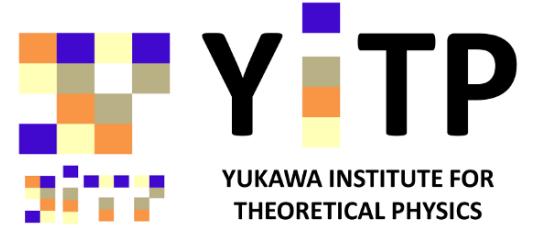


EPFL



Tuning **transduction** from hidden observables to optimize **information** harvesting

D. M. Busiello

in collaboration with G. Nicoletti (*EPFL*)

Workshop, July 8–19, 2024

Frontiers in Non-equilibrium Physics, YITP, Kyoto

Information from *inaccessible* degrees of freedom

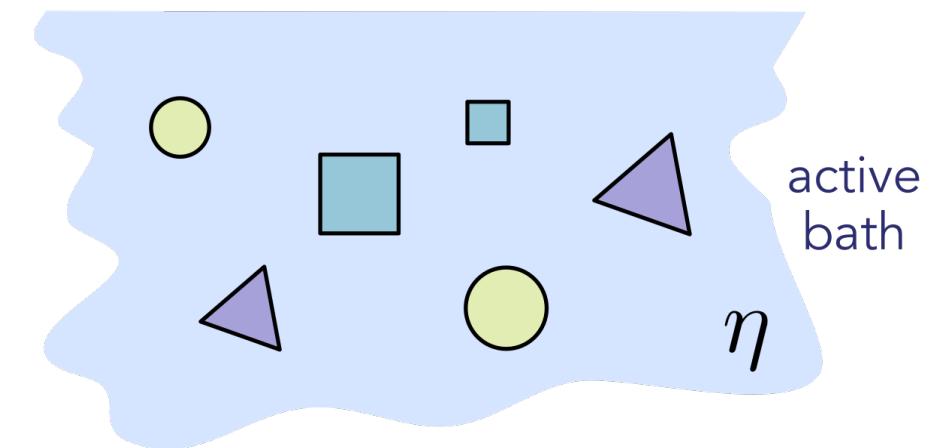
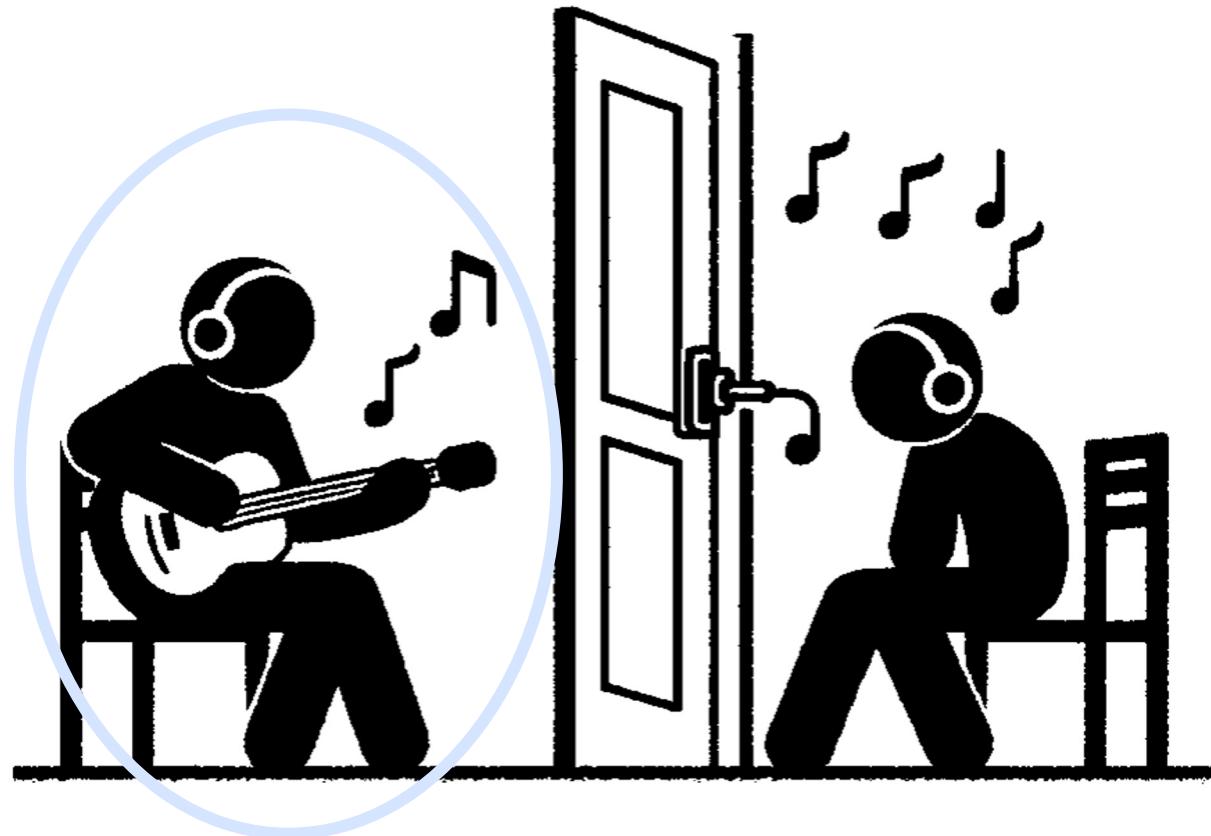


Who is playing the guitar?

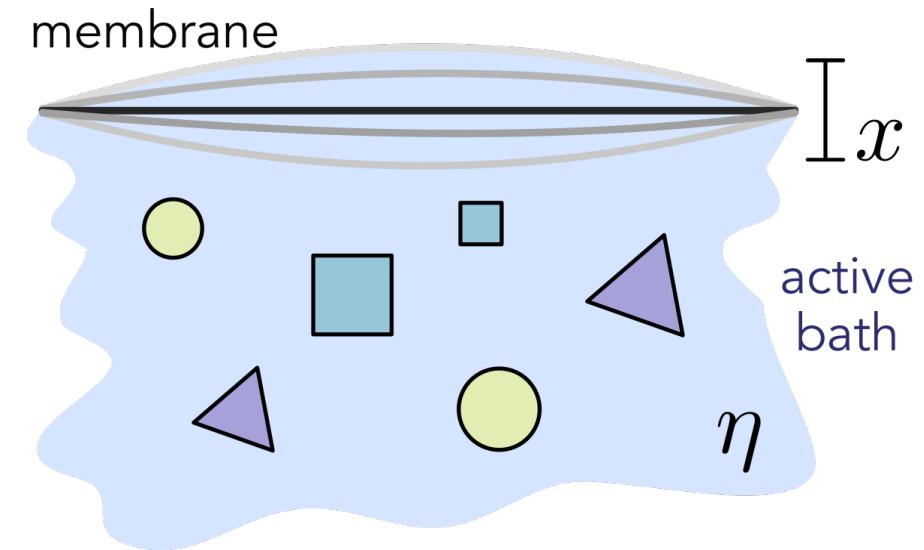
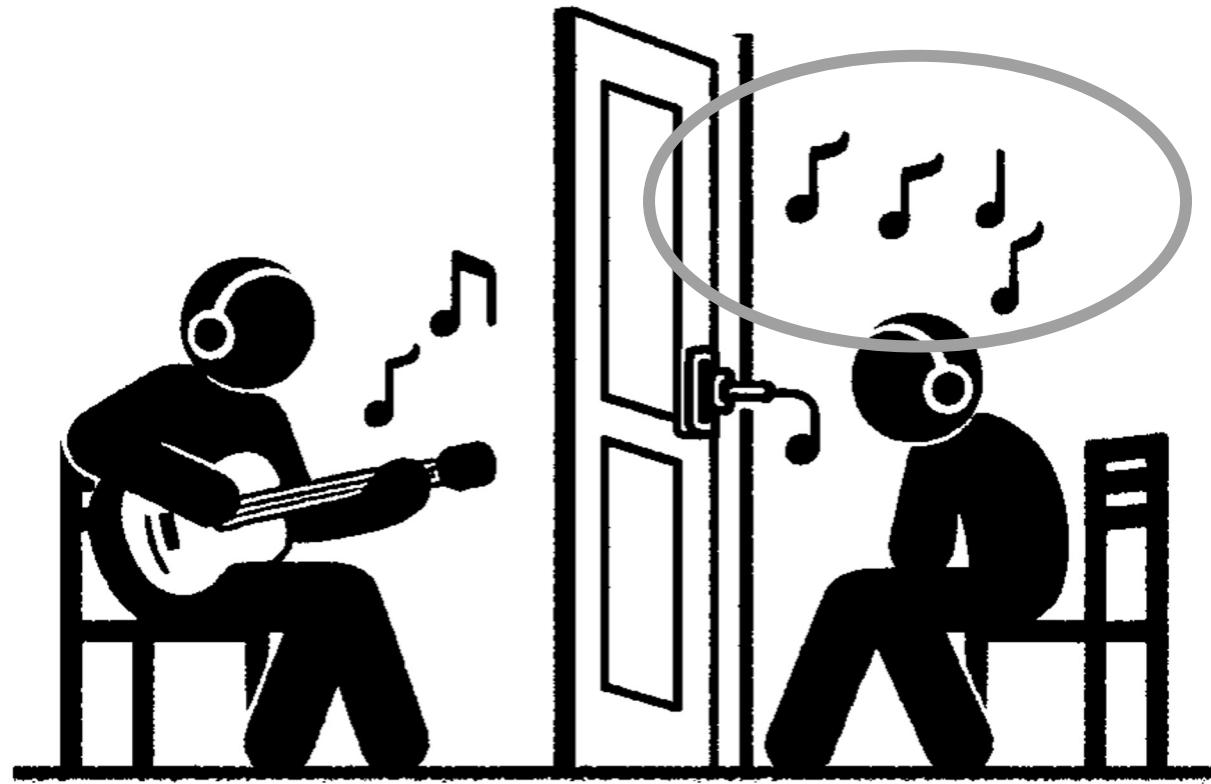
- Age
- Nationality
- ...

*Too many hidden variables
The lack of an underlying model*

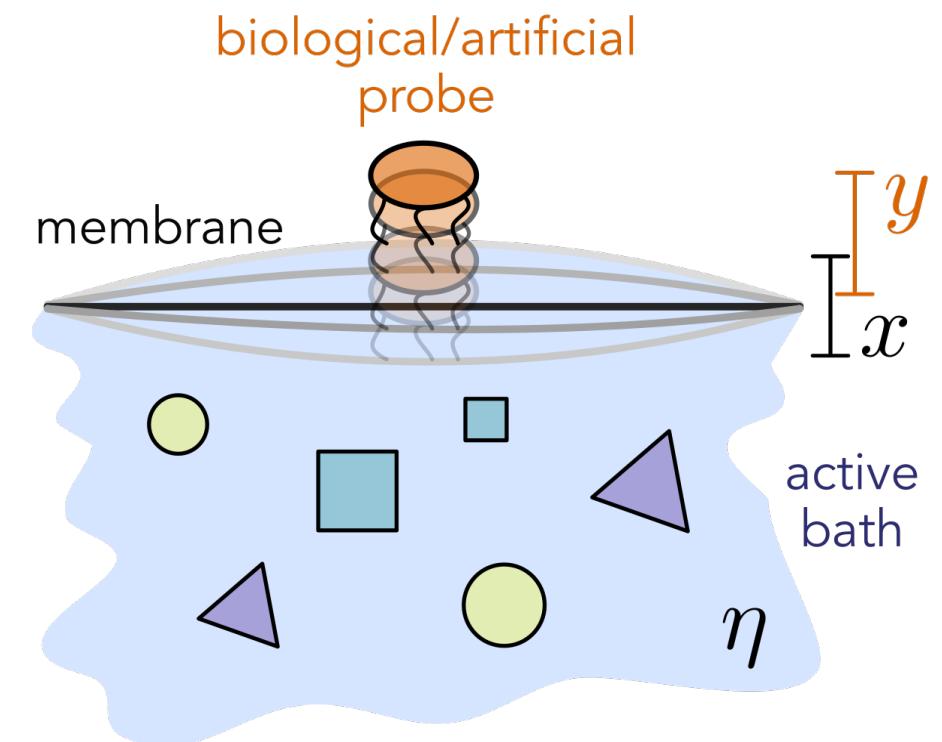
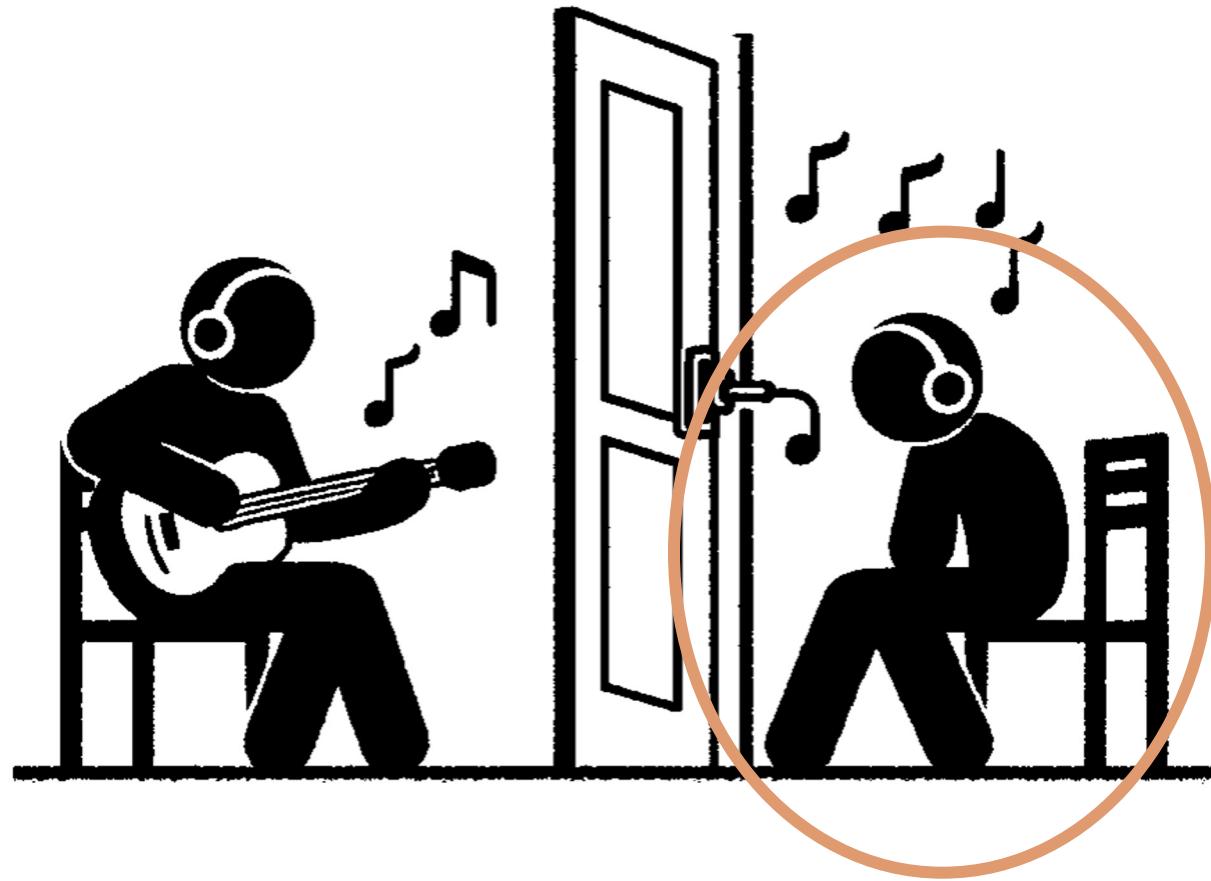
Information from *inaccessible* degrees of freedom



