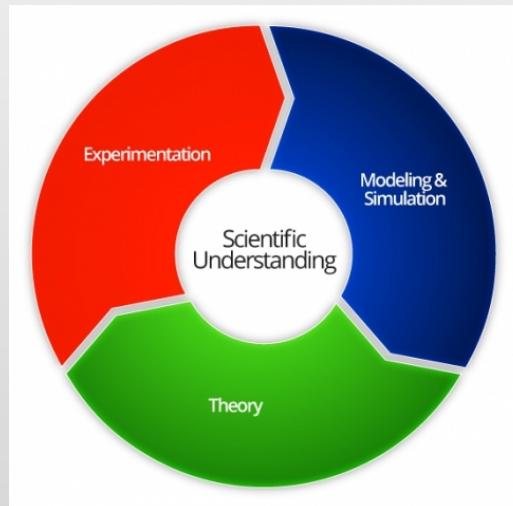
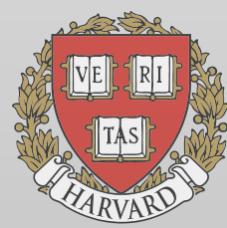


AC274: Computational Physics

The third route to scientific investigation



*Sauro Succi,
IAC Rome and IACS Harvard*



What is Computational Physics?

Interrogate Nature using computers

The **art** of filling the huge gap between the degrees of available to Nature and those affordable by our computers ...



Why an art?

(A. Avogadro, 1776-1856)

The Avogadro syndrome:

- **10^{23}** molecules in a cc of water vs $\sim 10^{16}$ bits in the largest computers
- About **10^6 atoms/bit**
- Our best e-computers can simulate a few billions “molecules” over milliseconds: 1 computational molecule $\sim \text{sqrt(Avogadro)}$ molecules.

YET IT WORKS!

WHY? Because Nature is **Hierarchical** and (sometimes) **Universale (detail insensitive)**
Hence: broad scope for clever proxies:

Mathematical Models&Algorithms

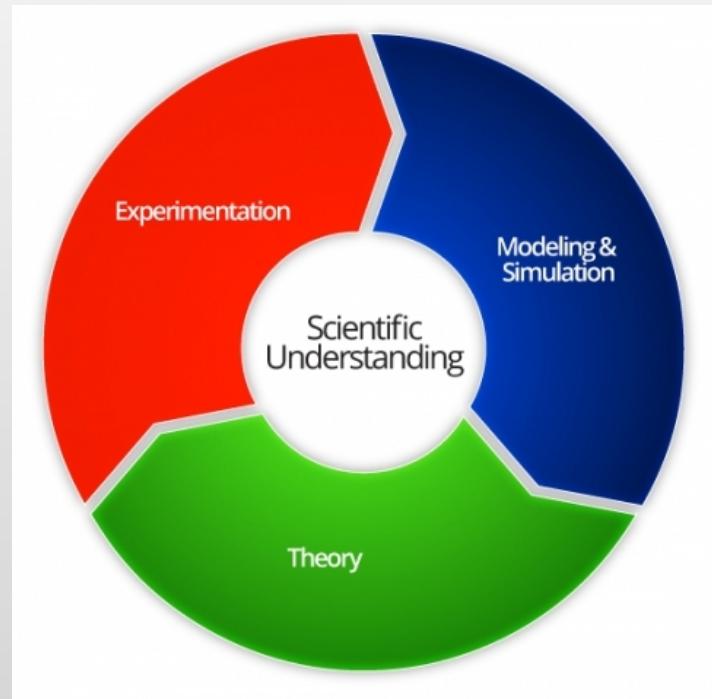
Why Computational Physics?

Experiment, problems with:

1. **Access (extreme conditions)**
2. **Costs**
3. **Safety**
4. **1+2+3...**

Theory: major barriers

1. **Non-linearity**
2. **Non-Locality**
3. **Large dimensions**
4. **Complex geometry**



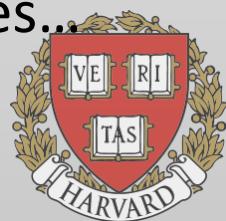
The Pillars of Computational Physics

The two pillars of computational physics: **hardware** and **software**



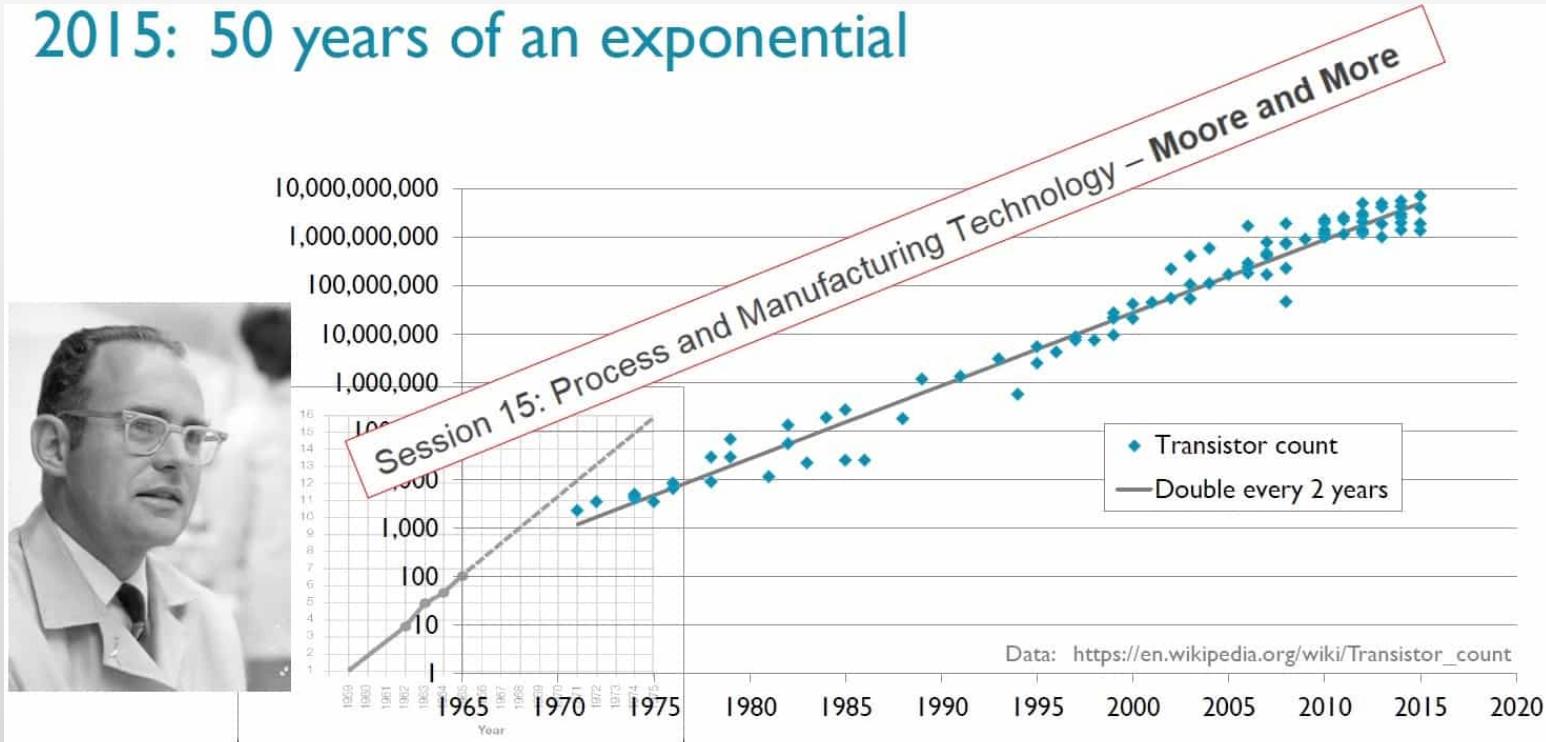
HW: Spectacular progress over the last 50 yrs: the compute power has **doubled every 1.5 yrs** (**Moore's law**): over a million in five decades!

SW: No less spectacular, proceeds by Revolution, see next slides...



Moore's law

2015: 50 years of an exponential



© 2015 ARM Greg Yeric, ARM

With the same trend in automotive, today we would drive faster than light!

Top current machine: Yian-He2, China: how many Flops?

5



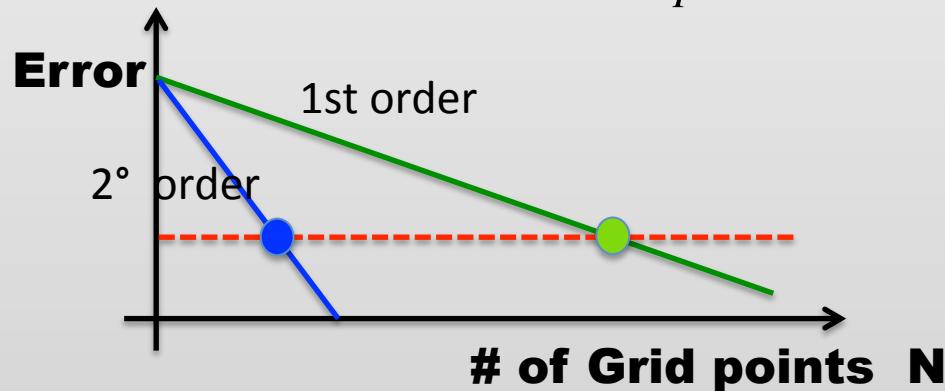
Methods&Algorithms

Algorithmic advances no less spectacular

SW: Computational methods proceed by **REVOLUTION.**

For instance, going from first to second order accuracy saves **decades** of Moore's time!!!

