Theoretical Foundations

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Part I

Introduction

Outline



Hierarchy

Problems

The biggest problem in scientific computing is programmability:

- Lack of usable implementations of modern algorithms
 - Unstructured Multigrid
 - Fast Multipole Method
- Lack of comparison among classes of algorithms
 - Meshes
 - Discretizations
- We should reorient thinking from
 - characterizing the solution (FEM)
 - "what is the convergence rate (in h) of this finite element?"
 - to
 - characterizing the computation (FErari)
 - "how many digits of accuracy per flop for this finite element?"

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Future Compilers

I think compilers are victims of their own success (ala Rob Pike)

Efforts to modularize compilers retain the same primtives

- compiling on the fly (JIT)
- Low Level Virtual Machine
- Raise the level of abstraction
 - Fenics Form Compiler (variational form compiler)
 - Mython (Domain Specific Language generator)





- Spiral Team, http://www.spiral.net
- Uses an intermediate language, SPL, and then generates C
- Works by circumscribing the algorithmic domain

FLAME & FLASH



- Robert van de Geijn, http://www.cs.utexas.edu/users/flame
- FLAME is an Algorithm-By-Blocks interface
- FLASH/SuperMatrix is a runtime system

Divide the work into levels:

- Model
- Algorithm
- Implementation

Divide the work into levels: Spiral Project:

- Algorithm
- Implementation •

Model

- Discrete Fourier Transform (DSP)
- Fast Fourier Transform (SPL)
- C Implementation (SPL Compiler)

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FLAME Project:

- Abstract LA (PME/Invariants)
- Basic LA (FLAME/FLASH)
- Scheduling (SuperMatrix)

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FEniCS Project:

- Navier-Stokes (FFC)
- Finite Element (FIAT)
- Integration/Assembly (FErari)

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- Implementation

Treecodes:

- Kernels with decay (Coulomb)
- Treecodes (PetFMM)
- Scheduling (PetFMM-GPU)

Divide the work into levels:

Treecodes:

• Kernels with decay (Coulomb)

Algorithm

Model

- Treecodes (PetFMM)
- Implementation
 Scheduling (PetFMM-GPU)

Each level demands a strong abstraction layer

Outline





Big Idea: Hierarchy

Multilevel Method

Solve local problems

- Locality of operations is key for efficient implementation
- Should enable reuse of serial implementation

Stitch together to form a global solution

- Manifold or Domain Decomposition idea: local pieces w/ overlap
- Global complexity is encoded in the (small) Overlap

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Hierarchy

Example: Manifold

Manifolds are locally homeomorphic to \mathbb{R}^n :



Transition maps provide a mechanism to connect the pieces.

M. Kneple	ey (AN	L,TTU)
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Example: FEM

The Finite Element Method does computation in a local basis:



The operator \mathcal{T} maps between the local and global bases.

Local (analytical)

- Discretization/Approximation
 - FEM integrals
 - FV fluxes
- Boundary conditions
- Largely dim dependent (e.g. quadrature)

- Data management
 - Sections (local pieces)
 - Completions (assembly)
- Boundary definition
- Multiple meshes
 Mesh hierarchies
- Largely dim independent (e.g. mesh traversal)

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Current algorithms do not efficiently utilize modern machines

- Processor flops are increasing much faster than bandwidth
- Multicore processors are the future
- Optimal multilevel solvers are necessary



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Claim: Hierarchical operations can be handled by a single interface
Why Optimal Algorithms?

- The more powerful the computer, the greater the importance of optimality
- Example:
 - Suppose Alg_1 solves a problem in time CN^2 , N is the input size
 - Suppose Alg₂ solves the same problem in time CN
 - Suppose Alg₁ and Alg₂ are able to use 10,000 processors
- In constant time compared to serial,
 - Alg1 can run a problem 100X larger
 - Alg2 can run a problem 10,000X larger
- Alternatively, filling the machine's memory,
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Sieve Overview

Hierarchy is the centerpiece

- Strip out unneeded complexity (dimension, shape, ...)
- Single relation, covering, handles all hierarchy
 Rich enough for FEM
- Single operation, completion, for parallelism
 - Enforces consistency of the relation

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Doublet Mesh



- Incidence/covering arrows •
- $cone(0) = \{2, 3, 4\}$

Doublet Mesh



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- $star(7) = \{7, 2, 3, 0\}$

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- $meet(0, 1) = \{4\}$

M. Knepley (ANL,TTU)

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Doublet Mesh Distribution



Doublet Mesh Distribution



Doublet Mesh Distribution



Restriction



Localization

- Restrict to patches (here an edge closure)
- Compute locally

Delta



• Delta

- Restrict further to the overlap
- Overlap now carries twice the data

Fusion



- Merge/reconcile data on the overlap
 - Addition (FEM)
 - Replacement (FD)
 - Coordinate transform (Sphere)
 - Linear transform (MG)

Update



• Update

- Update local patch data
- Completion = restrict \longrightarrow fuse \longrightarrow update, in parallel



- FEM accumulating integrals on shared faces
- **FVM** accumulating fluxes on shared cells
- FDM setting values on ghost vertices
 - distributing mesh entities after partition
 - redistributing mesh entities and data for load balance
 - accumlating matvec for a partially assembled matrix



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Benefits

Better mathematical abstractions bring concrete benefits

- Vast reduction in complexity
 - Declarative, rather than imperative, specification
 - Dimension independent code
- Opportunites for optimization
 - Higher level operations missed by traditional compilers
 - Single communication routine to optimize
- Expansion of capabilities
 - Easy model definition
 - Arbitrary elements
 - Complex geometries and embedded boundaries

Part II

Global Computation: Theory

Outline



- 4 Representing Topology
- 5 Representing Functions
- 6 Mapping Interpretation
- Connecting Sieves

Hierarchical Design

Big Idea: Hierarchy

Multilevel Method

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Claim: Hierarchical operations can be handled by a single interface

I will define *optimal* as an $\mathcal{O}(N)$ solution algorithm

These are generally hierarchical, so we need

- hierarchy generation
- assembly on subdomains
- restriction and prolongation

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We begin with a basic covering operation:

We begin with a basic <u>covering</u> operation: cone()

We begin with a basic <u>covering</u> operation: cone() and then add its dual:

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cone()
support()

We begin with a basic <u>covering</u> operation: and then add its dual: followed by the transitive closures:

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We begin with a basic covering operation: cone() and then add its dual: support() followed by the transitive closures: closure(), star()

We begin with a basic covering operation: and then add its dual: followed by the transitive closures: closure(), star() and finally lattice operations:

cone() support()

We begin with a basic <u>covering</u> operation: and then add its dual: followed by the transitive closures: and finally lattice operations:

cone()
support()
closure(),star()
meet(),join()

Sieve Definition

Definition

A <u>Sieve</u> consists of <u>points</u>, and <u>arrows</u>. Each arrow connects a point to another which it covers.

cone(p)	sequence of points which cover a given point p
closure(p)	transitive closure of cone
support(p)	sequence of points which are covered by a given point p
star(p)	transitive closure of support
meet(p,q)	minimal separator of closure(p) and closure(q)
join(p,q)	minimal separator of star(p) and star(q)



- Incidence/covering arrows
- cone(0) = {2,3,4}
 support(7) = {2,3}
 - M. Knepley (ANL,TTU)



- Incidence/covering arrows
- $cone(0) = \{2, 3, 4\}$
 - $support(7) = \{2$



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join(8,9) = {4
 M. Knepley (ANL,TTU)



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The Mesh Dual



(0.4)

3

38/214

Outline



Mesh Distribution

Distributing a mesh means

- distributing the topology (Sieve)
- distributing data (Section)

However, a Sieve can be interpreted as a Section of cone () s!
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Mesh Partition

- 3rd party packages construct a vertex partition
- For FEM, partition dual graph vertices
- For FVM, construct hyperpgraph dual with faces as vertices
- Assign closure (v) and star (v) to same partition

Doublet Mesh Distribution



Doublet Mesh Distribution



Doublet Mesh Distribution



Section Distribution

Section distribution consists of

- Creation of the local Section
- Distribution of the Atlas (layout Section)
- Completion of the Section

Construct local mesh from partition

- Construct partition overlap
- Omplete() the partition section
 - This distributes the cells
- Update Overlap with new points
- Oomplete() the cone section
 - This distributes the remaining sieve points
- Opdate local Sieves

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A simple triangular mesh



Sieve for the mesh



Local sieve on process 0



2D Example

Partition Overlap



Partition Section



Updated Sieve Overlap



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Theory

Cone Section



Distributed Sieve



Coordinate Section



Distributed Coordinate Section



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Theory

Distributed Mesh



A simple hexahedral mesh



3D Example

Sieve for the mesh



Its complicated!

