

Informational steady state and entropy production in continuously monitored systems

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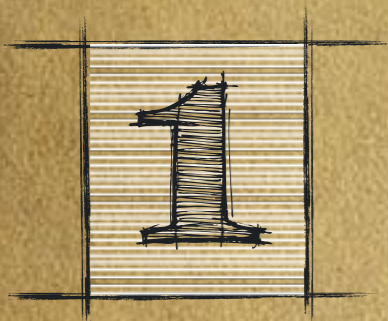


New trends in Quantum Thermodynamics

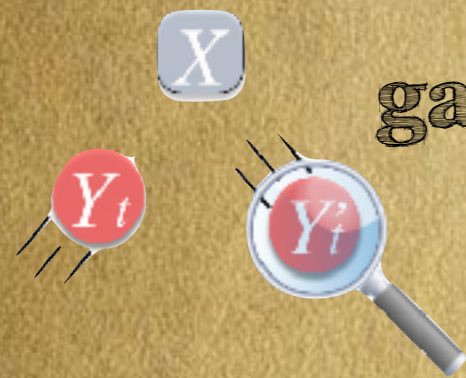
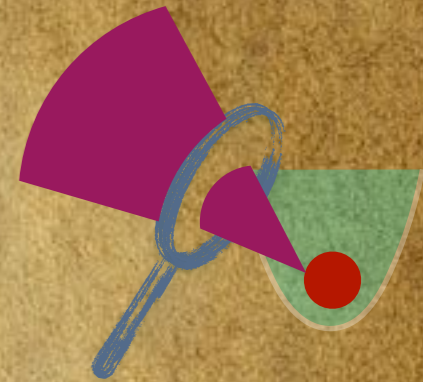
University of Surrey, 8 July 2024



STRUCTURE OF THE TALK



Formalism for entropy production
in continuously measured
quantum systems



Informational steady states:
gaining & losing through measurement



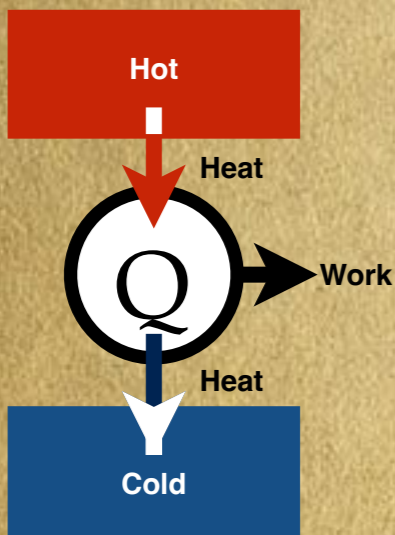
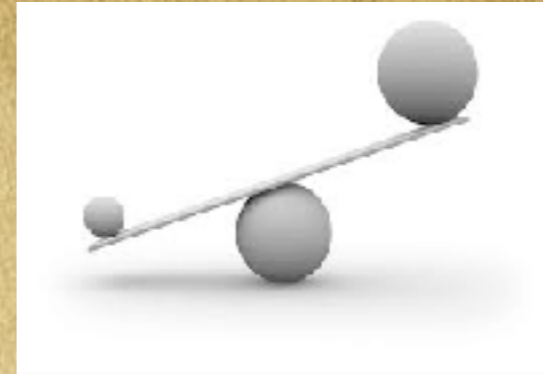
Observing irreversible
entropy in measured mesoscopic
quantum settings





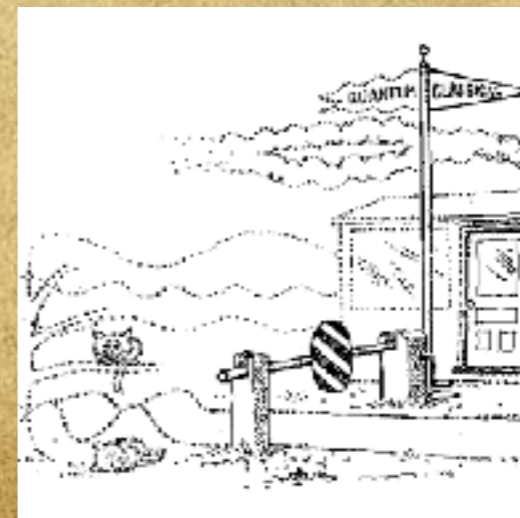
Why entropy production?

Non-equilibrium processes dissipate energy. This produces irreversible increase of entropy



Entropy production for estimating the performance of devices (**exergy** is reduced by irreversibility)

Fantastic framework for pinpointing the quantum-to-classical transition



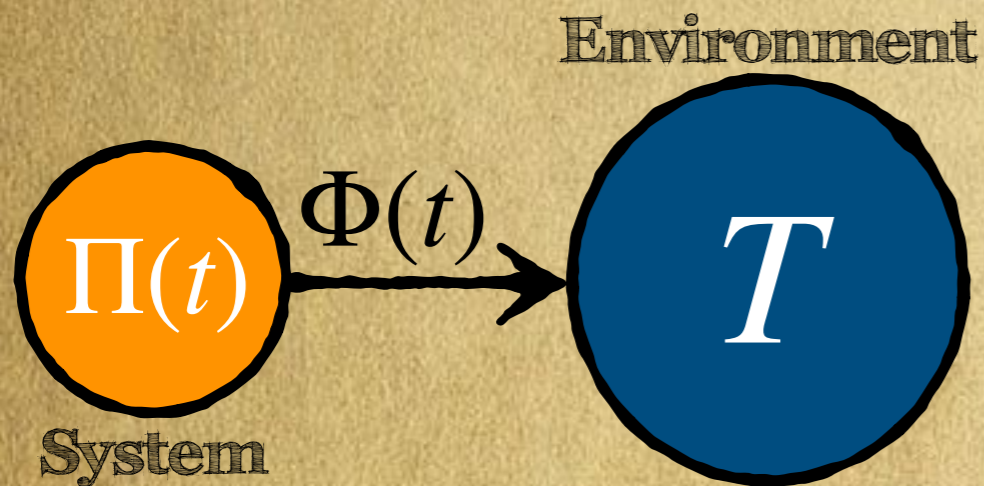


Entropy production

Second Law: $\Delta S \geq \int \frac{\delta Q}{T} \Rightarrow \Sigma = \Delta S - \int \frac{\delta Q}{T}$

Clausius: "Uncompensated transformation"

Entropy production



$$\frac{dS}{dt} = \Pi(t) + \Phi(t)$$

Entropy production rate

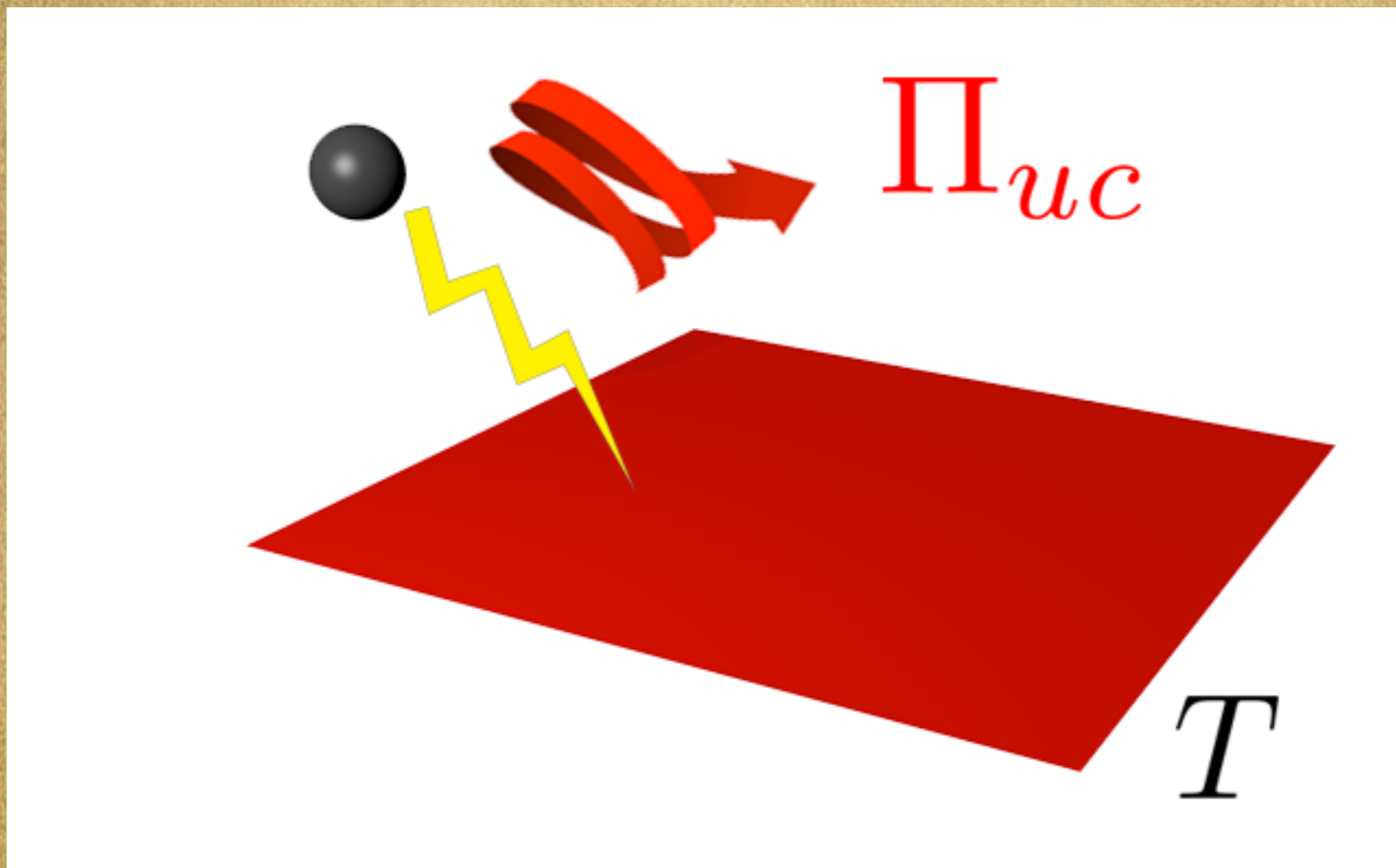
Entropy flux rate

Which is the role of quantum fluctuations on entropy production?

What happens if you plug in the effects of measuring?



