



Overview on Parallel Programming Paradigms

Ivan Girotto – igirotto@ictp.it

Information & Communication Technology Section (ICTS)
International Centre for Theoretical Physics (ICTP)





What Determines Performance?

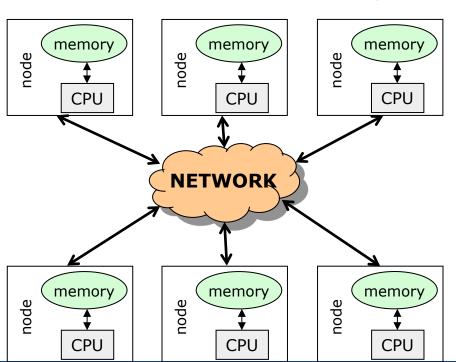
- How fast is my CPU?
- How fast can I move data around?
- How well can I split work into pieces?
 - Very application specific: never assume that a good solution for one problem is as good a solution for another
 - always run benchmarks to understand requirements of your applications and properties of your hardware
 - respect Amdahl's law



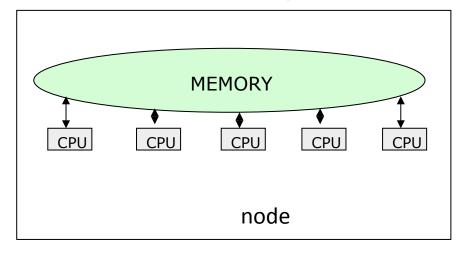


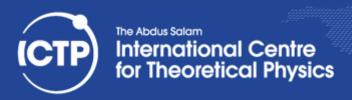
Parallel Architectures

Distributed Memory



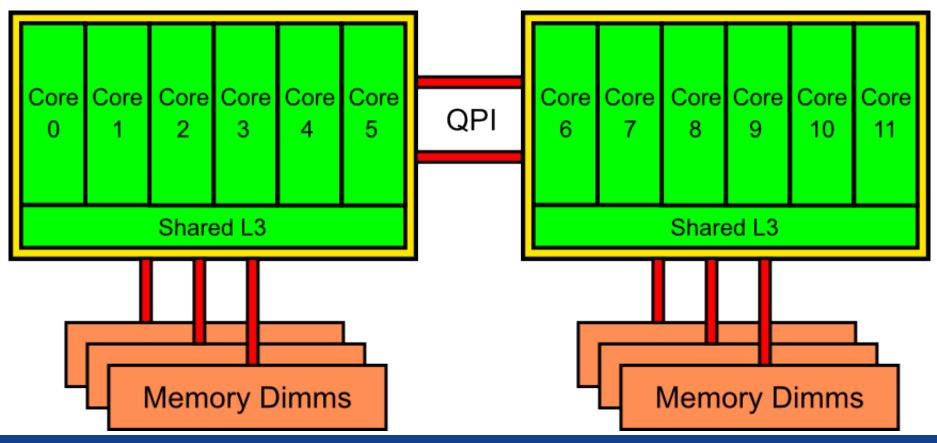
Shared Memory







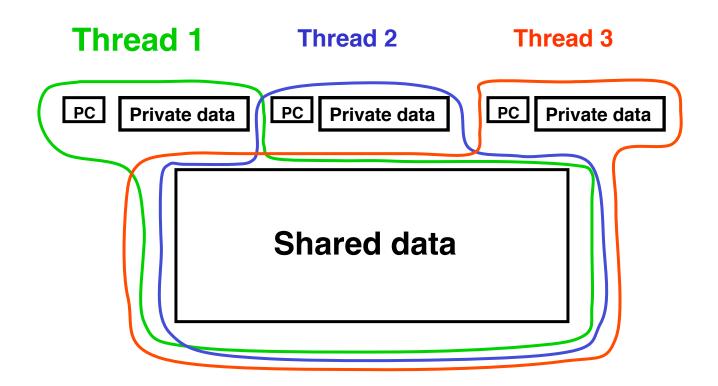
Multiple Socket CPUs

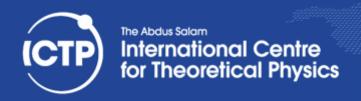






Paradigm at Shared Memory /1

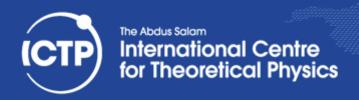






Paradigm at Shared Memory /2

- Usually indicated as Multithreading Programming
- Commonly implemented in scientific computing using the OpenMP standard (directive based)
- Thread management overhead
- Limited scalability
- Write access to shared data can easily lead to race conditions and incorrect data





Parallel Programming Paradigms

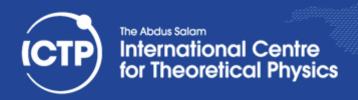
- MPI (Message Passing Interface)
 - A standard defined for portable message passing
