

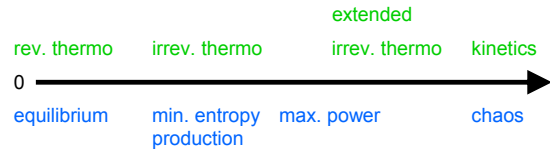
Finite-time optimization of chemical and unrelated reactions

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Finite-Time Thermodynamics

- Effect of finite **time**
- Effect of finite **size**
- Optimize **what**



Finite-Time Thermodynamic idea

Reversible thermo. procedures

- + simple rate processes
- + finite total duration

Macroscopic modeling

Generic

Any objective function

Finite-Time Thermodynamic concepts

- Bounds for finite time
- Generalized exergy for finite time
- Generalized thermo. potentials incorporating constraints
- Constant rate of entropy production for linear transfer laws
- What is being optimized?
- Optimal path
- Thermodynamic geometry
- ...

Entropy survives

Entropy applies equally in the chemistry lab, to the quantum computer, and to black holes.

Albert Einstein: "classical thermodynamics ... is the only physical theory of universal content that, within the framework of applicability of its basic concepts, will never be overthrown".

Cooling of atomic systems is not a question of removing the last bit of energy but entropy.

Entropy governs everything



Some basics

There are many ways to slice a kiwi

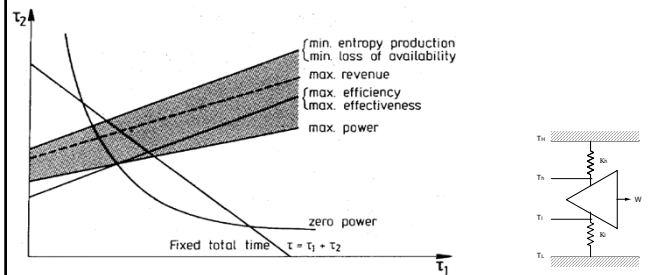


What makes you happy?

- you are in a hurry
- you want to conserve
- you want profit
- you have ulterior motives

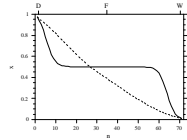
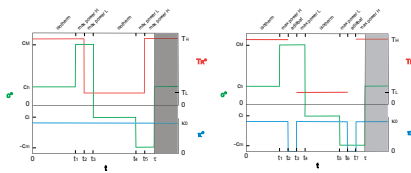
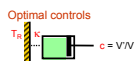
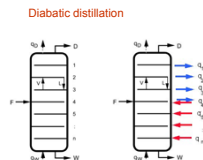
Criteria of performance

Every person has his own objective: max. power, max. efficiency, max. profit, ...
They all correspond to *different bounds* and *different optimal paths*.



Controls are needed

The operator, not Nature, decides.
All optimization requires controls to steer the process in the desired direction.



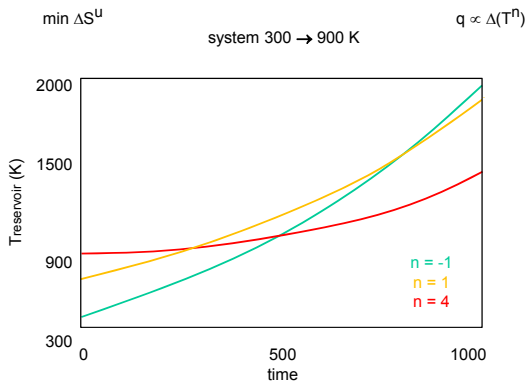
Simulated annealing
Annealing schedule $\frac{dT}{dt} = -\frac{\sqrt{T}}{\epsilon\sqrt{C}}$

M. H. Rubin, Phys. Rev. A 19, 1277-1289 (1979)

Constraints

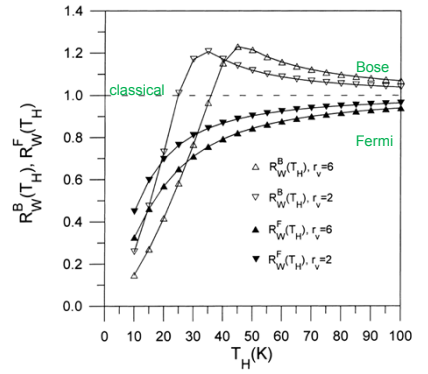
- you cannot use freeways
- you cannot drive at night
- you must meet district heating demands
- you have no use for waste heat
- limitations of your machinery
- purity of products
- fix production / consumption / duration