3D Example

Sieve for the mesh



Its complicated!

3D Example

Partition Overlap



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Theory

Partition Section



Distributed Mesh



Notice cells are ghosted

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Sections associate data to submeshes

- Name comes from section of a fiber bundle
 - Generalizes linear algebra paradigm
- **Define** restrict(),update()
- **Define** complete()
- Assembly routines take a Sieve and several Sections
 - This is called a Bundle

Basic Operations

We begin with a simple mapping operation:

Basic Operations

We begin with a simple mapping operation: restrictPoint()

Basic Operations

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updatePoint()

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 updatePoint()
restrictClosure()
 updateClosure()
 complete()



- Need picture of sieve (graph) <-> mesh (picture) maybe doublet
- Show both traversals (closure and restriction), perhaps an animated FEM integral



Section interface

- $restrict(0) = \{f_0\}$
- $restrict(2) = \{v_0\}$
- $restrict(6) = \{e_0, e_1\}$



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• Topological traversals: follow connectivity

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Since we have a single relation,

we can see all our objects merely as mappings:

- Section
 - point \longrightarrow real
- Sieve
 - point of S \longrightarrow {points of S}
- Overlap
 - point of S \longrightarrow {points of S'}

Composition

We may compose mappings to generate

- restrictClosure()
 - closure() restrictPoint()
- updateMeet()
 - meet() updatePoint()

and can even compose across an Overlap

- complete() looks like a
 - restriction to the overlap
 - copy between adjacent sieves
 - fusion of values in the overlap sections
 - update to original section

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Sieves of Sieves

- We can connect two Sieves by identifying points
 - This can be seen as nonlocal covering
- This relation is then encapsulated in an Overlap,
 - which is just another Sieve.
- Sections may be defined over the Overlap
 - Data movement follows the arrows
- Enforcing consistency across an Overlap gives completion ()

Restriction



Localization

- Restrict to patches (here an edge closure)
- Compute locally

Delta



• Delta

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Fusion



- Merge/reconcile data on the overlap
 - Addition (FEM)
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 - Linear transform (MG)

Update



• Update

- Update local patch data
- Completion = restrict \longrightarrow fuse \longrightarrow update, in parallel

Completion



- A ubiquitous parallel form of *restrict* \longrightarrow *fuse* \longrightarrow *update*
- Operates on Sections
 - Sieves can be "downcast" to Sections
- Based on two operations
 - Data exchange through overlap
 - Fusion of shared data



- **FEM** accumulating integrals on shared faces
- **FVM** accumulating fluxes on shared cells
- **FDM** setting values on ghost vertices
 - distributing mesh entities after partition
 - redistributing mesh entities and data for load balance
 - accumlating matvec for a partially assembled matrix



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Part III

Global Computation: Implementation

Outline



Mapping

10 Completion

Optimization and the Sieve Programming Model

12 Finite Elements

13 Boundary Conditions

Global/Local Dichotomy is the Heart of DD Software interfaces do not adequately reflect this

- PETSc DA is too specialized
 - Basically 1D methods applied to Cartesian products
- PETSc Index Sets and VecScatters are too fine
 - User "does everything", no abstraction
- PETSc Linear Algebra (Vec & Mat) is too coarse
 - No access to the underlying connectivity structure

Unstructured Interface (before)

• Explicit references to element type

- getVertices(edgeID), getVertices(faceID)
- getAdjacency(edgeID, VERTEX)
- getAdjacency(edgeID, dim = 0)
- No interface for transitive closure
 - Awkward nested loops to handle different dimensions
- Have to recode for meshes with different
 - o dimension
 - shapes

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 - dimension
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Go Back to the Math

Combinatorial Topology gives us a framework for geometric computing.

• Abstract to a relation, covering, on sieve points

- Points can represent any mesh element
- Covering can be thought of as adjacency
- Relation can be expressed in a DAG (Hasse Diagram)

• Simple query set:

- provides a general API for geometric algorithms
- leads to simpler implementations
- can be more easily optimized

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Unstructured Interface (after)

NO explicit references to element type

- A point may be any mesh element
- getCone(point): adjacent (d-1)-elements
- getSupport(point): adjacent (d+1)-elements
- Transitive closure
 - closure(cell): The computational unit for FEM

• Algorithms independent of mesh

- dimension
- shape (even hybrid)
- global topology
- data layout

Unstructured Interface (after)

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Hierarchy Abstractions

Generalize to a set of linear spaces

- Sieve provides topology, can also model Mat
- Section generalizes Vec
- Spaces interact through an Overlap (just a Sieve)
- Basic operations
 - Restriction to finer subspaces, restrict ()/update()
 - Assembly to the subdomain, complete()
- Allow reuse of geometric and multilevel algorithms



- The Quadtree is a Sieve
 - with optimized operations
- Multipoles are stored in Sections
- Two Overlaps are defined
 - Neighbors
 - Interaction List
- Completion moves data for
 - Neighbors
 - Interaction List



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Interfaces

Multigrid in Sieve



- Sieves represent coarse and fine meshes
- Sections represent coarse and fine fields
- An Overlap matches coarse and fine cells
- Interpolation and restriction are <u>completion</u> over the overlap
 - Fusion is a linear transformation

Outline





- 10 Completion
- Optimization and the Sieve Programming Model
- 12 Finite Elements
- 13 Boundary Conditions

Sequences:

- http://en.wikipedia.org/wiki/Iterator_pattern
- State is held by the iterator
- Special classes are unnecessary

```
const sequence& cells = mesh.heightStratum(0);
for(sequence::iterator c_iter = cells.begin();
   c_iter != cells.end(); ++c_iter) {
   point_type p = *c_iter;
}
```

Visitors:

- http://en.wikipedia.org/wiki/Visitor_pattern
- State is split between sieve and visitor
- User controls allocation

PrintVisitor pV;

sieve.cone(p, pV);

Visitor Composition

- \bullet Visitors can be composed by chaining <code>visit()</code> calls
 - Final template parameter is child visitor type
- closure() is accomplished by composition
 - Oriented traversal uses the variant visit (point, orient)
- Composition can also proceed by <u>slicing</u>
 - Discussed later by Dmitry

Outline



Mapping



Optimization and the Sieve Programming Model

12 Finite Elements

13 Boundary Conditions

Section Distribution

Section distribution consists of

- Creation of the local Section
- Distribution of the Atlas (layout Section)
- Completion of the Section

Completion can be broken into 4 phases:

- 1 restrict() to an overlap section
- Opy() data to the remote overlap section
- Ituse () data with existing point data
- update() remote section with fused overlap section data

It is common to combine phases 1 & 2, and also 3 & 4

 Data is moved directly between communication buffers and storage



Completion









Section Hierarchy

We have a hierarchy of section types of increasing complexity

- GeneralSection
 - An arbitrary number of values for each domain point
 - Constrain arbitrary values
 - Atlas is a UniformSection
- UniformSection
 - A fixed number of values for each domain point
 - Atlas is a ConstantSection
- ConstantSection
 - The same single value for all domain points
 - Only the domain must be completed

- like a power-law (small world) graph,
- MatMult() can be very unbalanced
 - since all edges for a vertex must be on one process.
- We can balance edges in local matrices
 - by leaving the partition boundary unassembled.
- We need only complete () the output section
 - due to the linearity of the operation

If a mesh has a highly graded degree sequence,

• like a power-law (small world) graph,

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Outline





0 Completion

Optimization and the Sieve Programming Model

- Automation
- Parallelism
- Completion
- Interval Sieves

2 Finite Elements

Outline



- Parallelism
- Completion
- Interval Sieves

Kernels Approach

Reducing operations to kernels is widespread in scientific computing:

- Facilitates code reuse
- Reduces code complexity
- Reduces work of optimization (?)
- Needs correct abstractions
- Dual to introducing common software structures
 - Kernels operate on common structures

Must enable automatic selection of algorithmic variants

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Dense Linear Algebra

Dense linear algebra is too rich:

- Rich structure allow many different organizations
- BLAS/LAPACK chooses certain kernel operations
 - Consider only reuse, not optimization
- LAPACK choose a single variant of each algorithm
- LAPACK fixes the structure implementation in the interface

FLAME allows new kernels to be created

Abstracts among implementations (layouts)



Spiral allows both reuse and optimization:

- Abstract model from algorithms
- Allows different implementations for common structures
- Automates algorithm selection
- Incorporates performance feedback

Unfortunately, DFT is simpler than our common operations.



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Unfortunately, DFT is simpler than our common operations.

