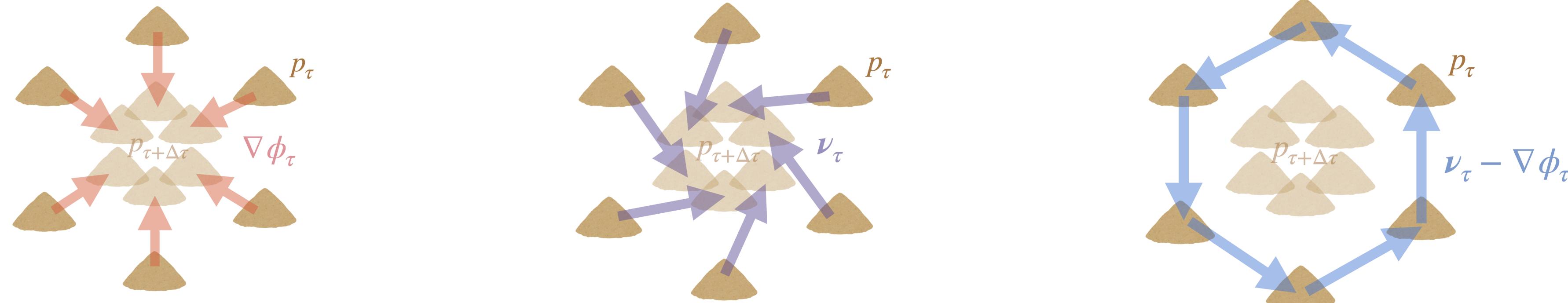


Geometric decomposition of entropy production rate:

Wisdom from optimal transport



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- M. Nakazato and SI. Phys. Rev. Res. **3**, 043093 (2021).
A. Dechant, S-I Sasa and SI. Phys. Rev. Res. **4**, L012034 (2022).
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K. Yoshimura, A. Kolchinsky, A. Dechant and SI. Phys. Rev. Res. **5**, 013017 (2023).
A. Kolchinsky, A. Dechant, K. Yoshimura and SI. arXiv:2206.14599 (2022).
SI, Information Geometry, 1-42 (2023).

Decomposition of entropy production rate

The entropy production rate (EPR) at time t : $\sigma_t(\geq 0)$

Hatano-Sasa (nonadiabatic/adiabatic) decomposition: $\sigma_t = \sigma_t^{\text{ex,HS}} + \sigma_t^{\text{hk,HS}}$
 $(\geq 0) (\geq 0)$

T. Hatano and S-I. Sasa, Phys. Rev. Lett. **86**. 3463 (2001). M. Esposito and C. Van den Broeck, Phys. Rev. E **82**. 011143 (2010).

The EPR in a non-equilibrium steady state (NESS) σ^{st} is characterized by cycle flows.

J. Schnakenberg, Rev. of Mod. Phys. **48**, 571 (1976).

In a NESS

$$\sigma_t^{\text{ex,HS}} = 0$$

$$\sigma_t^{\text{hk,HS}} = \sigma^{\text{st}}$$

A disadvantage of Hatano-Sasa decomposition

e.g.) Fokker-Planck equation

$$\partial_t p_t(\mathbf{x}) = - \nabla \cdot (\boldsymbol{\nu}_t(\mathbf{x}) p_t(\mathbf{x})) \quad \boldsymbol{\nu}_t(\mathbf{x}) = \mu \mathbf{F}_t(\mathbf{x}) - \mu T \nabla \ln p_t(\mathbf{x})$$

$$\text{NESS: } \partial_t p_t^{\text{st}}(\mathbf{x}) = - \nabla \cdot (\boldsymbol{\nu}_t^{\text{st}}(\mathbf{x}) p_t^{\text{st}}(\mathbf{x})) = 0 \quad \boldsymbol{\nu}_t^{\text{st}}(\mathbf{x}) = \mu \mathbf{F}_t(\mathbf{x}) - \mu T \nabla \ln p_t^{\text{st}}(\mathbf{x})$$

$$\sigma_t = \frac{1}{\mu T} \int d\mathbf{x} \|\boldsymbol{\nu}_t(\mathbf{x})\|^2 p_t(\mathbf{x}) \quad \sigma_t^{\text{ex,HS}} = \frac{1}{\mu T} \int d\mathbf{x} \|\boldsymbol{\nu}_t(\mathbf{x}) - \boldsymbol{\nu}_t^{\text{st}}(\mathbf{x})\|^2 p_t(\mathbf{x}) \quad \sigma_t^{\text{hk,HS}} = \frac{1}{\mu T} \int d\mathbf{x} \|\boldsymbol{\nu}_t^{\text{st}}(\mathbf{x})\|^2 p_t(\mathbf{x})$$

The existence of a unique and asymptotically stable NESS is assumed implicitly.

For a nonlinear rate equation, multiple NESSs generally exist and these NESSs can be unstable.

In chemical thermodynamics, this decomposition is only defined for a complex-balanced system with an asymptotically stable NESS.

Proposition

We propose another decomposition “geometric decomposition” $\sigma_t = \sigma_t^{\text{ex}} + \sigma_t^{\text{hk}}$
without considering a NESS.

The excess EPR: σ_t^{ex} (≥ 0)

- The dissipation by the same time evolution driven by a potential

The housekeeping EPR: σ_t^{hk} (≥ 0)

- The dissipation by cycle flows which do not affect the time evolution

An idea of this geometric decomposition is originated from
geometry in optimal transport theory.

Optimal transport

Optimal transport problem (Monge, 1781) C. Villani, Optimal transport: old and new, (Springer 2009).

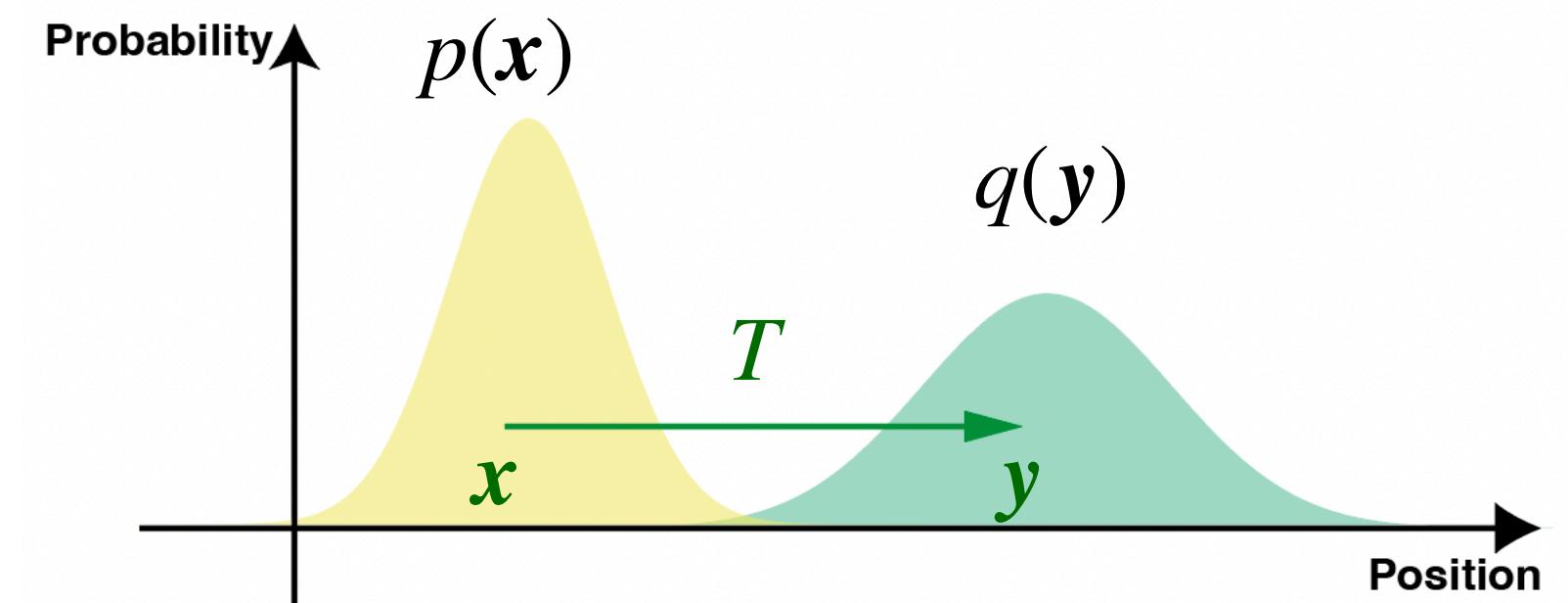
We would like to transport a probability distribution from $p(x)$ to $q(y) = \int dx \delta(y - T(x))p(x)$.

A transport cost for each point depends on $\|x - y\|^2$.

What is T to minimize all the cost?

L²-Wasserstein distance

$$\mathcal{W}(p, q) = \sqrt{\inf_T \int dx \|x - T(x)\|^2 p(x)}$$



For the Fokker-Planck equation, a relation between the free energy and L²-Wasserstein distance has been discussed.