Optimal heating sequence



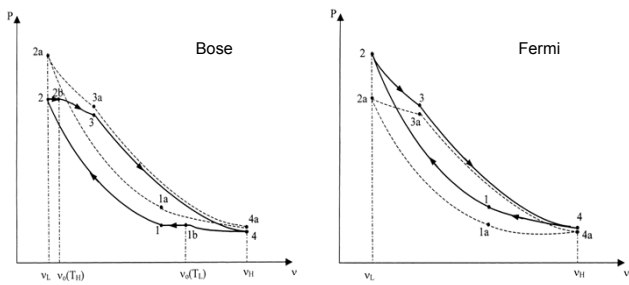
Bose/Fermi work per cycle

$T_L/T_H=0.4$
 $r_v=V_H/V_L$
 $P=nkT \times$
quantum factor
For all
 $\eta_c=1-T_L/T_H$



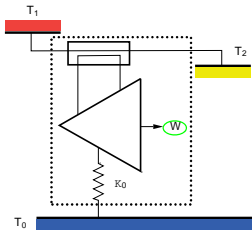
A. Sisman, H. Saygin, Applied Energy, 68, 367-376 (2001)

Carnot cycles



A. Sisman, H. Saygin, Applied Energy, 68, 367-376 (2001)

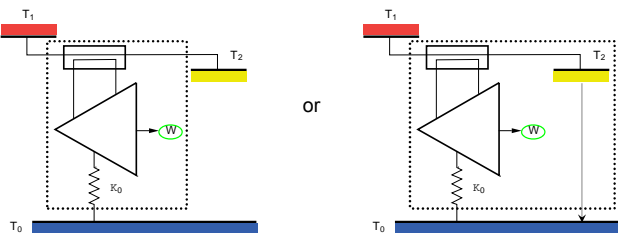
System boundary



$$\max W \leftrightarrow \min \Delta S^U$$

$$= \sigma_{\text{system}} ?$$

System boundary



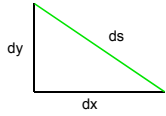
$$\max W \leftrightarrow \min \Delta S^U$$

$$= \sigma_{\text{system}} ?$$

$$\max W = \min \Delta S^U$$

Thermodynamic geometry

Length



$$ds = \sqrt{dx^2 + dy^2}$$

$$L = \int \sqrt{dx^2 + dy^2} = \int \sqrt{1 + \left(\frac{dy}{dx}\right)^2} dx = \int \sqrt{\left(\frac{dx}{dt}\right)^2 + \left(\frac{dy}{dt}\right)^2} dt$$

$$\text{In general } L = \int \sqrt{(dx, dy) \mathbf{M} \begin{pmatrix} dx \\ dy \end{pmatrix}}$$

$$\text{e.g. } L = \int_1^r \sqrt{(dx, dy) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} dx \\ dy \end{pmatrix}} = \int_1^r \sqrt{(dr, d\theta) \begin{pmatrix} 1 & 0 \\ 0 & r^2 \end{pmatrix} \begin{pmatrix} dr \\ d\theta \end{pmatrix}}$$

Thermodynamic metric

$$\text{Length: } L = \int \sqrt{d\mathbf{X}^T \mathbf{M} d\mathbf{X}} \quad \text{conjugate: } \mathbf{Y}_i = \left(\frac{\partial P}{\partial X_i} \right)$$

$$\text{Entropy picture: } \mathbf{S}(\mathbf{X}), \quad \mathbf{X} = (U, V, N_1, \dots), \quad \mathbf{Y} = \left(\frac{1}{T}, \frac{p}{T}, -\frac{\mu}{T}, \dots \right), \quad \mathbf{M}_S = \begin{pmatrix} \frac{\partial^2 S}{\partial X_i \partial X_j} \end{pmatrix}$$

