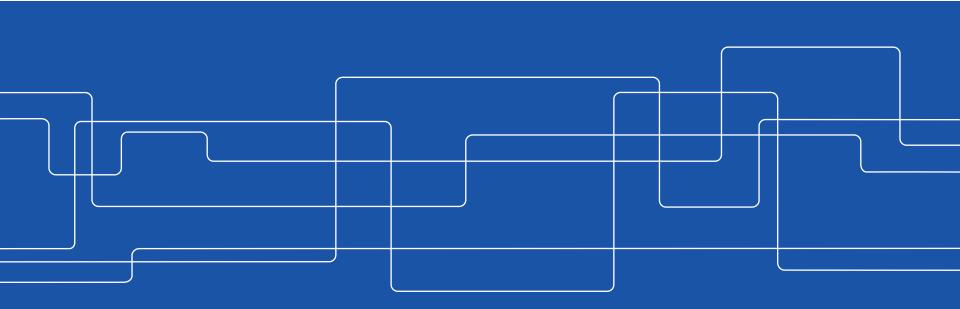


Implementation Costs and Information Flow in Kalman-Bucy Filters

Henrik Sandberg

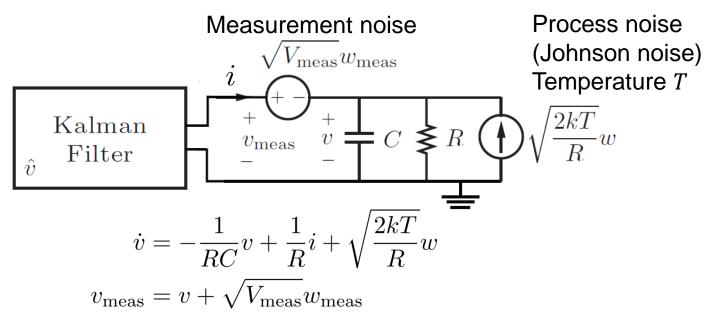
Joint work with

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





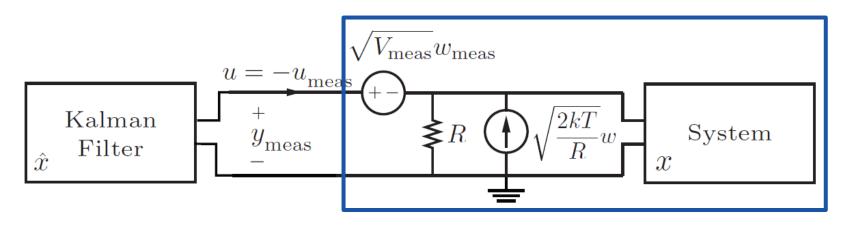
Motivating Example and Problem Formulation



- 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 Kalman-Bucy filter?



More General Class of Systems to Measure

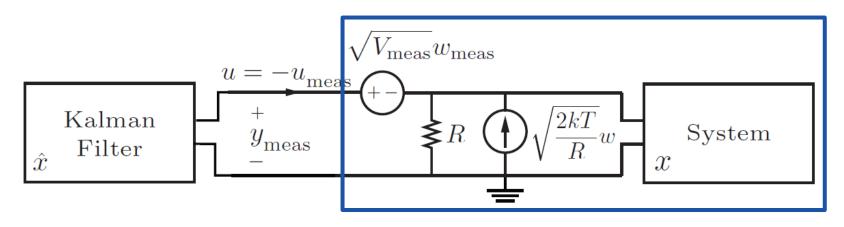


Linear passive system

$$\dot{x} = (J - GBB^T)Mx + Bu + B\sqrt{2kTG}w$$
$$y = B^T Mx$$
$$y_{\text{meas}} = B^T Mx + \sqrt{V_{\text{meas}}}w_{\text{meas}}$$
$$(J = -J^T, \quad G := 1/R, \quad M = M^T > 0)$$



More General Class of Systems to Measure

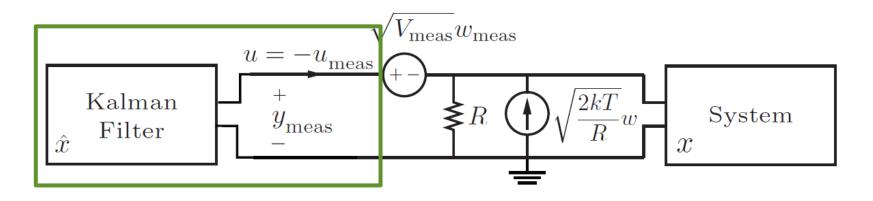


Signal-to-noise ratio (SNR):

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



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

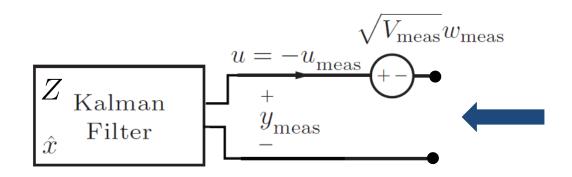
Assumption: Admit linear back action current

