

Ab Initio Molecular Dynamics

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Max Planck Institute for Polymer Research*

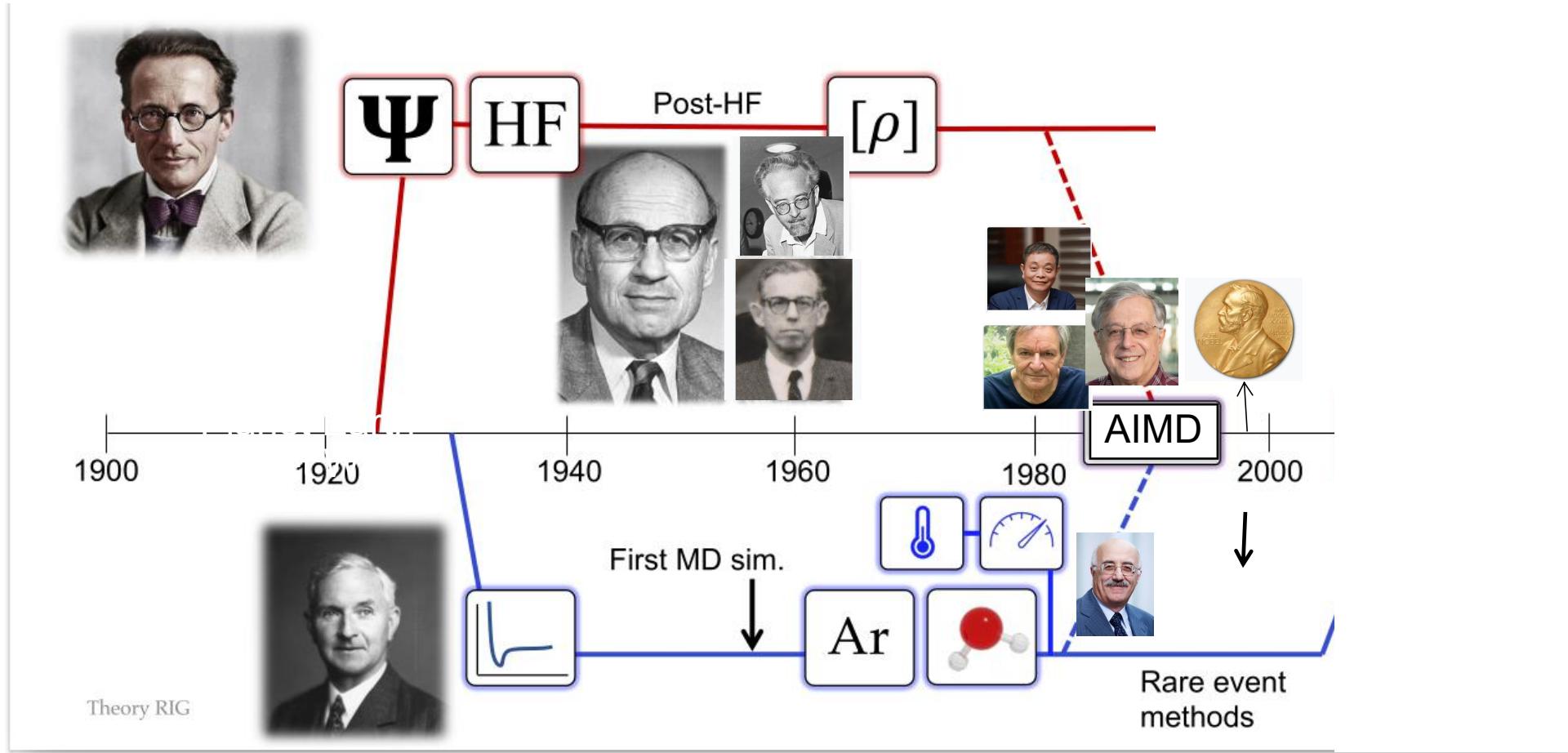
*Hands-On Workshop, Shanghai University
Nov 2025*

I've been in your place (many years ago)



Image of Neo being connected to the Matrix

Standing on the Shoulder of Giants

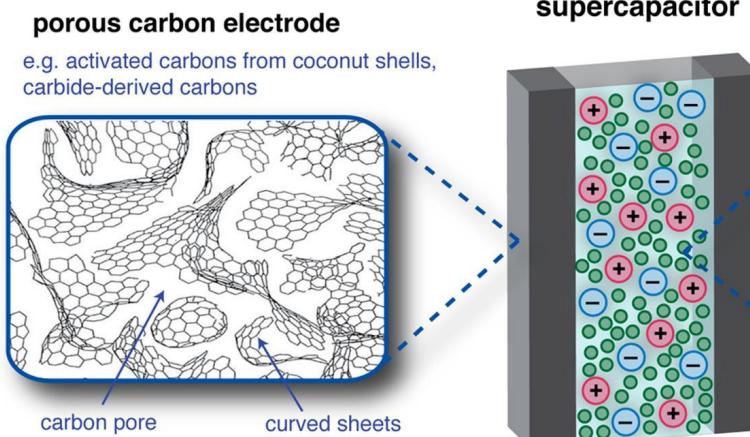


(Slide courtesy of Samuel Brookes@Cambridge)

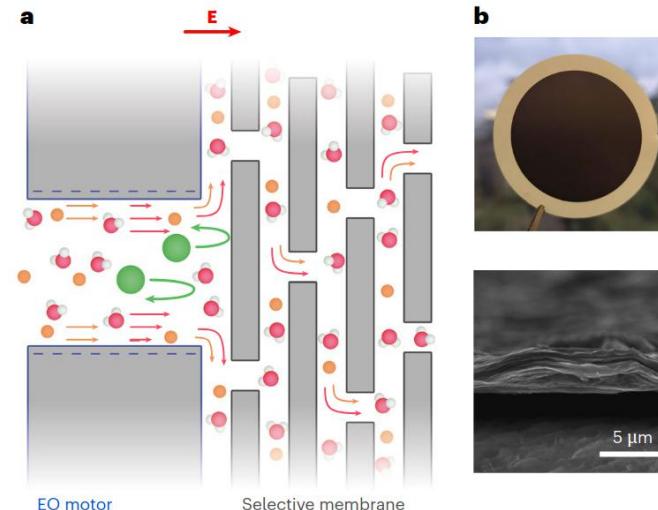
Aqueous Electrochemical Devices

Aqueous electrochemical devices play a critical role in the green energy transition

Supercapacitors



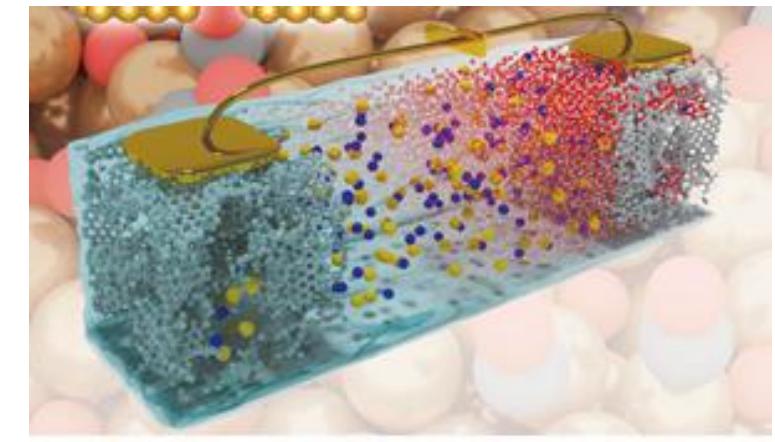
Water desalination/purification



Forse, et al. *JACS*, **138**, 5731-5744 (2016)

Abdelghani-Idrissi, et al. *Nat. Mat.* **24**, 1109-111 (2025)

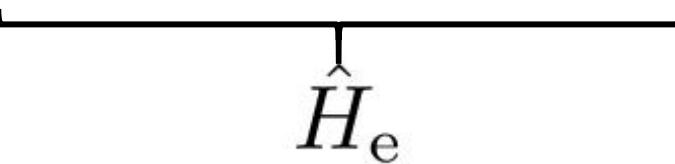
Aqueous batteries



Koper, et al. *Chem. Rev.* **122**, 10579 (2022)

The Hamiltonian that Rules Everything

“The underlying physical laws necessary for the mathematical theory of large parts of physics and the whole of chemistry are thus completely known ...” (Dirac, 1929)

$$\hat{H} = \hat{T}_{\text{nuc}} + \hat{V}_{\text{nuc-nuc}} + \hat{T}_{\text{e}} + \hat{V}_{\text{nuc-e}} + \hat{V}_{\text{e-e}}$$


$$\hat{T}_{\text{nuc}} = - \sum_I^N \frac{\vec{\nabla}_I^2}{2M_I}$$

$$\hat{V}_{\text{nuc-nuc}} = \frac{1}{2} \sum_{I \neq J}^N \frac{Z_I Z_J}{|\vec{R}_I - \vec{R}_J|}$$

$$\vec{V}_{\text{nuc-e}} = - \sum_i^n \sum_I^N \frac{Z_I}{|\vec{r}_i - \vec{R}_I|} \quad \hat{V}_{\text{e-e}} = \frac{1}{2} \sum_{j \neq i}^n \frac{1}{|\vec{r}_i - \vec{r}_j|}$$

$$\hat{T}_{\text{e}} = - \sum_i^n \frac{\vec{\nabla}_i^2}{2}$$

“.... the difficulty is only that the exact application of these laws leads to equations much too complicated to be soluble”

Born-Huang Expansion

We want to solve the time-independent Schrödinger equation

$$\hat{H}\Psi = E\Psi$$

Assuming we could solve first the electronic problem at fixed nuclei, we would get

$$\hat{H}_e\Phi_s(\mathbf{R}, \mathbf{r}) = E_s(\mathbf{R})\Phi_s(\mathbf{R}, \mathbf{r}),$$

for each nuclei position.

