

# Mimetic Discretizations and what they can do for you

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Computational Mathematics and Algorithms  
Sandia National Laboratories

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- Jonathan Hu
- Rich Lehoucq
- Denis Ridzal
- Allen Robinson
- John Shadid
- Chris Siefert
- Ray Tuminaro
- Max Gunzburger, Florida State
- Mac Hyman, Los Alamos
- Misha Shashkov, Los Alamos
- Pavel Solin, UTEP
- Kate Trapp, U. Richmond



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# Research Drivers

## Discretization

$$Lu = f \rightarrow Ax = b$$



model reduction, accompanied by **loss of information** that can be:

- 😊 **Acceptable** → **physically meaningful**, accurate and stable solutions.
- 💣 **Trivial** → **spectacular failure** that is easy to detect.
- ☠️ **Malicious** → **subtle failure**, imperceptible in the “eye ball” norm.

## Research goals:

- Develop “**compatible**” discretizations to manage information loss
- Use these discretizations and their properties to
  - a) Formulate and analyze new numerical methods for PDEs
  - b) **Support the development of better iterative solvers**
  - c) **Guide the design of better software tools for PDEs**

**Focus of this talk is on b) and c)**



# Impact

## External

- ❑ **2 short courses:** Von Karman Institute (2003), VA Tech (2005)
- ❑ **1 book:** Proceedings IMA workshop, Springer IMA series 142.  
(Arnold, Bochev, Lehoucq, Nicolaides, Shashkov, eds.)
- ❑ **2 book chapters**
- ❑ **14 papers** in peer reviewed journals
- ❑ **15 talks** (invited and plenary)
- ❑ **14 colloquium** talks
- ❑ **Originator and organizer:**
  - 2007 SIAM CS/E (with M. Shashkov)
  - 2007 FE Fluids (with M. Gunzburger)
  - 2006 CSRI PDE workshop (with R. Lehoucq and M. Gunzburger)
  - 2004 IMA Workshop on compatible discretizations (Arnold, Lehoucq, Nicolaides, Shashkov)
  - 2003 SIAM CS/E (with R. Tuminaro)

**Project inception: FY03**

## Internal

- ❑ **Compatible methods for x-MHD** - with J. Shadid, L. Chacon (LANL)
- ❑ **z-pinch modeling and simulation in Alegria** - with A. Robinson
- ❑ **Device modeling and simulation in CHARON** - with J. Shadid, R. Pawlovski
- ❑ **ML solvers for Maxwell's** - with R. Tuminaro, J. Hu, C. Siefert



# Research Approach

- ❑ **Use homological ideas** to identify formal mathematical structures (“**analytic home**”) that allow to encode a representative set of PDEs
- ❑ **Translate analytic structures** into “**compatible**” discrete structures (“**discrete home**”) that inherit their key properties
- ❑ **Manage loss of information** by translating PDEs into **compatible** discrete models that live in the “discrete home”

## Overview of my talk:

- ❑ **A compendium of failed discretizations**
- ❑ **Analytic → discrete translation**
  - Based on two fundamental operators: **Reduction and reconstruction**
- ❑ **Mimetic properties:**
  - Vector calculus and discrete cohomology
- ❑ **Payback:**
  - New infrastructure for interoperable software tools for FEM, FV, and FD discretizations
  - More efficient AMG solvers via reformulation of the discrete Maxwell’s equations



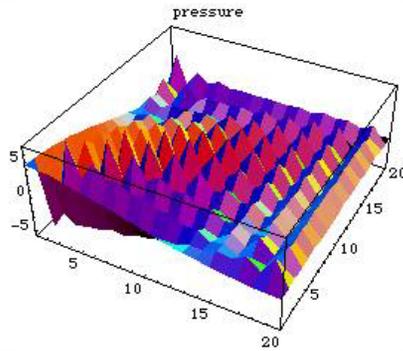
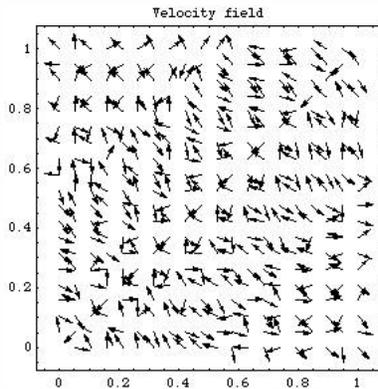
$$\begin{aligned} \nabla \cdot \mathbf{u} &= f \text{ in } \Omega \\ \nabla \phi + \mathbf{u} &= 0 \text{ in } \Omega \\ \phi &= 0 \text{ on } \Gamma \end{aligned}$$

# Deal or No Deal?

$$\begin{aligned} \sigma \mathbf{E} + \nabla \times \nabla \mu^{-1} \times \mathbf{E} &= 0 \text{ in } \Omega \\ \mathbf{n} \times \mathbf{E} &= 0 \text{ on } \Gamma \end{aligned}$$

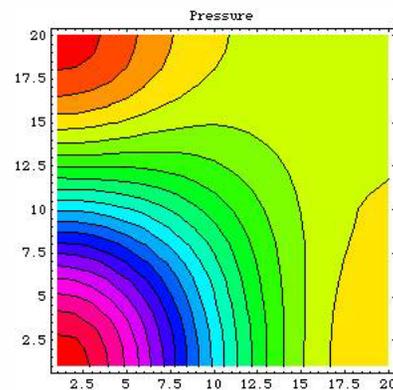
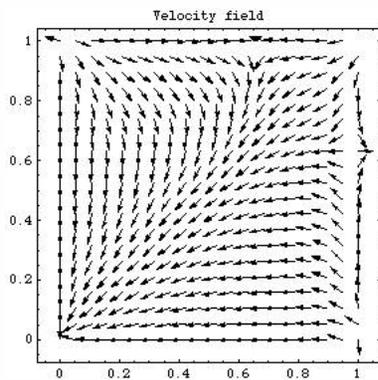
**Trivial failure: Mixed Galerkin and nodal (collocated) FEM**