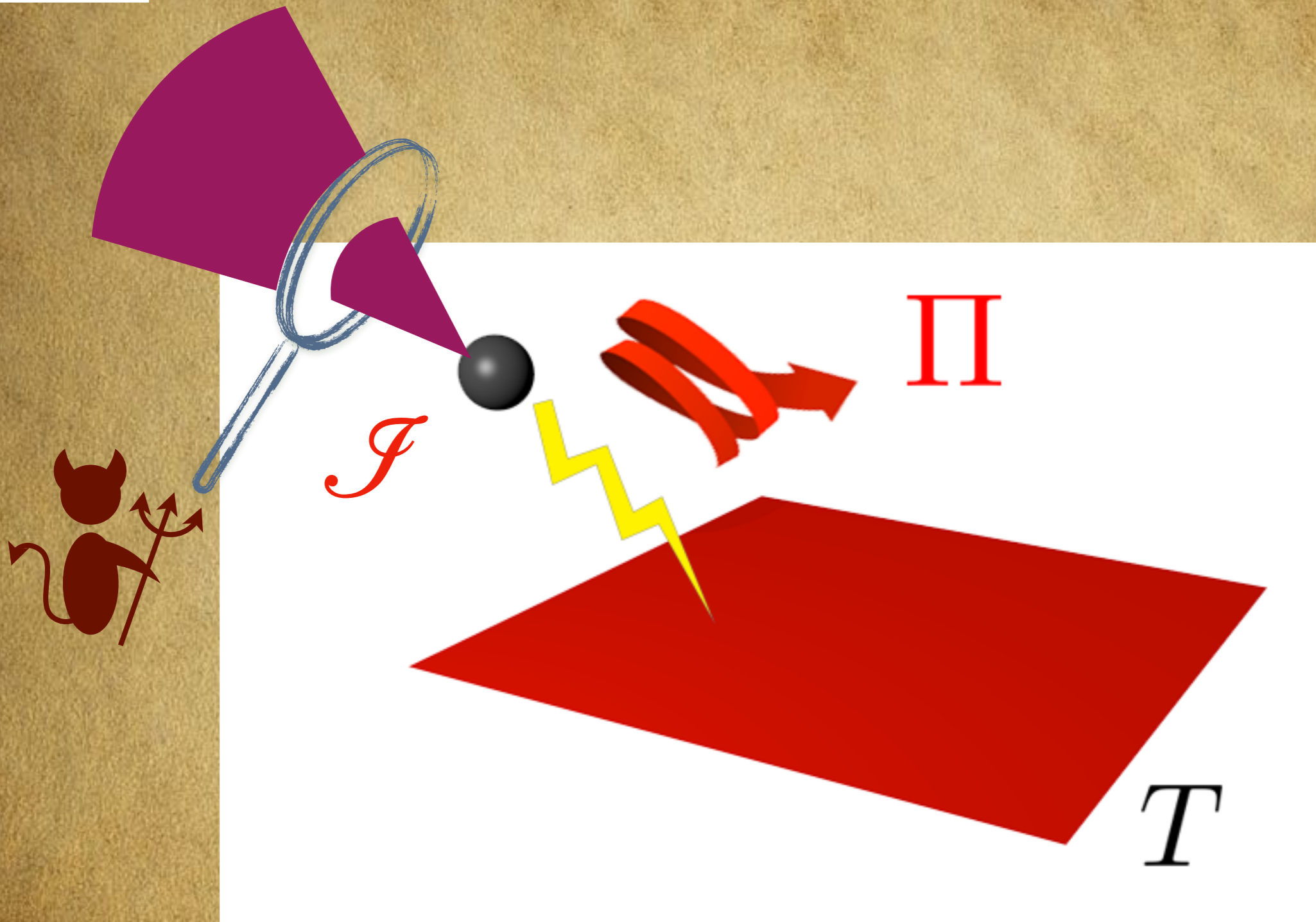
Don't look yet!!!



A Belenchia, L Mancino, G T Landi, and M Paternostro,
Nature Quantum Information 6, 97 (2019)

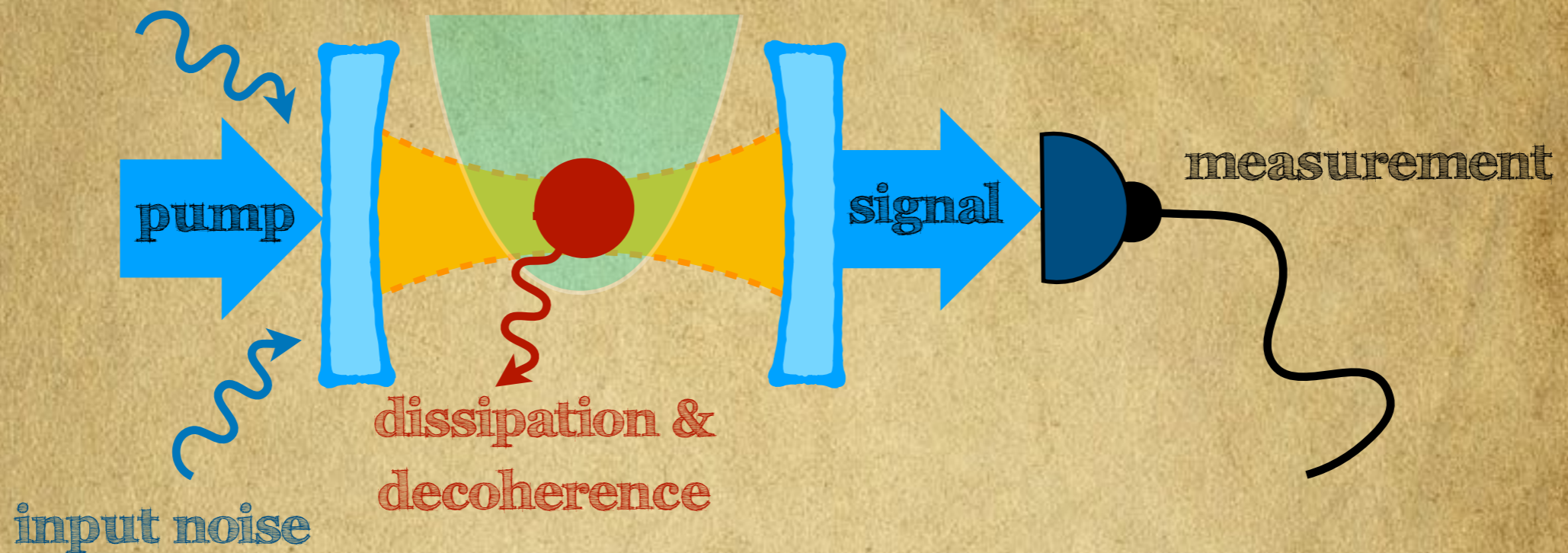


..now open your eyes..



A Belenchia, L Mancino, G T Landi, and M Paternostro,
Nature Quantum Information 6, 97 (2019)

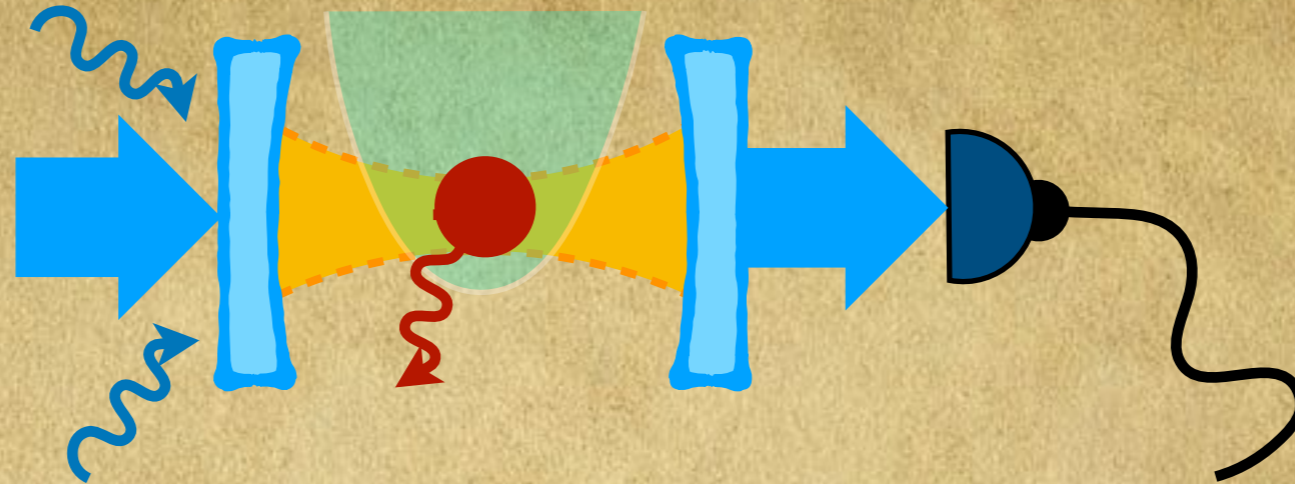
Let's fix the ideas



Now restrict the framework to quadratic evolution
and Gaussian states & measurements

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

A Belenchia, et al., Nature Quantum Information 6, 97 (2019)



Stochastic master equation

$$d\rho = \underbrace{-i[\hat{H}, \rho]dt + \sum_k \mathcal{D}[\hat{c}_k](\rho)dt}_{\text{Deterministic dynamics}} + \underbrace{\sum_k \sqrt{\eta_k} \mathcal{H}[\hat{c}_k](\rho)dw_k}_{\text{Stochastic terms}}$$

Deterministic dynamics

Stochastic terms

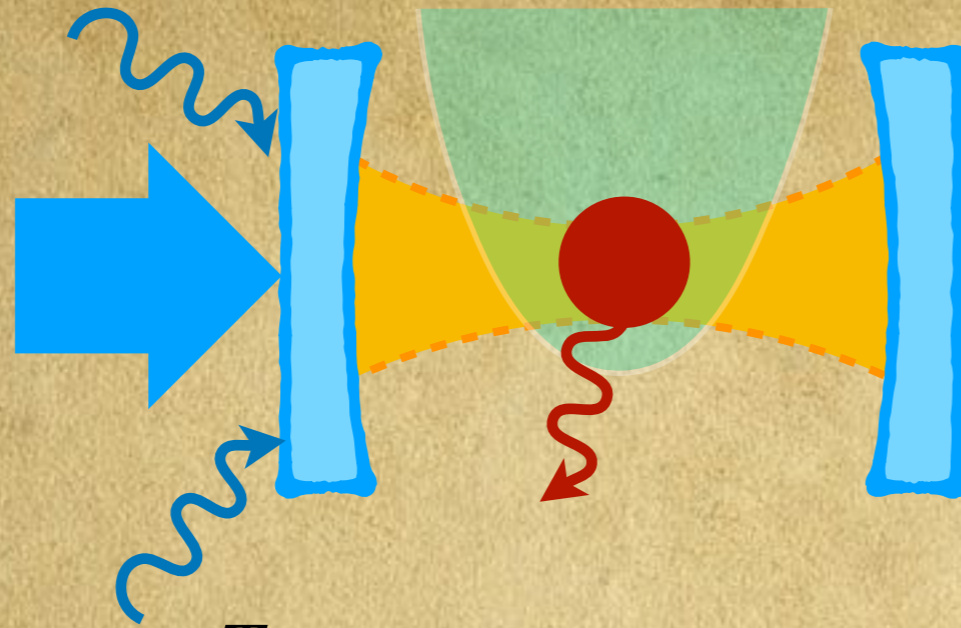
$$\mathcal{H}[\hat{c}]\rho = \hat{c}\rho + \rho\hat{c}^\dagger - \langle \hat{c} + \hat{c}^\dagger \rangle \rho$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

