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# OPTIMAL SUPPORT FOR THERMODYNAMIC PROCESSES

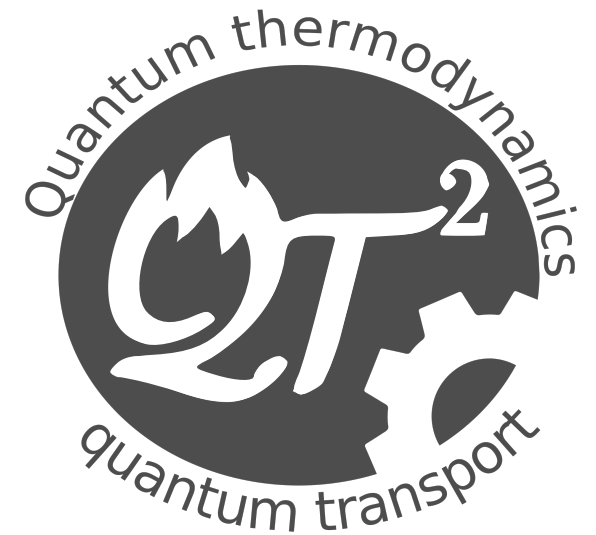
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IQOQI, Vienna.

Jan. 7th, 2020



IFUSP

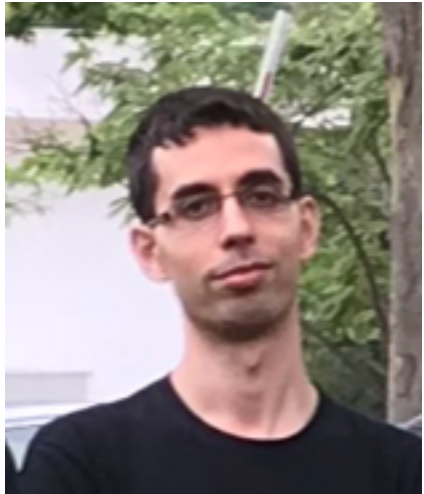
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# Summary

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André M. Timpanaro, Giacomo Guarnieri, John Goold, GTL,  
“**Thermodynamic uncertainty relations from exchange fluctuation theorems**”.  
*Phys. Rev. Lett.* **123**, 090604 (2019) (arXiv 1904.07574)

André M. Timpanaro, Jader P. Santos and GTL,  
“**Landauer’s principle at zero temperature**”  
(arXiv 1911.00910)

GTL, Giacomo Guarnieri, John Goold and André M. Timpanaro,  
“**Optimal support for thermodynamic processes**”  
(in preparation)

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# BOUNDS AND OPTIMIZATION

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- *MaxEnt principle*: which distribution maximizes the entropy for a fixed energy?

$$\tilde{S}[\mathbf{p}] = S[\mathbf{p}] + \alpha \left( 1 - \sum_n p_n \right) + \beta \left( U - \sum_n E_n p_n \right) \qquad S[\mathbf{p}] = - \sum_n p_n \ln p_n$$

- Answer: the equilibrium distribution:  $p_n^{\text{eq}} = e^{-\beta E_n} / Z$
- The optimization also leads to a *bound*:  $S[\mathbf{p}] \leq S[\mathbf{p}_{\text{eq}}]$



- *Landauer's principle*: what is the minimum amount of heat required to erase some information?

$$\Delta Q_{\text{env}} \geq -T \Delta S_{\text{sys}}$$

- The optimal way of accomplishing this is through a reversible process.
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# OPTIMAL SUPPORT

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The support of a distribution  $P(x)$  is defined as the set of points  $x_i$  at which  $P(x_i) \neq 0$ .

- Entropy does not care about support: it is a function only of the distribution.
- An example of a quantity that cares about the support is the variance

$$\text{var}(X) = \sum_i \left( x_i - \mathbb{E}(X) \right)^2 P(x_i)$$

- By changing the support, one can *increase* the variance and simultaneously *decrease* the entropy.
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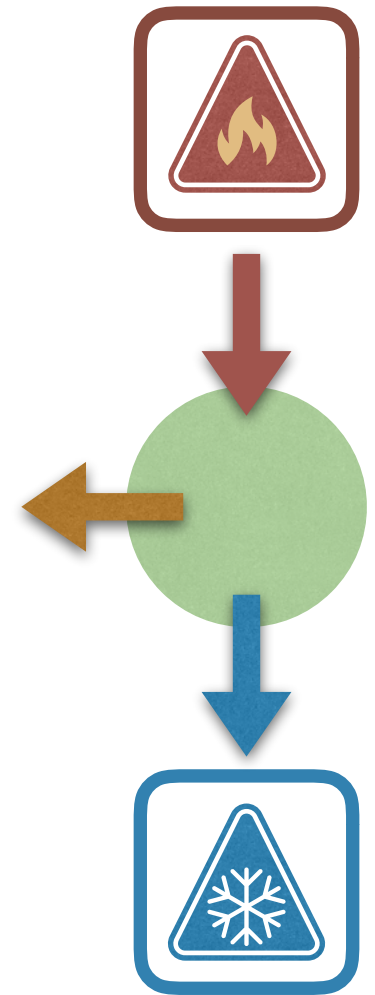


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# OPTIMIZATION vs. SUPPORT OPTIMIZATION

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- Suppose we wish to optimize a heat engine.
  - e.g., maximize the efficiency for a fixed output power.
- *Optimization* means tweaking the parameters of the engine.
- *Support optimization* means choosing the best working fluid.
  - i.e., change the energy levels of the system.



