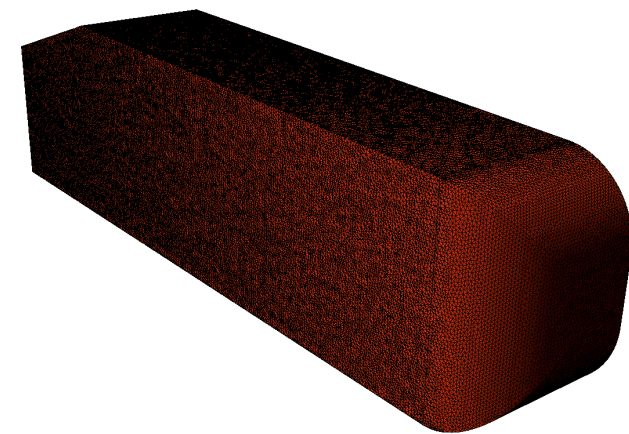
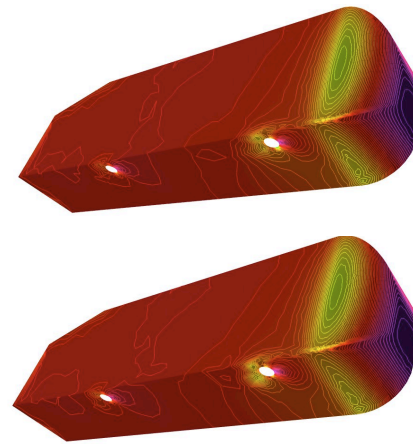
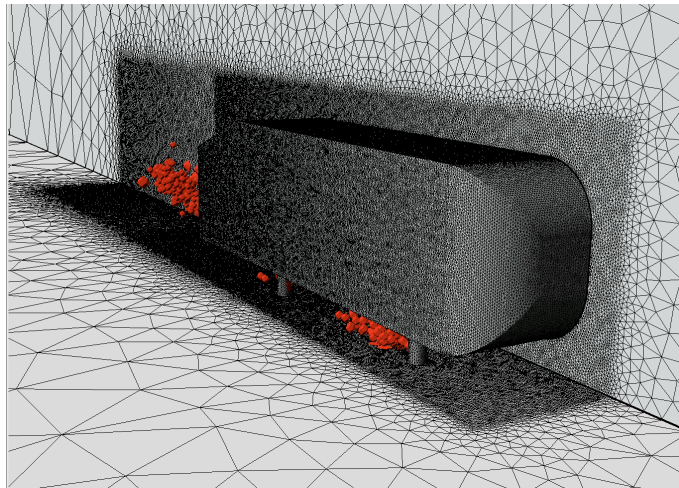


# Model Order Reduction in Albany & Trilinos



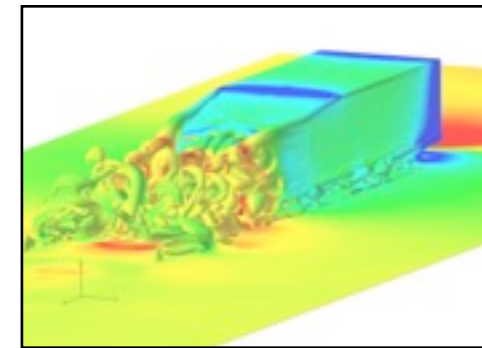
Julien Cortial and Kevin Carlberg  
Department 8954

