Fault Tolerance on Control Applications: Empirical Investigations of Impacts from Incorrect Calculations

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Outline

- Introduction
- System and Fault Model
- Studied Application with Fault Injection
- Empirical Approach for Obtaining (m,k)
- Experiment Results and Evaluation







Introduction (1/2)

High transient fault rates expected in future systems

- Corrupt execution state or cause soft-errors through bitflips
- Lead to catastrophic failures, e.g., JAXA satellite Hitomi

Software-based fault tolerance approaches

- No need for special hardware
- Higher utilization

Probability of faults is low

- Fully protecting everything is mostly over-provision
- Reduction of utilization is possible







Introduction (2/2)

Some errors can be tolerated in control applications

- If we allow to downgrade the quality of control
- High quality system with low utilization is desirable

Offline and online soft-error handling techniques

- Offline selective protection
- Online detection and compensation

We need an approach to

Quantify (m,k) robustness







Path Tracing Experiment¹

LegoNXT path-tracing robot

- Only one task with two light sensors
- On a circular track



¹Chen et al., "Compensate or Ignore? Meeting Control Robustness Requirements through Adaptive Soft-Error Handling", LCTES 2016





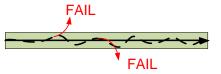
Path Tracing Experiment¹

LegoNXT path-tracing robot

- Only one task with two light sensors
- On a circular track

Go forward and follow the path

- Decision is made for jobs to have error
- More steering
- Leaving the track: mission failure



¹Chen et al., "Compensate or Ignore? Meeting Control Robustness Requirements through Adaptive Soft-Error Handling", LCTES 2016





System Model and Notation

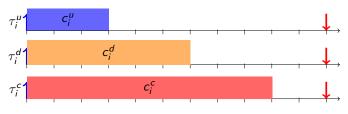
Robustness requirement

- Each task τ_i has a robustness requirement (m_i, k_i)
- m_i out of any k_i consecutive jobs have to be correct

Example for (m,k)

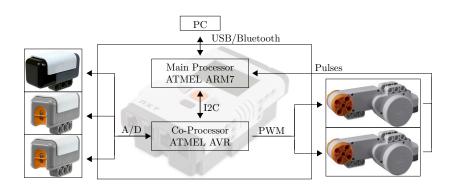
- The bit sequence $\{1, 1, 0, 1, 1, 0, 1, 1\}$ fulfills (2, 3)
- If $\{0,0,1\}$ occurs, (2,3) is not fulfilled

Soft-error (4) handling on task Level





Lego NXT Robot and nxtOSEK





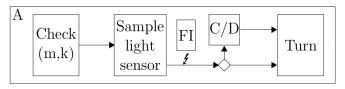




Original and Extended Fault Injection

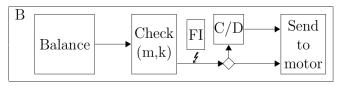
Original Design: Path tracing task

Fault injection only in light sensor values



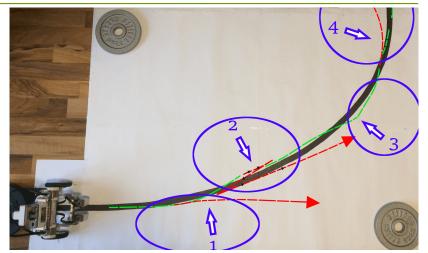
Extended Design: Balancing task

Fault injection in light sensor and motor input values





Experiment Setup: Path Tracing and Balance









General Experiment Flow: Obtaining (m,k)

Inject faults with f_{max}

Find (m,k) candidates

Verify (m,k) candidates

- Goal: show that (m,k) prevents mission failure
- Find the best (m,k) requirement: lowest ratio $\frac{m}{k}$
- Using (m,k) we can save utilization and protect the system





Fault Injection

Inject faults with f_{max}

Find (m,k) candidates

Verify (m,k) candidates

- f_{max} is the maximum tolerable fault rate
- Running system with f_{max} never causes mission failure
- Any $f_{fail} > f_{max}$ leads to mission failure
- Min. number of correct instances





Finding (m,k) Candidates

 $\begin{array}{c} \text{Inject faults} \\ \text{with } f_{max} \end{array}$

Find (m,k) candidates

Verify (m,k) candidates

- Goal: Min. number of correct instances in a sliding window
- Configure system with f_{max} and record bit sequence
- Example: {1,1,0,1,1,0,1,1}
 - k = 2: (1,2)
 - k = 3: (2,3)
 - k = 4: (2,4)
- List of (m,k) candidates: (2,4), (1,2), (2,3)



