#### Multilevel Resiliency for PDE Simulations

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#### **Overview**

We'll discuss:

- introduction to multilevel methods
- In multilevel methods on the extreme scale
- Multilevel checkpointing framework
- Multilevel error detection

With the goal of showing:

- algorithmic resiliency advantages of multigrid
- a basic scheme for multigrid-based checkpointing and error detection
- a way forward for extreme-scale algorithmic resiliency

## **Multigrid Preliminaries**

**Multigrid** is an O(n) method for solving linear algebra problems by defining a hierarchy of scale. A Multigrid method is constructed from:

- a series of discretizations
  - coarser approximations of the original problem
  - constructed algebraically or geometrically
- intergrid transfer operators
  - restriction **R** and injection  $\hat{\mathbf{R}}$  (fine to coarse)
  - prolongation P (coarse to fine)
- Smoothers (S)
  - correct the high frequency error components
  - Richardson, Jacobi, Gauss-Seidel, etc.
  - · Gauss-Seidel-Newton or optimization methods

## Multigrid

• Multigrid methods uses coarse correction for large-scale error



Algorithm  $MG(\mathbf{A}, \mathbf{b})$  for the solution of  $\mathbf{A}\mathbf{x} = \mathbf{b}$ :

$$\mathbf{x} = \mathbf{S}^m(\mathbf{x}, \mathbf{b})$$
pre-smooth $\mathbf{b}^H = \mathbf{R}(\mathbf{r} - \mathbf{A}\mathbf{x})$ restrict residual $\hat{\mathbf{x}}^H = MG(\mathbf{R}\mathbf{A}\mathbf{P}, \mathbf{b}^H)$ recurse $\mathbf{x} = \mathbf{x} + \mathbf{P}\hat{\mathbf{x}}^H$ prolong correction $\mathbf{x} = \mathbf{x} + \mathbf{S}^n(\mathbf{x}, \mathbf{b})$ post-smooth

**Resilient FAS** 

# Full Multigrid(FMG)



- start by going directly to coarse
- do number of V-cycles with each going one finer
- x is injected to finer levels as visited
- truncation error within one cycle
- highly efficient solution method



Algorithm  $FAS(\mathbf{F}, \mathbf{x}, \mathbf{b})$  for the solution of  $\mathbf{F}(\mathbf{x}) = \mathbf{b}$ :



#### au Correction

•  $\mathbf{F}^{H}(\mathbf{x}^{H}) = \mathbf{R}\mathbf{b} + [\mathbf{F}^{H}(\hat{\mathbf{R}}\mathbf{x}) - \mathbf{R}\mathbf{F}(\mathbf{x})]$  contains the term we call  $\tau$ 

 $\boldsymbol{\tau} = \mathbf{F}^{H}(\hat{\mathbf{R}}\mathbf{x}) - \mathbf{RF}(\mathbf{x})$ 

- $\bullet\,$  encodes the "difference" between problems F(x) and  $F^{H}(x^{H})$
- exact fine solution is solution to τ-corrected coarse problem
- au tells us how the fine problem can improve the coarse problem
- τ has same size as coarse solution
- au is the magic that makes this whole talk possible

### Extreme-Scale Multigrid: Redundant Coarse Problems

- simplest idea: local redundancy
- calculate coarser levels redundantly on subsets of processors
- requires more communication in fewer stages
- coarse problems must be duplicated; requires off-process restriction
- reduced synchronization



## Extreme-Scale Multigrid: Segmental Refinement

- more complicated idea: ficticious fine grid, au-corrected coarse
- originally for 70's very low memory; recently revived for extreme scale
- loop and "zoom" on subdomains
  - construct fine grid problem in cache
  - smooth locally
  - inject
- data dependencies vertical rather than horizontal between levels



## Basic resilience strategy

We assume the following simple model of checkpointing and recovery:



- control
- program stack
  - configuration
- essential
- time-dependent solution
- current optimization iterate
- ephemeral
- assembled matrices
- preconditioners
- residuals

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## **Essential State Recovery**

- <u>coarse</u> level checkpoints are orders of magnitude smaller
- can be stored at greater frequency
- quick recovery of local essential state from coarse history
- FMG recovery needs only nearest neighbor processors

We introduce FAS Checkpointing for rapid recovery of essential state

- minimal and lossy essential state storage
- whole state may be quickly recovered in total failure
- rapid local catch-up for failed processes

# FAS Checkpointing



- essential state: converged solutions at end of timesteps
- checkpoint converged state at level  $\ell_{CP}$
- $\ell_{CP}$  several levels down
- CP several orders of magnitude smaller than converged state

## **FAS Recovery**



- $\bullet~$  recover using FMG anchored at  $\ell_{cp}+1$
- needs only  $\ell_{cp}$  neighbor points
- τ correction is local
- FMG recovery only accesses levels finer than  $\ell_{CP}$
- Only failed processes and neighbors participate in recovery

### Other uses for coarse checkpoints

potential advantages to having coarse solutions around:

- lightweight high-time-resolution snapshots
- transient adjoint computation
- postprocessing
- coarse in-situ visualization

## Redundant Coarse-Grid Error Detection

The redundant coarse problem may be used to trivially check for errors:



However, this is uninteresting and doesn't exploit the algorithm; can we do anything better?

## au-Correction Error Detection



- x solving fine problem F(x) = b
- check residual of  $\hat{\mathbf{R}}\mathbf{x}$  on a  $\tau$ -corrected coarse grid
- As  $\hat{\mathbf{R}}\mathbf{x}$  solves the  $\tau$ -corrected coarse grid problem, residual should be small
  - incorrect result indicates error in fine grid residual evaluation
  - identifies the location of the error

### Conclusions

- multilevel methods allow efficient simulation
- can be leveraged to increase resiliency
- FAS Checkpointing allows for minimal overhead state reconstruction
- FAS Error Detection and Correction can be built into the solves
- other possibilities (Ensemble MG, etc.) loom

We've made progress towards having this working

- segmental refinement experiments and experience (Adams, 2012)
- FAS framework in place in PETSc