 - It available in the form of library which includes interfaces for expressing the data exchange among processes
 - A framework is provided for spawning the independent processes (i.e., mpirun)
 - Processes communication is via network
 - It works on either shared and distributed mem. architecture
 - ideal for distributing memory among compute nodes





MPI Program Design

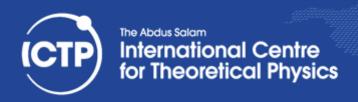
- Multiple and separate processes (can be local and remote) concurrently that are coordinated and exchange data through "messages" => a "share nothing" parallelization
- Best for coarse grained parallelization Distribute large data sets; replicate small data
- Minimize communication or overlap communication and computing for efficiency => Amdahl's law





What is MPI?

- A standard, i.e. there is a document describing how the API (constants & subroutines) are named and should behave; multiple "levels", MPI-1 (basic), MPI-2 (advanced), MPI-3 (new)
- A library or API to hide the details of low-level communication hardware and how to use it
- Implemented by multiple vendors
- Open source and commercial versions
- Vendor specific versions for certain hardware
- Not binary compatible between implementations





Programming Parallel Paradigms

- Are the tools we use to express the parallelism for on a given architecture
- They differ in how programmers can manage and define key features like:
 - parallel regions
 - concurrency
 - process communication
 - synchronism



