

Latticed k -Induction

with an Application to Probabilistic Programs

Mingshuai Chen

—Joint work with K. Batz, B. L. Kaminski, J.-P. Katoen, C. Matheja, P. Schröder—



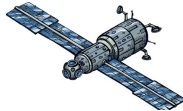
Cyber-Physical Systems (CPS)

An open, interconnected form of embedded systems that integrates capabilities of *computing*, *communication*, and *control*, among which many are **safety-critical**.



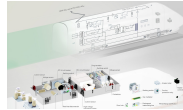
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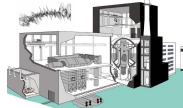
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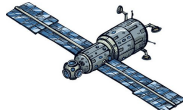
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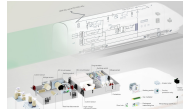
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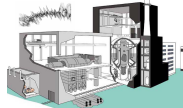
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"How can we provide people with CPS they can bet their lives on?"

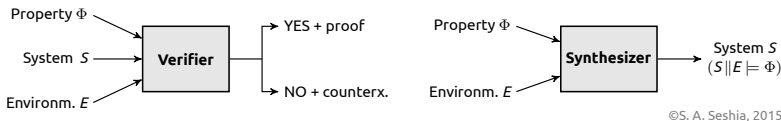
— Jeannette Wing

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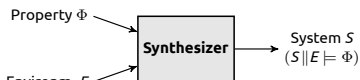
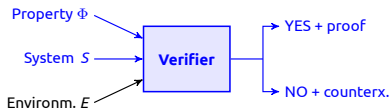
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Develop mathematically rigorous techniques for designing safety-critical CPS while pushing the limits of automation as far as possible.

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Is k -induction applicable to verifying **infinite-state probabilistic programs**?

⇓ **Latticed k -Induction**

Yes. It enables **fully automatic** verification of non-trivial properties.

k -Induction for Probabilistic Programs — Intuition

For a probabilistic loop C :

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

k -Induction for Transition Systems

Given : $TS = (S, I, T)$, invariant property $P \subseteq S$.

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

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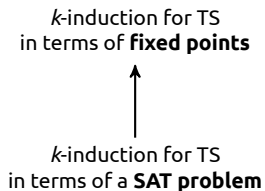
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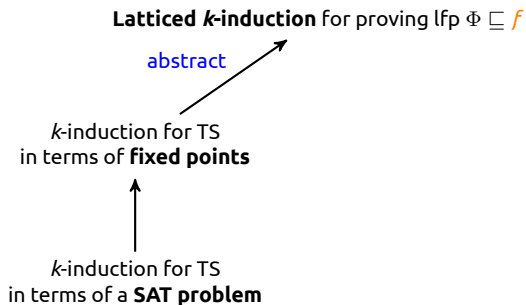
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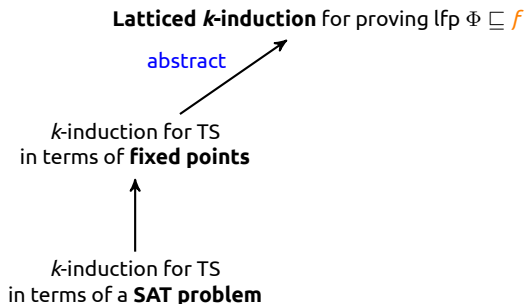
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- leave the Boolean domain and reason about **quantities**;
- reason about **sets of paths** rather than individual paths.

k -induction for TS
in terms of a **SAT problem**

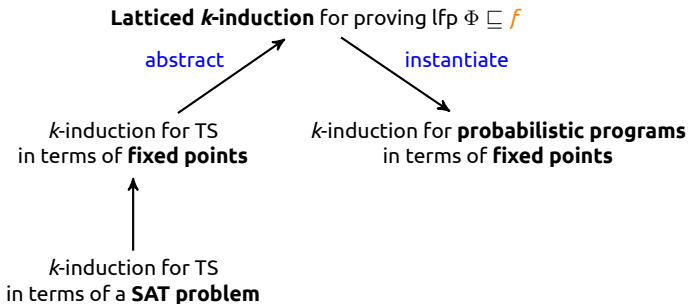




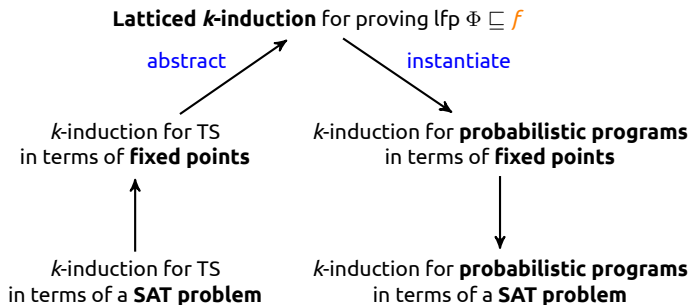


Whenever your problem boils down to verifying an upper bound on a lfp ,
latticed k-induction provides you with **inductive proof rules**!

