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## **xSDK User Manual**

Jeff Johnson, Alicia Klinvex, X. Sherry Li, Barry Smith, Ulrike Meier Yang

Prepared by  
Sandia National Laboratories  
Albuquerque, New Mexico 87185 and Livermore, California 94550

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## Abstract

**TODO: This needs work.**

Some application developers need to be able to use multiple DOE libraries together, such as PETSc, Hypre, SuperLU, Trilinos, and Alquimia. This is nontrivial because these libraries all expect the data to be stored in different ways, and the way that you call a PETSc KSP linear solver, for instance, looks fundamentally different from the way you would call a Trilinos Belos linear solver. The IDEAS software productivity project plans to address this problem with the Extreme-scale Scientific Software Development Kit (xSDK). The xSDK will provide an interoperability layer that enables easy installation and combined usage of the IDEAS libraries. This document describes the various interoperability layers and how to install and use the xSDK.

# Introduction

The following are the libraries included in the xSDK.

Two of the libraries, PETSc and Trilinos are comprehensive numerical software packages<sup>1</sup> that can solve a large variety of problems with both their own algorithmic implementations and wrappers to other numerical libraries. SuperLU and hypre are specialized libraries that provide sophisticated solvers for a restricted family of problems. SuperLU provides sparse direct linear solvers and ILU preconditioners while hypre provides iterative solvers for linear systems.

The xSDK provides a common system for installing and working with all of these packages so, in some sense, xSDK is meta-package.

## Alquimia

Alquimia is an API for providing mature geochemistry and biogeochemistry capabilities to reactive transport codes. It is not a geochemistry solver—it is a library comprising data structures and interfaces that wrap chemistry solvers from well-established codes like PFlotran (<http://www.pflotran.org>) and CrunchFlow (<http://www.csteefel.com/CrunchFlowIntroduction>) allowing developers of newer codes to use these solvers with a single interface. We refer to these chemistry solvers as *chemistry engines*.

Currently, Alquimia supports two chemistry engines: PFlotran and CrunchFlow. There are plans to support more in the future, depending on funding and levels of engagement in the DOE’s reactive transport research community.

You can find out more about installing and programming with Alquimia at <https://www.github.com/LAMMPS-MolSSI/xSDK>.

## hypre

The *hypre* software library provides high performance preconditioners and solvers for the solution of large sparse linear systems on massively parallel computers. It was created with the primary goal of providing users with advanced parallel preconditioners. The library

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<sup>1</sup>Note that the term “Trilinos package” has a specific meaning that, here we just mean package in the generic sense

features parallel multigrid solvers for both structured and unstructured grid problems. For ease of use, these solvers are accessed from the application code via *hypr*'s conceptual linear system interfaces, which allow a variety of natural problem descriptions and include a structured, a semi-structured interface, and a traditional linear-algebra based interface. The (semi-)structured interfaces are an alternative to the standard matrix-based interface that describes rows, columns, and coefficients of a matrix. Here, instead, matrices are described primarily in terms of stencils and logically structured grids. These interfaces give application users a more natural means for describing their linear systems, and provide access to methods such as structured multigrid solvers, which can take advantage of the additional information beyond just the matrix.

The library is written in C and has a Fortran interface. It supports MPI+OpenMP. More detailed information can be found at [?, ?].

TODO: Fix these citations.

## PETSc

PETSc is a suite of data structures and routines for the scalable solution of scientific applications modeled by partial differential equations. It includes linear solvers, preconditioners, nonlinear solvers, and ODE integrators. It also provides a variety of scalable constrained and unconstrained optimization solvers. PETSc utilizes the MPI program model and does not use threads. It does not support solving linear systems with multiple right-hand-sides. While it does not include eigensolvers, there is an eigensolver package called SLEPc built on top of PETSc with a very similar interface. The library libMesh and the framework MOOSE provide finite element solvers that utilize PETSc.

## SuperLU

SuperLU is a general purpose library for the direct solution of large, sparse, nonsymmetric systems of linear equations on high performance machines. The library routines will perform an LU decomposition with partial pivoting and triangular system solves through forward and back substitution. The LU factorization routines can handle non-square matrices but the triangular solves are performed only for square matrices. The matrix columns may be preordered (before factorization) either through library or user supplied routines. This reordering for sparsity is completely separate from the factorization. Working precision iterative refinement subroutines are provided for improved backward stability. Routines are also provided to equilibrate the system, estimate the condition number, calculate the relative backward error, and estimate error bounds for the refined solutions.

There are three separate versions of this code: SuperLU (for sequential machines), SuperLU-MT (for shared memory parallel machines with using OpenMP or Pthread), and

SuperLU\_DIST (for distributed memory machines using MPI).

The library is written in C, with Fortran interface. SuperLU\_DIST supports MPI+X, where X can be CUDA, OpenMP, or both.

## Trilinos

**TODO: This needs work. Maybe Jim or Mike should have a look at it.**

The Trilinos Project is an effort to develop algorithms and enabling technologies within an object-oriented software framework for the solution of large-scale, complex multi-physics engineering and scientific problems. Trilinos is organized into 66 different packages, each with a specific focus. These packages include linear and nonlinear solvers, preconditioners (including algebraic multigrid), graph partitioners, eigensolvers, and optimization algorithms, among other things. Users are only required to install the subset of packages related to the problems they are trying to solve.

Trilinos supports MPI+X, where X can be CUDA, OpenMP, etc.

# Chapter 1

## xSDK Installation

TODO: Ignore this. Barry is going to take care of the circular dependency issue; then we can revisit the installation procedure.

### Using PETSc configuration script

The easiest way to install the various xSDK libraries (Trilinos, PETSc, Hypre, and SuperLU) is the following three step process:

1. Download and unpack the PETSc tarball from <http://www.mcs.anl.gov/petsc/>, or pull from the repo via `git clone https://bitbucket.org/petsc/petsc`.
2. Run the PETSc configuration script.
3. Type `make`, then `make install`. Don't worry; the PETSc configuration script will remind you to do this.

The PETSc configuration script is very sophisticated, as you can tell by typing `./configure --help`. It will produce a list for you of all the different configuration options (organized into different categories). The most interesting of these configuration options is `--download-<PACKAGE>`, where `PACKAGE` can be `hypre`, `superlu`, `trilinos`, etc. If you install PETSc via the following line:

```
./configure --prefix=<HOME>/petsc-install --download-xsdk
```

PETSc will download and install `hypre`, `SuperLU`, `Trilinos`, as well as commonly needed dependent package for you, and PETSc itself will be installed to the directory `<HOME>/petsc-install`. If PETSc encounters any problems, such as being unable to find your MPI installation, it will output a helpful error message explaining what the problem is and how you can fix it.

If PETSc finds itself unable to download any packages you request (because you are behind a firewall, for instance), it will output the following message explaining how to fix the problem:

=====

```

Trying to download http://ftp.mcs.anl.gov/pub/petsc/externalpackages/hypre-2.10.0b-p1.tar.gz for HYPRE
=====
Trying to download ftp://ftp.mcs.anl.gov/pub/petsc/externalpackages/hypre-2.10.0b-p1.tar.gz for HYPRE
=====
-----
Unable to download package hypre from: http://ftp.mcs.anl.gov/pub/petsc/externalpackages/hypre-2.10.0b-p1.tar.gz
* If URL specified manually - perhaps there is a typo?
* If your network is disconnected - please reconnect and rerun ./configure
* Or perhaps you have a firewall blocking the download
* Alternatively, you can download the above URL manually, to /yourselectedlocation/hypre-2.10.0b-p1.tar.gz
  and use the configure option:
  --download-hypre=/yourselectedlocation/hypre-2.10.0b-p1.tar.gz
Unable to download package hypre from: ftp://ftp.mcs.anl.gov/pub/petsc/externalpackages/hypre-2.10.0b-p1.tar.gz
* If URL specified manually - perhaps there is a typo?
* If your network is disconnected - please reconnect and rerun ./configure
* Or perhaps you have a firewall blocking the download
* Alternatively, you can download the above URL manually, to /yourselectedlocation/hypre-2.10.0b-p1.tar.gz
  and use the configure option:
  --download-hypre=/yourselectedlocation/hypre-2.10.0b-p1.tar.gz
*****

