



Principles of compatible discretizations

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RWTH Aachen, November 12th, 2008
Chair for Computational Analysis of Technical Systems

Supported in part by



Sandia is a multiprogram laboratory operated by Sandia Corporation, a Lockheed Martin Company,
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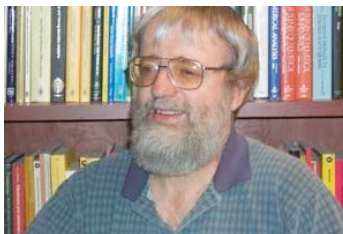
Part I

Where we learn about computational modeling, discretization of PDEs, and develop two simple discrete models (with mixed success)

Collaborators

Max Gunzburger

Florida State University
Tallahassee



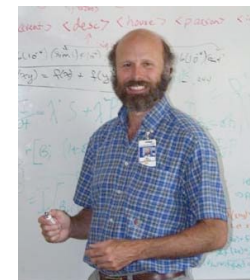
Misha Shashkov

Los Alamos National
Laboratory



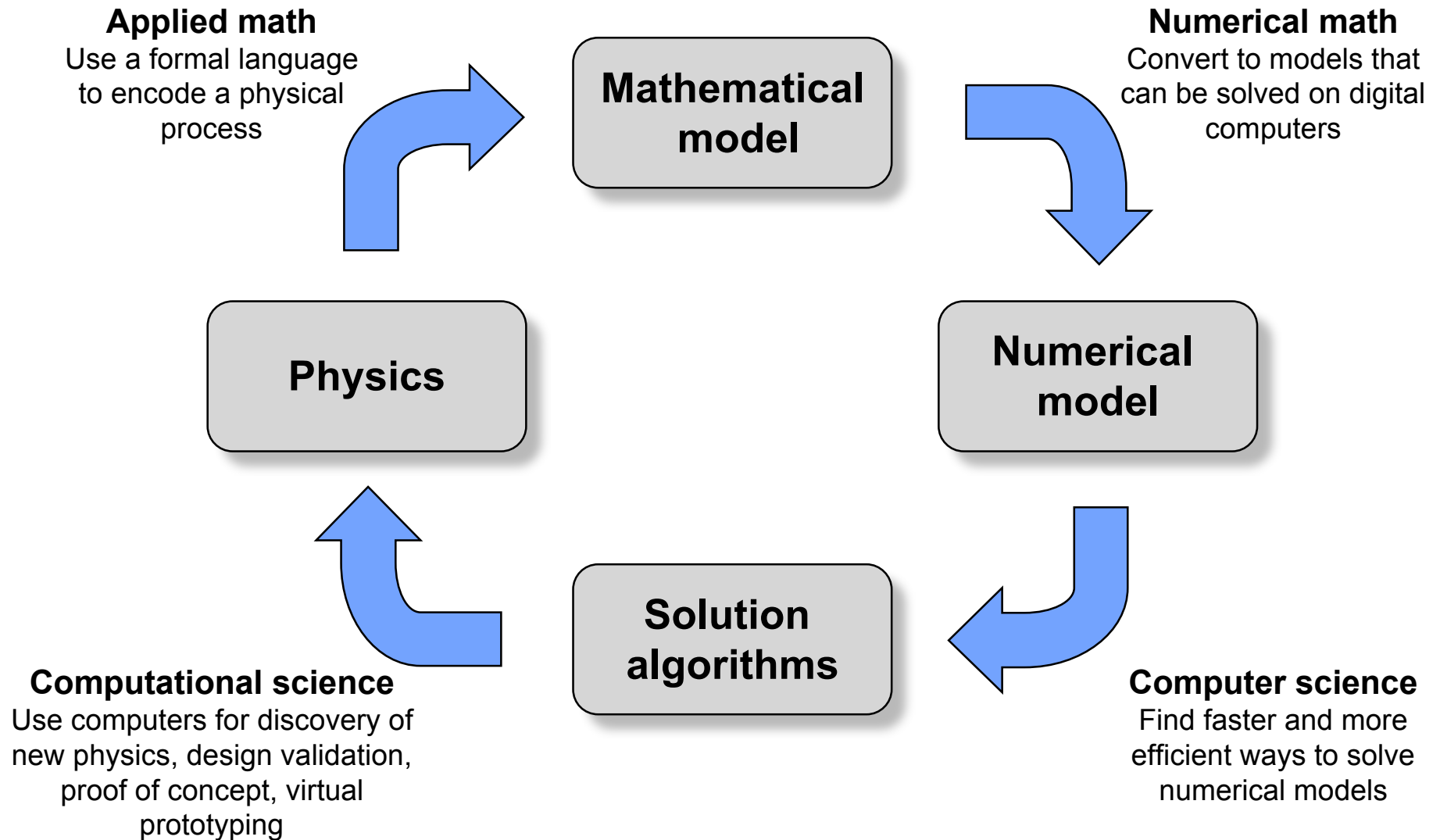
Mac Hyman

Los Alamos National
Laboratory





What is this talk about?





Focus on Numerical Math

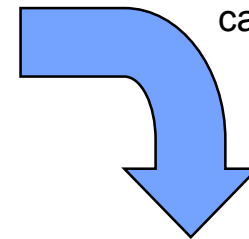
**Discretization
=
Model reduction**

$$\mathcal{A}u = f$$

mathematical model

Numerical math

Convert to models that
can be solved on digital
computers



$$\mathbf{A}_h \mathbf{u}_h = \mathbf{f}_h$$

**a parameterized family
of algebraic equations**

1. Is the sequence of algebraic equations well-behaved?

- are all problems **uniquely** and **stably** (in h) solvable?
- do solutions **converge** to the exact solutions as $h \rightarrow 0$?

2. Are physical and discrete models compatible?

- are solutions **physically** meaningful
- do they **mimic**, e.g., **invariants**, **symmetries**,
or **involutions** of actual states

3. How to make a compatible & accurate discretization?

- how to **choose** the **variables** and where to **place** them;
- how to avoid **spurious** solutions.



A toy problem

Boundary value problem

$$\begin{aligned}
 & -ay'' + by' + cy = f \quad \text{in } (0,1) \\
 & y(0) = y(1) = 0
 \end{aligned}$$

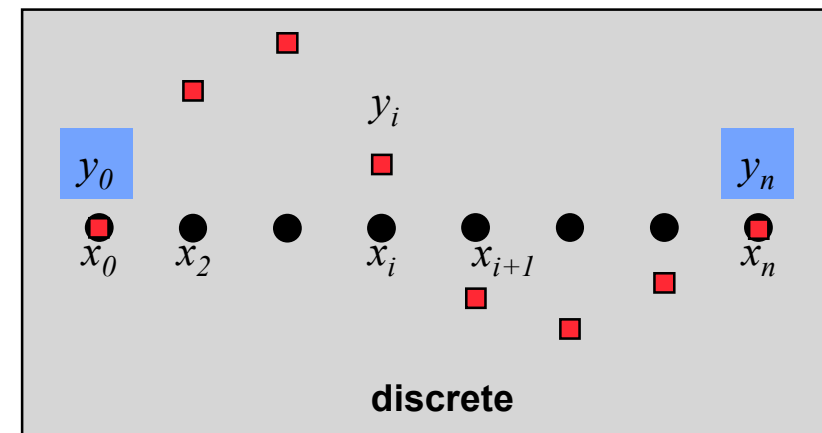
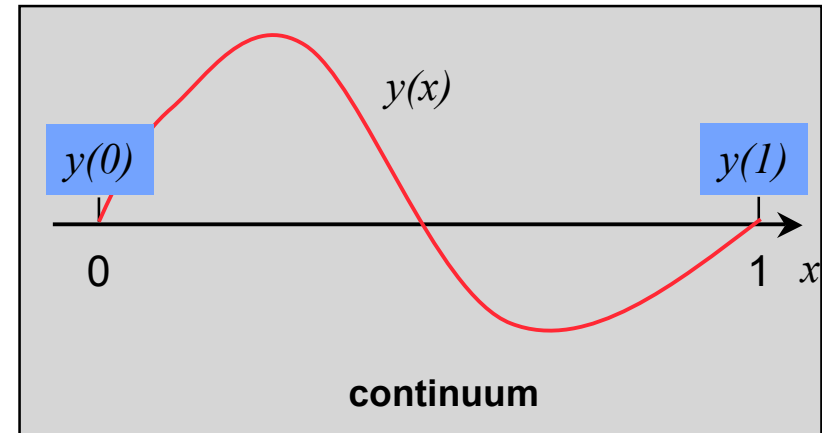
$$y''(x_i) \approx \frac{y_{i+1} - 2y_i + y_{i-1}}{h^2}$$

$$y(x_i) \approx \frac{y_{i+1} - y_{i-1}}{2h}$$

$$-a_i \frac{y_{i+1} - 2y_i + y_{i-1}}{h^2} + b_i \frac{y_{i+1} - y_{i-1}}{2h} + c_i y_i = f_i$$

