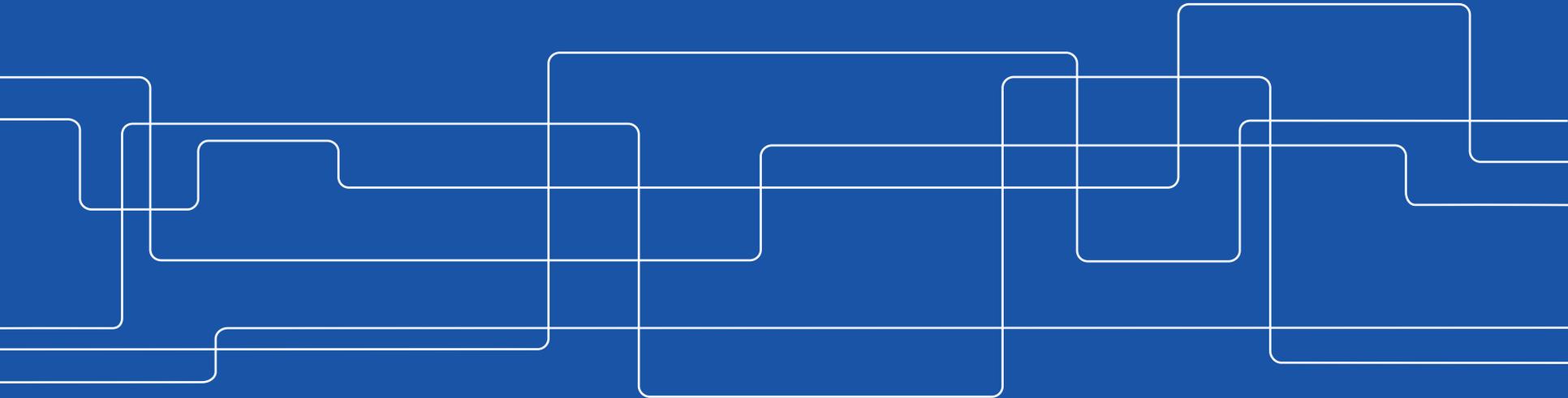




# Security and Limitations of Cyber-Physical Systems

## Lecture 4: Implementation Costs and Information Flow in Kalman-Bucy Filters

Henrik Sandberg,  
Jean-Charles Delvenne, Nigel J. Newton, Sanjoy K. Mitter





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# Course Outline

- Monday (8:30-10:30):
  - Lecture 1 (HS): Introduction, data attacks against non-dynamic systems, power network monitoring, security index, graph min cut
- Tuesday
  - 8:30-12:30:
    - Lecture 2 (HS): Attack space for cyber-physical systems: DoS, undetectable, stealth, covert, bias, replay attacks
    - Lecture 3 (AT): Defense mechanisms, risk management, anomaly detectors, watermarking
  - 15:30-16:30:
    - Exercise session (Graph min cut, security index)
- Wednesday (8:30-10:30):
  - **Lecture 4 (HS): Physical limits of control implementations**
  - Exercise session (open discussion, Q & A)



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# Key References for Lecture

- Henrik Sandberg, Jean-Charles Delvenne, Nigel J. Newton, Sanjoy K. Mitter: "Maximum work extraction and implementation costs for nonequilibrium Maxwell's demons". *Physical Review E*, 90, 042119, 2014.
  - Henrik Sandberg, Jean-Charles Delvenne, Nigel J. Newton, Sanjoy K. Mitter: "Thermodynamic Costs in Implementing Kalman-Bucy Filters". In *Proceedings of the 52nd Annual Allerton Conference on Communication, Control, and Computing*, Monticello, Illinois, October 2014.
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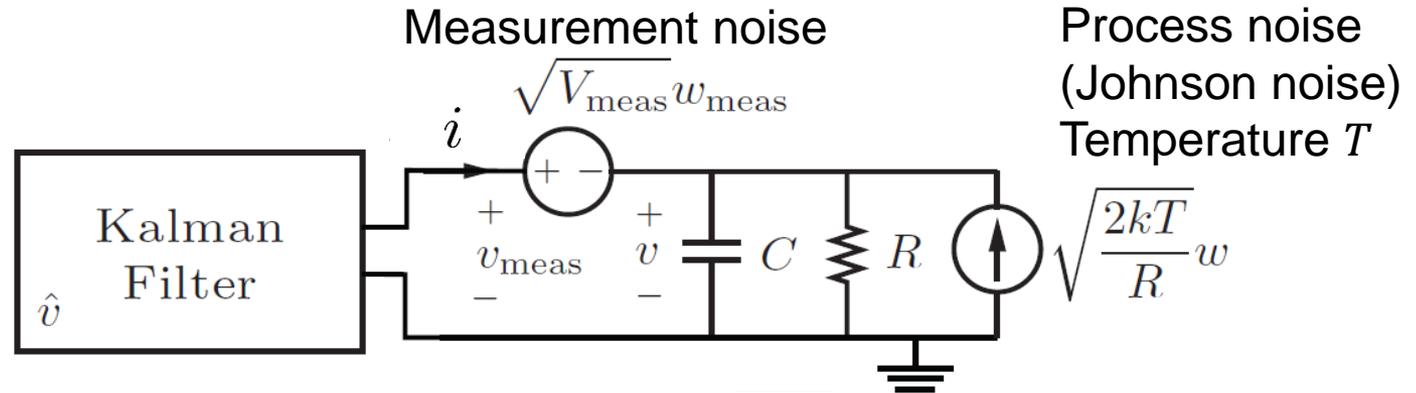


# Outline

- Motivating example
- Thermodynamics of the Kalman-Bucy filter
- Relation to the information flow and Landauer's principle
- (Maxwell's demon)

[Sandberg *et al.*, Allerton, 2014]

# Motivating Example and Problem Formulation

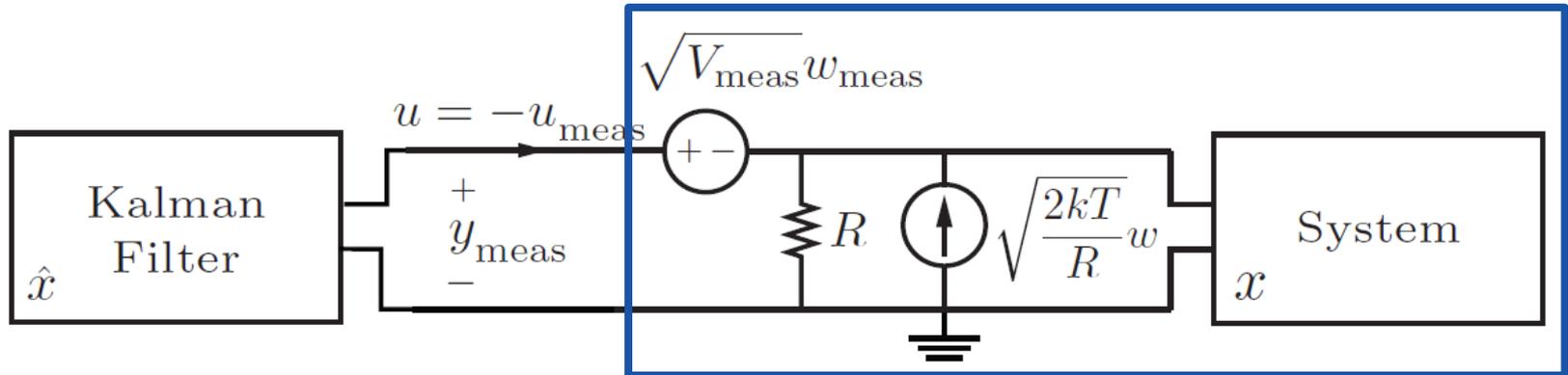


$$\dot{v} = -\frac{1}{RC}v + \frac{1}{R}i + \sqrt{\frac{2kT}{R}}w$$

$$v_{\text{meas}} = v + \sqrt{V_{\text{meas}}}w_{\text{meas}}$$

1. Is there a lower bound on external power supply to a physical implementation of the filter?
2. What is a simple cheap exact physical implementation of the Kalman-Bucy filter?

