Concurrency Testing with Data-Race Preemption Points

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Outline

What is Systematic Testing?

- Motivation
- Role of preemption points
- Challenges

Iterative Deepening

- Data race detection
- State space estimation
- Implementation in Landslide & Quicksand

Evaluation

Finding bugs in CMU OS student projects

Conclusion

Background: What is Systematic Testing?

Example

```
int thread_fork()
{
    thread_t *child = spawn_new_thread();
    add_to_runqueue(child);
    return child->tid;
}
```

Example

```
int thread_fork()
{
    thread_t *child = spawn_new_thread();
    add_to_runqueue(child);
    return child->tid; "child" gets freed!
}
```

- On exit, child's state is freed
- Forking thread does use-after-free
- Might return garbage instead of thread ID

Example

Thread 1	Thread 2
spawn_new_thread	
add_to_runqueue	
return child->tid	
	exit
	(TCB gets freed)

(new thread)
(yield to child)

Thread 1	Thread 2
spawn_new_thread	
add_to_runqueue	
	exit
	(TCB gets freed)
return child->tid	

(new thread + preempted)

(yield to parent)
(bad!)

Systematic Testing

Stress testing

- Common approach to finding concurrency bugs
- Relies on randomly-occurring context switches
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Systematic testing [Godefroid '97]

- Thread scheduling controlled by test framework
- Each iteration, test a different thread interleaving
- Goal: Exhaustively search "all possible" interleavings

Systematic Testing

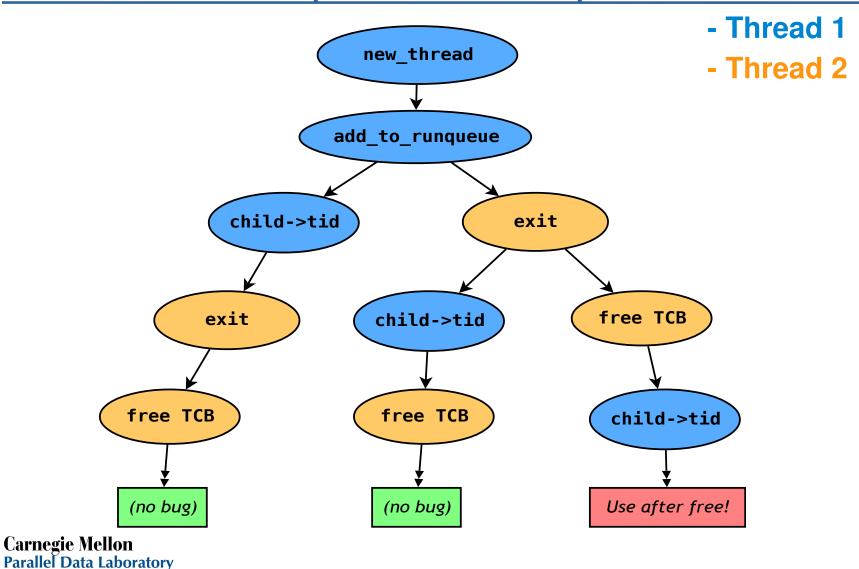
Stress testing

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Systematic testing [Godefroid '97]

- Thread scheduling controlled by test framework
- Each iteration, test a different thread interleaving
- Goal: Exhaustively search "all possible" interleavings
 - Wait... what?
 - In practice, interleavings are of coarse granularity.

Example - State Space



Preemption Points

The burning question of systematic testing: "Which **preemption points** (PPs) are important?"

State space of interleavings is *parameterized* by PPs.

- Preemption points everywhere: completion infeasible
- Too few preemption points: won't find bugs

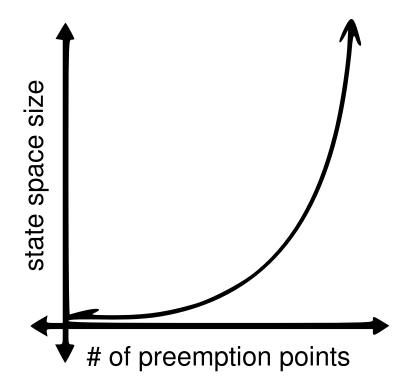
Prior work implementations hard-code a fixed set of PPs.

- dBug [Simsa '11]: pthread library calls
- Landslide [Blum '12]: kernel or user mutex lock/unlock

Challenges

Parameters of systematic tests must be kept small.

- Test program length / number of threads
- Number of preemption points used



Coping with Exponential Explosion

Some prior work focuses on improving *coverage* with...

- Reduction techniques (identifying equivalences)
 - [Flanagan '05, Guo '11, Huang '15]
- Determinizing runtimes [Cui '13]

Other prior systems give up on 100% completion...

