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Focus of our research

- Faults introduced by *silent data corruption* (SDC): Stored data is not changed but the result of a computation, e.g., for a = b = 2 we receive c = a + b = 5
- Node-losses which are a variation of hard faults/failures

Classical techniques

- Reliability in hardware (ECC protection etc.) too power-hungry
- Checkpoint-restart too memory-intensive (and too slow)
- Triple modular redundancy too power-hungry, but: can be more energy-efficient and applicable for large fault rates

Another approach

Algorithm-based fault-tolerance







Algorithm-based fault-tolerance

- Exploit algorithmic properties to detect and correct faults on-the-fly
- Can be more efficient than middleware-based solutions Benefit: Provable error bounds possible

Challenges

- Requires custom modifications for each class of methods
- Overhead in the fault-free scenario should be small
- False-positives should be rare without much impact on convergence
- MPI: Faults can result in node-losses
 ⇒ Once a rank dies, the universe is dead in current MPI3

Initial focus

Multigrid because it is an optimal solver for elliptic problems.









Observations (node-loss)

- A fault is comparable to a restart of the multigrid solver
- Multigrid converges always if the fault-rate is not to high
- Note: SDC and node-losses results in similar behaviour









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- A fault is comparable to a restart of the multigrid solver
- Multigrid converges always if the fault-rate is not to high
- Note: SDC and node-losses results in similar behaviour
- ⇒ Multigrid is provable self-stabilising: Starting from any state the algorithm comes back to a valid state and eventually converges afterwards







Ongoing projects

Compressed checkpointing

- Using compression techniques to decrease checkpoint size
- Locally restore lost or faulty data from compressed checkpoint
- Improve restoration by solving local auxiliary problems

O SDC-tolerant multigrid

- Increase the inherent robustness of multigrid with respect to bit-flips
- Apply a local smoothing stage protection to detect and repair soft faults

User level exception handling

- Propagate exceptions with MPI to always ensure same state on all ranks
- Necessary for an efficient implementation of the other projects
- User-friendly asynchronous C++ MPI interface for parallel exception handling







Research goal

Reduce size of checkpoints and restore lost data efficiently

Multigrid compression

- Use multigrid transfer operators to compress checkpoint
- Data reduction in d dimensions: $\sim 2^d$ per level (backup depth)
- Restore lost data with prolongated checkpoint









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SimTec





Limits of multigrid compression



- Discretisation error dominates at some point
- Dominates earlier for highly compressed data
- Factor between L^2 -quality and L^2 -error depends on amount of repaired data







Problem

- Late faults: Convergence cannot be restored with highly compressed backups
- Recurrent faults need even less-compressed checkpoints

Solution

- Solve an auxiliary problem with Dirichlet boundary to improve accuracy
- Use decompressed data as initial guess

Auxiliary problem (compare Huber et al.¹) Extend faulty indices $\mathcal{F} \subset \mathbb{N}$ by connectivity pattern of Operator A to \mathcal{J} and solve

$$\mathbf{A}(\mathcal{J}, \mathcal{J})\tilde{x}(\mathcal{J}) = b(\mathcal{J}) \qquad \text{in } \mathcal{F} \\ \tilde{x} = x \qquad \text{on } \mathcal{J} \backslash \mathcal{F}$$

iteratively with initial guess $\tilde{x} = x_{cp}$ in \mathcal{F} .



¹M. Huber, B. Gmeiner, U. Rüde, B. Wohlmuth, **Resilience for Massively Parallel Multigrid Solvers**, SIAM Journal on Scientific Computing, 2016







Summary: Multigrid compression

- Multigrid compressed checkpoints can be used to recover from faults
- Early fault: Highly compressed data is sufficient
- Late fault: Compression rate has to be decreased
- Eventually an auxiliary problem has to be solved to ensure convergence
- The decompressed data is a good initial guess for this auxiliary problem
- Same idea could be used with other hierarchic methods

But

Multigrid is good preconditioner, but rarely a standalone solver

D. Göddeke, M.A., D. Ribbrock, Fault-tolerant finite-element multigrid algorithms with hierarchically compressed asynchronous checkpointing, Parallel Computing, 2015







Aim: Extend the idea to other solvers and methods

Restore lost/faulty data from different kind of checkpoints

- Checkpoints based on different (lossy) compression techniques:
 - Multigrid compression
 - SZ compression
- Checkpoints with different frequency
- Different restoration approaches
 - Local restoration based on compressed checkpoint
 - Global roll-back to compressed checkpoint
 - Improved restoration by solving auxiliary problem