Error with p-th order method: $e_p \approx 1 / N^p$



$$N_p = N_1^{1/p}$$

From first to second order accuracy means 10^3 grid points instead of 100^3 saves *at least* a factor $1000=10$ **Moore years!!!**

The top 10 algorithms ...

Monte Carlo: hyperdimensional quadratures

Molecular Dynamics: condmat,biology

Comp. Fluid Dynamics: fluids

Fast Fourier Transform: signals

(AC274 will cover several
but not all (-:)

The **Fortran** compiler: all equations

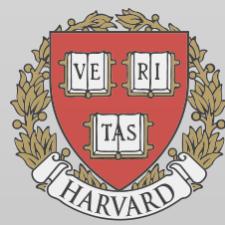
QR Decomposition: matrix algebra

Lanczos/Krylov iteration: iteration methods

Fast Multipole: long-range interactions

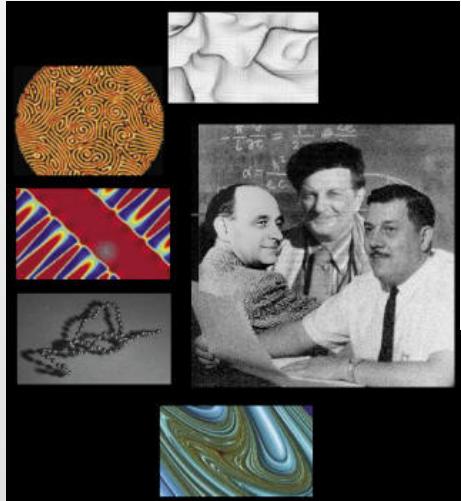
Simplex method: linear programming

Quicksort: search



The birth of Computational Physics

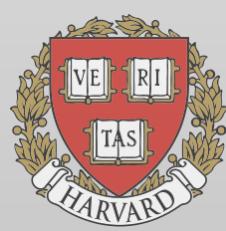
Fermi-Pasta-Ulam-Tsingou (Los Alamos, 1955)



CP at its zenith: not just “number crunching”
NEW model equations from THEORY!
New paradigms beyond the original problem

$$\partial_t u + u \partial_x u = \lambda \partial_{xxx} u$$

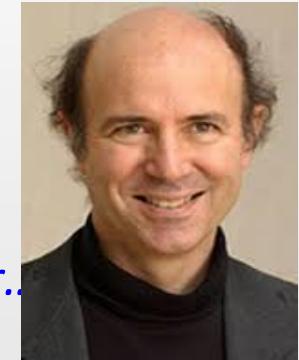
From heat propagation in solids to Tsunamis!



Computational State of Mind

A beautiful description of CP at its highest:

*The growing computational capacity **will change the nature of the questions we ask, the answers we seek and the investigations we pursue. Last but not least, it will change the nature of the investigator.***



Concepts and equations that computers can run will be powerfully leveraged, concepts and equations that cannot be turned into algorithms will be regarded as deficient.

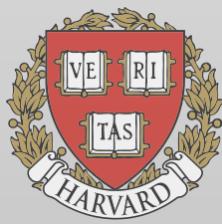
That does NOT mean that mindless number crunching will replace imaginative insight, on the contrary

(F. Wilczek, Phys Today April 2016)

Now to AC274

Four main parts:

1. Grid methods for classical and quantum fields
2. Advanced grid methods for complex geometry
3. Simulating complex states of matter
4. Data Analysis and Learning

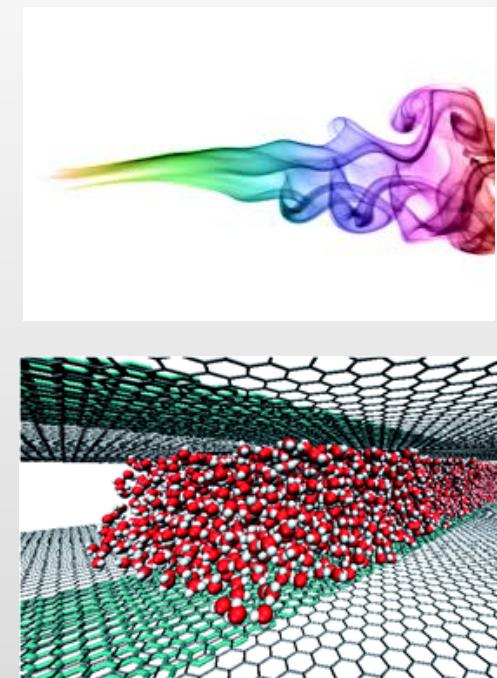


On Fields, Particles and Probabilities

FIELDS: based on the CONTINUUM assumption: granular/molecular aspects can be ignored. Typically modelled by **Partial Differential Equations (PDE's)**. Apparently the most economic description but often faces formidable computational complexity due to the four main barriers:

**NON-LINEARITY+COMPLEX GEOMETRY+NON-LOCALITY
+HIGH DIMENSIONS.**

Other representations may prove more efficient;
(PARTICLES, PROBABILITY DISTRIBUTIONS).



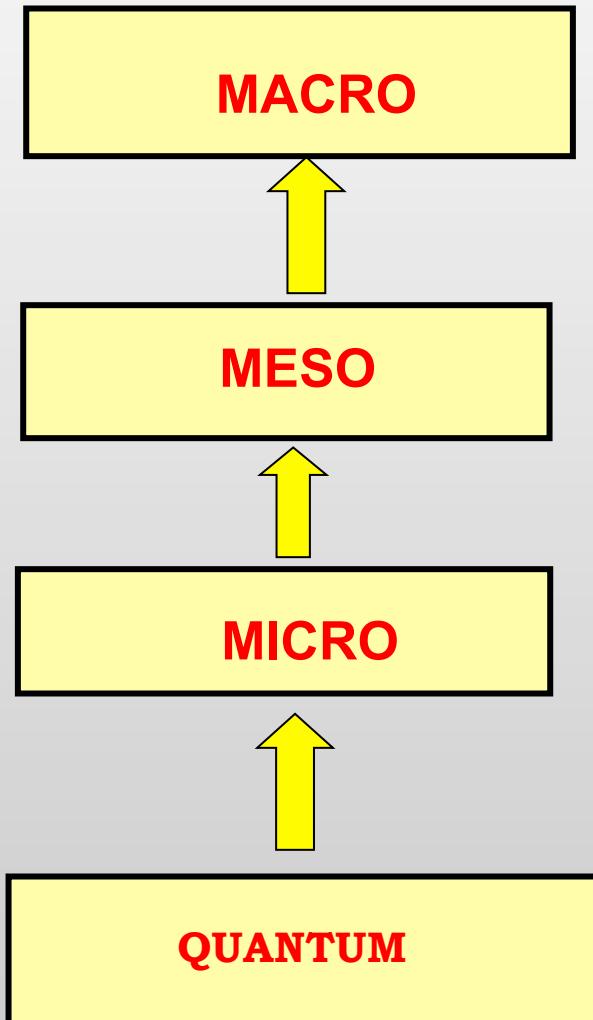
AC274 covers more than PDE's!

Modern physics reveals new states of matter besides the traditional triad **Gas/Liquid/Solid**: Amorphous (foams, emulsions, gels, colloids), and of course **Quantum Materials**, but also **Active Matter** (biomaterials, animal flocks) including **Social Systems**

The unifying conceptual framework is: **STATISTICAL MECHANICS**

Statistical Mechanics

Four basic levels to describe matter: (all covered by AC274)



Continuum Fields

$$\partial_t u + (u \cdot \nabla) u = -\frac{\nabla P}{\rho} + \nu \Delta u$$

Probability distribution functions

$$\partial_t f + (\mathbf{v} \cdot \nabla) f = -\frac{1}{\tau} (f - f^{(eq)})$$

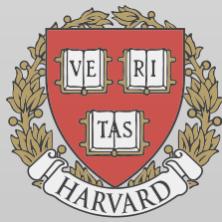
Particles (atoms/molecules)

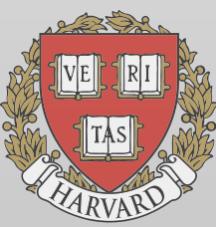
$$\frac{d^2 \mathbf{r}_i}{dt^2} = - \sum_{j>i} \nabla V_{ij}$$

Complex Fields

$$i\hbar \partial_t \Psi = H\Psi$$

Continuum Fields and PDE's





The Physics of Fluids

Why are fluids important ?

Pervasive!