# EXAMPLE: THERMOELECTRIC ENGINE

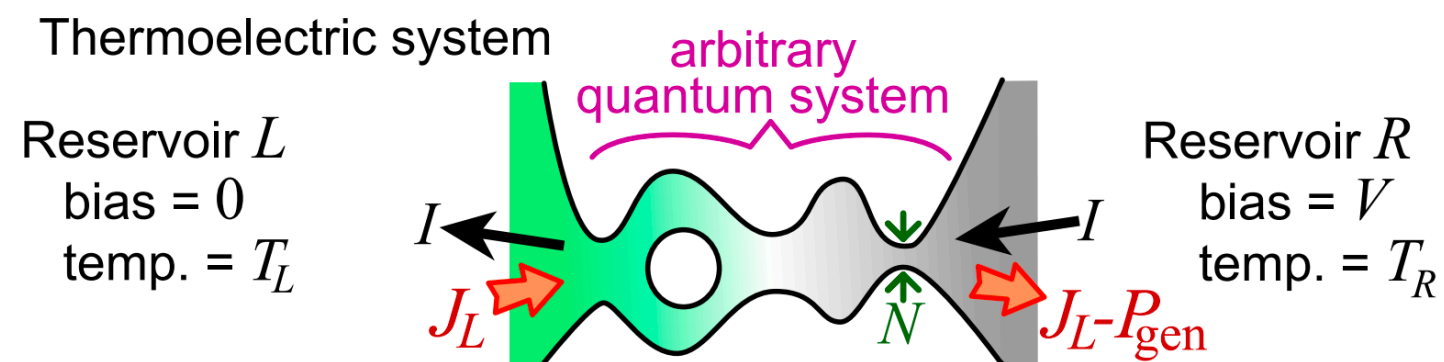
- In a thermoelectric the “working fluid” is characterized by a transmission function instead:

$$J_L = \int d\epsilon \mathcal{T}(\epsilon) \epsilon \left[ f_L(\epsilon) - f_R(\epsilon) \right] \quad P_{\text{gen}} = \int d\epsilon \mathcal{T}(\epsilon) V \left[ f_L(\epsilon) - f_R(\epsilon) \right]$$

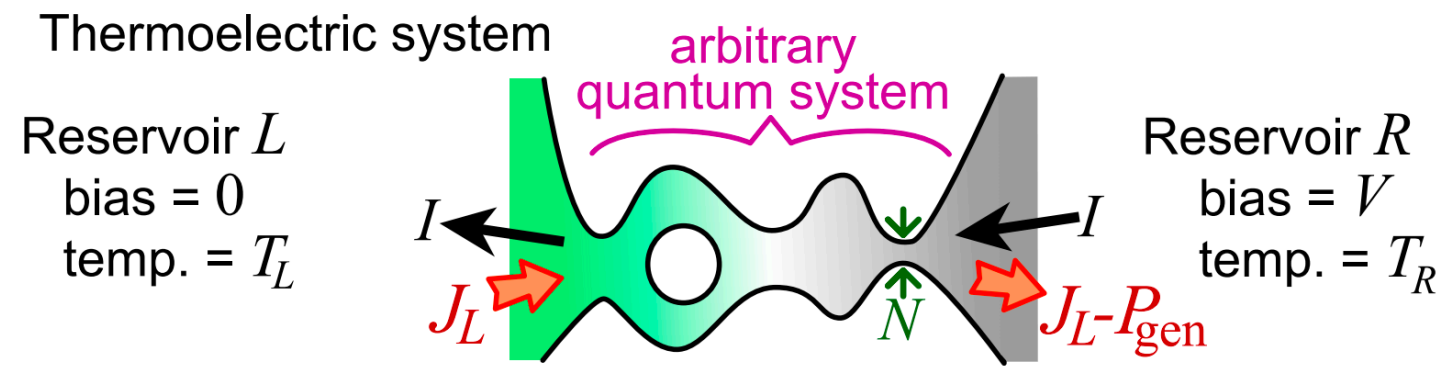
- Which thermoelectric has the best possible efficiency  $\eta = P_{\text{gen}}/J_L$ ?

- Answer: a delta transmission function  $\mathcal{T}(\epsilon) = N \delta\left(\epsilon - V(1 - T_R/T_L)\right)$

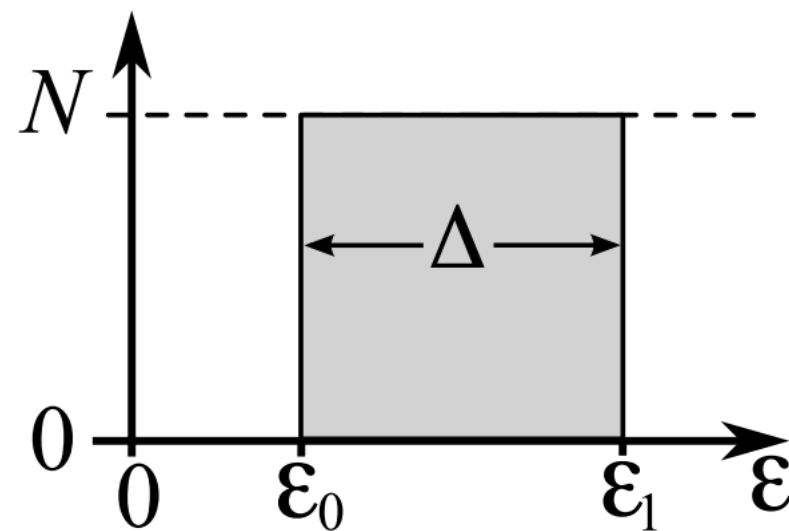
- It operates at Carnot efficiency,  $\eta = 1 - T_R/T_L$ , but with zero power.



