

Latticed k -Induction with an Application to Probabilistic Programs

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Programming Principles, Logic
and Verification Group

- ▶ SAT-based technique for **verifying** (invariant properties of) finite **transition systems**

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Yes. Enables *fully automatic* verification of non-trivial properties.

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$$\forall \text{ initial states } \sigma: \quad \text{wp}[[C]](x)(\sigma) \leq \sigma(x) + 1$$

is **not inductive** but *2-inductive*.

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Goal: Prove that P covers all **reachable states** of TS

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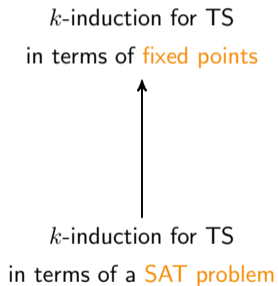
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... reason about **sets of paths** rather than individual paths

k -induction for TS
in terms of a SAT problem



latticed k -induction generalizes Park Induction for proving $\text{lfp } \Phi \sqsubseteq f$

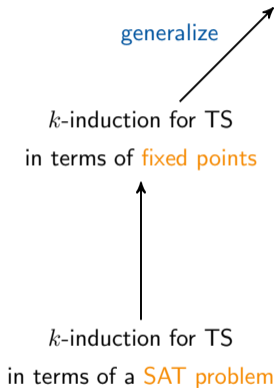
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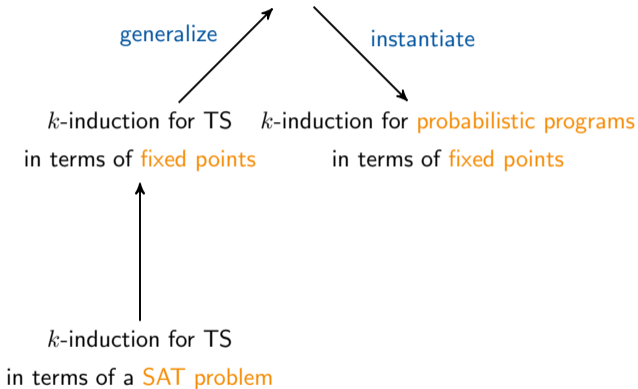
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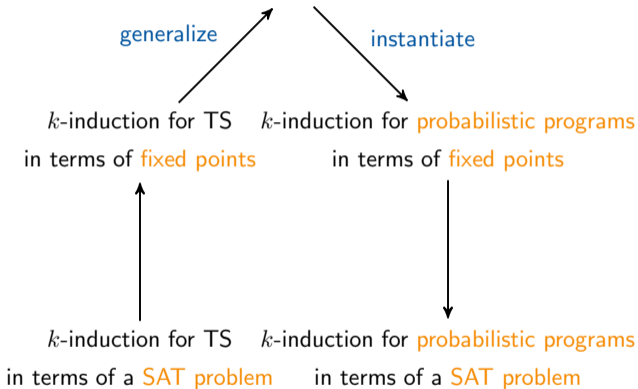
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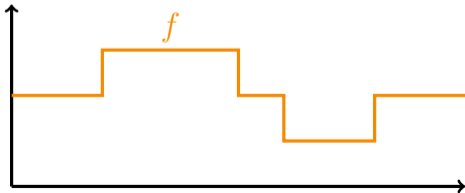
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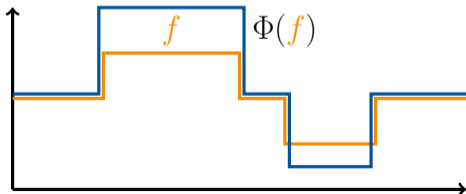
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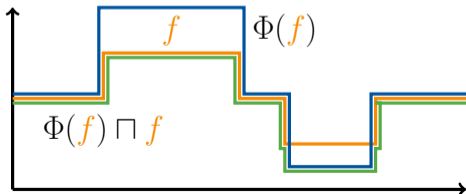
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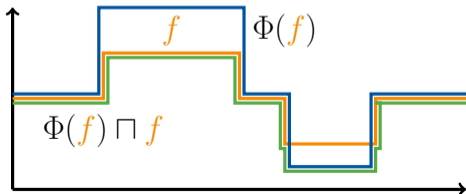
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Given: Complete lattice (E, \sqsubseteq) , monotonic $\Phi: E \rightarrow E$, and $f \in E$

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Define $\Psi_f: E \rightarrow E$ by

$$\Psi_f(g) = \Phi(g) \sqcap f .$$

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Theorem (Latticed k -Induction)

For every $k \geq 1$,

$$\Phi \left(\Psi_f^{k-1}(f) \right) \sqsubseteq f \text{ implies } \text{lfp } \Phi \sqsubseteq f .$$

We call such f k -inductive invariant.

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Can be generalized to transfinite κ -induction (not in this talk).