$$y_0 = y_n = 0$$

Parameterized linear system





Porous Media Flow

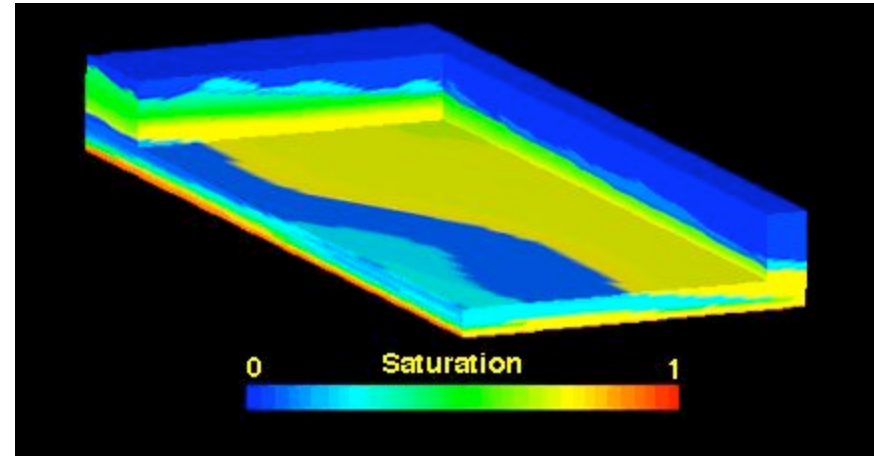
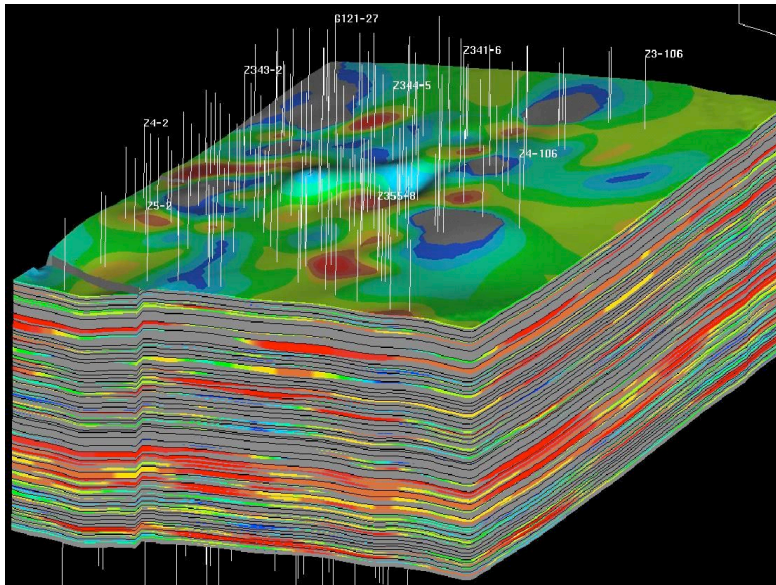
Boundary value problem

$$\begin{aligned} -\nabla \cdot \rho \nabla p &= f \text{ in } \Omega \\ p &= 0 \text{ on } \Gamma \end{aligned}$$

f - source term

p - pressure

ρ - permeability tensor



Steady state saturation in a site scale model of Yucca Mountain, Nevada. Model area is 1.7X4.2 square miles. Calculations like this are used for evaluating the suitability of **Yucca Mountain** as a potential repository for **high level nuclear waste**. *Courtesy LANL EES Division.*

Reservoir simulation can be used to forecast the **production of oil and gas fields**, optimize reservoir development, and evaluate the distribution of remaining oil. It is an **important tool** to improve the design of wells, the efficiency of reservoirs, and enhance oil and gas recovery. *Courtesy Prof. J. Chen, SMU*



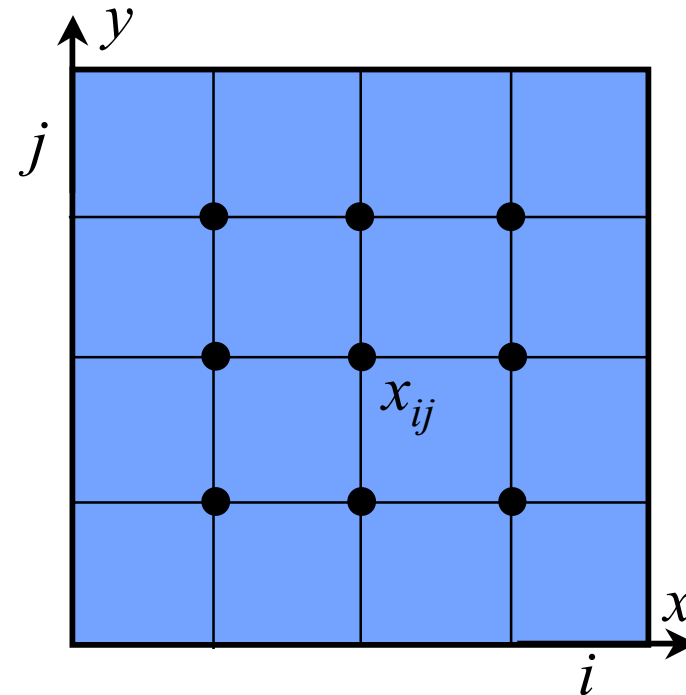
Darcy problem

In such applications, **velocity** \mathbf{u} , rather than the **pressure** p is the variable of primary interest, and direct approximation is desirable.

Equivalent 1st order form

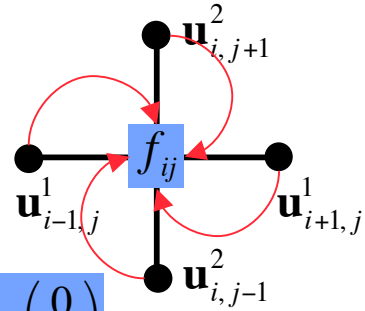
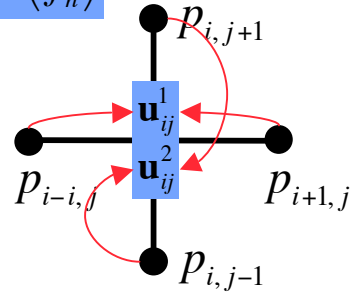
$$\begin{aligned}\nabla \cdot \rho \mathbf{u} &= f \text{ in } \Omega \\ \mathbf{u} + \nabla p &= 0 \text{ on } \Omega \\ p &= 0 \text{ on } \Gamma\end{aligned}$$

Discretized domain





PDE→Parameterized Linear System

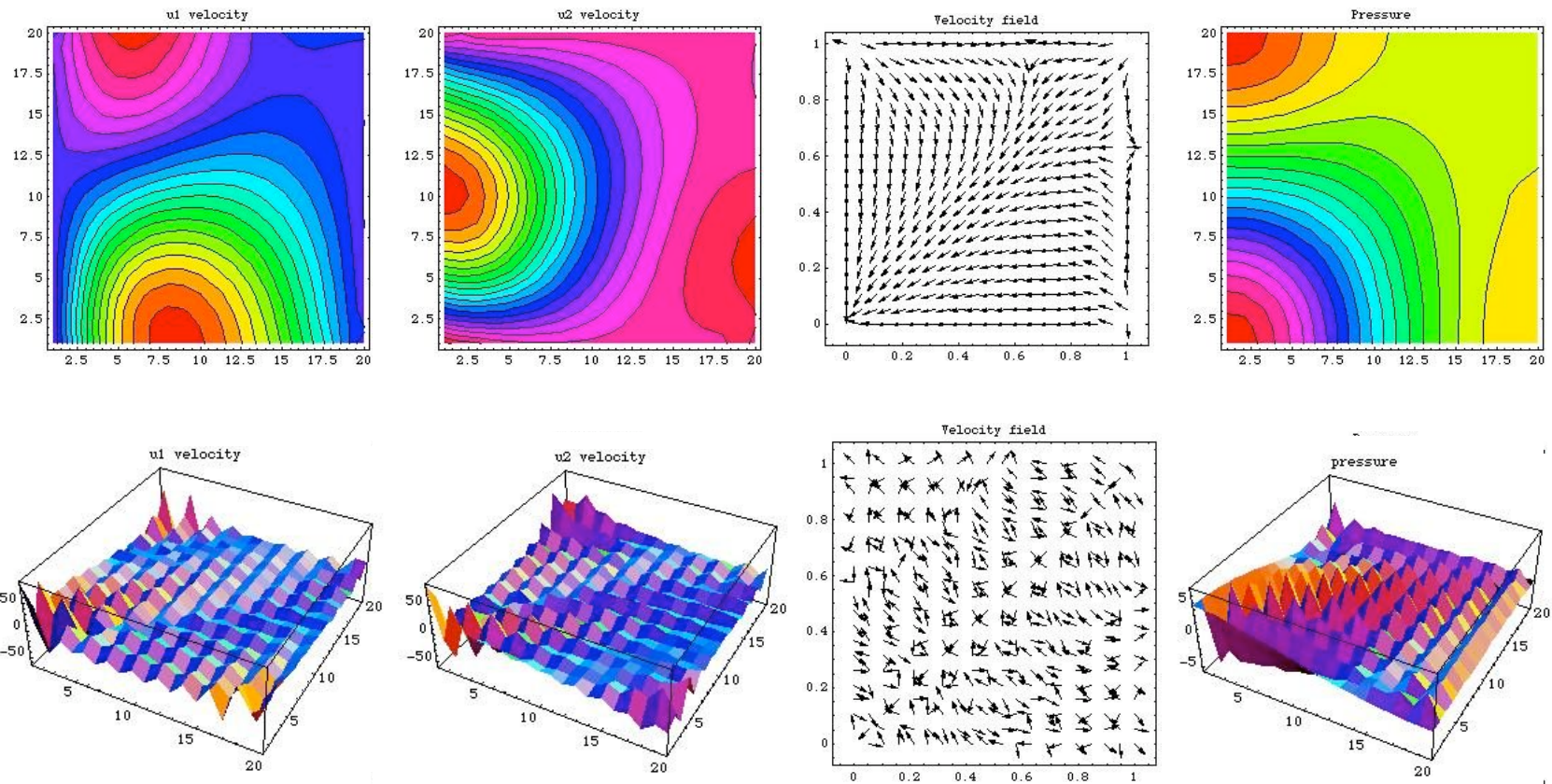
Continuum	Discrete	Stencil
$\nabla \cdot \rho \mathbf{u} \equiv \mathbf{u}_x^1 + \mathbf{u}_y^2 = f$	$\frac{\mathbf{u}_{i+1,j}^1 - \mathbf{u}_{i-1,j}^1}{2h} + \frac{\mathbf{u}_{i,j+1}^2 - \mathbf{u}_{i,j-1}^2}{2h} = f_{ij}$	
$\left\{ \begin{array}{l} \mathbf{u} + \nabla p = 0 \\ \nabla \cdot \mathbf{u} = f \end{array} \right\} \rightarrow \left\{ \begin{array}{l} A_h \mathbf{u}_h + B_h^T p_h = 0 \\ B_h \mathbf{u}_h = f_h \end{array} \right. \rightarrow \begin{pmatrix} A_h & B_h^T \\ B_h & 0 \end{pmatrix} \begin{pmatrix} \mathbf{u}_h \\ p_h \end{pmatrix} = \begin{pmatrix} 0 \\ f_h \end{pmatrix}$		
$\mathbf{u}^1 + p_x = 0$	$\mathbf{u}_{ij}^1 + \frac{p_{i+1,j} - p_{i-1,j}}{2h} = 0$	
$\mathbf{u}^2 + p_y = 0$	$\mathbf{u}_{ij}^2 + \frac{p_{i,j+1} - p_{i,j-1}}{2h} = 0$	