Sparse linear algebra has a single kernel (SpMV):

- Don't specify our algorithms at the FLAME level
 - Without a PME, cannot move between variants automatically
- Can be built from Sieve completion operations
 - Completion of operator gives assembled matrix
 - Completion of output gives matrix-free application
- VecScatter should be generalized to an Overlap

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Performance Insights

There are two key insights for automatic performance tuning:

- Memory layout controls performance (Goto)
 - Must be able to switch layouts for different algorithmic variants
 - Bad LAPACK interface truncates ATLAS search space
 - Example: GEPP kernel for DGEMM
- 2 Must understand data dependencies
 - OpenMP cannot express this
 - Can be encapsulated in a DAG
 - SuperMatrix
 - Sieve
 - Enables variants switching (loop fusion)

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Sieve and Overlap

Sieve and Overlap can structure computation by expression of

- Hierarchy
 - Reduces complexity and enables code reuse with
 - common components (sieve)
 - operations (completion)
 - Separates global and local concerns
 - Maps well to multiresolution algorithms
- Dependency
 - Allows tranformation between different algorithmic variants
 - Applies at many levels
 - algorithm selection
 - serial scheduling
 - parallel coordination
 - Key advance over Map-Reduce paradigm

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Outline

Optimization and the Sieve Programming Model

- Automation
- Parallelism
- Completion
- Interval Sieves

MPICH-G2

Early Attempt at Hierarchy

Communicator hierarchy, topology depth

Only exposed to the user through Comm attributes
 Still have to support flat model

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Hierarchy in MPI

MPI communicator should be imbued with hierarchy:

- Single relation is easy to add
 - Could be implemented using attributes
- Can easily code hierarchical algorithms
 - FMM, MG, ...
- Can express data dependencies
 - Communicator could represent a thread group
 - Scheduling could be done inside MPI interface (SuperMatrix)
- Enables large and small scale parallelism
 - Domain decomposition
 - Master-slave
- Could be proposed in MPI-3

Outline

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Completion

Completion Optimization

A Section with unchanged structure need not recomplete its Atlas

- The Overlap could store the packing information and buffers
 - A VecScatter could be created between buffers
- For simple fusers, the Overlap maps directly to section storage
 - A VecScatter could be created between the arrays

Outline

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Interval Sieves and Sections

We can demand that our chart be an interval:

- Membership is $\mathcal{O}(1)$
- cone() is $\mathcal{O}(1)$
- restrict() is $\mathcal{O}(1)$

Formerly, all point queries were $O(\log n)$

Moreover, no storage is needed for a search structure:

STL sets require 20 bytes/int

We can always achieve this in a static setting with local renumbering

ISieve

ISieve

- Separate AIJ structures for cones and supports
- Also store AIJ orientations
- Must call allocate () before setting cones
- Some support for dynamic insertion
- Cones and supports unconnected
 - Use symmetrize() to automate arrow reversal
- Has converter from standard Sieve
- Visitors for all traversals

ISection

ISection

- AIJ structure for values
- Same allocate() call before setting values
- Some support for dynamic insertion
- Completion must still send chart explicitly
 - Can amortize across similar completions

Outline

- 8 Interfaces
- Mapping
- 10 Completion
- Optimization and the Sieve Programming Model
- 12 Finite Elements
- Boundary Conditions

```
cells = mesh->heightStratum(0);
for(c = cells->begin(); c != cells->end(); ++c) {
  <Compute cell geometry>
  <Retrieve values from input vector>
  for (q = 0; q < numQuadPoints; ++q) {
    <Transform coordinates>
    for(f = 0; f < numBasisFuncs; ++f) {</pre>
      <Constant term>
      <Linear term>
      <Nonlinear term>
      elemVec[f] *= weight[q]*detJ;
    }
  <Update output vector>
}
<Aggregate updates>
```

```
cells = mesh->heightStratum(0);
for(c = cells->begin(); c != cells->end(); ++c) {
  coords = mesh->restrict(coordinates, c);
  v0, J, invJ, detJ = computeGeometry(coords);
  <Retrieve values from input vector>
  for (q = 0; q < numQuadPoints; ++q) {
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  for (q = 0; q < numQuadPoints; ++q) {
    realCoords = J*refCoords[q] + v0;
    for(f = 0; f < numBasisFuncs; ++f) {</pre>
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M. Knepley (ANL,TTU)

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      elemVec[f] += basis[q,f] *rhsFunc(realCoords);
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Theory

Simula '08

97/214
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      for (d = 0; d < \dim; ++d)
        for(e) testDerReal[d] += invJ[e,d]*basisDer[q,
      for (q = 0; q < numBasisFuncs; ++q) {
        for (d = 0; d < \dim; ++d)
          for(e) basisDerReal[d] += invJ[e,d]*basisDer
          elemMat[f,g] += testDerReal[d] * basisDerReal[
        elemVec[f] += elemMat[f,g]*inputVec[g];
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  }
  mesh->updateAdd(F, c, elemVec);
}
```

<Aggregate updates>

```
cells = mesh->heightStratum(0);
for(c = cells->begin(); c != cells->end(); ++c) {
  <Compute cell geometry>
  <Retrieve values from input vector>
  for (q = 0; q < numQuadPoints; ++q) {
    <Transform coordinates>
    for(f = 0; f < numBasisFuncs; ++f) {</pre>
      <Constant term>
      <Linear term>
      <Nonlinear term>
      elemVec[f] *= weight[q]*detJ;
    }
  <Update output vector>
}
<Aggregate updates>
```

```
cells = mesh->heightStratum(0);
for(c = cells->begin(); c != cells->end(); ++c) {
  <Compute cell geometry>
  <Retrieve values from input vector>
  for (q = 0; q < numQuadPoints; ++q) {
    <Transform coordinates>
    for(f = 0; f < numBasisFuncs; ++f) {</pre>
      <Constant term>
      <Linear term>
      <Nonlinear term>
      elemVec[f] *= weight[q]*detJ;
    }
  <Update output vector>
}
Distribution<Mesh>::completeSection(mesh, F);
```

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Outline

- 8 Interfaces
- Mapping
- Completion
- Optimization and the Sieve Programming Model
- Pinite Elements
- 13 Boundary Conditions

Dirichlet conditions may be expressed as

$$u|_{\Gamma}=g$$

and implemented by constraints on dofs in a Section

• The user provides a function.

Neumann conditions may be expressed as

$$\nabla u \cdot \hat{n}|_{\Gamma} = h$$

and implemented by explicit integration along the boundary

• The user provides a weak form.

Dual Basis Application

We would like the action of a dual basis vector (functional)

$$<\mathcal{N}_i,f>=\int_{\mathrm{ref}}N_i(x)f(x)dV$$

• Projection onto \mathcal{P}

• Code is generated from FIAT specification

- Python code generation package inside PETSc
- Common interface for all elements

Assembly with Dirichlet Conditions

The original equation may be partitioned into

- unknowns in the interior (I)
- unknowns on the boundary (Γ)

so that we obtain

$$\left(\begin{array}{cc}A_{II} & A_{I\Gamma}\\A_{\Gamma I} & A_{\Gamma\Gamma}\end{array}\right)\left(\begin{array}{c}u_{I}\\u_{\Gamma}\end{array}\right)=\left(\begin{array}{c}f_{I}\\f_{\Gamma}\end{array}\right)$$

However u_{Γ} is known, so we may reduce this to

$$A_{II}u_I=f_I-A_{I\Gamma}u_{\Gamma}$$

We will show that our scheme automatically constructs this extra term.

Assembly with Dirichlet Conditions Residual Assembly



Assembly with Dirichlet Conditions Residual Assembly



Assembly with Dirichlet Conditions Residual Assembly



Compute



M. Knepley (ANL,TTU)

Assembly with Dirichlet Conditions Residual Assembly



Compute



Assembly with Dirichlet Conditions Residual Assembly



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Dirichlet Values

- Topological boundary is marked during generation
- Cells bordering boundary are marked using markBoundaryCells()
- To set values:
 - Loop over boundary cells
 - 2 Loop over the element closure
 - Solution For each boundary point *i*, apply the functional N_i to the function *g*
- The functionals are generated with the quadrature information
- Section allocation applies Dirichlet conditions automatically
 - Values are stored in the Section
 - restrict() behaves normally, update() ignores constraints

Complex BC

We may want to constrain a dof not in the global basis:



For instance, no flow normal to a boundary

$$\hat{n} \cdot \mathbf{v} = \mathbf{0}$$

when the global basis follows the coordinate directions.