Albany Developer Meeting  
October 02, 2012

# Motivation and objectives

## *High-fidelity simulation*

- ◉ Detailed analysis of a few configurations



## barrier

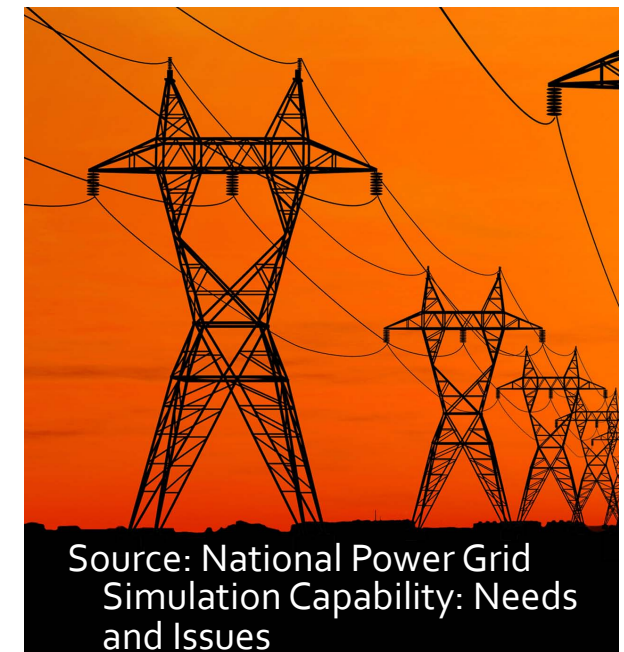
## *Time-critical applications*

- ◉ Many query: UQ, optimization
- ◉ Near-real-time analysis: control

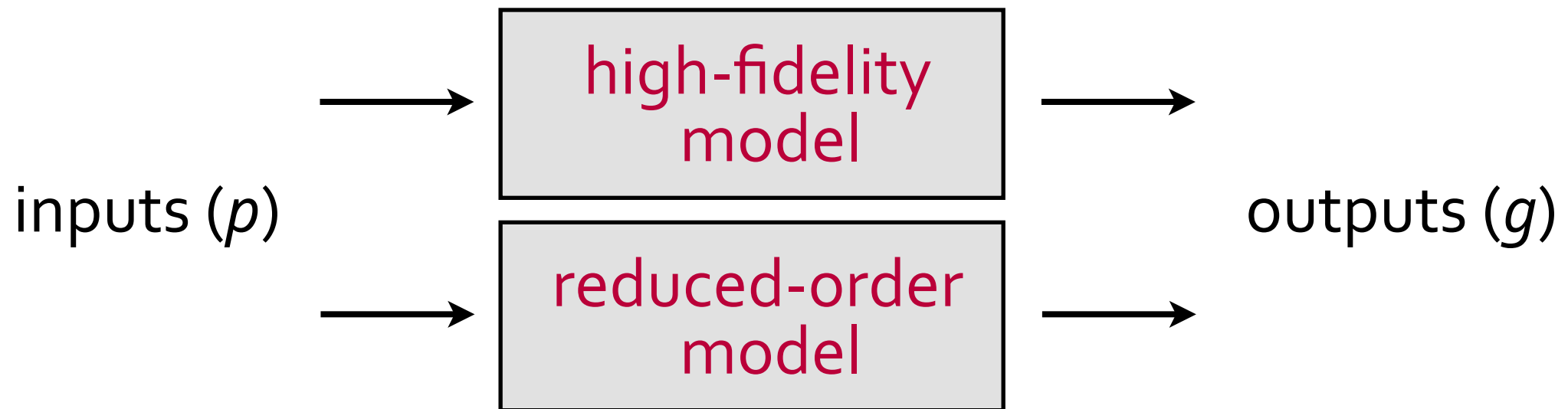
Goal: break barrier using model order reduction

## *Project outline*

- ◉ Improve the state of the art
- ◉ Develop software compatible with Sandia's infrastructure
  - experimental support for ongoing research
  - extension to broad range of applications and analyses