R. Jordan, D. Kinderlehrer and F. Otto, SIAM journal on mathematical analysis, **29**, 1-17 (1998).

Optimal transport based on the continuity equation

Benamou-Brenier formula

J. D. Benamou and Y. Brenier, *Numerische Mathematik* **84**, 375-393 (2000).

$$\mathcal{W}(p, q) = \sqrt{\inf_{(\nu_t, P_t)_{t \leq t \leq \tau + \Delta\tau}} (\Delta\tau) \int_{\tau}^{\tau + \Delta\tau} dt \int dx ||\nu_t(x)||^2 P_t(x)}$$

such that $\partial_t P_t(x) = -\nabla \cdot (\nu_t(x) P_t(x))$, $P_\tau(x) = p(x)$, $P_{\tau + \Delta\tau}(x) = q(x)$

An optimal solution:

$$\mathcal{W}(p, q) = \sqrt{(\Delta\tau) \int_{\tau}^{\tau + \Delta\tau} dt \int dx ||\nu_t^*(x)||^2 P_t(x)}$$

$$\partial_t P_t(x) = -\nabla \cdot (\nu_t^*(x) P_t(x)),$$

$$\nu_t^*(x) = \nabla \phi_t(x)$$

$$\partial_t \phi_t(x) + \frac{1}{2} \|\nabla \phi_t(x)\|^2 = 0$$

(Pressureless Euler equation in fluid mechanics without “vorticity”)

Minimum entropy production

Thermodynamic speed limit

E. Aurell *et al.* J. Stat. Phys. **147**, 487-505 (2012).

$$\int_{\tau}^{\tau+\Delta\tau} dt \sigma_t \geq \frac{\mathcal{W}(p_{\tau}, p_{\tau+\Delta\tau})^2}{\mu T \Delta\tau}$$

Infinitesimal time evolution

M. Nakazato and SI. Phys. Rev. Res. **3**, 043093 (2021).

$$\sigma_{\tau} \geq \frac{1}{\mu T} \left(\lim_{\Delta\tau \rightarrow 0} \frac{\mathcal{W}(p_{\tau}, p_{\tau+\Delta\tau})}{\Delta\tau} \right)^2$$

$$\underline{\sigma_{\tau}^{\text{rot}} = \sigma_{\tau} - \frac{1}{\mu T} \left(\lim_{\Delta\tau \rightarrow 0} \frac{\mathcal{W}(p_{\tau}, p_{\tau+\Delta\tau})}{\Delta\tau} \right)^2}$$
$$=: \sigma_{\tau}^{\text{hk}} (\geq 0)$$

$$\sigma_{\tau}^{\text{rot}} = 0 \quad (\nu_{\tau}(x) = v_{\tau}^*(x) = \nabla \phi_{\tau}(x))$$

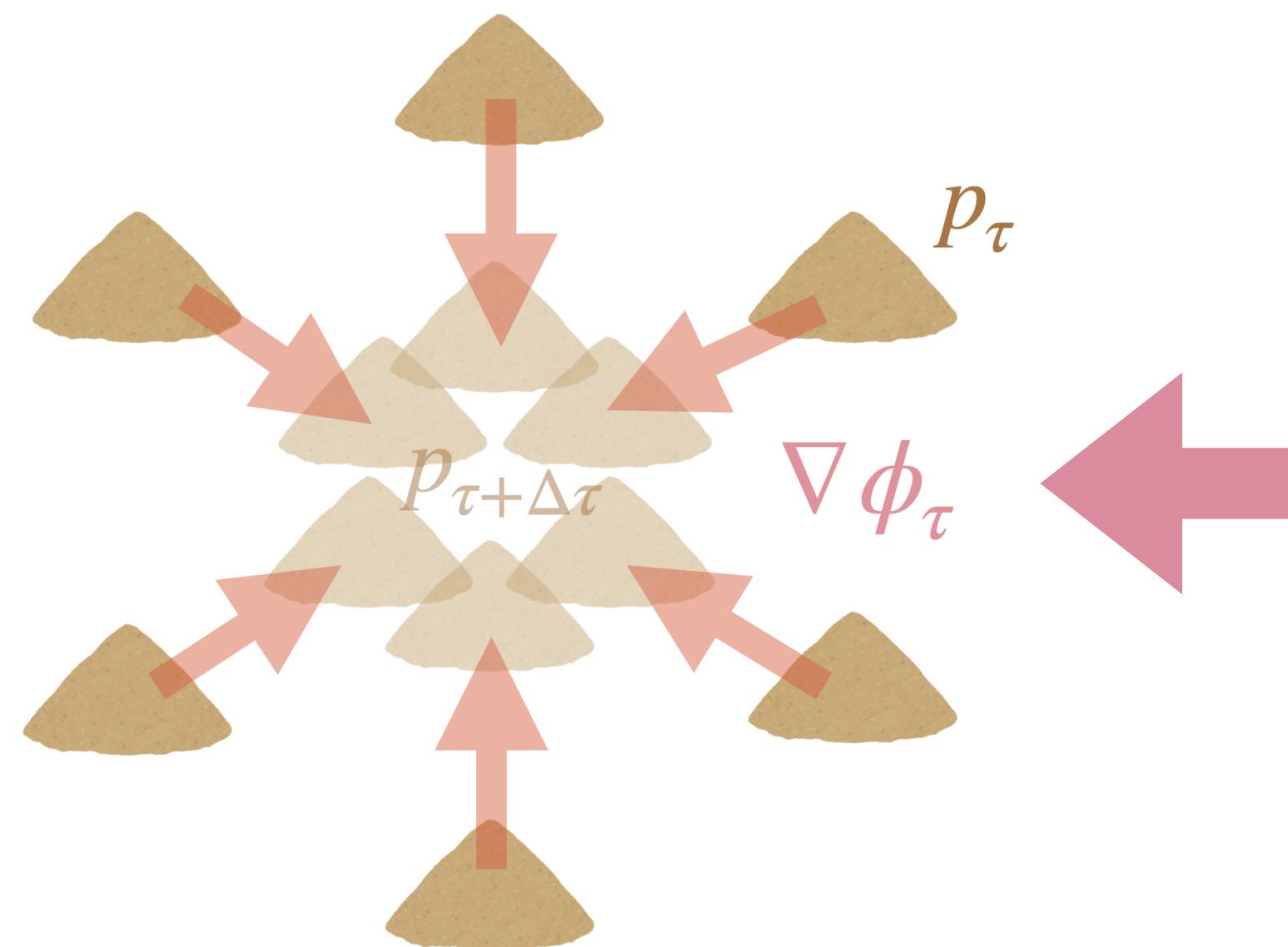
Vortices (cycle flows)

Optimal transport (potential)

