# Response property checking via distributed state space exploration

## **Brad Bingham** and Mark Greenstreet {*binghamb, mrg*}@*cs.ubc.ca*

Department of Computer Science University of British Columbia, Canada

October 24, 2014

FMCAD 2014

- $\bullet$  High-Level Models: use  ${\rm Mur}\varphi$  to describe a system
- Liveness: nice to verify, but challenging in practice
- Distributed Model Checking: memory and speed scalability
- Explicit-State: easy to distribute/parallelize
  - (Also outperforms symbolic methods for certain models)

**Our Goal:** Attack a practical liveness property called <u>response</u> with distributed, explicit-state model checking

#### Response and Fairness

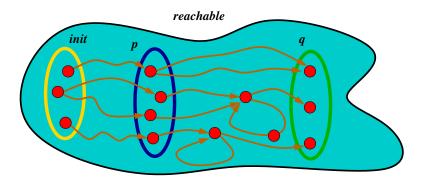
2 High Level Algorithm

#### 3 Our Implementation

- Distributed MC for Safety
- Adaptation for Response
- One Optimization (of many)

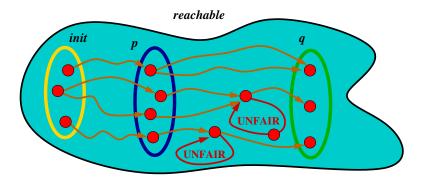
### 4 Results

### **Response Properties**



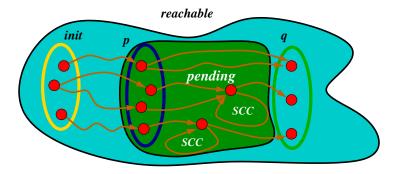
- "Will there always be a response?" ≡ "Does every fair path from each reachable *p*-state lead to a *q*-state?"
  - $p \equiv$  "request issued";  $q \equiv$  "request granted"
  - In LTL: fair  $\Rightarrow \Box(p \rightarrow \Diamond q)$
  - Most common/simplest notion of liveness

### **Response Properties**



- "Will there always be a response?" ≡ "Does every fair path from each reachable *p*-state lead to a *q*-state?"
  - $p \equiv$  "request issued";  $q \equiv$  "request granted"
  - In LTL: fair  $\Rightarrow \Box(p \rightarrow \Diamond q)$
  - Most common/simplest notion of liveness

### Response and Strongly Connected Components (SCCs)



- *pending*  $\equiv$  "states where the request is outstanding"
- The question fair ⇒ □(p → ◊q)? Is equivalent to asking "Is there a fair SCC within pending?"
  - Terminology: fair SCC  $\equiv$  FSCC

- In practice, we use fairness assumptions that reflect the underlying implementation
- Excludes unrealistic counterexamples
- We use action-based fairness:
  - An action *a* is a set of system transitions
  - a is called strongly-fair (aka compassionate; a ∈ C) if
     [a enabled ∞-often] ⇒ [a fires ∞-often]
  - *a* is called weakly-fair (aka just; *a* ∈ *J*) if
     [*a* presistently enabled] ⇒ [*a* fires]

- In practice, we use fairness assumptions that reflect the underlying implementation
- Excludes unrealistic counterexamples
- We use action-based fairness:
  - An action *a* is a set of system transitions
  - a is called strongly-fair (aka compassionate; a ∈ C) if
     [a enabled ∞-often] ⇒ [a fires ∞-often]
  - a is called weakly-fair (aka just; a ∈ J) if
     [a presistently enabled] ⇒ [a fires]
- Note: verifying *fair* ⇒ □(p → ◊q) with standard Büchi automata LTL MC approach will blow up
  - *i.e.*, property automata with size exponential in  $|\mathcal{C} \cup \mathcal{J}|$

#### 1 Response and Fairness

#### 2 High Level Algorithm

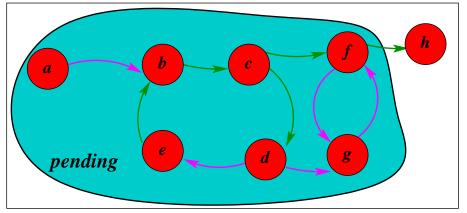
#### Our Implementation

- Distributed MC for Safety
- Adaptation for Response
- One Optimization (of many)