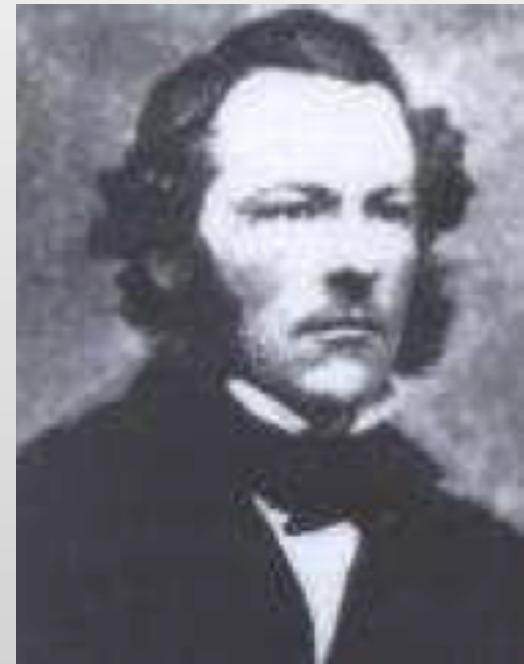


The heroes: Navier-Stokes

(L. Navier, 1785-1836))



(G. Stokes, 1819-1903)



The Navier-Stokes equations

$$\partial_t \rho + \nabla \cdot (\rho \vec{u}) = 0$$

Mass conservation

$$\partial_t (\rho \vec{u}) + \nabla \cdot (\rho \vec{u} \vec{u} + \overset{\leftrightarrow}{P}) = \vec{f}$$

Momentum conservation

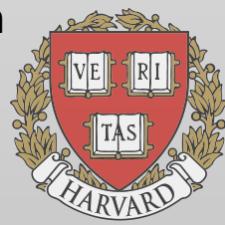
$$\overset{\leftrightarrow}{P} = p \overset{\leftrightarrow}{I} - \overset{\leftrightarrow}{\sigma}$$

$$p = f(\rho, T)$$

$$\overset{\leftrightarrow}{\sigma} = \lambda (\nabla \cdot \vec{u}) \overset{\leftrightarrow}{I} + \mu [\vec{\nabla} \vec{u} + (\vec{\nabla} \vec{u})^T]$$

Equation of State
(Ideal/Nonideal)

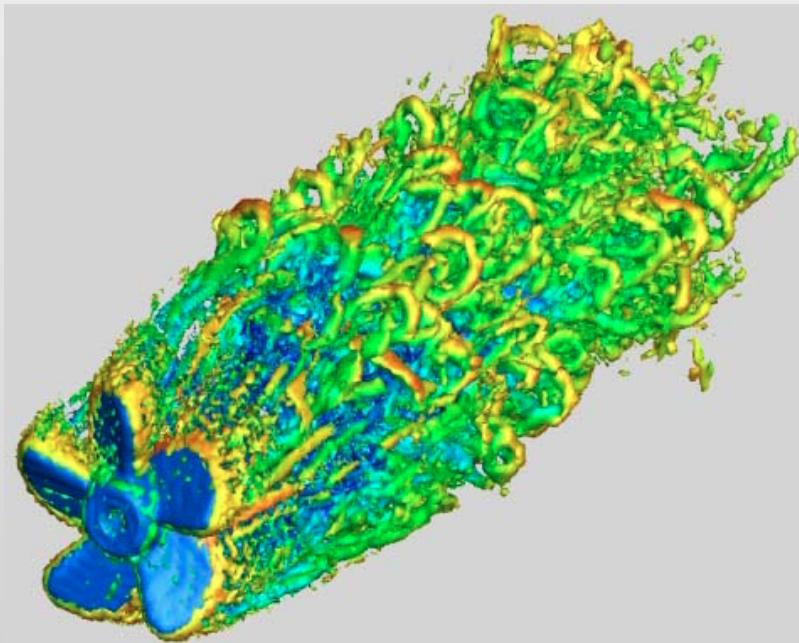
Stress-Strain Constitutive Relation
(Newtonian/Non-Newtonian)



Pandora's box

The NSE's look innocent but they are not!

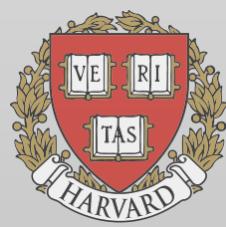
TURBULENCE!



(Covered by AC274, both in simulation and data analysis)

*The physics is not as
much in the equations
as it is in their solutions !!!*

*NEW physics from OLD
Equations!*



Solid/Wave Mechanics

How materials deform under stress (**RHEOLOGY**)

Displacements from the equilibrium configuration, **wave-propagation**

$$\xi(x, y, z) = r(x, y, z) - r_{eq}(x, y, z)$$

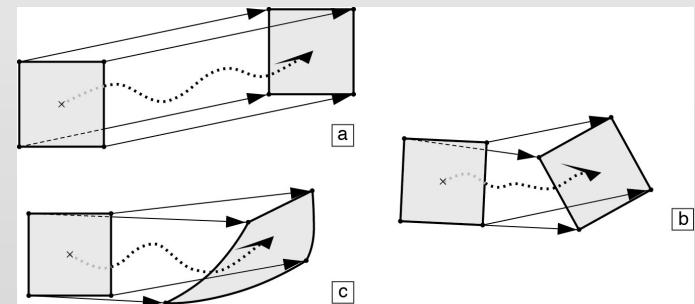
Obey a large set of Newton Equations:

$$M \ddot{\xi} = -K(\xi)\xi + F_{ext}$$

Inertia

Stiffness

External Load



Challenge: Large (non-linear) deformations

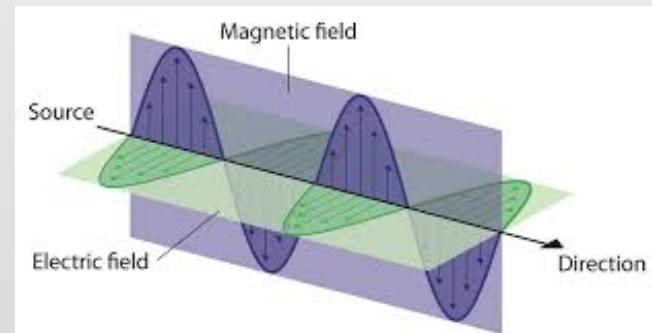
Electromagnetics

$$\nabla \times \vec{E} = -\frac{\partial \vec{B}}{\partial t}$$
$$\nabla \times \vec{H} = \vec{J} + \frac{\partial \vec{D}}{\partial t}$$
$$\nabla \cdot \vec{D} = \rho$$
$$\nabla \cdot \vec{B} = 0$$

ELECTROMAGNETICS

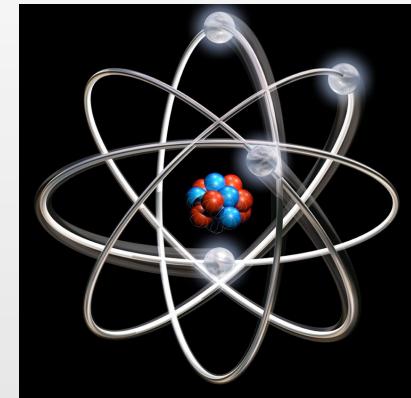
Branislav M. Notaroš

How em waves interact
with and propagate with matter
LIGHT-MATTER physics



Quantum Physics

Quantum physics is the language we speak with Nature,
paramount for applications and modern technology.
It deals with COMPLEX WAVEFUNCTIONS and PROBABILITIES



It comes in many forms:

- Single body (quantum mechanics)
- Many-body (quantum chemistry/biology)
- Quantum field theory (hep, condmat, exotic materials)
- Quantum computing, entanglement ...

AC274 will cover the basics (QM + many-body)

The hero

Schrödinger's Equation

$$i\hbar \frac{\partial}{\partial t} \psi(\mathbf{r}, t) = -\frac{\hbar^2}{2m} \nabla^2 \psi(\mathbf{r}, t) + V(\mathbf{r}, t) \psi(\mathbf{r}, t)$$

i is the imaginary number, $\sqrt{-1}$.

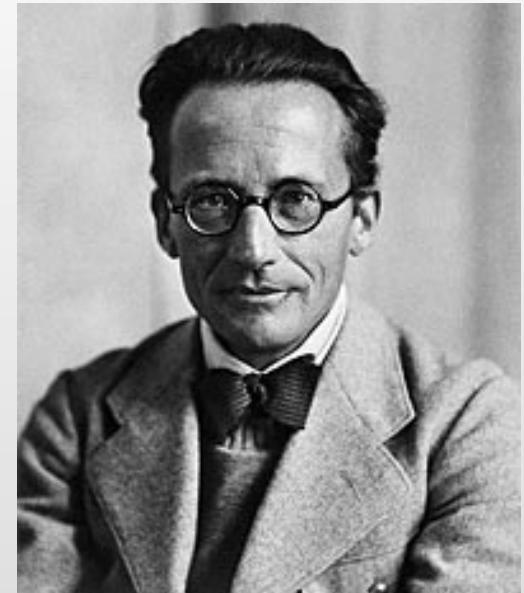
\hbar is Planck's constant divided by 2π : 1.05459×10^{-34} joule-second.

$\psi(\mathbf{r}, t)$ is the wave function, defined over space and time.

m is the mass of the particle.

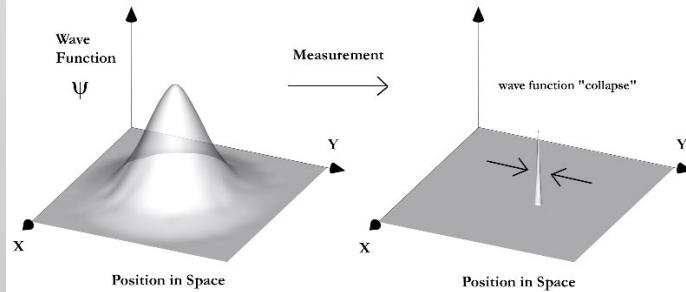
∇^2 is the Laplacian operator, $\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} + \frac{\partial^2}{\partial z^2}$.

$V(\mathbf{r}, t)$ is the potential energy influencing the particle.