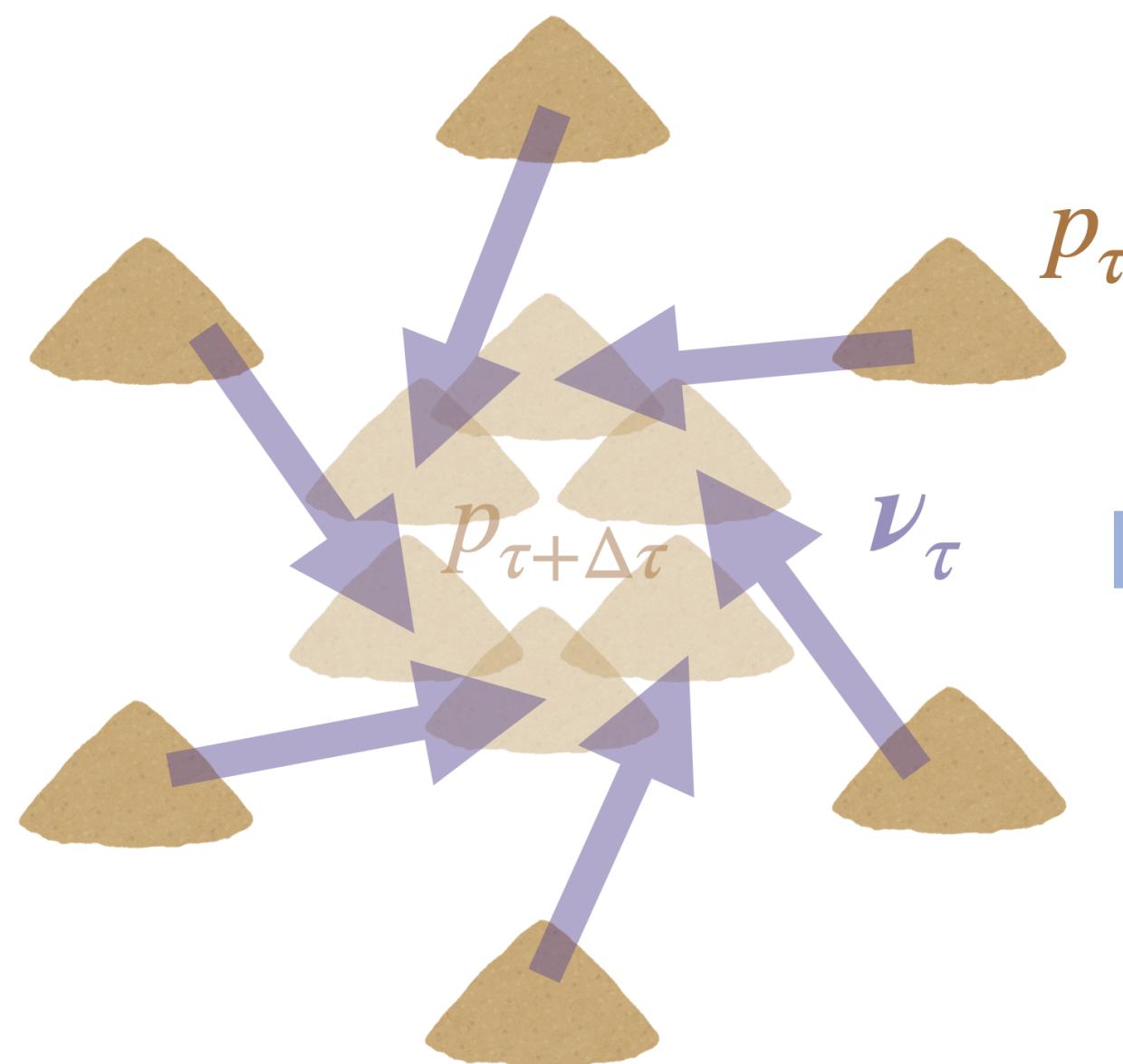
Schematic of geometric decomposition

$$(\Delta\tau \rightarrow 0)$$

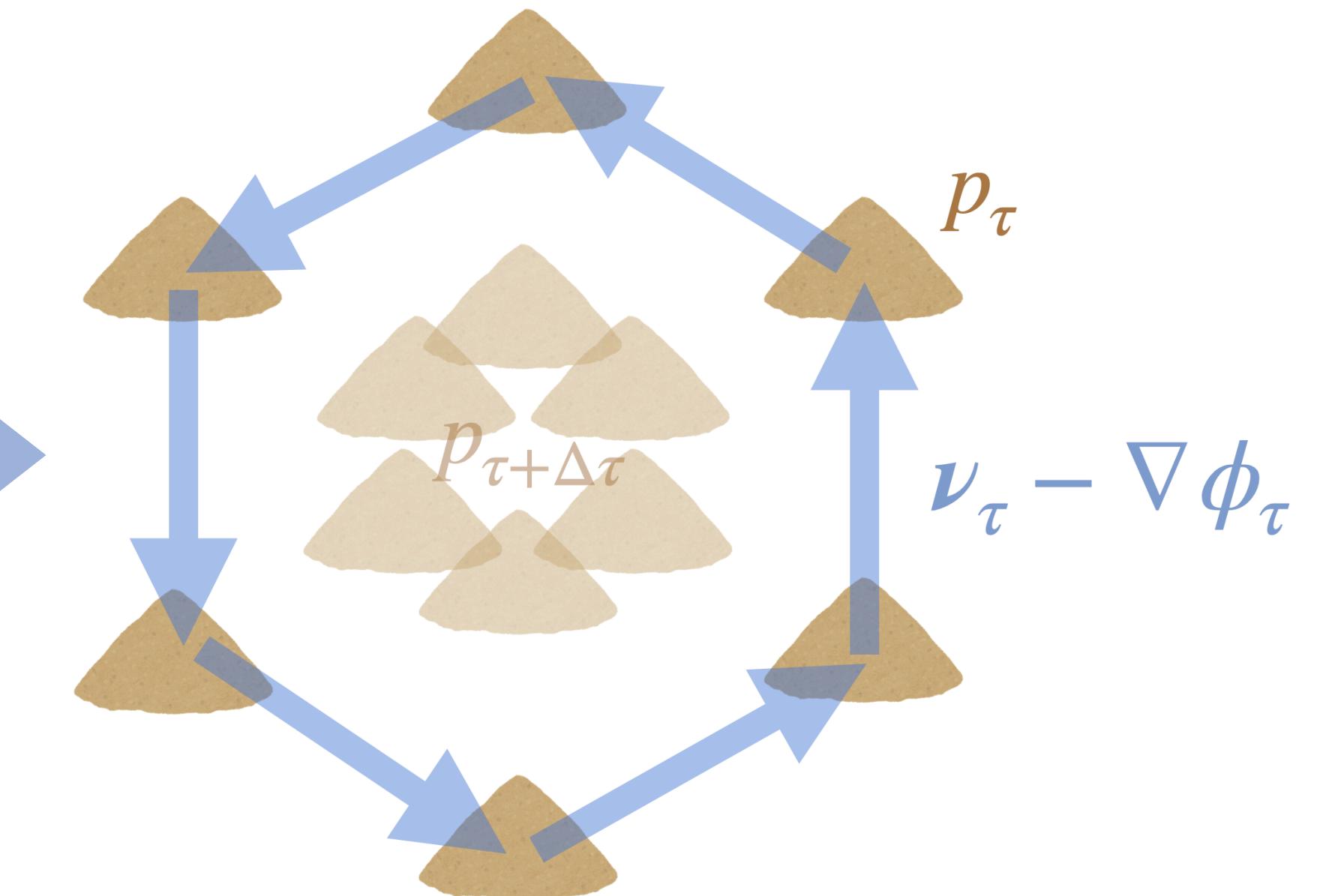
Optimal transport (potential)



(Non-optimal) time evolution



Vortices (cycle flows)



$$\partial_\tau p_\tau(x) = - \nabla \cdot [(\nabla \phi_\tau(x)) p_\tau(x)]$$

$$\partial_\tau p_\tau(x) = - \nabla \cdot [\nu_\tau(x) p_\tau(x)]$$

$$\nabla \cdot [(\nu_\tau(x) - \nabla \phi_\tau) p_\tau(x)] = 0$$

$$\sigma_\tau^{\text{ex}} := \frac{1}{\mu T} \int dx \|\nabla \phi_\tau(x)\|^2 p_\tau(x)$$

$$\sigma_\tau = \frac{1}{\mu T} \int dx \|\nu_\tau(x)\|^2 p_\tau(x)$$

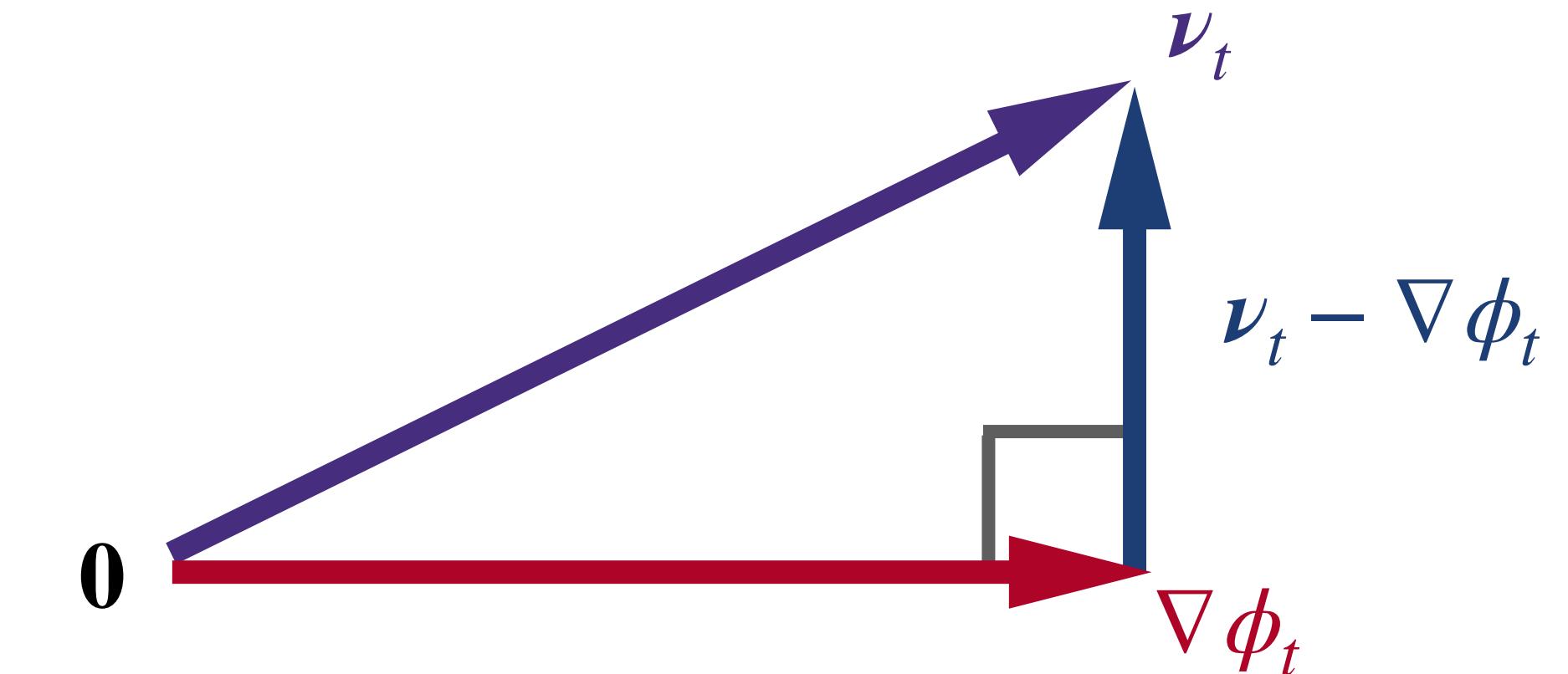
$$\sigma_\tau^{\text{hk}} := \frac{1}{\mu T} \int dx \|\nu_\tau(x) - \nabla \phi_\tau(x)\|^2 p_\tau(x)$$

