

993SM - Laboratory of Computational Physics lecture I - I part March 10, 2021

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Introduction

(1) Computational Physics

- Simulations and "what-if" experiments
- Deterministic and stochastic approaches
- -A few examples
- (2) This course

(3) Other Courses concerning computational Physics in our Physics training track

> M. Peressi – UniTS – Laurea Magistrale in Physics Laboratory of Computational Physics – Unit I – part I

(1) Computational Physics

Computers in physics

- control of instruments, data collection and analysis
- visualization
- symbolic manipulation

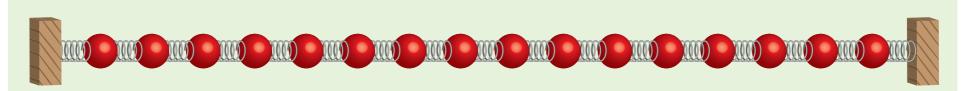


numerical analysis: to solve equations which could not be tackled by analytical methods. This allows to measure theories, in a similar way as natural phenomena are measured by experiments, the ultimate goal of science being the insight and understanding gained from the comparison of these two kinds of measures.

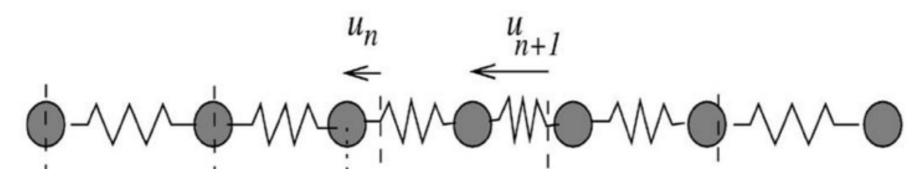
.....

• simulations: to model and study physical phenomena with numerical techniques. This means doing virtual experiments in which our representation of the physical reality, though necessarily schematic and simplified, can be tuned and varied at will.

PROBLEM: Fermi-Pasta-Ulam-Tsingou 1955



A chain of N particles linked by springs (one-dimensional analogue of atoms in a crystal)

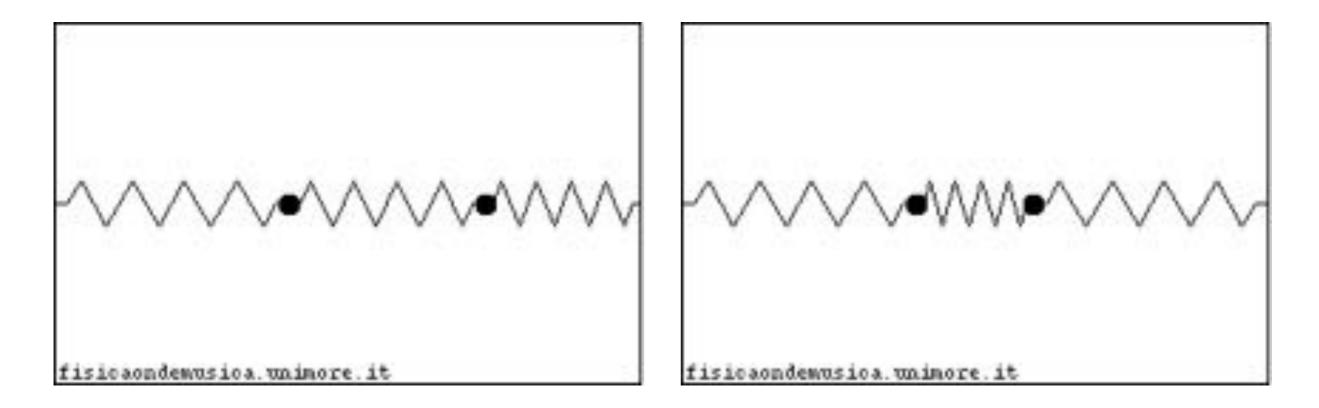


Linear interaction (Hooke's law):

there are N 'normal' modes

(i.e., patterns of motion in which all parts of the system oscillate with the same frequency and with a fixed phase relation)

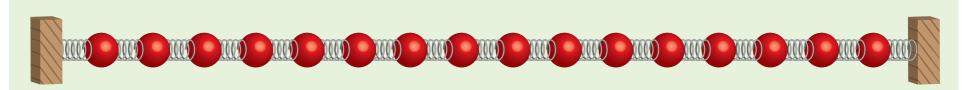
Example with 2 oscillators: 'normal' modes



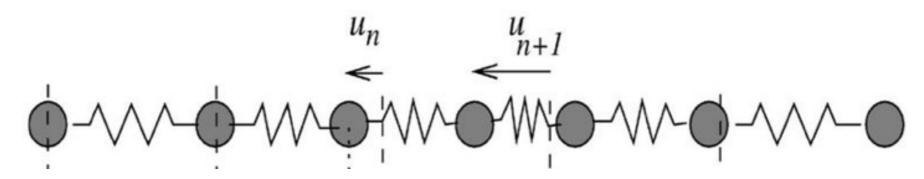
http://fisicaondemusica.unimore.it/Oscillatori_accoppiati.html