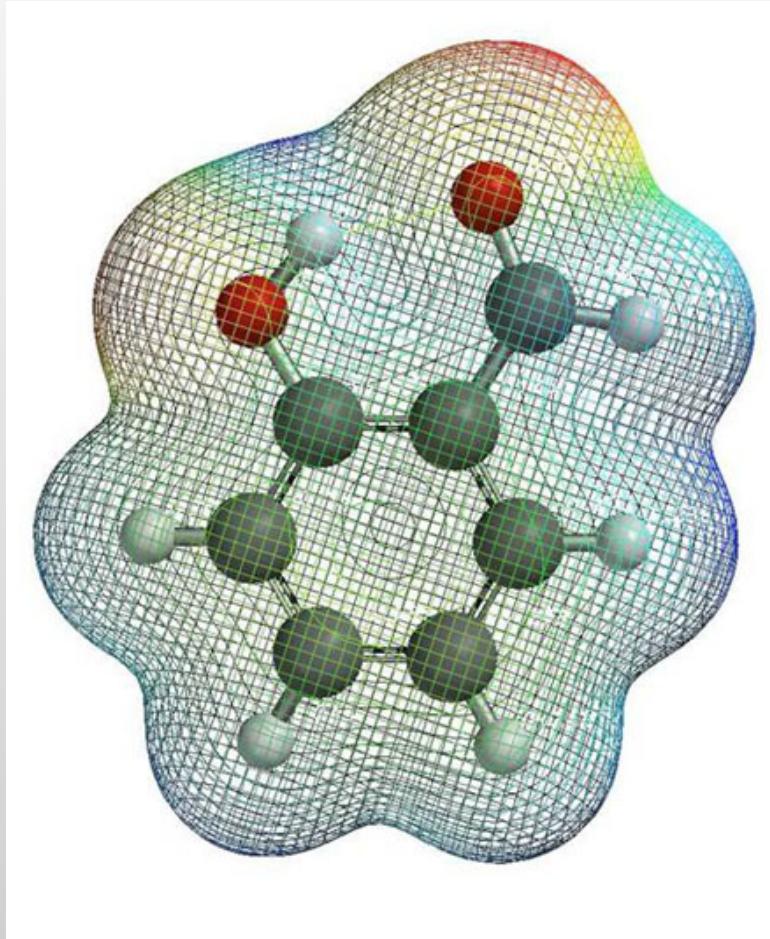


Erwin Schrödinger (1887-1961)

The Copenhagen Interpretation:



Computational Chem/Bio



Molecules and Biomolecules

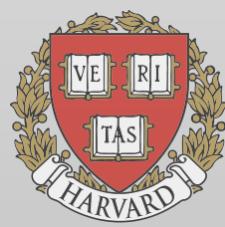
N=Nuclei+Electrons

The N-body wavefunction

$$\Psi(\vec{r}_1, \vec{r}_2, \dots \vec{r}_{Nel}; \vec{R}_1, \vec{R}_2, \dots \vec{R}_{Nuc})$$

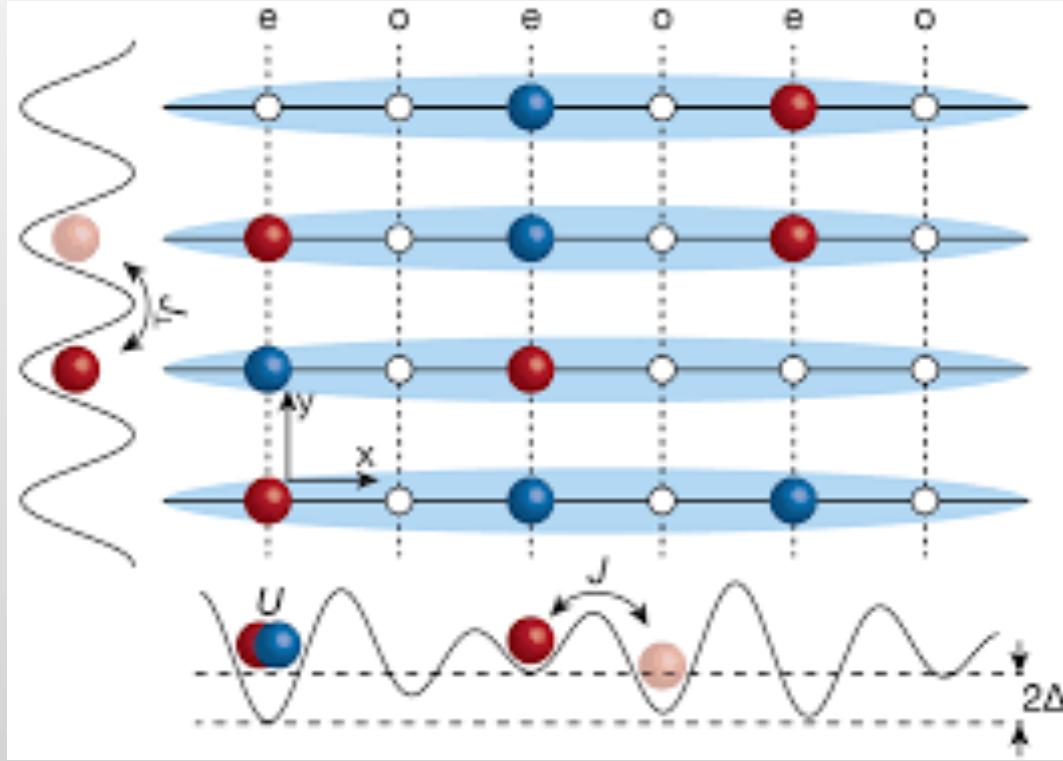
Lives in $3N$ -dimensional space!

$N \sim 10^5$: Ungriddable!
The kingdom of particle
methods (MonteCarlo)



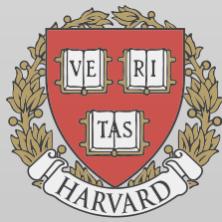
Quantum Field Theory

Each lattice site $i=1, 2 \dots N$ hosts a quantum field, then **send N to infinity!**



Promising Quantum Computing platforms

Complex geometries

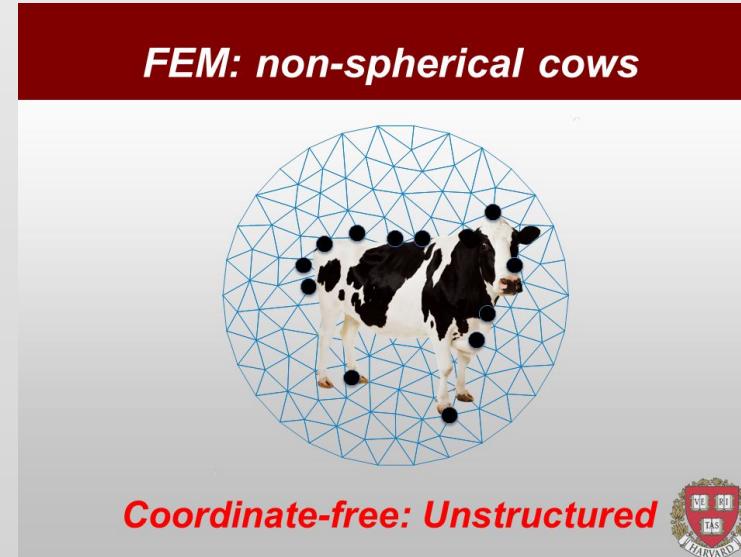
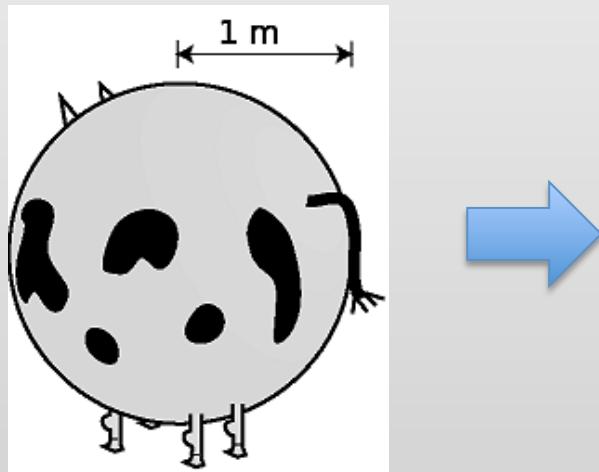


Non-spherical cows

Field theory and PDE's are subject to BOUNDARY CONDITIONS

Analytics is often at loss with complex geometries,
No smoothness!

