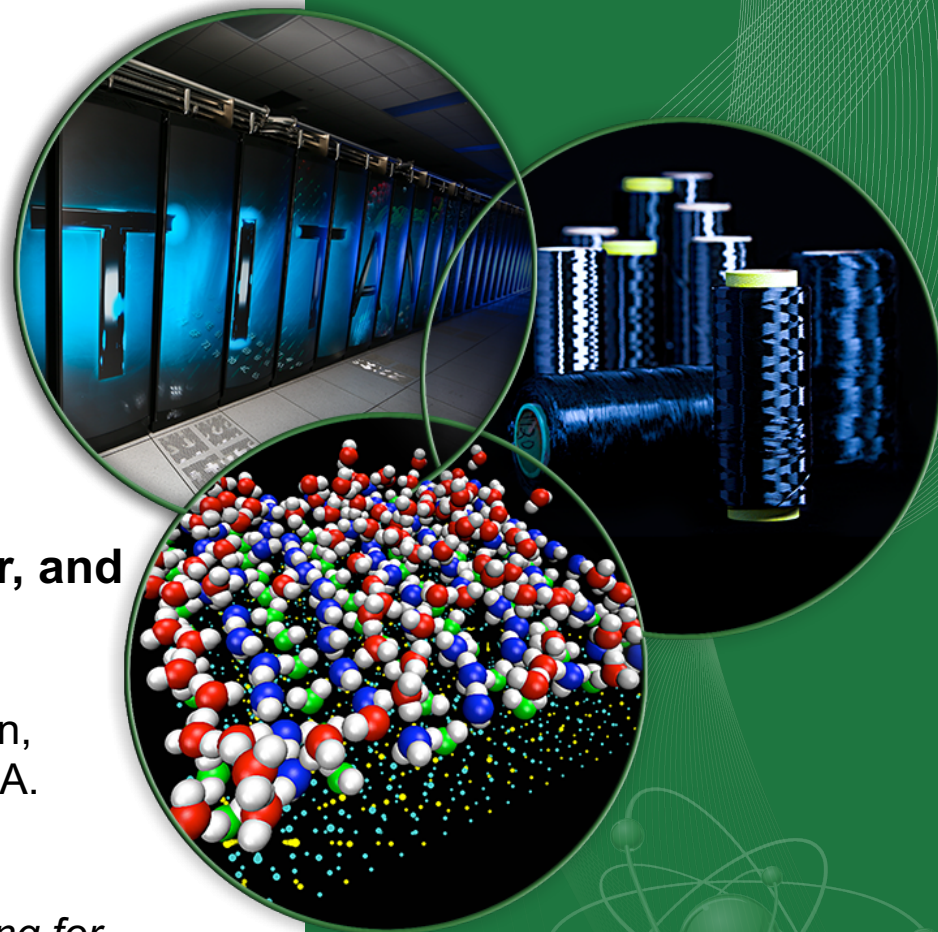


# Pattern-based Modeling of Fail-stop and Soft-error Resilience for Iterative Linear Solvers

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*18th SIAM Conference on Parallel Processing for  
Scientific Computing (PP), Tokyo, Japan, March 7-10, 2018*

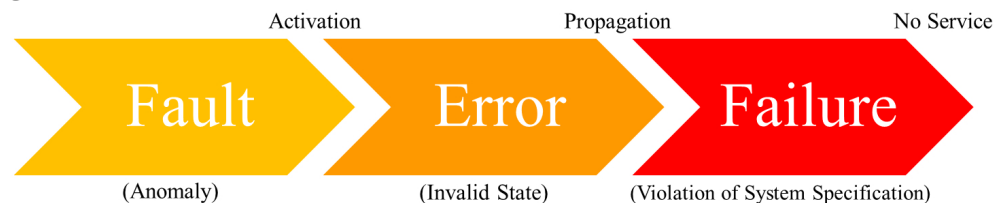


# Resilience, Why?

- **Resiliency** in high performance computing (HPC) applications: the ability to gracefully handle errors and recover from failures.
- Errors and failures are common place in HPC systems today.
  - Large-scale systems with a number of complex & diverse software and hardware components,
  - Technology scaling trends in hardware components,
  - Complex compute, memory, interconnect and storage architectures,
  - Cost (design, area, power, engineering) of achieving error-free large scale systems is too high.
- The situation is only expected to get worse, as we move towards the goal of achieving more computational power, i.e., *Exascale systems*.