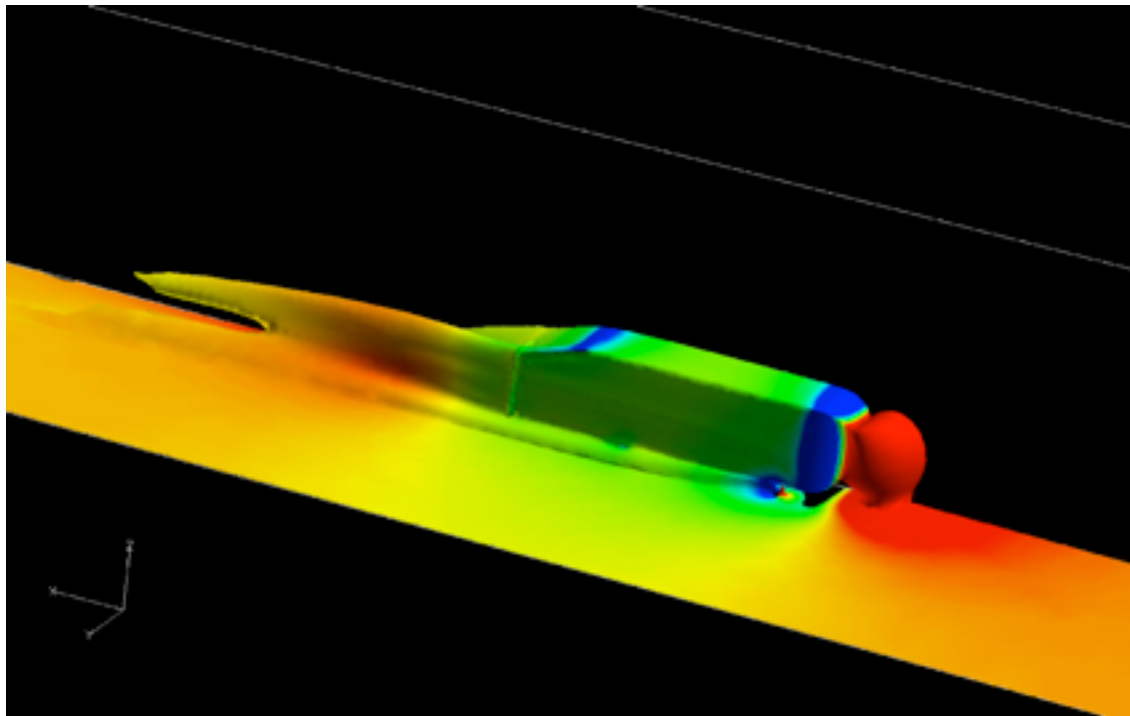
# Model order reduction



- A class of surrogate-modeling method
  - preparatory offline phase
  - inexpensive online computations
- Extracts *dominant physical behavior* from high-fidelity model
- Based on algebraic formulation  $F(x, \dot{x}, t, p) = 0$ 
  - basic methods are general, non-intrusive
- Better performance requires intrusiveness