Inner product and orthogonality

Inner product: $\langle \mathbf{a}, \mathbf{b} \rangle_{\frac{p_t}{\mu T}} = \frac{1}{\mu T} \int d\mathbf{x} \mathbf{a}(\mathbf{x}) \cdot \mathbf{b}(\mathbf{x}) p_t(\mathbf{x})$

Pythagorean theorem:

$$\begin{aligned} \sigma_t &= \langle \boldsymbol{\nu}_t, \boldsymbol{\nu}_t \rangle_{\frac{p_t}{\mu T}} = \underbrace{\langle \nabla \phi_t, \nabla \phi_t \rangle_{\frac{p_t}{\mu T}}}_{= \sigma_t^{\text{ex}}} + \underbrace{\langle \boldsymbol{\nu}_t - \nabla \phi_t, \boldsymbol{\nu}_t - \nabla \phi_t \rangle_{\frac{p_t}{\mu T}}}_{= \sigma_t^{\text{hk}}} \\ &= \sigma_t^{\text{ex}} + \sigma_t^{\text{hk}} \end{aligned}$$



Orthogonality: $\langle \boldsymbol{\nu}_t - \nabla \phi_t, \nabla \phi_t \rangle_{\frac{p_t}{\mu T}} = 0$
 $(\nabla \cdot [(\boldsymbol{\nu}_t(\mathbf{x}) - \nabla \phi_t)p_t(\mathbf{x})] = 0)$

A. Dechant, S-I Sasa and SI. Phys. Rev. Res. **4**, L012034 (2022).

Mathematically, the same decomposition had been proposed by Maes and Netočný based on $\nabla \cdot (\mathbf{u}_t p_t) = 0$.

$$\sigma_t \geq \sigma_t^{\text{hk}} = \langle \mathbf{u}_t, \mathbf{u}_t \rangle_{\frac{p_t}{\mu T}}$$

C. Maes and K. Netočný, J. Stat. Phys. **154**, 188-203 (2014).

Orthogonal complement

Orthogonality in geometric decomposition:

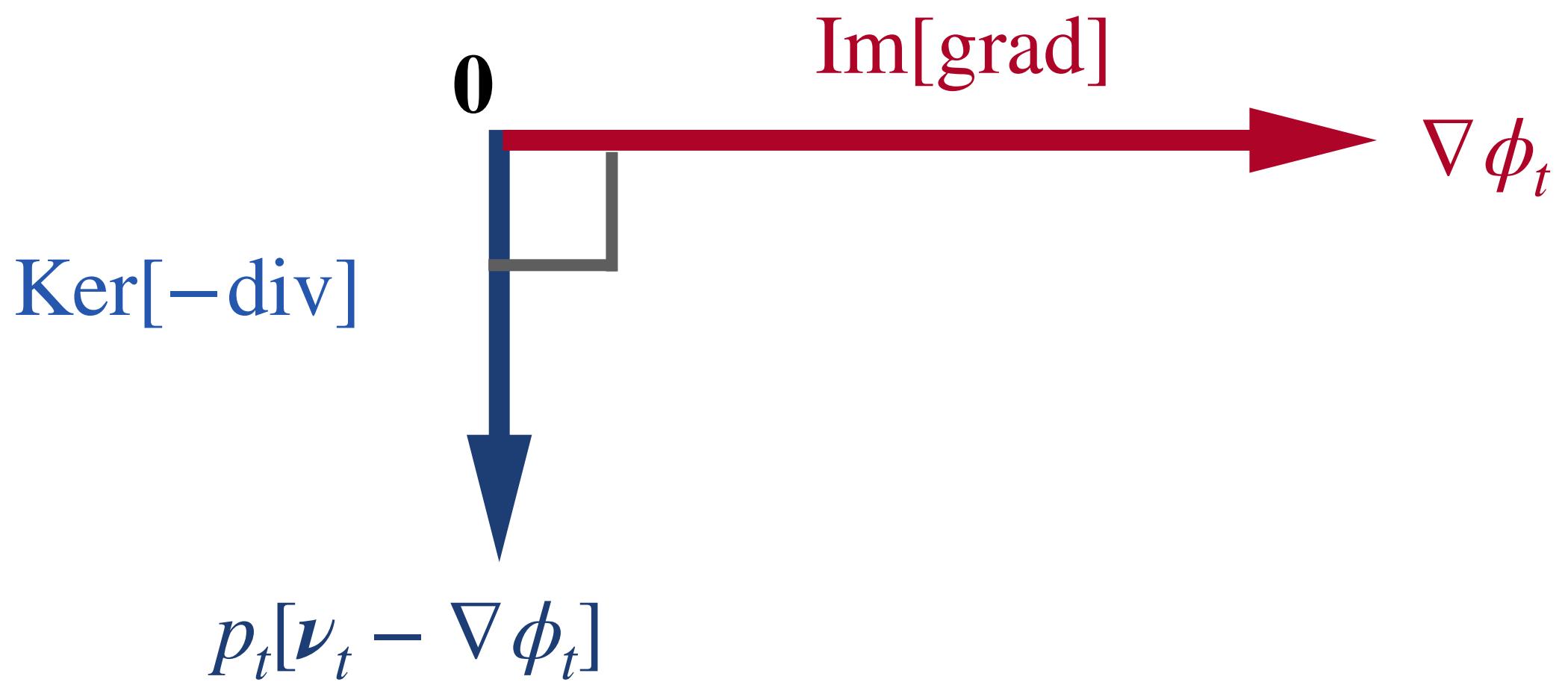
$$\langle \boldsymbol{\nu}_t - \nabla \phi_t, \nabla \phi_t \rangle_{\frac{p_t}{\mu T}} = 0$$

$$\nabla \phi_t \in \text{Im}[\text{grad}] = \{ \nabla \psi \mid \psi(x) \in \mathbb{R} \}$$

$$p_t[\boldsymbol{\nu}_t - \nabla \phi_t] \in \text{Ker}[-\text{div}] = \{ \mathbf{j}(x) \in \mathbb{R}^d \mid -\nabla \cdot \mathbf{j} = 0 \}$$

$$(-\nabla \cdot [(\boldsymbol{\nu}_\tau(x) - \nabla \phi_\tau)p_\tau(x)] = 0)$$

$$\text{Ker}[-\text{div}] \perp \text{Im}[\text{grad}]$$



Information-geometric orthogonality (Path probability)

SI, Information Geometry, 1-42 (2023)

Path probability for modified dynamics

$$\mathbb{P}_{\nu}(\mathbf{x}_t, \mathbf{x}_{t+dt}) = \mathbb{T}_{\nu}(\mathbf{x}_{t+dt} | \mathbf{x}_t) p_t(\mathbf{x}_t)$$

$$\mathbb{T}_{\nu}(\mathbf{x}_{t+dt} | \mathbf{x}_t) \propto \exp \left[-\frac{\|\mathbf{x}_{t+dt} - \mathbf{x}_t - (\mu \mathbf{F}_t(\mathbf{x}_t) + \nu'(\mathbf{x}_t) - \nu_t(\mathbf{x}_t)) dt\|^2}{4\mu T dt} \right]$$