**Malicious failure: Ritz-Galerkin and nodal (collocated) FEM**



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compatible



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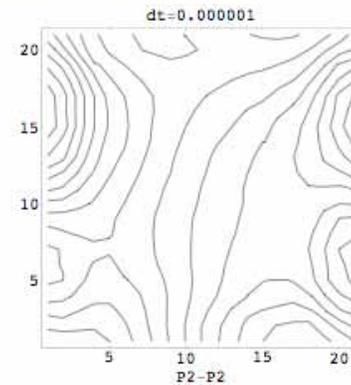
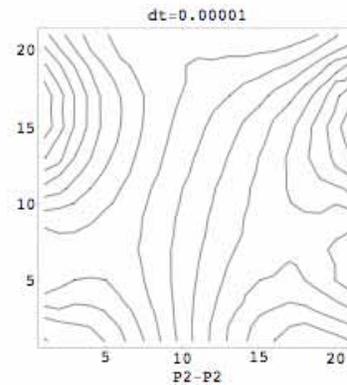
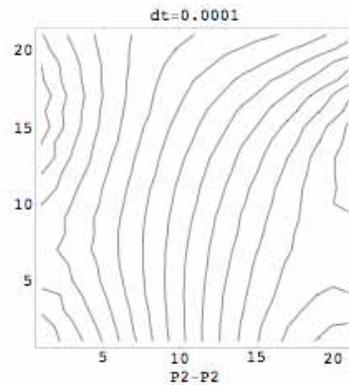
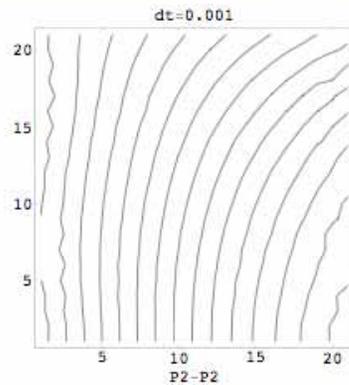
$$-\Delta \mathbf{u} + \nabla p = \mathbf{f} \text{ in } \Omega$$

$$\nabla \cdot \mathbf{u} = 0 \text{ in } \Omega$$

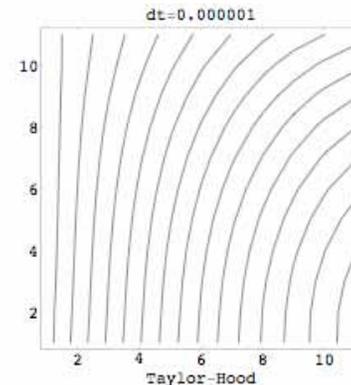
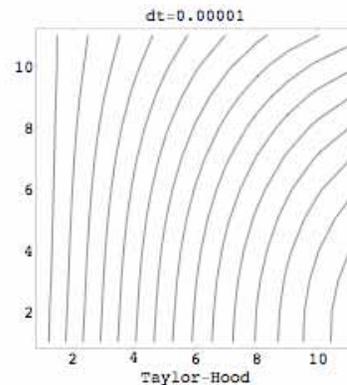
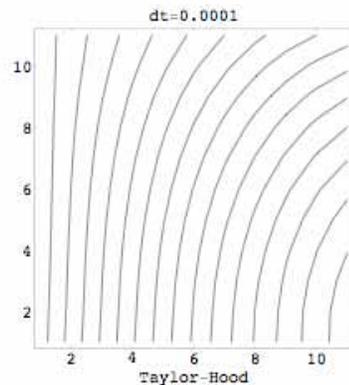
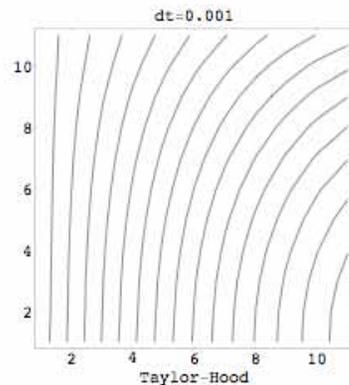
$$\mathbf{u} = 0 \text{ on } \Gamma$$

# Deal or No Deal?

Another **malicious** failure: **false transient** (spatially regularized nodal FE)



Common wisdom:  $\Delta t \rightarrow 0 \Rightarrow$  more accurate results



True solution is **time independent!**

Bochev, Gunzburger, Lehoucq, *IJNMF*, 2007



# Why Homological Ideas?

In the examples, there was nothing wrong with the **approximation** properties of the FEMs or the **formal consistency** of the methods.

However, key **relationships** between **differential operators** and **function spaces**, necessary for the **well-being** of the PDE, were **“lost in translation”**

We seek a discrete framework that **mimics** these relationships and provides mutually consistent notions of derivative, integral, inner product, Hodge theory, etc.

**Cohomology:** Describes **structural relationships** relevant to PDEs

**Differential forms:** Provide tools for **abstraction** of physical models leading to PDEs:

**Integration:** → an abstraction of the *measurement* process

**Differentiation:** → gives rise to *local invariants*

**Poincare Lemma:** → expresses *local geometric* relations

**Stokes Theorem:** → gives rise to *global relations*



# An (incomplete) Historical Survey

## In finite elements

1977 - **Fix, Gunzburger and Nicolaides**: **GDP** (a discrete Hodge decomposition) is **necessary** and **sufficient** for stable and optimally accurate mixed Galerkin discretization of the Poisson equation  
➔ **first (!) example of application of homological ideas in FEMs.**

1989 - **Bossavit**: reveals connection between Whitney forms and stable elements for mixed methods for diffusion and eddy currents

1997 - **Hiptmair**: uses exterior calculus to develop uniform definitions of FEM spaces

1999 - **Demkowicz, Ainsworth, et al**: develop *hp*-DeRham polynomial spaces

2002 - **Arnold et al.**: uses homological ideas to find stable FEMs for mixed elasticity

2003 - **White et al.**: FEMSTER, a software realization of polynomial differential forms

## Elsewhere: Discrete vector calculus structures

1980s - **Shashkov, Samarskii** - Support operator method

1992 - **Nicolaides** - direct covolume discretization for div-curl and incompressible flows

