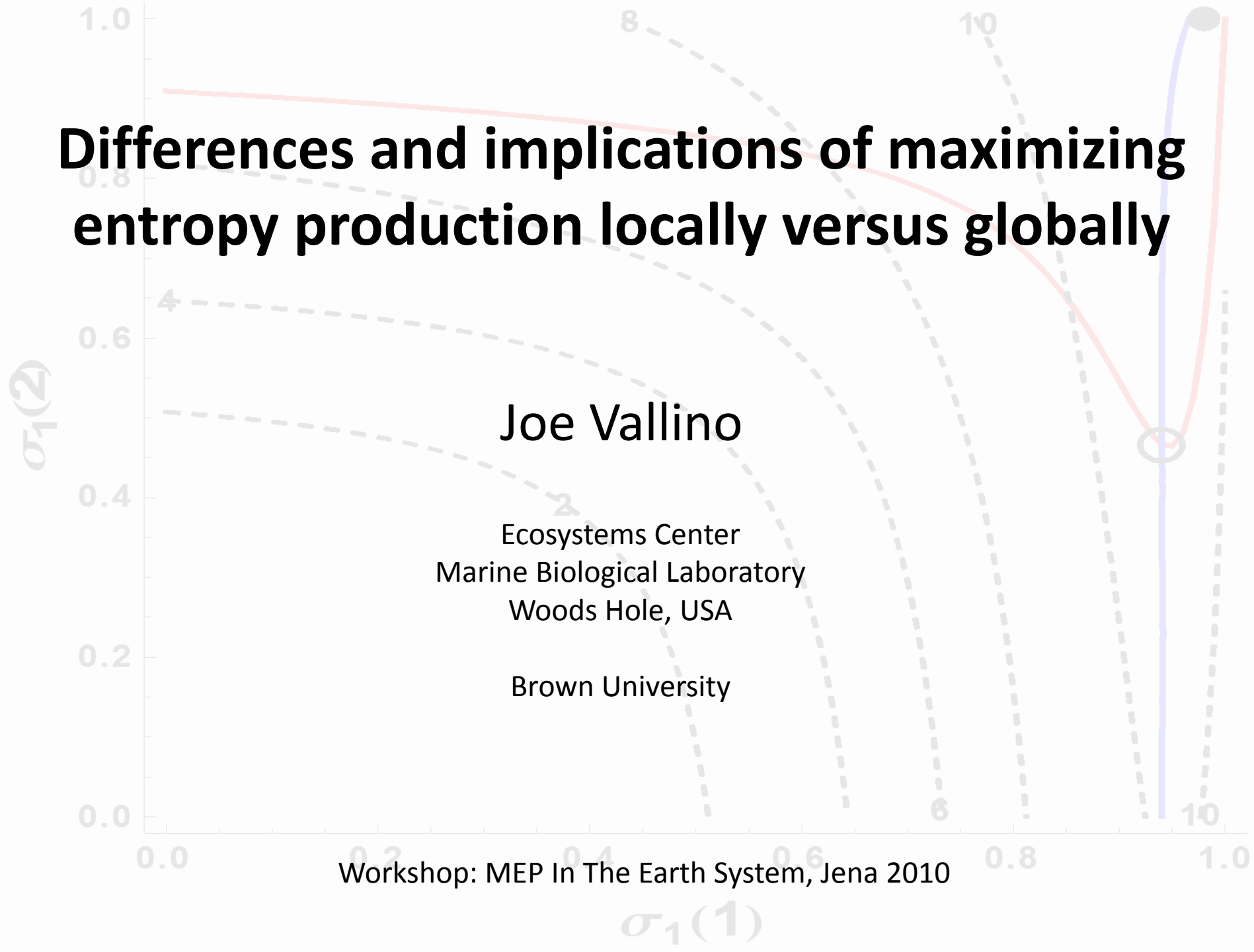


Differences and implications of maximizing entropy production locally versus globally



Objectives

- Develop a biogeochemistry (BGC) model based on the maximum entropy production (MEP) principle.
- *Hypothesis*: Models based on fundamental principles will have better predictive capabilities than those based on empirical relationships.
- *Alt-Hypothesis*: Forecast length will be greater with MEP-based models because model details are not needed.
- Demonstrate: Compare MEP-BGC and classic-BGC models to observations to test hypothesis.
- Longer term: Integrate MEP-BGC model into larger scale BGC models.

Today focus on: MEP and spatial scale

Emergent Systems Approach

Propose living systems organize around a dynamical attractor

- Concept attributable to at least Lotka (1922): maximum power

MEP as the dynamical attractor has many advantages

- Dewar (2003, 2005), and others, theoretical support
- Natural extension to classic equilibrium thermodynamics
- Thermodynamic entropy, S , is a well defined state variable:

$$r: A + B \rightarrow C + D \quad \frac{dS}{dt} = -\frac{\Delta G_r r(t)}{T} \quad \begin{array}{l} \text{Constant P and T} \\ \text{No mechanical energy storage} \end{array}$$

MEP paraphrasing:

If internal ordered structures facilitate energy dissipation, then the probability of a system with them is greater than a system without them.

