

Introduction to Numerical Libraries

Shaohao Chen

High performance computing @ Louisiana State University

Outline

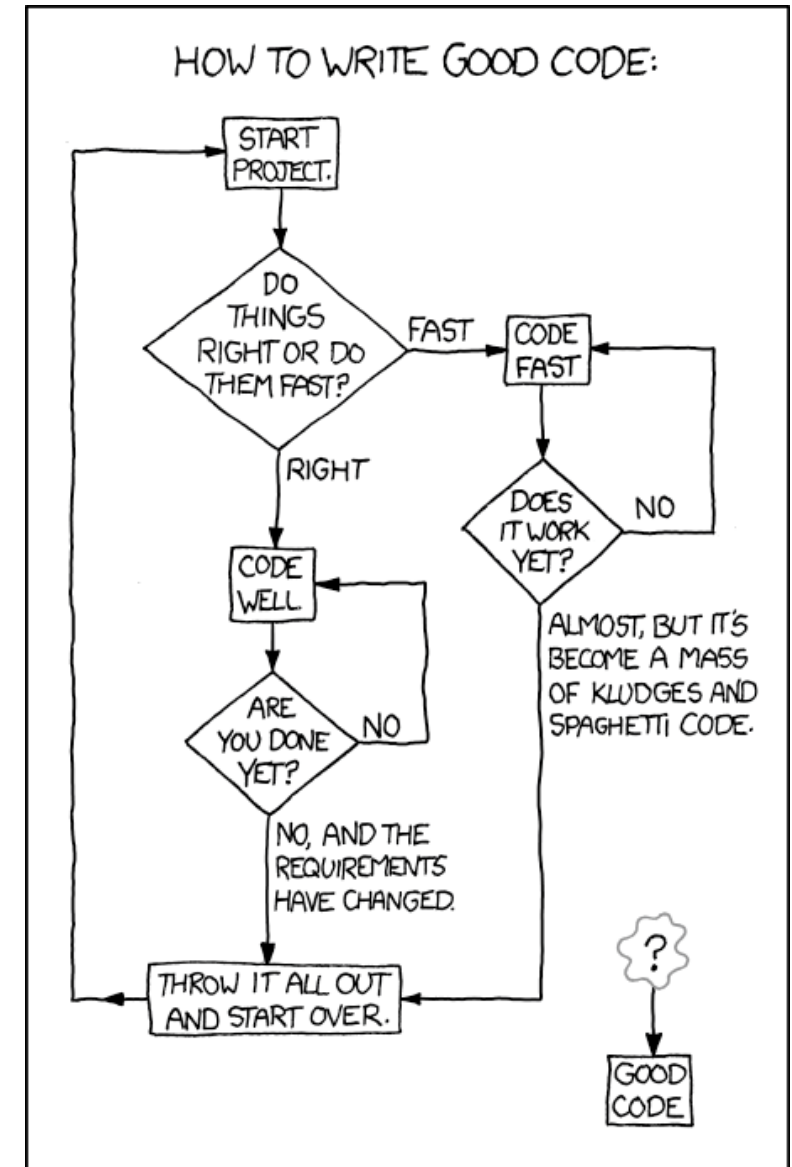
1. Introduction: why numerical libraries?
2. Fast Fourier transform: FFTw
3. Linear algebra libraries: LAPACK
4. Krylov subspace solver: PETSc
5. GNU scientific libraries: GSL

1. Introduction

It is very easy to write bad code!

- Loops : branching, dependencies
- I/O : resource conflicts,
- Memory : long fetch-time
- Portability : code and data portable?
- Readability : Can you still understand it?

A solution: look for existing libraries!



Why numerical libraries?

- Many functions or subroutines you need may have already been coded by others. Just use them.
- Not necessary to code every line by yourself.
- Check available libs before starting to write your program.
- Save your time and efforts!

Advantages of using numerical libraries:

- Computing optimizations
- Portability
- Easy to debug
- Easy to read

What you will learn in this training

- General knowledge of numerical libs.
- How to check available libs on HPC systems and the web.
- How to use numerical libs on HPC systems.
- Basic programming with numerical libs.