$$\text{Energy picture: } \mathbf{E}(\mathbf{X}), \quad \mathbf{X} = (S, V, N_1, \dots), \quad \mathbf{Y} = (T, -p, \mu, \dots), \quad \mathbf{M}_U = \begin{pmatrix} \frac{\partial^2 U}{\partial X_i \partial X_j} \end{pmatrix}$$

$$\text{Probability picture: } \mathbf{S}(\mathbf{X}) = -\sum p_i \ln p_i, \quad \mathbf{X} = (p_i), \quad \mathbf{Y} = -(\ln p_i + 1), \quad \mathbf{M}_S = \begin{pmatrix} 1 \\ p_i \end{pmatrix}$$

Thermodynamic metric

$$-D^2S = - \begin{pmatrix} \frac{\partial^2 S}{\partial U^2} & \frac{\partial^2 S}{\partial U \partial V} \\ \frac{\partial^2 S}{\partial V \partial U} & \frac{\partial^2 S}{\partial V^2} \end{pmatrix} = \begin{pmatrix} \frac{C_v}{U^2} & 0 \\ 0 & \frac{R}{V^2} \end{pmatrix}$$

Ideal gas
{U, S, V}

$$D^2U = \begin{pmatrix} \frac{\partial^2 U}{\partial S^2} & \frac{\partial^2 U}{\partial S \partial V} \\ \frac{\partial^2 U}{\partial V \partial S} & \frac{\partial^2 U}{\partial V^2} \end{pmatrix} = \begin{pmatrix} \frac{T}{C_v} & -\frac{p}{C_v} \\ -\frac{p}{C_v} & \frac{p}{V} \end{pmatrix}$$

Using: $dU = TdS - pdV$

$PV = RT$

$U = C_v T$

$TS = C_p T$

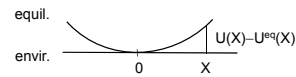
$R = C_p - C_v$

Curvature

$$U^{\text{eq}}(X) = U_0 + \frac{\partial U}{\partial X} \Delta X + \frac{1}{2} \frac{\partial^2 U}{\partial X^2} (\Delta X)^2 + \dots$$

$$U^{\text{env}}(X) = U_0 + \frac{\partial U}{\partial X} \Delta X$$

$$\text{Availability} = \text{exergy: } dA = dU - T_e dS + P_e dV$$



Thermodynamic metric bounds

Continuous process

$$\text{FTT bound: } \Delta S^U \geq \frac{L^2 \epsilon}{\tau} \quad -\Delta A^U \geq \frac{L_U^2 \epsilon}{\tau}$$

$$\text{Reversible bound: } \Delta S^U \geq 0 \quad -\Delta A^U \geq 0$$

$$\text{For slow processes } v = \frac{dL}{dt} = \text{const.} \Rightarrow \frac{dT}{dt} = \frac{-vT}{\epsilon \sqrt{C}}$$

Continuous adjustment to a moving target

Curvature ~ interaction strength

$$\text{e.g. } \sum p_i = 1$$

$$\Rightarrow \text{Gaussian curvature} = 4$$

Thermodynamic metric bounds

Continuous process

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Continuous adjustment to a moving target

Curvature ~ interaction strength

$$\text{e.g. } \sum p_i = 1$$

$$\Rightarrow \text{Gaussian curvature} = 4$$

Step process

$$\Delta S^U \geq \frac{L^2}{2N} \quad \Delta L = \text{const.}$$

Step processes

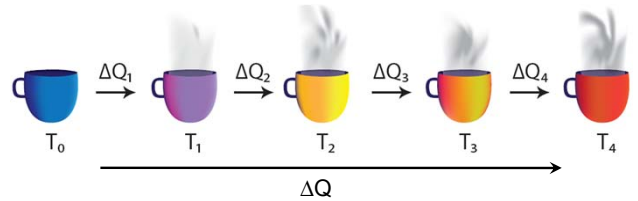
- Consecutive heat engines
- Information encoding
- Economic gain
- Distillation column
- Chemical reactions