# More General Class of Systems to Measure



**Linear passive system (cf. Darlington synthesis)**

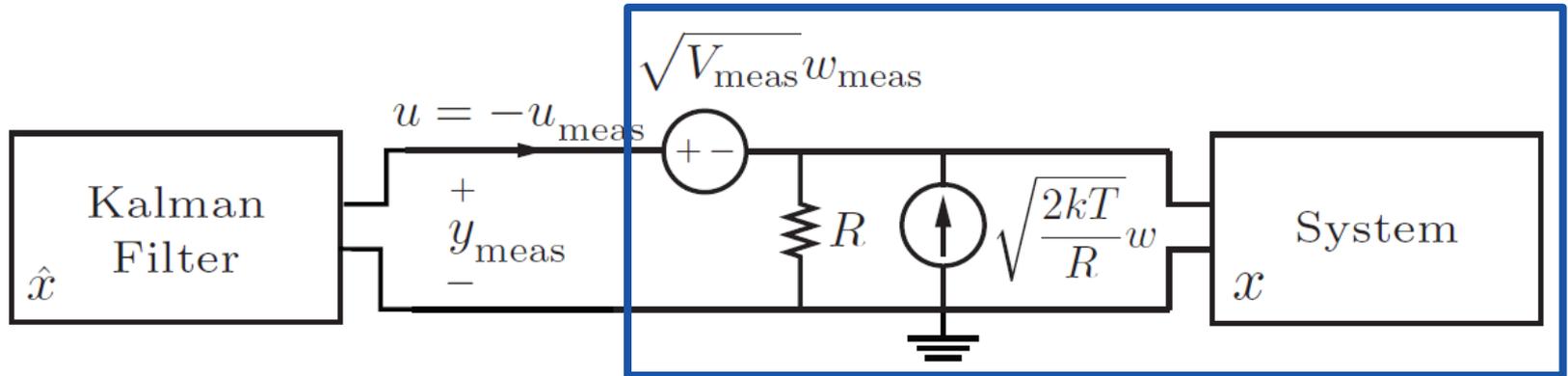
$$\dot{x} = (J - GBB^T)Mx + Bu + B\sqrt{2kTG}w$$

$$y = B^T Mx$$

$$y_{\text{meas}} = B^T Mx + \sqrt{V_{\text{meas}}}w_{\text{meas}}$$

$$(J = -J^T, \quad G := 1/R, \quad M = M^T > 0)$$

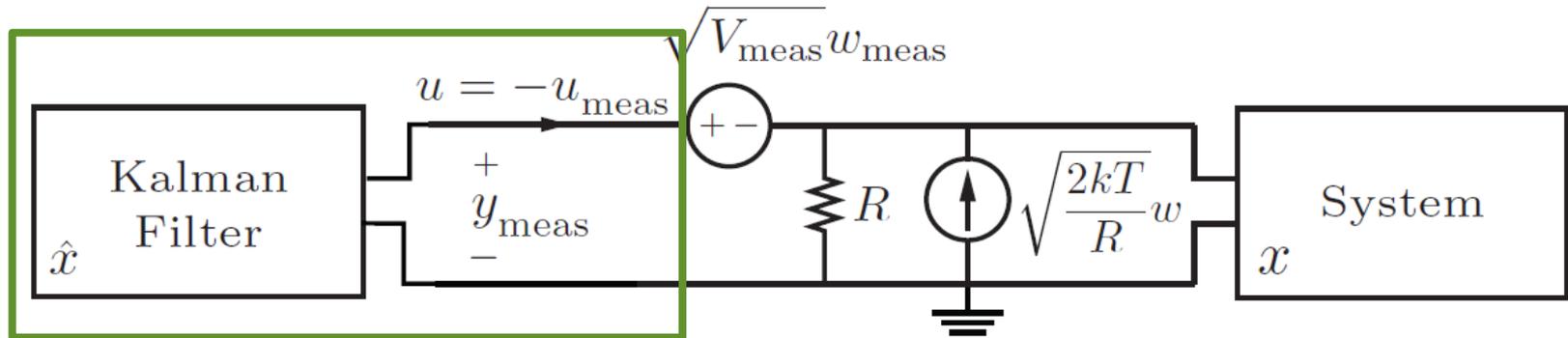
# More General Class of Systems to Measure



**Signal-to-noise ratio (SNR):**

$$\frac{(\text{process noise [V]})^2}{(\text{measurement noise [V]})^2} \equiv \frac{2kT}{GV_{\text{meas}}} =: \sigma$$

# Kalman-Bucy Filter



$$\frac{d}{dt}\hat{x} = (J - GBB^T)M\hat{x} + Bu + K(y_{\text{meas}} - B^T M\hat{x})$$

**Lemma:** Kalman gain is  
 $K = (\sqrt{1 + \sigma} - 1)GB \equiv g_K B$



# Kalman-Bucy Filter is Passive

**Assumption:** Admit linear back action/feedback current

$$\underbrace{u_{\text{meas}}}_{\text{current}} = -u = g \underbrace{B^T M \hat{x}}_{\text{voltage}}, \quad g = \text{gain} \in (0, \infty), \text{ free parameter}$$

**Theorem:** A realization of Kalman-Bucy filter is

$$\frac{d}{dt} \hat{x}_s = (J - Z B_s B_s^T) M \hat{x}_s + B_s y_{\text{meas}}$$

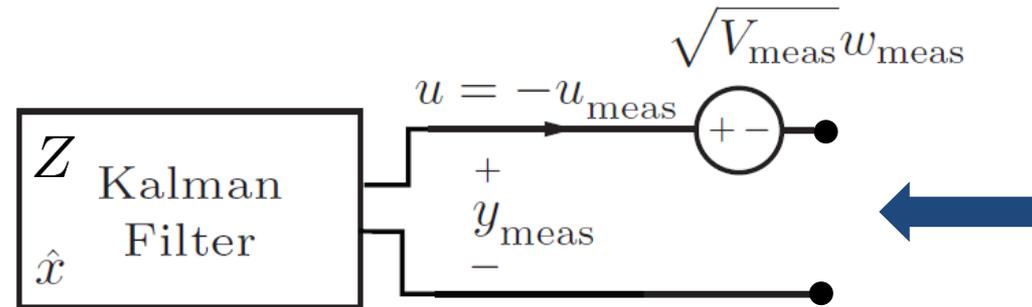
$$u_{\text{meas}} = B_s^T M \hat{x}_s$$

$$\hat{x}_s = \sqrt{g/g_K} \hat{x}$$

with effective resistance

$$Z = \frac{\sqrt{\sigma + 1}}{(\sqrt{\sigma + 1} - 1)g} + \frac{1}{(\sqrt{\sigma + 1} - 1)G} > 0$$