List of notable numerical libraries

- **Linear Algebra Package (LAPACK)**: computes matrix and vector operations. [Fortran]
- **Fastest Fourier Transform in the West (FFTW)**: computes Fourier and related transforms. [C/C++]
- **GNU Scientific Library (GSL)**: provides a wide range of mathematical routines. [C/C++]
- **Portable, Extensible Toolkit for Scientific Computation (PETSc)**: a suite of data structures and routines for the scalable (parallel) solution of scientific applications modeled by partial differential equations. [C/C++]
- **NumPy**: adds support for the manipulation of large, multi-dimensional arrays and matrices; also includes a large collection of high-level mathematical functions. [Python]
- For more: http://en.wikipedia.org/wiki/List_of_numerical_libraries

Prerequisite knowledge

Compilers: compile source codes

- **Intel**: icc, icpc, ifort
- **GNU**: gcc, g++, gfortran
- **PGI**: pgcc, pgc++, pgf90

MPI implementations: enable parallel computation with MPI standard

- **mpich** :
- **mvapich2** :
- **openmpi** :
- **impi** :

Installing libs

A typical three-step installation:

- **configure** : configure machine-dependent environments
- **make** : compiles source codes based on settings in the makefile
- **make install** : copy installed libs and binaries to a destination

Other types of installation:

- **manually modify makefile**, then make
- **cmake** : machine-dependent make

✓ Most HPC users do not have to install numerical libs.

Available libs for LSU HPC and LONI users

- **LSU HPC website**

<http://www.hpc.lsu.edu/docs/guides/index.php>

- **module**: for SuperMIC and QB2

module av

module whatis

module list

- **softenv**: for SuperMike, phlip and eric

softenv -a

softenv -k

How to use libs

In your source codes:

- Call functions or subroutines provided by the libs.
- Include head files if necessary.

In a Linux shell:

- Set up environmental variables related to the libs (using module or softenv, or set up manually).
- Link your own codes to the precompiled libs.

Notes

- In most cases, the compilers and the MPI implementations should be the same for your own codes and for the libs.

Static libs

- **libname.a files**: an archive of a bunch of name.o files.

- **Set up environmental valuables**

For bash: `export LIBRARY_PATH=/path/to/lib`

For csh/tcsh: `setenv LIBRARY_PATH /path/to/lib`

Use module: `module load`

Use softenv: add keys to the ~/.soft file, then `resoft`, or `add soft`

- **Compile and link**

`${compiler} -c -I/path/to/include name.c (or name.f)` `# compile your own source code`

`${compiler} name.o -l${libname} -o name` `# link to the lib`

`${compiler} name.o -L/path/to/lib -l${libname} -o name` `# if LIBRARY_PATH is not set`

Dynamic libs

- `libname.so` or `libname.so.*` files
- Set up environmental variables for run-time access
 - For bash: `export LD_LIBRARY_PATH=/path/to/lib`
 - For csh/tcsh: `setenv LD_LIBRARY_PATH /path/to/lib`
 - Use module: `module load`
 - Use softenv: add keys to the `~/.soft` file, then `resoft`, or `add soft`

Notes

- Make sure the executable binary can “see” the dynamic libs.
- If a parallel job runs on multi nodes, `LD_LIBRARY_PATH` has to be set for every node. Set it in the PBS batch script.

2. Fast Fourier Transform: FFTw

Main features:

- C subroutine library for computing the discrete Fourier transform (DFT)
- Both C and Fortran interfaces
- One or more dimensions
- Arbitrary input size
- Both real and complex data
- Even/odd data, i.e. the discrete cosine/sine transforms
- Efficient handling of multiple, strided transforms
- Parallel transforms: parallelized code for platforms with SMP machines with some flavor of threads (e.g. POSIX) or OpenMP. An MPI version for distributed-memory transforms is also available in FFTW 3.3.

FFTw basics

Data type

- `fftw_complex`
- `fftw_plan`

Allocate and deallocate data

- `fftw_malloc`
- `fftw_free`

FFT plan and execution

- FFT plan functions (see next pages)
- `fftw_execute` // execute FFT plan
- `fftw_destroy_plan`



FFTw plan functions I

❑ Complex DFT:
$$X_k \stackrel{\text{def}}{=} \sum_{n=0}^{N-1} x_n \cdot e^{-i2\pi kn/N}, \quad k \in \mathbb{Z}$$

❑ Inverse Complex DFT:
$$x_n = \frac{1}{N} \sum_{k=0}^{N-1} X_k \cdot e^{i2\pi kn/N}, \quad n \in \mathbb{Z}$$

One dimensional

- `fftw_plan_dft_1d`(int n, fftw_complex *in, fftw_complex *out, int sign, unsigned flags);
 sign: either `FFTW_FORWARD` (-1) or `FFTW_BACKWARD` (+1).
 flags: either `FFTW_MEASURE` or `FFTW_ESTIMATE`

Multi dimensional

- `fftw_plan_dft_2d` // two dimensions
- `fftw_plan_dft_3d` // three dimensions
- `fftw_plan_dft` // arbitrary dimensions

