

Multi-level checkpointing and silent data corruption

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Fail-stop errors

Characteristics

- ► Component failure (node, network, power, ...)
- Application fails and data is lost

Fault rate proportional to number of components

- ▶ 2013: *Preprod.* Blue Waters requires repairs \approx 4 hours [2, 1]
- ▶ 2014: Titan loses a node every \approx 1.5 days [2, 3, 1]
- ▶ 2014: Blue Waters loses \approx 2 nodes per day [1]



Coping with fail-stop errors

Instantaneous error detection

Standard approach: Periodic checkpoint, rollback, and recovery:

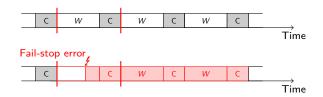




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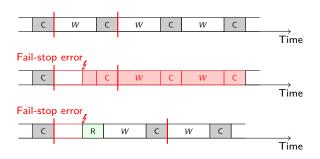




Coping with fail-stop errors

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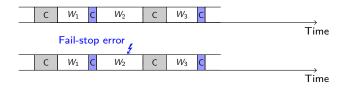


- Different kinds of checkpoints: local disk storage, partner-copy, Reed-Solomon encoding technique, file system
- ▶ Different kinds of errors: node failure, router failure, etc.
- ► Each checkpoint has a cost and some resilience capabilities



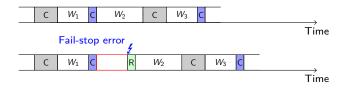


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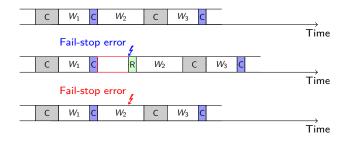


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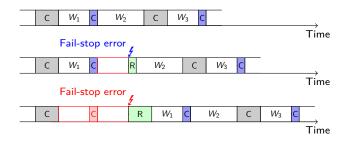


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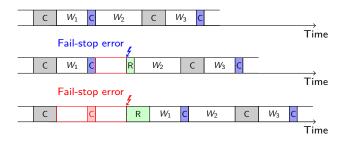


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When should we checkpoint? Using which mechanism?



Two-level checkpointing: assumptions

Two types of faults

- ▶ Type-1: follow an exponential distribution of failure rate λ_1
- ▶ Type-2: follow an exponential distribution of failure rate λ_2

Two types of checkpoints

- ► Type-2 checkpoints take time C₂ (recovery R₂) Enables recovery from type-1 and type-2 faults
- ► Type-1 checkpoints take time C₁ (recovery R₁) Enables recovery from type-1 faults



Two-level checkpointing: assumptions

Two types of faults

- ▶ Type-1: follow an exponential distribution of failure rate λ_1
- Type-2: follow an exponential distribution of failure rate λ₂
 More dramatic faults

Two types of checkpoints

- ► Type-2 checkpoints take time C₂ (recovery R₂) Enables recovery from type-1 and type-2 faults More expensive checkpoints
- ▶ Type-1 checkpoints take time C_1 (recovery R_1) Enables recovery from type-1 faults Cheap checkpoints



Two-level checkpointing: assumptions

Two types of faults

- ▶ Type-1: follow an exponential distribution of failure rate λ_1
- ► Type-2: follow an exponential distribution of failure rate λ_2 More dramatic faults

Two types of checkpoints

- ► Type-2 checkpoints take time C₂ (recovery R₂) Enables recovery from type-1 and type-2 faults More expensive checkpoints
- ► Type-1 checkpoints take time C_1 (recovery R_1)
 Enables recovery from type-1 faults Cheap checkpoints

Other assumptions

- ► Fault of type-i is followed by a *downtime* and a type-i recovery
- No faults during recoveries



Execution time of a pattern

▶ Pattern: work of some size W divided in K chunks



Objective: overhead minimization

Overhead(Pattern
$$(K, W, w_1, ..., w_K)$$
) =
$$\frac{\mathbb{E}(\text{Pattern}(K, W, w_1, ..., w_K))}{W} - 1$$

► First property:

Execution time is minimized when all chunks have same size



Unknown job length: optimal solution

Chunks have size w_{opt} where:

$$N(w_{opt})\ln(N(w_{opt})) = \lambda L w_{opt}(e^{\lambda(w_{opt}+C_1)}-1)$$

▶ There are *K* chunks in a pattern where:

$$\beta \lambda K w_{opt} e^{\lambda (w_{opt} + C_1)} (1 + L(e^{\lambda (w_{opt} + C_1)} - 1))^{K-1} = \alpha + \frac{\beta}{L} (1 + L(e^{\lambda (w_{opt} + C_1)} - 1))^{K}$$

Missing notations $N(w)=1+L(e^{\lambda(w+C_1)}-1),\ L=rac{\lambda_2}{\lambda},\ \lambda=\lambda_1+\lambda_2,$ $lpha=\mathcal{R}(e^{\lambda C_2}-1)-rac{\beta}{L},\ eta=\mathcal{R}(1+L(e^{\lambda C_2}-1)),$ $\mathcal{R}=rac{1+\lambda_1R_1+\lambda_2R_2}{\lambda}+D$

▶ Ugly implicit equations: solve them numerically!



Known job length: optimal solution

- lacktriangle Total size of job: \mathcal{W}_{total}
- Chunks have same w_{opt} size than previously
- ▶ There are p^* patterns where:

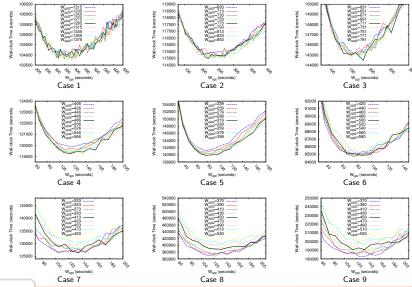
$$p^* = rac{\mathcal{W}_{total} \ln(\mathit{N}(w_{opt}))}{\left(\mathbb{L}\left(rac{lpha \mathit{L}}{eta e}
ight) + 1
ight) \mathit{w}_{opt}}$$

with the same notations as previously and $\mathbb{L}(z) = x$ if $xe^x = z$.

Ugly implicit equations: solve them numerically!



Assessment through simulations





Conclusion so far

 We know how to use efficiently two-level checkpointing under fail-stop failures

What about silent data corruption?



Second kind of errors: silent data corruption

Characteristics

- ▶ Bit flip (Disk, RAM, Cache, Bus, ...)
- ▶ Problems: detection latency, potentially wrong results

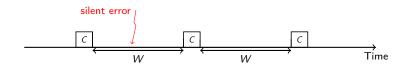
Cosmic rays do produce errors

- ▶ 2002: Unprotected address bus ASCI Q at Los Alamos National Laboratory could not run more than one hour [3]
- ▶ 2003: *No ECC* Virginia Tech 1,100 Apple Power Mac G5 supercomputer could not boot [3]
- ▶ 2010: ECC protected Jaguar saw 350 bit-flips/min [3]
- ➤ 2010: ECC protected Jaguar saw 1 double-bit error/day [3]
- ▶ 2014: Titan: reported > 1 double-bit error per week [4]



Main problem: detection latency

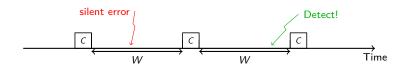
Question: can we follow the same approach?





Main problem: detection latency

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Main problem: detection latency

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Keep multiple checkpoints?



Main problem: detection latency

Question: can we follow the same approach?



Keep multiple checkpoints?

Which checkpoint to recover from?



Main problem: detection latency

Question: can we follow the same approach?



Keep multiple checkpoints?

Which checkpoint to recover from?

Need an active method to detect silent errors!



Existing Methods for Detecting Silent Errors

General-purpose approaches

 Replication [Fiala et al. 2012] or triple modular redundancy and voting [Lyons and Vanderkulk 1962]

Application-specific approaches

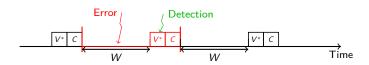
- Algorithm-based fault tolerance (ABFT): checksums in dense matrices Limited to one error detection and/or correction in practice [Huang and Abraham 1984]
- Partial differential equations (PDE): use lower-order scheme as verification mechanism [Benson, Schmit and Schreiber 2014]
- Generalized minimal residual method (GMRES): inner-outer iterations [Hoemmen and Heroux 2011]
- Preconditioned conjugate gradients (PCG): orthogonalization check every k iterations, re-orthogonalization if problem detected [Sao and Vuduc 2013, Chen 2013]

Data-analytics approaches

- Dynamic monitoring of HPC datasets based on physical laws (e.g., temperature limit, speed limit) and space or temporal proximity [Bautista-Gomez and Cappello 2014]
- Time-series prediction, spatial multivariate interpolation [Di et al. 2014]



Solution: coupling checkpointing with verification



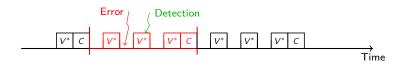
- Before each checkpoint, run some verification mechanism or error detection test
- Silent error, if any, is detected by verification
- Last checkpoint is always valid

Problem solved! But can do better than that!