# Effective Temperature of Kalman-Bucy Filter



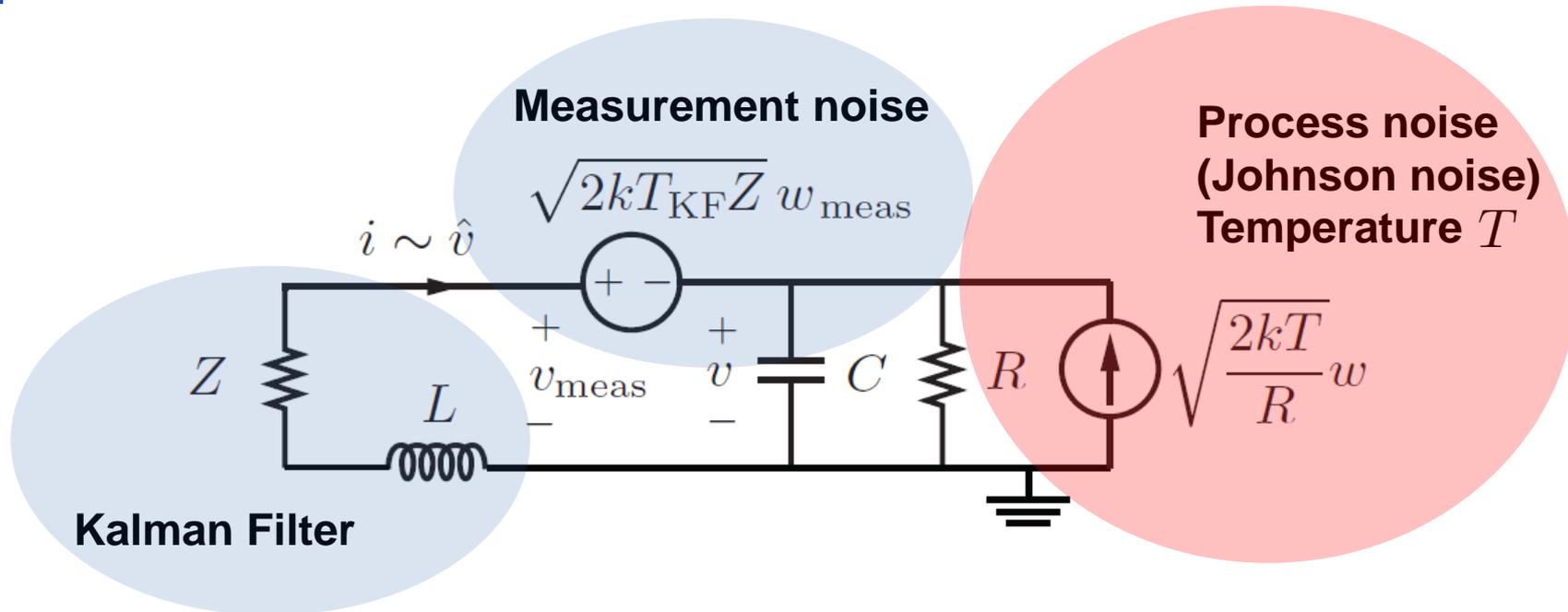
**Fluctuation-dissipation theorem:**  $2kT_{\text{KF}}Z = V_{\text{meas}}$

**Effective filter temperature:**

$$T_{\text{KF}} = \frac{T}{\sqrt{1 + \sigma} + 1} \frac{g}{g + g_K + G} < \frac{1}{2}T$$

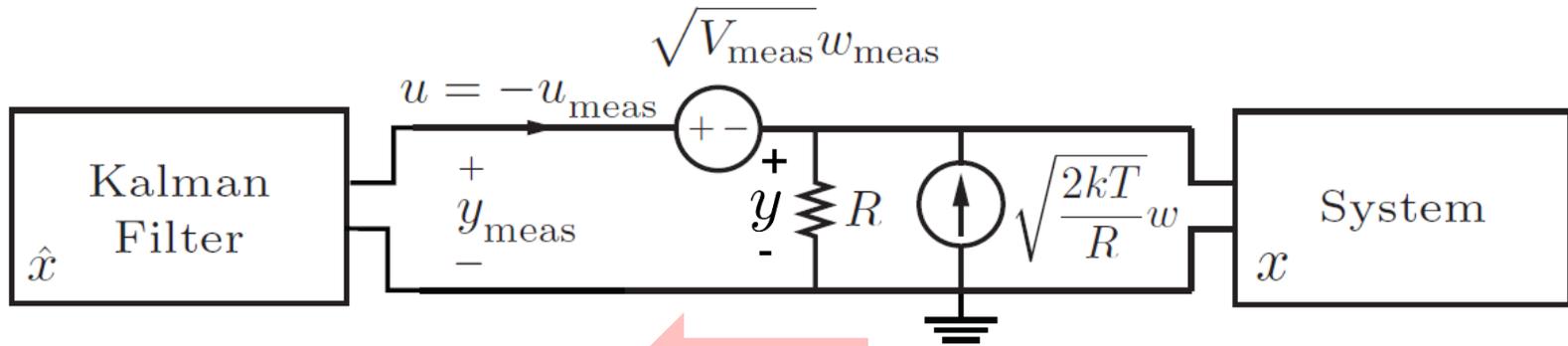
**Insight:** As back action/feedback tends to zero, filter temperature tends to zero

# Exact Filter Circuit for Motivating Example



- Externally equivalent implementation using passive non-zero temperature components
- $T_{KF} < \frac{1}{2}T$  (non-equilibrium thermodynamic system)

# Heat Flow



$$\dot{Q} := -\mathbf{E}[uy]$$

Expected stationary energy flow from system to filter:

$$\dot{Q} := -\mathbf{E}[uy] = \frac{g}{g + G} \dot{Q}_{\max}$$

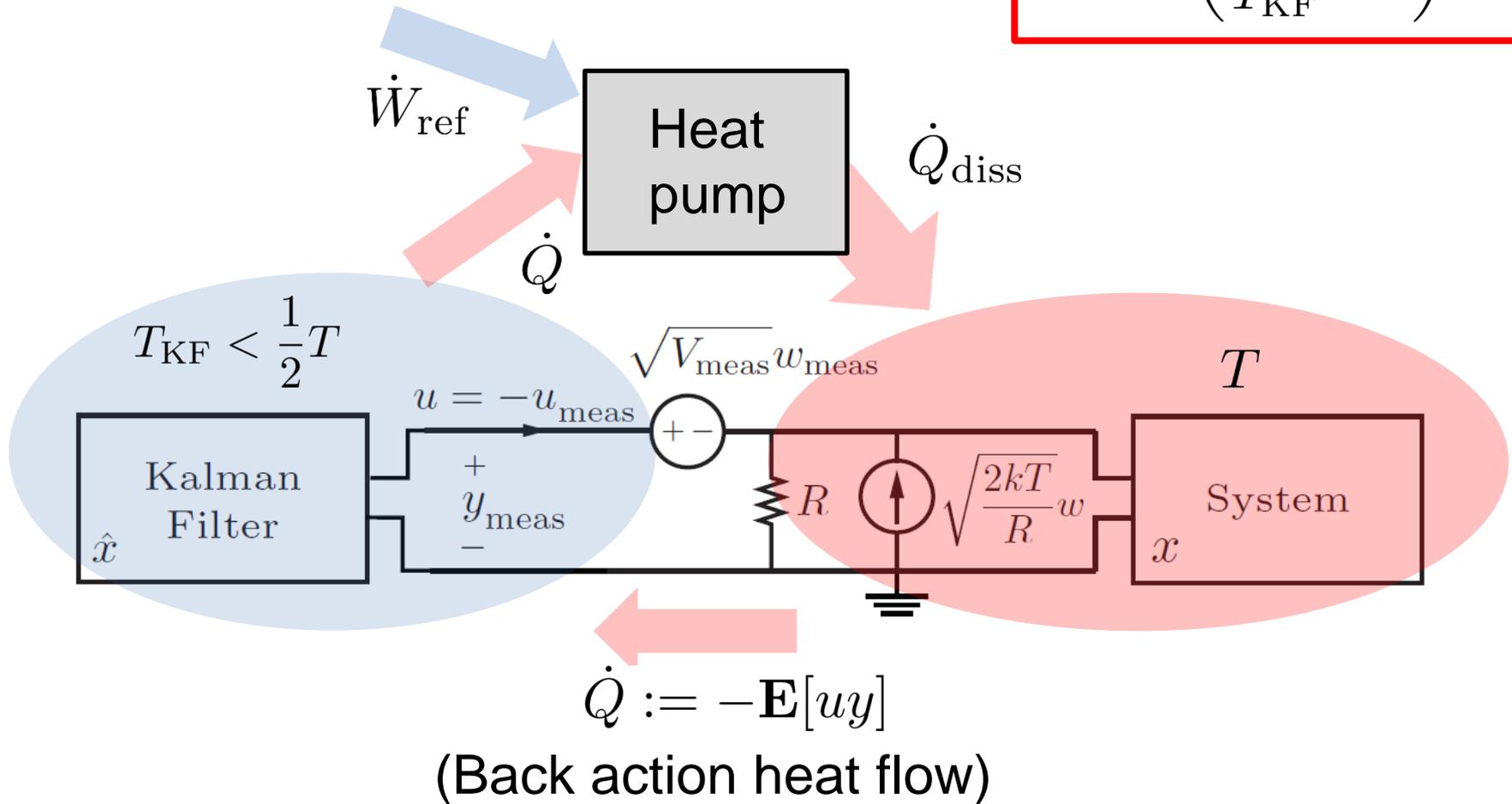
Maximum possible energy flow ( $g \rightarrow \infty$ ):