Horse-Carrot processes

Use thermodynamic geometry to determine optimal operation

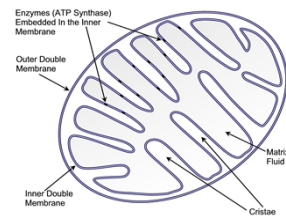
Stepwise heating



$$\Delta Q = \sum_i \Delta Q_i$$

$$\Delta S = \int_{T_0}^{T_4} dQ \left(\frac{1}{T} - \frac{1}{T_4} \right) > \sum_i \Delta S_i = \sum_i \int_{T_{i-1}}^{T_i} dQ \left(\frac{1}{T} - \frac{1}{T_i} \right)$$

Mitochondrion



1×10^{-6} m

Fuel cell

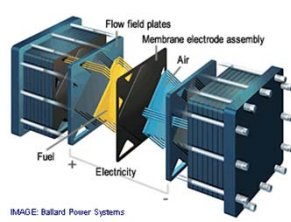
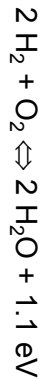
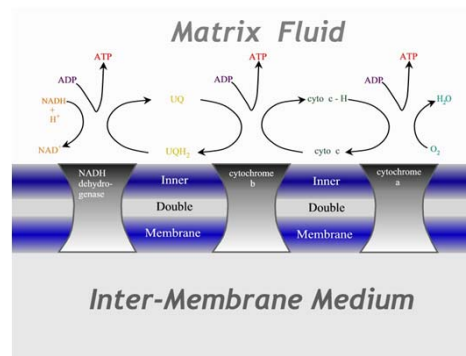


IMAGE: Ballard Power Systems

1×10^{-1} m

Cytochrome chain



Thermodynamic metric

Length: $L = \int \sqrt{d\mathbf{X}^T \mathbf{M} d\mathbf{X}}$ conjugate: $Y_i = \left(\frac{\partial P}{\partial X_i} \right)$

Entropy picture: $S(\mathbf{X}), \mathbf{X} = (U, V, N_1, \dots), \mathbf{Y} = \left(\frac{1}{T}, \frac{p}{T}, -\frac{\mu}{T}, \dots \right), \mathbf{M}_S = \left(\frac{\partial^2 S}{\partial X_i \partial X_j} \right)$

Energy picture: $E(\mathbf{X}), \mathbf{X} = (S, V, N_1, \dots), \mathbf{Y} = (T, -p, \mu, \dots), \mathbf{M}_U = \left(\frac{\partial^2 U}{\partial X_i \partial X_j} \right)$

Probability picture: $S(\mathbf{X}) = -\sum p_i \ln p_i, \mathbf{X} = (p_i), \mathbf{Y} = -(\ln p_i + 1), \mathbf{M}_S = \left(\frac{1}{p_i} \right)$

Energy picture – ideal gas

$U(S, V, n_i)$ derived from:

$$U = NC_v T$$

$$pV = NkT$$

$$\mu_i = kT \left[\ln \frac{n_i}{V} - \frac{C_v}{k} \ln(m_i kT) + a \right]$$

with $T = \frac{\partial U}{\partial S}, p = -\frac{\partial U}{\partial V}, \mu_i = \frac{\partial U}{\partial n_i}, N = \sum_i n_i$

$$U(S, V, n_i) = be^{\frac{S}{NC_v}} N \prod_j \left[\left(\frac{n_j}{V} \right)^{\frac{k}{C_v}} \frac{1}{m_j} \right]^{\frac{n_j}{N}}$$

Energy picture – ideal gas

$U(S, V, n_i)$ derived from:

$$U = NC_v T$$

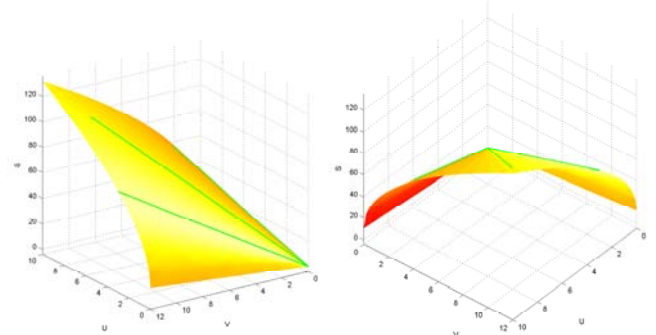
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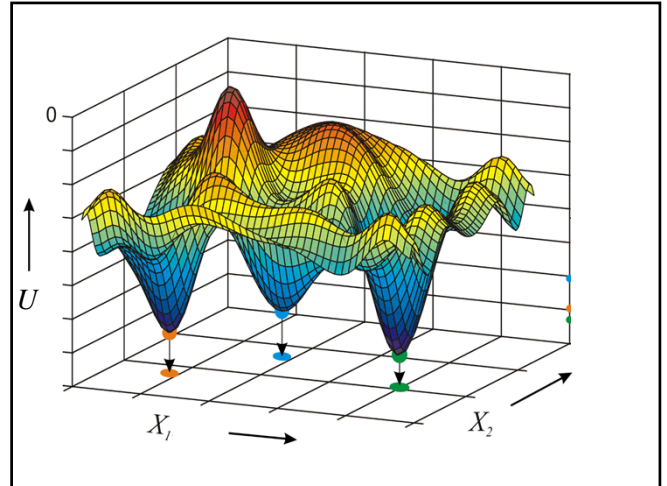
Ideal gas equilibrium surface



Water equilibrium surface



Gift from James Clerk Maxwell to J. Willard Gibbs 1874.



Metric – ideal gas

$$M_U = \begin{pmatrix} \frac{\partial^2 U}{\partial S^2} & \frac{\partial^2 U}{\partial S \partial V} & \frac{\partial^2 U}{\partial S \partial n_i} \\ \frac{\partial^2 U}{\partial V \partial S} & \frac{\partial^2 U}{\partial V^2} & \frac{\partial^2 U}{\partial V \partial n_i} \\ \frac{\partial^2 U}{\partial n_i \partial S} & \frac{\partial^2 U}{\partial n_i \partial V} & \frac{\partial^2 U}{\partial n_i \partial n_j} \end{pmatrix}$$

$$= \begin{pmatrix} \frac{T}{NC_v} & -\frac{p}{NC_v} & \frac{\mu_i}{NC_v} - \frac{T}{N} \\ -\frac{p}{NC_v} & \left(1 + \frac{k}{C_v}\right) \frac{p}{V} & -\frac{k}{C_v} \frac{\mu_i}{V} \\ \frac{\mu_i}{NC_v} - \frac{T}{N} & -\frac{k}{C_v} \frac{\mu_i}{V} & \mu_i \left(\frac{\mu_j}{U} - \frac{1}{N} \right) - \frac{1}{N} \mu_j + \frac{1}{N^2} U + \frac{RT}{N_j} \delta_{ij} \end{pmatrix}$$

Variables

variable	condition
S	T const.
V	p const.
N_{H^+}	pH const.
N_P	[P] const.
N_{ATP}	$N_{ATP} + N_{ADP}$ const. [ATP] const.
N_{NADH}	$N_{NADH} + N_{NAD^+}$ const. [NADH] const.
N_{CoQ}	$N_{CoQ} + N_{CoQH_2}$ const.
N_{cyto-c}	$N_{cyto-c} + N_{cyto-c-H}$ const.
N_{H_2O}	[H ₂ O] const.
N_{O_2}	[O ₂] const.

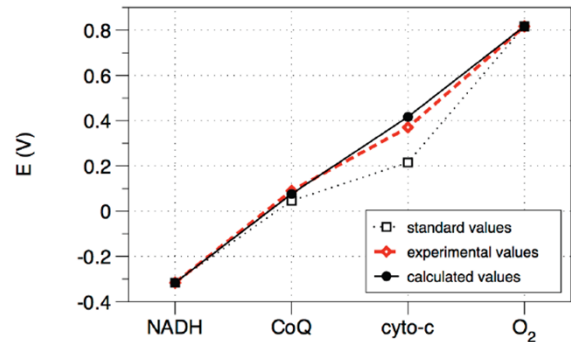
Dimension reduction

- Many dimensions: S, V, N₁, N₂, ...
- Conservation equations and common T, p relate these
- Eventually just one, e.g. ζ or electrochemical potential
- Formally equivalent to distillation column in T