Form-Function paradigm

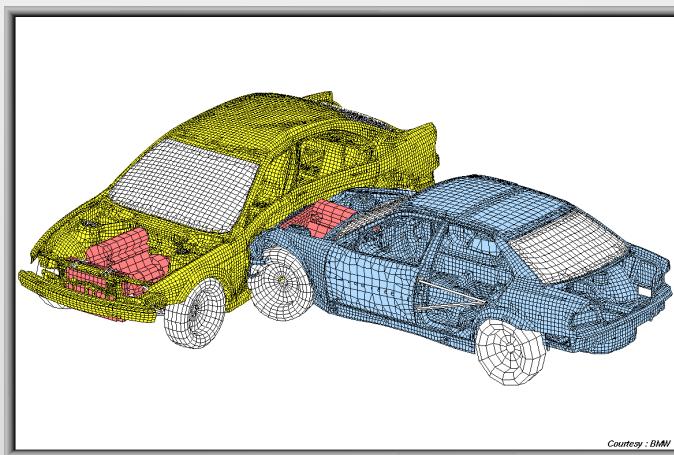


Major methods: **Finite Volumes and Finite Elements**

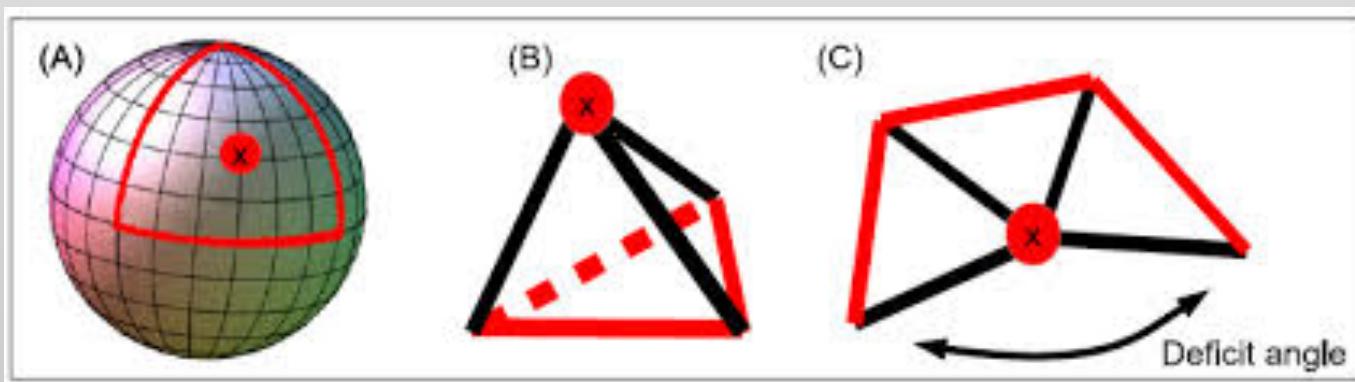
Finite Volumes/Elements

Computing without coordinates: Unstructured Data

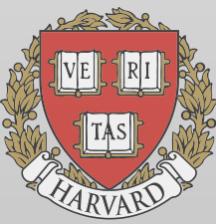
From engineering to quantum gravity!



(Tullio Regge, 1931-2014)

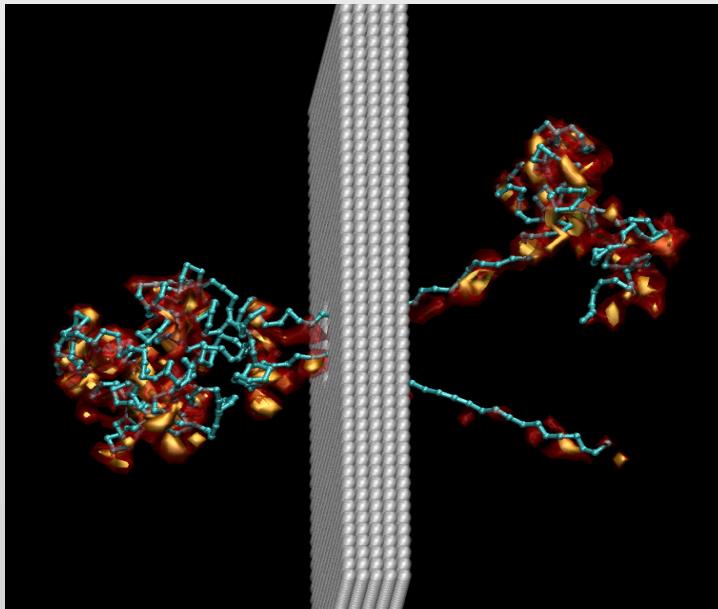


From fields to Particles: PDE's to ODE's



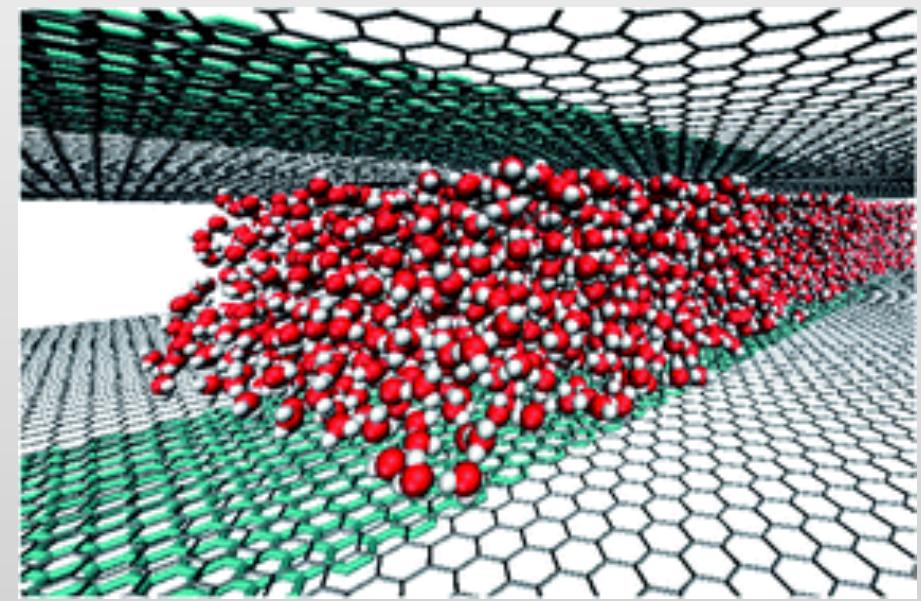
Molecular physics

Under the drive towards increasing miniaturization, micro and nanofluids are becoming more and more important in modern science. Granular nature of matter is exposed: **thermal fluctuations**



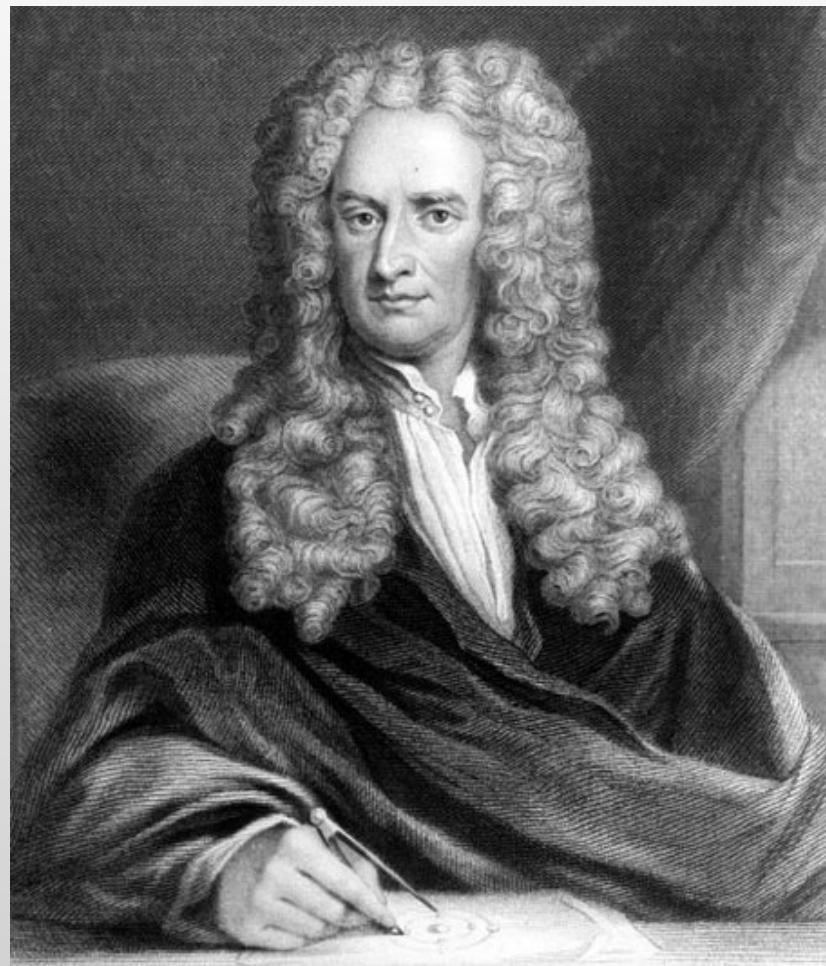
DNA Translocation

(See Lectures on Lattice Boltzmann and Molecular Dynamics)