M. G. Genoni, L. Lami, and A. Serafini, Contemp. Phys. 57, 331 (2016)



Un-Conditioned Gaussian dynamics



$$\dot{\sigma} = A\sigma + \sigma A^T + D$$

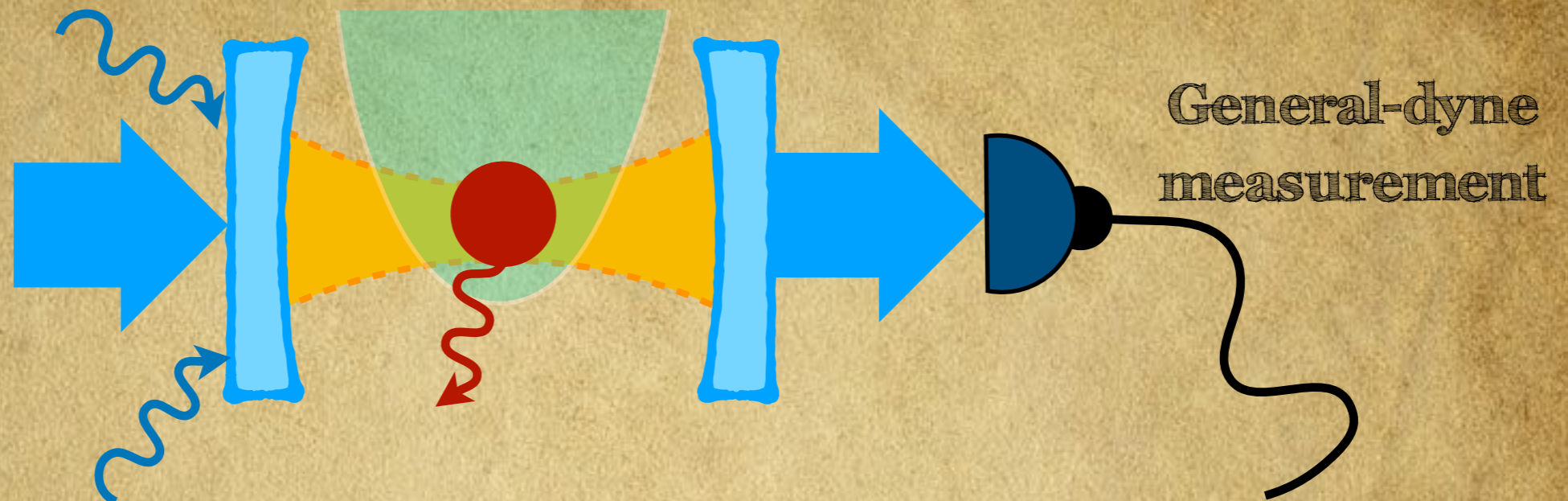
$$d\mathbf{x} = (A\mathbf{x} + \mathbf{b}) dt$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

A Belenchia, et al., Nature Quantum Information 6, 97 (2019)



Conditioned Gaussian dynamics



$$\dot{\sigma} = A\sigma + \sigma A^T + D$$



$$\dot{\sigma} = \tilde{A}\sigma + \sigma\tilde{A}^T + \tilde{D} - \sigma BB^T\sigma = A\sigma + \sigma A^T + D - \chi(\sigma)$$

contains terms depending on the measurement

$$d\mathbf{x} = (A\mathbf{x} + \mathbf{b}) dt$$



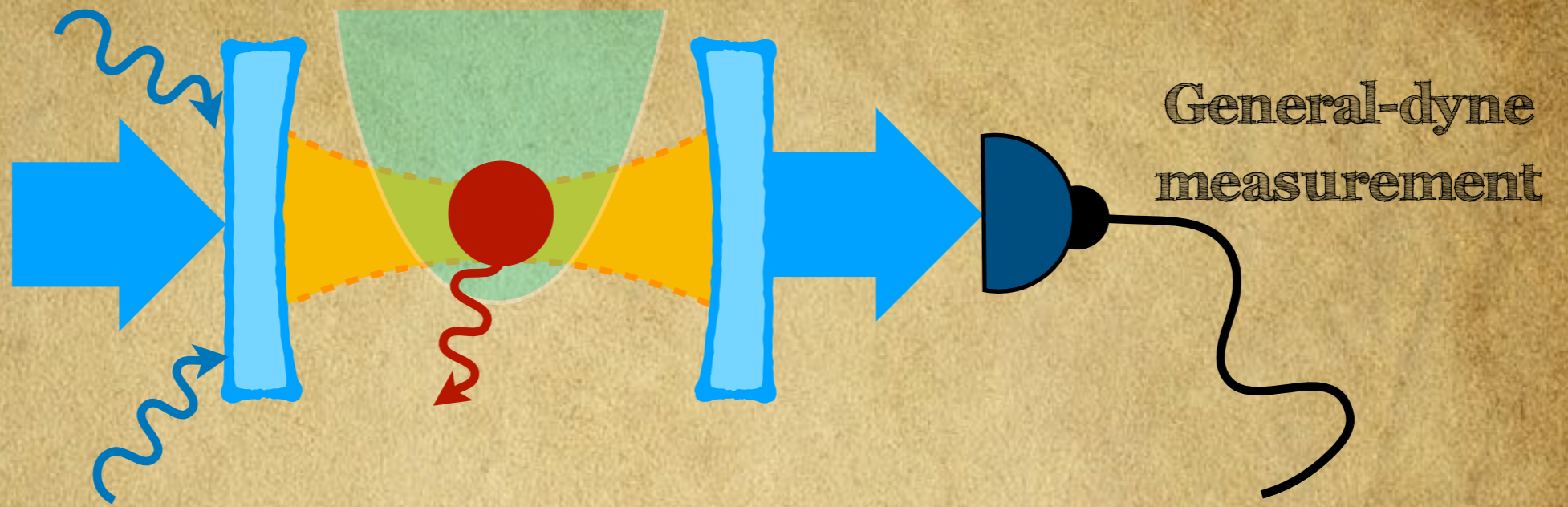
$$d\bar{\mathbf{x}} = (A\bar{\mathbf{x}} + \mathbf{b})dt + \mu(\sigma)d\mathbf{w}$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

A Belenchia, et al., Nature Quantum Information 6, 97 (2019)



Conditioned Gaussian dynamics



$$dS = d\Phi_{\bar{x}} + d\Sigma_{\bar{x}}$$

deterministic
(only depends on CM)

stochastic
(depend also on
1st moments)

$$\phi = \mathbb{E} [d\Phi_{\bar{x}}/dt]$$

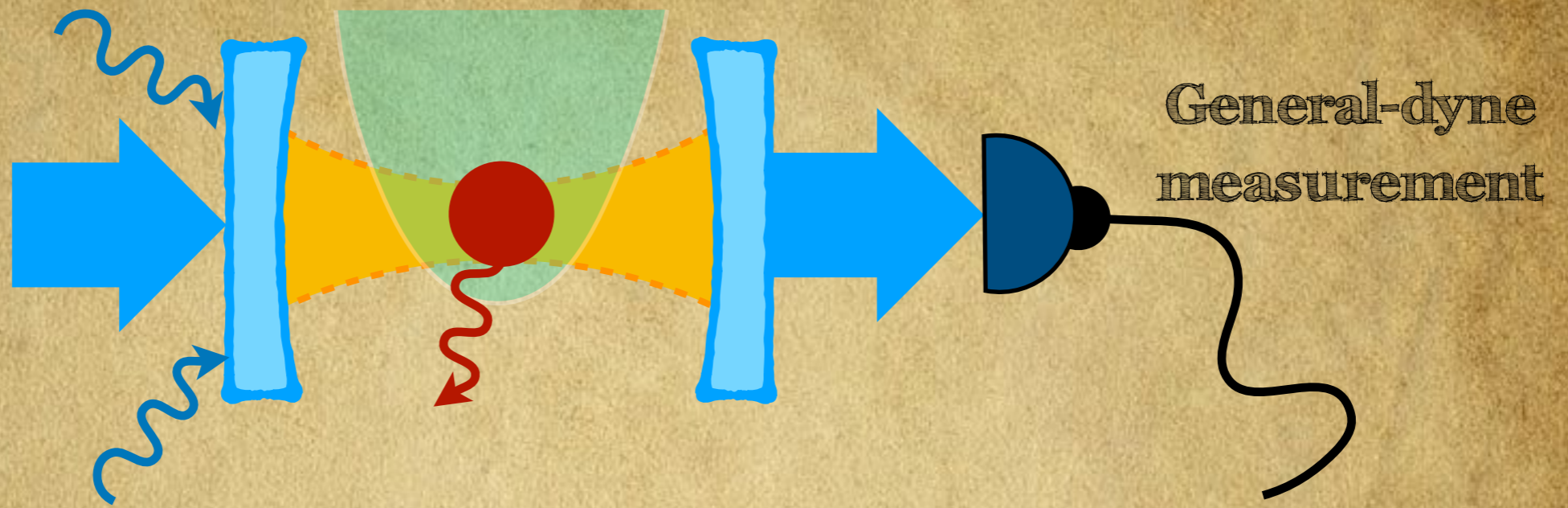
$$\Pi = \mathbb{E} [d\Sigma_{\bar{x}}/dt]$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

A Belenchia, et al., Nature Quantum Information 6, 97 (2019)



Conditioned Gaussian dynamics



$$dS = dS_{uc} + \dot{\mathcal{J}} dt$$

$$\dot{\mathcal{J}} = \frac{1}{2} \text{Tr}[\sigma^{-1} D - \sigma^{-1} \chi(\sigma)] - \frac{1}{2} \text{Tr}[\sigma_{uc}^{-1} D]$$

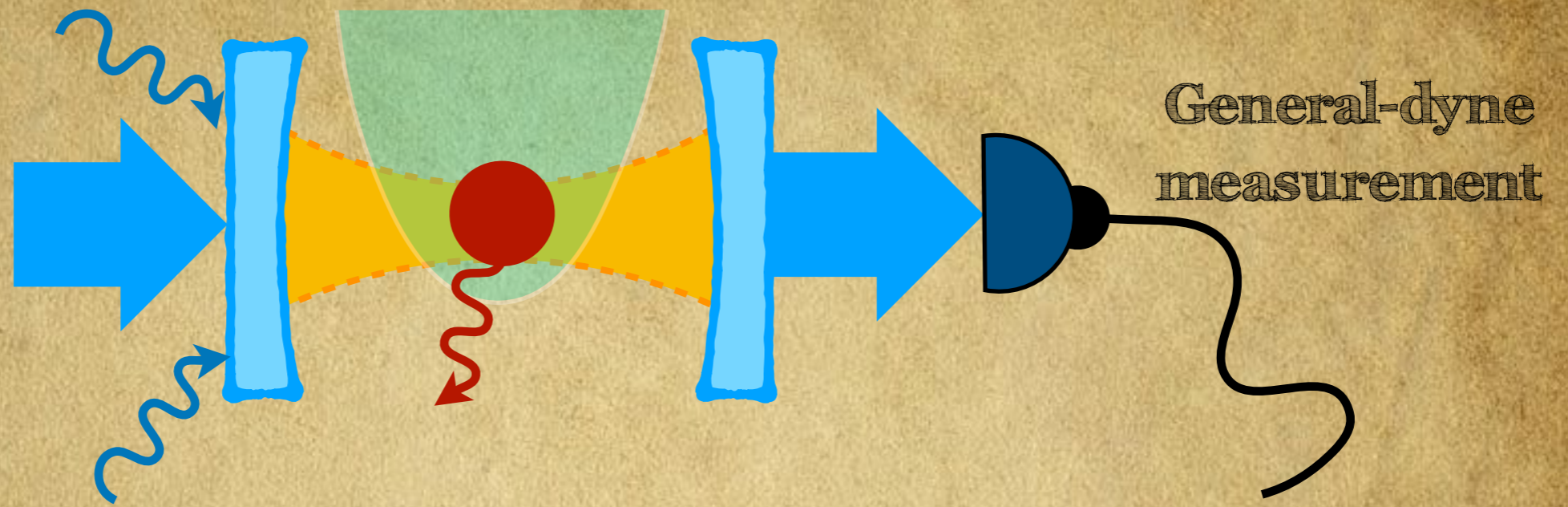
$$\phi = \mathbb{E} [d\Phi_{\bar{x}}/dt]$$

$$\Pi = \mathbb{E} [d\Sigma_{\bar{x}}/dt]$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