FFTw plan functions II

Real DFTs

- `fftw_plan_r2r_1d(int n, double *in, double *out, fftw_r2r_kind kind, unsigned flags)`
kind: `FFTW_REDFT00`, `FFTW_RODFT00`, etc. For different types of even or odd transforms.
- `fftw_plan_r2r_2d`, `fftw_plan_r2r_3d`, `fftw_plan_r2r`

Real input, complex output, always `FFTW_FORWARD`

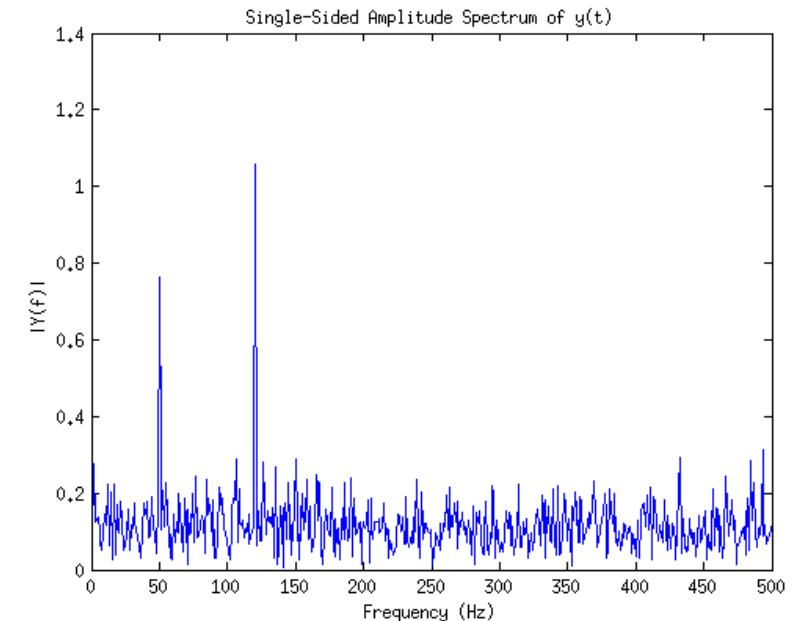
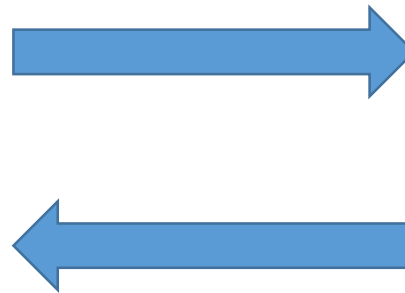
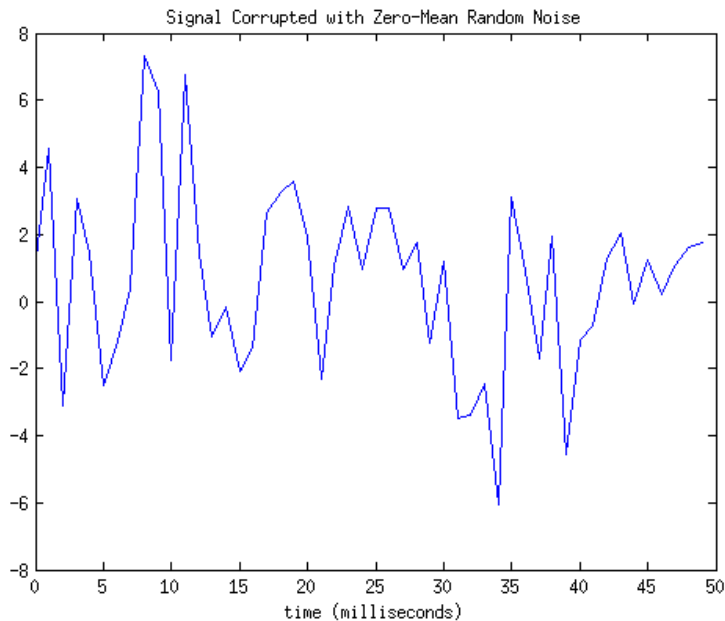
- `fftw_plan_dft_r2c_1d`, `fftw_plan_dft_r2c_2d`
- `fftw_plan_dft_r2c_3d`, `fftw_plan_dft_r2c`

Complex input, real output, always `FFTW_BACKWARD`

- `fftw_plan_dft_c2r_1d`, `fftw_plan_dft_c2r_2d`
- `fftw_plan_dft_c2r_3d`, `fftw_plan_dft_c2r`

Exercise 1: Fourier transform with FFTw

- **Task:** Compute the Fourier transform of a one-dimensional complex array, and compute the inverse Fourier transform of the output, which should be the same as the original input data.