# Offline data collection

## 1. Collect snapshots of the state vector



$$S = \text{[Red vertical bar]}$$

## 2. Compression

### a. compute singular value decomposition

$$S = U \Sigma V^T$$

### b. truncate

$$U \rightarrow \Phi$$

# POD-Galerkin reduced-order model

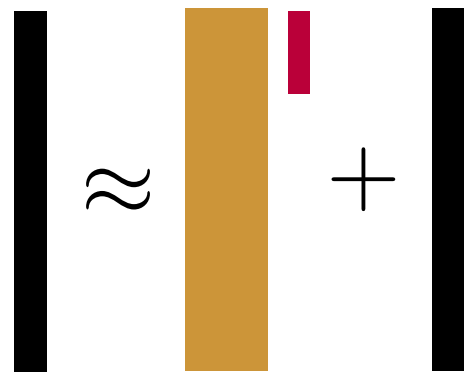
- High-fidelity model

$$F(x, \dot{x}, t, p) = 0$$

- Galerkin projection

reduce # unknowns

$$x \approx \Phi y + \bar{x}$$



reduce # equations

$$\Phi^T F(\Phi y + \bar{x}, \Phi \dot{y}, t, p) = 0$$

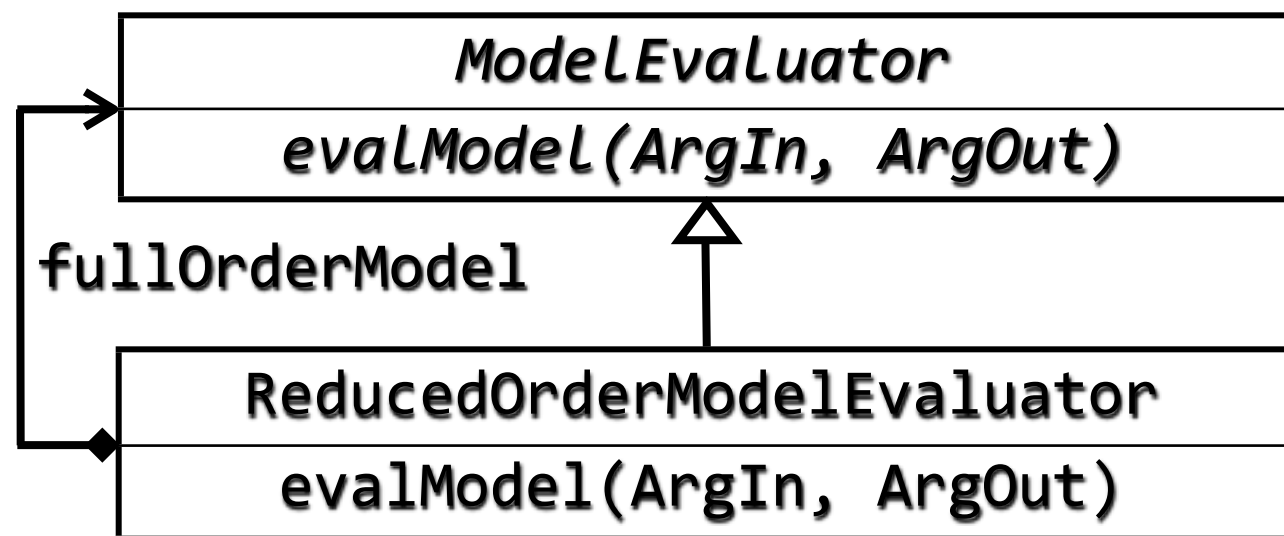


- Reduced-order model

$$F_r(y, \dot{y}, t, p) = 0$$

# Non-intrusive implementation

- Exploit intrinsic modularity, standard interfaces in Trilinos / Albany
  - Loose coupling
  - Textbook use case for Composite / Decorator patterns
- ModelEvaluator hierarchy



$$F(x, \dot{x}, t, p) = 0$$

$$\Phi^T F(\Phi y + \bar{x}, \Phi \dot{y}, t, p) = 0$$

$$F_r(y, \dot{y}, t, p) = 0$$

- Observer hierarchies
  - snapshot collection
  - output for the reduced order model run



# Complexity reduction

- Small dimension  $\neq$  small cost

- Newton iterations 
$$\left[ \Phi^T J^{(k)} \Phi \right] p^{(k)} = -\Phi^T F^{(k)}$$

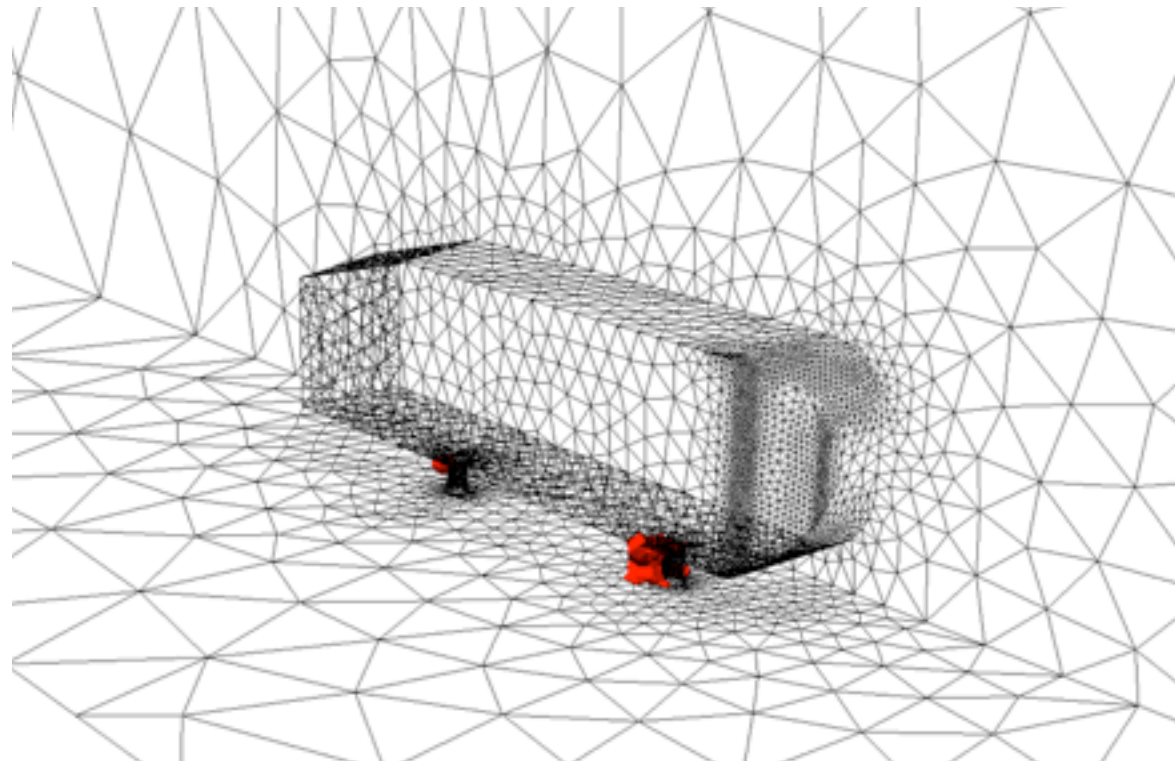
- system of small dimension: inexpensive solve
- system assembly scales with full-order dimension !