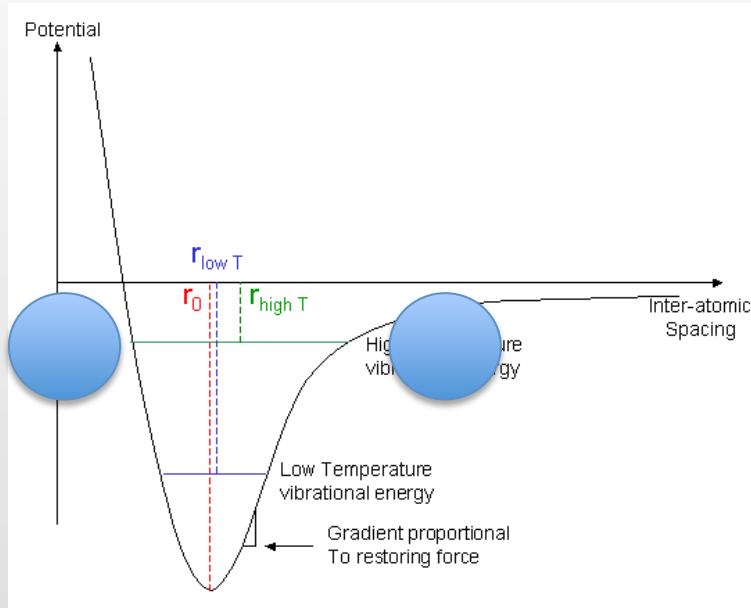
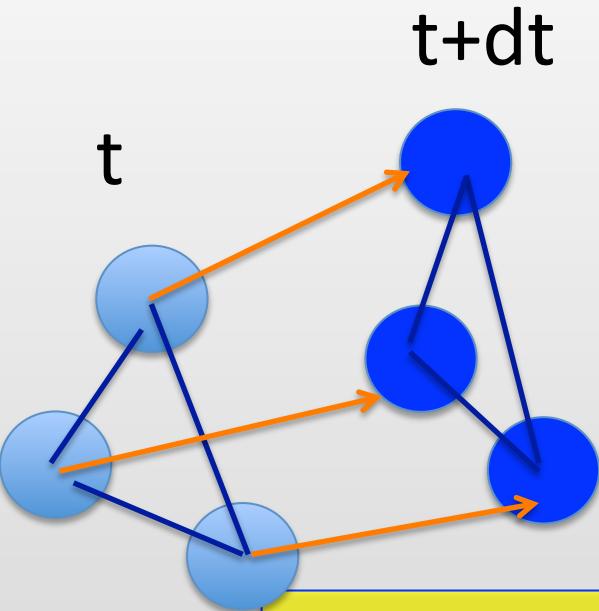
Water flow in nano-ribbons

The hero



Isaac Newton (1642 – 1727)

Molecular Dynamics

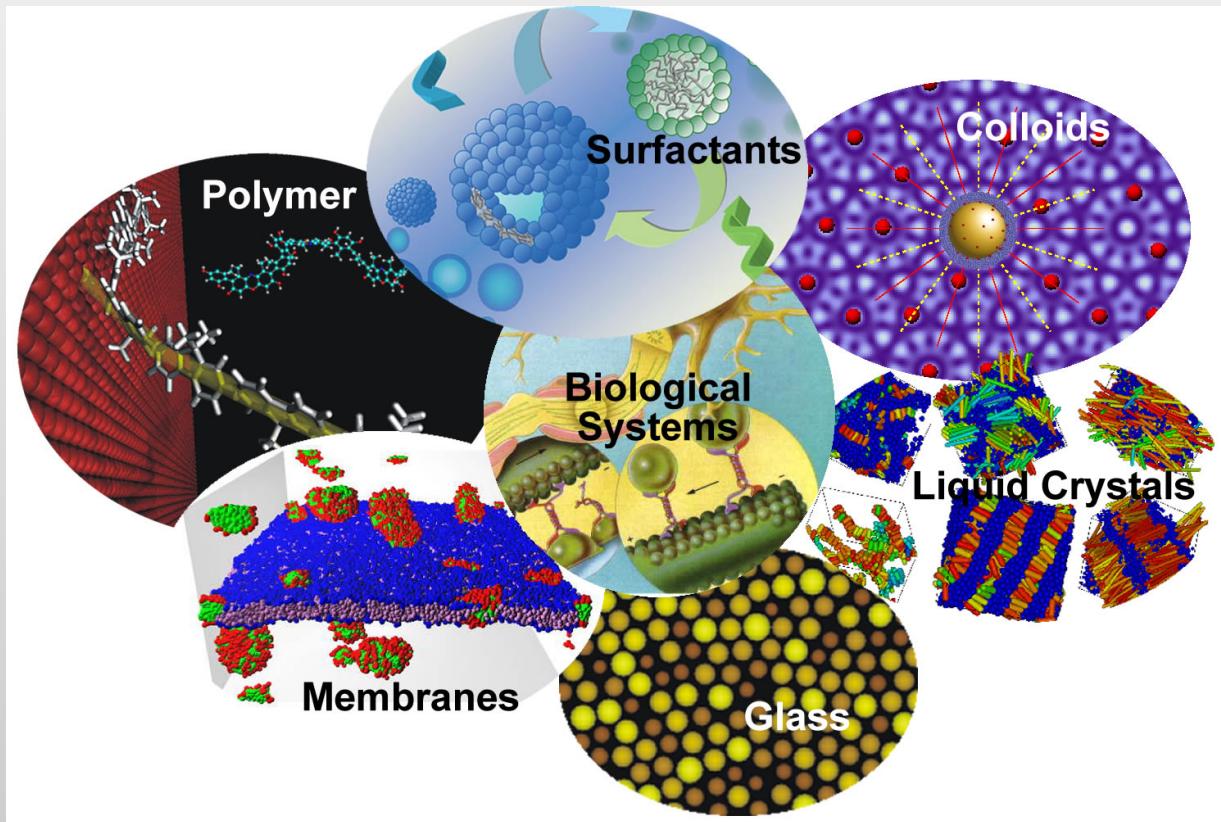


$$m \frac{d^2 \vec{r}_i}{dt^2} = - \sum_{j>i}^N \nabla_i V_{ij} + \sum_{k>j>i}^{N^2} \nabla_i V_{ijk} + \dots$$

"Just " a huge bag of ODE's: but the bag is **Avogadro-huge** and the time-integration goes on for billions/trillions time-steps. Femtoseconds to seconds: the **TIME-GAP** problem

Complex states of matter

Molecules with *internal structure*, interferes with external motion, very complex rheology



An ocean of applications
In Material Science,
Biology, Medicine...
you_name_it...

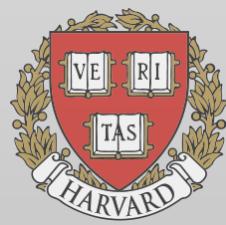
Boltzmann kinetic theory



“The man who
trusted atoms”

Ludwig Boltzmann (1844-1906)

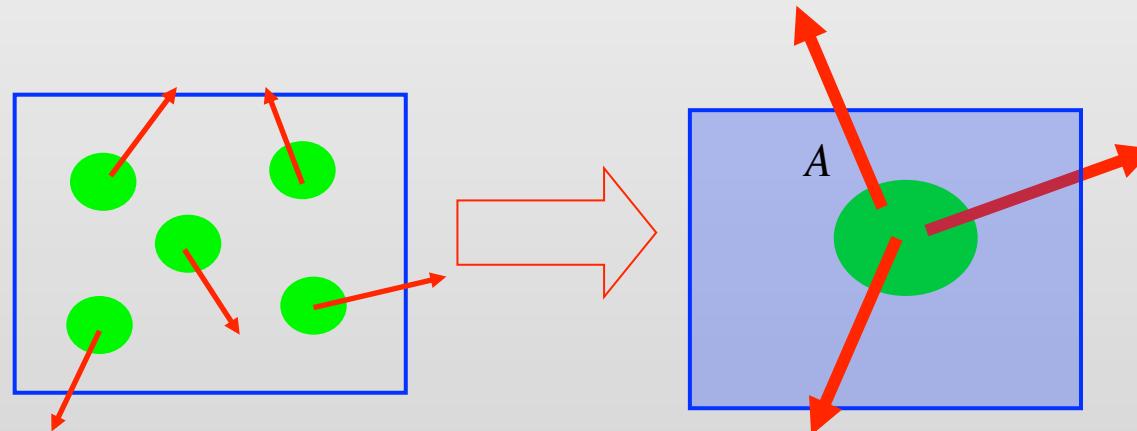
33



Boltzmann: Probability Distribution Function

The **average** number of molecules around r at time t with velocity v

$$\Delta N = f(\vec{r}, \vec{v}; t) \Delta \vec{r} \Delta \vec{v}$$

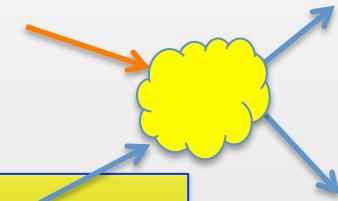


f lives in a 6-dimensional world
(phase-space)

The Boltzmann equation

Fluid in 6-dim phase-space:

$$\partial_t f + \vec{v} \cdot \nabla_r f + (\vec{F} / m) \cdot \nabla_v f = C(f, f)$$



$$\frac{d\vec{r}}{dt} = \vec{v}$$

$$m \frac{d\vec{v}}{dt} = \vec{F}$$

Collisions

A very tough computational cookie!
6+1 dimensional, nonlinear, integro-differential...

