Laboratory of Computational Physics

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- (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

(1) Computational Physics

Computers in physics

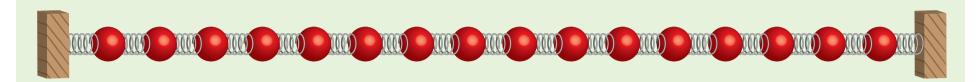
- control of instruments, data collection and analysis
- visualization
- symbolic manipulation



• 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 tupod and varied at will

OBLEM: Fermi-Pasta-Ulam-Tsingou 1955



Linear interaction (Hooke's law):

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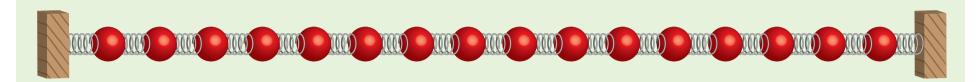
Example with 2 oscillators: 'normal' modes

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

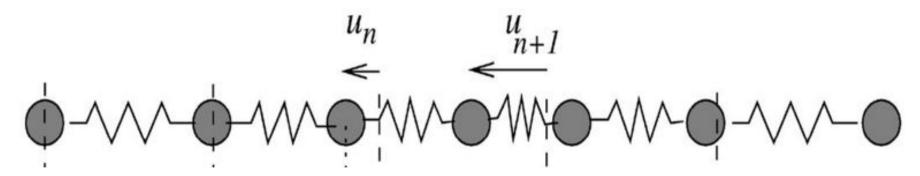
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OBLEM: Fermi-Pasta-Ulam-Tsingou 1955



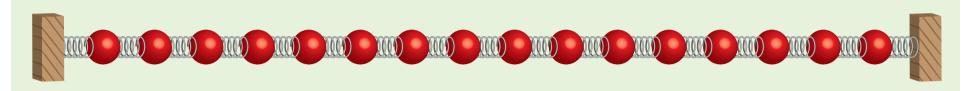
A chain of N particles linked by springs (one-dimensional analogue of atoms in a crystal)



Linear interaction (Hooke's law): analytical solution

The end of the strength of the strength of the second of t

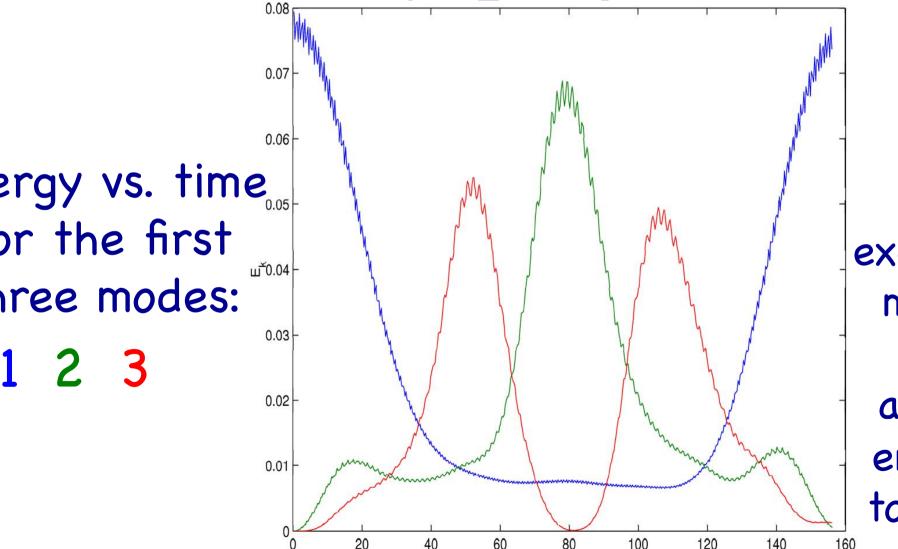
OBLEM: Fermi-Pasta-Ulam-Tsingou 1955



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

xpected behavior based on *the equipartition theorem*: the energy will be equally distributed among all the degrees of freedom of the system.