Then we use these solutions as a basis,

$$\Psi_\nu = \sum_s \Lambda_s^\nu(\mathbf{R})\Phi_s(\mathbf{R}, \mathbf{r}),$$

So far we haven't made any additional approximation, so if we introduce the previous equation into the first one, we obtain

Born-Oppenheimer Approximation

... this beauty

$$E_t \Lambda_t^\nu(\mathbf{R}) = [\hat{T}_{\text{nuc}} + \hat{V}_{\text{nuc-nuc}} + E_t(\mathbf{R})] \Lambda_t^\nu(\mathbf{R}) - \sum_s \sum_I \frac{1}{2M_I} [2\vec{A}_{ts,I}^{(1)}(\mathbf{R}) \vec{\nabla}_I + A_{ts,I}^{(2)}(\mathbf{R})] \Lambda_s^\nu(\mathbf{R})$$

where

$$A_{ts,I}^{(2)}(\mathbf{R}) = \int d\mathbf{r} \Phi_t^*(\mathbf{R}, \mathbf{r}) \vec{\nabla}_I^2 \Phi_s(\mathbf{R}, \mathbf{r}) \quad \vec{A}_{ts,I}(\mathbf{R}) = \int d\mathbf{r} \Phi_t^*(\mathbf{R}, \mathbf{r}) \vec{\nabla}_I \Phi_s(\mathbf{R}, \mathbf{r})$$

If we neglect the last two terms, (i.e. movement of nuclei can't induce electronic excitations), we obtain

$$\hat{H}_{\text{nuc}} \Lambda_0^\nu(\mathbf{R}) = [\hat{T}_{\text{nuc}} + V_{\text{nuc},0}^{\text{BO}}(\mathbf{R})] \Lambda_0^\nu(\mathbf{R})$$

where

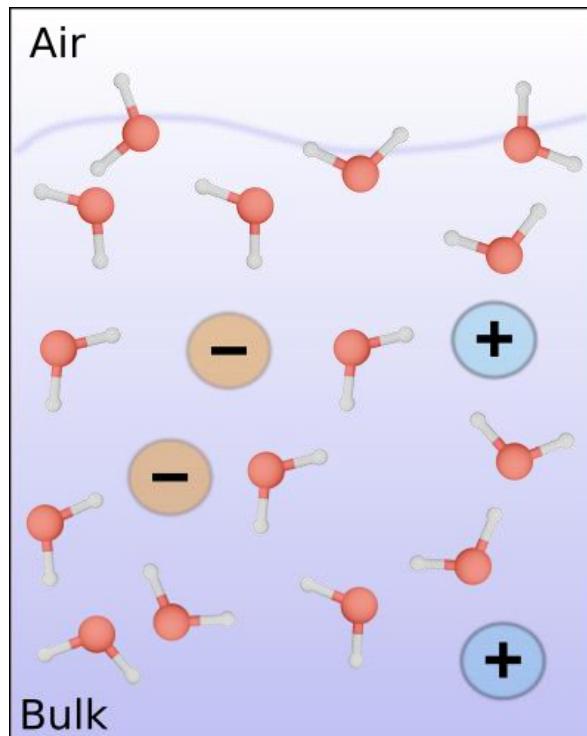
$$V_{\text{nuc},s}^{\text{BO}}(\mathbf{R}) := E_s(\mathbf{R}) + V_{\text{nuc-nuc}}(\mathbf{R})$$

We effectively separated the nuclei and electronic problems!

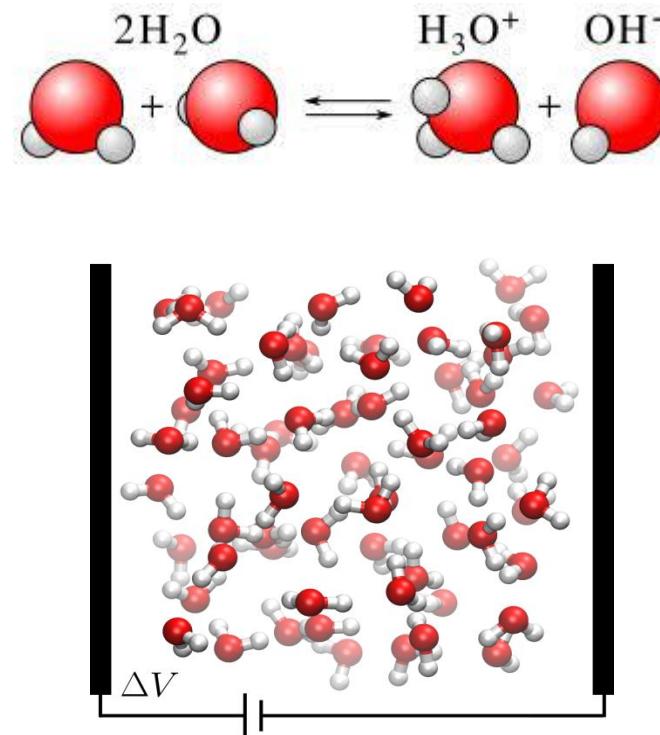
Note that nuclei are still quantum particles!

Chemistry/Physics doesn't Stop at the Minimum

Interfacial structure of simple
electrolyte at aqueous
solutions



Water dissociation
under electric fields



Equilibrium Properties at Finite Temperature

We first consider the classical nuclei case. The canonical partition function is defined as

$$Q^{\text{CL}}(\beta) = \frac{1}{(2\pi\hbar)^{3N}} \int d\mathbf{R} d\mathbf{P} e^{-\beta H_{\text{nuc}}(\mathbf{R}, \mathbf{P})}$$

$$\beta = \frac{1}{k_{\text{B}}T}$$

$$k_{\text{B}}T/\hbar\omega \gg 1$$

Equilibrium averages are computed as

$$\langle A \rangle_{\text{CL}} = \frac{1}{(2\pi\hbar)^{3N} Q^{\text{CL}}(\beta)} \int d\mathbf{R} d\mathbf{P} A(\mathbf{R}, \mathbf{P}) e^{-\beta H_{\text{nuc}}(\mathbf{R}, \mathbf{P})}$$

Evaluating the previous integral on a grid is a bad idea (why?), so we normally ‘sample’ it.

if we sample from the Boltzmann distribution, $e^{-\beta H_{\text{nuc}}(\mathbf{R}, \mathbf{P})}$, we can compute thermal averages as

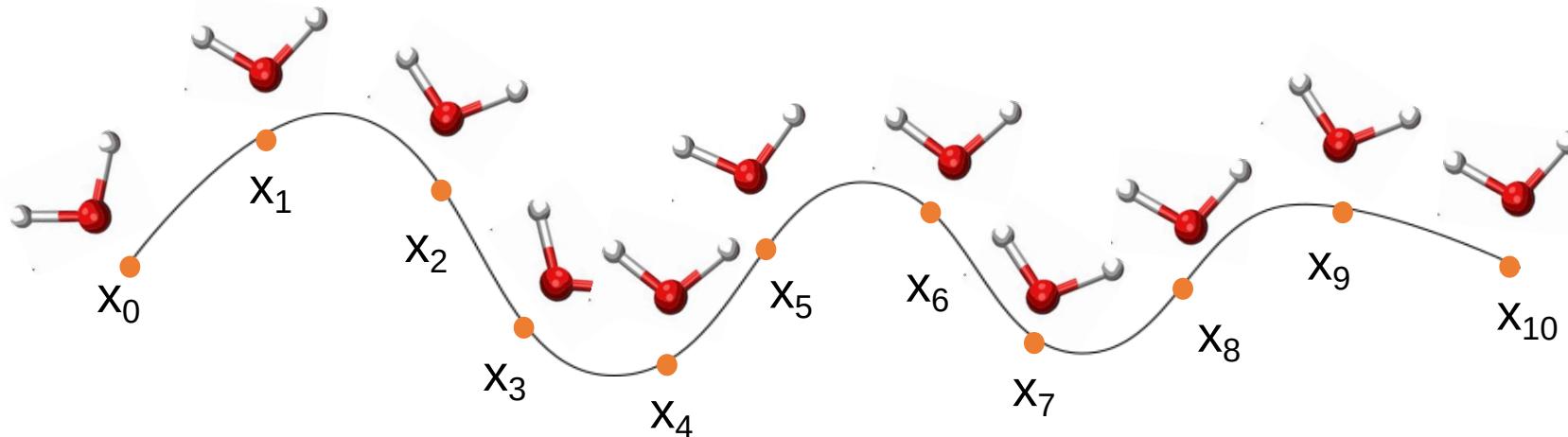
$$\langle A(R) \rangle = \frac{1}{N} \sum_i^N A(R^i)$$

Molecular Dynamics

Hamilton's equations of motion (EOM)

$$\dot{\vec{R}}_I = \frac{\partial H_{\text{nuc}}}{\partial \vec{P}_I} = \frac{\vec{P}_I}{2M}$$

$$\dot{\vec{P}}_I = -\frac{\partial H_{\text{nuc}}}{\partial \vec{R}_I} = -\nabla V_{\text{nuc}}^{\text{BO}}$$



Ergodic hypothesis: *The system will visit every accessible region of phase space, given enough time.*

$$\langle A \rangle = \lim_{\tau \rightarrow \infty} \frac{1}{\tau} \int_0^\tau dt A(t)$$

EOM in Practice

The integration of the equations of motion (EOM) is normally done using the velocity Verlet algorithm

velocity update

$$\dot{R}\left(t + \frac{\Delta t}{2}\right) = \dot{R}(t) + \frac{F(t)}{M} \frac{\Delta t}{2}$$

position update

$$R(t + \Delta t) = R(t) + \dot{R}\left(t + \frac{\Delta t}{2}\right) \Delta t$$

force update

Evaluate $F(t + \Delta t)$ from $R(t + \Delta t)$

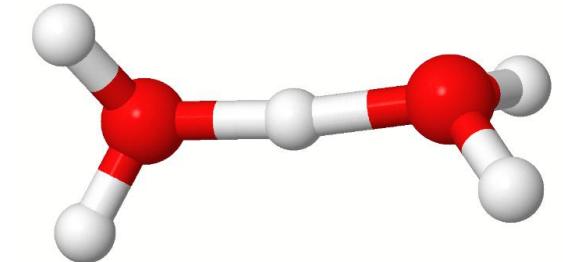
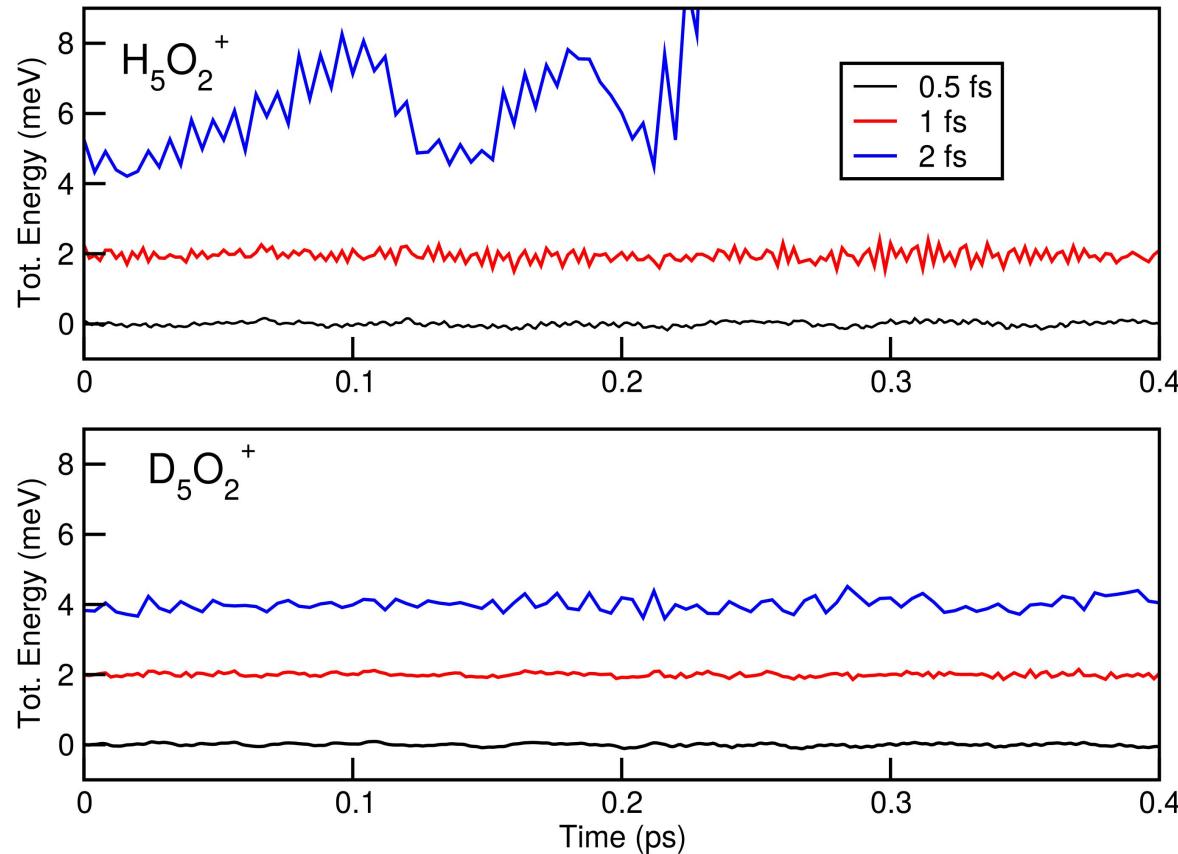
velocity update

$$\dot{q}(t + \Delta t) = \dot{q}\left(t + \frac{\Delta t}{2}\right) + \frac{F(t + \Delta t)}{M} \frac{\Delta t}{2}$$