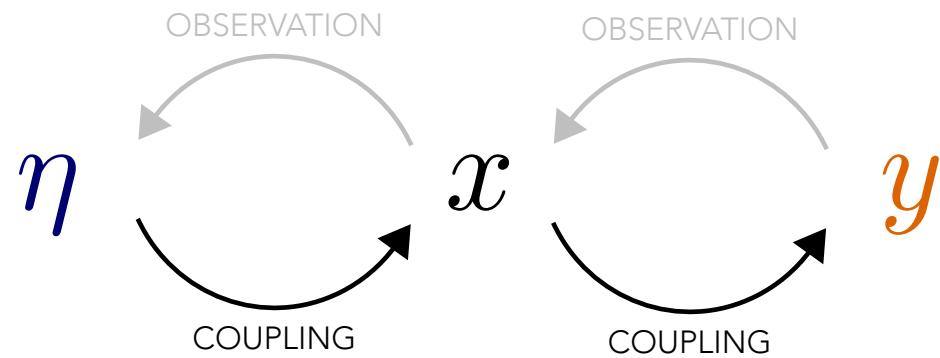
Information from *inaccessible* degrees of freedom



Information from *inaccessible* degrees of freedom



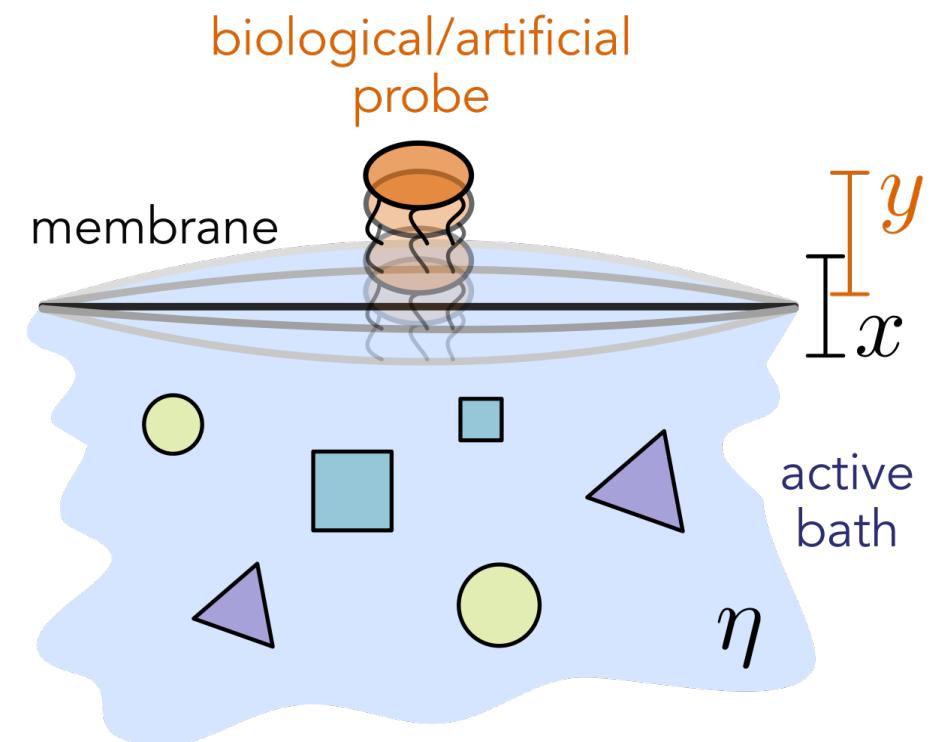
Information from *inaccessible* degrees of freedom



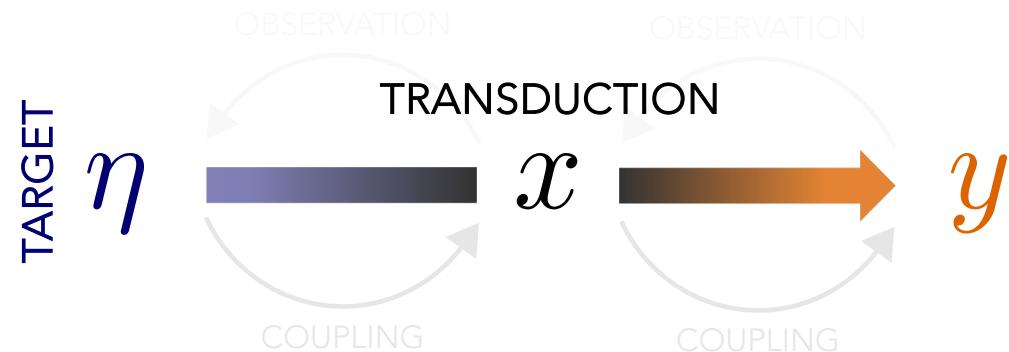
$$\tau_y \dot{y} = -y + ax + \sqrt{2\tau D_y} \xi_y(t)$$

$$\tau_x \dot{x} = -x + \sigma\eta + \sqrt{2\tau D_x} \xi_x(t)$$

$$\tau_\eta \dot{\eta} = -\eta + \sqrt{2\tau_\eta D_\eta} \xi_\eta(t)$$



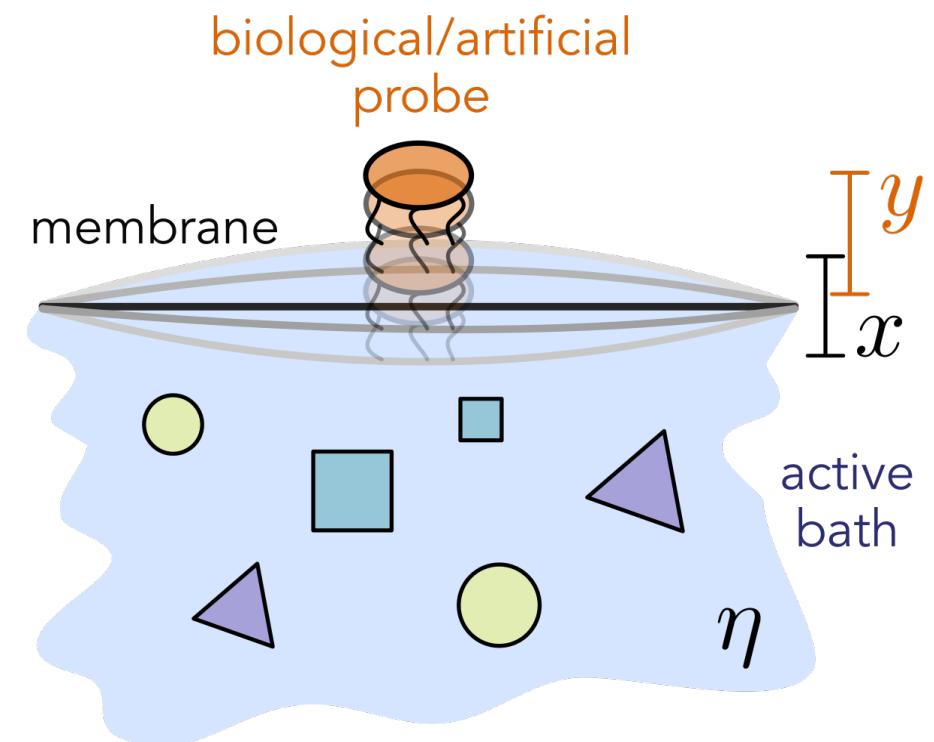
Information from *inaccessible* degrees of freedom



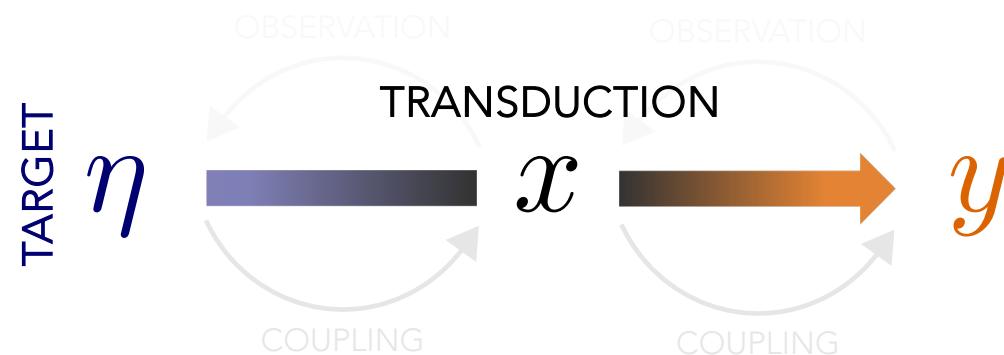
$$\tau_y \dot{y} = -y + ax + \sqrt{2\tau D_y} \xi_y(t)$$

$$\tau_x \dot{x} = -x + \sigma\eta + \sqrt{2\tau D_x} \xi_x(t)$$

$$\tau_\eta \dot{\eta} = -\eta + \sqrt{2\tau_\eta D_\eta} \xi_\eta(t)$$



Ideal but unrealizable scenario: *all-knowing probe*



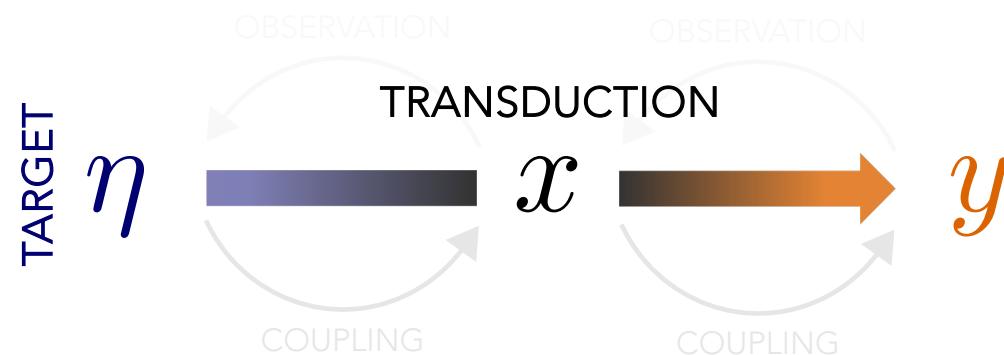
Probe-membrane coupling to be tuned

$$\tau_y \dot{y} = -y + \cancel{ax} + \sqrt{2\tau D_y} \xi_y(t)$$

$$\tau_x \dot{x} = -x + \sigma \eta + \sqrt{2\tau D_x} \xi_x(t)$$

$$\tau_\eta \dot{\eta} = -\eta + \sqrt{2\tau_\eta D_\eta} \xi_\eta(t)$$

Ideal but unrealizable scenario: *all-knowing probe*



Probe-membrane coupling to be tuned

$$\tau_y \dot{y} = -y + \cancel{ax} + \sqrt{2\tau D_y} \xi_y(t)$$

$$\tau_x \dot{x} = -x + \sigma \eta + \sqrt{2\tau D_x} \xi_x(t)$$

$$\tau_\eta \dot{\eta} = -\eta + \sqrt{2\tau_\eta D_\eta} \xi_\eta(t)$$

$$\max_a \underbrace{\left(\lambda I_{y\eta} - (1 - \lambda) T \dot{S}_a \right)}_{\mathcal{L}(a)}$$

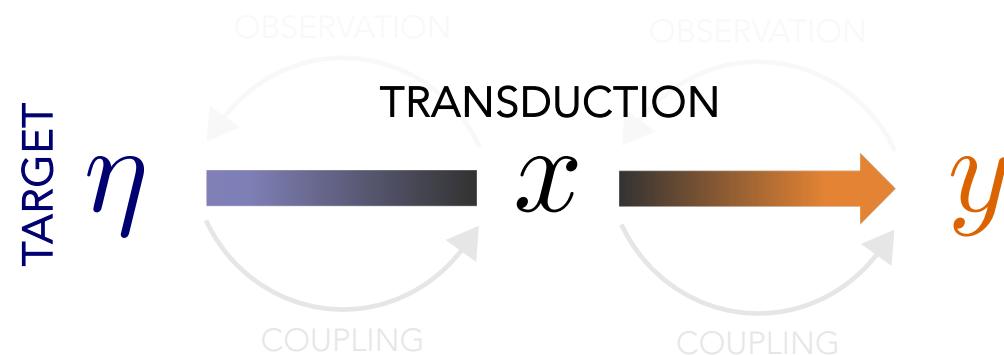
Probe-bath mutual information

$$I_{y\eta} = \int p_{y\eta} \log(p_{y|\eta}/p_y) dy d\eta$$

Entropy production due to the probe

$$\dot{S}_{\text{tot}} - \dot{S}_{\text{tot}}(a=0)$$

Ideal but unrealizable scenario: *all-knowing probe*



Probe-membrane coupling to be tuned

$$\tau_y \dot{y} = -y + \cancel{ax} + \sqrt{2\tau D_y} \xi_y(t)$$

$$\tau_x \dot{x} = -x + \sigma \eta + \sqrt{2\tau D_x} \xi_x(t)$$

$$\tau_\eta \dot{\eta} = -\eta + \sqrt{2\tau_\eta D_\eta} \xi_\eta(t)$$

$$\max_a \underbrace{\left(\lambda I_{y\eta} - (1 - \lambda) T \dot{S}_a \right)}_{\mathcal{L}(a)}$$

Probe-bath mutual information

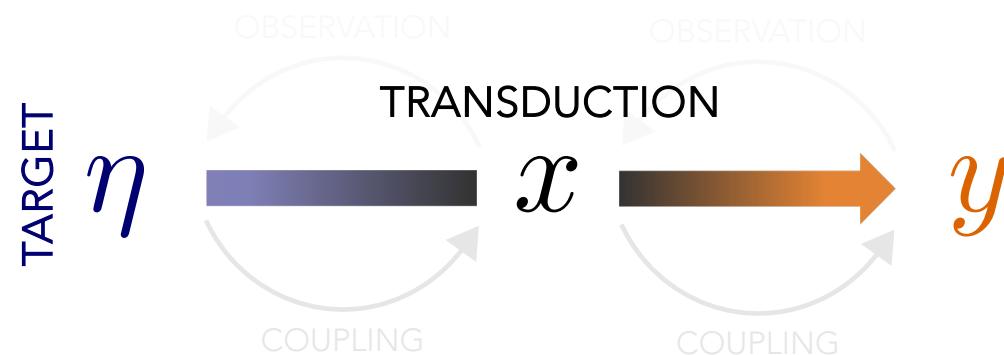
$$I_{y\eta} = \int p_{y\eta} \log(p_{y|\eta}/p_y) dy d\eta$$