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

Theorem: A realization of Kalman-Bucy filter is $\frac{d}{dt}\hat{x}_{s} = (J - ZB_{s}B_{s}^{T})M\hat{x}_{s} + B_{s}y_{\text{meas}}$ $u_{\text{meas}} = B_{s}^{T}M\hat{x}_{s}$ $\hat{x}_{s} = \sqrt{g/g_{K}}\hat{x}$ with effective resistance $Z = \frac{\sqrt{\sigma+1}}{(\sqrt{\sigma+1}-1)g} + \frac{1}{(\sqrt{\sigma+1}-1)G} > 0$



Temperature of Kalman-Bucy Filter



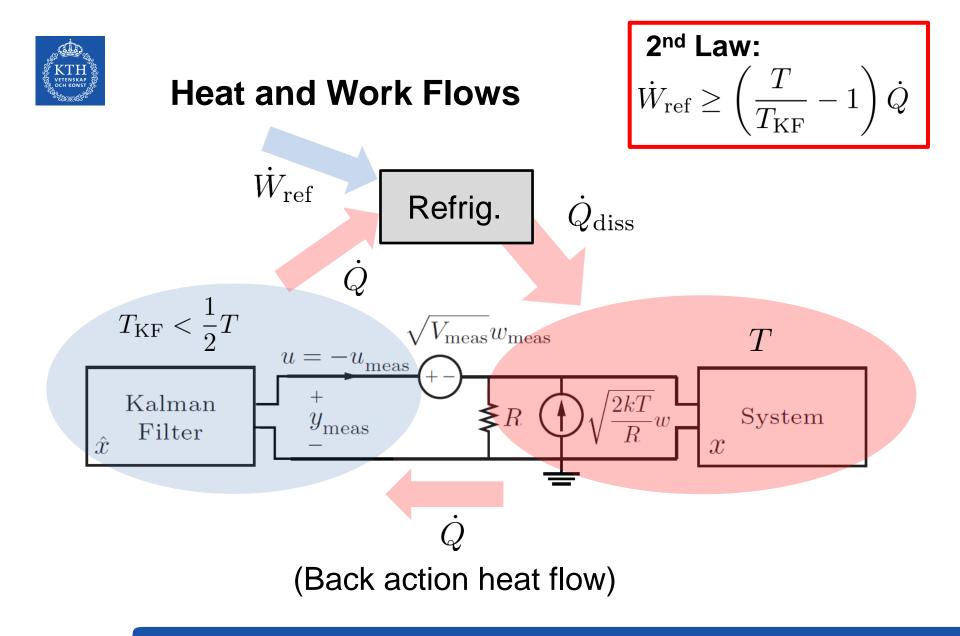
Fluctuation-dissipation theorem: $2kT_{\rm KF}Z = V_{\rm meas}$

Effective filter temperature:

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

Insight:

As back action tends to zero, filter temperature tends to zero





Power Supply Required by the 2nd Law

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

$$\frac{\text{No back action (BA \to 0)}}{\dot{W}_{\text{ref}} \ge \kappa T (1 + \sigma - \sqrt{1 + \sigma})} \approx \begin{cases} \frac{\kappa T}{2} \times \text{SNR (low SNR)} \\ \kappa T \times \text{SNR (high SNR)} \end{cases}$$

$$\frac{\text{Max back action (BA \to 1)}}{\dot{W}_{\text{ref}} \ge \kappa T \frac{1 + \sigma - \sqrt{1 + \sigma}}{1 + \sqrt{1 + \sigma}} \approx \begin{cases} \frac{\kappa T}{4} \times \text{SNR (low SNR)} \\ \kappa T \times \sqrt{\text{SNR}} \text{ (high SNR)} \end{cases}$$

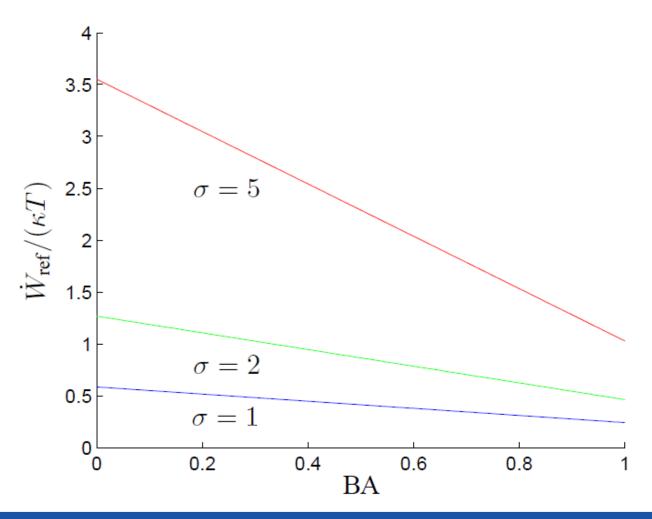


Power Supply Required by the 2nd Law

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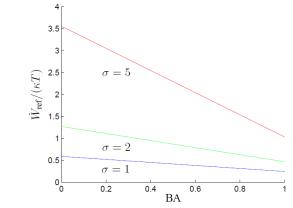


