

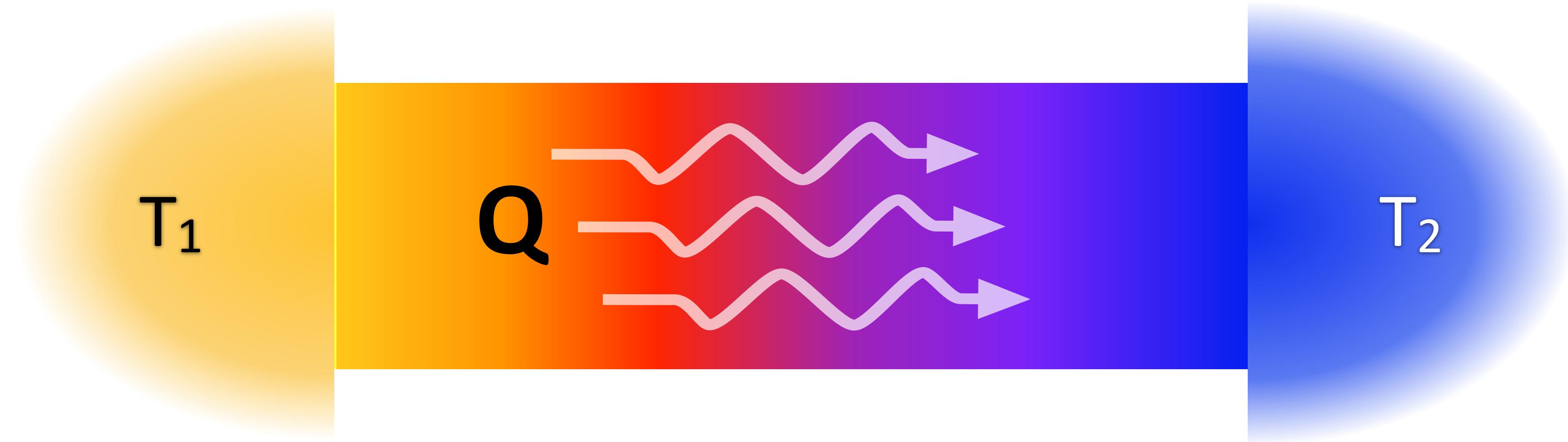


gauge invariance of heat transport coefficients

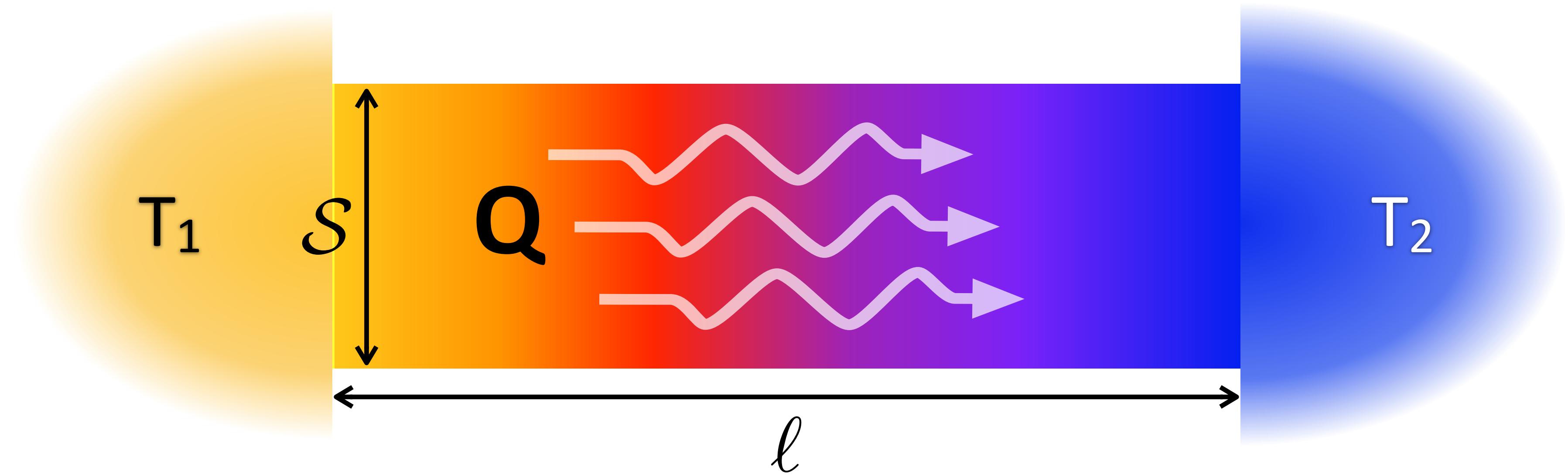
fathoming heat transport from the struggle to simulate it

Stefano Baroni
Scuola Internazionale Superiore di Studi Avanzati
Trieste — Italy

what heat transport is all about

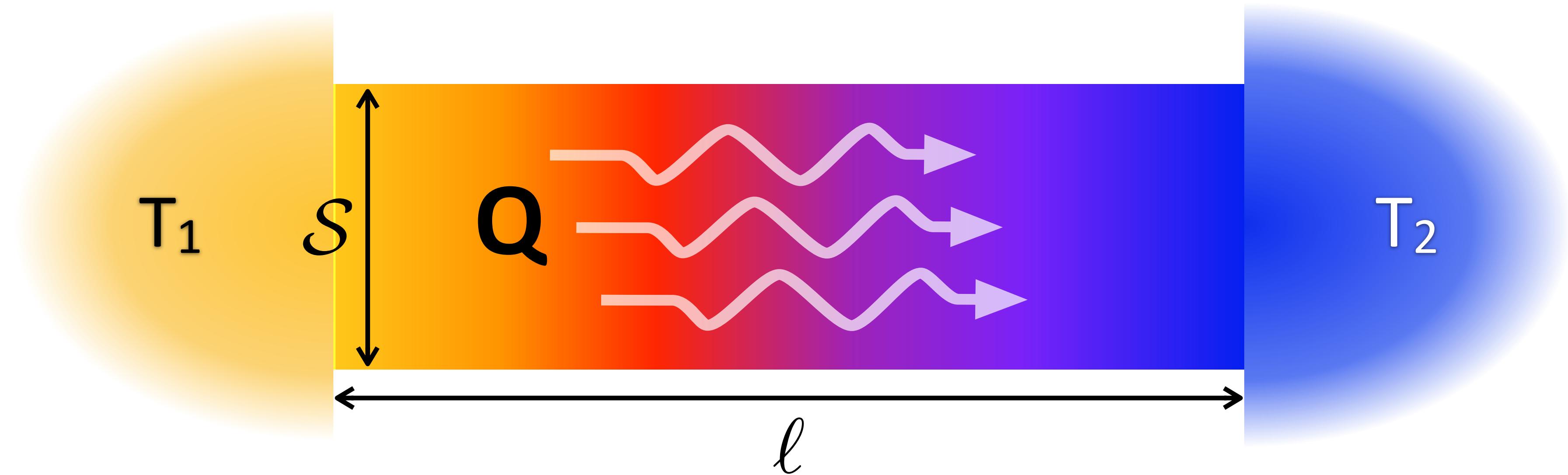


what heat transport is all about



$$\frac{1}{S} \frac{dQ}{dt} = -\kappa \frac{(T_2 - T_1)}{\ell}$$

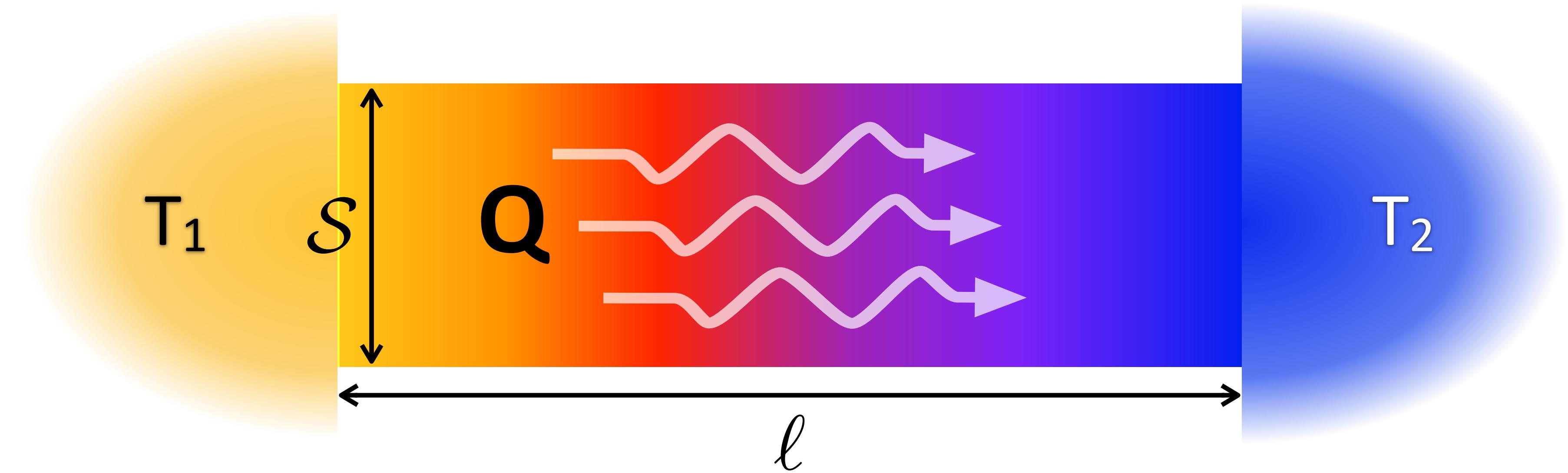
what heat transport is all about



$$\frac{1}{S} \frac{dQ}{dt} = -\kappa \frac{(T_2 - T_1)}{\ell}$$

$$\mathbf{j}_Q(\mathbf{r}, t) = -\kappa \nabla T(\mathbf{r}, t)$$

what heat transport is all about



$$\frac{1}{S} \frac{dQ}{dt} = -\kappa \frac{(T_2 - T_1)}{\ell}$$

$$\mathbf{j}_Q(\mathbf{r}, t) = -\kappa \nabla T(\mathbf{r}, t)$$

$$\frac{\partial T}{\partial t} = \frac{\kappa}{\rho c_p} \Delta T$$



what heat transport is all about



heat flows from warmth to chill as
time flows from the past to the future

$$\frac{1}{\kappa} \frac{dQ}{dt} = -\frac{(T_2 - T_1)}{l}$$

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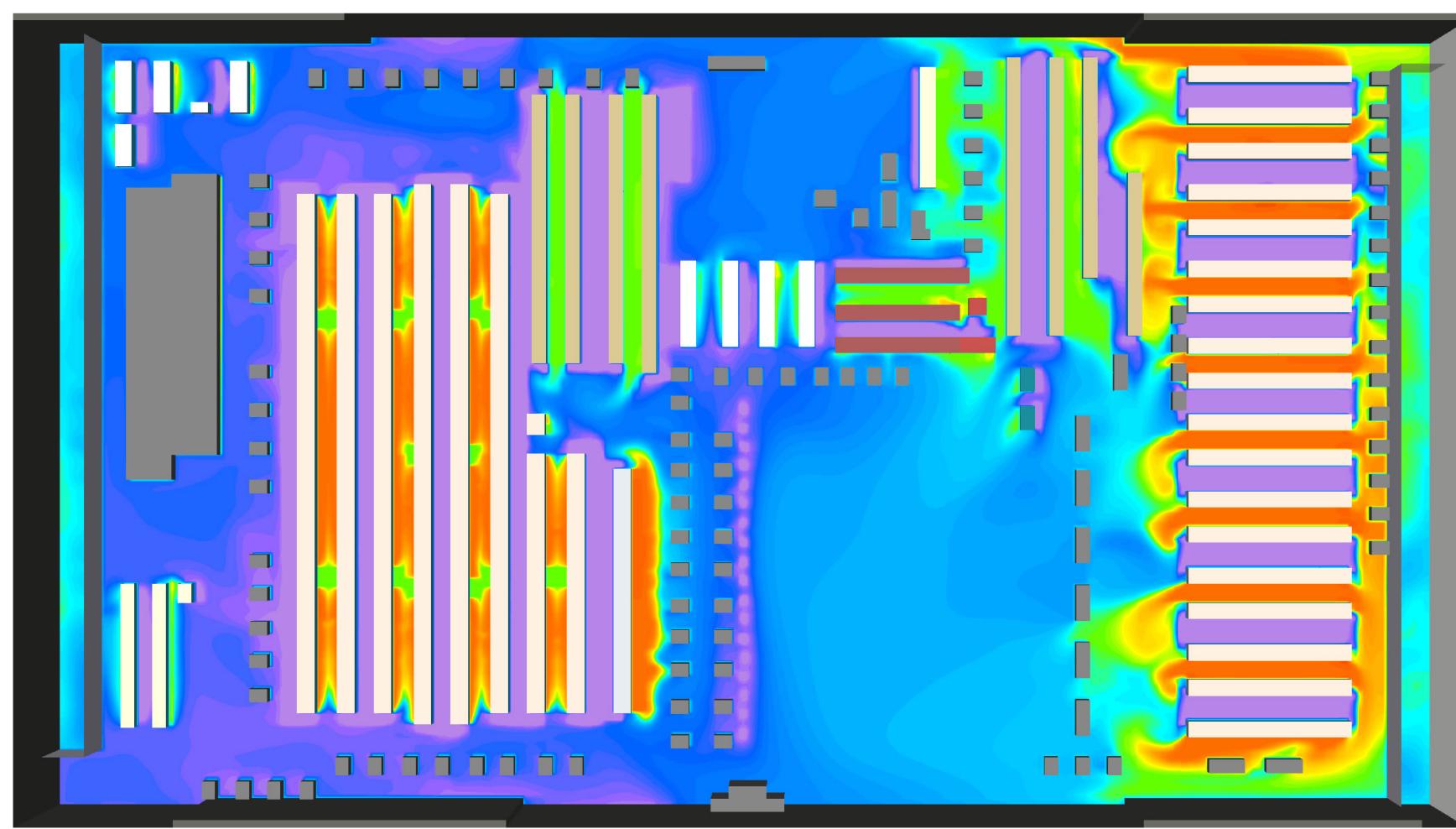
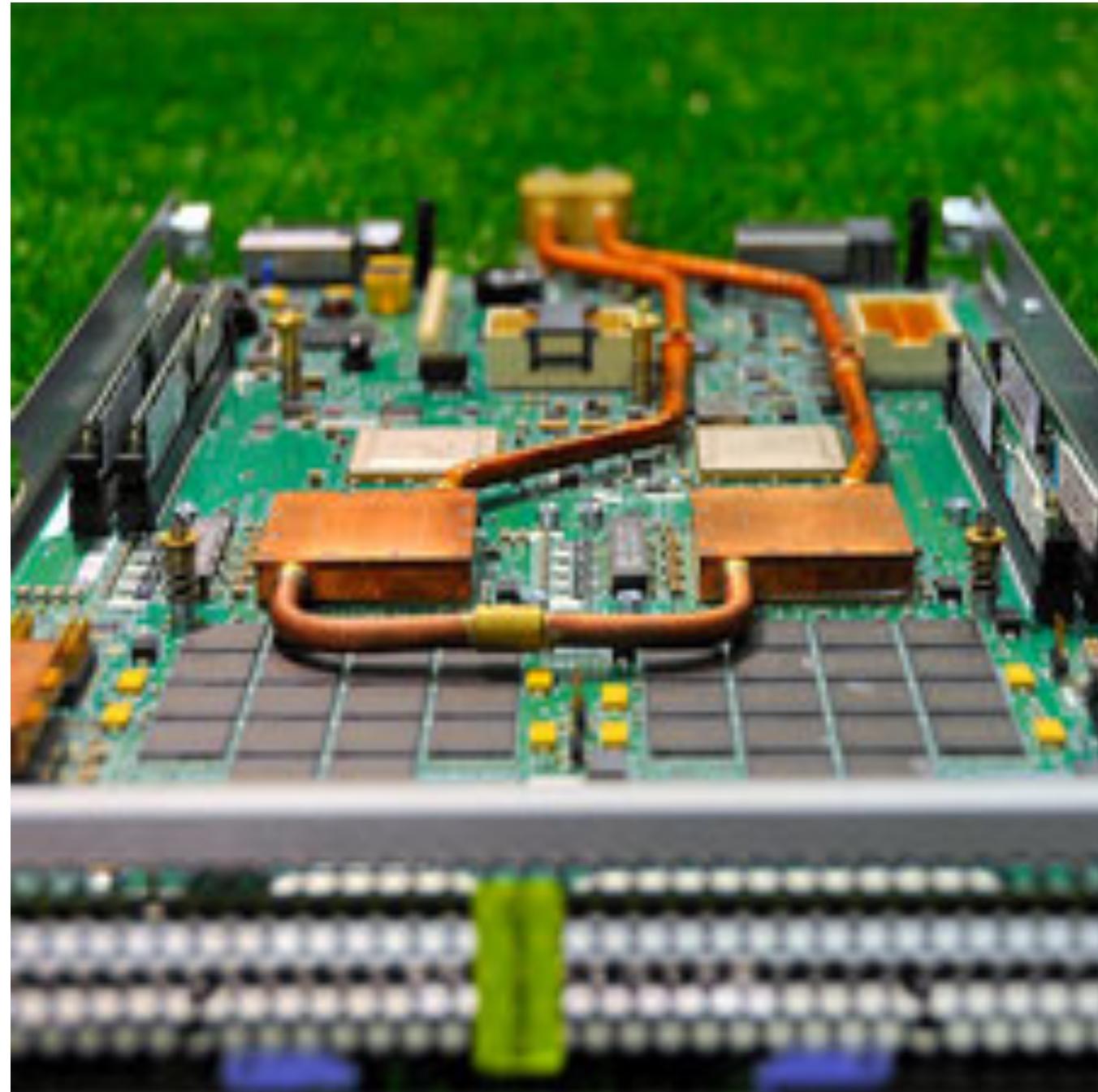
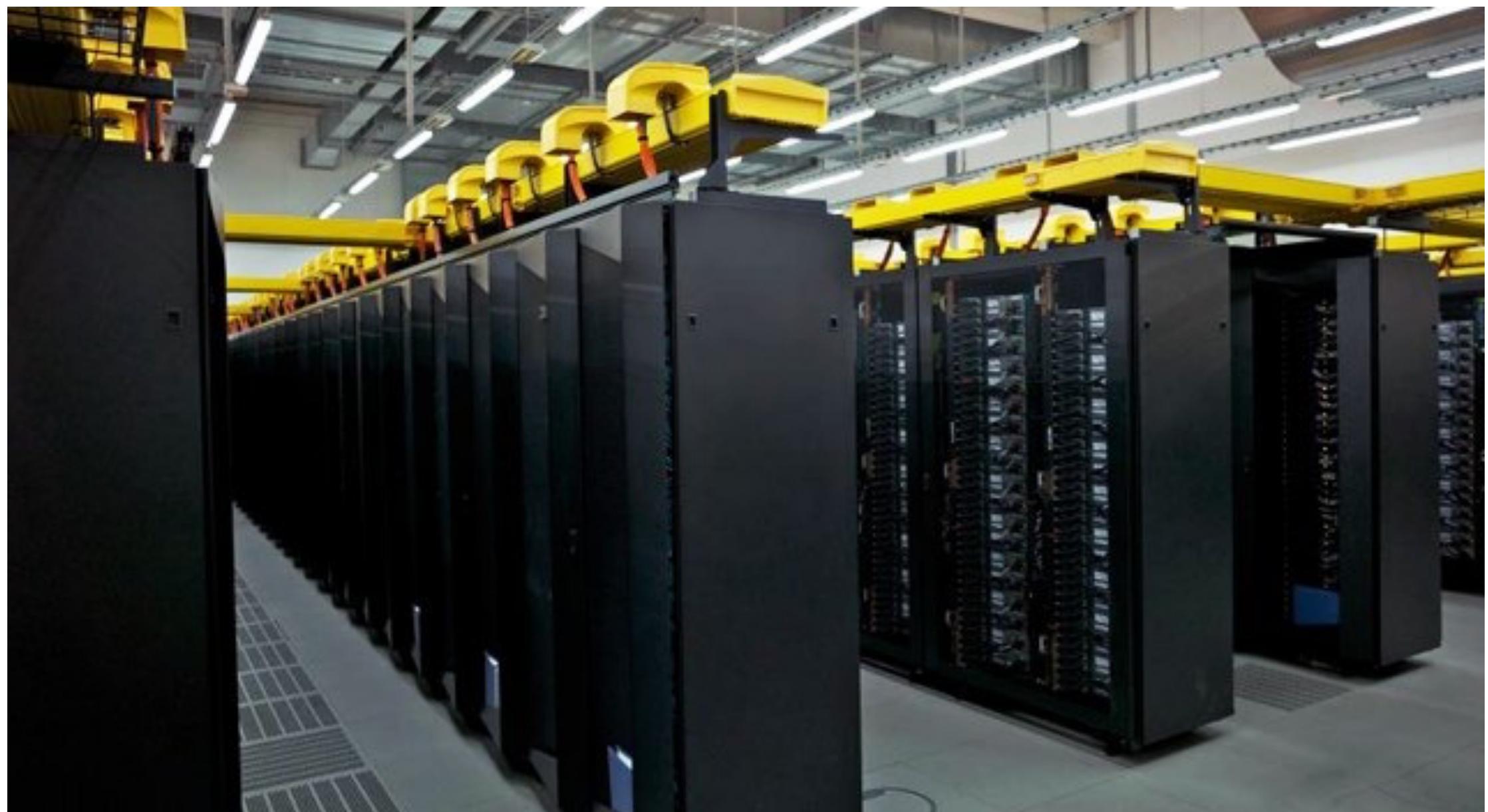
why should we care?



energy saving



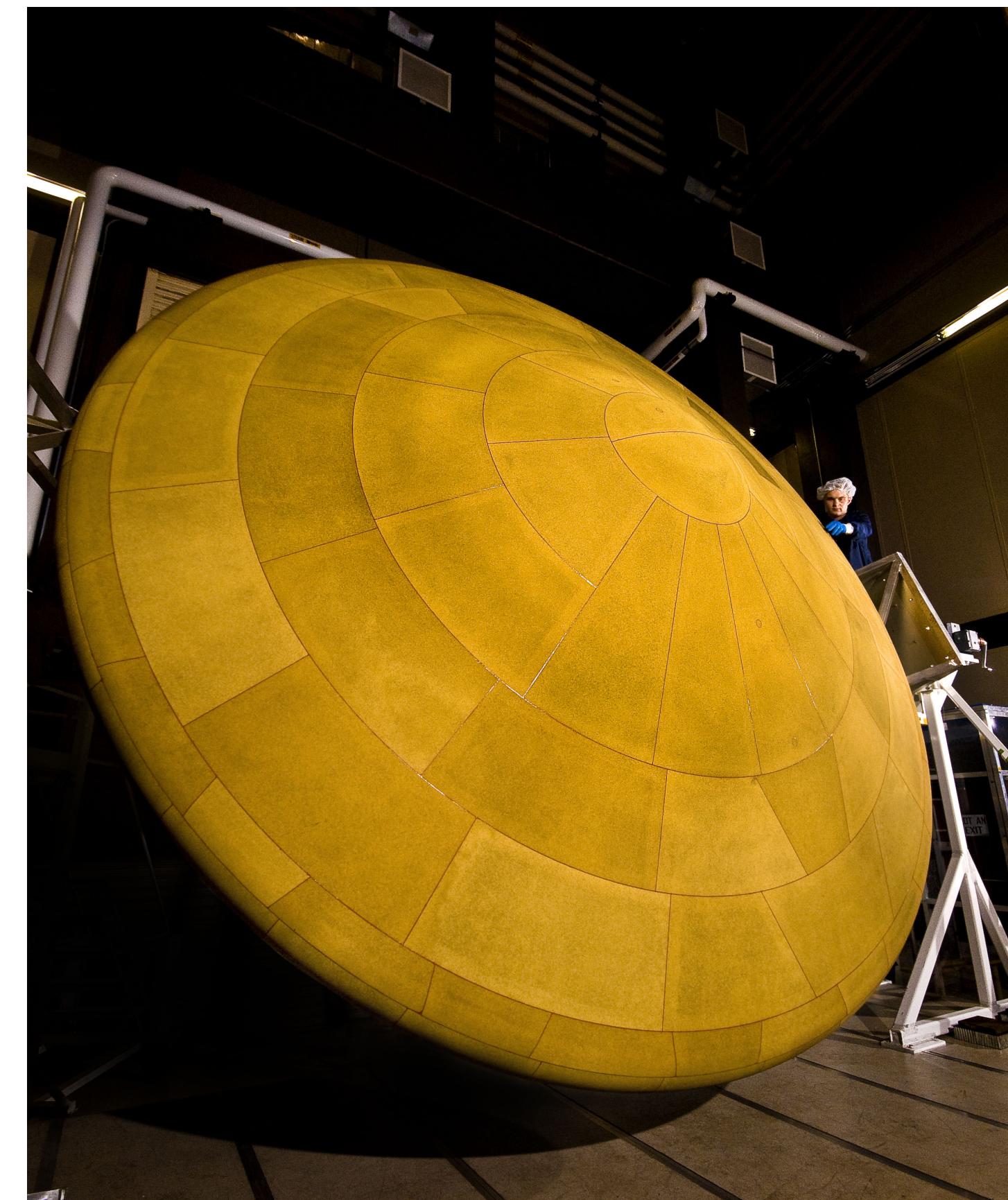
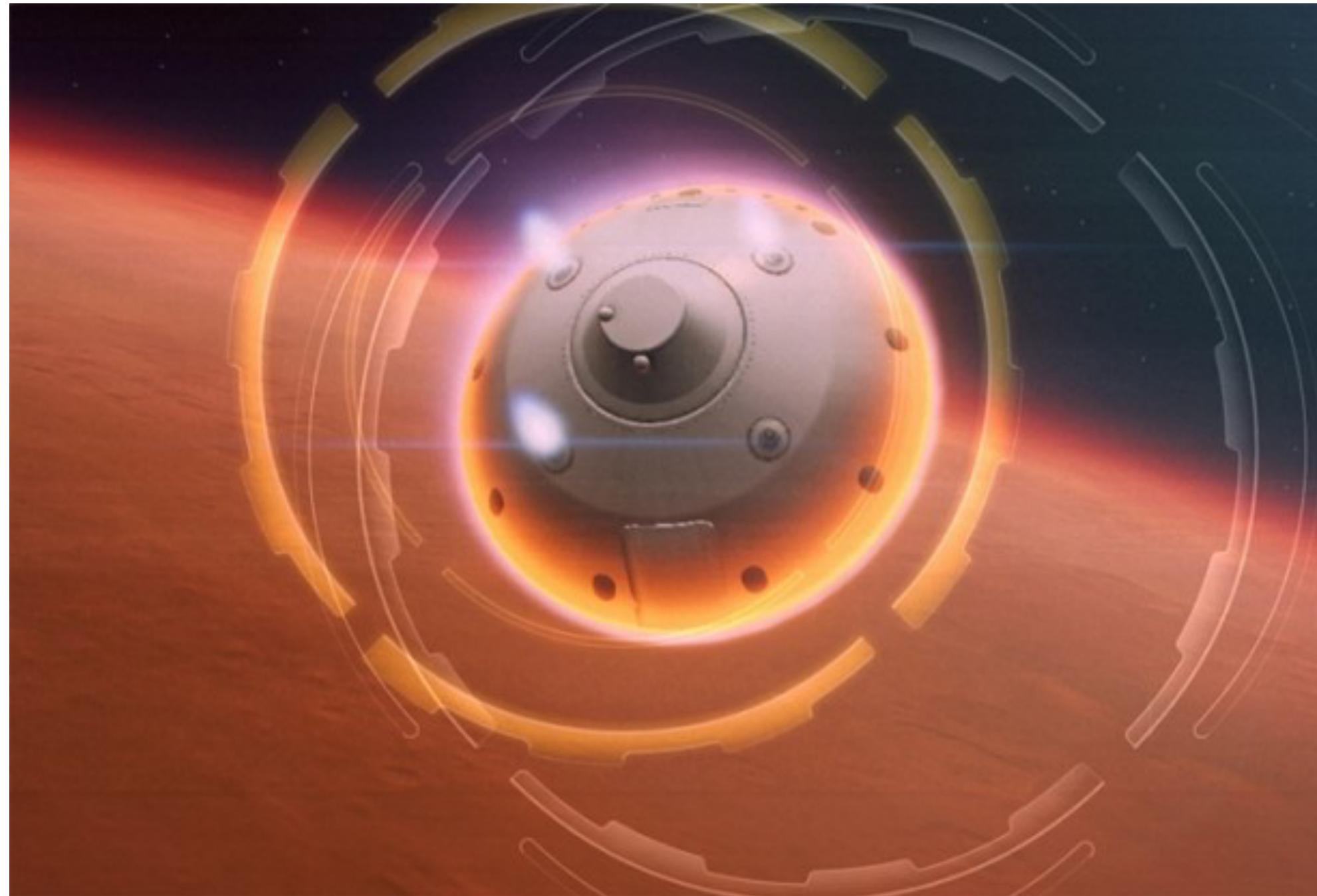
heat dissipation



