

Periodic Orbits with “4D”

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Space-Time “4D” Approach

Transient Simulation of: $B\dot{x} = f(x, \lambda)$

First solve: $B\frac{x_1 - x_0}{\Delta t} - f(x_1, \lambda) = 0$

Then solve: $B\frac{x_2 - x_1}{\Delta t} - f(x_2, \lambda) = 0$

Then solve: $B\frac{x_3 - x_2}{\Delta t} - f(x_3, \lambda) = 0$

Instead, solve for all solutions at once:

$$g(y, \lambda) = 0$$

where $y = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$

$$g_i = Bx_i - Bx_{i-1} - \Delta t f(x_i, \lambda)$$

... and with Newton solve:

$$\begin{pmatrix} (B - \Delta t J) & 0 & 0 & 0 & 0 \\ -B & (B - \Delta t J) & 0 & 0 & 0 \\ 0 & -B & (B - \Delta t J) & 0 & 0 \\ 0 & 0 & -B & (B - \Delta t J) & 0 \\ 0 & 0 & 0 & -B & (B - \Delta t J) \end{pmatrix} \begin{pmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \\ \Delta x_4 \\ \Delta x_5 \end{pmatrix} = \begin{pmatrix} -g_1 \\ -g_2 \\ -g_3 \\ -g_4 \\ -g_5 \end{pmatrix}$$

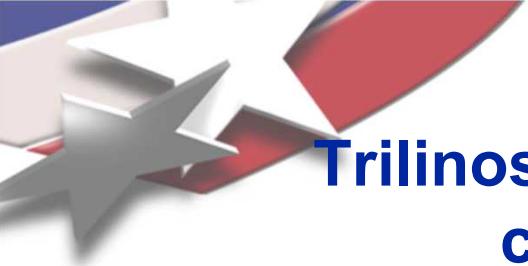
$$"K\Delta y = -g"$$



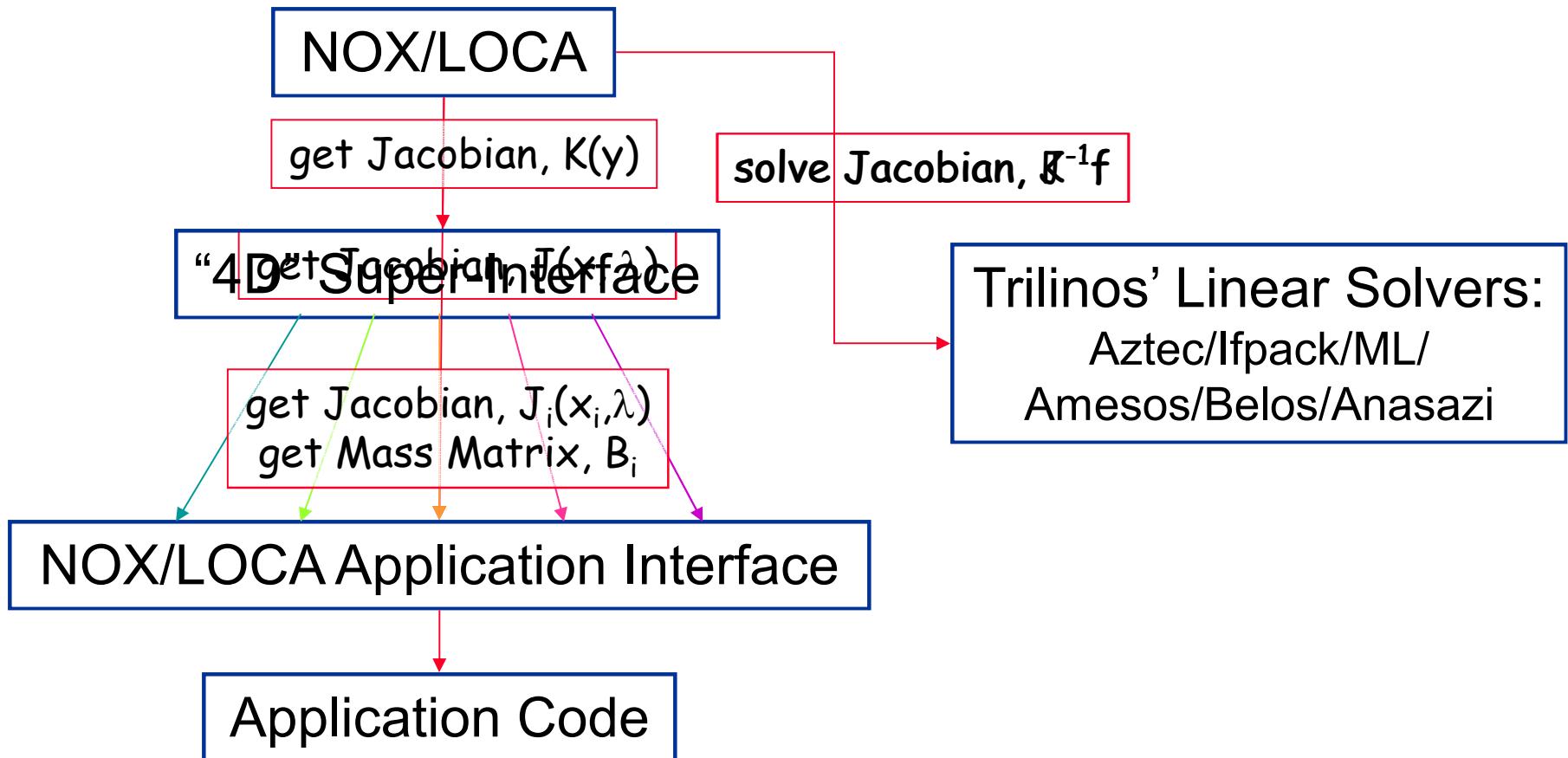
Why do a Space-Time “4D” Approach?