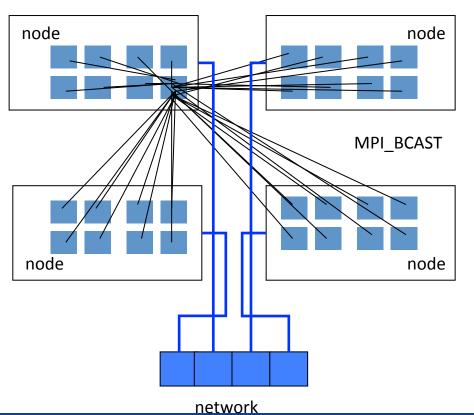






MPI inter process communications

MPI on Multi core CPU



1 MPI proces / core Stress network Stress OS

Many MPI codes (QE) based on ALLTOALL Messages = processes * processes

We need to exploit the hierarchy



Re-design applications



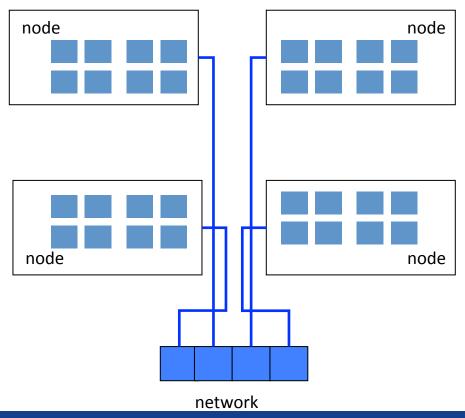
Mix message passing And multi-threading







The Hybrid Mode

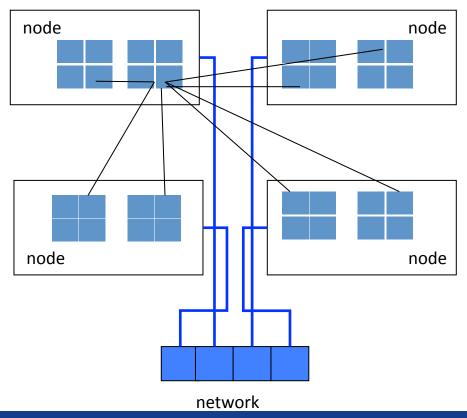








The Hybrid Mode











~ 8 GBytes



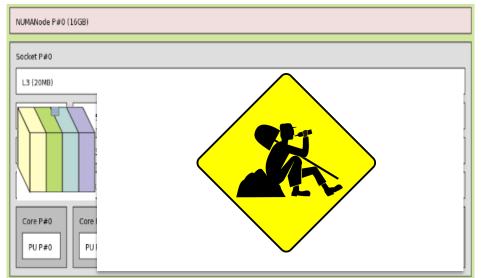
The Intel Xeon E5-2665 Sandy Bridge-EP 2.4GHz

mpirun -np 8 pw-gpu.x -inp input file













The Intel Xeon E5-2665 Sandy Bridge-EP 2.4GHz

mpirun -np 1 pw-gpu.x -inp input file













The Intel Xeon E5-2665 Sandy Bridge-EP 2.4GHz

export OMP_NUM_THREADS=4
export OPENBLAS_NUM_THREADS=\$OMP_NUM_THREADS
mpirun -np 2 pw-gpu.x -inp input file







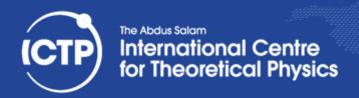
Workload Management: system level, High-throughput

Python: Ensemble simulations, workfows

MPI: Domain partition

OpenMP: Node Level shared mem

CUDA/OpenCL/OpenAcc: floating point accelerators







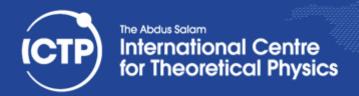
Type of Parallelism

• <u>Functional (or task) parallelism</u>: different people are performing different task at the same time



• <u>Data Parallelism</u>: different people are performing the same task, but on different equivalent and independent objects





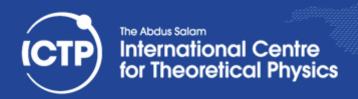


Process Interactions

- The effective speed-up obtained by the parallelization depend by the amount of overhead we introduce making the algorithm parallel
- There are mainly two key sources of overhead:
 - 1. Time spent in inter-process interactions (communication)
 - 2. Time some process may spent being idle (synchronization)