M. Knepley	(ANL,TTU)
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Complex BC

In order to constrain the value we

- rotate the storage coordinates to the $n-\tau$ frame
- project out the normal coordinate (freeze the value)
- This rotation is also needed for restriction
 - and any action accessing section storage
- In general, we need
 - a transformation to BC coordinates
 - a projection onto free variables (trivial)
- Transformation might involve all element variables
 - which would be an action on the closure

Part IV

Local Computation: Theory

Outline



15 Models of Local Computation

16 Dof Kinds

- Boundary Conditions
- 18 Weak Form Languages

FIAT

Outline

14 FIAT



Models of Local Computation

16 Dof Kinds

- Boundary Conditions
- 18 Weak Form Languages

Form Decomposition

Element integrals are decomposed into <u>analytic</u> and <u>geometric</u> parts:

$$\int_{\mathcal{T}} \nabla \phi_i(\mathbf{x}) \cdot \nabla \phi_j(\mathbf{x}) d\mathbf{x}$$
(1)

$$= \int_{\mathcal{T}} \frac{\partial \phi_i(\mathbf{x})}{\partial x_{\alpha}} \frac{\partial \phi_j(\mathbf{x})}{\partial x_{\alpha}} d\mathbf{x}$$
(2)

$$= \int_{\mathcal{T}_{ref}} \frac{\partial \xi_{\beta}}{\partial x_{\alpha}} \frac{\partial \phi_{i}(\xi)}{\partial \xi_{\beta}} \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} \frac{\partial \phi_{j}(\xi)}{\partial \xi_{\gamma}} |J| d\mathbf{x}$$
(3)

$$= \frac{\partial \xi_{\beta}}{\partial x_{\alpha}} \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} |J| \int_{\mathcal{T}_{ref}} \frac{\partial \phi_i(\xi)}{\partial \xi_{\beta}} \frac{\partial \phi_j(\xi)}{\partial \xi_{\gamma}} d\mathbf{x}$$
(4)
$$= \frac{G^{\beta\gamma}(\mathcal{T}) \mathcal{K}_{\beta\gamma}^{ij}}{G^{\beta\gamma}}$$
(5)

Coefficients are also put into the geometric part.

Form Decomposition

Additional fields give rise to multilinear forms.

$$\int_{\mathcal{T}} \phi_i(\mathbf{x}) \cdot \left(\phi_k(\mathbf{x}) \nabla \phi_j(\mathbf{x}) \right) \, dA \tag{6}$$

$$= \int_{\mathcal{T}} \phi_i^{\beta}(\mathbf{x}) \left(\phi_k^{\alpha}(\mathbf{x}) \frac{\partial \phi_j^{\beta}(\mathbf{x})}{\partial x_{\alpha}} \right) dA$$
(7)

$$= \int_{\mathcal{T}_{ref}} \phi_i^{\beta}(\xi) \phi_k^{\alpha}(\xi) \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} \frac{\partial \phi_i^{\beta}(\xi)}{\partial \xi_{\gamma}} |J| dA$$
(8)

$$= \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} |J| \int_{\mathcal{T}_{ref}} \phi_{i}^{\beta}(\xi) \phi_{k}^{\alpha}(\xi) \frac{\partial \phi_{j}^{\beta}(\xi)}{\partial \xi_{\gamma}} dA \qquad (9)$$
$$= \frac{G^{\alpha \gamma}(\mathcal{T}) \mathcal{K}_{\alpha \gamma}^{ijk}}{G^{\alpha \gamma}(\mathcal{T})} \qquad (10)$$

The index calculus is fully developed by Kirby and Logg in A Compiler for Variational Forms.

Form Decomposition

Isoparametric Jacobians also give rise to multilinear forms

$$\int_{\mathcal{T}} \nabla \phi_i(\mathbf{x}) \cdot \nabla \phi_j(\mathbf{x}) dA$$
(11)

$$= \int_{\mathcal{T}} \frac{\partial \phi_i(\mathbf{x})}{\partial x_{\alpha}} \frac{\partial \phi_j(\mathbf{x})}{\partial x_{\alpha}} dA$$
(12)

$$= \int_{\mathcal{T}_{ref}} \frac{\partial \xi_{\beta}}{\partial x_{\alpha}} \frac{\partial \phi_{i}(\xi)}{\partial \xi_{\beta}} \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} \frac{\partial \phi_{j}(\xi)}{\partial \xi_{\gamma}} |J| dA$$
(13)

$$= |J| \int_{\mathcal{T}_{ref}} \phi_k J_k^{\beta \alpha} \frac{\partial \phi_i(\xi)}{\partial \xi_{\beta}} \phi_l J_l^{\gamma \alpha} \frac{\partial \phi_j(\xi)}{\partial \xi_{\gamma}} dA$$
(14)

$$= J_{k}^{\beta\alpha} J_{l}^{\gamma\alpha} |J| \int_{\mathcal{T}_{ref}} \phi_{k} \frac{\partial \phi_{l}(\xi)}{\partial \xi_{\beta}} \phi_{l} \frac{\partial \phi_{j}(\xi)}{\partial \xi_{\gamma}} dA$$
(15)
$$= G_{kl}^{\beta\gamma}(\mathcal{T}) K_{\beta\gamma}^{ijkl}$$
(16)

A different space could also be used for Jacobians

M. Knepley (ANL,TTU)	Theory	Simula '08	111/214

Outline





16 Dof Kinds

- Boundary Conditions
- 18 Weak Form Languages

Outline



15 Models of Local Computation

10 Dof Kinds

- 1 Boundary Conditions
- 18 Weak Form Languages

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We will show that our scheme automatically constructs this extra term.

Assembly with Dirichlet Conditions Residual Assembly



Assembly with Dirichlet Conditions Residual Assembly



Assembly with Dirichlet Conditions Residual Assembly



Compute



M. Knepley (ANL,TTU)

Assembly with Dirichlet Conditions **Residual Assembly**





Compute



Assembly with Dirichlet Conditions Residual Assembly



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Dirichlet Values

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when the global basis follows the coordinate directions.

M. Knep	lev (Al	NL,TTU)
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Complex BC

In order to constrain the value we

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- Transformation might involve all element variables
 - which would be an action on the closure

14 FIAT

15 Models of Local Computation

16 Dof Kinds

Boundary Conditions



FFC is a compiler for variational forms by Anders Logg.

Here is a mixed-form Poisson equation:

$$a((au, w), (\sigma, u)) = L((au, w)) \qquad orall (au, w) \in V$$

where

$$a((\tau, w), (\sigma, u)) = \int_{\Omega} \tau \sigma - \nabla \cdot \tau u + w \nabla \cdot u \, dx$$
$$L((\tau, w)) = \int_{\Omega} wf \, dx$$

```
shape = "triangle"
BDM1 = FiniteElement("Brezzi-Douglas-Marini",shape,1)
DG0 = FiniteElement("Discontinuous Lagrange",shape,0)
element = BDM1 + DG0
(tau, w) = TestFunctions(element)
(sigma, u) = TrialFunctions(element)
a = (dot(tau, sigma) - div(tau)*u + w*div(sigma))*dx
f = Function(DG0)
L = w*f*dx
```

FFC

Here is a discontinuous Galerkin formulation of the Poisson equation:

$$a(v, u) = L(v) \qquad \forall v \in V$$

where

$$\begin{aligned} a(v,u) &= \int_{\Omega} \nabla u \cdot \nabla v \, dx \\ &+ \sum_{S} \int_{S} -\langle \nabla v \rangle \cdot [[u]]_{n} - [[v]]_{n} \cdot \langle \nabla u \rangle - (\alpha/h) v u \, dS \\ &+ \int_{\partial \Omega} -\nabla v \cdot [[u]]_{n} - [[v]]_{n} \cdot \nabla u - (\gamma/h) v u \, dS \\ L(v) &= \int_{\Omega} v f \, dx \end{aligned}$$

- DG1 = FiniteElement("Discontinuous Lagrange", shape, 1)
- v = TestFunctions(DG1)
- u = TrialFunctions(DG1)
- f = Function (DG1)
- g = Function (DG1)
- n = FacetNormal("triangle")
- h = MeshSize("triangle")
- a = dot(grad(v), grad(u)) * dx
 - dot(avg(grad(v)), jump(u, n))*dS
 - dot(jump(v, n), avg(grad(u))) * dS
 - + alpha/h * dot(jump(v, n) + jump(u, n)) * dS
 - dot(grad(v), jump(u, n))*ds
 - dot(jump(v, n), grad(u))*ds
 - + gamma/h * v * u * ds
- L = v * f * dx + v * g * ds

Part V

Local Computation: Implementation







22 Scheduling and Asynchronous Computation

STREAM Benchmark

Simple benchmark program measuring sustainable memory bandwidth

- Protoypical operation is Triad (WAXPY): $\mathbf{w} = \mathbf{y} + \alpha \mathbf{x}$
- Measures the memory bandwidth bottleneck (much below peak)
- Datasets outstrip cache

Machine	Peak (MF/s)	Triad (MB/s)	MF/MW	Eq. MF/s
Matt's Laptop	1700	1122.4	12.1	93.5 (5.5%)
Intel Core2 Quad	38400	5312.0	57.8	442.7 (1.2%)
Tesla 1060C	984000	102000.0*	77.2	8500.0 (0.8%)

Table: Bandwidth limited machine performance

http://www.cs.virginia.edu/stream/

Analysis of Sparse Matvec (SpMV)

Assumptions

- No cache misses
- No waits on memory references

Notation

- m Number of matrix rows
- nz Number of nonzero matrix elements
 - V Number of vectors to multiply

We can look at bandwidth needed for peak performance

$$\left(8+\frac{2}{V}\right)\frac{m}{nz}+\frac{6}{V}$$
 byte/flop (17)

or achieveable performance given a bandwith BW

$$\frac{Vnz}{(8V+2)m+6nz}BW \text{ Mflop/s}$$
(18)

Towards Realistic Performance Bounds for Implicit CFD Codes, Gropp, Kaushik, Keyes, and Smith.