Taken from  
R. Whitney, *PRL*  
**112**, 130601 (2014)



- Which thermoelectric has the best possible efficiency for a fixed output power  $P_{\text{gen}}$ ?
- The answer is a boxcar function



$$\mathcal{T}(\epsilon) = N \theta(\epsilon - \epsilon_0) \theta(\epsilon_1 - \epsilon)$$

$$\epsilon_0 = V(1 - T_R/T_L)$$

$$\epsilon_1 \text{ determined by } P_{\text{gen}}$$

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# EXAMPLE: OPTIMAL WORK EXTRACTION

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- Consider the *single-shot* extraction of work from a thermodynamic system.
- This process is stochastic and thus must be characterized by a work distribution  $P(W)$ , restricted by the Jarzynski equality

$$\mathbb{E}(e^{-\beta W}) = e^{-\beta \Delta F}$$

- *Which kind of system/process allows me to extract the maximum amount of work?*
- Since we are at the single-shot level, this is best formulated as:
  - Which process maximizes  $P(W \geq \Lambda)$ , for some given  $\Lambda$ ?

- 
- This question turns out to be ill-posed:
    - The Jarzynski equality is not a strong enough constraint.
    - You can always extract an infinite amount of work in principle.
  - But suppose we add the additional constraint that  $P(W < W_{\min}) = 0$

- Cavina *et. al.* found that

$$P(W \geq \Lambda) = \frac{e^{-\beta\Delta F} - e^{\beta W_{\min}}}{e^{\beta\Lambda} - e^{\beta W_{\min}}}$$

- In the unrestricted regime  $W_{\min} \rightarrow \infty$  this reduces to

$$P(W \geq \Lambda) = e^{-\beta(\Delta F - \Lambda)}$$

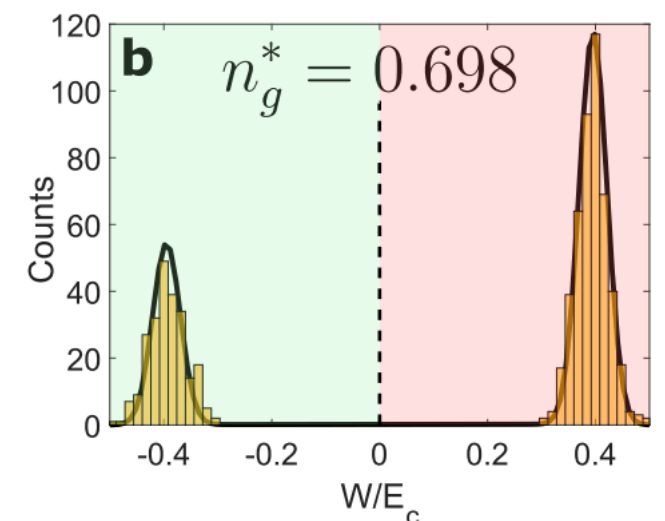
$$P(W \geq \Lambda) = e^{-\beta(\Delta F - \Lambda)}$$

- This result is pretty cool.
  - It implies that in the single-shot scenario, the probability of extracting work above  $\Delta F$  is exponentially small.
- This bound was already known since 2008. The novel result of Cavina *et. al.* is that they also found the *optimal process*:

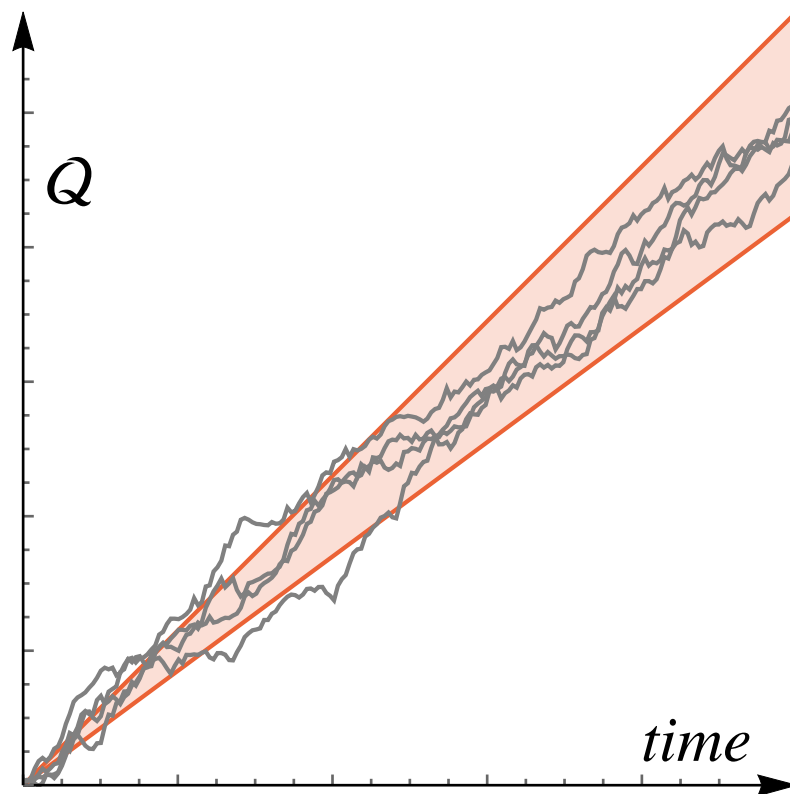
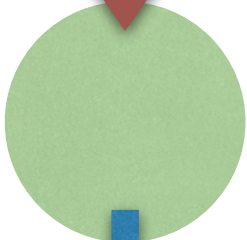
$$P(W) = p\delta(W - \Lambda) + (1 - p)\delta(W - W_{\min})$$

$$p = \frac{e^{-\beta\Delta F} - e^{\beta W_{\min}}}{e^{\beta\Lambda} - e^{\beta W_{\min}}}$$