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



After initial excitation of t mode k=1 and 157 periods, almost all th energy is bac to this mode

Simulations as "virtual experiments"

A few similarities between experiments:

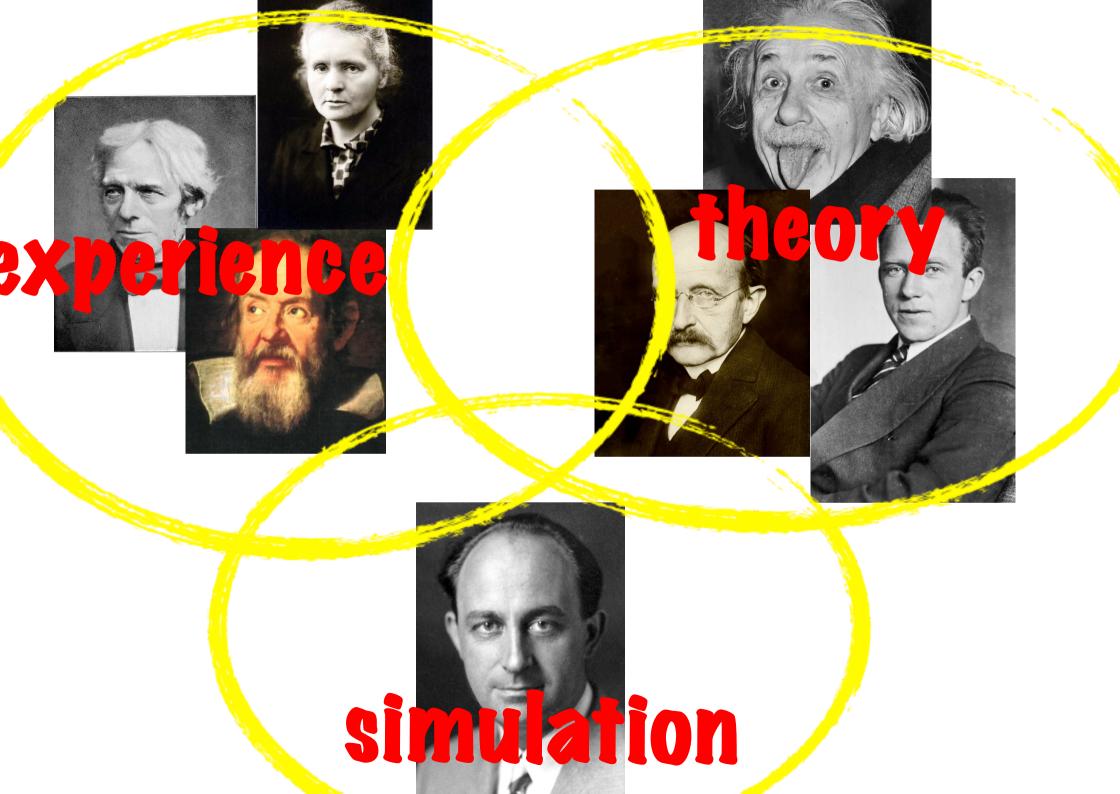
al" (in lab)	"virtual" (computatio
nple	model and algorithms
vsical apparatus	code
bration of instruments	test of the code
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a analysis	data analysis

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- <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



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" (V. 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

ine aeterministic approacn

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)

ine aeterministic approacn

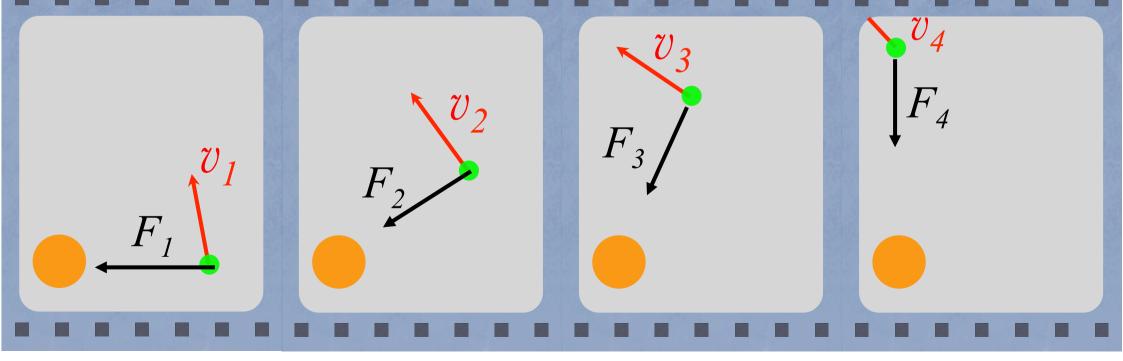
Numerical integration of the eqs. of motion: discretization and iteration