#### 4 Results

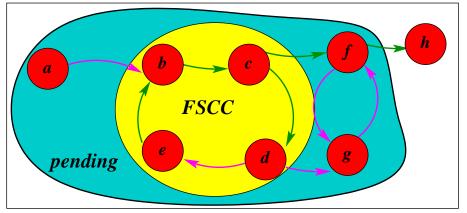
FSCCs

#### Both green actions and pink actions are strongly fair

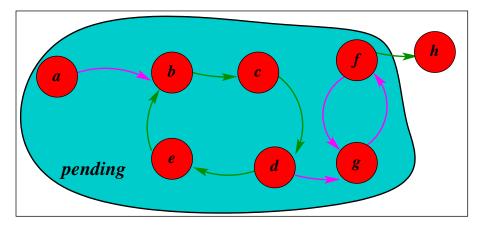


FSCCs

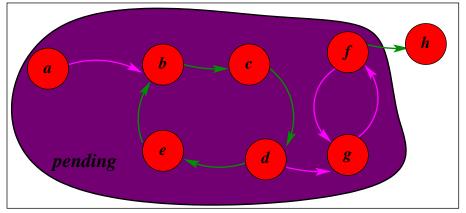
#### Both green actions and pink actions are strongly fair



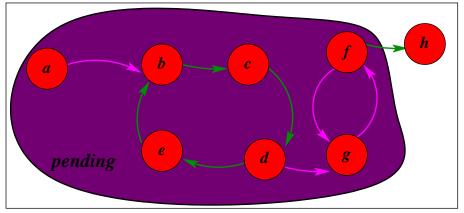
#### Both green actions and pink actions are strongly fair



#### Both green actions and pink actions are strongly fair Purple Blob $\equiv$ MaybeFair

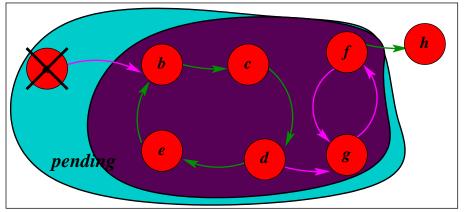


#### Both green actions and pink actions are strongly fair Purple Blob $\equiv$ MaybeFair



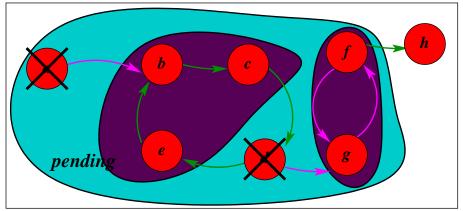
**Idea**: find unfair states by looking at previous actions within (*MaybeFair*)

#### Both green actions and pink actions are strongly fair Purple Blob $\equiv$ MaybeFair



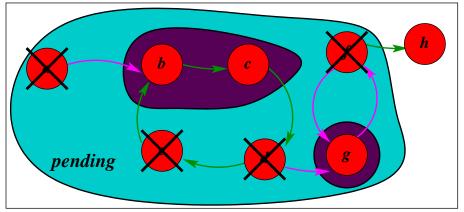
**Idea**: find unfair states by looking at previous actions within (*MaybeFair*)

#### Both green actions and pink actions are strongly fair Purple Blob $\equiv$ MaybeFair



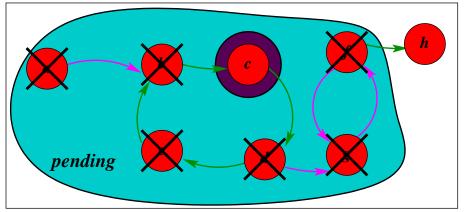
**Idea**: find unfair states by looking at previous actions within (*MaybeFair*)

#### Both green actions and pink actions are strongly fair Purple Blob $\equiv$ MaybeFair



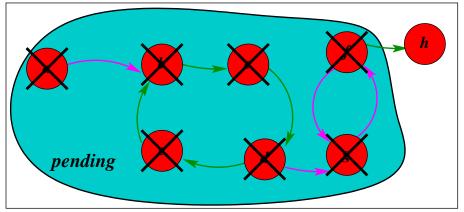
**Idea**: find unfair states by looking at previous actions within (*MaybeFair*)

