

Introduction to Numerical Libraries

Theory, Methods and Libraries.

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WELCOME!!

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The goal of this course is to show you how to get advantage of some of the most important numerical libraries for improving the performance of your HPC applications. We will focus on:



2DECOMP O

FFTW, a subroutine library for computing the discrete Fourier transform (DFT) in one or more dimensions, of arbitrary input size, and of both real and complex data (as well as of even/odd data, i.e. the discrete cosine/sine transforms or DCT/DST)

The 2DECOMP&FFT library is a software framework in Fortran to build large-scale parallel applications. It is designed for applications using three-dimensional structured mesh and spatially implicit numerical algorithms. At the foundation it implements a general-purpose 2D pencil decomposition for data distribution on distributed-memory platforms.

ScaLAPACK

FFT C

A good number of libraries for Linear Algebra operations, including BLAS, LAPACK, SCALAPACK, MKL and MAGMA



PETSc, a suite of data structures and routines for the scalable (parallel) solution of scientific applications modeled by partial differential equations



Part 0:

Before to start....





This first lecture won't be about numerical libraries...

Its purpose is to teach you the very basics of how to interact with CINECA's HPC cluster, where exercises will take place.

You will learn how to access to our system, how to compile, how to launch batch jobs, and everything you need in order to complete the exercises succesfully



...don't worry, it won't last long!! ;-)

WORK ENVIRONMENT

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Workstation: User \rightarrow corsi Password \rightarrow corsi_2013! Open a terminal: **ssh username@login.galileo.cineca.it**

Once you're logged on a cluster, you are on your **home** space. It is best suited for **programming** environment (compilation, small debugging sessions...) Environment variable: \$HOME

Another space you can access to is your **scratch** space. It is best suited for **production** environment (launch your jobs from there) Environment variable: \$CINECA_SCRATCH WARNING: is active a **cleaning procedure**, that deletes your files older than 30 days!

Use the command "cindata" for a quick briefing about your space occupancy



As an user, you have access to a limited number of CPU hours to spend. They are not assigned to users, but to **projects** and are shared between the users who are working on the same project (i.e. your research partners). Such projects are called **accounts** and are a different concept from your username.

You can check the state of your account with the command "saldo -b", which tells you how many CPU hours you have already consumed for each account you're assigned at (a more detailed report is provided by "saldo -r").

[amarani0@fen08 ~]\$ saldo -b							
account	start	end	total (local h)	localCluster Consumed(local h)	totConsumed (local h)	totConsumed %	
cin staff	20110323	20200323	1000000000	30365762	30527993	3.1	
cin totview	20130123	20130213	50000	0	0	0.0	
train sc32013	20130211	20130411	1250000	87458	87458	7.0	
train_cnl12013	20130311	20130411	100000	0	0	0.0	
19999							

ACCOUNTING

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The account provided for this course is "**train_cnl2016**" (you have to specify it on your job scripts). It expires Monday the 15th and is shared between all the students; there are plenty of hours for everybody, but don't waste them!





CINECA's work environment is organized with modules, a set of installed compilers, libraries, tools and applications available for all users.

"loading" a module means defining all the environment variables that point to the path of what you have loaded.

After a module is loaded, an environment variable is set of the form "MODULENAME_HOME"

[amarani0@fen07 ~]\$ module load namd [amarani0@fen07 ~]\$ ls \$NAMD_HOME backup flipbinpdb flipdcd namd2 namd2_plumed namd2_remd psfgen sortreplicas

MODULE COMMANDS

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> module available (or just "> module av")

Shows the full list of the modules available in the profile you're into, divided by: environment, libraries, compilers, tools, applications

> module load <module_name>

Loads a specific module

> module show <module_name>
Shows the environment variables set by a specific module

> module help <module_name>
Gets all informations about how to use a specific module

> module list Shows the loaded modules

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The Numerical Libraries you will learn about and use during the course are also available via module system

/cineca/prod/modulefiles	/base/libraries
blas/2007bgq-xl1.0(default)	libjpeg/8dbgq-gnu4.4.6
essl/5.1	mass/7.3bgq-xl1.0
fftw/2.1.5bgq-xl1.0	mpi4py/1.3bgq-gnu4.4.6
fftw/3.3.2bgq-xl1.0	netcdf/4.1.3bgq-xl1.0
<pre>fftw/3.3.3bgq-xl1.0(default)</pre>	numpy/1.6.2bgq-gnu4.4.6
gsl/1.15bgq-xl1.0	papi/4.4.0bgq-gnu4.4.6
hdf5/1.8.9_parbgq-xl1.0	petsc/3.3-p2bgq-xl1.0
hdf5/1.8.9_serbgq-xl1.0	<pre>scalapack/2.0.2bgq-xl1.0(default)</pre>
<pre>lapack/3.4.1bgq-xl1.0(default)</pre>	szip/2.1bgq-xl1.0
<pre>libint/2.0bgq-xl1.0(default)</pre>	zlib/1.2.7bgq-gnu4.4.6

Once loaded, they set the environment variable LIBRARYNAME_LIB .

If needed, there is also LIBRARYNAME_INC for the header files.

More on that during the course...



COMPILING ON GALILEO

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In EURORA you can choose between three different compiler families: gnu, intel and pgi

You can take a look at the versions available with "*module av*" and then load the module you want. Defaults are: gnu 4.9.2, intel xe 2015, pgi 16.3 *module load intel #* loads default intel compilers suite *module load intel/pe-xe-2016--binary #*loads specific compilers suite

Compiler's name	GNU	INTEL	PGI	
Fortran	gfortran	ifortran	pgf77	
С	gcc	icc	pgcc	
C++	g++	icpc	pgCC	

Get a list of the compilers flags with the command *man*

PARALLEL COMPILING ON GALILEO

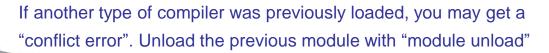
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For parallel programming, two families of compilers are available: openmpi (recommended) and intelmpi .

There are different versions of openmpi, depending on which compiler has been used for creating them. Default is openmpi/1.8.4--gnu--4.9.2 module load autoload openmpi # loads default openmpi compilers suite module load autoload openmpi/1.8.4--intel--cs-xe-2015--binary # loads specific compilers suite

Warning: mpi compiler needs to be loaded after the corresponding basic compiler suite. You can load both compilers at the same time with "autoload"

[cin0955a@node342 ~]\$ module load openmpi WARNING: openmpi/1.4.4--gnu--4.5.2 cannot be loaded due to missing prereq. HINT: the following modules must be loaded first: gnu/4.5.2 [cin0955a@node342 ~]\$ module load autoload openmpi ### auto-loading modules gnu/4.5.2



Once you have loaded the proper library module, specify its linking by adding a reference in the compiling command.

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Two ways to link a library: -L\$LIBRARY_LIB -lname - or - -L\$LIBRARY_LIB/libname.a For some libraries, it may be necessary to include the header path -I\$LIBRARY_INC

\$ mpicc -I\$HDF5_INC input.c -L\$HDF5_LIB -lhdf5 \
 -L\$SZIP_LIB -lsz -LZLIB_LIB -lz
\$ mpicc -I\$HDF5_INC input.c -L\$HDF5_LIB/libhdf5.a \
 -L\$SZIP_LIB/libsz.a -L\$ZLIB_LIB/libz.a



COMPILING WITH LIBRARIES

Galileo lets you choose between static and dynamic linking, with the latter one as a default.

Static linking means that the library references are resolved at compile time, so the necessary functions and variables are already contained in the executable produced. It means a bigger executable but no need for linking the library paths at runtime.

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Dynamic linking means that the library references are resolved at run time, so the executable searches for them in the paths provided. It means a lighter executable and no need to recompile the program after every library update, but need a lot of environment variables to define at runtime.

For enabling static linking: -static (gnu), -intel-static (intel), -Bstatic (pgi)



LAUNCHING JOBS

Now that we have our GALILEO program, it's time to learn how to prepare a job for its execution

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GALILEO uses a scheduler called **PBS**.