Time and Space Considerations

Time*

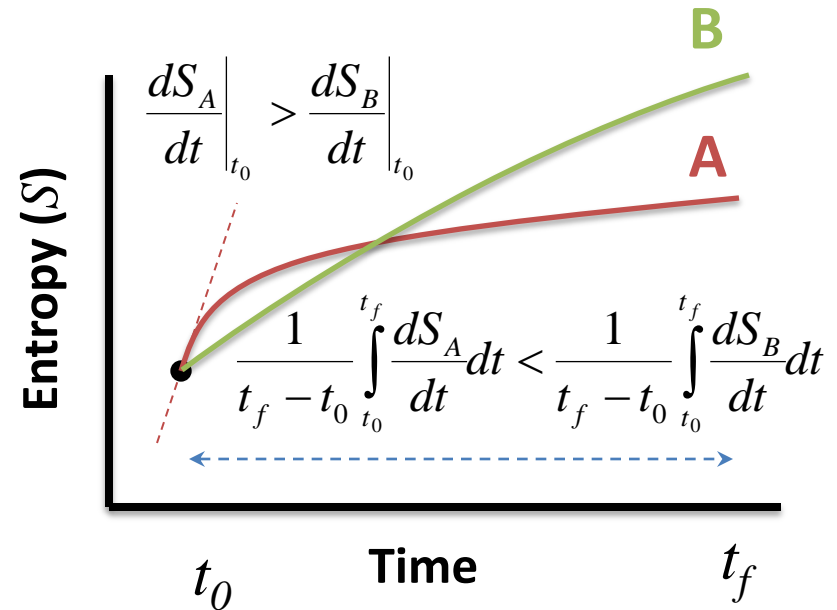
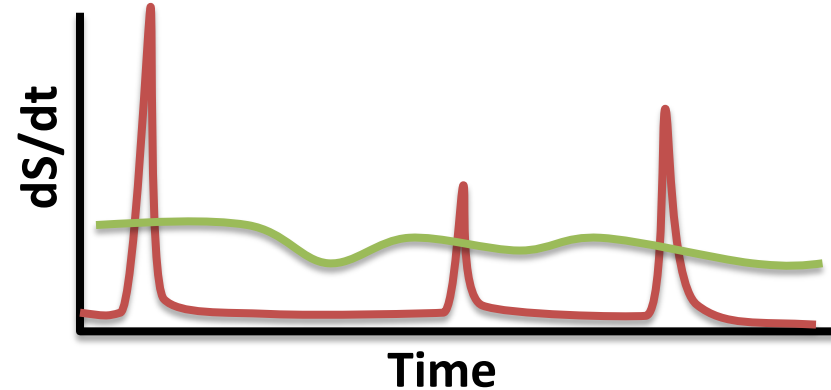
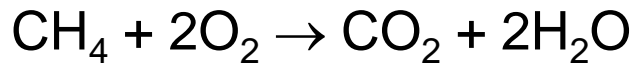
- Maximize entropy production *Instantaneously?*
- Maximize entropy production *Integrated Over Time?*

Space

- Maximize entropy production *At a Point?*
- Maximize entropy production *Integrated Over Space?*

*Vallino (2010) *Phil. Trans. R. Soc. B*

Average vs Instantaneous EP

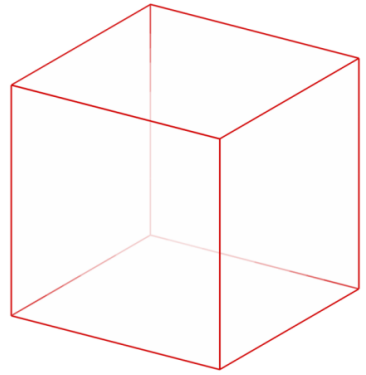


- “Useful information” stored in the metagenome allows living systems to predict future states and proceed along pathways that result in greater averaged entropy production than abiotic systems.
- **This is the only difference biotic systems and abiotic systems**
- However, pathways for maximal averaged entropy production may be flanked by pathways of steepest descent (e.g., forest fires, invasive species).

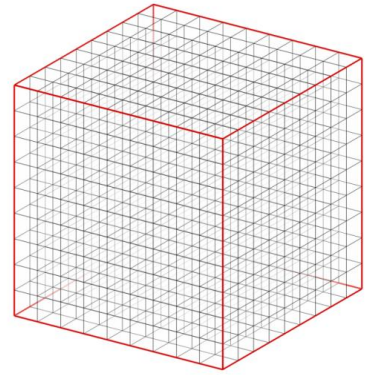
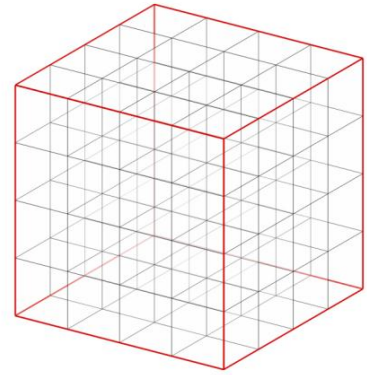
Entropy Production and Discretization?

