

INFORMATIONAL STEADY-STATES AND
CONDITIONAL ENTROPY PRODUCTION
IN CONTINUOUSLY MONITORED SYSTEMS

Mauro Paternostro

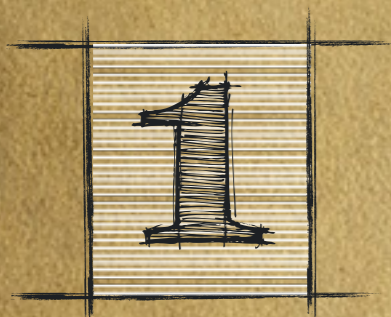
School of Mathematics & Physics, Queen's University Belfast



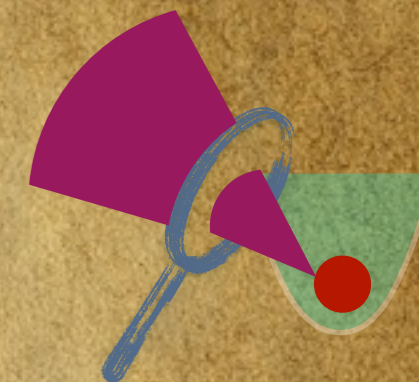
52th Symposium on Mathematical Physics

"Channels, Maps and All That"

16 June 2021

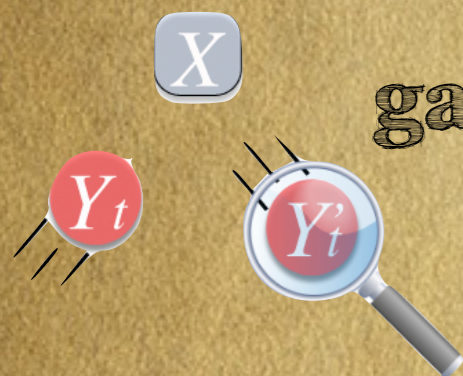


Formalism for entropy production
in continuously measured
quantum systems



Informational steady states:

gaining & losing through measurement



Observing irreversible
entropy in measured mesoscopic
quantum settings





DUE CREDIT

Alessio Belenchia (Tubingen)



Luca Mancino (Belfast)



Massimiliano Rossi (ETH Zürich)

Gabriel T. Landi (Sao Paulo)

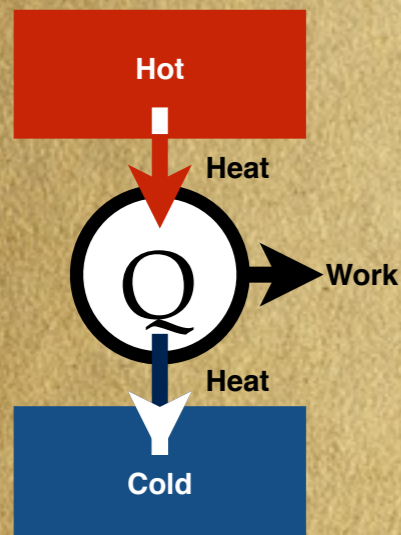
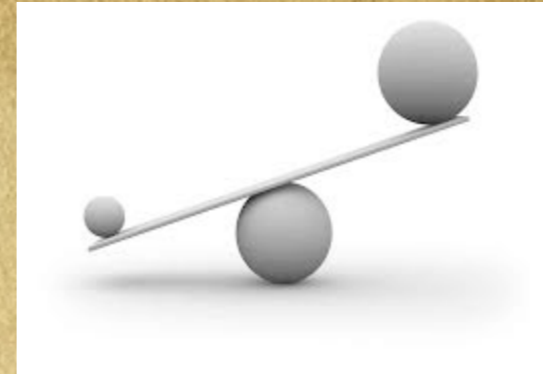


Albert Schliesser (Copenhagen)



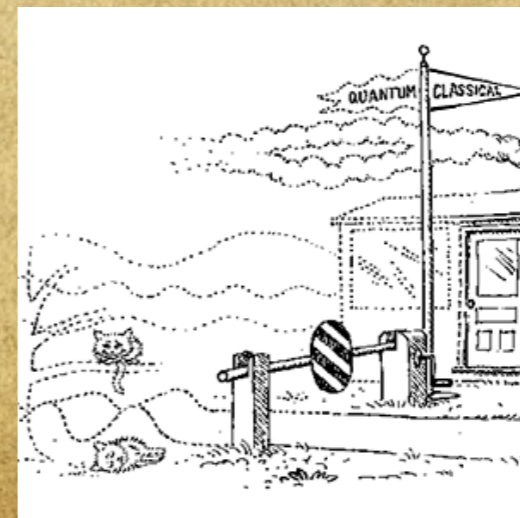
Why entropy production?

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

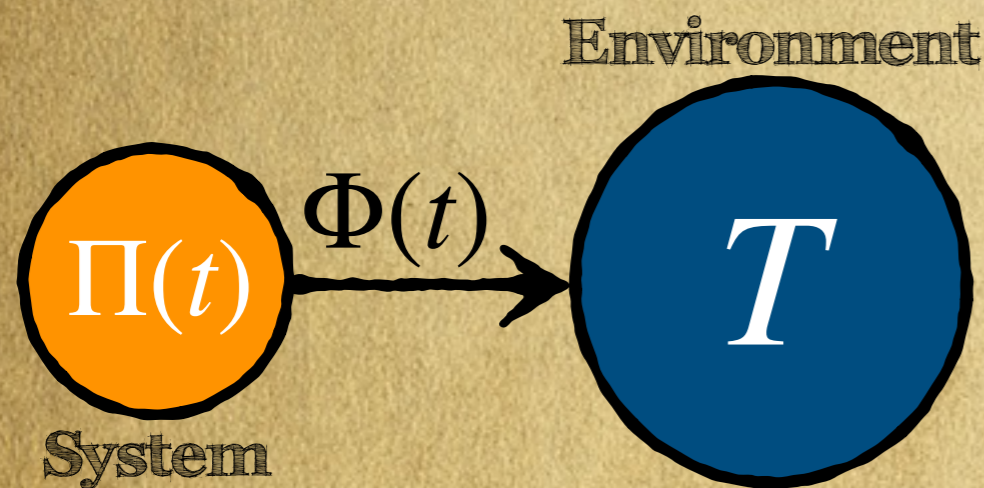


Entropy production

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

Clausius: "Uncompensated transformation"

Entropy production



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

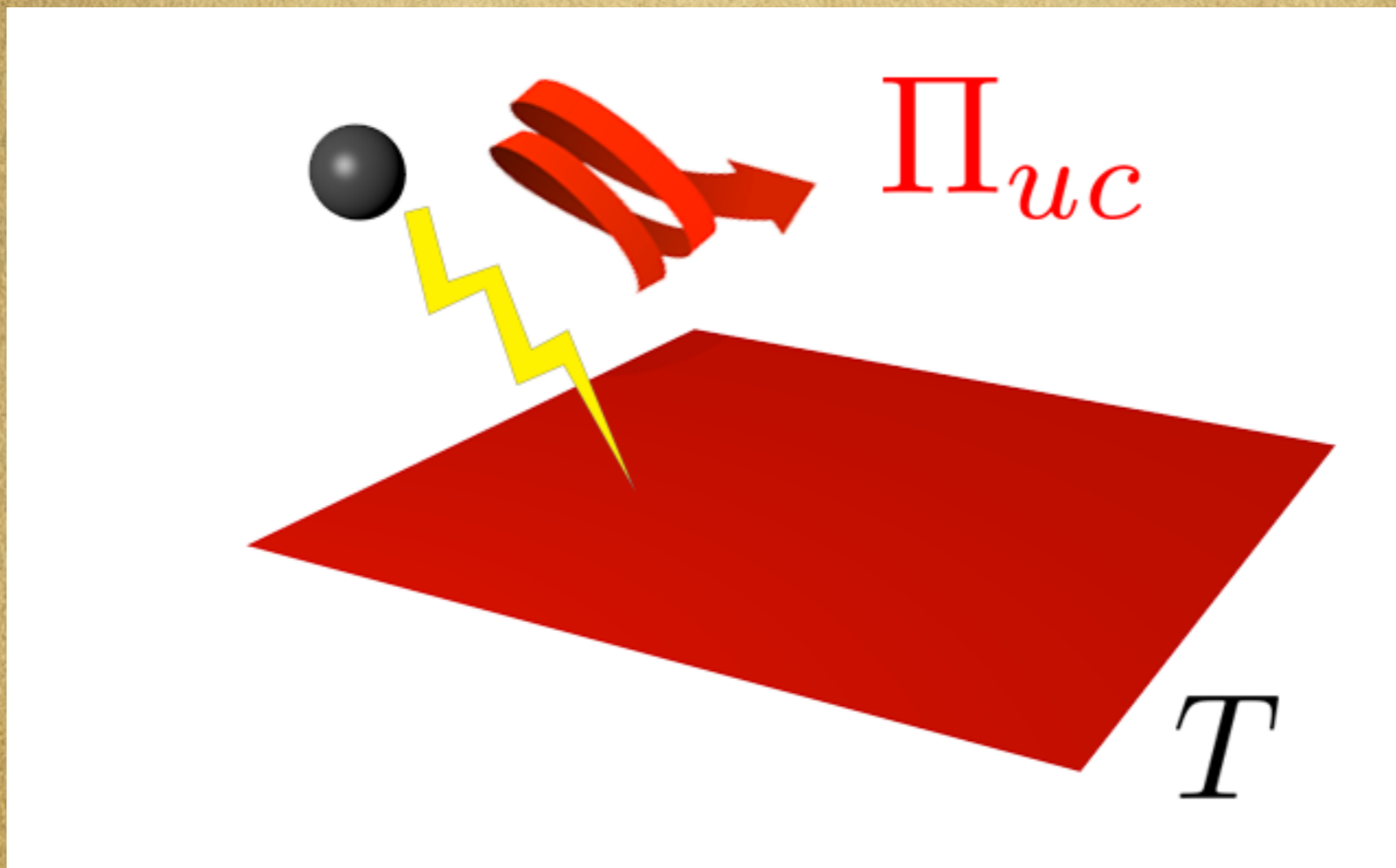
Entropy production rate

Entropy flux rate

Which is the role of quantum fluctuations on entropy production?