$$\dot{Q}_{\max} := kGB^T MBT \frac{\sqrt{1 + \sigma} - 1}{\sqrt{1 + \sigma} + 1}$$

# Heat and Work Flows

**2<sup>nd</sup> Law (Carnot):**

$$\dot{W}_{\text{ref}} \geq \left( \frac{T}{T_{\text{KF}}} - 1 \right) \dot{Q}$$



## Power Supply Required by the 2<sup>nd</sup> Law

$$\dot{W}_{\text{ref}} \geq \kappa T (1 + \sigma - \sqrt{1 + \sigma}) \left( 1 - \frac{\sqrt{1 + \sigma}}{1 + \sqrt{1 + \sigma}} \text{BA} \right)$$

Heat conductivity of system

$$kGB^T MB$$

Normalized back action

$$\frac{g}{g + G}$$

- Kalman-Bucy filter with no feedback:  $\text{BA} \rightarrow 0$
- Minimum-variance controller:  $\text{BA} \rightarrow 1$

# Power Supply Required by the 2<sup>nd</sup> Law

$$\dot{W}_{\text{ref}} \geq \kappa T (1 + \sigma - \sqrt{1 + \sigma}) \left( 1 - \frac{\sqrt{1 + \sigma}}{1 + \sqrt{1 + \sigma}} \text{BA} \right)$$

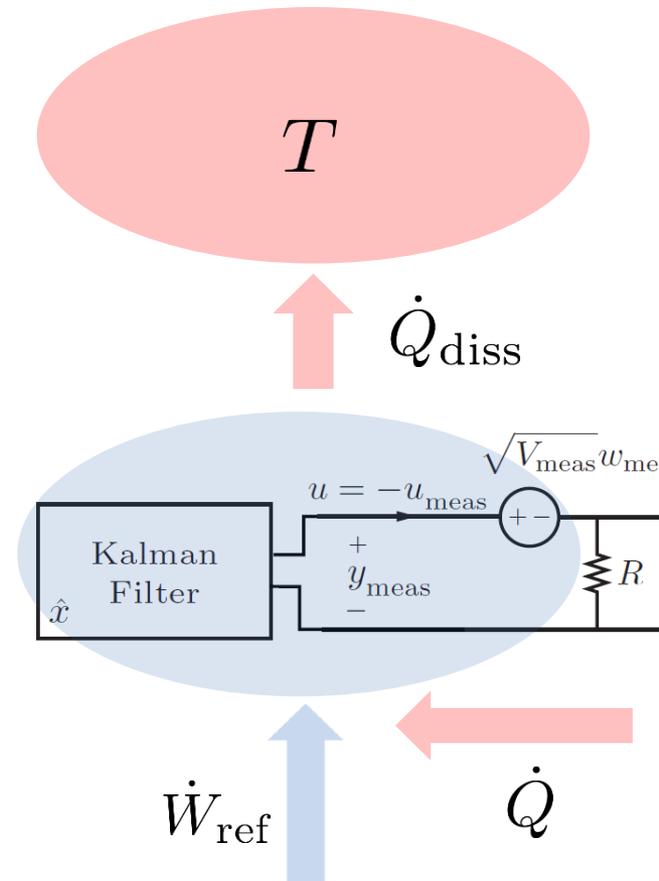
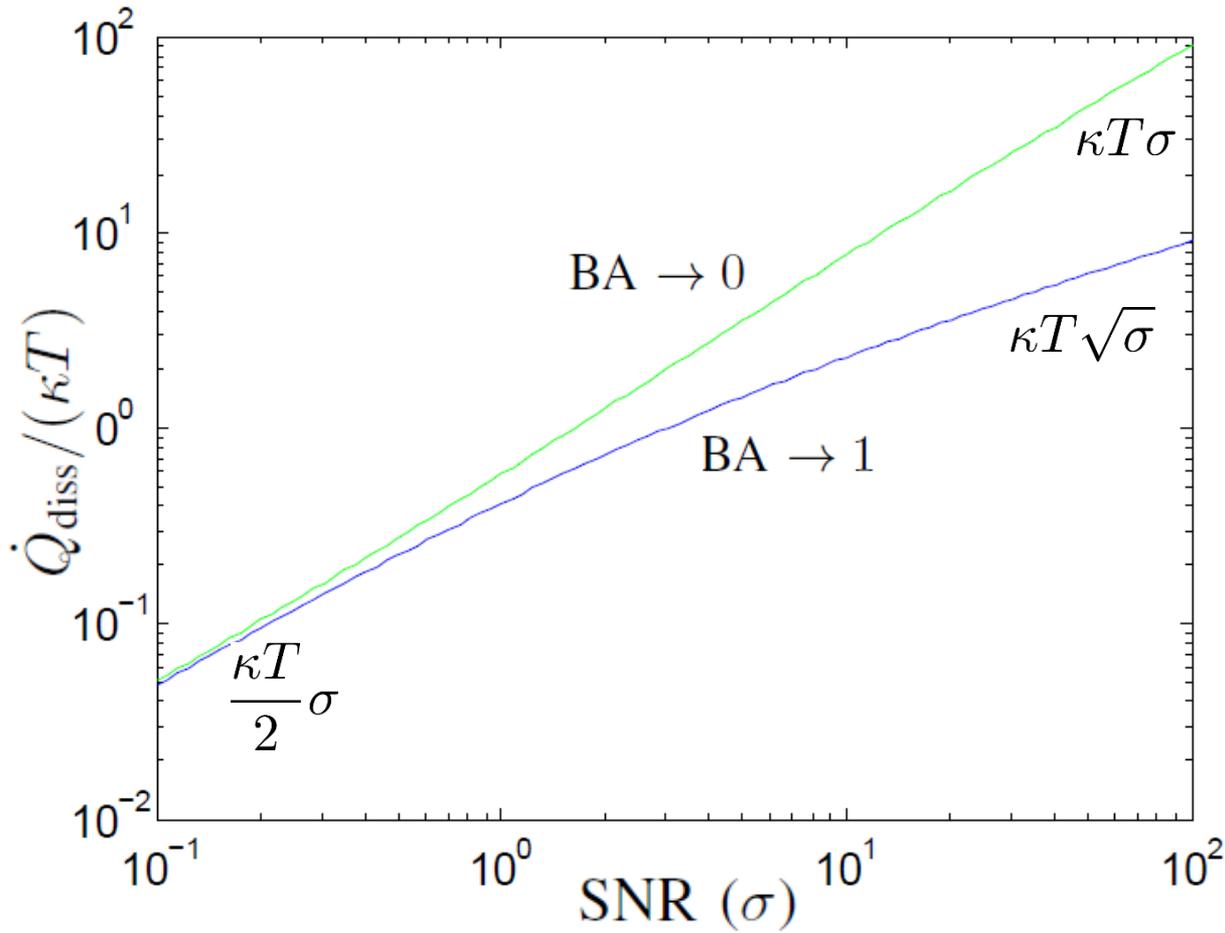
## No feedback (BA→0)

