# Automatically Repairing Concurrency Bugs with ARC

MUSEPAT 2013 • Saint Petersburg, Russia

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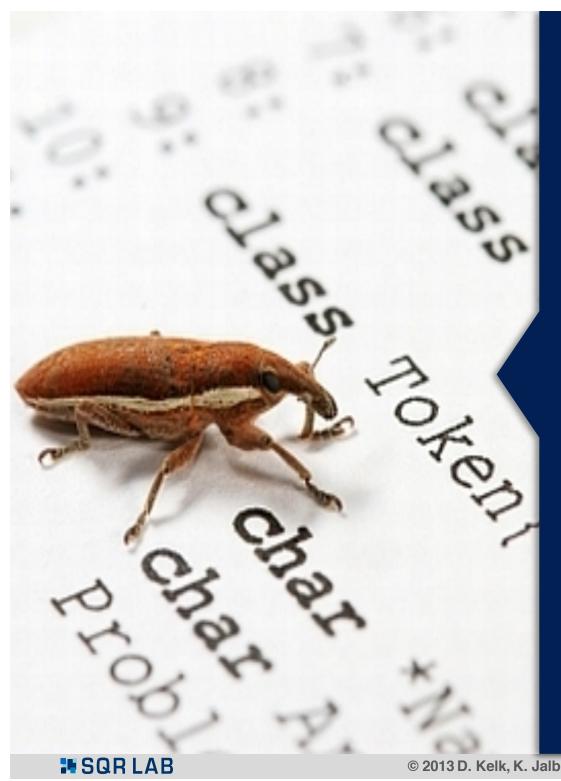
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What do we mean by concurrency bugs?

- There are many different kinds of concurrency bugs
- We focus on two of the most common kinds - data races and deadlocks

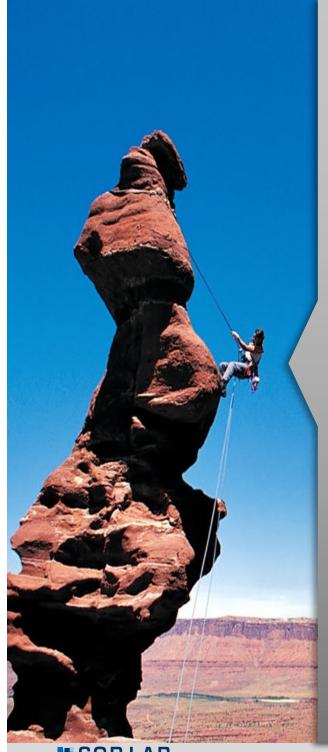
# What do we mean by bug repair?

 We view bug repair as a source code modification that fixes a concurrency bug while minimizing the effect on performance



# How do we approach bug repair?

- We use Search-Based Software Engineering (SBSE)
- Many software engineering problems can be expressed as optimization problems
- SBSE positions these problems in a search-based context
  - Applies meta-heuristic optimization techniques



Example SBSE techniques include hill climbing, particle swarm optimizations, genetic algorithms (GAs)...

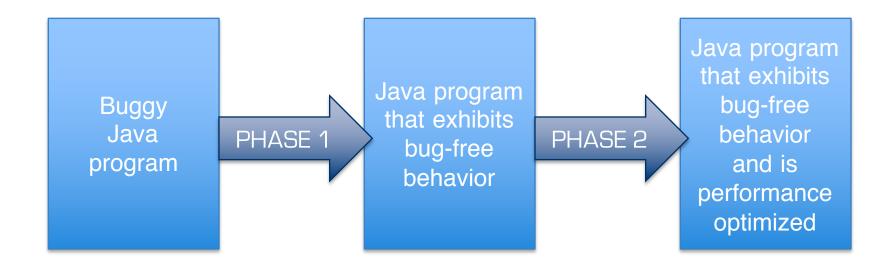


# **Automatic Repair of Concurrency Bugs**

- Several SBSE approaches have been proposed to fix bugs in single threaded programs [LDFW12, Arc11]
  - genetic programming is used to evolve patches, while testing evaluates fitness
- These techniques cannot be applied directly to fix concurrency bugs due to the nondeterministic nature of thread scheduling
- We adapt this work to handle concurrency bugs by modifying the fitness function and it's evaluation

[LDFW12] C. Le Goues, M. Dewey-Vogt, S. Forrest, and W. Weimer, "A systematic study of automated program repair: Fixing 55 out of 105 bugs for \$8 each," in Proc. of ICSE 2012, Jun. 2012. [Arc11] A. Arcuri, "Evolutionary repair of faulty software," in Applied Soft. Computing, vol. 11, 2011, pp. 3494–3514.