Entropy production due to the probe

$$\dot{S}_{\text{tot}} - \dot{S}_{\text{tot}}(a=0)$$

λ represents the “strategy” of the probe
and sets the *energy budget* of the system
Invested into information transduction

Ideal but unrealizable scenario: *all-knowing probe*



Probe-membrane coupling to be tuned

$$\tau_y \dot{y} = -y + ax + \sqrt{2\tau D_y} \xi_y(t)$$

$$\tau_x \dot{x} = -x + \sigma\eta + \sqrt{2\tau D_x} \xi_x(t)$$

$$\tau_\eta \dot{\eta} = -\eta + \sqrt{2\tau_\eta D_\eta} \xi_\eta(t)$$

Timescales: $\theta_\eta = \frac{\tau_\eta}{\tau} \gg \frac{\tau_x}{\tau} = \frac{\tau_y}{\tau} = 1$

$$\max_a \underbrace{\left(\lambda I_{y\eta} - (1 - \lambda) T \dot{S}_a \right)}_{\mathcal{L}(a)}$$

Probe-bath mutual information

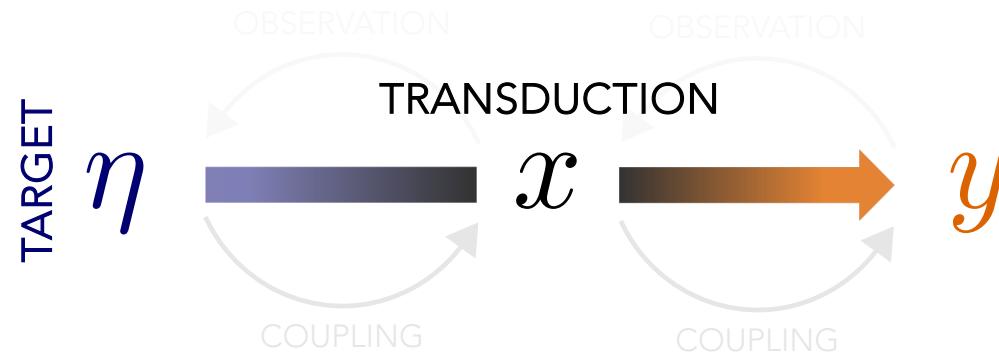
$$I_{y\eta} = \int p_{y\eta} \log(p_{y|\eta}/p_y) dy d\eta$$

Entropy production due to the probe

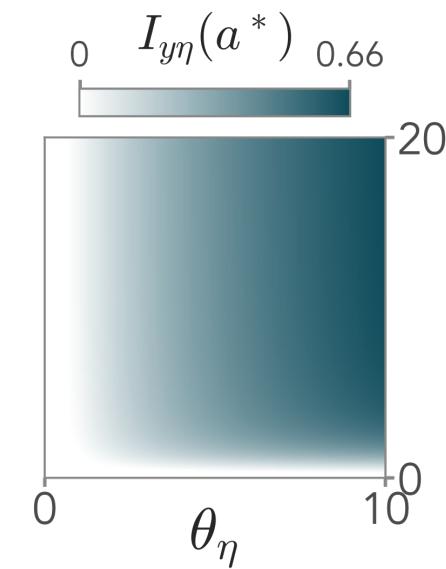
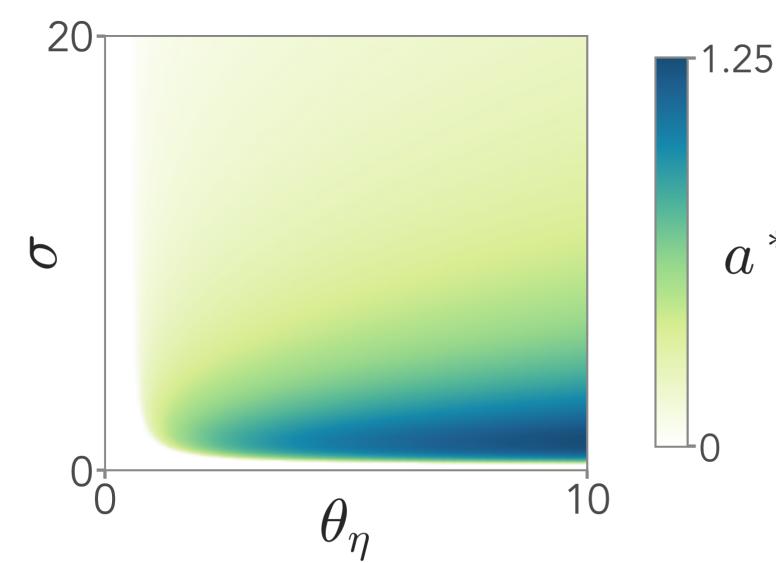
$$\dot{S}_{\text{tot}} - \dot{S}_{\text{tot}}(a=0)$$

λ represents the “strategy” of the probe
and sets the *energy budget* of the system
Invested into information transduction

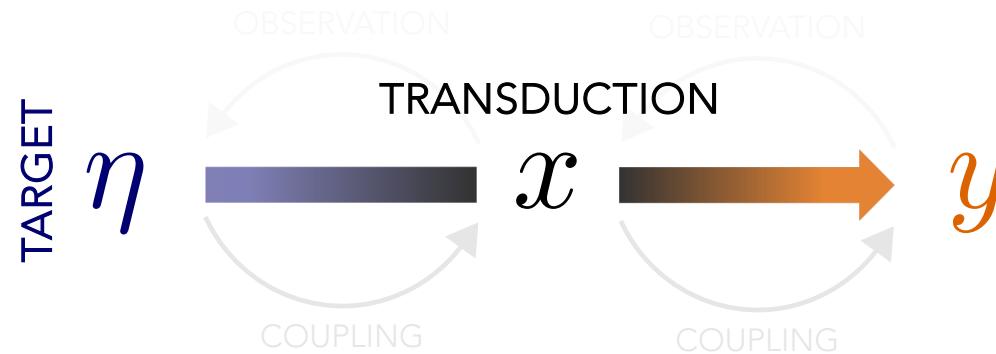
Ideal but unrealizable scenario: *all-knowing probe*



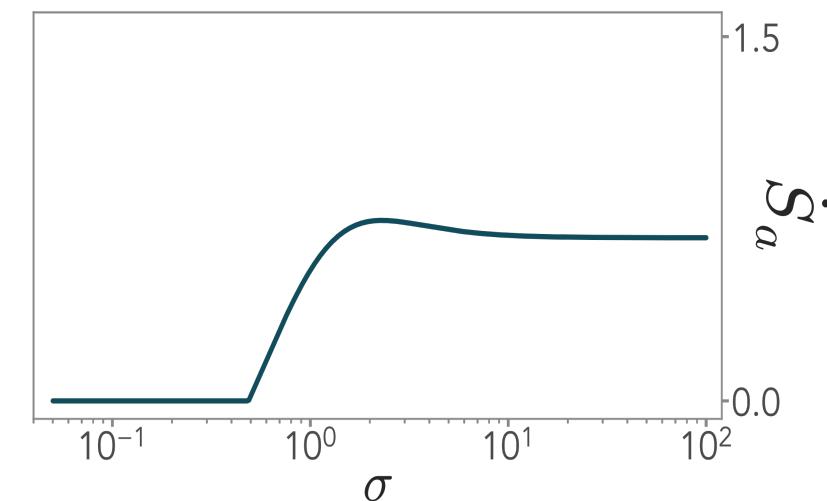
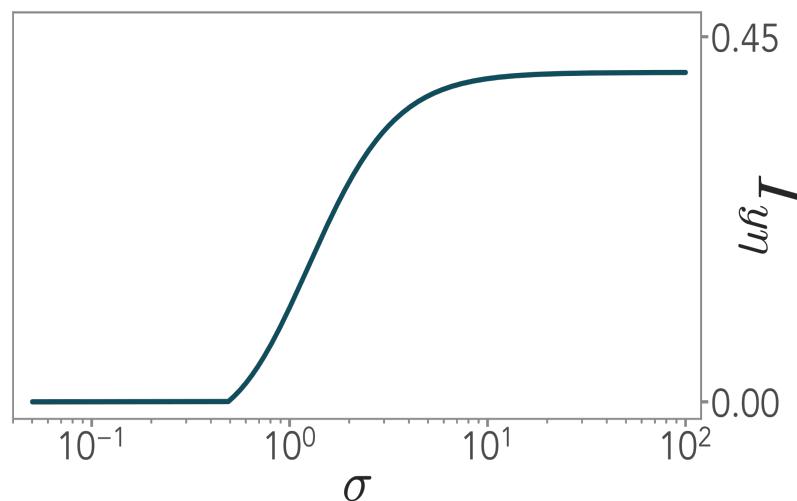
$$\max_a \underbrace{\left(\lambda I_{y\eta} - (1 - \lambda)T \dot{S}_a \right)}_{\mathcal{L}(a)}$$



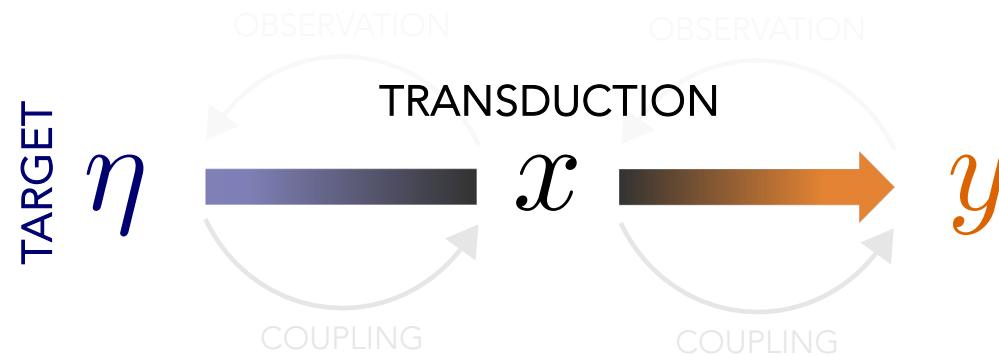
Ideal but unrealizable scenario: *all-knowing probe*



$$\max_a \underbrace{\left(\lambda I_{y\eta} - (1 - \lambda) T \dot{S}_a \right)}_{\mathcal{L}(a)}$$

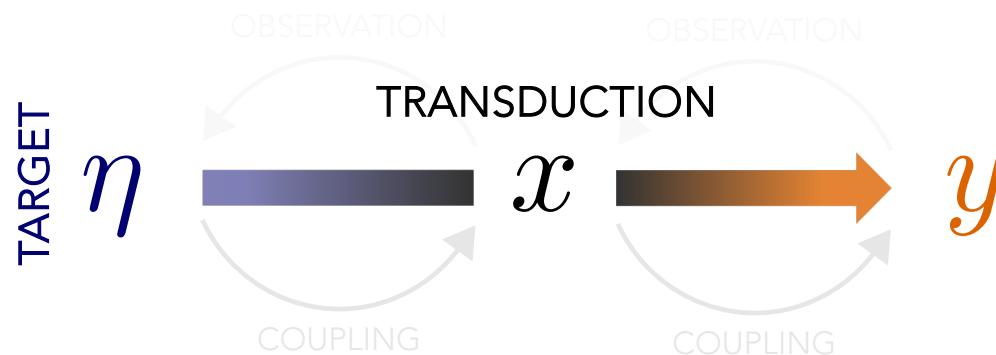


Realistic scenario: *partial information available*



$$\max_a \underbrace{\left(\lambda I_{xy} - (1 - \lambda) T \dot{S}_{xy} \right)}_{\mathcal{L}_{\text{eff}}(a)}$$

Realistic scenario: *partial information available*



$$\max_a \underbrace{\left(\lambda I_{xy} - (1 - \lambda) T \dot{S}_{xy} \right)}_{\mathcal{L}_{\text{eff}}(a)}$$

Probe-membrane mutual information

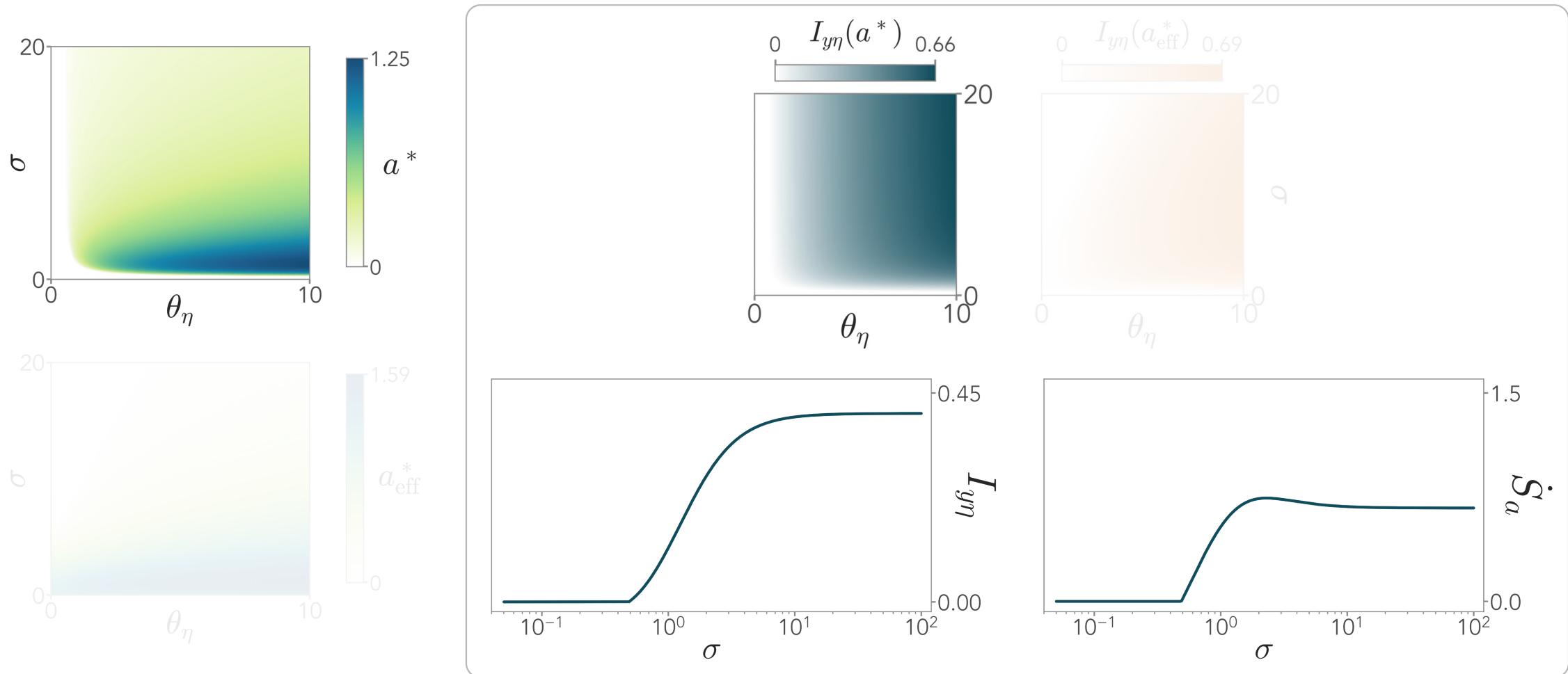
$$I_{xy} = \frac{1}{2} \log \frac{\Sigma_{xx} \Sigma_{yy}}{\det(\Sigma_{xy})}$$