$$\dot{W}_{\text{ref}} \geq \kappa T (1 + \sigma - \sqrt{1 + \sigma}) \approx \begin{cases} \frac{\kappa T}{2} \times \text{SNR} & (\text{low SNR}) \\ \kappa T \times \text{SNR} & (\text{high SNR}) \end{cases}$$

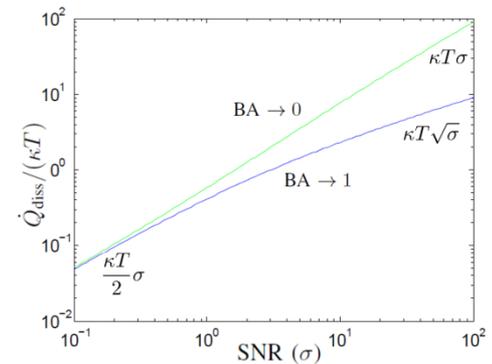
## Min-variance control (BA→1)

$$\dot{W}_{\text{ref}} \geq \kappa T \frac{1 + \sigma - \sqrt{1 + \sigma}}{1 + \sqrt{1 + \sigma}} \approx \begin{cases} \frac{\kappa T}{4} \times \text{SNR} & (\text{low SNR}) \\ \kappa T \times \sqrt{\text{SNR}} & (\text{high SNR}) \end{cases}$$

# Trade-Off: Power Dissipation vs. SNR

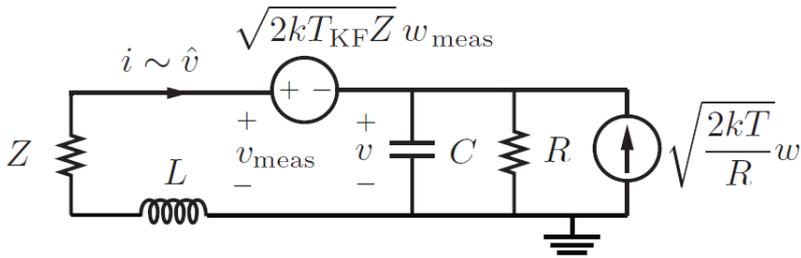


# Observations



- Large back action/feedback  $\Rightarrow$  smaller power dissipation (“energy harvesting”)
- **Explanation:** Temperature ratio  $T/T_{KF}$  smaller  $\Rightarrow$  2<sup>nd</sup> law less restrictive
- Difference significant for high SNRs
- No back action/feedback costs a factor  $\sqrt{\text{SNR}}$  more than min-variance control in high SNR regime