Solution for Exercise 1

❑ Source code at `/home/shaohao/hpc_train/numlibs/fftw/fftw3_prb.c`

- Include fftw head file: `# include <fftw3.h>`
- Call fftw functions: `fftw_plan_dft_1d`, `fftw_execute`, etc.

❑ Compile and run

<code>module load fftw/3.3.3/INTEL-140-MVAPICH2-2.0</code>	<code># load fftw by moudle</code>
<code>module show fftw/3.3.3/INTEL-140-MVAPICH2-2.0</code>	<code># show fftw-related environments</code>
<code>icc -c -I\${LHPC_ROOTFFTW}/include fftw3_prb.c</code>	<code># compile</code>
<code>icc fftw3_prb.o -lfftw3 -o fftw3_prb</code>	<code># link</code>
<code>./fftw3_prb</code>	<code># run</code>

3. Linear Algebra libs

History:

- **LINPACK (LINear algebra PACKage)**: since 1974
based on level-1 BLAS
- **LAPACK (Linear Algebra PACKage)**: since 1989
based on level-3 BLAS, vectorized and threaded in Intel MKL
- **ScaLAPACK (Scalable LAPACK)**: since 1995
parallel with MPI, for distributed memory, only a subset of LAPACK functions
- **DPLASMA (Distributed Parallel Linear Algebra Software for Multicore Architectures)**: 2000's
parallel for shared memory
- **MAGMA (Matrix Algebra for GPUs and Multicore Architectures)**: 2000's
parallel for GPU
- **Matlab**: a commercial software developed from LINPACK.

$$\begin{bmatrix} L & A & P & A & C & K \\ L & -A & P & -A & C & -K \\ L & A & P & A & -C & -K \\ L & -A & P & -A & -C & K \\ L & A & -P & -A & C & K \\ L & -A & -P & A & C & -K \end{bmatrix}$$

LAPACK

- provides routines for solving systems of linear equations, linear least squares, eigenvalue problems, and singular value decomposition.
- also includes routines to implement the associated matrix factorizations such as LU, QR, Cholesky and Schur decomposition.
- was originally written in FORTRAN 77, but moved to Fortran 90 in version 3.2 (2008).
- can be seen as the successor to the linear equations and linear least-squares routines of LINPACK and the eigenvalue routines of EISPACK.

❑ Usage of LAPACK

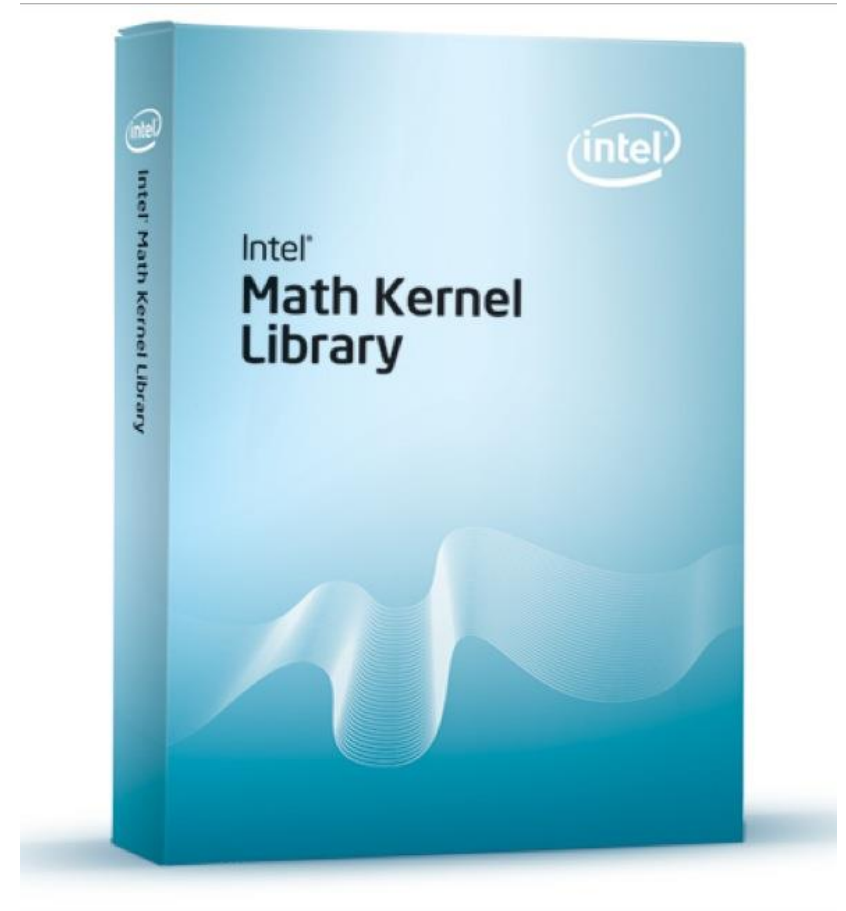
module load lapack/3.4.2/INTEL-140-MVAPICH2-2.0

icc -llapack name.c -o name

ifort -llapack name.f -o name

Intel Math Kernel Library (MKL)