Collocated discretization: variables share same grid location

Divergence and **gradient** discretized by the **same stencil**



Computational example



Complete DISASTER (💀 💀 💀)
Our discrete model is **incompatible**

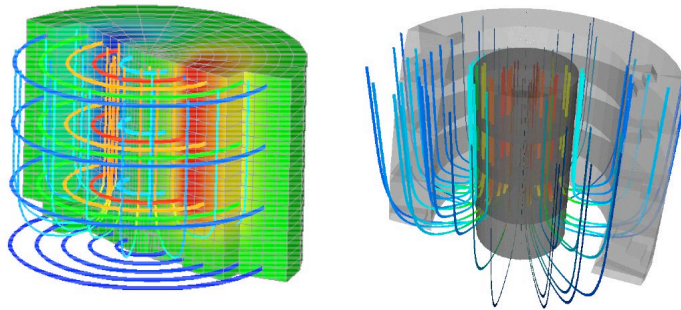


Coupled multiscale, nonlinear physics is even more challenging: Z-Pinch simulation in ALEGRA

Scales:

PULSE DURATION	10^{-9} sec
TIME SCALE	10^{-3} sec
CURRENT POWER	20×10^6 A
X-RAY POWER	10^{12} W
X-RAY ENERGY	1.9×10^6 J

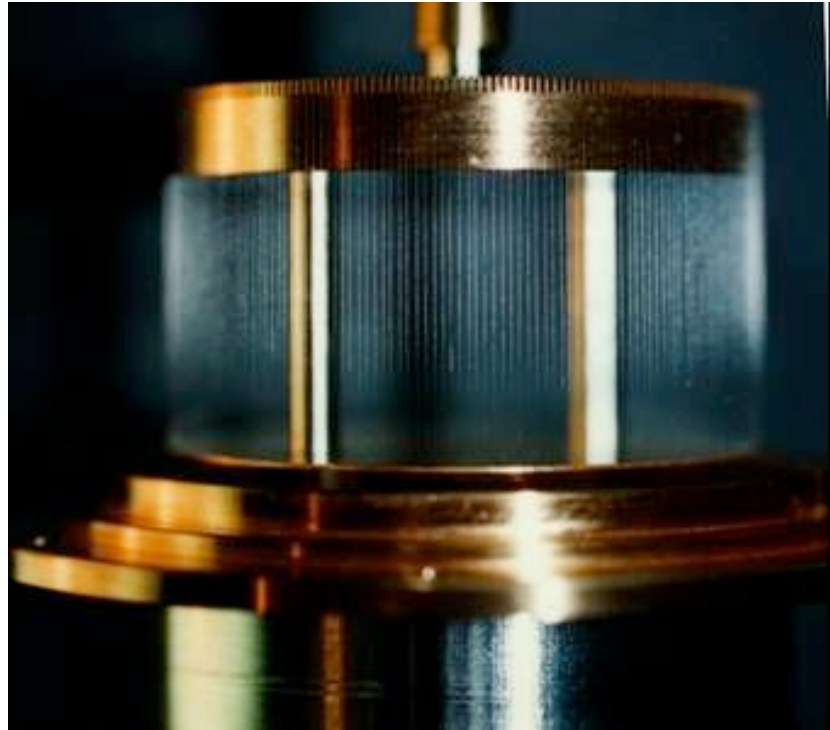
C. Garasi, A. Robinson



MHD MODEL

=

Hydrodynamics + Magnetic Diffusion

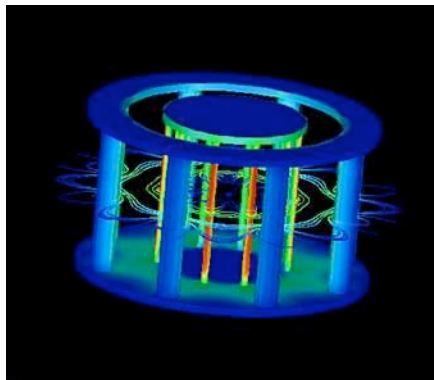
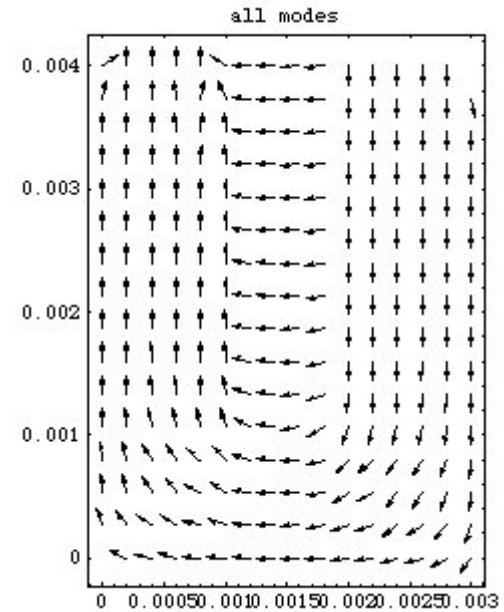
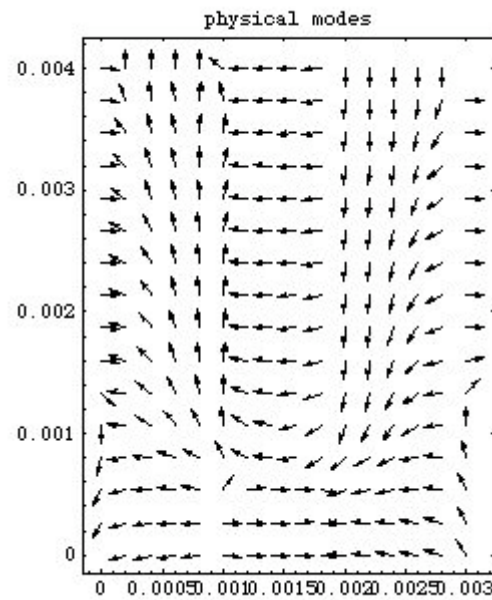
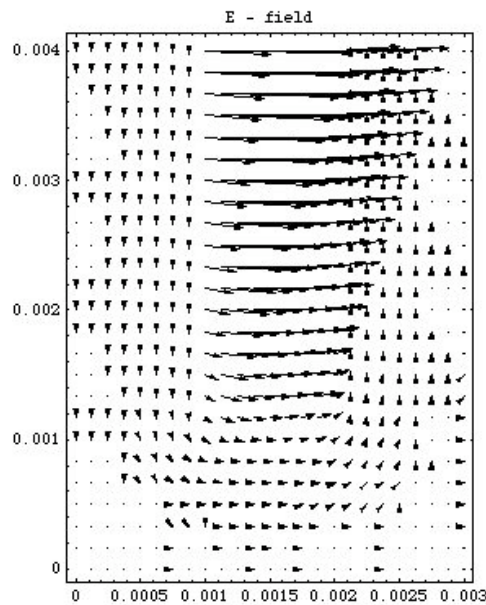


Z-machine: (**Ostensibly used by Ocean's 11**)

Electric currents are used to produce an ionized gas by vaporizing a spool-of-thread sized array of 100-400 wires of diameter $\approx 10\mu\text{m}$



Magnetic diffusion



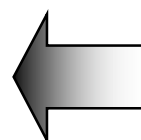
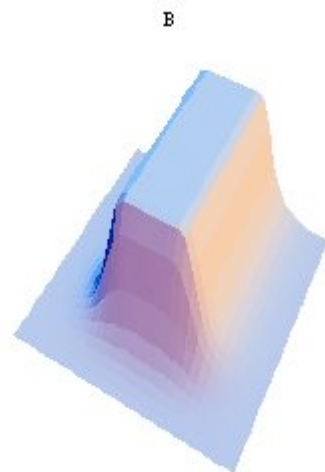
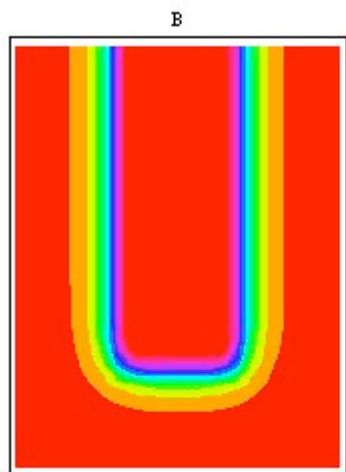
$$\nabla \times \frac{1}{\mu} \mathbf{B} = \sigma \mathbf{E} \quad \text{Ampere}$$

$$\nabla \times \mathbf{E} = -\frac{\partial \mathbf{B}}{\partial t} \quad \text{Faraday}$$

