIDATION: OBSERVING THE DYNAMICS OF $\alpha$ IN BACTERIAL CELLS



# CONCLUSION

- ▶ Is considering balanced-growth a critical assumption to understand growth control strategies?
  - ▶ Yes, because strategies are equivalent at steady state
- ▶ Can we gain additional information by extending growth rate studies to dynamical environments?
  - ▶ Yes, because they become distinguishable in dynamic conditions
  - ▶ Complex strategies are beneficial during growth transitions
  - ▶ The widespread ppGpp system might actually be a simple way for the cell to gain information on several variables

# PERSPECTIVE

- ▶ Can we observe experimentally an oscillatory pattern of ribosome synthesis during transitions?
- ▶ Is there a fundamental relation between environment dynamics and complexity of regulations?
- ▶ Can we apply this approach to maximize industrial production yields?