Entropy production due to the probe

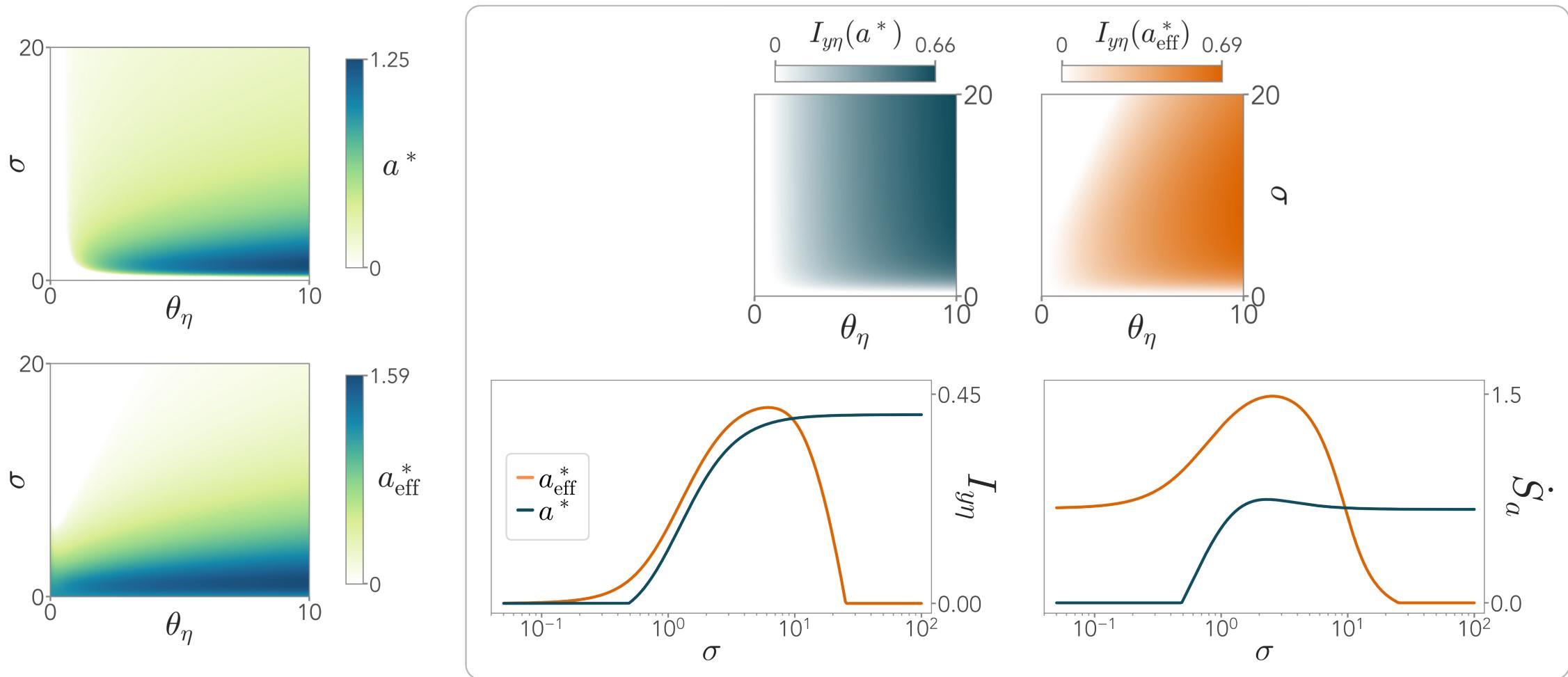
$$\dot{S}_{xy} = \text{Tr} \left[\mathbf{D}_{xy}^{-1} \mathbf{A}_{xy}^{\text{red}} \boldsymbol{\Sigma}_{xy} (\mathbf{A}_{xy}^{\text{red}})^T \right] - \text{Tr} (\mathbf{A}_{xy}^{\text{red}})$$

Reduced interaction matrix marginalizing
over non-accessible DOFs

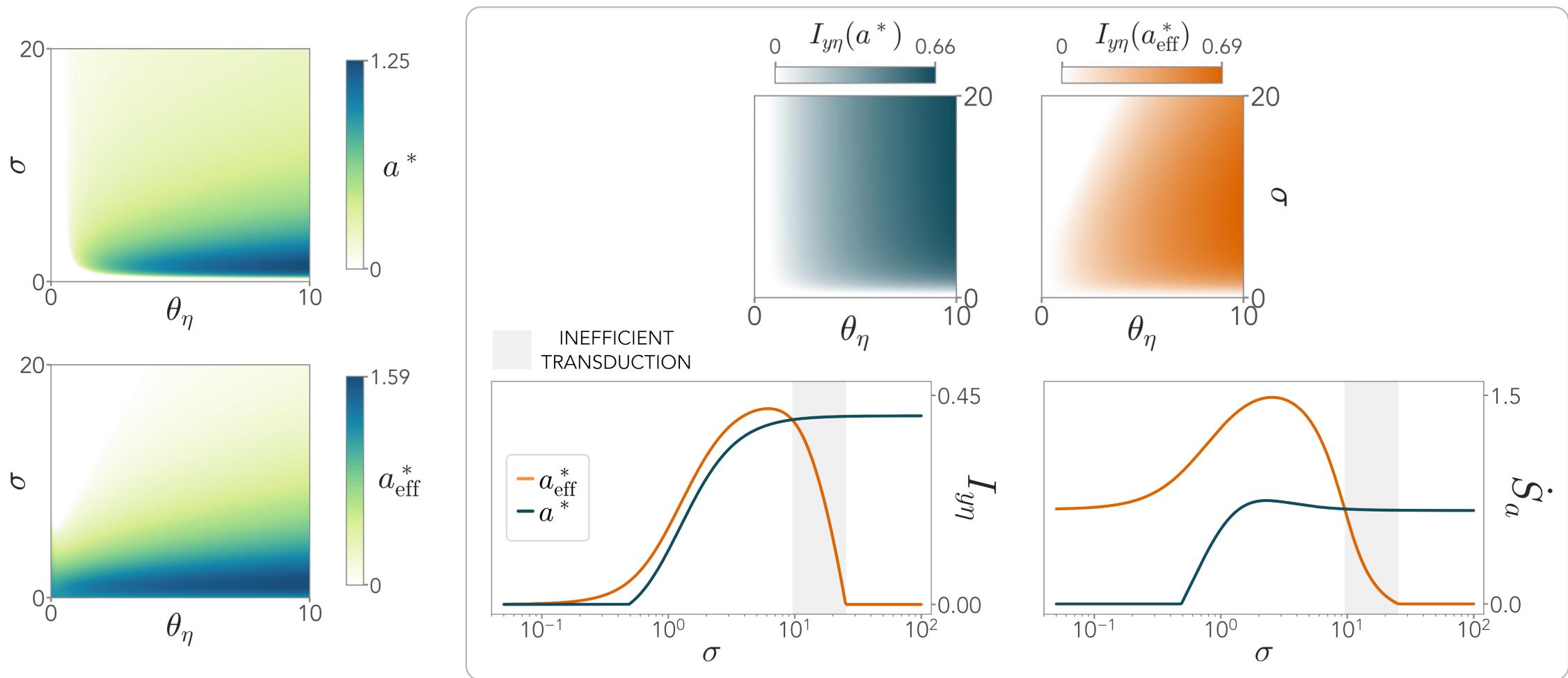
Information transduction can be efficient



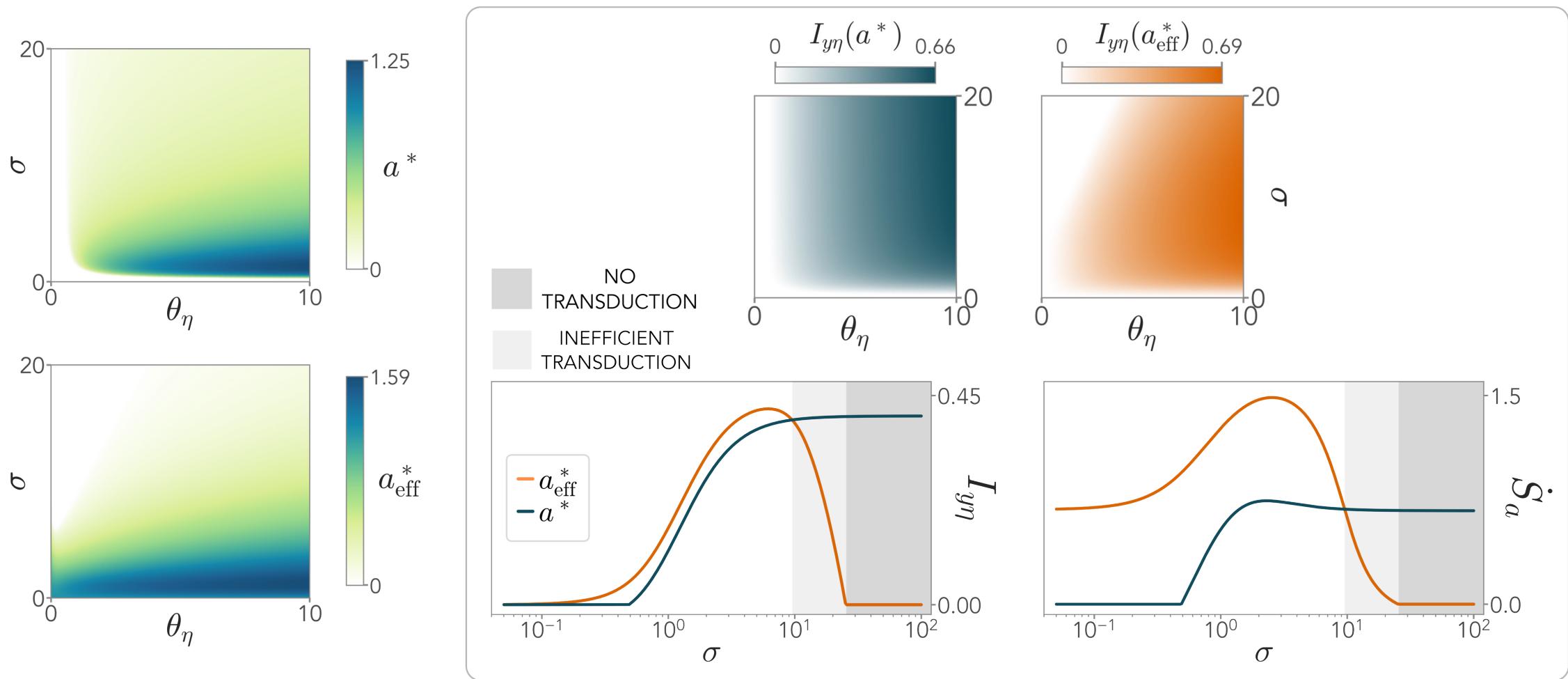
Information transduction can be efficient



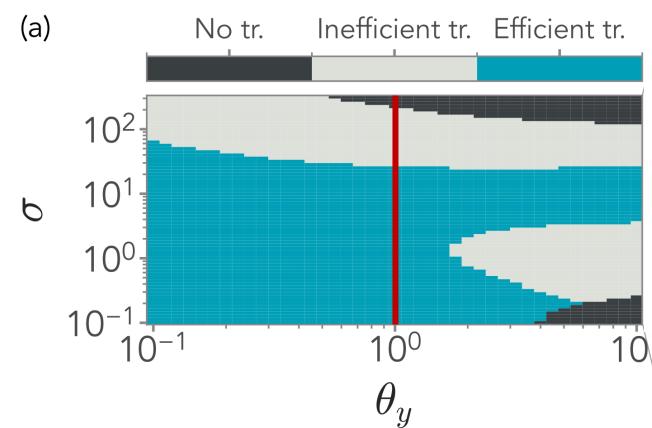
Information transduction can be efficient



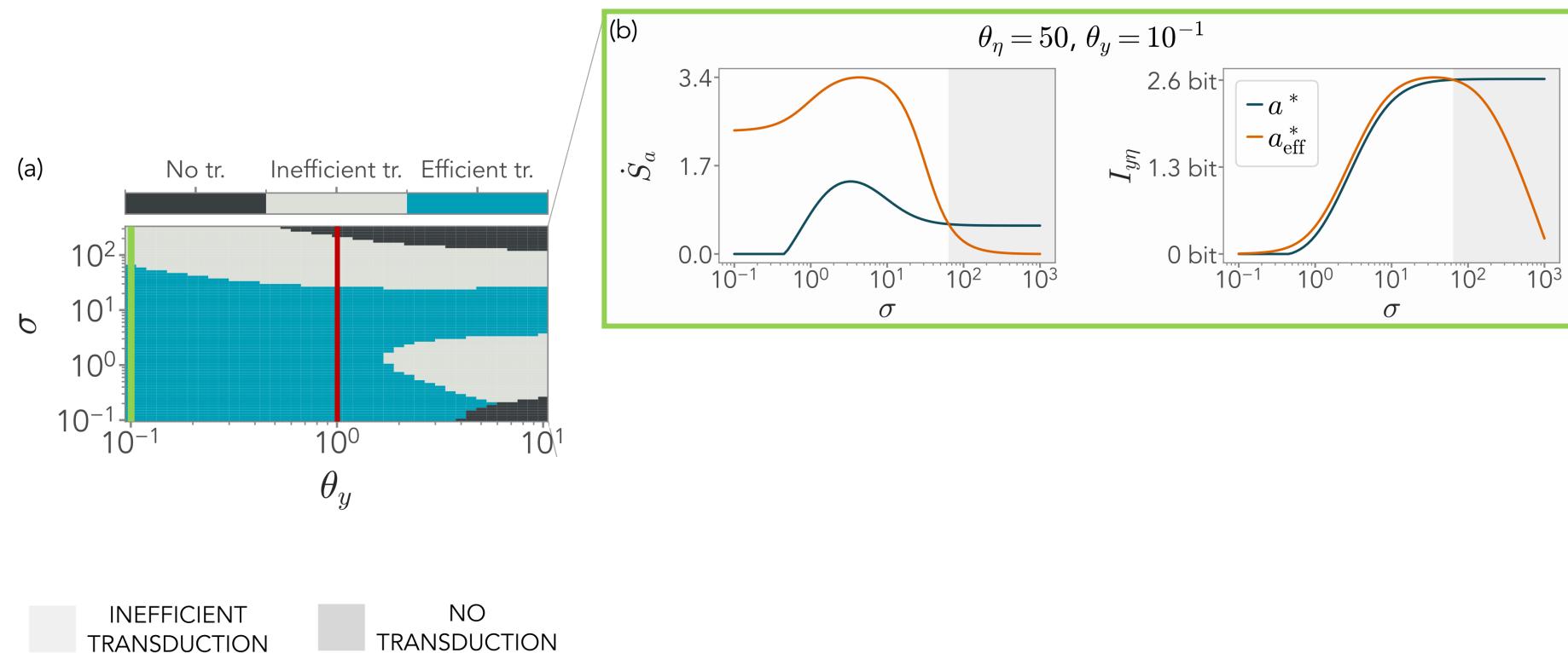
Information transduction can be efficient