# Example: Power Supply for RC Circuit



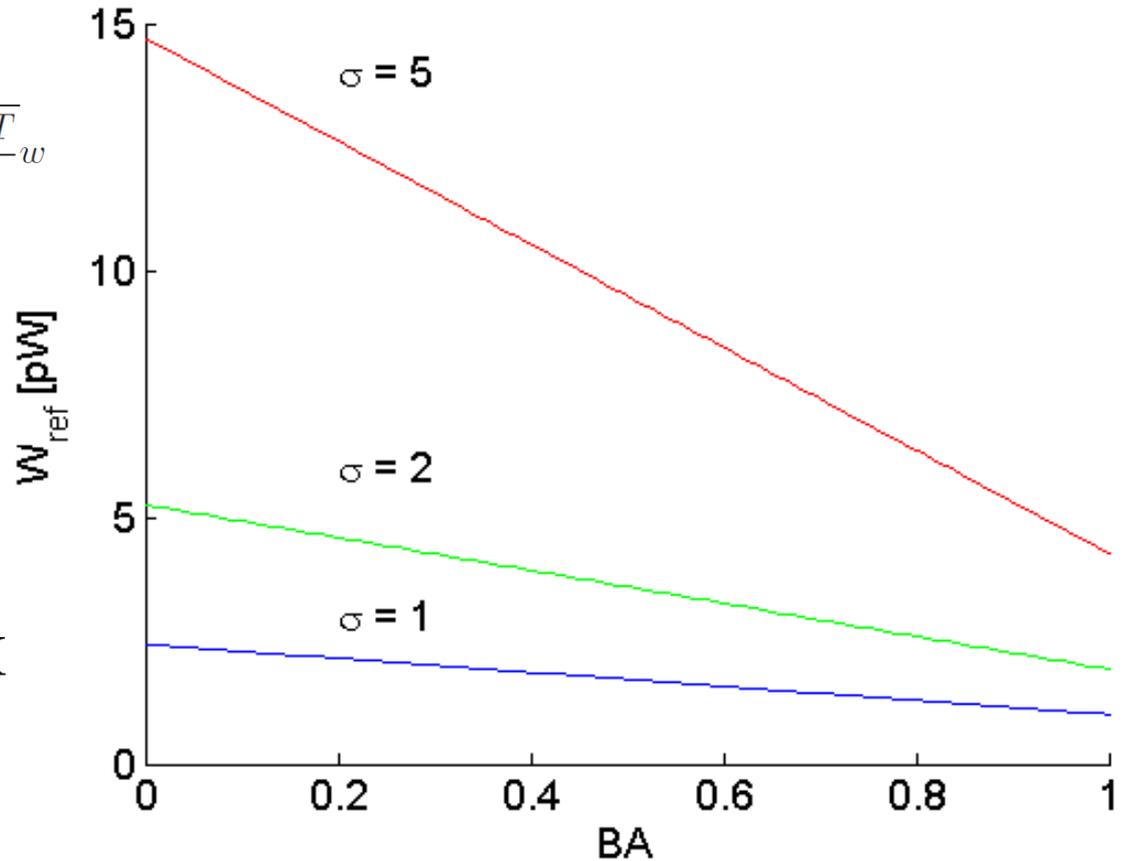
$$R = 1 \Omega$$

$$C = 1 \text{ nF}$$

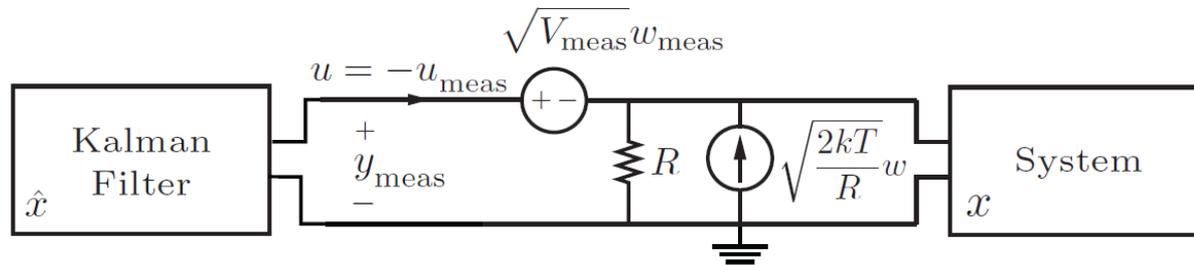
$$k = 1.38 \cdot 10^{-23} \text{ J/K}$$

$$\kappa = \frac{k}{RC} = 1.38 \cdot 10^{-14} \text{ W/K}$$

$$T = 300 \text{ K}$$



# Landauer's Principle and Directed Information

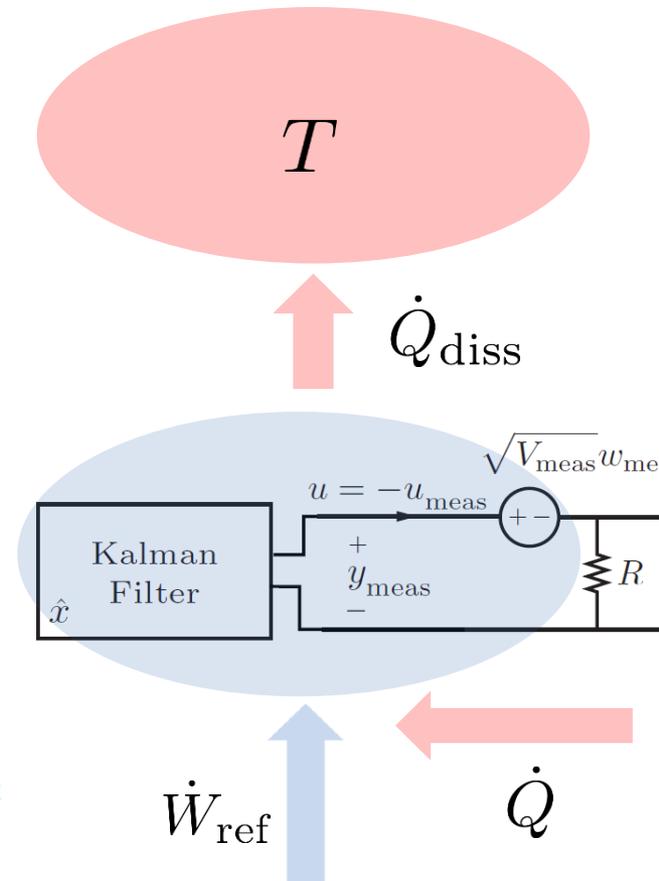
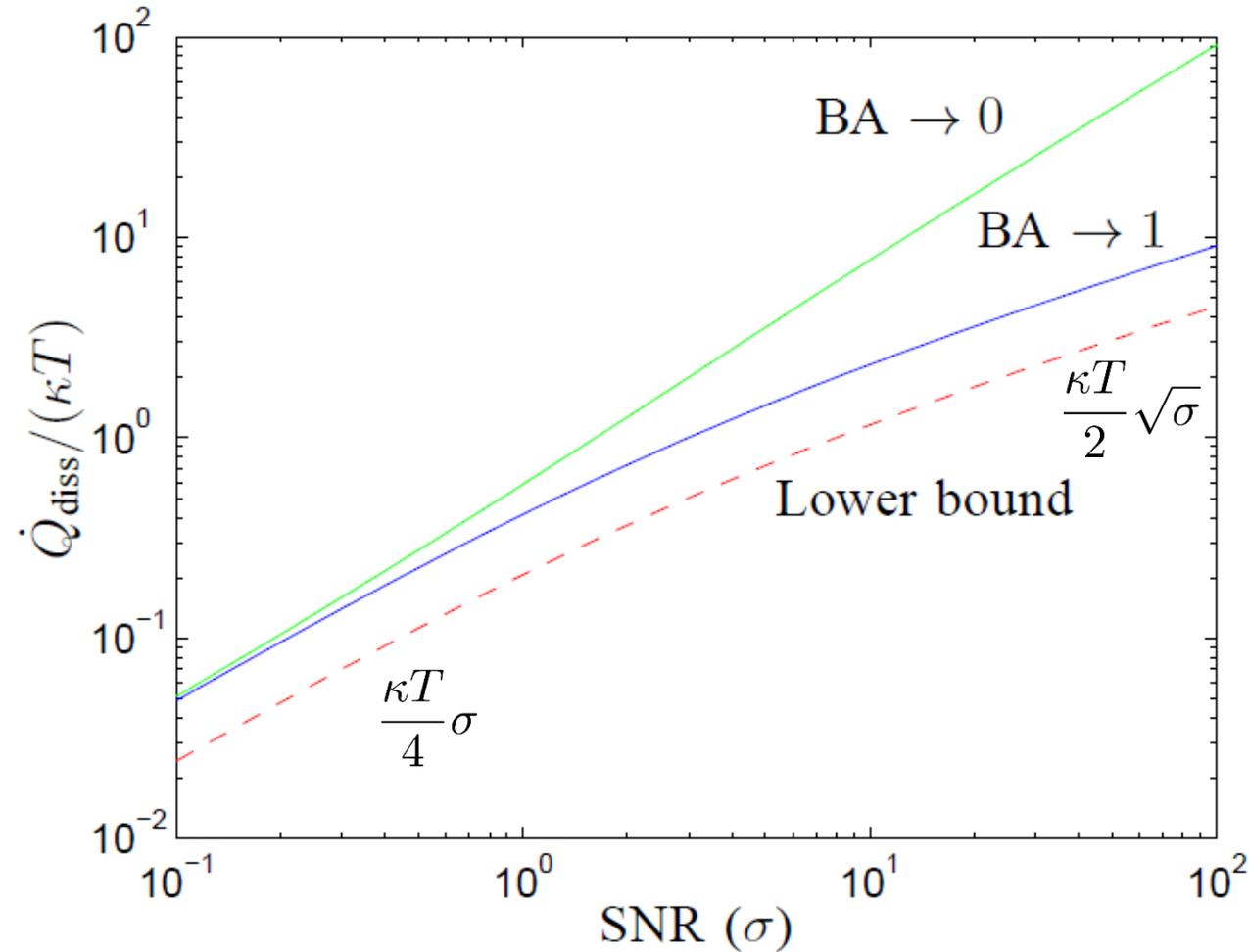


- **Landauer's principle (1961):** Need to spend at least work  $kT \ln 2$  to erase one bit of information
- Directed information flow (system to filter):

$$\dot{I}_c := \frac{d}{dt} I((w_0^t, x(0)); (y_{\text{meas}})_0^t)$$

$$\dot{W}_{\text{memo}} \geq kT \dot{I}_c = \frac{\kappa T}{2} (\sqrt{1 + \sigma} - 1)$$

# Absolute Lower Bound Compared to Physical Implementations





# Observations

- “Passive” implementation at least a factor 2 more dissipation than required by lower bound
- **Possible explanations:**
  - Landauer’s principle holds for **infinitely slow erasure**. Here finite erasure rate, which costs more
  - Factor 2 often appears in maximum-power relations. Compare with impedance matching
  - Directed information rate is a **lower bound** on **entropy rate** of memory in filter. Entropy rate can be a factor 2 larger

[Sandberg *et al.*, Phys. Rev. E, 2014]



# Outline

- Motivating example
- Thermodynamics of the Kalman-Bucy filter
- Relation to the information flow and Landauer's principle
- **(Maxwell's demon)**

# James Clerk Maxwell (1831-1879)

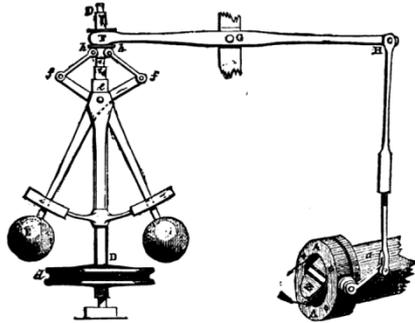
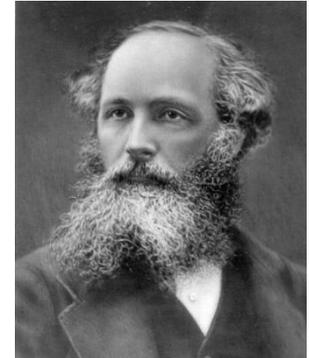


FIG. 4.—Governor and Throttle-Valve.