Gap modeled as a
heterogeneous
conductor

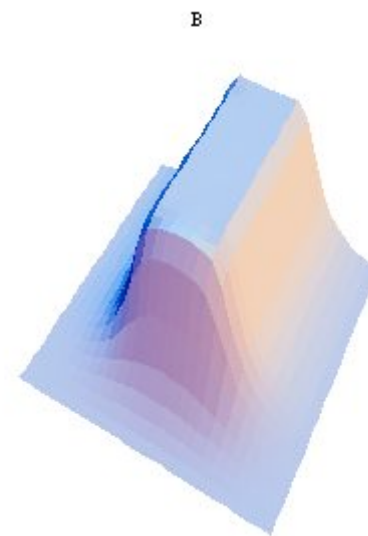
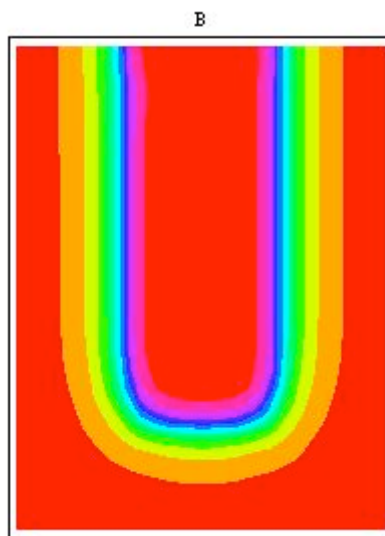
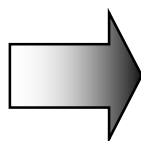


Compatible vs. Collocated: B-field



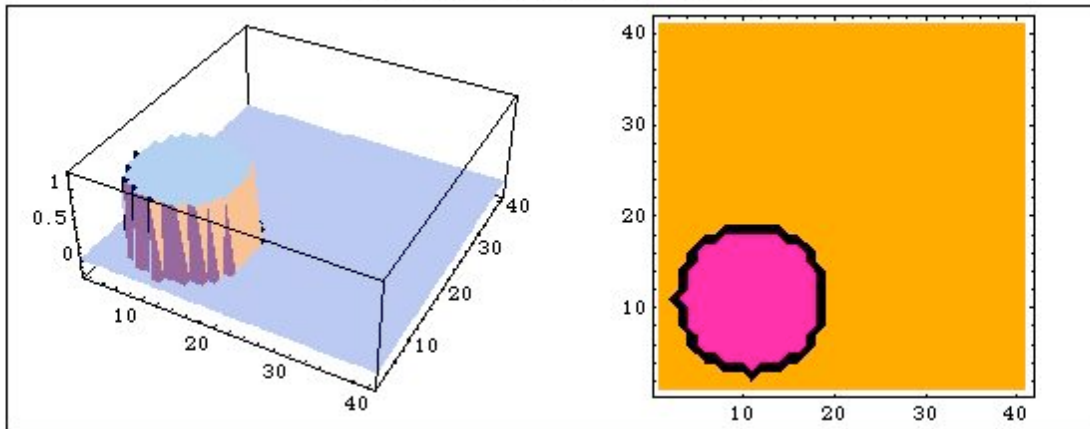
Compatible
 $\text{Ker}(\text{curl}) = \{\text{grad } p\}$

Collocated
 $\text{Ker}(\text{curl}) = \{0\}$



Incompatibility arises in other contexts as well, which are not discussed here!

Courant Number = 0.012, Pe=Infinity



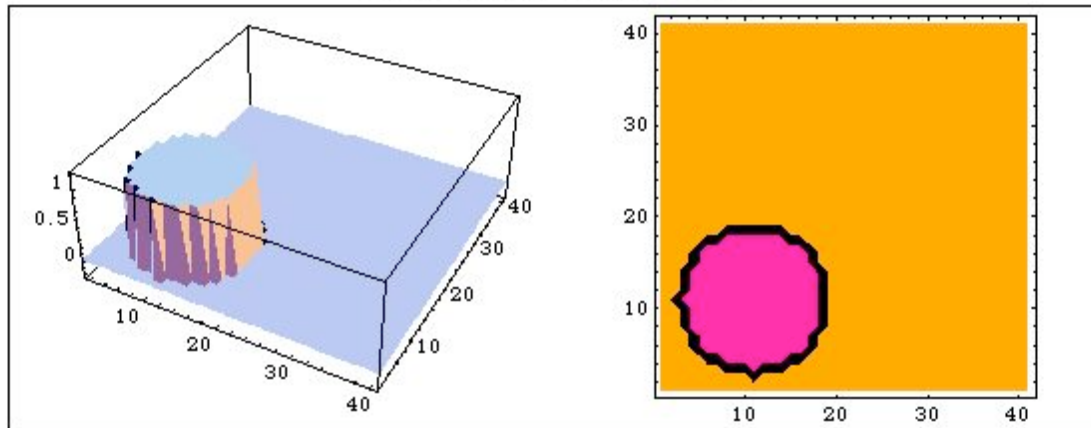
Galerkin

Advection of a scalar quantity

$$\nabla \cdot (\mathbf{a}\psi) = f$$

An extremely simplified case of a transport problem

Courant Number = 0.012, Pe=11.517



Adjoint Stabilized Galerkin

Seemingly reasonable fixes can make things even worse.



Part II

Where we learn about duality, its role in
PDE structure and what it can tell us about
compatible discretizations

All you ever wanted to know about duality (but were afraid to ask)

In mathematics “duality” is used in two contexts

1. Duality pairing

The process of combining **two objects** to generate a **scalar**

$$\mathbf{a}^T \mathbf{b} \rightarrow \langle \mathbf{a}, \mathbf{b} \rangle$$

Vector \mathbf{a} dual to co-vector \mathbf{b}

$$f(x) \rightarrow \langle f, x \rangle$$

Function f dual to *its argument* x

$$\int_{\Omega} f \rightarrow \langle f, \Omega \rangle$$

Function f dual to *integration domain* Ω

$$\int_{\Omega} fg \rightarrow \langle f, g \rangle$$

Function f dual to *distribution* g

$$\langle f | g \rangle \rightarrow \langle f, g \rangle$$

Bra vector f dual to *ket vector* g

$$Fd \rightarrow \langle F, d \rangle$$

Force F dual to *displacement* d



Leads to the fundamental notions of adjoint and self-adjoint operators

$$\langle Af, g \rangle = \langle f, A^* g \rangle \quad A = A^*$$

$$\mathbf{b}^T \mathbf{A} \mathbf{a} = \mathbf{a}^T \mathbf{A}^T \mathbf{b}$$

$$\langle \mathbf{A} \mathbf{a}, \mathbf{b} \rangle = \langle \mathbf{a}, \mathbf{A}^T \mathbf{b} \rangle$$

Adjoint of a matrix \mathbf{A} is the **transpose**

$$\int_{\Omega} \nabla \cdot \mathbf{v} = \int_{\partial\Omega} \mathbf{n} \cdot \mathbf{v}$$

$$\langle \nabla \cdot \mathbf{v}, \Omega \rangle = \langle \mathbf{v}, \partial\Omega \rangle$$

Adjoint of **divergence** is **boundary**

$$\int_{\Omega} u \nabla \cdot \mathbf{v} = - \int_{\Omega} \nabla u \cdot \mathbf{v}$$

$$\langle u, \nabla \cdot \mathbf{v} \rangle = \langle -\nabla u, \mathbf{v} \rangle$$