- Random distribution of PPs [Fonseca '14]
- Heuristic exploration ordering (ICB) [Musuvathi '08]

Coping with Exponential Explosion

Current systematic testing model not user-friendly.

- Test framework: "I want to use these PPs, but can't predict how long until completion."
- User: "I have 16 CPUs and 24h to test my program."

Stress testing offers this... can we offer it too?

Our Technique: Iterative Deepening

Goal: Run the best tests for a given CPU budget.

- User studies with 15-410 (operating systems)
 - Students worked best with an iterative process
 - "Start small, then add more preemption points"

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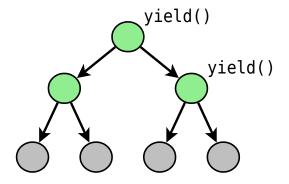
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Named after analogous technique in chess Al.

- Chess search is DFS limited by max # of moves (ply).
- Chess Als repeat DFS, increasing ply, until timeout.

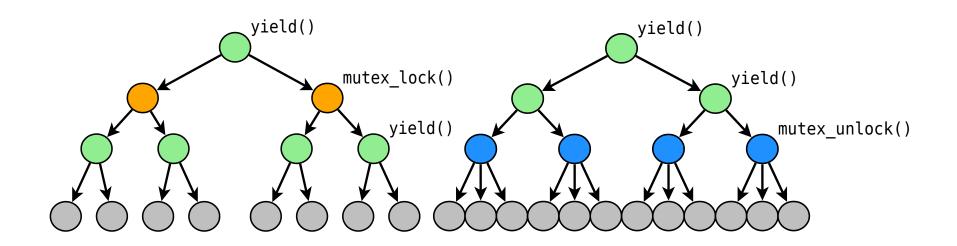
"Minimal" state space == mandatory thread switches only

- yield()
- cond_wait()



Different PPs can produce state spaces of different sizes.

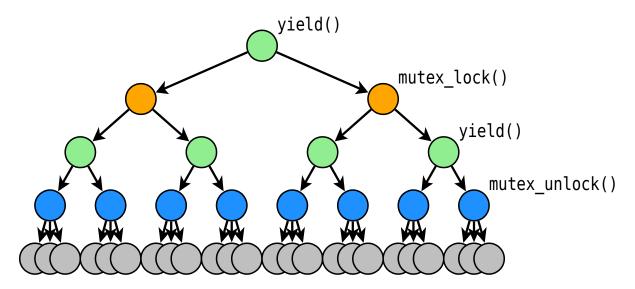
Testing them in parallel hedges our bets.



If time allows, we combine PPs to produce larger tests.

All PPs enabled = "maximal" state space

Prior work tools explore this state space only.



Choosing Preemption Points

How do we choose good preemption points?

- Start with "hardcoded" PPs as candidates
- Dynamic analysis during run can find new candidates:
 - Ad-hoc yield or busy loops
 - » Avoid cyclic state spaces
 - Data races [Savage '97]

Our main contribution

» Including atomic operations enables testing locks, lock-free queues, etc.

Data-Race Preemption Points

Data race: 2 threads access the same memory, and:

- At least one access is a write
- Sets of locks held by each thread do not overlap
- No sync-enforced ordering (e.g. cond_signal())

Let's preempt threads during these accesses!

- Preempting only on API boundaries can miss bugs
- 1-pass data-race analysis has false positive problem
 - Using data races as PPs automatically verifies or refutes each data race candidate.

Landslide

Landslide [Blum '12]: simulator-based tester

- Originally built to test 15-410 Pebbles kernels
 - (a landslide tests the stability of pebbles)
- Wind River Simics: Full-system x86 simulator
 - Supports instruction/memory tracing, backtracking
 - No additional overhead for data race analysis

Landslide's features

- DPOR [Flanagan '05] prunes equivalences
- State space estimation guesses total time [Simsa '12]
- Bug detection: heap checking, deadlock, infinite loop...

Landslide & Quicksand

Quicksand: wrapper program for Iterative Deepening

- Manages work queue of jobs with different PP configs
- Each job is a new state space for Landslide to explore
- Prioritizes which jobs are important / likely to finish
 - Based on nature of PPs (data races? mutexes?)
 - Based on estimated completion time

Only required argument is user's CPU budget

Evaluation

PLDI Evaluation Plan

Comparing to single-state-space (SSS) testing

- Finding bugs faster, while SSS times out
- Finding new bugs with data-race PPs

Comparing to 1-pass data-race analysis

No details now, happy to explain at poster session

Testing OS projects from CMU, Berkeley, and U Chicago

- CMU: 79 "P2" thread libraries
- Berkeley & U Chicago: 79 "Pintos" kernels
- Under construction, feedback much appreciated!