```

## Without PETSc configuration script

If you would like to manually install Trilinos and enable the PETSc, hypre, and SuperLU interfaces, you may specify those options in the Trilinos configuration script (which I will call do-configure). Note that Trilinos requires all third party libraries to be installed before the configuration process. An example configuration script is below:

### Program 1.1. do-configure

```

1  #!/bin/bash
2  #
3
4  TRILINOS_HOME=/home/amklnv/TrilinosDir/Trilinos
5  PETSC_DIR=/home/amklnv/PETSc/petsc-3.6.0
6  PETSC_ARCH=arch-linux2-cxx-debug
7  PETSC_LIB="-Wl,-rpath,/home/amklnv/PETSc/petsc-3.6.0/arch-linux2-cxx-debug/lib -L/home/amklnv/PETSc/petsc-3.6.0/arch-
  linux2-cxx-debug/lib -lpetsc -Wl,-rpath,/home/amklnv/PETSc/petsc_install/lib -L/home/amklnv/PETSc/petsc_install/lib
  -lsuperlu_4.3 -lsuperlu_dist_4.0 -lHYPRE -Wl,-rpath,/home/amklnv/lapack-3.5.0 -L/home/amklnv/lapack-3.5.0 -llapack
  -lrefblas -lparmetis -lmetis -L/projects/install/rhel6-x86_64/sems/compiler/gcc/4.7.2/openmpi/1.6.5/lib -L/projects/
  install/rhel6-x86_64/sems/compiler/gcc/4.7.2/base/lib/gcc/x86_64-unknown-linux-gnu/4.7.2 -L/projects/install/rhel6-
  x86_64/sems/compiler/gcc/4.7.2/base/lib64 -L/projects/install/rhel6-x86_64/sems/compiler/gcc/4.7.2/base/lib -lmpi_f90
  -lmpi_f77 -lgfortran -lm -lgfortran -lm -lgfortran -lm -lgfortran -lm -lquadmath -lm -lmpi_cxx -lstlcl++ -L/projects/
  install/rhel6-x86_64/sems/compiler/gcc/4.7.2/openmpi/1.6.5/lib -L/projects/install/rhel6-x86_64/sems/compiler/gcc
  /4.7.2/base/lib/gcc/x86_64-unknown-linux-gnu/4.7.2 -L/projects/install/rhel6-x86_64/sems/compiler/gcc/4.7.2/base/
  lib64 -L/projects/install/rhel6-x86_64/sems/compiler/gcc/4.7.2/base/lib -ldl -lmpi -lrt -lnsl -lutil -lgcc_s -
  lpthread -ldl"
8
9  PETSC_INCLUDE_PATH="${PETSC_DIR}/${PETSC_ARCH}/include;${PETSC_DIR}/include;${PETSC_DIR}"
10
11 HYPRE_LIBRARY_DIRS="/home/amklnv/PETSc/petsc-3.6.0/arch-linux2-cxx-debug/externalpackages/hypre-2.10.0b-p1/src/hypre/lib"
12 HYPRE_INCLUDE_DIRS="/home/amklnv/PETSc/petsc-3.6.0/arch-linux2-cxx-debug/externalpackages/hypre-2.10.0b-p1/src/hypre/
  include"
13
14 rm -rf CMakeFiles CMakeCache.txt
15
16 cmake \
17   -D CMAKE_BUILD_TYPE=DEBUG \
18   -D Trilinos_ENABLE_STRONG_C_COMPILE_WARNINGS:BOOL=OFF \
19   -D CMAKE_INSTALL_PREFIX:PATH="/home/amklnv/TrilinosDir/trilinos-install" \
20   -D TPL_ENABLE_MPI:BOOL=ON \
21   -D Trilinos_ENABLE_EXPLICIT_INSTANTIATION:BOOL=ON \
22   -D Trilinos_ENABLE_ALL_OPTIONAL_PACKAGES:BOOL=OFF \
23   -D Trilinos_ENABLE_Amesos:BOOL=ON \
24   -D Trilinos_ENABLE_Amesos2:BOOL=ON \
25   -D Trilinos_ENABLE_Anasazi:BOOL=ON \
26   -D Trilinos_ENABLE_Aztec00:BOOL=ON \
27   -D Trilinos_ENABLE_Belos:BOOL=ON \

```



```

28 -D Trilinos_ENABLE_Epetra:BOOL=ON \
29 -D Trilinos_ENABLE_EpetraExt:BOOL=ON \
30 -D Trilinos_ENABLE_Galeri:BOOL=ON \
31 -D Trilinos_ENABLE_Ipack:BOOL=ON \
32 -D Trilinos_ENABLE_Ipack2:BOOL=ON \
33 -D Trilinos_ENABLE_Isorropia:BOOL=ON \
34 -D Trilinos_ENABLE_Kokkos:BOOL=ON \
35 -D Trilinos_ENABLE_ML:BOOL=ON \
36 -D Trilinos_ENABLE_Tpetra:BOOL=ON \
37 -D Trilinos_ENABLE_TrilinosCouplings:BOOL=ON \
38 -D Trilinos_ENABLE_Triutils:BOOL=ON \
39 -D Trilinos_ENABLE_NOX:BOOL=ON \
40 -D Trilinos_ENABLE_Zoltan:BOOL=ON \
41 -D Trilinos_ENABLE_Zoltan2:BOOL=ON \
42 \
43 -D EpetraExt_USING_PETSC:BOOL=ON \
44 -D NOX_ENABLE_PETSC:BOOL=ON \
45 -D NOX_ENABLE_ABSTRACT_IMPLEMENTATION_PETSC:BOOL=ON \
46 -D TPL_ENABLE_PETSC:BOOL=ON \
47 -D PETSC_LIBRARY_DIRS:FILEPATH="${PETSC_LIB}" \
48 -D PETSC_INCLUDE_DIRS:FILEPATH="${PETSC_INCLUDE_PATH}" \
49 -D TPL_PETSC_LIBRARIES:STRING="${PETSC_LIB}" \
50 -D TPL_PETSC_INCLUDE_DIRS:STRING="${PETSC_INCLUDE_PATH}" \
51 \
52 -D Amesos2_ENABLE_KLU2:BOOL=ON \
53 \
54 -D EpetraExt_ENABLE_HYPRE:BOOL=ON \
55 -D Ipack_ENABLE_HYPRE:BOOL=ON \
56 -D TPL_ENABLE_HYPRE:BOOL=ON \
57 -D HYPRE_LIBRARY_DIRS:FILEPATH="${HYPRE_LIBRARY_DIRS}" \
58 -D HYPRE_INCLUDE_DIRS:FILEPATH="${HYPRE_INCLUDE_DIRS}" \
59 \
60 -D Belos_ENABLE_TESTS:BOOL=ON \
61 -D Belos_ENABLE_EXAMPLES:BOOL=ON \
62 -D Ipack2_ENABLE_TESTS:BOOL=ON \
63 -D Ipack2_ENABLE_EXAMPLES:BOOL=ON \
64 -D DART_TESTING_TIMEOUT:STRING=300 \
65 -D BLAS_LIBRARY_NAMES:STRING="libf77blas.so.3" \
66 -D BLAS_LIBRARY_DIRS:PATH="/usr/lib64/atlas" \
67 -D LAPACK_LIBRARY_NAMES:STRING="liblapack.so.3" \
68 -D LAPACK_LIBRARY_DIRS:PATH="/usr/lib64/atlas" \
69 \
70 -D Trilinos_EXTRA_REPOSITORIES:STRING=preCopyrightTrilinos \
71 -D Trilinos_ENABLE_xSDKTrilinos:BOOL=ON \
72 -D xSDKTrilinos_USING_PETSC:BOOL=ON \
73 -D xSDKTrilinos_USING_HYPRE:BOOL=ON \
74 \
75 ${TRILINOS_HOME}

```

Let's examine this script in more detail. Since we are interested in the PETSc interface, we must first define `PETSC_DIR` and `PETSC_ARCH`. `PETSC_LIB` looks intimidating at first, but you can generate it automatically by typing `make getlinklibs` in `PETSC_DIR`. We then define the PETSc and hypre include paths. Everything else looks pretty normal until we reach lines 43–50, which enable the Epetra-based PETSc interface. Lines 54–58 enable the Epetra-based hypre interface. Then lines 70–73 enable the Tpetra-based interfaces living in the pre-Copyright package `xSDKTrilinos`.