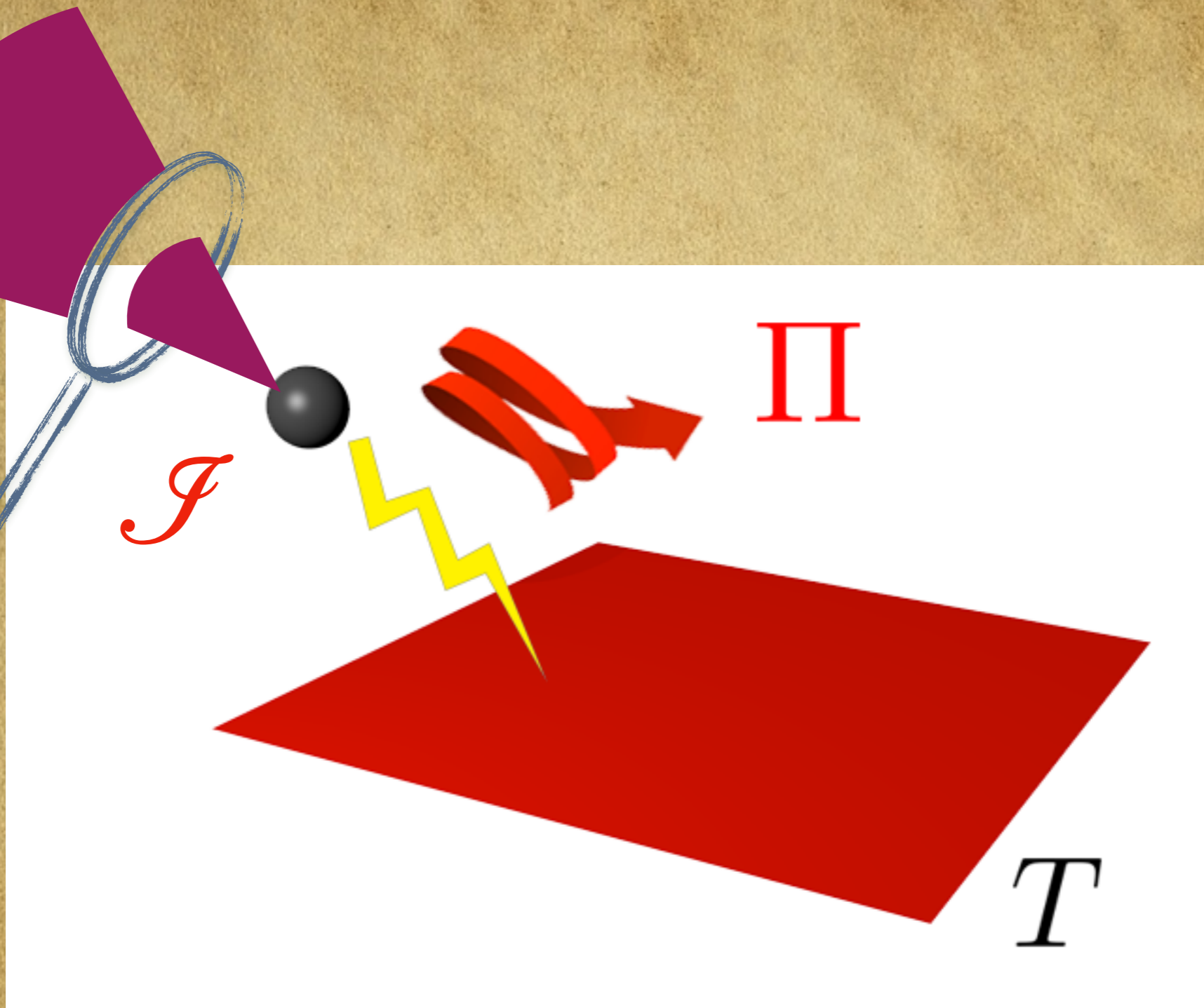
What happens if you plug in the effects of measuring?

Don't look yet!!!



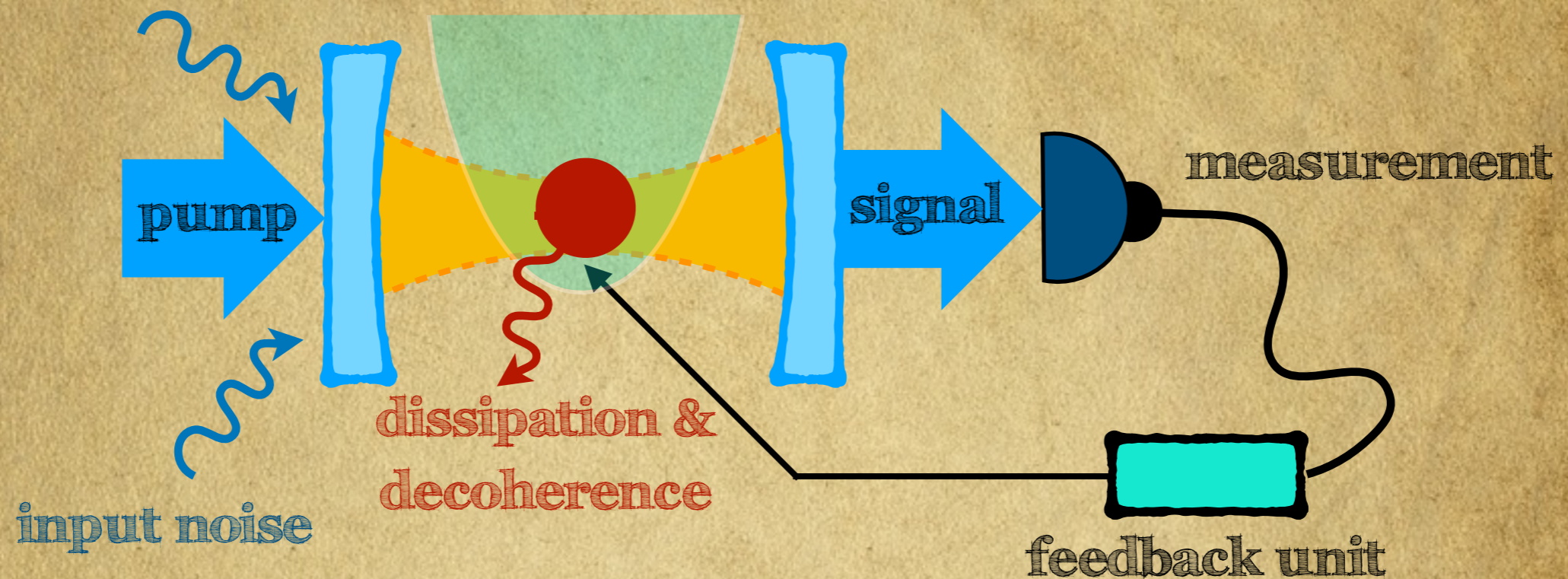
A Belenchia, L Mancino, G T Landi, and M Paternostro,
Nature Quantum Information 6, 97 (2019)

..now open your eyes..



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

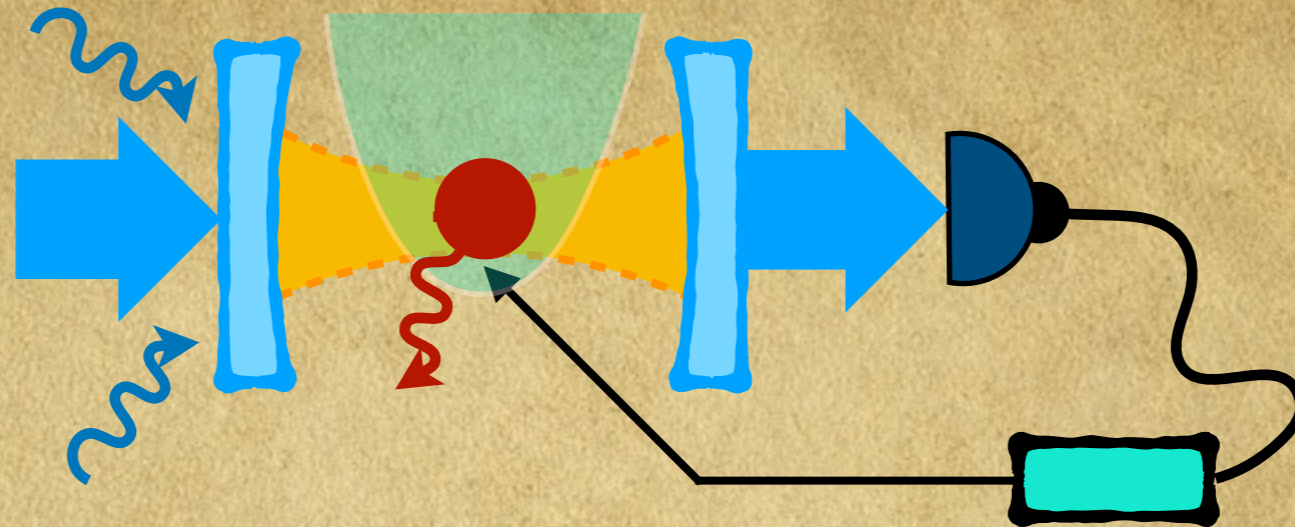
Let's fix the ideas



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

A Belenchia, M Paternostro, and G T Landi, arXiv:2105.12518 (2021)

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



Stochastic master equation

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

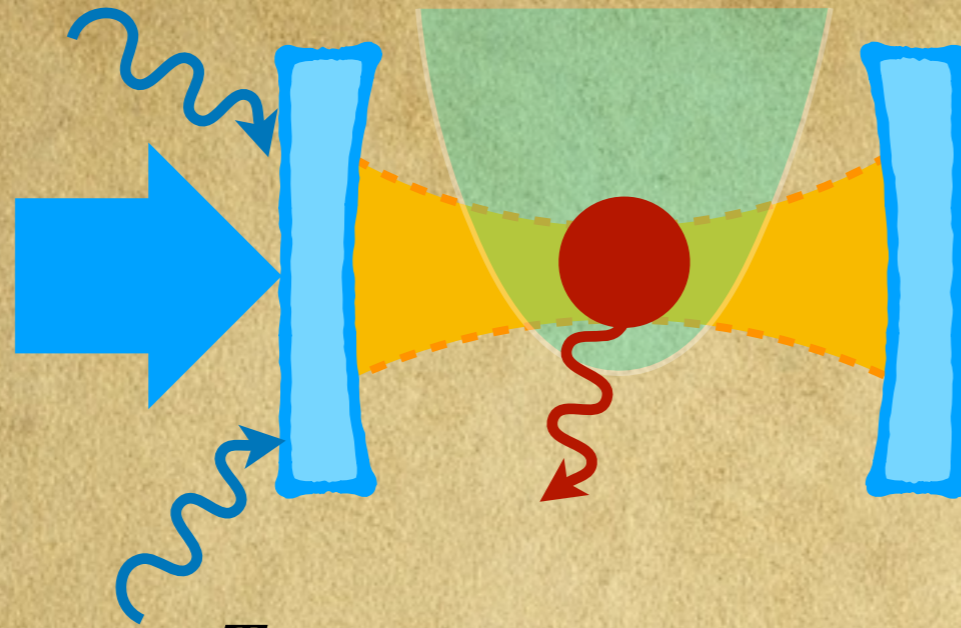
Un-monitored dynamics

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, arXiv:2105.12518 (2021)