Test Suite

15-410 – Undergrad Operating Systems at CMU

- P2: pthread-like userspace thread library
 - Sync primitives: mutexes, cvars, sems, rwlocks
 - Thread API: thr_create, thr_join, thr_exit
- 6 test cases (50 LOC average):
 - thr_exit_join, broadcast_test,
 paraguay, rwlock_test, paradise_lost
 - » Test of everything built on top of mutexes
 - mutex test
 - » Test of low-level lock implementation

mutex test is special...

Normally, we give a lock's internal accesses a "free pass".

This enables productive data-race analysis elsewhere.

```
lock() while (xchg(&lock->held,1)==1)
    yield(lock->owner);
    lock->owner = gettid();
              // in critical section...
              // access protected data...
unlock() { lock->owner = -1; lock->held = 0;
```

All accesses in this range are considered protected by the lock.

mutex test is special...

In mutex_test, we wish to verify the implicit assumption that other tests make that the locks are correct.

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Experimental Setup

79 P2s × 6 test cases × 10 CPU-hours ≈ 200 CPU-days

18 12-core (HT) 3.2 GHz Xeon machines, 12 GB RAM

Compare to a "control" experiment, also given 10 CPU-hours

- 1 state space, all "fixed" PPs enabled
 - mutex_lock()
 - mutex_unlock()
 - yield()
 - (cond_wait(), etc, have mutex_lock() PPs inside)

Do we find the same bugs faster?

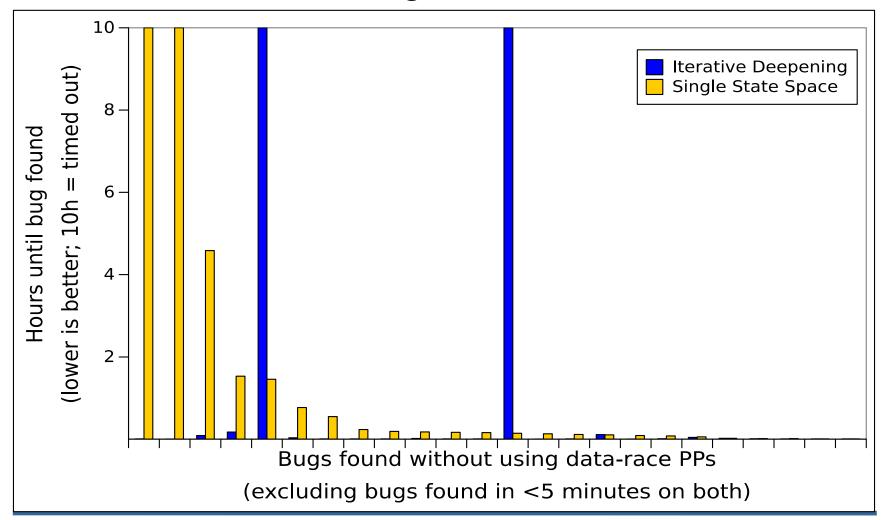
Control experiment with fixed PPs

- 70 nondeterministic bugs
- 90 timeouts

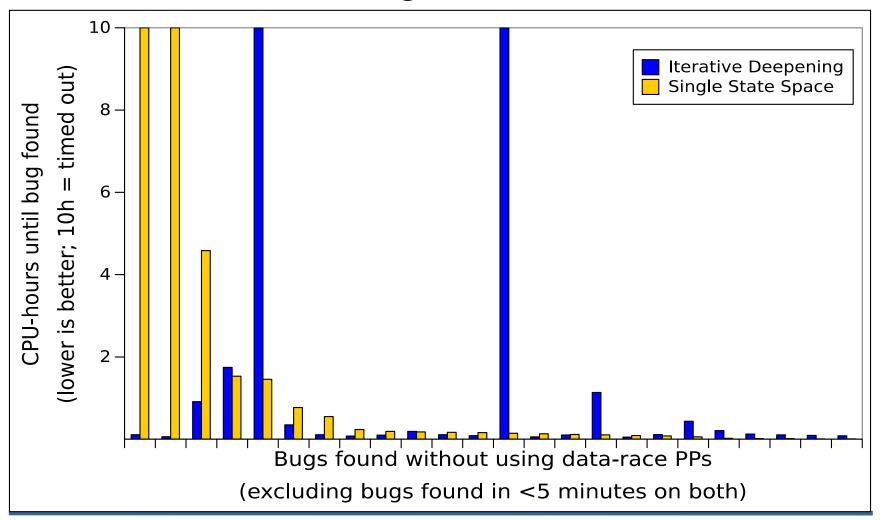
Quicksand with iterative deepening

- Found 68 of those 70
 - Maximal state space's ETA was too high
- Found 2 more (no DR PPs needed) in control time-outs
 - Found much faster in subset state spaces

Do we find the same bugs faster?



Do we find the same bugs faster?