How should domain be discretized?

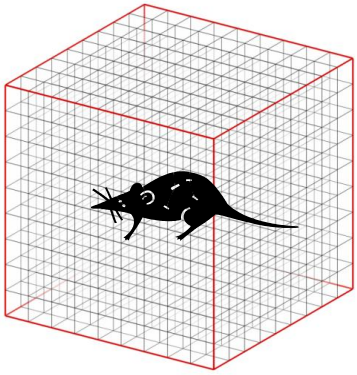
$$\iiint_{\Omega} \frac{dS}{dt} d\Omega$$



The standard approach for fluids:



How about if “Organized Structures” are in domain; Discretize them too?



If yes, discretize how?

- Into small cubes?
- Molecule?
- Cellular organelle?
- Cell?
- Whole organism?
- Community, Ecosystem?

Maximize Entropy How?

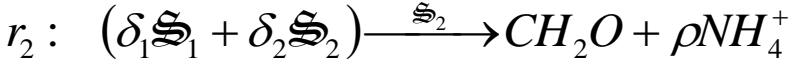
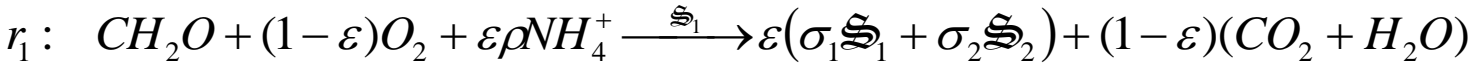
Locally $\sum_i \max \left(\frac{dS_i}{dt} \right)$

or

Globally $\max \left(\sum_i \frac{dS_i}{dt} \right)$

Example: “Two Layer Ocean”

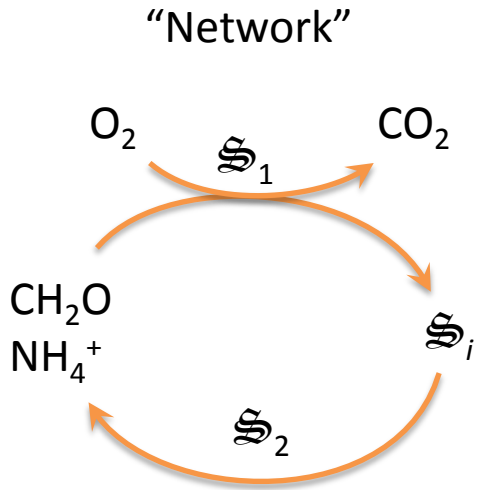
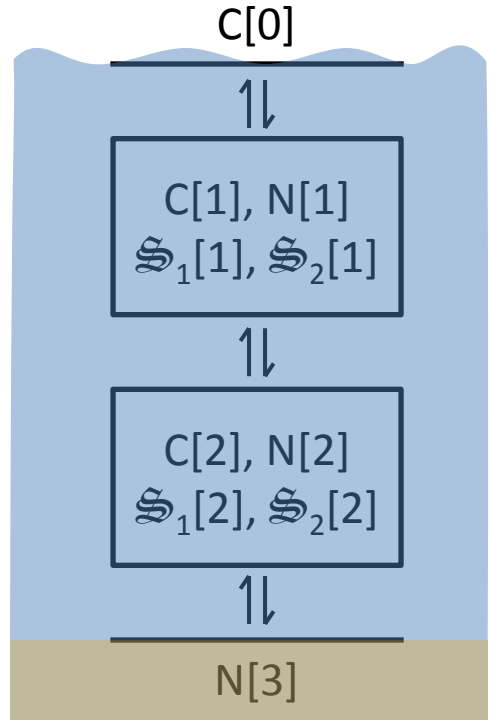
Metabolic Network:



Kinetics:

$$r_1[j] = v_1 \mathfrak{S}_1[j] \frac{C[j]}{C[j] + k_{C,\mathfrak{S}}} \frac{N[j]}{N[j] + k_{N,\mathfrak{S}}}$$

$$r_2[j] = v_2 \mathfrak{S}_2[j] \frac{\mathfrak{S}_T[j]}{\mathfrak{S}_T[j] + k_{\mathfrak{S},CN}}$$



Oxygen and other elements in excess

Example: State Equations

Surface layer: [1]

$$\frac{dC[1]}{dt} = \frac{D}{\ell^2} ((C[0] - C[1]) - (C[1] - C[2])) - r_1[1] + r_2[1]$$

$$\frac{dN[1]}{dt} = -\frac{D}{\ell^2} (N[1] - N[2]) - \varepsilon \rho r_1[1] + \rho r_2[1]$$

$$\frac{d\mathcal{S}_1[1]}{dt} = -\frac{D}{\ell^2} (\mathcal{S}_1[1] - \mathcal{S}_1[2]) + \varepsilon \sigma_1[1] r_1[1] - \delta_1[1] r_2[1]$$

$$\frac{d\mathcal{S}_2[1]}{dt} = -\frac{D}{\ell^2} (\mathcal{S}_2[1] - \mathcal{S}_2[2]) + \varepsilon \sigma_2[1] r_1[1] - \delta_2[1] r_2[1]$$