# Chapter 2

## xSDK Interface Usage

This section describes the individual interfaces and their usage.

### Trilinos-PETSc

There is a two-way interface between PETSc and Trilinos which allows users to use PETSc datatypes with Trilinos and vice-versa.

### Using PETSc Mat and Vec with Trilinos solvers

Trilinos has two new interfaces to support using PETSc Mat anywhere a Tpetra::RowMatrix or Tpetra::CrsMatrix can be used. For packages requiring a Tpetra::RowMatrix or Tpetra::Operator, such as Anasazi and Belos, you may wrap a PETSc Mat in a Tpetra::PETScAIJMatrix; otherwise, you can copy it to a Tpetra::CrsMatrix. We will demonstrate each of those functions in the examples below.

Our first example (Program 2.1) shows how to compute the smallest eigenpairs of a PETSc Mat, specifically Poisson2D, using Trilinos' Anasazi package.

**Program 2.1.** PETSc\_AnasaziEx.cpp

```
1  #include "petscksp.h"
2  #include "AnasaziBasicEigenproblem.hpp"
3  #include "AnasaziConfigDefs.hpp"
4  #include "AnasaziTpetraAdapter.hpp"
5  #include "AnasaziRTSOLMgr.hpp"
6  #include "Teuchos_ParameterList.hpp"
7  #include "Tpetra_PETScAIJMatrix.hpp"
8
9  int main(int argc, char **args)
10 {
11     using Teuchos::RCP;
12     using Teuchos::rcp;
13     using std::cout;
14     using std::endl;
15
16     typedef Tpetra::PETScAIJMatrix<>          PETScAIJMatrix;
17     typedef PETScAIJMatrix::scalar_type      Scalar;
18     typedef PETScAIJMatrix::local_ordinal_type L0;
19     typedef PETScAIJMatrix::global_ordinal_type G0;
20     typedef PETScAIJMatrix::node_type        Node;
21     typedef Tpetra::Vector<Scalar, L0, G0, Node> Vector;
22     typedef Tpetra::Map<L0, G0, Node>        Map;
```

```

23 typedef Tpetra::Operator<Scalar,LO,GO,Node> OP;
24 typedef Tpetra::MultiVector<Scalar,LO,GO,Node> MV;
25 typedef Anasazi::RTRSolMgr<Scalar,MV,OP> SolMgr;
26 typedef Anasazi::BasicEigenproblem<Scalar,MV,OP> Problem;
27 typedef Anasazi::OperatorTraits<Scalar,MV,OP> OPT;
28 typedef Anasazi::MultiVecTraits<Scalar,MV> MVT;
29
30 Mat A;
31 PetscInt m = 50, n = 50;
32 PetscInt nev = 4;
33 PetscErrorCode ierr;
34 MPI_Comm comm;
35 PetscInt Istart, Iend, Ii, i, j, J, rank;
36 PetscScalar v;
37
38 PetscInitialize(&argc,&args,NULL,NULL);
39 ierr = PetscOptionsGetInt(PETSC_NULL,"-mx",&m,PETSC_NULL);CHKERRQ(ierr);
40 ierr = PetscOptionsGetInt(PETSC_NULL,"-my",&n,PETSC_NULL);CHKERRQ(ierr);
41 ierr = PetscOptionsGetInt(PETSC_NULL,"-nev",&nev,PETSC_NULL);CHKERRQ(ierr);
42
43 ierr = MatCreate(PETSC_COMM_WORLD,&A);CHKERRQ(ierr);
44 ierr = MatSetSizes(A,PETSC_DECIDE,PETSC_DECIDE,m*n,m*n);CHKERRQ(ierr);
45 ierr = MatSetType(A,MATAIJ);CHKERRQ(ierr);
46 ierr = MatMPIAIJSetPreallocation(A,5,PETSC_NULL,5,PETSC_NULL);CHKERRQ(ierr);
47 ierr = MatSetUp(A);CHKERRQ(ierr);
48 ierr = PetscObjectGetComm((PetscObject)A,&comm);CHKERRQ(ierr);
49 ierr = MPI_Comm_rank(comm,&rank);CHKERRQ(ierr);
50
51 ierr = MatGetOwnershipRange(A,&Istart,&Iend);CHKERRQ(ierr);
52 for (Ii=Istart; Ii<Iend; Ii++) {
53     v = -1.0; i = Ii/n; j = Ii - i*n;
54     if (i>0) {J = Ii - n; ierr = MatSetValues(A,1,&Ii,1,&J,&v,INSERT_VALUES);CHKERRQ(ierr);}
55     if (i<m-1) {J = Ii + n; ierr = MatSetValues(A,1,&Ii,1,&J,&v,INSERT_VALUES);CHKERRQ(ierr);}
56     if (j>0) {J = Ii - 1; ierr = MatSetValues(A,1,&Ii,1,&J,&v,INSERT_VALUES);CHKERRQ(ierr);}
57     if (j<n-1) {J = Ii + 1; ierr = MatSetValues(A,1,&Ii,1,&J,&v,INSERT_VALUES);CHKERRQ(ierr);}
58     v = 4.0; ierr = MatSetValues(A,1,&Ii,1,&Ii,&v,INSERT_VALUES);CHKERRQ(ierr);
59 }
60
61 ierr = MatAssemblyBegin(A,MAT_FINAL_ASSEMBLY);CHKERRQ(ierr);
62 ierr = MatAssemblyEnd(A,MAT_FINAL_ASSEMBLY);CHKERRQ(ierr);
63
64 RCP<PETScAIJMatrix> tpetraA = rcp(new PETScAIJMatrix(A));
65
66 RCP<MV> initGuess = rcp(new MV(tpetraA->getDomainMap(),4,false));
67 initGuess->randomize();
68
69 RCP<Problem> problem = rcp(new Problem(tpetraA,initGuess));
70 problem->setNEV(nev);
71 problem->setHermitian(true);
72 problem->setProblem();
73
74 Teuchos::ParameterList pl;
75 pl.set("Verbosity", Anasazi::IterationDetails + Anasazi::FinalSummary);
76 pl.set("Convergence Tolerance", 1e-6);
77 RCP<SolMgr> solver = rcp(new SolMgr(problem, pl));
78
79 Anasazi::ReturnType returnCode = solver->solve();
80 Anasazi::Eigensolution<Scalar,MV> sol = problem->getSolution();
81 std::vector<Anasazi::Value<Scalar>> evals = sol.Evals;
82 RCP<MV> evecs = sol.Evecs;
83
84 ierr = PetscFinalize();CHKERRQ(ierr);
85 return 0;
86 }

```

**Lines 1–7** Include statements

**Lines 11–28** Typedefs and using statements to make the code more readable

**Lines 30–36** PETSc variables

**Lines 38–41** Get the command line arguments using PETSc. This example has three of them: the number of mesh points in the x direction, number of mesh points in the y direction, and the number of desired eigenpairs.

**Lines 43–62** Create the PETSc Mat and set its values.

**Line 64** Wrap the PETSc Mat in a `Tpetra::PETScAIJMatrix`. Since Anasazi only requires a `Tpetra::Operator`<sup>1</sup>, we do not have to deep copy the data to a `Tpetra::CrsMatrix`.