The job script scheme is:

- #!/bin/bash
- PBS keywords
- variables environment
- execution line



PBS KEYWORDS

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#PBS -N jobname # name of the job
#PBS -o job.out # output file
#PBS -e job.err # error file
#PBS -l select=1:ncpus=8:mpiprocs=2:mem=8GB # resources requested *
#PBS -l walltime=1:00:00 # max 24h, depending on the queue
#PBS -q parallel # queue desired
#PBS -A <my_account> # name of the account
#PBS -m abe # mail events
#PBS -M <email1,email2,...> # list of emails

 * select = number of chunks (not exactly the nodes) requested ncpus = number of cpus per chunk requested mpiprocs = number of mpi tasks per chunk mem = amount of RAM per chunk
 For pure MPI jobs, ncpus = mpiprocs
 For OpenMP jobs, mpiprocs < ncpus



KEYWORDS SPECIFIC FOR THE COURSE

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#PBS -A train_cnl2016 # your account name #PBS -q R1620674 # special queue reserved for you #PBS -W group_list=train_cnl2016 # needed for entering in private queue

"R1620674" queue is a reserved queue composed by a node equipped with 2 GPUs and a node equipped with 2 MICs.

In order to grant fast runs to all the students, we ask you to not launch too big jobs (you won't need them, anyways). Please don't request more than half a node at a time!

ENVIRONMENT SETUP AND EXECUTION

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The command that "split" the executable on the processes is mpirun: mpirun -n 8 ./myexe arg_1 arg_2 -n is the number of cores you want to use.

In order to use mpirun, openmpi (or intelmpi) has to be loaded. Also, if you linked dynamically, you have to remember to load every library module you need.

The environment setting usually start with "cd \$PBS_O_WORKDIR". That's because by default you are launching on your home space and may not find the executable you want to launch. \$PBS_O_WORKDIR points at the folder you're submitting the job from.

Galileo JOB SCRIPT EXAMPLE

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#!/bin/bash
#PBS -I walltime=0:10:00
#PBS -I select=1:ncpus=4:mpiprocs=4:mem=4GB
#PBS -o job.out
#PBS -e job.err
#PBS -q R1620674
#PBS -A train_cnl2016
#PBS -W group_list=train_cnl2016

cd \$PBS_O_WORKDIR module load autoload openmpi module load somelibrary

mpirun ./myprogram < myinput > myoutput

For GPU accelerators add this to the select line: :ngpus=2

For MIC accelerators add this to the select line: :nmics=2

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PBS COMMANDS

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qsub

qsub <job_script>
Your job will be submitted to the PBS scheduler and executed
when there will be nodes available (according to your priority and the
queue you requested)

qstat

qstat -u \$USER

Shows the list of all your scheduled jobs, along with their status (idle, running, closing,...)

Also, shows you the job id required for other qstat options



PBS COMMANDS

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qstat -f <job_id>
Provides a long list of informations for the job requested.
In particular, if your job isn't running yet, you'll be notified about its
estimated start time or, if you made an error on the job script, you will
learn that the job won't ever start

qdel

qdel <job_id>
Removes the job from the scheduler, killing it



USEFUL DOCUMENTATION

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Exercises on Numerical Librearies:

Slides:

Bologna/HPC_Numerical_Libraries/





Part 1:

Introduction to Numerical Fourier Transforms



Fourier Transforms

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 $H(f) = \int_{-\infty}^{\infty} h(t) e^{2\pi i f t} dt$ $h(t) = \int_{-\infty}^{\infty} H(f) e^{-2\pi i f t} df$ **Frequency Domain Time Domain Real Space Reciprocal Space**



$$H_n = \sum_{k=0}^{N-1} h_k e^{2\pi i kn/N}$$

$$h_{k} = \frac{1}{N} \sum_{n=0}^{N-1} H_{n} e^{-2\pi i k n/N}$$

frequencies from 0 to fc (maximum frequency) are mapped in the values with index from 0 to N/2-1, while negative ones are up to -fc mapped with index values of N / 2 to N

Scale like N*N



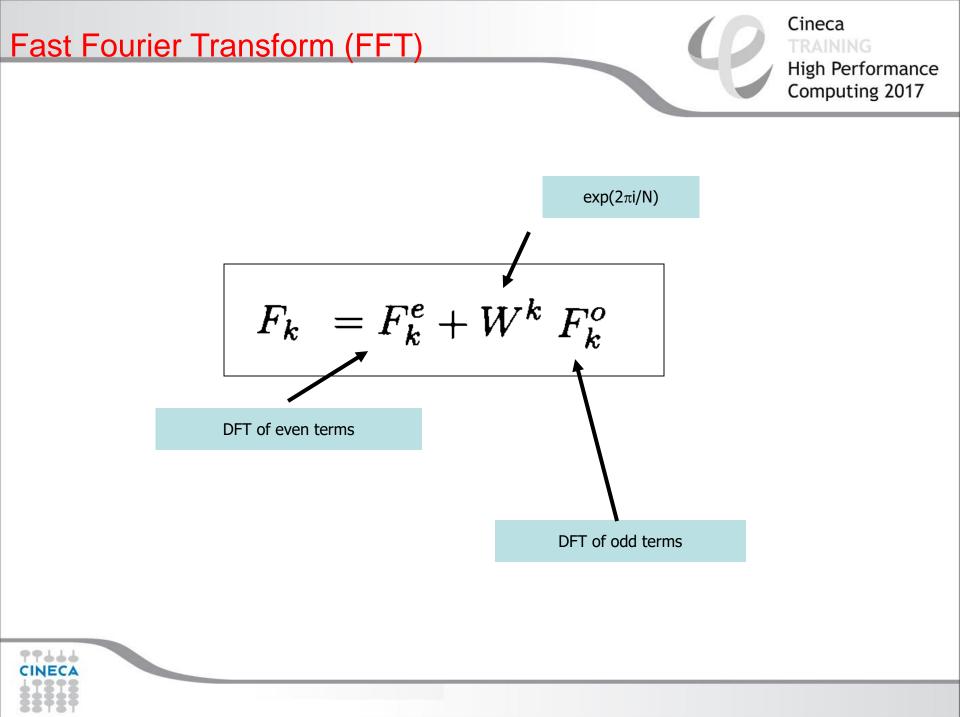
Fast Fourier Transform (FFT)

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The DFT can be calculated very efficiently using the algorithm known as the FFT, which uses symmetry properties of the DFT s

$$\begin{split} F_k &= \sum_{j=0}^{N-1} e^{2\pi i j k/N} f_j \\ &= \sum_{j=0}^{N/2-1} e^{2\pi i k(2j)/N} f_{2j} + \sum_{j=0}^{N/2-1} e^{2\pi i k(2j+1)/N} f_{2j+1} \\ &= \sum_{j=0}^{N/2-1} e^{2\pi i k j/(N/2)} f_{2j} + W^k \sum_{j=0}^{N/2-1} e^{2\pi i k j/(N/2)} f_{2j+1} \\ &= F_k^e + W^k F_k^o \end{split}$$

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Now Iterate:

$$F^{e} = F^{ee} + W^{k/2} F^{eo}$$
$$F^{o} = F^{oe} + W^{k/2} F^{oo}$$

You obtain a series for each value of fn

Scale like N*logN (binary tree)



Parallel Domain Decomposition

How to compute a FFT on a distributed memory system

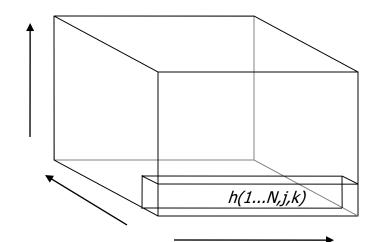


- On a 1D array:
 - Algorithm limits:
 - All the tasks must know the whole initial array
 - No advantages in using distributed memory systems
 - Solutions:
 - Using OpenMP it is possible to increase the performance on shared memory systems
- On a Multi-Dimensional array:
 - It is possible to use distributed memory systems





$H(n_1, n_2) = \text{FFT-on-index-1} \left(\text{FFT-on-index-2} \left[h(k_1, k_2) \right] \right)$ $= \text{FFT-on-index-2} \left(\text{FFT-on-index-1} \left[h(k_1, k_2) \right] \right)$