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

ine aeterministic approacn

Discretization of the equation of motion and iteration:

$$a(t) = m \frac{d^2 x(t)}{dt^2} = \frac{F(t)}{m} \frac{1}{2} \frac{F(t)}{t^2}$$
$$x(t) = x(0) + v(0)t + \frac{1}{2}a(t)\Delta t^2$$
$$x(t + \Delta t) = x(t) + v(t)\Delta t + \frac{1}{2}a(t)\Delta t^2$$



(1) (1) E(1) (0) (0) E(0) (2) (2) (2)

The stochastic approach

- seful to model:
- Some physical processes which are herently probabilistic.
-) Many large classical systems which ave so many variables, or degrees of reedom, that an exact treatment is tractable and not useful.

Prodadilistic physical processes

- e attempt to follow the `time dependence a model for which change, or growth, doe t proceed in some rigorously predefined shion (e.g. according to Newton's uations of motion) but rather in a ochastic manner which depends on a quence of random numbers which is nerated during the simulation.
- g.: radioactive decay

with many degrees of freedom

- .: Thermodynamic properties of gases
- possible and not useful to know the exact positions a ocities of all molecules.
- e<mark>ful properties are statistical averages:</mark> average ener particles (temperature), average momentum change m collisions with walls of container (pressure), etc.
- e error in the averages decreases as the number of ticles increases. Macroscopic volume of gas has 0^23) molecules. Thus a statistical approach works ve

monte cario

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

Nonte 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

to calculate integrals

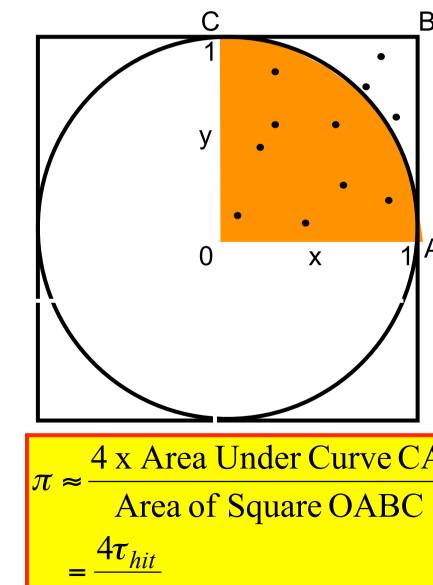
"Hit or Miss" Method: How much is π ?

Igorithm:

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

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

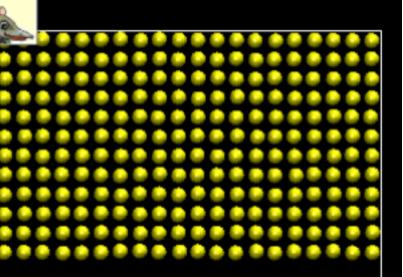
•If $d \leq 1$, $\tau_{hit} = \tau_{hit} + 1$



A few selected examples of applications

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

rom normal scales...