A Belenchia, et al., Nature Quantum Information 6, 97 (2019)

Conditioned Gaussian dynamics



$$dS = dS_{uc} + \dot{\mathcal{I}} dt$$

$$\Pi_c(t) = \Pi_{uc}(t) + \dot{\mathcal{I}}$$

$$\phi = \mathbb{E} \left[d\Phi_{\bar{x}} / dt \right]$$

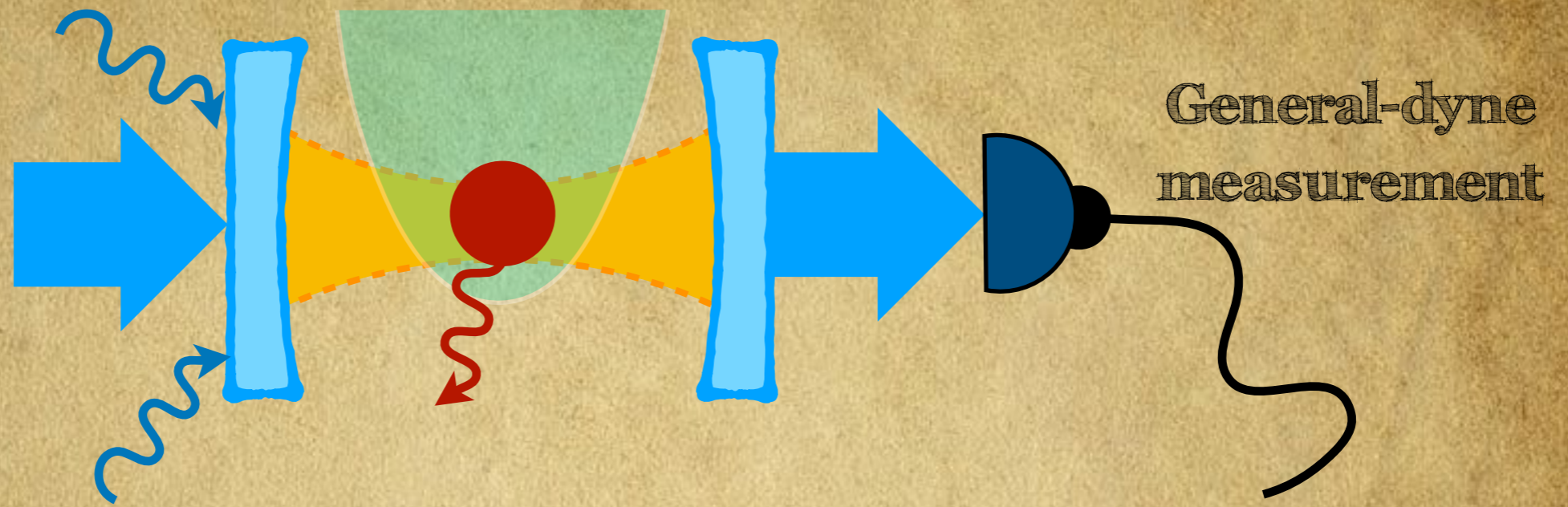
$$\Pi = \mathbb{E} \left[d\Sigma_{\bar{x}} / dt \right]$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

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Conditioned Gaussian dynamics



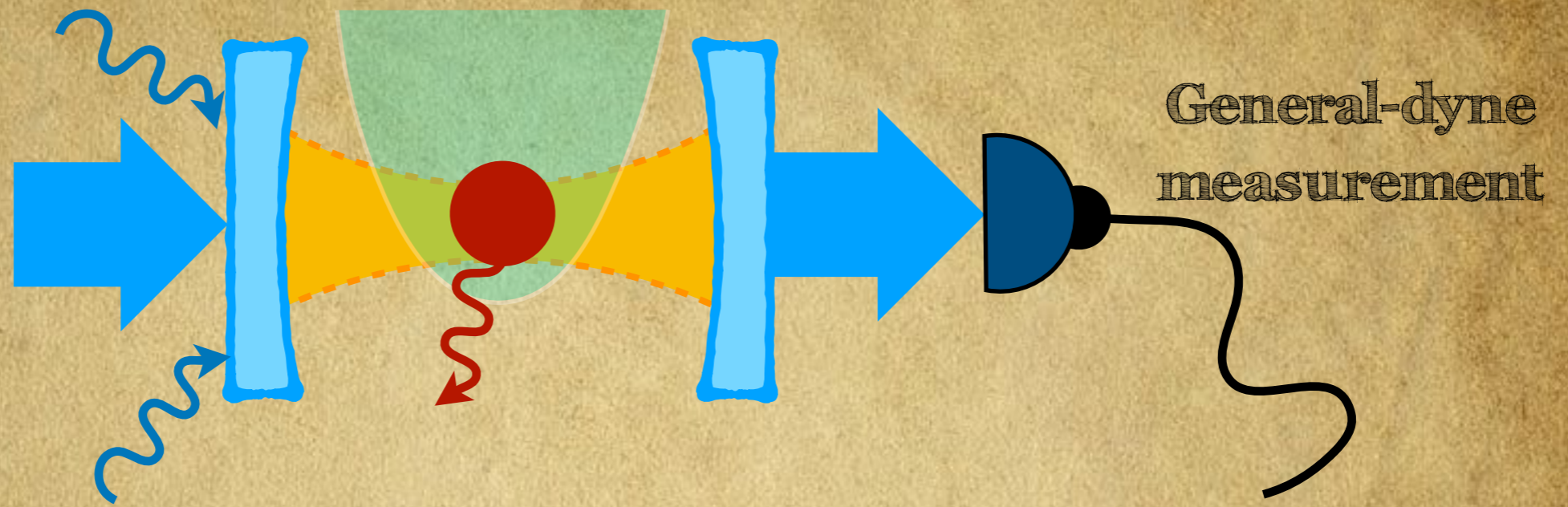
$\Pi_{uc}(t) \geq 0$ second law for un-conditioned dynamics

$$\Pi_c(t) = \Pi_{uc}(t) + \dot{\mathcal{J}}$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

A Belenchia, et al., Nature Quantum Information 6, 97 (2019)

Conditioned Gaussian dynamics



$\Pi_c(t) \geq \dot{\mathcal{I}}$ second law for conditioned dynamics

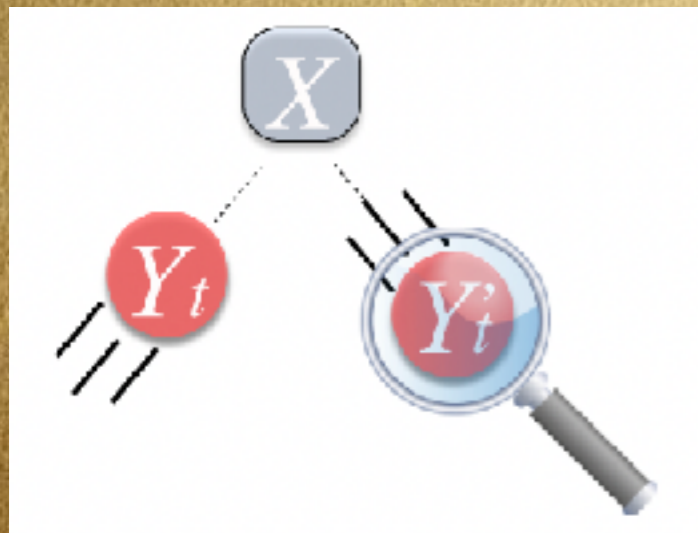
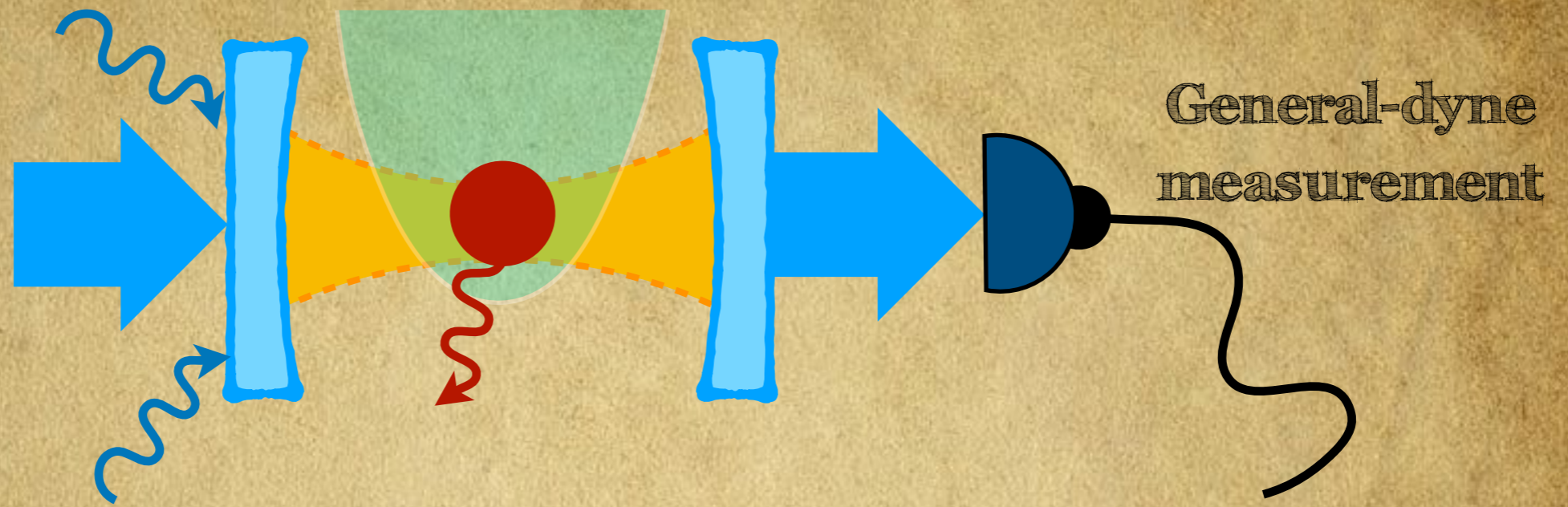
$$\Pi_c(t) = \Pi_{uc}(t) + \dot{\mathcal{I}}$$

A Belenchia, M Paternostro, and G T Landi, PRX Quantum 3, 010303 (2022)