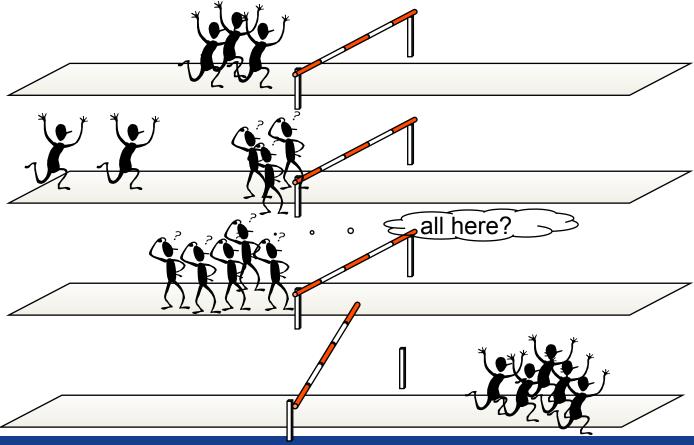








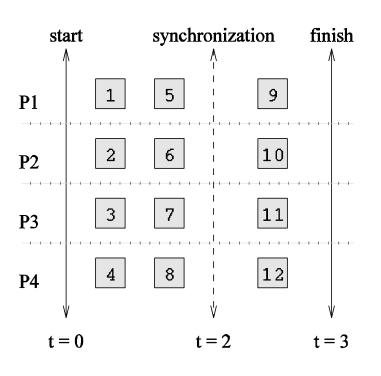
Effect of load-unbalancing

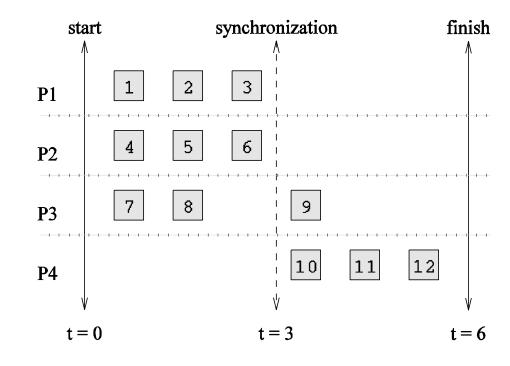






Mapping and Synchronization





(a)

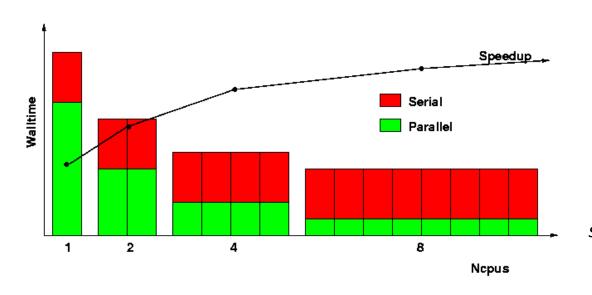
(b)

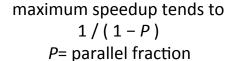




Amdahl's law

In a massively parallel context, an upper limit for the scalability of parallel applications is determined by the fraction of the overall execution time spent in non-scalable operations (Amdahl's law).

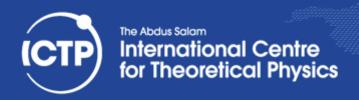




1000000 core

P = 0.999999

serial fraction= 0.000001

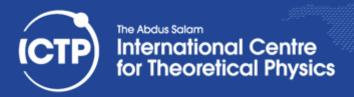




How do we evaluate the improvement?

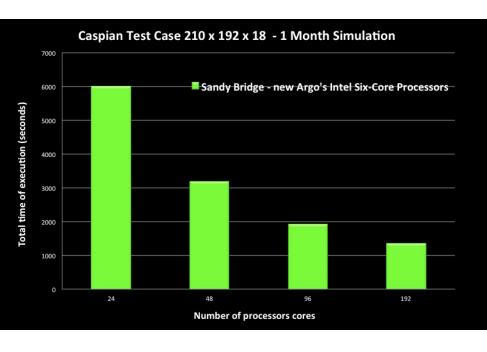
- We want estimate the amount of the introduced overhead => T_o = n_{pes}T_P - T_S
- But to quantify the improvement we use the term Speedup:

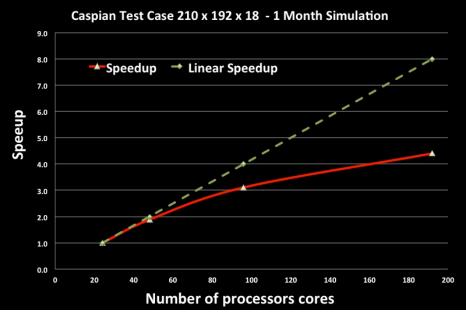
$$S_P = \frac{T_S}{T_P}$$





Speedup









Efficiency

- Only embarrassing parallel algorithm can obtain an ideal Speedup
- The Efficiency is a measure of the fraction of time for which a processing element is usefully employed:

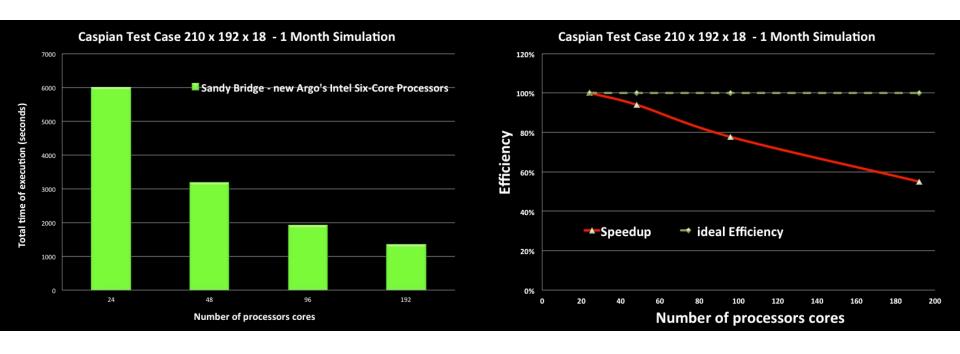
$$E_p = \frac{S_p}{p}$$







Efficiency

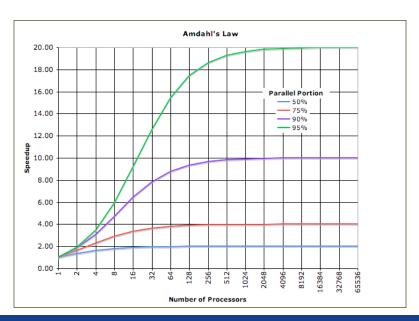


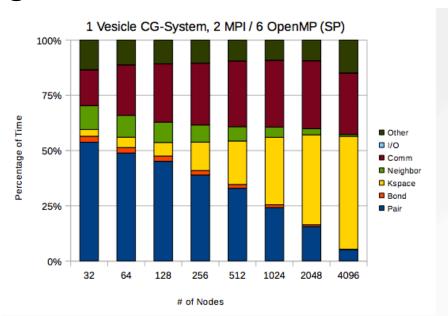




Amdal's Law And Real Life

- The speedup of a parallel program is limited by the sequential fraction of the program
- This assumes perfect scaling and no overhead

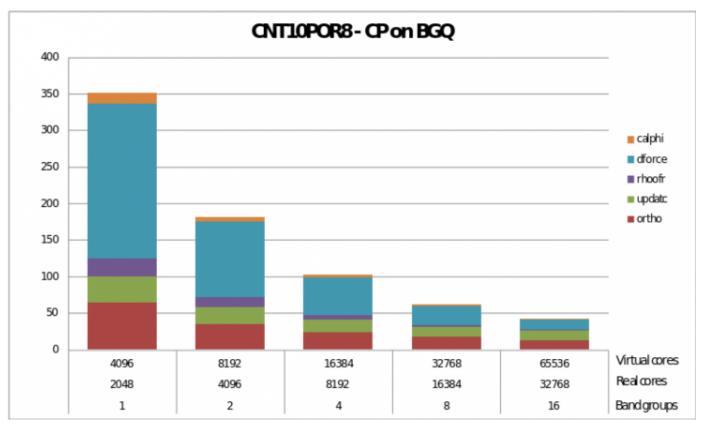


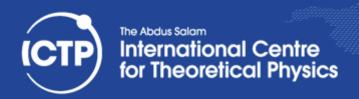






Scaling - QE-CP on Fermi BGQ @ CINECA

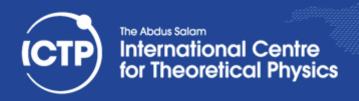






Easy Parallel Computing

- Farming, embarrassingly parallel
 - Executing multiple instances on the same program with different inputs/initial cond.
 - Reading large binary files by splitting the workload among processes
 - Searching elements on large data-sets
 - Other parallel execution of embarrassingly parallel problem (no communication among tasks)
- Ensemble simulations (weather forecast)
- Parameter space (find the best wing shape)





