Boosting *k*-Induction with Continuously-Refined Invariants

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Joint work with Dirk Beyer and Philipp Wendler

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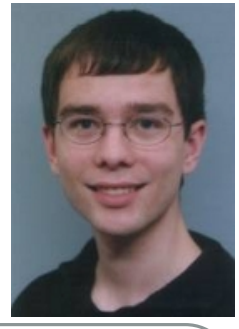

UNIVERSITÄT PASSAU

Fakultät für Informatik und Mathematik

CPA

SoSy-Lab

Software Systems

k-Induction for Software Verification

- Bounded Model Checking (BMC) is successful for finding bugs
- But not all loop bounds are small enough or even known/computable
- BMC is good for falsification, but often cannot prove absence of bugs

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- But not all loop bounds are small enough or even known/computable
- BMC is good for falsification, but often cannot prove absence of bugs
- (k-)Induction extends BMC towards unbounded safety proofs

1-Induction

- 1-Induction:
  - Check that the safety property holds in the first loop iteration: $P(1)$
  - Equivalent to BMC with loop bound 1
  - Check that the safety property is 1-inductive: $\forall n : P(n) \implies P(n+1)$

k-Induction

- k-Induction generalizes the induction principle:
  - Check that the property holds in the first k iterations:
    $\bigwedge_{i=1}^{k} P(i)$
  - Equivalent to BMC with loop bound k
  - Check that the safety property is k-inductive:
    $\forall n : \left(\bigwedge_{i=1}^{k} P(n+i-1) \right) \implies P(n+k)$
  - Stronger hypothesis is more likely to succeed [Wahl'13]
- Iteratively increase k

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- No!

Example

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int main() {
  unsigned int x1 = 0, x2 = 0;
  int s = 1;

  while (nondet()) {
    if (s == 1) x1++;
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    if (s == 5) s = 1;

    if (s == 1) assert(x1 == x2);
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- BMC?
  Unbounded loop.
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- k-Induction
  Hypothesis too weak!
  Needs $s > 0$

Further Strengthening

- Proofs still fail too often
- Introduce auxiliary invariants to strengthen the hypothesis:

$$\forall n : \left(\mathbf{Inv(n)} \wedge \bigwedge_{i=1}^{k} P(n+i-1) \right) \implies P(n+k)$$

- Auxilary invariants must hold
- Auxiliary invariants must be inductive
- Where do these invariants come from?

Auxiliary Invariants

- An additional component provides auxiliary invariants: The invariant generator
- Should be strong enough so that the proof succeeds
- Should not waste more resources than necessary

Experimental Results for k-Induction with static Invariant Generation by Abstract Interpretation

2 814 verification tasks taken from SV-COMP'15

Approach	KI	KI⇐AI		
		weakest	weak	strongest
Correct results	1 082	1 900	**1 934**	1 861
CPU time (h)	380	190	**180**	200
k-Values for correct safe results only:				
Max. final k	101	101	100	**86**

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Continuously-Refined Invariants

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- Should not waste more resources than necessary
- But no single fixed-precision configuration can provide this!
- Invariant generator can be run in parallel and provide invariants continuously
- Invariant generator improves invariants continuously over time
- Pick up current set of auxiliary invariants in each k-Induction iteration

Algorithm

Induction:

1: $k = 0$
2: **while** !finished **do**
3: BMC(k)
4: Induction(k, invariants)
5: k++

Invariant generation:

1: prec = <weak>
2: invariants = $\emptyset$
3: **while** !finished **do**
4: invariants = GenInv(prec)
5: prec = RefinePrec(prec)

Invariant Generation

How to generate invariants?

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- Option 3: Policy Iteration (see next talk)

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- ...

Experimental Results for *k*-Induction with Continuously-Refined Invariants

- 2 814 verification tasks taken from SV-COMP'15
- Best static configuration solved **1 934** tasks in **180** CPU hours

Approach	KI	KI⟲AI	KI⟲KI	KI⟲KI⟲AI
Correct Results	1 082	1 984	1 690	**2 005**
CPU Time (h)	380	170	240	**170**

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k-Induction in Other Tools: Comparison

Tool	CBMC	ESBMC		CPACHECKER	
Configuration		sequential	parallel	KI⟲	KI⟲AI
Correct results	1 216	2 214	2 137	2 005	
Wrong proofs	261	184	137	4	
Wrong alarms	4	28	24	25	
CPU time (h)	350	100	130	170	

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- Unsound approaches are not worth their trouble
- Bounded model checkers can easily be extended to provide proofs

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- Auxiliary invariants should be continously refined
- Combinations of KI and AI techniques are successful
- Unsound approaches are not worth their trouble
- Bounded model checkers can easily be extended to provide proofs
- Read the upcoming paper:
  Boosting k-Induction with Continuously-Refined Invariants [CAV'15]
  ... or email me at dangl@fim.uni-passau.de