Task

Compare quality and efficiency of repair using different combinations







SZ compression

- Save initial point value (with reduced accuracy): unpredictable data
- Predict next point based on previous processed points via an interpolation based on *n* layers
- Compare predicted value with real value and improve it through a *Huffman*-like coding
- If still not 'close enough' data is stored as unpredictable and compressed via binary-representation analysis

Advantages and disadvantages

- Adaptive controllable compression rate (via parameter)
- More computational overhead than inherent multigrid compression
- No random-access to single decompressed values

D. Tao, S. Di, Z. Chen and F. Cappello, Significantly Improving Lossy Compression for Scientific Data Sets Based on Multidimensional Prediction and Error-Controlled Quantization, Computing Research Repository, 2017







Numerical tests

• Anisotropic diffusion in 2D

$$-\nabla \cdot \begin{pmatrix} 1 & 0 \\ 0 & 0.01 \end{pmatrix} \nabla u = b$$

- Linear finite elements on grid with 1500×1500 degrees of freedom
- Solver: 4-way parallel CG without overlap (slices)
- Preconditioner: Algebraic multigrid (one V-cycle)
- MG compression of 3 levels; SZ compression tolerance is adaptive:

locale_def_norm_at_backup_time / def.size() $\ast\,10^{-3}$

Auxiliary solver reduction to

global_def_norm_at_backup_time / $\sqrt{(\texttt{#cores})}*10^{-(\texttt{age_of_backup}+1)}$

Challenge

Recover from a fault in the solver with a compressed checkpoint















Summary (early fault)

- Early on the advantage of using a backup is small ($\leq 10\%$ runtime)
- No backup: Global restart is better than local restoration with zeros
- With backup: Local restoration seems to be superior
- Lower backup frequency (\rightarrow older backups) makes restoration worse
- MG compression is more robust with respect to age/delay
- Auxiliary problem can nearly restore convergence behaviour:
 - MG and SZ compressed initial solution reduces iteration count significantly
 - Age of backup is important
- In case of multiple data-losses a lower iteration count could be important















Summary (late fault)

- Multigrid compression seems not competitive in this scenario
- SZ compression works good and maintains a reasonable compression factor
- Local restoration works better than global roll-back
- A delay can deteriorate the quality of repair
 But: Local SZ restoration with frequency 3 is better than global roll-back with frequency 1 and has less communication and compression overhead
- Auxiliary solver improves quality significantly but increases runtime
- Greater delay/depth/compression still increases iteration count of auxiliary solver

Open task

Develop a performance model to find the most effective combination







Research goal

Increase the robustness of multigrid with respect to silent data corruption

Observation Most time is spent within the smoothing stage



Idea

- Don't ensure correctness value by value
- Only verify if output of the smoothing stage is 'good enough'
- Use invariants of Full Approximation Scheme multigrid (FASMG) to test output
- Protect remaining part (transfer phase and coarse grid correction) with traditional checksums







STMG algorithm

Call : STMG $(k, \mathcal{A}, b, u^{(0)})$ 1 $u^{(\nu)} = S^{\nu} (u^{(0)}, b)$ // pre-smoothing 2 $r_k = b - \mathbf{A}_k u^{(\nu)}$ 3 check_and_repair_res (r_k, k) 4 $r_{k-1} = \mathbf{R}_{k-1}^{k} r_{k}$ 5 $\tilde{v}_{h-1}^{(0)} = \mathbf{I}_{h-1}^{k} u^{(\nu)}$ 6 $r_{k-1} = r_{k-1} + \mathbf{A}_{k-1} \tilde{v}_{k-1}^{(0)}$ 7 $\tilde{v}_{k-1} = \text{STMG}(k-1, \mathcal{A}, r_{k-1}, \tilde{v}_{k-1}^{(0)})$ // coarse grid correction 8 $c = \tilde{v}_{k-1} - \tilde{v}_{k-1}^{(0)}$ 9 check_and_repair_cor(c, k-1) 10 $\tilde{u}^{(\nu)} = u^{(\nu)} + \mathbf{P}_{L}^{k-1}(c)$ 11 $u = S^{\mu} (\tilde{u}^{(\nu)}, b)$ // post-smoothing 12 if on fine grid then check_and_repair_res $(b - \mathbf{A}_L u, k)$ 13 14 end

- *k* denotes the current grid level
- \mathbf{P}_{k}^{k-1} is the prolongation operator
- S^ν is the smoother which is applied ν times
- Direct solver on coarsest grid







Check and repair algorithm (correction)

- Check output of smoother through element-wise comparison
- Threshold based on residual/correction norm (scaled by tolerance factor): Converges monotonously to zero if operator is s.p.d.
- Transfer (scale) to next level grid with transfer operator norm
- Store 'faulty' indices in set ${\cal L}$ and repair:

- Residual check and repair works similar but easier
- Assumption: Coarse grid solver output is fault-free







Numerical tests

- V-cycle multigrid with 4 + 4 Jacobi smoothing steps
- 1 million degrees of freedom, Q_1 Lagrange Finite Elements
- 4000 different fault scenarios per test problem
- Fault probability of 10^{-7} per degree of freedom
 - \Rightarrow Approximately once every 10th smoothing step on fine grid
 - \Rightarrow Approximately twice every multigrid iteration

	diff	diff-conv	andiff	andiff-conv-reac
fault-free	4	6	14	7
MG (div.)	4.225 (6.8%)	6.268 (8.4%)	15.111 (21.3%)	7.466 (11%)
STMG	4.038	6.007	14.007	7.017

• Nearly no false-positives: Approximately 15 in 4000 runs

M.A. and D. Göddeke, **Soft fault detection and correction for multigrid**, International Journal of High Performance Computing Applications, 2017







Numerical overhead

- Overhead of FASMG is approximately 20%
- Smoother protection itself results in an overhead of 4%
- Checksums lead to additional 5% (8× Jacobi smoothing)

	unprotected (MG)	unprotected (FASMG)	defect correction (checksums)	smoothing stage (new algorithm)	STMG (both)
time	35.49	43.02	45.23	44.76	46.18
factor	0.825	1	1.051	1.040	1.073
factor	1	1.212	1.274	1.261	1.301

 \Rightarrow Overall overhead of less than 30% compared to classical MG







Applicability

- Geometric and algebraic multigrid (AMG)
- Standalone and as preconditioner
- Serial and parallel:

#it	17	18	19	20	21	25	34	41	div	avg
AMG	97	1			2	1	2	1	87	17.72
STAMG	179	4	6	2					0	17.12

Parallel execution of protected algorithm on 4 procs with CG and AMG preconditioner.

• All cycle types:







User level exception handling

Research goal

Extend the functionality of MPI for fault-tolerant algorithms

O User level exception handling

Challenges

- Detect locally thrown exceptions
- Inform all processes of the error
- Wrap it into a user-friendly C++ compliant interface
- Support asynchronous communication (similar to C++ future concept)

Code Example

```
try{ // scope to be protected
Guard guard(communicator);
Future f = init_communication();
do_some_computation();
f.get(); // MPI::Wait()
}catch(...) {
   // handle thrown exceptions
}
```

- Cheap guard object protects *try* block
- Is destructed during stack unwinding
- Propagate exception across communicator (uses std::uncaught_exception)







Output User level exception handling

MPI-4 variant

- Interface is using the User-level failure-mitigation extension (ULFM)
- Provides functionality for
 - Hard fault detection
 - Communicator revocation
 - *Shrinking* of faulty communicator (i.e. excluding faulty processors)

MPI-3 variant

- Fall-back library which creates additional communication channel for exceptions
- Drawback: cannot interrupt MPI collectives, no hard fault protection



https://gitlab.dune-project.org/exadune/blackchannel-ulfm

C. Engwer, M. A., N. Dreier, D. Göddeke, A High-Level C++ Approach to Manage Local Errors, Asynchrony and Faults in an MPI Application, Proceedings of PDP 2018, 2018







Summary and Outlook

Summary

- We developed three 'orthogonal' approaches to increase fault-tolerance, especially for multigrid-type algorithms:
 - Restoration with compressed checkpoints: MG and SZ compression both have their (dis-)advantages
 - Efficient SDC protection with built-in in properties
 - Exception-propagation to ensure same state in MPI programs
- 'User level exception handling' can be used for many algorithms to develop MPI-4 ready fault-tolerant algorithms already in MPI-3
- Interface is ready for asynchronous algorithms (future concept):
 - Asynchronous checkpointing/repair
 - Local-failure local-recovery
 - ...







What's next?

- Integrating the new MPI interface into DUNE¹
- Improving features/functionality of the interface for wider applicability
- Local restoration for non-linear solvers
- Evaluating and combining developed concepts:
 - Switch between different compression and repair techniques: Adaptively select the most efficient one
 - Asynchronous checkpointing/repair
 - Asynchrony in multigrid with ideas from abstract Schwarz theory: Local smoothing while restoring lost processors?

• ...

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Thinking about ideas for fault-tolerance and asynchrony in remaining PDE solver parts, not only linear solver

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Assumption: Multigrid is self-stabilising

Self-stabilising

Starting from any state the solver comes back to a valid state.