C. Jarzynski, *European Physical Journal B*, **64**, 331 (2008).  
 O. Maillet, *PRL*, **122**, 150604 (2019)



# Thermodynamic Uncertainty Relations (TURs)



$$\frac{\text{var}(Q)}{\mathbb{E}(Q)^2} \geq \frac{2}{\mathbb{E}(\Sigma)}$$

$$\Sigma = \delta\beta Q \text{ (in the simplest case)}$$

- Simple, elegant and powerful.
- Counterintuitive: To reduce the fluctuations, the process should be *more irreversible*.



- Derived only for the steady-state of classical Markov chains.
- Can be violated in many relevant scenarios (e.g. thermoelectrics).

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# FLUCTUATIONS IN A HEAT ENGINE

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- As an example of the applicability of TURs, Pietzonka and Seifert showed that the output power in a heat engine is bounded by

$$\text{var}(P_{\text{gen}}) \geq 2T_c P_{\text{gen}} \frac{\eta}{\eta_C - \eta}$$

- For a fixed *average* power, the fluctuations go up if we approach Carnot efficiency.
  - To reduce fluctuations, one should operate away from Carnot efficiency.

**Fluctuations therefore appear as an additional property to take into account when optimizing heat devices.**



# TUR from fluctuation theorems

André M. Timpanaro, Giacomo Guarnieri, John Goold, GTL,  
“**Thermodynamic uncertainty relations from exchange fluctuation theorems**”.  
*Phys. Rev. Lett.* **123**, 090604 (2019) (arXiv 1904.07574)



# EXCHANGE FLUCTUATION THEOREM

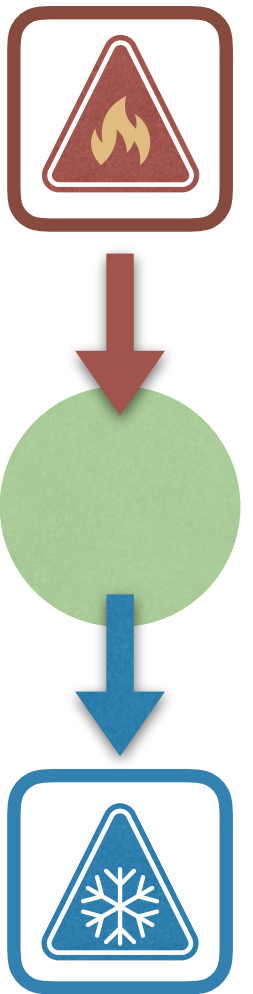
- Fluctuation theorems for thermodynamic processes usually have the form

$$\frac{P_F(\Sigma)}{P_B(-\Sigma)} = e^\Sigma$$

- e.g. Crooks theorem for work:  $\Sigma = \beta(W - \Delta F)$
- FTs, however, compare a *forward* with a *backward* process.
- In some systems, both coincide. These are called *Exchange FTs*:

$$\frac{P(\Sigma)}{P(-\Sigma)} = e^\Sigma$$

- This is *much stronger*: it is a symmetry on a single probability distribution.
- Example: direct heat exchange:  $\Sigma = \delta\beta Q$



- Motivated by this, we proved the following theorem:

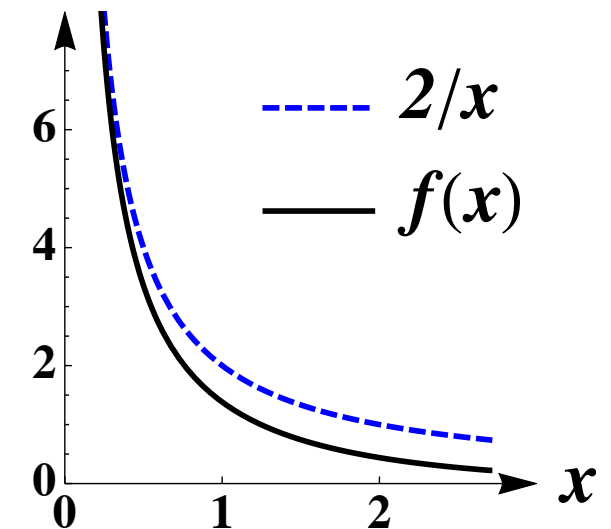
**Theorem** (“TUR de force”). For fixed finite  $\mathbb{E}(\Sigma)$ , the probability distribution  $P(\Sigma)$  satisfying  $P(\Sigma)/P(-\Sigma) = e^\Sigma$ , with the smallest possible variance (the minimal distribution) is

$$P_{min}(\Sigma) = \frac{1}{2 \cosh(a/2)} \left\{ e^{a/2} \delta(\Sigma - a) + e^{-a/2} \delta(\Sigma + a) \right\},$$

where the value of  $a$  is fixed by  $\mathbb{E}(\Sigma) = a \tanh(a/2)$ .  
For this distribution

$$\text{Var}(\Sigma)_{min} = \mathbb{E}(\Sigma)^2 f(\mathbb{E}(\Sigma)),$$

where  $f(x) = \text{csch}^2(g(x/2))$ ,  $\text{csch}(x)$  is the hyperbolic cosecant and  $g(x)$  is the function inverse of  $x \tanh(x)$ .