M. Knepley (ANL,TTU)

Improving Serial Performance

For a single matvec with 3D FD Poisson, Matt's laptop can achieve at most

$$\frac{1}{(8+2)\frac{1}{7}+6} \text{ bytes/flop(1122.4 MB/s)} = 151 \text{ MFlops/s},$$
(19)

which is a dismal 8.8% of peak.

Can improve performance by

- Blocking
- Multiple vectors

but operation issue limitations take over.

Improving Serial Performance

For a single matvec with 3D FD Poisson, Matt's laptop can achieve at most

$$\frac{1}{(8+2)\frac{1}{7}+6}$$
 bytes/flop(1122.4 MB/s) = 151 MFlops/s, (19)

which is a dismal 8.8% of peak.

Better approaches:

- Unassembled operator application (Spectral elements, FMM)
 - N data, N² computation
- Nonlinear evaluation (Picard, FAS, Exact Polynomial Solvers)
 - *N* data, *N^k* computation

Performance Tradeoffs

We must balance storage, bandwidth, and cycles

- Assembled Operator Action
 - Trades cycles and storage for bandwidth in application
- Unassembled Operator Action
 - Trades bandwidth and storage for cycles in application
 - For high orders, storage is impossible
 - Can make use of FErari decomposition to save calculation
 - Could storage element matrices to save cycles
- Partial assembly gives even finer control over tradeoffs
 - Also allows introduction of parallel costs (load balance, ...)



20 FIAT

- Implementation
- Optimization

21 FErari

22 Scheduling and Asynchronous Computation

Finite Element Integrator And Tabulator by Rob Kirby

http://fenicsproject.org/

FIAT

FIAT understands

- Reference element shapes (line, triangle, tetrahedron)
- Quadrature rules
- Polynomial spaces
- Functionals over polynomials (dual spaces)
- Derivatives

Can build arbitrary elements by specifying the Ciarlet triple (K, P, P')

FIAT is part of the FEniCS project

FIAT

Finite Element Integrator And Tabulator by Rob Kirby

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FIAT understands

- Reference element shapes (line, triangle, tetrahedron)
- Quadrature rules
- Polynomial spaces
- Functionals over polynomials (dual spaces)
- Derivatives

Can build arbitrary elements by specifying the Ciarlet triple (K, P, P')

FIAT is part of the FEniCS project

FIAT Integration

The quadrature.fiat file contains:

• An element (usually a family and degree) defined by FIAT

FIAT

A quadrature rule

It is run

- automatically by make, or
- independently by the user

It can take arguments

- -element_family and -element_order, or
- make takes variables ELEMENT and ORDER

Then make produces quadrature.h with:

- Quadrature points and weights
- Basis function and derivative evaluations at the quadrature points
- Integration against dual basis functions over the cell
- Local dofs for Section allocation

M. Knepley (ANL,TTU)

Theory





Optimization



0 FIAT



FErari

- Problem Statement
- Plan of Attack
- Results
- Mixed Integer Linear Programming

22 Scheduling and Asynchronous Computation

Finite Element rearragement to automaically reduce instructions

- Open source implementation http://www.fenics.org/wiki/FErari
- Build tensor blocks $K_{m,m'}^{ij}$ as vectors using FIAT
- Discover dependencies
 - Represented as a DAG
 - Can also use hypergraph model
- Use minimal spanning tree to construct computation



Problem Statement

- Plan of Attack
- Results
- Mixed Integer Linear Programming

Element integrals are decomposed into <u>analytic</u> and <u>geometric</u> parts:

FErari

$$\int_{\mathcal{T}} \nabla \phi_i(\mathbf{x}) \cdot \nabla \phi_j(\mathbf{x}) d\mathbf{x}$$
(20)

Problem Statement

$$= \int_{\mathcal{T}} \frac{\partial \phi_i(\mathbf{x})}{\partial x_{\alpha}} \frac{\partial \phi_j(\mathbf{x})}{\partial x_{\alpha}} d\mathbf{x}$$
(21)

$$= \int_{\mathcal{T}_{ref}} \frac{\partial \xi_{\beta}}{\partial x_{\alpha}} \frac{\partial \phi_{i}(\xi)}{\partial \xi_{\beta}} \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} \frac{\partial \phi_{j}(\xi)}{\partial \xi_{\gamma}} |J| d\mathbf{x}$$
(22)

$$= \frac{\partial \xi_{\beta}}{\partial x_{\alpha}} \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} |J| \int_{\mathcal{T}_{ref}} \frac{\partial \phi_i(\xi)}{\partial \xi_{\beta}} \frac{\partial \phi_j(\xi)}{\partial \xi_{\gamma}} d\mathbf{x}$$
(23)
$$= G^{\beta \gamma}(\mathcal{T}) \mathcal{K}^{ij}_{\beta \gamma}$$
(24)

Coefficients are also put into the geometric part.

Form Decomposition

Additional fields give rise to multilinear forms.

$$\int_{\mathcal{T}} \phi_i(\mathbf{x}) \cdot \left(\phi_k(\mathbf{x}) \nabla \phi_j(\mathbf{x}) \right) \, dA \tag{25}$$

$$= \int_{\mathcal{T}} \phi_i^{\beta}(\mathbf{x}) \left(\phi_k^{\alpha}(\mathbf{x}) \frac{\partial \phi_j^{\beta}(\mathbf{x})}{\partial x_{\alpha}} \right) dA$$
(26)

$$= \int_{\mathcal{T}_{ref}} \phi_i^{\beta}(\xi) \phi_k^{\alpha}(\xi) \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} \frac{\partial \phi_i^{\beta}(\xi)}{\partial \xi_{\gamma}} |J| dA$$
(27)

$$= \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} |J| \int_{\mathcal{T}_{ref}} \phi_{i}^{\beta}(\xi) \phi_{k}^{\alpha}(\xi) \frac{\partial \phi_{j}^{\beta}(\xi)}{\partial \xi_{\gamma}} dA \qquad (28)$$
$$= \frac{G^{\alpha \gamma}(\mathcal{T}) \mathcal{K}_{\alpha \gamma}^{ijk}}{G^{\alpha \gamma}} \qquad (29)$$

The index calculus is fully developed by Kirby and Logg in A Compiler for Variational Forms. Isoparametric Jacobians also give rise to multilinear forms

$$\int_{\mathcal{T}} \nabla \phi_i(\mathbf{x}) \cdot \nabla \phi_j(\mathbf{x}) dA \tag{30}$$

$$= \int_{\mathcal{T}} \frac{\partial \phi_i(\mathbf{x})}{\partial x_{\alpha}} \frac{\partial \phi_j(\mathbf{x})}{\partial x_{\alpha}} dA$$
(31)

$$= \int_{\mathcal{T}_{\text{ref}}} \frac{\partial \xi_{\beta}}{\partial x_{\alpha}} \frac{\partial \phi_{i}(\xi)}{\partial \xi_{\beta}} \frac{\partial \xi_{\gamma}}{\partial x_{\alpha}} \frac{\partial \phi_{j}(\xi)}{\partial \xi_{\gamma}} |J| dA$$
(32)

$$= |J| \int_{\mathcal{T}_{ref}} \phi_k J_k^{\beta \alpha} \frac{\partial \phi_i(\xi)}{\partial \xi_\beta} \phi_l J_l^{\gamma \alpha} \frac{\partial \phi_j(\xi)}{\partial \xi_\gamma} dA$$
(33)

$$= J_{k}^{\beta\alpha} J_{l}^{\gamma\alpha} |J| \int_{\mathcal{T}_{ref}} \phi_{k} \frac{\partial \phi_{l}(\xi)}{\partial \xi_{\beta}} \phi_{l} \frac{\partial \phi_{j}(\xi)}{\partial \xi_{\gamma}} dA \qquad (34)$$
$$= G_{kl}^{\beta\gamma}(\mathcal{T}) K_{\beta\gamma}^{ijkl} \qquad (35)$$