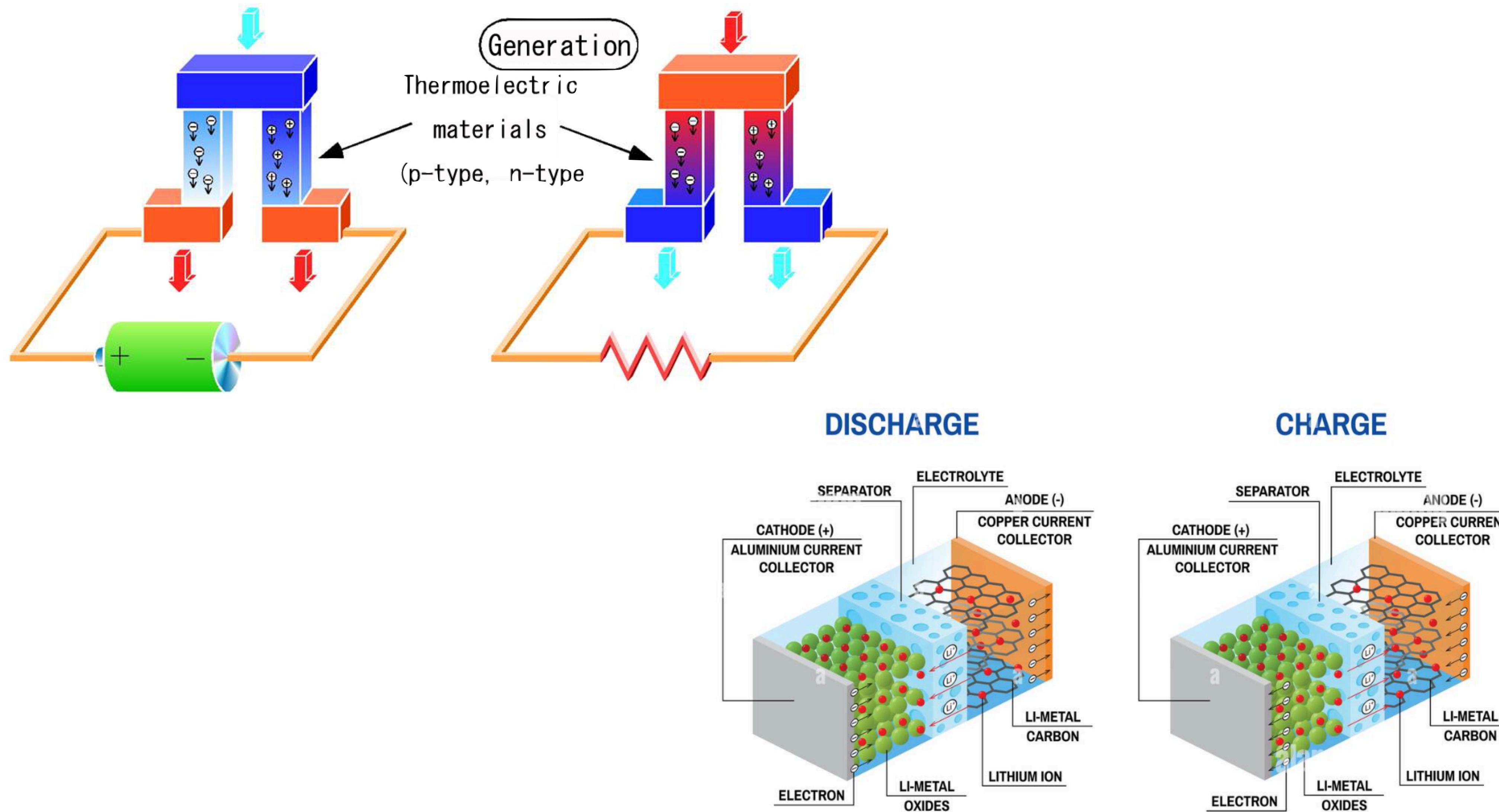
heat shielding



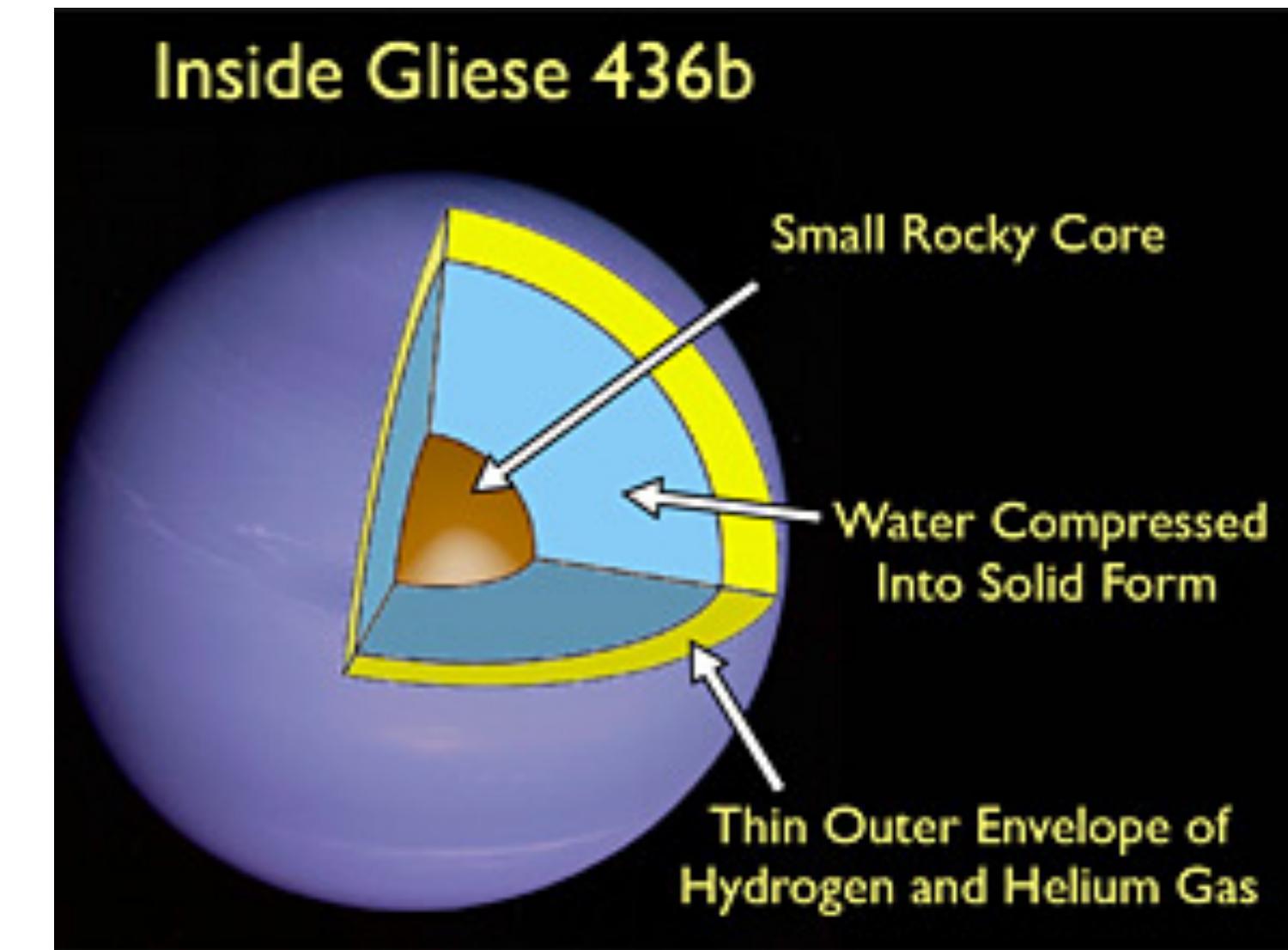
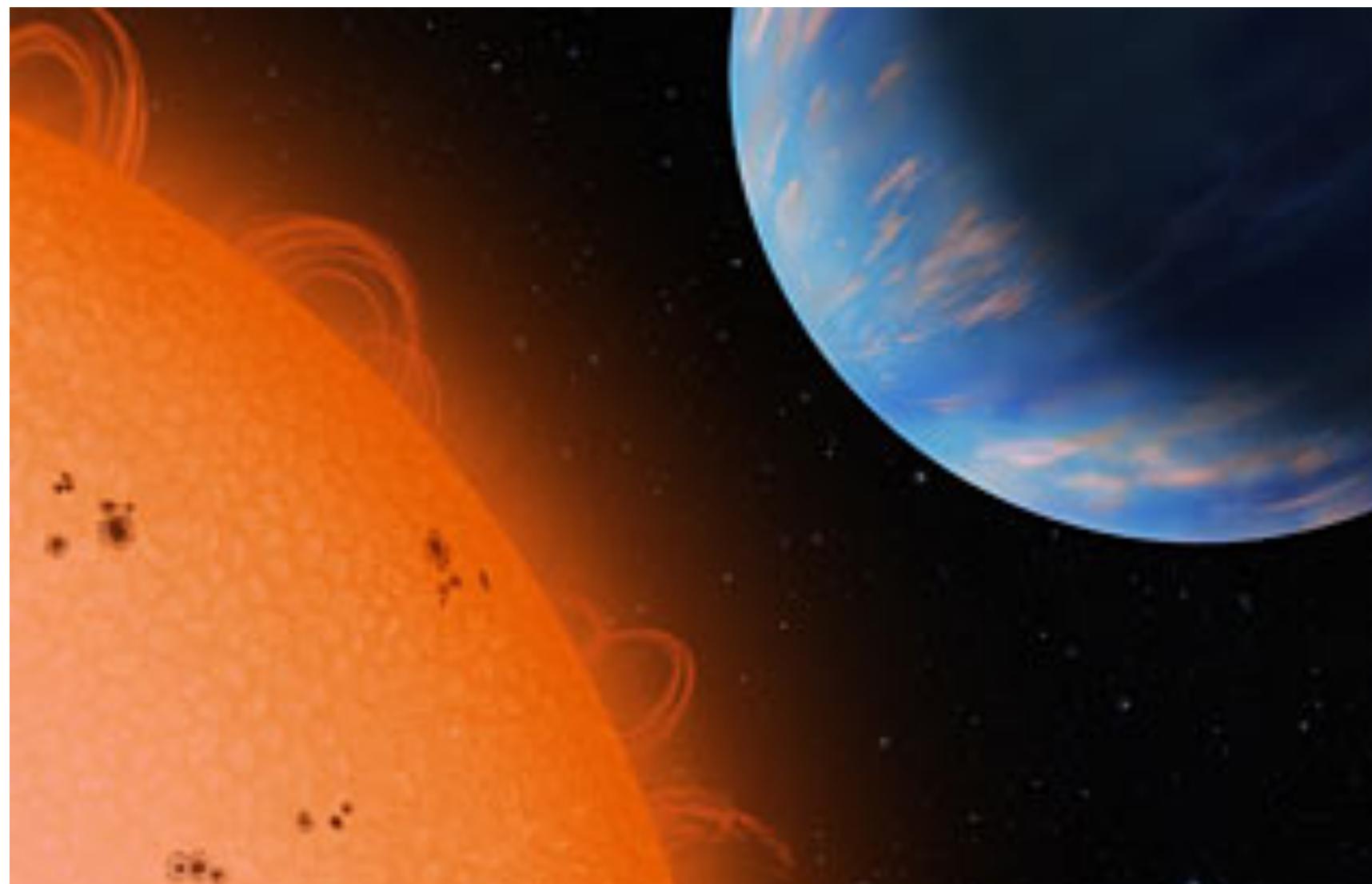
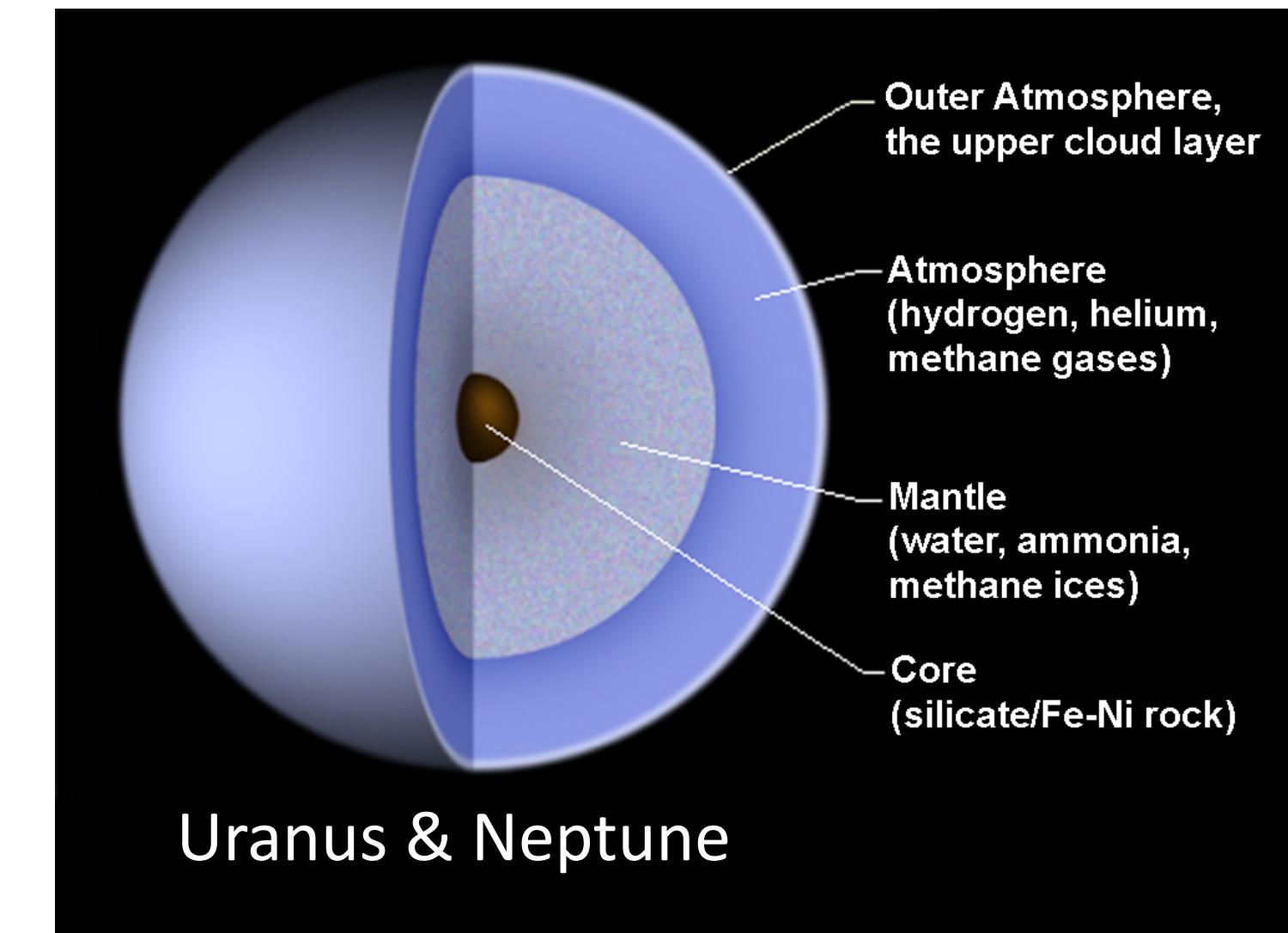
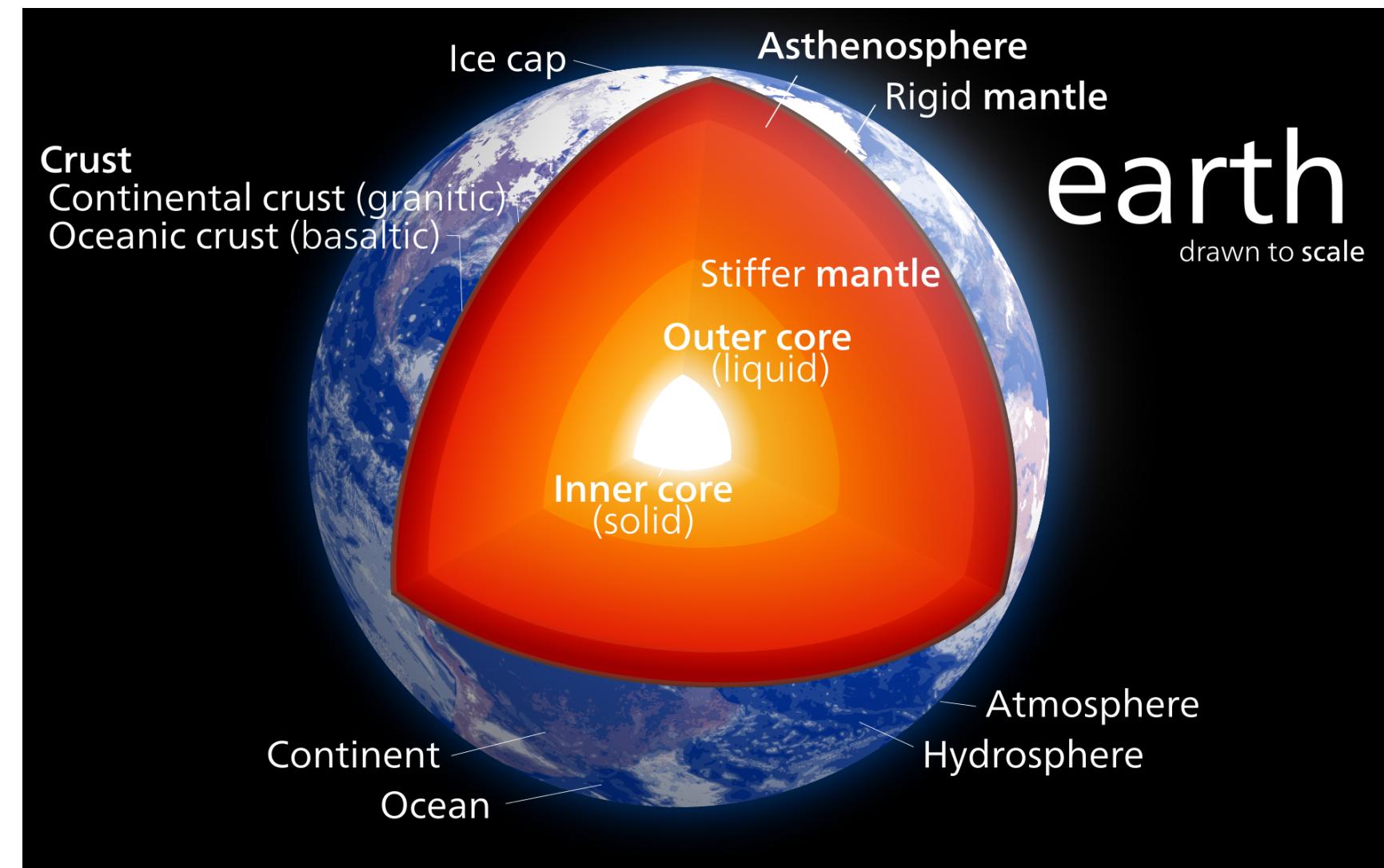
heat shielding



energy conversion & storage



planetary sciences



why should we care?

- energy saving and heat dissipation
- heat shielding
- energy harvesting, scavenging, and storage
- earth and planetary sciences
- ...



why should we care?



- ... because it is important and still poorly understood

the linear-response theory of transport

$$J = \lambda F$$



the linear-response theory of transport

$$\mathbf{J} = \lambda \mathbf{F}$$

charge transport

$$\mathbf{J}_Q = \sum_I q_I \mathbf{V}_I$$

$$\mathbf{F}_Q = -\nabla\phi$$

λ = electric conductivity



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energy transport

$$\mathbf{J}_{\mathcal{E}} = \sum_I e_I \mathbf{V}_I + \frac{1}{2} \sum_{I \neq J} (\mathbf{V}_I \cdot \mathbf{F}_{IJ}) (\mathbf{R}_I - \mathbf{R}_J)$$

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$$\begin{aligned}\sum_I e_I &= E \\ \mathbf{J}_{\mathcal{E}} &= \sum_I e_I \mathbf{V}_I + \frac{1}{2} \sum_{I \neq J} (\mathbf{V}_I \cdot \mathbf{F}_{IJ})(\mathbf{R}_I - \mathbf{R}_J) \\ \mathbf{F}_{\mathcal{E}} &= -\nabla T\end{aligned}$$

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λ = heat conductivity

$$\lambda \propto \int_0^\infty \langle \mathbf{J}(t) \mathbf{J}(0) \rangle dt$$

Green-Kubo



the linear-response theory of transport

$$\mathbf{J} = \lambda \mathcal{F}$$

Green-Kubo

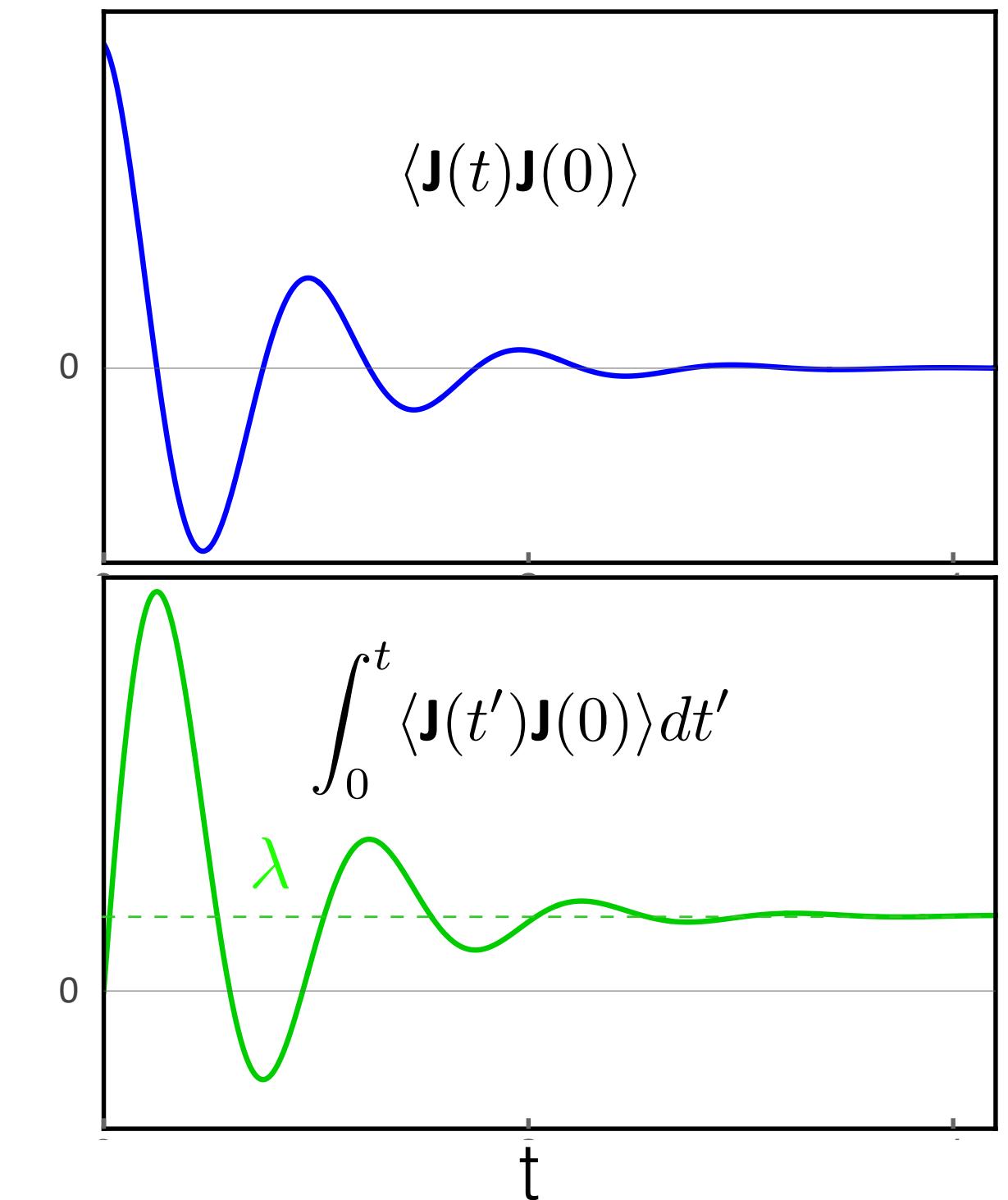
$$\begin{aligned}\lambda &\propto \int_0^\infty \langle \mathbf{J}(t) \mathbf{J}(0) \rangle dt \\ &= \int P^\circ(\Gamma_0) \left[\int_0^\infty \mathbf{J}(\Gamma_t) \mathbf{J}(\Gamma_0) dt \right] d\Gamma_0 \\ &= \lim_{T \rightarrow \infty} \int_0^T \left[\frac{1}{T-t} \int_0^{T-t} \mathbf{J}(\Gamma_{t+t'}) \mathbf{J}(\Gamma_{t'}) dt' \right] dt\end{aligned}$$



the linear-response theory of transport

Green-Kubo

$$J = \lambda F$$
$$\lambda \propto \underbrace{\int_0^\infty \langle \mathbf{J}(t)\mathbf{J}(0) \rangle dt}_{\langle \mathbf{J}^2 \rangle \tau}$$

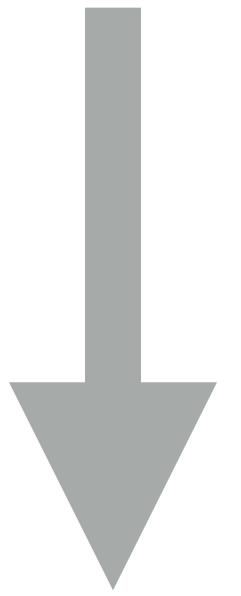


the linear-response theory of transport

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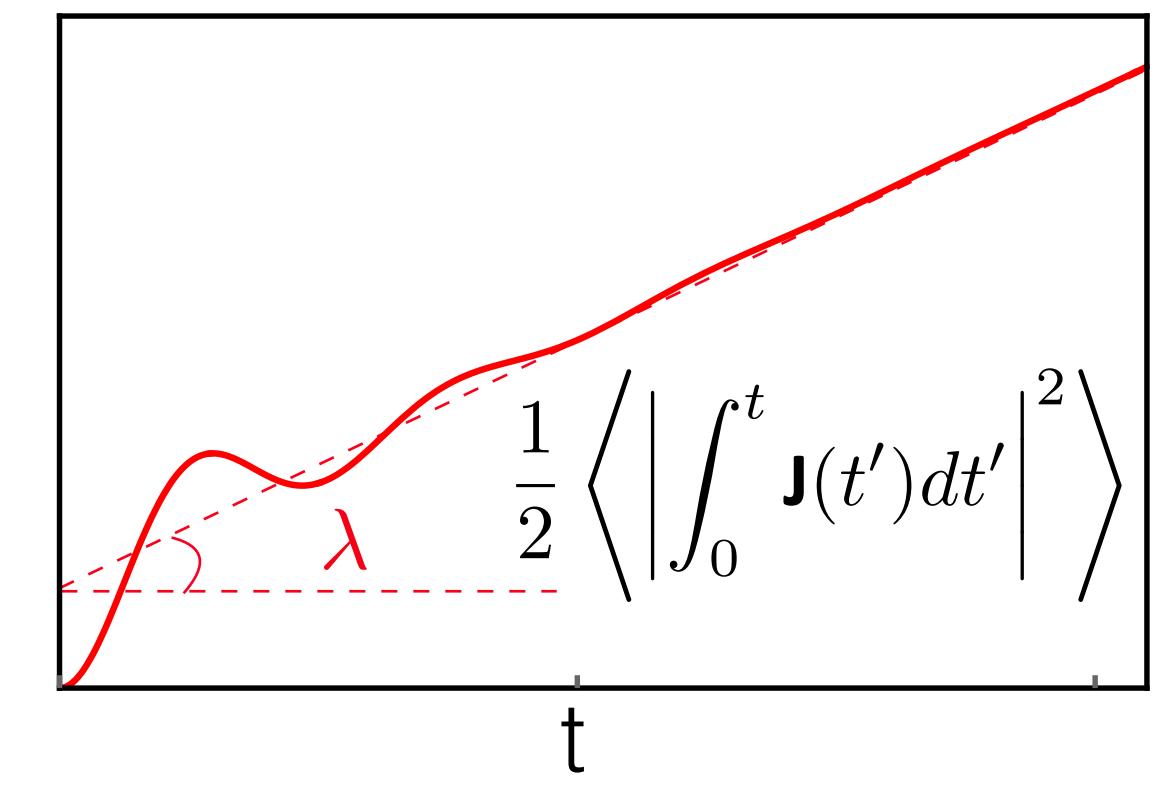
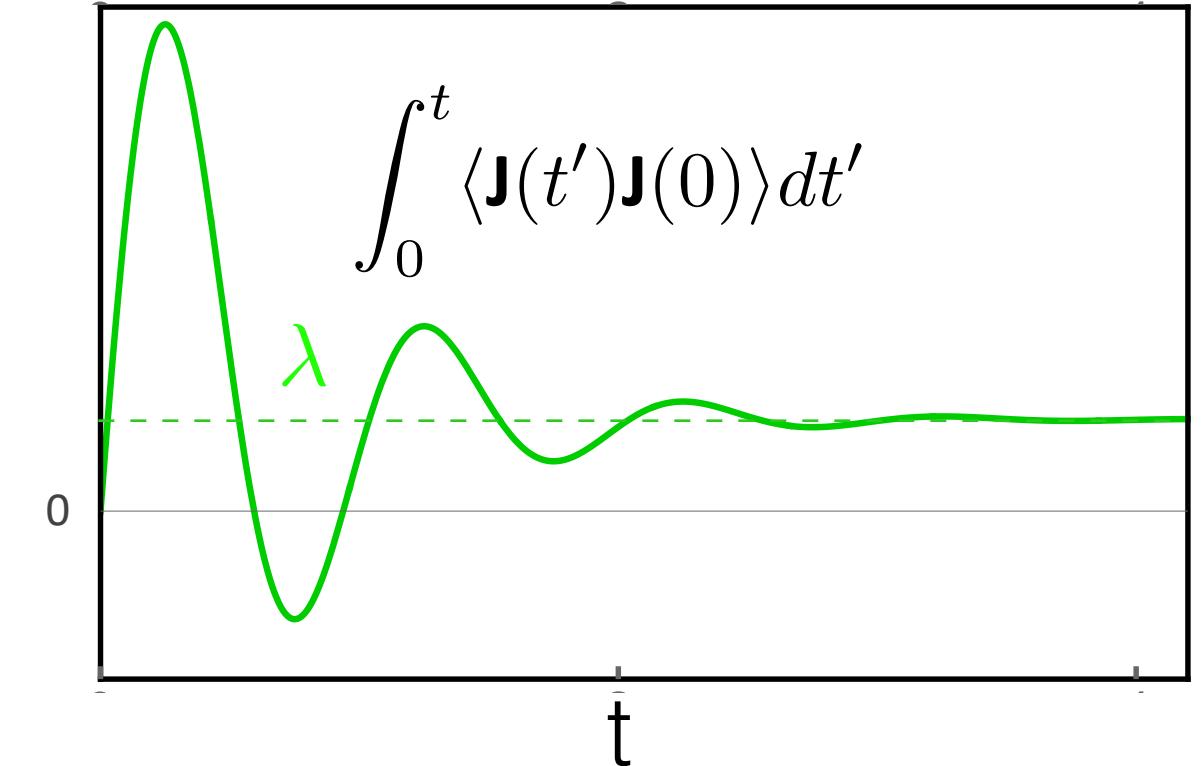
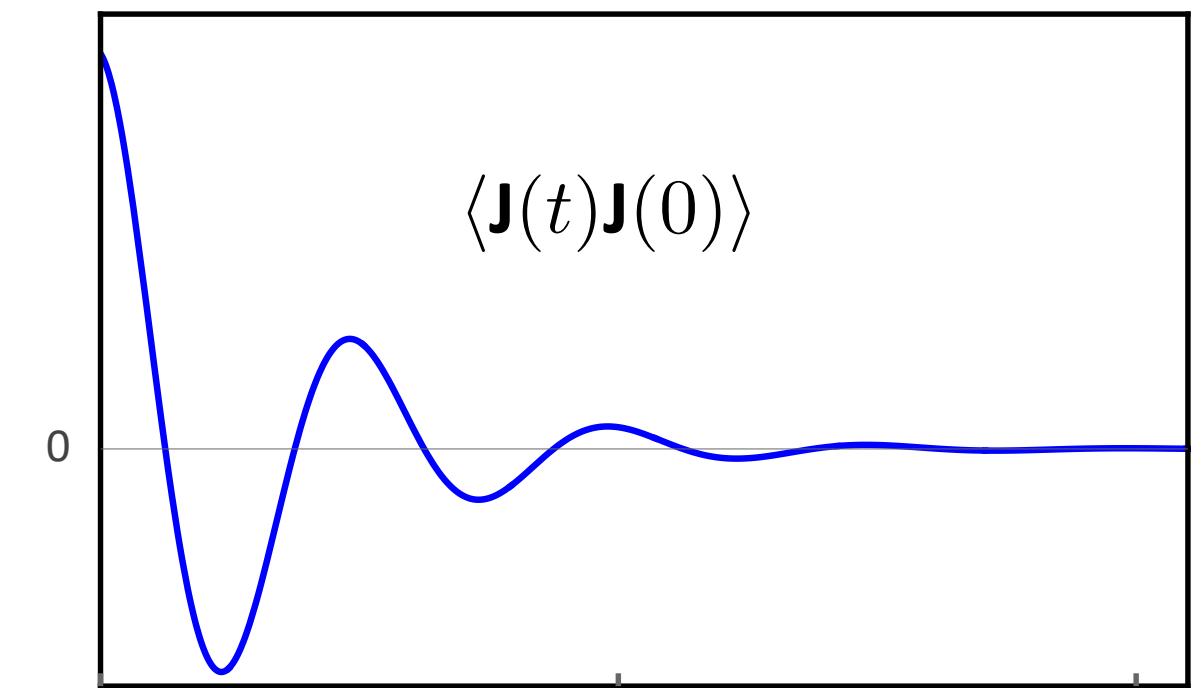
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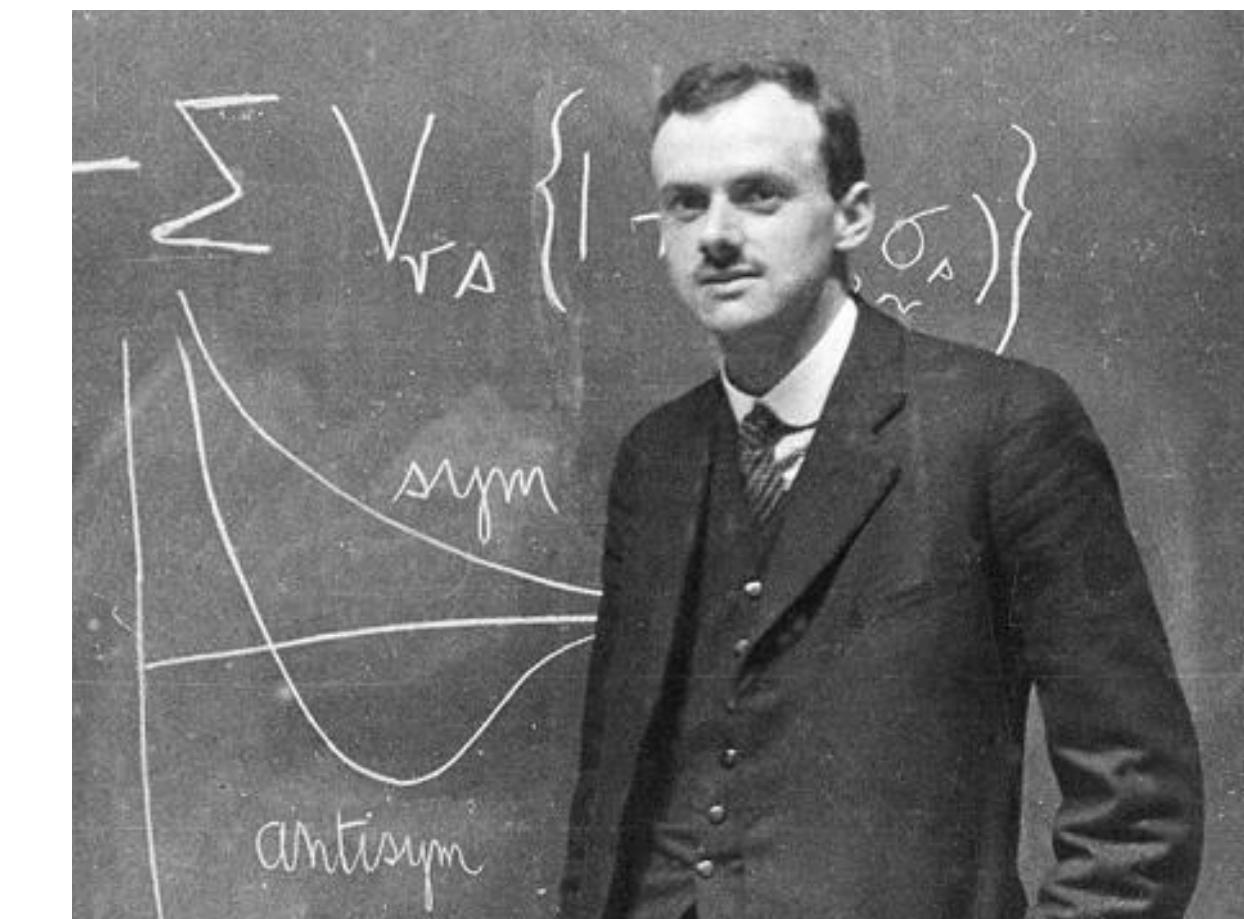
Einstein-Helfand

$$\lambda \propto \lim_{t \rightarrow \infty} \frac{1}{2t} \text{var} \left[\int_0^t \mathbf{J}(t') dt' \right]$$



materials properties from first principles

The underlying physical laws necessary for a large part of physics and all of chemistry are completely known, and the difficulty is only that the exact application of these laws leads to equations much too complicated to be soluble.