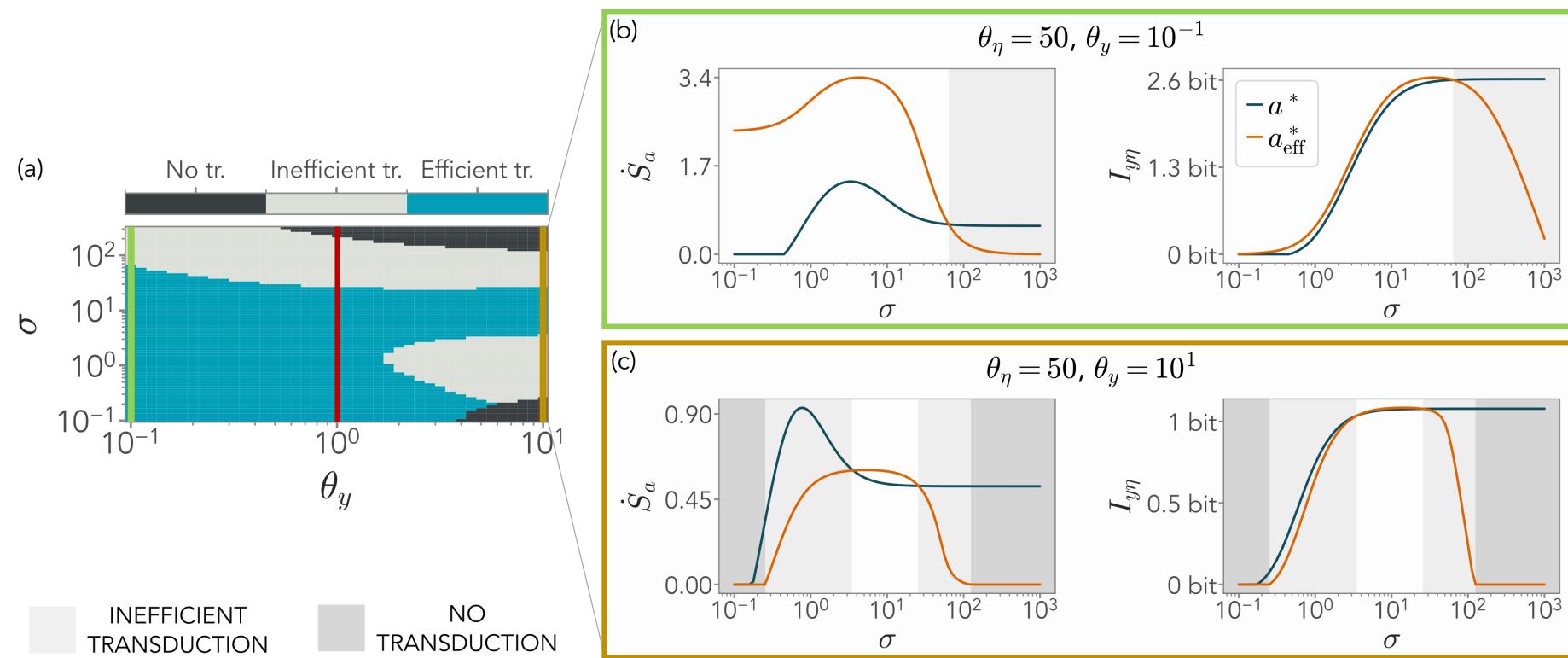
Timescales and *transduction efficiency*



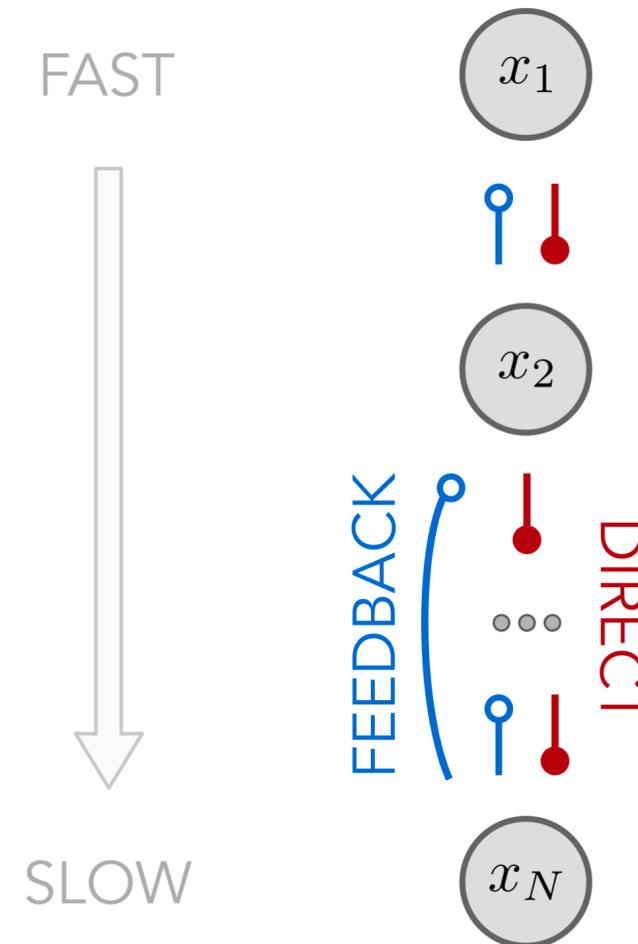
Timescales and *transduction efficiency*



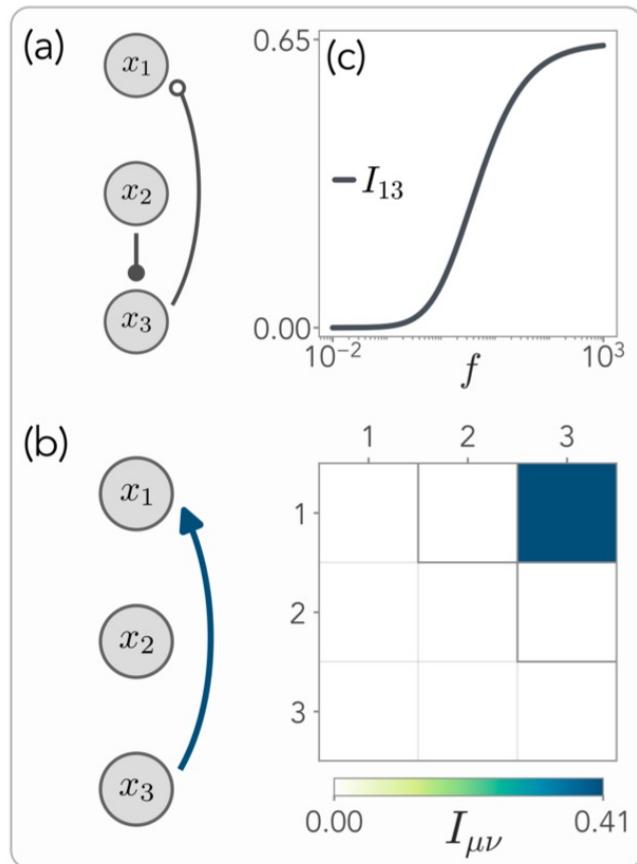
Timescales and *transduction efficiency*



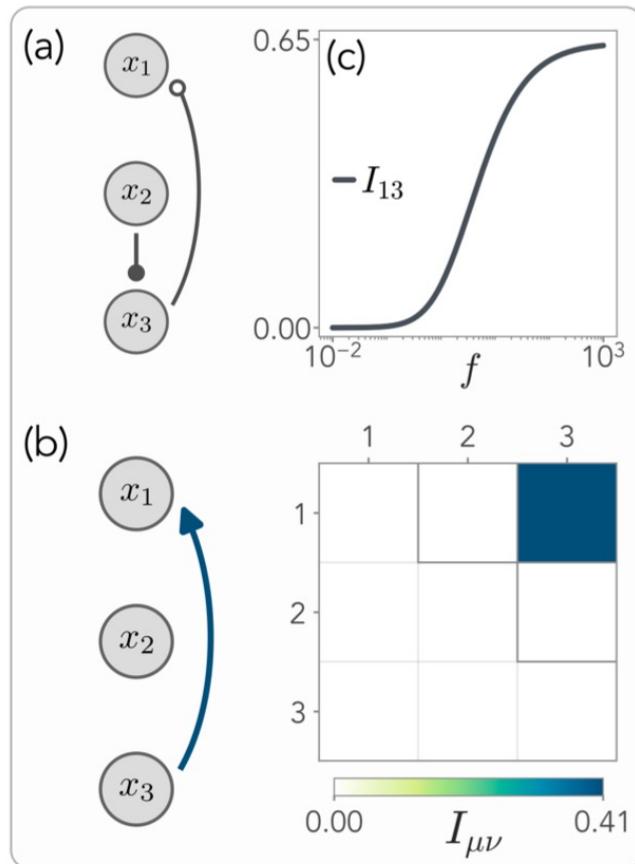
Information propagation across scales



Information propagation across scales



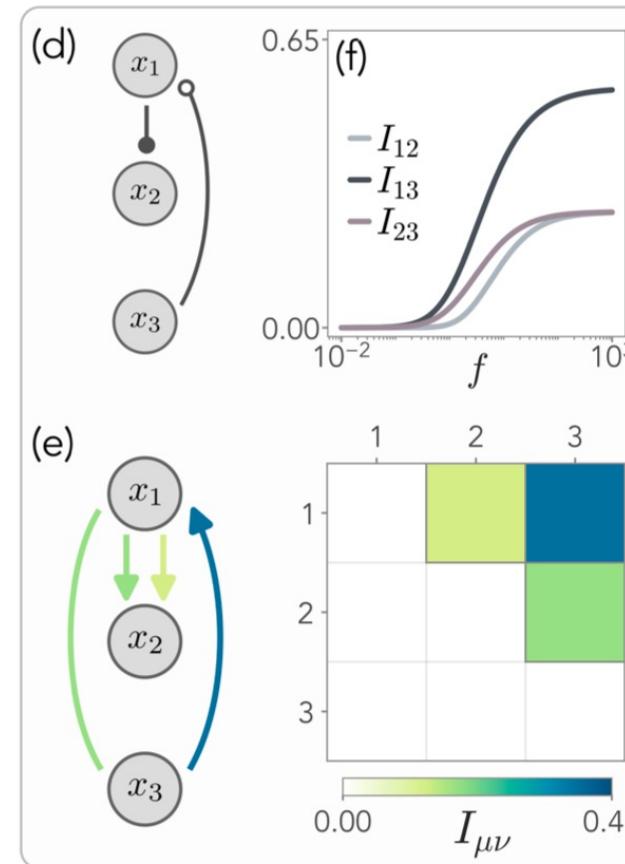
Information propagation across scales



The only contribution to the Mutual Information Matrix for Multiscale Observables (MIMMO) is from 3 to 1

- (i) Feedback interactions generate information from slow to fast layers
- (ii) Direct interaction alone do not generate information

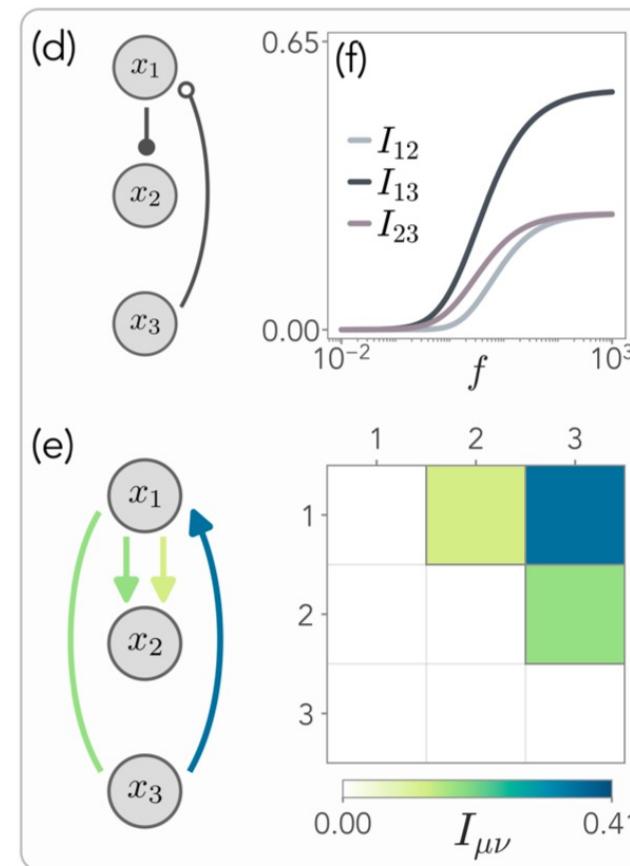
Information propagation across scales



Information propagation across scales

Direct contribution to the MIMMO from 3 to 1

Indirect contribution due to propagation through 1

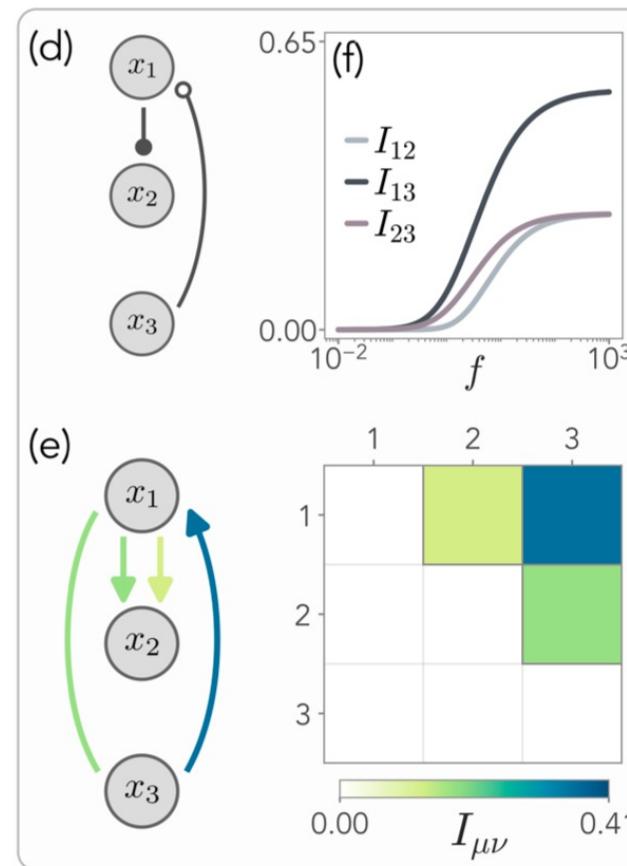


(iii) Information created by a slow layer can be propagated to faster layers via direct links