Usually solved by Monte Carlo, but many simplified models (see Lattice Boltzmann)

The Boltzmann equation



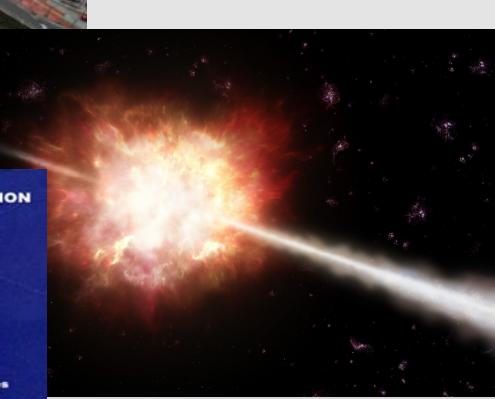
*Transport in dilute media
far from local equilibrium*



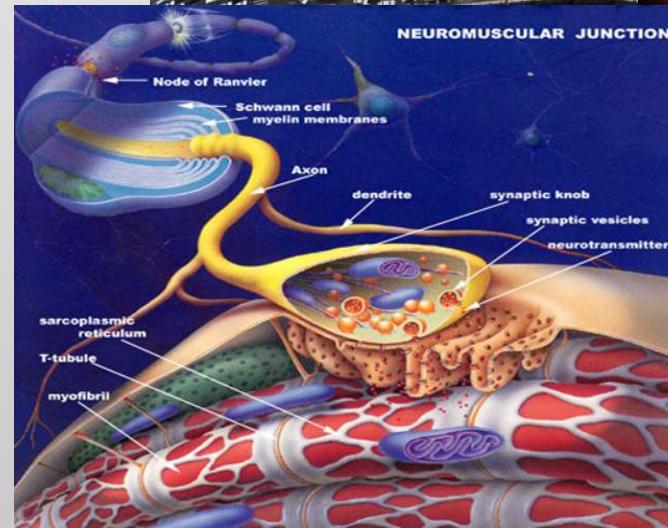
Neutron transport



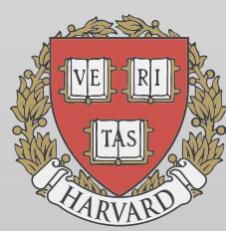
Gamma rays transport



Shuttle re-entry



Electron flows



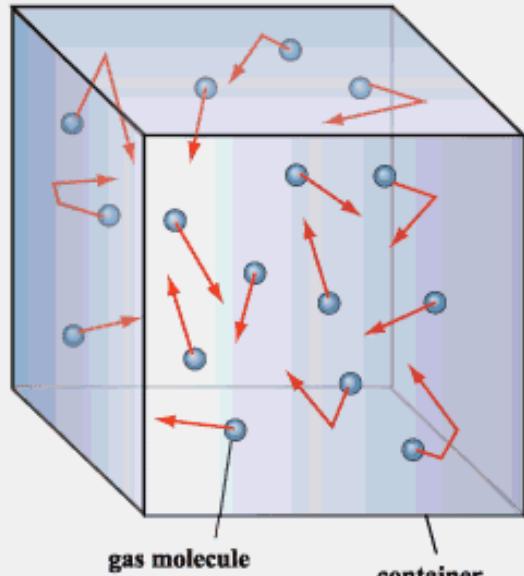
Traffic flows

Lattice Boltzmann: Platonic hydrodynamics

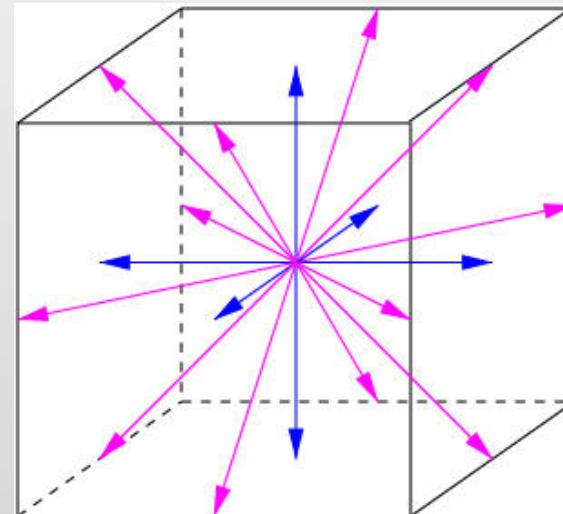
$$f(x, v; t) = \sum_{i=0}^b f_i(x, t) \delta(v - c_i)$$

$i = 0, b$

Triple infinity to just 19!

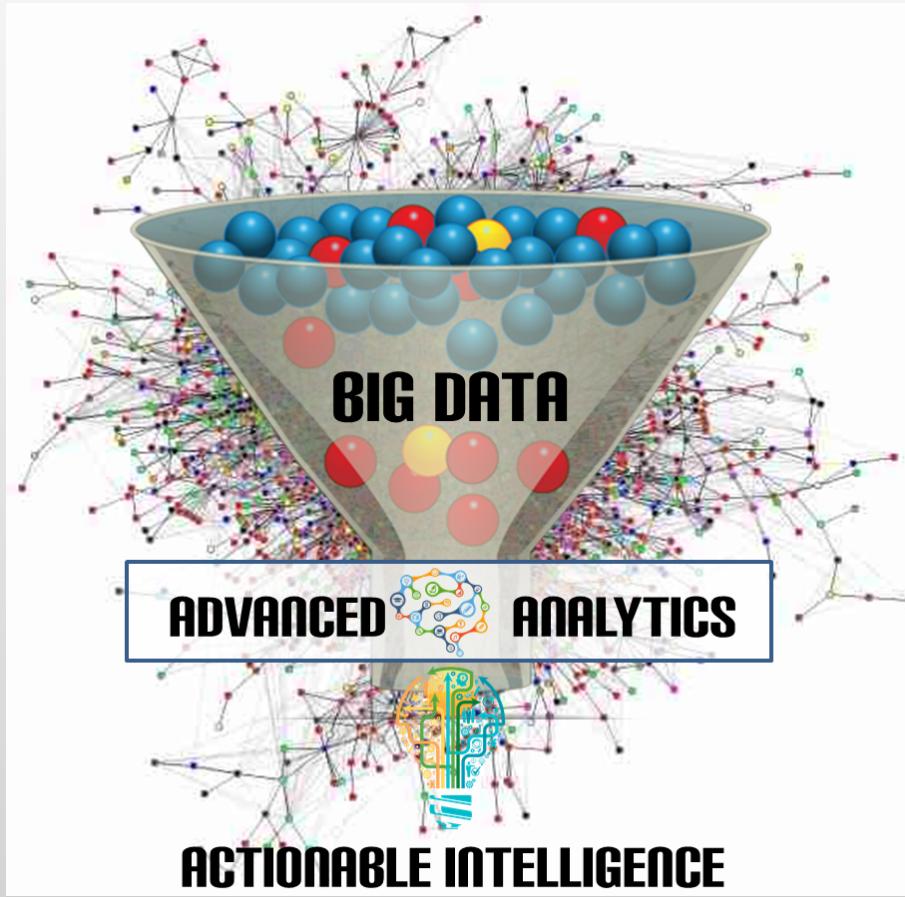


Magic speeds!



Continuum spacetime --> Crystal

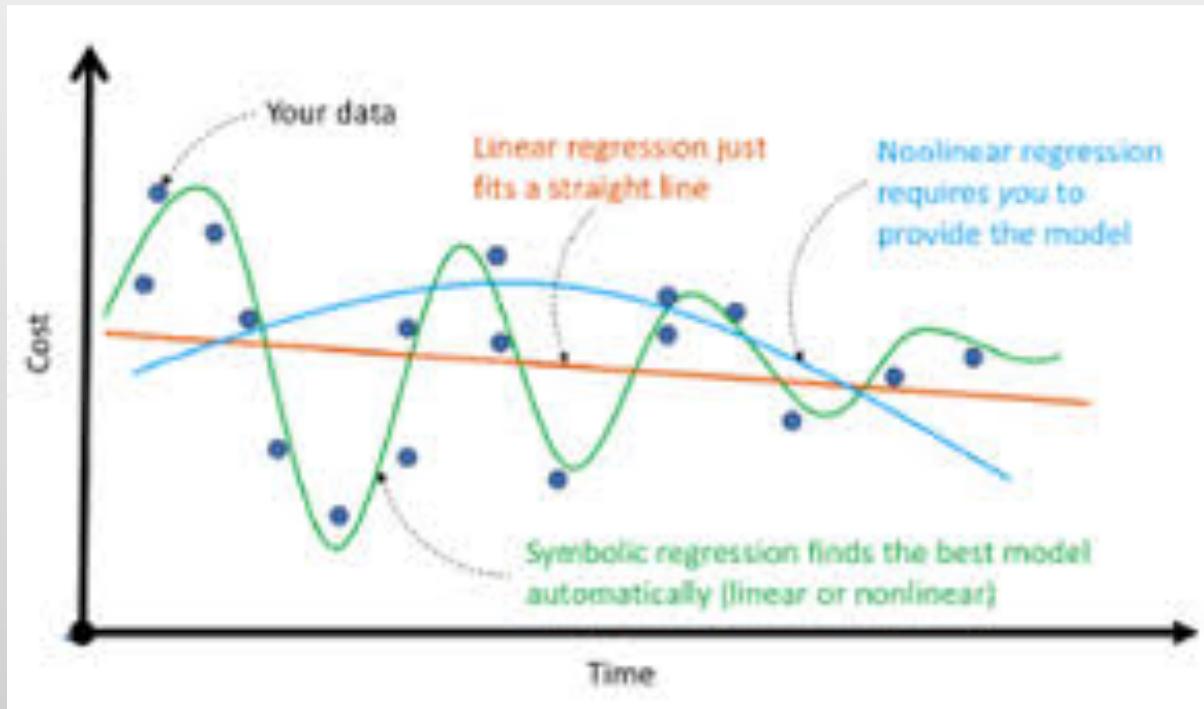
Data Science



Numerical and Symbolic regression

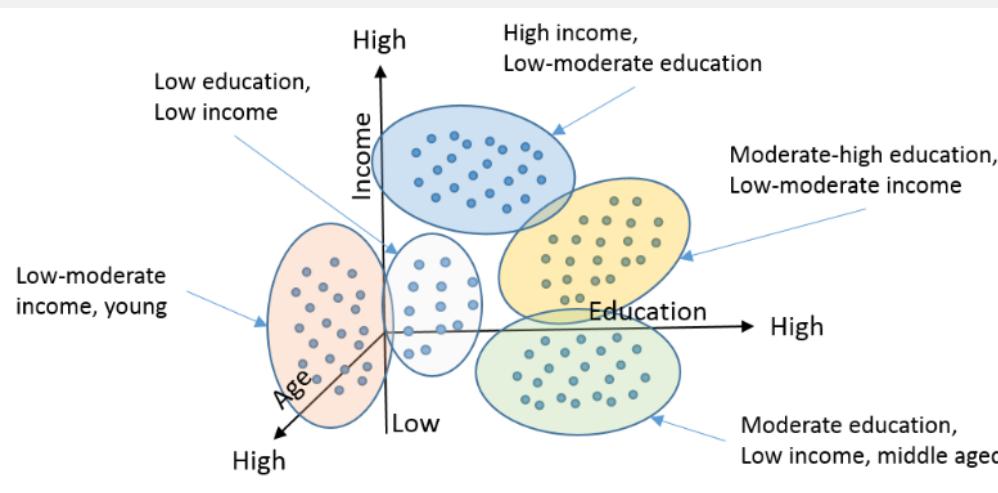
Numerical regression: best (linear or nonlinear fit)