More in: http://fisicaondemusica.unimore.it/Catena_di_Fermi_Pasta_Ulam.html

PROBLEM: Fermi-Pasta-Ulam-Tsingou 1955



A chain of N particles linked by springs (one-dimensional analogue of atoms in a crystal)



Linear interaction (Hooke's law): analytical solution

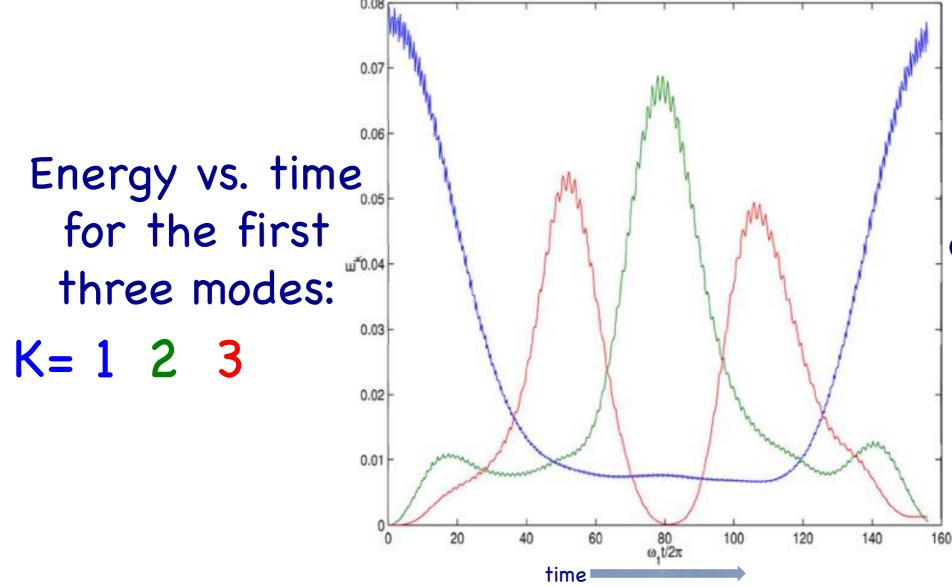
The energy given to a single 'normal' mode always remains in that mode.

PROBLEM: Fermi-Pasta-Ulam-Tsingou 1955

in presence of a weak non linear coupling (quadratic or cubic correction to the linear term), *which modes will be excited after a long enough time?*

Expected behavior based on *the equipartition theorem*: the energy will be equally distributed among all the degrees of freedom of the system. **However: analytical solution impossible**

PROBLEM: Fermi-Pasta-Ulam-Tsingou 1955 Numerical solution (originally: calculations for N=5)



After initial excitation of the mode k=1 and 157 periods, almost all the energy is back to this mode !!!

Pictures from: Thierry Dauxois and Stefano Ruffo (2008), Scholarpedia, 3(8):5538. <u>doi:10.4249/scholarpedia.5538</u> <u>http://www.scholarpedia.org/article/Fermi-Pasta-Ulam_nonlinear_lattice_oscillations</u> and other web sites

Simulations as "virtual experiments"

A few similarities between experiments:

"real" (in lab)	"virtual" (computational)
sample physical apparatus	model and algorithms code
calibration of instruments	test of the code
measurements data analysis	numerical results data analysis