- Optimization for intel processors.
- accelerates math processing routines that increase application performance and reduce development time.
- includes highly vectorized and threaded Lapack, FFT, Vector Math and Statistics functions.
- Xeon-phi enabled.



MKL LAPACK subroutines I

Routine	Description
?geev	Computes the eigenvalues and, optionally, the left and/or right eigenvectors of a general matrix.
?gels	Uses <i>QR</i> or <i>LQ</i> factorization to solve an overdetermined or underdetermined linear system with a full rank matrix.
?gelsd	Computes the minimum norm solution to a linear least squares problem using the singular value decomposition of <i>A</i> and a divide and conquer method.
?gesdd	Computes the singular value decomposition of a general rectangular matrix using a divide and conquer algorithm.
?gesv	Computes the solution to the system of linear equations with a square matrix <i>A</i> and multiple right-hand sides.
?gesvd	Computes the singular value decomposition of a general rectangular matrix.
?heev	Computes all the eigenvalues and, optionally, the eigenvectors of a Hermitian matrix.
?heevd	Computes all the eigenvalues and, optionally, all the eigenvectors of a complex Hermitian matrix using a divide and conquer algorithm.

? could be: s – single precision; d – double precision; c – single-precision complex; z – double-precision complex.

MKL LAPACK subroutines II

<u>?heevr</u>	Computes the selected eigenvalues and, optionally, the eigenvectors of a Hermitian matrix using the Relatively Robust Representations.
<u>?heevx</u>	Computes the selected eigenvalues and, optionally, the eigenvectors of a Hermitian matrix.
<u>?hesv</u>	Computes the solution to the system of linear equations with a Hermitian matrix A and multiple right-hand sides.
<u>?posv</u>	Computes the solution to the system of linear equations with a symmetric or Hermitian positive definite matrix A and multiple right-hand sides.
<u>?syev</u>	Computes all the eigenvalues and, optionally, the eigenvectors of a real symmetric matrix.
<u>?syevd</u>	Computes all the eigenvalues and, optionally, all the eigenvectors of a real symmetric matrix using a divide and conquer algorithm.
<u>?syevr</u>	Computes the selected eigenvalues and, optionally, the eigenvectors of a real symmetric matrix using the Relatively Robust Representations.
<u>?syevx</u>	Computes the selected eigenvalues and, optionally, the eigenvectors of a symmetric matrix.
<u>?sysv</u>	Computes the solution to the system of linear equations with a real or complex symmetric matrix A and multiple right-hand sides.

Exercise 2: Solve a linear system with LAPACK subroutines in MKL

❑ **Task:** Compute the solution to the system of linear equations $AX=B$ with a square matrix A and multiple right-hand sides B.