1990s - **Hyman, Scovel, Shashkov, Steinberg** - Mimetic finite difference methods

1997 - **Mattiussi** - connection between FV and FEM

2004 - **Bochev and Hyman** - **Algebraic topology** approach: includes FV, FD and FEM



# Analytic $\rightarrow$ Discrete

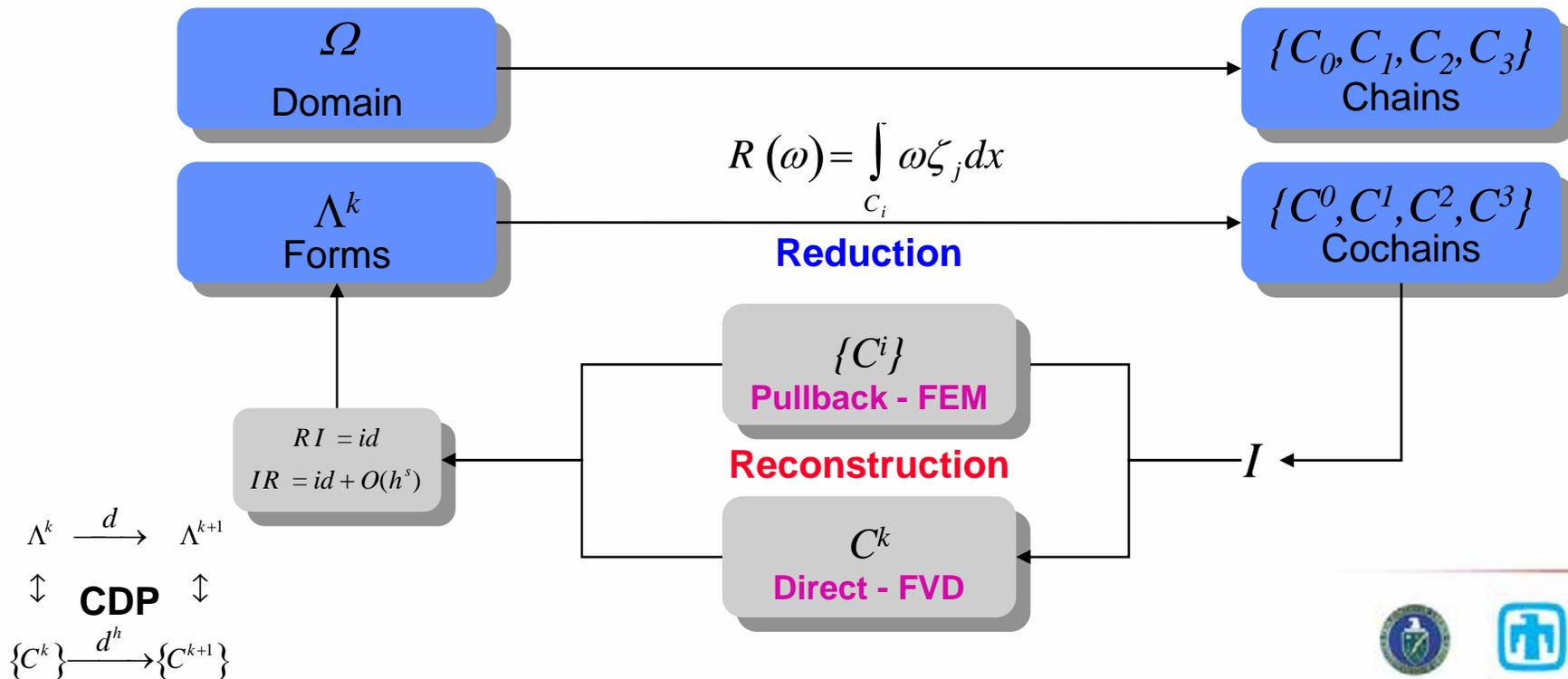
## Framework for mimetic discretizations (IMA Proceedings, 2006)

- Exterior Derivative
- Metric structure
- Adjoint derivative



- Natural operations
- Discrete inner product
- Derived operations

induced by 2 basic operations



# Discrete Operations for $R: \Lambda^k \rightarrow C^k$

**Natural derivative**

$$\delta: C^k \mapsto C^{k+1}$$

$$\langle \delta a, \sigma \rangle = \langle a, \partial \sigma \rangle$$

**Natural inner product**

$$(\cdot, \cdot)_k: C^k \times C^k \rightarrow \mathbf{R}$$

$$(a, b)_k = (Ia, Ib)_k$$

**Adjoint derivative**

$$\delta^*: C^{k+1} \mapsto C^k$$

$$(\delta^* a, b)_k = (a, \delta b)_{k+1}$$

Provides a second set of **grad**, **div** and **curl** operators.

Derivative choice depends on encoding:



**scalars**  $\rightarrow$  0 or 3-forms

**vectors**  $\rightarrow$  1 or 2-forms.

**Discrete Laplacian**

$$D: C^k \mapsto C^k$$

$$D = \delta^* \delta + \delta \delta^*$$

**Natural wedge product**

$$\wedge: C^k \times C^l \mapsto C^{k+l}$$

$$a \wedge b = R(Ia \wedge Ib)$$

**Flat** and **sharp**

can be defined using the inner product

Derived operations help to avoid **internal inconsistencies** between the discrete operations:

- I is only **approximate inverse** of R and natural definitions will clash.