Information propagation across scales

Direct contribution to the MIMMO from 3 to 1

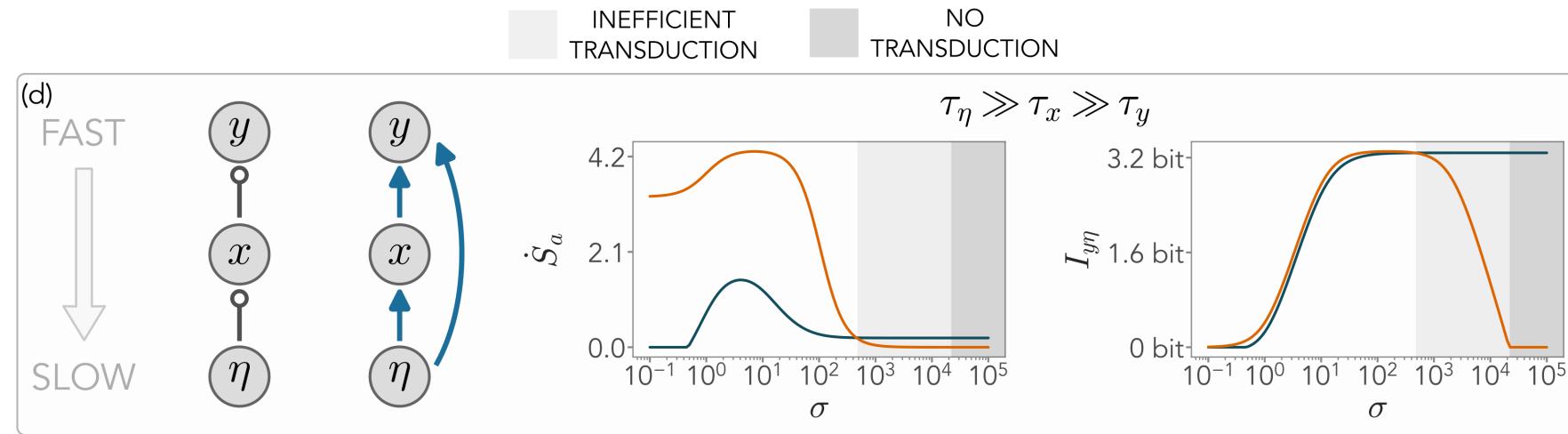
Indirect contribution due to propagation through 1



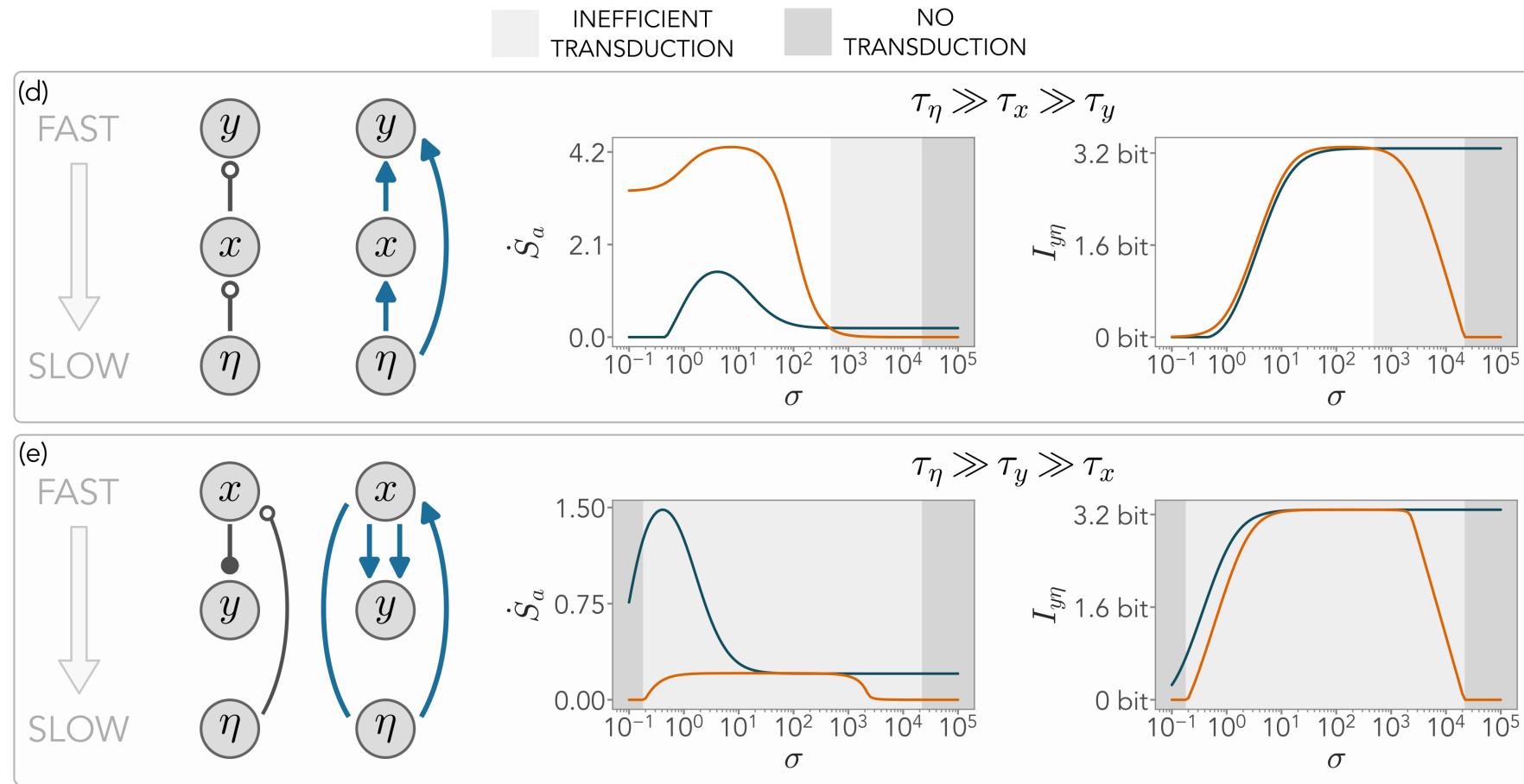
(iii) Information created by a slow layer can be propagated to faster layers via direct links

To propagate information to the other variables, the active bath has to be the slowest DOF

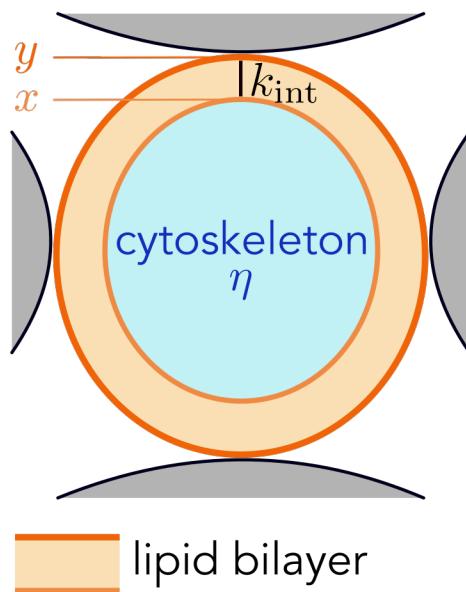
Information propagation across scales



Information propagation across scales



Information transduction in *red blood cells*



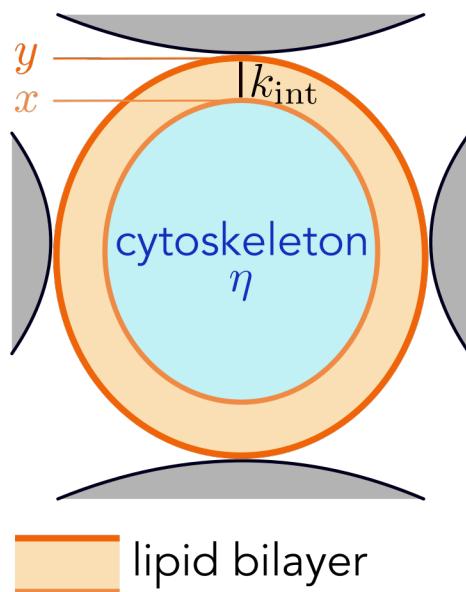
Membrane flickering is measured and it contains information on hidden dissipative degree of freedom

$$\dot{y} = -\mu_y (k_y y - k_{\text{int}} x) + \sqrt{2k_B T \mu_y} \xi_y(t)$$

$$\dot{x} = -\mu_x (k_x x - k_{\text{int}} y + \eta) + \sqrt{2k_B T \mu_x} \xi_x(t)$$

$$\dot{\eta} = -1/\tau_\eta \eta + \sqrt{2\epsilon^2/\tau_\eta} \xi_\eta(t)$$

Information transduction in *red blood cells*



Membrane flickering is measured and it contains information on hidden dissipative degree of freedom

$$\dot{y} = -\mu_y (k_y y - k_{\text{int}} x) + \sqrt{2k_B T \mu_y} \xi_y(t)$$

$$\dot{x} = -\mu_x (k_x x - k_{\text{int}} y + \eta) + \sqrt{2k_B T \mu_x} \xi_x(t)$$

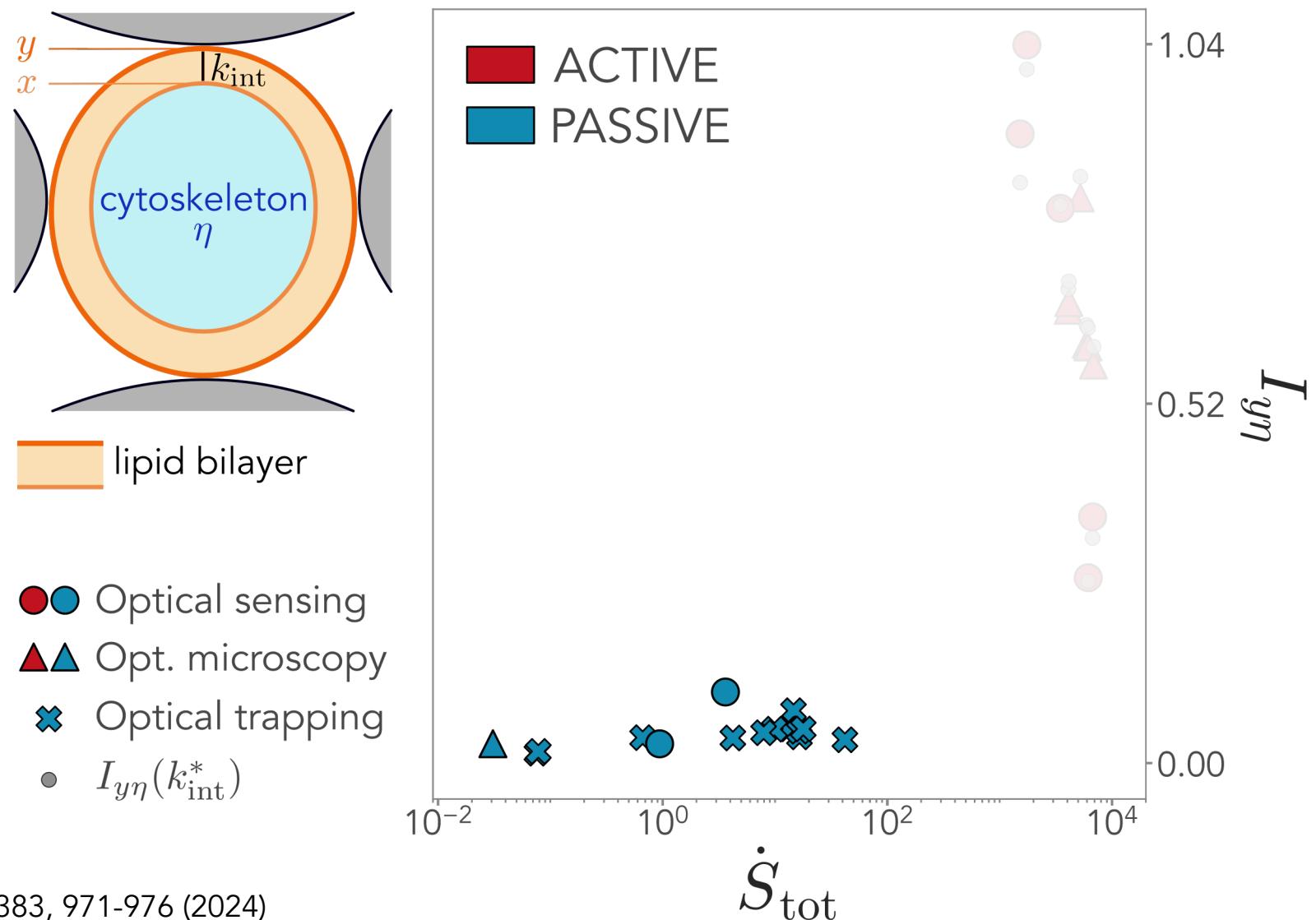
$$\dot{\eta} = -1/\tau_\eta \eta + \sqrt{2\epsilon^2/\tau_\eta} \xi_\eta(t)$$

$$k_{\text{int}}^{\text{exp}} = \arg \max_{k_{\text{int}}} \mathcal{L}_{\text{eff}}(\lambda, k_{\text{int}})$$

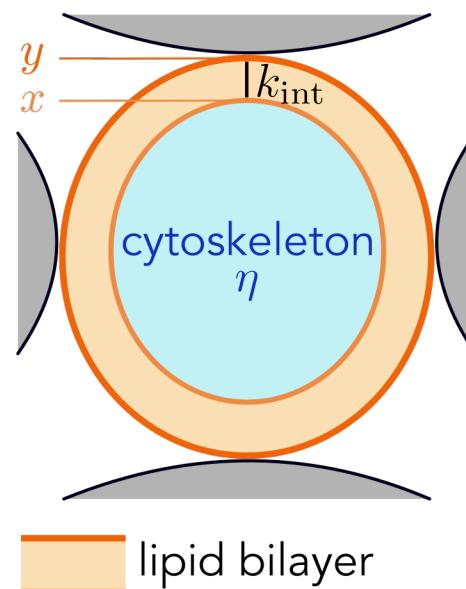
$$\lambda^{\text{exp}} : \max_{k_{\text{int}}} \mathcal{L}_{\text{eff}}(\lambda^{\text{exp}}, k_{\text{int}}) = \mathcal{L}_{\text{eff}}(\lambda^{\text{exp}}, k_{\text{int}} = k_{\text{int}}^{\text{exp}})$$

Given the estimated dissipation, how efficiently is the information on cytoskeleton transduced?