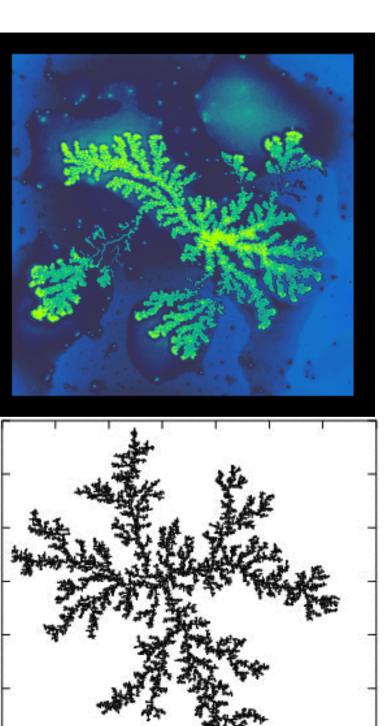
(classical)

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

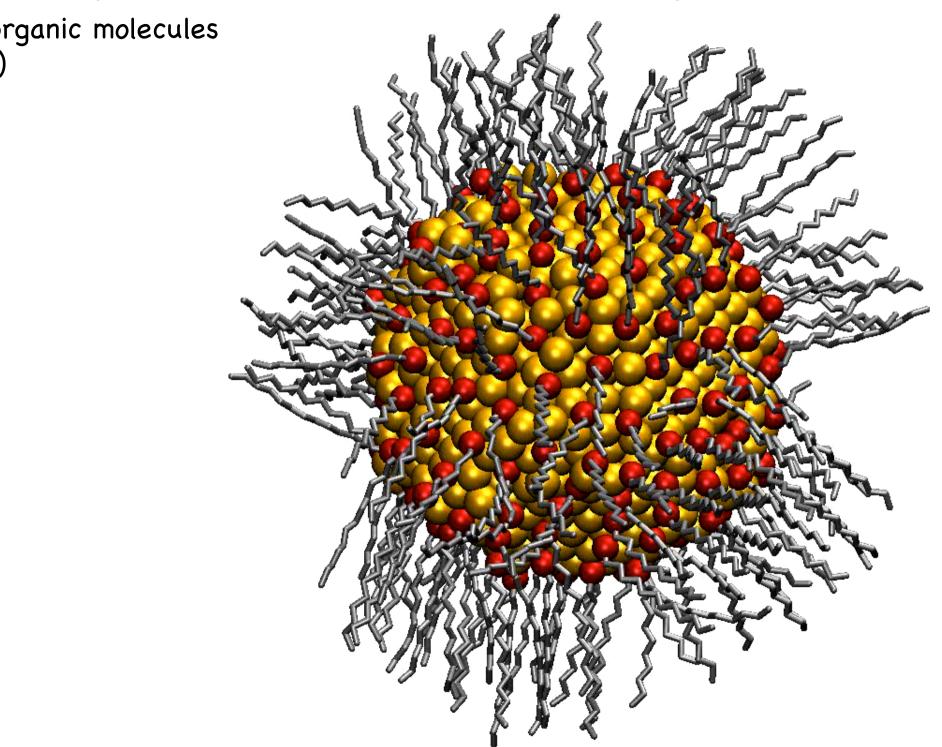
substrate...



