Phonon thermal transport in advanced materials

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It is everywhere!





Outline

- \square Why thermal transport?
- \Box Why atomistic-scale simulations?
- $\hfill\square$ Predicting thermal transport in nuclear materials
- $\hfill\square$ Switchable thermal conductivity in metal–organic frameworks
- $\hfill\square$ Uncertainty quantification in the Green–Kubo method



Heat transfer mechanisms



Conduction

Fourier's law

 $\mathbf{J} = -\mathbf{k} \, \nabla T$





Conduction at the atomic scale

Carriers:

- Electrons
- Phonons

$$\mathbf{J} = -\mathbf{k} \, \nabla T$$





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Conduction at the atomic scale

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$$\mathbf{J} = -\mathbf{k} \,
abla \, \mathcal{T}$$





- Macroscopic transport
- Defined over a continuum
- Assumes diffusive transport
- Ignores the (micro)structure of materials
- Solves a classical problem



Conduction at the atomic scale

Carriers:

- Electrons
- Phonons

$$\mathbf{J} = -\mathbf{k} \, \nabla \mathcal{T}$$





 $\frac{\partial f_{\omega}}{\partial t} + \mathbf{v}_{\omega} \nabla f_{\omega} = -\frac{f_{\omega} - f_{\omega}^{0}}{\tau_{\omega}}, f_{\omega}^{0}(T, \omega) = \frac{\hbar \omega D(\omega) \langle n \rangle_{BE}(T, \omega)}{4\pi}$

 Ab initio calculations e.g. density functional theory (DFT)

$$\kappa = \int_{0}^{\omega \max} \frac{\hbar \omega D(\omega)}{\partial T} \frac{\partial \langle n \rangle_{BE}}{\partial T} v_{g}^{2} \tau d\omega$$

Matthiessen's rule





Calculating intrinsic materials properties

Ab initio (but also classical MD)

► E.g. silicon



- LA longitudinal acoustic mode
- TA transverse acoustic mode
- LO longitudinal optical mode
- TO transverse optical mode

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$$\kappa = \int_{0}^{\omega_{max}} \hbar \omega D(\omega) \frac{\partial \langle n \rangle_{BE}}{\partial T} v_{g}^{2} \tau d\omega$$

In the scale of heat transfer simulations



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- Evolution of defects in nuclear graphite while in service as a direct result of radiation
- In materials which undergo fission (e.g. UO2), isotopic byproducts affect heat transfer at a microscopic level







The insight we gain can be incorporated into approaches for quantitatively predicting the lattice thermal conductivity based on solving the Boltzmann transport equation

Molecular dynamics simulations

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15

13

The rate of heat transfer

Molecular dynamics — The Green–Kubo method







Natural heat fluctuations around equilibrium



The rate of heat transfer

Molecular dynamics — The Green–Kubo method



 $\textit{HCACF}_{J}(\tau) = \langle J(t)J(t+\tau) \rangle$

$$\kappa = rac{V}{k_B T^2} \int_0^\infty HCACF_{
m J}(au) d au$$

$$HCACF_{J}(\tau) \approx HCACF_{J_m} \equiv \sum_{n=0}^{N-m} \frac{J_n J_{m+n}}{N-m}$$







Results



(de S. Oliveira, et al., 2015)



Thermal resistance along the C-axis

$$rac{1}{ au} = rac{1}{ au_{ ext{impurities}}} + rac{1}{ au_{ ext{lattice}}} + rac{1}{ au_{ ext{defects}}} + \cdots$$

 $r_{\rm defective} = r_{\rm perfect} + r_{\rm defects}$



Thermal resistance along the C-axis





 $r_{6 \text{ interstitials}} = 0.16 \pm 0.03 (mK/W)$

 $6 \cdot r_{1 \text{ interstitial}} = 0.04 \pm 0.17 \ (mK/W)$



Thermal resistance along the C-axis

Actual:

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Predicting thermal properties in nuclear materials Gray approximation



► Neutron transport code (Rattlesnake) modified for phonon transport

$$\begin{split} \Lambda \hat{\boldsymbol{\Omega}} \cdot \nabla \left[I(\mathbf{r}, \hat{\boldsymbol{\Omega}}) \right] &= I^0(\mathbf{r}) - I(\mathbf{r}, \hat{\boldsymbol{\Omega}}) \\ &\frac{1}{4\pi} \int_{4\pi} \int_{4\pi} I(\mathbf{r}, \hat{\boldsymbol{\Omega}}) = I^0(\mathbf{r}) \\ I^0(T, \mathbf{r}) &= \frac{1}{4\pi} \sum_p \int_0^{\omega_{max}} v_g \hbar \langle n \rangle_{BE}(\omega, T) \omega D(\omega) d\omega \end{split}$$



Informing the Boltzmann Transport equation

UO₂ 2-D phonon dispersion

Xenon properties from MC





Informing the Boltzmann Transport equation







(Harter, et al., 2015)



Conclusions

- Tease apart κ from defects
- Not all defects are created equal:



- Clustered interstitial defects are stable and strongly detrimental to thermal transport along the *c*-axis
- ► Thermal resistance of platelets is approximately 4 times larger than their constituent number of interstitials long the *c*-axis
- Xenon lowers thermal conductivity at a constant ratio with temperature change
- Developed an efficient and versatile method for predicting phonon transport that can be informed by atomistic simulations

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Metal-organic frameworks





Heat transfer in metal–organic frameworks

- Large surface area convenient for gas sorption applications
- Poor thermal conductivity
- Heat dissipation for sorption applications