1) For each value of **j** and **k**

Apply FFT to h(1...N,j,k)

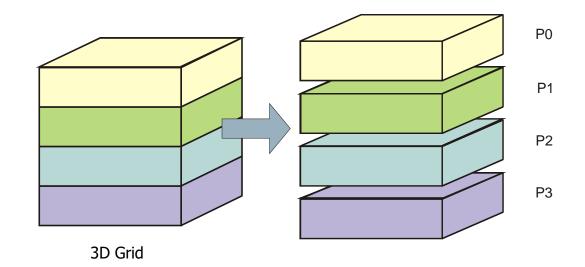
2) For each value of **i** and **k**

Apply FFT to h(i,1...N,k)

3) For each value of **i** and **j** Apply FFT to h(i,j,1...N)

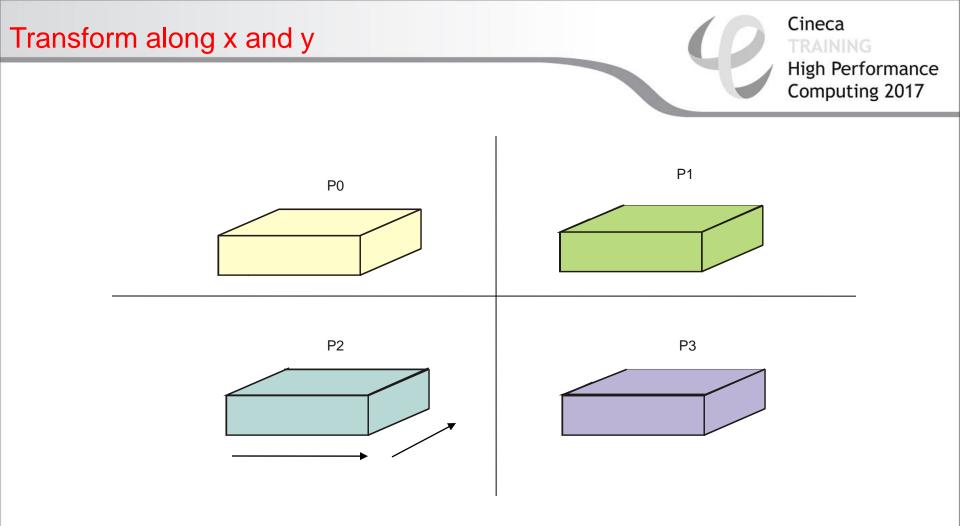
Parallel FFT Data Distribution

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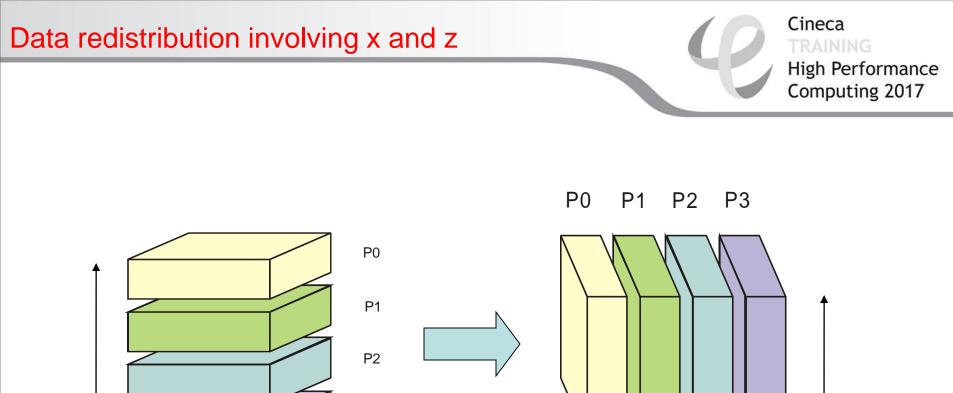
Distribute data along one coordinate (e.g. Z)

This is know as "Slab Decomposition" or 1D Decomposition



each processor trasform its own sub-grid along the x and y independently of the other





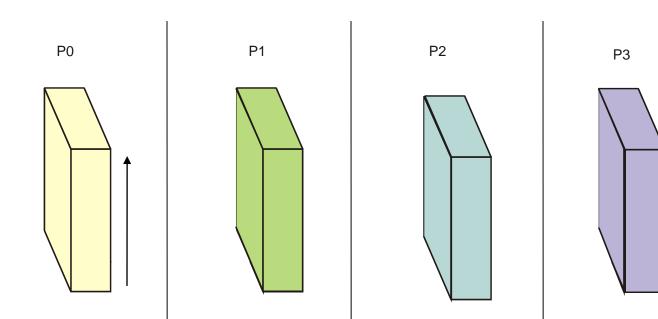
P3

The data are now distributed along x



FFT along z

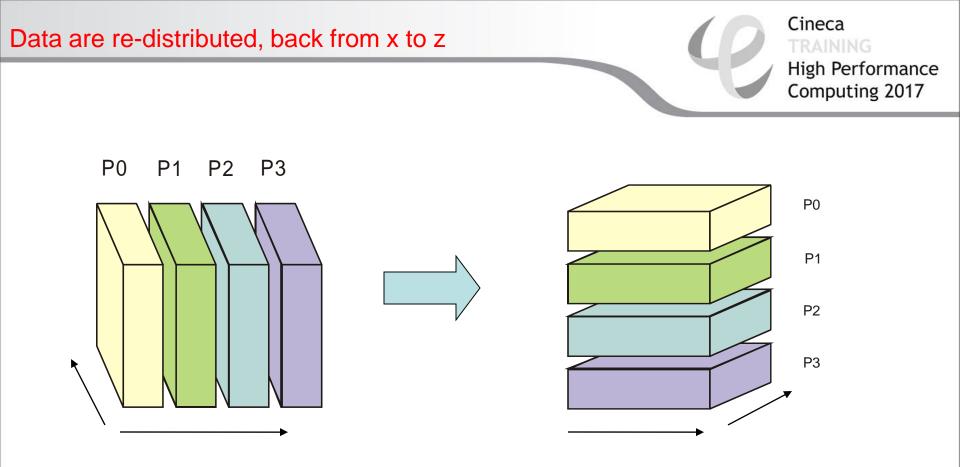
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each processor transform its own sub-grid

along the z dimension independently of the other





The 3D array now has the original layout, but each element

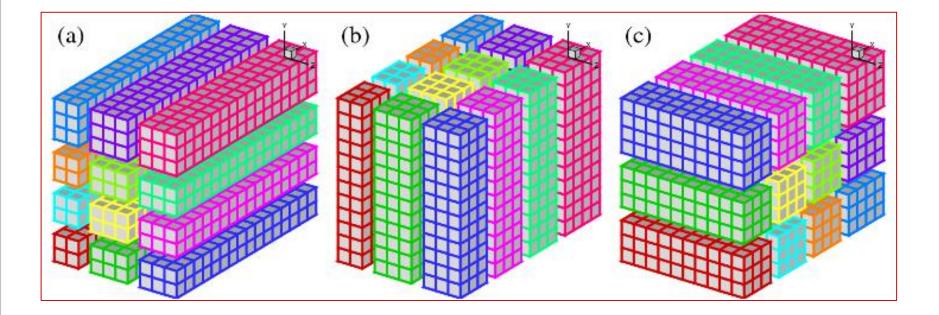
Has been substituted with its FFT.



Pro:

- Simply to implement
- Moderate communications
- Con:
 - Parallelization only along one direction
 - Maximum number of MPI tasks bounded by the size of the larger array index
- Possible Solutions:
 - 2D (Pencil) Decomposition







- Slab (1D) decomposition:
 - Faster on a limited number of cores
 - Parallelization is limited by the length of the largest axis of the 3D data array used
- Pencil (2D) decomposition:
 - Faster on massively parallel supercomputers
 - Slower using large size arrays on a moderate number of cores (more MPI communications)



FFT Numerical Libraries

The simplest way to compute a FFT on a modern HPC system





Introduction

FFTW is a C subroutine library for computing the Discrete Fourier Transform (DFT) in one or more dimensions, of both real and complex data, and of arbitrary input size. We believe that FFTW, which is <u>free software</u>, should become the FFT library of choice for most applications. Our <u>benchmarks</u>, performed on on a variety of platforms, show that FFTW's performance is typically superior to that of other publicly available FFT software. Moreover, FFTW's performance is *portable*: the program will perform well on most architectures without modification.