#### Both green actions and pink actions are strongly fair Purple Blob $\equiv$ MaybeFair



**Idea**: find unfair states by looking at previous actions within (*MaybeFair*)

#### Both green actions and pink actions are strongly fair Purple Blob $\equiv$ MaybeFair



**Idea**: find unfair states by looking at previous actions within (*MaybeFair*)

- Suppose H ⊆ pending. Let ⟨H⟩ be the subgraph of the transition graph induced by H
- The Predecessor Actions for state s ∈ H, are actions appearing on some path that
  - **1** is contained within  $\langle H \rangle$ ; and
  - ends at s
- **Observe**: If *s* lies on a FSCC in  $\langle H \rangle$ , then all enabled strongly-fair actions at *s* are PAs
- Contrapositive: If there ∃ a strongly-fair action enabled at s that isn't a PA, then s does NOT lie on a FSCC in ⟨H⟩

- Suppose H ⊆ pending. Let ⟨H⟩ be the subgraph of the transition graph induced by H
- The Predecessor Actions for state s ∈ H, are actions appearing on some path that
  - **1** is contained within  $\langle H \rangle$ ; and
  - ends at s
- **Observe**: If *s* lies on a FSCC in  $\langle H \rangle$ , then all enabled strongly-fair actions at *s* are PAs
- Contrapositive: If there ∃ a strongly-fair action enabled at s that isn't a PA, then s does NOT lie on a FSCC in ⟨H⟩

#### ...and $\therefore$ remove *s* from consideration!

#### Response and Fairness

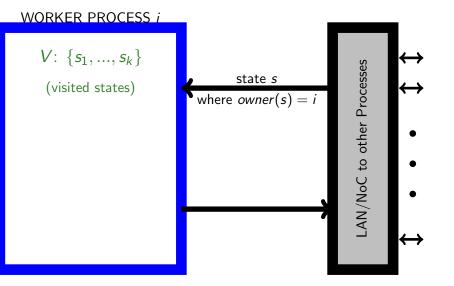
#### 2 High Level Algorithm

#### 3 Our Implementation

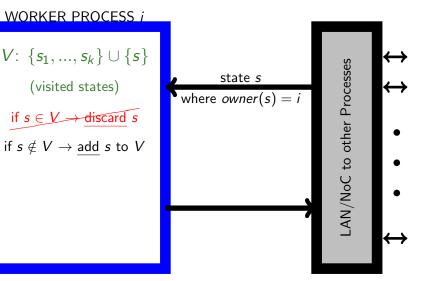
- Distributed MC for Safety
- Adaptation for Response
- One Optimization (of many)

#### 4 Results

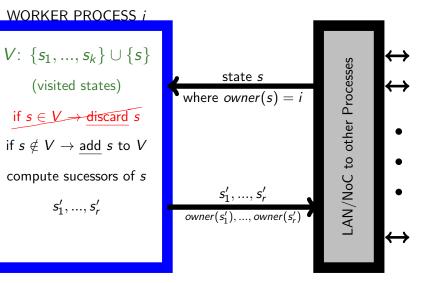
- Simple approach to distributing explicit-state model checking (for safety)
  - Use uniform random hash function *owner* : *States*  $\rightarrow$  *PIDs*
  - PID *i* only stores states *s* such that owner(s) = i.
- Each PID maintains two data structures:
  - V: Set of (owned) states visited so far
  - WQ: List of states waiting to be expanded
- Start: compute initial states and send to their owners
- Iterate: state sucessors are sent to their respective owners
- Termination: when each WQ is empty and no messages are in flight



### Message Flow



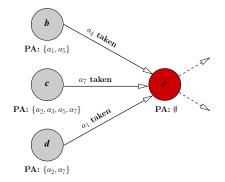
### Message Flow

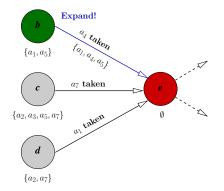


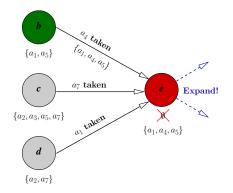
- For safety: use a  ${\rm Mur}\varphi$  hash table implementation that stores visited states as 40-bit values
  - Chance of a missed state, but typically it's a tiny chance  $(pprox 10^{-10})$
  - Once a state is inserted, it can't be recovered from its hash value
- For response: necessary to track extra information about states, for example
  - Is it a *pending*-state?
  - Is it in MaybeFair?
  - What are its predecessor actions, relative to (*MaybeFair*)?
- We use  $pprox 16 + |\mathcal{C} \cup \mathcal{J}|$  extra bits per state