❑ **Solution:** source code at `/home/shaohao/hpc_train/numlibs/mkl/dgesv_ex.c`

- Call the MKL LAPACK function: `dgesv`

❑ **Compile and run**

```
icc -mkl dgesv_ex.c -o dgesv_ex  
./dgesv
```

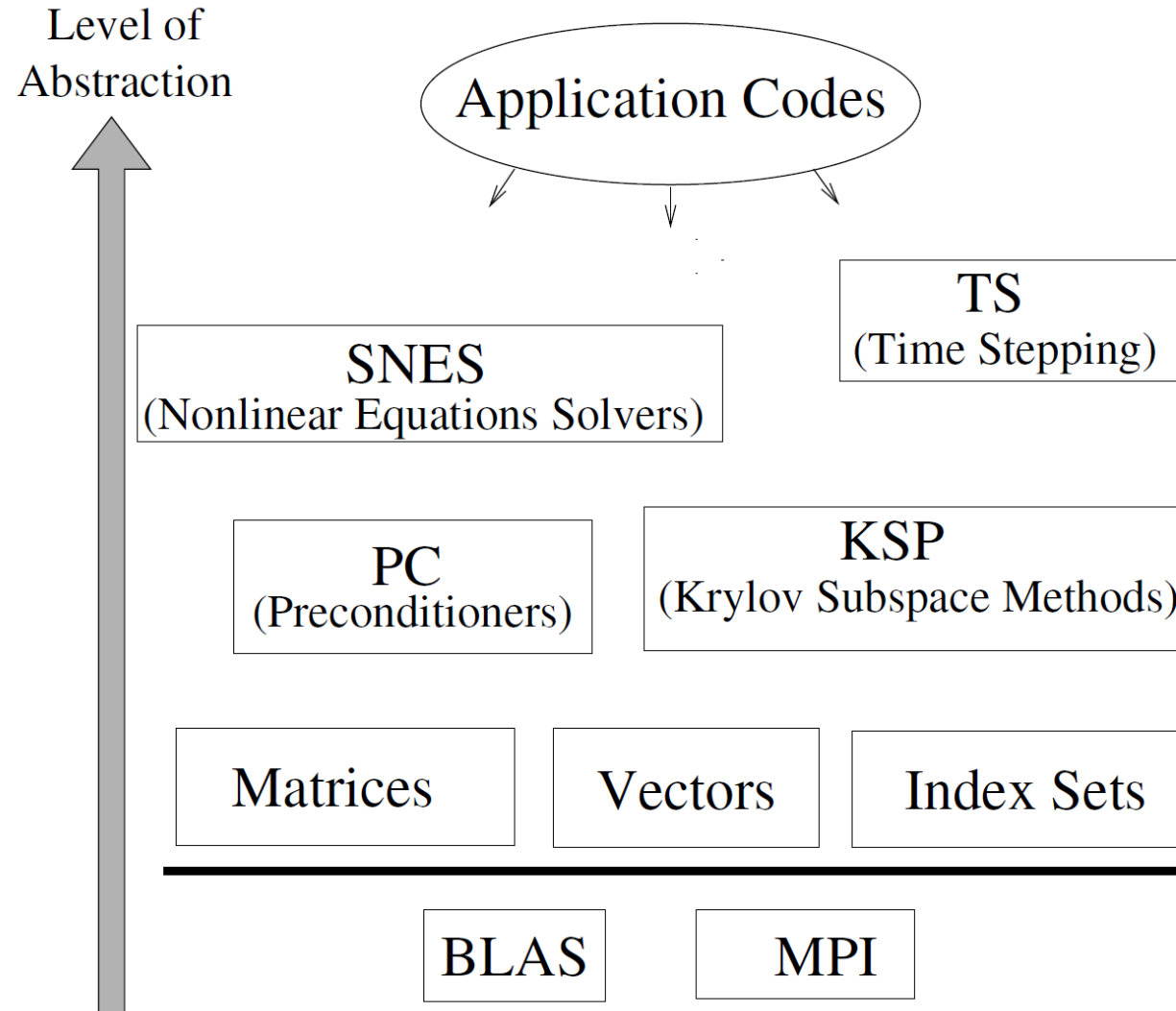

4. PETSc

- ❑ PETSc, pronounced PET-see (the S is silent), is a suite of data structures and routines for the **scalable (parallel)** solution of scientific applications modeled by partial differential equations.
- ❑ It supports MPI, shared memory pthreads, and GPUs through CUDA or OpenCL, as well as hybrid MPI-shared memory pthreads or MPI-GPU parallelism.
- ❑ **Efficient for sparse-matrix problems**

Main contents:

- **distributed data structure**: vectors, matrices, index sets
- **vector operations, matrix operations**
- **Krylov subspace solver (KSP)**: solve linear systems, Laplacian solver, etc.
- **SNES**: Nonlinear Solvers
- **TS**: Scalable ODE and DAE Solvers

Organization of the PETSc Libraries



Parallel Numerical Components of PETSc

Nonlinear Solvers				Time Steppers			
Newton-based Methods		Other	Euler	Backward Euler	Pseudo Time Stepping	Other	
Line Search	Trust Region						
Krylov Subspace Methods							
GMRES	CG	CGS	Bi-CG-STAB	TFQMR	Richardson	Chebyshev	Other
Preconditioners							
Additive Schwartz	Block Jacobi	Jacobi	ILU	ICC	LU (Sequential only)	Others	
Matrices							
Compressed Sparse Row (AIJ)	Blocked Compressed Sparse Row (BAIJ)		Block Diagonal (BDIAG)	Dense	Other		
Vectors		Index Sets					
		Indices	Block Indices	Stride	Other		