It is difficult to summarize in a few words all the complexities that arise when testing many programs, and there is no "best" or "fastest" program. However, FFTW appears to be the fastest program most of the time for in-order transforms, especially in the multi-dimensional and real-complex cases (Kasparov is the best chess player in the world even though he loses some games). Hence the name, "FFTW," which stands for the somewhat whimsical title of "Fastest Fourier Transform in the West." Please visit the <u>benchFFT</u> home page for a more extensive survey of the results.

The FFTW package was developed at <u>MIT</u> by <u>Matteo Frigo</u> and <u>Steven G. Johnson</u>.



•Written in C

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Fortran wrapper is also provided

•FFTW adapt itself to your machines, your cache, the size

of your memory, the number of register, etc...

•FFTW doesn't use a fixed algorithm to make DFT

•FFTW chose the best algorithm for your machines

Computation is split in 2 phases:

PLAN creation

Execution

•FFTW support transforms of data with arbitrary length,

rank, multiplicity, and memory layout, and more....



•Many different versions:

•FFTW 2:

•Released in 2003

•Well tested and used in many codes

Includes serial and parallel transforms for both shared

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and distributed memory system

•FFTW 3:

• Released in February 2012

•Includes serial and parallel transforms for both shared

and distributed memory system

•Hybrid implementation MPI-OpenMP

Last version is FFTW 3.3.3



Some Useful Instructions



How can I compile a code that uses FFTW?

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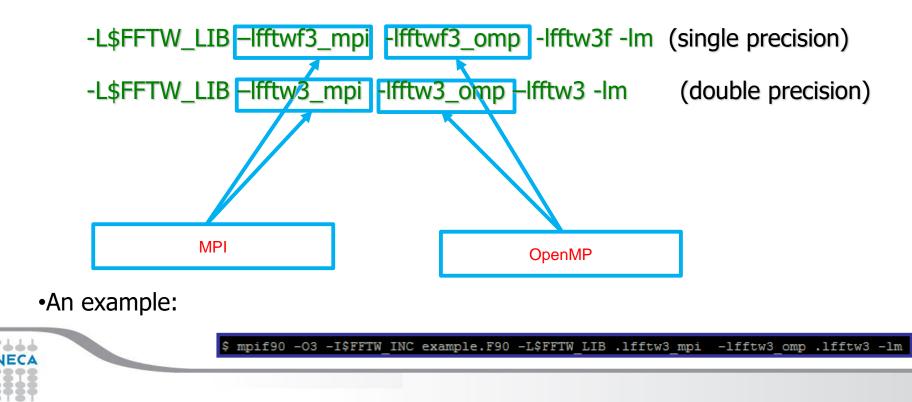
•Module Loading:

module load autoload fftw/3.3.5--intelmpi--2017-binary

Including header:

•-I\$FFTW_INC

•Linking:



Some important Remarks for FORTRAN users Cineca

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• Function in C became function in FORTRAN if they have a return value, and subroutines otherwise.

- All C types are mapped via the iso_c_binning standard.
- FFTW plans are type(C_PTR) in FORTRAN.

•The ordering of FORTRAN array dimensions must be reversed when they are passed to the FFTW plan creation



Initialize FFTW

Including FFTW Lib:

- C:
- Serial:

#include <fftw.h>

• MPI:

#include <fftw-mpi.h>

- FORTRAN:
 - Serial:

include 'fftw3.f03'

•MPI:

include 'fftw3-mpi.f03'

MPI initializzation:

• C:

```
void fftw_mpi_init(void)
```

• FORTRAN:

```
fftw_mpi_init()
```



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Array creation

```
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С:
                                                                                      Computing 2017
• Fixed size array:
      fftx_complex data[n0][n1][n2]
• Dynamic array:
      data = fftw alloc complex(n0*n1*n2)

    MPI dynamic arrays:

      fftw_complex *data
      ptrdiff t alloc local, local no, local no start
      alloc_local = fftw_mpi_local_size_3d(n0, n1, n2, MPI_COMM_WORLD, &local_n0,&local_n0_start)
     data = fftw alloc complex(alloc local)
FORTRAN:

    Fixed size array (simplest way):

     complex(C_DOUBLE_COMPLEX), dimension(n0,n1,n2) :: data
• Dynamic array (simplest way):
     complex(C_DOUBLE_COMPLEX), allocatable, dimension(:, :, :) :: data
     allocate (data(n0, n1, n2))

    Dynamic array (fastest method):

     complex(C_DOUBLE_COMPLEX), pointer :: data(:, :, :) )
     type(C_PTR) :: cdata
     cdata = fftw_alloc_complex(n0*n1*n2)
     call c_f_pointer(cdata, data, [n0,n1,n2])

    MPI dynamic arrays:

     complex(C_DOUBLE_COMPLEX), pointer :: data(:, :, :)
     type(C PTR) :: cdata
     integer(C INTPTR T) :: alloc local, local n2, local n2 offset
     alloc local = fftw mpi local size 3d(n_2, n_1, n_0, MPI COMM WORLD, local n_2, local n_2 offset)
     cdata = fftw_alloc_complex(alloc_local)
     call c_f_pointer(cdata, data, [n0,n1,local_n2])
```

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Plan Creation (C2C)

Cineca **High Performance** 1D Complex to complex DFT: Computing 2017 • C: fftw plan = fftw plan dft 1d(int nx, fftw complex *in, fftw complex *out, fftw direction dir, unsigned flags) •FORTRAN: plan = ftw_plan_dft_1d(nz, in, out, dir, flags) FFTW FORWARD FFTW_ESTIMATE FFTW_BACKWARD 2D Complex to complex DFT: FFTW_MEASURE • C: fftw_plan = fftw_plan_dft_2d(int nx, int ny, fftw_complex *in, fftw_complex *out, fftw_direction dir, unsigned flags) fftw_plan = fftw_mpi_plan_dft_2d(int nx, int ny, fftw_complex *in, fftw_complex *out,MPI_COMM_WORLD, fftw_direction dir, int flags) •FORTRAN: plan = ftw_plan_dft_2d(ny, nx, in, out, dir, flags) plan = ftw_mpi_plan_dft_2d(ny, nx, in, out, MPI_COMM_WORLD, dir, flags)

3D Complex to complex DFT: • C:

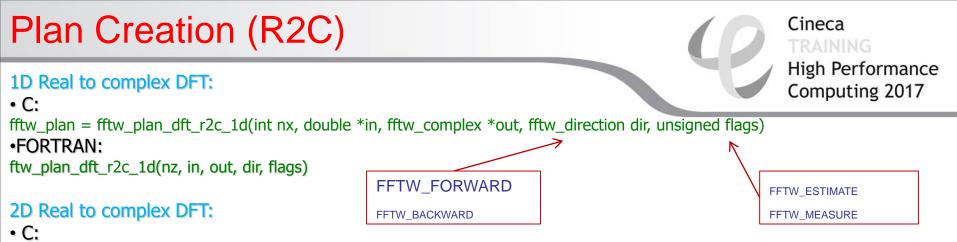
999999

fftw_plan = fftw_plan_dft_3d(int nx, int ny, int nz, fftw_complex *in, fftw_complex *out, fftw_direction dir, unsigned flags)

fftw_plan = fftw_mpi_plan_dft_3d(int nx, int ny, int nz, fftw_complex *in, fftw_complex *out,MPI_COMM_WORLD, fftw_direction dir, int flags)