EOM in Practice

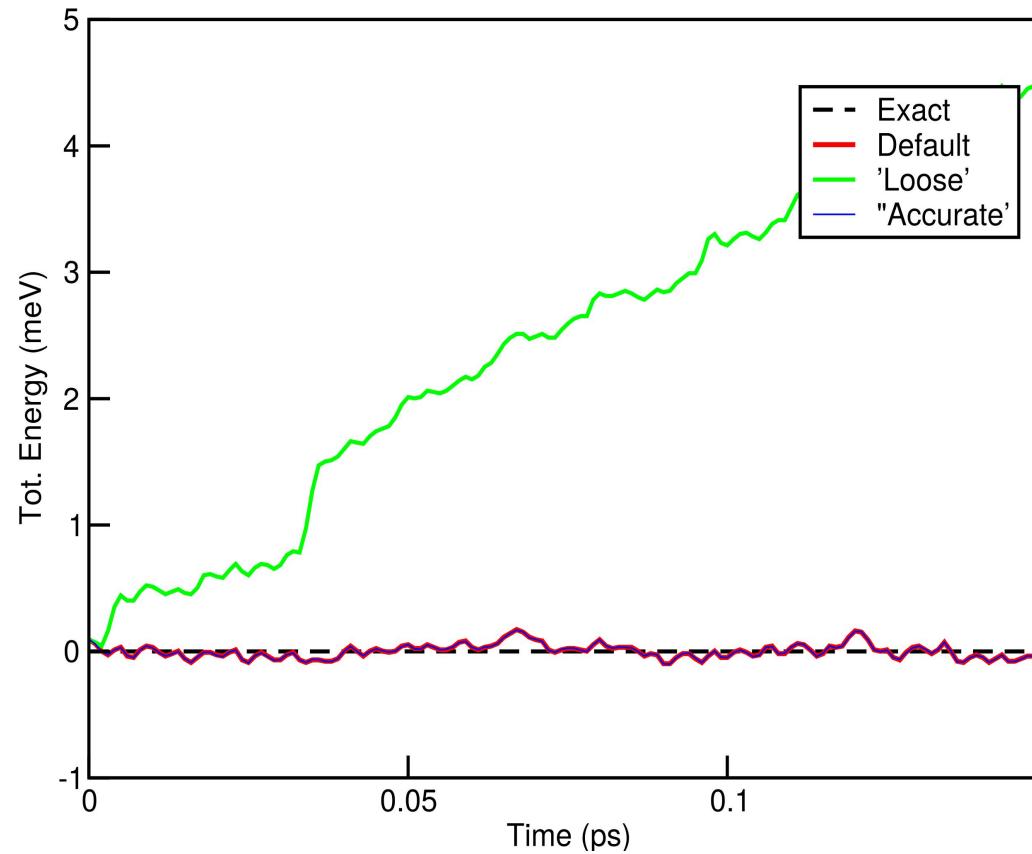
Selection of time step:



Rule of thumb: time step $\approx 1/10$ of the maximum frequency

SCF Accuracy

Selection of SCF accuracy (in FHI-aims)



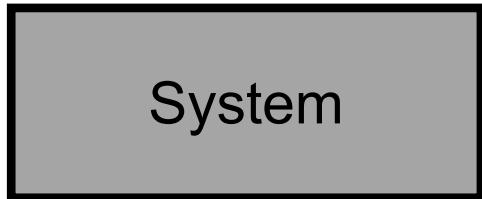
Stability of MD simulations depends critically on the accuracy of the forces

Ensembles

We can perform simulations on different ensembles. The most common are

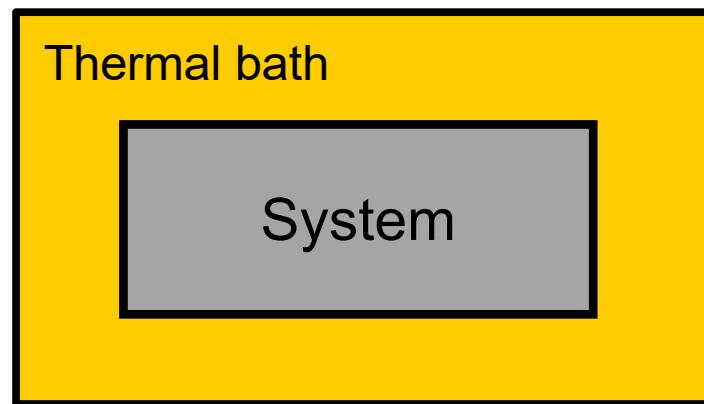
Microcanonical
ensemble (NVE)

$N, V, E = \text{constant}$



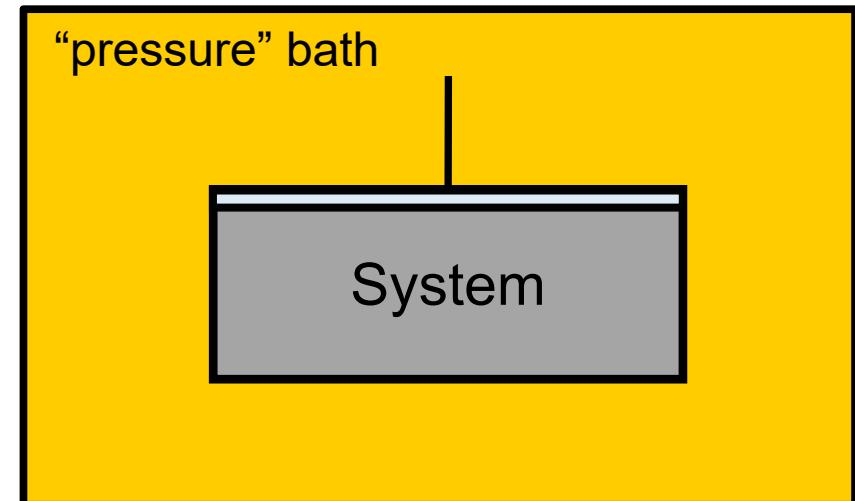
Canonical
ensemble (NVT)

$N, V, T = \text{constant}$



Isothermal-Isobaric
ensemble (NPT)

$N, P, T = \text{constant}$



Thermostat

Barostat and
Thermostat (not shown)

Thermostat

Extended Lagrangian approach (deterministic):

Langevin Thermostats (stochastic):

Seminal idea from Nose

Simplest formulation:

$$H^{\text{Nose}} = H_{\text{nuc}} + \frac{P_s^2}{2Q} + gkT \ln(s)$$

= additional degree
of freedom

More complex and reliable implementations:

- Nose-Hoover
- Nose-Hoover chains

More complex and reliable implementations:

- Stochastic velocity rescaling
- Colored noise thermostat

Nose, J. Chem. Phys. 81, 511 (1984)

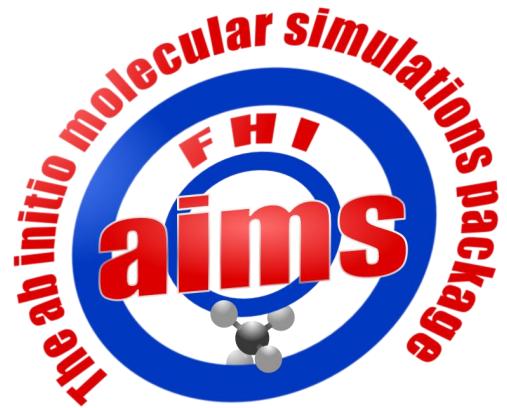
Hoover, Phys. Rev. A. 31, 1695 (1985)

Martyna, Klein, Tuckerman, J. Chem. Phys. 97 2635 (1992)

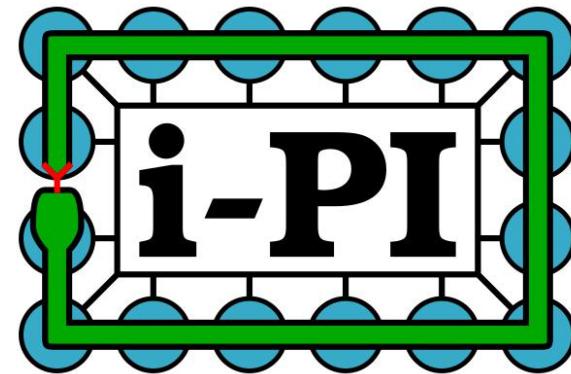
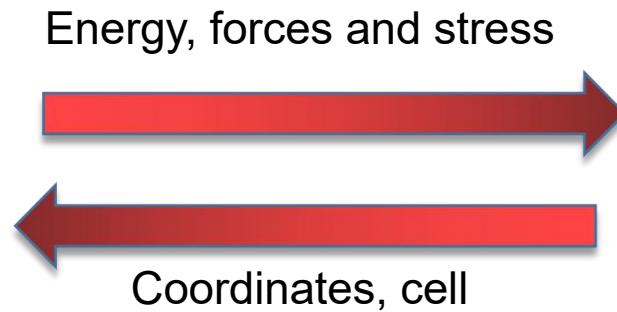
Bussi, Donadio, Parrinello, J. Chem. Phys. **126**, 014101 (2007)

Ceriotti, Bussi, Parrinello, J. Chem. Theory Comput. 6, 1170 (2010)

A Universal Force Engine



Compute forces and potential energy



Evolve atomic positions and velocities

Apply thermostat

Apply barostat

....