PETSc Basics I

- `PetscInitialize` `// call MPI_Initialized`
- `PetscFinalize` `// call MPI_Finalize`
- Data types:
`PetscInt, PetscScalar, Vec, Mat`
- Create objects:
`VecCreate(MPI_Comm comm, Vec *vec)`
`MatCreate(MPI_Comm comm, Mat *mat)`
- Destroy objects
`VecDestroy(Vec *vec)`
`MatDestroy(Mat *mat)`

PETSc Basics II

- Set sizes of objects

VecSetSizes(Vec v, PetscInt n, PetscInt N) // local size n, global size N

MatSetSizes(Mat A, PetscInt m, PetscInt n, PetscInt M, PetscInt N) // local size m, n, global size M, N

- Set values of objects

VecSetValues(Vec x, PetscInt ni, const PetscInt ix[], const PetscScalar y[], InsertMode mode)

MatSetValues(Mat mat, PetscInt m, const PetscInt idxm[], PetscInt n, const PetscInt idxn[], const PetscScalar v[], InsertMode mode) // Set values of a block. Unset blocks are filled with zero.

mode: either **INSERT_VALUES** or **ADD_VALUES**

PETSc Basics III

- Assembly

`VecAssemblyBegin(Vec vec)`

`VecAssemblyEnd(Vec vec)`

`MatAssemblyBegin(Mat mat, MatAssemblyType type)`

`MatAssemblyEnd(Mat mat, MatAssemblyType type)`

type: either `MAT_FLUSH_ASSEMBLY` or `MAT_FINAL_ASSEMBLY`

Vector and matrix are ready to use only after the assembly functions have been called.

- Vector operations (see next slides)

- Matrix operations (see next slides)

- PETSc documentation: <http://www.mcs.anl.gov/petsc/documentation/index.html>

PETSc vector operations

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

PETSc matrix operations

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

PETSc Krylov subspace solver

- **KSP**: Krylov subspace solver
- **PC**: preconditioner

Basic KSP functions:

- **KSPCreate**(MPI_Comm comm, KSP *ksp)
- **KSPSetOperators**(KSP ksp, Mat Amat, Mat Pmat) // assign the linear system to a KSP solver
- **KSPSetType**(KSP ksp, KSPType type) // KSP type: see next slides
- **KSPGetPC**(KSP ksp, PC *pc)
- **PCSetType**(PC pc, PCType type) // PC type: see next slides
- **KSPSetTolerances**(KSP ksp, PetscReal rtol, PetscReal abstol, PetscReal dtol, PetscInt maxits)
- **KSPDestroy**(KSP *ksp)

PETSc KSP types

Method	KSPType	Options Database Name
Richardson	KSPRICHARDSON	richardson
Chebyshev	KSPCHEBYSHEV	chebyshev
Conjugate Gradient [12]	KSPCG	cg
BiConjugate Gradient	KSPBICG	bicg
Generalized Minimal Residual [16]	KSPGMRES	gmres
Flexible Generalized Minimal Residual	KSPFGMRES	fgmres
Deflated Generalized Minimal Residual	KSPDGMRES	dgmres
Generalized Conjugate Residual	KSPGCR	gcr
BiCGSTAB [19]	KSPBCGS	bcgs
Conjugate Gradient Squared [18]	KSPCGS	cgs
Transpose-Free Quasi-Minimal Residual (1) [8]	KSPTFQMR	tfqmr
Transpose-Free Quasi-Minimal Residual (2)	KSPTCQMR	tcqmr
Conjugate Residual	KSPCR	cr
Least Squares Method	KSPLSQR	lsqr
Shell for no KSP method	KSPPREONLY	preonly

PETSc PC types

Method	PCType	Options Database Name
Jacobi	PCJACOBI	jacobi
Block Jacobi	PCBJACOBI	bjacobi
SOR (and SSOR)	PCSOR	sor
SOR with Eisenstat trick	PCEISENSTAT	eisenstat
Incomplete Cholesky	PCICC	icc
Incomplete LU	PCILU	ilu
Additive Schwarz	PCASM	asm
Algebraic Multigrid	PCGAMG	gamg
Linear solver	PCKSP	ksp
Combination of preconditioners	PCCOMPOSITE	composite
LU	PCLU	lu
Cholesky	PCCHOLESKY	cholesky
No preconditioning	PCNONE	none
Shell for user-defined PC	PCSHELL	shell

Exercise 3: Solves a linear system in parallel with PETSc

❑ **Task:** Compute the solution of a sparse-matrix linear system $Ax=b$, using a KSP solver (e.g. MINRES).

❑ **Solution:** C source code at `/home/shaohao/hpc_train/numlibs/petsc/ex42.c`

- Include petsc head file: `#include <petscksp.h>`
- Call petsc functions: `KSPSetOperators`, `KSPSolve`, `KSPSetType`, etc.