•FORTRAN: plan = ftw_plan_dft_3d(nz, ny, nx, in, out, dir, flags)

plan = ftw_mpi_plan_dft_3d(nz, ny, nx, in, out, MPI_COMM_WORLD, dir, flags)



fftw_plan = fftw_plan_dft_r2c_2d(int nx, int ny, double *in, fftw_complex *out, fftw_direction dir, unsigned flags)

fftw_plan = fftw_mpi_plan_dft_r2c_2d(int nx, int ny, double *in, fftw_complex *out,MPI_COMM_WORLD, fftw_direction dir, int
flags)
•FORTRAN:

ftw_plan_dft_r2c_2d(ny, nx, in, out, dir, flags)

ftw_mpi_plan_dft_r2c_2d(ny, nx, in, out, MPI_COMM_WORLD, dir, flags)

3D Real to complex DFT:

• C:

999999

fftw_plan = fftw_plan_dft_r2c_3d(int nx, int ny, int nz, fftw_complex *in, fftw_complex *out, fftw_direction dir, unsigned flags)

fftw_plan = fftw_mpi_plan_dft_r2c_3d(int nx, int ny, int nz, fftw_complex *in, fftw_complex *out,MPI_COMM_WORLD,
fftw_direction dir, int flags)
•FORTRAN:
ftw_plan_dft_r2c_3d(nz, ny, nx, in, out, dir, flags)

ftw_mpi_plan_dft_r2c_3d(nz, ny, nx, in, out, MPI_COMM_WORLD, dir, flags)

```
Complex to complex DFT:
```

• C:

void fftw_execute_dft(fftw_plan plan, fftw_complex *in, fftw_complex *out)
void fftw_mpi_execute_dft (fftw_plan plan, fftw_complex *in, fftw_complex *out)

•FORTRAN: fftw_execute_dft (plan, in, out) fftw_mpi_execute_dft (plan, in, out)

Real to complex DFT:

• C:

void fftw_execute_dft (fftw_plan plan, double *in, fftw_complex *out)
void fftw_mpi_execute_dft (fftw_plan plan, double *in, fftw_complex *out)

•FORTRAN: fftw_execute_dft (plan, in, out) Fftw mpi execute dft (plan, in, out)



Destroying PLAN:

```
• C:
```

```
void fftw_destroy_plan(fftw_plan plan)
```

```
• FORTRAN:
```

```
fftw_destroy_plan(plan)
```

```
FFTW MPI cleanup:
• C:
void fftw_mpi_cleanup ()
• FORTRAN:
fftw_mpi_cleanup ()
```

Deallocate data:

```
C:
void fftw_free (fftw_complex data)
FORTRAN:
fftw_free (data)
```





Some Useful Examples



1D Serial FFT - Fortran

program FFTW1D

```
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```

```
use, intrinsic :: iso_c_binding
     implicit none
     include 'fftw3.f03'
     integer(C_INTPTR_T):: L = 1024
     integer(C INT) :: LL
     type(C_PTR) :: plan1
     complex(C_DOUBLE_COMPLEX), dimension(1024) :: idata, odata
     integer :: i
     character(len=41), parameter :: filename='serial_data.txt'
    LL = int(L,C INT)
!! create MPI plan for in-place forward DF
      plan1 = fftw plan dft 1d(LL, idata, odata, FFTW FORWARD, FFTW ESTIMATE)
!! initialize data
      do i = 1, L
        if (i .le. (L/2)) then
           idata(i) = (1.,0.)
        else
           idata(i) = (0.,0.)
        endif
      end do
!! compute transform (as many times as desired)
      call fftw_execute_dft(plan1, idata, odata)
!! deallocate and destroy plans
      call fftw_destroy_plan(plan1)
   end
```

1D Serial FFT - C

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```
# include <stdlib.h>
# include <stdio.h>
# include <math.h>
# include <fftw3.h>
int main (void)
 ptrdiff t i;
 const ptrdiff t n = 1024;
 fftw complex *in;
 fftw complex *out;
 fftw_plan plan_forward;
/* Create arrays. */
  in = fftw_malloc ( sizeof ( fftw_complex ) * n );
  out = fftw_malloc ( sizeof ( fftw_complex ) * n );
/* Initialize data */
  for ( i = 0; i < n; i++ ) {
    if (i <= (n/2-1)) {
      in[i][0] = 1.;
      in[i][1] = 0.;
     ļ
    else {
      in[i][0] = 0.;
      in[i][1] = 0.;
     }
/* Create plans. */
  plan_forward = fftw_plan_dft_1d ( n, in, out, FFTW_FORWARD, FFTW_ESTIMATE );
/* Compute transform (as many times as desired) */
  fftw_execute ( plan_forward );
/* deallocate and destroy plans */
  fftw_destroy_plan ( plan_forward );
  fftw_free ( in );
  fftw free (out);
 return 0;
```

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2D Parallel FFT – Fortran (part1)

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program FFT MPI 3D use, intrinsic :: iso c binding implicit none include 'mpif.h' include 'fftw3-mpi.f03' integer(C INTPTR T), parameter :: L = 1024 integer(C INTPTR T), parameter :: M = 1024 type(C_PTR) :: plan, cdata complex(C_DOUBLE_COMPLEX), pointer :: fdata(:,:) integer(C_INTPTR_T) :: alloc_local, local_M, local_j_offset integer(C_INTPTR_T) :: i, j complex(C DOUBLE COMPLEX) :: fout integer :: ierr, myid, nproc

! Initialize

call mpi_init(ierr) call MPI COMM SIZE(MPI COMM WORLD, nproc, ierr) call MPI COMM RANK(MPI COMM WORLD, myid, ierr) call fftw_mpi_init()

get local data size and allocate (note dimension reversal)

alloc_local = fftw_mpi_local_size_2d(M, L, MPI_COMM_WORLD, local_M, local_j_offset) cdata = fftw_alloc_complex(alloc_local)

call c_f_pointer(cdata, fdata, [L,local_M])

create MPI plan for in-place forward DFT (note dimension reversal)

plan = fftw_mpi_plan_dft_2d(M, L, fdata, fdata, MPI_COMM_WORLD, FFTW_FORWARD, FFTW_MEASURE)

2D Parallel FFT – Fortran (part2)

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2D Parallel FFT – C (part1)

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include <stdlib.h>
include <stdlib.h>
include <stdlib.h>
include <math.h>
include <mpi.h>
include <fftw3-mpi.h>

{

```
int main(int argc, char **argv)
```

```
const ptrdiff_t L = 1024, M = 1024;
fftw_plan plan;
fftw_complex *data ;
ptrdiff_t alloc_local, local_L, local_L_start, i, j, ii;
double xx, yy, rr, r2, t0, t1, t2, t3, tplan, texec;
const double amp = 0.25;
/* Initialize */
MPI_Init(&argc, &argv);
fftw mpi init();
```

/* get local data size and allocate */
alloc_local = fftw_mpi_local_size_2d(L, M, MPI_COMM_WORLD, &local_L, &local_L_start);
data = fftw_alloc_complex(alloc_local);
/* create plan for in-place forward DFT */
plan = fftw_mpi_plan_dft_2d(L, M, data, data, MPI_COMM_WORLD, FFTW_FORWARD, FFTW_ESTIMATE);

2D Parallel FFT – C (part2)

444

```
/* initialize data to some function my_function(x,y) */
/*.....*/
/* compute transforms, in-place, as many times as desired */
    fftw_execute(plan);
/* deallocate and destroy plans */
    fftw_destroy_plan(plan);
    fftw_mpi_cleanup();
    fftw_free ( data );
    MPI_Finalize();
    }
```

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2DECOMP & FFT &

The most important FFT Fortran Library that use 2D (Pencil) Domain Decomposition



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•General-purpose 2D pencil decomposition module to support building large-scale parallel applications on distributed memory systems.

•Highly scalable and efficient distributed Fast Fourier Transform module, supporting three dimensional FFTs (both complex-to-complex and real-to-complex/complex-to-real).

•Halo-cell support allowing explicit message passing between neighbouring blocks.

•Parallel I/O module to support the handling of large data sets.

•Shared-memory optimisation on the communication code for multi-code systems.