1. Design: Formulating transient problems as “steady” problems in space-time allows for use:
 - Continuation
 - Adding Design Constraints
 - Optimization (Optimal control)
2. Scalability: Can implement parallelism over time domain
3. Periodic Orbits: Only way with analytic Jacobian
4. Verification: Can do mesh refinement in 1D time axis

➤ There are many potential hurdles and liabilities with this approach (high memory usage, poor scalability, incompatibility with mesh adaptivity, time step control,...)



Trilinos Implementation of Space-Time Capability: create a “Super” NOX/LOCA interface





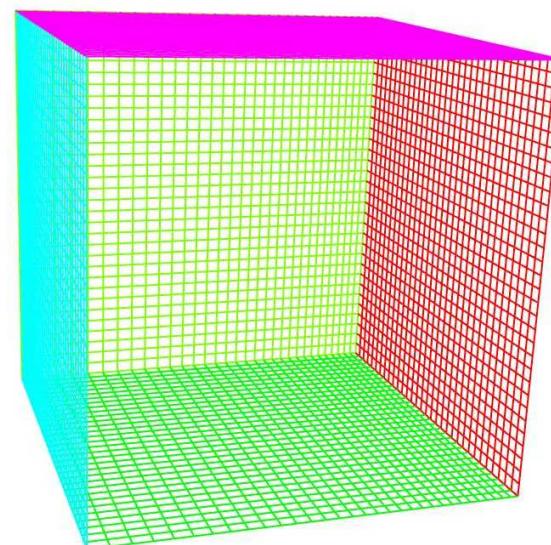
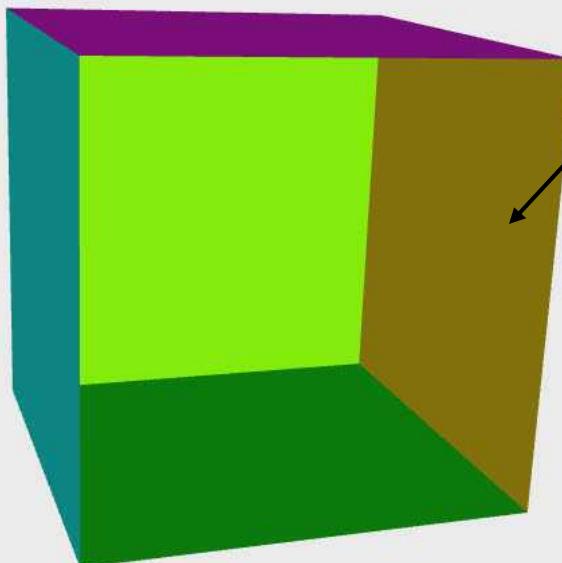
Periodic Solutions of Large Problems