# Multiresilience in HPC Applications

- HPC Applications are affected by multiple types of events which hinders with their ability to make forward progress and their correctness.
  - Soft Errors: Silent data corruptions (**SDC**),
  - Hard Errors: process failures (**Crash**),
  - Broad categorization which covers the affects of various types of faults excluding *performance faults*.
- Most works provide resilience to only a single type of error mechanism.
- Need to *systematically* integrate multiple techniques to detect and handle multiple error events, without sacrificing performance.

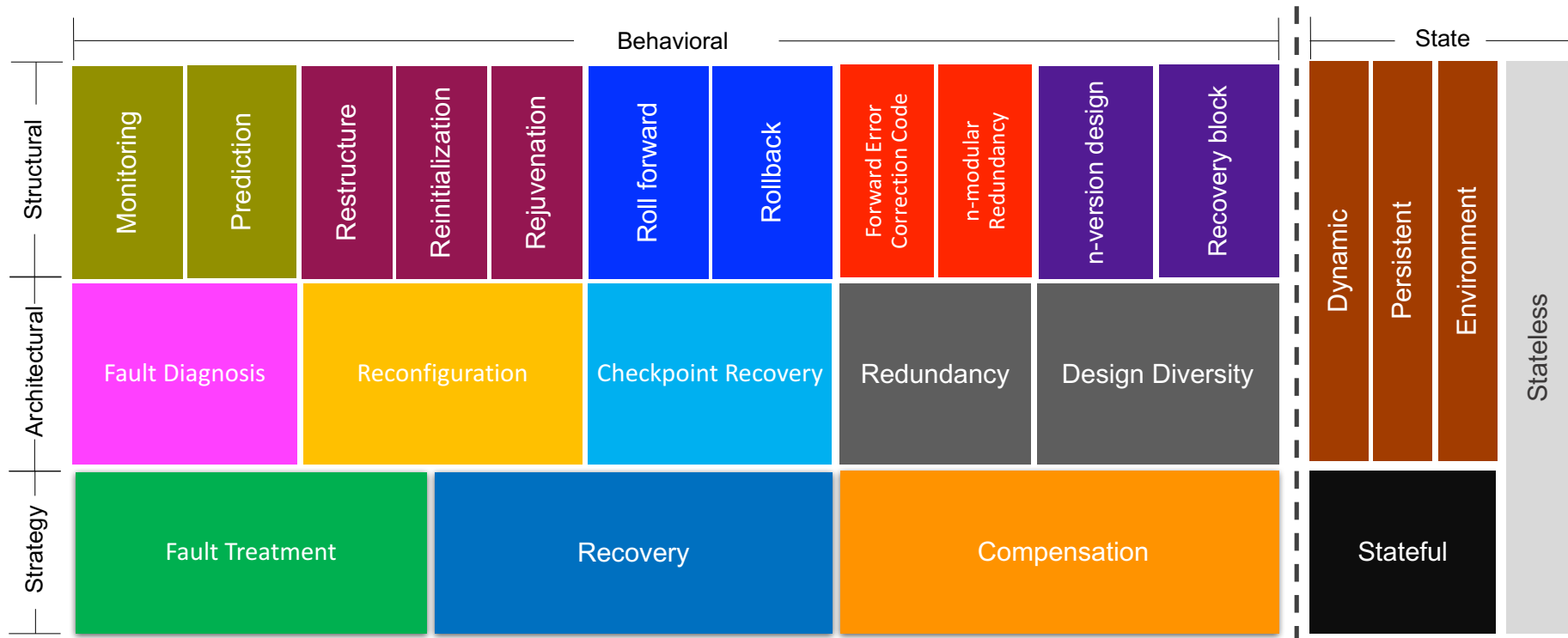


# Design Patterns for Resilience

- Patterns provide a *generalizable solution* to a recurring problem.
- The solution is *formalized* with a set of activation and response interfaces, and a behavior specification.
- Patterns *do not* provide concrete solutions, instead focus on a reproducible strategy which may be used many times, implemented in different manners.
- *State Patterns*, provide encapsulation of application's state:
  - Static/Persistent State, Dynamic State, and Environment State.
- *Behavioral Patterns*, provide detection, containment and mitigation techniques:
  - Strategy, Architecture, and Structural patterns.

# Resilience Patterns Catalog

- See “*Resilience Design Patterns: A Structured Approach to Resilience at Extreme Scale*,” ORNL Technical Report v1.2, December 2017.



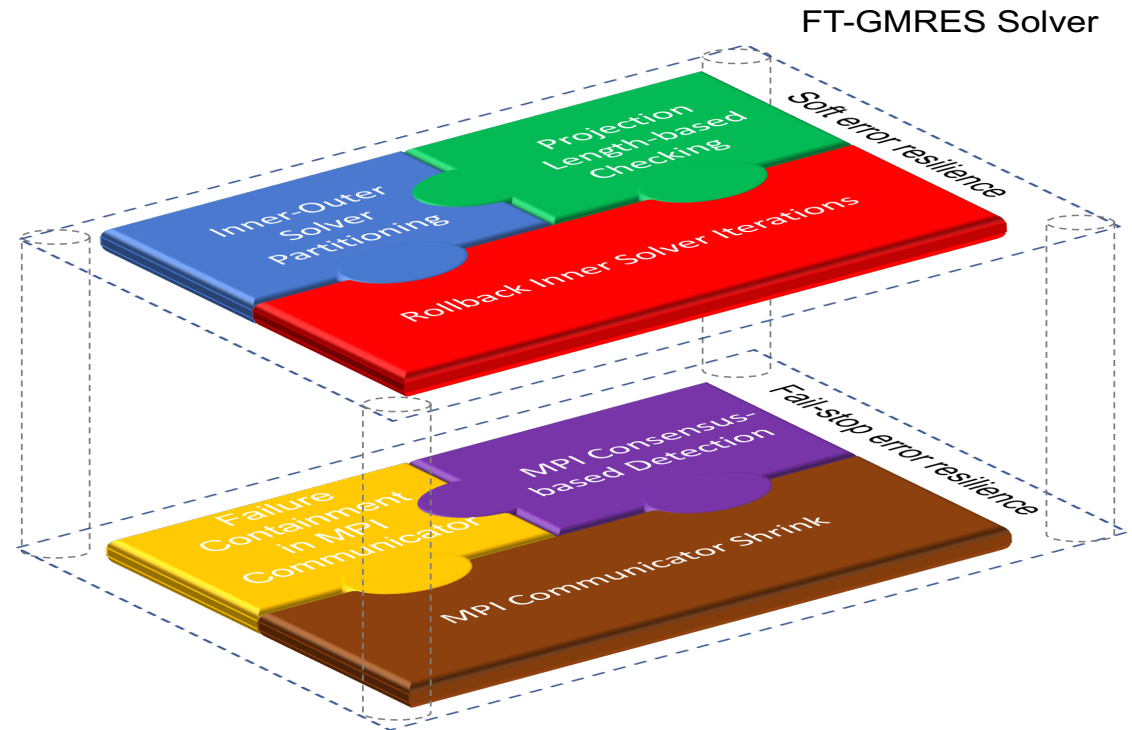
# Pattern-based Modeling of Multiresilience

- Multiple patterns are instantiated across layers of the system stack, interlinked using a *building blocks approach*.
- Coordination among multiple patterns designed to provide optimal *end-to-end application performance*.
  - Interfaces are standardized,
  - Systematic software and hardware layer coordination.
- Navigate the *performance resilience tradeoff* space by evaluating multiple solutions.
  - Each pattern has significantly different performance and implementation characteristics.
- Naïve stacking can lead to overprotection resulting in degradation of application performance.