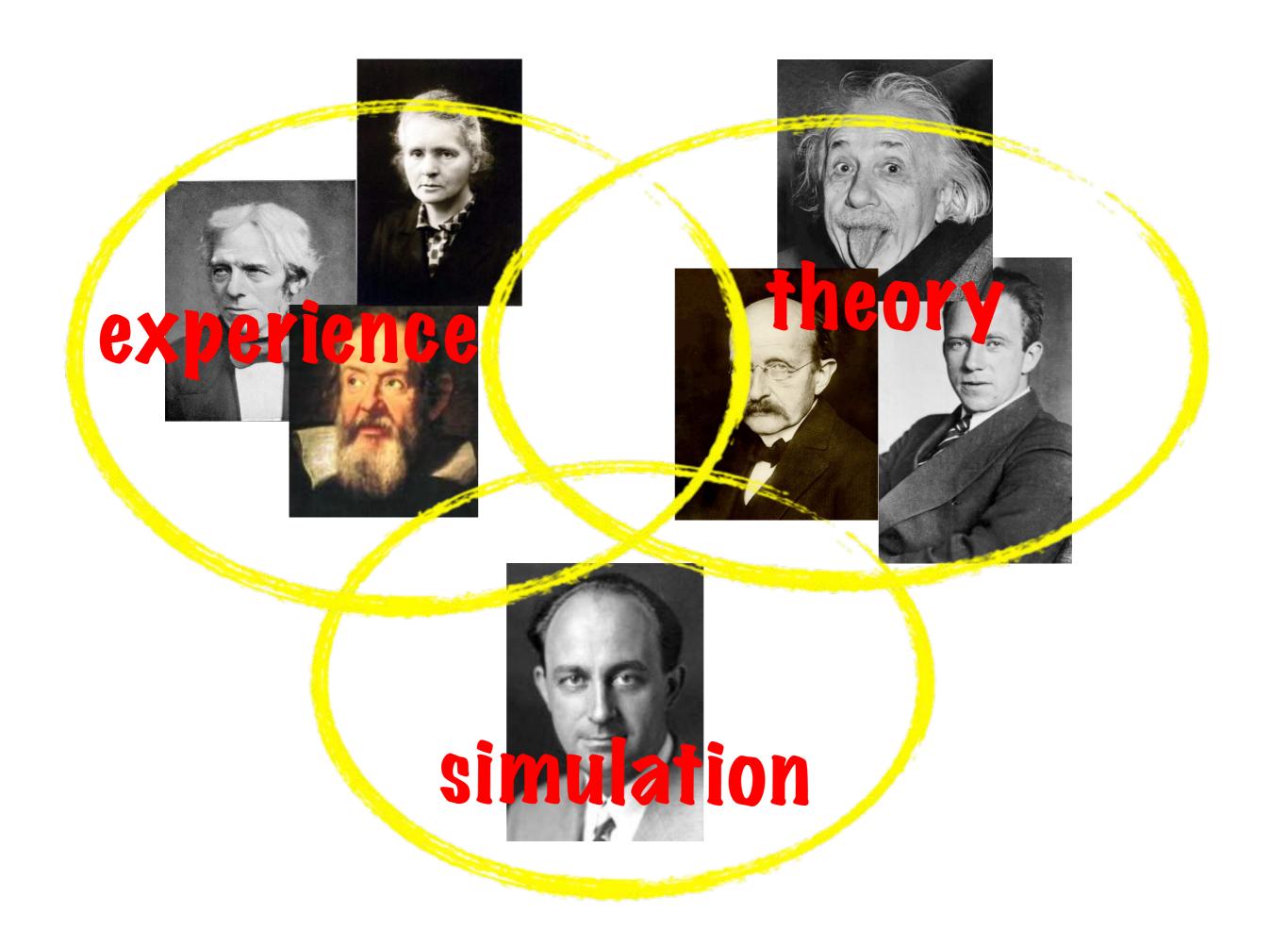
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With errors!!!

- <u>Importance of simulations</u>: "what-if" experiments (large flexibility in varying parameters; e.g. material properties can be studied also under conditions not accessible in real labs) ; predictions, not just description.
- <u>Use of simulations:</u> not "final goal", but "instruments" to study and shed light on complex phenomena and/or systems with many degrees of freedom or many variables and parameters
- in the last ~4 decades simulation has emerged as the third fundamental paradigm of science, beside theory and experiment



The purposes of the scientific calculus

- "The computer is a tool for clear thinking" (Freeman J. Dyson)
- "... whose [of the calculations] purpose is insight, not numbers" (Richard W. Hamming)

TWO different approaches for numerical simulations

deterministic

Info can be obtained both on the equilibrium properties and on the dynamics of the system

 stochastic (Monte Carlo, MC)
 Typically to simulate random processes, and/or sampling of most likely events

The deterministic approach

We can write the equations of motion (Classical => Newton; Quantum => Schroedinger)

and we know the initial condition

the problem is related to the <u>numerical integration of differential equations</u> (or integral-differential in quantum problems)

(like the FPUT problem)

The deterministic approach

Numerical integration of the eqs. of motion: discretization and iteration

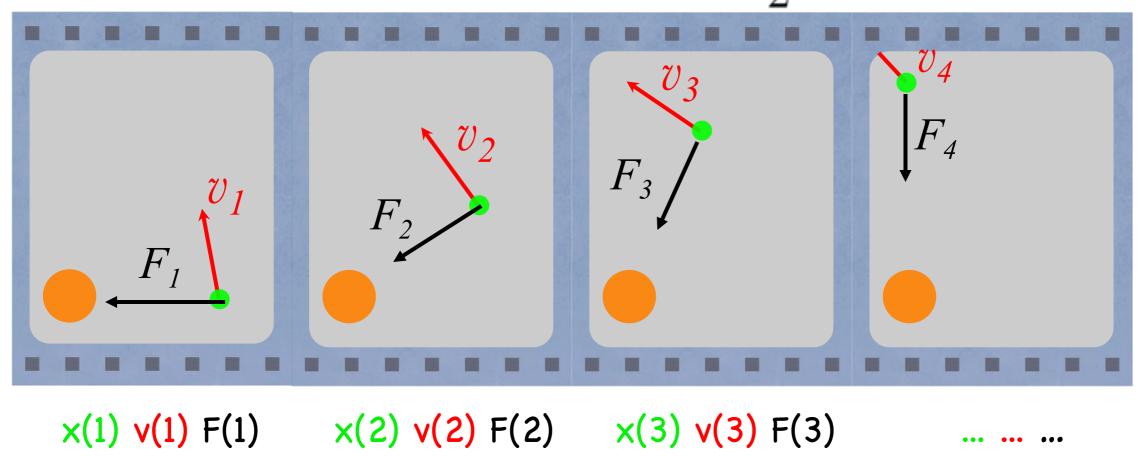
Different algorithms according whether the equation is 1st, 2nd order... (the equation for the velocity is 1st order), whether the force is dependent or not on the velocity, to which order...