Written in Fortran

- •Best performance using Fortran 2003 standard
- No C wrapper is already provided
- •Structure: Plan Creation Execution Plan Destruction
- •Uses FFTW lib (or ESSL) to compute 1D transforms
- •More efficient on massively parallel supercomputers.
- Well tested
- Additional features





Parallel Three-Dimensional Fast Fourier Transforms (P3DFFT)



•General-purpose 2D pencil decomposition module to support building large-scale parallel applications on distributed memory systems.

•Highly scalable and efficient distributed Fast Fourier Transform module, supporting three dimensional FFTs (both complex-to-complex and real-to-complex/complex-to-real).

Sine/cosine/Chebyshev/empty transform

•Shared-memory optimisation on the communication code for multi-code systems.

- •Written in Fortran 90
- •C wrapper is already provided
- •Structure: Plan Creation Execution Plan Destruction
- •Uses FFTW lib (or ESSL) to compute 1D transforms
- •More efficient on massively parallel supercomputers.
- •Well tested but not stable as 2Decomp&FFT

Additional features

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Some useful papers

- Auto-tuning of the FFTW Library for Massively Parallel Supercomputers.
 - M. Guarrasi, G. Erbacci, A. Emerson;
 - > 2012, PRACE white paper;
 - Available at this link;
- Scalability Improvements for DFT Codes due to the Implementation of the 2D Domain Decomposition Algorithm.
 - M. Guarrasi, S. Frigio, A. Emerson, G. Erbacci
 - 2013, PRACE white paper;
 - Available at this link
- Testing and Implementing Some New Algorithms Using the FFTW Library on Massively Parallel Supercomputers.
 - M. Guarrasi, N. Li, S. Frigio, A. Emerson, G. Erbacci;
 - Accepted for ParCo 2013 conference proceedings.
- > 2DECOMP&FFT A highly scalable 2D decomposition library and FFT interface.
 - N. Li, S. Laizet;
 - 2010, Cray User Group 2010 conference;
 - Available at this link
- > P3DFFT: a framework for parallel computations of Fourier transforms in three dimensions.
 - D. Pekurovsky;
 - > 2012, SIAM Journal on Scientific Computing, Vol. 34, No. 4, pp. C192-C209
- The Design and Implementation of FFTW3.
 - M. Frigio, S. G. Johnson;
 - 2005, Proceedings of the IEEE.



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Part 3:

Introduction to PETSc

(Portable, Extensible Toolkit for Scientific Computation)



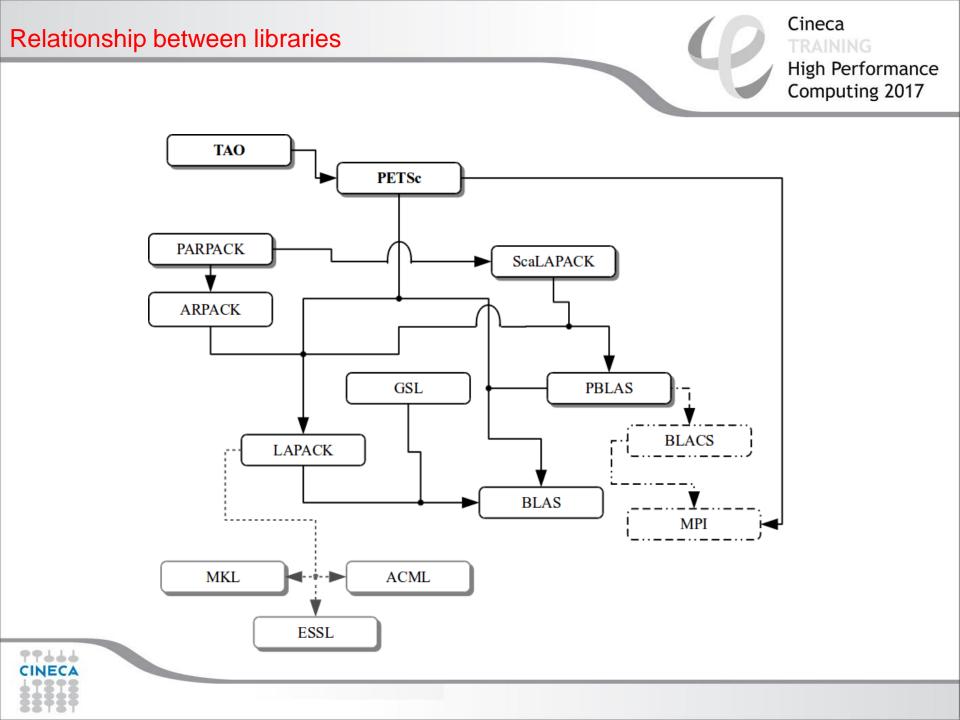
PETSc – Portable, Extensible Toolkit for Scientific Computation

Is a suite of data structures and routines for the scalable (parallel) solution of scientific applications mainly modelled by partial differential equations.

- > **ANL** Argonne National Laboratory
- Begun September 1991
- Uses the MPI standard for all message-passing communication
- C, Fortran, and C++
- Consists of a variety of libraries; each library manipulates a particular family of **objects** and the operations one would like to perform on the objects
- > PETSc has been used for modelling in all of these **areas**:

Acoustics, Aerodynamics, Air Pollution, Arterial Flow, Brain Surgery, Cancer Surgery and Treatment, Cardiology, Combustion, Corrosion, Earth Quakes, Economics, Fission, Fusion, Magnetic Films, Material Science, Medical Imaging, Ocean Dynamics, PageRank, Polymer Injection Molding, Seismology, Semiconductors, ...





PETSc programming model

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Goals

- Portable
- Performance
- Scalable parallelism

Approach

- Variety of libraries
 - Objects (One interface One or more implementations)
 - Operations on the objects

Benefit

- Code reuse
- Flexibility
- Hide within objects the details of the communication



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Nonlinear Solvers			Time Steppers					
Newton-based Methods		Other		Backward	Pseudo-Time			
Line Search	Trust Region	Other	Euler	Euler	Stepping	Other		

Krylov Subspace Methods								
GMRES	CG	CGS	Bi-CG-Stab	TFQMR	Richardson	Chebychev	Other	

Preconditioners						
Additive Schwarz	Block Jacobi	Jacobi	ILU	ICC	LU (sequential only)	Other

Matrices						
Compressed Sparse Row	Block Compressed Sparse Row	Block Diagonal	Dense	Other		
(AIJ)	(BAIJ)	(BDiag)				

	Index Sets					
Vectors	Indices	Block Indices	Stride	Other		

Writing PETSc programs: initialization and finalization

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PetscInitialize(int *argc, char ***args, const char file[], const char help[])

- Setup static data and services
- Setup MPI if it is not already

PetscFinalize()

- Calculates logging summary
- Finalize MPI (if PetscInitialize() began MPI)
- Shutdown and release resources



#include "petsc.h"

```
#undef __FUNCT__
#define __FUNCT_ "main"
int main(int argc,char **args)
{
    PetscErrorCode ierr;
    PetscMPIInt rank;
```

PetscInitialize(&argc, &args,(char *)0, PETSC NULL);

```
ierr = PetscFinalize();
return 0;
```



program main

integer :: ierr, rank
character(len=6) :: num
character(len=30) :: hello

#include "finclude/petsc.h"

call **PetscInitialize**(PETSC NULL CHARACTER, ierr)

```
call MPI_Comm_rank( PETSC_COMM_WORLD, rank, ierr )
write(num,*) rank
hello = 'Hello by process '//num
call PetscPrintf( PETSC_COMM_SELF, hello//achar(10), ierr )
```

```
call PetscFinalize(ierr)
```

end program



Vec and Mat



Vectors

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What are PETSc vectors?