Verifying (m,k) Candidates

 $\begin{array}{c} \text{Inject faults} \\ \text{with } f_{max} \end{array}$

Find (m,k) candidates

Verify (m,k) candidates

- Goal: Verify that (m,k) prevents mission failure
- Configure system with $f_{fail} > f_{max}$ and first (m,k) in list
- Run system with (2,4) and f_{fail} : system executes $\{0,0,1,1\}$
 - ↓ Runs without mission failure: Success!
 - 4 If mission failure, try next best requirement (1,2)





Finding (m,k) for Path Tracing and Balancing

Path tracing (m,k)

Fault rate %	20	25	30	35	40
(i,j)	(5,16)	(5,16)	(5,20)	(5,20)	(5,20)

Balancing (m,k)

Fault rate $\%$	10	12	14	16
(i,j)	(11,16)	(10,16)	(10,16)	(8,16)





Verifying (m,k) for Path Tracing and Balancing

Path tracing (m,k)

(m,k)	(5,20)	(4,20)
f = 60%	Υ	Υ
f = 80%	Υ	N
f = 100%	Υ	N

Balancing (m,k)

(m,k)	(16,16)	(14,16)	(11,16)	(10,16)
f = 25%	Y	Y	Υ	N
f = 50%	Υ	Υ	N	N
f = 100%	Υ	N	N	N





Overall Utilization (1/4): Compensation Techniques

 $\begin{array}{c} \text{Inject faults} \\ \text{with } f_{\max} \end{array}$

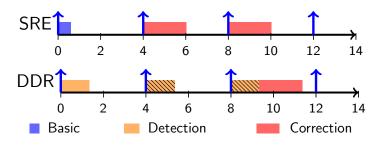
Find (m,k) candidates

Verify (m,k) candidates

Evaluate softerror handling techniques

Static Reliable Execution, Dynamic Detection and Recovery

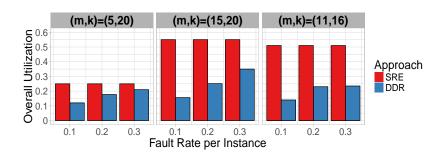
ullet Example: specify system with (2,3) requirement and $\{0,1,1\}$







Overall Utilization (2/4): Task Configurations



Task execution times are specified as:

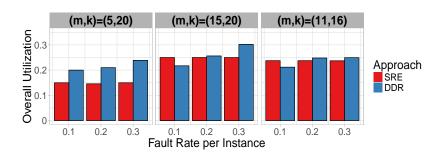
- $c_i^c = 700 \ \mu \text{s}, \ c_i^d = 120 \ \mu \text{s}, \ \text{and} \ c_i^u = 100 \ \mu \text{s}$
- DDR always outperforms SRE







Overall Utilization (3/4): Task Configurations



Task execution times are specified as:

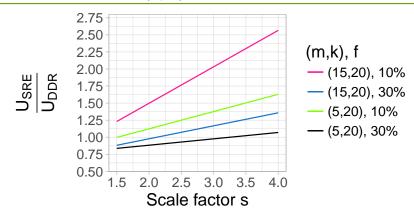
- $c_i^c=300~\mu \mathrm{s},~c_i^d=200~\mu \mathrm{s},~\mathrm{and}~c_i^u=100~\mu \mathrm{s}$
- In some cases SRE dominates DDR







Overall Utilization (4/4): Scale Factor



The scale is the deciding factor

• If the scale factor $s=\frac{c_i^c}{c_i^d}$ is low, SRE can be better than DDR







Conclusion and Outlook

Limitations

- No mathematical theory for deriving (m,k)
- Soft-errors are ideally detectable and recoverable
- Only one platform for experiments

Future Work

- Implement fault tolerance techniques in RTEMS
- Obtain (m,k) on Raspberry PI running RTEMS
- Mathematical theory to derive (m,k) for predefined controllers

Thank you!





Concurrent fault injection into Motors and Sensors

Inject faults in path tracing and balance task realistic

- (5, 20) and f = 60%, (11, 16) and f = 25%4. Robot fails in 10% of all runs
- Same experiment setting with (8, 20) and (13, 16) ↓ Success
- If (m, k) requirements are enforced concurrently, m may need to be increased to prevent mission failure





Overall Utilization Equations

Static Reliable Execution

•
$$\mathbb{U}_{\mathsf{SRE}} = \frac{u \cdot c_i^u + r \cdot c_i^c}{p \cdot T_i}$$

Dynamic Detection and Recovery

•
$$\mathbb{U}_{\text{DDR}} = \frac{c_i^d \cdot (d-r) + c_i^c \cdot (d+r)}{p \cdot T_i}$$