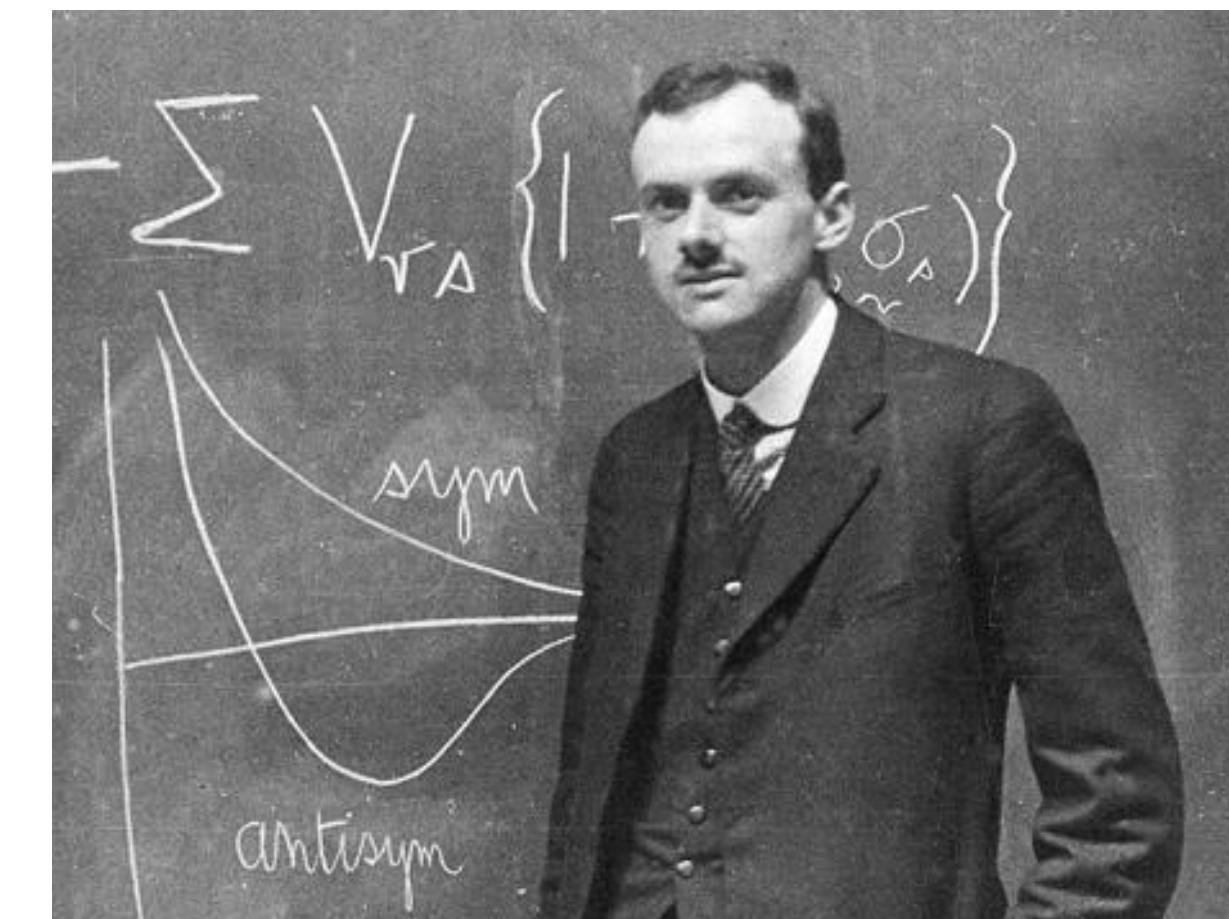


P.A.M. Dirac, 1929



materials properties from first principles

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P.A.M. Dirac, 1929

Dirac's challenge has been answered in [our] field [... using] new physical models [... and] computers.



M.L. Cohen, 2015

hurdles toward an ab initio Green-Kubo theory

$$\mathbf{J}_{\mathcal{E}} = \sum_I e_I \mathbf{v}_I + \frac{1}{2} \sum_{I \neq J} (\mathbf{v}_I \cdot \mathbf{F}_{IJ})(\mathbf{R}_I - \mathbf{R}_J)$$

PRL 104, 208501 (2010)

PHYSICAL REVIEW LETTERS

week ending
21 MAY 2010

Thermal Conductivity of Periclase (MgO) from First Principles

Stephen Stackhouse*

Department of Geological Sciences, University of Michigan, Ann Arbor, Michigan, 48109-1005, USA

Lars Stixrude†

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Bijaya B. Karki‡

*Department of Computer Science, Louisiana State University, Baton Rouge, Louisiana 70803, USA
and Department of Geology and Geophysics, Louisiana State University, Baton Rouge, Louisiana 70803, USA*

sensitive to the form of the potential. The widely used Green-Kubo relation [14] does not serve our purposes, because in first-principles calculations it is impossible to uniquely decompose the total energy into individual contributions from each atom.



hurdles toward an ab initio Green-Kubo theory

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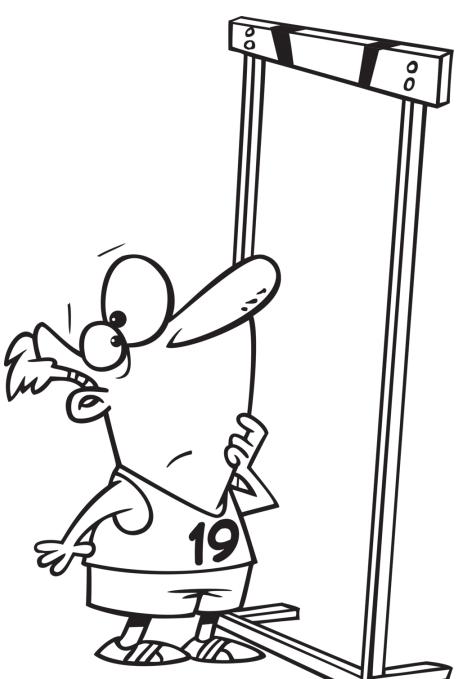
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sensitive to the form of the potential. The widely used Green-Kubo relation [14] does not serve our purposes, because in first-principles calculations it is impossible to uniquely decompose the total energy into individual contributions from each atom.



how come?



how come?



how is it that a formally exact theory of
the electronic ground state cannot predict
all measurable adiabatic properties?

insights from classical mechanics

$$E = \sum_I \epsilon_I(\mathbf{R}, \mathbf{V}) \\ = \text{cnst}$$

$$\epsilon_I(\mathbf{R}, \mathbf{V}) = \frac{1}{2} M_I V_I^2 + \frac{1}{2} \sum_{J \neq I} v(|\mathbf{R}_I - \mathbf{R}_J|)$$

$$\mathbf{J}_e = \sum_I \epsilon_I \mathbf{V}_I + \frac{1}{2} \sum_{I \neq J} (\mathbf{V}_I \cdot \mathbf{F}_{IJ})(\mathbf{R}_I - \mathbf{R}_J)$$



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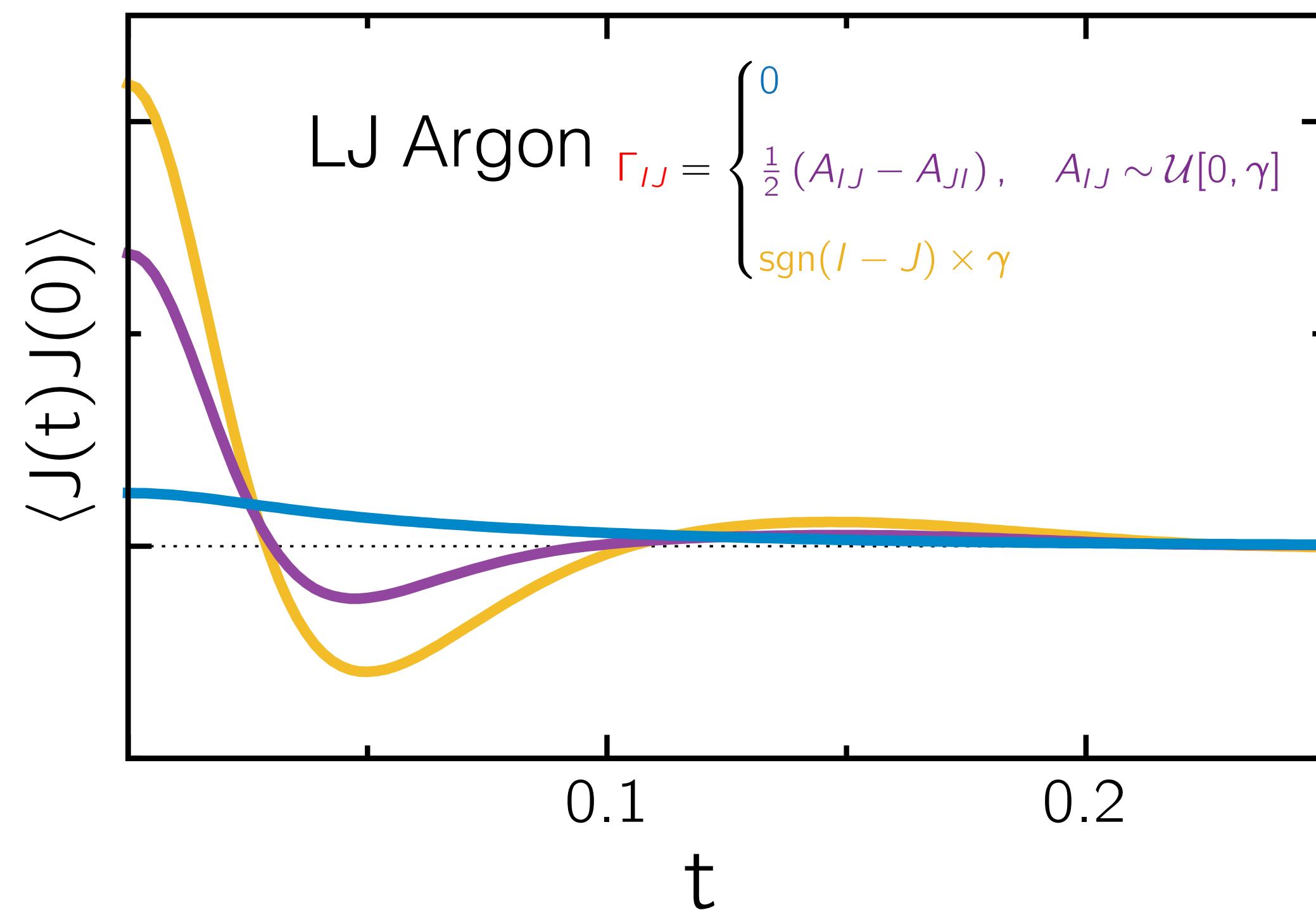
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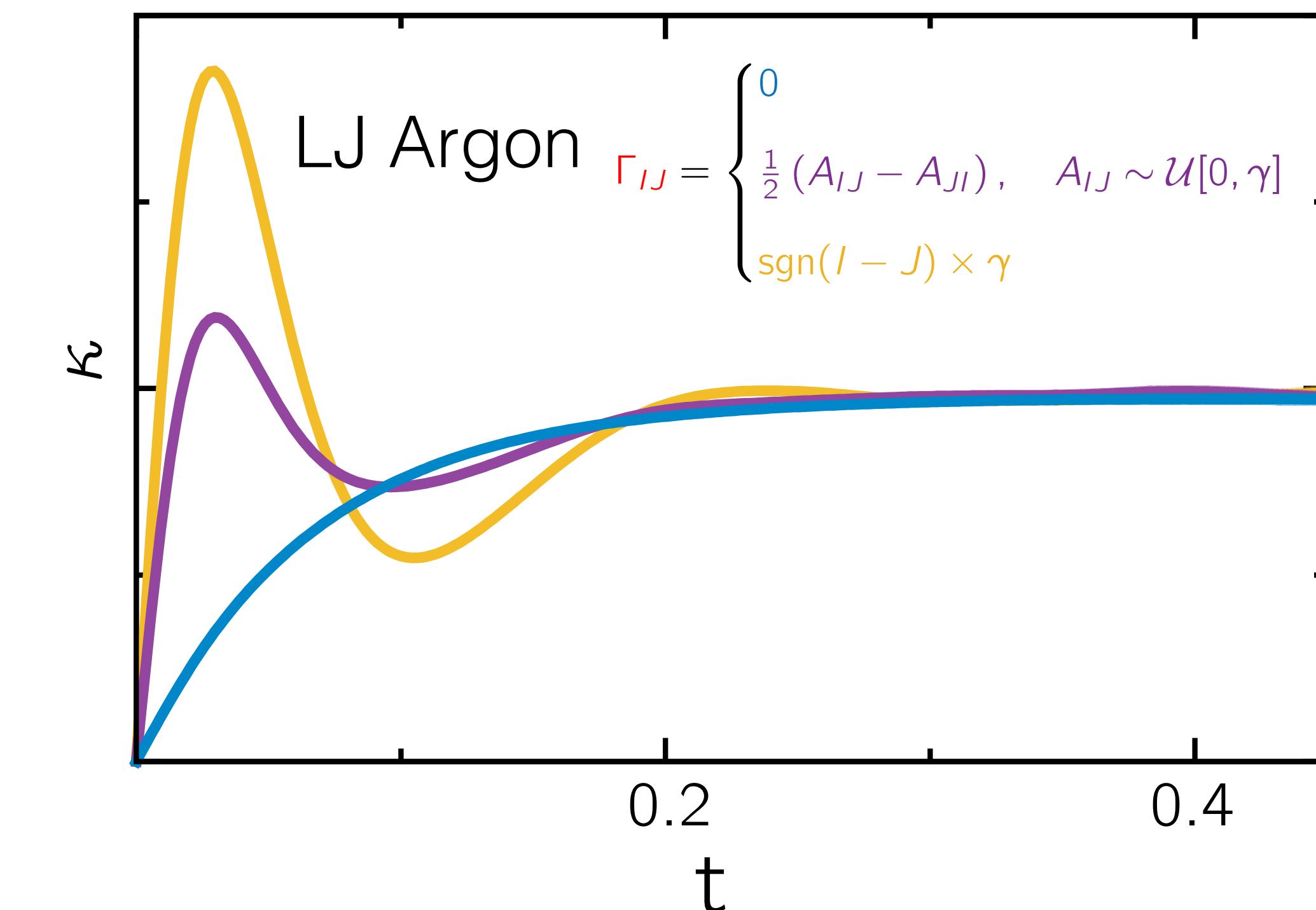
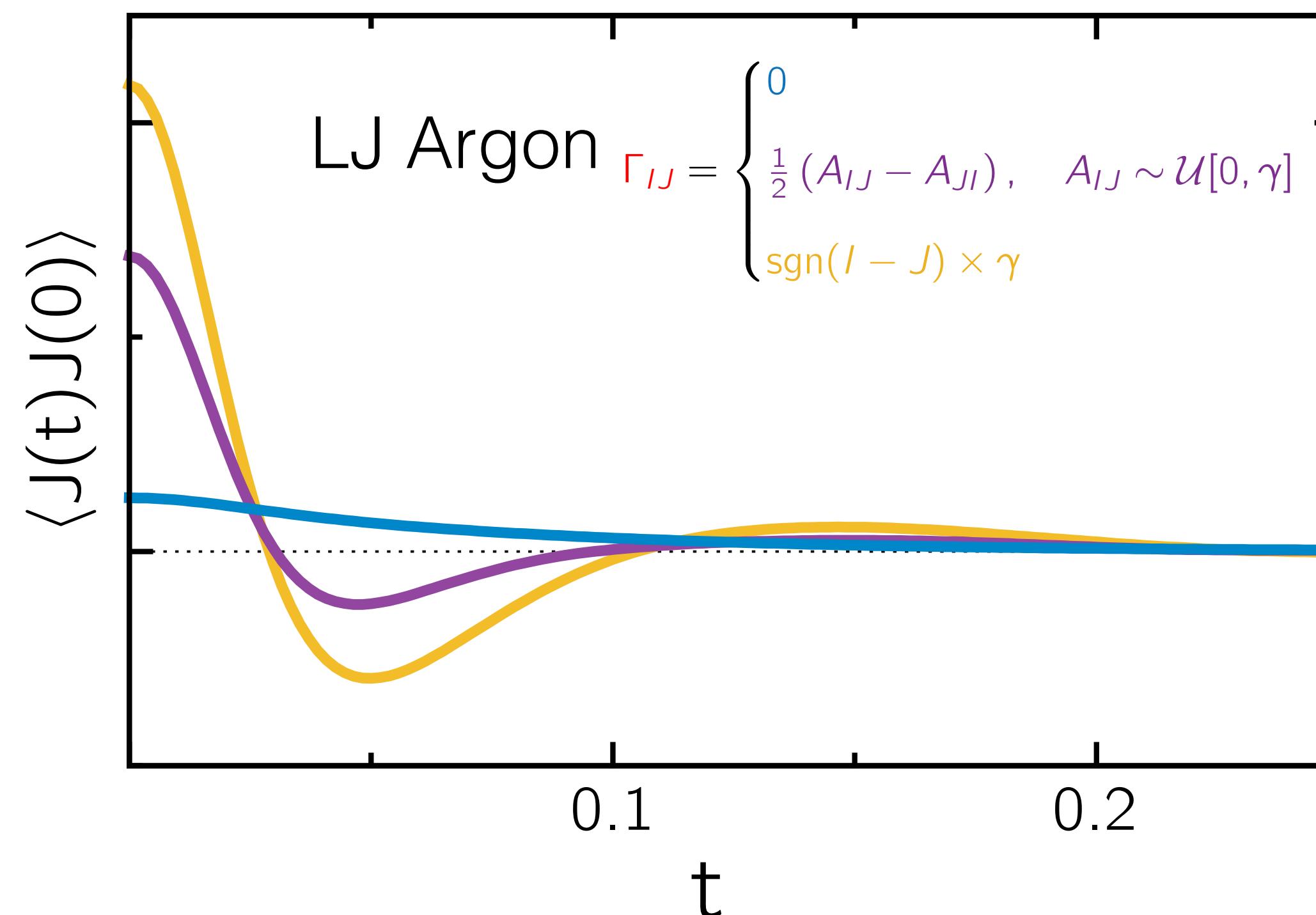
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insights from classical mechanics

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$$\dot{\mathbf{P}} = \frac{d}{dt} \frac{1}{4} \sum_{I \neq J} \Gamma_{IJ} v(|\mathbf{R}_I - \mathbf{R}_J|)(\mathbf{R}_I - \mathbf{R}_I)$$



insights from classical mechanics

$$\mathbf{J}' = \mathbf{J} + \dot{\mathbf{P}}$$



insights from classical mechanics

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$$\kappa \sim \frac{1}{2t} \text{var}[\mathbf{D}(t)] \quad \mathbf{D}(t) = \int_0^t \mathbf{J}(t') dt'$$



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$$\text{var}[\mathbf{D}'(t)] = \text{var}[\mathbf{D}(t)] + \text{var}[\Delta \mathbf{P}(t)] + 2\text{cov}[\mathbf{D}(t) \cdot \Delta \mathbf{P}(t)]$$



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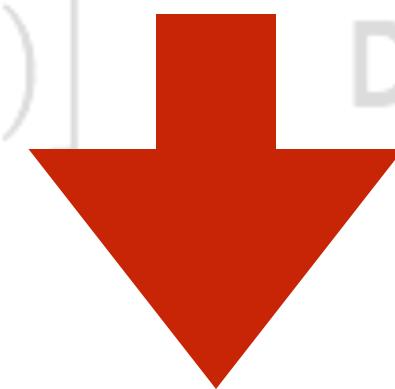
$$\text{var}[\mathbf{D}'(t)] = \underbrace{\text{var}[\mathbf{D}(t)]}_{\mathcal{O}(t)} + \cancel{\text{var}[\Delta \mathbf{P}(t)]} + \cancel{2\text{cov}[\mathbf{D}(t) \cdot \Delta \mathbf{P}(t)]}$$
$$+ \underbrace{2\text{cov}[\mathbf{D}(t) \cdot \mathbf{P}(t)]}_{\mathcal{O}(1)} + \underbrace{2\text{cov}[\mathbf{P}(t) \cdot \mathbf{P}(t)]}_{\mathcal{O}(t^{\frac{1}{2}})}$$



insights from classical mechanics

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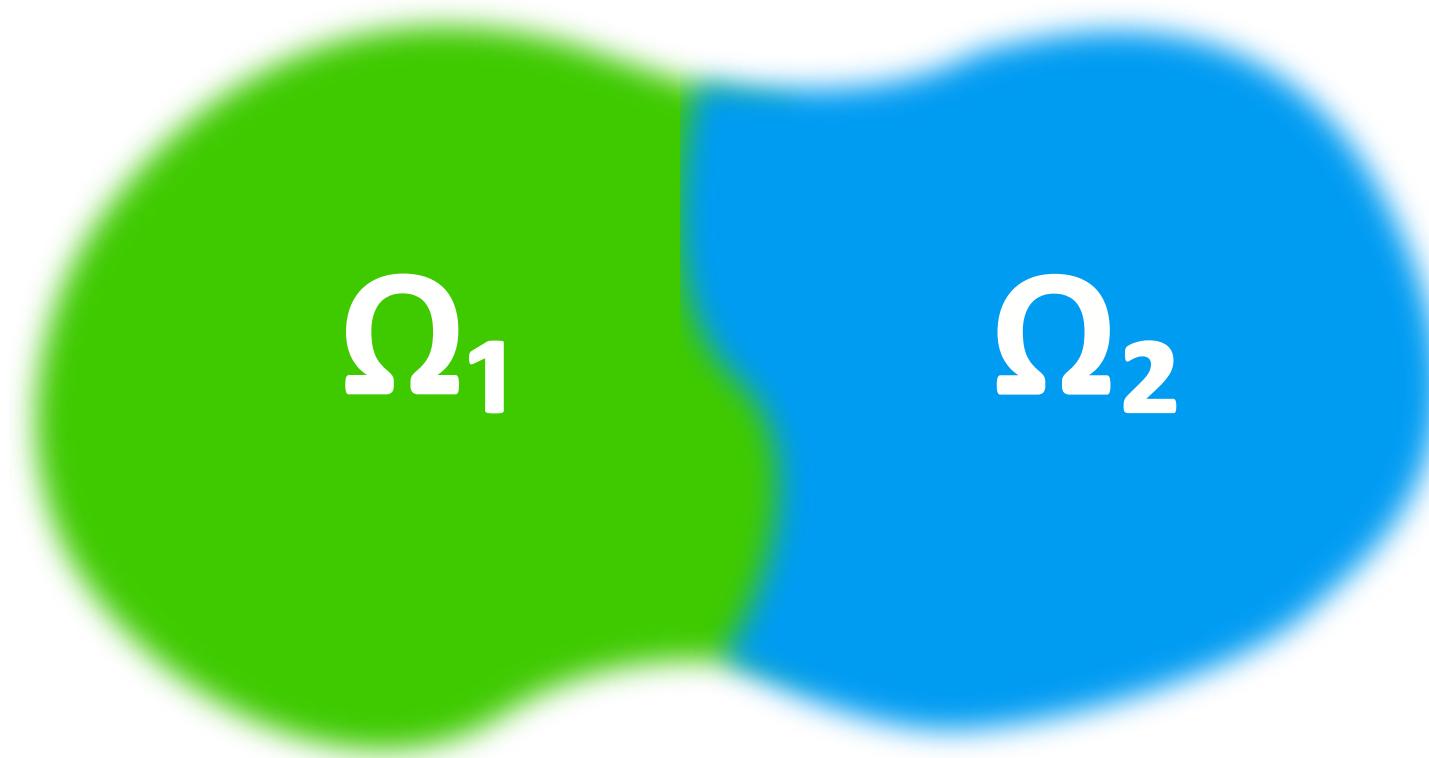
$$\mathbf{D}(t) = \mathbf{D}'(t) + \mathbf{P}(t) - \mathbf{P}(0)$$

$$\kappa' = \kappa$$

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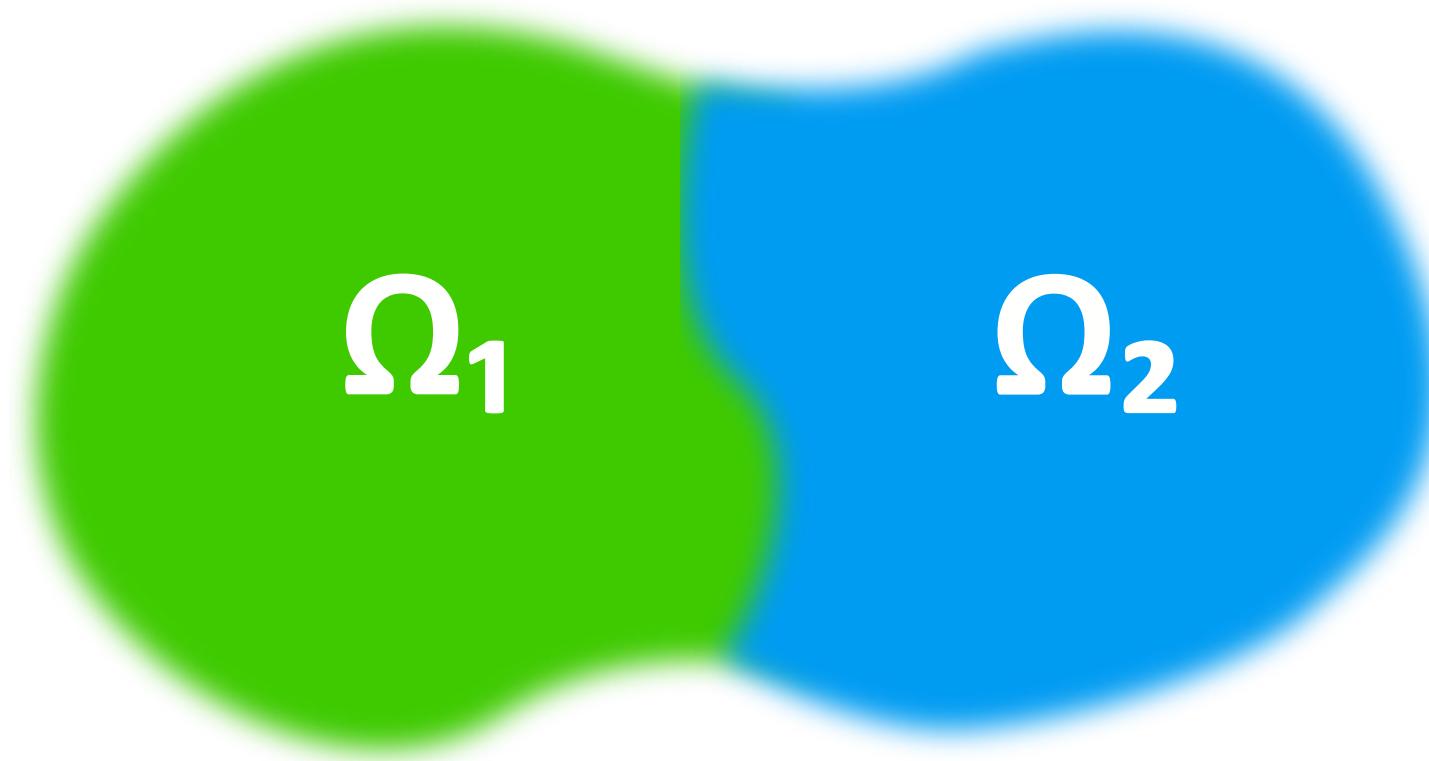


gauge invariance of transport coefficients



$$E[\Omega_1 \cup \Omega_2] = E[\Omega_1] + E[\Omega_2]$$

gauge invariance of transport coefficients



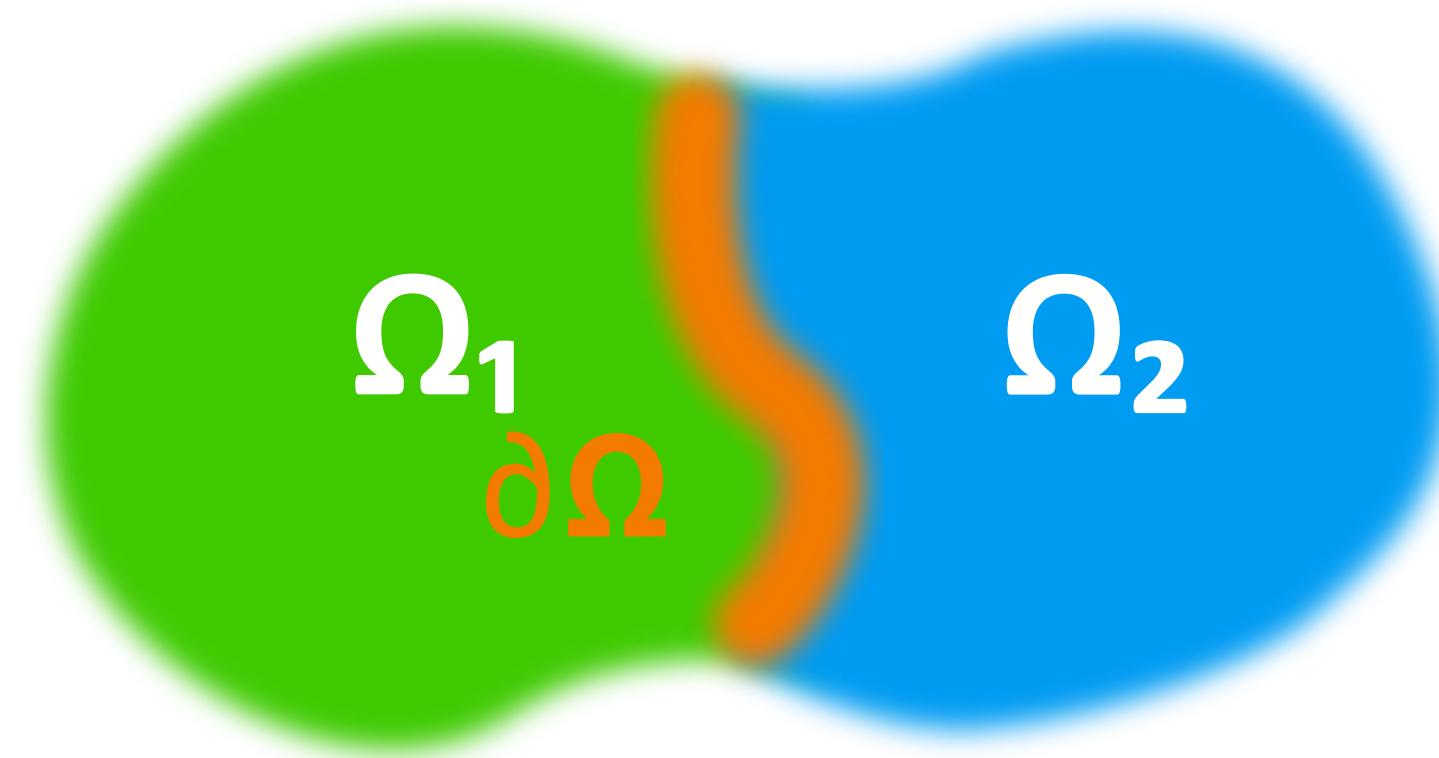
$$E[\Omega_1 \cup \Omega_2] = E[\Omega_1] + E[\Omega_2]$$