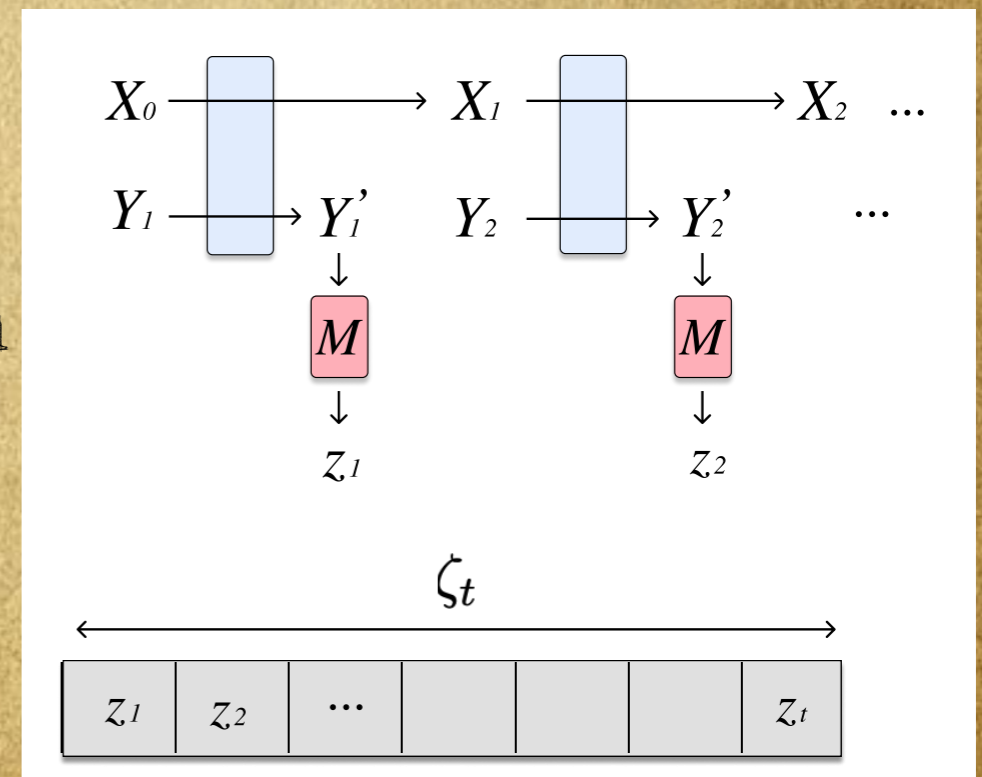
A Belenchia, et al., Nature Quantum Information 6, 97 (2019)



Generalising it



Interpretation through collisional model



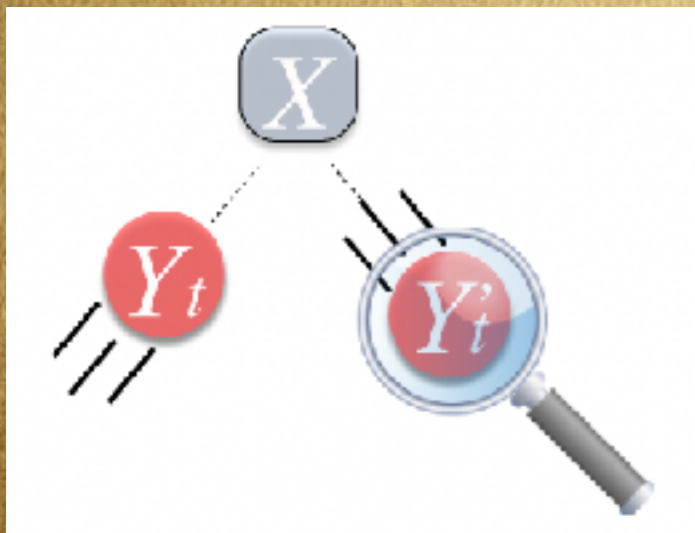
Can we generalise?

$$P(\zeta_t) = \text{tr}_{XY_1 \dots Y_t} \left\{ M_{z_t} \dots M_{z_1} \rho_{XY_1 \dots Y_t} M_{z_1}^\dagger \dots M_{z_t}^\dagger \right\}$$

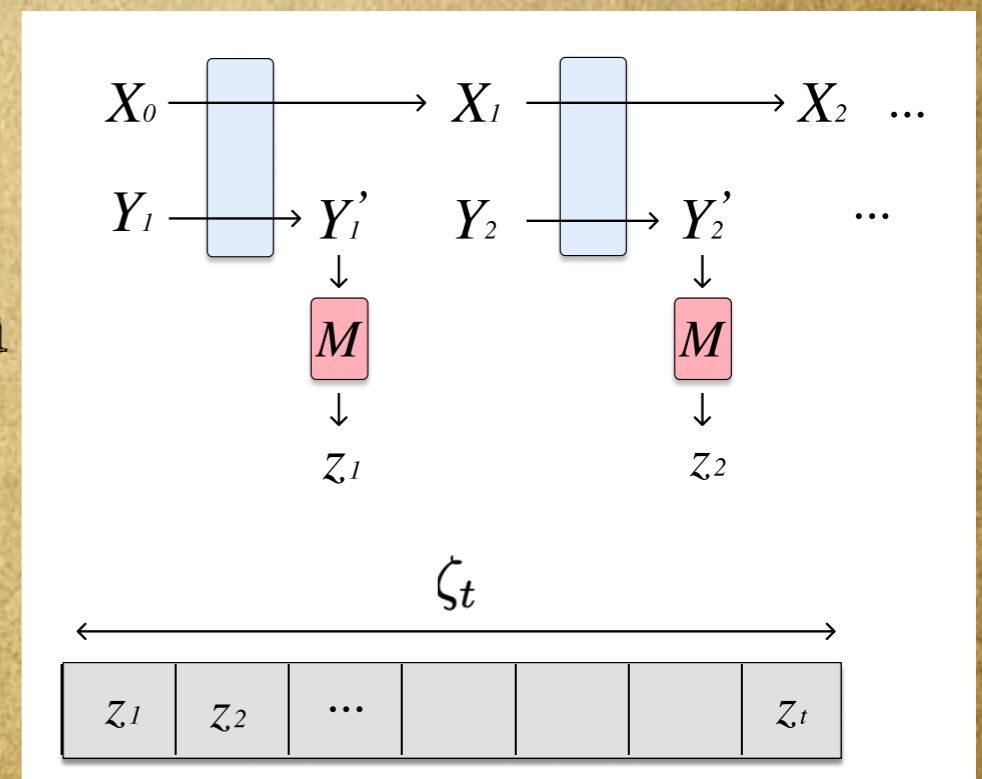
$$\rho_{XY_1 \dots Y_t} = \left(\prod_{k=1}^t U_k \right) \left(\rho_{X_0} \otimes_{j=1}^t \rho_{Y_j} \right) \left(\prod_{k=1}^t U_k \right)^\dagger$$

$$\rho_{X_t | \zeta_t} = \frac{1}{P(\zeta_t)} \text{tr}_{Y_1 \dots Y_t} \left\{ \left(\prod_{k=1}^t M_{z_k} \right) \rho_{XY_1 \dots Y_t} \left(\prod_{k=1}^t M_{z_k} \right)^\dagger \right\}$$

Conditional state



Interpretation through collisional model





Can we generalise?

Information rate

$$\Delta I_t := I(X_t : \zeta_t) - I(X_{t-1} : \zeta_{t-1})$$

can take any sign

Holevo information: info on X contained in ζ_t

$$I(X_t : \zeta_t) := S(X_t) - S(X_t | \zeta_t)$$

strictly positive



$$I(X_t : \zeta_t) - I(X_t : \zeta_{t-1})$$

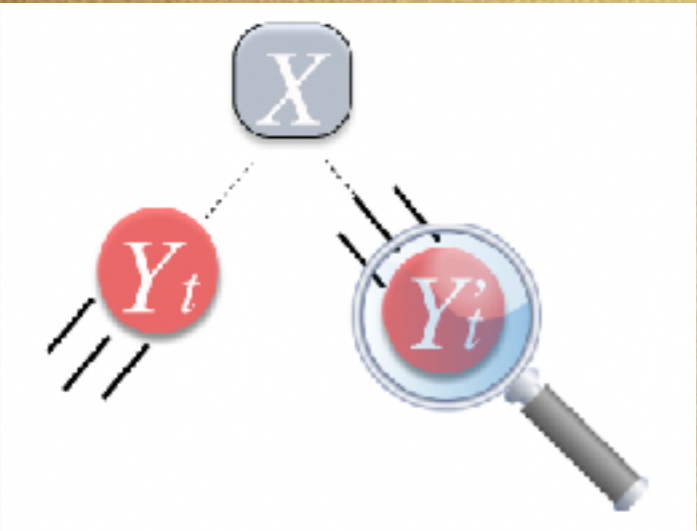
$$\text{differential} = S(X_t | \zeta_{t-1}) - S(X_t | \zeta_t)$$

information gain

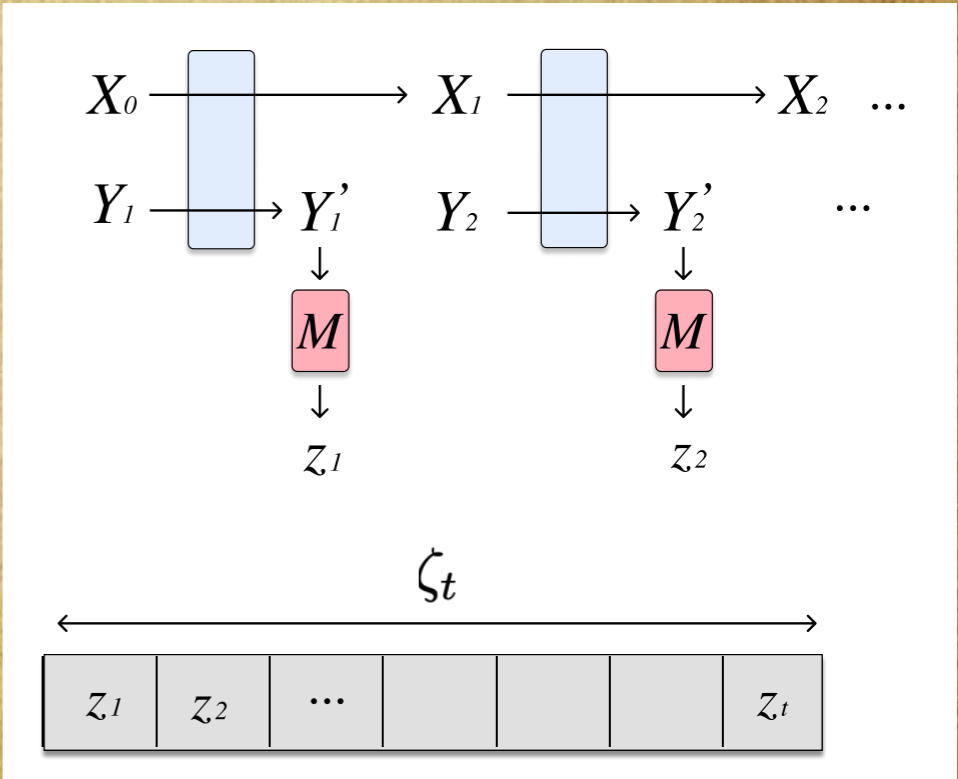
information

$$L_t := I(X_{t-1} : \zeta_{t-1}) - I(X_t : \zeta_{t-1}) \quad \text{loss}$$

non-negative



Interpretation through collisional model





Can we generalise?