**Lines 66–67** Create a random initial guess for the eigensolver. Note that we can treat `tpetraA` just like any other `Tpetra::RowMatrix` and obtain its domain map via `getDomainMap()`.

**Lines 69–72** Create the eigenproblem for Anasazi. We provide the operator  $A$  as well as our initial guess for the set of desired eigenvectors to the constructor. We then request a certain number of eigenvectors and inform the eigensolver that our problem is Hermitian<sup>2</sup>.

**Lines 74–77** Create the parameter list for the Riemannian Trust Region eigensolver. We have elected to have the solver print out the list of approximate eigenvalues at each iteration, along with their associated residuals. We also set our convergence tolerance here.

**Lines 79–82** Solve the eigenvalue problem. After it is solved, we may grab the eigenvalues and eigenvectors via `getSolution()`.

The second example (Program 2.2) demonstrates how to use PETSc datatypes with Trilinos packages that require a `Tpetra::CrsMatrix`. One such package is Amesos2, which contains a variety of direct solvers (and interfaces to external direct solvers). In this example, we will solve a linear system  $Ax = b$  where  $A$  is the 2D discretization of the Poisson operator on a unit cube, and  $b$  is a random vector.

### Program 2.2. PETSc\_Amesos2Ex.cpp

```
1  #include <Teuchos_ScalarTraits.hpp>
2  #include <Teuchos_RCP.hpp>
3  #include <Tpetra_Map.hpp>
4  #include <Tpetra_MultiVector.hpp>
5  #include <Tpetra_CrsMatrix.hpp>
6  #include <Tpetra_Vector.hpp>
7  #include "Amesos2.hpp"
8  #include "Amesos2_Version.hpp"
9  #include "Amesos2_KLU2.hpp"
10
11 #include "petscksp.h"
12 #include "Tpetra_PETScAIJMatrix.hpp"
13
14 int main(int argc, char **args)
15 {
16     using Teuchos::RCP;
```

---

<sup>1</sup>RowMatrix is a specific type of Operator, and CrsMatrix is a specific type of RowMatrix. Therefore, you can use a RowMatrix anywhere an Operator is accepted, but you can't necessarily use a RowMatrix anywhere a CrsMatrix is expected.

<sup>2</sup>Some eigensolvers are optimized for use on Hermitian eigenproblems. Others do not work on non-Hermitian problems at all, so it is important to specify this.

```

17 using Teuchos::rcp;
18 using Teuchos::ArrayView;
19
20 typedef Tpetra::PETScAIJMatrix<> PETScAIJMatrix;
21 typedef PETScAIJMatrix::scalar_type Scalar;
22 typedef PETScAIJMatrix::local_ordinal_type LO;
23 typedef PETScAIJMatrix::global_ordinal_type GO;
24 typedef PETScAIJMatrix::node_type Node;
25 typedef Tpetra::CrsMatrix<Scalar, LO, GO> CrsMatrix;
26 typedef Tpetra::Vector<Scalar, LO, GO> Vector;
27 typedef Tpetra::Map<LO, GO> Map;
28 typedef Tpetra::Operator<Scalar, LO, GO> OP;
29 typedef Tpetra::MultiVector<Scalar, LO, GO> MV;
30 typedef Amesos2::Solver<CrsMatrix, MV> Solver;
31
32 Vec x, b;
33 Mat A;
34 PetscRandom rctx;
35 PetscInt i, j, li, j, Istart, Iend;
36 PetscInt m = 50, n = 50;
37 PetscErrorCode ierr;
38 PetscScalar v;
39 PetscInt rank=0;
40 MPI_Comm comm;
41
42 PetscInitialize(&argc, &argv, NULL, NULL);
43 ierr = PetscOptionsGetInt(PETSC_NULL, "-mx", &m, PETSC_NULL); CHKERRQ(ierr);
44 ierr = PetscOptionsGetInt(PETSC_NULL, "-my", &n, PETSC_NULL); CHKERRQ(ierr);
45
46 ierr = MatCreate(PETSC_COMM_WORLD, &A); CHKERRQ(ierr);
47 ierr = MatSetSizes(A, PETSC_DECIDE, PETSC_DECIDE, m*n, m*n); CHKERRQ(ierr);
48 ierr = MatSetType(A, MATAIJ); CHKERRQ(ierr);
49 ierr = MatMPIAIJSetPreallocation(A, 5, PETSC_NULL, 5, PETSC_NULL); CHKERRQ(ierr);
50 ierr = MatSetUp(A); CHKERRQ(ierr);
51 PetscObjectGetComm((PetscObject)A, &comm);
52 ierr = MPI_Comm_rank(comm, &rank); CHKERRQ(ierr);
53
54 ierr = MatGetOwnershipRange(A, &Istart, &Iend); CHKERRQ(ierr);
55 for (Ii=Istart; Ii<Iend; Ii++) {
56     v = -1.0; i = Ii/n; j = Ii - i*n;
57     if (i>0) {j = Ii - n; ierr = MatSetValues(A, 1, &Ii, 1, &j, &v, INSERT_VALUES); CHKERRQ(ierr);}
58     if (i<m-1) {j = Ii + n; ierr = MatSetValues(A, 1, &Ii, 1, &j, &v, INSERT_VALUES); CHKERRQ(ierr);}
59     if (j>0) {j = Ii - 1; ierr = MatSetValues(A, 1, &Ii, 1, &j, &v, INSERT_VALUES); CHKERRQ(ierr);}
60     if (j<n-1) {j = Ii + 1; ierr = MatSetValues(A, 1, &Ii, 1, &j, &v, INSERT_VALUES); CHKERRQ(ierr);}
61     v = 4.0; ierr = MatSetValues(A, 1, &Ii, 1, &Ii, &v, INSERT_VALUES); CHKERRQ(ierr);
62 }
63
64 ierr = MatAssemblyBegin(A, MAT_FINAL_ASSEMBLY); CHKERRQ(ierr);
65 ierr = MatAssemblyEnd(A, MAT_FINAL_ASSEMBLY); CHKERRQ(ierr);
66
67 ierr = VecCreate(PETSC_COMM_WORLD, &x); CHKERRQ(ierr);
68 ierr = VecSetSizes(x, PETSC_DECIDE, m*n); CHKERRQ(ierr);
69 ierr = VecDuplicate(x, &b); CHKERRQ(ierr);
70 ierr = PetscRandomCreate(PETSC_COMM_WORLD, &rctx); CHKERRQ(ierr);
71 ierr = PetscRandomSetFromOptions(rctx); CHKERRQ(ierr);
72 ierr = VecSetRandom(b, rctx); CHKERRQ(ierr);
73 ierr = PetscRandomDestroy(&rctx); CHKERRQ(ierr);
74
75 RCP<CrsMatrix> tpetraA = xSDKTrilinos::deepCopyPETScAIJMatrixToTpetraCrsMatrix<Scalar, LO, GO, Node>(A);
76 RCP<Vector> tpetraX = xSDKTrilinos::deepCopyPETScVecToTpetraVector<Scalar, LO, GO, Node>(x);
77 RCP<Vector> tpetraB = xSDKTrilinos::deepCopyPETScVecToTpetraVector<Scalar, LO, GO, Node>(b);
78
79 RCP<Solver> solver = Amesos2::create<CrsMatrix, MV>("KLU2", tpetraA, tpetraX, tpetraB);
80 solver->symbolicFactorization();
81 solver->numericFactorization();
82 solver->solve();
83
84 ierr = PetscFinalize(); CHKERRQ(ierr);
85 return 0;
86 }

```

**Lines 1–65** These lines are very similar to the previous example. We set up convenient typedefs, read in the command line arguments with PETSc (in this case, the  $x$  and  $y$  dimensions of the grid), and set up the Poisson2D matrix.