Adjoint of **divergence** is **-gradient**

$$\int_{\Omega} -\Delta u v = \int_{\Omega} -\Delta v u$$

$$\langle -\Delta u, v \rangle = \langle u, -\Delta v \rangle$$

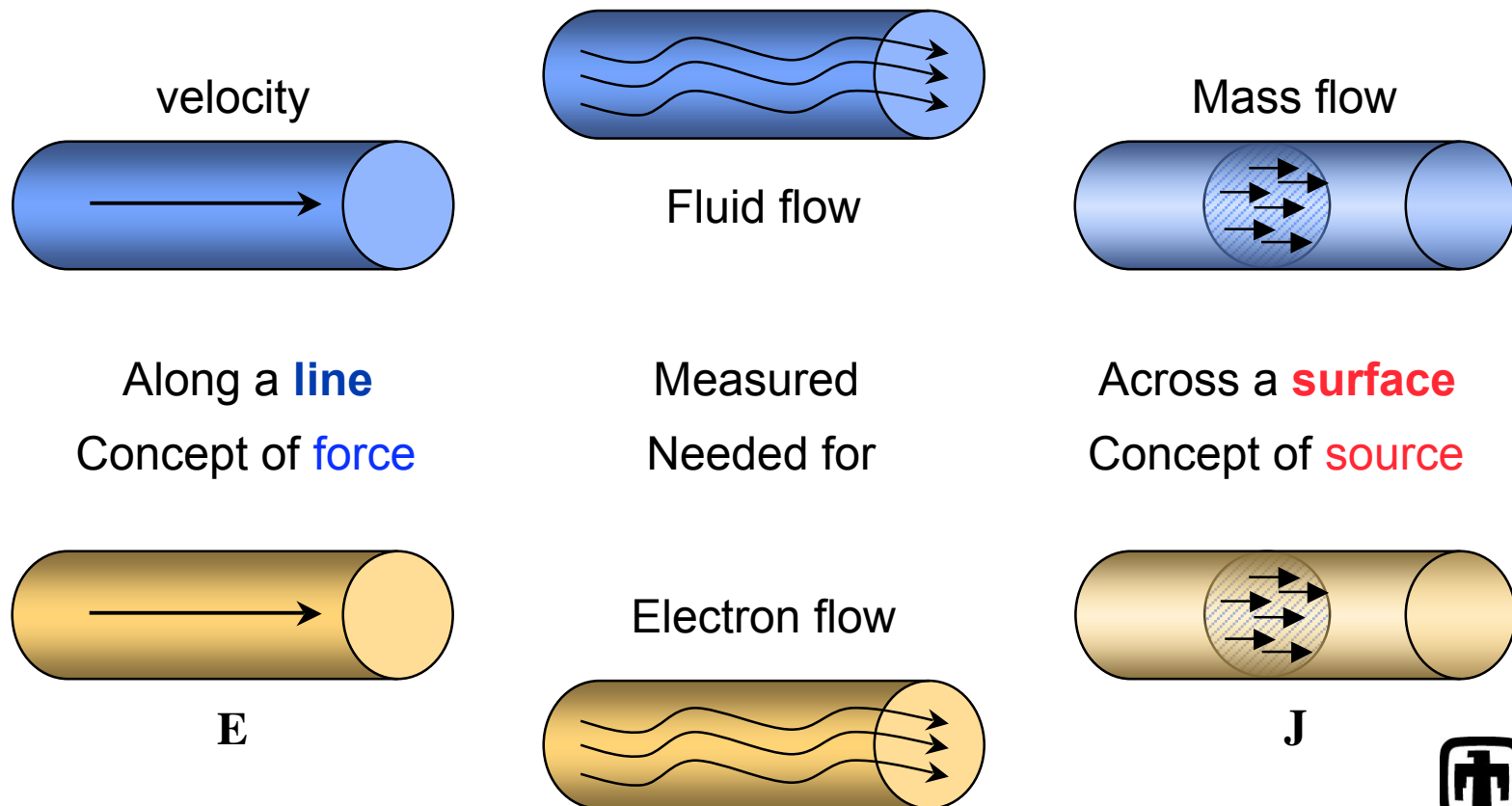
Laplacian is **self-adjoint**



A second duality concept

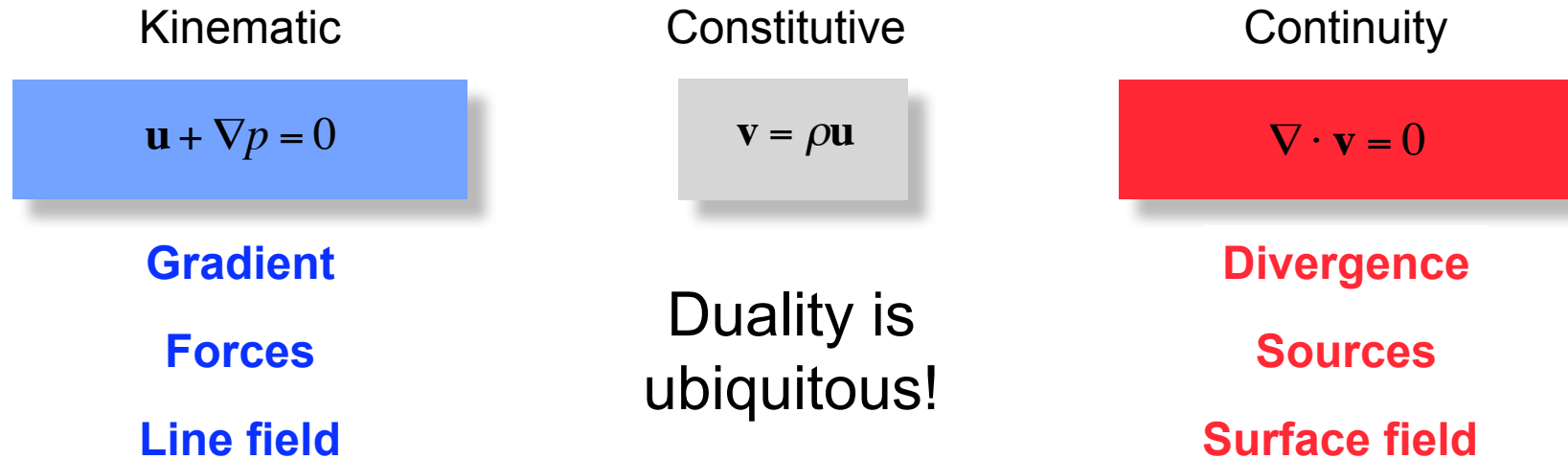
2. Duality of representations

The process of using complementary descriptions of the same process

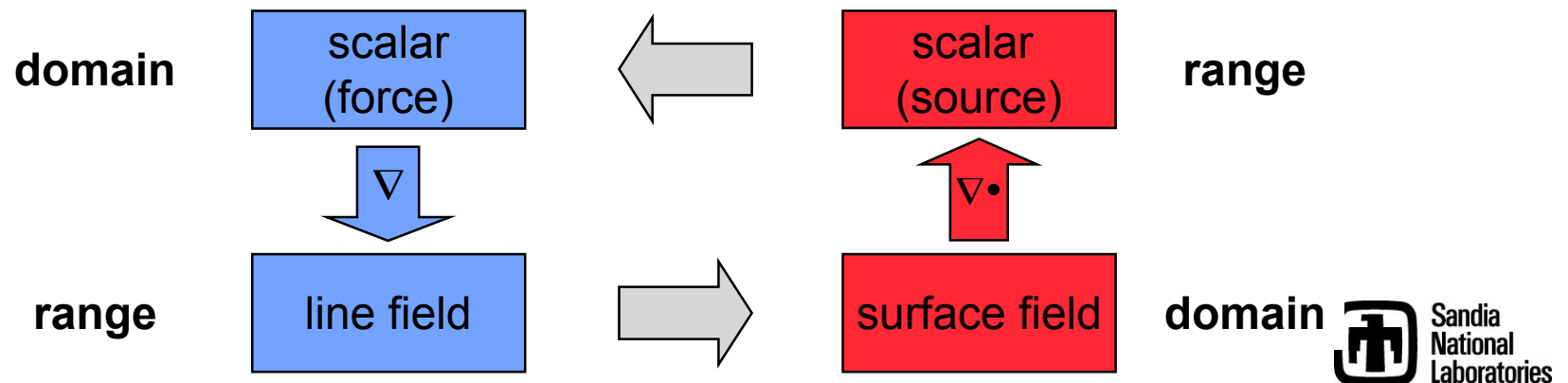




Darcy problem deconstructed

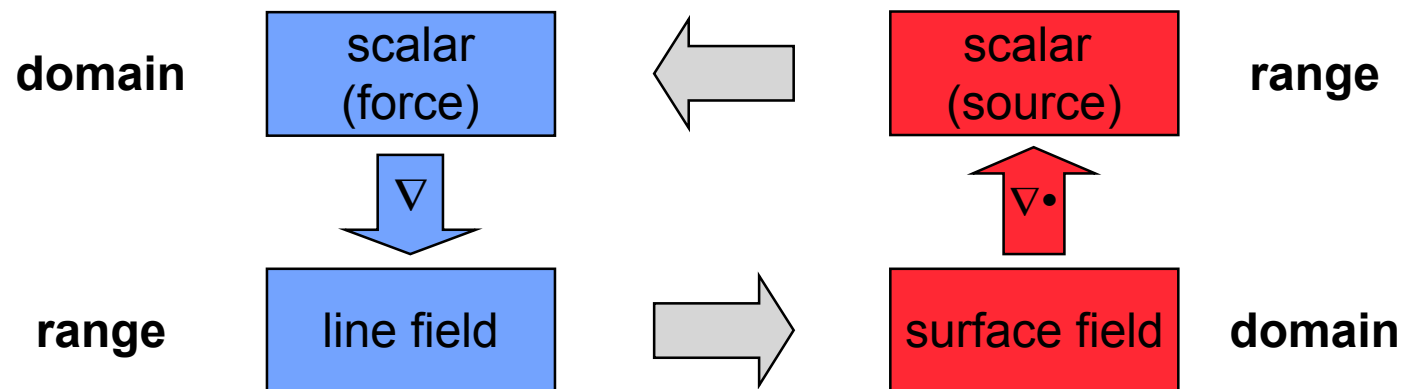


Duality structure can be encoded by the following diagram



Compatible discretizations

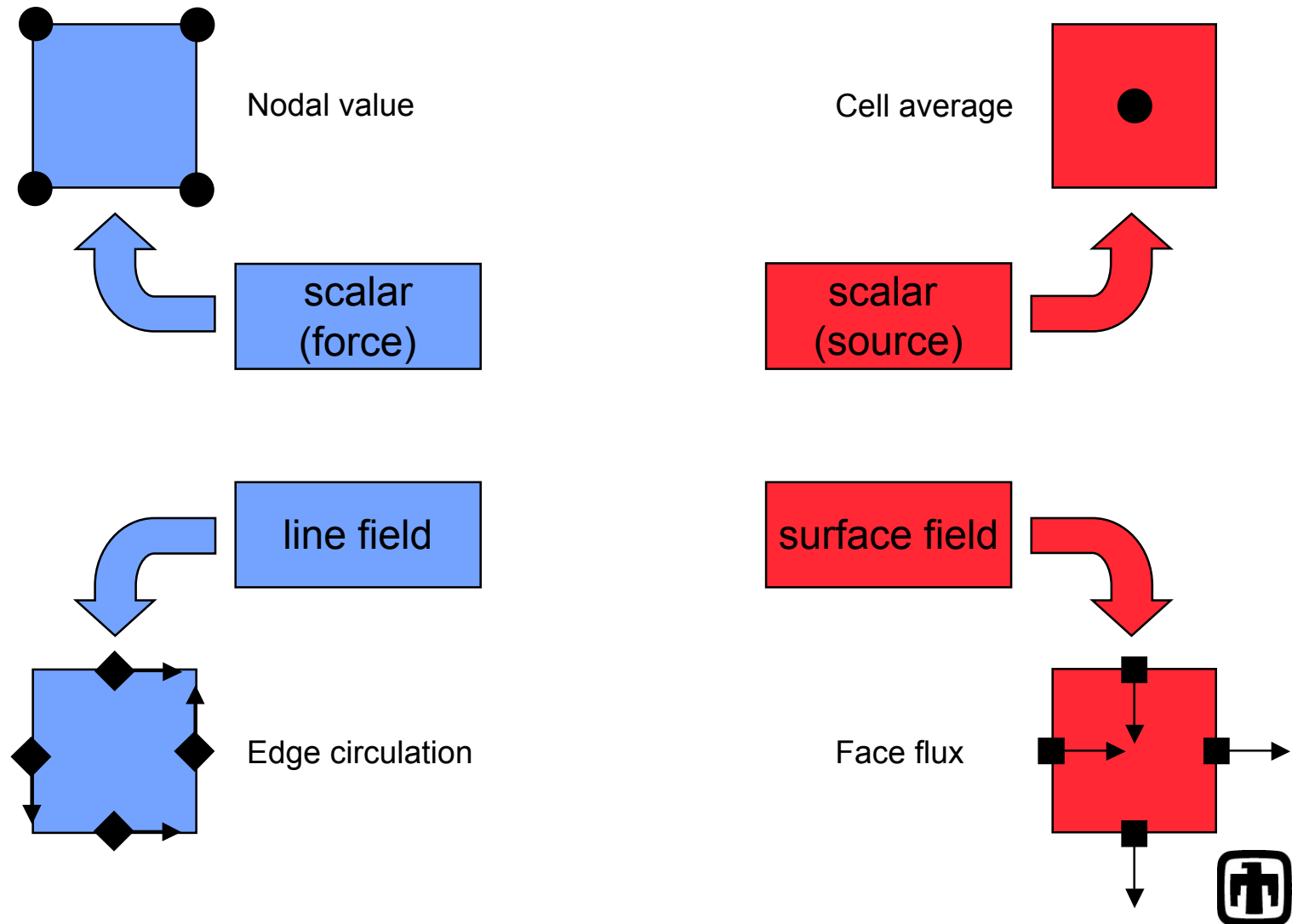
A compatible discretization should mimic the duality structure of the PDE



We need to build a discrete analogue of this diagram!