Example (classical):

The deterministic approach

Discretization of the equation of motion and iteration:

$$a(t) = m \frac{d^2 x(t)}{dt^2} = \frac{F(t)}{m}$$
$$x(t + \Delta t) = x(t) + v(t)\Delta t + \frac{1}{2}a(t)\Delta t^2$$



The stochastic approach

Useful to model:

1) Some physical processes which are inherently probabilistic.

2) Many large classical systems which have so many variables, or degrees of freedom, that an exact treatment is intractable and not useful.

Probabilistic physical processes

We attempt to follow the `time dependence' of a model for which change, or growth, does not proceed in some rigorously predefined fashion (e.g. according to Newton's equations of motion) but rather in a stochastic manner which depends on a sequence of random numbers which is generated during the simulation.

E.g.: radioactive decay

Systems with many degrees of freedom

E.g.: Thermodynamic properties of gases

Impossible and not useful to know the exact positions and velocities of all molecules.

Useful properties are statistical averages: average energy of particles (temperature), average momentum change from collisions with walls of container (pressure), etc.

The error in the averages decreases as the number of particles increases. Macroscopic volume of gas has $O(10^23)$ molecules. Thus a statistical approach works very well!

Monte Carlo

Monte Carlo refers to any procedure which makes use of random numbers (*)

Monte Carlo is used in:

- -Numerical analysis
- -Statistical Mechanics Simulation

(*) a sequence of random numbers is a set of numbers which looks unpredictable but with well defined statistical properties

Monte Carlo Methods: to calculate integrals

"Hit or Miss" Method: How much is π ?

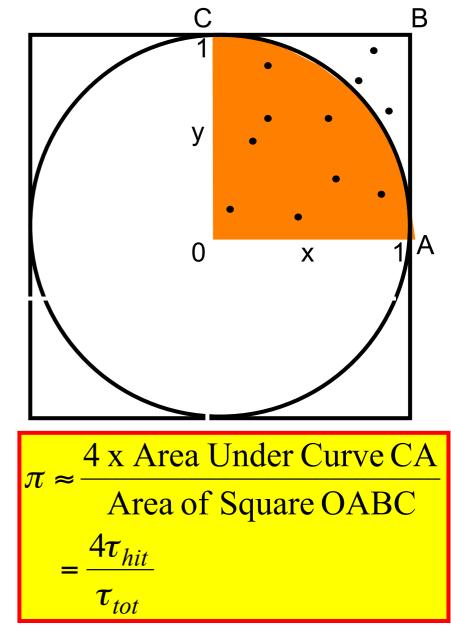
Algorithm:

•Generate uniform, random x and y between 0 and 1

 Calculate the distance from the origin: d=(x²+y²)^{1/2}

•If
$$d \leq 1$$
, $\tau_{hit} = \tau_{hit} + 1$

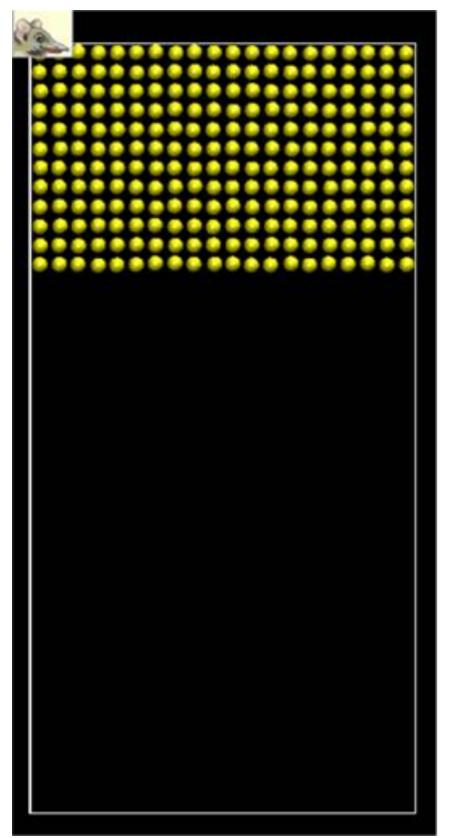
•Repeat for τ_{tot} trials



A few selected examples of applications

(here: atomistic simulations in condensed matter...)

From "normal" scales...