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



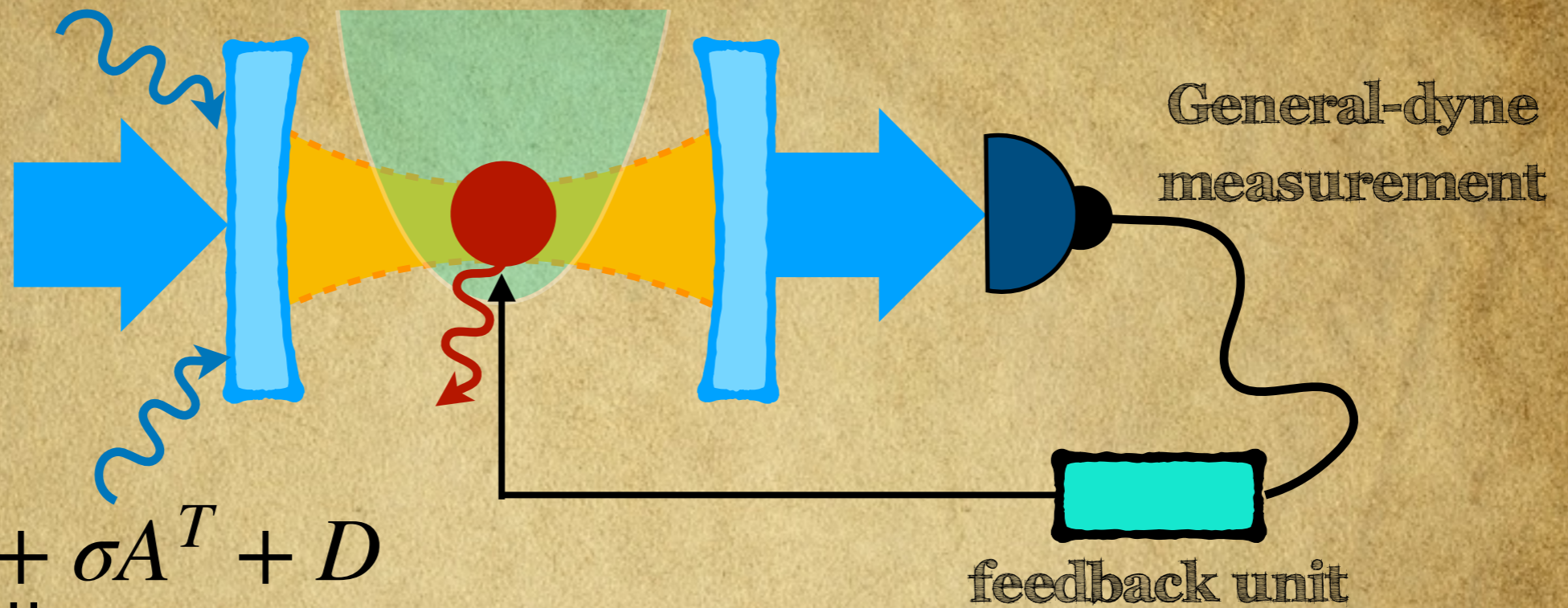
$$\dot{\sigma} = A\sigma + \sigma A^T + D$$

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

A Belenchia, M Paternostro, and G T Landi, arXiv:2105.12518 (2021)

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

Conditioned Gaussian dynamics



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

\Downarrow

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

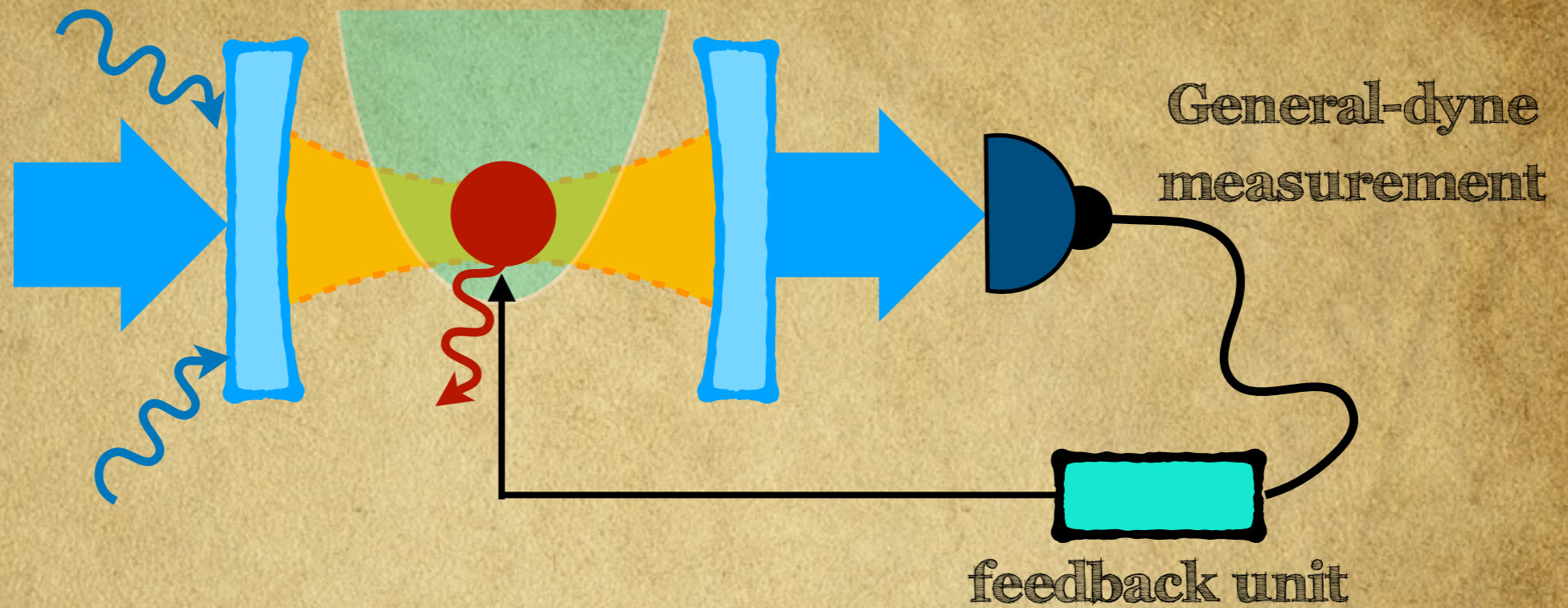
contains terms depending on the measurement

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

\Downarrow

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

Conditioned Gaussian dynamics



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

deterministic (only depends on CM) stochastic (depend also on 1st moments)

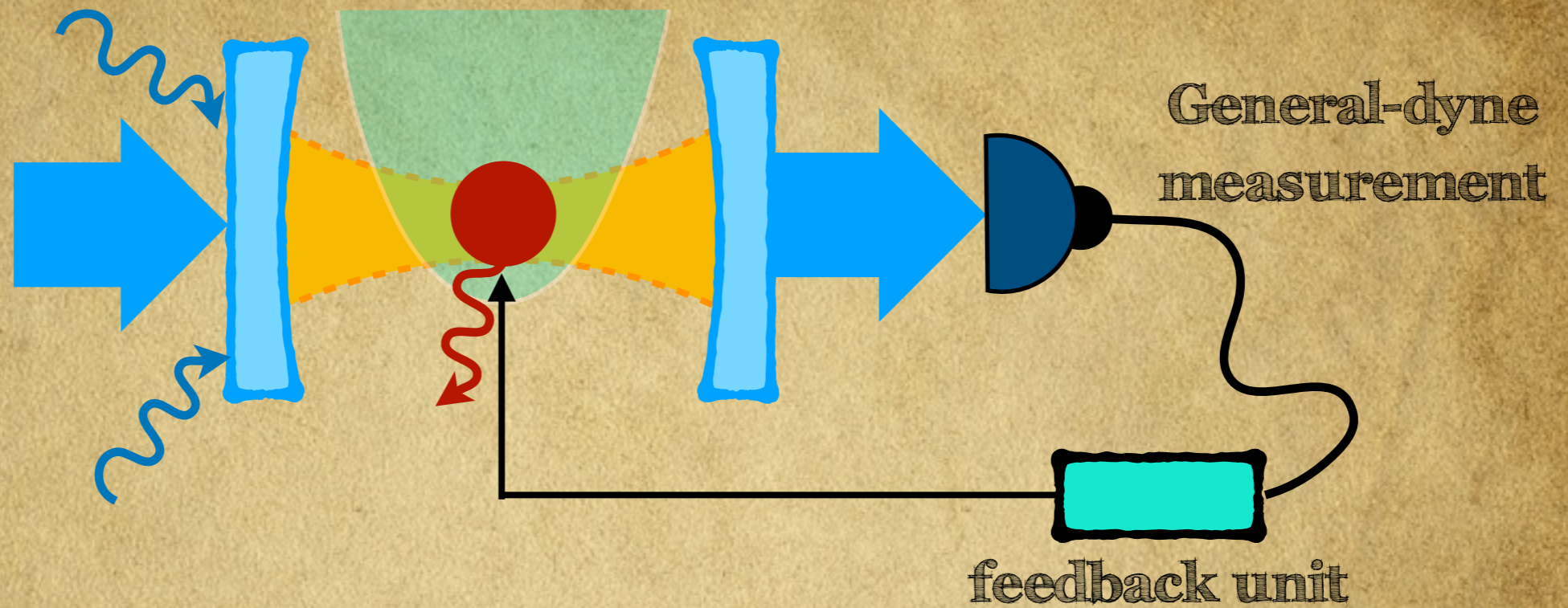
$$\phi = \mathbb{E} [d\Phi_{\bar{x}}/dt]$$