# Use Case: Linear Solver

- GMRES minimal residual method for solving non-symmetric linear systems.
  - Solve:  $Ax = b$
  - Iterative algorithm
- Resilience patterns provide *detection*, *containment*, and *mitigation* for soft and fail-stop errors.



# Patterns for Soft Error Resilience

- SDCs can cause unbounded numerical errors, which propagate across parallel processes, resulting in slow convergence of the solver.
- **State patterns:** segregation enables exploration of detection and recovery patterns, reduces overheads in most cases.
  - *Static state:* Matrix  $A$  and Right-hand vector  $b$ ,
  - *Dynamic state:* Solution vector  $x$ ,
  - *Environment state:* Data-structure indices, pointers, loop counters, etc.
- **Detection patterns:** utilize properties/characteristics of the algorithm/application/state patterns to detect presence of SDCs.
  - *Monitoring pattern:* checks progress by keeping track of the *quality metric*. Depending on the application, the overhead of calculating the quality metric can vary widely.
  - Application specific instantiation of the monitoring pattern: *bounded-compute pattern*, verifies outputs produced during critical computations against a-priori known bounds.
- **Mitigation patterns:** ensures forward progress of the algorithm and application.
  - *Compensation strategy pattern:* modular redundancy, results in high overheads.
  - *Rollback recovery pattern:* preserve dynamic state in local memory (checkpoints).