**Lines 67–73** Create the initial guess for  $x$  and the random RHS  $b$ .

**Lines 75–77** Copy the PETSc Mat and Vecs to Tpetra::CrsMatrix and Tpetra::Vector, so that we can use them with the Amesos2 linear solvers.

**Lines 79–82** Create an Amesos2 linear solver, specifically the native solver KLU2. Perform the symbolic and numeric factorizations, then solve the linear system.

### Is the data copied or wrapped?

If you are using a part of Trilinos that requires Operator or RowMatrix, the data is wrapped. If you need a CrsMatrix specifically, the data is deep-copied.

## Using Trilinos datatypes with PETSc KSP solvers

If you would like to use Trilinos datatypes, such as Tpetra::Operator and Tpetra::MultiVector, with a PETSc KSP linear solver, you may use the new Belos<sup>3</sup> interface: PETScSolMgr. This interface is very similar to that of the other native Belos linear solvers, which makes solving linear systems such as  $AX = B$  a simple process.

1. (Optional) Create a Tpetra::Operator for the preconditioner  $M \approx A$ . You may use the preexisting preconditioners of Ifpack2 and MueLu, or you may create your own custom preconditioner. Alternatively, you may choose not to use a preconditioner at all.
2. Create a Belos::LinearProblem containing the operator  $A$ , the initial guess  $X$ , the right-hand side  $B$ , and the preconditioner  $M$  (if you have one).
3. Create a Teuchos::ParameterList containing the parameters you wish to set. These parameters are summarized in Table 2.1.
4. Create a Belos::PETScSolMgr with the LinearProblem and ParameterList from the previous steps.
5. Call solve()

The following example (Program 2.3) illustrates this process in greater detail. Note that this example does not contain a single of PETSc code.

### Program 2.3. Tpetra\_KSPEx.cpp

---

<sup>3</sup>Belos is the iterative solver package of Trilinos.

Parameter	Description	Default Value
Maximum Iterations	integer defining the maximum number of iterations to be performed.	1000
Solver	string defining the linear solver to be used. A list of all valid linear solver options can be found at <a href="http://www.mcs.anl.gov/petsc/petsc-current/docs/manualpages/KSP/KSPType.html">http://www.mcs.anl.gov/petsc/petsc-current/docs/manualpages/KSP/KSPType.html</a>	KSPGMRES
Verbosity	Belos::MsgType defining the amount of output the program should produce. Options include Belos::Errors, Belos::Warnings, Belos::IterationDetails, Belos::TimingDetails, and Belos::StatusTestDetails	Belos::Errors
Convergence Tolerance	double defining the tolerance of the linear solver	$10^{-8}$

**Table 2.1.** Belos::PETScSolMgr parameters

```

1  #include "BelosTpetraAdapter.hpp"
2  #include "BelosPETScSolMgr.hpp"
3  #include "Ifpack2_Factory.hpp"
4  #include "Teuchos_CommandLineProcessor.hpp"
5  #include "Teuchos_ParameterList.hpp"
6  #include "Tpetra_CrsMatrix.hpp"
7  #include "Tpetra_DefaultPlatform.hpp"
8  #include "Tpetra_MultiVector.hpp"
9  #include "MatrixMarket_Tpetra.hpp"
10
11 int main(int argc, char *argv[]) {
12     typedef double ST;
13     typedef Teuchos::ScalarTraits<ST> SCT;
14     typedef SCT::magnitudeType MT;
15     typedef Tpetra::MultiVector<> MV;
16     typedef Tpetra::Operator<> OP;
17     typedef Belos::MultiVecTraits<ST,MV> MVT;
18     typedef Belos::OperatorTraits<ST,MV,OP> OPT;
19     typedef Tpetra::CrsMatrix<> CrsMatrix;
20     typedef Ifpack2::Preconditioner<> Prec;
21
22     using Teuchos::ParameterList;
23     using Teuchos::RCP;
24     using Teuchos::rcp;
25
26     Teuchos::oblackholestream blackhole;
27     Teuchos::GlobalMPISession mpiSession (&argc, &argv, &blackhole);
28     RCP<const Teuchos::Comm<int>> comm = Tpetra::DefaultPlatform::getDefaultPlatform().getComm();
29     const int myRank = comm->getRank();
30
31     double tol = 1e-6;
32     std::string filename("/home/amklinv/matrices/cage4.mtx");
33     std::string ksptype("gmres");
34     Teuchos::CommandLineProcessor cmdp(false,false);
35     cmdp.setOption("filename",&filename,"Filename for test matrix.");
36     cmdp.setOption("tol",&tol,"Relative residual tolerance.");
37     cmdp.setOption("ksptype",&ksptype,"Type of linear solver to be used.");
38     if (cmdp.parse(argc,argv) != Teuchos::CommandLineProcessor::PARSE_SUCCESSFUL) {
39         return -1;
40     }
41
42     RCP<CrsMatrix> A = Tpetra::MatrixMarket::Reader<CrsMatrix>::readSparseFile(filename,comm);
43     RCP<MV> B = rcp(new MV(A->getRowMap(),1,false));
44     RCP<MV> X = rcp(new MV(A->getRowMap(),1,false));
45     MVT::MvInit(*X);
46     MVT::MvInit(*B,1);
47
48     Ifpack2::Factory factory;
49     RCP<Prec> M = factory.create("RELAXATION", A.getConst());
50     ParameterList ifpackParams;
51     ifpackParams.set("relaxation: type","Jacobi");

```



```

52  M->setParameters(ifpackParams);
53  M->initialize();
54  M->compute();
55
56  ParameterList beloSList;
57  beloSList.set( "Maximum Iterations", 100 );
58  beloSList.set( "Convergence Tolerance", tol );
59  beloSList.set( "Solver", ksptype );
60
61  RCP<Belos::LinearProblem<double,MV,OP> > problem
62  = rcp( new Belos::LinearProblem<double,MV,OP>( A, X, B ) );
63  problem->setLeftPrec( M );
64  problem->setProblem();
65
66  RCP< Belos::PETScSolMgr<double,MV,OP> > solver
67  = rcp( new Belos::PETScSolMgr<double,MV,OP>(problem, rcp(&beloSList,false)) );
68  solver->solve();
69 }

```

**Lines 1–9** Include statements

**Lines 12–24** Typedefs and using statements to make the code more readable

**Lines 26–29** Set up MPI

**Lines 31–40** Parse command line arguments. This program allows the user to specify the filename for the matrix, the tolerance for the linear solve, and which linear solver is used. A list of all valid linear solver options can be found at <http://www.mcs.anl.gov/petsc/petsc-current/docs/manualpages/KSP/KSPType.html>.

**Lines 42–46** Set up the linear system by reading the matrix from a file, setting the initial guess for the solution as  $\vec{0}$  and setting the right hand side as  $\vec{1}$ .

**Lines 48–54** Set up the Ifpack2 Jacobi preconditioner.

**Lines 56–59** Set the maximum number of iterations, convergence tolerance, and which PETSc KSP solver is being used.

**Lines 61–64** Set up the linear problem for the Belos solver.

**Lines 66–68** Solve the linear system.

## Can I use this to solve linear systems with multiple right-hand sides?

Yes. Unfortunately, PETSc has no support for multivectors at this time, so each of the right hand sides will be processed independently. If you want block or pseudo-block linear solvers, those are available within Trilinos.

## Is the data copied or wrapped?

The raw Tpetra matrix (or operator) data is wrapped rather than deep copied. The same applies to the preconditioner, if you are using a preconditioner.

## Trilinos-hypre

Trilinos also has a new Tpetra-based interface to hypre. This interface lives in the Ifpack2 library with the native Trilinos preconditioners. The following examples will demonstrate how to take advantage of this exciting new functionality.

In our first example (Program 2.4), we will examine how to use hypre solvers and preconditioners with Tpetra objects.