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

A Belenchia, M Paternostro, and G T Landi, arXiv:2105.12518 (2021)

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

Conditioned Gaussian dynamics



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

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

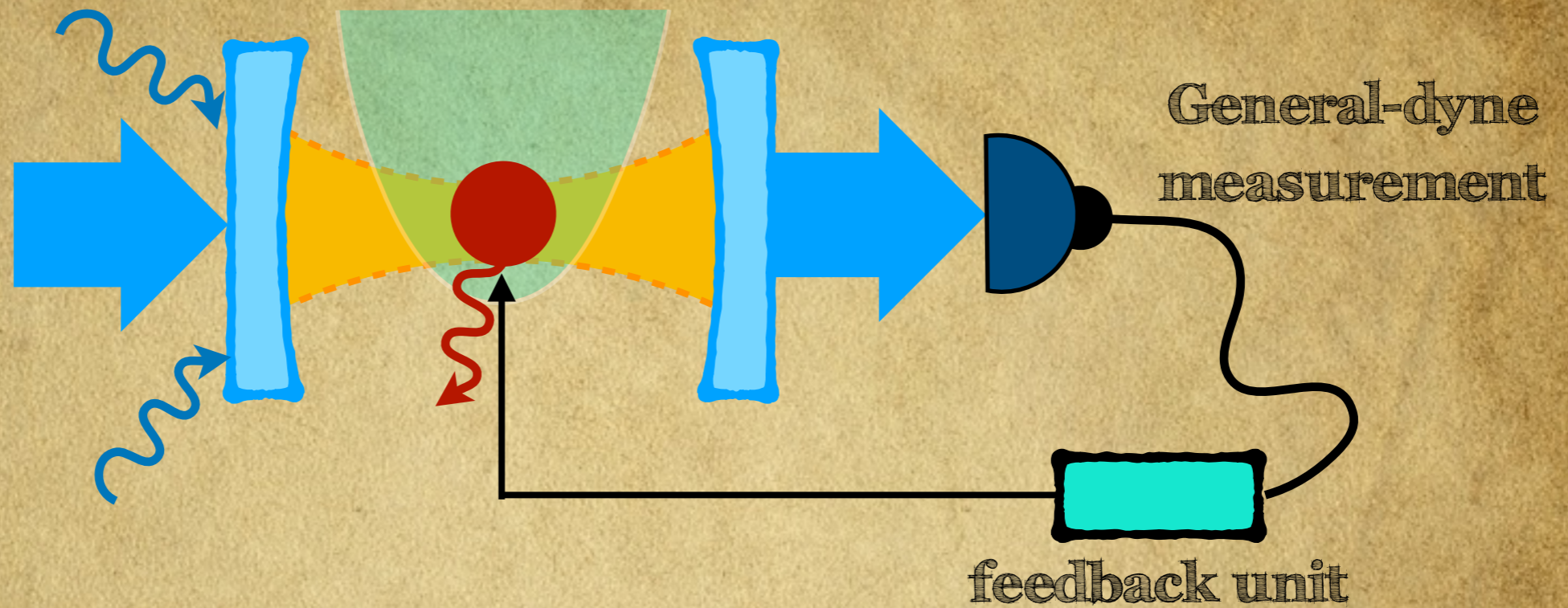
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A Belenchia, et al., Nature Quantum Information 6, 97 (2019)

Conditioned Gaussian dynamics



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

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

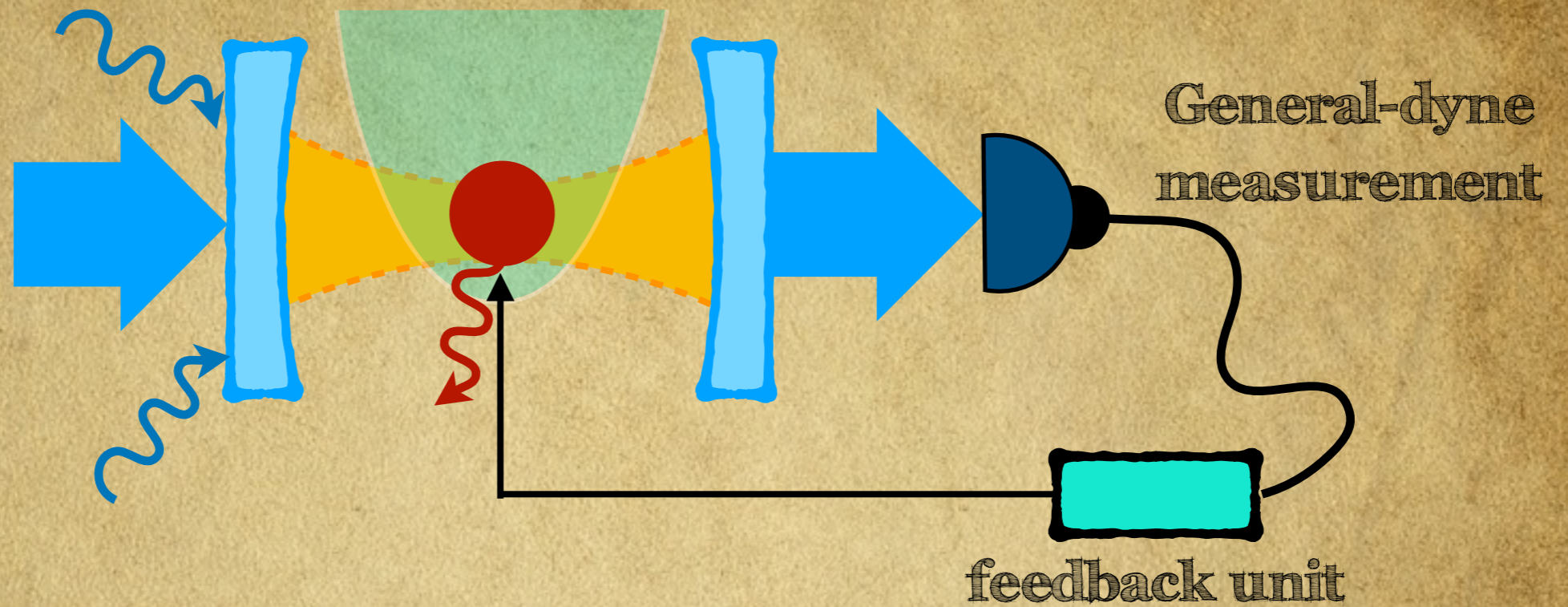
$$\phi = \mathbb{E} \left[d\Phi_{\bar{x}} / dt \right]$$

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

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Conditioned Gaussian dynamics



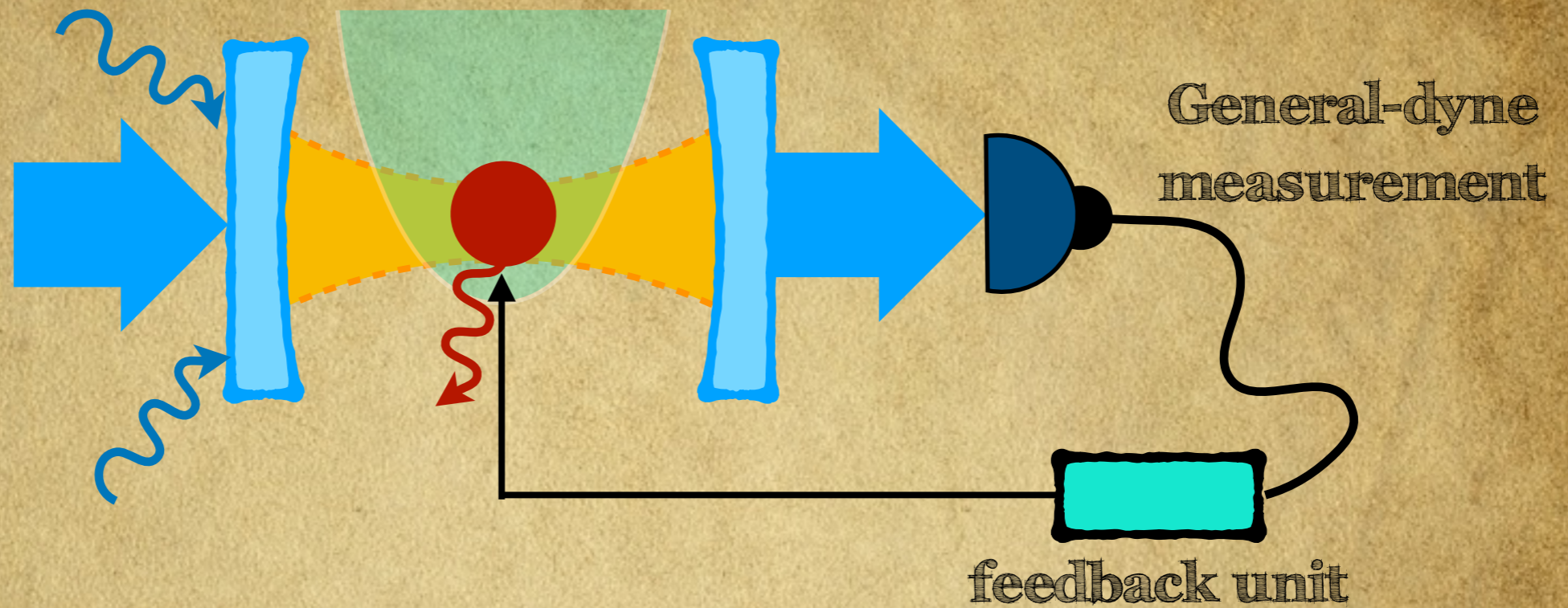
$\Pi_{uc}(t) \geq 0$ second law for un-conditioned dynamics

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

A Belenchia, M Paternostro, and G T Landi, arXiv:2105.12518 (2021)