Deep layer: [2]

$$\frac{dC[2]}{dt} = \frac{D}{\ell^2} (C[1] - C[2]) - r_1[2] + r_2[2]$$

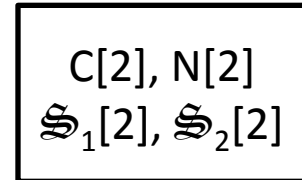
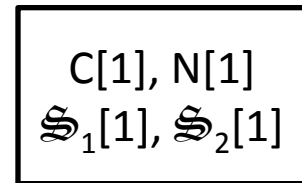
$$\frac{dN[2]}{dt} = \frac{D}{\ell^2} ((N[1] - N[2]) - (N[2] - N[3])) - \varepsilon \rho r_1[2] + \rho r_2[2]$$

$$\frac{d\mathcal{S}_1[2]}{dt} = \frac{D}{\ell^2} (\mathcal{S}_1[1] - \mathcal{S}_1[2]) + \varepsilon \sigma_1[2] r_1[2] - \delta_1[2] r_2[2]$$

$$\frac{d\mathcal{S}_2[2]}{dt} = \frac{D}{\ell^2} (\mathcal{S}_2[1] - \mathcal{S}_2[2]) + \varepsilon \sigma_2[2] r_1[2] - \delta_2[2] r_2[2]$$

No N or \mathcal{S} flux

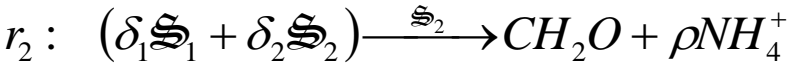
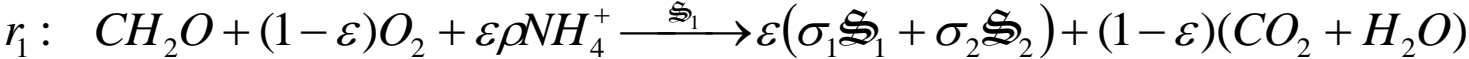
C[0]



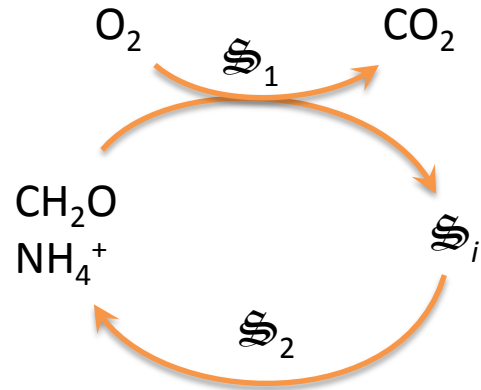
N[3]

No C or \mathcal{S} flux

Example: Entropy Production



Entropy Production:



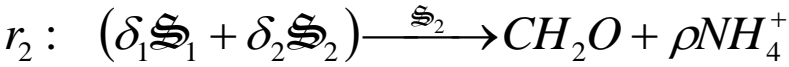
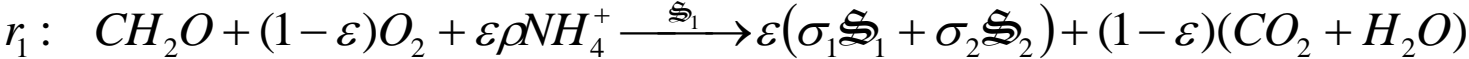
$$\frac{dS}{dt} = -\frac{1}{T}(\Delta G_{r_1} r_1 + \Delta G_{r_2} r_2)$$

Free energy of biosynthesis is small

$$\Delta G_{r_1} = (1-\varepsilon)\Delta G_{CH_2Ox}^o + \varepsilon\cancel{\Delta G_{\mathfrak{S}}^o}^0 + RT \ln\left(\frac{[H_2CO_3]^{(1-\varepsilon)}[\mathfrak{S}_T]^\varepsilon}{[CH_2O][O_2]^{(1-\varepsilon)}[NH_4^+]^{\varepsilon\rho}}\right)$$

$$\Delta G_{r_2} = -\cancel{\Delta G_{\mathfrak{S}}^o}^0 + RT \ln\left(\frac{[CH_2O][NH_4^+]^\rho}{[\mathfrak{S}_T]}\right)$$

Example: Control Variables



Parameters:

$\varepsilon, \rho, \nu_1, \nu_2, k_{C,\mathfrak{S}}, k_{N,\mathfrak{S}}, k_{\mathfrak{S},CN}, D, \ell$