# ARC - Automatic Repair of Concurrency



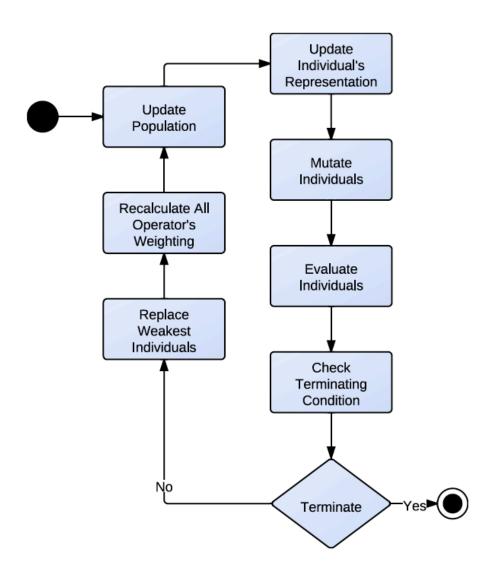
#### PHASE 1:

Repairing Deadlocks and Data Races

#### PHASE 2:

 Optimizing the Performance of Repaired Source Code

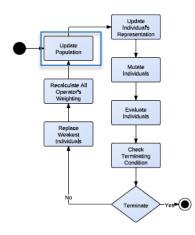
# Repairing Deadlocks and Data Races



#### **INPUT**

- 1. Java program with concurrency bugs
- 2. Set of JUnit tests
  - The test suite is the oracle (hence, the approach is only as good as the tests!)

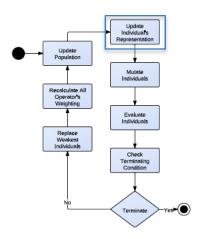
## Repairing Deadlocks and Data Races



#### 1. Update Population

- Create the population for the genetic algorithm (GA)
- The first generation is a set of copies of the original buggy program
- Subsequent generations will be updated based on the GA (described in future steps)

## Repairing Deadlocks and Data Races



#### 2. Generate mutants

- Use mutation operators to generate mutants for all members of the population
- The generated mutants are optimized using the static analysis tool Chord [NA07]
  - Allows mutation operators to target specific shared classes, methods and variables when generating mutants

In: Proc. of ACM SIGPLAN-SIGACT Symp. on Principles of Programming Languages (POPL 2007), pp. 327–338, Jan. 2007.



Mutation Operator <sup>1</sup> Description	Acronym
Add a synchronized block around a statement	ASAT
Add the synchronized keyword to the method header	ASIM
Add a synchronized block around a method	ASM
Change the order of two synchronized blocks order	CSO
Expand synchronized region after	EXSA
Expand synchronized region before	EXSB
Remove synchronized statement around a synchronized statement	RSAS
Remove synchronization around a variable	RSAV
Remove synchronized keyword in method header	RSIM
Remove synchronization block around method	RSM
Shrink synchronization block after	SHSA
Shrink synchronization block before	SHSB

 $^{1}\text{All mutation operators are written in the TXL source transformation language} - \underline{\text{http://www.txl.ca}}.$ 



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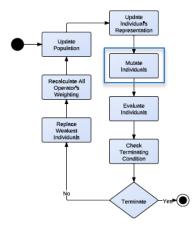
```
Program P:

obj.write(var1);
synchronized(lock) {
    myHash.remove(var1);
}

program P':

synchronized(lock) {
    obj.write(var1);
    myHash.remove(var1);
}
```

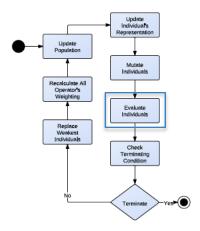
# ARC PHASE 1 Repairing Deadlocks and Data Races



#### 3. Apply mutation to an individual in population

- During execution the GA selects a type of mutation (i.e., a mutation operator) – random on first generation
- From the set of mutations created a random instance is used