### Program 2.4. Hypre\_SolveEx.cpp

```
1  #include "Ifpack2_Preconditioner.hpp"
2  #include "Ifpack2_Hypre.hpp"
3  #include "Teuchos_CommandLineProcessor.hpp"
4  #include "Tpetra_CrsMatrix.hpp"
5  #include "Tpetra_DefaultPlatform.hpp"
6
7  int main(int argc, char *argv[]) {
8      using Teuchos::Array;
9      using Teuchos::RCP;
10     using Teuchos::rcp;
11     using Teuchos::ParameterList;
12     using Ifpack2::FunctionParameter;
13     using Ifpack2::Hypre::Prec;
14     using Ifpack2::Hypre::Solver;
15
16     typedef Tpetra::CrsMatrix<>::scalar_type Scalar;
17     typedef Tpetra::CrsMatrix<>::local_ordinal_type LO;
18     typedef Tpetra::CrsMatrix<>::global_ordinal_type GO;
19     typedef Tpetra::CrsMatrix<>::node_type Node;
20     typedef Tpetra::DefaultPlatform::DefaultPlatformType Platform;
21     typedef Tpetra::CrsMatrix<Scalar> CrsMatrix;
22     typedef Tpetra::MultiVector<Scalar> MV;
23     typedef Ifpack2::Preconditioner<Scalar> Preconditioner;
24     typedef Tpetra::Map<> Map;
25
26     // Initialize MPI
27     Teuchos::oblackholestream blackhole;
28     Teuchos::GlobalMPISession mpiSession(&argc,&argv,&blackhole);
29     Platform &platform = Tpetra::DefaultPlatform::getDefaultPlatform();
30     RCP<const Teuchos::Comm<int>> comm = platform.getComm();
31
32     // Get parameters from command-line processor
33     int nx = 10;
34     Scalar tol = 1e-6;
35     Teuchos::CommandLineProcessor cmdp(false,true);
36     cmdp.setOption("nx",&nx, "Number of mesh points in x direction.");
37     cmdp.setOption("tolerance",&tol, "Relative residual used for solver.");
38     if(cmdp.parse(argc,argv) != Teuchos::CommandLineProcessor::PARSE_SUCCESSFUL) {
39         return -1;
40     }
```

```

41
42 // Create the 2D Laplace operator
43 int n = nx*nx;
44 RCP<Map> map = rcp(new Map(n,0,comm));
45 RCP<CrsMatrix> A = rcp(new CrsMatrix(map,5));
46 for(L0 i = 0; i<nx; i++) {
47     for(L0 j = 0; j<nx; j++) {
48         GO row = i*nx+j;
49         if(!map->isNodeGlobalElement(row))
50             continue;
51
52         Array<L0> indices;
53         Array<Scalar> values;
54
55         if(i > 0) {
56             indices.push_back(row - nx);
57             values.push_back(-1.0);
58         }
59         if(i < nx-1) {
60             indices.push_back(row + nx);
61             values.push_back(-1.0);
62         }
63         indices.push_back(row);
64         values.push_back(4.0);
65         if(j > 0) {
66             indices.push_back(row-1);
67             values.push_back(-1.0);
68         }
69         if(j < nx-1) {
70             indices.push_back(row+1);
71             values.push_back(-1.0);
72         }
73         A->insertGlobalValues(row,indices,values);
74     }
75 }
76 A->fillComplete();
77
78 // Create the vectors
79 RCP<MV> X = rcp(new MV(A->getRowMap(),1));
80 RCP<MV> B = rcp(new MV(A->getRowMap(),1,false));
81 B->randomize();
82
83 // Create the parameters for hypre
84 RCP<FunctionParameter> functs[10];
85 functs[0] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetPrintLevel, 1)); // print AMG solution info
86 functs[1] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetCoarsenType, 6)); // Falgout coarsening
87 functs[2] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetRelaxType, 6)); // Sym GS/Jacobi hybrid
88 functs[3] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetNumSweeps, 1)); // Sweeps on each level
89 functs[4] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetTol, 0.0)); // Conv tolerance zero
90 functs[5] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetMaxIter, 1)); // Do only one iteration!
91 functs[6] = rcp(new FunctionParameter(Solver, &HYPRE_PCGSetMaxIter, 1000)); // Maximum iterations
92 functs[7] = rcp(new FunctionParameter(Solver, &HYPRE_PCGSetTol, tol)); // Convergence tolerance
93 functs[8] = rcp(new FunctionParameter(Solver, &HYPRE_PCGSetTwoNorm, 1)); // Use the two-norm as the stopping
94 // criteria
95 functs[9] = rcp(new FunctionParameter(Solver, &HYPRE_PCGSetPrintLevel, 2)); // Print solve info
96
97 // Create the hypre solver and preconditioner
98 RCP<Preconditioner> prec = rcp(new Ifpack2::Ifpack2_Hypre<Scalar,L0,GO,Node>(A));
99 ParameterList hypreList;
100 hypreList.set("SolveOrPrecondition", Solver);
101 hypreList.set("Preconditioner", Ifpack2::Hypre::PCG);
102 hypreList.set("SetPreconditioner", true);
103 hypreList.set("NumFunctions", 10);
104 hypreList.set<RCP<FunctionParameter>*>("Functions", functs);
105 prec->setParameters(hypreList);
106 prec->compute();
107
108 // Perform solve
109 prec->apply(*B,*X);
110
111 return 0;
112 }

```

**Lines 1–5** Include statements

**Lines 8–24** Typedefs and using statements to make the code more readable

**Lines 26–30** Set up MPI

**Lines 32–40** Parse command line arguments. This program allows the user to specify how large the problem should be and how accurately the linear system should be solved.

**Lines 42–76** Set up the 2D Laplace operator.

**Lines 78–81** Create a random right-hand-side and initialize the solution vector to 0.

**Lines 83–94** Set hybre options (documented in the hybre user and reference manuals found at [http://computation.llnl.gov/project/linear\\_solvers/software.php](http://computation.llnl.gov/project/linear_solvers/software.php)). In this example, we have elected to use the conjugate gradient method with an algebraic multigrid preconditioner. We have specified a particular coarsening and relaxation type. The most important thing to note about BoomerAMG is that if you would like to use it as a preconditioner, you must set its maximum number of iterations to 1; otherwise, hybre will assume you meant to use it as a linear solver. We then set the tolerance and maximum number of iterations for hybre’s conjugate gradient solver.

**Lines 96–106** Create the hybre solver. Line 99 specifies that we will be using a hybre linear solver, and line 100 says it will be the conjugate gradient method. Line 101 says we would also like to use BoomerAMG. Remember that you must also set “SetPreconditioner” to true, or the preconditioner will not be used. Lines 103 and 104 specify the hybre parameters such as print level, tolerance, and maximum number of iterations.

**Lines 108–109** Solve the linear system. The function “apply” actually calls the hybre linear solve routine PCG with a BoomerAMG preconditioner, as we specified above.

In the next example (Program 2.5), we will examine how to use hybre preconditioners with Belos solvers.