Theorem (Park Induction from k -Induction)

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Iterating Ψ_f on f yields a descending chain, i.e.,

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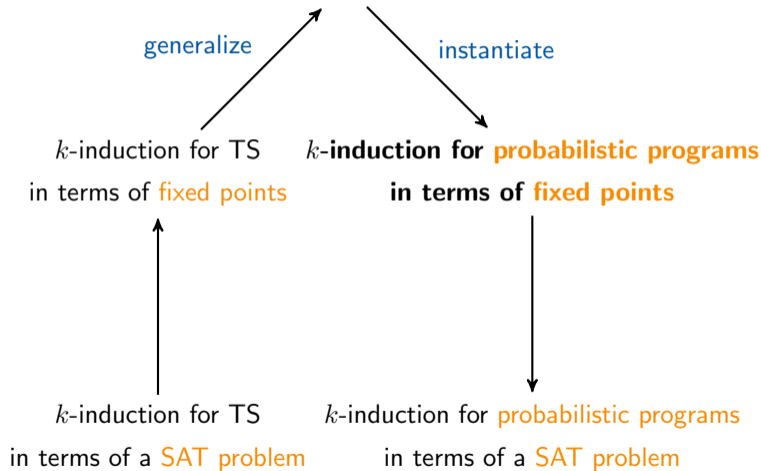
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- ▶ $\Psi_f^{k-1}(f)$ is an inductive invariant,
- ▶ which is stronger than f .

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Hence, latticed k -induction applies:

Corollary

For every $k \geq 1$,

$$\Phi \left(\Psi_f^{k-1}(f) \right) \leq f \quad \text{implies} \quad \text{wp}[[C]](g) \leq f .$$

Given: Loop $C = \text{while}(\varphi) \{ C' \}$ and $f, g \in \mathbb{E}$

Goal: Prove $\text{wp}[[C]](g) \leq f$

We have

$$\text{wp}[[C]](g) = \text{lfp } \Phi \quad \text{with } \Phi: \mathbb{E} \rightarrow \mathbb{E} \text{ monotonic .}$$

Hence, latticed k -induction applies:

Corollary

For every $k \geq 1$,

$$\Phi \left(\Psi_f^{k-1}(f) \right) \leq f \quad \text{implies} \quad \text{wp}[[C]](g) \leq f .$$

Here

$$\Psi_f(h) = \Phi(h) \sqcap f \quad \text{where for } h, h' \in \mathbb{E}, \quad h \sqcap h' = \lambda \sigma. \min\{h(\sigma), h'(\sigma)\} .$$

kipro2: k -Induction for PRObabilistic PROgrams



Given *linear* $C = \text{while}(\varphi)\{C'\}$ and *piecewise linear* f, g , kipro2 *semi-decides* by SMT:

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Given *linear* $C = \text{while}(\varphi)\{C'\}$ and *piecewise linear* f, g , kipro2 *semi-decides* by SMT:

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If $\text{wp}[[C]](g) \not\leq f$, KIPRO2 finds via *bounded model checking* some $\sigma \in \Sigma$ with

$$\text{wp}[[C]](g)(\sigma) > f(\sigma) .$$

For C given by

$$\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \} ,$$

the property

$$\forall \text{ initial states } \sigma: \quad \text{wp}[[C]](x)(\sigma) \leq \sigma(x) + 1$$

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also hold? **No; counterexample by BMC:** $\sigma(c) = 1, \sigma(x) = 6$.

Sampling uniformly from $\{elow, \dots, ehigh\}$ using fair coin flips only [Lumbroso 2013]:

```
while(running = 0){  
  
  v := 2*v;  
  {c := 2*c+1}[0.5]{c := 2*c};  
  if((not (v<n))){  
    if((not (n=c)) & (not (n<c))){ # terminate  
      running := 1  
    }{  
      v := v-n;  
      c := c-n;  
    }  
  }{  
    skip  
  }  
  
  # On termination, determine correct index  
  if((not (running = 0))){  
    c := elow + c;  
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    skip  
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For *arbitrary* array of fixed size

$n = \{2, 3, 4, 5\}$, we verify

$$\Pr(\text{"sample fixed element"}) \leq \frac{1}{n} .$$

Table 2: Empirical results for the first benchmark set (time in seconds).

	postexpectation	variant	result	k	#formulae	formulae_t	sat_t	total_t
geo	c	1	ind	2	18	0.01	0.00	0.08
		2	ref	11	103	0.04	0.01	0.09
		3	ref	46	1223	0.39	0.04	0.48
unif_gen	$[c = i]$	1	ind	2	267	0.27	0.02	0.56
		2	ind	3	1402	1.45	0.10	1.81
		3	ind	3	1402	1.48	0.11	1.86
		4	ind	5	40568	47.31	15.70	63.28
		5	TO	—	—	—	—	—
⋮								

- ▶ k -Induction for transition systems in terms of fixed points

Conclusion

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Thank you!