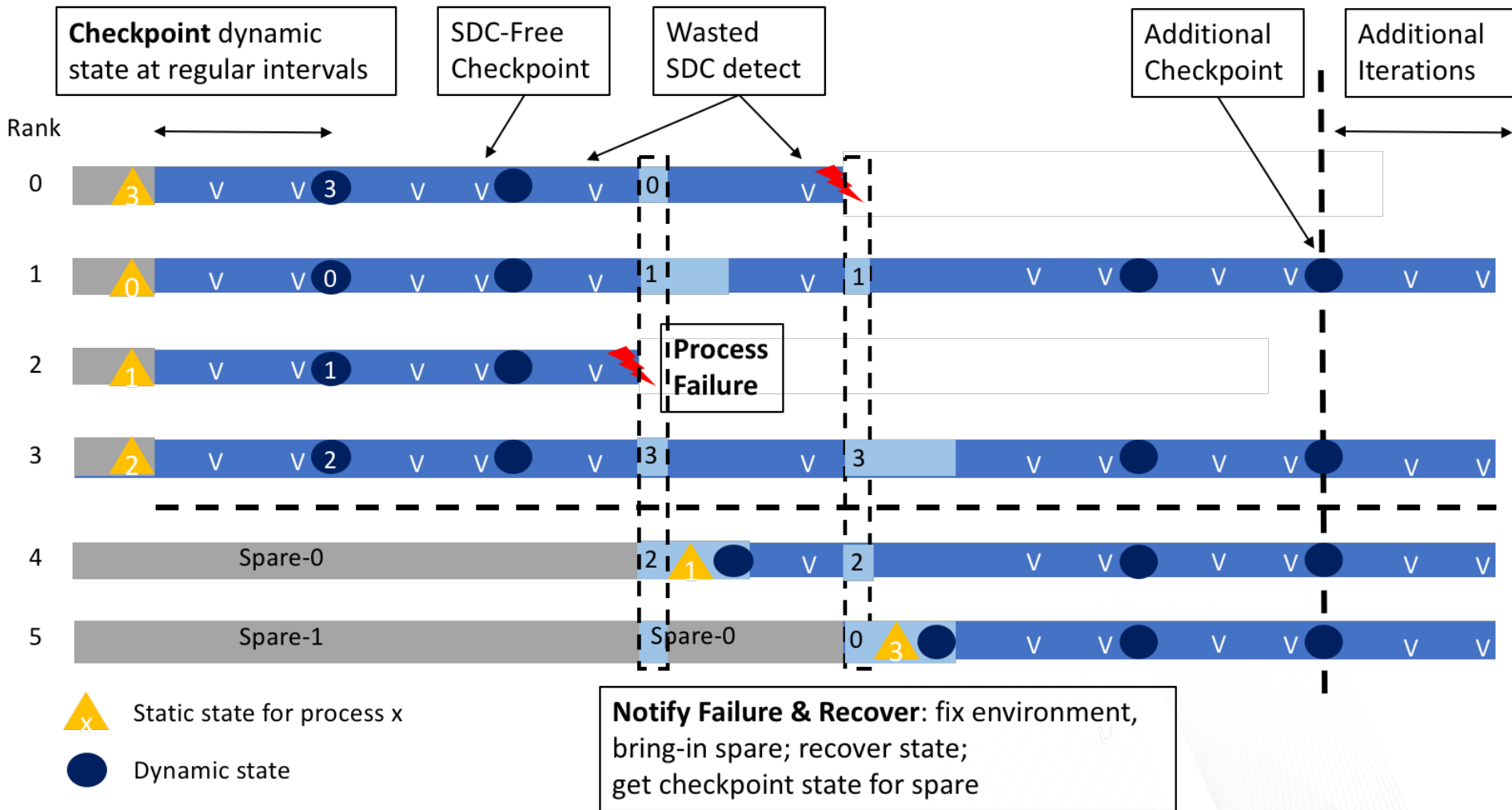
# Patterns for Hard Error Resilience

- Process failures make a parallel application to stall indefinitely or result in fatal crash, within the distributed memory model, losing all work done.
- **State patterns:** encapsulate the application state to facilitate recovery of lost state after process failure.
  - *Environment state:* Objects in parallel runtime environment,
  - *Static & Dynamic state:* Distributed across parallel processes.
- **Detection patterns:** instantiated in the environment state pattern, for robust detection and identification of failed processes.
  - *Consensus structural pattern:* proactive or reactive approach to failure detection.
  - Proactive strategy: Collective operations (all process communication) can be strategically placed to enable runtime environment to detect failures regularly.
- **Mitigation patterns:** recover lost static and dynamic state, and mitigate environment state for forward progress of parallel application.
  - *Reconfiguration pattern:* rejuvenate parallel runtime environment by removing failed processes and refreshing parallel runtime objects for future communications.
  - *Compensation strategy pattern:* maintain a pool of spare processes for replacement.
  - *Checkpoint restart pattern:* remote in-memory checkpoints of static and dynamic state.

# Pattern-based Modeling

Fault Model	Pattern Class	Choices	Selection
<i>Soft Error</i>	State	Dynamic, static, environment	Need scoping of <u>dynamic and static state</u> only, since environment corruption results in process failure.
	Detection	Monotonicity, bounded compute, checksums	<u>Bounded compute</u> results in 14x lower overhead compared to monotonicity monitoring pattern.
	Recovery	Checkpoint, checksums	<u>Local in-memory checkpoints</u> due to less computational overhead.
<i>Hard Error</i>	State	Dynamic, static, environment	<u>All</u> , since process failures are fatal for application.
	Detection	Proactive, reactive	<u>Proactive</u> , since collectives present in every iteration, prevent propagation.
	Recovery	Environment: warm spares, rejuvenate	<u>Spares</u> to avoid re-allocation of state.
		Checkpoint-restart, diskless checkpoints, linear interpolation	<u>In-memory checkpoint-restart</u> because of least overhead and no effect on convergence of solver.