### Program 2.5. Hybre\_BelosEx.cpp

```
1  #include "BelosTpetraAdapter.hpp"
2  #include "BelosPseudoBlockCGSolMgr.hpp"
3  #include "Ifpack2_Preconditioner.hpp"
4  #include "Ifpack2_Hybre.hpp"
5  #include "Teuchos_CommandLineProcessor.hpp"
6  #include "Tpetra_CrsMatrix.hpp"
7  #include "Tpetra_DefaultPlatform.hpp"
8
9  int main(int argc, char *argv[]) {
10     using Teuchos::Array;
11     using Teuchos::RCP;
12     using Teuchos::rcp;
13     using Teuchos::ParameterList;
14     using Ifpack2::FunctionParameter;
15     using Ifpack2::Hybre::Prec;
16
17     typedef Tpetra::CrsMatrix<>::scalar_type Scalar;
18     typedef Tpetra::CrsMatrix<>::local_ordinal_type LO;
19     typedef Tpetra::CrsMatrix<>::global_ordinal_type GO;
20     typedef Tpetra::CrsMatrix<>::node_type Node;
21     typedef Tpetra::DefaultPlatform::DefaultPlatformType Platform;
22     typedef Tpetra::CrsMatrix<Scalar> CrsMatrix;
```

```

23 typedef Tpetra::MultiVector<Scalar> MV;
24 typedef Tpetra::Operator<Scalar> OP;
25 typedef Ifpack2::Preconditioner<Scalar> Preconditioner;
26 typedef Tpetra::Map<> Map;
27 typedef Belos::PseudoBlockCGSolMgr<Scalar,MV,OP> Solver;
28
29 // Initialize MPI
30 Teuchos::oblackholestream blackhole;
31 Teuchos::GlobalMPISession mpiSession(&argc,&argv,&blackhole);
32 Platform &platform = Tpetra::DefaultPlatform::getDefaultPlatform();
33 RCP<const Teuchos::Comm<int>> > comm = platform.getComm();
34
35 // Get parameters from command-line processor
36 int nx = 10;
37 Scalar tol = 1e-6;
38 bool verbose = false;
39 Teuchos::CommandLineProcessor cmdp(false,true);
40 cmdp.setOption("nx",&nx, "Number of mesh points in x direction.");
41 cmdp.setOption("tolerance",&tol, "Relative residual used for solver.");
42 cmdp.setOption("verbose","quiet",&verbose, "Whether to print a lot of info or a little bit.");
43 if(cmdp.parse(argc,argv) != Teuchos::CommandLineProcessor::PARSE_SUCCESSFUL) {
44     return -1;
45 }
46
47 // Create the 2D Laplace operator
48 int n = nx*nx;
49 RCP<Map> map = rcp(new Map(n,0,comm));
50 RCP<CrsMatrix> A = rcp(new CrsMatrix(map,5));
51 for(L0 i = 0; i<nx; i++) {
52     for(L0 j = 0; j<nx; j++) {
53         G0 row = i*nx+j;
54         if(!map->isNodeGlobalElement(row))
55             continue;
56
57         Array<L0> indices;
58         Array<Scalar> values;
59
60         if(i > 0) {
61             indices.push_back(row - nx);
62             values.push_back(-1.0);
63         }
64         if(i < nx-1) {
65             indices.push_back(row + nx);
66             values.push_back(-1.0);
67         }
68         indices.push_back(row);
69         values.push_back(4.0);
70         if(j > 0) {
71             indices.push_back(row-1);
72             values.push_back(-1.0);
73         }
74         if(j < nx-1) {
75             indices.push_back(row+1);
76             values.push_back(-1.0);
77         }
78         A->insertGlobalValues(row,indices,values);
79     }
80 }
81 A->fillComplete();
82
83 // Create the vectors
84 RCP<MV> X = rcp(new MV(A->getRowMap(),1));
85 RCP<MV> B = rcp(new MV(A->getRowMap(),1,false));
86 B->randomize();
87
88 // Create the parameters for hypre
89 RCP<FunctionParameter> functs[6];
90 functs[0] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetPrintLevel, 1)); // print AMG solution info
91 functs[1] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetCoarsenType, 6)); // Falgout coarsening
92 functs[2] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetRelaxType, 6)); // Sym GS/Jacobi hybrid
93 functs[3] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetNumSweeps, 1)); // Sweeps on each level
94 functs[4] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetTol, 0.0)); // Conv tolerance zero
95 functs[5] = rcp(new FunctionParameter(Prec, &HYPRE_BoomerAMGSetMaxIter, 1)); // Do only one iteration!
96
97 // Create the hypre preconditioner
98 RCP<Preconditioner> prec = rcp(new Ifpack2::Ifpack2_Hypre<Scalar,L0,G0,Node>(A));
99 ParameterList hypreList;
100 hypreList.set("SolveOrPrecondition", Prec);
101 hypreList.set("Preconditioner", Ifpack2::Hypre::BoomerAMG);
102 hypreList.set("NumFunctions", 6);
103 hypreList.set<RCP<FunctionParameter>*>("Functions", functs);
104 prec->setParameters(hypreList);
105 prec->compute();
106
107 // Create the linear problem
108 RCP< Belos::LinearProblem<Scalar,MV,OP> > problem = rcp(new Belos::LinearProblem<Scalar,MV,OP>(A,X,B));
109 problem->setHermitian();
110 problem->setLeftPrec(prec);
111 problem->setProblem();
112
113 // Create the Belos linear solver
114 RCP<ParameterList> belosList = rcp(new ParameterList());

```

```

115     beloSList->set("Convergence Tolerance", tol);
116     if(verbose)
117         beloSList->set("Verbosity", Belos::Errors + Belos::Warnings + Belos::TimingDetails + Belos::StatusTestDetails);
118     else
119         beloSList->set("Verbosity", Belos::Errors + Belos::Warnings);
120     RCP<Solver> newSolver = rcp(new Solver(problem,belosList));
121
122     // Perform solve
123     newSolver->solve();
124
125     return 0;
126 }

```

**Lines 1–86** These lines are not substantially different from the previous example. We defined convenient typedefs, then set up our operator, solution vector, and right hand side.

**Lines 88–95** This time, we have elected not to use a hypr linear solver, so we only set the parameters related to the AMG preconditioner. Again, it is very important to set the maximum number of iterations if you wish to use AMG as a preconditioner.

**Lines 97–105** Create the hypr preconditioner. This time, we specify that we would like to precondition rather than solve, since we will be using a Belos linear solver.

**Lines 107–111** Create a Belos::LinearProblem that encapsulates the operator, solution vector, right-hand side, and preconditioner. We also specify that our operator is Hermitian so that Belos allows us to use PCG.

**Lines 113–120** Create a Belos linear solver. The Belos solvers have many parameters, but we only specify the convergence tolerance and verbosity (what information will be printed).

**Lines 122–123** Solve the linear system using a Belos pseudo-block conjugate gradient solver with hypr’s BoomerAMG preconditioner.

## Is the data wrapped or copied?

The Tpetra matrix data is deep-copied to a hypr matrix.

## Trilinos-SuperLU

Trilinos has a Tpetra-based interface to SuperLU-Dist in the sparse factorization package Amesos2. All Amesos2 solvers can be used the following way:

1. Create the solver using the Amesos2 solver factory, which takes as input a string denoting which solver is to be used (KLU2, MUMPS, PARDISO, etc) and the matrix to be factored.
2. Set the parameters for that solver (optional)<sup>4</sup>
3. Perform a symbolic factorization based on the sparsity pattern of the matrix
4. Perform a numeric factorization based on the entries of the matrix. If the matrix's values changed, this factorization must be performed again.
5. Set the initial guess and right-hand side vectors.
6. Solve the linear system. Note that multiple solves can be done without needing to refactor the matrix.

There are numerous examples on the Amesos2 Doxygen page (<https://trilinos.org/docs/dev/packages/amesos2/doc/html/examples.html>) We now present one such example demonstrating how to use this interface to solve a sparse linear system.