extensivity

$$\mathcal{E}[\Omega] = \int_{\Omega} e(\mathbf{r}) d\mathbf{r}$$



gauge invariance of transport coefficients



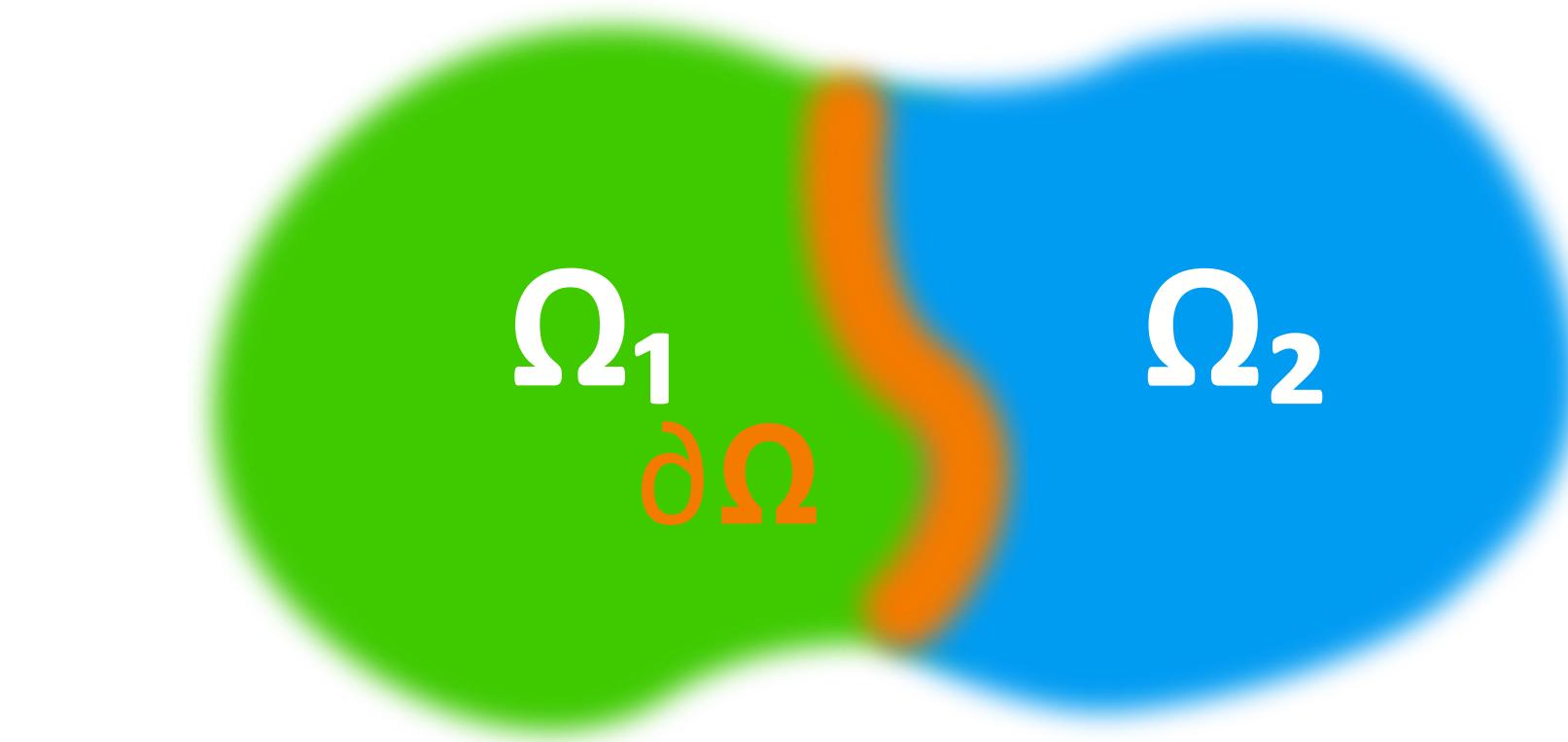
extensivity

$$E[\Omega_1 \cup \Omega_2] = E[\Omega_1] + E[\Omega_2] + W[\partial\Omega]$$

$$\mathcal{E}[\Omega] = \int_{\Omega} e(\mathbf{r}) d\mathbf{r}$$



gauge invariance of transport coefficients



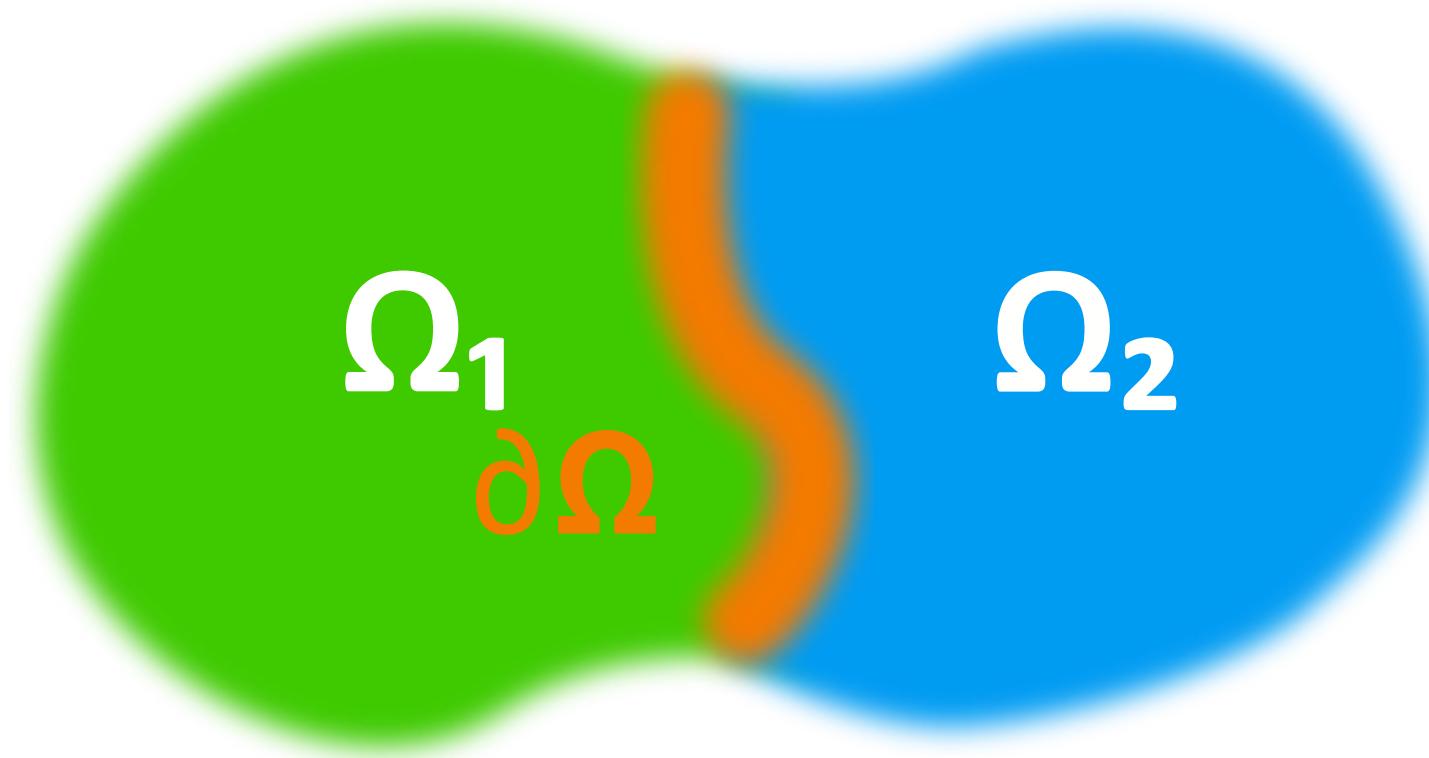
extensivity

$$\begin{aligned} E[\Omega_1 \cup \Omega_2] &= E[\Omega_1] + E[\Omega_2] + W[\partial\Omega] \\ &\stackrel{?}{=} \mathcal{E}[\Omega_1] + \mathcal{E}[\Omega_2] \end{aligned}$$

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gauge invariance of transport coefficients



extensivity

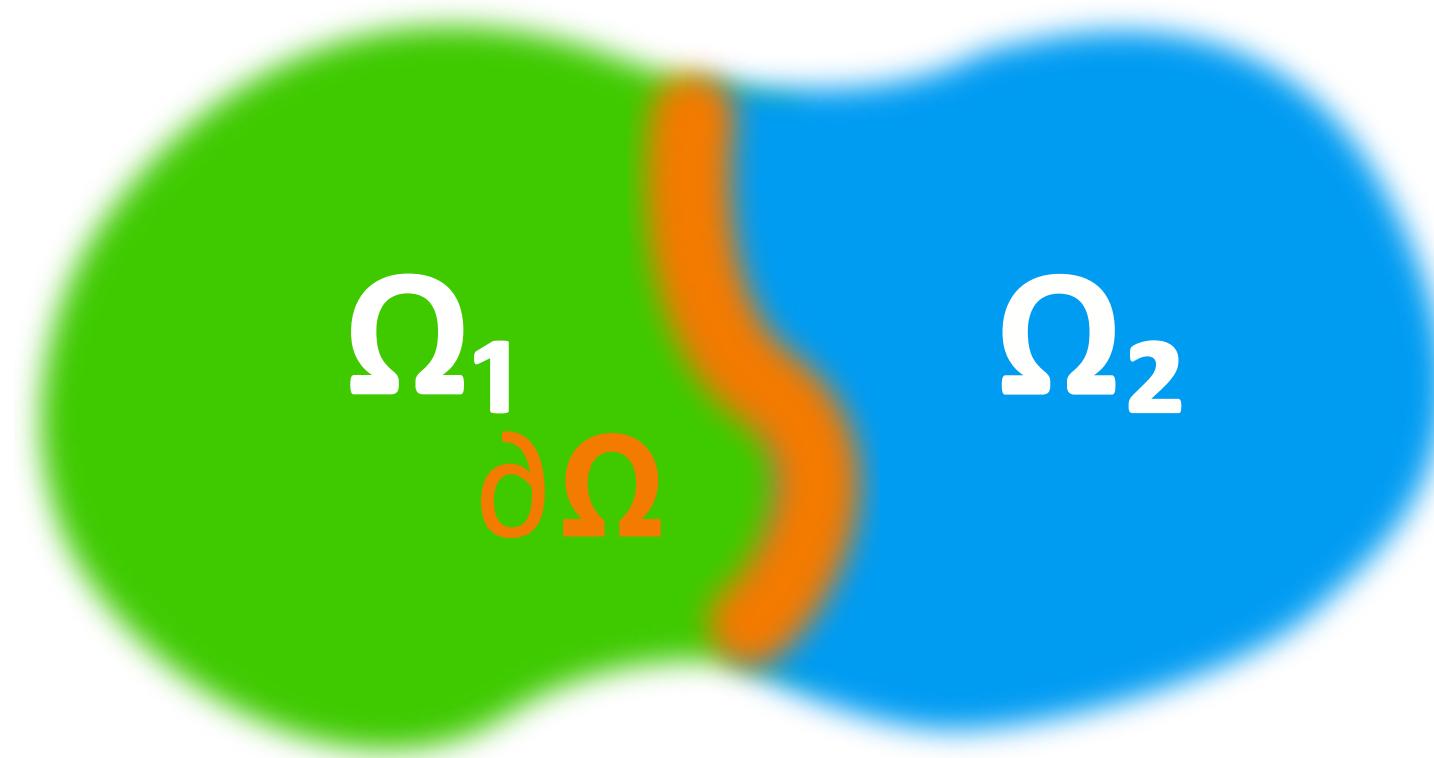
thermodynamic invariance

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gauge invariance of transport coefficients



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thermodynamic invariance

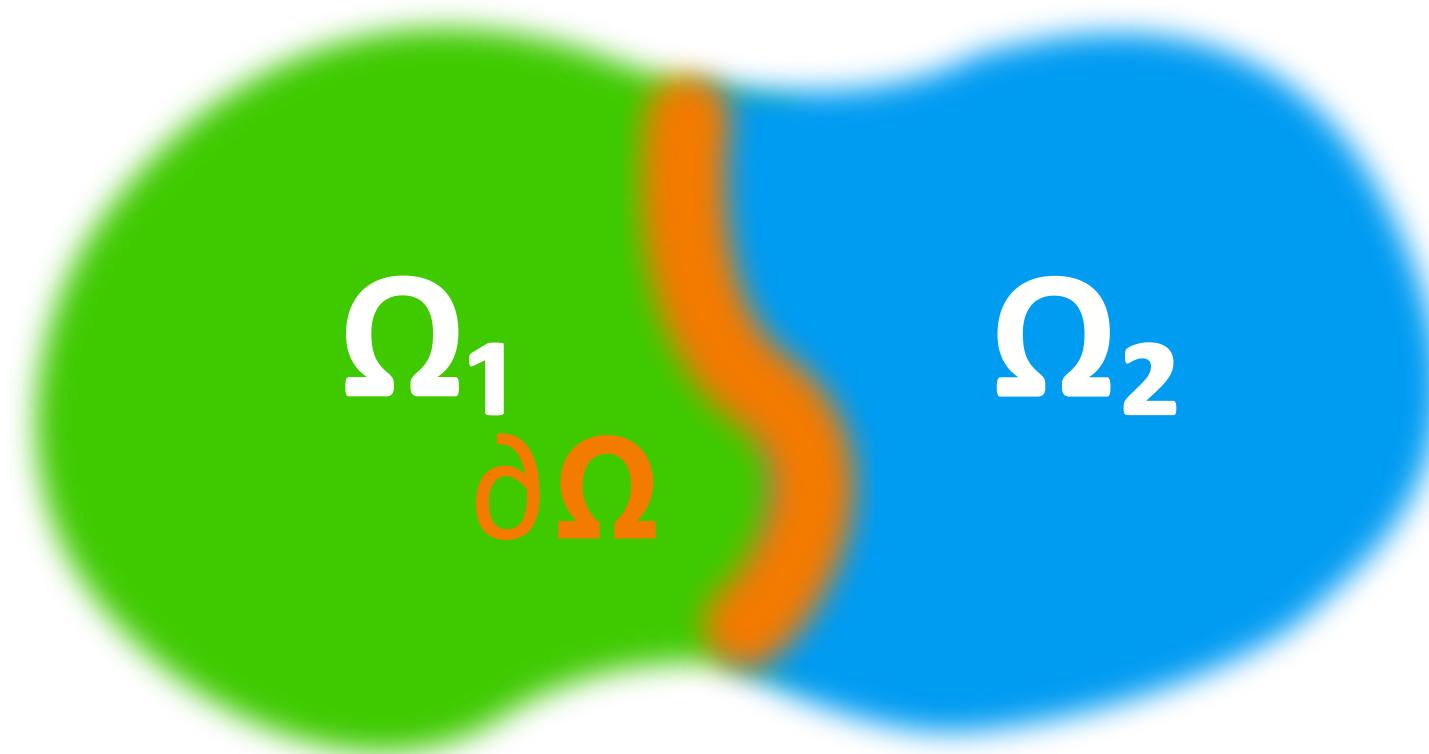
$$\mathcal{E}[\Omega] = \int_{\Omega} e(\mathbf{r}) d\mathbf{r}$$

gauge invariance

$$\mathcal{E}'[\Omega] = \mathcal{E}[\Omega] + \mathcal{O}[\partial\Omega]$$

$$e'(\mathbf{r}) = e(\mathbf{r}) - \nabla \cdot \mathbf{p}(\mathbf{r}) \quad \int_{\Omega} e'(\mathbf{r}) d\mathbf{r} = \int_{\Omega} e(\mathbf{r}) d\mathbf{r} + \mathcal{O}(\partial\Omega)$$

gauge invariance of transport coefficients



extensivity

$$\begin{aligned} E[\Omega_1 \cup \Omega_2] &= E[\Omega_1] + E[\Omega_2] + W[\partial\Omega] \\ &\stackrel{?}{=} \mathcal{E}[\Omega_1] + \mathcal{E}[\Omega_2] \end{aligned}$$

thermodynamic invariance

$$\mathcal{E}[\Omega] = \int_{\Omega} e(\mathbf{r}) d\mathbf{r}$$

gauge invariance

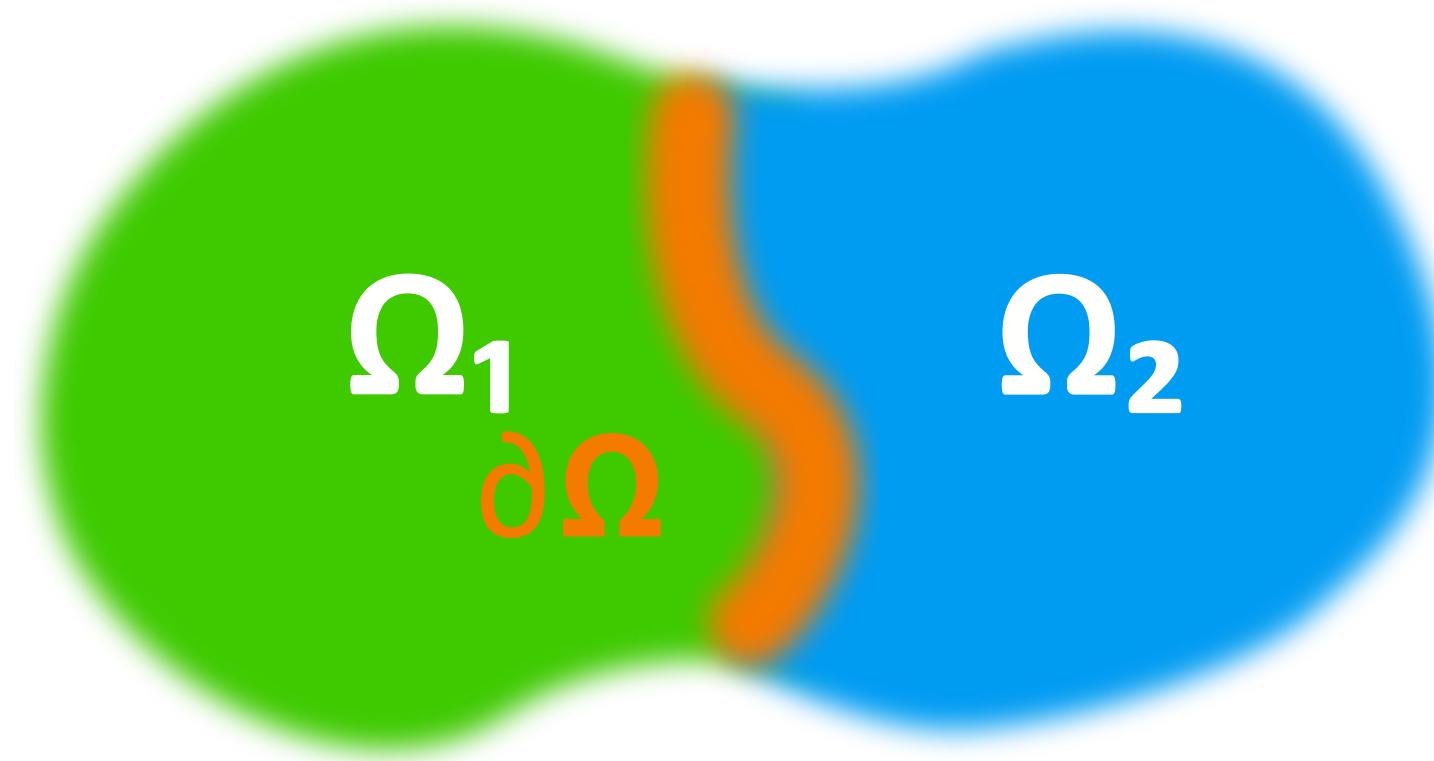
$$\mathcal{E}'[\Omega] = \mathcal{E}[\Omega] + \mathcal{O}[\partial\Omega]$$

conservation

$$\dot{e}(\mathbf{r}, t) = -\nabla \cdot \mathbf{j}(\mathbf{r}, t)$$

$$e'(\mathbf{r}) = e(\mathbf{r}) - \nabla \cdot \mathbf{p}(\mathbf{r}) \quad \int_{\Omega} e'(\mathbf{r}) d\mathbf{r} = \int_{\Omega} e(\mathbf{r}) d\mathbf{r} + \mathcal{O}(\partial\Omega)$$

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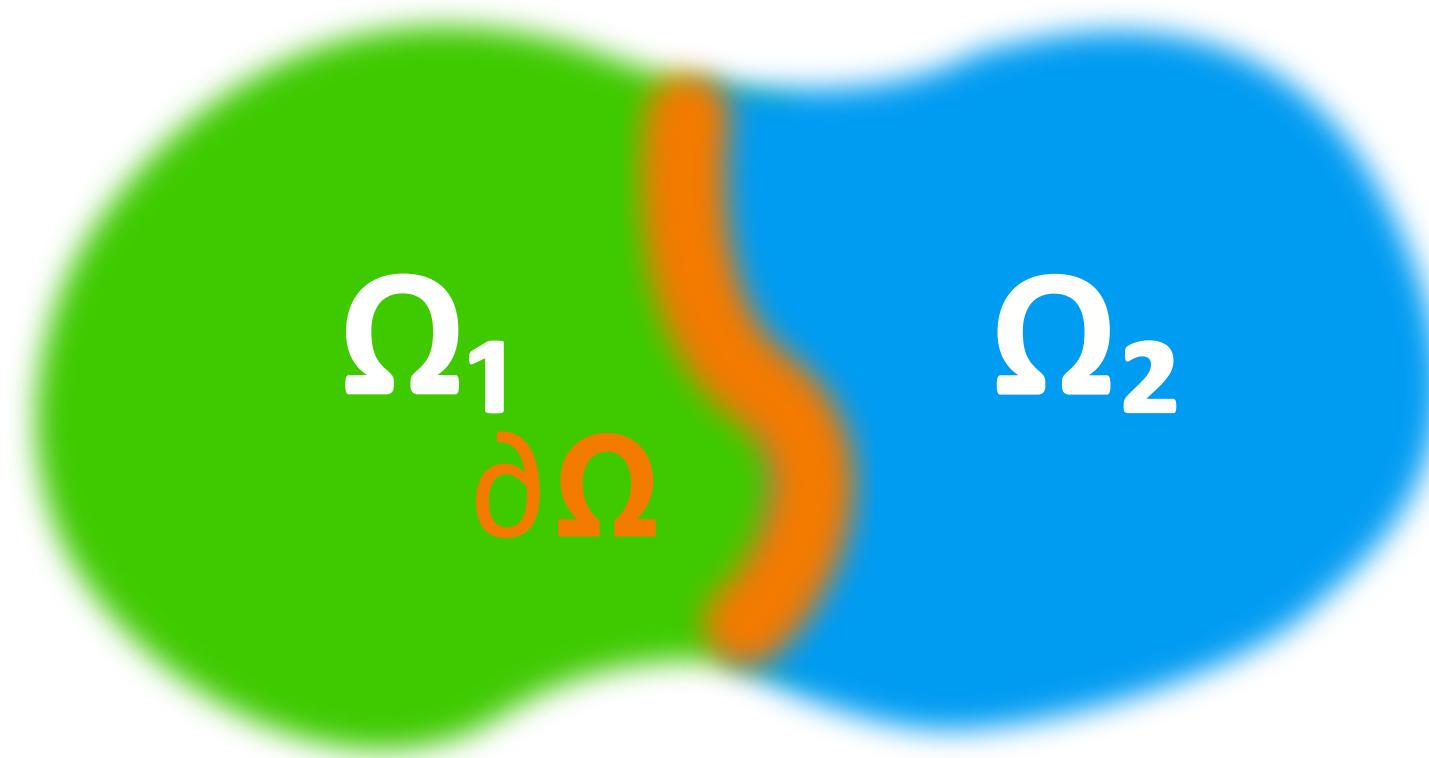
$$\begin{aligned} e'(\mathbf{r}) &= e(\mathbf{r}) - \nabla \cdot \mathbf{p}(\mathbf{r}) & \int_{\Omega} e'(\mathbf{r}) d\mathbf{r} &= \int_{\Omega} e(\mathbf{r}) d\mathbf{r} + \mathcal{O}(\partial\Omega) \\ \mathbf{j}'(\mathbf{r}, t) &= \mathbf{j}(\mathbf{r}, t) + \dot{\mathbf{p}}(\mathbf{r}, t) \end{aligned}$$

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$$\mathbf{J}(t) = \frac{1}{\Omega} \int \mathbf{j}(\mathbf{r}, t) d\mathbf{r}$$

$$\mathbf{P}(t) = \frac{1}{\Omega} \int \mathbf{p}(\mathbf{r}, t) d\mathbf{r}$$

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$$\mathbf{J}'(t) = \mathbf{J}(t) + \dot{\mathbf{P}}(t)$$

conservation

$$\dot{e}(\mathbf{r}, t) = -\nabla \cdot \mathbf{j}(\mathbf{r}, t)$$

gauge invariance of transport coefficients

any two conserved densities that differ by the divergence of a (bounded) vector field are physically equivalent

the corresponding conserved fluxes differ by a total time derivative, and the transport coefficients coincide

gauge invariance
conservation

nature
physics

ARTICLES

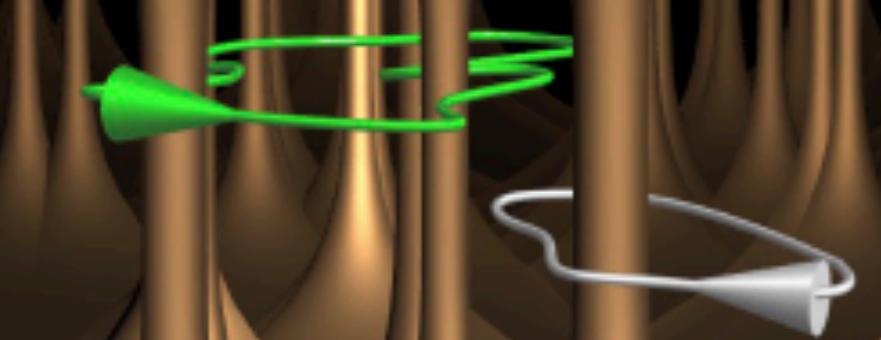
PUBLISHED ONLINE: 19 OCTOBER 2015 | DOI: 10.1038/NPHYS3509

Microscopic theory and quantum simulation of atomic heat transport

Aris Marcolongo¹, Paolo Umari² and Stefano Baroni^{1*}



Topology, oxidation states, and charge transport in ionic conductors



Paolo Pegolo

Scuola Internazionale Superiore di Studi Avanzati - SISSA - Trieste, Italy



SISSA

Mixed-Gen Season 3 – Session 2: Theory and numerical simulation of transport processes in condensed matter

gauge invariance of heat transport

PRL 104, 208501 (2010)

PHYSICAL REVIEW LETTERS

week ending
21 MAY 2010



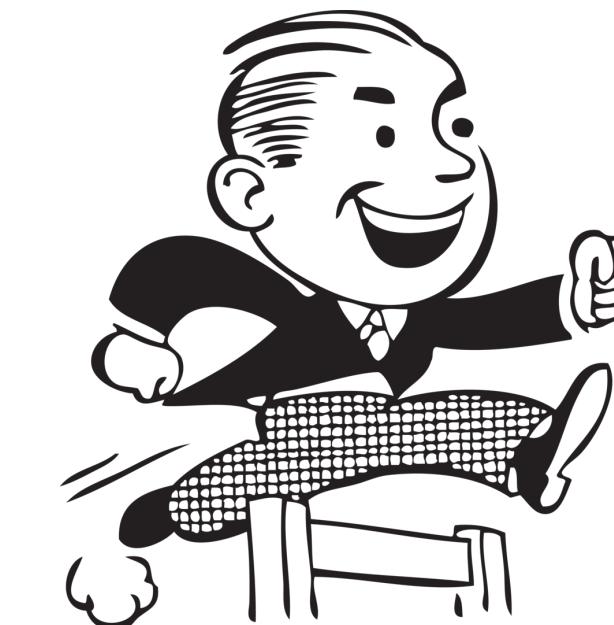
Thermal Conductivity of Periclase (MgO) from First Principles

Stephen Stackhouse, Lars Stixrude, and Bijaya B. Karki

sensitive to the form of the potential. The widely used Green-Kubo relation [14] does not serve our purposes, because in first-principles calculations it is impossible to uniquely decompose the total energy into individual contributions from each atom.

solution:

choose *any* local representation of the energy that integrates to the correct value and whose correlations decay at large distance — the conductivity computed from the resulting current will be *independent* of the chosen representation.