For any other distribution we must then have:

$$\frac{\text{var}(\Sigma)}{\mathbb{E}(\Sigma)^2} \geq f(\mathbb{E}(\Sigma))$$

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# TUR de force IS TIGHT

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- Our TUR is the tightest (saturable) bound for this scenario.
- And we know which thermodynamic process saturates it.
- This is relevant because, around the same time, similar papers appeared.
  - But all derived a looser bound with

$$f(x) = \frac{2}{e^x - 1}$$

- This bound, however, is never tight.

Hasegawa & Vu 1902.06376.

Proesman & Horowitz 1902.07008.

Potts & Samuelsoon 1904.04913.

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# EXTENSION TO MULTIPLE CHARGES

- We can also generalize our framework to Exchange FTs involving multiple charges:

$$\frac{P(Q_1, \dots, Q_n)}{P(-Q_1, \dots, -Q_n)} = e^{\sum_i A_i Q_i}$$

- The entropy production in this case is  $\Sigma = \sum_i A_i Q_i$

- ex: heat engine FT:

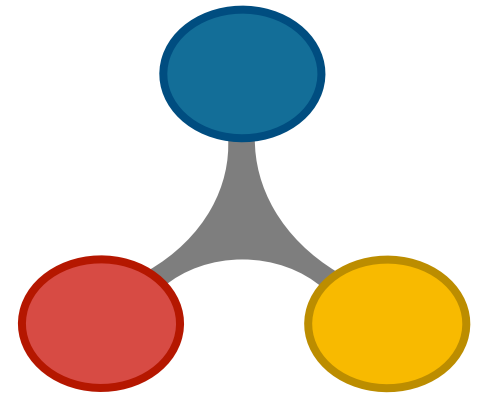
$$\frac{P(Q_h, W)}{P(-Q_h, -W)} = e^{(\beta_h - \beta_c)Q_h + \beta_c W}$$

- In this case we obtain the matrix bound

$$\mathcal{C} - f(\mathbb{E}(\Sigma))\mathbf{q}\mathbf{q}^T \geq 0$$

$$q_i = \mathbb{E}(Q_i)$$

$$C_{ij} = \text{cov}(Q_i, Q_j)$$



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$$\mathcal{C} - f(\mathbb{E}(\Sigma))\mathbf{q}\mathbf{q}^T \geq 0$$

$$q_i = \mathbb{E}(Q_i)$$

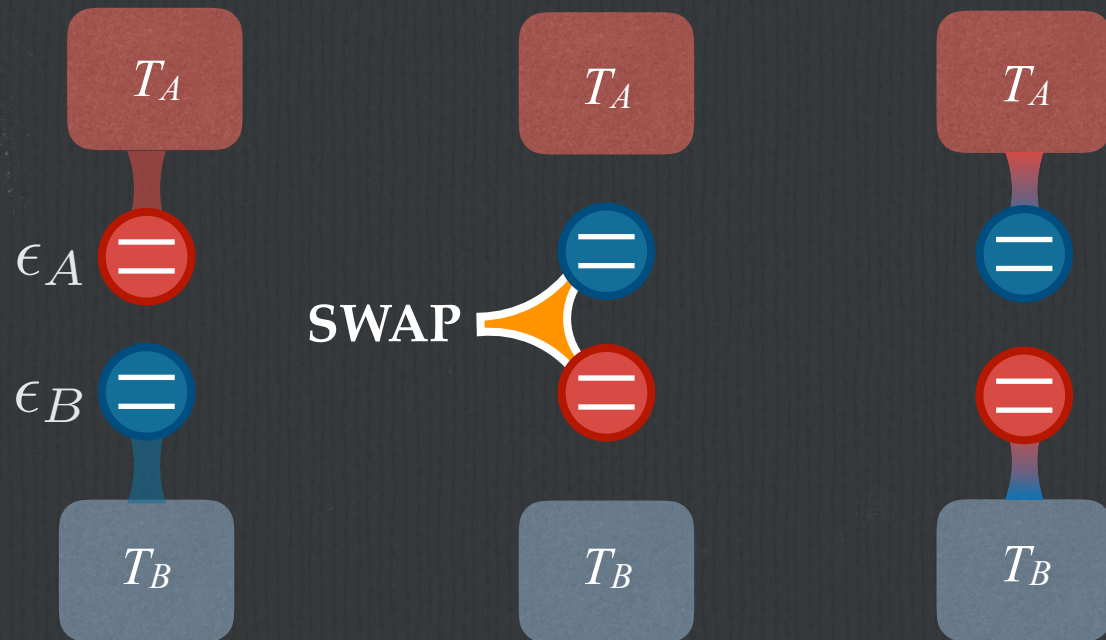
$$C_{ij} = \text{cov}(Q_i, Q_j)$$

- This says that the matrix above is positive semi-definite.
- As a consequence, all diagonal entries must be positive, which implies an individual TUR for each charge:

$$\frac{\text{var}(Q_i)}{\mathbb{E}(Q_i)^2} \geq f(\mathbb{E}(\Sigma))$$

- In addition, it also places restrictions on the covariances:
    - If  $G$  is psd then  $G_{ij}^2 \leq G_{ii}G_{jj}$
  - Correlations between thermodynamic quantities has so far been largely unexplored.
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# SWAP engine