Symbolic regression: finds the best model by searching symbolic **space of equations** (algebraic, ODE's, PDE's ...).
Goes more and more into AI territory (**Machine Learning**)



DS for decision-making

Classify (A,E,I) and give a score:



Identify a few qualifiers (order parameters in physics parlance) and **cluster** the data accordingly.

Decision-making: each individual gets a score and produces a binary (yes/no) output beyond a given score threshold.

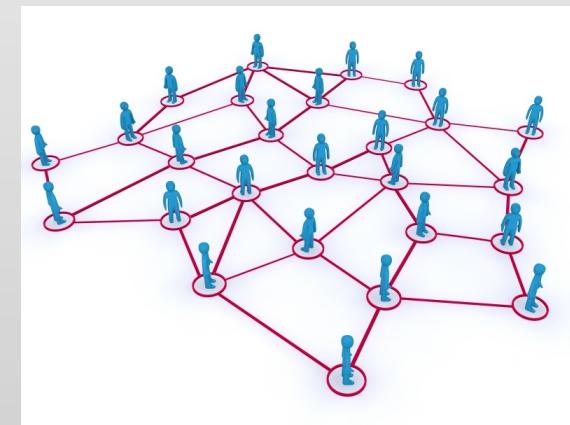
The binary output is the **DECISION:**
(SOCIAL/PSYCHO PHYSICS...)

$$S_i = f(A_i, E_i, I_i)$$

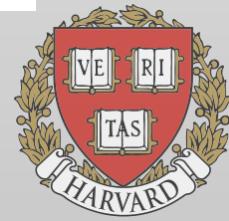
Decision:

Fix a bar and filter 0,1 = NO/YES

$$y_i = \text{sign}(S_i - S^*)$$

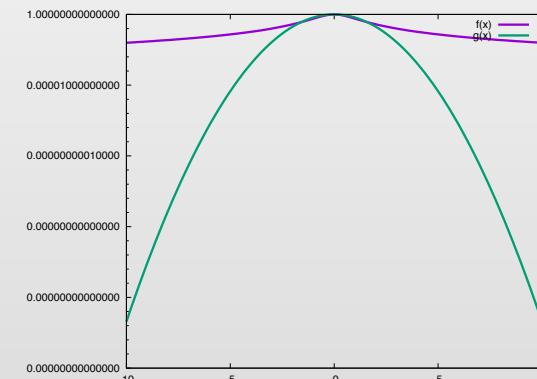
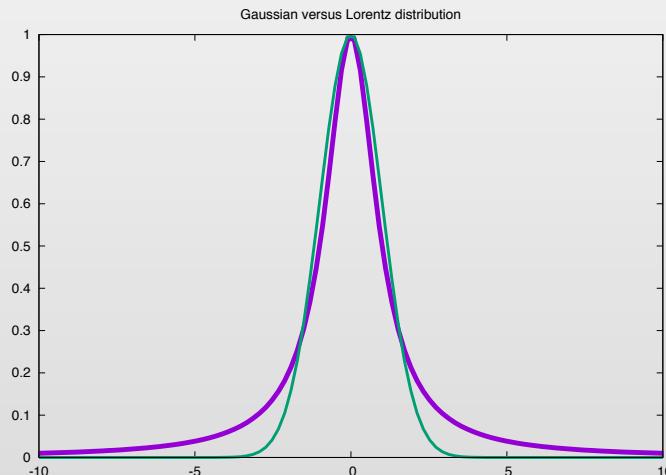


Sentiment Analysis:
(steer and manipulate?)
40



Wild fluctuations and rare events

Outliers (**rare events**) are heavily suppressed in the Gaussian world.
But the real world (complex systems) is hardly gaussian!



x

$$I = \int_{-\infty}^{+\infty} p(x)i(x)dx$$

x

Turbulence
Weather
Earthquakes
Finance

...

If $p(x)$ does not decay fast enough, the overall Intensity I
Is dominated by rare (small p) but intense (high i) events; divergent!

The Gauss distribution $\exp(-x^2/2)$ has finite moments at *all orders*
The Lorentz distribution $1/(1+x^2)$ has infinite variance

Why is Bigdata so sexy?

1. Extracting **patterns/trends** from Big Datasets can be MUCH faster than modeling/simulating the phenomena which produce the data
2. It applies to **all disciplines**, including those allegedly not math-friendly (too complex to be modeled) Biology, Medicine and esp Social, Communications and ... Marketing...

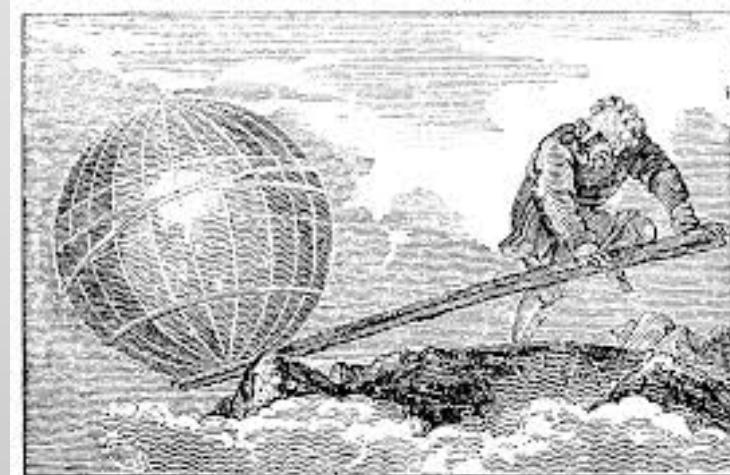


The Grand-Question:

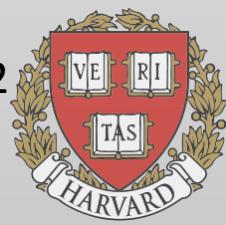
Is BigData the

- i) New Archimedes lever?
- ii) The graveyard of Theory?
- iii) The End of Insight?

See Chomsky vs Norvig, Strogatz...



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The end of Insight?



Is Bacon defeating Galileo?



Predict/Control/(Manipulate) without theoretical underpinning/understanding?

But...is it all blue-sky?

Big-Data depends on EXTRA-polation (from in_sample to out_of_sample).
Highly complex and/or singular behavior does not converge in the
for any reasonable N, **in fact ML can be much**
slower/less accurate than modeling!
Bogus correlations, Overfitting,

Can ML help with Quantum Gravity or Turbulence?

Not yet: Till then, down-toning and cooperation is healthier

AC274: examples of ML-assisted simulations in materials and turbulence

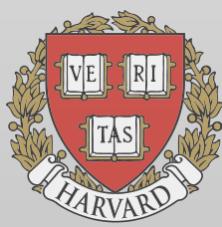
Summary

Computational Physics provides a third effective avenue for scientific exploration
Driven by hw advances and even more so, by methodological breakthroughs

Traditionally targeted to **continuum fields** (PDEs), is getting more and more
into **complex states of matter**, including **quantum, active matter and social systems**
More general mathematical structures (*fields, particles, probabilities,*
New synergies between Computational Modeling and Data Science
are key to future applications involving complex systems in science and society.

AC274 will help grasping this exciting
forefront of modern science:

ENJOY!



References

P. Moin, *Fundamentals of Engineering Analysis*, Cambridge U.P., 2001

(<https://www.amazon.com/Fundamentals-Engineering-Numerical-Analysis-Parviz/dp/0521711231>)

Excellent for grid methods (especially finite differences)

T. Pang, *Computational Physics*, Cambridge U. P., 4° ed., 2010

(<https://www.amazon.com/Introduction-Computational-Physics-Tao-Pang/dp/0521532760>)

Nice coverage of a very broad range of topics in classical and quantum physics

S. Succi, *The lattice Boltzmann Equation*, Oxford Univ. Press, 2001

(<https://www.amazon.com/Boltzmann-Numerical-Mathematics.../dp/0198503989>)

Lattice Boltzmann method for fluids and beyond

Y. Abu-Mostafa et al, *Learning from Data*, AMLbook.com, 2012

(<https://www.goodreads.com/book/show/15706459-learning-from-data>)

Nice introduction to Data Science and Machine Learning

End of the lecture