Procedure:

calculate total length L
 each step is $L_n=L/N$
 $L_n \Rightarrow \zeta_n$ and thus concentrations

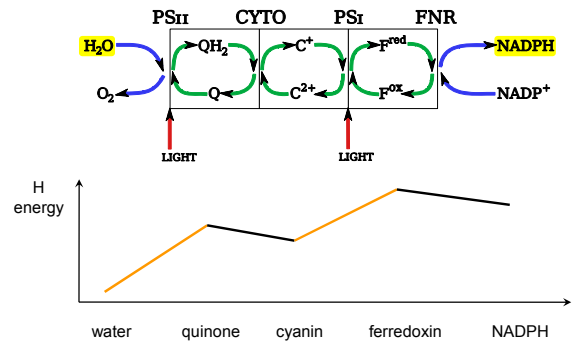
Optimal electrochemical path



Conclusions

- The cytochrome chain is a step process
- sharing a common medium
- at constant T and pH
- with the electro-chemical potential as the control.
- The sequence is logical, not physical.

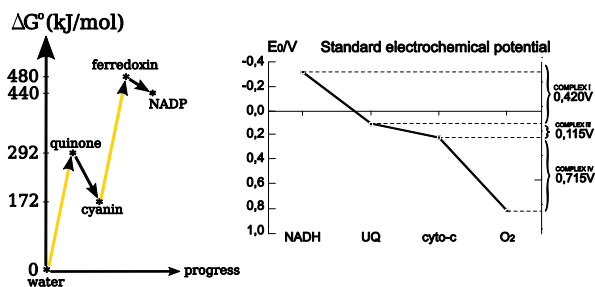
Electron transport chain



Energy path

photosynthesis

cytochrome chain



How to handle light

A + photon → B

- Does the photon count as a reactant?
- Is there equilibrium between the 3 components?
- How does light interact with matter?

Length elements

length element :

$$(dL)^2 = kT \left[\frac{\tilde{N}_O}{x_O(1-x_O)} dx_O^2 + \frac{\tilde{N}_C}{x_C(1-x_C)} dx_C^2 + \frac{\tilde{N}_F}{x_F(1-x_F)} dx_F^2 \right]$$

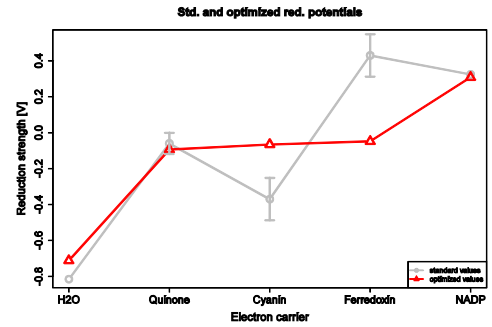
reduction potential :

$$E - E_i^0 = \frac{RT}{n_i F} \ln \frac{[A_{red}^i]}{[A_{ox}^i]} = \frac{kT}{n_i e} \ln \frac{1-x_i}{x_i}$$

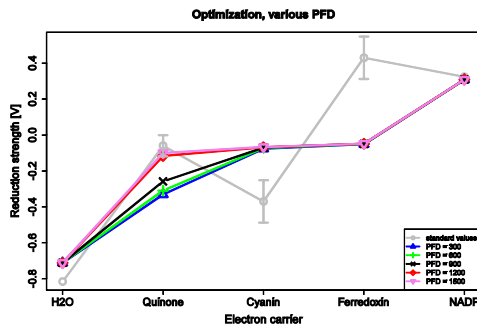
total length :

$$L = \int dL = \int_{E_1}^{E_2} \sqrt{\sum_i \frac{\tilde{N}_i e^2 n_i^2 / kT}{1 + e^{(E-E_i^0)n_i/kT}}} dE$$