- Fundamental objects for storing field solutions, right-hand sides, etc.
- Each process locally owns a subvector of contiguously numbered global indices

Features

- Has a direct interface to the values
- Supports all vector space operations
 - VecDot(), VecNorm(), VecScale(), ...
- Also unusual ops, e.g. VecSqrt(), VecInverse()
- Automatic communication during assembly
- Customizable communication (scatters)

VecCreate(MPI_Comm comm, Vec *v)

- Vector types: sequential and parallel (MPI based)
- Automatically generates the appropriate vector type (sequential or parallel) over all processes in comm
- VecSetSizes(Vec v, int m, int M)
 - Sets the local and global sizes, and checks to determine compatibility
- VecSetFromOptions (Vec v)
 - Configures the vector from the options database

VecDuplicate(Vec old, Vec *new)

- Does not copy the values



VecGetSize(Vec v, int *size)
VecGetLocalSize(Vec v, int *size)
VecGetOwnershipRange(Vec vec, int *low, int *high)

VecView(Vec x, PetscViewer v)

VecCopy (Vec x, Vec y)

VecDestroy(Vec *x)



Once all of the values have been inserted with VecSetValues(), one must call

```
VecAssemblyBegin(Vec x)
```

```
VecAssemblyEnd(Vec x)
```

to perform any needed message passing of nonlocal components.

A three step process

- Each process tells PETSc what values to set or add to a vector component. Once *all* values provided,
- begin communication between processes to ensure that values end up where needed (allow other operations, such as some computation, to proceed).
- Complete the communication



```
VecGetSize(x, &N); /* Global size */
MPI Comm rank(PETSC COMM WORLD, &rank);
```

```
if (rank == 0) {
  for (i=0; i<N; i++)
     VecSetValues(x, 1, &i, &i, INSERT_VALUES);
}</pre>
```

/* These two routines ensure that the data is
distributed to the other processes */
VecAssemblyBegin(x);
VecAssemblyEnd(x);

Vector – Example 2 (Do it in Parallel!)

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VecGetOwnershipRange(x, &low, &high);

```
for (i=low; i<high; i++)</pre>
```

VecSetValues(x, 1, &i, &i, INSERT_VALUES);

/* These routines must be called in case some other
process contributed a value owned by another process
*/

VecAssemblyBegin(x);

```
VecAssemblyEnd(x);
```



Numerical vector operations

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Function Name	Operation
VecAXPY(Vec y,PetscScalar a,Vec x);	y = y + a * x
VecAYPX(Vec y,PetscScalar a,Vec x);	y = x + a * y
VecWAXPY(Vec w,PetscScalar a,Vec x,Vec y);	w = a * x + y
VecAXPBY(Vec y,PetscScalar a,PetscScalar b,Vec x);	y = a * x + b * y
VecScale(Vec x, PetscScalar a);	x = a * x
VecDot(Vec x, Vec y, PetscScalar *r);	$r = \bar{x}' * y$
VecTDot(Vec x, Vec y, PetscScalar *r);	r = x' * y
VecNorm(Vec x,NormType type, PetscReal *r);	$r = x _{type}$
VecSum(Vec x, PetscScalar *r);	$r = \sum x_i$
VecCopy(Vec x, Vec y);	y = x
VecSwap(Vec x, Vec y);	y = x while $x = y$
VecPointwiseMult(Vec w,Vec x,Vec y);	$w_i = x_i * y_i$
VecPointwiseDivide(Vec w,Vec x,Vec y);	$w_i = x_i/y_i$
<pre>VecMDot(Vec x,int n,Vec y[],PetscScalar *r);</pre>	$r[i] = \bar{x}' * y[i]$
<pre>VecMTDot(Vec x,int n,Vec y[],PetscScalar *r);</pre>	r[i] = x' * y[i]
VecMAXPY(Vec y,int n, PetscScalar *a, Vec x[]);	$y = y + \sum_{i} a_i * x[i]$
VecMax(Vec x, int *idx, PetscReal *r);	$r = \max x_i$
<pre>VecMin(Vec x, int *idx, PetscReal *r);</pre>	$r = \min x_i$
VecAbs(Vec x);	$x_i = x_i $
VecReciprocal(Vec x);	$x_i = 1/x_i$
VecShift(Vec x,PetscScalar s);	$x_i = s + x_i$
VecSet(Vec x,PetscScalar alpha);	$x_i = \alpha$

It is sometimes more efficient to directly access the storage for the local part of a PETSc ${\tt Vec}.$

E.g., for finite difference computations involving elements of the vector

VecGetArray(Vec, double *[])

Access the local storage

VecRestoreArray(Vec, double *[])

– You must return the array to PETSc when you finish

Allows PETSc to handle data structure conversions

- For most common uses, these routines are inexpensive and do *not* involve a copy of the vector.



Vector – Example 3

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```
Vec vec;
Double *avec;
[...]
VecCreate(PETSC COMM WORLD, &vec);
VecSetSizes(vec, PETSC DECIDE, n);
VecSetFromOptions(vec);
[...]
VecGetArray(vec, &avec);
/* compute with avec directly, e.g.: */
PetscPrintf(PETSC COMM WORLD,
            "First element of local array of vec in
             each process is %f\n", avec[0] );
```

VecRestoreArray(vec, &avec);



2_petsc_vec.c

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```
[...]
PetscViewer viewer fd;
Vec va;
[...]
ierr = PetscViewerBinaryOpen(PETSC COMM WORLD, "data/va 200.bin",
                       FILE MODE READ, &viewer fd ); CHKERRQ(ierr);
ierr = VecCreate(PETSC COMM WORLD, &va);
                                                     CHKERRQ(ierr);
ierr = VecLoad(va, viewer fd);
                                                      CHKERRQ(ierr);
ierr = PetscViewerDestroy(&viewer fd);
                                                      CHKERRO(ierr);
CHKMEMO;
VecView(va, PETSC VIEWER STDOUT WORLD);
VecGetSize(va, &size global);
                                                 CHKERRQ (ierr);
VecGetLocalSize(va, &size local);
                                                 CHKERRQ (ierr);
VecGetOwnershipRange(va, &low idx, &high idx); CHKERRQ(ierr);
[...]
VecDestroy(&va);
[...]
```

Matrices

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What are PETSc matrices?

- Fundamental objects for storing linear operators
- Each process locally owns a submatrix of contiguous rows

Features

- Supports many data types
 - AIJ, Block AIJ, Symmetric AIJ, Block Diagonal, etc.
- Supports structures for many packages
 - Spooles, MUMPS, SuperLU, UMFPack, DSCPack
- A matrix is defined by its interface, the operations that you can perform with it, not by its data structure

Creating a matrix

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MatCreate(MPI_Comm comm, Mat *A)

- Matrices types: sequential and parallel (MPI based).
- Automatically generates the appropriate matrix type (sequential or parallel) over all processes in comm.

MatSetSizes(Mat A, int m, int n, int M, int N)

 Sets the local and global sizes, and checks to determine compatibility

MatSetFromOptions (Mat A)

- Configures the matrix from the options database.

MatDuplicate(Mat B, MatDuplicateOption op, Mat *A)

– Duplicates a matrix including the non-zero structure.



MatView(Mat A, PetscViewer v)

MatGetOwnershipRange(Mat A, PetscInt *m, PetscInt* n)
MatGetOwnershipRanges(Mat A, const PetscInt **ranges)

 Each process locally owns a submatrix of contiguously numbered global rows.