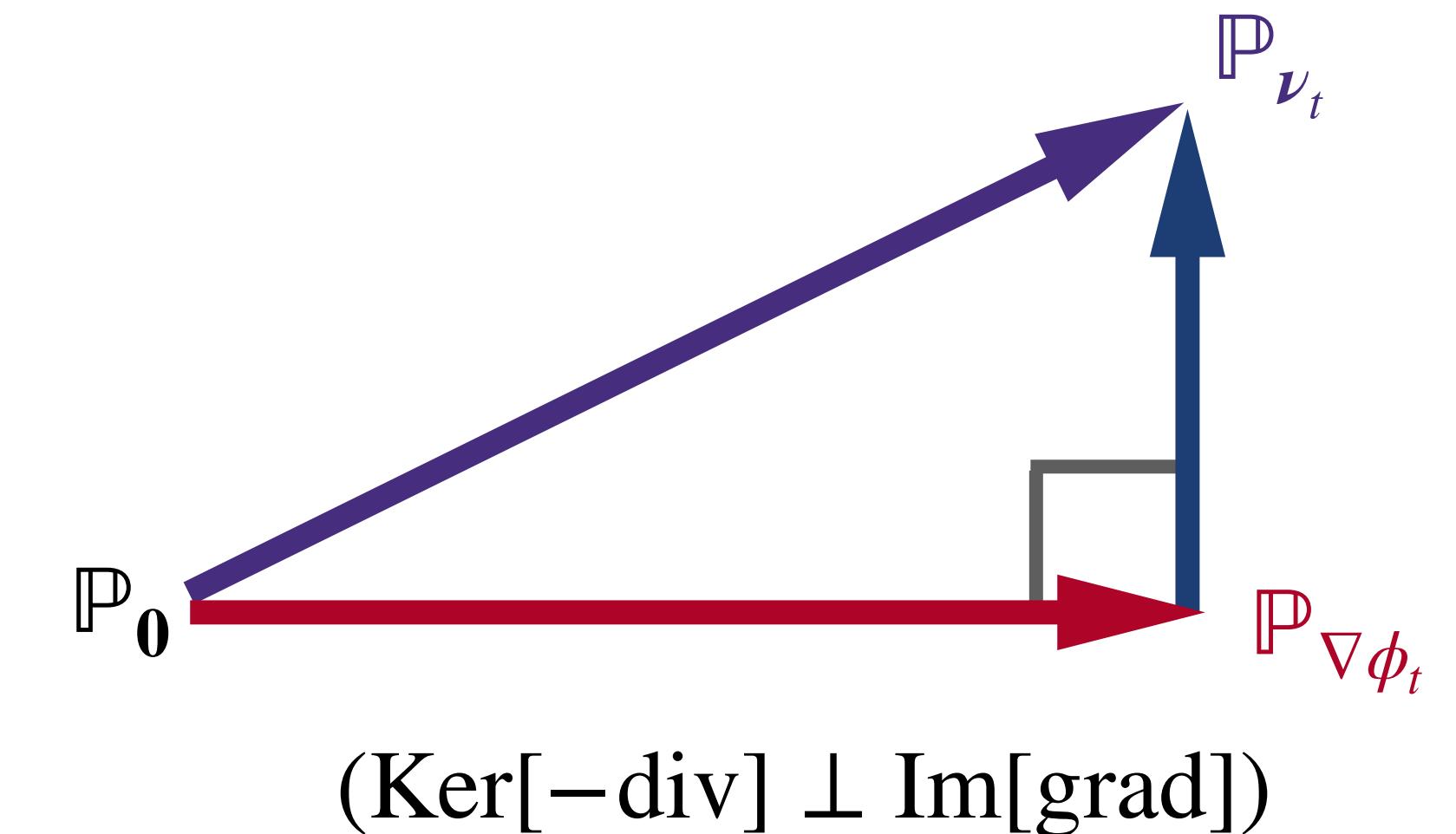
$$\partial_t p_t(\mathbf{x}) = - \nabla \cdot (\nu'(\mathbf{x}) p_t(\mathbf{x}))$$

Kullback-Leibler divergence:

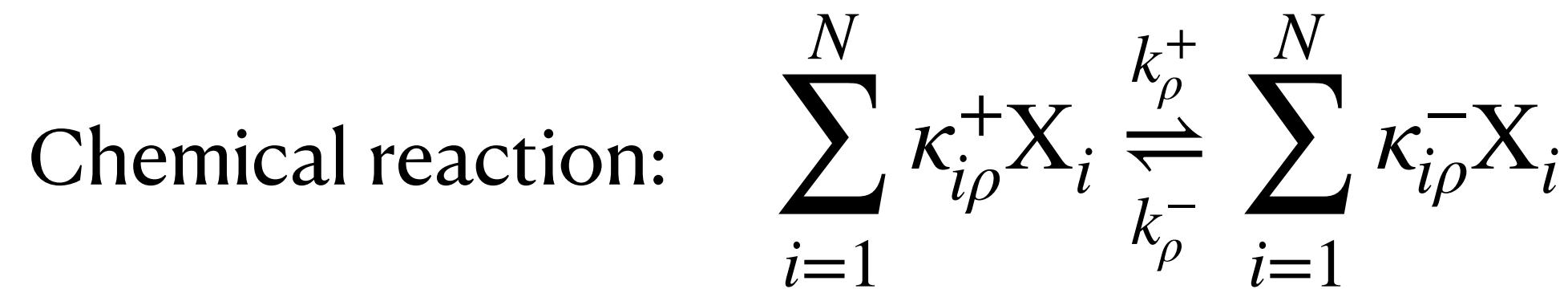
$$D_{\text{KL}}(\mathbb{P}_{\nu'} \| \mathbb{P}_{\nu''}) = \int d\mathbf{x}_t d\mathbf{x}_{t+dt} \mathbb{P}_{\nu'}(\mathbf{x}_t, \mathbf{x}_{t+dt}) \ln \frac{\mathbb{P}_{\nu'}(\mathbf{x}_t, \mathbf{x}_{t+dt})}{\mathbb{P}_{\nu''}(\mathbf{x}_t, \mathbf{x}_{t+dt})}$$

Generalized Pythagorean theorem (Information-geometric orthogonality):

$$\begin{aligned} D_{\text{KL}}(\mathbb{P}_{\nu_t} \| \mathbb{P}_0) &= D_{\text{KL}}(\mathbb{P}_{\nu_t} \| \mathbb{P}_{\nabla \phi_t}) + D_{\text{KL}}(\mathbb{P}_{\nabla \phi_t} \| \mathbb{P}_0) \\ &= \sigma_t dt / 4 \quad = \sigma_t^{\text{hk}} dt / 4 \quad = \sigma_t^{\text{ex}} dt / 4 \end{aligned}$$



Rate/master equation



Rate equation (nonlinear): $d_t c_i = \sum_{\rho} S_{i\rho} J_{\rho}$ Stoichiometric matrix: $S_{i\rho} = \kappa_{i\rho}^- - \kappa_{i\rho}^+$

Flow: $J_{\rho} = J_{\rho}^+ - J_{\rho}^-$ e.g., (Ideal dilute solution) $J_{\rho}^{\pm} = k_{\rho}^{\pm} \prod_i (c_i)^{K_{i\rho}^{\pm}}$

Master equation: $d_t p_i = \sum_{\rho} S_{i\rho} J_{\rho} = \sum_j [W_{ij} p_j - W_{ji} p_i]$

Flow: $J_{\rho} = W_{ij} p_j - W_{ji} p_i$ ($\rho = (i, j)$, $W_{ij} \neq 0, W_{ji} \neq 0$)

Incidence matrix: $S_{k\rho} = \delta_{ki} - \delta_{kj}$ ($\rho = (i, j)$)

Geometric decomposition for rate/master equation

$$\text{EPR: } \sigma_t = \sum_{\rho} F_{\rho} J_{\rho} = \sum_{\rho} (J_{\rho}^+ - J_{\rho}^-) \ln \frac{J_{\rho}^+}{J_{\rho}^-}$$

Orthogonality: $\text{Ker}[S] \perp \text{Im}[S^T]$



Geometry I (Inner product given by the Onsager matrix):

$$\sigma_t = \langle F, F \rangle_L$$

$$\langle F, F \rangle_L \leftrightarrow \langle \nu_t, \nu_t \rangle_{\frac{p_t}{\mu T}}$$

K. Yoshimura, A. Kolchinsky, A. Dechant and SI. Phys. Rev. Res. **5**, 013017 (2023).

Geometry II (generalized Kullback-Leibler divergence):

$$\sigma_t = D(\mathbf{j}(f) \parallel \mathbf{j}(\mathbf{0}))$$

$$D(\mathbf{j}(f) \parallel \mathbf{j}(\mathbf{0})) \leftrightarrow 4D_{\text{KL}}(\mathbb{P}_{\nu_t} \parallel \mathbb{P}_0)/dt$$

Geometry I: Onsager-projective excess/housekeeping EPR

K. Yoshimura, A. Kolchinsky, A. Dechant and SI. Phys. Rev. Res. **5**, 013017 (2023).

Inner product: $\langle A, B \rangle_{\mathsf{L}} = A^T \mathsf{L} B$

Onsager matrix: $\mathsf{L}_{\rho\rho'} = \frac{J_{\rho}^{+} - J_{\rho}^{-}}{\ln J_{\rho}^{+} - \ln J_{\rho}^{-}} \delta_{\rho\rho'} \quad \mathsf{L}_{\rho\rho'} = J_{\rho}^{+} \delta_{\rho\rho'} \quad (J_{\rho}^{+} = J_{\rho}^{-})$

