## Solvers in PyLith

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## How Do We Solve a System of Linear Equations?

For a set of linear equations,



we can get the solution x using

- Gaussian elimination (LU)
- QR factorization and backsolve (QR)
- SVD factorization (SVD)

However, these methods use lots of memory and time.

For any new problem, always begin with a small, serial version which can be solved by LU factorization using the options below:

--petsc.ksp\_type preonly --petsc.pc\_type lu

This eliminates the solver as a variable, so that non-convergent or inaccurate solutions do not complicate interpretation of the model. You can move to slightly larger and parallel problems by adding

--petsc.pc\_factor\_mat\_solver\_package mumps

### How Do We Solve a System of Linear Equations?

We can reduce the problem size using subspace projection, where V is the basis for a small subspace. We solve a small system,  $V^T AV$ , and then project that solution into the full solution space.



A Krylov Method uses matrix vector products with the system matrix A to generate a subspace V in which the problem is solved.

$$V = \operatorname{span} \left\{ b, Ab, A^2b, \dots, A^{k-1}b \right\}$$

GMRES is a Krylov method which generates an approximate solution  $\hat{x}$  that minimizes the residual,  $||b - A\hat{x}||$ , over all choices  $\hat{x} \in V$ .

Krylov methods are not robust solvers. They are generally used to accelerate other methods, called preconditioners,

- Relaxation (Jacobi, Gauss-Seidel, Successive Over-Relaxation)
- Additive Schwarz Method and Block-Jacobi
- Multigrid (MG)
- Incomplete Factorization (Cholesky, LU)

which can stagnate. Thus Krylov solvers are used to make existing solvers more robust.

Schwarz Methods, a particular kind of Domain Decomposition, divide the domain into overlapping subdomains and repeatedly solve the equations over each subdomain. In terms of linear algebra, using only two subdomains, we have

$$\begin{bmatrix} R_0^T A R_0 \\ R_1^T A R_1 \end{bmatrix} \begin{bmatrix} x_0 \\ x_1 \end{bmatrix} = \begin{bmatrix} R_0^T b \\ R_1^T b \\ R_1^T b \end{bmatrix}$$

The solution is repeatedly updated using  $x = x_0 + x_1$ . The Block-Jacobi method is the degenerate case where we have no overlap between subdomains.

- The traditional preconditioners we have described perform poorly for PyLith. They are designed for
- scalar problems,
- which model a single physics,
- and are dominated by local interactions.
- These three limitations must be overcome to achieve efficient, scalable solutions in PyLith.

Elliptic equations, the kind we solve for the quasistatic problem in PyLith, have nonlocal behavior. For example, a small perturbation to a boundary condition changes not just the solution around that part of the boundary, but the entire solution. Prototypical elliptic equations are electrostatics

$$\Delta \phi = -\rho$$

and linear elasticity

$$abla \cdot \left( 
abla \mathbf{u} + 
abla^T \mathbf{u} \right) = -\mathbf{f}$$

Multigrid uses the traditional preconditioners to solve problems on a series of coarse grids. Since coarse grids are small, they can be solved much faster than the original problem. The coarse grid solutions provide global information, which is used to speed up the original solve. However, this requires knowledge of the user's mesh, and the ability to coarsen it.



Algebraic Multigrid (AMG) also constructs coarse problems in order to generate global information. However, it does not use coarse meshes. Instead, AMG builds coarse linear systems using only information from the original system. Thus, it can be applied to any problem, although it is only effective on certain classes of problems.

In PyLith, we use AMG to solve single components of the momentum balance equation using the ML package from Sandia, which can be automatically installed and used through PETSc.

# What is PCFIELDSPLIT?

The FieldSplit preconditioner in PETSc solves individual systems, such as different components of momentum balance, and then combines these solves to produce the overall solution.



By combining traditional preconditioners, effective for simple scalar equations, we can construct a very effective overall solver.

### What is Recommended for PyLith?

Two examples, examples/3d/tet4/step02.cfg and examples/3d/tet4/step04.cfg, use the field-split solver.

| Property                       | Value          | Description  |
|--------------------------------|----------------|--|
| fs_pc_type                     | field_split    | Precondition fields separately   |
| fs_pc_fieldsplit_real_diagonal |                | Use diagonal blocks from the true operator, rather than the preconditioner                                 |
| fs_pc_fieldsplit_type          | multiplicative | Apply each field preconditioning in sequence, which is stronger than all-at-once (additive)                |
| fs_fieldsplit_n_pc_type        | ml             | Multilevel algebraic multigrid pre-<br>conditioning using Trilinos/ML via<br>PETSc, n=0,,N+1               |
| fs_fieldsplit_n_ksp_type       | jacobi         | Jacobi is sometimes the best op-<br>tion for fault preconditioning, how-<br>ever we often use ML, n=0,,N+1 |

Table: PETSc Options which activate PCFIELDSPLIT

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- The PyLith Manual Section on PETSc has a thorough discussion of PETSc solver options for PyLith, as well as default values.
- The PETSc Homepage contains a user manual, FAQ, webpage documentation for each function, and hundreds of example codes.