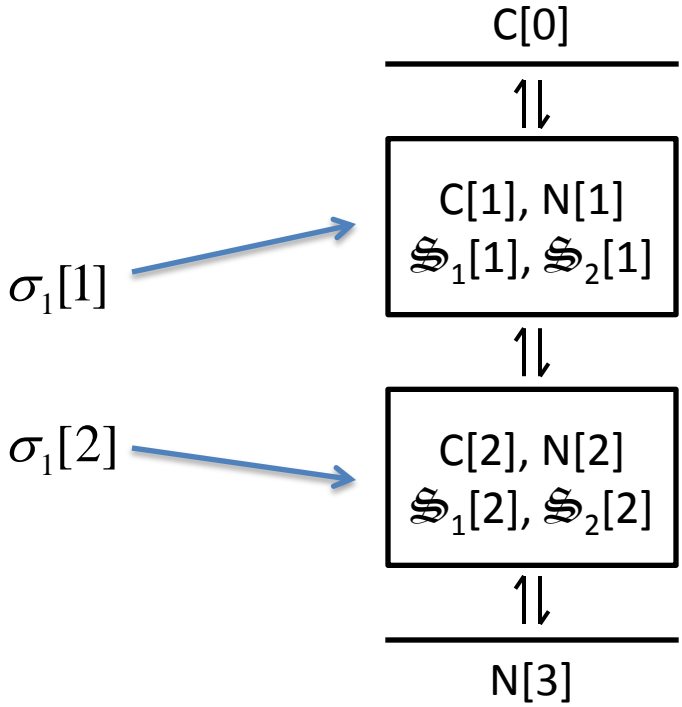
$$\delta_i[j] = \frac{\mathfrak{S}_i[j]}{\mathfrak{S}_1[j] + \mathfrak{S}_2[j]} = \frac{\mathfrak{S}_i[j]}{\mathfrak{S}_T[j]}$$

Control Variables:

$\sigma_1[j], \sigma_2[j], j=1, 2$

$\sigma_1[j] + \sigma_2[j] = 1$

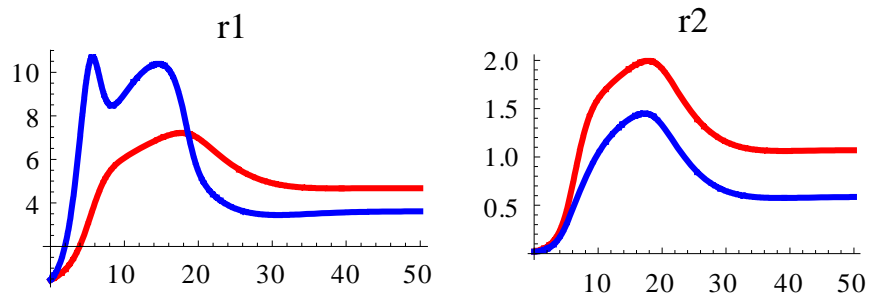
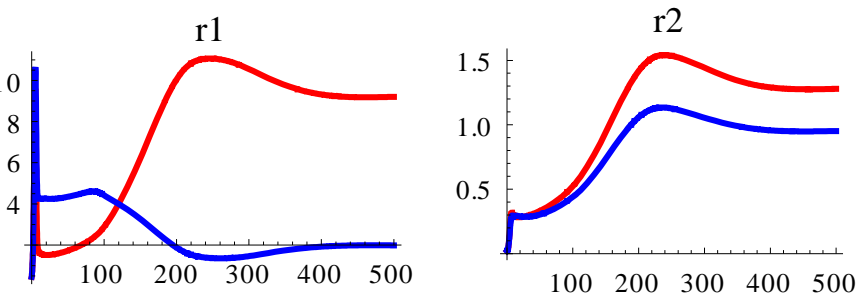
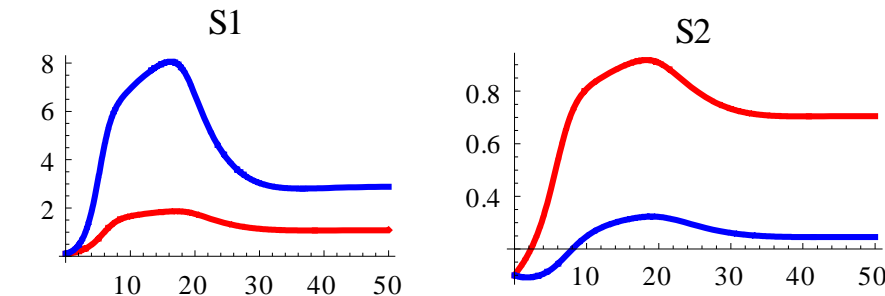
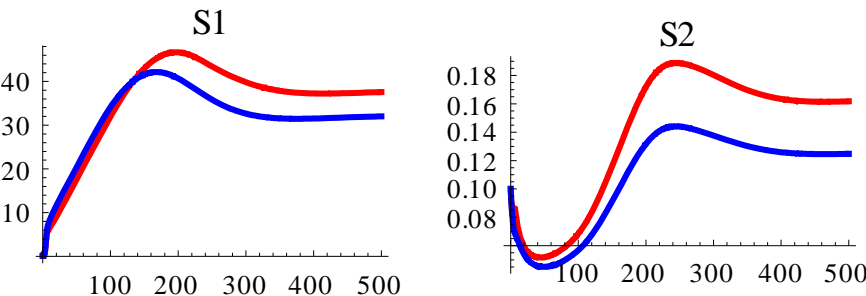
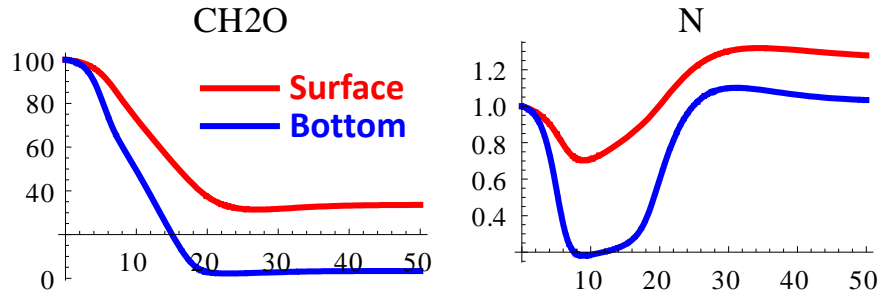
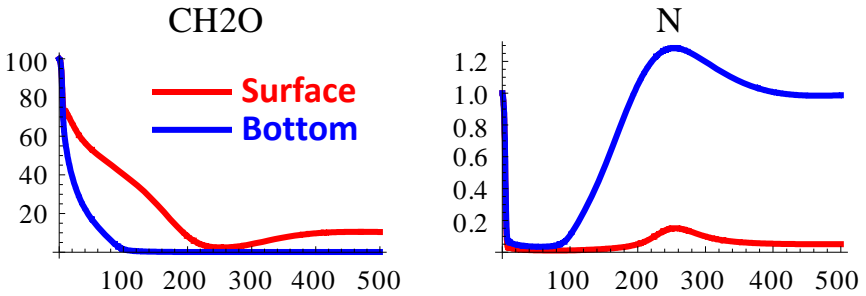
$\Rightarrow \sigma_2[j] = (1 - \sigma_1[j])$