$$\begin{vmatrix} (B - \Delta t J) & 0 & 0 & 0 & -B \\ -B & (B - \Delta t J) & 0 & 0 & 0 \\ 0 & -B & (B - \Delta t J) & 0 & 0 \\ 0 & 0 & -B & (B - \Delta t J) & 0 \\ 0 & 0 & 0 & -B & (B - \Delta t J) \end{vmatrix} \begin{vmatrix} \Delta x_1 \\ \Delta x_2 \\ \Delta x_3 \\ \Delta x_4 \\ \Delta x_5 \end{vmatrix} = \begin{vmatrix} -g_1 \\ -g_2 \\ -g_3 \\ -g_4 \\ -g_5 \end{vmatrix}$$

Demonstration Problem

3D Buoyancy-driven flow in a box with periodic thermal forcing

$$T(1.0, y, z, t) = \sin(2\pi t)$$

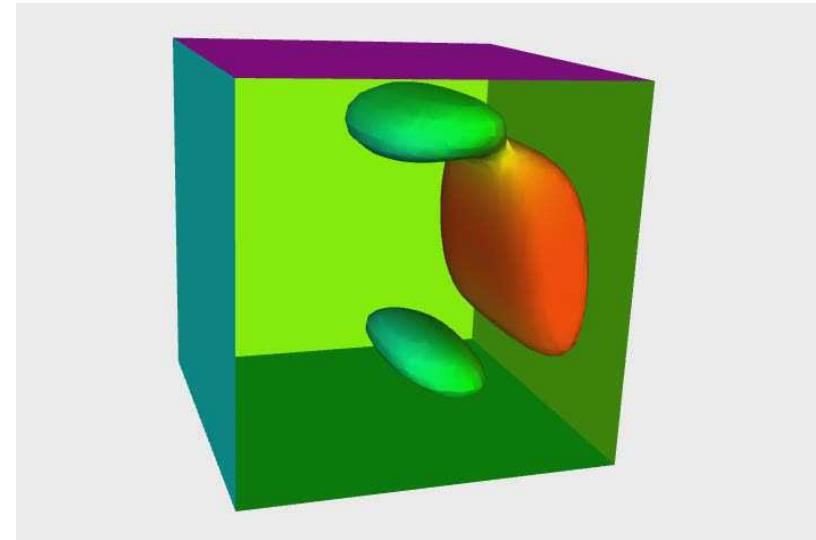
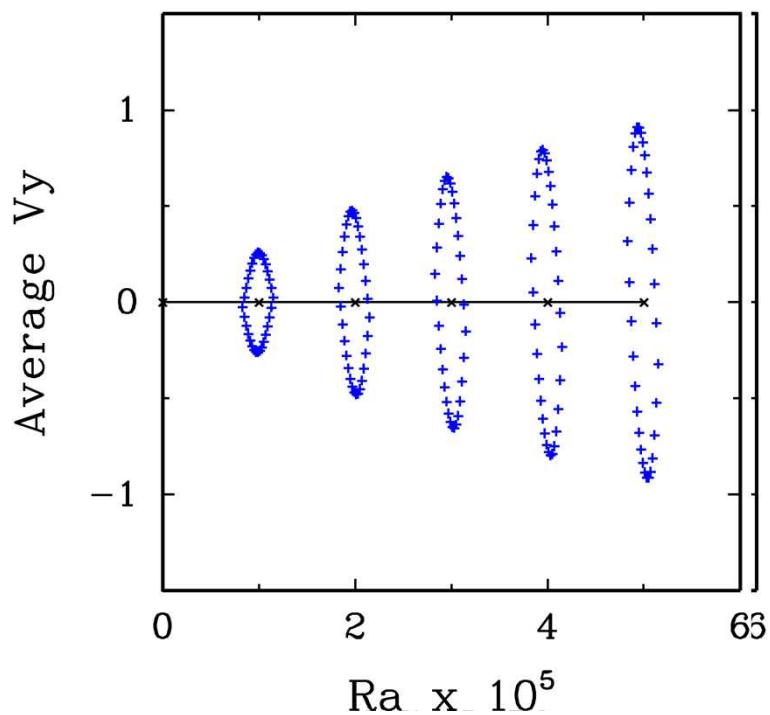