Do we find new bugs with data-race PPs?

Quicksand with iterative deepening

- 107 nondeterministic bugs in total
- 33 required data-race PPs to expose!
 - 10 where control timed-out
 - 23 where control completed w/ no bug found
- mutex test
 - 12 bugs (to control's 1)
 - All 12 others found with data-race PPs

Conclusion

Future Work

Immediate future

- Ongoing user study with 15-410 students
 - Evaluate effectiveness for education/grading
- Finish PLDI evaluation plan, obviously
 - Have any experiment design ideas to share?

More uncertain future

- Smarter policies for too-large state spaces (ICB)
- Incorporate parallel DPOR to saturate CPUs better
- Applying to production, not student, code

Conclusion

Systematic Testing

- Controls scheduling rather than rely on randomness
- More reliable bug-finding than stress testing
- State space depends on preemption points (PPs)
- Completion time is exponential and unpredictable

Iterative deepening with Landslide & Quicksand

- Automatically searches for optimal set of PPs to use
- User need only supply CPU time budget
- Data race PPs uncover 50% more bugs
- PLDI submission soon; want evaluation feedback

References

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 Jiri Simsa. Runtime Estimation and Resource Allocation for Concurrency Testing. CMU-PDL-12-113. December 2012.

[Cui '13]

 Heming Cui et al. Parrot: A Practical Runtime for Deterministic, Stable, and Reliable Threads. SOSP 2013.

[Fonseca '14]

 Pedro Fonseca et al. SKI: Exposing Kernel Concurrency Bugs through Systematic Schedule Exploration. OSDI 2014.

Related Work

Systematic testing

- MaceMC (NSDI '07) liveness, random walking
- CHESS (PLDI '07) iterative context bounding (ICB)
- MoDist (NSDI '09) network/disk model checking
- dBug (SSV '10/SPIN '11) dynamic partial order reduction
- SimTester (VEE '12) interrupt injection, drivers
- Parrot (SOSP '13) dBug + deterministic userspace scheduler
- SKI (OSDI '14) randomly-chosen PPs for testing Linux
- SAMC (OSDI '14) annotated sync primitives, better reduction

Data race detection

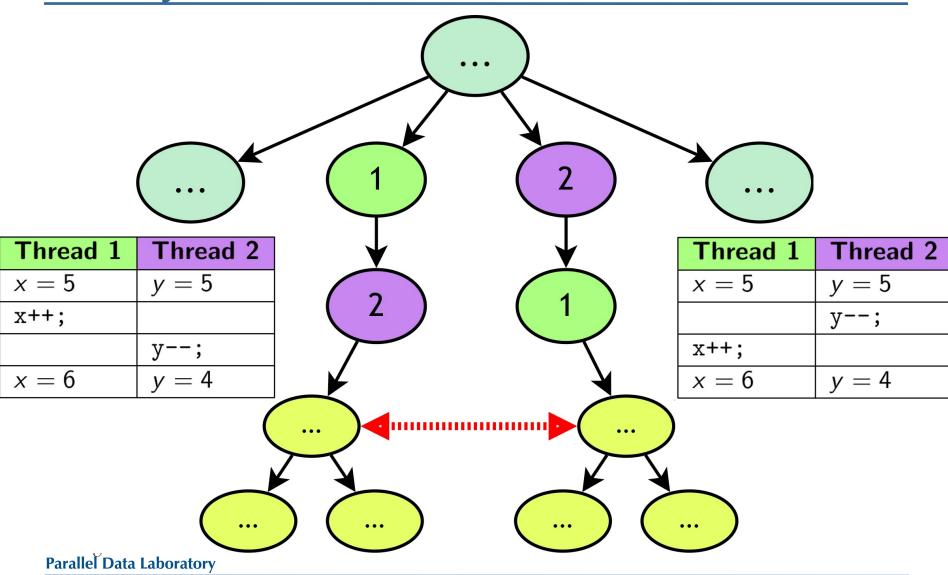
- Eraser (TOCS '97) lock-set tracking, annotations
- DataCollider (OSDI '10) random sampling, kernel
- RacePro (SOSP '11) inter-process races
- Portend (ASPLOS '12) alternate interleavings, symbolic execution

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Bonus Slides

Dynamic Partial Order Reduction



Among 474 total tests (P2+unit test pairs)...

• 23 deterministic bugs (e.g. use-after-free); fixed by hand

Control experiment with fixed PPs

- 70 nondeterministic bugs
- 90 timeouts

Quicksand with iterative deepening

- 107 nondeterministic bugs
 - 37 requiring data-race PPs to expose
 - mutex_test: 13 bugs, 12 requiring data-race PPs
- 2 bugs found without data-race PPs (control timeout)
- 2 bugs missed that control found (explain why)