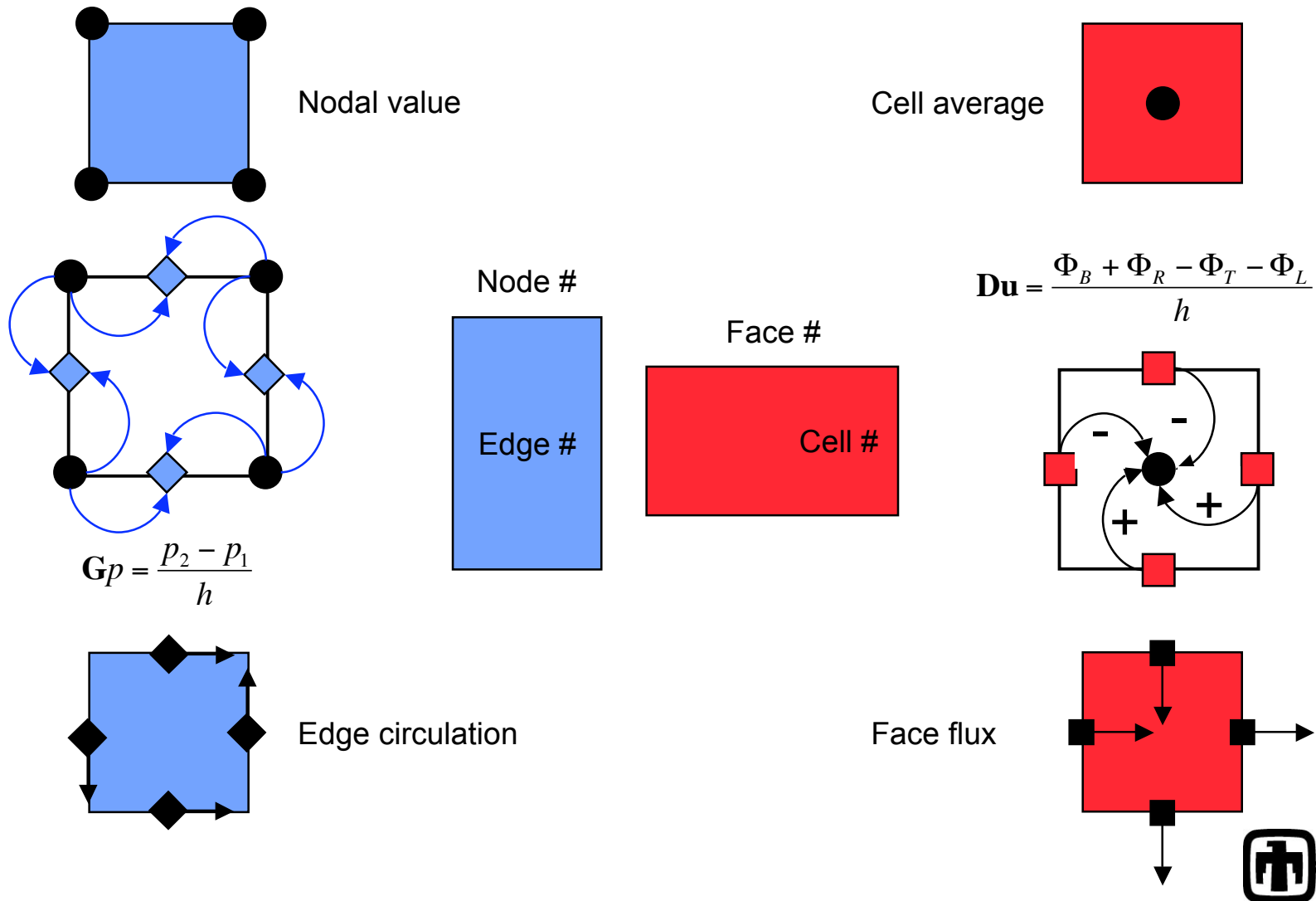


1. Compatible representations



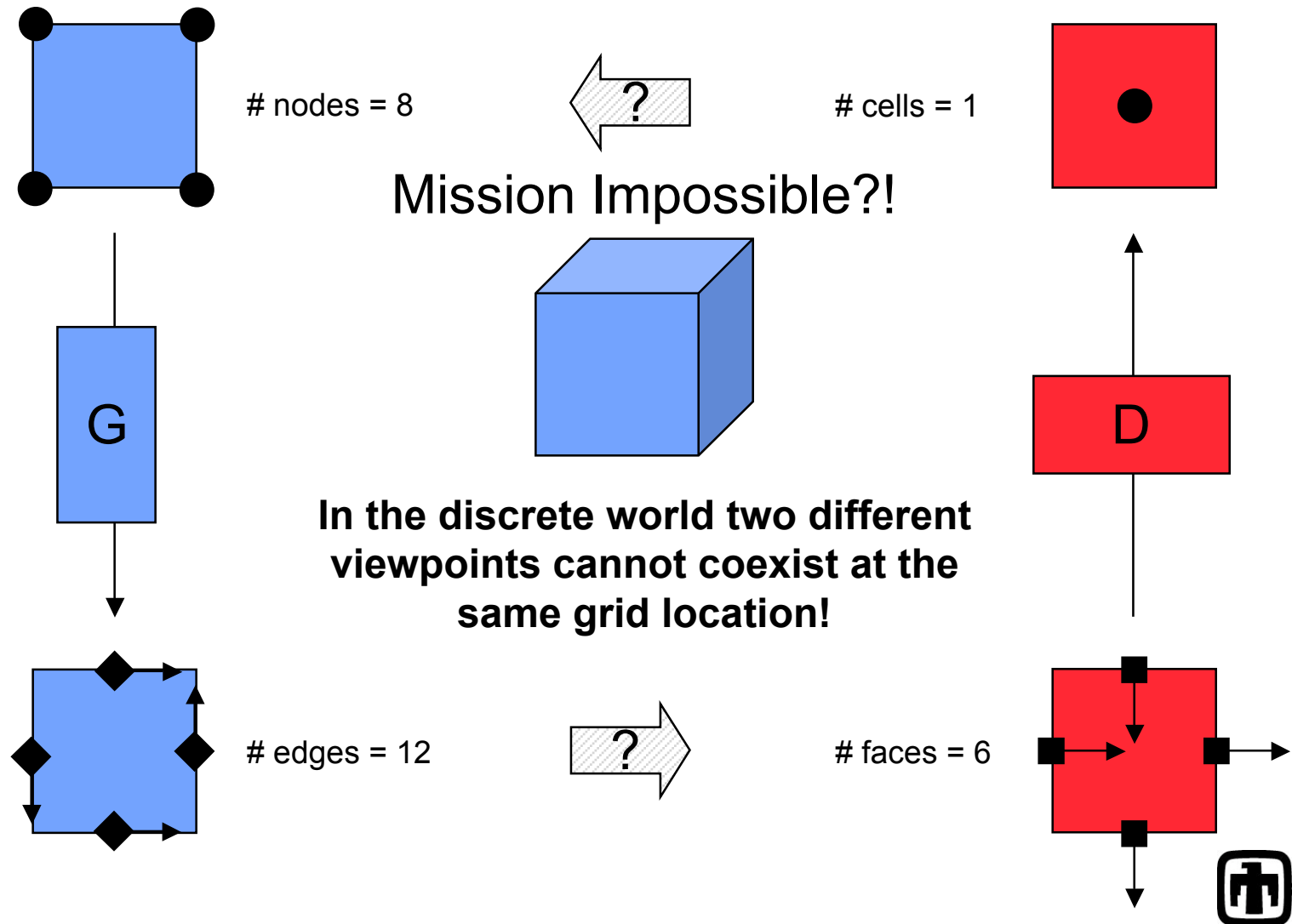


2. Compatible discrete operators



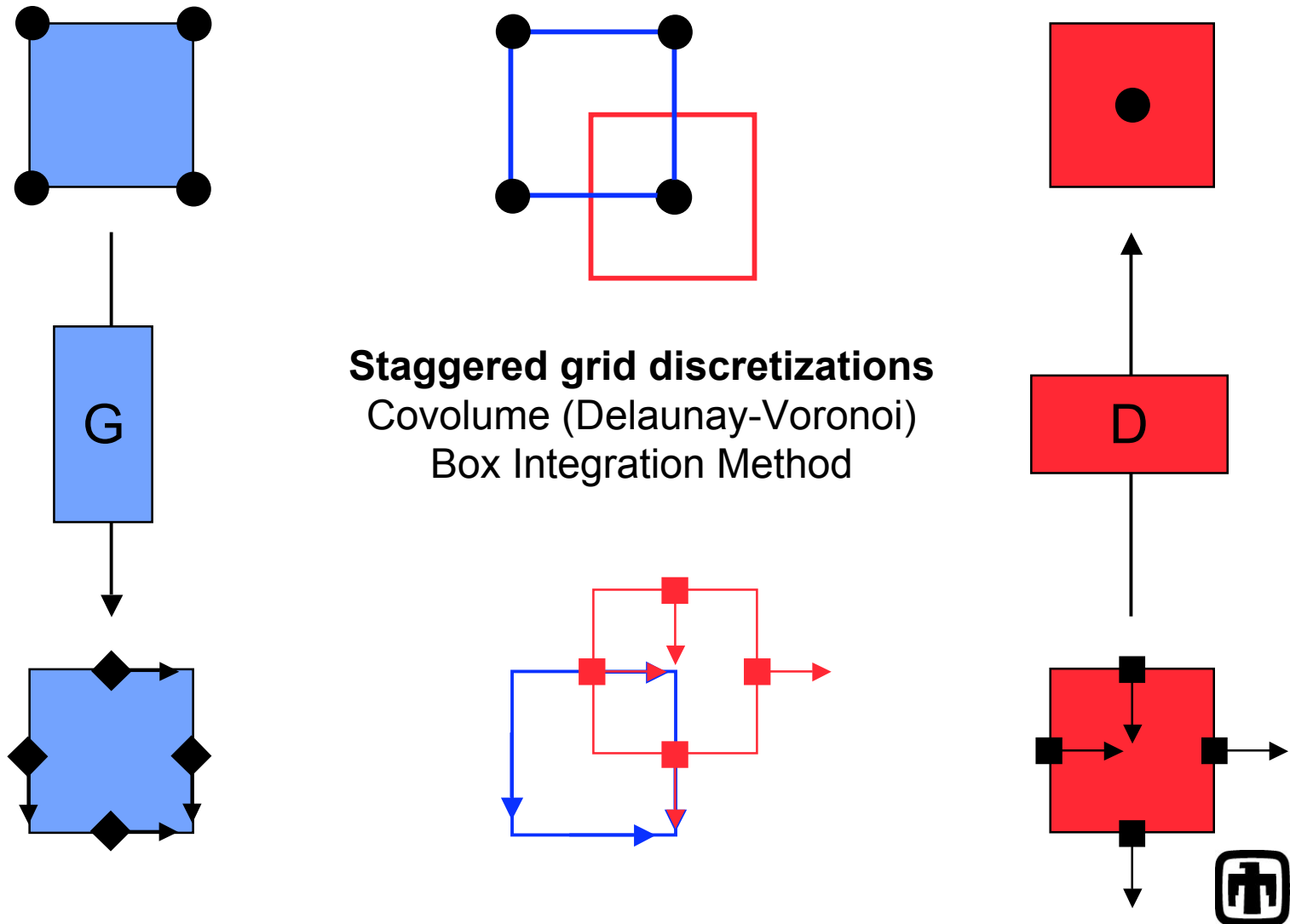


3. Putting it together