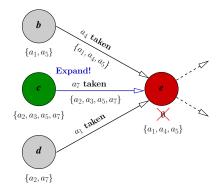
- Suppose  $C = \{a_1, ..., a_k\}$
- "Tag" each hash table entry with PAs, which is a subset of C
  (plus a few other bookkeeping bits)
- For states in  $s \in MaybeFair$ : initialize PA(s) to  $\emptyset$
- Message Passing:
  - Expand state s: if  $(s,s') \in a_i$ , send msg  $[s', PA(s) \cup \{a_i\}]$  to  $\mathit{owner}(s')$
  - Receive msg [s', F]: PA(s') := PA(s') ∪ F; expand state s' if PA(s') changed.
  - Continue until no further expansions.

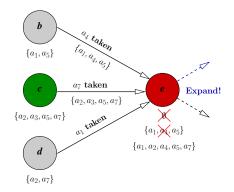
- Suppose  $C = \{a_1, ..., a_k\}$
- "Tag" each hash table entry with PAs, which is a subset of C
  (plus a few other bookkeeping bits)
- For states in  $s \in MaybeFair$ : initialize PA(s) to  $\emptyset$
- Message Passing:
  - Expand state s: if  $(s,s') \in a_i$ , send msg  $[s', PA(s) \cup \{a_i\}]$  to  $\mathit{owner}(s')$
  - Receive msg [s', F]: PA(s') := PA(s') ∪ F; expand state s' if PA(s') changed.
  - Continue until no further expansions.
- (A similar idea works for weakly-fair actions)

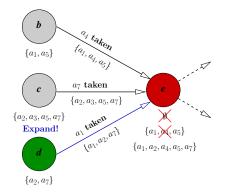


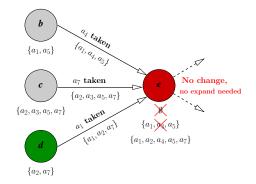






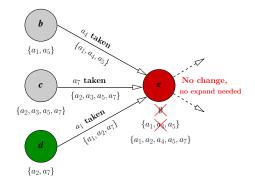






• Strongly-fair actions 
$$C = \{a_1, ..., a_7\}$$

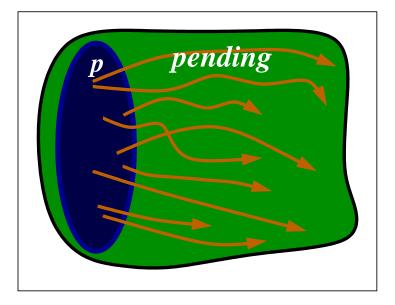
#### PA Propagation Example

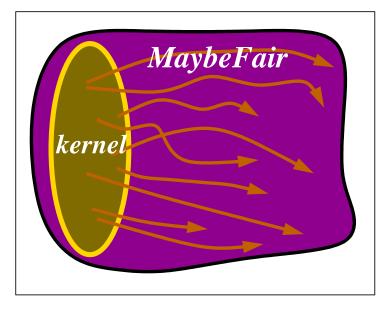


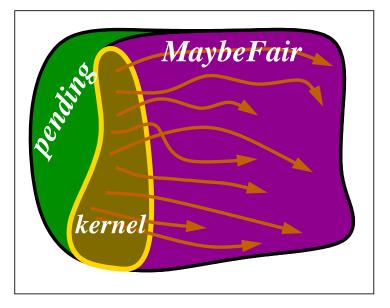
- Strongly-fair actions  $C = \{a_1, ..., a_7\}$
- (once PAs reach a fixpoint, remove unfair states from *MaybeFair*, clear the PAs and compute them again)

**Idea:** save set of states K to disk so that *MaybeFair* can be generated through reachability starting with K

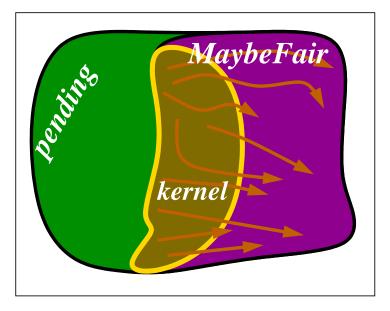
- Call K a kernel if  $MaybeFair \subseteq Reach(K)$ 
  - *i.e.*, *MaybeFair* is reachable starting from K
- Note: both initial states *I* and *p*-states are kernels for all subsets of *pending*
- To maintain *K*:
  - Initialize K to p-states;
  - If  $s \in K$  is removed from *MaybeFair*, then
    - Remove *s* from *K*;
    - Insert  $successors(s) \cap MaybeFair$  into K