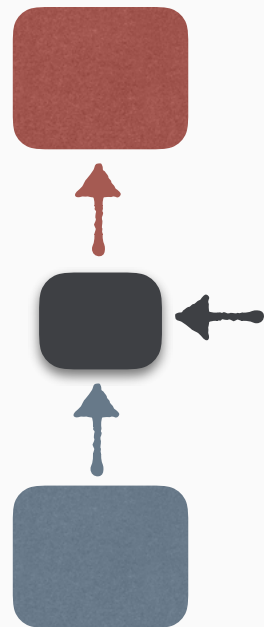


$$\langle Q_h \rangle = \epsilon_A (f_A - f_B)$$

$$\langle Q_c \rangle = -\epsilon_B (f_A - f_B)$$

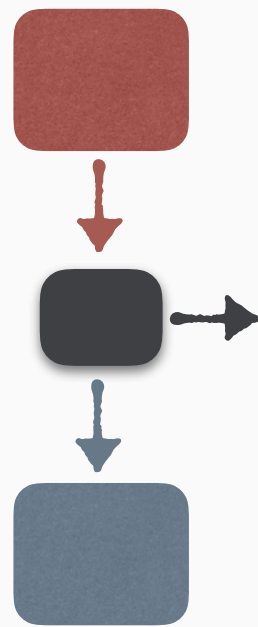
$$\langle W \rangle = -(\epsilon_A - \epsilon_B)(f_A - f_B)$$