*hurdles toward an *ab initio* Green-Kubo theory*

PRL 104, 208501 (2010)

PHYSICAL REVIEW LETTERS

week ending
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PRL 118, 175901 (2017)

PHYSICAL REVIEW LETTERS

week ending
28 APRIL 2017

***Ab Initio* Green-Kubo Approach for the Thermal Conductivity of Solids**

Christian Carbogno, Rampi Ramprasad, and Matthias Scheffler

ulations: Because of the limited time scales accessible in aiMD runs, thermodynamic fluctuations dominate the HFACF, which in turn prevents a reliable and numerically stable assessment of the thermal conductivity via Eq. (2).



spectral analysis

$$\begin{aligned} J &= \int_V j(r) d\mathbf{r} \\ &= \sum_i \int_{V_i} j(r) d\mathbf{r} \end{aligned}$$



spectral analysis

$$\begin{aligned} \mathbf{J} &= \int_V \mathbf{j}(\mathbf{r}) d\mathbf{r} \\ &= \sum_i \int_{V_i} \mathbf{j}(\mathbf{r}) d\mathbf{r} \end{aligned}$$

if $\langle \mathbf{j}(\mathbf{r})\mathbf{j}(\mathbf{r}') \rangle$ is short-range, $\int_{V_i} \mathbf{j}(\mathbf{r}) d\mathbf{r}$ and $\int_{V_j} \mathbf{j}(\mathbf{r}) d\mathbf{r}$ for $i \neq j$ are independent stochastic variables and, by the central-limit theorem,

$\mathbf{J}(t)$ is a Gaussian process



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stationarity implies:

$$\langle \tilde{\mathbf{J}}_T(\omega) \tilde{\mathbf{J}}_T(-\omega') \rangle \sim \frac{1}{T} \quad \text{for } \omega \neq \omega'$$



spectral analysis

$$\begin{aligned}\lambda &= \int_0^\infty \langle \mathbf{J}(t) \mathbf{J}(0) \rangle dt \quad S(\omega) \\ &= \frac{1}{2} \left. \int_{-\infty}^\infty \langle \mathbf{J}(t) \mathbf{J}(0) \rangle e^{i\omega t} dt \right|_{\omega=0} \\ &= \frac{1}{2} S(0)\end{aligned}$$



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Wiener-Kintchine theorem

$$S(\omega) = \lim_{T \rightarrow \infty} \frac{1}{T} \left\langle |\tilde{\mathbf{J}}_T(\omega)|^2 \right\rangle$$
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 &\sim \frac{1}{2} S(\omega_k) \hat{\chi}_2^2 \quad \text{sample spectrum} \\
 &\quad \text{aka "periodogram"} \\
 &\doteq S(\omega_k) \hat{\xi}_k
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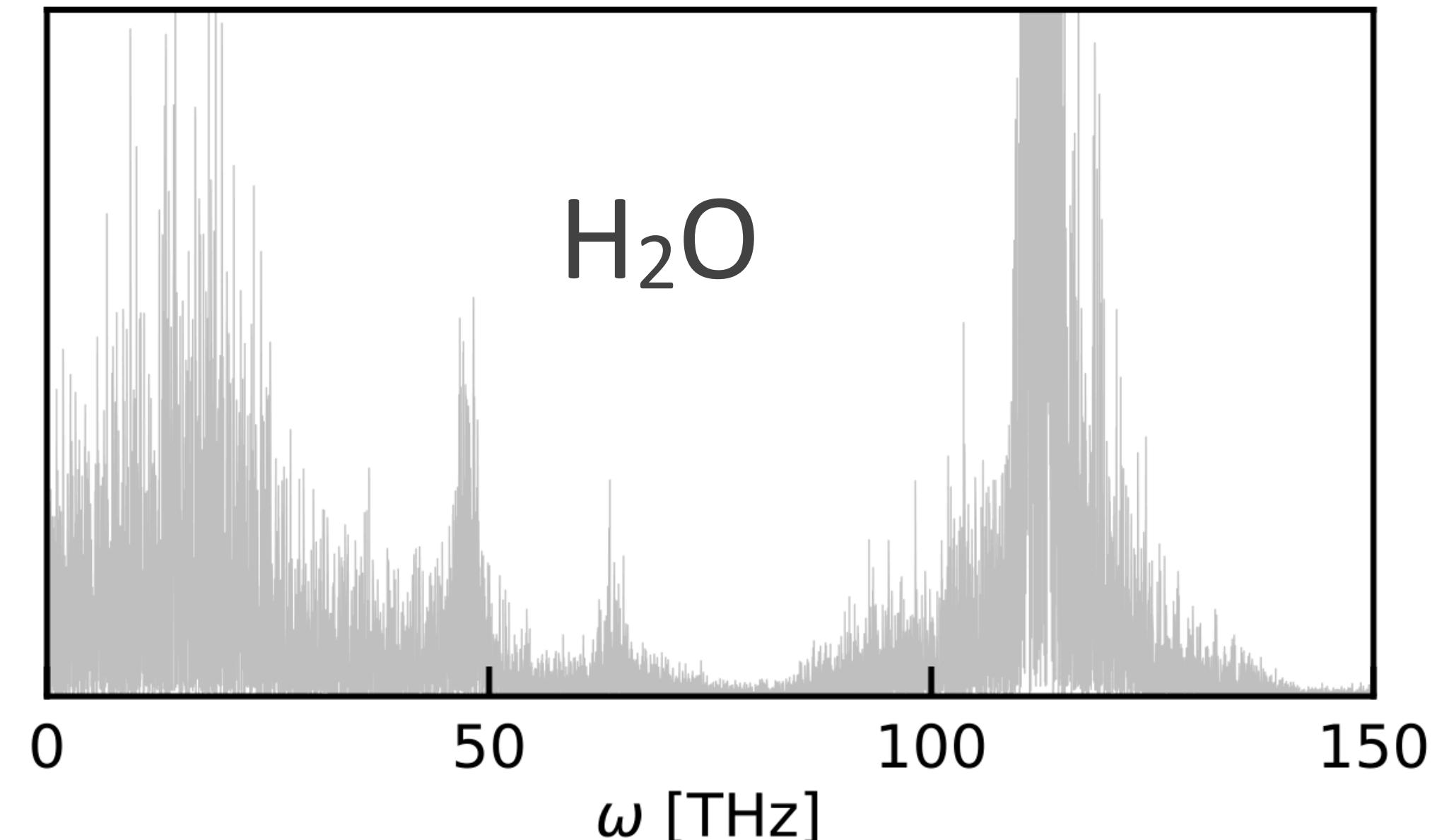
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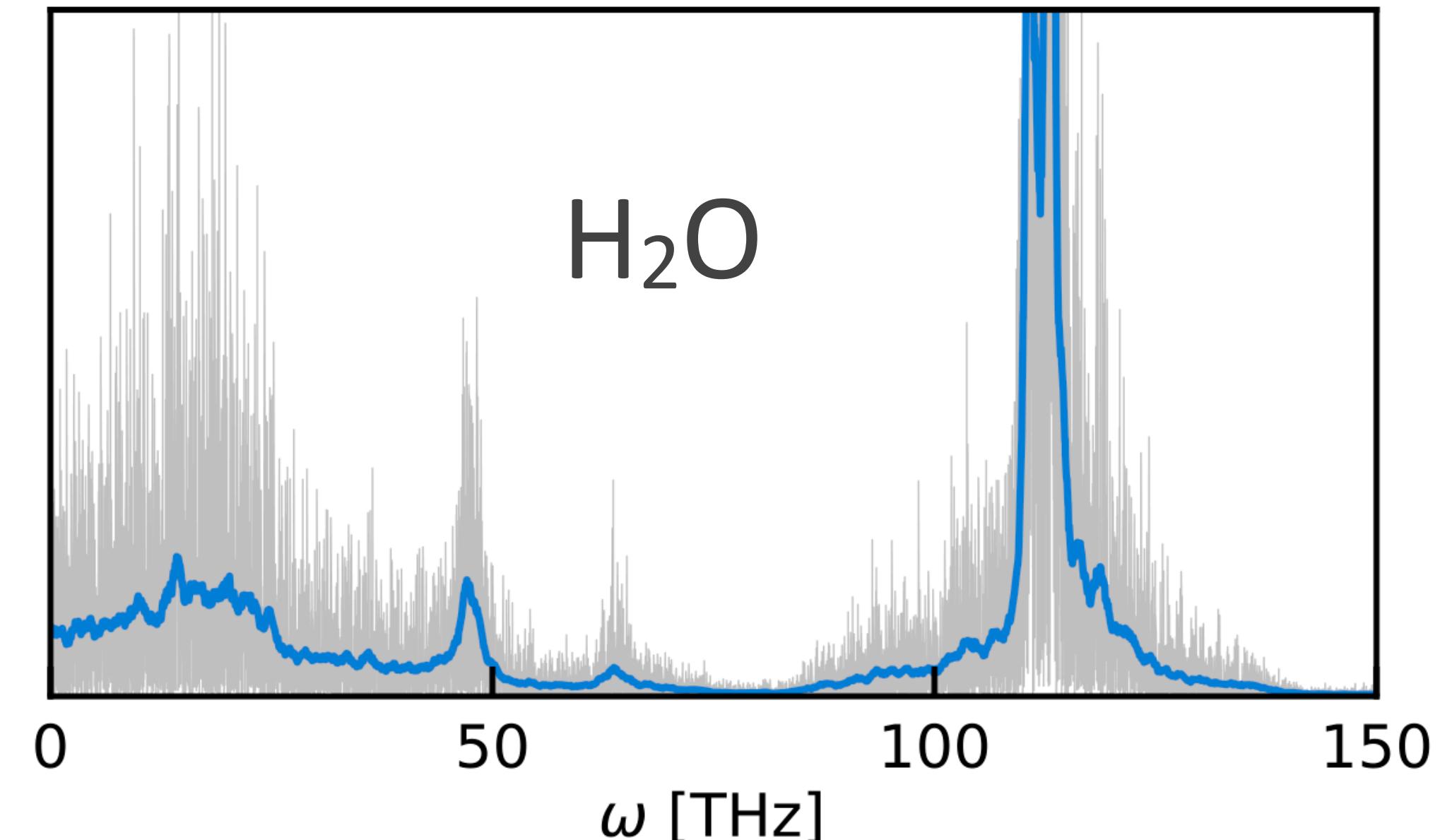
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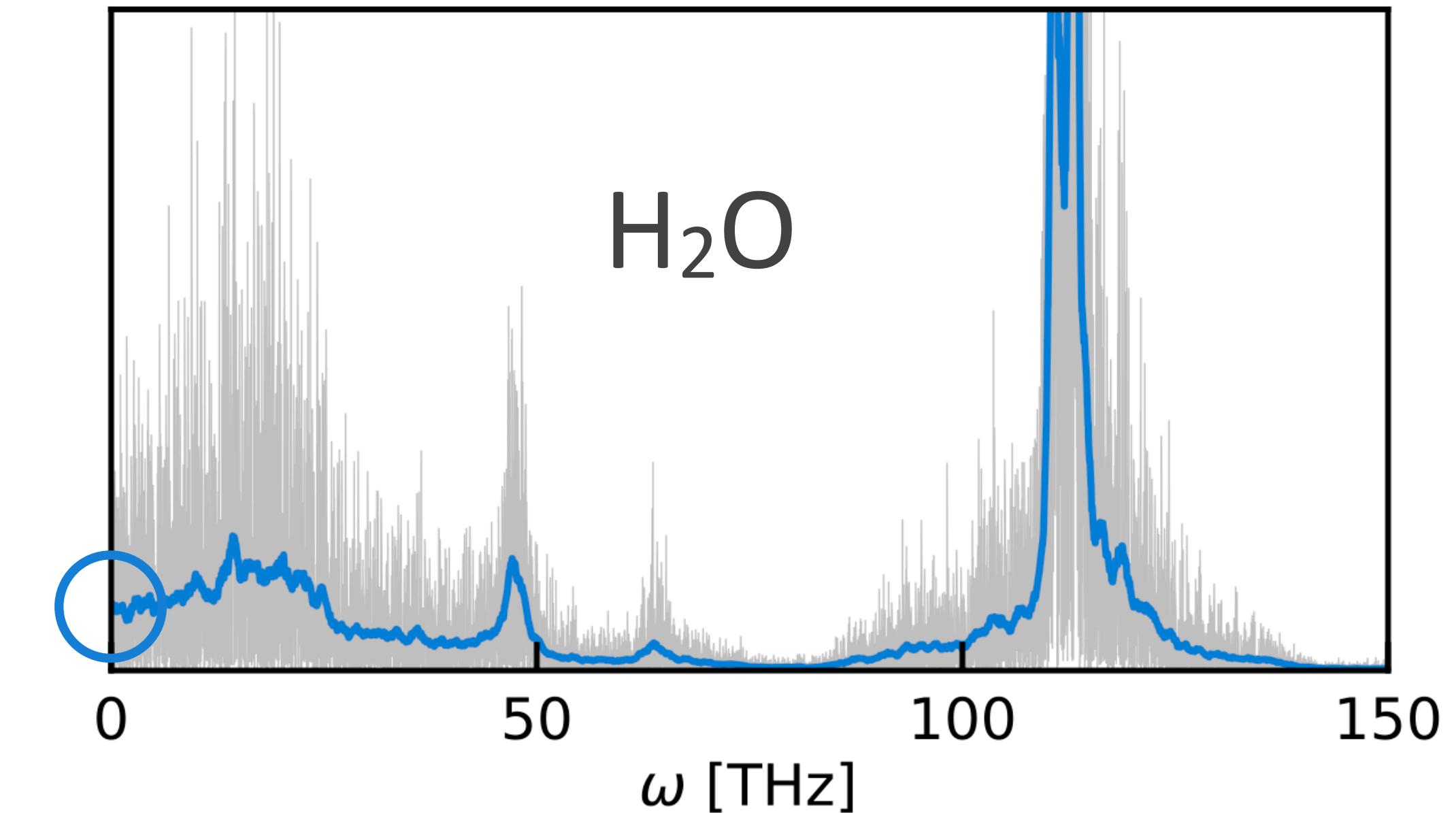
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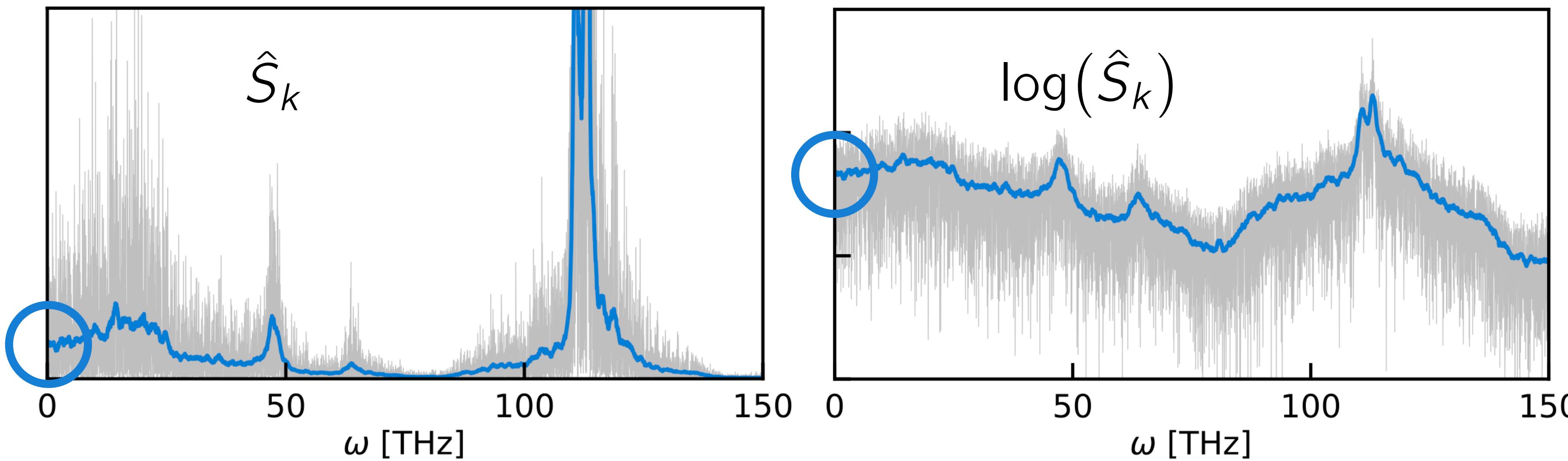
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separating flour from bran

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“cepstral coefficients”
(J.W. Tukey, 1963)

$$\begin{aligned}\hat{C}_n &\doteq \frac{1}{N} \sum_{k=0}^{N-1} \log(\hat{S}_k) e^{2\pi i \frac{kn}{N}} \\ &= C_n + \lambda \delta_{n0} + \hat{w}_n\end{aligned}$$

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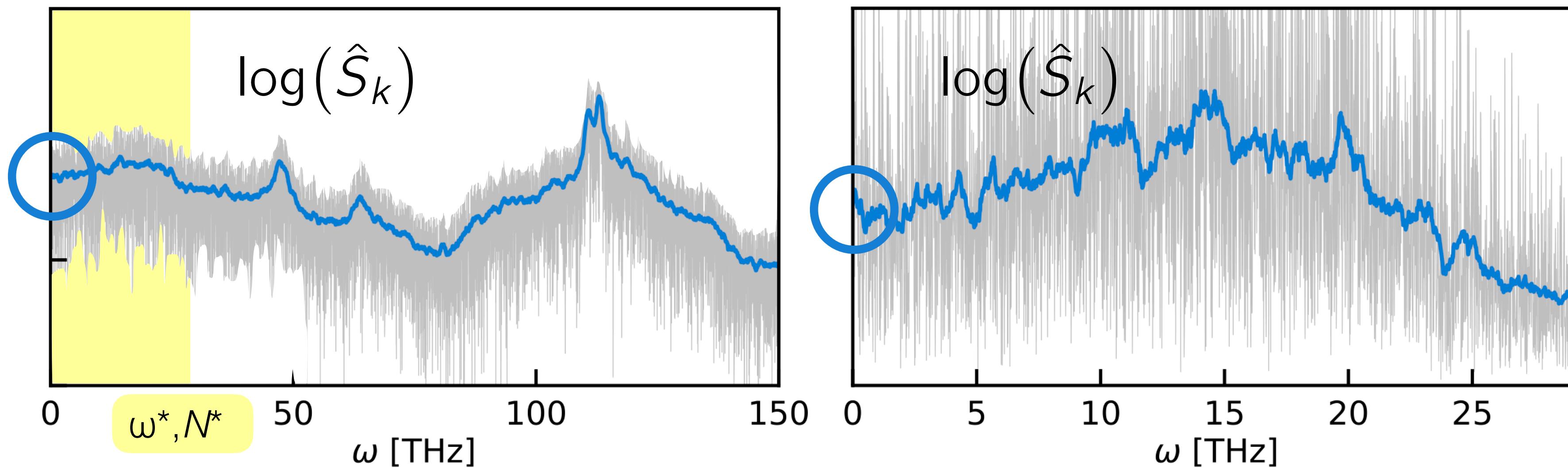
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separating flour from bran

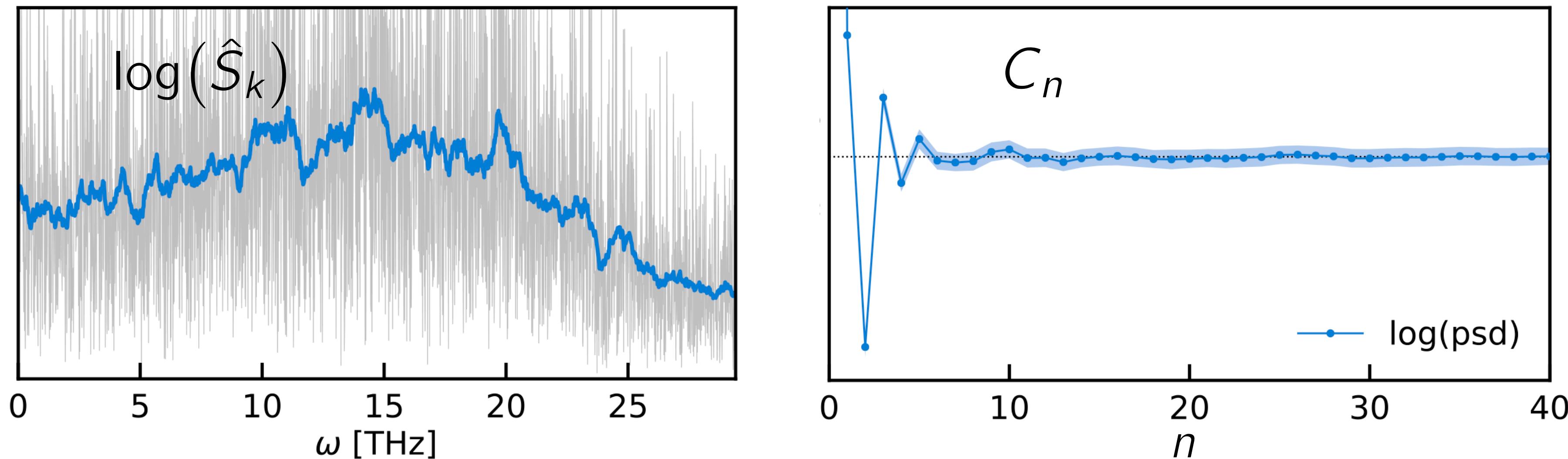
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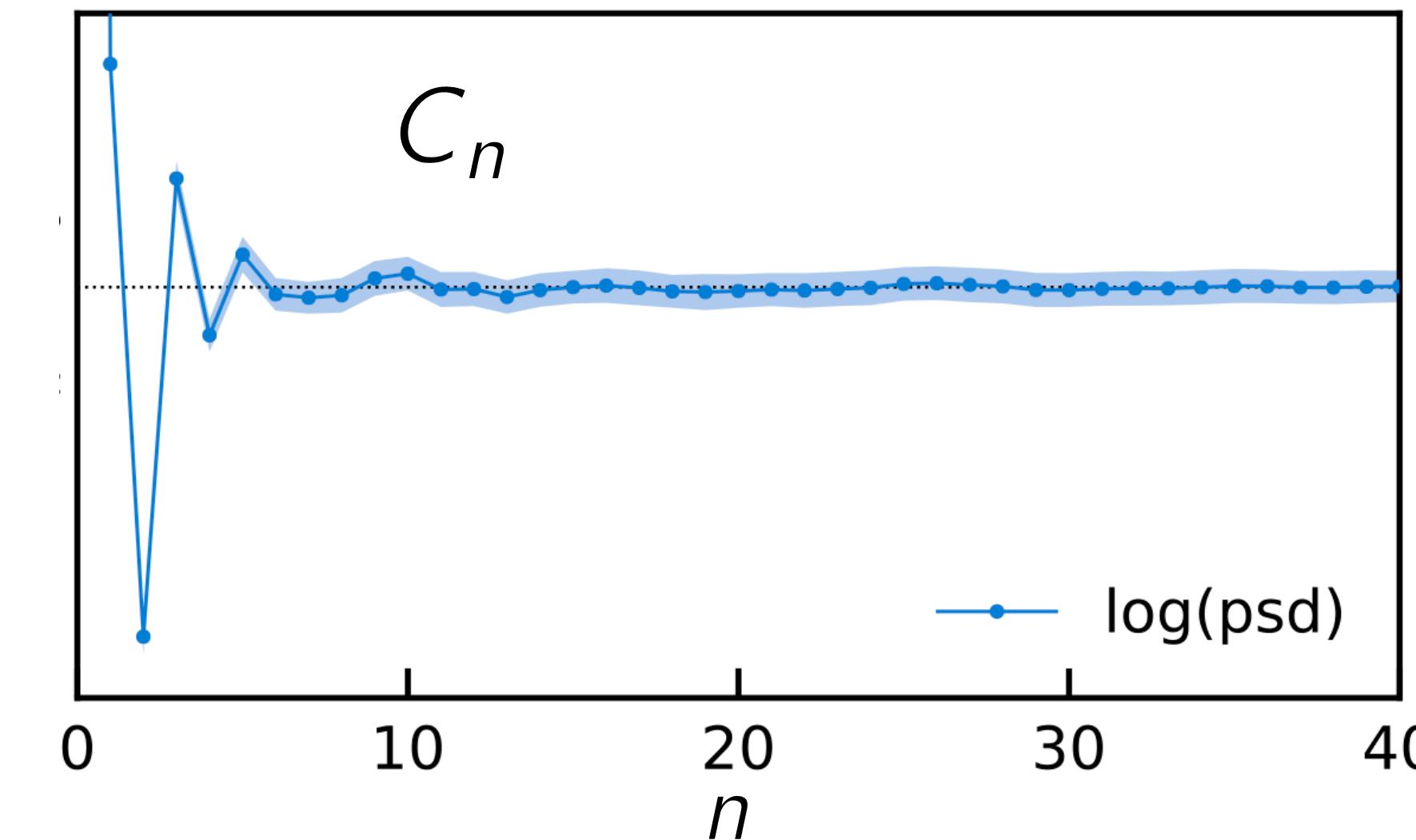
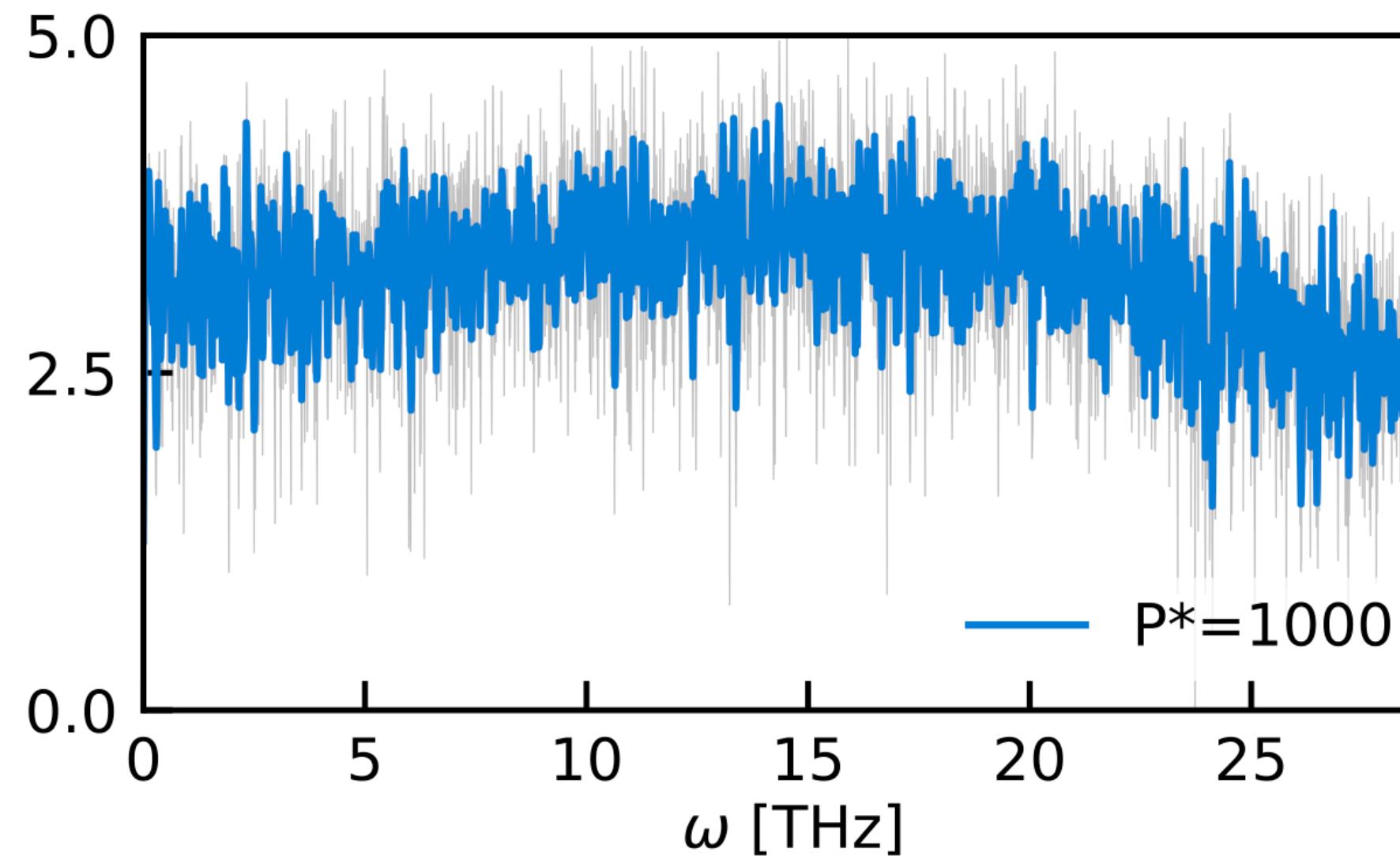


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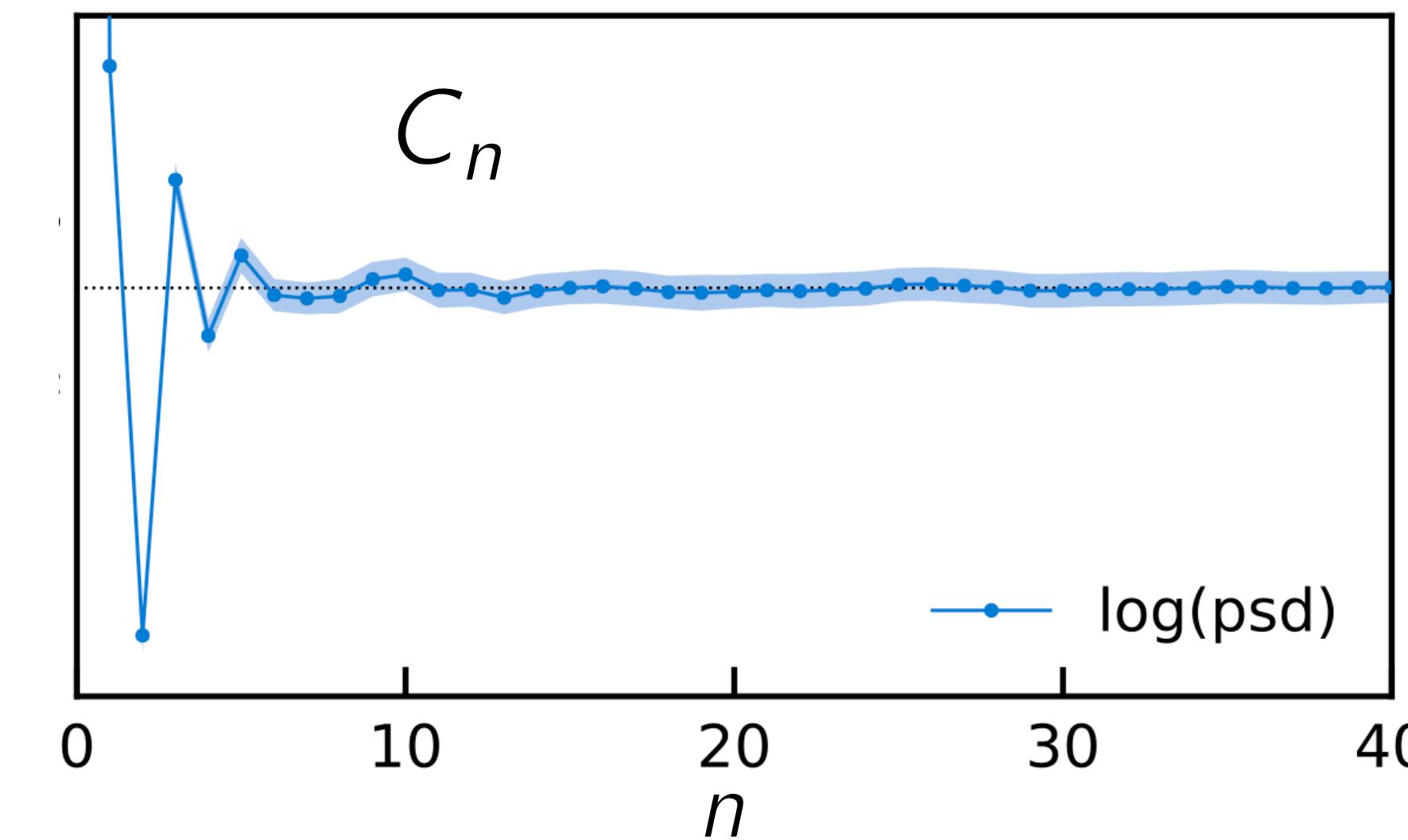
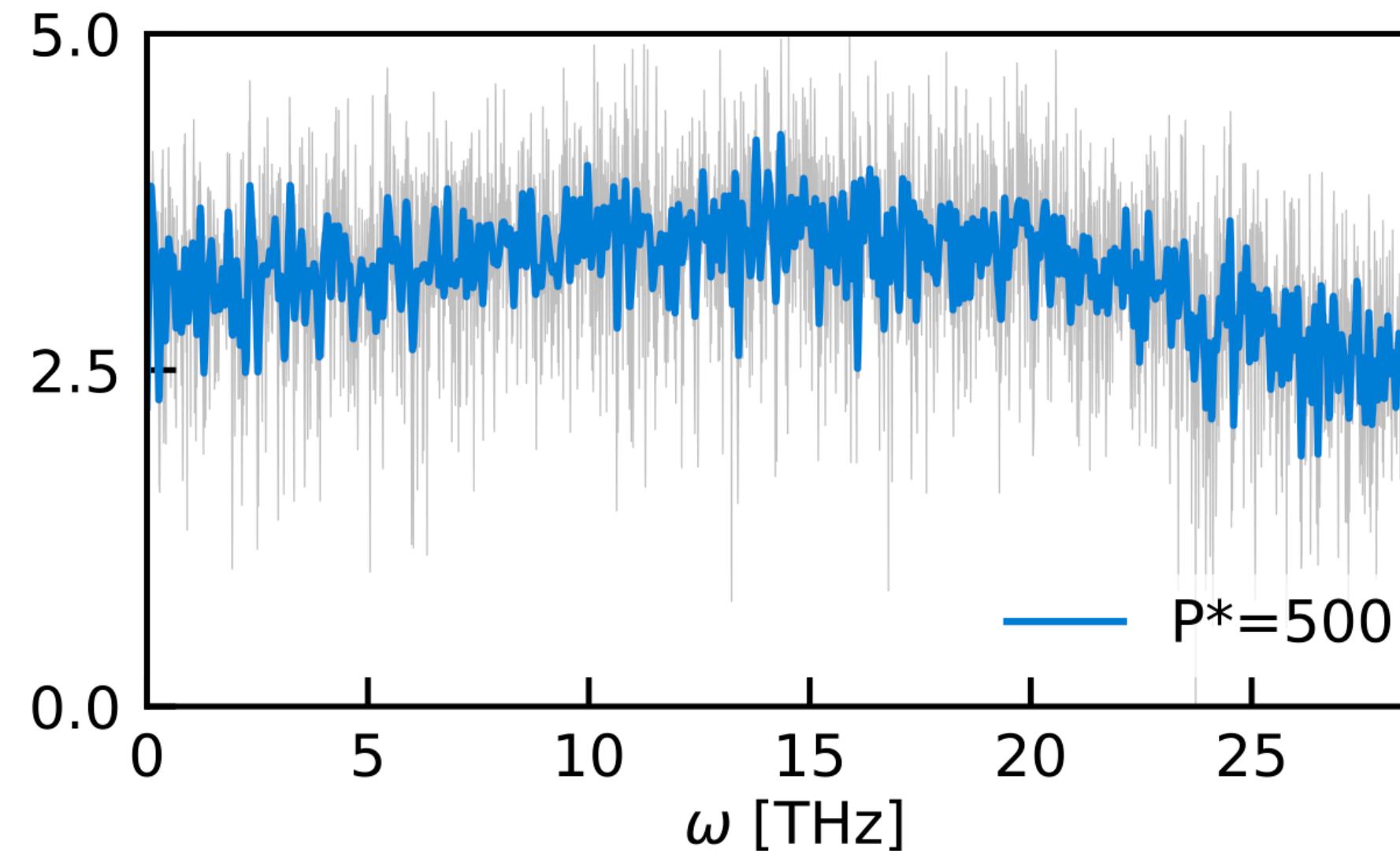