# Multiresilience - Pattern coordination



# Experimental Setup

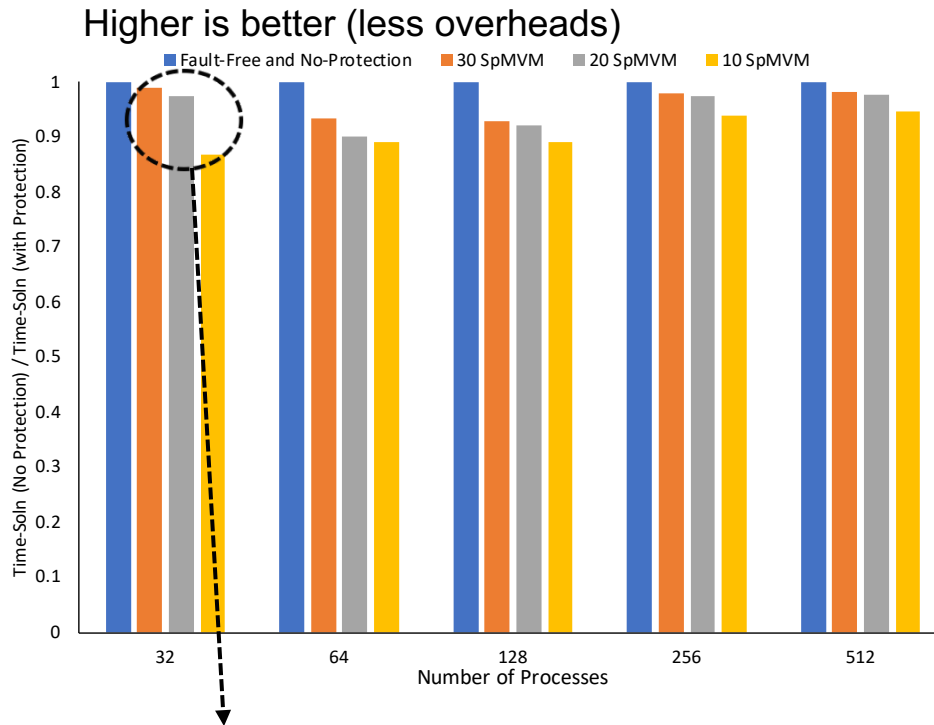
- FT-GMRES implemented using Trilinos 12.6.4 framework, <https://trilinos.org/>.
  - Tpetra package for parallel linear algebra using MPI.
- Parallel Environment: ULFM release 1.1, based on Open MPI 1.7.1 <http://fault-tolerance.org/>. ULFM provides:
  - Process failure detection,
  - Parallel environment reconfiguration capabilities (remove failed process)
- Test problem: Discretization of 3D mesh. Sparse Matrix with about 7 million rows and 186 Million non-zeros.
- 40-node Linux cluster with AMD Opteron processors.
  - Cores/node: 24 (Total: 960 cores),
  - Memory/node: 64 GB,
  - Point-to-point bandwidth: 215 MB/s.

# Goals of Experiments

- **Goal # 1:** Evaluate the resiliency and performance characteristics of individual patterns.
  - Design reproducible error injection experiments, e.g., coefficient of variation for all experiments ranges between 0.01 and 0.05.
- **Goal # 2:** Quantify the performance interactions between soft error and process failure resilience patterns in a multiresilience solution.
  - Identify constraints of combining patterns.

# Results – Soft Error Resilience

- Soft Error injected after every 30 (less frequent), 20 and 10 (more frequent) Sparse Matrix Vector Multiplication (SpMVM) operations.
- In all cases, the solver converged to a correct solution in allotted time.