❑ **Compile and run**

- `module load mvapich2/2.0/INTEL-14.0.2` `# set up MPI`
- `module load petsc/3.5.0-real/INTEL-140-MVAPICH2-2.0` `# set up PETSc`
- `make ex42` `# compile and link`
- `mpirun -n 20 ./ex42 -m 120` `# run job on a single node`
- `mpirun_rsh -n $NPROC -hostfile $PBS_NODEFILE ./ex42 -m 120` `# run job on multi nodes`

PETSc-dependent packages

- SLEPc:

Scalable Library for Eigenvalue Problems

- MOOSE:

Multiphysics Object-Oriented Simulation Environment finite element framework,
built on top of libMesh and PETSc

For more:

<http://www.mcs.anl.gov/petsc/index.html>

<http://www.mcs.anl.gov/petsc/publications/index.html>

5. GNU Scientific Libs: GSL

Main features:

- A numerical library for C and C++ programmers
- Provides a wide range of mathematical routines such as random number generators, special functions and least-squares fitting
- Uses an object-oriented design. Different algorithms can be plugged-in easily or changed at run-time without recompiling the program.
- It is intended for ordinary scientific users. Anyone who knows some C programming will be able to start using the library straight-away.
- Serial

Complete GSL subjects

- **Mathematical Functions**

- Complex Numbers
- Polynomials
- Special Functions
- Vectors and Matrices
- Permutations
- Combinations
- Multisets
- Sorting
- BLAS Support
- **Linear Algebra**
- Eigensystems

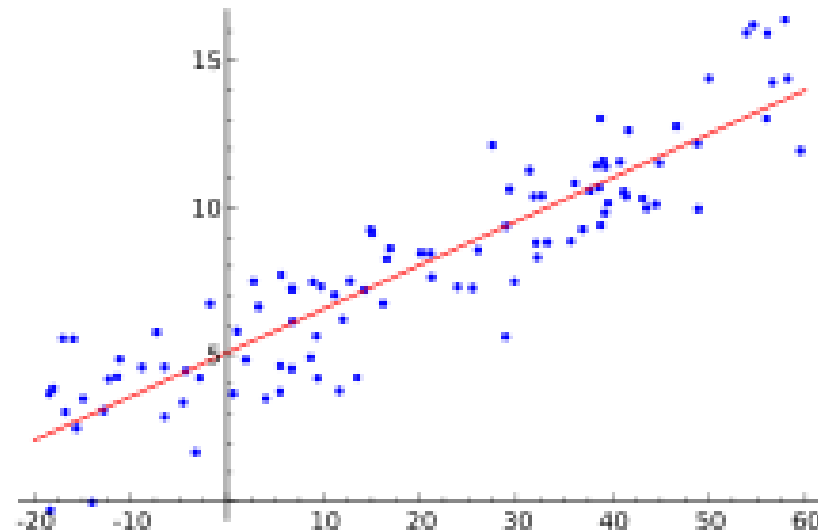
- **Fast Fourier Transforms**

- Numerical Integration
- Random Number Generation
- Quasi-Random Sequences
- Random Number Distributions
- Statistics
- Histograms
- N-tuples
- Monte Carlo Integration
- Simulated Annealing
- Ordinary Differential Equations
- Interpolation
- Numerical Differentiation

- Chebyshev Approximations
- Series Acceleration
- Wavelet Transforms
- Discrete Hankel Transforms
- One dimensional Root-Finding
- One dimensional Minimization
- Multidimensional Root-Finding
- Multidimensional Minimization
- **Least-Squares Fitting**
- Nonlinear Least-Squares Fitting
- Basis Splines
- Physical Constants

Exercise 4: Computes linear fit with GSL

- ❑ **Task:** computes a least squares straight-line fit to a simple dataset, and outputs the best-fit line and its associated one standard-deviation error bars.



Solution for Exercise 4

❑ C source code at `/home/shaohao/hpc_train/numlibs/gsl/linear_fit.c`

- Include gsl head file: `#include <gsl/gsl_fit.h>`
- Call gsl function: `gsl_fit_linear_est`

❑ Compile and run

```
module load gsl/1.16/INTEL-14.0.2    # set up gsl environmens
module show gsl/1.16/INTEL-14.0.2   # show gsl environments
icc -c -I${LHPC_ROOTGSL}/include linear_fit.c  # compile
icc linear_fit.o -lgsl -o linear_fit          # link
./linear_fit                                # run
```

Next training: Introduction to R

March 18, 2015

9:30 AM - 11:30 AM

- R is a free software environment for statistical computing and graphics.