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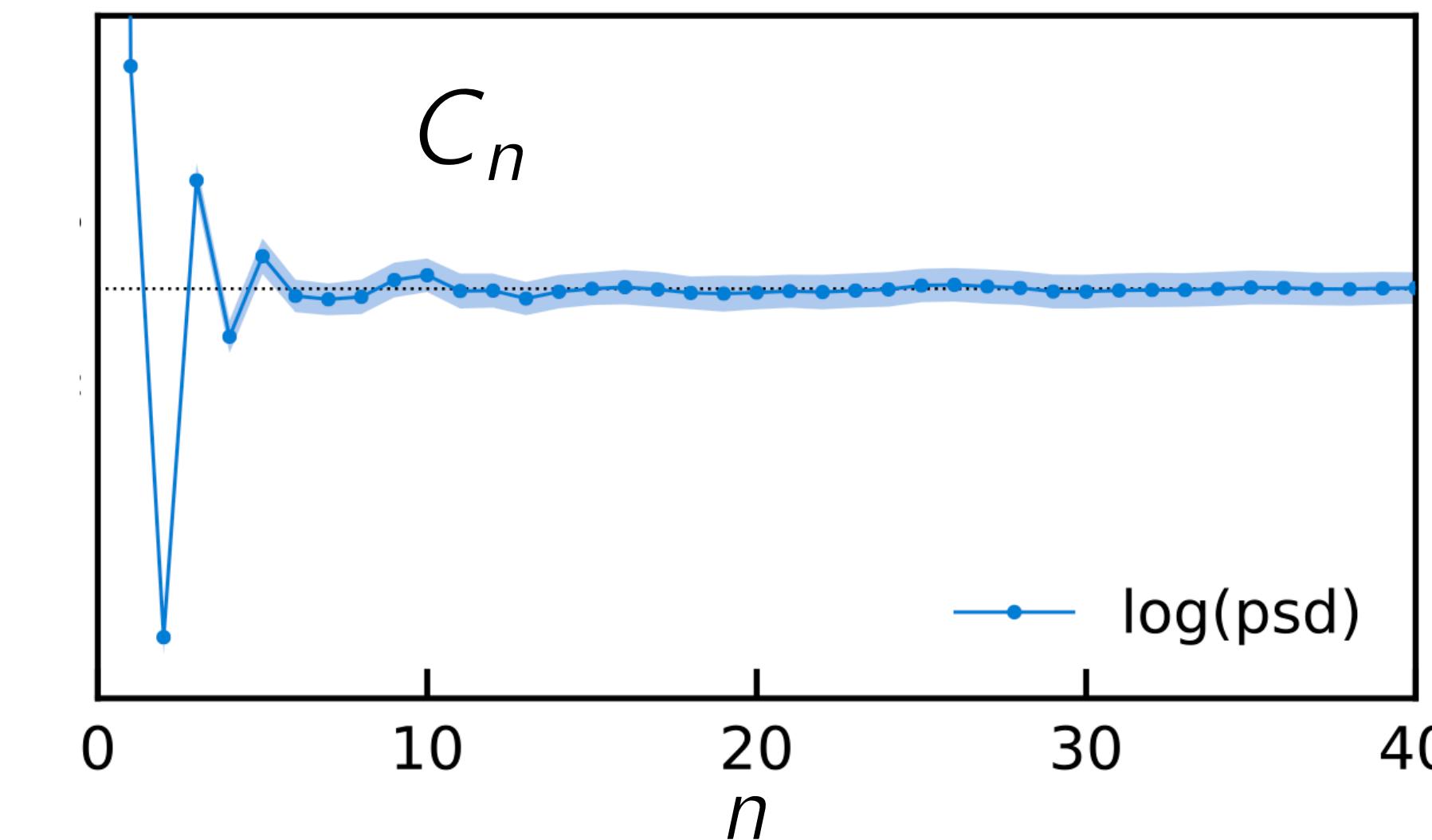
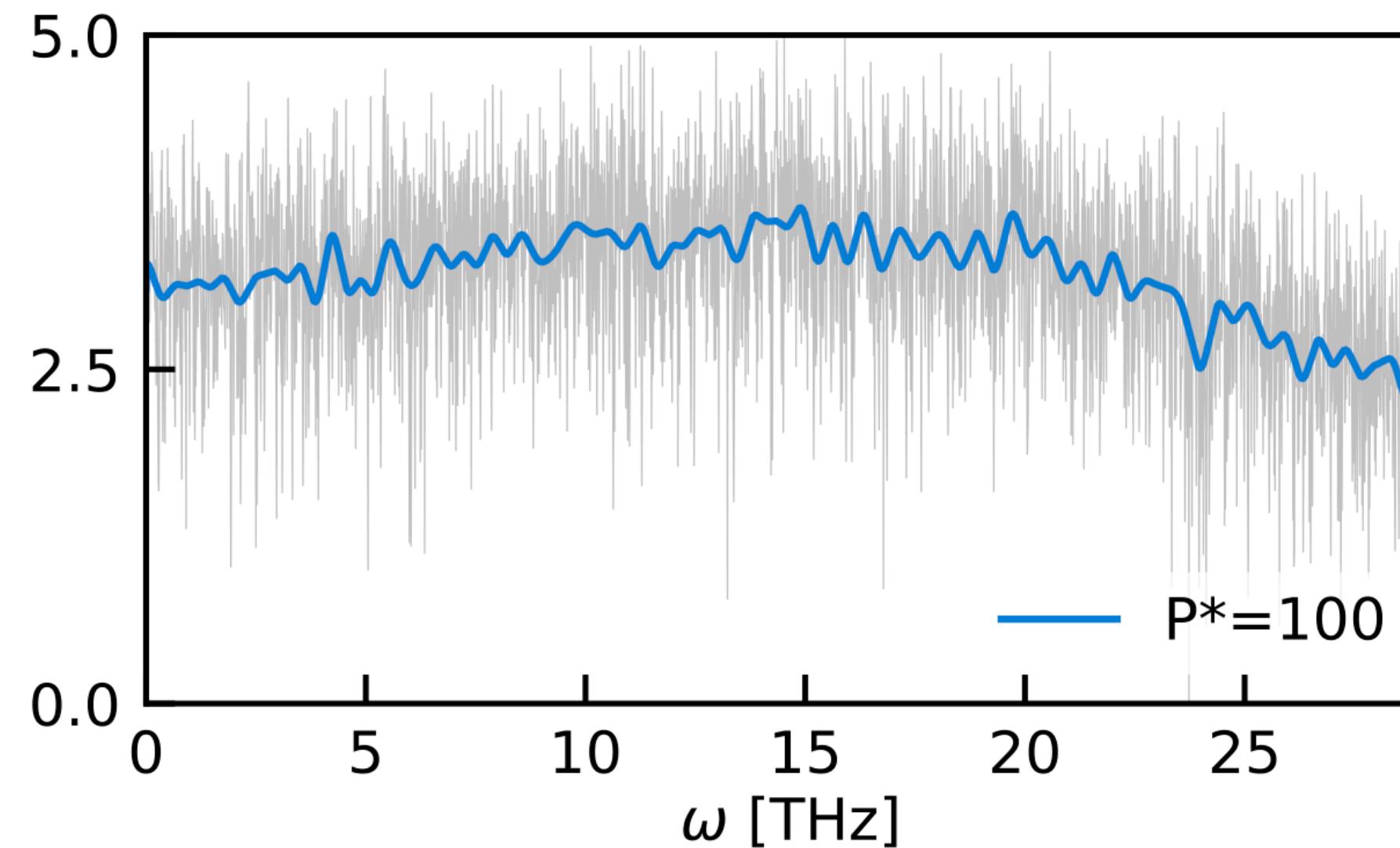


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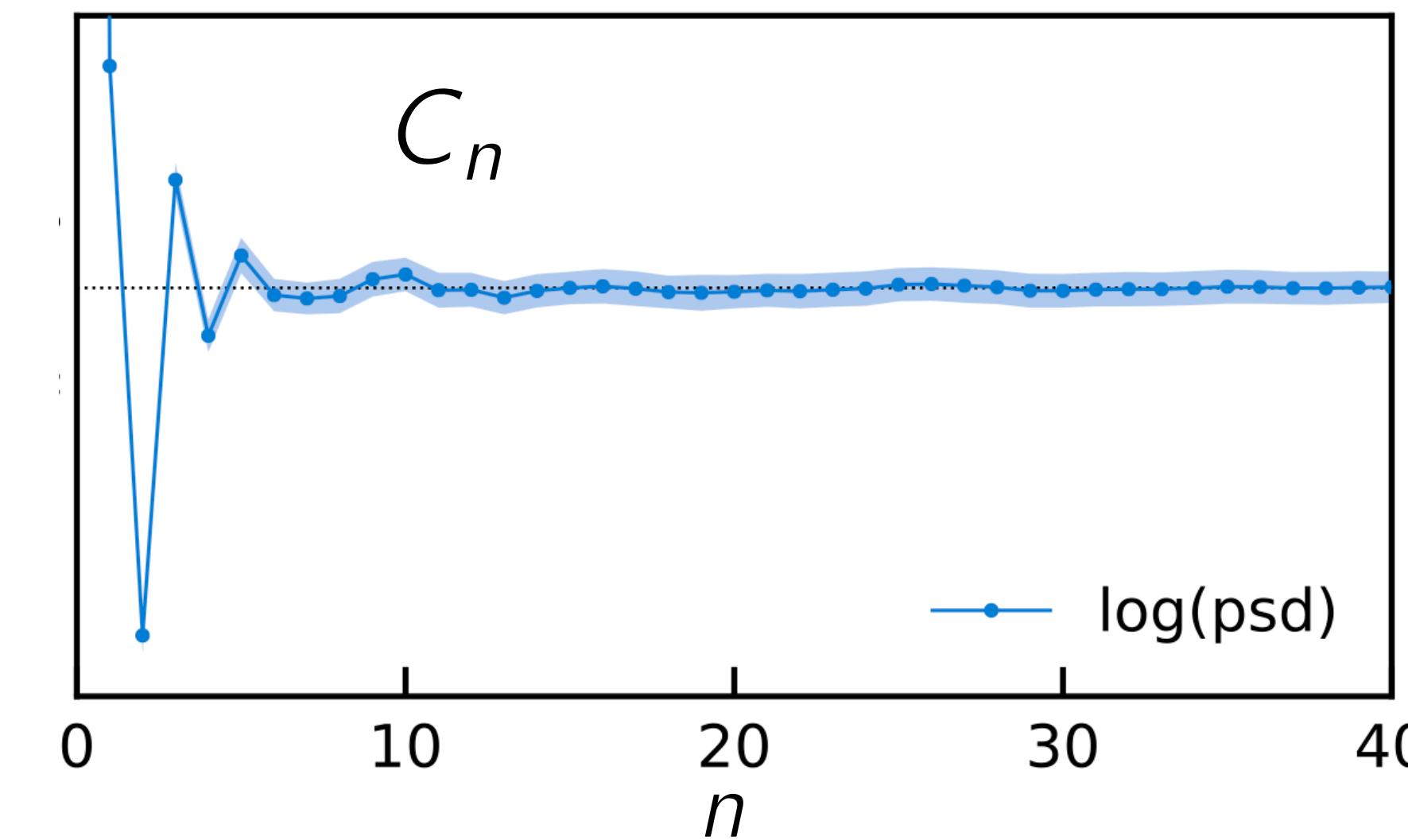
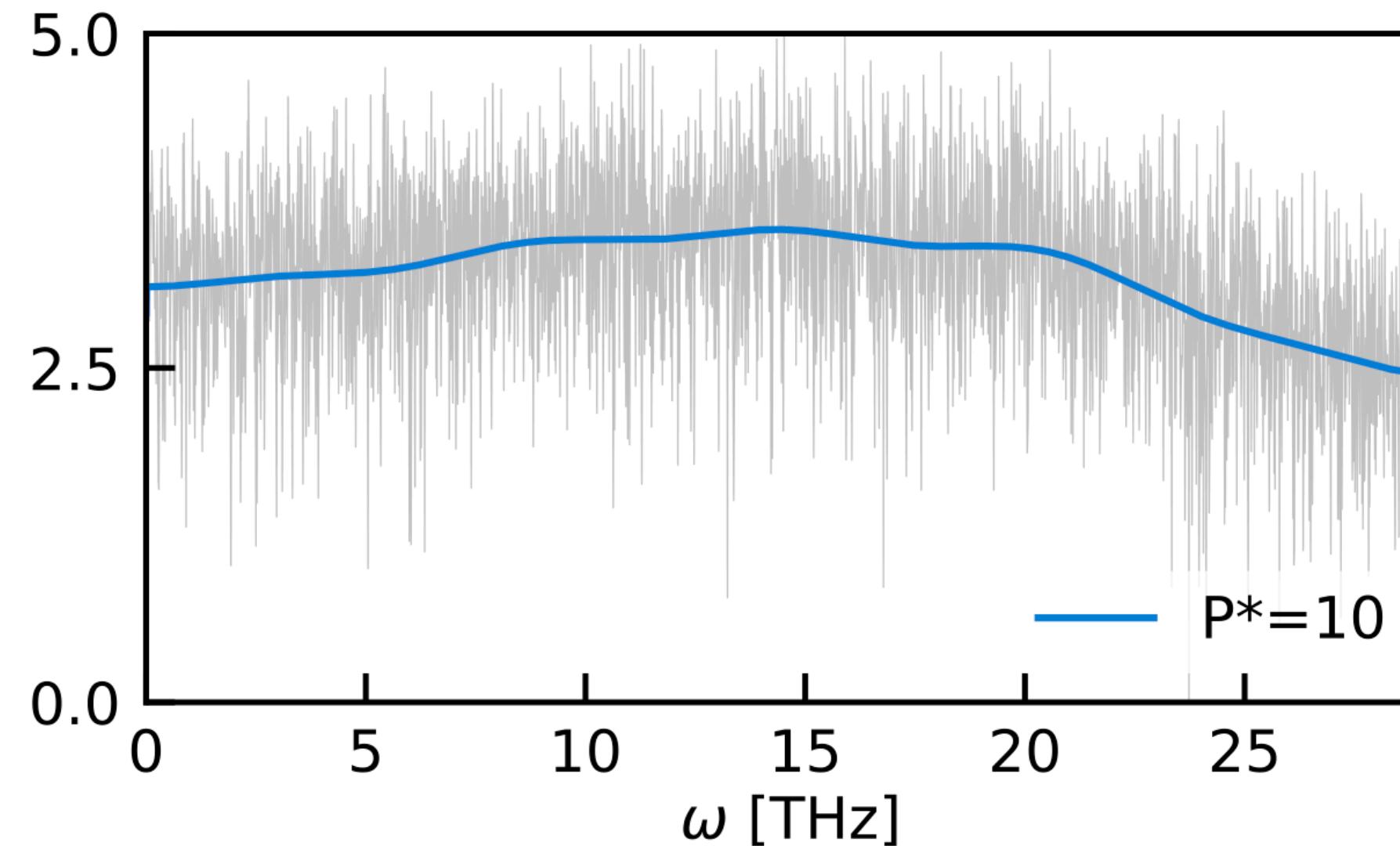


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separating flour from bran

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constants independent of the time series being sampled



determining the optimal number of cepstral coefficients

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Bayes: $P(M, D) = \mathcal{P}(M|D)P(D) = \mathcal{L}(D|M)P(M)$



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Optimal model, maximum of:

$$\mathcal{P}(M|D) = \frac{\mathcal{L}(D|M)P(M)}{P(D)}$$



determining the optimal number of cepstral coefficients

cepstral analysis amounts to assuming that the logarithm of the power spectrum can be modelled by a smooth Fourier series:

Model: $\{P, C_0, C_1, \dots, C_{P-1}\}$; Data: $\{\hat{C}_0, \hat{C}_1, \dots, \hat{C}_{N/2}\}$.

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$$-2 \log(\mathcal{L}) \sim \frac{N}{\sigma^2} \left[\frac{1}{2} (C_0 + \lambda - \hat{C}_0)^2 + \sum_{n=1}^{P-1} (C_n - \hat{C}_n)^2 + \sum_{n=P}^{\frac{N}{2}} \hat{C}_n^2 \right]$$

$$P(M) \propto e^{-\alpha P}$$



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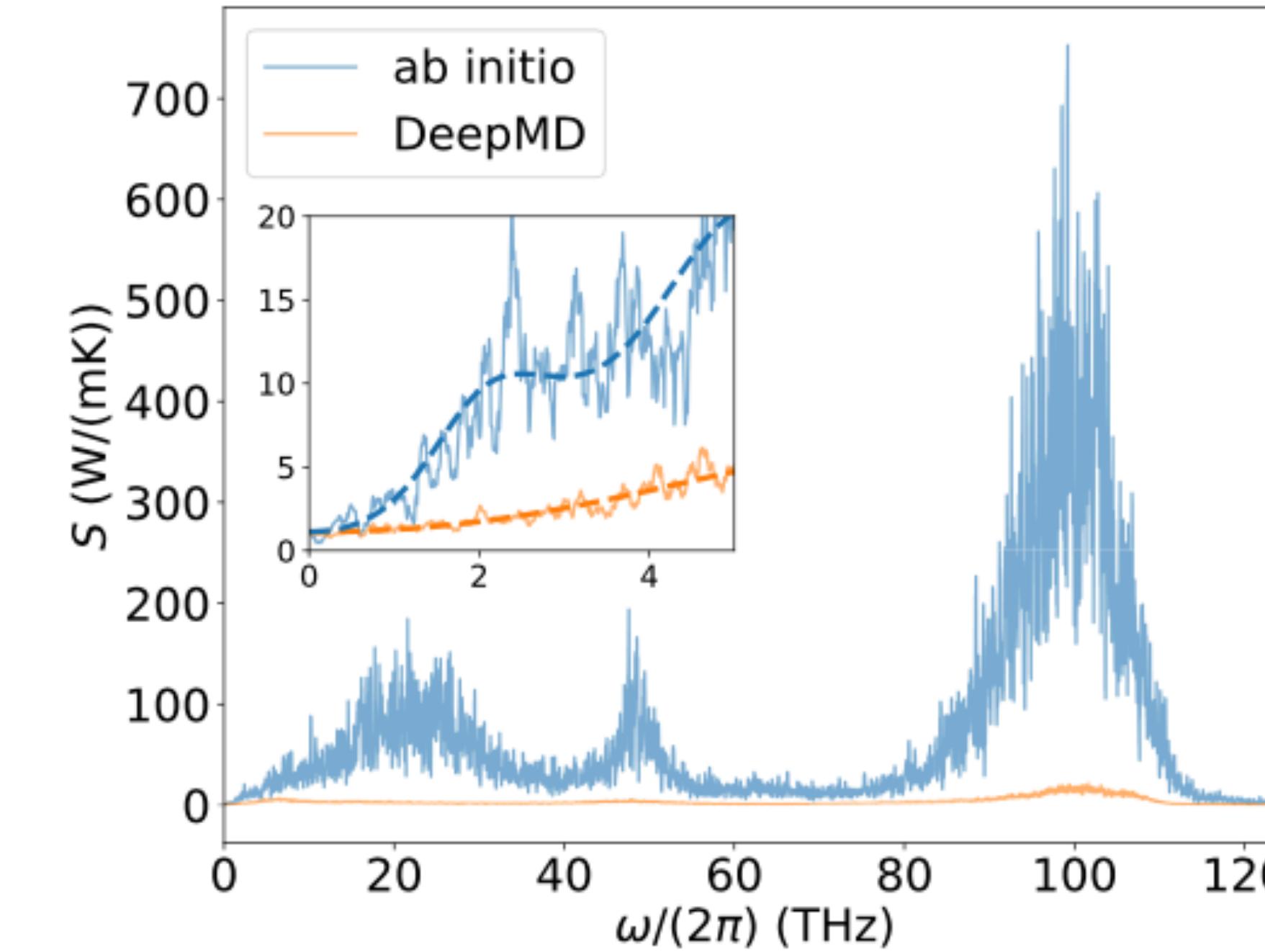
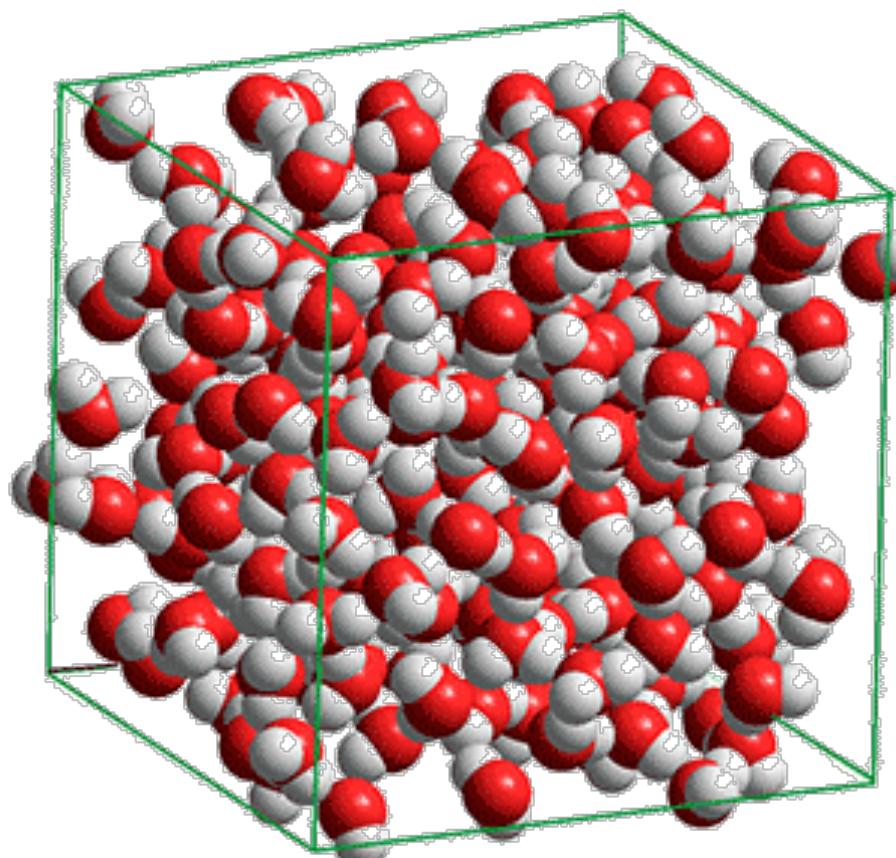
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$$\text{AIC: } \alpha = 2$$



thermal conductivity of liquid water from DFT



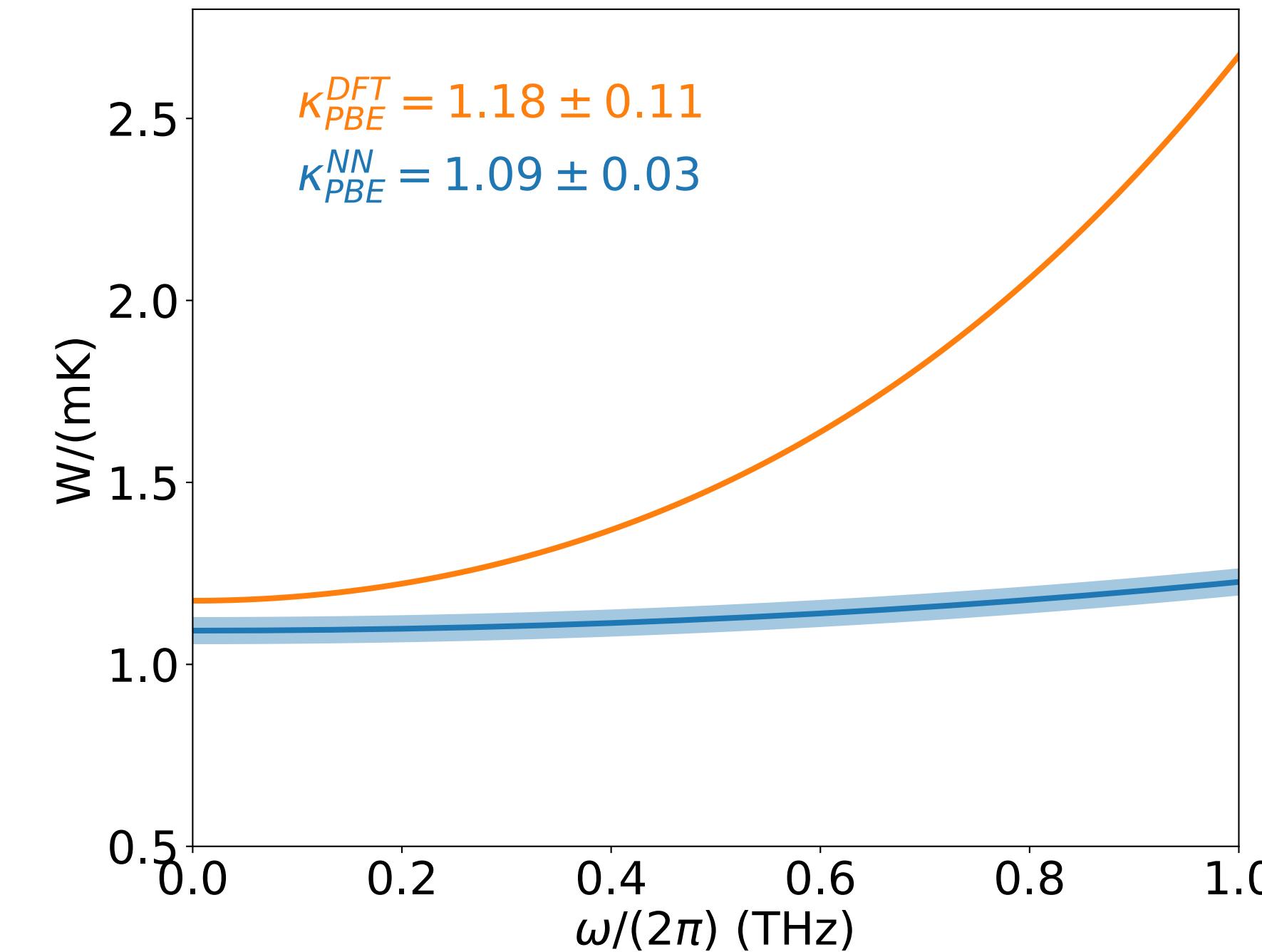
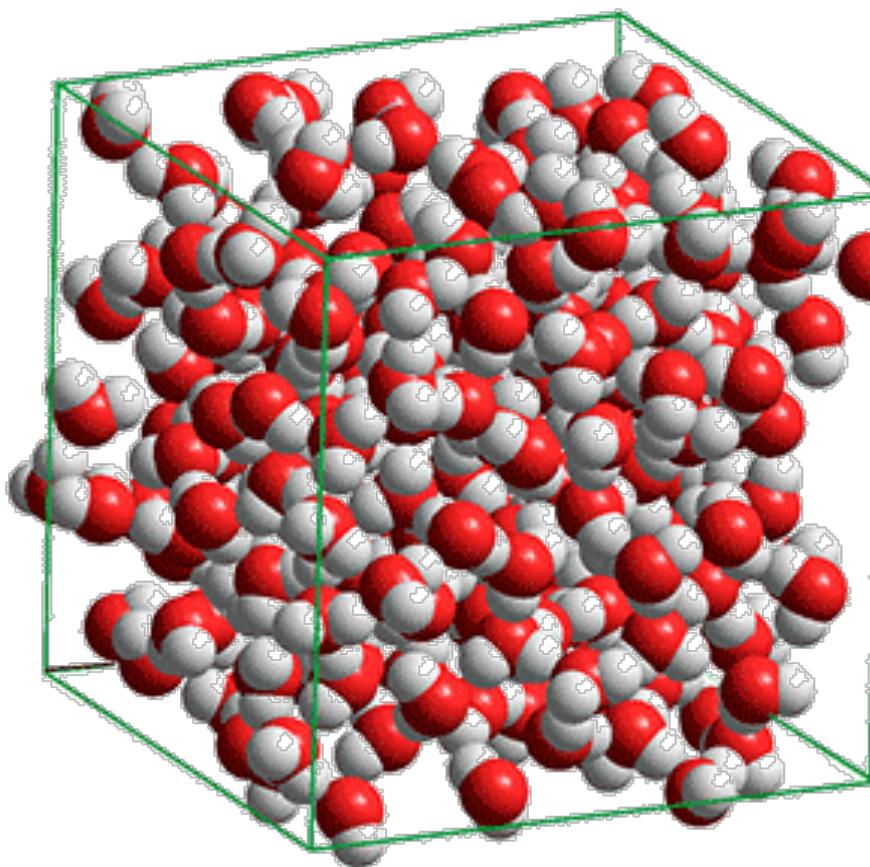
PHYSICAL REVIEW B **104**, 224202 (2021)

Heat transport in liquid water from first-principles and deep neural network simulations