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

Conditioned Gaussian dynamics



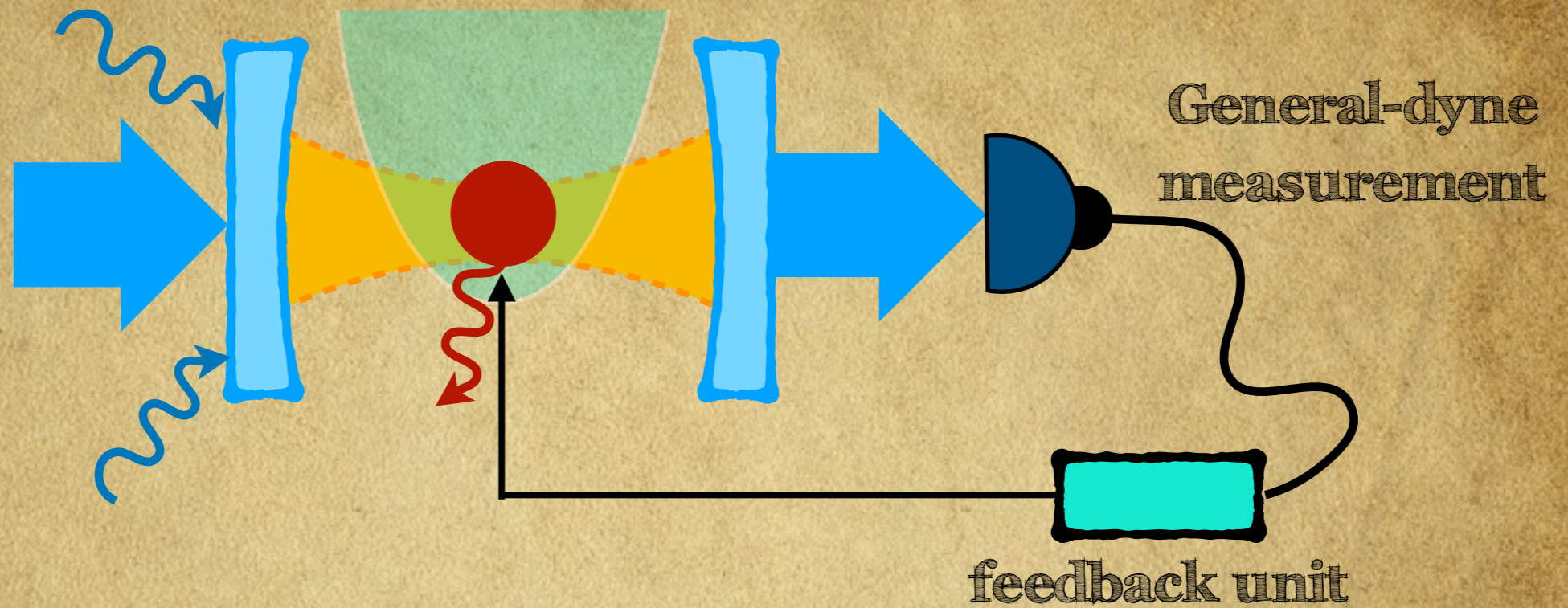
$\Pi_c(t) \geq \dot{\mathcal{J}}$ second law for conditioned dynamics

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

A Belenchia, M Paternostro, and G T Landi, arXiv:2105.12518 (2021)

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

Conditioned Gaussian dynamics



$\Sigma_c \geq \mathcal{F}$ integral second law for conditioned dynamics

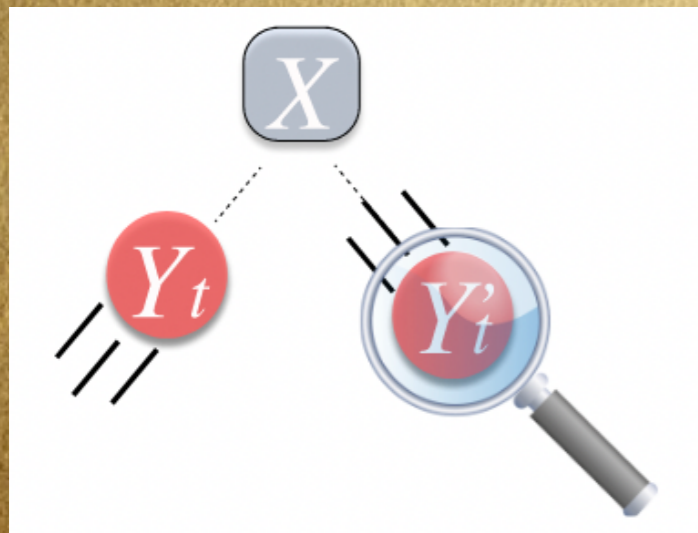
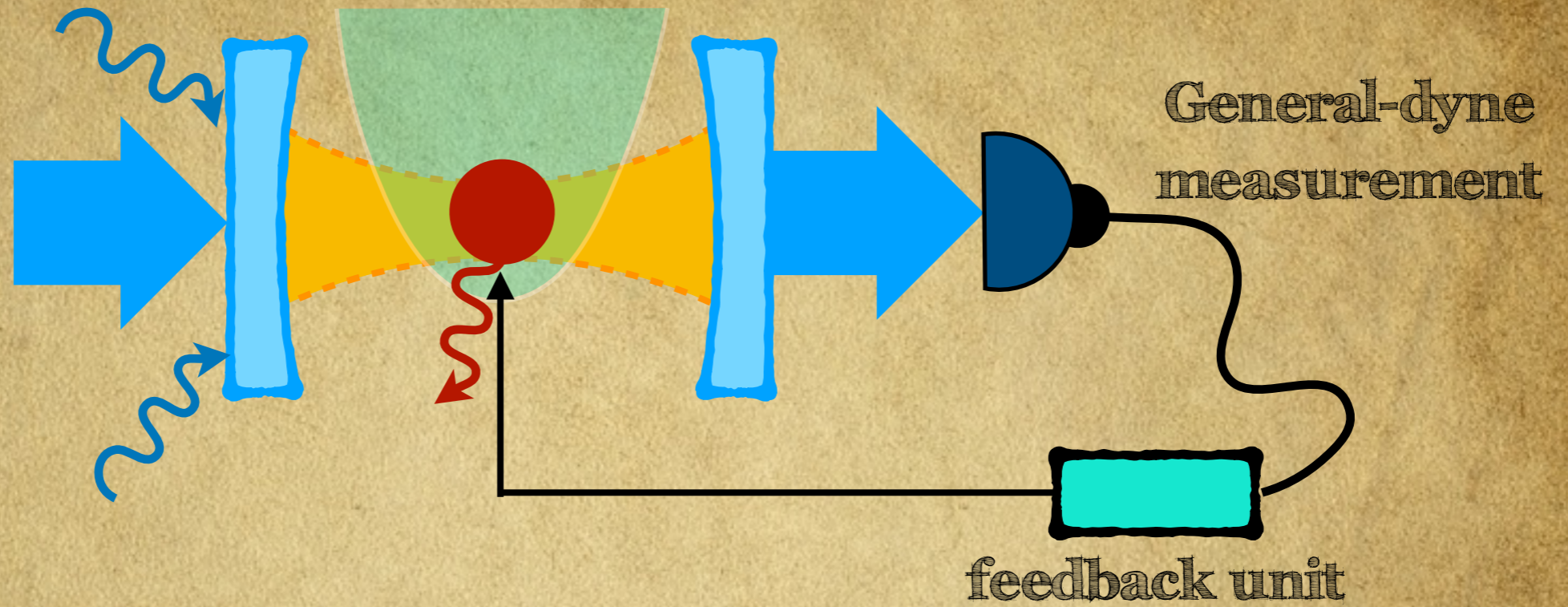
$$\Pi_c(t) = \Pi_{uc}(t) + \dot{\mathcal{F}}$$

Observable (SEE PART 3)

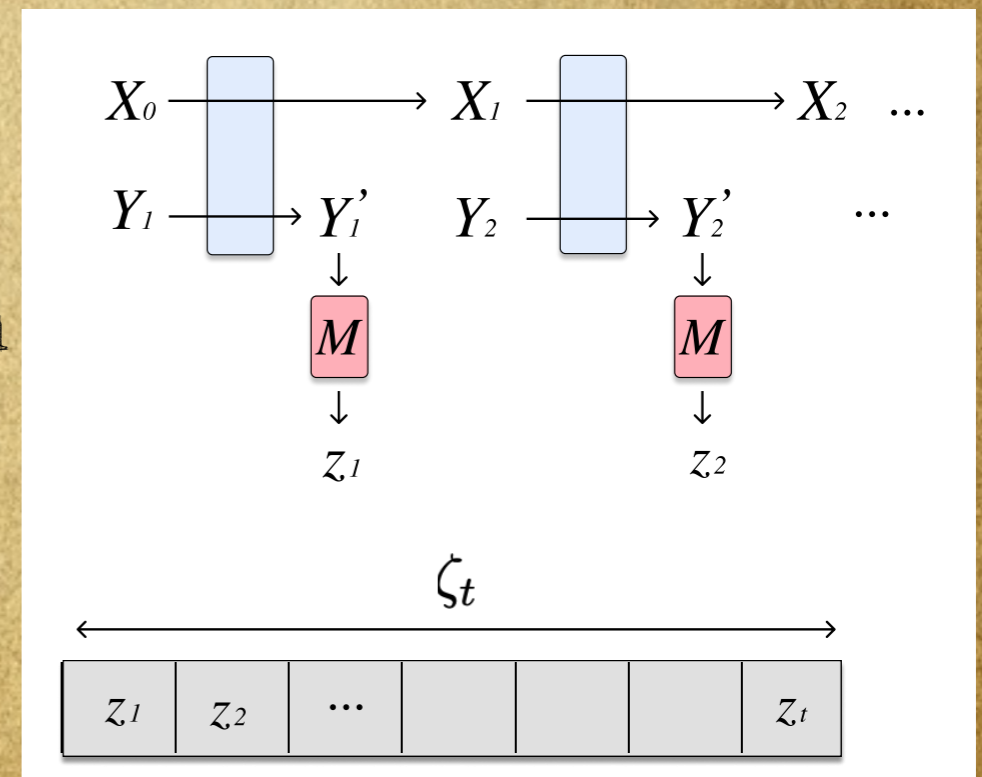
A Belenchia, M Paternostro, and G T Landi, arXiv:2105.12518 (2021)

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

Generalising it



Interpretation through collisional model



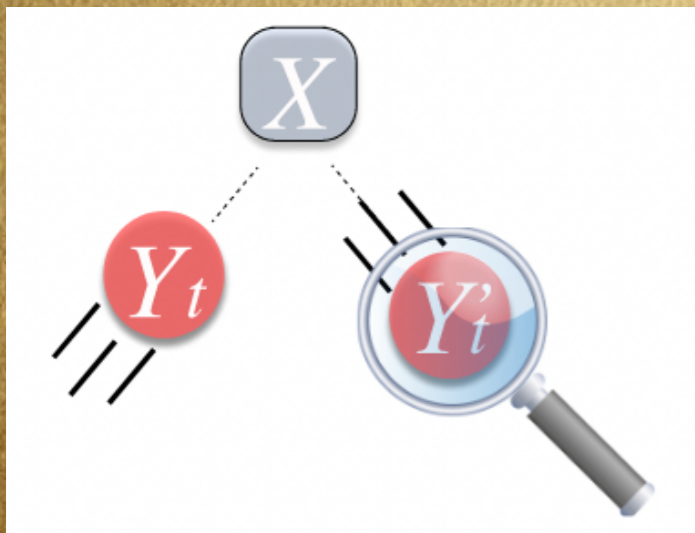
Can we generalise?