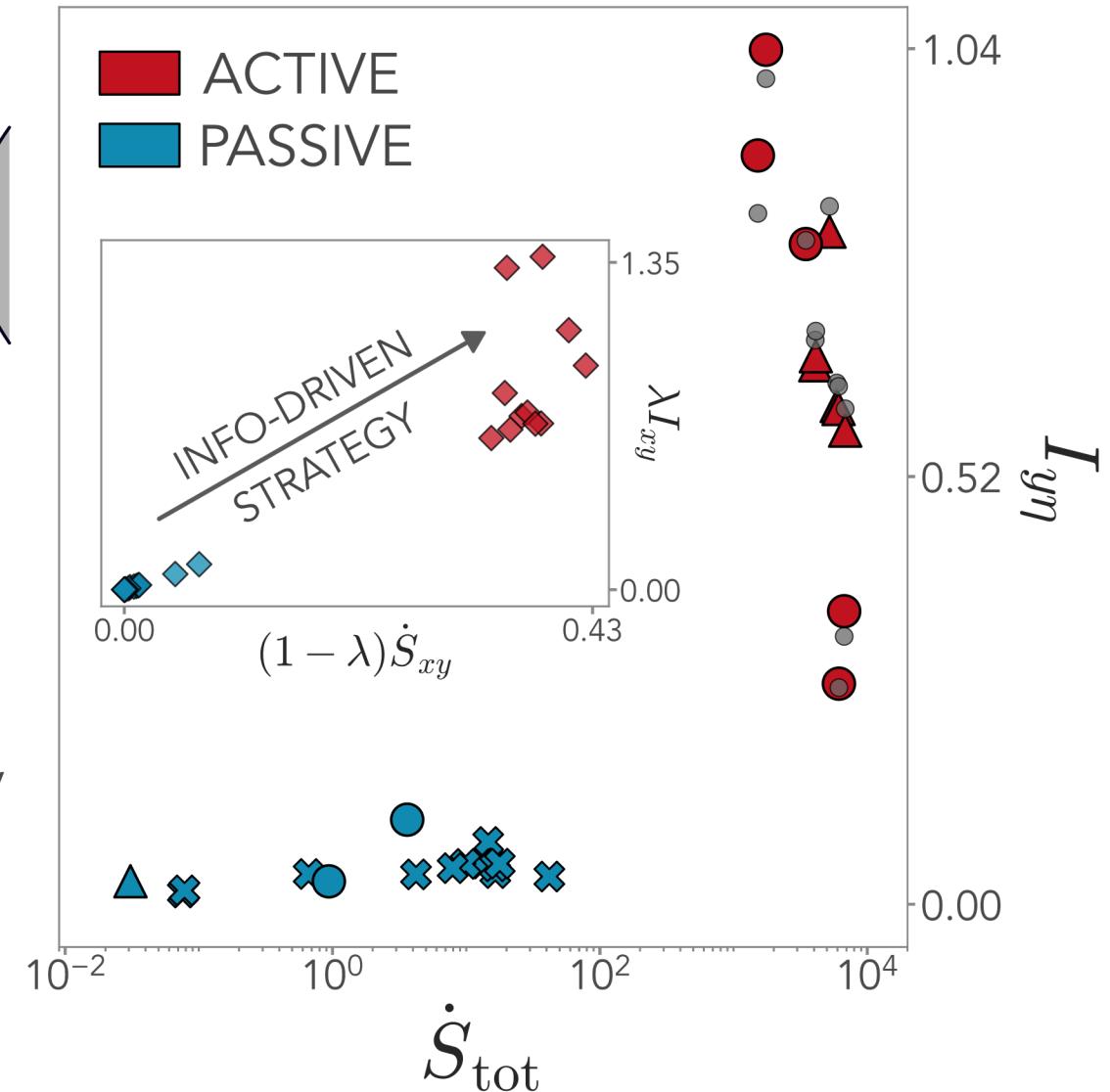
Information transduction in *red blood cells*



Information transduction in *red blood cells*



- Optical sensing
- ▲△ Opt. microscopy
- * Optical trapping
- $I_{y\eta}(k_{\text{int}}^*)$



Take-home message

- Information harvesting on hidden (dissipative) degrees of freedom is possible by tuning the coupling to accessible variables
- Transduced information can overcome the ideal case, even starting from finite-time stochastic trajectories (and empirical distributions)
- Interesting connection between transduction efficiency and mechanical conditions can be highlighted from membrane flickering in red blood cells

What's next? Estimation of forces, Multi-variables data, Pareto fronts in parameter space

- [1] G Nicoletti, DMB, Physical Review Letters 127, 228301 (2021)
- [2] G Nicoletti, DMB, Physical Review X 14, 021007 (2024)
- [3] G Nicoletti, DMB, arXiv:2403.04709 (2024)

ARXIV:2403.04709
(2024)



Thanks for your attention!

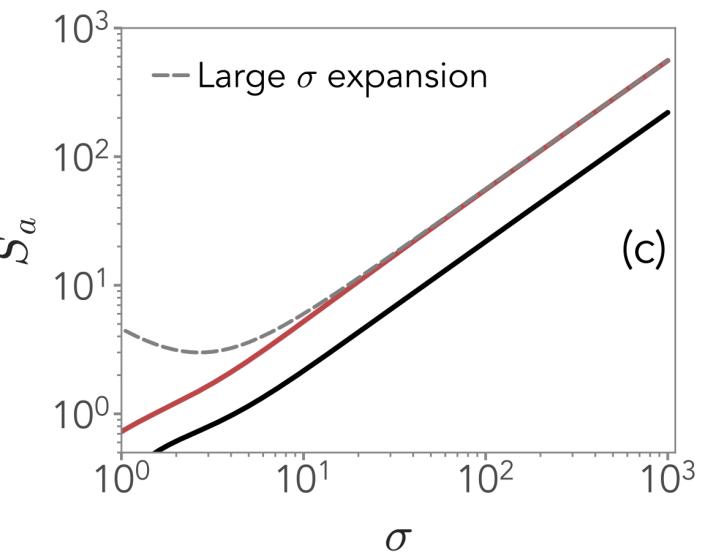
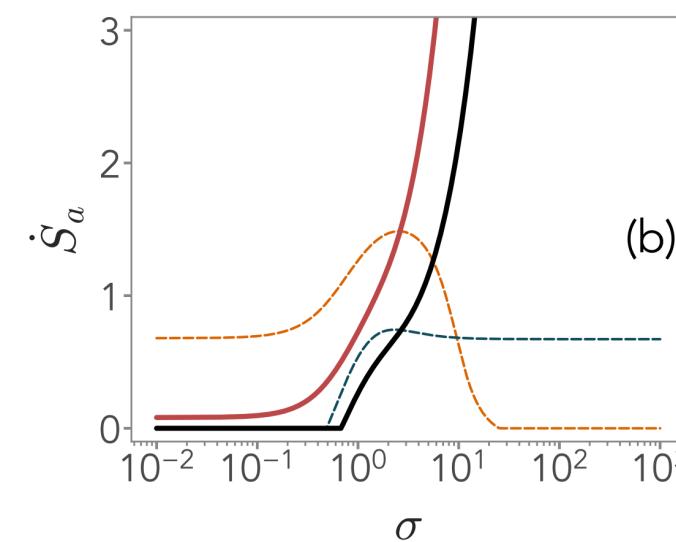
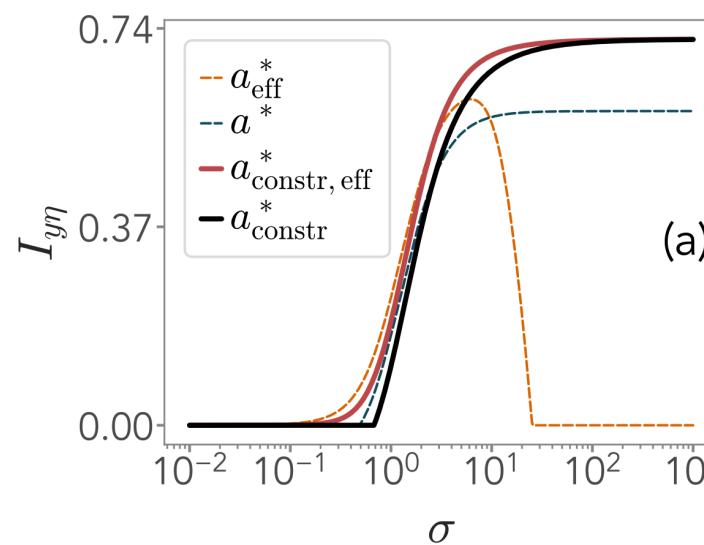
BONUS SLIDES

An *energetic constraint* is necessary

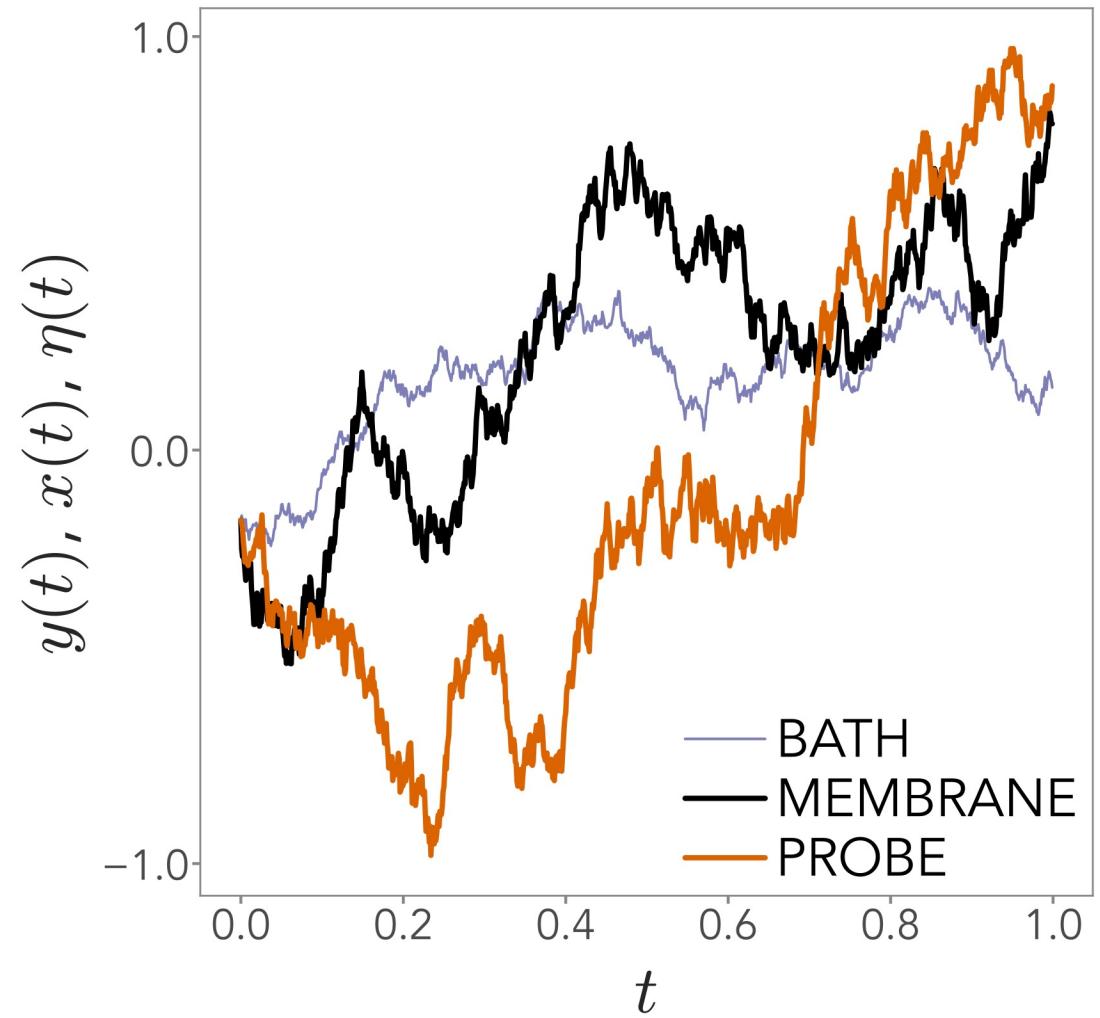
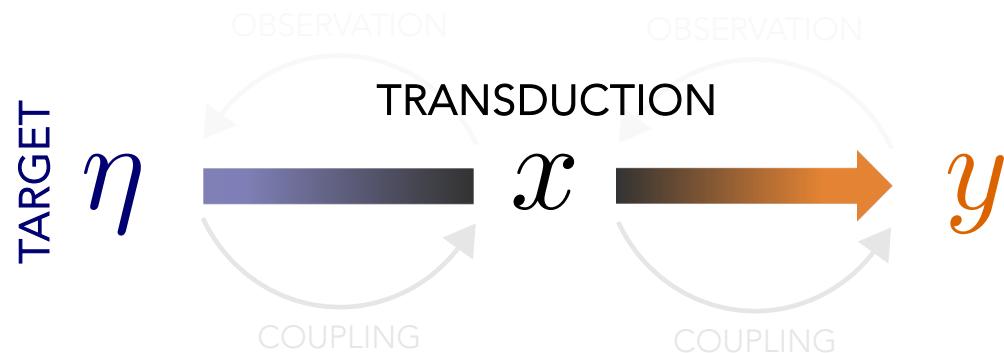
$$\max_a (\lambda I_{y\eta} - (1 - \lambda)a^2)$$

The cost of maintaining an efficient transduction at increasing σ **diverges** without an energetic constraint

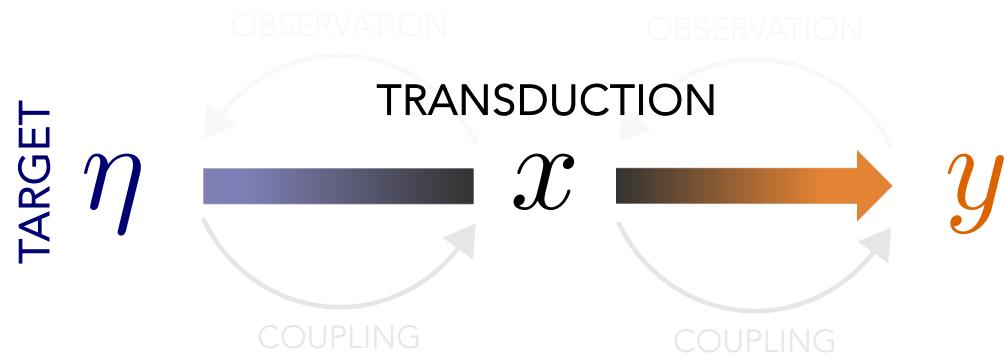
$$\dot{S}_a \approx \frac{\lambda(2 + \theta_\eta + \sigma^2)}{2\sqrt{\theta_\eta(1 - \lambda)} \sigma}$$