Single Program on Multiple Data

- performing the same program (set of instructions) among different data
- Same model adopted by the MPI library
- A parallel tool is needed to handle the different processes working in parallel
- The MPI library provides the mpirun application to execute parallel instances of the same program

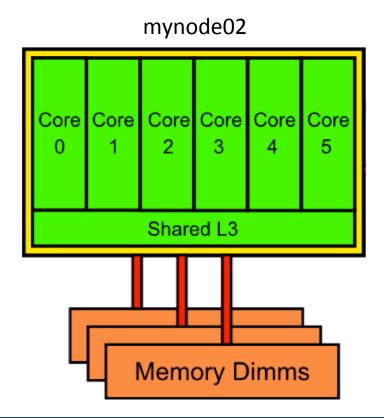






\$ mpirun -np 12 my program.x

mynode01 Core Core Core Core Core Shared L3 **Memory Dimms**









[igirotto@mynode01 ~]\$ mpirun -np 12 /bin/hostname

mynode01

mynode02

mynode01

mynode02

mynode01

mynode02

mynode01

mynode02

mynode01

mynode02

mynode01

mynode02

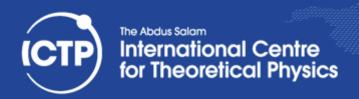






Parallel Operations in Practice

- Parallel reading and computing in parallel is always allowed
- Parallel writing is extremely dangerous!
- To control the parallel flow each process should be unique and identifiable (ID)
- The OpenMPI implementation of the MPI library provides a series of environment variables defined for each MPI process





OMPI_COMM_WORLD_SIZE - the number of processes in this process' MPI Comm World

OMPI_COMM_WORLD_RANK - the MPI rank of this process

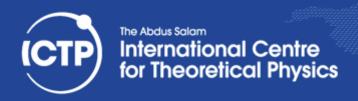
OMPI_COMM_WORLD_LOCAL_RANK - the relative rank of this process on this node within its job. For example, if four processes in a job share a node, they will each be given a local rank ranging from 0 to 3.

OMPI_UNIVERSE_SIZE - the number of process slots allocated to this job. Note that this may be different than the number of processes in the job.

OMPI_COMM_WORLD_LOCAL_SIZE - the number of ranks from this job that are running on this node.

OMPI_COMM_WORLD_NODE_RANK - the relative rank of this process on this node looking across ALL jobs.

http://www.open-mpi.org





In Python

```
import os
myid = os.environ['OMPI_COMM_WORLD_RANK']
[...]
```

In BASH

```
#!/bin/bash
myid=${OMPI_COMM_WORLD_RANK}
[...]
```

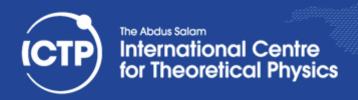
[igirotto@mynode01 ~]\$ mpirun ./myprogram.[py/sh...]





Possible Applications

- Executing multiple instances on the same program with different inputs/initial cond.
- Reading large binary files by splitting the workload among processes
- Searching elements on large data-sets
- Other parallel execution of embarrassingly parallel problem (no communication among tasks)





Conclusions

- Task Farming is a simple model to parallelize simple problems that can be divided in independent task
- The mpirun application aids to easily perform multiple processes, includes environment setting
- Load balancing remains a main problem, but moving from serial to parallel processing can substantially speed-up time of simulation





Task Farming

- Many independent programs (tasks) running at once
 - each task can be serial or parallel
 - "independent" means they don't communicate directly
 - Processes possibly driven by the mpirun framework

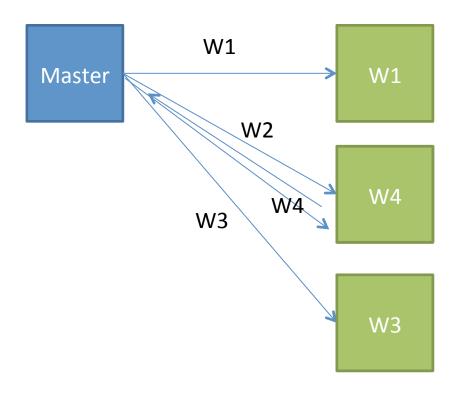
```
[igirotto@localhost]$ more my_shell_wrapper.sh
#!/bin/bash
#example for the OpenMPI implementation
./prog.x --input input_${OMPI_COMM_WORLD_RANK}.dat
[igirotto@localhost]$ mpirun -np 400 ./my_shell_wrapper.sh
```