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

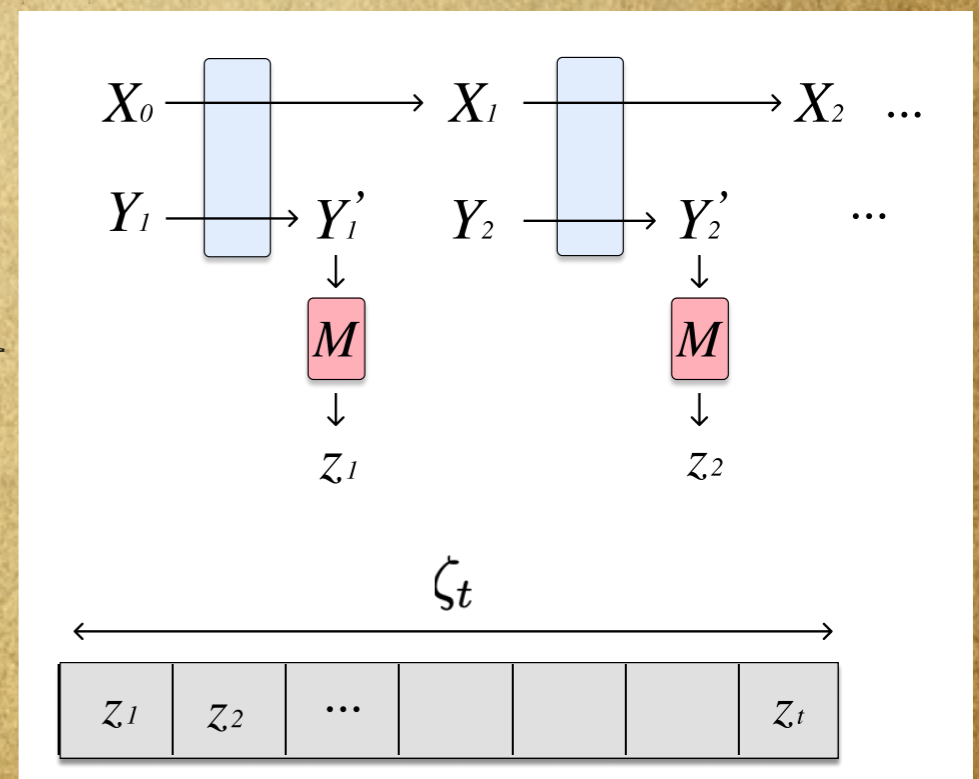
$$\rho_{XY_1 \dots Y_t} = \left(\prod_{k=1}^t U_k \right) \left(\rho_{X_0} \otimes_{j=1}^t \rho_{Y_j} \right) \left(\prod_{k=1}^t U_k \right)^\dagger$$

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

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

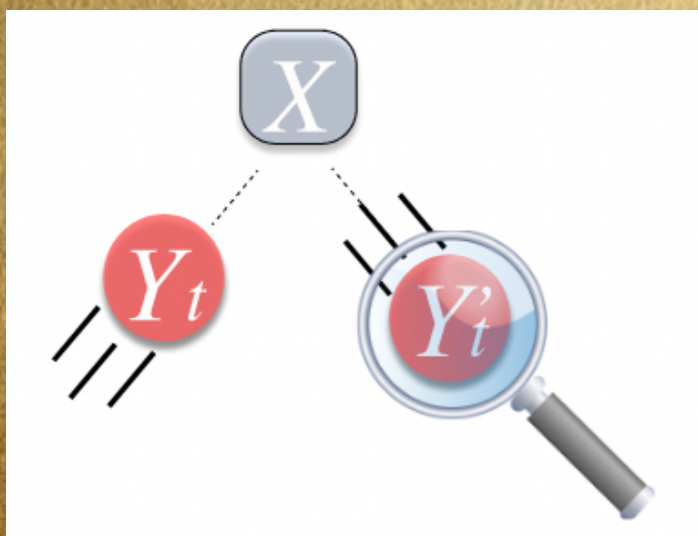


$$I(X_t : \zeta_t) - I(X_t : \zeta_{t-1})$$

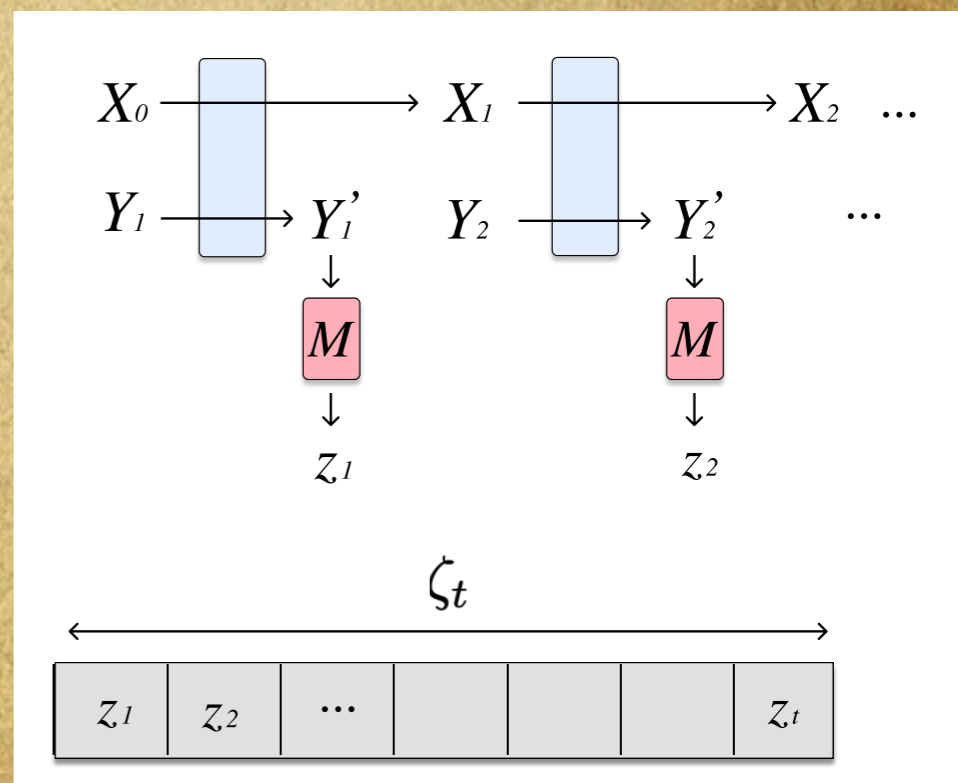
differential = $S(X_t | \zeta_{t-1}) - S(X_t | \zeta_t)$
information gain

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

information 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

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

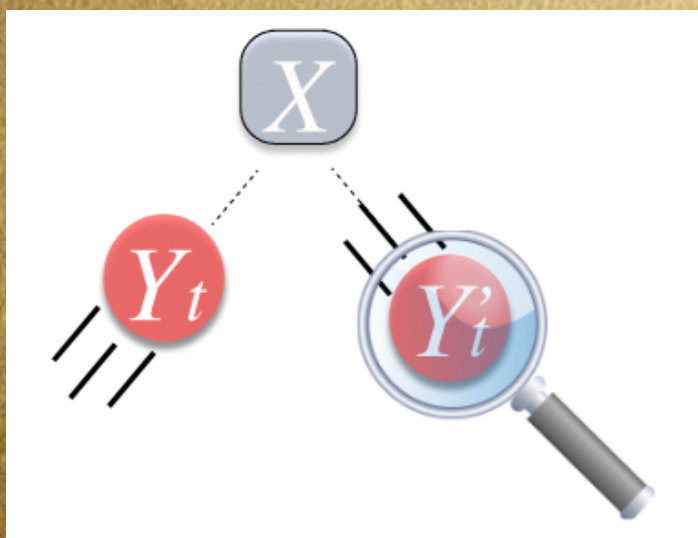
Informational steady state:

balance between gain & loss

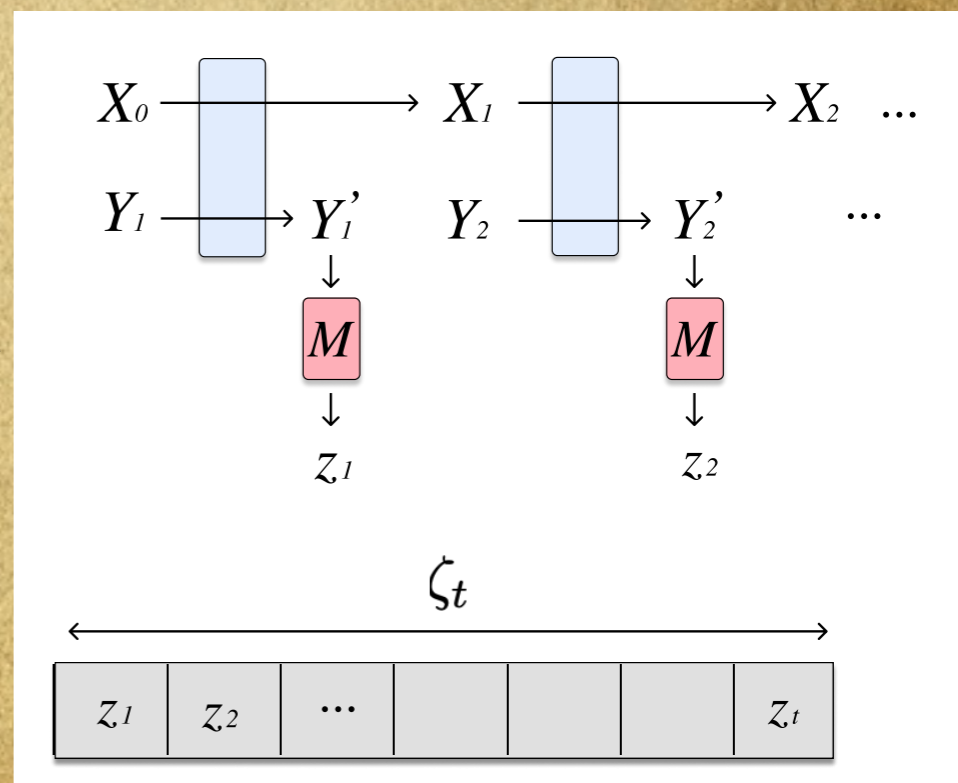
measurements sustain the state

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

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



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

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

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

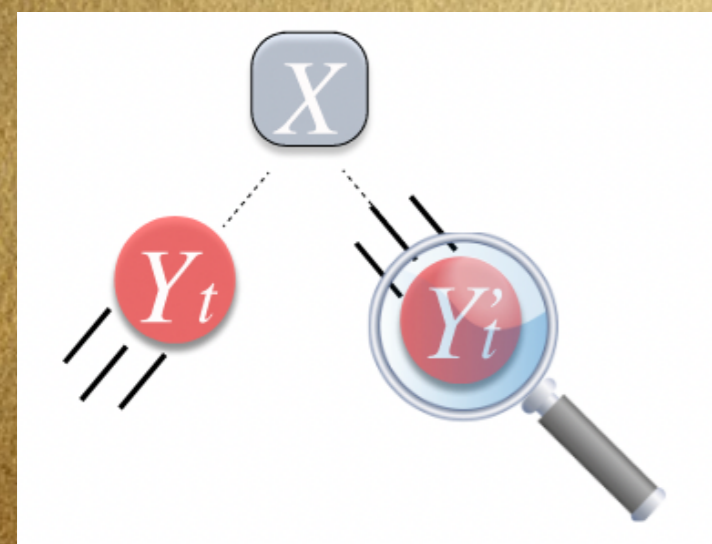
Informational steady state:

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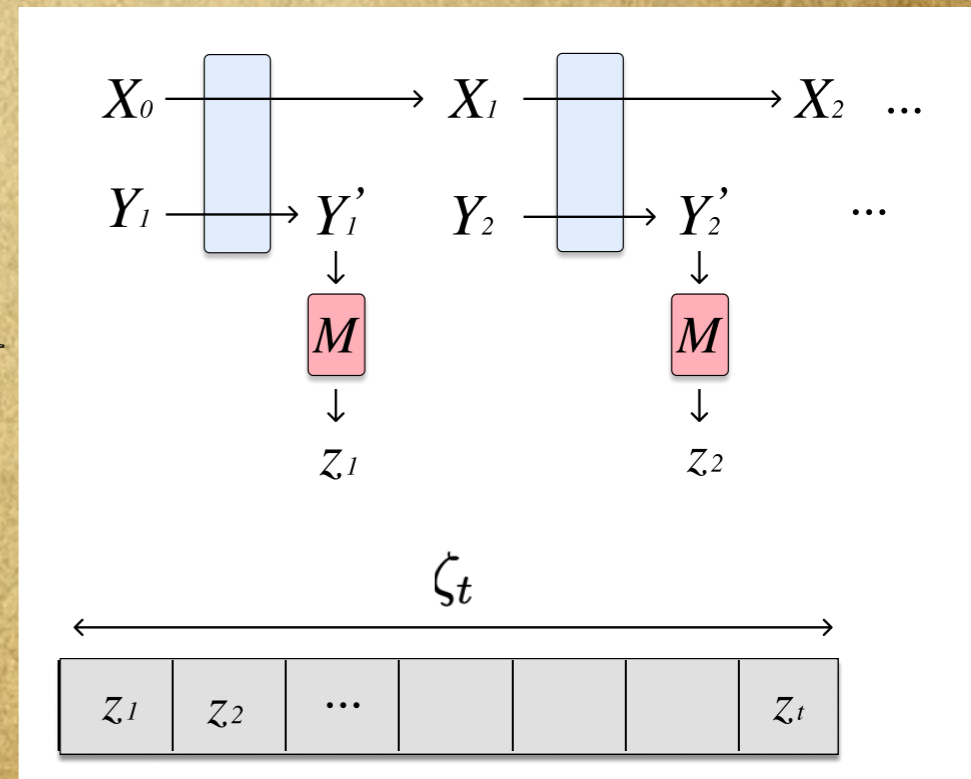
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$$\Delta\Sigma_t^u = S(X_t) - S(X_{t-1}) + \Delta\Phi_t^u$$

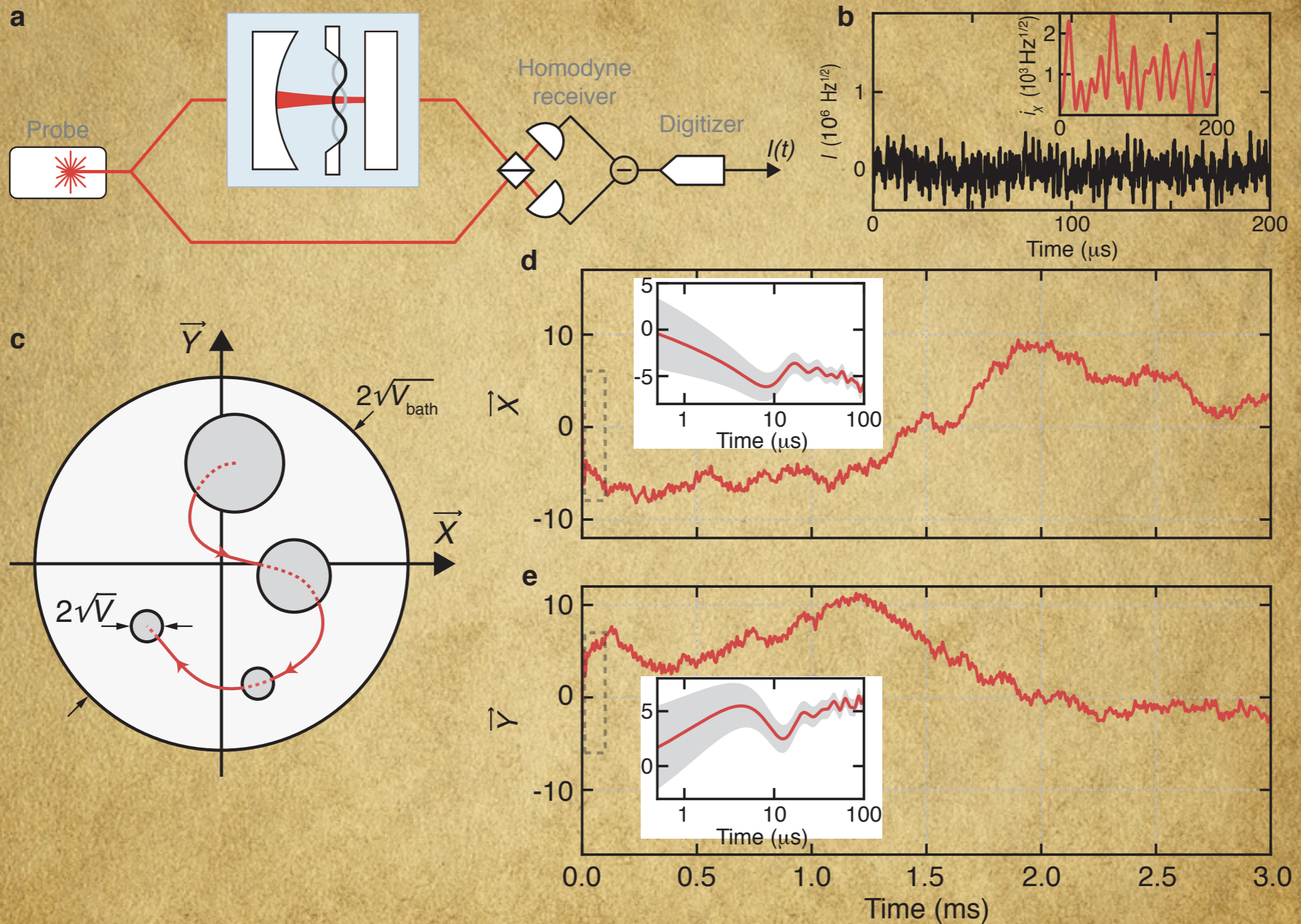
$$\Delta\Sigma_t^c = S(X_t|\zeta_t) - S(X_{t-1}|\zeta_{t-1}) + \Delta\Phi_t^c$$



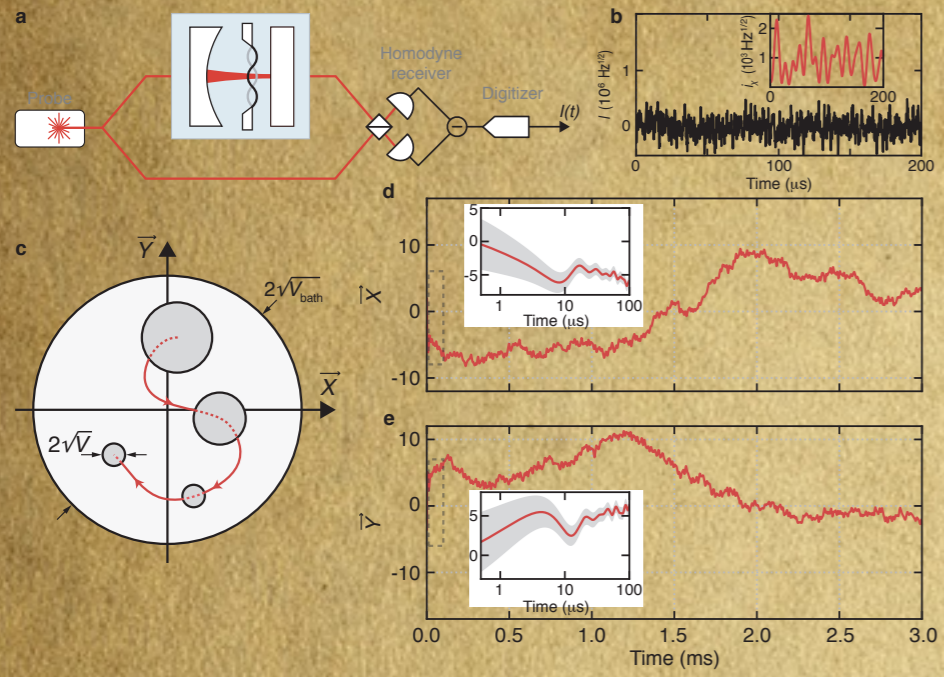
Interpretation through collisional model