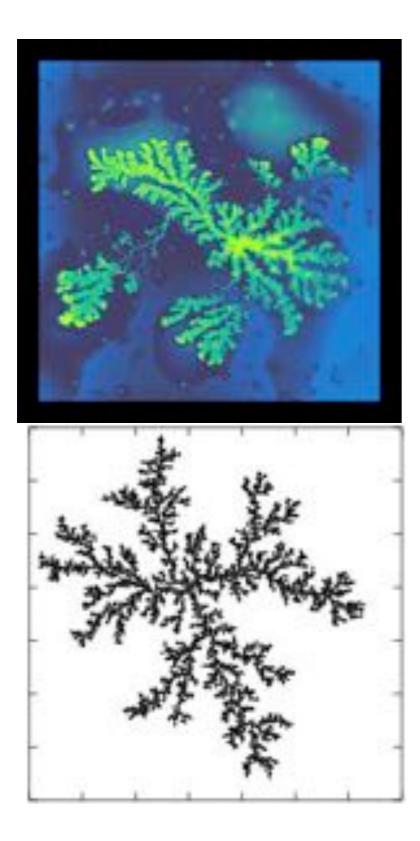
(classical)

SIMULATION of the Brownian motion

Sedimentation of hard spheres in a 2D system with walls. Included interactions with smaller particles (not shown here) representing the thermohydrodynamic solvent

(deterministic, classical simulation)

... colloidal systems growth on a substrate...



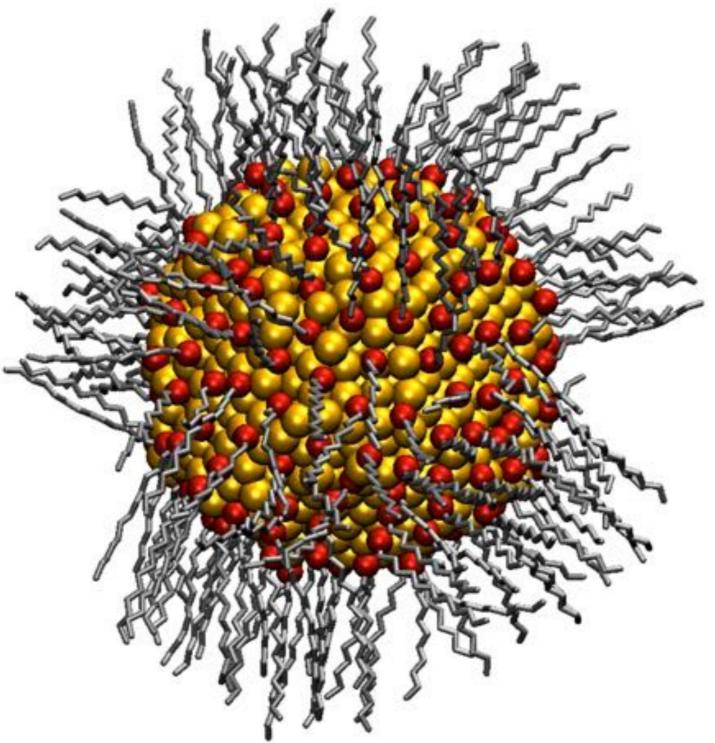
REAL IMAGE (by Atomic Field Microscopy) of a gold colloid of about 15 nm on a mica substrate

SIMULATION of a diffusion-limited auto-aggregation model (fractal)

(stochastic, classical simulation)

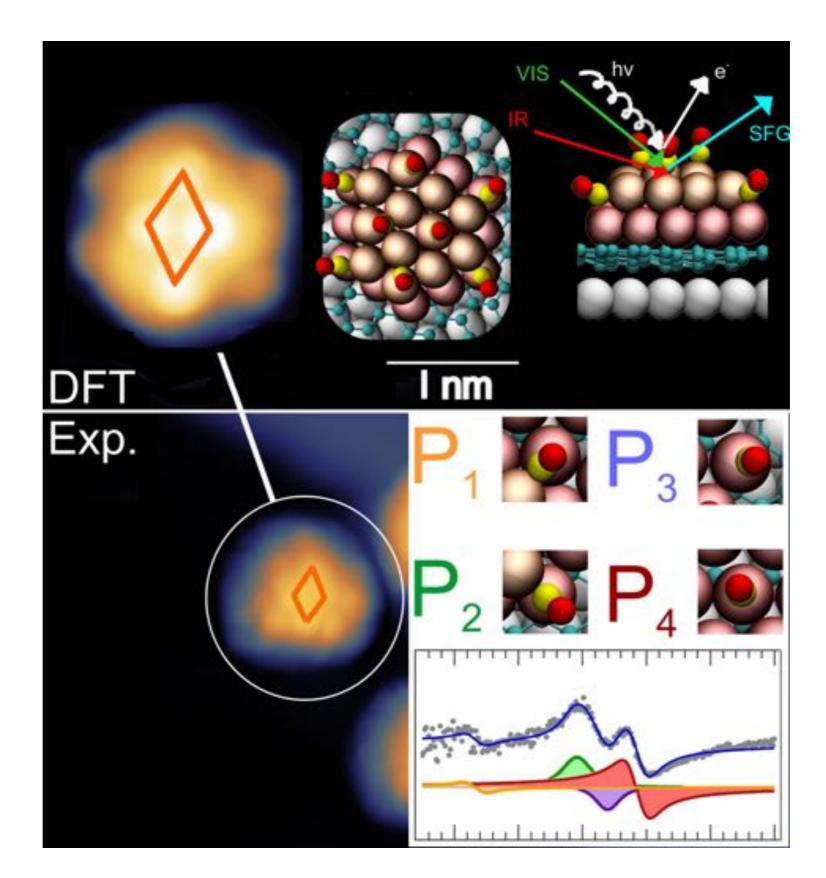
... to the nanoscale: passivation of nanoparticles