### Program 2.6. SuperLU\_Amesos2Ex.cpp

```

1  #include <Teuchos_ScalarTraits.hpp>
2  #include <Teuchos_RCP.hpp>
3  #include <Teuchos_GlobalMPISession.hpp>
4  #include <Teuchos_Tuple.hpp>
5  #include <Teuchos_VerboseObject.hpp>
6  #include <Teuchos_ParameterList.hpp>
7
8  #include <Tpetra_DefaultPlatform.hpp>
9  #include <Tpetra_Map.hpp>
10 #include <Tpetra_MultiVector.hpp>
11 #include <Tpetra_CrsMatrix.hpp>
12
13 #include "Amesos2.hpp"
14 #include "Amesos2_Version.hpp"
15
16
17 int main(int argc, char *argv[]) {
18     typedef double Scalar;
19     typedef Teuchos::ScalarTraits<Scalar>::magnitudeType Magnitude;
20
21     typedef double Scalar;
22     typedef int LO;
23     typedef int GO;
24
25     typedef Tpetra::CrsMatrix<Scalar,LO,GO> MAT;
26     typedef Tpetra::MultiVector<Scalar,LO,GO> MV;
27
28     using Tpetra::global_size_t;
29     using Teuchos::tuple;
30     using Teuchos::RCP;
31     using Teuchos::rcp;
32
33     Teuchos::GlobalMPISession mpiSession(&argc,&argv);
34     Teuchos::RCP<const Teuchos::Comm<int>> comm = Tpetra::DefaultPlatform::getDefaultPlatform().getComm();
35     size_t myRank = comm->getRank();
36
37     RCP<Teuchos::FancyOStream> fos = Teuchos::fancyOStream(Teuchos::rcpFromRef(std::cout));
38     if( myRank == 0 ) *fos << Amesos2::version() << std::endl << std::endl;
39
40     // create a Map
41     global_size_t nrows = 6;
42     RCP<Tpetra::Map<LO,GO>> map = rcp( new Tpetra::Map<LO,GO>(nrows,0,comm) );
43     RCP<MAT> A = rcp( new MAT(map,3) ); // max of three entries in a row
44

```

---

<sup>4</sup>Parameters are documented at [https://trilinos.org/docs/dev/packages/amesos2/doc/html/group\\_\\_amesos2\\_\\_solver\\_\\_parameters.html](https://trilinos.org/docs/dev/packages/amesos2/doc/html/group__amesos2__solver__parameters.html)

```

45  /*
46  * We will solve a system with a known solution, for which we will be using
47  * the following matrix:
48  *
49  * [ [ 7,  0, -3, 0, -1,  0 ]
50  *   [ 2,  8,  0, 0,  0,  0 ]
51  *   [ 0,  0,  1, 0,  0,  0 ]
52  *   [-3,  0,  0, 5,  0,  0 ]
53  *   [ 0, -1,  0, 0,  4,  0 ]
54  *   [ 0,  0,  0, -2,  0,  6 ] ]
55  *
56  */
57  // Construct matrix
58  if( myRank == 0 ){
59      A->insertGlobalValues(0,tuple<G0>(0,2,4),tuple<Scalar>(7,-3,-1));
60      A->insertGlobalValues(1,tuple<G0>(0,1),tuple<Scalar>(2,8));
61      A->insertGlobalValues(2,tuple<G0>(2),tuple<Scalar>(1));
62      A->insertGlobalValues(3,tuple<G0>(0,3),tuple<Scalar>(-3,5));
63      A->insertGlobalValues(4,tuple<G0>(1,4),tuple<Scalar>(-1,4));
64      A->insertGlobalValues(5,tuple<G0>(3,5),tuple<Scalar>(-2,6));
65  }
66  A->fillComplete();
67
68  // Create random X
69  const size_t numVectors = 1;
70  RCP<MV> X = rcp(new MV(map,numVectors));
71  X->randomize();
72
73  /* Create B
74  *
75  * Use RHS:
76  *
77  * [[-7]
78  *  [18]
79  *   [ 3]
80  *  [17]
81  *  [18]
82  *  [28]]
83  */
84  RCP<MV> B = rcp(new MV(map,numVectors));
85  int data[6] = {-7,18,3,17,18,28};
86  for( int i = 0; i < 6; ++i ){
87      if( B->getMap()->isNodeGlobalElement(i) ){
88          B->replaceGlobalValue(i,0,data[i]);
89      }
90  }
91
92  // Check first whether SuperLU is supported
93  if( Amesos2::query("SuperLU_DIST") ){
94
95      // Constructor from Factory
96      RCP<Amesos2::Solver<MAT,MV> > solver = Amesos2::create<MAT,MV>("SuperLU_DIST", A, X, B);
97
98      solver->symbolicFactorization();
99      solver->numericFactorization();
100      solver->solve();
101
102      /* Print the solution
103      *
104      * Should be:
105      *
106      * [[1]
107      *   [2]
108      *   [3]
109      *   [4]
110      *   [5]
111      *   [6]]
112      */
113      X->describe(*fos,Teuchos::VERB_EXTREME);
114  } else {
115      *fos << "SuperLU solver not enabled. Exiting..." << std::endl;
116  }
117 }

```

**Lines 1–31** Include statements and typedefs

**Lines 33–35** Initialize MPI



**Lines 41–66** Create a map describing the parallel distribution of the matrix rows. Then, create the matrix, specifying that each row will have at most three entries. We then set the entries of the matrix by calling `insertGlobalValues`.

**Lines 68–90** Create the initial guess and right-hand side vector. The initial guess  $X$  will be overwritten by the solution computed by `SuperLU_Dist`.

**Lines 92–116** Ask `Amesos2` whether `SuperLU_Dist` has been enabled. If so, create a `SuperLU_Dist` solver using the `Amesos2` solver factory. Note that the interface is the same regardless of the solver; if you wished to use `Amesos2`’s native `KLU2` solver, you would simply replace “`SuperLU_DIST`” in lines 93 and 96 with “`KLU2`”. Perform a symbolic factorization, numeric factorization, then a linear solve.

## PETSc to Other Packages

PETSc uses run-time binding of data structures and solver algorithms which means that in most cases you do not need to change your simulation code to switch between solvers in `hypre`, `SuperLU`, and `Trilinos`. You use the PETSc options database (for example command line arguments) to select your solvers; it is also possible to hardwire particular solvers into the source code but we do not recommend this approach. The solvers in the other xSDK packages are interfaced to PETSc in the PETSc abstract PC preconditioners interface<sup>5</sup>. To select a particular preconditioner one uses

```
-pc_type typename [-pc_* other potential options related to the solver]
```

### PETSc utilizing SuperLU

PETSc can use both `SuperLU` and `SuperLU_Dist` solvers. Since the `SuperLU_Dist` solvers are parallel and in general are faster and require less memory than `SuperLU` we will only discuss the details of utilizing `SuperLU_Dist`. It is utilized with

```
-pc_type lu -pc_factor_mat_solver_package superlu_dist [-mat_superlu_dist_*  
SuperLU_Dist options]
```

All the options will be printed if you run with `-pc_type lu -pc_factor_mat_solver_package superlu_dist -help | grep superlu_dist`. We also display many of the options in Table 2.2.

---

<sup>5</sup>This same interface is used for direct solvers

Option	Arguments	Default	Description
-mat_superlu_dist_statprint	Yes/No	No	Prints statistics on factorization
TODO: Someone needs to with -help and add the rest.			

**Table 2.2.** PETSc SuperLU\_Dist Options

Option	Arguments	Default	Description
-pc_hypre_boomeramg_max_levels	1	Problem dependent	Sets maximum number of levels
TODO: Someone needs to with -help and add the rest.			

**Table 2.3.** PETSc hypre BoomerAMG Options

## PETSc utilizing hypre

PETSc utilizes several of the solvers/preconditioners in hypre via the command

```
-pc_type hypre -pc_hypre_type [boomeramg (default),pilut,parasails,ams,ads]
```

Since BoomerAMG is the crown jewel of hypre we recommend its use whenever possible and do not discuss here the use of the other solvers. Running with `-pc_type hypre -help | grep hypre` for example will show the options for the `pilut` preconditioner. Many of the options are described in Table 2.3.

## PETSc utilizing Trilinos

PETSc can utilize the ML solver in Trilinos.

```
-pc_type ml [-pc_ml_* options for ML]
```

Running with `-pc_type ml -help | grep pc_ml` will show the available options, many of which are detailed in Table 2.4.

TODO: Is this the only interface PETSc has to Trilinos?

Option	Arguments	Default	Description
-pc_ml_maxNlevels	1	10	Sets maximum number of levels
<p>TODO: Someone needs to with -help and add the rest.</p>			

**Table 2.4.** PETSc ML Options