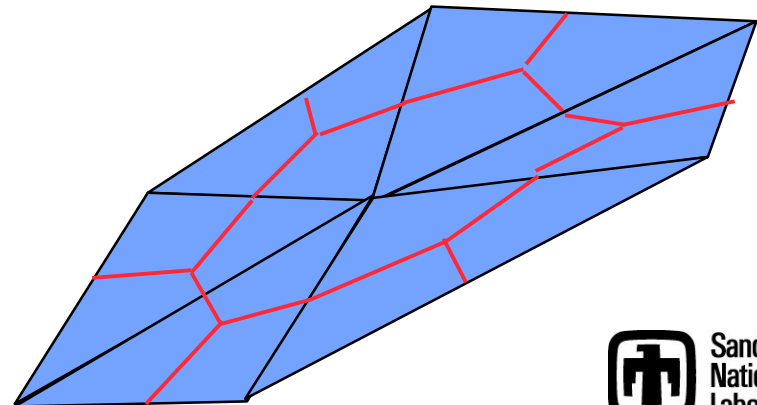
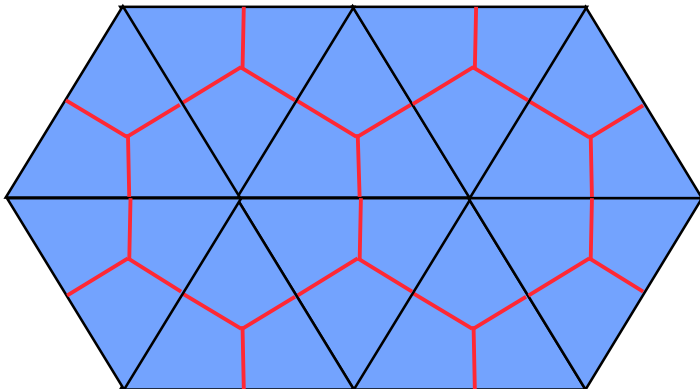
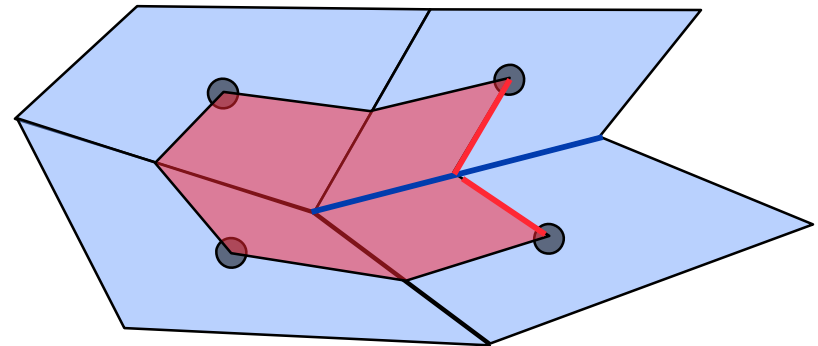
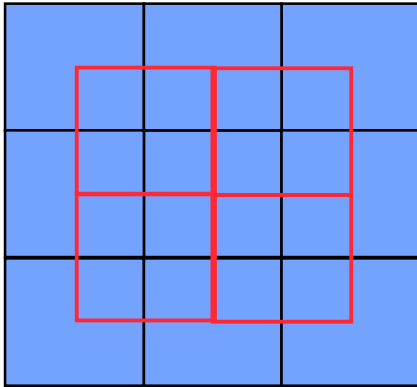
Solution #1 Primal-Dual Grid Complex