with organic molecules (thiols) Au S

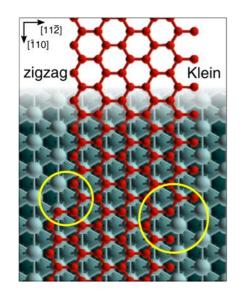


Credits: J. Olmos-Asar

... to the atomic scale



(M.P. in collaboration with E. Vesselli and G. Comelli)

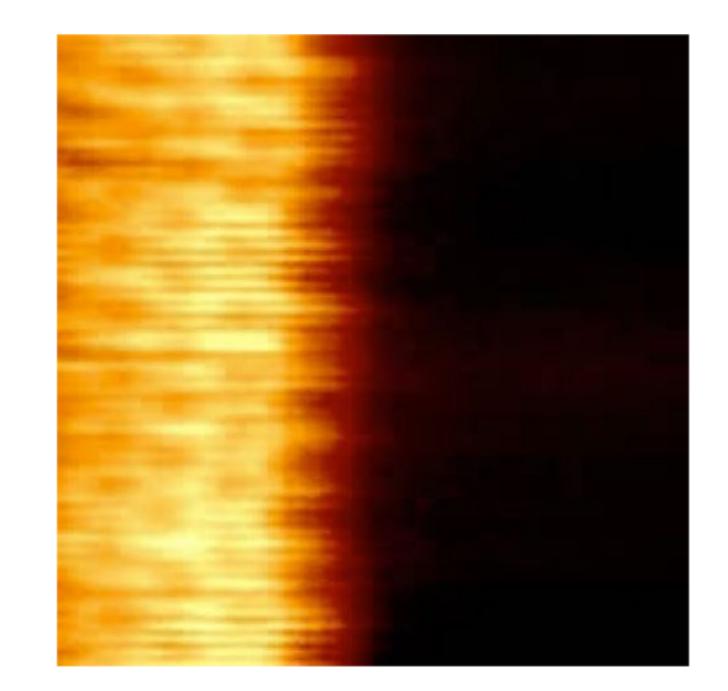


Grafene @Ni(111)

(M.P. in collaboration with TASC; Science, 2018)