A more realistic scenario: *finite-time trajectories*

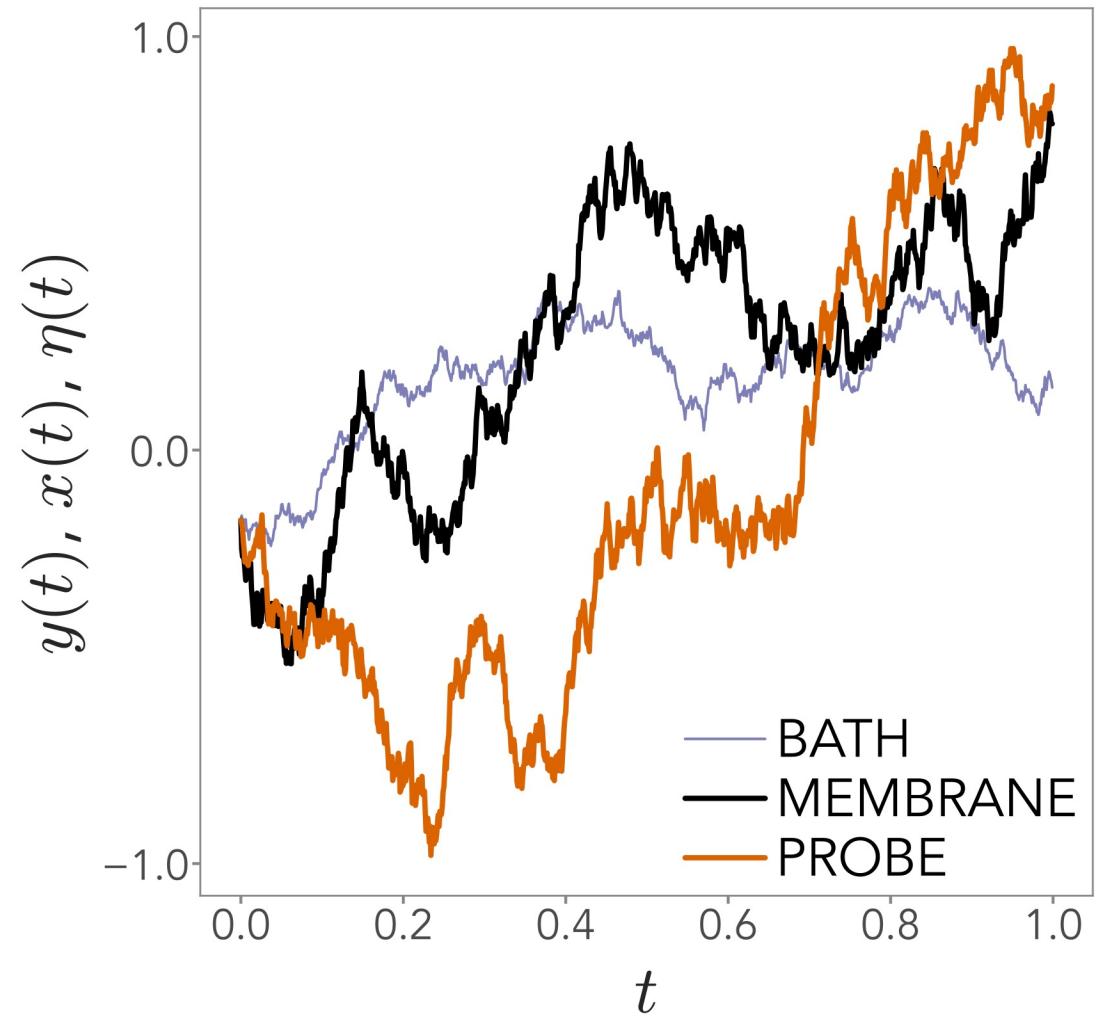


A more realistic scenario: *finite-time trajectories*

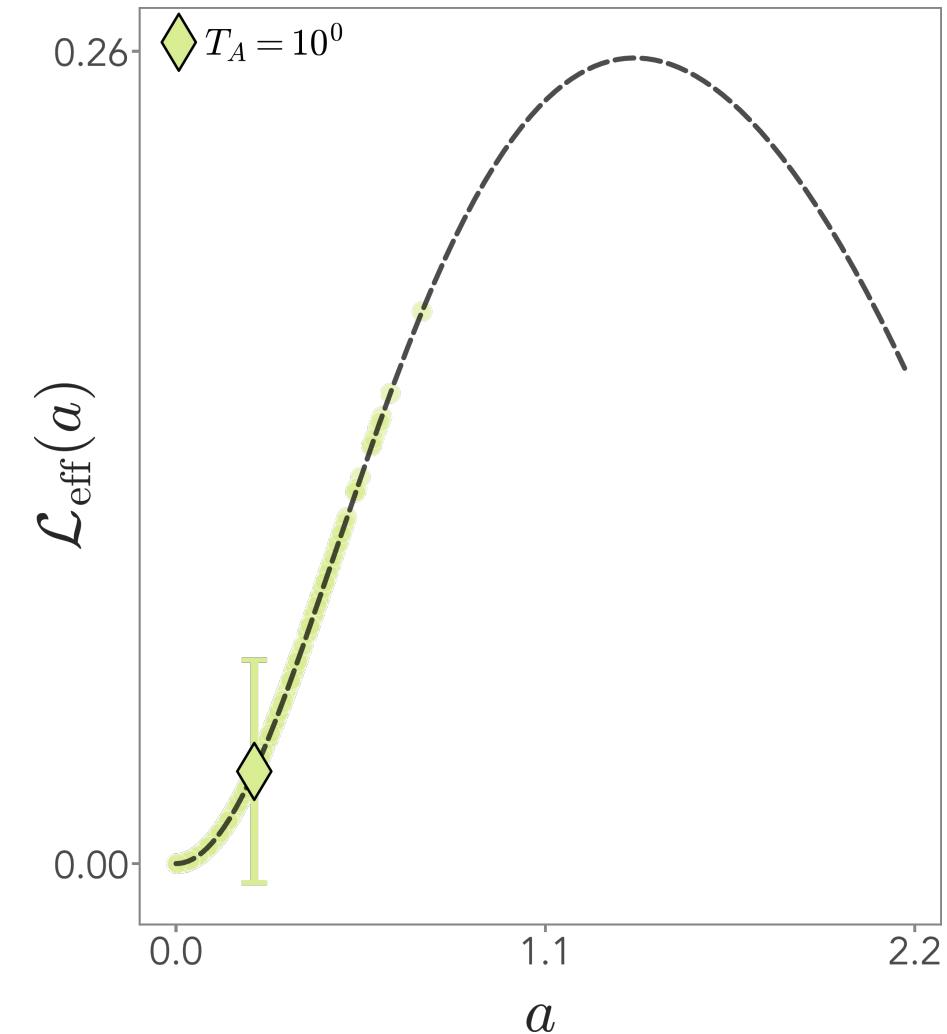
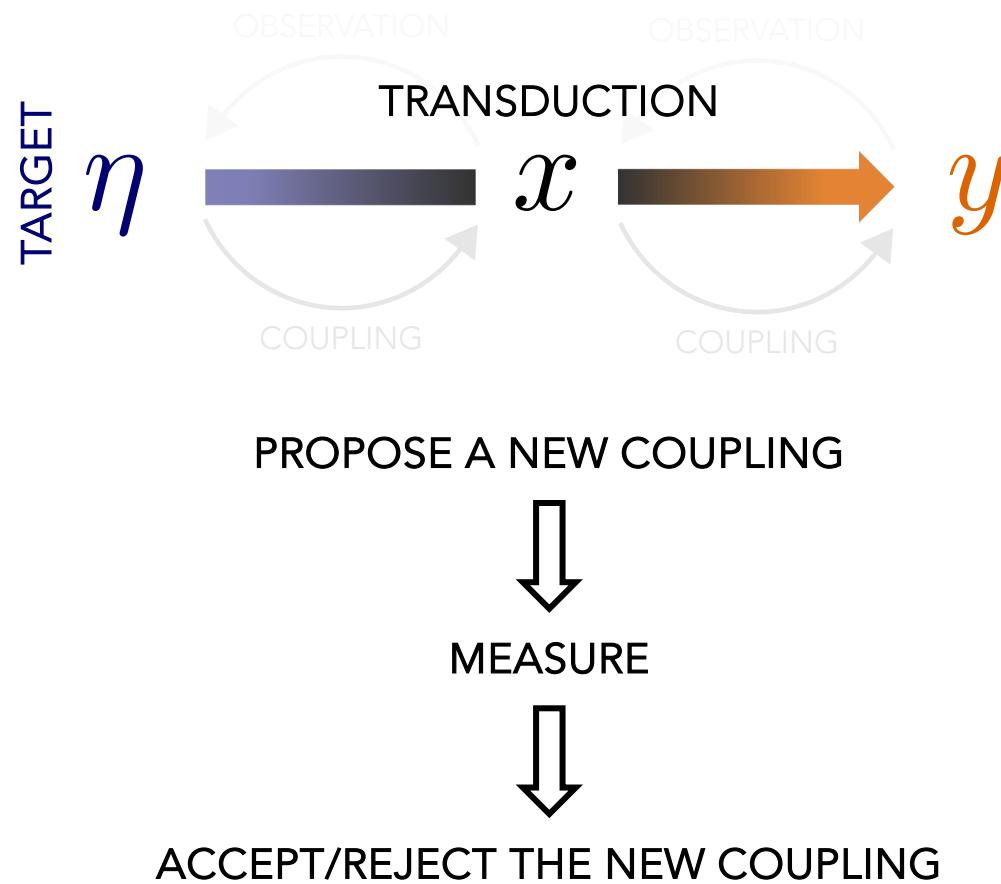


$$I_{xy} \rightarrow i_{xy}(t) = \log \frac{p(x(t), y(t))}{p(x(t))p(y(t))}$$

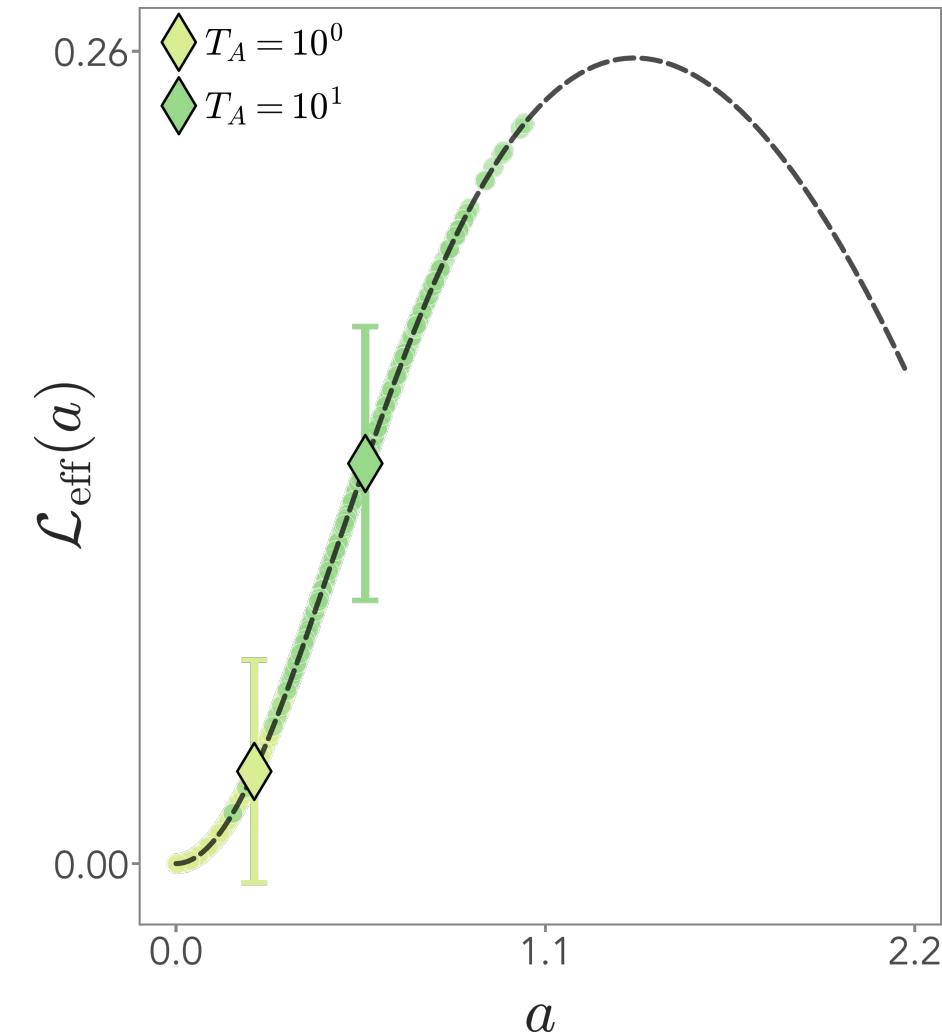
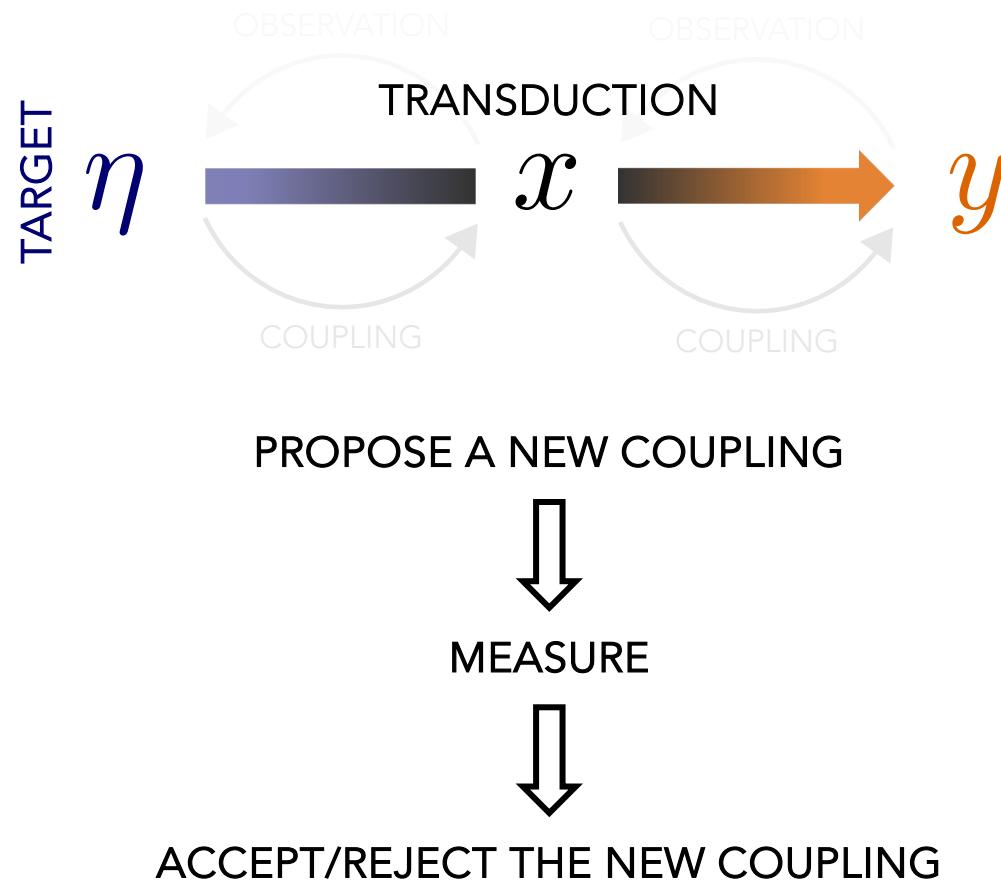
$$\dot{S}_{xy} \rightarrow \dot{s}_{xy}(t) = \mathbf{A}_{xy}^{\text{red}} \circ (\dot{x}(t), \dot{y}(t))^T$$



A more realistic scenario: *finite-time trajectories*



A more realistic scenario: *finite-time trajectories*



A more realistic scenario: *finite-time trajectories*