MatGetSize(Mat A, PetscInt *m, PetscInt* n)

Once all of the values have been inserted with MatSetValues(), one must call

MatAssemblyBegin (Mat A, MatAssemblyType type) MatAssemblyEnd (Mat A, MatAssemblyType type)

to perform any needed message passing of nonlocal components.



```
Mat A;
int column[3], i;
double value[3];
```

```
[...]
```

MatCreate(PETSC_COMM_WORLD, PETSC_DECIDE, PETSC_DECIDE, n, n, &A); MatSetFromOptions(A);

```
value[0] = -1.0; value[1] = 2.0; value[2] = -1.0;
if (rank == 0) {
    for (i=1; i<n-2; i++) {
        column[0] = i-1; column[1] = i; column[2] = i+1;
        MatSetValues(A,1,&i,3,column,value,INSERT_VALUES);
}}
MatAssemblyBegin(A,MAT_FINAL_ASSEMBLY);
MatAssemblyEnd(A,MAT_FINAL_ASSEMBLY);
```

```
Matrix – Example 2
                                                          Cineca
                                                           High Performance
                                                           Computing 2017
  Mat A;
   int column[3], i, start, end, istart, iend;
  double value[3];
   [...]
  MatCreate(PETSC COMM WORLD, PETSC DECIDE, PETSC DECIDE, n, n, &A);
  MatSetFromOptions (A);
  MatGetOwnershipRange(A, &istart, &iend);
  value[0] = -1.0; value[1] = 2.0; value[2] = -1.0;
   for (i=istart; i<iend; i++) {</pre>
       column[0] = i-1; column[1] = i; column[2] = i+1;
       MatSetValues (A, 1, &i, 3, column, value, INSERT VALUES);
   }
```

MatAssemblyBegin(A,MAT_FINAL_ASSEMBLY);
MatAssemblyEnd(A,MAT_FINAL_ASSEMBLY);

Numerical matrix operations

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Function Name	Operation
MatAXPY(Mat Y, PetscScalar a, Mat X, MatStructure);	Y = Y + a * X
MatMult(Mat A, Vec x, Vec y);	y = A * x
MatMultAdd(Mat A,Vec x, Vec y,Vec z);	z = y + A * x
MatMultTranspose(Mat A, Vec x, Vec y);	$y = A^T * x$
MatMultTransposeAdd(Mat A, Vec x, Vec y, Vec z);	$z = y + A^T * x$
<pre>MatNorm(Mat A,NormType type, double *r);</pre>	$r = A _{type}$
MatDiagonalScale(Mat A, Vec l, Vec r);	$A = \operatorname{diag}(l) * A * \operatorname{diag}(r)$
MatScale(Mat A,PetscScalar a);	A = a * A
<pre>MatConvert(Mat A,MatType type,Mat *B);</pre>	B = A
MatCopy(Mat A,Mat B,MatStructure);	B = A
MatGetDiagonal(Mat A, Vec x);	$x = \operatorname{diag}(A)$
MatTranspose(Mat A,MatReuse,Mat* B);	$B = A^T$
MatZeroEntries(Mat A);	A = 0
MatShift(Mat Y,PetscScalar a);	Y = Y + a * I

Preallocation of memory is critical for achieving **good performance** during matrix assembly, as this reduces the number of allocations and copies required.

PETSc sparse matrices are dynamic data structures. Can **add additional nonzeros freely**.

Dynamically adding many nonzeros

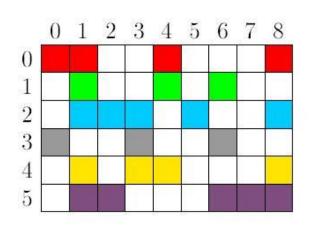
- requires additional memory allocations
- requires copies
- can kill performance

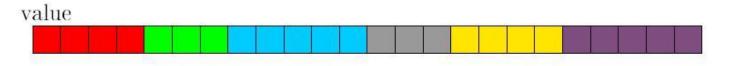
Memory pre-allocation provides the freedom of dynamic data structures plus good performance



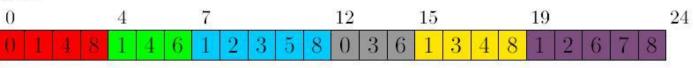
Matrix AIJ format

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row pointer





of sequential sparse matrix (1/2)

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- 1. If $(nz == 0 \&\& nnz == PETSC_NULL)$
- \rightarrow PETSc to control all matrix memory allocation
- 1. Set **nz** = <*value*>
 - \rightarrow Specify the expected number of nonzeros for each row.
 - Fine if the number of nonzeros per row is roughly the same throughout the matrix
 - Quick and easy first step for pre-allocation



3. Set nnz[0] = <nonzeros in row 0>

nnz [m] = <nonzeros in row m>

- indicate (nearly) the exact number of elements intended for the various rows
- If one **underestimates** the actual number of nonzeros in a given row, then during the assembly process PETSc will **automatically allocate additional needed space**.

This extra memory allocation can **slow** the

computation!

Each process locally owns a submatrix of contiguously numbered global rows.

Each submatrix consists of **diagonal** and **off-diagonal** parts.

P0 P1 P2



Pre-allocation

of parallel sparse matrix (1/2)

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MatCreateMPIAIJ(MPI_Comm comm,

int m, int n, int M, int N, int d_nz, int *d_nnz, int o_nz, int *o_nnz, Mat *A)

1. If $(d_nz == o_nz == 0 \&\& d_nnz == o_nnz == PETSC_NULL)$

→ PETSc to control dynamic allocation of matrix memory space

```
1. Set d_nz = <value> and o_nz = <value>
```

→ Specify nonzero information for the diagonal (d_nz) and off-diagonal (o_nz) parts of the matrix.

of parallel sparse matrix (2/2)

Pre-allocation

MatCreateMPIAIJ (MPI Comm comm,

int m, int n, int M, int N, int d_nz, int *d_nnz, int o_nz, int *o_nnz, Mat *A)

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3. Set d_nnz[0] = <nonzeros in row 0, diagonal part>

d_nnz[m] = <nonzeros in row m, diagonal part >
o_nnz[0] = <nonzeros in row 0, off-diagonal part>

o_nnz[m] = <nonzeros in row m , off-diagonal part >

→ Specify nonzero information for the diagonal (d_nnz) and off-diagonal (o_nnz) parts of the matrix.

MatGetInfo(Mat mat, MatInfoType flag, MatInfo *info)
 Or
 Runtime option: -info -mat view info

typedef struct {
 PetscLogDouble block_size;
 PetscLogDouble nz_allocated, nz_used, nz_unneeded;
 PetscLogDouble memory;
 PetscLogDouble assemblies;
 PetscLogDouble mallocs;
 PetscLogDouble fill_ratio_given, fill_ratio_needed;
 PetscLogDouble factor_mallocs;

MatInfo;

[...]

MatInfo info; Mat A; double numMal, nz_a, nz_u;

[...]

MatGetInfo(A, MAT_LOCAL, &info);

```
numMal = info.mallocs;
nz_a = info.nz_allocated;
nz_u = info.nz_used;
```

[...]

```
3_petsc_mat.c
```

```
[...]
PetscViewer viewr fd;
Mat mC;
[...]
ierr = PetscViewerBinaryOpen(PETSC COMM WORLD, "data/mC.bin",
                       FILE_MODE_READ, &viewr fd ); CHKERRQ(ierr);
ierr = MatCreate(PETSC COMM WORLD, &mC);
                                                      CHKERRO(ierr);
ierr = MatSetType(mC, MATAIJ);
                                                      CHKERRQ (ierr);
ierr = MatLoad(mC, viewr fd);
                                                      CHKERRQ(ierr);
ierr = PetscViewerDestroy(&viewr fd);
                                                      CHKERRQ (ierr);
CHKMEMO;
                                                      CHKERRQ (ierr);
```

```
MatGetSize(mC,&row_global,&col_global); CHKERRQ(ierr)
MatGetOwnershipRange(mC,&row_local_min,&row_local_max);
[...]
MatDestroy(&mC);
[...]
```



KSP and SNES



The **object KSP** provides uniform and efficient access to all of the package's **linear system solvers**

KSP is intended for solving nonsingular systems of the form Ax = b.

PETSc KSP methods

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		Options Database	Default Convergence
Method	КЅРТуре	Name	Monitor †
Richardson	KSPRICHARDSON	richardson	true
Chebychev	KSPCHEBYCHEV	chebychev	true
Conjugate Gradient [11]	KSPCG	cg	true
BiConjugate Gradient	KSPBICG	bicg	true
Generalized Minimal Residual [15]	KSPGMRES	gmres	precond
BiCGSTAB [18]	KSPBCGS	bcgs	precond
Conjugate Gradient Squared [17]	KSPCGS	cgs	precond
Transpose-Free Quasi-Minimal Residual (1) [7]	KSPTFQMR	tfqmr	precond
Transpose-Free Quasi-Minimal Residual (2)	KSPTCQMR	tcqmr	precond
Conjugate Residual	KSPCR	cr	precond
Least Squares Method	KSPLSQR	lsqr	precond
Shell for no KSP method	KSPPREONLY	preonly	precond

†true - denotes true residual norm, precond - denotes preconditioned residual norm