EXPERIMENTAL FAST-STM MOVIE

... another example

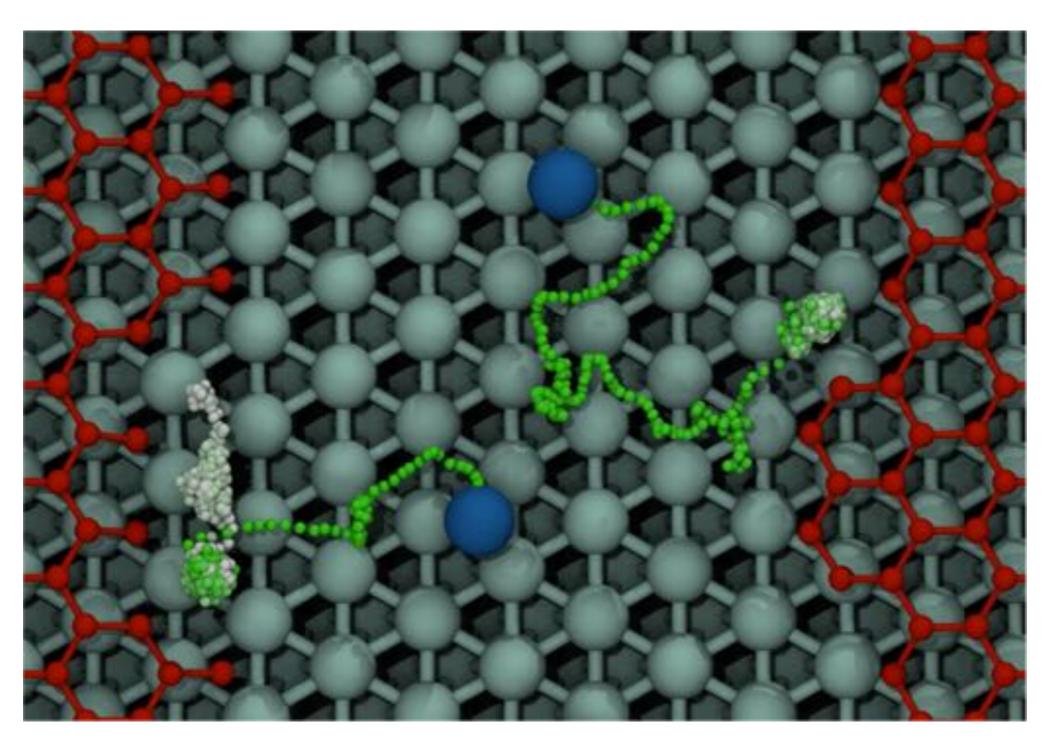


... another example

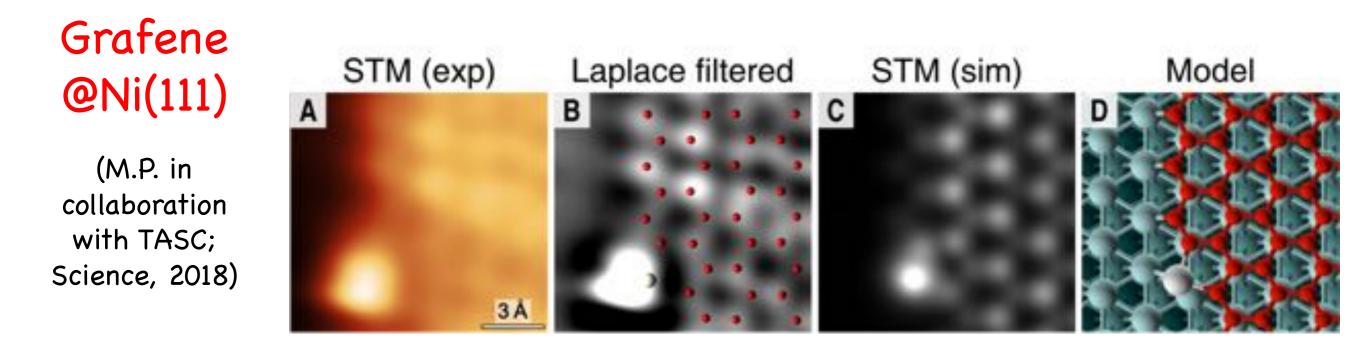


(M.P. in collaboration with TASC; Science, 2018)

> CLASSICAL MOLECULAR DYNAMICS SIMULATION

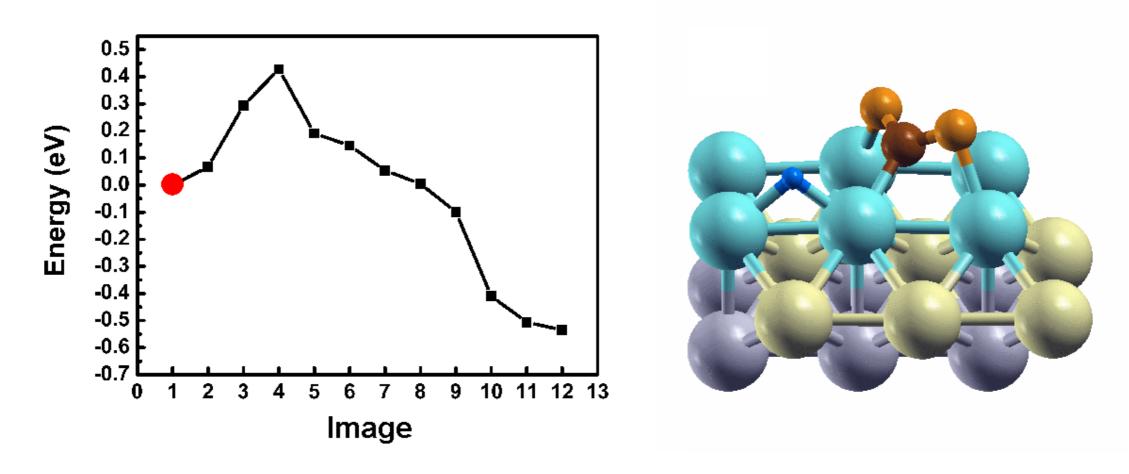


... another example



QUANTUM MECHANICAL SIMULATION

...including chemical reactions



CO₂ + H -> HCOO @ Ni(110)

(deterministic, quantum mechanical simulation)

even within the condensed matter:

- wide range of length scales: ≈12 orders of magnitude (nuclei/electrons/atoms/chemical bonds ~ 10-12 m, fracture/macroscopic mechanical phenomena ~ 100 m; nano / micro / meso / macroscopic scales)
- wide range of time scales: ≈12 orders of magnitude (nuclei/electrons/atoms/chemical bonds ~ 10-12 s, fracture/macroscopic mechanical phenomena ~ year)
- wide range of chemical-physical properties: structural, elastic, vibrational, electronic, dielectric, magnetic, optical, thermal . . .
- wide range of materials: different phases, traditional materials (crystalline / amorphous , metals/ semiconductors / insulators . . .), new materials. . .

different kind of interactions

- Classic
- Quantum

different approaches

- Deterministic
- Stochastic

...and also different specific techniques

corresponding to different size/time scales:

- continuous models (for macroscopic systems)
- atomistic simulations
- ab initio techniques (or "first-principles"): up to ~10^3 atoms, 10 ps
- Semiempirical techniques: up to 10~7 atoms, 1 ms
- models at different levels

...and different computational workload

Some techniques and systems are not computationally very demanding (our experiments will be quite small and simple)

others, they are...