$$f_i = \frac{1}{e^{\beta_i \epsilon_i} + 1}$$



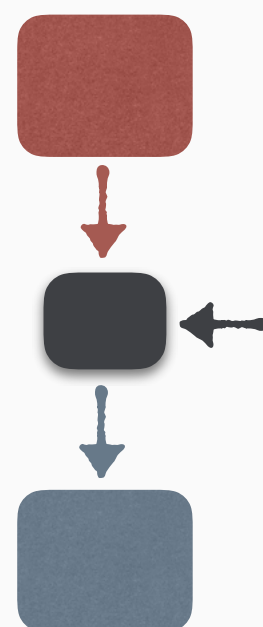
**Refrigerator**

$$\frac{\epsilon_B}{\epsilon_A} < \frac{\beta_A}{\beta_B}$$



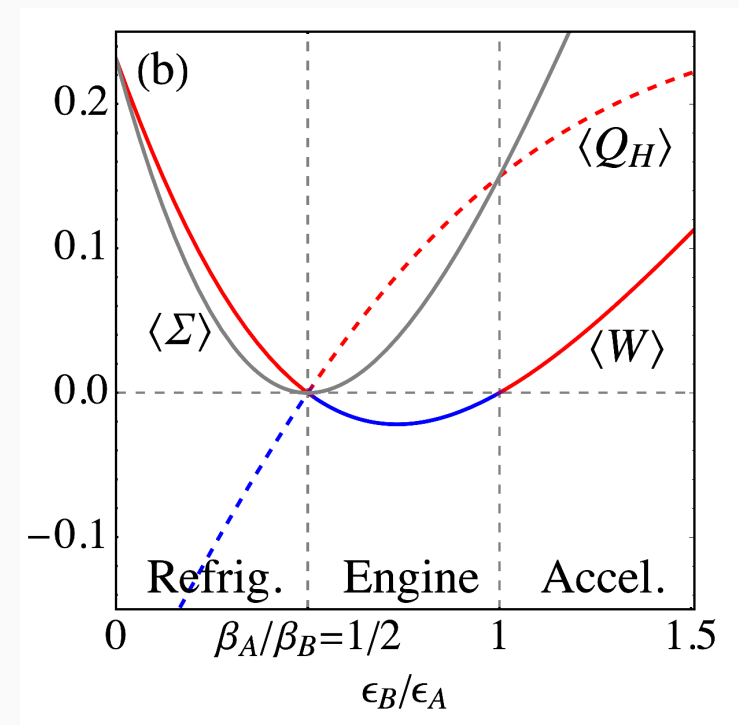
**Engine**

$$\frac{\beta_A}{\beta_B} < \frac{\epsilon_B}{\epsilon_A} < 1$$



**Accelerator**

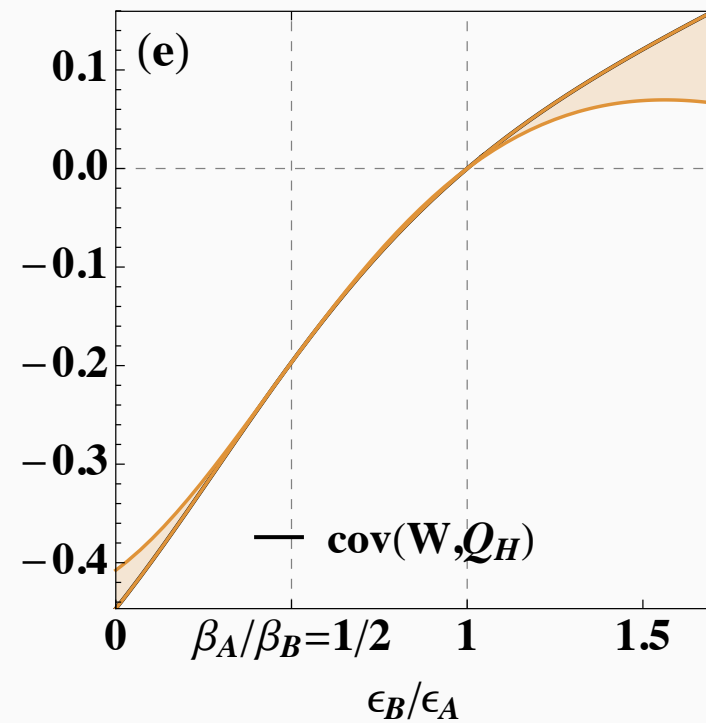
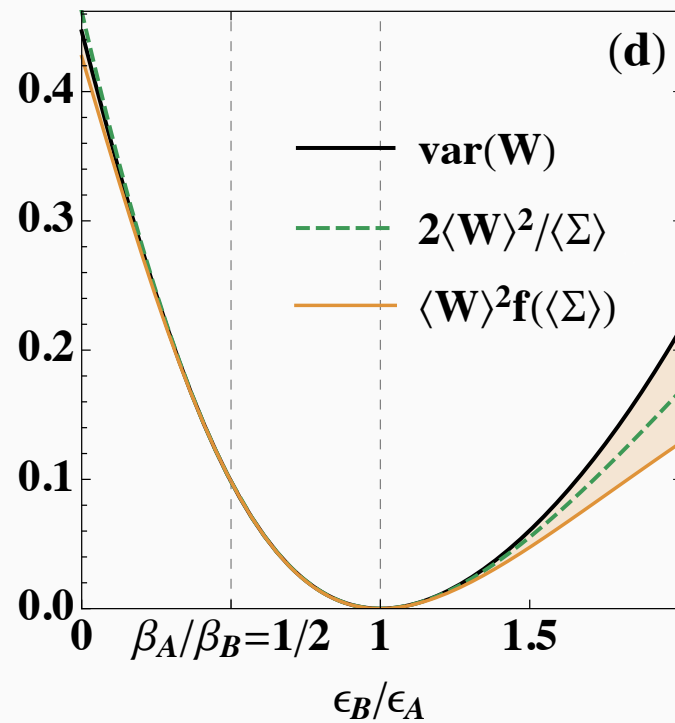
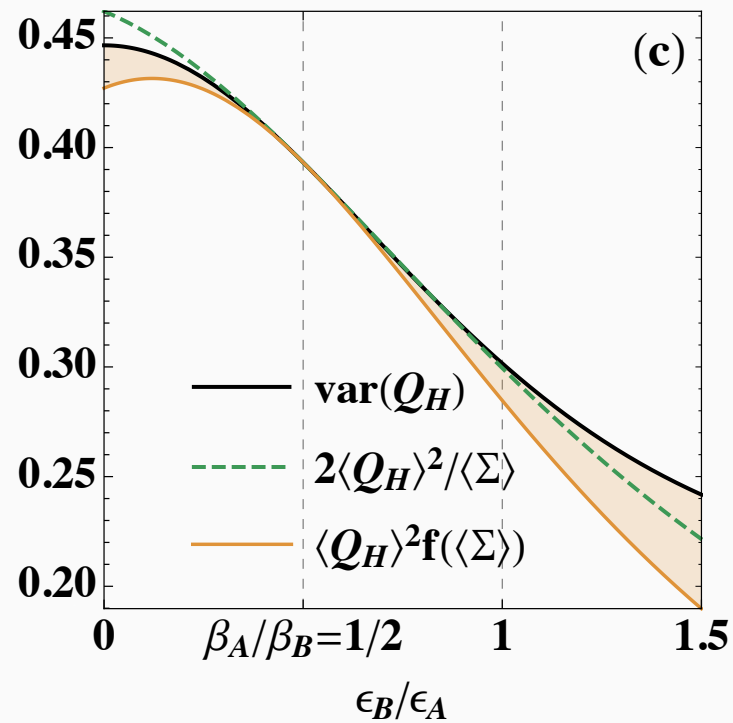
$$1 < \frac{\epsilon_B}{\epsilon_A}$$





# SWAP engine

$$\frac{P(Q_H, W)}{P(-Q_H, -W)} = e^{(\beta_B - \beta_A)Q_H + \beta_B W}$$





# Landauer's principle at zero temperature

André M. Timpanaro, Jader P. Santos and GTL,  
“**Landauer's principle at zero temperature**”  
(arXiv 1911.00910)



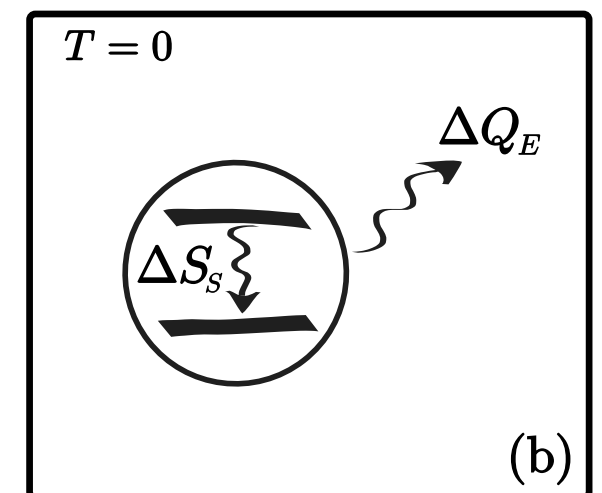
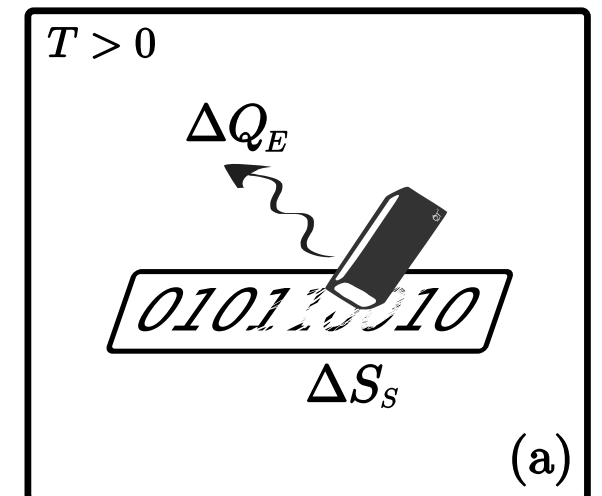
- *Landauer's principle*: what is the minimum amount of heat required to erase some information?

$$\Delta Q_{\text{env}} \geq -T \Delta S_{\text{sys}}$$

- If the environment is at zero temperature, the bound becomes trivial:

$$\Delta Q_{\text{env}} \geq 0$$

- But is this tight?
- Consider, for instance, *spontaneous emission*:
  - Can we really erase information about a system at zero heat cost?