Information rate

$$\Delta I_t := I(X_t : \zeta_t) - I(X_{t-1} : \zeta_{t-1})$$

can take any sign

Holevo information: info on X contained in ζ_t

$$I(X_t : \zeta_t) := S(X_t) - S(X_t | \zeta_t)$$

strictly positive

$$\Delta I_t = G_t - L_t \xrightarrow{\text{steady state}} \Delta I_\infty = 0 \quad \text{with} \quad G_\infty = L_\infty \neq 0$$

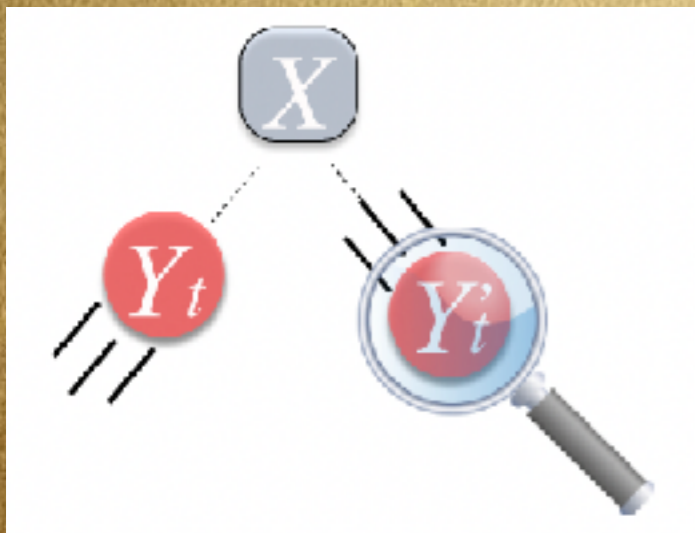
Informational steady state:

balance between gain & loss

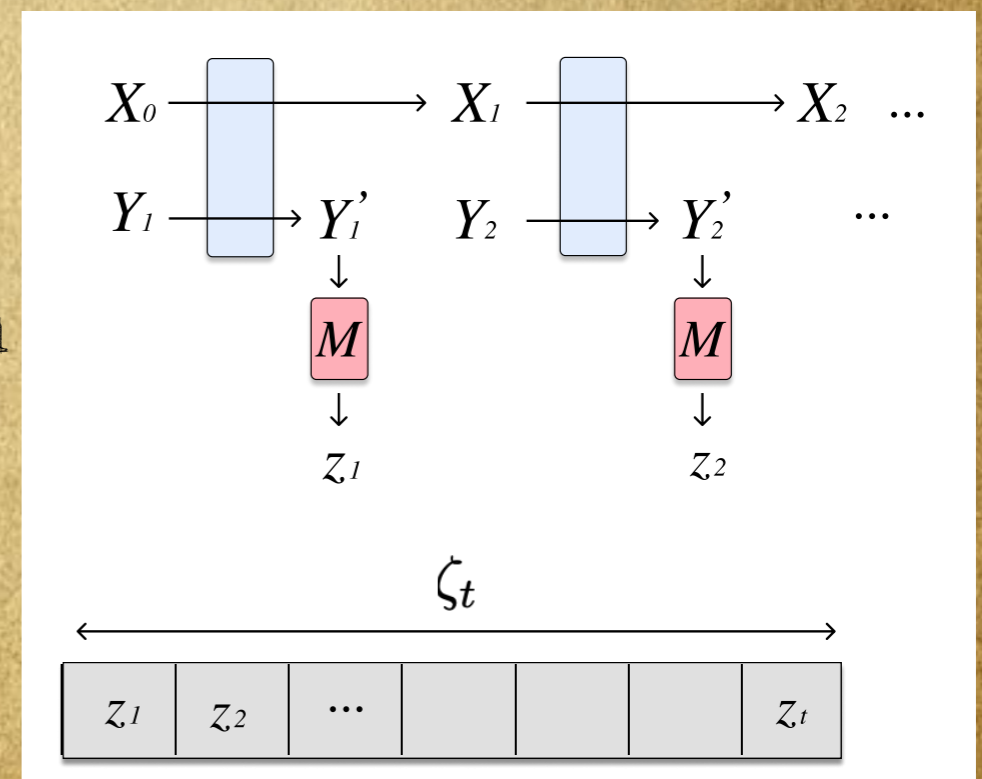
measurements sustain the state

$$\Delta \Sigma_t^u = S(X_t) - S(X_{t-1}) + \Delta \Phi_t^u$$

$$\Delta \Sigma_t^c = S(X_t | \zeta_t) - S(X_{t-1} | \zeta_{t-1}) + \Delta \Phi_t^c$$



Interpretation through collisional model





Can we generalise?

Conditioning on the outcomes is subjective (I decide to read outcomes or not...)

No influence on flux of entropy into the ancillae

$$\Delta\Phi_t^c = \Delta\Phi_t^u$$

$$\Delta\Sigma_t^c = \Delta\Sigma_t^u + \Delta I_t$$

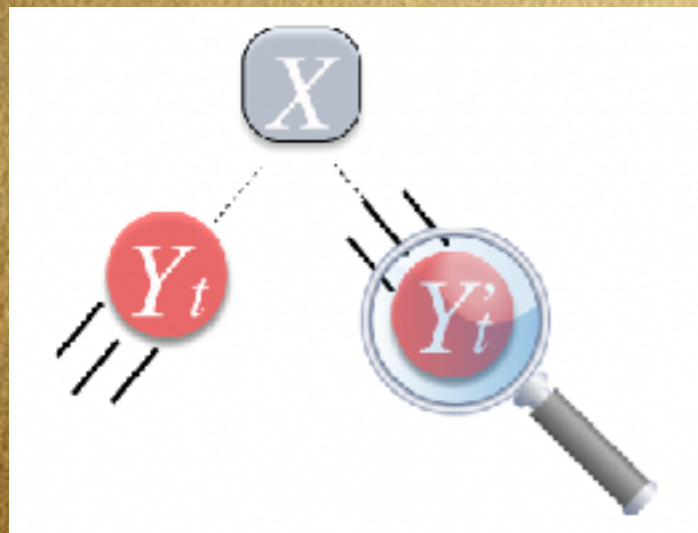
Informational steady state:

balance between gain & loss

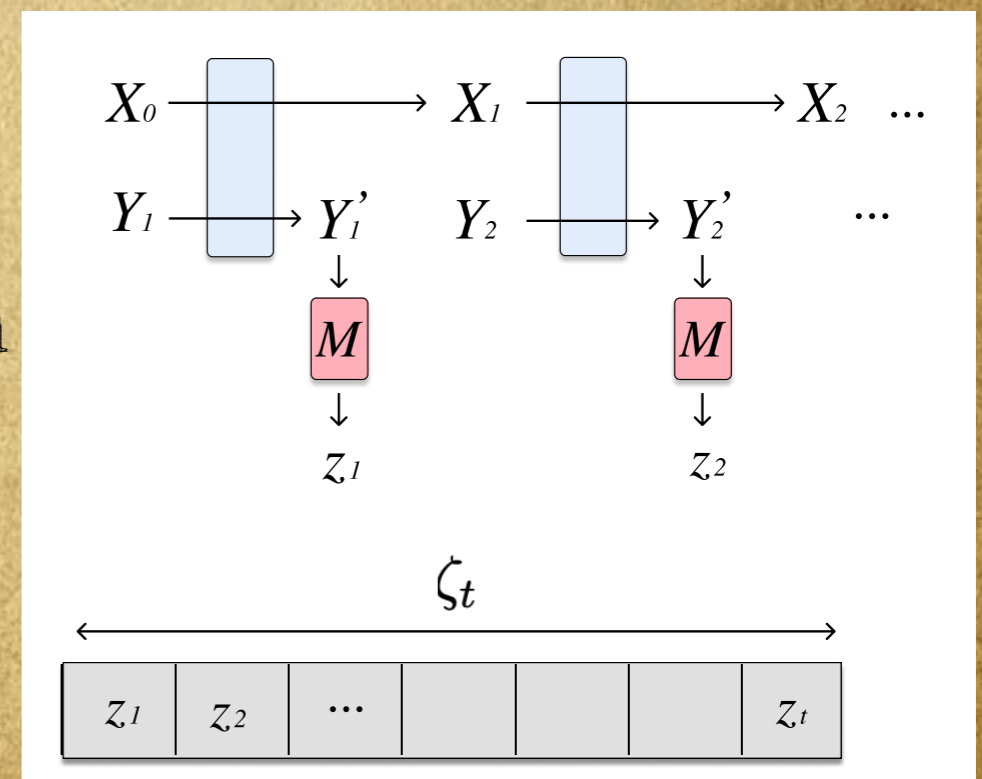
measurements sustain the state

$$\Delta\Sigma_t^u = S(X_t) - S(X_{t-1}) + \Delta\Phi_t^u$$

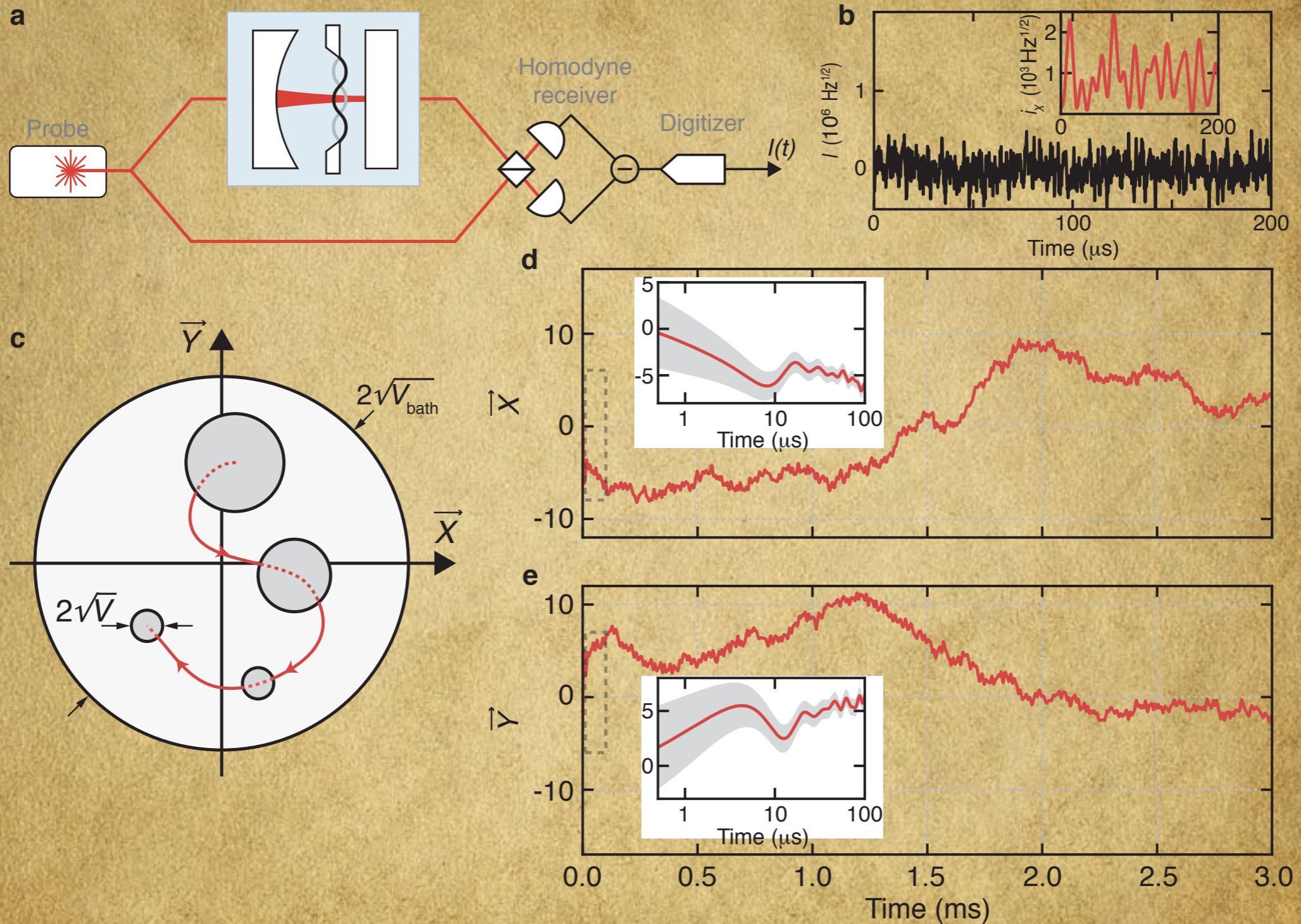
$$\Delta\Sigma_t^c = S(X_t|\zeta_t) - S(X_{t-1}|\zeta_{t-1}) + \Delta\Phi_t^c$$