EPR: $\sigma_t = \langle F, F \rangle_{\mathsf{L}}$

“Optimal transport”: $d_t c = S J = S J^*$

$J - J^* \in \text{Ker}[S]$

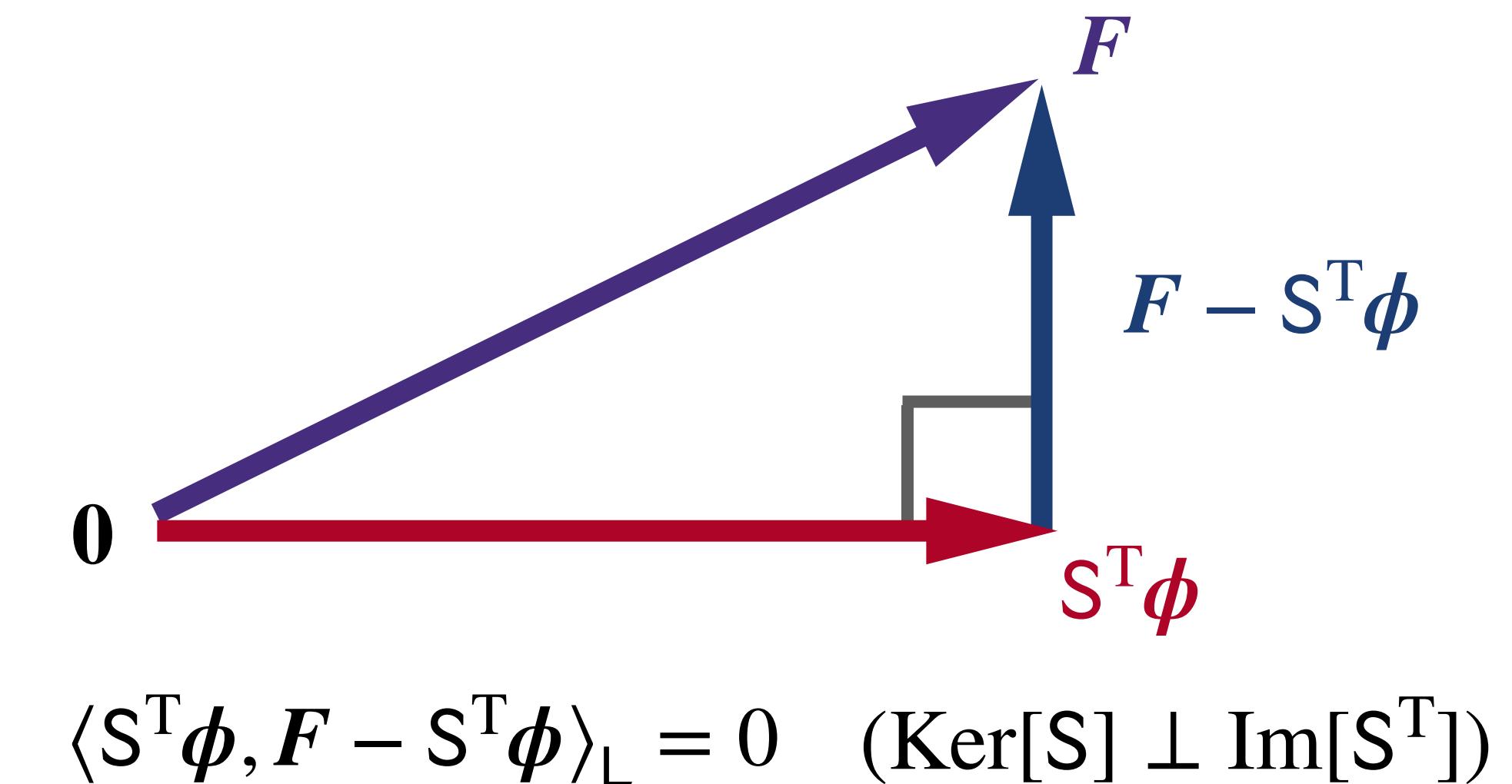
$\mathsf{L} S^T \phi \leftrightarrow p_t(x)(\nabla \phi_t(x))$

$\mathsf{L} S^T \phi = J^*$

$S^T \phi \in \text{Im}[S^T] \quad (\perp \text{Ker}[S])$

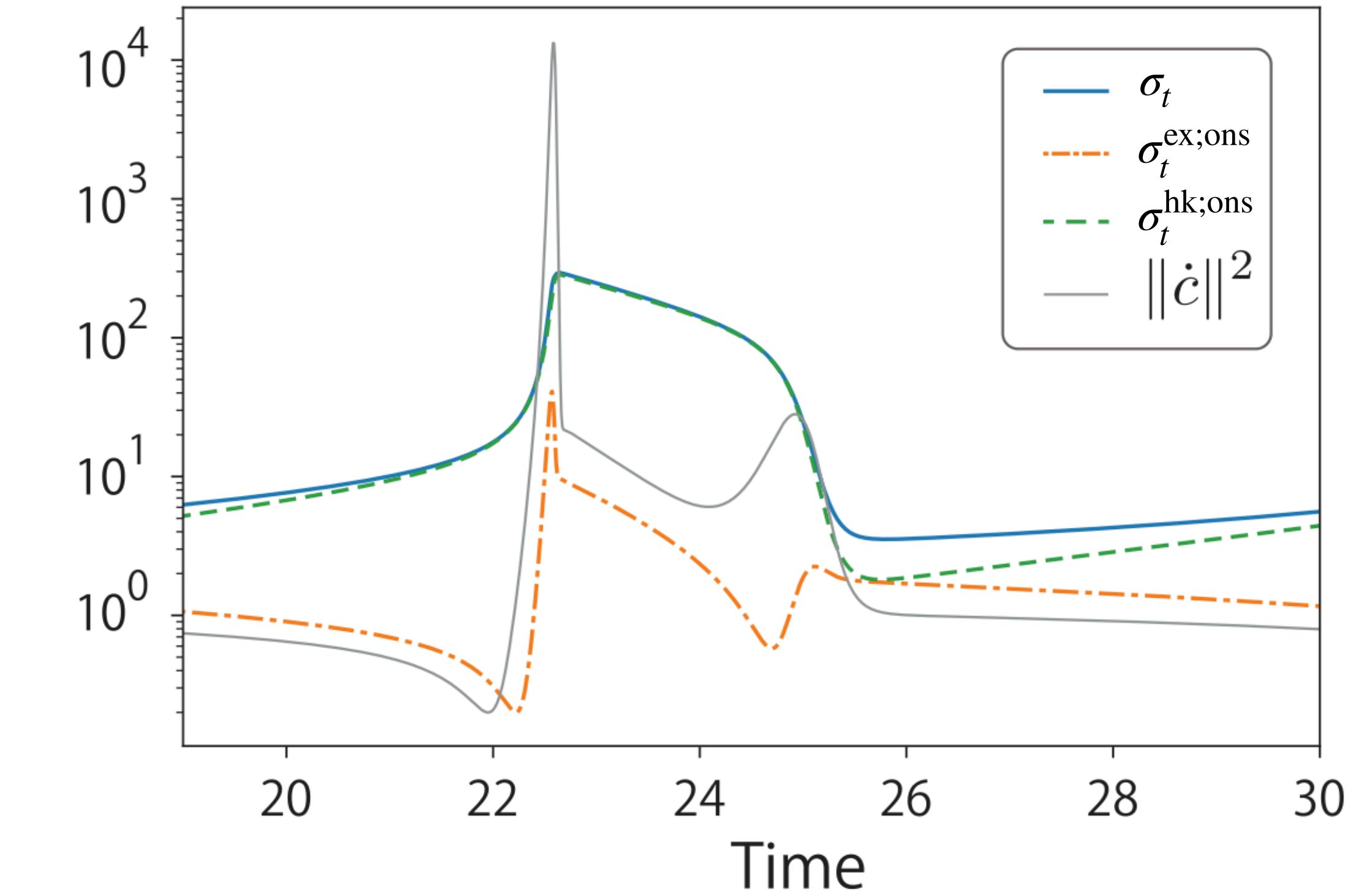
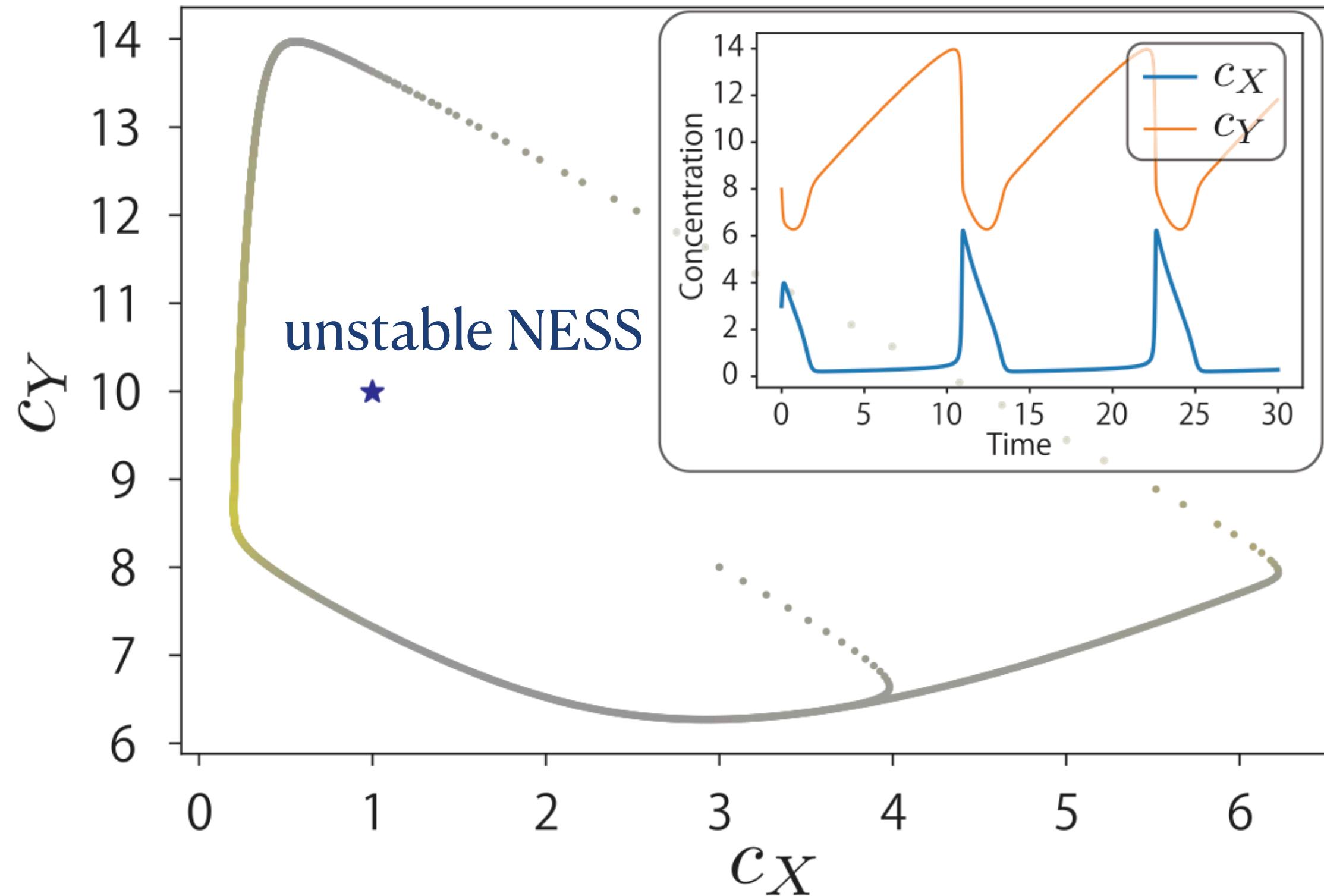
Onsager-projective decomposition:

$$\begin{aligned} \sigma_t &= \langle F, F \rangle_{\mathsf{L}} = \underbrace{\langle S^T \phi, S^T \phi \rangle_{\mathsf{L}}}_{=: \sigma_t^{\text{ex;ons}} (\geq 0)} + \underbrace{\langle F - S^T \phi, F - S^T \phi \rangle_{\mathsf{L}}}_{=: \sigma_t^{\text{hk;ons}} (\geq 0)} \end{aligned}$$

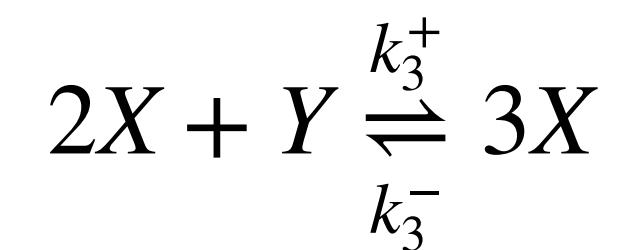
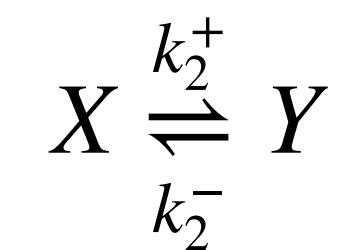
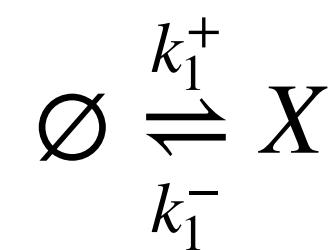


Numerics: Brusselator

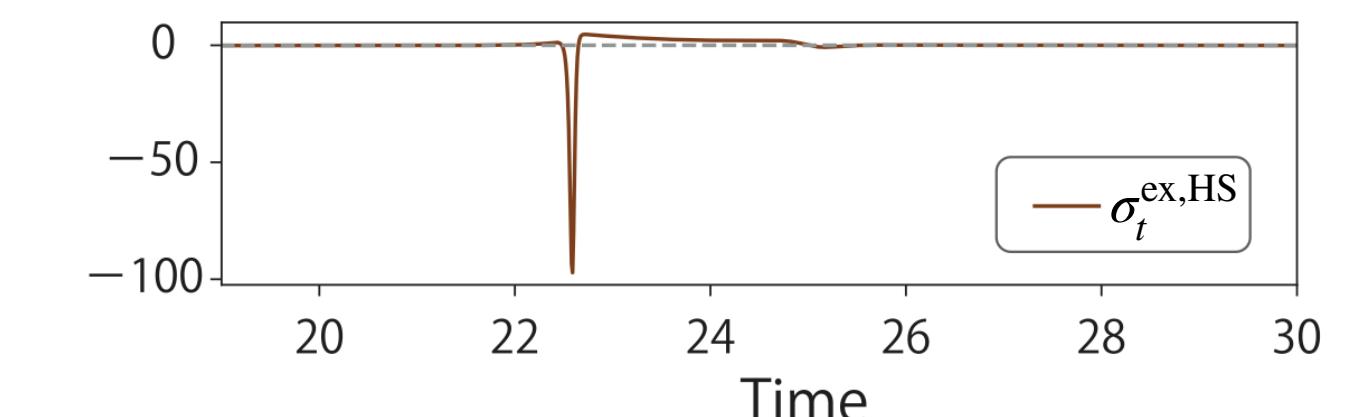
K. Yoshimura, A. Kolchinsky, A. Dechant and SI. Phys. Rev. Res. **5**, 013017 (2023).



The Hatano-Sasa excess EPR can be negative.



$$\|\dot{c}\|^2 = (dc_X)^2 + (dc_Y)^2$$



Geometry II: Information-geometric excess/housekeeping EPR

A. Kolchinsky, A. Dechant, K. Yoshimura and SI. arXiv:2206.14599 (2022).

Generalized KL divergence:

$$D(\mathbf{j}(f') \parallel \mathbf{j}(f'')) = \sum_e \left[j_e(f') \ln \frac{j_e(f')}{j_e(f'')} - j_e(f') + j_e(f'') \right]$$

EPR:

$$\sigma_t = D(\mathbf{j}(f) \parallel \mathbf{j}(\mathbf{0}))$$

$$\mathbf{j}^+ = (J_1^+, \dots, J_{|\rho|}^+, J_1^-, \dots, J_{|\rho|}^-)^T$$

$$\mathbf{j}^- = (J_1^-, \dots, J_{|\rho|}^-, J_1^+, \dots, J_{|\rho|}^+)^T$$

$$j_e(\boldsymbol{\theta}) = j_e^+ \exp(\theta_e - f_e)$$

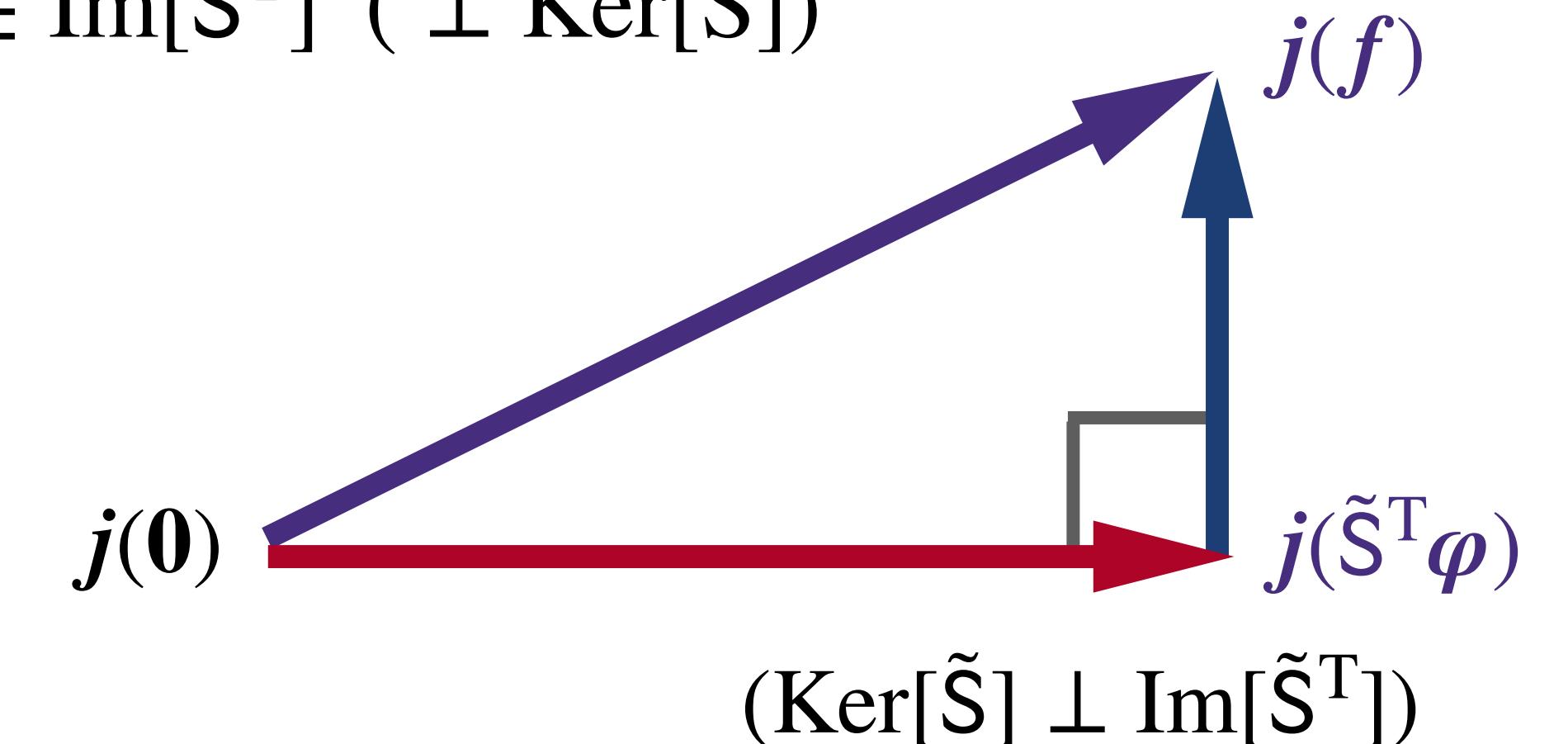
$$f_e = \ln[j_e^+ / j_e^-]$$

“Optimal transport”: $d_t \mathbf{c} = \mathbf{S} \mathbf{J} = \tilde{\mathbf{S}} \mathbf{j}(f) = \tilde{\mathbf{S}} \mathbf{j}(\tilde{\mathbf{S}}^T \boldsymbol{\varphi})$ $\mathbf{j}(f) - \mathbf{j}(\tilde{\mathbf{S}}^T \boldsymbol{\varphi}) \in \text{Ker}[\tilde{\mathbf{S}}]$ $\mathbf{j}(\tilde{\mathbf{S}}^T \boldsymbol{\varphi}) \leftrightarrow \mathbb{P}_{\nabla \phi_t}$

$$\tilde{\mathbf{S}}^T \boldsymbol{\varphi} \in \text{Im}[\tilde{\mathbf{S}}^T] \ (\perp \text{Ker}[\tilde{\mathbf{S}}])$$

Information-geometric decomposition:

$$\begin{aligned} \sigma_t &= D(\mathbf{j}(f) \parallel \mathbf{j}(\mathbf{0})) = \frac{D(\mathbf{j}(f) \parallel \mathbf{j}(\tilde{\mathbf{S}}^T \boldsymbol{\varphi}))}{=: \sigma_t^{\text{hk;IG}} (\geq 0)} + \frac{D(\mathbf{j}(\tilde{\mathbf{S}}^T \boldsymbol{\varphi}) \parallel \mathbf{j}(\mathbf{0}))}{=: \sigma_t^{\text{ex;IG}} (\geq 0)} \\ &\quad (\sigma_t^{\text{hk;IG}} \neq \sigma_t^{\text{hk;ons}}, \sigma_t^{\text{ex;IG}} \neq \sigma_t^{\text{ex;ons}}) \end{aligned}$$



Applications of geometric decompositions to TURs

Thermodynamic uncertainty relations (TURs):

$R(x), \mathbf{R}$: Observable

For Fokker-Planck Eq.: $\sigma_t^{\text{ex}} = \langle \nabla \phi_t, \nabla \phi_t \rangle_{\frac{p_t}{\mu T}} \geq \frac{|\langle \nabla \phi_t, \nabla R \rangle_{\frac{p_t}{\mu T}}|^2}{\langle \nabla R, \nabla R \rangle_{\frac{p_t}{\mu T}}} = \frac{|d_t \langle R \rangle_{p_t}|^2}{\langle \nabla R, \nabla R \rangle_{\frac{p_t}{\mu T}}} \quad \langle R \rangle_{p_t} = \int dx R(x) p_t(x)$

A. Dechant, S-I Sasa and SI. Phys. Rev. E **106**, 024125 (2022).

For rate Eq.: $\sigma_t^{\text{ex;ons}} = \langle \mathbf{S}^T \boldsymbol{\phi}, \mathbf{S}^T \boldsymbol{\phi} \rangle_L \geq \frac{|\langle \mathbf{S}^T \mathbf{R}, \mathbf{S}^T \boldsymbol{\phi} \rangle_L|^2}{\langle \mathbf{S}^T \mathbf{R}, \mathbf{S}^T \mathbf{R} \rangle_L} = \frac{|d_t(\mathbf{c}^T \mathbf{R})|^2}{\langle \mathbf{S}^T \mathbf{R}, \mathbf{S}^T \mathbf{R} \rangle_L}$

K. Yoshimura, A. Kolchinsky, A. Dechant and SI. Phys. Rev. Res. **5**, 013017 (2023).

$\sigma_t^{\text{ex;IG}}$ also provides a TUR for a highly irreversible process.

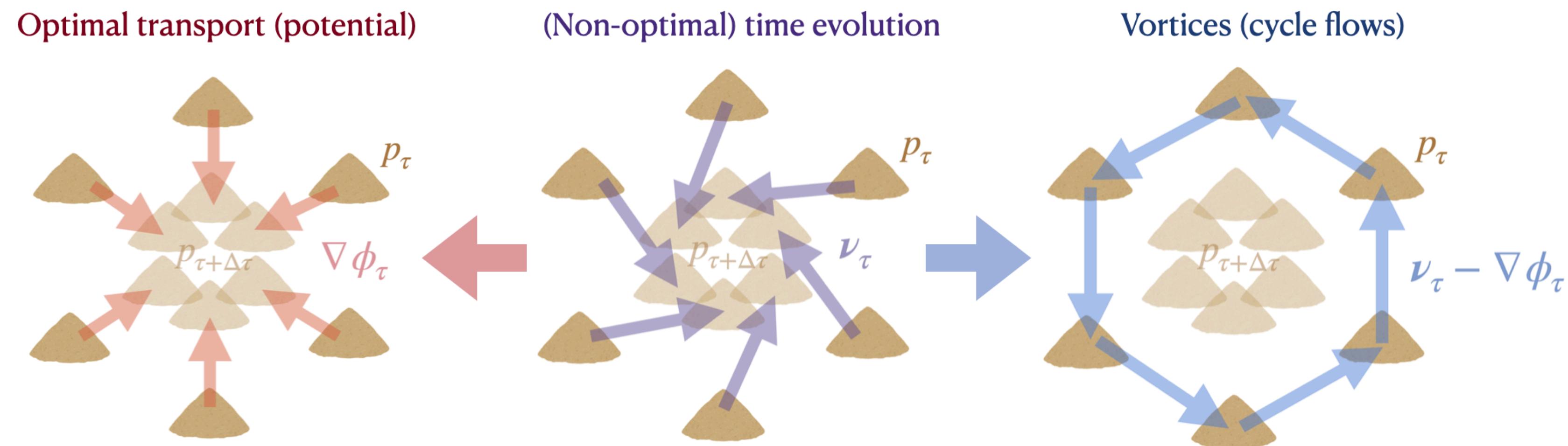
$$\sigma_t^{\text{ex;IG}} \geq 2 |d_t(\mathbf{c}^T \mathbf{R})| \tanh^{-1} \frac{|d_t(\mathbf{c}^T \mathbf{R})|}{\sum_e j_e^+ |(\tilde{\mathbf{S}}^T \mathbf{R})_e|} \quad \left(\sup_e |(\tilde{\mathbf{S}}^T \mathbf{R})_e| \leq 1 \right)$$

A. Kolchinsky, A. Dechant, K. Yoshimura and SI. arXiv:2206.14599 (2022).

Summary

We propose geometric decompositions of EPR $\sigma_t = \sigma_t^{\text{ex}} + \sigma_t^{\text{hk}}$. ($\sigma_t = \sigma_t^{\text{ex;ons}} + \sigma_t^{\text{hk;ons}}$, $\sigma_t = \sigma_t^{\text{ex;IG}} + \sigma_t^{\text{hk;IG}}$)

- The excess EPR σ_t^{ex} means the dissipation by the same time evolution driven by a potential.
- The housekeeping EPR σ_t^{hk} means the dissipation by cycle flows which do not affect the time evolution.



We generalize geometric decomposition for a nonlinear rate equation, where a unique and stable NESS does not exist generally. Geometric decompositions may be useful, at least, to derive TURs.