# Discrete Vector Calculus

**Poincare lemma** (existence of potentials in contractible domains)

$$d\omega_k = 0 \Rightarrow \omega_k = d\omega_{k+1} \quad \longrightarrow \quad \delta c^k = 0 \Rightarrow c^k = \delta c^{k+1}$$

**Stokes Theorem**

$$\langle d\omega_{k-1}, c_k \rangle = \langle \omega_{k-1}, \hat{c}_k \rangle \quad \longrightarrow \quad \langle \delta c^{k-1}, c_k \rangle = \langle c^{k-1}, \hat{c}_k \rangle$$

**Vector Calculus**

$$\begin{aligned} dd &= 0 & \delta\delta &= \delta^* \delta^* = 0 \\ \omega \wedge \eta &= (-1)^{kl} \eta \wedge \omega & a \wedge b &= (-1)^{kl} b \wedge a \\ d(\omega \wedge \eta) &= d\omega \wedge \eta + (-1)^k \omega \wedge d\eta & \delta(a \wedge b) &= \delta a \wedge b + (-1)^k a \wedge \delta b \end{aligned}$$

**Mimetic** = Key properties of the analytic structures inherited by the discrete structures. First used by Hyman and Scovel (1988)



# Discrete Cohomology

**R** is a chain map: preserves co-boundaries and co-cycles

$$d\omega = 0 \quad \Rightarrow \quad \delta R \omega = 0$$

$$\text{Co-cycles of } (\Lambda^0, \Lambda^1, \Lambda^2, \Lambda^3) \xrightarrow{R} \text{co-cycles of } (C^0, C^1, C^2, C^3)$$

**Natural inner product induces combinatorial Hodge theory on cochains:**

## Discrete Harmonic forms

$$H^k(\Omega) = \{\eta \in \Lambda^k(\Omega) \mid d\eta = d^*\eta = 0\} \longrightarrow H^k(K) = \{c^k \in C^k \mid \delta c^k = \delta^* c^k = 0\}$$

## Discrete Hodge decomposition

$$\omega = d\rho + \eta + d^*\sigma \longrightarrow a = \delta b + h + \delta^*c$$

**Theorem** (*IMA Proc.*, 2006)

$$\dim \ker(\Delta) = \dim \ker(D)$$

Remarkable property of the mimetic  $D$  - kernel size is a **topological invariant!**

Not used

# Inventory

Non-local

$$\omega = df + h + d^*g$$

“Roof”

$$\Lambda^k(\Omega) = \text{Range}(d_{k-1}) \oplus H^k \oplus \text{Range}(d_{k+1}^*)$$

$$H^k = \{\omega \in \Lambda^k \mid d\omega = d^*\omega = 0\} \quad \ker(\Delta_k) = H^k \quad H^k = \ker(d_k) / \text{Range}(d_{k-1})$$

“Bricks”

$$d : \Lambda^k(\Omega) \rightarrow \Lambda^{k+1}(\Omega)$$

exterior derivative

$$(\cdot, \cdot)_k : \Lambda^k(\Omega) \times \Lambda^k(\Omega) \rightarrow \mathbf{R}$$

inner product

$$\int : \Lambda^k(\Omega) \rightarrow \mathbf{R}$$

integral

$$\wedge : \Lambda^k(\Omega) \times \Lambda^l(\Omega) \rightarrow \Lambda^{k+l}(\Omega)$$

wedge product

$$\Delta : \Lambda^k(\Omega) \rightarrow \Lambda^k(\Omega)$$

Hodge Laplacian

“Foundation”

$$\Lambda^k(\Omega) : x \rightarrow \omega(x) \in \text{Alt}^k(T_x\Omega) \quad (\Lambda^0(\Omega), \Lambda^1(\Omega), \Lambda^2(\Omega), \Lambda^3(\Omega)) \quad \text{Smooth differential forms}$$

# The Tenants

## “Laplacians”

$$\min_{\Lambda^k} \frac{1}{2} \left( \|du\|^2 + \|d^*u\|^2 \right) - (f, u) \longrightarrow d^*du + dd^*u = f \longrightarrow \begin{cases} -\Delta u = f \\ \nabla \times \nabla \times u - \nabla \nabla \cdot u = f \end{cases}$$

## “Incomplete Laplacians”

$$\left. \begin{aligned} \min_{\Lambda^k} \frac{1}{2} \left( \|u\|^2 + \|du\|^2 \right) - (f, u) \\ \min_{\Lambda^k} \frac{1}{2} \left( \|u\|^2 + \|d^*u\|^2 \right) - (f, u) \end{aligned} \right\} \longrightarrow \begin{cases} dd^*u + u = f \\ d^*du + u = f \end{cases} \longrightarrow \begin{cases} -\Delta u + u = f \\ \nabla \times \nabla \times u + u = f \\ -\nabla \nabla \cdot u + u = f \end{cases}$$

## “Div-curl systems”

$$\min_{\Lambda^k} \frac{1}{2} \left( \|du\|^2 + \|d^*u\|^2 \right) - (f, u) \longrightarrow \begin{aligned} du + d^*p &= f \\ d^*u &= 0 \\ d^*u + dp &= f \\ du &= 0 \end{aligned} \longrightarrow \begin{cases} \nabla \times u + \nabla p = f \\ \nabla \cdot u = f \end{cases}$$