A Universal Force Engine

Molecular Dynamics

Geometry Optimization

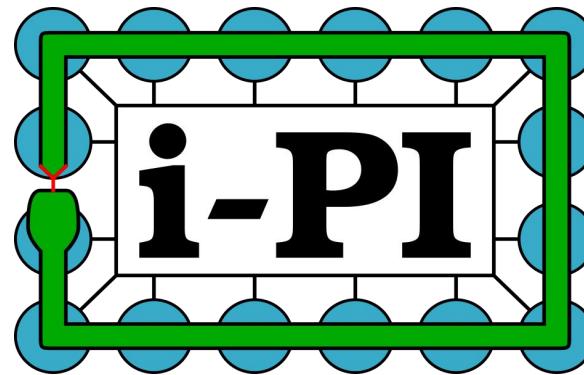
Phonons

Nudge Elastic Band

Replica Exchange Molecular
Dynamics

Metadynamics

Committee Models



Path Integral Molecular
Dynamics

Ring Polymer
Molecular Dynamics

Centroid Molecular
Dynamics

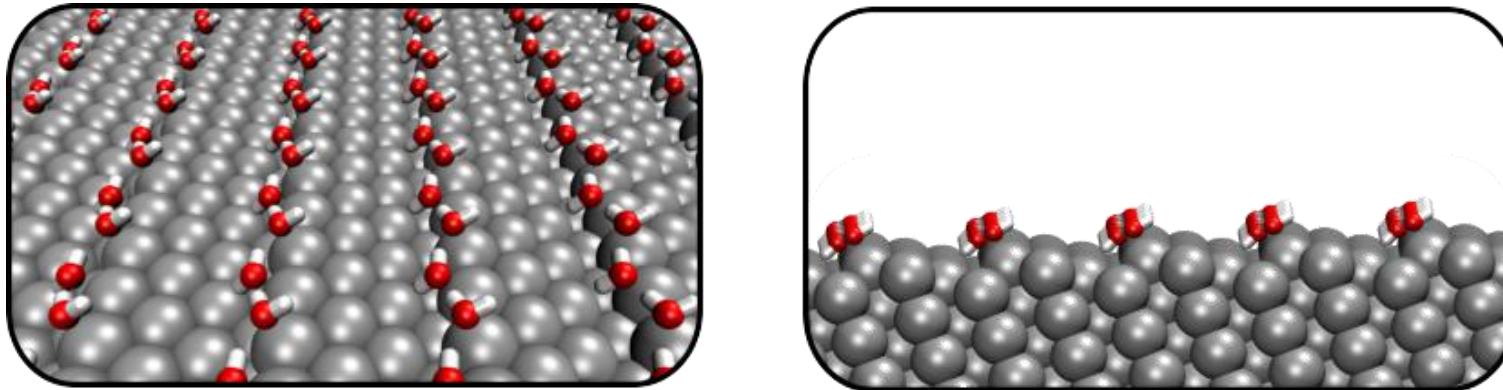
Ring Polymer
Instantons

See full list of features at: <https://docs.ipi-code.org/features.html>

Litman, et al. J. Chem. Phys. **161**, 062504 (2024)

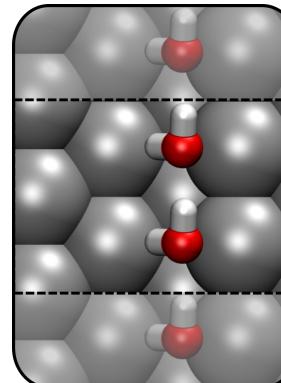
Water Dissociation in Pt(221)

Model system to study water dissociation at metallic interfaces.

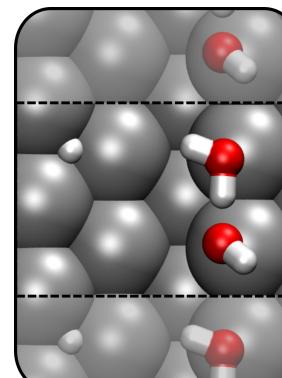


Geometry optimization leads to:

Intact



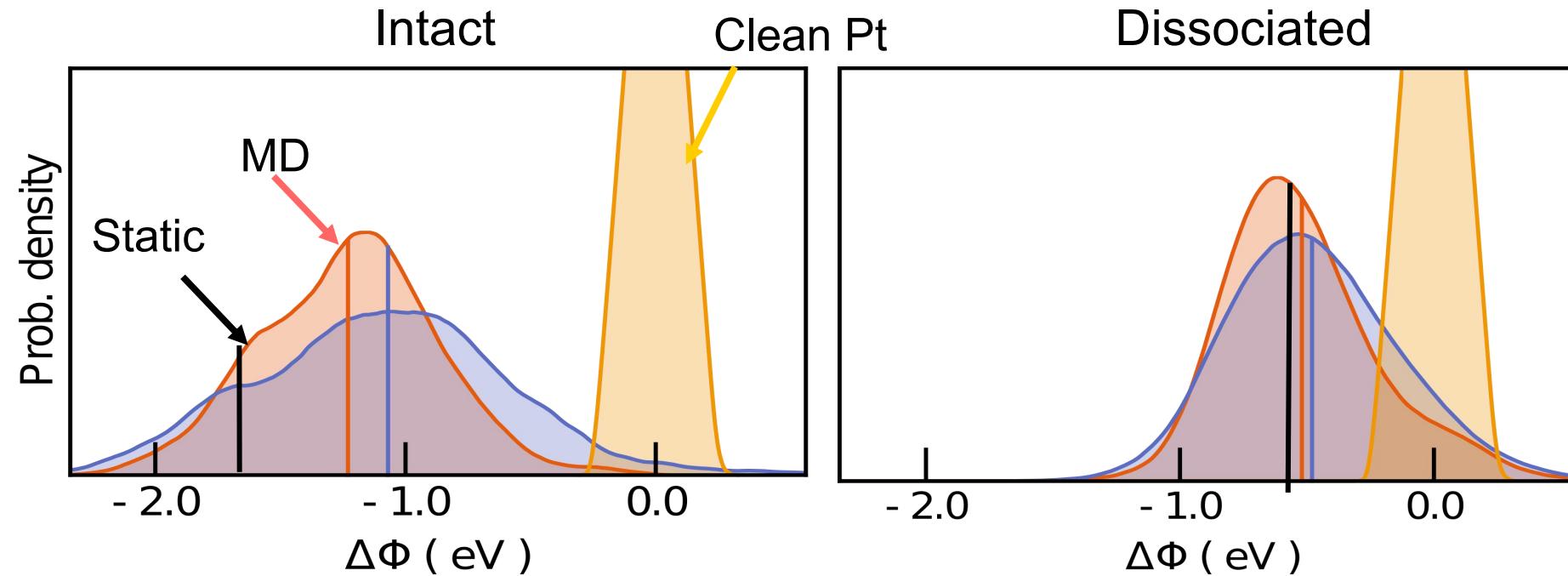
Dissociated



Water Dissociation in Pt(221)

PBE+TS, 300K

Work function change: $\Delta\Phi = \Phi_{\text{FullSystem}} - \Phi_{\text{CleanPt}}$



Temperature effects change work function by more than 0.5 eV!

But, sometimes static estimates are really good!

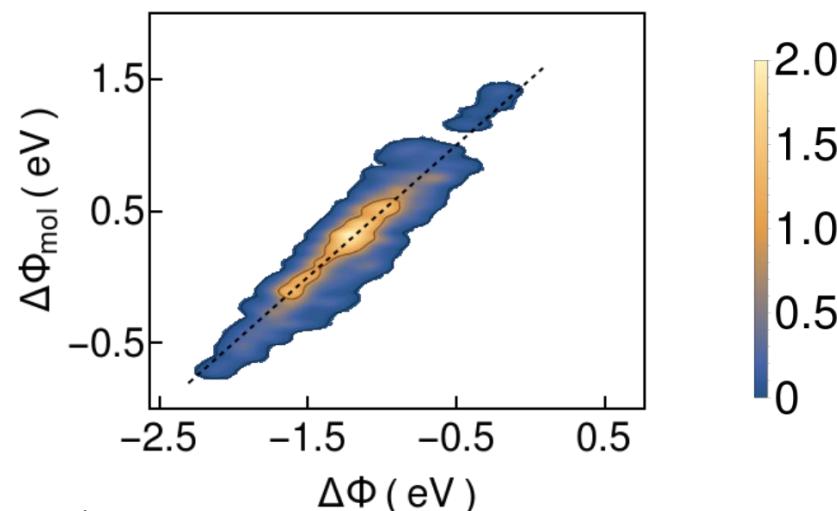
Water Dissociation in Pt(221)

Work function change decomposition:

$$\Delta\Phi = \Delta\Phi_{\text{BD}} + \Delta\Phi_{\text{mol}}$$

↓ ↓ ↓

Total work function change Caused by charge rearrangement on the interface (“Bond dipole”) Caused by molecular dipole perpendicular to the surface

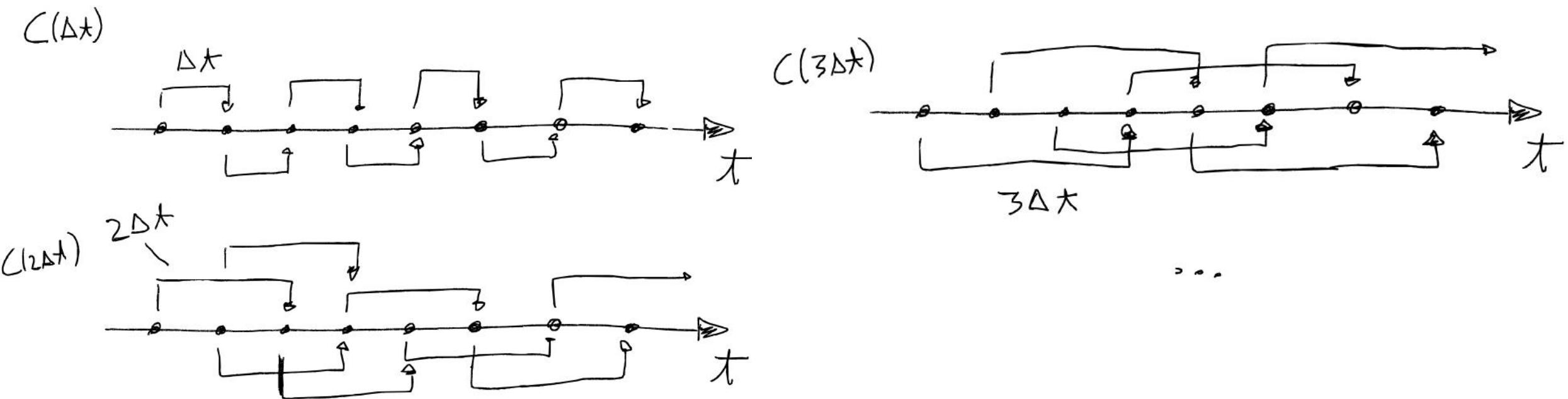


Litman, et al. J. Chem. Phys. 148, 102320 (2018)

Dynamical Properties

Molecular dynamics simulations enable the calculation of time correlations functions (TCF)

$$c_{AB}^{\text{CL}}(t) = \frac{1}{2\pi\hbar Q^{\text{CL}}(\beta)} \int dR dP A(0)B(t) e^{-\beta H(R,P)} = \langle A(0)B(t) \rangle$$



Time Correlation Functions

$$c_{AB}^{\text{CL}}(t) = \frac{1}{2\pi\hbar Q^{\text{CL}}(\beta)} \int dR dP A(0) B(t) e^{-\beta H(R, P)}$$

Spectroscopy:

<i>IR</i>	<i>Raman</i>	<i>Sum Frequency Generation</i>
$I_{\text{IR}}(\omega) \propto \int_0^\infty dt e^{-i\omega t} \langle \mu(t) \mu(0) \rangle$	$I_{\text{Raman}}(\omega) \propto \int_0^\infty dt e^{-i\omega t} \langle \alpha(t) \alpha(0) \rangle$	$\chi^{(2)}(\omega) \propto i \int_0^\infty dt e^{-i\omega t} \langle \alpha(t) \mu(0) \rangle$