## ON GOVERNORS

J.C. MAXWELL

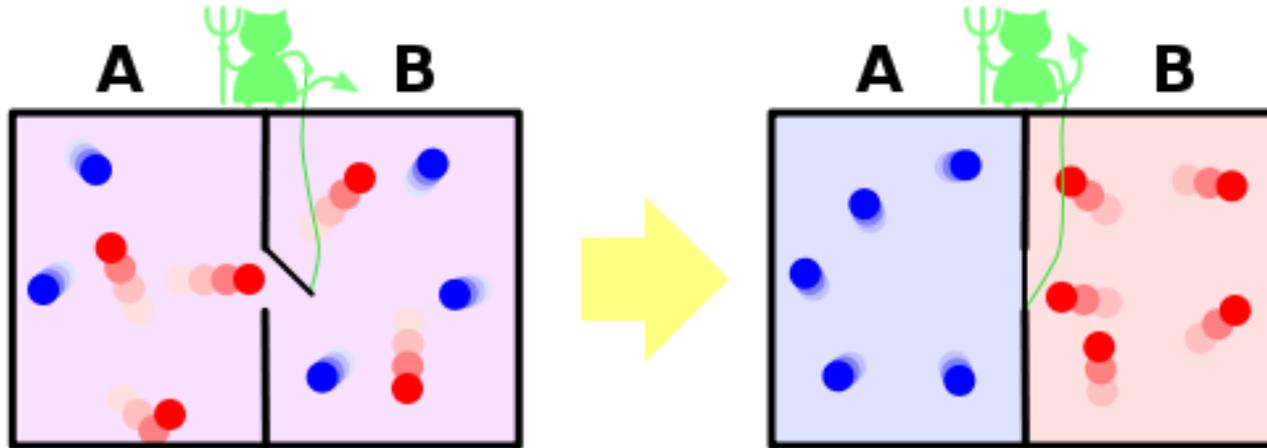


*From the Proceedings of the Royal Society, No.100, 1868.*

A GOVERNOR is a part of a machine by means of which the velocity of the machine is kept nearly uniform, notwithstanding variations in the driving-power or the resistance.

Most governors depend on the centrifugal force of a piece connected with a shaft of the machine. When the velocity increases, this force increases, and either increases the pressure of the piece against a surface or moves the piece, and so acts on a break or a valve.

# Maxwell's Demon (1867, 1871)

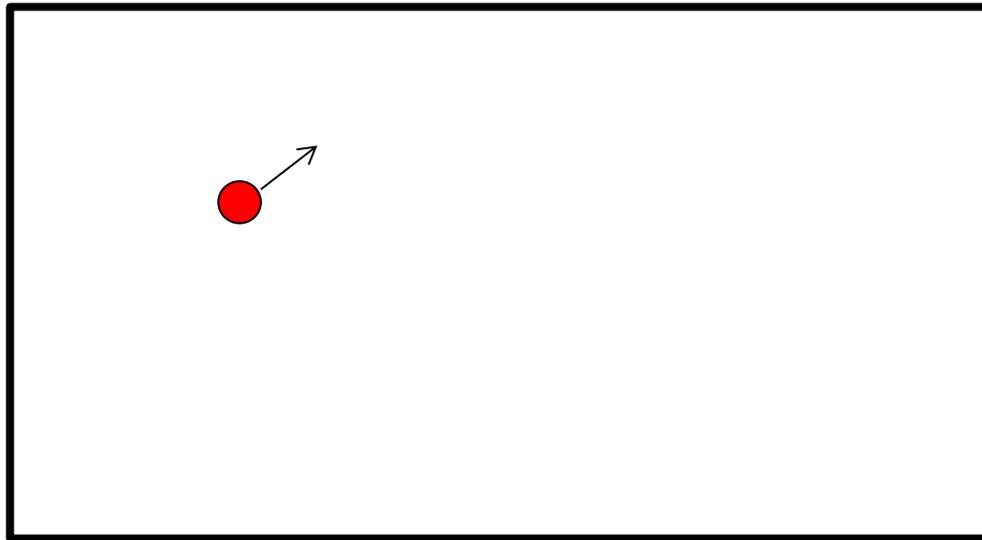


“... Now let us suppose that such a vessel is divided into two portions, A and B, by a division in which there is a small hole, and that a being, who can see the individual molecules, opens and closes this hole, so as to allow only the swifter molecules to pass from A to B, and only the slower molecules to pass from B to A. He will thus, without expenditure of work, raise the temperature of B and lower that of A, in contradiction to the second law of thermodynamics ...”