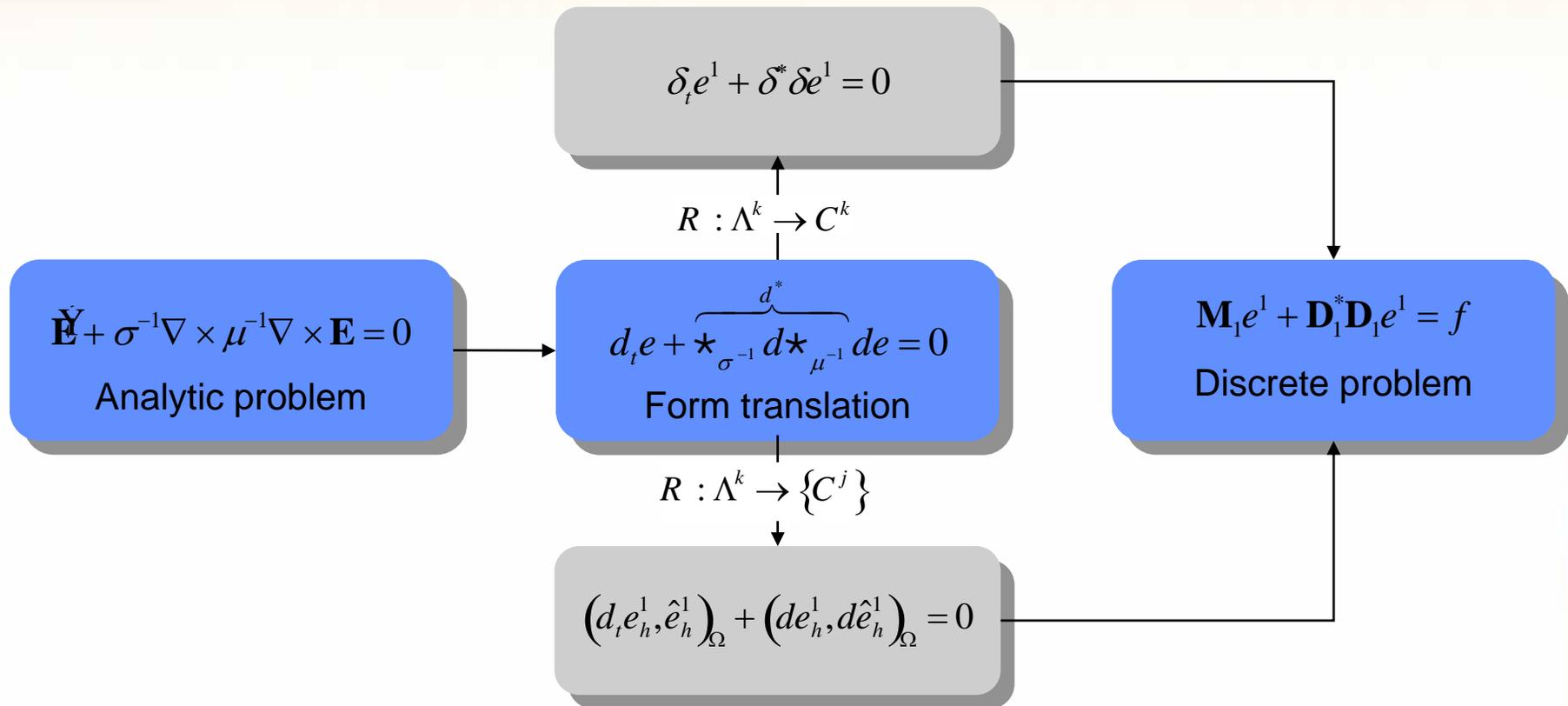
subject to  $du = 0$  or  $d^*u = 0$

## Div of a vector field

$$\longrightarrow d^*(\bar{u}) = (*d*)(\bar{u}) \longrightarrow \nabla \cdot u$$



# Placing a PDE in the Discrete Home



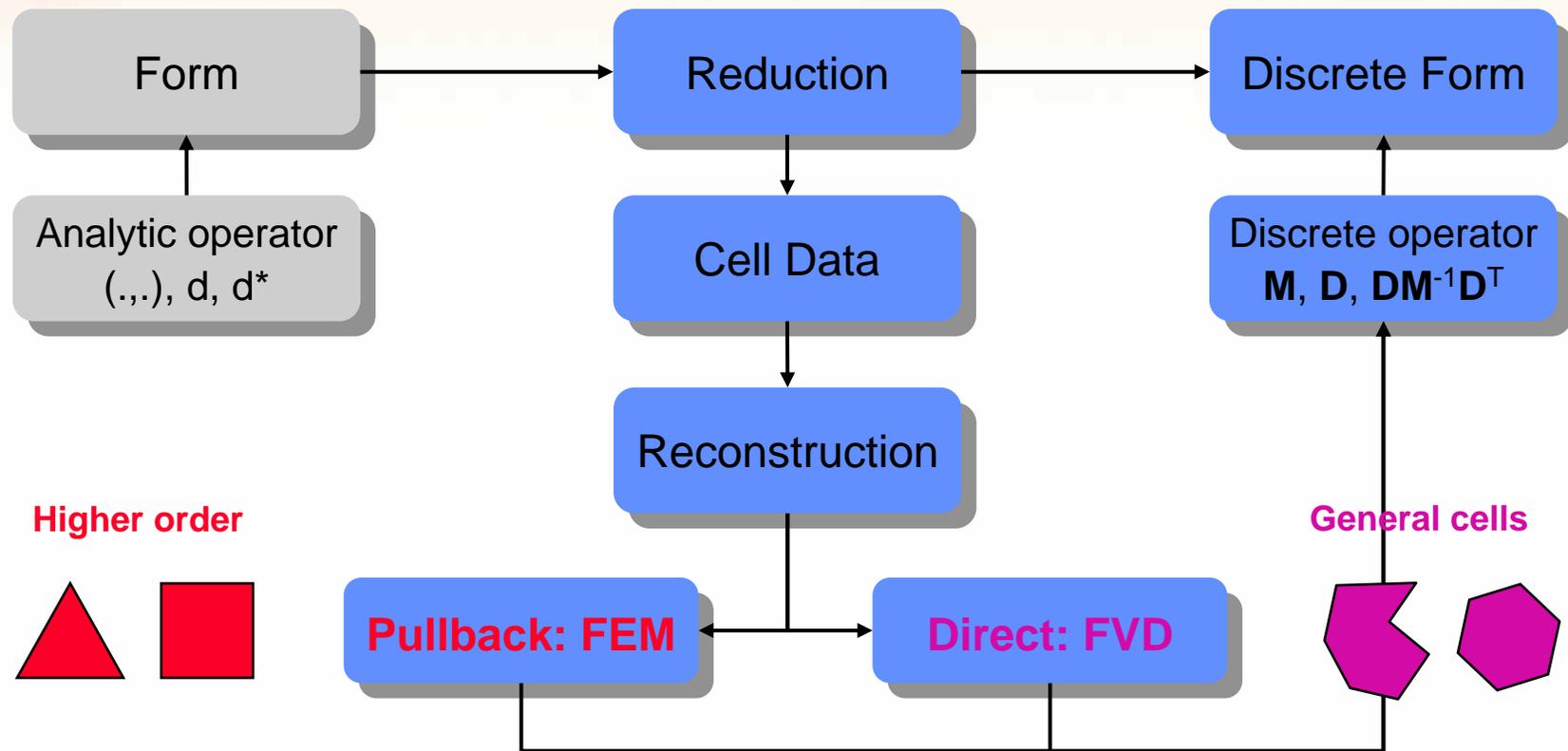
**Theorem** (IMA Proc., 2006)

Let  $R : \Lambda^k \rightarrow C^k$ . **Direct** and **pullback** reconstructions yield equivalent methods.