Processes	Detect + Recovery Overheads		Num. of Additional Iterations	
	30 SpMVM	10 SpMVM	30 SpMVM	10 SpMVM
32	2.1%	8.9%	30 [75]	36 [150]
64	6.5%	4.8%	27 [50]	34 [75]
128	8.1%	8.8%	25 [25]	33 [125]
256	1.1%	1.9%	32 [50]	36 [125]
512	0.7%	0.7%	28 [50]	35 [100]

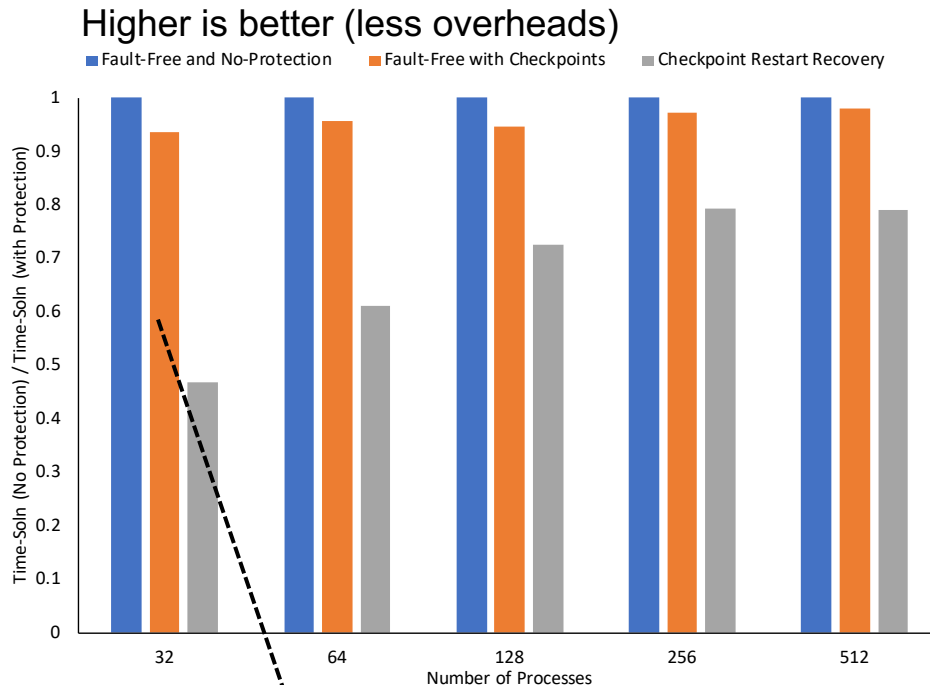
Overheads: Detection, recovery, additional iterations.

Tradeoff: High detection overhead, less additional iterations.



# Results – Process Failure Resilience

- Processes terminated based on exponential distribution with constant failure rate of time to complete 75 iterations. Up to four independent process failures.
- Pre-selected processes for termination gives results with low standard deviation.

