Holistic Root Cause Analysis of Node Failures in Production HPC



Investigate:

✓ What are the

exact root causes?

Background and Motivation



Computer Science

NC STATE UNIVERSIT

- > Waste compute capacity and energy in HPC systems
- User job disruptions and System Wide Outages (SWOs)



considered, lead time analysis not well researched [4]



Research Goals

How well are spatial/temporal correlations indicative of the root cause?

By how much can the lead times increase if external factors are considered?

Node-related Logs

How to correlate?

System-wide Logs



- Processor corruptions in a single day can be
- caused by jobs scheduled on nodes, spatially apart How to infer real cause based on multiple tangible events?

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Solution Design

- Statistical analysis of root causes for node failures in HPC
- Estimate inaccurate predictions (false positives and false negatives) over a sample time-frame
- What are the conditions of trivial faults not leading to failures?
- Quantify increase in lead times w.r.t. the case when environmental influences are not considered
- Uncover insights to suggest mitigation approaches (proactive/reactive) for longevity of healthy conditions
- Correlation based on not just symptoms but related implications of events
- **Figure 7** Track down the root causes and their propagation across layers/components
- Will prevalent mitigation approaches fix the diagnosis result? Analyze what is the solution for the interpreted root cause?

Production Logs Studied:





SMW/Event Logs *Comprehensive understanding of how nodes fail?*

Non-trivial implications of vendor specific low-level system logs

Consultation with system administrators and cluster management team

Results



Observations: Figure 1 shows that lead times of node failures increase by ~5 times compared to node failure analysis in isolation

Figure 2 shows that with ~5 times increased lead times, the false positive (FP) rate do not rise with external correlations (they are lower than the FP rate with only node-specific events)



Results (cont.)



- Internal Causes (console/message/consumer logs)

- External Causes (Controller/Environment/Event)

• Lead time enhancements not possible

• Application-triggered, do not have early external indicators

• Fan speeds & voltage often operate below the min threshold



Observations:

- In systems C1 and C2, > 50% node failures have early indicators related to external causes, in a sample time-frame of 10 days

- Internal causes:

- Job triggered resource exhaustion
- Processor corruptions followed by kernel oops

Conclusions

- Root cause diagnosis enhances lead times to node failures by ~5 times
- With external environmental correlations, false positive rates are lower w.r.t. the cases without any external correlation.
- Several failures have unknown root causes (intangible in logs): - Generic algorithm impractical, automation got the goal
- Measurement-driven statistical analysis, insights to potential causes
- Holistic understanding of how failures happen enhance awareness of what actions to take for long-term system health.
- \blacktriangleright Results suggest that more than 20% of the sensor reading deviation messages do not lead to eventual failures.

Further investigation

- Pin-point conditions when typical software traps and hardware faults do not result in failures.
- Analyze inaccuracy in failure predictions with system-wide environmental considerations.
- Quantify inter-node correlations in the context of resource sharing and components influencing them (file system, interconnect).

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Acknowledgments: This work was supported in part by DOE subcontracts from Lawrence Berkeley National Lab and Lawrence Livermore National Lab, and NSF grants 1525609 and 0958311.