Davide Tisi  ¹, Linfeng Zhang  ², Riccardo Bertossa  ¹, Han Wang  ³, Roberto Car  ^{2,4}, and Stefano Baroni  ^{1,5}



thermal conductivity of liquid water from DFT



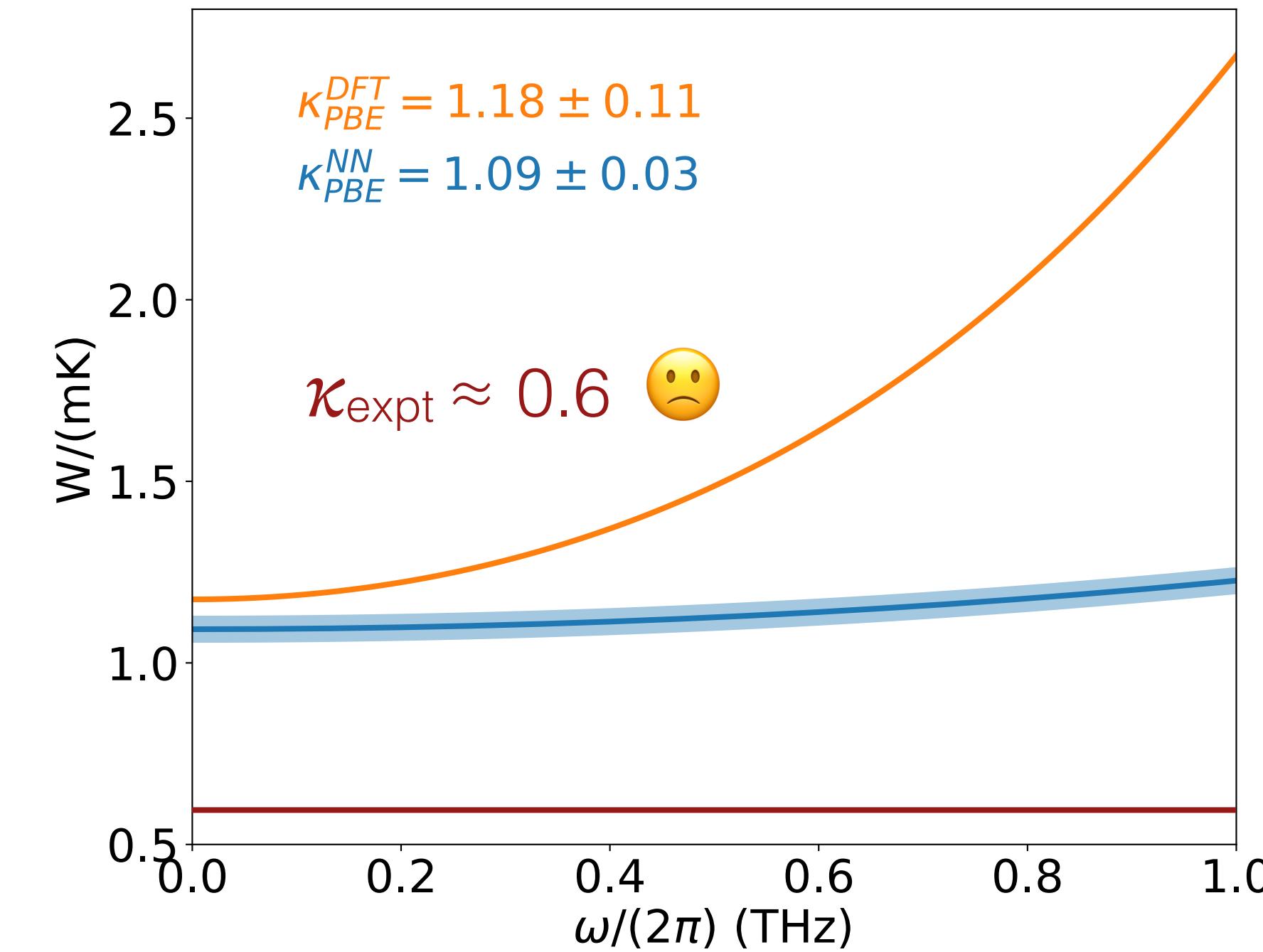
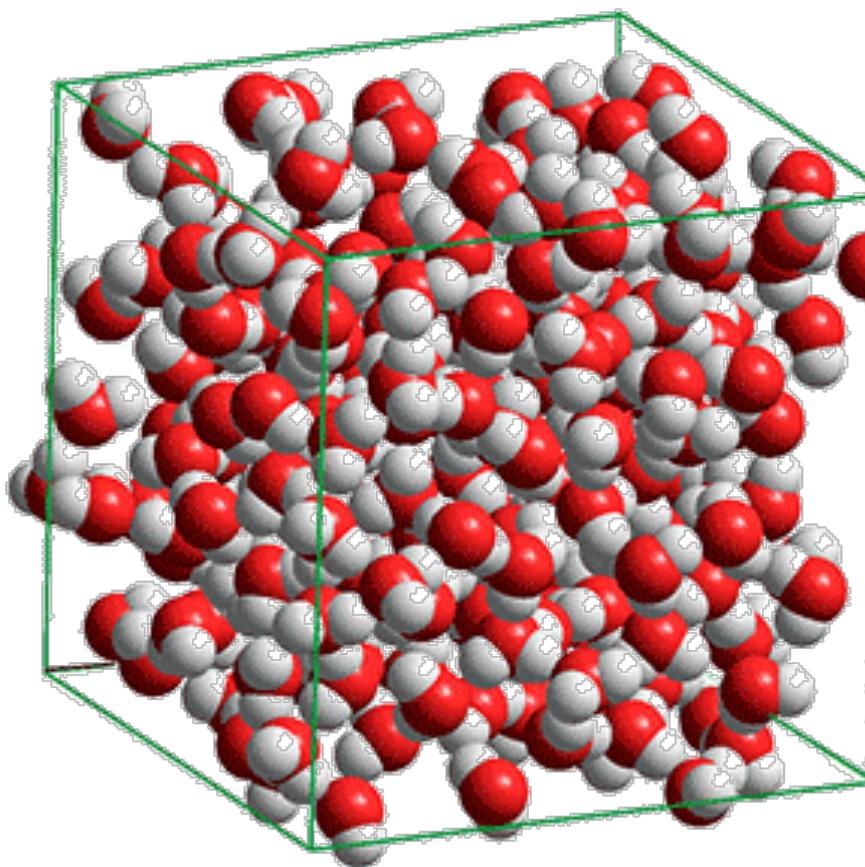
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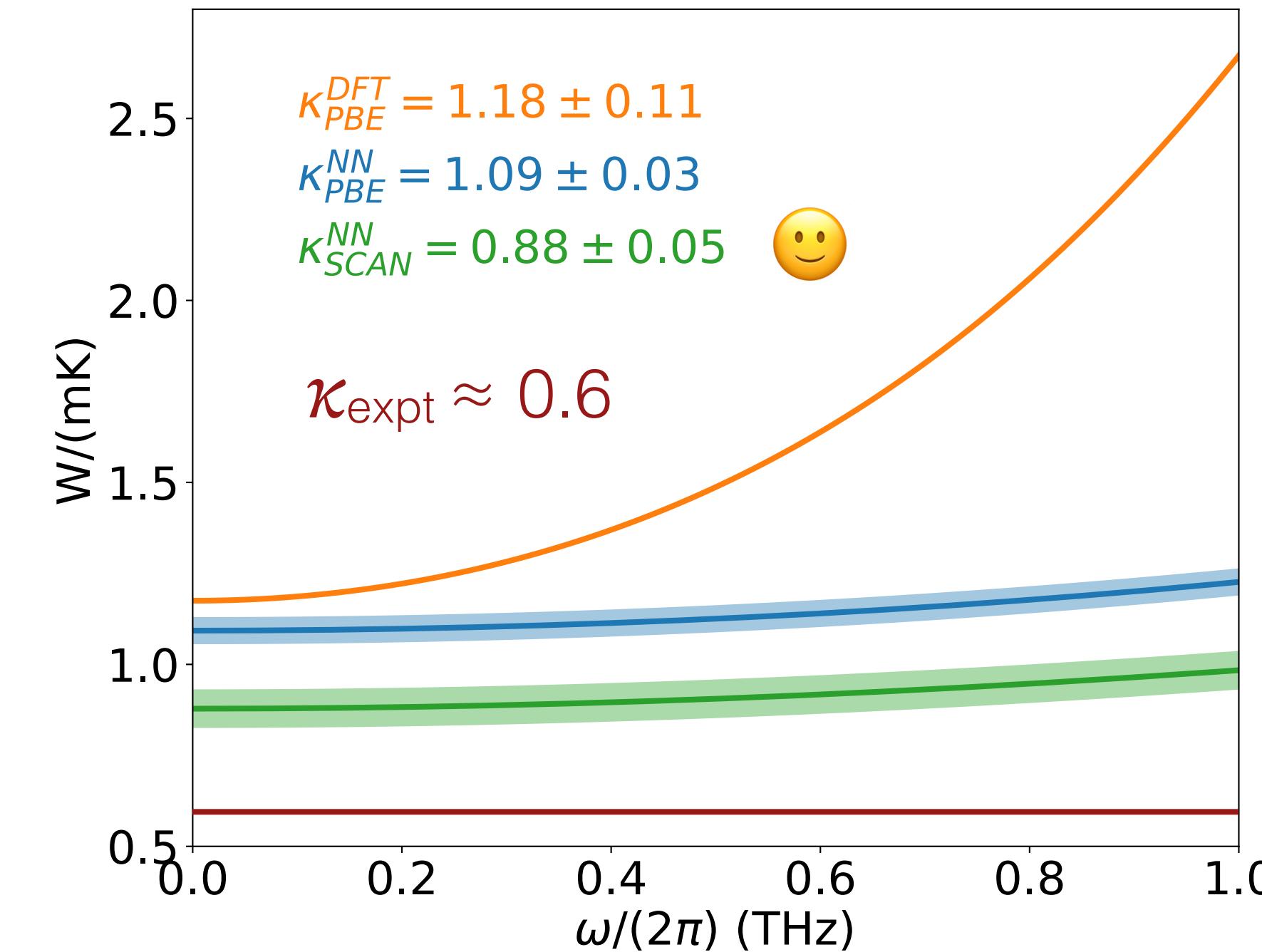
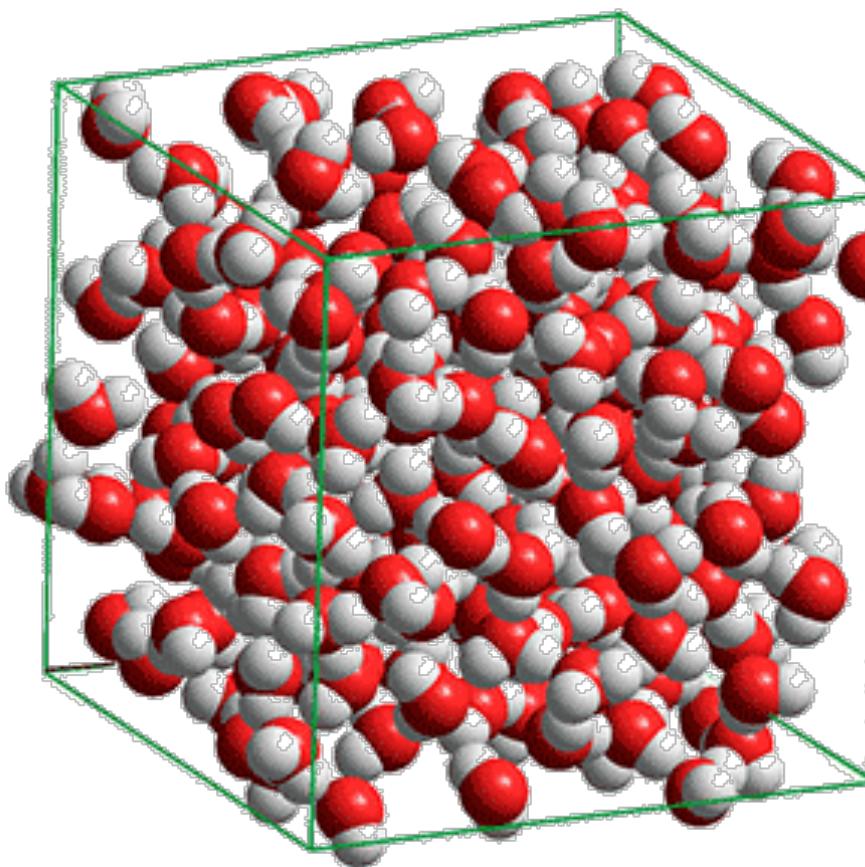
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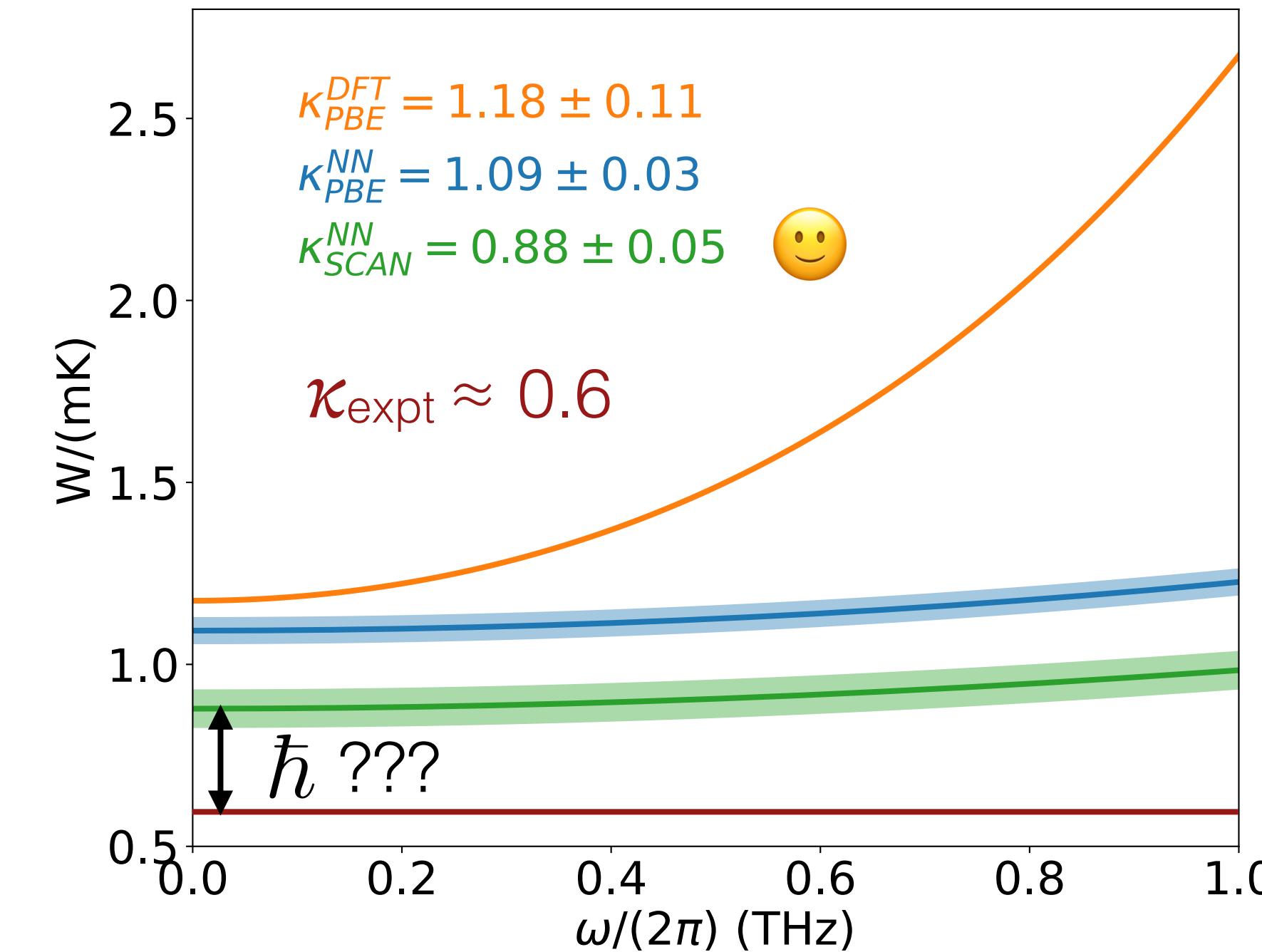
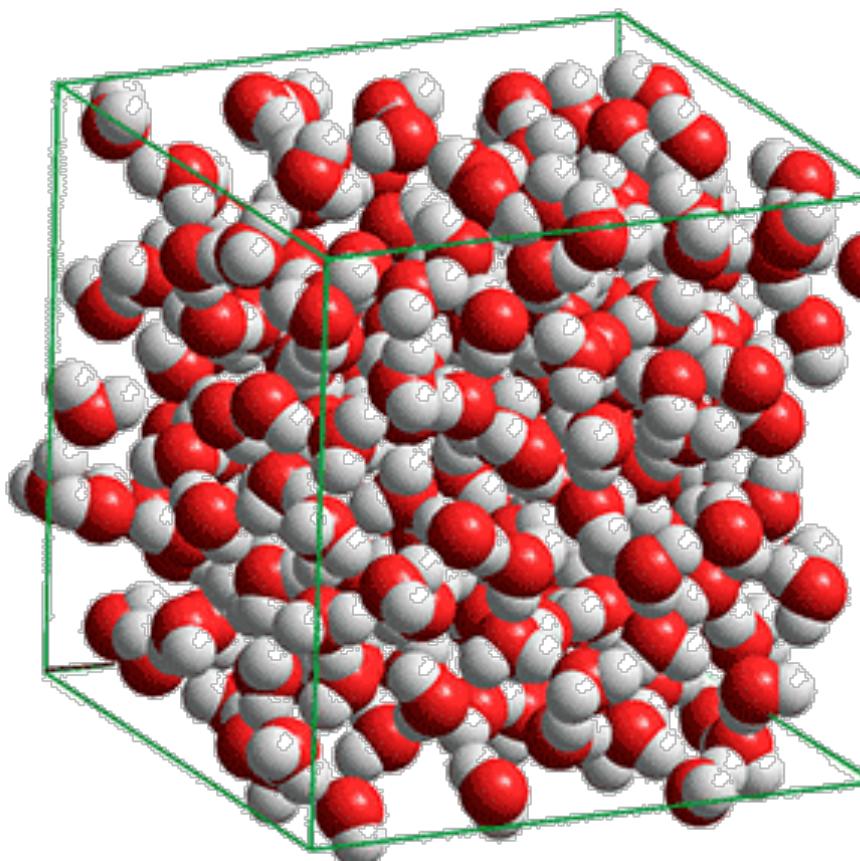
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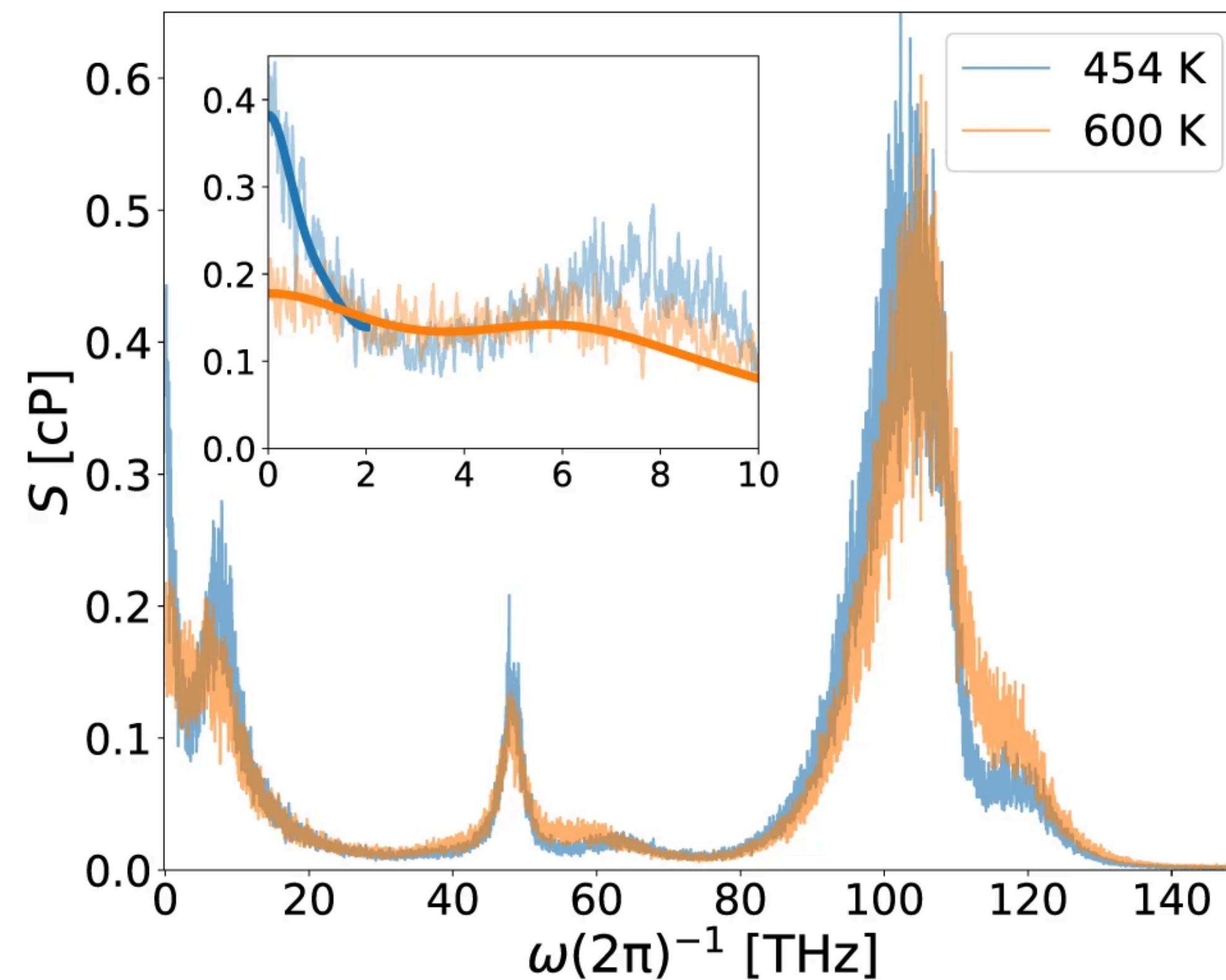
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ab initio shear viscosity in water



npj | computational materials

Viscosity in water from first-principles and deep-neural-network simulations

[Cesare Malosso](#), [Linfeng Zhang](#), [Roberto Car](#), [Stefano Baroni](#) & [Davide Tisi](#)

[npj Computational Materials](#) **8**, Article number: 139 (2022) | [Cite this article](#)



checking normality

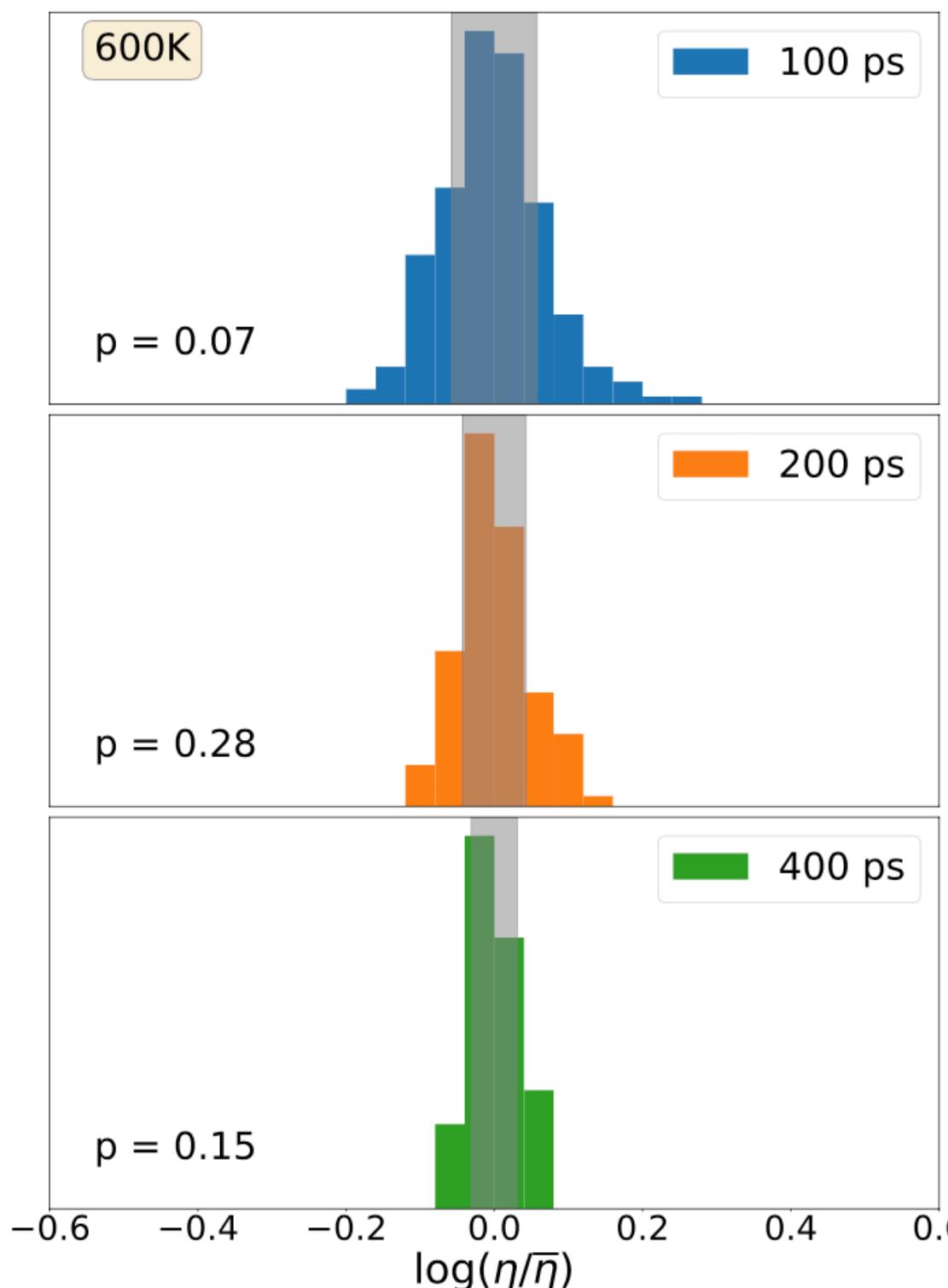
$$\eta = \frac{V}{k_B T} \int_0^\infty \langle \sigma_s(t) \sigma_s(0) \rangle dt$$



checking normality

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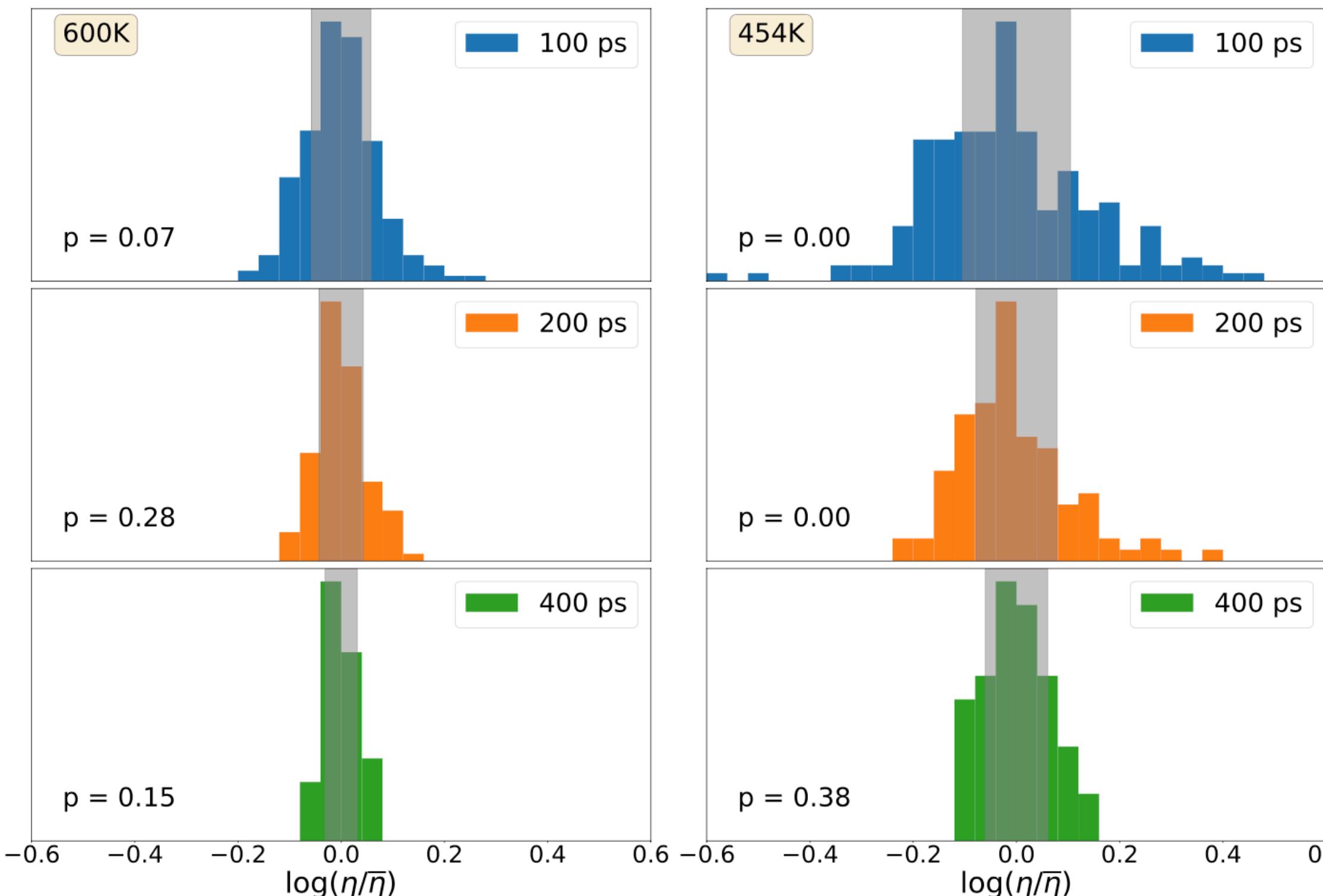
viscosity of water computed for different temperatures
and using trajectory segments of different lengths



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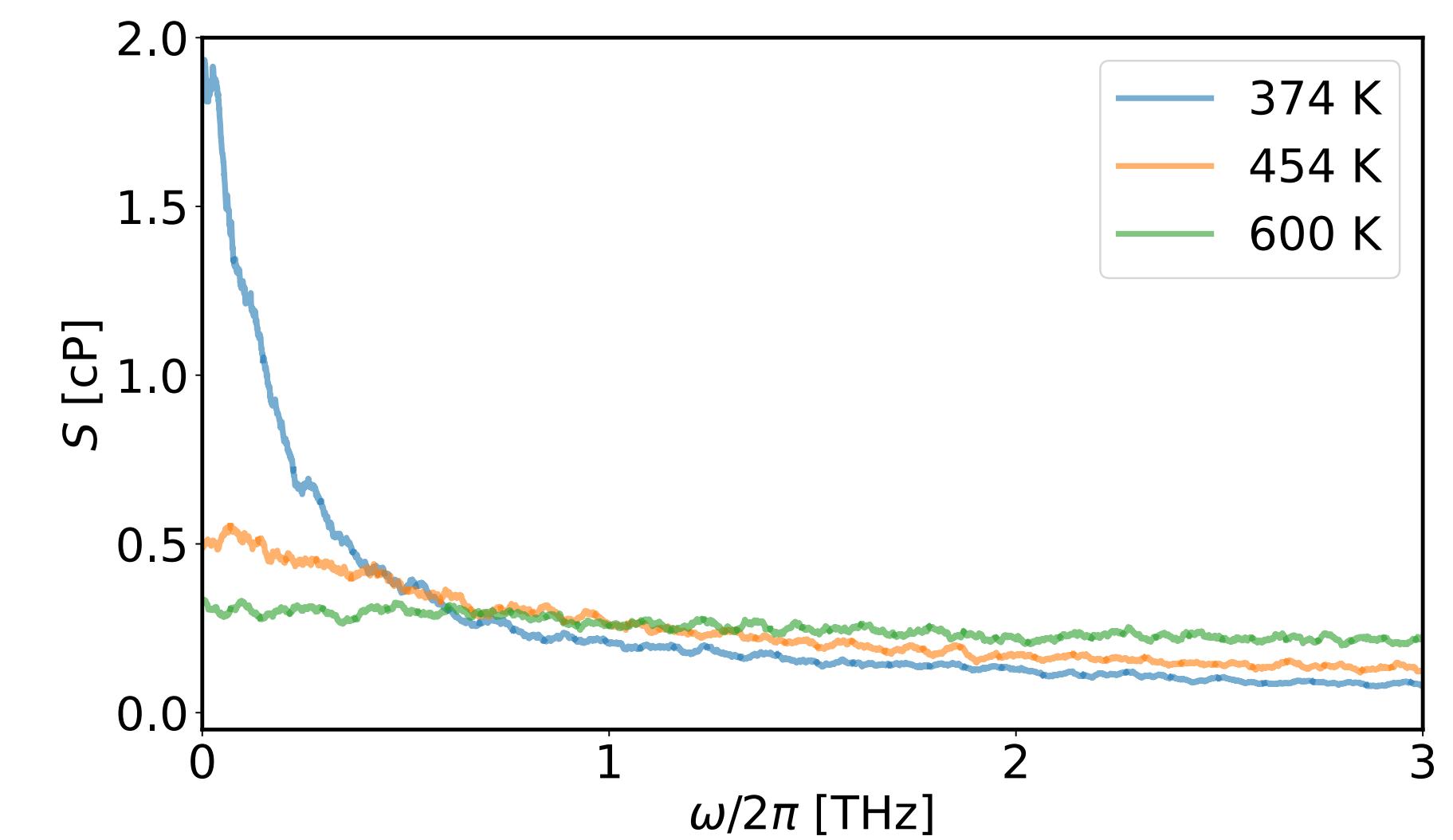
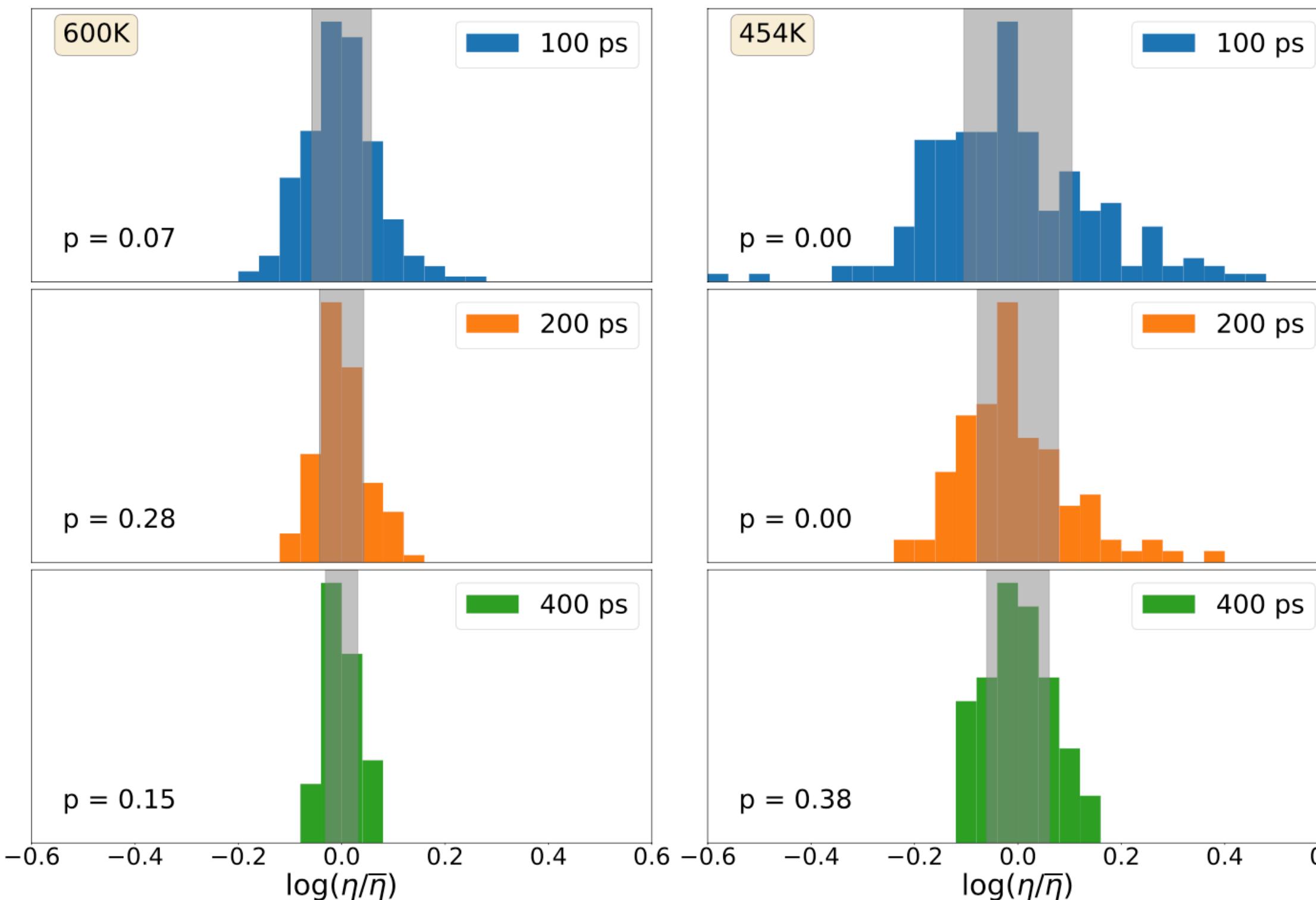
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the cepstral cavobulary