## Repairing Deadlocks and Data Races



#### 4. Evaluate individuals

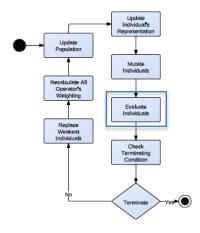
 Fitness function is used to evaluate mutants selected in previous step

```
functional fitness(P) = (s x sw) + (t x tw)

where:
    s = # of successful executions
    sw = success weighting
    t = # of timeout executions
    tw = timeout weighting
```

 The mutated individual is maintained only if the fitness function is improved

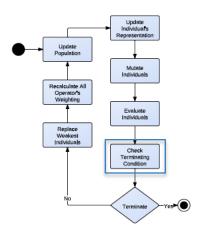
# ARC PHASE 1 Repairing Deadlocks and Data Races



#### 4. Evaluate individuals

- In order to evaluate the fitness function for a given individual we need to evaluate the function over many different interleavings/executions
- We use IBM's ConTest [EFN+02], which instruments the program with noise, to ensure that many interleavings are evaluated

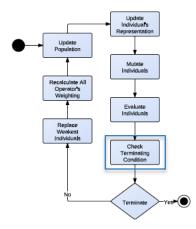
### Repairing Deadlocks and Data Races



#### 4. Check terminating condition

- An individual that produces 100% successful executions is a potential fix
- However, we perform an additional step to increase confidence that the individual is in fact correct
  - We evaluate the individual with ConTest again using a safety multiplier (e.g., 20) to increase the number of interleavings explored – a fix is only accepted if we achieved 100% success for this additional evaluation

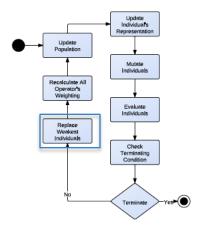
## Repairing Deadlocks and Data Races



#### 4. Check terminating condition

- Our approach will not always find a successful solution that repairs all of the concurrency bugs in a program
- If after a user-defined number of generations a solution is not reached our algorithm will terminate

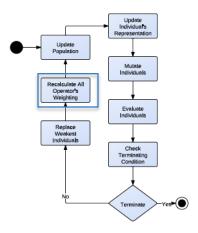
# ARC PHASE 1 Repairing Deadlocks and Data Races



#### 5. Replace Weakest Individuals (Optional)

- To encourage individuals to explore more fruitful areas of state space we can replace individuals
  - We can restart with original population
  - We can replace (e.g., 10%) of underperforming individuals with random high-performance individuals or with original program

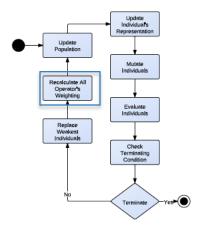
## Repairing Deadlocks and Data Races



#### 6. Calculate Operator Weighting

- We leverage historic information from previous generations to weight the operators and increase the likelhood that useful operators are selected first/more frequently
- Strategy 1: weight based on % of dead locks/data races uncovered
- Strategy 2: Weight based on a mutation operator's fitness function success

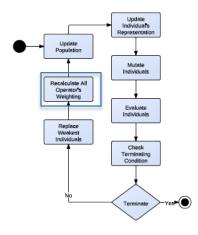
# ARC PHASE 1 Repairing Deadlocks and Data Races



#### 6. Calculate Operator Weighting

- Strategy 1: weight based on % of dead locks/data races uncovered
  - For example, some operators are geared towards fixing deadlocks, others data races and some both.
  - We increase the likelihood that specific operators are selected based on the number of deadlock and data races in our historic evaluations

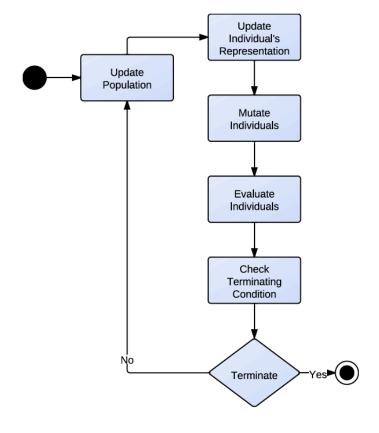
# ARC PHASE 1 Repairing Deadlocks and Data Races