- 
- We have recently derived a new Landauer bound which remains valid even at  $T = 0$ .
  - Our bound follows exactly the same spirit of the original one:
    - It says absolutely nothing about the state of the system;
    - nor the system-environment interaction.
    - All it assumes is that the environment is in thermal equilibrium.

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- The result is:

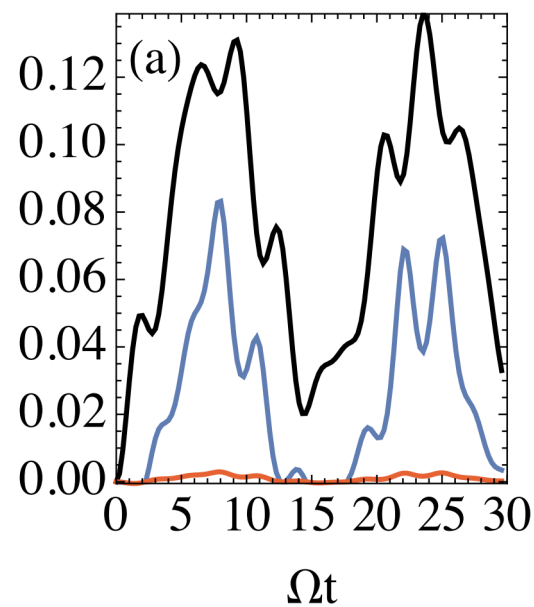
$$\Delta Q_{\text{env}} \geq \mathcal{U}(\mathcal{S}^{-1}(-\Delta S_{\text{sys}}))$$

- where  $\mathcal{U}$  and  $\mathcal{S}$  are proportional to the *equilibrium* energy and entropy of the environment.
- Our bound does require more information than the original one (which only requires knowledge of the temperature  $T$ ).
- However, it is also *always tighter than the original one*.
- Moreover, it tends to it in the high temperature limit.

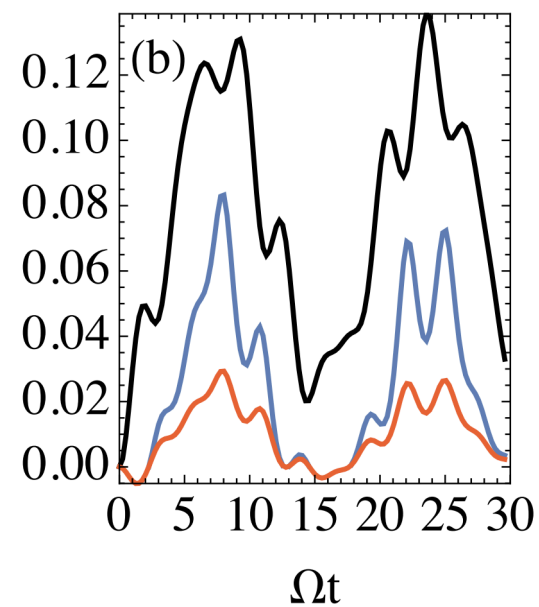
- Most importantly, our bound remains useful at zero temperature.
- Ex: when the bath is a  $1D$  waveguide

$$\Delta Q_{\text{env}} \geq -T \Delta S_{\text{sys}} + \frac{3\hbar c}{\pi L} \Delta S_{\text{sys}}^2$$

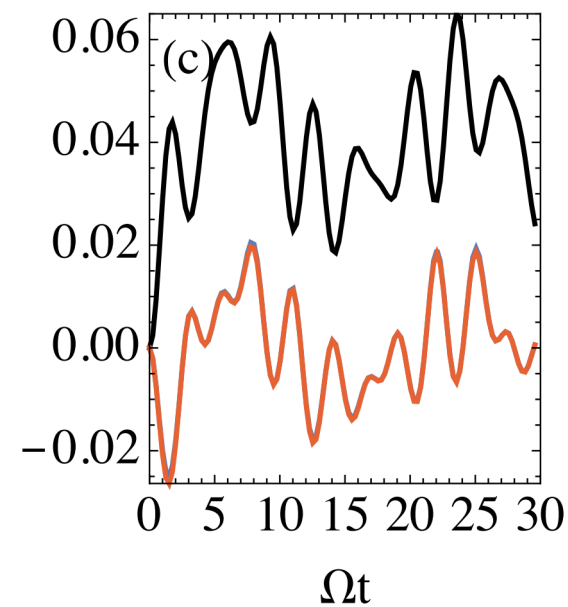
- where  $L$  is the length of the waveguide.



T cold



T medium



T hot



# Conclusions

Acknowledgements:  
IFUSP, FAPESP, CNPq

- In this talk I tried to discuss the idea of **optimizing the support in thermodynamic processes**.
- I feel that this is important because it sheds light on:
  - a. What are the ultimate limits;
  - b. Which sorts of processes are possible;
  - c. What are the ideal processes and machines;



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