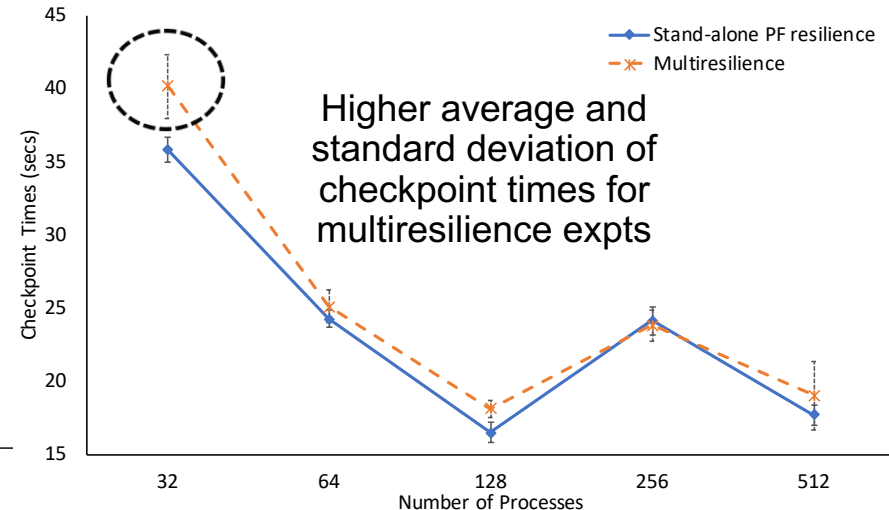
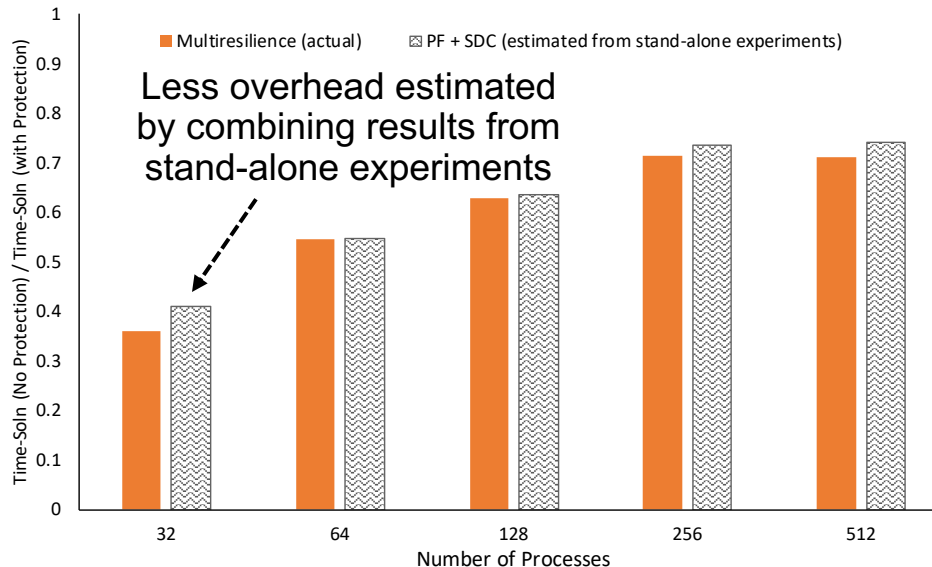


Overheads: **[No fail:]** Checkpoint of static (once) and dynamic states; **[Failures:]** No fail + state recovery, recompute, checkpoint (static, dynamic) state spares.

Processes	Reconfig	Recover State	Checkpoint Static + Dynamic [%dynamic]	Re-compute Overhead
32	0.02%	17.1%	28.1% [25.6%]	10.9%
64	0.03%	9.4%	18.5% [22.9%]	13.4%
128	0.04%	5.4%	12.9% [14.7%]	12.9%
256	0.02%	1.9%	7.5% [16.7%]	13.5%
512	0.05%	1.2%	5.1% [12.2%]	16.2%

# Results – Multiresilience

- Multiresilience to soft errors injected after every 10 SpMVMs (multiple times in a checkpoint interval) and up to four process failures.
- More overhead of dynamic state checkpoints due to increase in convergence time as compared to stand-alone process failure experiments.
  - On average, additional time is less than expected failure time, otherwise difference will be significant since checkpoint of static state is expensive.
- Overhead of wasted soft error detections is negligible in our experiments.



# Conclusions

- A *pattern-oriented design and implementation* approach for gracefully handling multiple error modes.
- Iterative refinement of pattern relationships to *optimize end-to-end application performance*.
- A generalizable approach to *architect resilience or multiresilience solutions* by composing patterns from multiple layers of system stack.
- *Experimental evaluation* for an iterative linear solver application exploiting algorithmic patterns.
- Highlighted the importance of considering *interactions between patterns* when designing a multiresilience solution.