Reaction rates:

$$k_{\text{Q}}(T) = c_{\text{fs}}(t) \Big|_{t>\tau} = \frac{1}{(2\pi\hbar)^F} \int dp \int dq e^{-\beta H} \delta[s(R)] v_s(R) h[s(R(t))]$$

Transport properties:

$$D = \frac{1}{6N} \lim_{t \rightarrow \infty} \frac{d}{dt} \sum_{i=1}^N \langle [R_i(t) - R_i(0)]^2 \rangle \quad \kappa \propto \int_0^\infty dt \langle J(t) J(0) \rangle$$

Harmonic Approximation

A very powerful (first) approximation to the potential energy surface can be obtained by Taylor expanding the energy:

Harmonic approximation

$$V^{\text{BO}} = V(R_{\min}) + \sum_I \frac{\partial V}{\partial R_I} \bigg|_{R_{\min}} \Delta R_I + \frac{1}{2} \sum_{I,J} \frac{\partial^2 V}{\partial R_I \partial R_J} \bigg|_{R_{\min}} \Delta R_I \Delta R_J$$

\downarrow

Static Equilibrium
Energy
from DFT

\downarrow

Hessian matrix

It is convenient to define the mass-weighted Hessian

$$H_{IJ} = \frac{1}{\sqrt{M_I M_J}} \frac{\partial^2 V}{\partial R_I \partial R_J} \bigg|_{R_{\min}}$$

which leads to the eigenvalue problem

$$HQ = \omega^2 Q$$

\downarrow

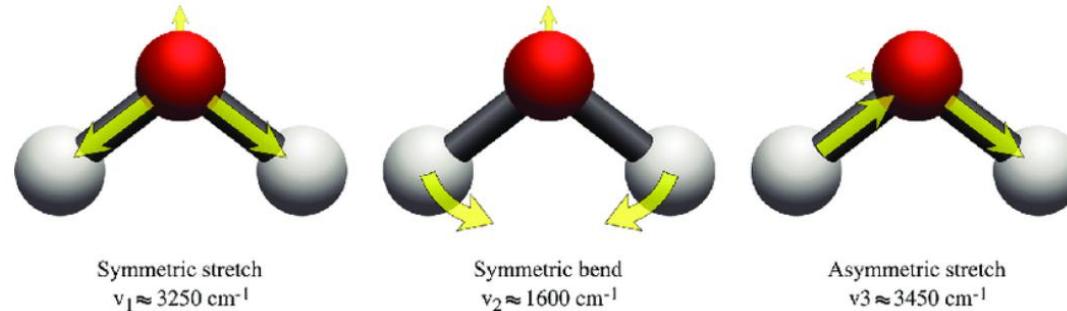
Normal modes

Normal mode frequencies

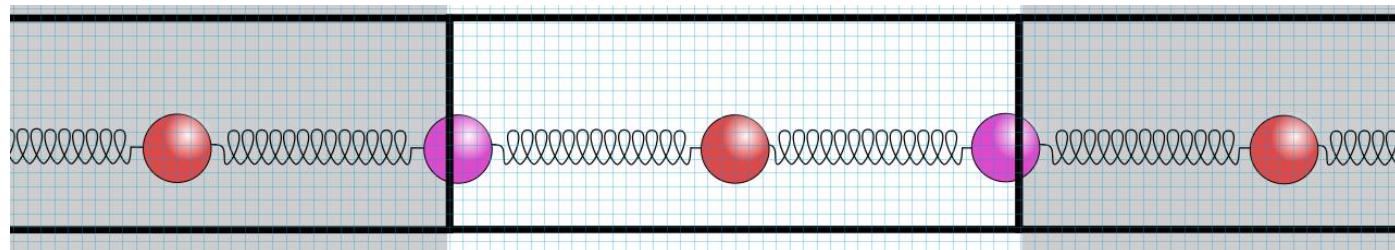
Harmonic Approximation

In (non-linear) molecules we have $3N-6$ vibrations

Example water molecule



However in periodic solids we have infinitely many atoms,



So, we have to go to reciprocal space and define the dynamical matrix as

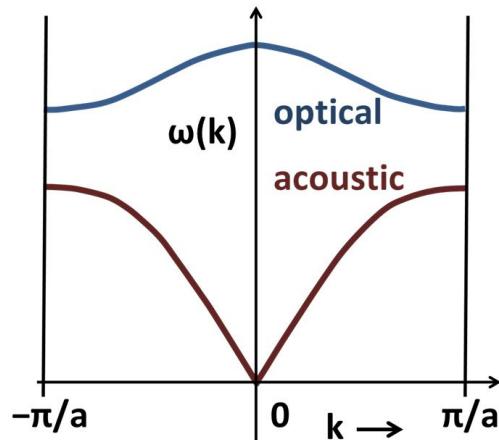
$$D_{IJ}(q) = e^{i\mathbf{q}(R_I^0 - R_J^0)} H_{IJ}$$

Vibrations in Extended Systems

$$D_{IJ}(q) = e^{i\mathbf{q}(R_I^0 - R_J^0)} H_{IJ}$$

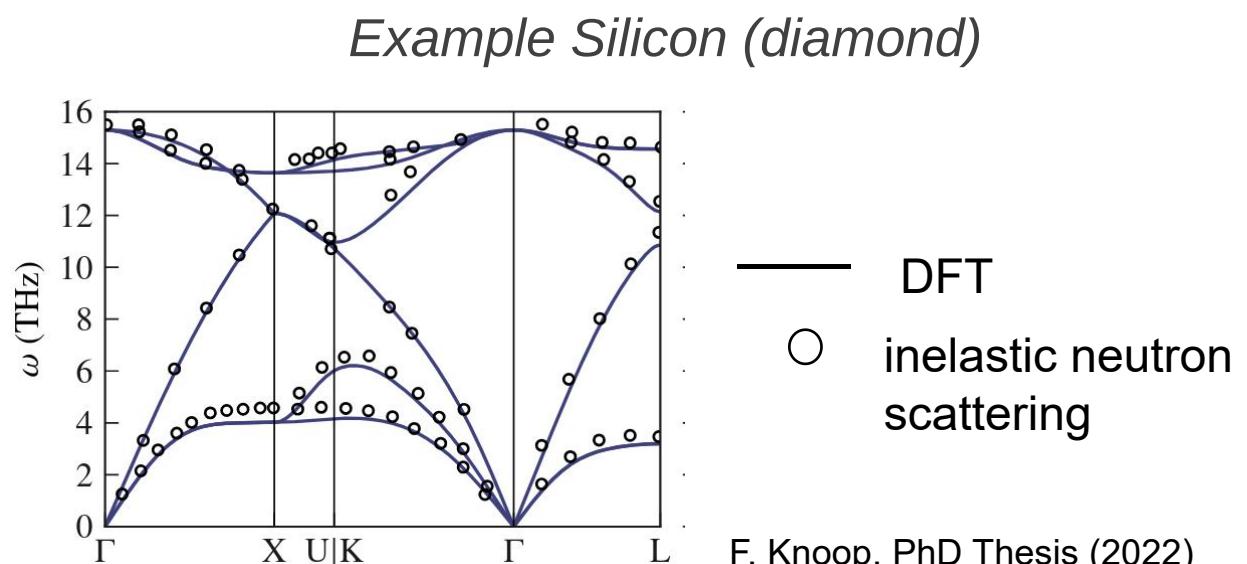
Now, we have to solve an eigenvalue problem for each value of wave vector, q

$$D(q)\nu(q) = \omega^2(q)\nu(q) \rightarrow \begin{array}{l} \text{phonon modes} \\ \downarrow \\ \text{phonon frequencies} \end{array}$$



3 acoustic modes (vanish at Γ)

$3N_p - 3$ optical modes (for a primitive cell with N_p atoms)



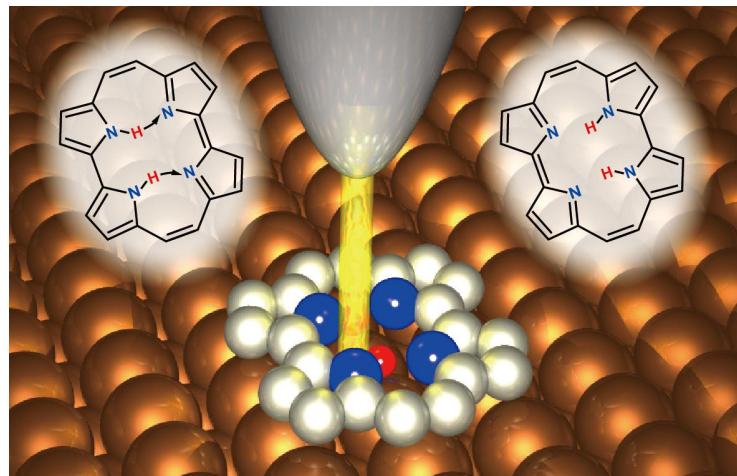
F. Knoop, PhD Thesis (2022)

Test case: Porphycene Molecule

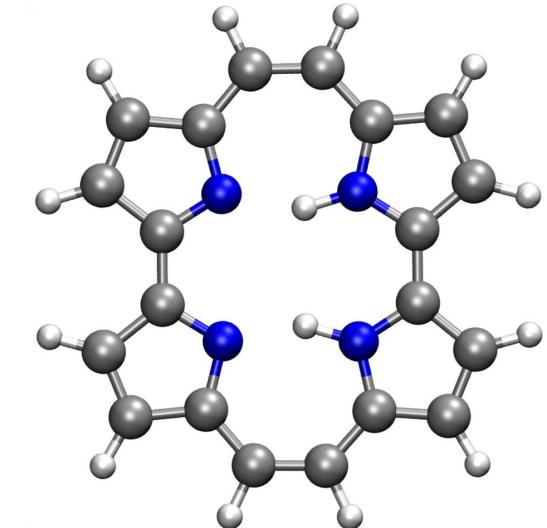
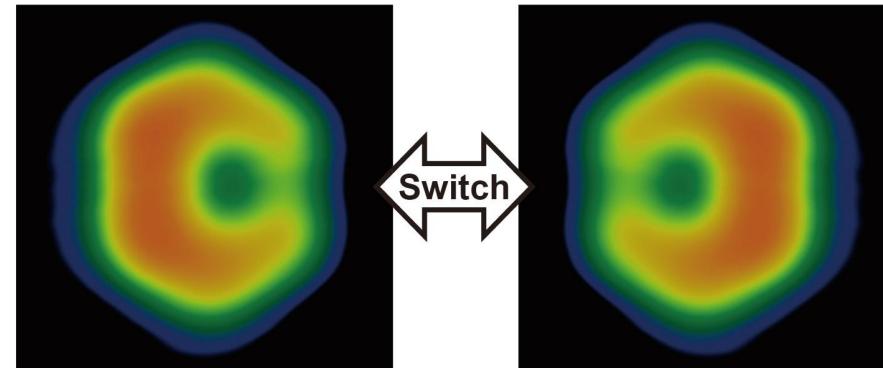
Well studied model of intramolecular hydrogen transfer processes in both ground and excited electronic states