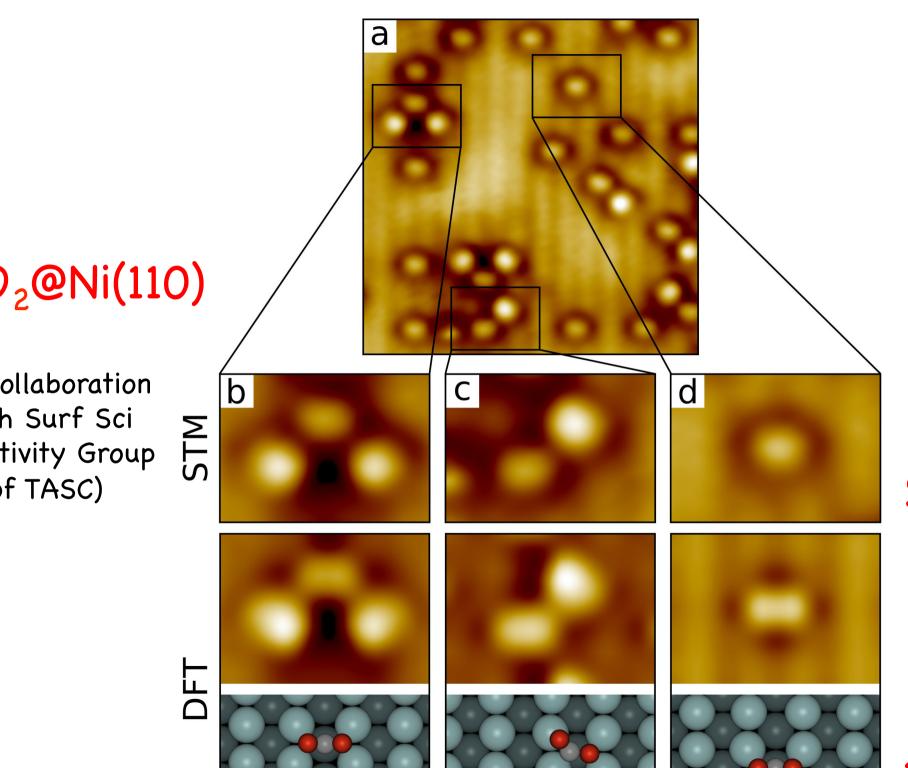
REAL IMAGE (by Atomic Field Microscopy) of a gold colloid of about 15 nm on a mica substrate

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

passivation of nanoparticles



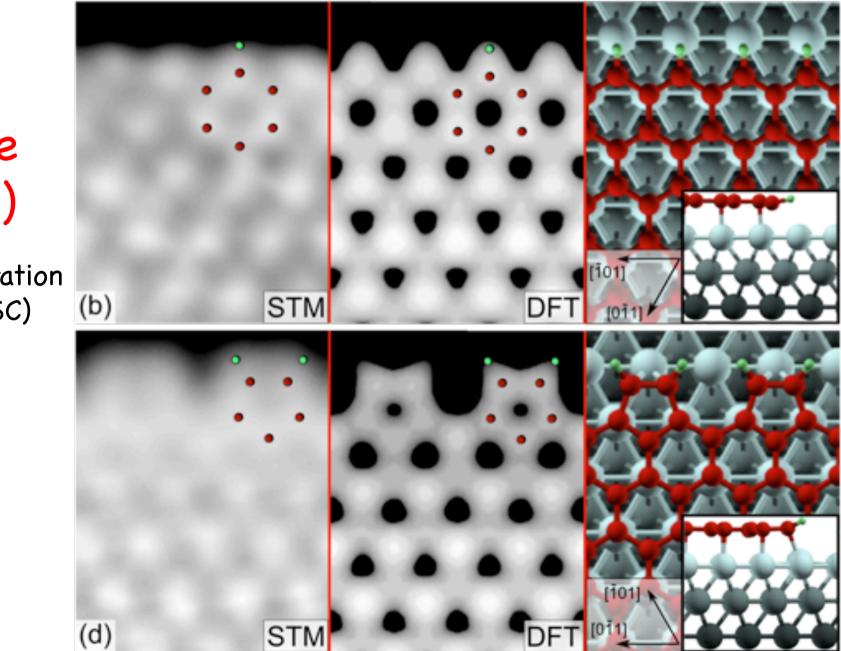
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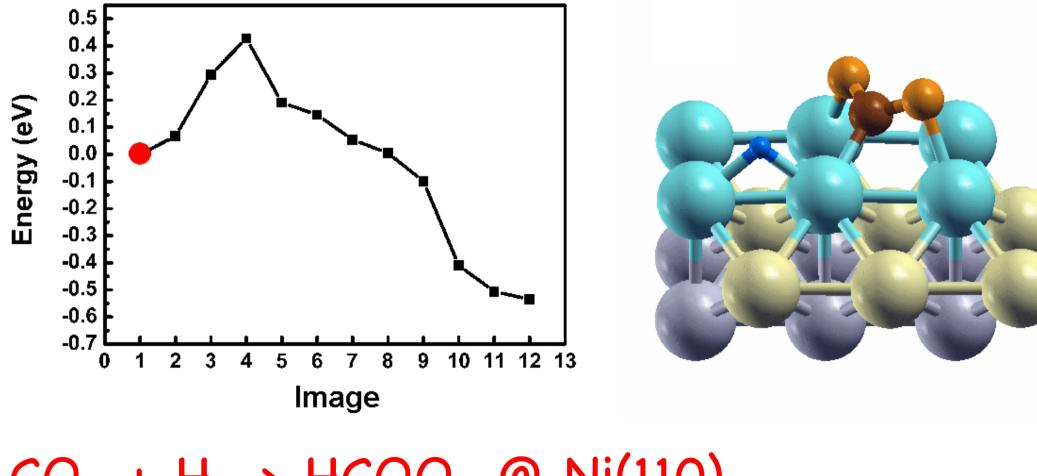
RT (H-terminated edges)