**⇒ There's only "one" low-order compatible method**



# Abstraction for OO Software Design



**This prompts a fresh look at software design for compatible discretizations:**

⇒ Different **methods** are defined by choosing a specific **reconstruction** operator  $I$ :

**Direct:**  $I$  is **low order** but more easily extendable to **arbitrary cells**

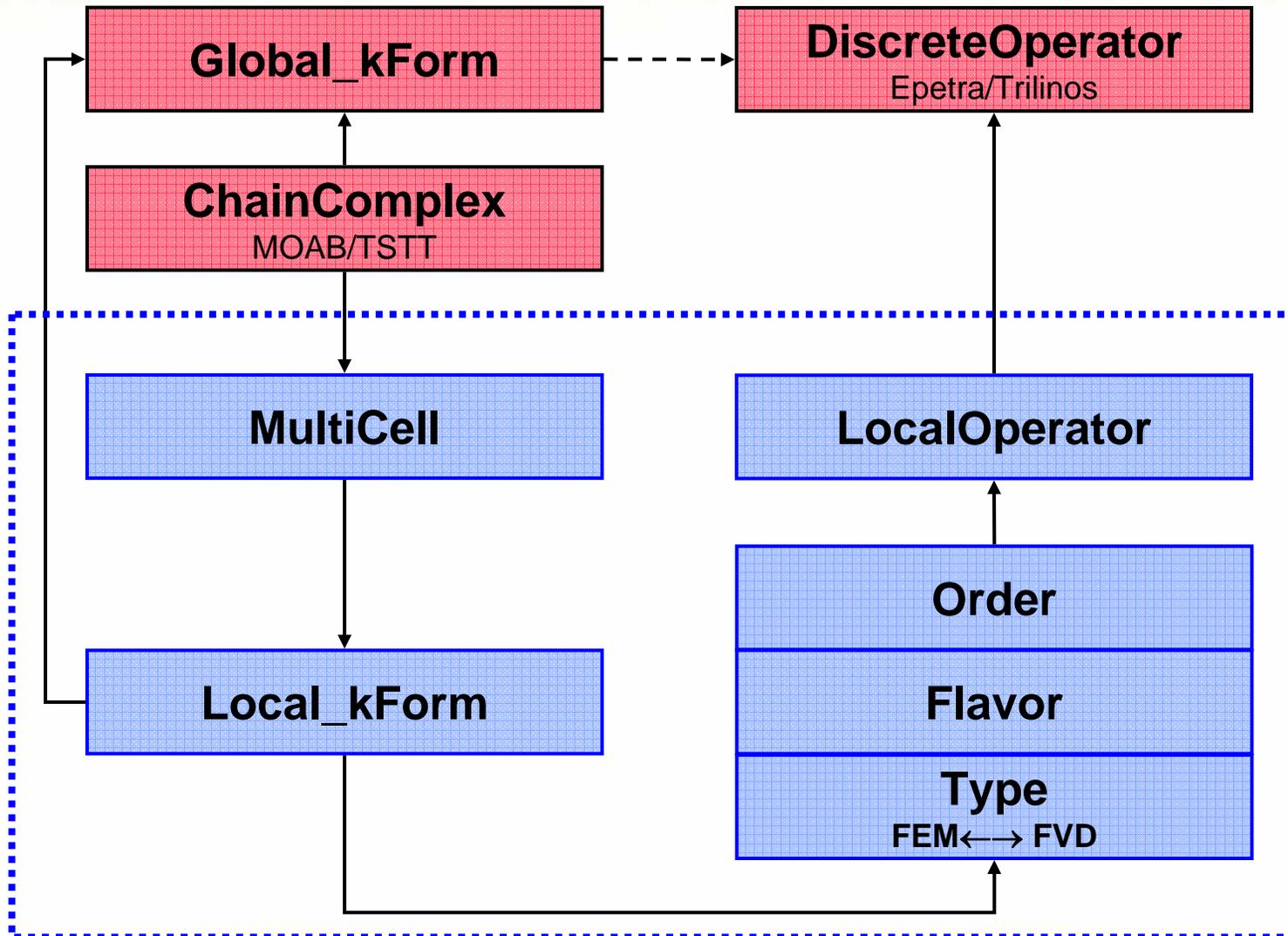
**Pullback:**  $I$  is **high order** but not easy to extend beyond **standard cells**

⇒ **There's no fundamental reason not to have simultaneous access to both...**

# Intrepid\*

Joint work with D. Ridzal, D. Day

*IN*teroperable *T*ools for *R*apid *dE*velo*P*ment of *compa*t*I*ble *D*iscretizations



# Anticipated Applications

## CHARON - X-MHD

**Intrepid** will enable side by side comparisons of FV and mimetic div free methods and FEM using vector potential and B-projection, and discretization tools for extended MHD modeling and simulation (Shadid, Banks, Chacon).

## CHARON - DEVICE

**Intrepid** will be used to test compatible discretizations for device modeling, prototype optimization and control problems, and as a discretization library (Pawlovski, Shadid, Bartlet)

## ALEGRA

**Intrepid** will provide discretization tools for multimaterial ALE modeling and simulation on general polyhedral cells (Robinson, Shashkov, Lipnikov)

## Org.1641 (HEDP Theory)

**Intrepid** will provide discretization tools for Sandia's Pulsed Power modeling and simulation effort (Hanshaw, Brunner, Robinson)