One step further

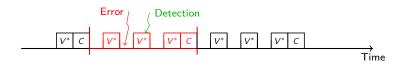
Perform several verifications before each checkpoint:



- ▶ Pro: silent error detected earlier in pattern
- ► Con: additional overhead in error-free executions
- Need to find the best trade-off

One step further

Perform several verifications before each checkpoint:



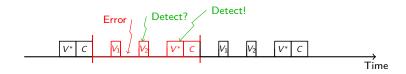
- ▶ Pro: silent error detected earlier in pattern
- Con: additional overhead in error-free executions
- Need to find the best trade-off
- ► Not all verification mechanisms have 100% accuracy! Should we use partial detectors? How?



Partial verification

Guaranteed/perfect verifications (V^*) can be very expensive! Partial verifications (V) are available for some HPC applications!

- ▶ Lower accuracy: recall $r = \frac{\text{#detected errors}}{\text{#total errors}} < 1$
- ▶ Lower cost, i.e., $V < V^*$





The optimization problem

Two types of checkpoints

- Disk checkpoint: stable storage (slow but resilient)
- Memory checkpoint: local copy (fast but lost on fail-stop)

Checkpoint only done after guaranteed verification

Two types of responses to errors

- ► Fail-stop error ⇒ rollback to last disk checkpoint
- Silent errors ⇒ rollback to last memory checkpoint

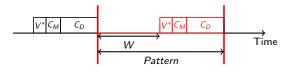
Goal:

- ► Combine everything into a single periodic pattern
- Minimize the overhead due to faults and to fault-tolerance



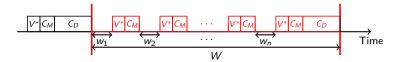
Resilience patterns (1/2)

Starting with base pattern



Pattern à la Young-Daly

Adding verified memory checkpoints

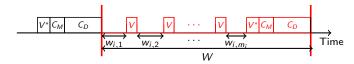


Pattern with *n* segments



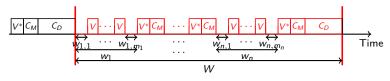
Resilience patterns (2/2)

Adding intermediate verifications between memory checkpoints



Segment w_i has m_i chunks

Putting everything together



Full pattern

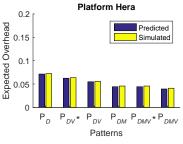


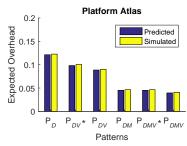
The optimal solution (first order approximation)