P. Sao, R. Vuduc, Self-stabilizing Iterative Solvers, 2013

- Original defined by Dijkstra in 1974 for systems of distributed control
- Examples: Newton- and Jacobi-methods







Assumption: Multigrid is self-stabilising

Sketch of the proof

- Multigrid is a defect correction procedure, i.e., a fixed point iteration
- Hackbusch's multigrid convergence proof is based on contraction arguments:

If the contraction property holds for a given iteration operator, then convergence of the corresponding iteration procedure is guaranteed for any initial guess

- Basically Banach's fixed point theorem
- The new initial guess is simply the last iterate with some faulty entries
- Matrices and grid transfer operators are fault-free
 ⇒ Contraction property is not affected







SZ compression (version 1.4.2, 2D)

 $2^{m-1} - 2$

- Predict values row by row (top to bottom, left to right)
- $\mathcal{V} = \{V(i, j)\}$: set of already compressed point values
- Interpolation based first-phase prediction f(i, j)

 $2^{m-1} - 1$

1-Layer	V(i, j - 1) + V(i - 1, j) - V(i - 1, j - 1)
2-Layer	$\begin{array}{l} 2V(i,j-1)+2V(i-1,j)-4V(i-1,j-1)\\ -V(i,j-2)-V(i-2,j)+2V(i-2,j-1)\\ +2V(i-1,j-2)-V(i-2,j-2) \end{array}$

• 2^m intervals with size of $2 \times \texttt{error_bound}$ around f(i, j)

2 xerror bound

3m-1



real value

9m

• Store index *p* or and *p* = 0 and compressed binary-representation if the real value is not in any second-phase prediction interval

 $2^{m-1} + 1$

 $2^{m-1} + 2$

• Data is decompressed via interpolation and shifted by the Huffman-code























Differences between MG and FASMG

• Multigrid's correction problem is given by

$$\mathbf{A}_k(u_k + v_k) = b_k$$

• Classic MG uses linearity and searches on the next coarser level for the correction v_k only

$$\mathbf{A}_{k-1}v_{k-1} = \mathbf{R}_{k-1}^k(b_k - \mathbf{A}_k u_k)$$

• FASMG searches always for the full solution $\tilde{v}_k := u_k + v_k$

$$\mathbf{A}_{k-1}\tilde{v}_{k-1} = \mathbf{R}_{k-1}^k(b_k - \mathbf{A}_k u_k) + \mathbf{A}_{k-1}\mathbf{I}_{k-1}^k u_k$$

- \tilde{v}_{k-1} is an approximation to the fine grid problem but with lower resolution
- We can interpret \tilde{v}_{k-1} as a compressed backup of \tilde{v}_k
- \mathbf{R}_{k-1}^k and \mathbf{I}_{k-1}^k are different restriction operators





Justification of soft fault detection mechanism

• FAS correction problem

$$\mathbf{A}_{k-1}\tilde{v}_{k-1} = \mathbf{R}_{k-1}^k r_k + \mathbf{A}_{k-1}\tilde{v}_{k-1}^{(0)}$$

• Using linearity of the operator and $c_{k-1} = \tilde{v}_{k-1} - \tilde{v}_{k-1}^{(0)}$ yields $\|c_{k-1}\| < \|(\mathbf{A}_{k-1})^{-1}\| \| \mathbf{R}_{k-1}^k \| \|r_k\|$

• $r_k = b_k - \mathbf{A}_k u_k^{(\nu)}$ on fine grid converges monotonously to zero if \mathbf{A} is s.p.d.

• On coarser grid levels $b_k = \mathbf{R}_k^{k+1}(b_{k+1} - \mathbf{A}_{k+1}u_{k+1}^{(\nu)}) + \mathbf{A}_k\mathbf{I}_k^{k+1}u_{k+1}^{(\nu)}$ gives

$$\begin{aligned} r_k &= \mathbf{R}_k^{k+1}(b_{k+1} - \mathbf{A}_{k+1}u_{k+1}^{(\nu)}) + \mathbf{A}_k \mathbf{I}_k^{k+1}u_{k+1}^{(\nu)} - \mathbf{A}_k u_k^{(\nu)} \\ &= \mathbf{R}_k^{k+1}(b_{k+1} - \mathbf{A}_{k+1}u_{k+1}^{(\nu)}) + \mathbf{A}_k(u_k^{(0)} - u_k^{(\nu)}) \\ \Rightarrow \qquad \|r_k\| \leq \|\mathbf{R}_k^{k+1}\| \|b_{k+1} - \mathbf{A}_{k+1}u_{k+1}^{(\nu)}\| + \|\mathbf{A}_k\| \|u_k^{(0)} - u_k^{(\nu)}\| \end{aligned}$$

• Operators are bounded, multigrid converges, smoothing property holds: $||r_k|| \to 0$ and by this $||c_k|| \to 0$