Serves as a playground to test methods

Prototype of a molecular switch



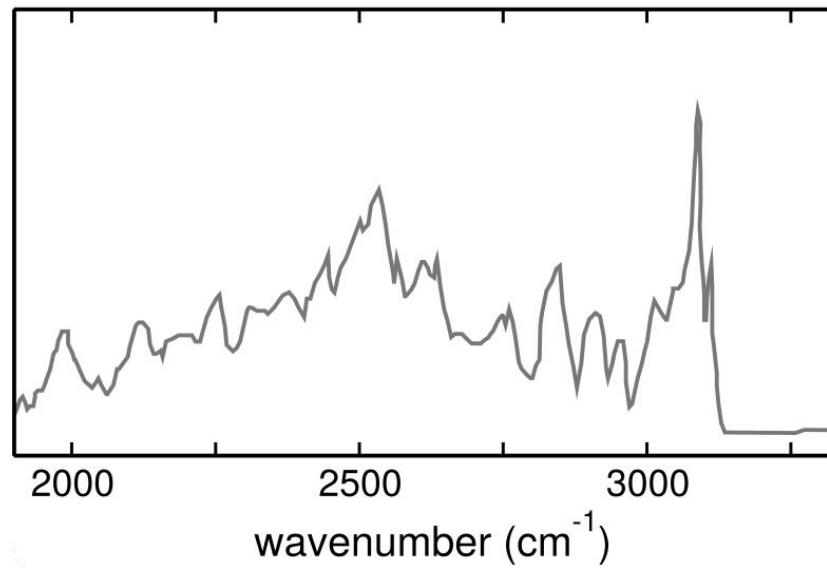
STM image of single porphycene molecule



Pictures taken from http://www.fhi-berlin.mpg.de/pc/kumagai/nsc_research.html

Anharmonic Vibrational Dynamics

Experiment
(KBr matrix, 290K)



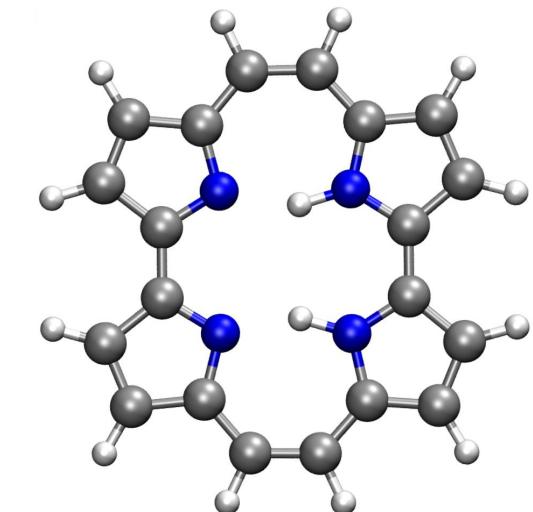
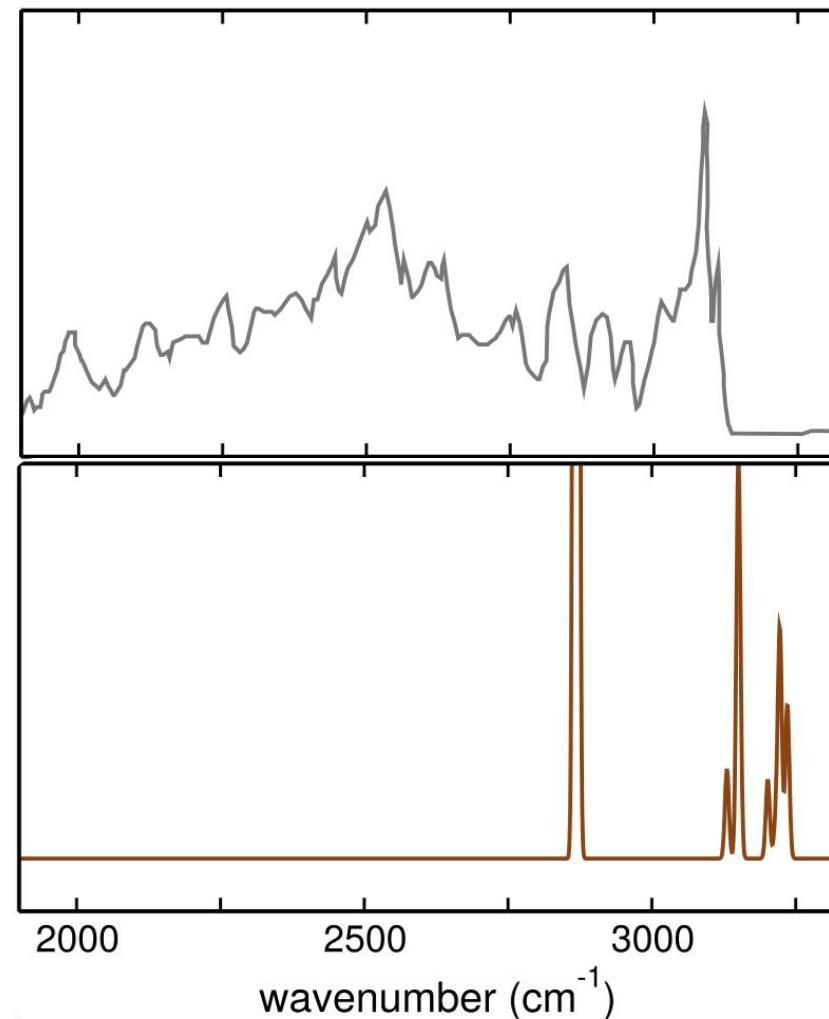
Gawinkowski, et al. Phys. Chem. Chem. Phys., 14, 5489 (2012)
Litman et al. J. Am. Chem. Soc. 141, 2526 (2019)

Anharmonic Vibrational Dynamics

$$I_{\text{IR}}(\omega) \propto \int_0^{\infty} dt e^{-i\omega t} \langle \mu(t) \mu(0) \rangle$$

Experiment
(KBr matrix, 290K)

Harmonic



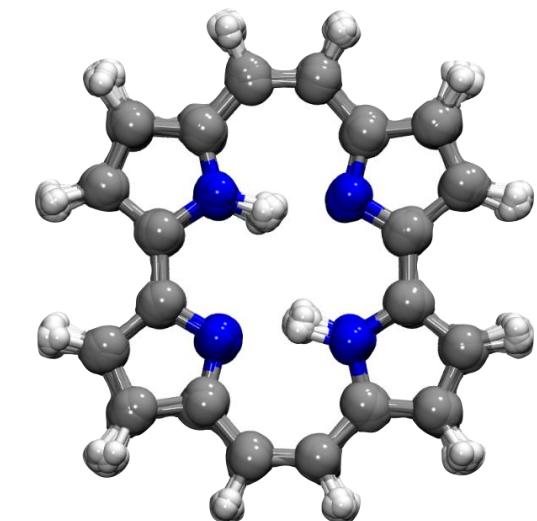
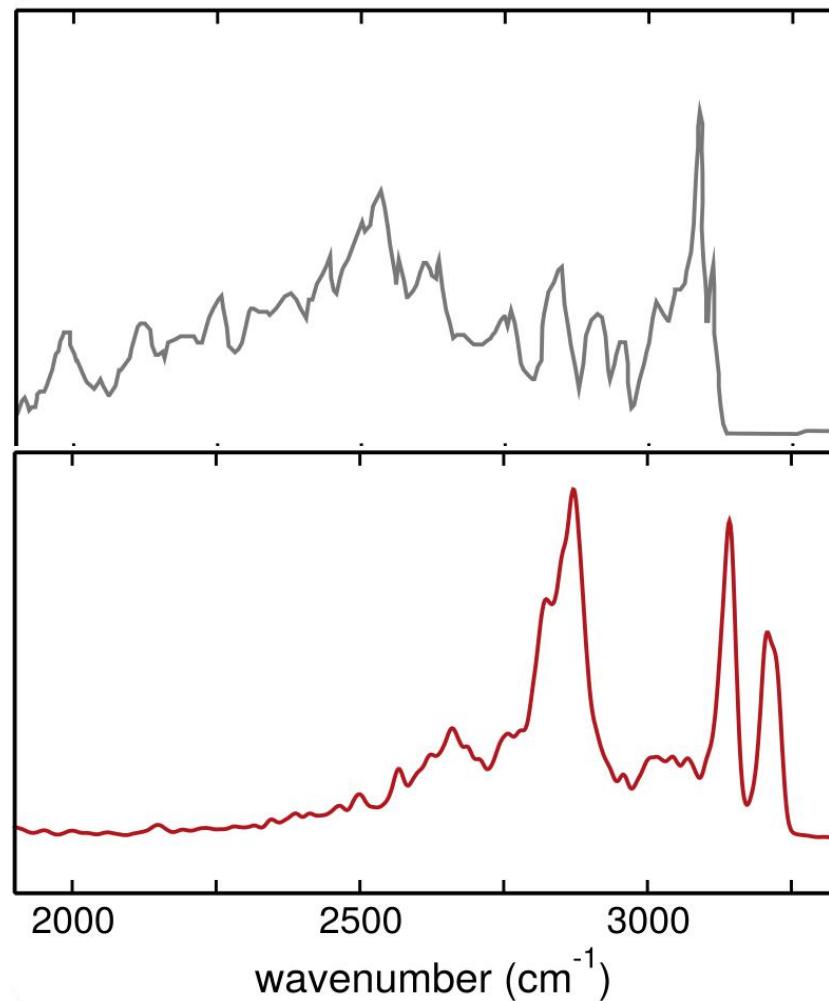
Gawinkowski, et al. Phys. Chem. Chem. Phys., 14, 5489 (2012)

Litman et al. J. Am. Chem. Soc. 141, 2526 (2019)

Anharmonic vibrational dynamics

$$I_{\text{IR}}(\omega) \propto \int_0^{\infty} dt e^{-i\omega t} \langle \mu(t) \mu(0) \rangle$$

Experiment
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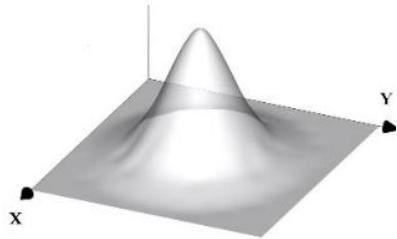


Gawinkowski, et al. Phys. Chem. Chem. Phys., 14, 5489 (2012)

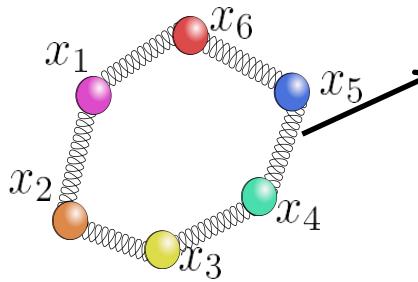
Litman et al. J. Am. Chem. Soc. 141, 2526 (2019)

Path Integral Isomorphism

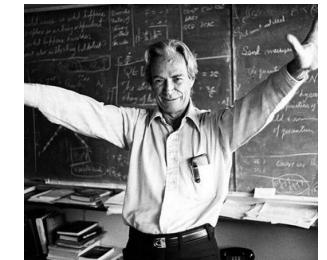
Path integral formulation of quantum mechanics



Quantum particle



$$\omega_N = \frac{Nk_B T}{\hbar}$$



Feynman

Ring Polymer (RP) made of **classical** particles

Sum over replicas

$$\hat{H} = \frac{\hat{P}}{2M} + \hat{V}(R)$$

$$H_N = \sum_{k=1}^N \left[\underbrace{\frac{P^{(k)}}{2M} + V(R^{(k)})}_{\text{classical Hamiltonian}} + \underbrace{\frac{1}{2}m\omega_N^2(R^{(k+1)} - R^{(k)})^2}_{\text{-springs-}} \right]$$

This mapping is known as the *path integral isomorphism*

Algorithm: path integral molecular dynamics (PIMD)

Exact results for time-independent equilibrium properties

Chandler, Wolynes, J. Chem. Phys. **74**, 4078 (1981)

Parrinello, Rahman, J. Chem. Phys. **80**, 860 (1984)

Want to become an expert in modeling nuclear quantum effects?