Trade-Off: Power Supply vs. Back Action





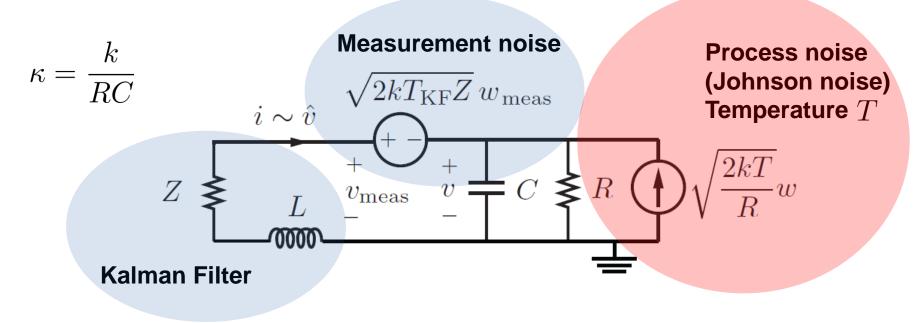
Observations



- Large back action ⇒ lower power supply
- **Explanation**: Temperature ratio T/T_{KF} smaller $\Rightarrow 2^{nd}$ law less restrictive
- Trade-off more significant for high SNRs
- No back action costs a factor $\sqrt{\text{SNR}}$ more than max back action in high SNR regime



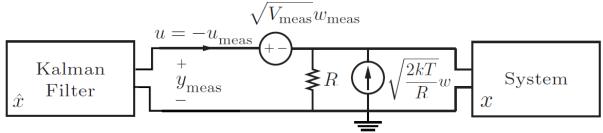
Exact Filter Circuit for Motivating Example



- Implementation using passive components
- $T_{KF} < T$ (non-equilibrium thermodynamic system)



Landauer's Principle and (Directed) Information Flow



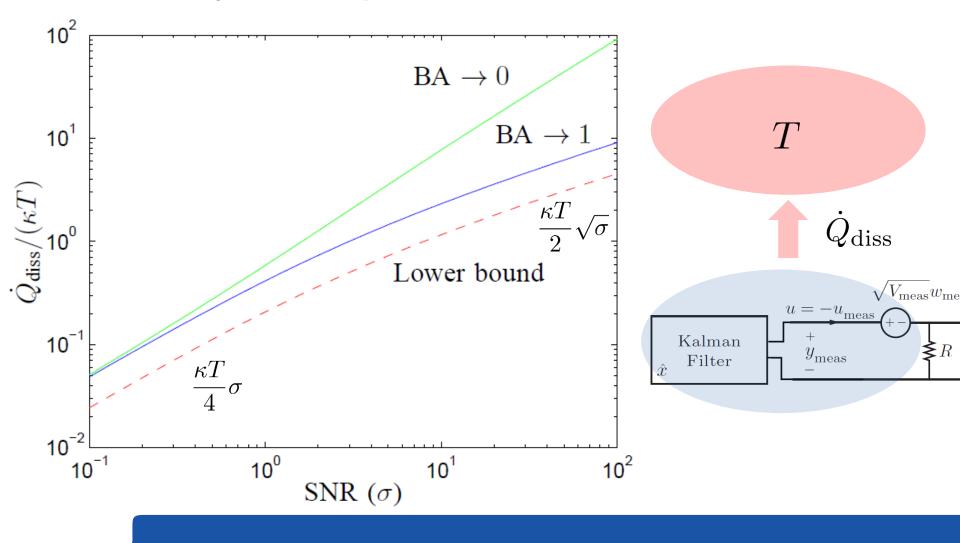
- Landauer's principle (1961): Need to spend at least work $kT \ln 2$ to erase one bit of information
- Directed information flow (system to filter):

$$\begin{split} \dot{I}_c &:= \frac{d}{dt} I((w_0^t, x(0)); (y_{\text{meas}})_0^t) \\ &= \frac{\kappa T}{2} (\sqrt{1 + \sigma} - 1) \end{split} \begin{aligned} & \text{[Sandberg et al.,} \\ & \text{Phys. Rev. E, 2014]} \end{split}$$

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



Absolute Lower Bound Compared to Physical Implementations





Observations

- "Passive" implementation at least a factor 2 more dissipation than required by lower bound
- Explanations:
 - Landauer's principle holds for infinitely slow erasure. Here finite erasure rate, which costs more
 - 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]



Summary

- Class of systems with "passive" Kalman-Bucy filters found. Passive but active cooling required (unless we own a cold heat bath...)
- Trade-off identified: Allow for back action to reduce required power supply
- Physical implementations are a factor 2 away from absolute lower bound. In fact optimal?
- Possible applications: nonequilibrium thermodynamics, synthetic biology, energy harvesting...



Related References

- H. Sandberg, J.-C. Delvenne, N.J. Newton, S.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.
- H. Sandberg, J.-C. Delvenne, N.J. Newton, S.K. Mitter: "Maximum work extraction and implementation costs for nonequilibrium Maxwell's demons". *Physical Review E*, 90, 042119, 2014.
- S.K. Mitter and N.J. Newton: "Information and entropy flow in the Kalman-Bucy filter". *Journal of Statistical Physics*, 2005.