afene li(111)

ollaboration ith TASC)

...including chemical reactions



CO₂ + H -> HCOO @ Ni(110)

- 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

techniques

- corresponding to different size/time scales:
- continuous models (for macroscopic systems)
- atomistic simulations
- ab initio techniques (or "first-principles"): up ~10^3 atoms, 10 ps
- Semiempirical techniques: up to 10~7 atoms, 1 m
- 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...





November 2015

Tianhe-2 (MilkyWay-2) NUDT

ional Super Computer Center in Guangzhou)

000 nodes, with 3,120,000 cores with CPU+GPU ch node: different Intel Xeon types of processors d GPU, like those for videogames!!!) ak performances: up to 34 PFlop ta = 10¹⁵)

op = "floating-point" operation / second)







November 2015

FERMI - BlueGene/QIBM, CPU+GPU

e with grants; agreement also with UniTS)



163.840 computing cores ar peak performace of 2,1 PFI



(2) This course

inis 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...

inis 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)

p://www-dft.ts.infn.it/~peressi/comp-phys.ht

With:

- nportant announcements
- etailed contents of each lecture
- ectures notes
- kercises
- fo about textbooks
- ks, tutorials (for surviving with Linux/Unix, rtran90, gnuplot...)
- fo about exams

- carto sinutanon or Random Walks.
- rical integration in 1 dimension: deterministic and stocastic algorithms;
- e Carlo algorithms.•
- estimate and reduction of the variance methods.
- polis algorithm for arbitrary random number generation.
- polis method in the canonical ensamble.
- model and Metropolis-Monte Carlo simulation.
- cal fluids: Monte Carlo and Molecular Dynamics simulation of hard spheres and ennard-Jones fluids.
- states and macrostates: efficient algorithm for the numerical calculation of ent
- tional Monte Carlo in quantum mechanics (basics).•
- e gas: vacancy diffusion in a solid.
- and determinism: classical billiards and caotic billiards, logistic maps; Lyapunov
- also diffusion and agarogation models for surface arowth simulation. Dercelation

Available computational resources: INFIS

p://df2.units.it/?q=it/modulistica

p://df2.units.it/?q=it/node/2919/ DATTICA => SERVIZI AGLI STUDENTI

> Remote connection: \$ssh username@w01.infis.units.it

<u>Your address at INFIS:</u> username@infis.univ.trieste.it

You can copy the source codes <u>lecture by lecture (wait</u> updates!!! Do not copy everything today!!!) from:

<u>tp://www.infis.units.it/~peressi/</u> (PUBLIC ACCE

- on INFIS, you find the <mark>source codes</mark> in the directory (you need to have an account on INFIS):
- nome/peressi/comp-phys
- d subdirectories (I-basic, etc. etc...)
- m your directory, do:
- p /home/peressi/comp-phys/I-basics/* .

- g95 or gfortran (free): ([] for optional)
 - \$ g95 [-o test.o] test.f90 or
 - \$ gfortran [-std=f95] [-o test.o] test.f90
- OTHERS (NOT SUPPORTED ON INFIS):
 - ifort (Fortran Intel compiler, NOT free)
 - F (free; useful options: -ieee=full for floating point exception manipulation)
- 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)

computational Physics in our Physics training track

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