A different space could also be used for Jacobians

M. Knepley	(ANL,TTU)
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Element Matrix Formation

• Element matrix K is now made up of small tensors

FErari

• Contract all tensor elements with each the geometry tensor $G(\mathcal{T})$

Problem Statement

3	0	0	-1	1	1	-4	-4	0	4	0	0
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0
-1	0	0	3	1	1	0	0	4	0	-4	-4
1	0	0	1	3	3	-4	0	0	0	0	-4
1	0	0	1	3	3	-4	0	0	0	0	-4
-4	0	0	0	-4	-4	8	4	0	-4	0	4
-4	0	0	0	0	0	4	8	-4	-8	4	0
0	0	0	4	0	0	0	-4	8	4	-8	-4
4	0	0	0	0	0	-4	-8	4	8	-4	0
0	0	0	-4	0	0	0	4	-8	-4	8	4
0	0	0	-4	-4	-4	4	0	-4	0	4	8

Element Matrix Computation

- Element matrix K can be precomputed
 - FFC
 - SyFi
- Can be extended to nonlinearities and curved geometry
- Many redundancies among tensor elements of *K*
 - Could try to optimize the tensor contraction...

Given vectors $v_i \in \mathbb{R}^m$, minimize $flops(v^Tg)$ for arbitrary $g \in \mathbb{R}^m$

- The set *v_i* is not at all random
- Not a traditional compiler optimization
- How to formulate as an optimization problem?

Plan of Attack

Outline



Problem Statement

Plan of Attack

Results

Mixed Integer Linear Programming

Complexity Reducing Relations

If
$$v_i^T g$$
 is known, is $flops(v_i^T g) < 2m - 1$?

We can use binary relations among the vectors:

Equality

• If
$$v_j = v_i$$
, then $flops(v_j^T g) = 0$

Colinearity

• If
$$v_j = \alpha v_i$$
, then $flops(v_j^T g) = 1$

- Hamming distance
 - If $dist_H(v_j, v_i) = k$, then $flops(v_j^T g) = 2k$

Algorithm for Binary Relations

Construct a weighted graph on v_i

- The weight w(i, j) is $flops(v_i^T g)$ given $v_i^T g$
- With the above relations, the graph is symmetric
- Find a minimum spanning tree
 - Use Prim or Kruskal for worst case $O(n^2 \log n)$
- Traverse the MST, using the appropriate calculation for each edge
 - Roots require a full dot product

Coplanarity

Ternary relation

- If $v_k = \alpha v_i + \beta v_j$, then $flops(v_k^T g) = 3$
- Does not fit our undirected graph paradigm

SVD for vector triples is expensive

- Use a randomized projection into a few \mathbb{R}^3s
- Use a hypergraph?
 - MST algorithm available
- Appeal to geometry?
 - Incidence structures



FErari

- Problem Statement
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Results

Preliminary Results

Order	Entries	Base MAPs	FErari MAPs
1	6	24	7
2	21	84	15
3	55	220	45
4	120	480	176
5	231	924	443
6	406	1624	867



FErari

- Problem Statement
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Modeling the Problem

- Objective is cost of dot products (tensor contractions in FEM)
 Set of vectors V with a given arbitrary vector g
- The original MINLP has a nonconvex, nonlinear objective

EErari

Reformulate to obtain a MILP using auxiliary binary variables
Modeling the Problem

Variables

- $\alpha_{ij} = Basis expansion coefficients$
- y_i = Binary variable indicating membership in the basis
- s_{ij} = Binary variable indicating nonzero coefficient α_{ij}
- z_{ij} = Binary variable linearizes the objective function (equivalent to $y_i y_j$)
- U =Upper bound on coefficients

Constraints

- Eq. (36b) : Basis expansion
- Eq. (36c) : Exclude nonbasis vector from the expansion
- Eq. (36d) : Remove offdiagonal coefficients for basis vectors
 - Eq. (7c) : Exclude vanishing coefficients from cost

Original Formulation

MINLP Model

minimize
$$\sum_{i=1}^{n} \left\{ y_i(2m-1) + (1-y_i) \left(2\sum_{j=1, j\neq i}^{n} y_j - 1 \right) \right\}$$
(36a)
subject to $v_i = \sum_{j=1}^{n} \alpha_{ij} v_j$
 $- Uy_j \le \alpha_{ij} \le Uy_j$
 $i, j = 1, \dots, n$
(36b)
 $i, j = 1, \dots, n$
(36c)

$$- U(1 - y_i) \le \alpha_{ij} \le U(1 - y_i)$$
 $i, j = 1, ..., n,$
(36d)
$$y_i \in \{0, 1\}$$
 $i = 1, ..., n.$

Original Formulation

Equivalent MILP Model: $z_{ii} = y_i \cdot y_i$ minimize $2m \sum_{i=1}^{n} y_i + 2 \sum_{i=1}^{n} \sum_{j=1, j \neq i}^{n} (y_j - z_{ij}) - n$ (36a) subject to $v_i = \sum \alpha_{ij} v_j$ $i=1,\ldots,n$ (36b) $-Uy_i \leq \alpha_{ii} \leq Uy_i$ i, j = 1, ..., n(36c) $i, j = 1, ..., n, i \neq j$ $-U(1-y_i) < \alpha_{ii} < U(1-y_i)$ (36d) $z_{ii} \leq y_i, \quad z_{ij} \leq y_i, \quad z_{ij} \geq y_i + y_j - 1,$ i, j = 1, ..., n(36e) $y_i \in \{0, 1\}, \quad z_{ii} \in \{0, 1\}$ $i, j = 1, \ldots, n.$ 152 / 214 Simula '08 M. Knepley (ANL,TTU) Theory

Sparse Coefficient Formulation

- Take advantage of sparsity of α_{ij} coefficient
- Introduce binary variables s_{ij} to model existence of α_{ij}
- Add constraints $-Us_{ij} \le \alpha_{ij} \le Us_{ij}$

Sparse Coefficient Formulation

MINLP Model minimize $\sum_{i=1}^{n} \left\{ y_i(2m-1) + (1-y_i) \left(2 \sum_{i=1}^{n} s_{ij} - 1 \right) \right\}$ (37a) subject to $v_i = \sum_{i=1}^{n} \alpha_{ij} v_j$ i = 1, ..., n(37b) $-Us_{ii} \leq \alpha_{ii} \leq Us_{ii}$ i, j = 1, ..., n(37c) $-U(1-y_i) \leq \alpha_{ii} \leq U(1-y_i)$ i, j = 1, ..., n(37d) i, j = 1, ..., n $s_{ii} \leq y_i$ (37e) $y_i \in \{0, 1\}, \quad s_{ii} \in \{0, 1\}$ i, j = 1, ..., nSimula '08 153 / 214 M. Knepley (ANL,TTU) Theory

Sparse Coefficient Formulation

Equivalent MILP Model

minimize $2m\sum_{i=1}^{n}y_i+2\sum_{i=1}^{n}\sum_{j=1}^{n}(s_{ij}-z_{ij})-n$ (37a) i=1 $i=1, i\neq i$ subject to $v_i = \sum \alpha_{ij} v_j$ i = 1, ..., n(37b) i, j = 1, ..., n $-Us_{ii} \leq \alpha_{ii} \leq Us_{ii}$ (37c) $i, j = 1, ..., n, i \neq i$ $-U(1-y_i) \leq \alpha_{ii} \leq U(1-y_i)$ (37d) i, j = 1, ..., n $z_{ii} \leq y_i, \quad z_{ii} \leq s_{ii}, \quad z_{ii} \geq y_i + s_{ii} - 1,$ (37e) $y_i \in \{0, 1\}, \quad z_{ii} \in \{0, 1\}, \quad s_{ii} \in \{0, 1\}$ $i, j = 1, \ldots, n$. 071

Initial Formulation

- Initial formulation only had sparsity in the α_{ij}
- MINTO was not able to produce some optimal solutions
 - Report results after 36000 seconds

	Default		MILP		Spa	arse Coef. N	ЛILР
Element	Flops	Flops	LPs	CPU	Flops	LPs	CPU
<i>P</i> ₁ 2D	42	42	33	0.10	34	187	0.43
<i>P</i> ₂ 2D	147	147	2577	37.12	67	6030501	36000
P ₁ 3D	170	166	79	0.49	146	727	3.97
P ₂ 3D	935	935	25283	36000	829	33200	36000