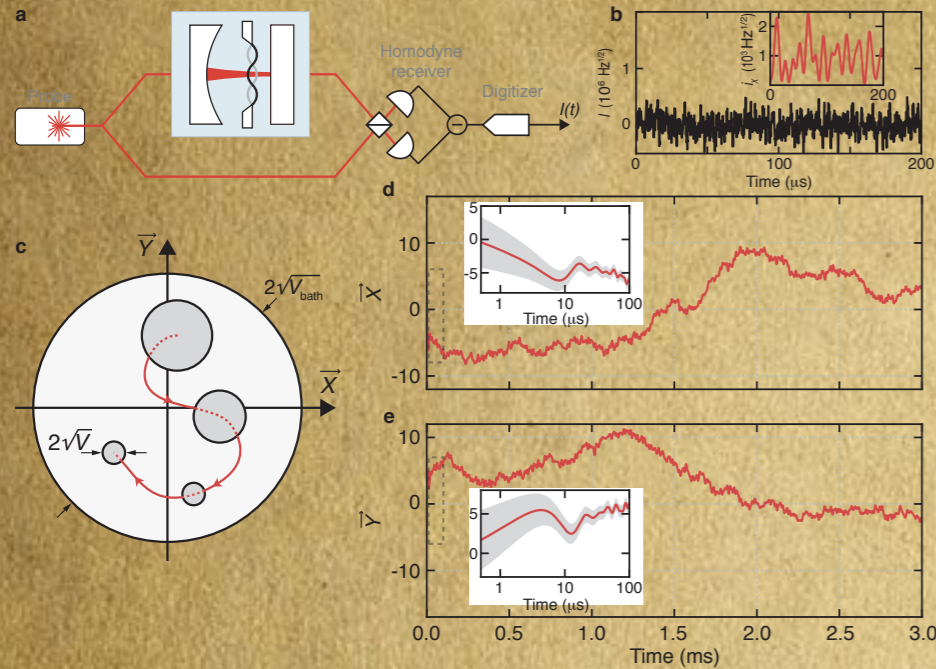
Interpretation through collisional model



Observing trajectories of mechanical systems



Observing trajectories of mechanical systems



$$d\mathbf{r}(t) = -\frac{\Gamma_m}{2} \mathbf{r} dt + \sqrt{4\eta_{\text{det}} \Gamma_{\text{qba}} V(t)} d\mathbf{W},$$

$$\dot{V}(t) = \Gamma_m (V_{\text{uc}} - V(t)) - 4\eta_{\text{det}} \Gamma_{\text{qba}} V(t)^2$$

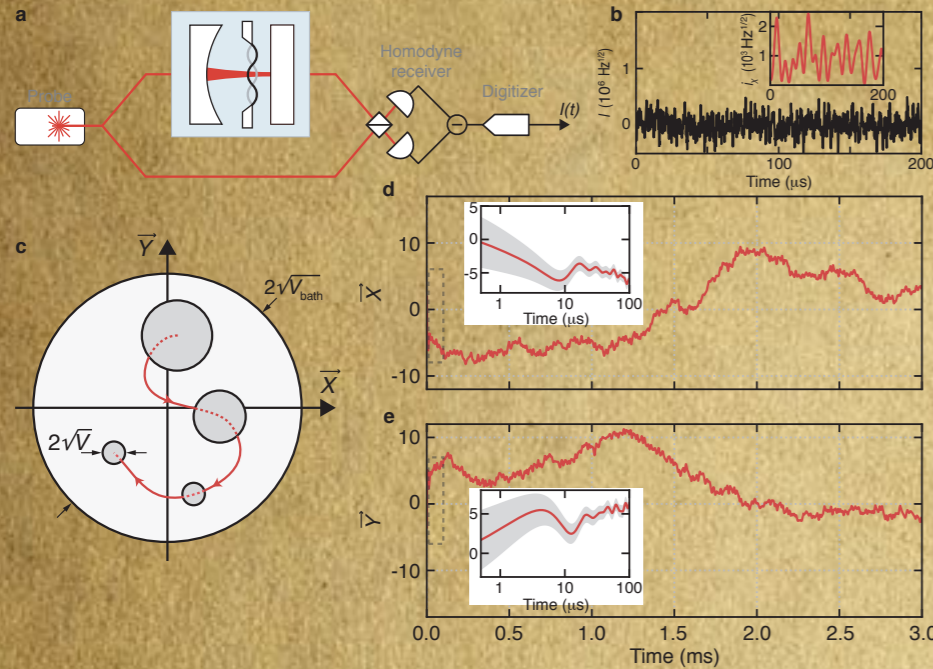
dynamics of the experiment

$$d\mathbf{r}(t) = A\mathbf{r}(t)dt + (V(t)C^T + \Gamma^T)d\mathbf{W},$$

$$\dot{V}(t) = AV(t) + V(t)A^T + D - \underbrace{(V(t)C^T + \Gamma^T)(CV(t) + \Gamma^T)}_{\chi(V(t))}$$

theoretical counterpart

Observing trajectories of mechanical systems



Initial state: equilibrium state
at environment temperature



Steady state of the unconditional
dynamics: NESS very close to equilibrium

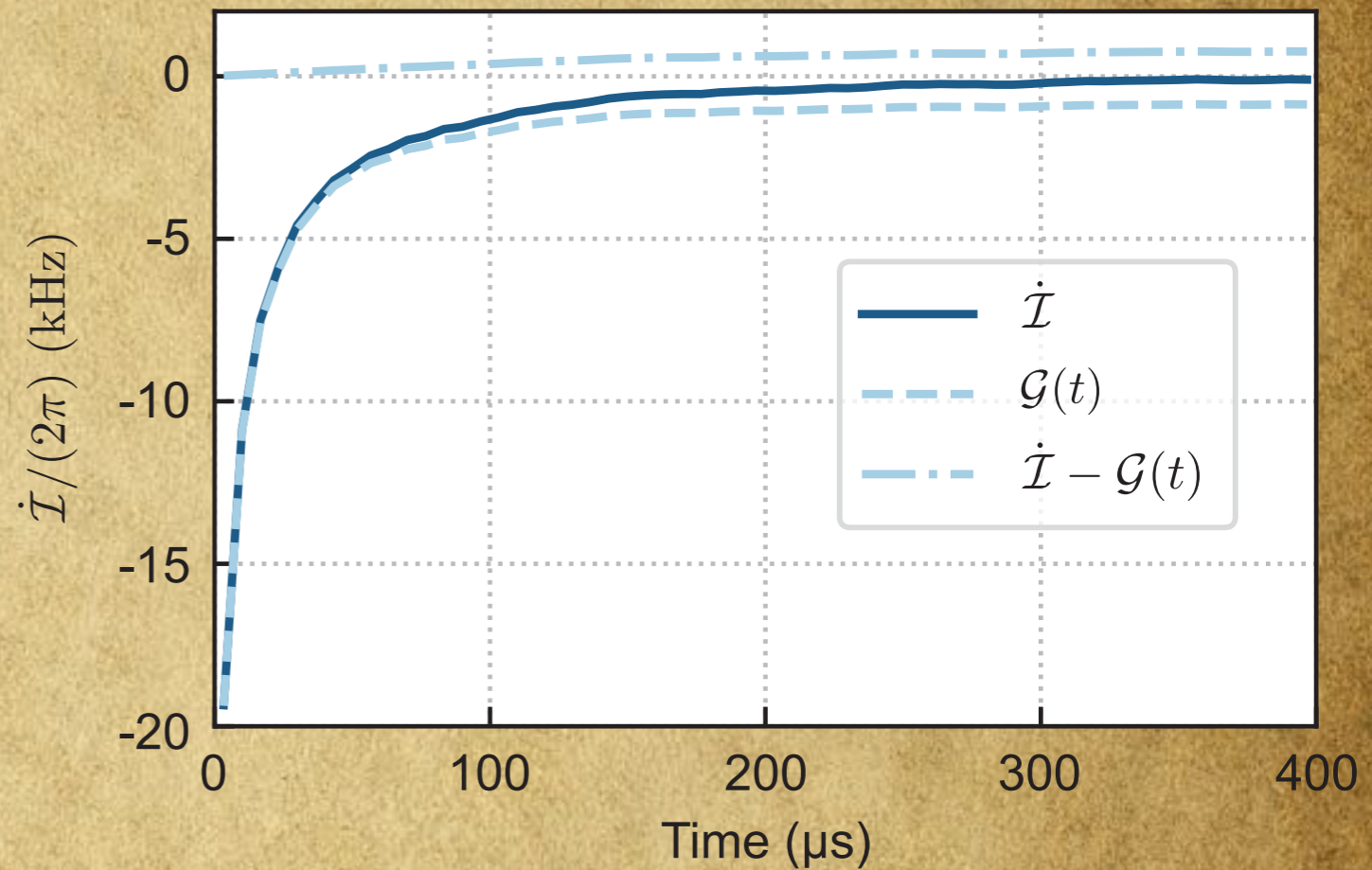
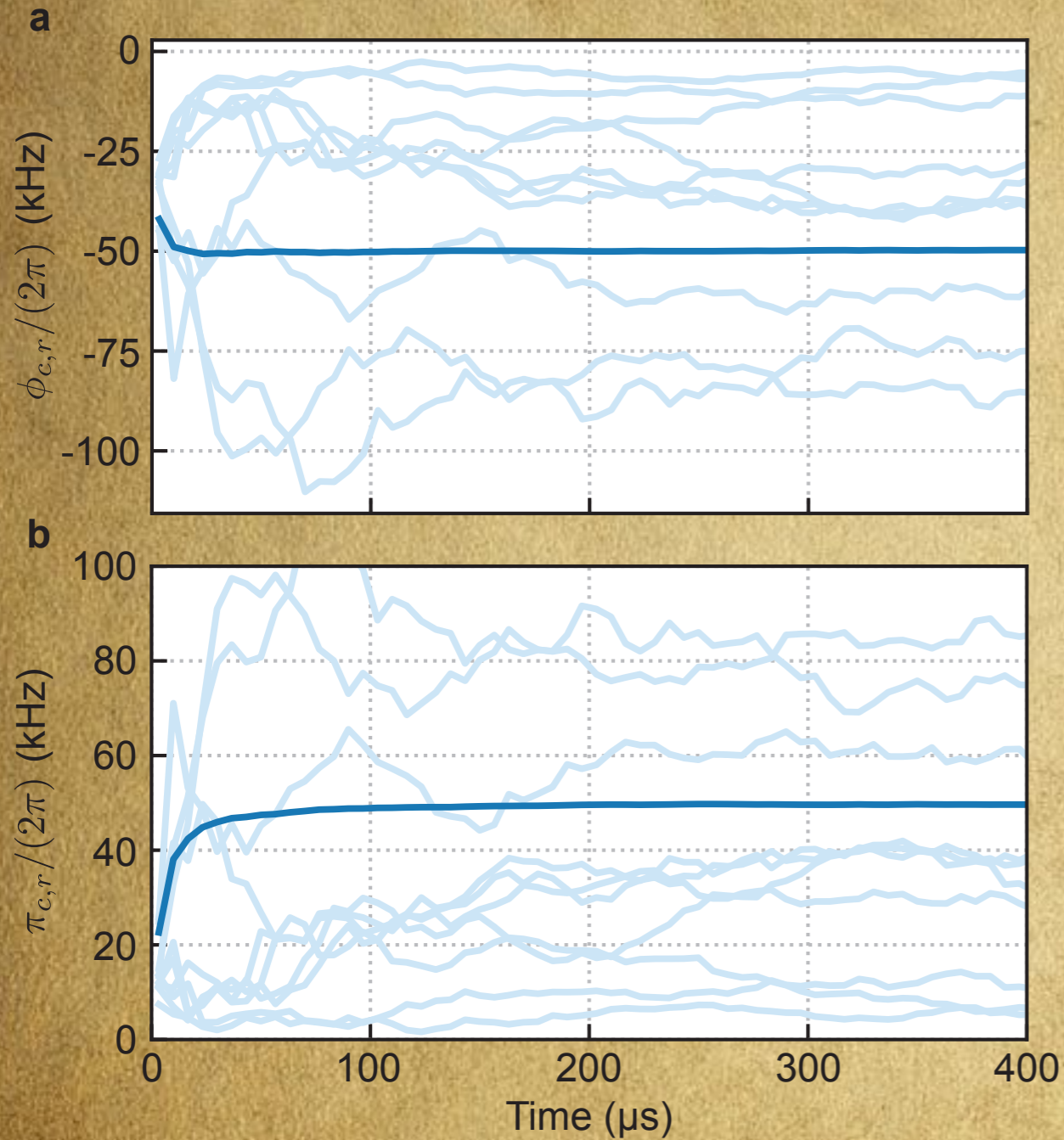
$$\Pi_{uc}(t) = \Gamma_m \left[V_{uc} / (n_{th} + 1/2) - 1 \right] + 4\Gamma_{qba} V_{uc}$$