The Quefrency Alanysis of Time Series for
Echoes: Cepstrum, Pseudo-Autocovariance,
Cross-Cepstrum and Saphe Cracking

Bruce P. Bogert, M. J. R. Healy,* John W. Tukey†
Bell Telephone Laboratories and Princeton University

Proceedings of the Symposium on Time
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spectrum	cepstrum
frequency	quefrency
analysis	alansys
period	repiod
filtering	liftering
phase	saphe



SporTran: a code to estimate transport coefficients from current time series

Riccardo Bertossa¹, Loris Ercole^{3,1}, Stefano Baroni^{1,2}

¹SISSA, Italy, ²CNR-IOM, Italy, ³EPFL, Switzerland

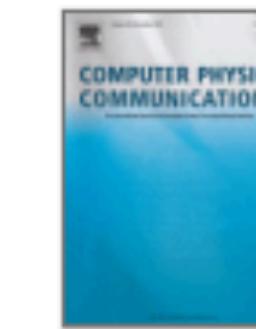


github.com/sissaschoool/sportran



Computer Physics Communications

Volume 280, November 2022, 108470



SporTran: A code to estimate transport coefficients from the cepstral analysis of (multivariate) current time series ★, ★★

<https://doi.org/10.1016/j.cpc.2022.108470>

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*hurdles toward an *ab initio* Green-Kubo theory*

PRL 104, 208501 (2010)

PHYSICAL REVIEW LETTERS

week ending
21 MAY 2010

Thermal Conductivity of Periclase (MgO) from First Principles

Stephen Stackhouse, Lars Stixrude, and Bijaya B. Karki

sensitive to the form of the potential. The widely used Green-Kubo relation [14] does not serve our purposes, because in first-principles calculations it is impossible to uniquely decompose the total energy into individual contributions from each atom.



PRL 118, 175901 (2017)

PHYSICAL REVIEW LETTERS

week ending
28 APRIL 2017

Ab Initio Green-Kubo Approach for the Thermal Conductivity of Solids

Christian Carbogno, Rampi Ramprasad, and Matthias Scheffler

ulations: Because of the limited time scales accessible in aiMD runs, thermodynamic fluctuations dominate the HFACF, which in turn prevents a reliable and numerically stable assessment of the thermal conductivity via Eq. (2).



heat transport from lattice dynamics

$$\mathbf{J} = \sum_n (\dot{\mathbf{R}}_n e_n + \mathbf{R}_n \dot{e}_n)$$



heat transport from lattice dynamics

$$\begin{aligned}\mathbf{R}_n &= \mathbf{R}_n^\circ + \mathbf{u}_n \\ \mathbf{J} &= \sum_n (\dot{\mathbf{R}}_n e_n + \mathbf{R}_n \dot{e}_n) \\ &= \sum_n \mathbf{R}_n^\circ \dot{e}_n + \frac{d}{dt} \sum_n \mathbf{u}_n e_n\end{aligned}$$



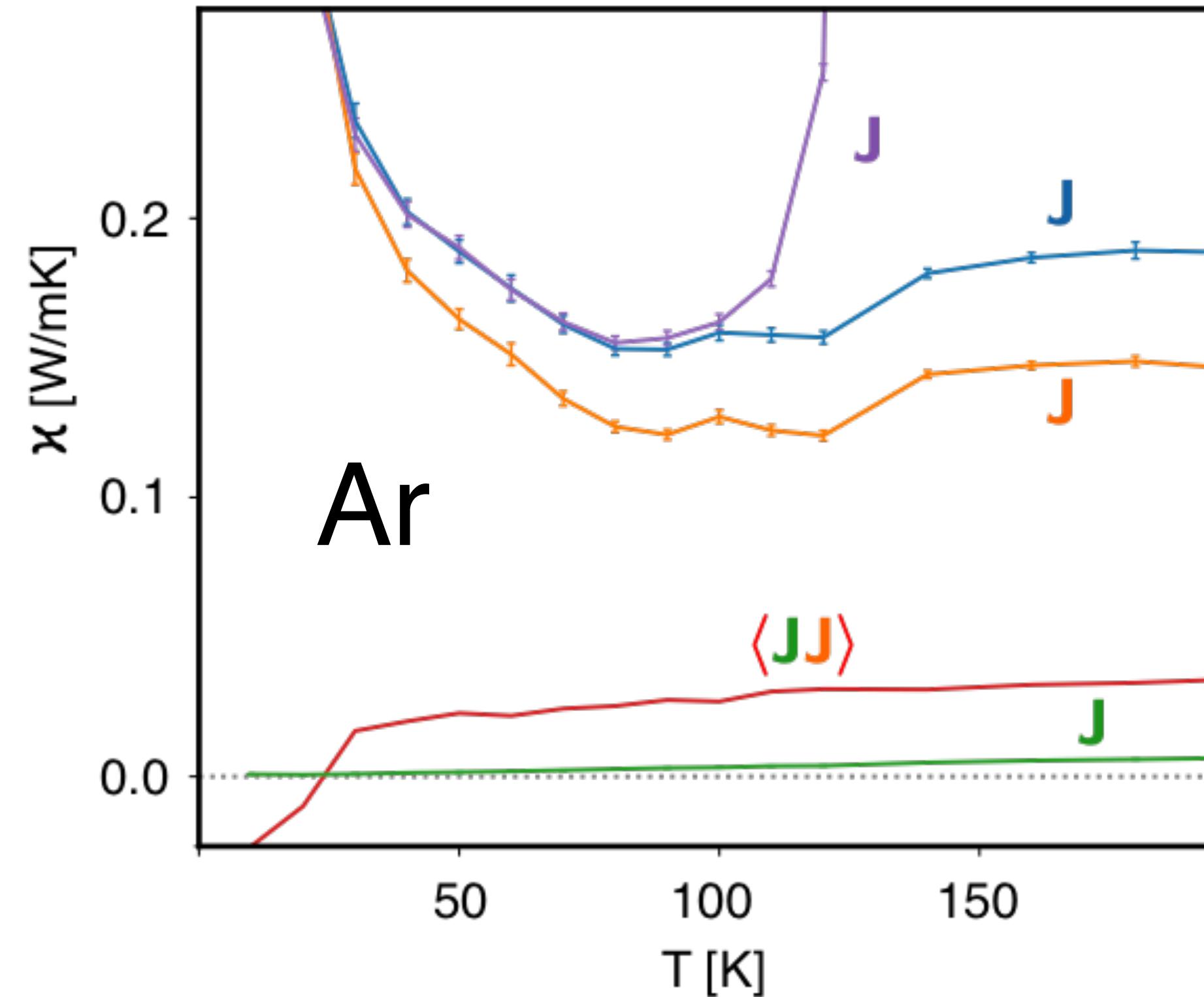
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$$J_\alpha = \frac{1}{2} \sum_{ij\beta\gamma} (R_{i\alpha}^\circ - R_{j\alpha}^\circ) \Phi_{i\beta}^{j\gamma} u_{i\beta} \dot{u}_{j\gamma}, \quad \Phi_{i\beta}^{j\gamma} = \left. \frac{\partial^2 E}{\partial u_{i\beta} \partial u_{j\gamma}} \right|_{u=0}$$



heat transport from lattice dynamics

$$J_\alpha = \frac{1}{2} \sum_{ij\beta\gamma} (R_{i\alpha}^\circ - R_{j\alpha}^\circ) \Phi_{i\beta}^{j\gamma} u_{i\beta} \dot{u}_{j\gamma},$$

$$\kappa \propto \int_0^\infty dt \int du_\circ d\dot{u}_\circ \underbrace{J(u_t \dot{u}_t) J(u_\circ \dot{u}_\circ)}_{\text{4-th order polynomial}} \underbrace{e^{-\beta H(u_\circ \dot{u}_\circ)}}_{\text{Gaussian}}$$

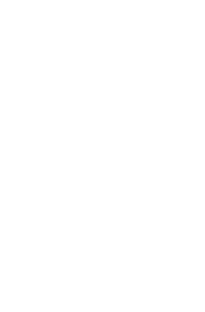


heat transport from lattice dynamics

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[Gaussian integral \mapsto Wick theorem]

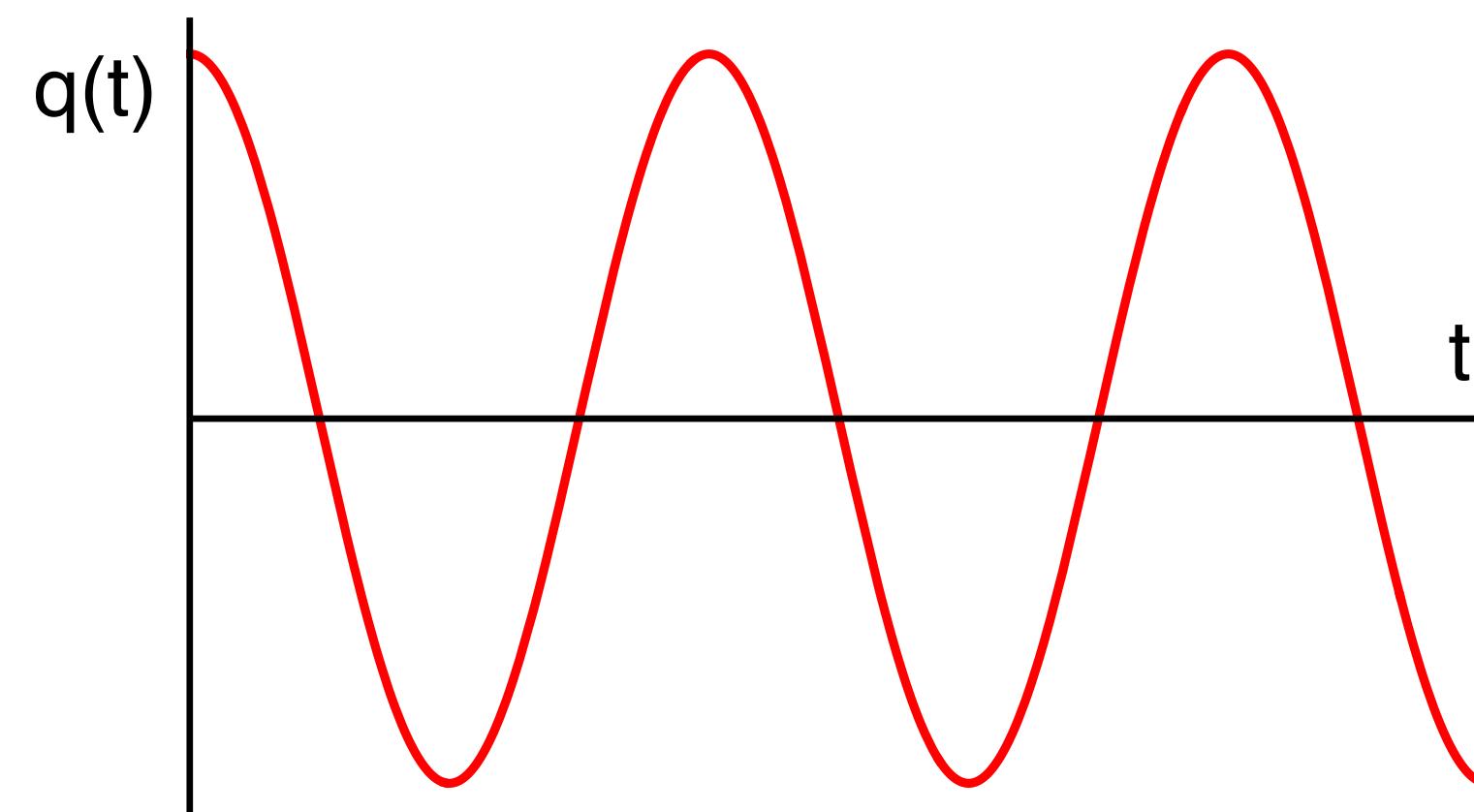


heat transport from lattice dynamics

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Gaussian integral \mapsto Wick theorem



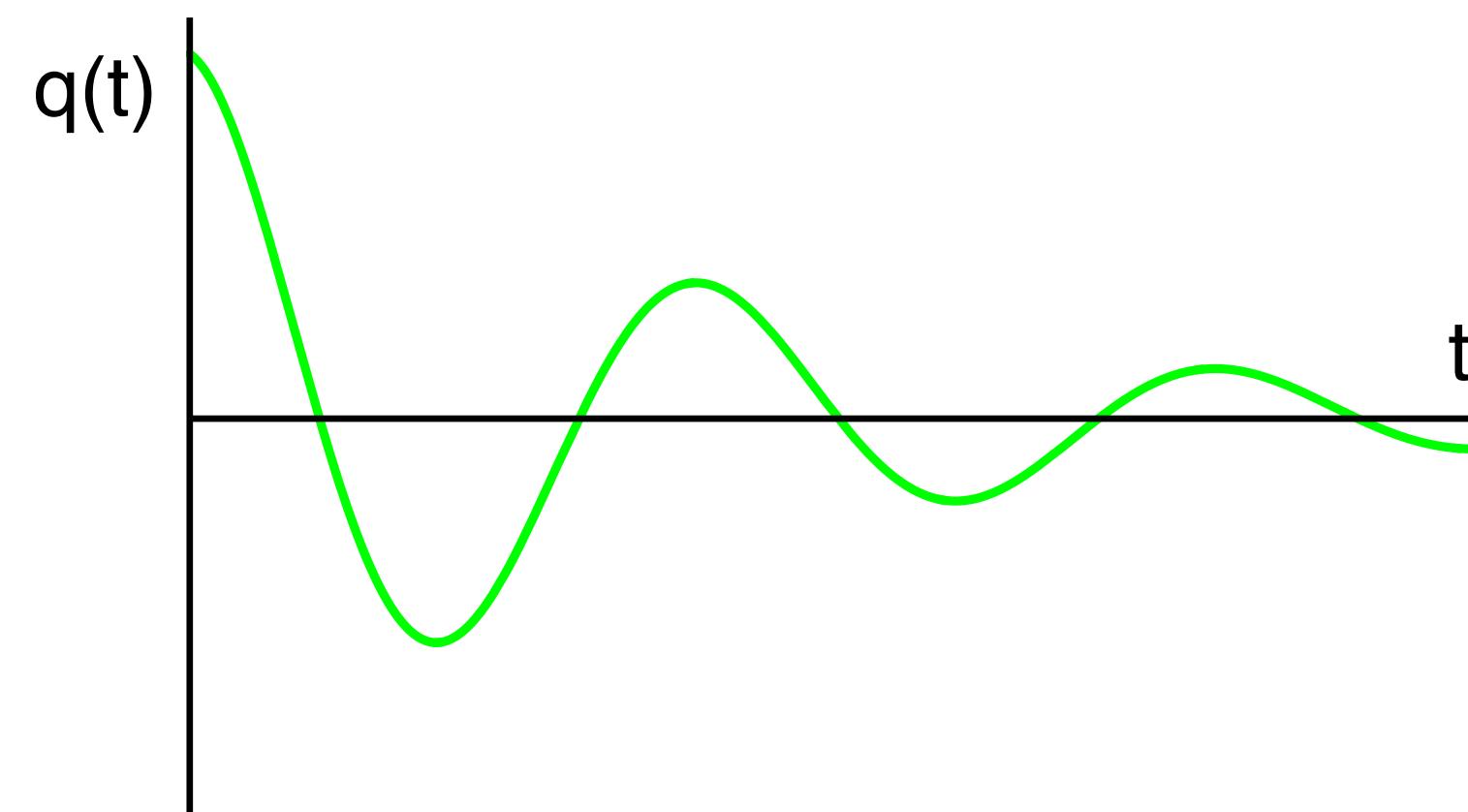
$$\kappa = \infty$$

heat transport from lattice dynamics

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Gaussian integral \mapsto Wick theorem



$$\omega \mapsto \omega + i\gamma$$

\longrightarrow

$$\kappa < \infty$$

heat transport from lattice dynamics

$$\begin{aligned} J_\alpha &= \frac{1}{2} \sum_{ij\beta\gamma} (R_{i\alpha}^\circ - R_{j\alpha}^\circ) \Phi_{i\beta}^{j\gamma} u_{i\beta} \dot{u}_{j\gamma} \\ &= \sum_{nm} v_{nm}^\alpha \sqrt{\omega_n \omega_m} \xi_n \pi_m \\ v_{nm}^\alpha &= \frac{1}{2\sqrt{\omega_n \omega_m}} \sum_{ij\beta\gamma} \frac{R_{i\alpha}^\circ - R_{j\alpha}^\circ}{\sqrt{M_i M_j}} \Phi_{i\beta}^{j\gamma} e_n^{i\beta} e_m^{j\gamma} \end{aligned}$$

$$\kappa = \frac{1}{V} \sum_{nm} c_{nm} (v_{nm})^2 \tau_{nm}^\circ$$

$$\begin{aligned} \tau_{nm}^\circ &= \frac{\gamma_n + \gamma_m}{(\gamma_n + \gamma_m)^2 + (\omega_n - \omega_m)^2} \\ c_{nm} &= \frac{\hbar \omega_m \omega_n}{T} \frac{n(\omega_n) - n(\omega_m)}{\omega_m - \omega_n} \approx k_B \left(\frac{\hbar \omega}{k_B T} \right)^2 \frac{1}{\left(e^{\frac{\hbar \omega}{k_B T}} - 1 \right)^2} \end{aligned}$$



heat transport from lattice dynamics

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In a periodic system

$$v_{nn'} = \delta_{\nu\nu'} \delta_{qq'}$$



heat transport from lattice dynamics

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heat transport from lattice dynamics

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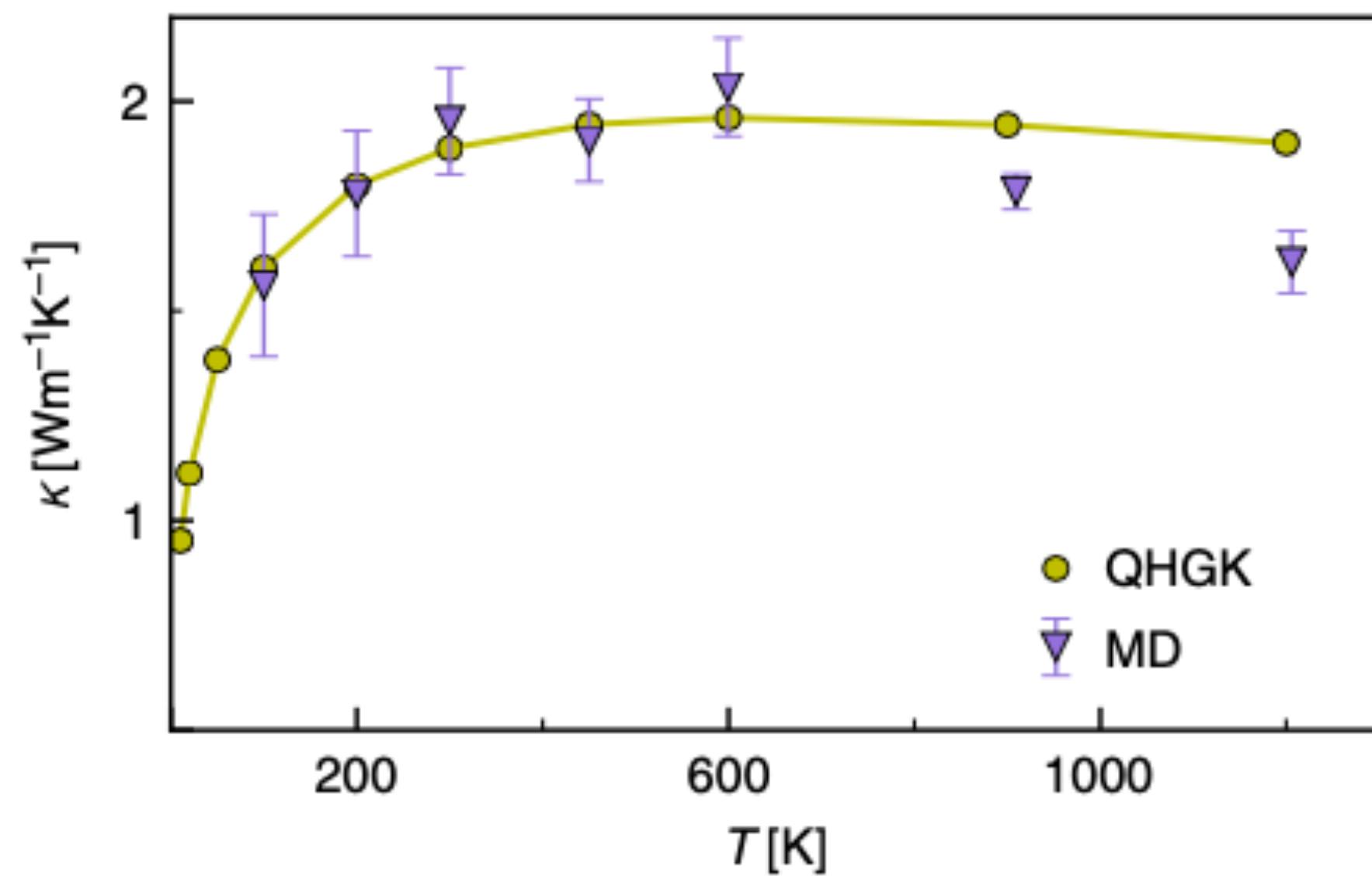
$$\kappa = \frac{1}{V} \sum_{\mathbf{q}\nu} c_\nu(\mathbf{q}) v_\nu(\mathbf{q})^2 \tau_\nu(\mathbf{q})$$

BTE

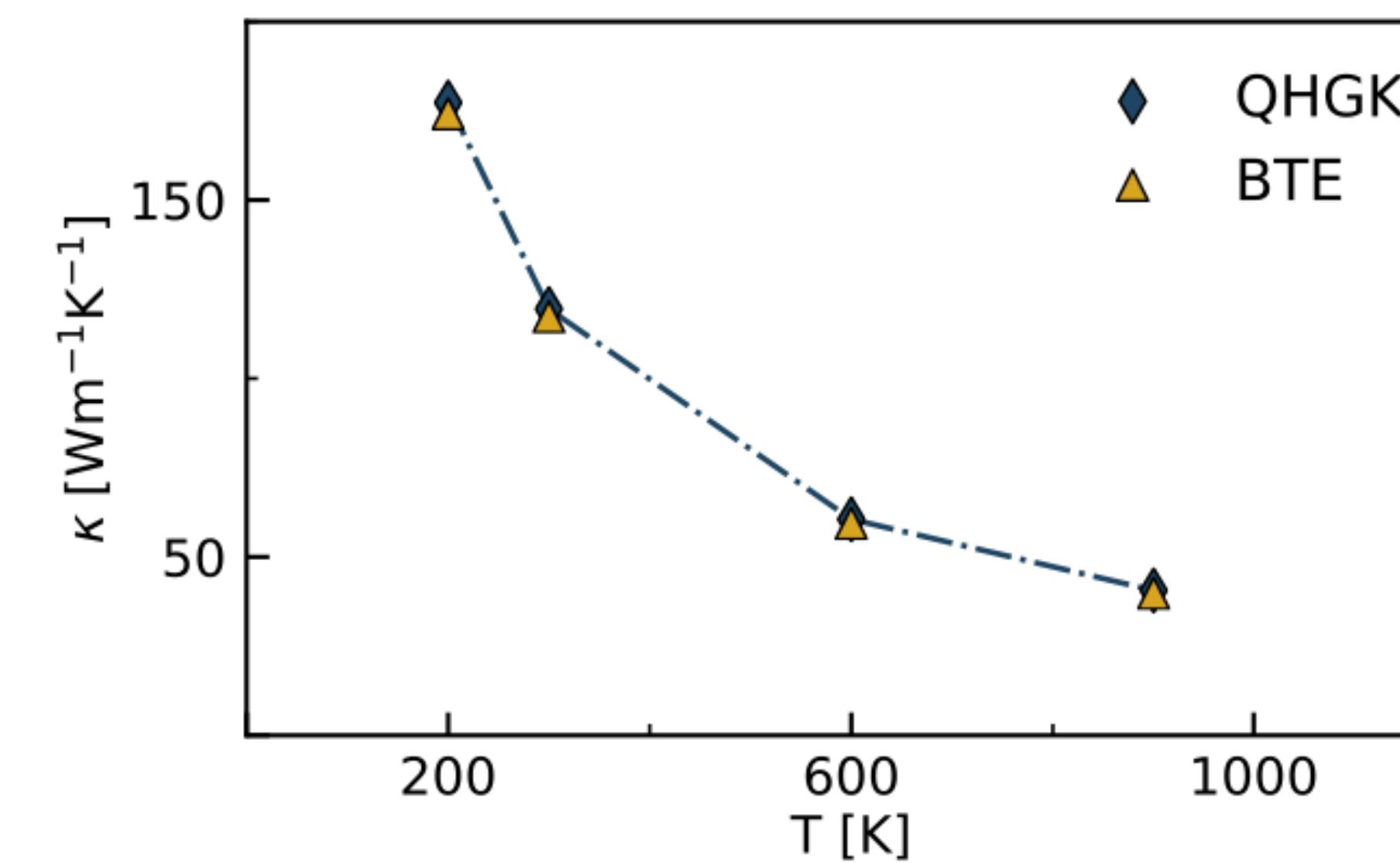


heat transport in c- and a-Si

a-Si



c-Si





ARTICLE

<https://doi.org/10.1038/s41467-019-11572-4>

OPEN

Modeling heat transport in crystals and glasses from a unified lattice-dynamical approach

Leyla Isaeva¹, Giuseppe Barbalinardo², Davide Donadio² & Stefano Baroni^{1,3}

**From Green-Kubo to the full Boltzmann kinetic approach to heat transport
in crystals and glasses**

Alfredo Fiorentino¹, Stefano Baroni^{1,2}

¹ SISSA, Trieste, Italy, ² CNR-IOM, Trieste, Italy



conclusions

- conserved currents are intrinsically ill-defined at the atomic scale
- conservation and extensiveness make transport coefficients independent of the specific microscopic representation of the conserved densities and currents
- this *gauge invariance* of transport coefficients makes it possible to compute thermal transport coefficients from DFT using equilibrium AIMD and the Green-Kubo formalism
- cepstral analysis of the current time series generated by (AI) MD substantially shorten the length of simulations needed to achieve a given target accuracy
- gauge invariance can be leveraged to unify the Green-Kubo and lattice Boltzmann approaches to heat transport in crystalline and amorphous solids



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Ö
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+
to



Aris Marcolongo



Loris Ercole



Riccardo Bertossa



Cesare Malosso



Leyla Isaeva



Giuseppe Barbalinardo



Davide Donadio



Alfredo Fiorentino



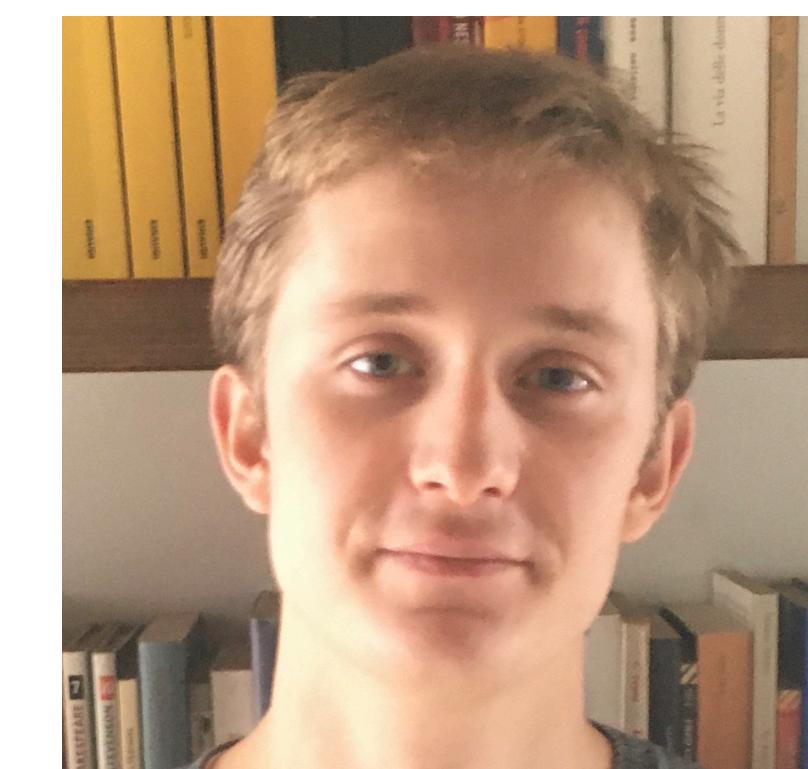
Federico Grasselli



Paolo Pegolo



Davide Tisi



Enrico Drigo

That's all Folks!

these slides soon available at
<http://talks.baroni.me>