PIQM 2026

Path Integral Quantum Mechanics in the Era of Machine Learning

Jul 12, 2026 - Jul 16, 2026

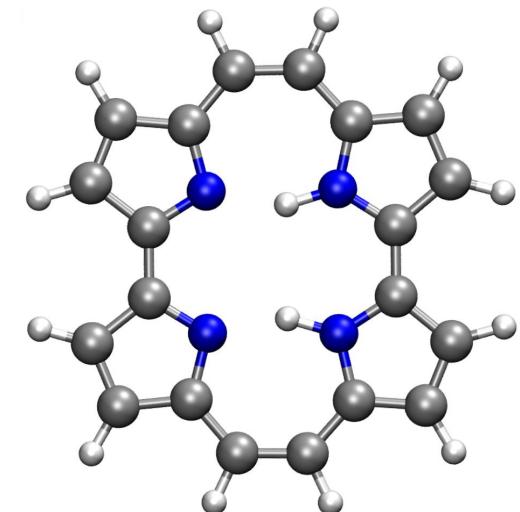
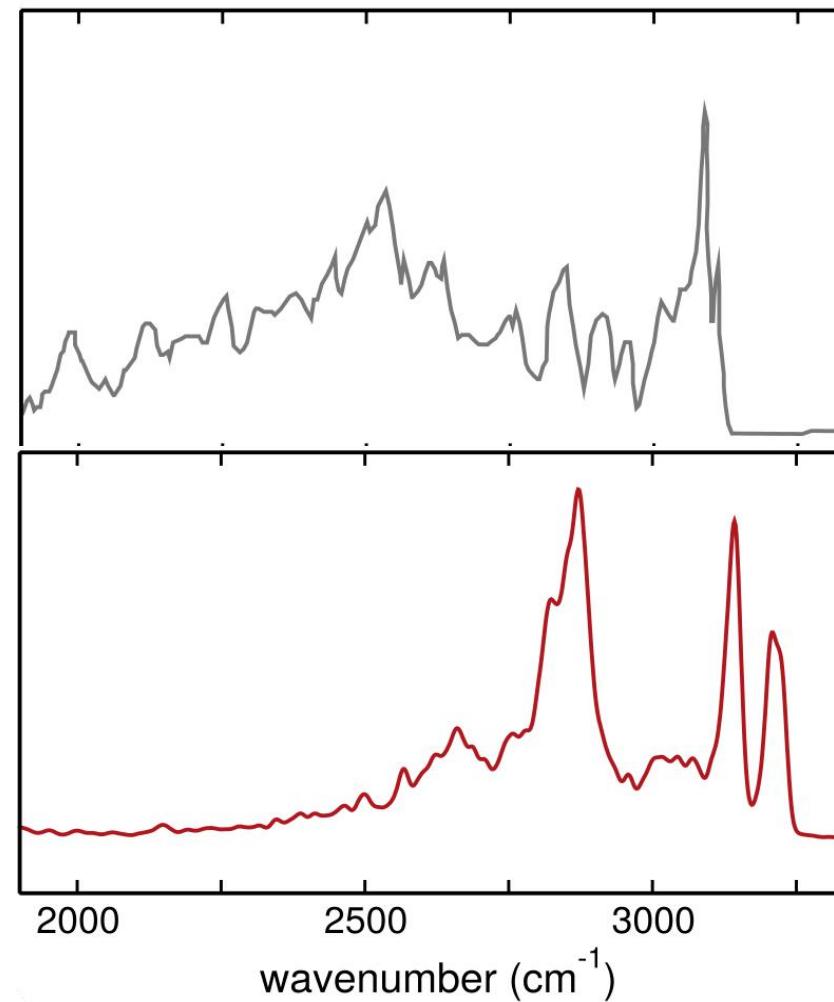
Shanghai, China

Anharmonic vibrational dynamics

$$I_{\text{IR}}(\omega) \propto \int_0^{\infty} dt e^{-i\omega t} \langle \mu(t) \mu(0) \rangle$$

Experiment
(KBr matrix, 290K)

MD
(290K, 4 x 10ps)



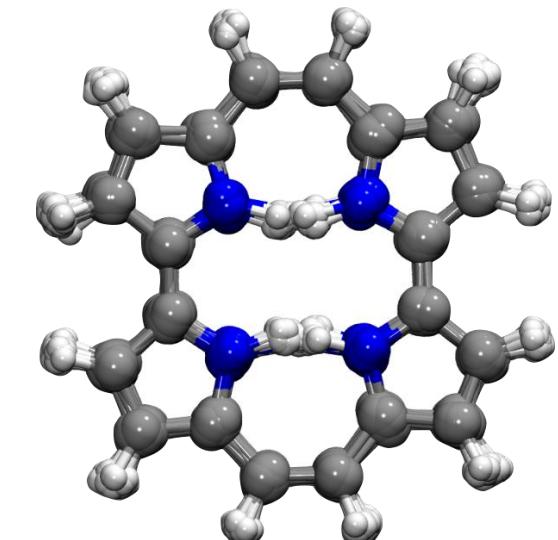
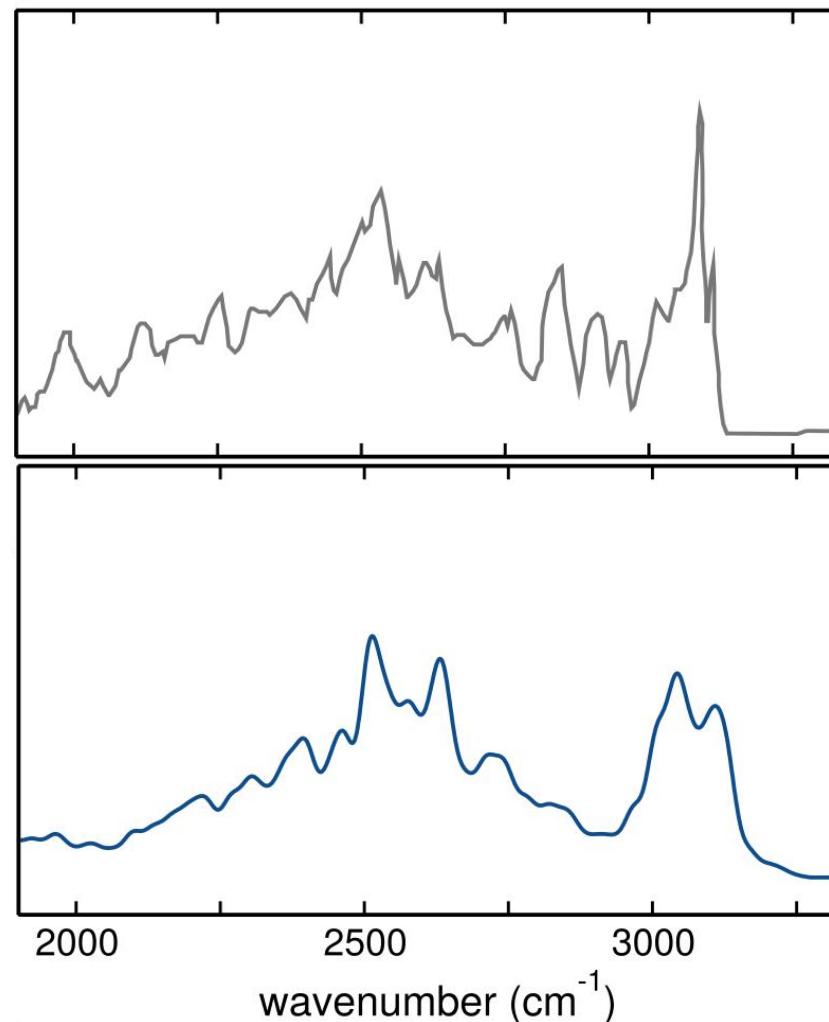
Gawinkowski, et al. Phys. Chem. Chem. Phys., 14, 5489 (2012)
Litman; et al. J. Am. Chem. Soc. 141, 2526 (2019)

Anharmonic vibrational dynamics

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Experiment
(KBr matrix, 290K)

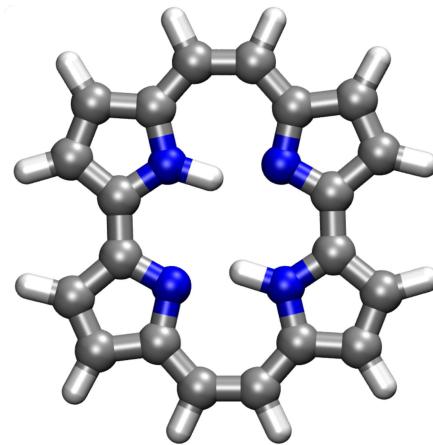
TRPMD,
(290K, 7 x 10ps,
16 beads)



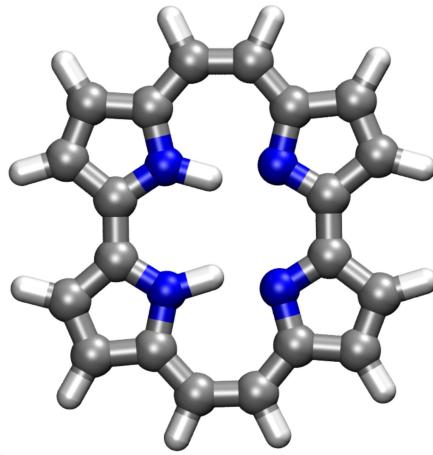
Gawinkowski, et al. *Phys. Chem. Chem. Phys.*, 14, 5489 (2012)
Litman; et al. *J. Am. Chem. Soc.* **141**, 2526 (2019)

Mind the XC Functional!