Observing trajectories of mechanical systems



Observing trajectories of mechanical systems



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

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

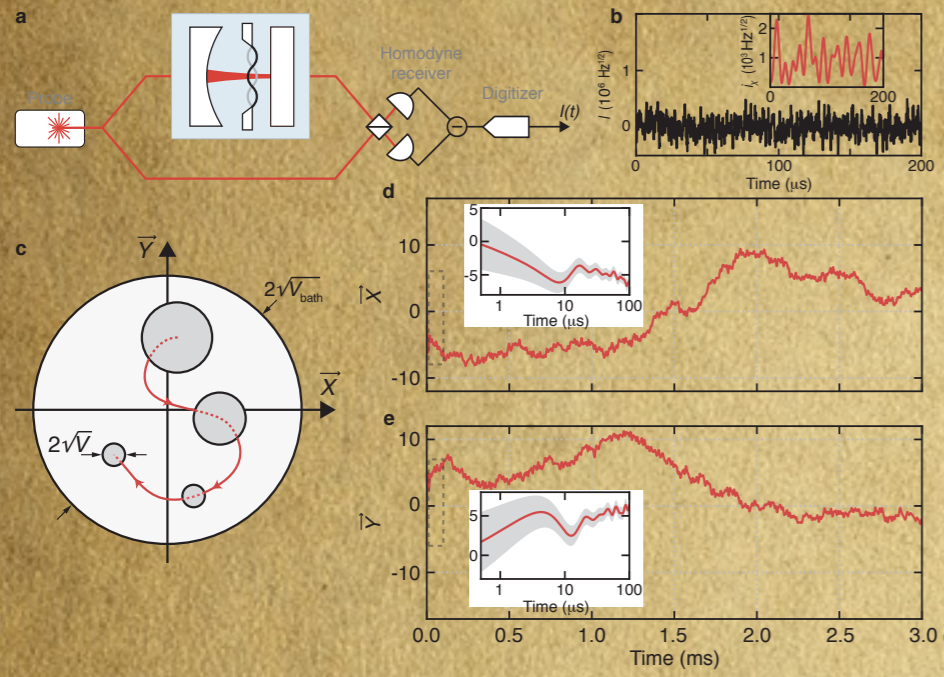
dynamics of the experiment

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

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

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

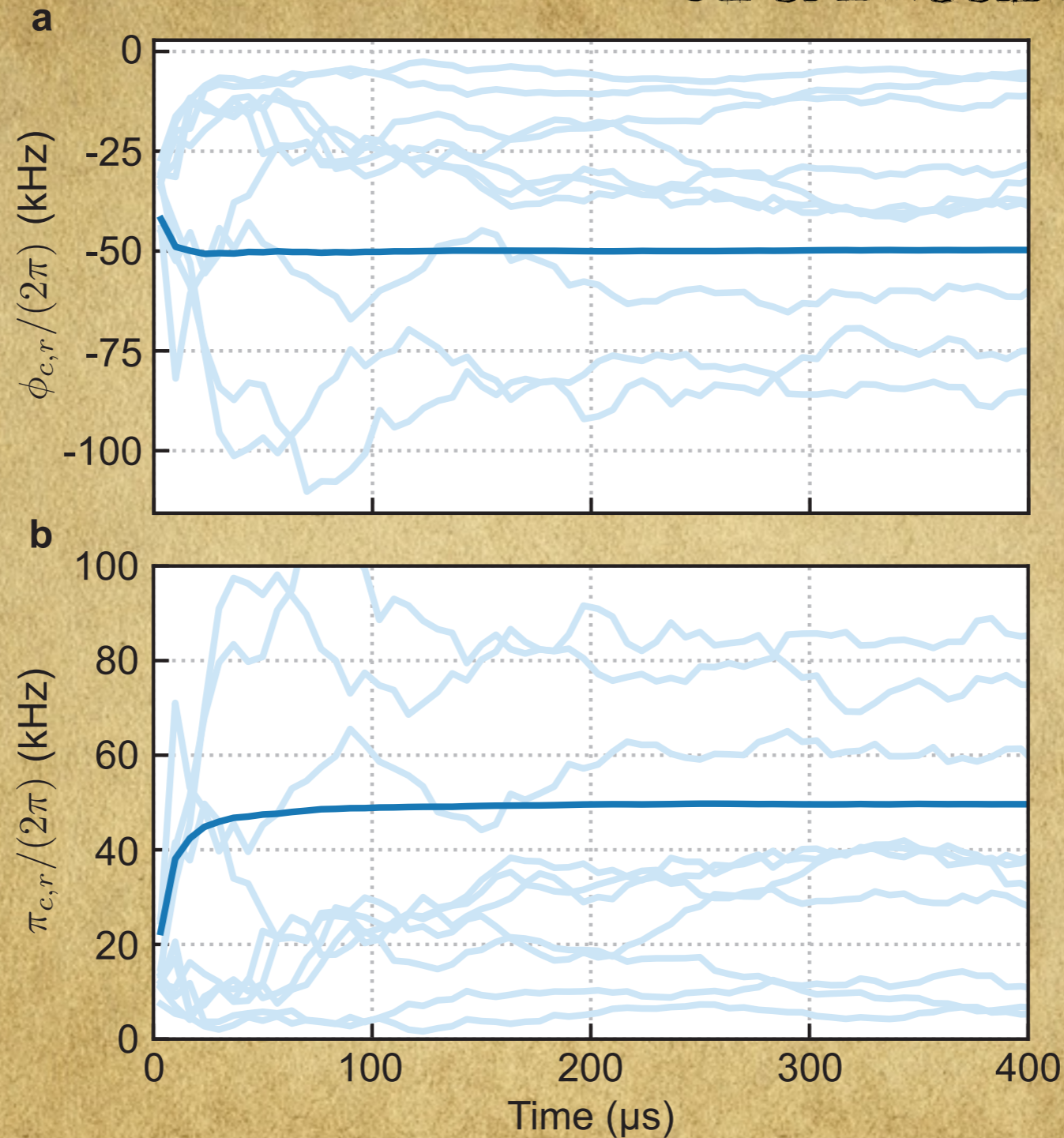
$$\Pi_{uc}(t) = \Gamma_m \left[V_{uc} / (n_{th} + 1/2) - 1 \right] + 4\Gamma_{qba} V_{uc}$$

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

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



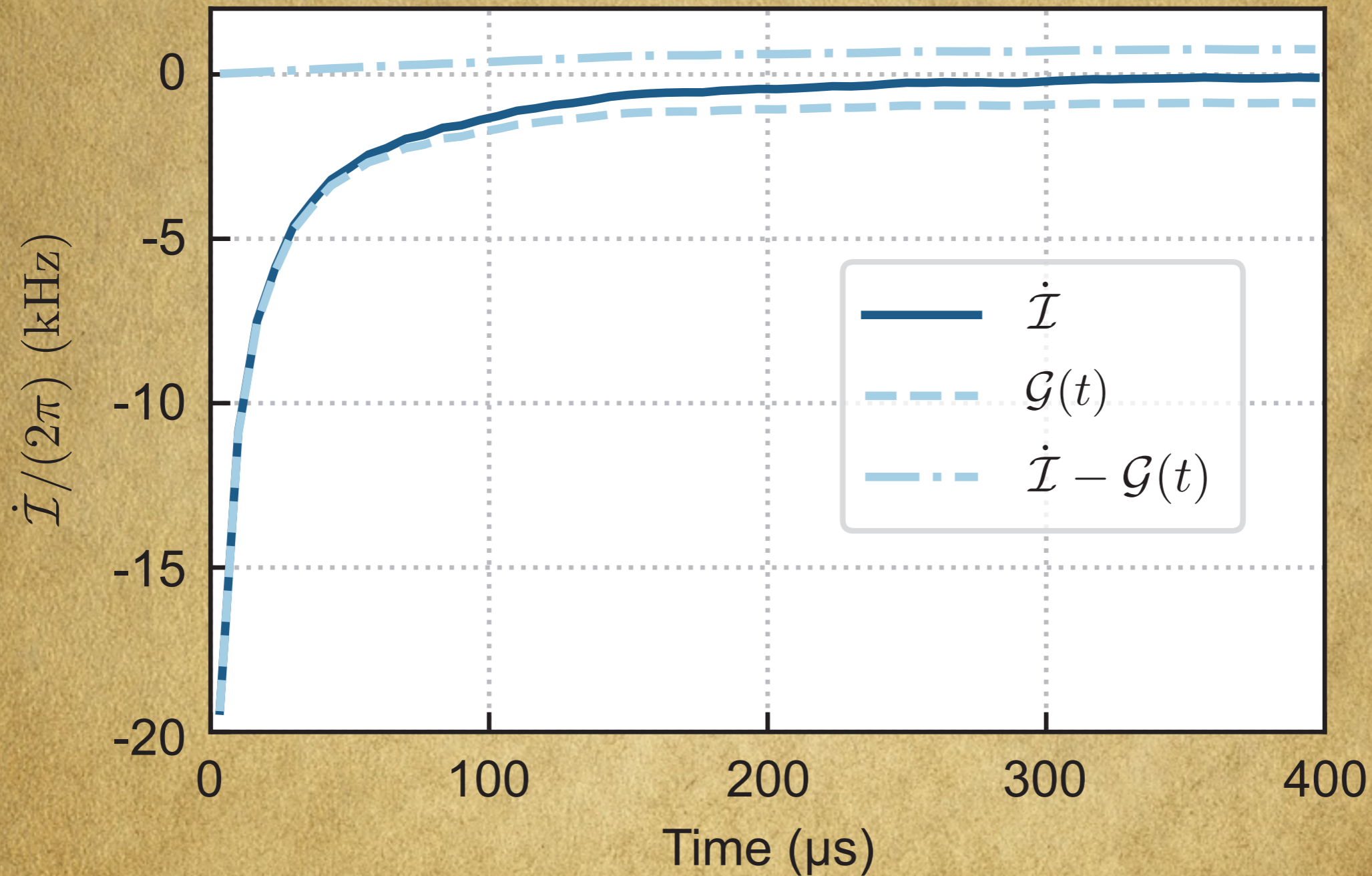
Observing entropy production rates of a measured system



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)



QTEQ
QUANTUM TECHNOLOGY at QUEEN'S

The Belfast crew





Bread on tables..



THANK YOU