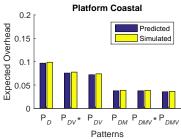
Pattern	W*	n*	m*	Overhead (Pattern)
P _D	$\sqrt{\frac{V^* + C_M + C_D}{\lambda_s + \frac{\lambda_f}{2}}}$	-	-	$2\sqrt{\left(\lambda_s + \frac{\lambda_f}{2}\right)\left(V^* + C_M + C_D\right)}$
P _{DV} *	$\sqrt{\frac{m^*V^*+C_M+C_D}{\frac{1}{2}\left(1+\frac{1}{m^*}\right)\lambda_s+\frac{\lambda_f}{2}}}$	-	$\sqrt{\frac{\lambda_s}{\lambda_s + \lambda_f} \cdot \frac{c_M + c_D}{V^*}}$	$\sqrt{2(\lambda_s + \lambda_f)C_M + C_D} + \sqrt{2\lambda_s}V^*$
P _{DV}	$\sqrt{\frac{(m^*-1)V+V^*+C_M+C_D}{\frac{1}{2}(1+\frac{2-r}{(m^*-2)r+2})\lambda_s+\frac{\lambda_f}{2}}}$		$2-\frac{2}{r}+\sqrt{\frac{\lambda_s}{\lambda_s+\lambda_f}}$	$\sqrt{2(\lambda_s + \lambda_f)\left(V^* - \frac{2-r}{r}V + C_M + C_D\right)}$
	$\sqrt{\frac{1}{2}\left(1+\frac{2-r}{(m^*-2)r+2}\right)\lambda_s+\frac{\lambda_f}{2}}$	-	$\times \sqrt{\frac{2-r}{r}\left(\frac{V^*+C_M+C_D}{V}-\frac{2-r}{r}\right)}$	$+\sqrt{2\lambda_s \frac{2-r}{r} V}$
P _{DM}	$\sqrt{\frac{n^*(V^*+C_M)+C_D}{\frac{\lambda_s}{n^*}+\frac{\lambda_f}{2}}}$	$\sqrt{\frac{2\lambda_s}{\lambda_f}\cdot\frac{C_D}{V^*+C_M}}$	-	$2\sqrt{\lambda_s(V^*+C_M)}+\sqrt{2\lambda_fC_D}$
P _{DMV} *	$\sqrt{\frac{n^*m^*V^* + n^*C_M + C_D}{\frac{1}{2}\left(1 + \frac{1}{m^*}\right)\frac{\lambda_s}{n^*} + \frac{\lambda_f}{2}}}$	$\sqrt{\frac{\lambda_s}{\lambda_f} \cdot \frac{c_D}{c_M}}$	$\sqrt{\frac{C_M}{V^*}}$	$\sqrt{2\lambda_f C_D} + \sqrt{2\lambda_s C_M} + \sqrt{2\lambda_s V^*}$
P _{DMV}	$\sqrt{\frac{n^*(m^*-1)V + n^*(V^* + C_M) + C_D}{\frac{1}{2}\left(1 + \frac{2 - r}{(m^*-2)r + 2}\right)\frac{\lambda_s}{n^*} + \frac{\lambda_f}{2}}}$	λ _s	$2 - \frac{2}{r}$	$\sqrt{2\lambda_f C_D} + \sqrt{2\lambda_s \left(V^* - \frac{2-r}{r}V + C_M\right)}$
	$\sqrt{\frac{1}{2}\left(1+\frac{2-r}{(m^*-2)r+2}\right)\frac{\lambda_s}{n^*}+\frac{\lambda_f}{2}}$	$\bigvee^{\lambda_f} V^* - \frac{2-r}{r} V + C_M$	$2 - \frac{2}{r}$ $+\sqrt{\frac{2-r}{r}\left(\frac{V^* + c_M}{V} - \frac{2-r}{r}\right)}$	$+\sqrt{2\lambda_s \frac{2-r}{r} V}$

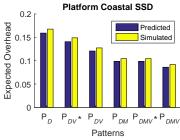


Simulations











Conclusion so far

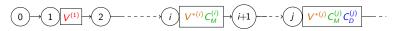
We know how to use efficiently two-level checkpointing under fail-stop failures and silent data corruption with guaranteed verifications and partial verifications

Caveat: we assumed full freedom to place checkpoints and verifications (divisible load) Question: What about task graphs?

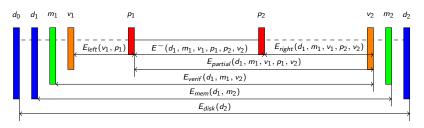


The optimization problem

- Application modeled as a linear task graph
- Checkpoints and verifications are performed in between tasks



Question: when to take which checkpoint and verification in order to minimize the execution time?



▶ Optimal solution: $O(n^6)$ dynamic programming algorithm



Conclusion and perspectives

Pros

- Mix of silent and fail-stop errors
- Mix of partial and guaranteed verifications

Cons

- Results limited to 2 levels...
 - ... but upcoming generalization for any number of levels!
- Exponential failure distribution



All details can be found in

- S. Di, Y. Robert, F. Vivien, and F. Cappello.
 Toward an Optimal Online Checkpoint Solution under a Two-Level HPC Checkpoint Model. *IEEE Transactions on Parallel and Distributed Systems*, 2016. To appear.
- A. Benoit, A. Cavelan, Y. Robert, and H. Sun.
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- ▶ A. Benoit, A. Cavelan, Y. Robert, and H. Sun. Two-Level Checkpointing and Verifications for Linear Task Graphs. In The 17th IEEE International Workshop on Parallel and Distributed Scientific and Engineering Computing (PDSEC 2016), May 2016.



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ANY QUESTIONS?