Formulation with Sparse Basis

- We can also take account of the sparsity in the basis vectors
- Count only the flops for nonzero entries
 - Significantly decreases the flop count

	Sparse Coefficient	Sparse Basis
Elements	Flops	Flops
<i>P</i> ₁ 2D	34	12
P ₁ 3D	146	26

Outline







22 Scheduling and Asynchronous Computation

Part VI

Fast Methods

Outline



The Fast Multipole Method

- Spatial Decomposition
- Data Decomposition
- Serial Implementation
- Parallel Spatial Decomposition
- Parallel Performance



Outline



The Fast Multipole Method

- Spatial Decomposition
- Data Decomposition
- Serial Implementation
- Parallel Spatial Decomposition
- Parallel Performance



- The Quadtree is a Sieve
 - with optimized operations
- Multipoles are stored in Sections
- Two Overlaps are defined
 - Neighbors
 - Interaction List
- Completion moves data for
 - Neighbors
 - Interaction List



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Quadtree Implementation

- We use binary scheme to label cells (or vertices)
- Relevant relations can be determined implicitly
 - cone()
 - neighbors
 - parent
 - interaction list
- When vertices are not used, we can directly connect cells
 - cone () becomes neighbor method

Tree Interface

- locateBlob(blob)
 - Locate point in the tree
- fillNeighbors()
 - Compute the neighbor section
- findInteractionList()
 - Compute the interaction list cell section, allocate value section
- fillInteractionList(level)
 - Compute the interaction list value section
- fill(blobs)
 - Compute the blob section
- dump()
 - Produces a verifiable repesentation of the tree

Outline



The Fast Multipole Method

Spatial Decomposition

Data Decomposition

- Serial Implementation
- Parallel Spatial Decomposition
- Parallel Performance

FMM requires data over the Quadtree distributed by:

- box
 - Box centers, Neighbors
- box + neighbors
 - Blobs
- box + interaction list
 - Interaction list cells and values
 - Multipole and local coefficients

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Notice this is multiscale since data is divided at each level

Outline



The Fast Multipole Method

- Spatial Decomposition
- Data Decomposition
- Serial Implementation
- Parallel Spatial Decomposition
- Parallel Performance

Evaluator Interface

• initializeExpansions(tree, blobInfo)

- · Generate multipole expansions on the lowest level
- Requires loop over cells
- *O*(*p*)
- upwardSweep(tree)
 - Translate multipole expansions to intermediate levels
 - Requires loop over cells and children (support)
 - O(p²)
- downwardSweep(tree)
 - Convert multipole to local expansions and translate local expansions on intermediate levels
 - Requires loop over cells and parent (cone)
 - $O(p^2)$

Evaluator Interface

• evaluateBlobs(tree, blobInfo)

- Evaluate direct and local field interactions on lowest level
- Requires loop over cells and neighbors (in section)
- *O*(*p*²)
- evaluate(tree, blobs, blobInfo)
 - Calculate the complete interaction (multipole + direct)

Kernel Interface

Method	Description
P2M(t)	Multipole expansion coefficients
L2P(t)	Local expansion coefficients
M2M(t)	Multipole-to-multipole translation
M2L(t)	Multipole-to-local translation
L2L(t)	Local-to-local translation
evaluate(blobs)	Direct interaction

- Evaluator is templated over Kernel
- There are alternative kernel-independent methods
 - kifmm3d

Outline



The Fast Multipole Method

- Spatial Decomposition
- Data Decomposition
- Serial Implementation

Parallel Spatial Decomposition

Parallel Performance

Parallel Tree Implementation

- Divide tree into a root and local trees
- Distribute local trees among processes
- Provide communication pattern for local sections (overlap)
 - Both neighbor and interaction list overlaps
 - Sieve generates MPI from high level description

Parallel Tree Implementation

How should we distribute trees?

- Multiple local trees per process allows good load balance
- Partition weighted graph
 - Minimize load imbalance and communication
 - Computation estimate:

Leaf $N_i p$ (P2M) + $n_i p^2$ (M2L) + $N_i p$ (L2P) + $3^d N_i^2$ (P2P) Interior $n_c p^2$ (M2M) + $n_i p^2$ (M2L) + $n_c p^2$ (L2L)

• Communication estimate:

Diagonal $n_c(L-k-1)$ Lateral $2^{d} \frac{2^{m(L-k-1)}-1}{2^m-1}$ for incidence dimesion *m*

Leverage existing work on graph partitioning

ParMetis

Parallel Tree Implementation Why should a good partition exist?

Shang-hua Teng, Provably good partitioning and load balancing algorithms for parallel adaptive N-body simulation, SIAM J. Sci. Comput., **19**(2), 1998.

- Good partitions exist for non-uniform distributions
 2D O (√n(log n)^{3/2}) edgecut
 3D O (n^{2/3}(log n)^{4/3}) edgecut
- As scalable as regular grids
- As efficient as uniform distributions
- ParMetis will find a nearly optimal partition

Parallel Tree Implementation Will ParMetis find it?

George Karypis and Vipin Kumar, Analysis of Multilevel Graph Partitioning, Supercomputing, 1995.

- Good partitions exist for non-uniform distributions 2D $C_i = 1.24^i C_0$ for random matching 3D $C_i = 1.21^i C_0$?? for random matching
- 3D proof needs assurance that averge degree does not increase
- Efficient in practice

The Fast Multipole Method Parallel Spatial Decomposition

Parallel Tree Implementation Advantages

Simplicity

• Complete serial code reuse

Provably good performance and scalability
The Fast Multipole Method Parallel Spatial Decomposition

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The Fast Multipole Method Parallel Spatial Decomposition

Parallel Tree Implementation

- Simplicity
- Complete serial code reuse
- Provably good performance and scalability

Parallel Tree Interface

- fillNeighbors()
 - Compute neighbor overlap, and send neighbors
- findInteractionList()
 - Compute the interaction list overlap
- fillInteractionList(level)
 - Complete and copy into local interaction sections
- fill(blobs)
 - Now must scatter blobs to local trees
 - Uses scatterBlobs() and gatherBlobs()

Parallel Data Movement

Complete neighbor section

Opward sweep

- Upward sweep on local trees
- Ø Gather to root tree
- Opward sweep on root tree
- Omplete interaction list section
- Oownward sweep
 - Downward sweep on root tree
 - ② Scatter to local trees
 - Ownward sweep on local trees

Parallel Evaluator Interface

• initializeExpansions(local trees, blobInfo)

- Evaluate each local tree
- upwardSweep(local trees, partition, root tree)
 - Evaluate each local tree and then gather to root tree
- downwardSweep(local trees, partition, root tree)
 - Scatter from root tree and then evaluate each local tree
- evaluateBlobs(local trees, blobInfo)
 - Evaluate on all local trees
- evaluate(tree, blobs, blobInfo)
 - Identical

Outline



The Fast Multipole Method

- Spatial Decomposition
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Recursive Parallel

- For large problems, a single root can be a bottleneck
- We can recursively assign roots to subtrees
 - Bandwidth to root is controlled
 - What about utilization?
- Root computation is similar to MG coarse solve

Outline





Multigrid

- Structured
- Unstructured





Structured

A DMDA is more than a Mesh

A DMDA contains topology, geometry, and (sometimes) an implicit Q1 discretization.

It is used as a template to create

- Vectors (functions)
- Matrices (linear operators)

DMDA Global vs. Local Numbering

- Global: Each vertex has a unique id belongs on a unique process
- Local: Numbering includes vertices from neighboring processes
 - These are called ghost vertices

F	Proc 2	Proc 3				
Х	Х	Х	Х	Х		
Х	Х	Х	Х	Х		
12	13	14	15	Х		
8	9	10	11	Х		
4	5	6	7	X		
0	1	2	3	Х		
Proc 0			Proc 1			
Local numbering						

Proc 2			Proc 3			
21	22	23	28	29		
18	19	20	26	27		
15	16	17	24	25		
6	7	8	13	14		
3	4	5	11	12		
0	1	2	9	10		
Proc 0			Proc 1			
Global numbering						

DMDA Local Function

User provided function calculates the nonlinear residual (in 2D)

(* If)(DMDALocalInfo *info, PetscScalar**x, PetscScalar **r, void *ctx)

info: All layout and numbering information

- x: The current solution (a multidimensional array)
- r: The residual
- ctx: The user context passed to DMDASNESSetFunctionLocal()

The local DMDA function is activated by calling

DMDASNESSetFunctionLocal(dm, INSERT_VALUES, lfunc, &ctx)