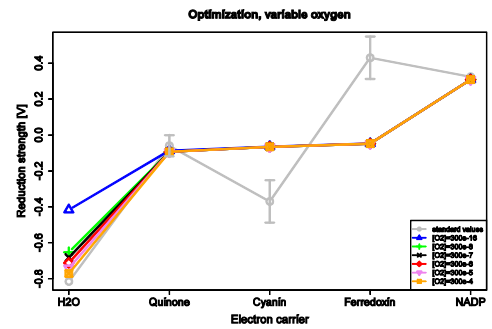
Optimal path



Light intensity



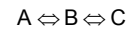
Oxygen partial pressure



Conclusions

- Photosynthesis is a step process
- sharing a common medium
- at constant T and pH
- with the electro-chemical potential as the control.
- Alternating light and dark processes necessary to avoid recombination.
- Sunlight intensity optimal. However ...
- Low sensitivity on oxygen pressure.

Chemical reaction



$$k_i = s_i e^{-U_i/kT} \quad u = e^{-1/kT}$$

maximize [B] at $t = \tau$

Constraints :

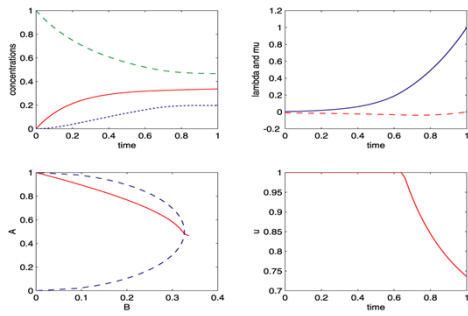
rate equations $\frac{dA}{dt}, \frac{dB}{dt}$

conjugate λ, μ

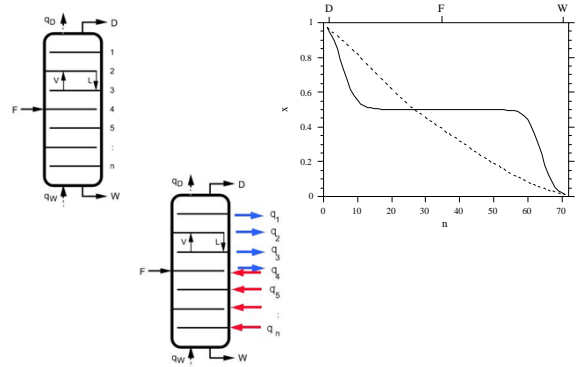
initial conds. $A(0) = 1, B(0) = 0, C(0) = 0$

At final time $\frac{dB^*}{dt} = \frac{dB}{dt} \Big|_{t=\tau}$

Chemical reaction optimization



Distillation column



Column optimization

Step process: $\Delta S^U \geq \frac{L^2}{2N}$ $\Delta L = \text{const.}$

$\mathbf{X} = (S_V, V_V, N_{1V}, N_{2V}, S_L, V_L, N_{1L}, N_{2L})$

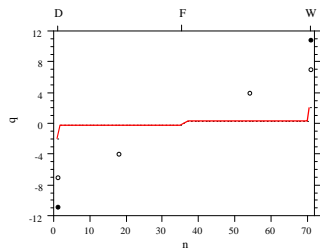
$\mathbf{M}_S = (8 \times 8)$

...but simplifies using equilibrium equations to

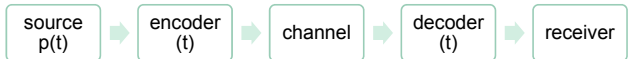
$$L_S = \sum_{k=1}^N D_S^k$$

$$D_S^k = \frac{\sqrt{C_r^k}}{T^k} \Delta T$$

Optimum: all D_S^k equal

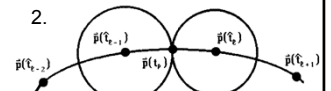


Min losses in adaptive coding



1. How many code changes?
2. When to change?
3. Which code in segment?

3. E.g. Huffman code



1. $K \propto \sqrt[3]{rtL^2}$

$$C = \frac{3}{2} C_{\text{switch}} + \frac{1}{3} C_{\text{excess bits}}$$

Probabilistic description

Metric: $S(p_i) = -\sum p_i \ln p_i$, $\mathbf{Y} = -(\ln p_i + 1)$, $\mathbf{M}_S = \left(\frac{1}{p_i} \right)$