# Szilard's Engine (1929)

A

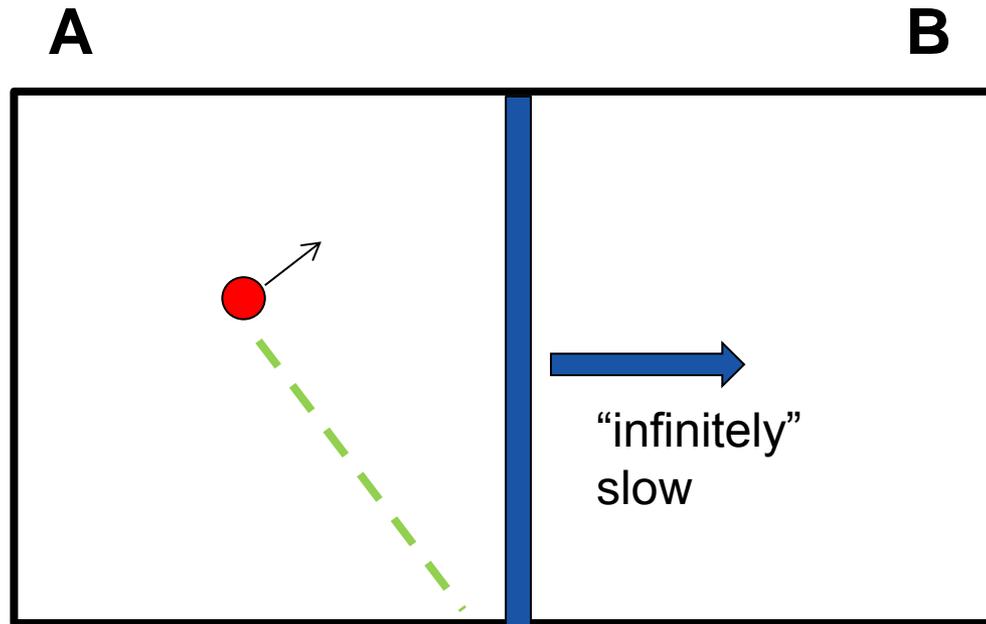
B



Heat-  
reservoir  
of temp.  $T$



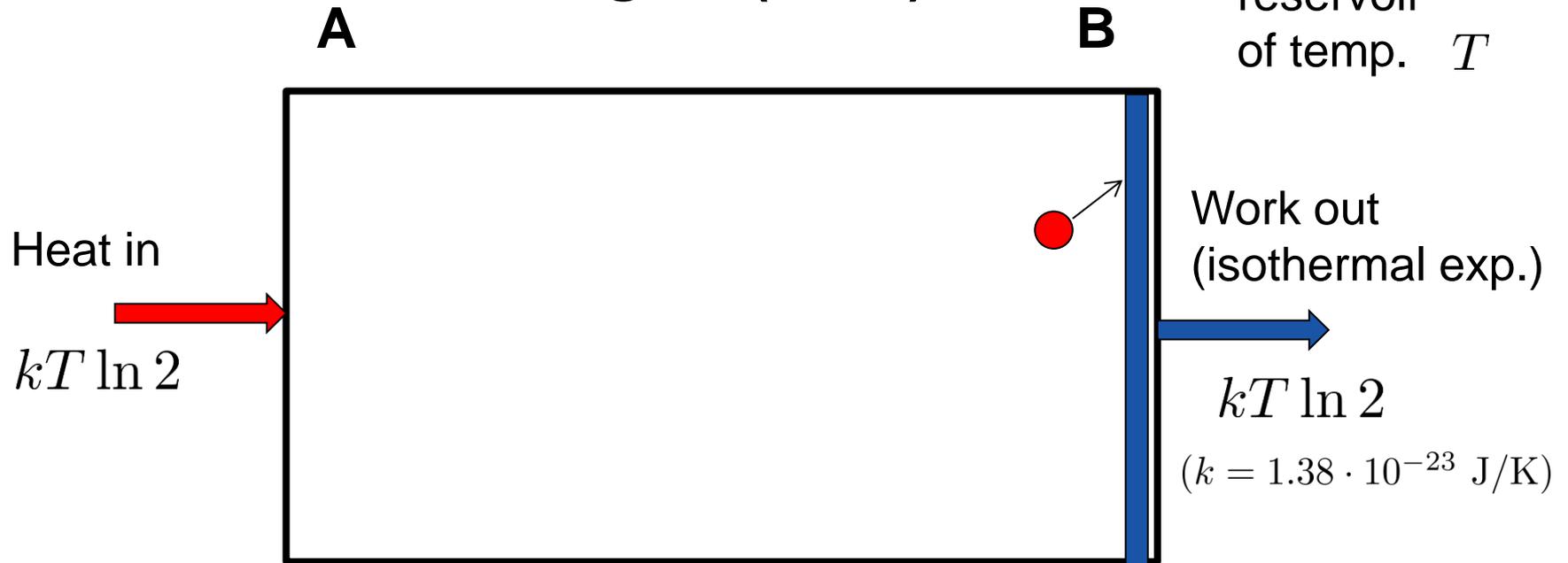
# Szilard's Engine (1929)



Heat-  
reservoir  
of temp.  $T$



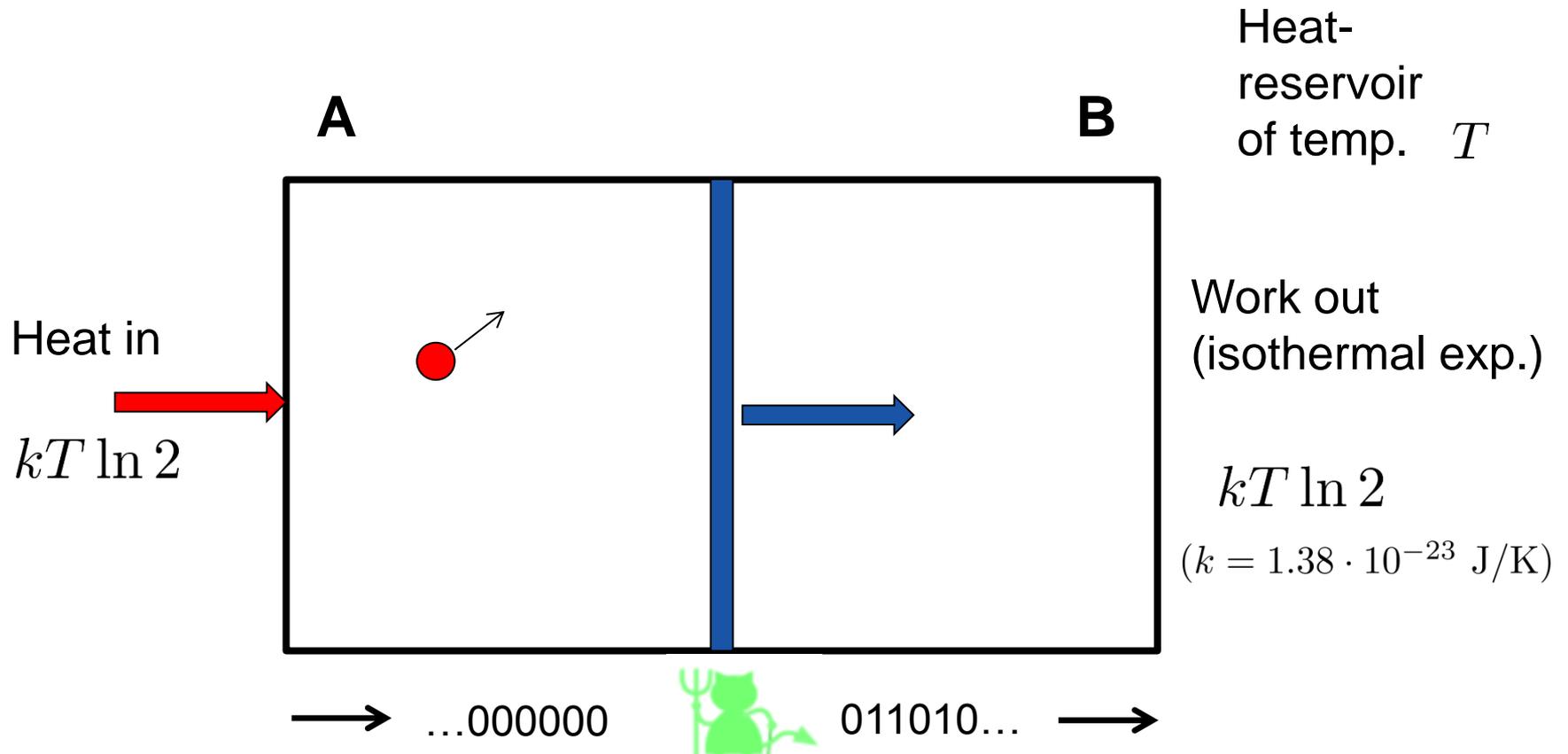
## Szilard's Engine (1929)



Seemingly in violation of the Kelvin-Planck statement of the 2<sup>nd</sup> law:

“It is impossible to construct an engine which will work in a complete cycle, and produce *no effect* except the raising of a weight and the cooling of a heat-reservoir.”  
(Planck)

# Solution by Landauer, Penrose, and Bennett



Demon obtains information (1 bit/cycle), which is stored in its memory.

## Landauer's principle:

Erasing 1 bit of information requires at least the work  $kT \ln 2$  (IBM J. Res. Dev., 1961)



# Summary

- Class of systems with “passive” Kalman-Bucy filters found. Passive but active cooling required (unless we own a cold heat bath...)
- Back action/feedback reduces required power supply
- Physical implementations are a factor 2 away from Landauer’s lower bound. In fact optimal implementation?
- Possible applications: stochastic thermodynamics, synthetic biology, energy harvesting...