Structured

Bratu Residual Evaluation

 $\Delta u + \lambda e^u = \mathbf{0}$

ResLocal(DMDALocalInfo *info, PetscScalar **x, PetscScalar **f, void *ctx)
for(j = info->ys; j < info->ys+info->ym; ++j) {
 for(i = info->xs; i < info->xs+info->xm; ++i) {
 u = x[j][i];
 if (i==0 || j==0 || i == M || j == N) {
 f[j][i] = 2.0*(hydhx+hxdhy)*u; continue;
 }
 u_xx = (2.0*u - x[j][i-1] - x[j][i+1])*hydhx;
 u_yy = (2.0*u - x[j-1][i] - x[j+1][i])*hxdhy;
 f[j][i] = u_xx + u_yy - hx*hy*lambda*exp(u);
}}

\$PETSC_DIR/src/snes/examples/tutorials/ex5.c

User provided function calculates the Jacobian (in 2D)

(* ljac)(DMDALocalInfo *info, PetscScalar**x, Mat J, void *ctx)

info: All layout and numbering information

- x: The current solution
- J: The Jacobian
- ctx: The user context passed to DASetLocalJacobian()

The local DMDA function is activated by calling

DMDASNESSetJacobianLocal(dm, ljac, &ctx)

Structured

Setting Values on Regular Grids

PETSc provides

MatSetValuesStencil(Mat A, m, MatStencil idxm[], n, MatStencil idxn[], PetscScalar values[], InsertMode mode)

- Each row or column is actually a MatStencil
 - This specifies grid coordinates and a component if necessary
 - Can imagine for unstructured grids, they are vertices
- The values are the same logically dense block in row/col

Two-step process enables overlapping computation and communication

- DMGlobalToLocalBegin(da, gvec, mode, lvec)
 - gvec provides the data
 - mode is either INSERT_VALUES or ADD_VALUES
 - lvec holds the local and ghost values
- DMGlobalToLocalEnd(da, gvec, mode, lvec)
 - Finishes the communication

The process can be reversed with DALocalToGlobalBegin/End().

DM Integration with SNES

DM supplies global residual and Jacobian to SNES

- User supplies local version to DM
- The \mathtt{Rhs}_\star () and \mathtt{Jac}_\star () functions in the example
- Allows automatic parallelism
- Allows grid hierarchy
 - Enables multigrid once interpolation/restriction is defined
- Paradigm is developed in unstructured work
 - Solve needs scatter into contiguous global vectors (initial guess)
- Handle Neumann BC using KSPSetNullSpace()

Multigrid with DM

Allows multigrid with some simple command line options

- -pc_type mg, -pc_mg_levels
- -pc_mg_type, -pc_mg_cycle_type, -pc_mg_galerkin
- -mg_levels_1_ksp_type, -mg_levels_1_pc_type
- -mg_coarse_ksp_type, -mg_coarse_pc_type
- -da_refine, -ksp_view

Interface also works with GAMG and 3rd party packages like ML

Outline



Unstructured

Unstructured Meshes

- Same DMMG options as the structured case
- Mesh refinement
 - Ruppert algorithm in Triangle and TetGen
- Mesh coarsening
 - Talmor-Miller algorithm in PETSc
- More advanced options
 - -dmmg_refine
 - -dmmg_hierarchy
- Current version only works for linear elements

A Priori refinement

For the Poisson problem, meshes with reentrant corners have a length-scale requirement in order to maintain accuracy:

$$egin{aligned} C_{\mathit{low}} r^{1-\mu} &\leq h \leq C_{\mathit{high}} r^{1-\mu} \ \mu &\leq rac{\pi}{ heta} \end{aligned}$$



Multigrid Unstructured

The Folly of Uniform Refinement

uniform refinement may fail to eliminate error



M. Knepley (ANL,TTU)

Geometric Multigrid

- We allow the user to refine for fidelity
- Coarse grids are created automatically
- Could make use of AMG interpolation schemes



Requirements of Geometric Multigrid

Sufficient conditions for optimal-order convergence:

- $|M_c| < 2|M_f|$ in terms of cells
- any cell in *M_c* overlaps a bounded # of cells in *M_f*
- monotonic increase in cell length-scale

• Each *M_k* satisfies the **quasi-uniformity** condition:

 $C_1 h_k \leq h_K \leq C_2 \rho_K$

- h_K is the length-scale (longest edge) of any cell K
- h_k is the maximum length-scale in the mesh M_k
- ρ_K is the diameter of the inscribed ball in K

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Function Based Coarsening

- (Miller, Talmor, Teng; 1997)
- triangulated planar graphs \equiv disk-packings (Koebe; 1934)
- define a spacing function S() over the vertices
- obvious one: $S(v) = \frac{dist(NN(v),v)}{2}$



• pick a subset of the vertices such that $\beta(S(v) + S(w)) > dist(v, w)$

Unstructured

Multiarid

- for all $v, w \in M$, with $\beta > 1$
- dimension independent
- provides guarantees on the size/quality of the resulting meshes



Loop over the vertices

• include a vertex in the new mesh

- remove colliding adjacent vertices from the mesh
- remesh links of removed vertices
- repeat until no vertices are removed.
- Eventually we have that
 - every vertex is either included or removed
 - bounded degree mesh $\Rightarrow O(n)$ time

- Iocal Delaunay remeshing can be done in 2D and 3D
- faster to connect edges and remesh later

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- Eventually we have that
 - every vertex is either included or removed
 - bounded degree mesh $\Rightarrow O(n)$ time
- Remeshing may be performed either during or after coarsening
 - local Delaunay remeshing can be done in 2D and 3D
 - faster to connect edges and remesh later

Implementation in *Sieve* Peter Brune, 2008

- vertex neighbors: $cone(support(v)) \setminus v$
- vertex link: $closure(star(v)) \setminus star(closure(v))$
- connectivity graph induced by limiting sieve depth
- remeshing can be handled as local modifications on the sieve
- meshing operations, such as cone construction easy



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3D Test Problem

- Reentrant corner
- $-\Delta u = f$
- $f(x, y, z) = 3\sin(x + y + z)$
- Exact Solution: $u(x, y, z) = \sin(x + y + z)$



Multigrid Unstructured

GMG Performance

Linear solver iterates are nearly as system size increases:



KSP Iterates on Reentrant Domains

M. Knepley (ANL,TTU)

GMG Performance

Coarsening work is nearly constant as system size increases:



Vertex Comparisons on Reentrant Domains

Quality Experiments

Table: Hierarchy quality metrics - 2D

Pacman Mesh, $\beta = 1.45$						
level	cells	vertices	$\frac{\min(h_{\kappa})}{h_{k}}$	max $\frac{h_{\kappa}}{\rho_{k}}$	$\min(h_{\mathcal{K}})$	max. overlap
0	19927	10149	0.020451	4.134135	0.001305	-
1	5297	2731	0.016971	4.435928	0.002094	23
2	3028	1572	0.014506	4.295703	0.002603	14
3	1628	856	0.014797	5.295322	0.003339	14
4	863	464	0.011375	6.403574	0.003339	14
5	449	250	0.022317	6.330512	0.007979	13

Part VII

Sample Application: Fault Mechanics



26 Mesh Handling

27 Parallelism

28 Fault Handling

29 Coupling

Formulation

Reverse-slip Benchmark





26 Mesh Handling

7 Parallelism



29 Coupling

Mesh Handling

Multiple Mesh Types







26 Mesh Handling



28 Fault Handling

29 Coupling

Parallelism

- Function and Operator Assembly
 - Parallel element integration over multiple materials/models
 - Assembly uses completion for functions and PETSc Mat for operators
- Algebraic sovlers
 - Use MUMPS for small problems
 - PETSc ASM/ILU for large problems
 - Hope to use unstructured MG when fault support is implemented
- Parallel data movement routines do not change for
 - Different dimension
 - Different element shapes
 - Different discretization
 - Fault inclusion



- 26 Mesh Handling
- 27 Parallelism



29 Coupling

Fault Handling

Cohesive Cells



Cohesive cells are used to enforce slip conditions on a fault

- Demand complex mesh manipulation
 - We allow specification of only fault vertices
 - Must "sew" together on output
- Use Lagrange multipliers to enforce constraints
 - Forces illuminate physics
- Allow different fault constitutive models
 - Simplest is enforced slip
 - Now have fault constitutive models

• In order to create a fault, the generator provides

a set of fault vertices, or

a set of fault faces.

• Fault vertices, unlike fault faces, must be

- o combined into faces on a fault mesh, and
- oriented

• The fault mesh is used to

- split vertices along the fault
- introduce prism elements between adjacent fault faces
- Sieve code works for
 - any dimension
 - any element shape

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- 25 Formulation
- 26 Mesh Handling
- 27 Parallelism
- 28 Fault Handling