$$\Pi_{uc}(t) = \text{const.} \text{ and } \Pi_c(t) = \dot{\mathcal{J}} + \text{const.}$$

$$\dot{\mathcal{J}} = \Gamma_m (V_{uc} / V(t) - 1) - 4\eta_{det} \Gamma_{qba} V(t)$$



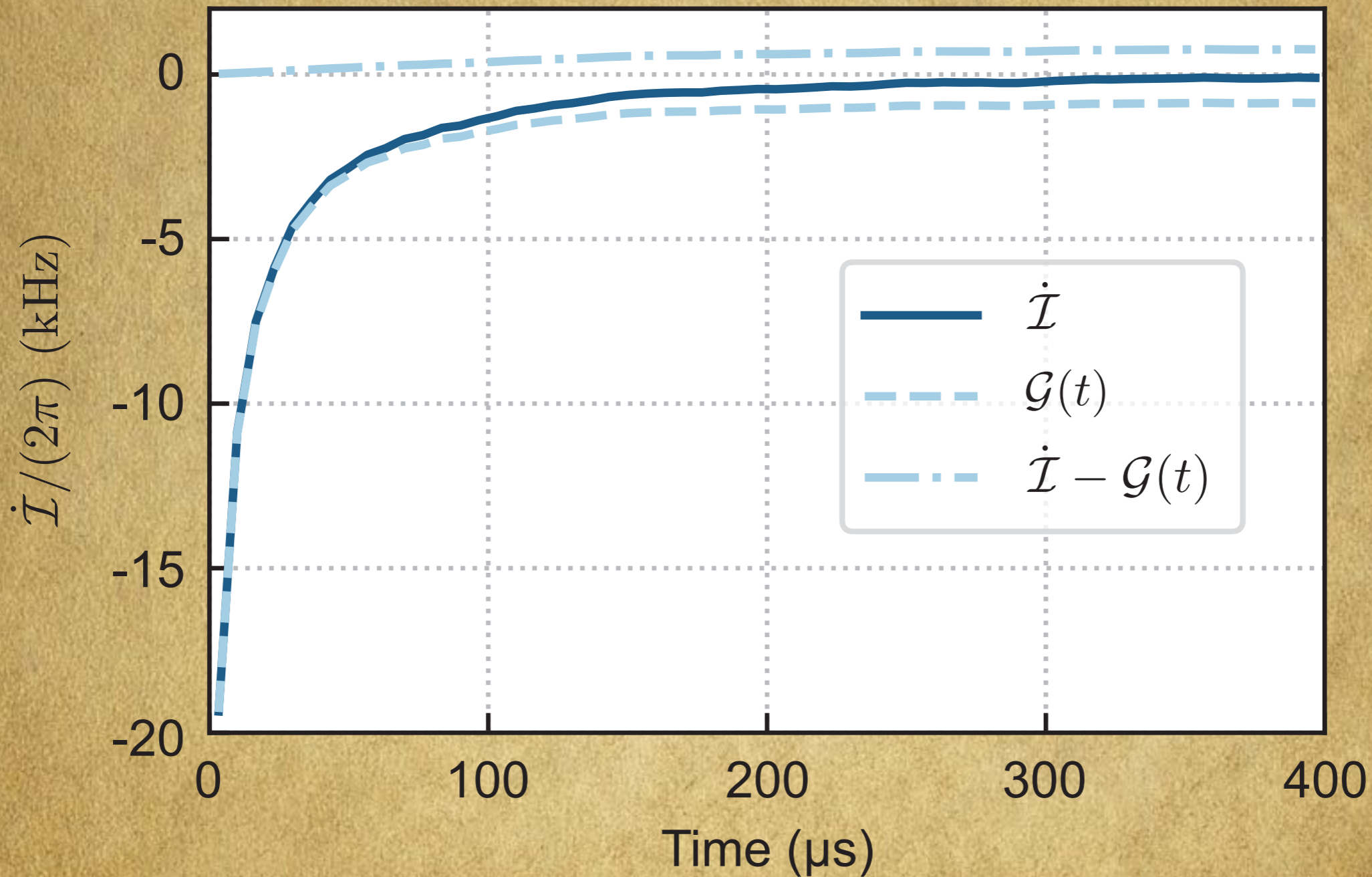
Observing entropy production rates of a measured system



M. Rossi, L. Mancino, G. T. Landi, M. Paternostro, A. Schliesser,
and A. Belenchia, Phys. Rev. Lett. 125, 080601 (2020)



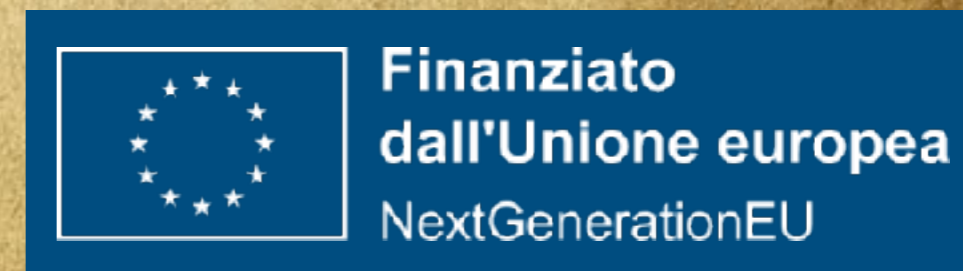
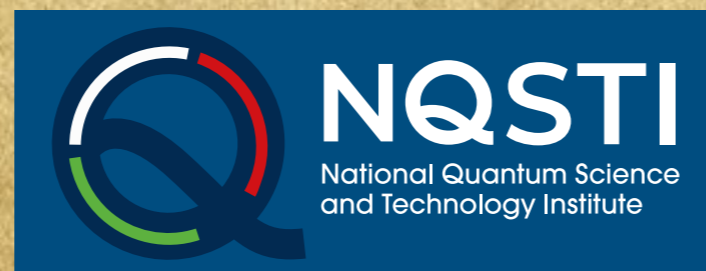
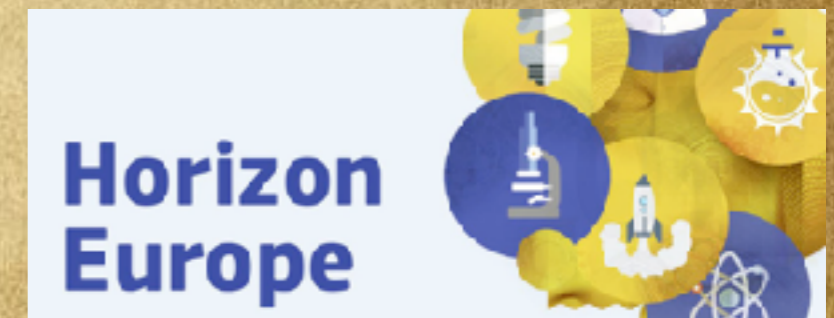
Observing entropy production rates of a measured system



M. Rossi, L. Mancino, G. T. Landi, M. Paternostro, A. Schliesser, and A. Belenchia, Phys. Rev. Lett. 125, 080601 (2020)



Bread on tables..





The crews



in (very) warm places..

..& (slightly) colder ones.



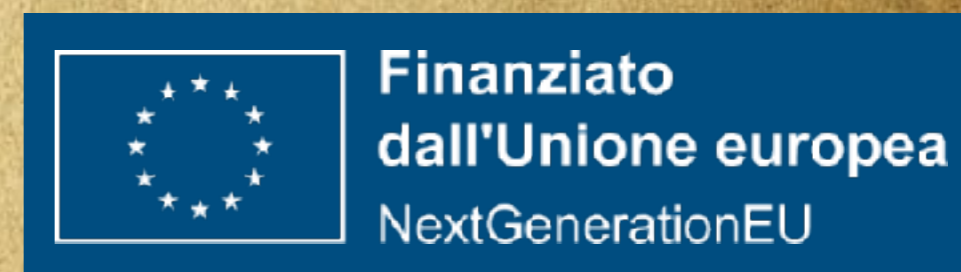
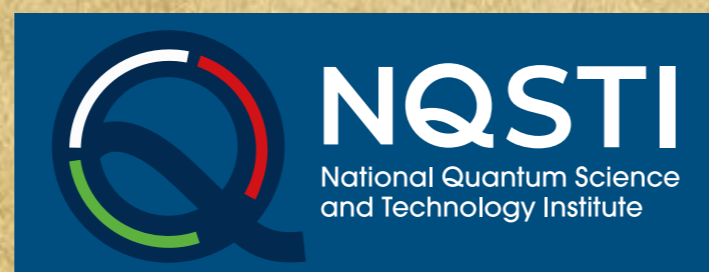


Shameless advertisement

At least 1 PhD position



Several Postdoc positions soon to be open to work on quantum thermodynamics for quantum computation



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THANK YOU