MPSalsa

$32 \times 32 \times 32 \times 32$ grid, with 5 PDEs
F.E. F.D.

Demonstration Problem: Continuation Run

Ra (Rayleigh Number) = Buoyancy force compared to dissipation





Demonstration Problem: Timings

32 x 32 x 32 x 32 grid, with 5 PDEs = 5.75M unknowns,
(with 270 nonzeros/row)

```
-----XYZT Partition Info-----
  NumProcs          = 64
  Spatial Decomposition = 2
  Number of Time Domains = 32
  Time Steps on Domain 0 = 1
  Number of Time Steps = 32
  -->Solving for a Periodic Orbit!
-----
```

Preconditioner calc (ifpack): 31 sec

GMRES: 23 – 130 iters, 9 – 75 secs

Newton iters: 3-5

Continuation steps: 4

Total time: 30 min



Floquet Analysis of Periodic Orbit

Floquet Multipliers:

- Given a periodic solution
- Integrate perturbations through 1 period
- Eigenvalues of this operator are called Floquet Multipliers σ_i
- Orbits are stable if all $\|\sigma_i\| < 1$

- Arnoldi iteration of eigensolver:

$$\begin{vmatrix} (B - \Delta t J) & 0 & 0 & 0 & 0 & \cdots & Bv_i \\ -B & (B - \Delta t J) & 0 & 0 & 0 & \cdots & 0 \\ 0 & -B & (B - \Delta t J) & 0 & 0 & \cdots & 0 \\ 0 & 0 & -B & (B - \Delta t J) & 0 & \cdots & 0 \\ 0 & 0 & 0 & -B & (B - \Delta t J) & \cdots & 0 \end{vmatrix} v_{i+1} = \begin{vmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{vmatrix}$$



Where does the code currently reside?

packages/epetraext/src/block:

- EpetraExt::BlockVector
- EpetraExt::BlockCrsMatrix
- EpetraExt::MultiMpiComm

packages/nox/src-loca/src-epetra:

- LOCA::Epetra::Interface::xyzt
- LOCA::Epetra::xyztPrec
- LOCA::Epetra::AnasaziOperator::Floquet*

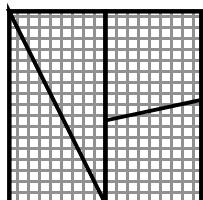
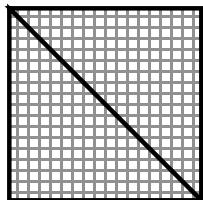
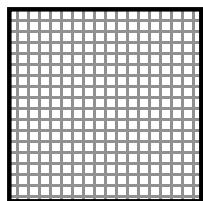
packages/nox/examples/epetra/LOCA_Brusselator_xyzt:

- Example.C

*Floquet capability only in development branch

Space and Time can be partitioned independently, Ex: 4 Time Steps on 4 Procs

Spatial Partition

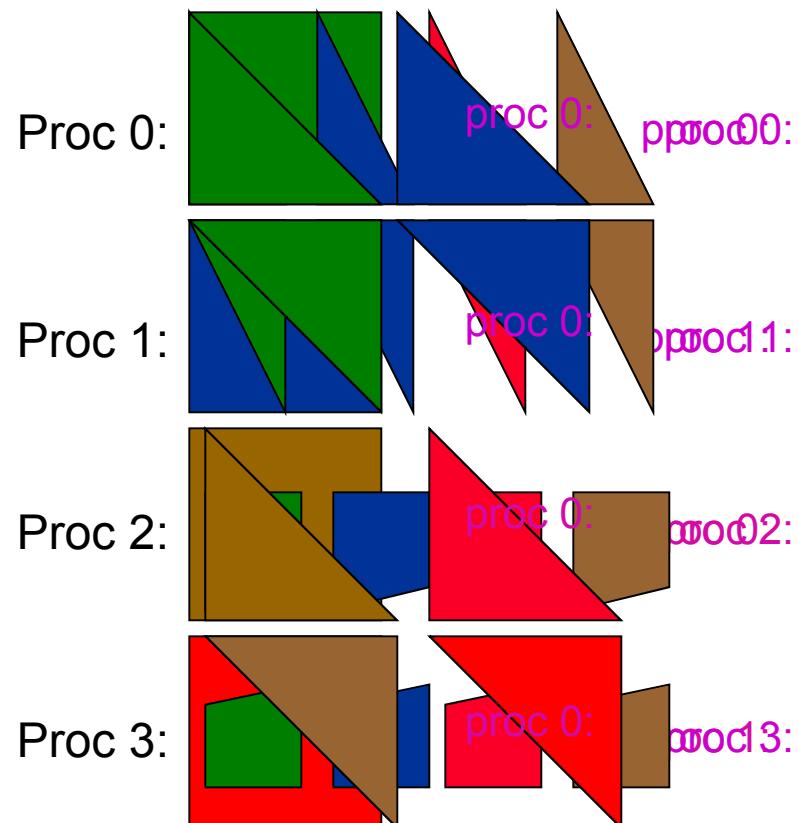


Each processor owns 1 time step for the entire spatial domain

Each processor owns 2 time steps for $\frac{1}{2}$ of the spatial domain

Each processor owns 4 time steps for $\frac{1}{4}$ of the spatial domain

Space-Time Partition



mpirun -np 4 salsa infile 4 4

Numerical Experiment #1

How much can parallelism in time speed up the solve?

- **Fixed Number of Spatial Domains (4)**

- **Processors:** 4 8 16 32 64 128
- **Time Domains:** 1 2 4 8 16 32

MPSalsa:

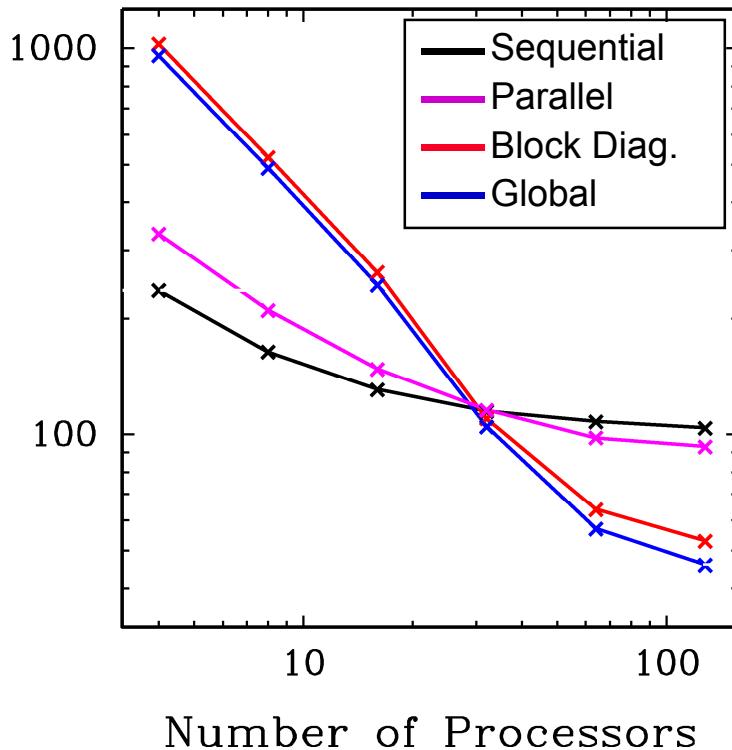
PDEs: 5
FEM: 2D, 64 x 48 elements
Time steps: 32
Unknowns: 509,600