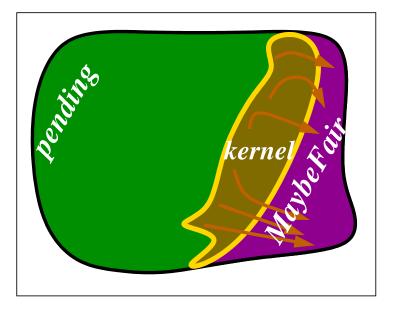


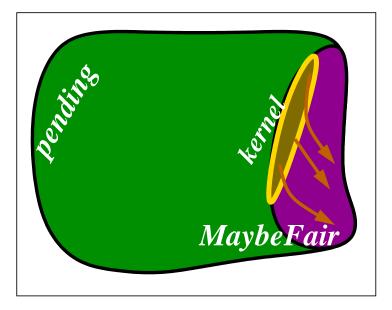


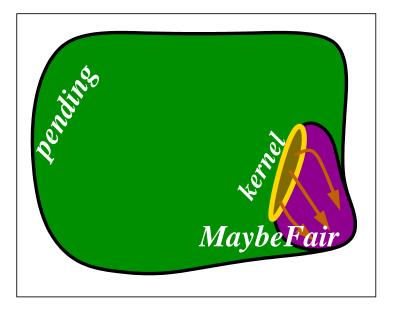


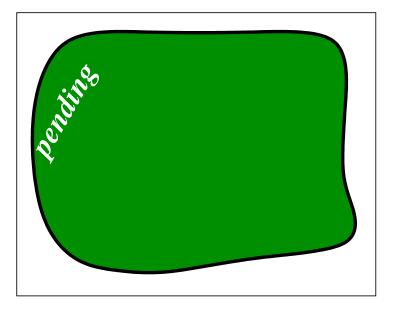
Bingham/Greenstreet (UBC)











#### Response and Fairness

2 High Level Algorithm

#### Our Implementation

- Distributed MC for Safety
- Adaptation for Response
- One Optimization (of many)

#### 4 Results

#### Performance

model	runtime*	states <sup>†</sup>	pending  <sup>†</sup>	exp/state
german5_sf	189	15.8	4.9	3.48
german6_sf	4253	316.5	95.3	3.33
peterson6_wf	820	13.8	12.1	12.91
peterson7_wf	26957	380.3	340.5	14.19
snoop2_sf	160	2.6	1.3	12.71
saw20_sf	323	0.3	0.3	44.06
gbn3_2_sf	369	12.8	7.9	6.44
swp4_2_sf	503	18.6	11.7	6.58
intelsmall_sf	285	0.5	0.3	6.36
intelmed_sf	1,015	2.7	1.9	8.59
intelbig_sf	13,872	51.8	29.9	11.92

• \*runtime is in seconds; <sup>†</sup>state counts in millions

- Blue: 40 processes running on 20 Core i7 machines (UBC)
- Green: 16 processes running on Xeon machines (Intel)

- Our approach does well in practice expands each state a small number of times (modest overhead compared with safety <sup>(i)</sup>)
  - (in the worst case, could expand each state O(mn<sup>2</sup>) times where m is # of fair rules and n number of states)
- Optimizations improve the performance by more than a factor of 2 on average
- Our tool is massively scalable can use on industrial problems

- Our approach does well in practice expands each state a small number of times (modest overhead compared with safety <sup>(i)</sup>)
  - (in the worst case, could expand each state O(mn<sup>2</sup>) times where m is # of fair rules and n number of states)
- Optimizations improve the performance by more than a factor of 2 on average
- Our tool is massively scalable can use on industrial problems
  Thank-you!

- Our approach does well in practice expands each state a small number of times (modest overhead compared with safety <sup>(i)</sup>)
  - (in the worst case, could expand each state O(mn<sup>2</sup>) times where m is # of fair rules and n number of states)
- Optimizations improve the performance by more than a factor of 2 on average
- Our tool is massively scalable can use on industrial problems Thank-you! Questions?