Example: Transient Solutions

$\sigma_1[1] = 0.995, \sigma_1[2] = 1.0$

$\sigma_1[1] = 0.5, \sigma_1[2] = 1.0$



(mJ L⁻¹ d⁻¹ °K)

dSdt 1	11.4734
dSdt 2	2.45696
dSdt 1 ; dSdt 2	13.9304

Total N 12.68

(mJ L⁻¹ d⁻¹ °K)

dSdt 1	5.87634
dSdt 2	4.47857
dSdt 1 ; dSdt 2	10.3549

Total N 3.13137

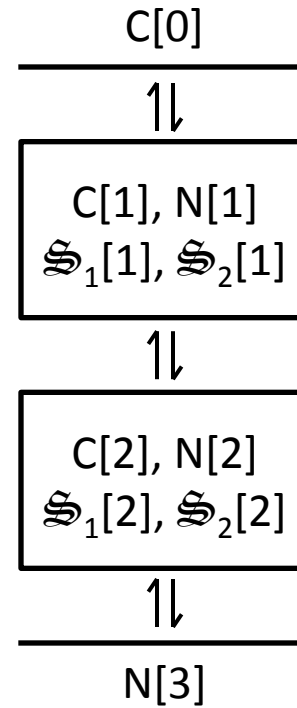
Example: Optimization

Local Optimization

$$\text{Max}_{\sigma_1[1]} \frac{dS[1]}{dt} \quad \text{and} \quad \text{Max}_{\sigma_1[2]} \frac{dS[2]}{dt}$$