Master/Slave

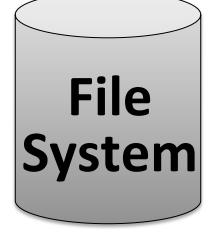


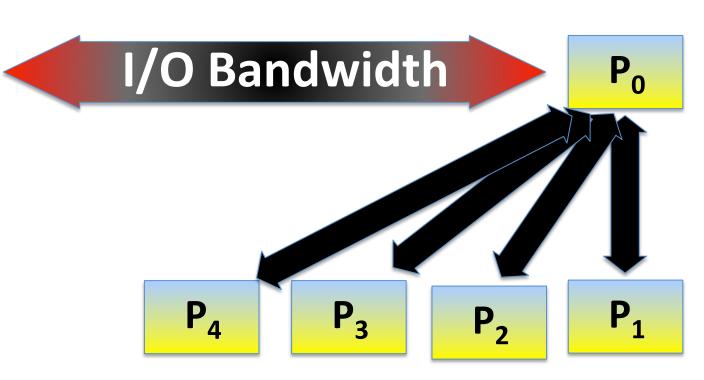


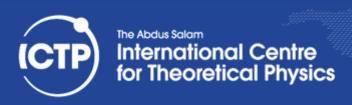




Parallel I/O











Parallel I/O

/O Bandwidth File **System**

Bandwidt File **System**

Bandwidt File **System**

/O Bandwidth File System







Parallel I/O

P₀



 P_1



 P_2



P₃



MPI I/O & Parallel I/O Libraries (Hdf5, Netcdf, etc...)

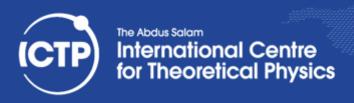
Parallel File System













What If You Want to Learning How to Program All This?!

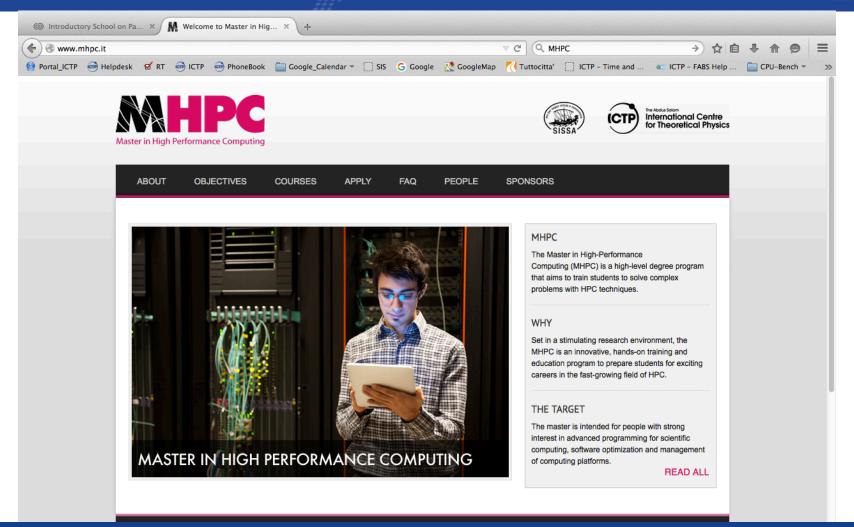
- Introductory School on Parallel Programming and Parallel Architecture for High Performance Computing | (smr 2877)
- 3 October 2016 14 October 2016

What If You Want to Master All This?!















Thanks for your attention!!