Trilinos:

Newton (NOX) : 3-6 iterations
GMRES (Aztec) : <200 iters
ILUk (Ifpack) : overlap=0, fill=1
Continuation steps (LOCA): 1

5x Speedup

Time (sec)



Numerical Experiment #2

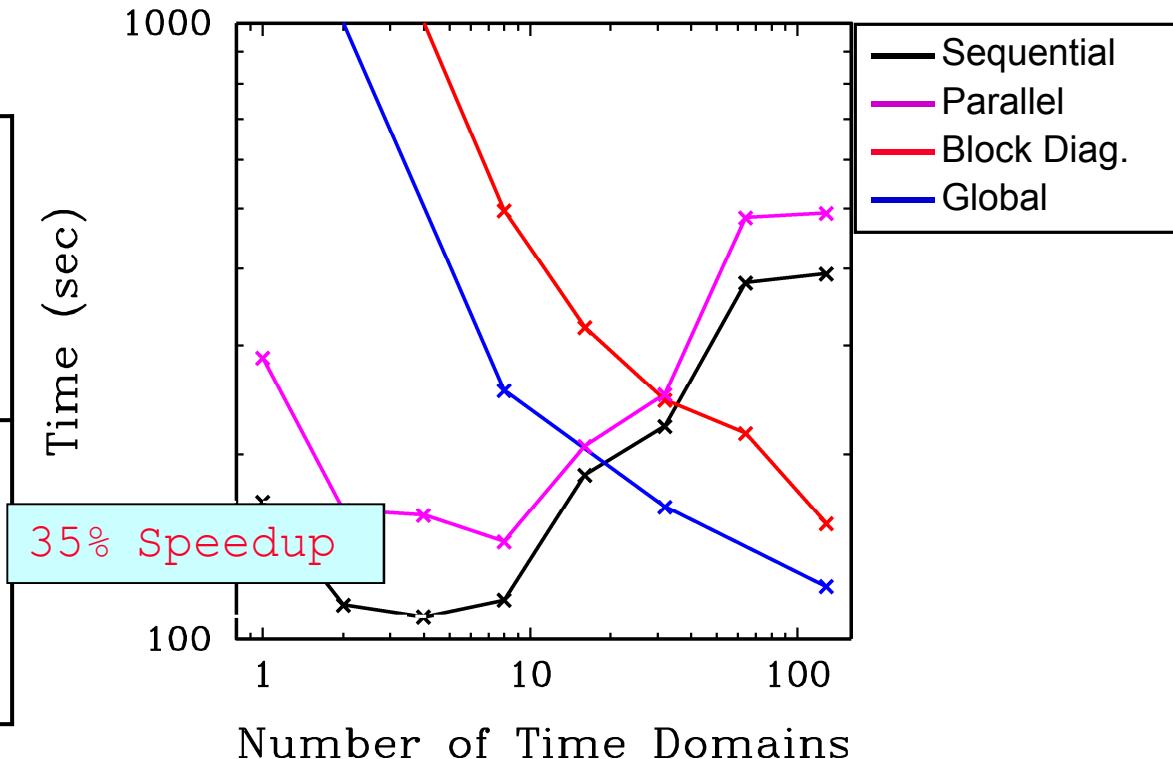
Can space-time parallelism be more effective than just spatial parallelism?

- **Fixed Number of Processors (128)**

– Spatial domains:	1	2	4	8	16	32	64	128
– Time domains:	128	64	32	16	8	4	2	1

MPSalsa:
PDEs: 5
FEM: 2D, 64 x 48 elements
Time steps: 128
Unknowns: 2,038,400

Trilinos:
Newton (NOX) : 4–6 iterations
GMRES (Aztec) : <200 iters
ILUk (Ifpack) : overlap=0, fill=1
Continuation steps (LOCA): 1





Space-Time **capability is integrated** with Trilinos' design and analysis tools