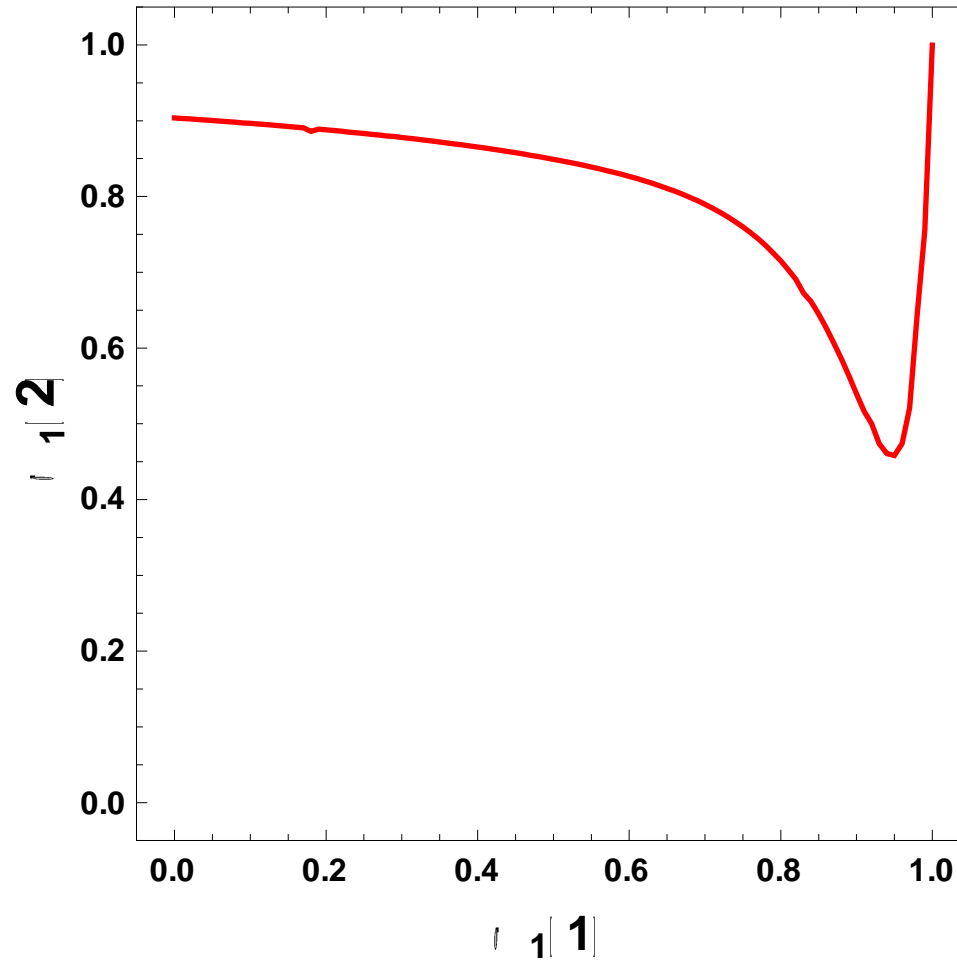
Global Optimization

$$\text{Max}_{\sigma_1[1], \sigma_1[2]} \frac{dS[1]}{dt} + \frac{dS[2]}{dt}$$



Example: Results, Local Optimum [2]

$$\text{Max}_{\sigma_1[2]} \frac{dS[2]}{dt}$$

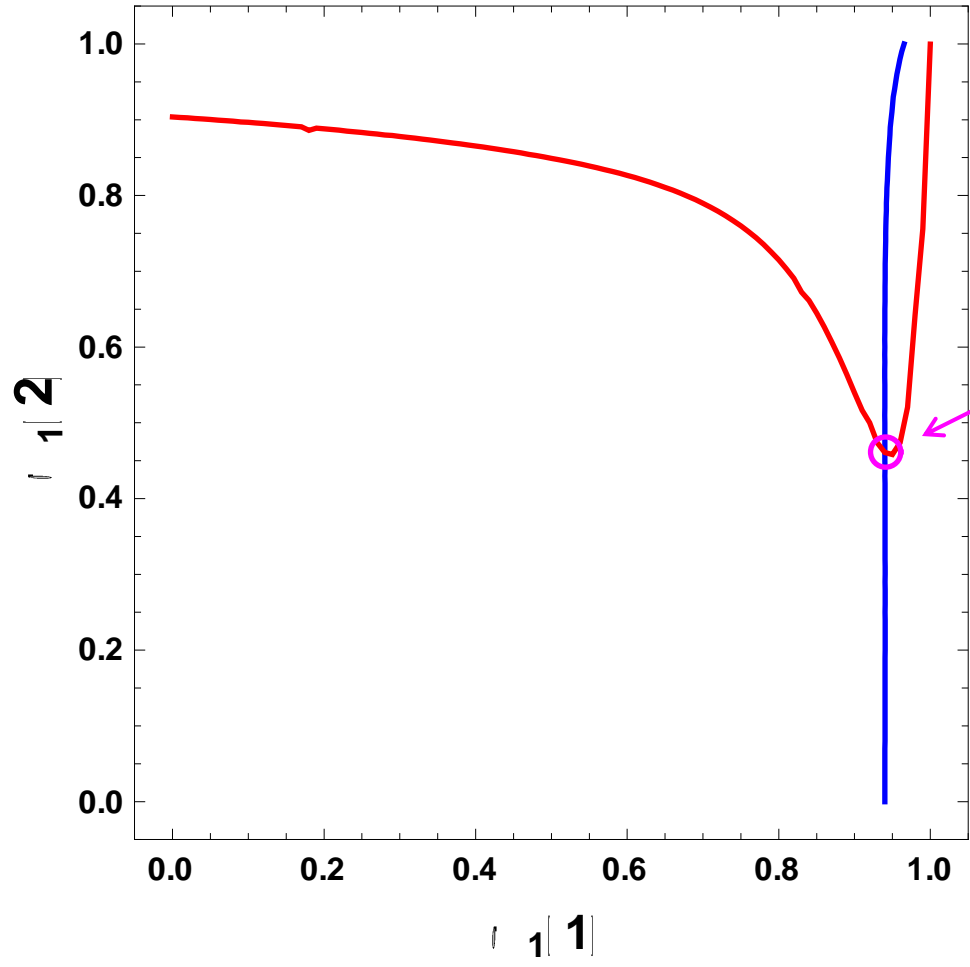


Example: Results, Local Optimum [2] and [1]