- Compute only **a few entries** of the full-order residual

$$\Phi^T F^{(k)} = \sum_{i=1}^N \Phi_i^T F_i^{(k)} \approx 0 \Leftrightarrow \sum_{i_s} \Phi_{i_s}^T F_{i_s}^{(k)} = 0$$

# Sample mesh implementation

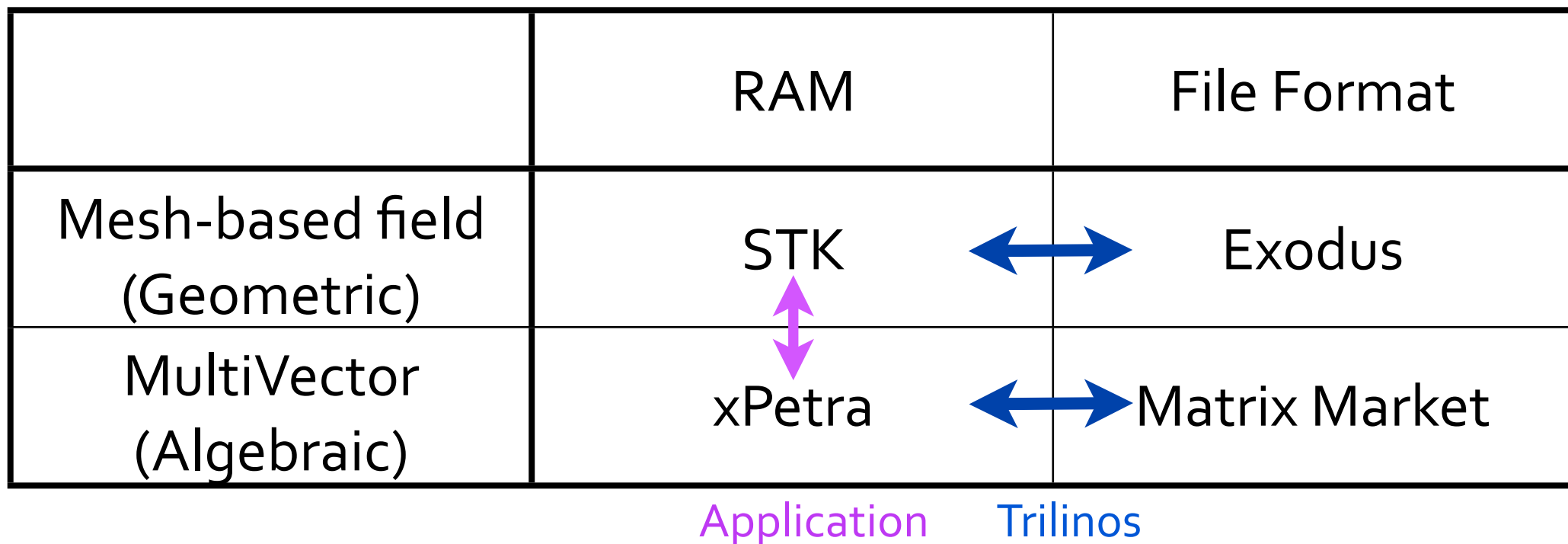
- ◉ *Goals:*
  - reuse existing simulation codes
  - minimize computing cores
  - scalability
- ◉ *Key:* evaluate **only a few entries** of the residual online
- ◉ *Idea:* sample a **small subset** of mesh





# Data representation & storage

- Reduced-order basis representation



- Towards a mesh-based representation
  - + Physical interpretation
  - + Flexible computational topology
  - Complexity of implementation
  - Application dependent

# Experience with Albany

- Enable MOR for a broad range of applications & types of analyses
  - **versatility** of the Albany platform
- Minimally intrusive Trilinos-based MOR software
  - general algebraic formulation  $\longleftrightarrow$  abstract Trilinos interfaces
  - **rapid development** in Albany with loose coupling
- Towards more efficient, specialized, intrusive implementations
  - existing software (e.g. STK): both constraint and leverage
  - Albany is a prime **testing ground**