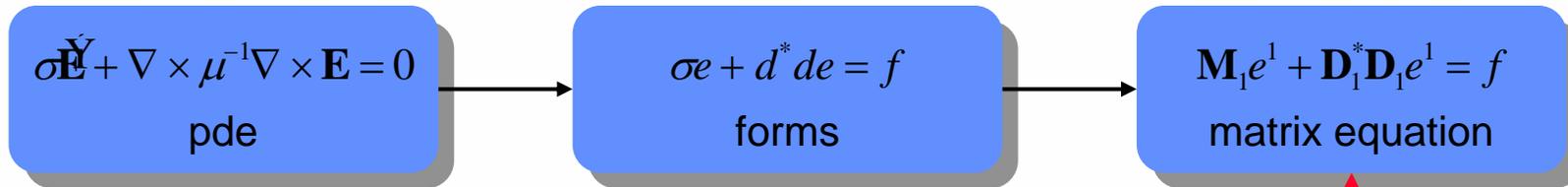
## External:

- ➔ LANL Theoretical Division T-7 (Shashkov)
- ➔ Center for computation & technology, Louisiana State University
- ➔ HERMES project, UT El Paso (Solin)



# Reformulation of Maxwell's equations

Recall the mimetic discretization of the primal equation



Relevant operators acting on 1-cochains:

$$\mathbf{D}_1^* \mathbf{D}_1 = \mathbf{D}_1^T \mathbf{M}_2 \mathbf{D}_1 \quad \text{A curl-curl operator}$$

$$\mathbf{D}_0 \mathbf{D}_0^* = \mathbf{M}_1 \mathbf{D}_0 \mathbf{M}_0^{-1} \mathbf{D}_0^T \mathbf{M}_1 \quad \text{A grad-div operator}$$

$$\left. \begin{array}{l} \mathbf{D}_1^* \mathbf{D}_1 \\ + \\ \mathbf{D}_0 \mathbf{D}_0^* \end{array} \right\} = \left\{ \begin{array}{l} \mathbf{D}_1^T \mathbf{M}_2 \mathbf{D}_1 \\ + \\ \mathbf{M}_1 \mathbf{D}_0 \mathbf{M}_0^{-1} \mathbf{D}_0^T \mathbf{M}_1 \end{array} \right. \quad \text{A Hodge Laplacian}$$

$$e^1 = \mathbf{D}_0 p^0 + \mathbf{D}_1^* b^2 \quad \text{A Hodge decomposition}$$

Requires specialized AMG solvers to deal with Ker(curl)

Sandia app: Z-pinch

QuickTime™ and a TIFF (LZW) decompressor are needed to see this picture.



# Why Reformulate?

ML methods work well for Laplacians  $\Rightarrow$  make curl-curl more “Laplace”-like

- ❑ **Reformulate and then discretize:** first add **grad div** and then discretize

**Misconception:** reformulation allows to use **collocated** methods, e.g., **nodal FE**

**Major issue:** scaling of the Laplacian when  $\sigma$  varies orders of magnitude

$$\nabla \times \mu^{-1} \nabla \times - \sigma \nabla \nabla \cdot \sigma \approx \mathbf{C}_h + \mathbf{G}_h$$

curl curl completely dominates **grad div** when  $\sigma \approx 0$

- ❑ **Discretize and then reformulate:** our approach - add **discrete grad div**

**Key idea:** use different inner product for the Hodge decomposition of 1-cochains

$$e^1 = \mathbf{D}_0 p^0 + \tilde{\mathbf{D}}_1^* b^2 = \mathbf{D}_0 p^0 + \tilde{\mathbf{M}}_1^{-1} \mathbf{D}_1^T \mathbf{M}_2 b^2$$

$$\mathbf{D}_1^T \mathbf{M}_2 \mathbf{D}_1 + \tilde{\mathbf{M}}_1 \mathbf{D}_0 \mathbf{M}_0^{-1} \mathbf{D}_0^T \tilde{\mathbf{M}}_1 \approx \nabla \times \mu^{-1} \nabla \times - \nabla \gamma^{-1} \nabla \cdot$$

$\tilde{\mathbf{M}}_1$  is scaled by 1!

$\mathbf{M}_0$  is scaled by  $\gamma = \mu$

- ❑ **Issue:** does this “mismatched” Laplacian have the same null-space as the true one?

# Why not Reformulate and Then Discretize?

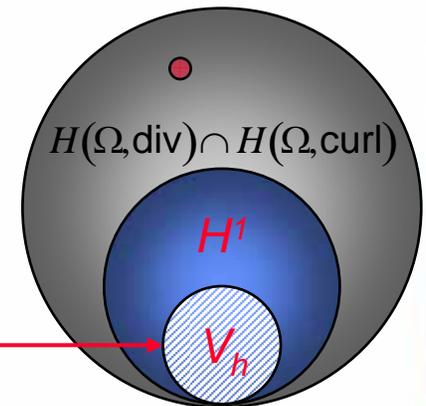
Assume a general unstructured grid without a topologically dual

Reformulated problem

$$(\nabla \times \mathbf{E}, \nabla \times \hat{\mathbf{E}})_{\mu^{-1}} + (\nabla \cdot \sigma \mathbf{E}, \nabla \cdot \sigma \hat{\mathbf{E}})_{\gamma^{-1}} = 0 \quad \forall \hat{\mathbf{E}} \in H(\Omega, \text{div}) \cap H(\Omega, \text{curl})$$

Conforming discretization

$$\begin{array}{l}
 \begin{array}{l}
 \rightarrow V_h \subset H(\Omega, \text{curl}) \Rightarrow [\mathbf{E}_h \times \mathbf{n}] = 0 \\
 \rightarrow V_h \subset H(\Omega, \text{div}) \Rightarrow [\mathbf{E}_h \cdot \mathbf{n}] = 0
 \end{array} \\
 \left. \begin{array}{l}
 H(\Omega, \text{div}) \cap H(\Omega, \text{curl}) \supset V_h \\
 \rightarrow V_h \subset H(\Omega, \text{curl}) \Rightarrow [\mathbf{E}_h \times \mathbf{n}] = 0 \\
 \rightarrow V_h \subset H(\Omega, \text{div}) \Rightarrow [\mathbf{E}_h \cdot \mathbf{n}] = 0
 \end{array} \right\} \begin{array}{l}
 \rightarrow [\mathbf{E}_h] = 0 \Rightarrow V_h \subset H^1(\Omega)
 \end{array}
 \end{array}$$



The problem: in 3D  $H^1$  can have **infinite co-dimension** in  $H(\text{div}) \cap H(\text{curl})$

Reformulate and discretize approaches that work need additional structure:

- **Single mesh:** Manteuffel et. al. - using potentials for E, J, B, H (potentials are more regular)
- **Primal-dual:** Haber et al. - using Yee scheme (curl on primal, div on dual)

# Discretize and Then Reformulate:

## Theorem

Assume that  $e^1$  solves the discrete Maxwell's equation and let  $e^1 = \mathbf{D}_0 p^0 + \tilde{\mathbf{D}}_1^* b^2$ .