High performance computing

https://www.top500.org/lists/top500/2020/11/

Rank	System	Cores	Rmax (TFlop/s)	Rpeak (TFlop/s)	Power (kW)
1	Supercomputer Fugaku - Supercomputer Fugaku, A64FX 48C 2.2GHz, Tofu interconnect D, Fujitsu RIKEN Center for Computational Science Japan	7,630,848	442,010.0	537,212.0	29,899
2	Summit - IBM Power System AC922, IBM POWER9 22C 3.07GHz, NVIDIA Volta GV100, Dual-rail Mellanox EDR Infiniband, IBM DOE/SC/Oak Ridge National Laboratory United States	2,414,592	148,600.0	200,794.9	10,096
3	Sierra - IBM Power System AC922, IBM POWER9 22C 3.1GHz, NVIDIA Volta GV100, Dual-rail Mellanox EDR Infiniband, IBM / NVIDIA / Mellanox DOE/NNSA/LLNL United States	1,572,480	94,640.0	125,712.0	7,438

High performance computing https://www.top500.org/lists/top500/2020/11/

8	HPC5 - PowerEdge C4140, Xeon Gold	669,760	35,450.0	51,720.8	2,252
	6252 24C 2.1GHz, NVIDIA Tesla V100,				
	Mellanox HDR Infiniband, Dell EMC				
	Eni S.p.A.				
	Italy				

1	11	Marconi-100 - IBM Power System AC922, IBM POWER9 16C 3GHz, Nvidia Volta V100,	347,776	21,640.0	29,354.0	1,476	
		Dual-rail Mellanox EDR Infiniband, IBM					
		CINECA					
		Italy					

(usage with grants; agreement also with UniTS)

(2) This course

This course

- IS NOT a course on Information Technology, Computer Science, Programming languages...
- BUT a **PHYSICS LAB**.
- focusing on modeling, problem solving and algorithms
- Not exhaustive, of course...



- Stochastic approach, classical interactions (mainly)
- + <u>basic</u> ingredients of the deterministic approach (Molecular Dynamics) and quantum mechanics (Variational Monte Carlo) (1 week each topic)

TENTATIVE LIST OF ARGUMENTS

- Properties and generation of Random Numbers with different distributions.
- Monte Carlo simulation of Random Walks.
- Numerical integration in 1 dimension: deterministic and stocastic algorithms;
- Monte Carlo algorithms.
- Error estimate and reduction of the variance methods.
- Metropolis algorithm for arbitrary random number generation.
- Metropolis method in the canonical ensamble.
- Ising model and Metropolis-Monte Carlo simulation.
- Classical fluids: Monte Carlo and Molecular Dynamics simulation of hard spheres and Lennard-Jones fluids.
- Microstates and macrostates: efficient algorithm for the numerical calculation of entropy.
- Variational Monte Carlo in quantum mechanics (basics).
- Lattice gas: vacancy diffusion in a solid.
- Caos and determinism: classical billiards and caotic billiards, logistic maps; Lyapunov exponents.
- Fractals: diffusion and aggregation, models for surface growth simulation. Percolation.

Course on moodle2

You can find the material <u>lecture by lecture</u>

- -Important announcements
- -Detailed contents of each lecture
- -Lectures notes
- -Exercises
- -Info about textbooks
- -links, tutorials (for surviving with Linux/Unix, Fortran90, gnuplot...)
 -Info about exams

SUBSCRIBE TO THE COURSE on moodle2! https://moodle2.units.it Then select: ⇒Dipartimento di Fisica \Rightarrow Laurea Magistrale \Rightarrow SM23 - FISICA ⇒A.A. 2020-2021

Or point directly to: <u>https://moodle2.units.it/course/view.php</u> <u>?id=7238</u>

Fortran compiler on the virtual Lab

- gfortran (free): ([] for optional)
- \$ gfortran [-std=f95] [-o test.o] test.f90

The option -std=f95 allows to obtain, after compilation, supplementary info about the commands you wrote in your code (the syntax, whether it is standard fortran or not...)

OPTIONS ARE IMPORTANT AND USEFUL!

Other possible Fortran compilers

• g95 (free)

ifort (Fortran Intel compiler, NOT free)

F (free; useful options: -ieee=full for floating point exception manipulation)

- To run the executables (e.g. test.o or a.out by default):
 - \$./a.out (or \$bash a.out)

A few useful UNIX (Linux, MacOSx,...) commands:

Check your space!

\$ quota

or "du" (displays disk usage statistics):

\$ du ~ | more

(if "-k" flag is specified, the number of 1024-byte blocks used by the file is displayed):

 $du - k \sim |$ more (Last line shows the total)

\$ find . -size +20000 -print (to identify big files)

3) Other Courses concerning computational Physics in our Physics training track

- complementary to "Classical simulations of many body systems" (E. Smargiassi, I semester) (deterministic, classical)
- complementary to "Numerical Methods of Quantum Mechanics" (P. Giannozzi, II semester) (deterministic, quantum)