A MOF-5 based H_2 adsorption tank developed at Oregon State University



Simple κ model



$$J_{xx} = -\frac{2C_{\text{linker}}\Delta T}{L_y L_z}$$
$$= -2C_{\text{linker}}\frac{L_x}{L_y L_z}\frac{\partial T}{\partial x}$$
$$= -\kappa_{xx}\frac{\partial T}{\partial x}$$

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Then:

$$\dot{Q} = -2C_{\text{linker}}\Delta T, \ C_{\text{linker}} = \frac{1}{2(R_N + R_L)} \qquad \kappa = 2C_{\text{linker}} \begin{pmatrix} \frac{L_x}{L_y}L_z & 0 & 0\\ 0 & \frac{C_{\text{chain}}}{C_{\text{linker}}}\frac{L_y}{L_x}L_z & 0\\ 0 & 0 & \frac{L_z}{L_x}L_y \end{pmatrix}$$



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MIL-53

Simple κ model — MD calculations





$$\kappa = 2C_{\text{linker}} \begin{pmatrix} \frac{L_x}{L_y L_z} & 0 & 0\\ 0 & \frac{C_{\text{chain}}}{C_{\text{linker}}} \frac{L_y}{L_x L_z} & 0\\ 0 & 0 & \frac{L_z}{L_x L_y} \end{pmatrix}$$

 $\begin{aligned} R_n + R_l &= 0.083 \text{ K/W (along a)} \\ \text{and } 0.138 \text{ K/W (along c)} \\ R_N &= 0.015 \text{ K/W} \end{aligned}$



Ab initio calculations — dispersion relations





Ab initio calculations — dispersion relations





"Radiant flux"

$$\phi(\mathbf{k},\omega,T) \propto |\mathbf{v}_g(\mathbf{k},\omega)| \hbar \omega(\mathbf{k}) D(\mathbf{k},\omega) \langle n \rangle_{BE}(\omega,T)$$





C

"Radiant flux"

$$\phi(\mathbf{k},\omega,T) \propto |\mathbf{v}_g(\mathbf{k},\omega)| \hbar \omega(\mathbf{k}) D(\mathbf{k},\omega) \langle n \rangle_{BE}(\omega,T)$$





- A first look at κ in MIL-53
- Switchable κ conductivity
- Observed rattler modes
- Phonon focusing effects



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The equilibrium molecular dynamics approach



$$\mathit{HCACF}_{\mathsf{J}}(\tau) = \langle \mathsf{J}(t) \mathsf{J}(t+\tau) \rangle$$

$$\kappa = \frac{V}{k_B T^2} \int_0^\infty HCACF_{\rm J}(\tau) d\tau$$

$$HCACF_{\mathbf{J}}(\tau) \approx HCACF_{\mathbf{J}_m} \equiv \sum_{n=0}^{N-m} \frac{\mathbf{J}_n \mathbf{J}_{m+n}}{N-m}$$



Applications of the Green-Kubo

Thermal conductivity

$$\kappa_{
m xx} = rac{V}{k_B T^2} \int_0^\infty \; \left< {f J}_{
m xx}(t) {f J}_{
m xx}(t+ au)
ight> \; d au$$

Diffusivity

$$D_{
m xx} = \int_0^\infty ig \langle {f v}_{
m xx}(t) {f v}_{
m xx}(t+ au) ig \, d au$$

Viscosity

$$\eta_{xy} = rac{V}{k_{B}T} \int_{0}^{\infty} \left< \mathbf{P}_{xy}(t) \mathbf{P}_{xy}(t+ au) \right> d au$$



FIG. 1. Heat flux autocorrelation functions of q, q⁰, and q⁰ for N = 108 and $T = 0.546 e/k_B$. Only the initial portions of the correlation functions are shown.

(Ladd, Moran & Hoover, 1986)



0.45

FIG. 1. The correlation function $C_{c}(t)$ in a lattice-ps model of the OW(110) system at $\theta = 0.4$ stan t = 0.8337. This intermediates the time instruction of the order of the order



(Chen, Smit & Bell, 2009)







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Approaches commonly used

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10648 atom system

Approaches commonly used





Approaches commonly used





Approaches commonly used

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Approach examples



Fig. 1 Influence of the simulation durations on the determination of the GK viscosity for four different values.

(Van-Oanh, Hourier & Rousseau, 2009)

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Fig. 4. (a) Decay of the normalized HCACF for MOF-5 and (b) its integral (the thermal conductivity) at a temperature of 300 K.

(Huang, McGaughey & Kaviany, 2006)





Approaches commonly used



(d. S. Oliveira & Greaney, 2015)

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The heat current autocorrelation function





The heat current autocorrelation function





Random walk



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Random walk



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Random walk





Random walk in the integrated autocorrelation function



Random walk in the integrated autocorrelation function



(d. S. Oliveira & Greaney, 2017)



 $\sigma_n \sqrt{t \, dt}$

Random walk in the integrated autocorrelation function



(d. S. Oliveira & Greaney, 2017)



Random walk in the integrated autocorrelation function

 Fit error distribution to determine if slow decay processes are present in the ACF



(d. S. Oliveira & Greaney, 2017)

Noise coupled with overall trend of heat flux

Conclusions

- ▶ The noise of the flux autocorrelation grows as a random walk
- Noise coupled with overall trend of heat flux
- One simulation suffices to quantify the error
- Determine optimal simulation time on-the-fly
- Trade smaller sampling step for longer simulation
- Fit error distribution to determine if slow decay processes are present in the ACF



Collaborators

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