Length: $L = \int \sqrt{d\mathbf{p}^T \mathbf{M} d\mathbf{p}}$

Loss: $R = r \left[\sum_{j=1}^K \sum_{i=1}^m p_i^j (-\ln p_i^{-j}) - \int_0^r H(\mathbf{p}(t)) dt \right]$

Delayed adjustment in economics

Utility fcn. moves faster than the market

Loss $W \geq L^2/T$

L total length of change, T duration

Example:

Cobb-Douglas utility fcn. $U(x,y) = x^a y^b$

Geometry flat (spiral) in (r,θ)

$$L[(x_1, y_1), (x_2, y_2)] = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2} \quad \text{if } |\theta_1 - \theta_2| \leq \pi$$

$$L[(x_1, y_1), (x_2, y_2)] = \sqrt{x_1^2 + y_1^2} + \sqrt{x_2^2 + y_2^2} \quad \text{if } |\theta_1 - \theta_2| > \pi$$

Simulated annealing

State space / universe: $\Omega = \{\omega\}$

Energy / objective function: $E(\omega)$

Neighborhood: $N(\omega)$

Temperature schedule / control: $T(t)$

Metropolis algorithm:

1. choose prospective $\omega' \in N(\omega)$

2. accept ω' with probability $P_{\text{acc}} = \begin{cases} 1 & \text{if } \Delta E \leq 0 \\ e^{-\Delta E/T} & \text{if } \Delta E > 0 \end{cases}$

Annealing schedule

Originally suggested:

$$T(t) = a \exp(-t/b)$$

$$T(t) = a/(b+t)$$

$$T(t) = a/\ln(b+t)$$

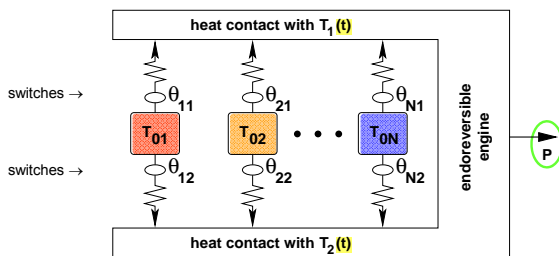
Adaptive schedule: $\frac{dT}{dt} = -\frac{vT}{\varepsilon\sqrt{C}}$ or $\frac{\langle E \rangle - E_{\text{eq}}(T)}{\sigma} = v$

From ensemble $Z(T) = \sum_i p_i \exp(-E_i/T)$

$$E(T) = T^2 \frac{d \ln Z}{dT}$$

$$C(T) = \frac{dE}{dT} = \frac{\langle (\Delta E)^2 \rangle}{T^2}$$

Multi-source endoreversible engine



Averaged non-linear optimization

- Optimal controls are piecewise constant fcn.s. taken from
- $m+1$ possible values where
- m number of constraints

Thus:

- T_1 and T_2 are time-independent
- $m=1 \Rightarrow$ just 2 possible values for T_1 and T_2 (T_h and T_c) and for $\theta_{i\alpha}$ (0 and 1)

Possible solutions

$\theta_{i\alpha} = 1$ if $T_h < T_{0i}$ HOT

$\theta_{i\alpha} = 0$ if $T_c < T_{0i} < T_h$ temperate

$\theta_{i\alpha} = 1$ if $T_{0i} < T_c$ COLD

i.e. hot and cold reservoirs are constantly used, temperate ones never

Recap of some finite-time thermo tools

- Curzon-Ahlborn engine
- Generalized potentials
- Criteria of performance
- Thermodynamic length
- Optimal control
- Averaged non-linear optimization

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The hard workers

- Kristian Bødker Frederiksen (cytochrome chain)



- Nikolaj Becker (photosynthesis)



Thanks

Thanks to

- Peter Salamon
- R. Stephen Berry
- Karl Heinz Hoffmann
- Jeff Gordon
- Anatoly Tsirlin
- ... and a bunch of others

FTT problem

Find the metric matrix $-D^2P = -\frac{\partial^2 S}{\partial X_i \partial X_j}$

in statistical mechanics, i.e. based on

$$S = -\sum_i p_i \ln p_i$$