The pair  $(a^1, p^0)$ , where  $a^1 = \tilde{\mathbf{D}}_1^* b^2$ , solves the *reformulated* Maxwell's equation

$$\begin{bmatrix} \mathbf{M}_1 + \mathbf{D}_1^T \mathbf{M}_2 \mathbf{D}_1 + \tilde{\mathbf{M}}_1 \mathbf{D}_0 \mathbf{M}_0^{-1} \mathbf{D}_0^T \tilde{\mathbf{M}}_1 & \mathbf{M}_1 \mathbf{D}_0 \\ \mathbf{D}_0^T \mathbf{M}_1 & \mathbf{D}_0^T \mathbf{M}_1 \mathbf{D}_0 \end{bmatrix} \begin{bmatrix} a^1 \\ p^0 \end{bmatrix} = \begin{bmatrix} f \\ g \end{bmatrix}$$

## Theorem

Kernels of the mismatched and standard Laplacian have the same dimension

$$\dim \ker(\mathbf{D}_1^T \mathbf{M}_2 \mathbf{D}_1 + \tilde{\mathbf{M}}_1 \mathbf{D}_0 \mathbf{M}_0^{-1} \mathbf{D}_0^T \tilde{\mathbf{M}}_1) = \dim \ker(\mathbf{D}_1^T \mathbf{M}_2 \mathbf{D}_1 + \mathbf{M}_1 \mathbf{D}_0 \mathbf{M}_0^{-1} \mathbf{D}_0^T \mathbf{M}_1) = 0$$

Proof uses that mimetic spaces *inherit the cohomology* of the analytic spaces and so:  $\dim \ker(\Delta) = \dim \ker(D)$  for contractible domains.

**Exercise:** try proving this directly using only linear algebra!

## Related approaches:

⇒ Hiptmair, Xu, Kolev, Vassilevski: **auxiliary space preconditioners** use the so-called **regular decomposition** of  $H(\text{curl})$  instead of the Hodge decomposition;

⇒ Bossavit: same edge inner product, uses lumped mass over dual volumes



# Solver Performance

Because the blocks of the reformulated system are the **edge Laplacian** and **node Laplacian**, we can apply a standard AMG for the Laplace eq. to solve the problem (after applying edge to node interpolant to 1-1 block).

QuickTime™ and a TIFF (LZW) decompressor are needed to see this picture.

## A-slot regression test problem: ALEGRA (C. Siefert)

- ➔ mesh refinement 1,4,8 times
- ➔ conductivity:  $\sigma=1$  (void);  $\sigma=6 \cdot 10^6$  (material))

METH DOF	ML-edge elements		Reformulated	
	2 Chev.	3 Chev.	2 Chev.	3 Chev.
2,300	28	18	21(15%)	17(6%)
140,528	43	35	33(28%)	26(28%)
1,123,696	66	53	54(12%)	41(23%)

QuickTime™ and a TIFF (LZW) decompressor are needed to see this picture.

- ➔ ML = specialized, **highly tuned** AMG for edge elements (Trilinos)
- ➔ Reformulated = off the shelf AMG for Poisson equation, **few tricks!**



# Solver Performance

## $\sigma$ sensitivity:

Grid	cplx	$\sigma_2$				
		$10^0$	$10^{-2}$	$10^{-4}$	$10^{-6}$	$10^{-8}$
$9^2$	1.07	7	7	7	7	7
$27^2$	1.20	12	12	12	12	12
$81^2$	1.25	15	16	16	16	16
$243^2$	1.27	17	18	18	18	18

## $\mu$ sensitivity:

Grid	cplx	$\mu_2$						
		$10^0$	$10^{-1}$	$10^{-2}$	$10^{-3}$	$10^1$	$10^2$	$10^3$
$9^2$	1.07	7	7	7	7	7	8	9
$27^2$	1.34	12	12	13	12	12	13	13
$81^2$	1.24	15	18	19	20	19	21	21
$243^2$	1.27	17	22	25	26	24	29	31

## ML trivia

The new solver has been run in parallel to ~2000 processors with about a 65% (weak scaling) efficiency on a model problem.

## Under the hood

- ➔ The edge Laplacian has the right null-space but lives on the edges - needs to be transferred to a nodal Laplacian before we apply OTS AMG.
- ➔ **Trick 1:** piecewise edge constants on first fine level only (theory “says” that’s OK) are used to define a cheap grid transfer to nodes to avoid complexity.
- ➔ **Trick 2:** the fine grid smoother ignores the discrete gauge term! Hence we never need to form it explicitly, effectively it gauges the coarse grid operator.



# Conclusions

- ❑ **Compatible** discretizations inherit key structural properties of analytic spaces & operators
  - **discrete models are physical**  $\Rightarrow$  **have intrinsic control over information loss**
- ❑ We presented a **framework for compatible discretizations** where:
  - All operations are defined by **two mappings**: **reduction**  $R$  and **reconstruction**  $I$
  - The central concept is the natural inner product
- ❑ **The framework has two basic operation types**
  - **Natural** derivative, inner product, wedge product,...
  - **Derived** adjoint derivative, Hodge Laplacian,...
- ❑ **The framework has important mimetic properties:**
  - **discrete vector calculus**
  - **combinatorial Hodge theory**
- ❑ **The framework helped us to**
  - Recognize that differences between FV, FD and FE are largely **superficial**
  - Derive a powerful **abstraction of the discretization process** and use it to develop new software design for interoperable discretization tools
  - Reformulate the discrete Maxwell's equations so as to make them better suited for ML solvers