#### 6. Calculate Operator Weighting

- Strategy 2: Weight based on a mutation operator's fitness function success
  - For example, operators that have historically increased the fitness are weighted proportional to their success

# ARC PHASE 2 Optimizing Repaired Source Code



- ARC may introduce unnecessary synchronization during Phase 1.
- If Phase 1 is successful, an optional second phase attempts to improve the running time of the program-under-repair by shrinking and removing unnecessary synchronization blocks
- A new non-functional fitness function and a subset of the TXL operators (e.g., RSAS, SHSA) are used

# **Evaluation - Setup**

- We selected a set of 8 programs from the IBM Concurrency Benchmark [EU04] that have deadlock or data race bugs
  - 6 programs that exhibit bugs ARC was designed to fix
  - 2 programs that ARC was not designed to fix (sanity check)
- Each program was analyzed using 5 executions of ARC

### Evaluation - Results<sup>1</sup>

Program	Bug Type	Bug Repaired?	# Generations to Repair Bug (Avg.)	Time Required to Repair Bug (Avg.)
account	Data Race	<b>✓</b>	5.0	08m 08s
accounts	Data Race	<b>✓</b>	1.0	44m 00s
bubblesort2	Data Race	<b>✓</b>	2.2	1h 40m 20s
deadlock	Deadlock	<b>✓</b>	1.0	02m 12s
lottery	Data Race	<b>✓</b>	2.4	38m 00s
pingpong	Data Race	<b>✓</b>	1.0	12m 32s
airline	Data Race	×	-	-
buffer	Data Race	×	-	-



<sup>&</sup>lt;sup>1</sup>Our evaluation was conducted on a Linux PC with a 2.33 GHz processor, 4 gigabytes of RAM running Linux Mint 13.

#### Related Work

- The use of ConTest to heal data races [LVK08] healing influences the schedule and does not repair the source code
- AFix [JSZ+11] fixes single-variable atomicity violations in C++ programs – combines dynamic bug analysis, patch creation and merging and dynamic testing
- Axis [LZ12] uses branch discrete control theory and invariants to fix any number of correlated atomicity violations

[LVK08] Letko, Z., Vojnar, T., Krena, B.: AtomRace: Data Race and Atomicity Violation Detector and Healer. In: *Proc. of Work. on Parallel and Distributed Sys.: Testing, Analysis, and Debugging (PADTAD 2008)*, 2008.

[JSZ+11] Jin, Guoliang, Linhai Song, Wei Zhang, Shan Lu, and Ben Liblital: Automated atomicity-violation fixing. In: *Proc. of ACM SIGPLAN Conf. on Prog. Lang. Design and Implementation (PLDI 2011)*, pp. 389–400, 2011.

[LZ12] Liu, P., Zhang, C.: Axis: automatically fixing atomicity violations through solving control constraints. In: *Proc. of Int. Conf. on Soft. Eng. (ICSE 2012)*, pp. 299–309, 2012.



#### **Conclusions**

- We automatically repair deadlocks and data races in concurrent Java programs using ARC
  - The goal of ARC is not only to ensure that a concurrency bug is repaired but also to optimize the performance of the program once the bug has been fixed
- To evaluate ARC, we conducted experiments using a set of 8 programs from the IBM Concurrency Benchmark
  - ARC was able to fix the data races and deadlocks in all 6 of the fixable programs

# Challenges & Future Work

- Scalability Although ARC was successful with the IBM Concurrency Benchmark programs we still need to evaluate ARC on larger projects
- Performance The use of testing with noise making to evaluate the variants of programs produced during ARC is costly
  - further heuristics and optimizations need to be explored

# Challenges & Future Work

- Flexibility ARC is currently only capable of fixing deadlocks and data races
  - We place to explore other mutation operators that will increase the kinds of bugs that can be fixed
- Readability [FLW12] automatic repair always has the potential to decrease the readability and maintainability of the source code
  - We have not studied the readability of the fixes produced by ARC

[FLW12] Zachary P. Fry, Bryan Landau, and Westley Weimer. "A Human Study of Patch Maintainability." In *Proc. of the International Symposium on Software Testing and Analysis (ISSTA)*, 177–187, 2012.

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