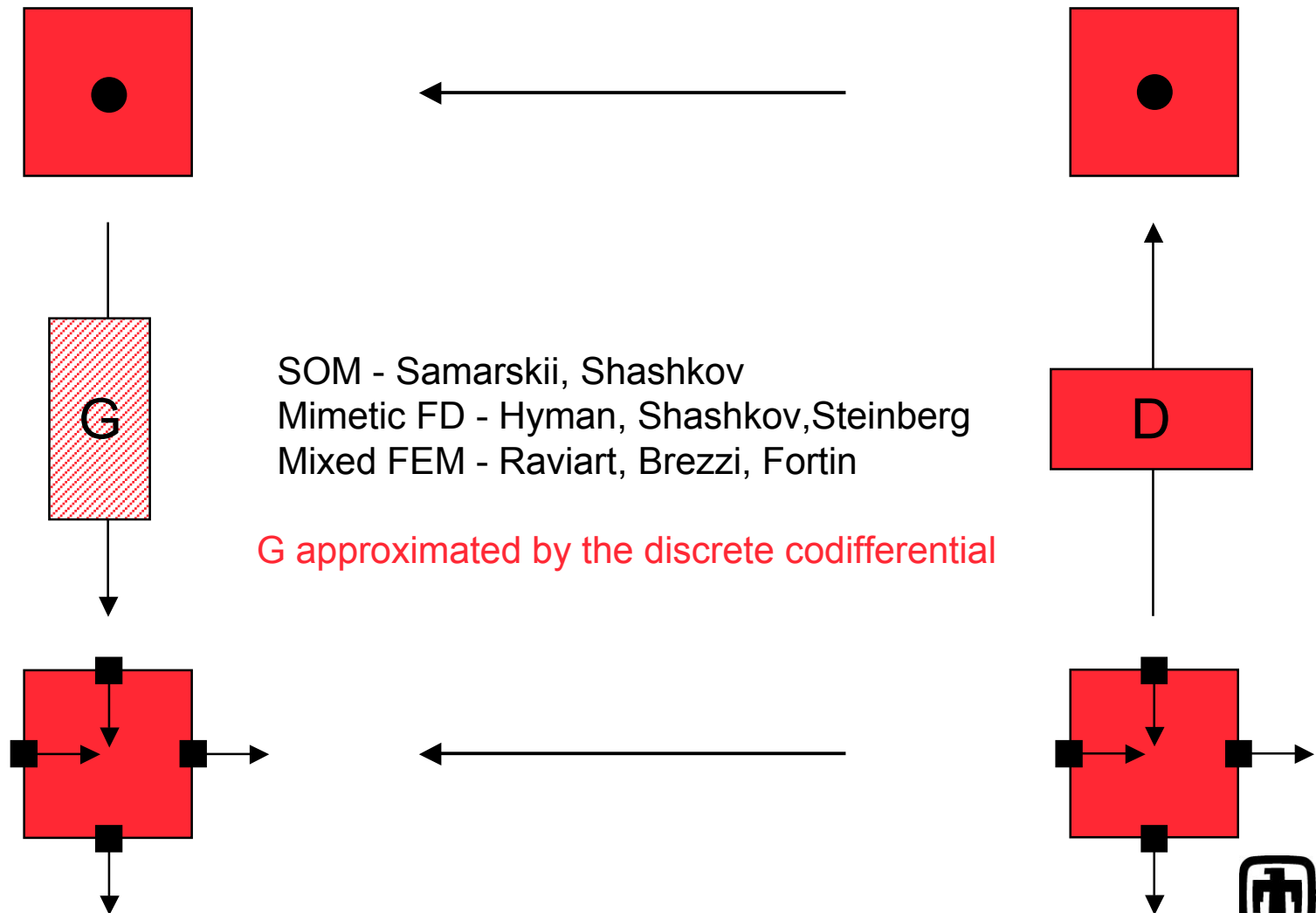
Not always feasible

Topologically dual grids hard to maintain for unstructured meshes
such as arising in ALE computations





Solution #2: cheat cleverly





So what went wrong earlier?



1. We used the same discrete representation for **all (!)** fields which was **incompatible** with their physical **meaning** and their **places** in the **domain** and the **range** of the gradient and divergence operators



The loss of coexistence

In the continuum world line and surface fields **can coexist** at the same point in space:

⇒ Only one vector field can be used in the model

⇒ The other can be eliminated:

$$\begin{array}{ccc}
 \nabla \cdot \mathbf{v} = 0 & & \\
 \mathbf{u} + \nabla p = 0 & \longrightarrow & \mathbf{v} = \rho \mathbf{u} \\
 \mathbf{v} - \rho \mathbf{u} = 0 & & \longrightarrow
 \end{array}
 \begin{array}{l}
 \nabla \cdot \rho \mathbf{u} = 0 \\
 \mathbf{u} + \nabla p = 0
 \end{array}$$

Unfortunately, in the discrete world line and surface fields **cannot coexist** at the same grid location

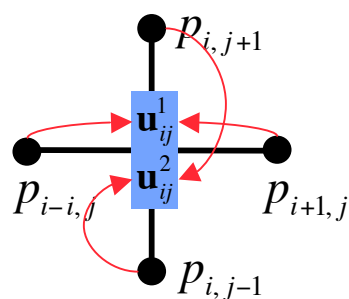
⇒ Either both types must be retained (requires primal-dual grid)

⇒ Or one of the operators must be modified



Collocated discrete operators can't work properly!

Discrete gradient



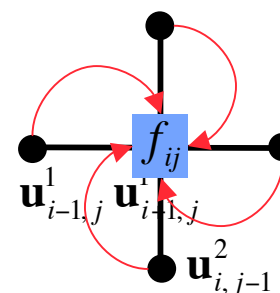
node \rightarrow node

Adjoint of discrete
divergence not boundary

$$\int_{\Omega} \nabla \cdot \mathbf{v} = \int_{\partial\Omega} \mathbf{n} \cdot \mathbf{v}$$

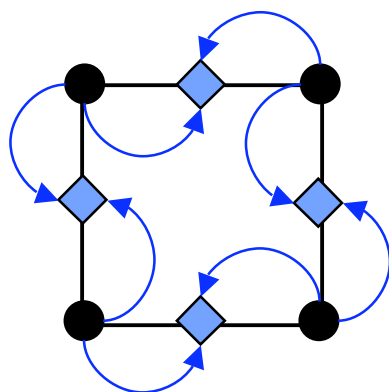
Non-conservative method

Discrete divergence



node \rightarrow node

node \rightarrow edge

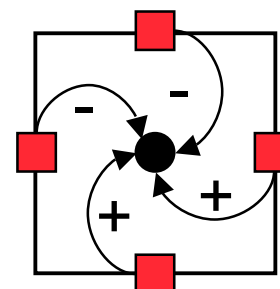


Adjoint of discrete
divergence is boundary

$$\int_{\Omega} \nabla \cdot \mathbf{v} = \int_{\partial\Omega} \mathbf{n} \cdot \mathbf{v}$$

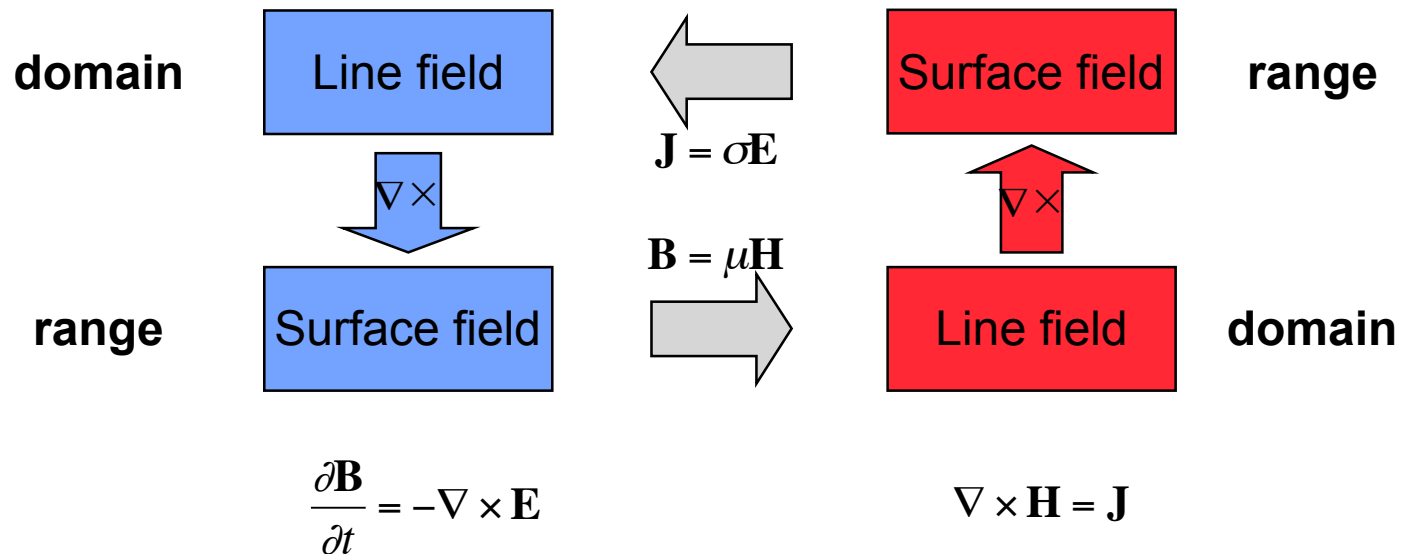
Conservative method

face \rightarrow center



Is this stuff useful in other cases?

Many other important models have identical duality structure



E & J provide dual description of electric phenomena
B & H provide dual description of magnetic phenomena
Curl is self-adjoint.



Conclusions

- ❑ Discretization is a **model reduction**
 - ❑ Careless discretization causes **unphysical behavior**
 - ❑ Compatible discretizations **mimic** continuum structures
 - ❑ Our discussion can be formalized using Differential Geometry and Algebraic Topology
 - ❑ Differential geometry provides the **tools to encode** this structure
 - ❑ Algebraic topology provides the **tools to copy** the structure to discrete models
 - ❑ Further details:
-
- Springer IMA Lecture Notes 142, **Spatial Compatible Discretizations**. Edited by D. Arnold, P. Bochev, R. Lehoucq, R. Nicolaides, M. Shashkov
 - <http://www.ima.umn.edu/talks/workshops/5-11-15.2004>
 - <http://www.sandia.gov/~pbboche/index.html>