Trans



Cis

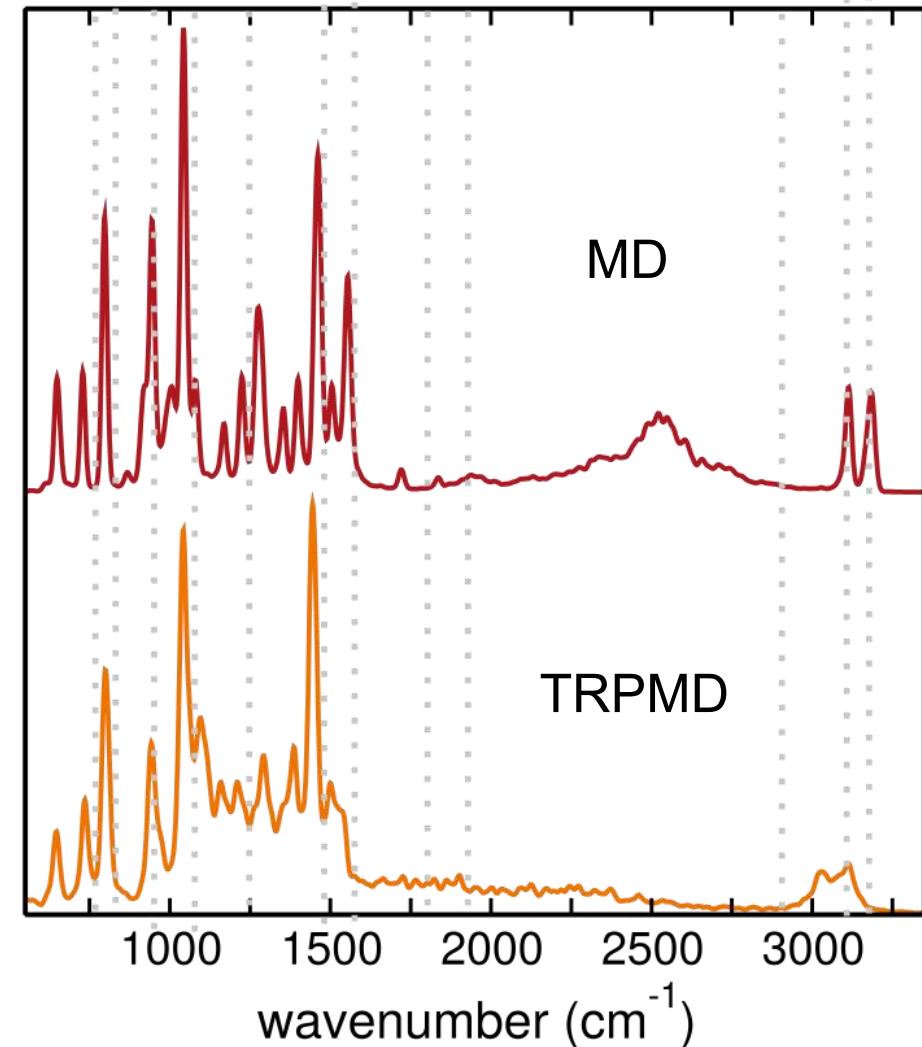


Energy in meV

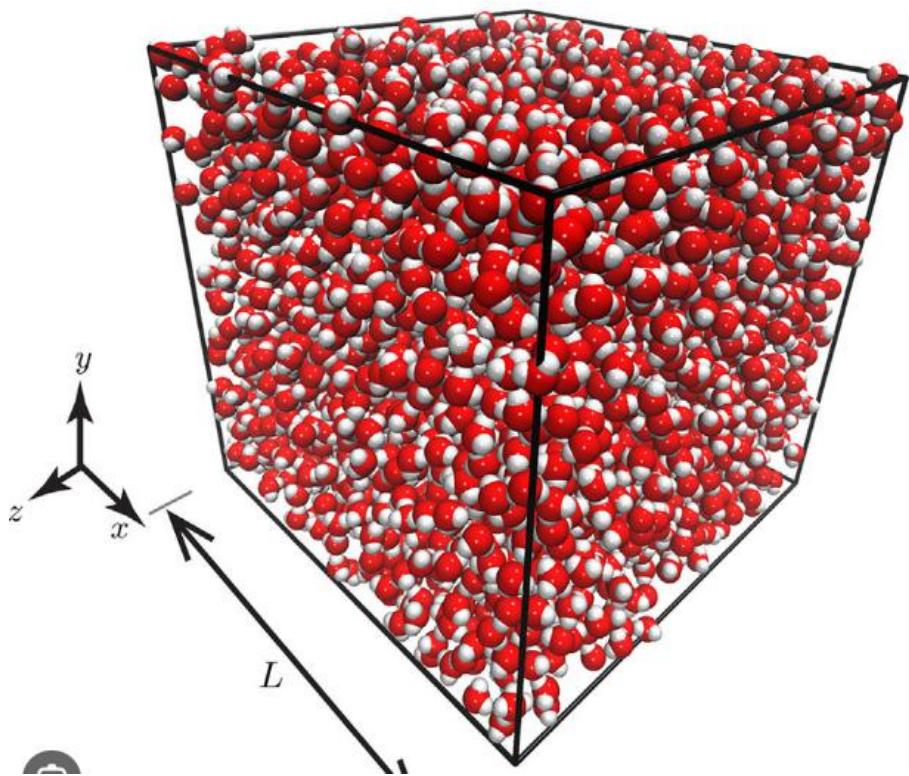
	PBE+TS	PBE0+TS	B3LYP+TS	MP2	CCSD(T)
cis	72 (35)	89 (60)	93 (72)	104	160
SP1	106 (-14)	155 (29)	189 (62)	144	218
SP2	150 (-61)	228 (-1)	290 (56)	194	322

Alkylammonium

TS: Tkatchenko.; Scheffler, Phys. Rev. Lett. 102, 073005 (2009)

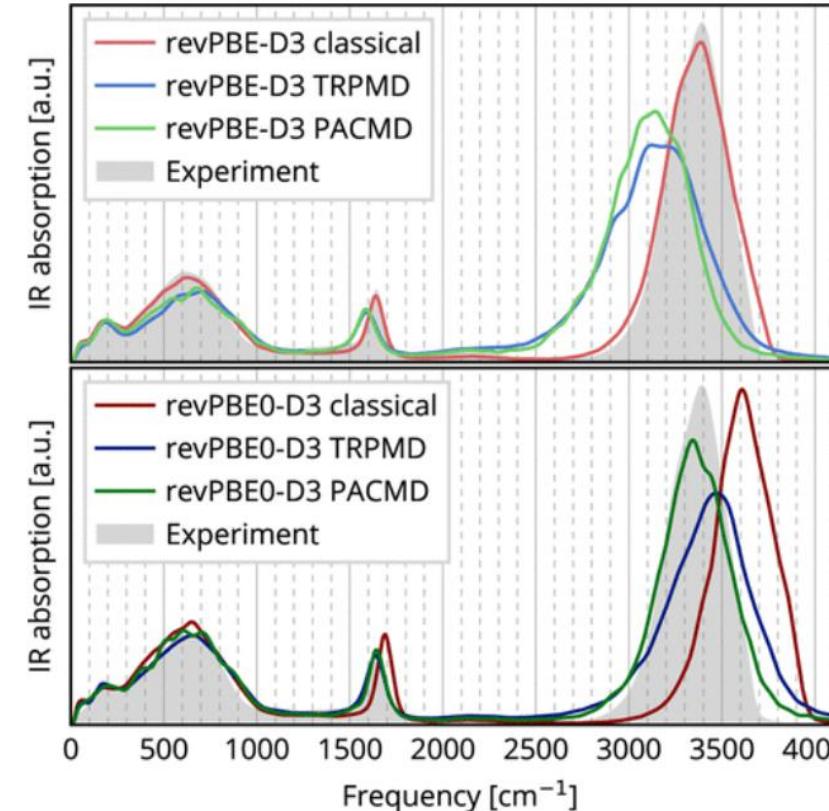


Mind the XC Functional!



GGA XC functionals tend to over estimate the strength of hydrogen-bonds

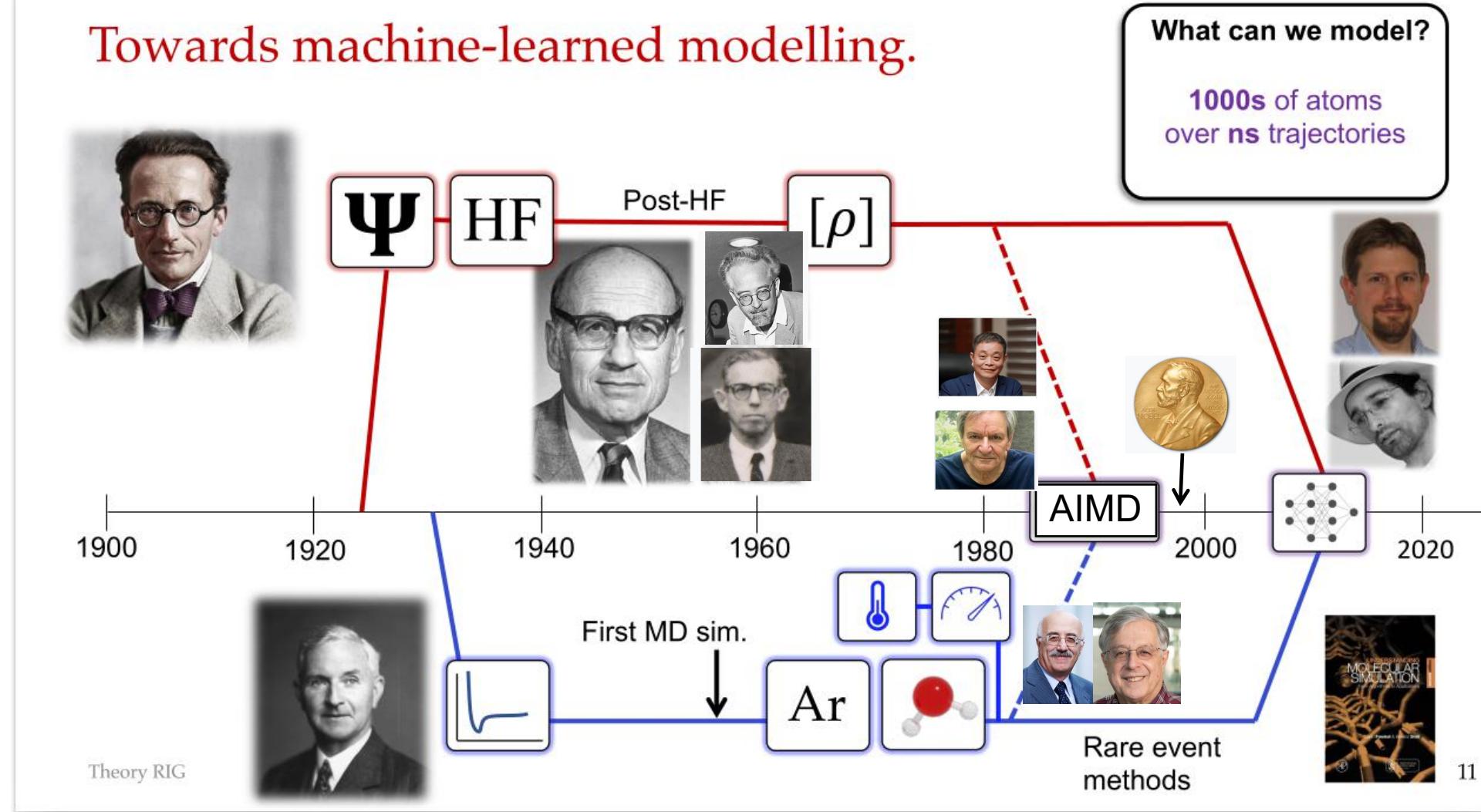
Marsalek, Markland, J. Phys Chem. Lett. **8**, 1545 (2017)



DFT can predict the right results for the wrong reasons

The Future is Here

Towards machine-learned modelling.



(Slide courtesy of Samuel Brookes@Cambridge)