$$\text{Max}_{\sigma_1[2]} \frac{dS[2]}{dt}$$

and

$$\text{Max}_{\sigma_1[1]} \frac{dS[1]}{dt}$$



Local

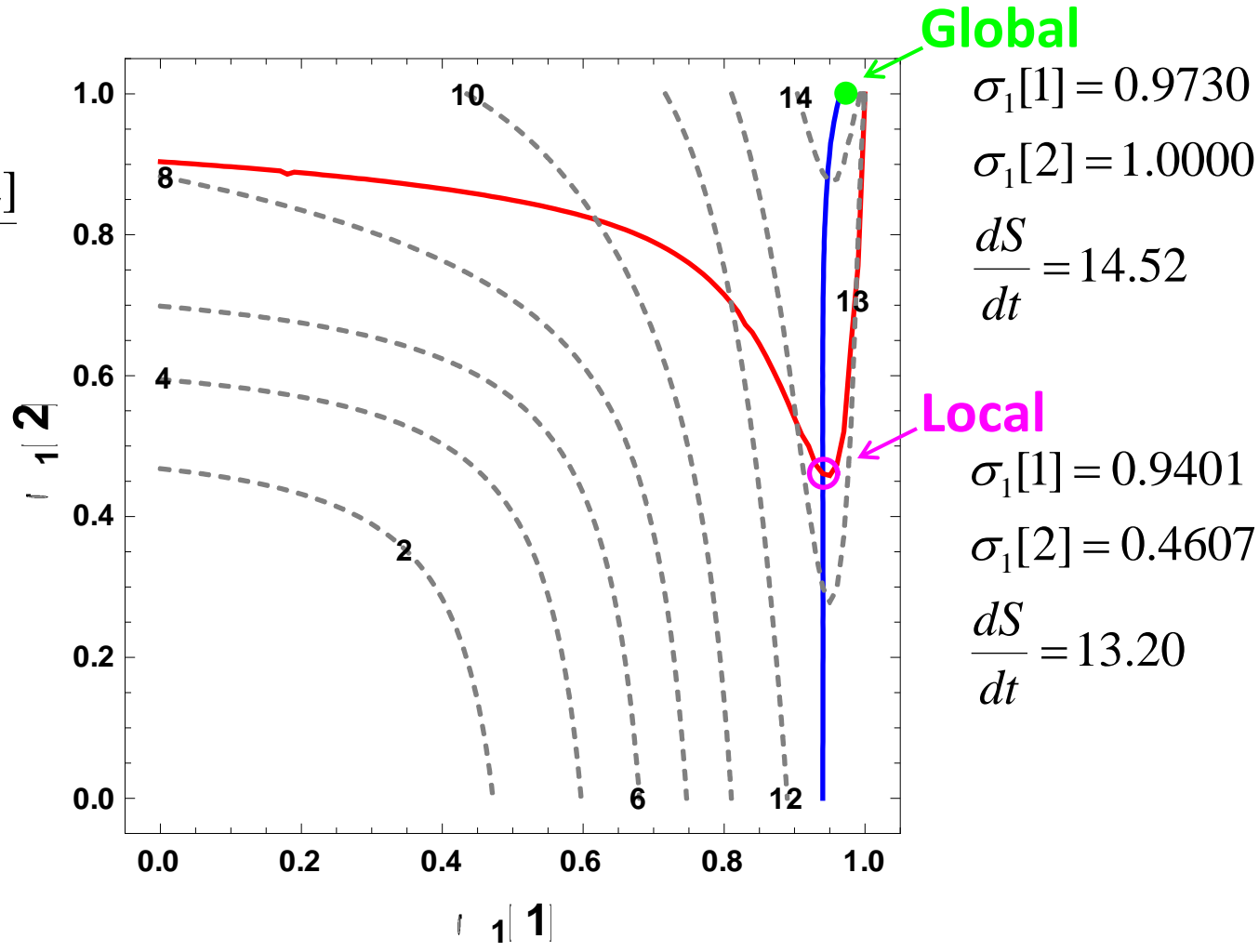
$$\sigma_1[1] = 0.9401$$

$$\sigma_1[2] = 0.4607$$

$$\frac{dS}{dt} = 13.20$$

Example: Results, Global Optimum

$$\text{Max}_{\sigma_1[1], \sigma_1[2]} \frac{dS[1]}{dt} + \frac{dS[2]}{dt}$$



CONCLUSIONS (CONJECTURES)

- Entropy production is greater when maximized globally rather than locally.
- Systems should organize/evolve to maximize entropy production globally then.
- Implies local “sacrifice” to maximize EP globally (altruism*).
- Systems must exhibit cooperation for entropy to be maximized globally.
- The extent of information exchange defines the domain of a system (extent of cooperation).
- Does information (aka Shannon entropy) facilitate thermodynamic entropy production over space and time?

*Ref: Wilson and Wilson (2008) *American Scientist*

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