Idea Sketch

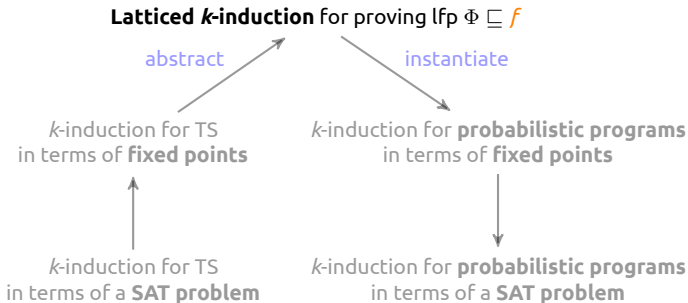


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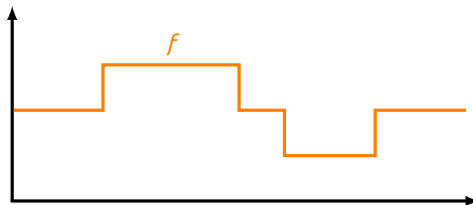
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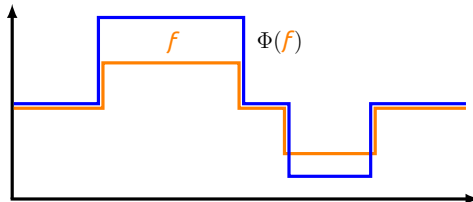
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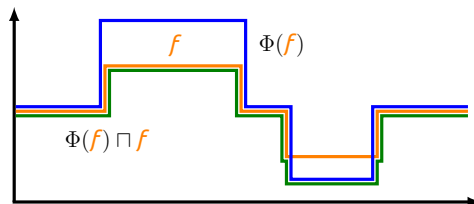
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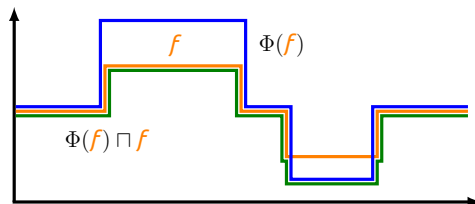
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Theorem (Latticed k -induction)

For every $k \geq 1$,

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We call such f k -inductive invariant.

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

Key Insights of Soundness

Lemma (Descending chain)

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

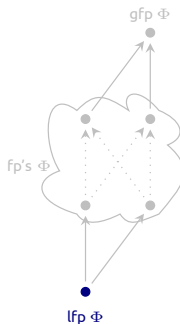
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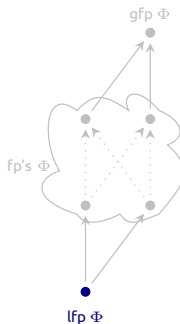


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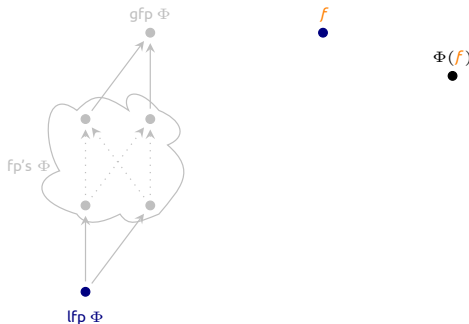


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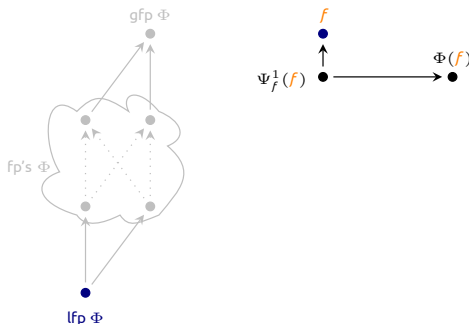


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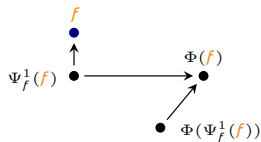
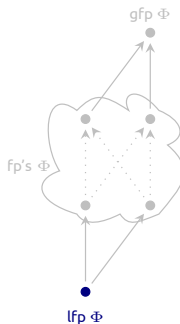


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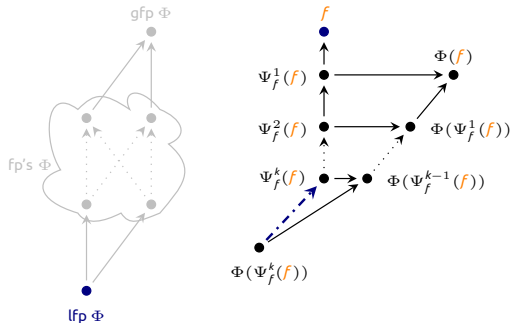


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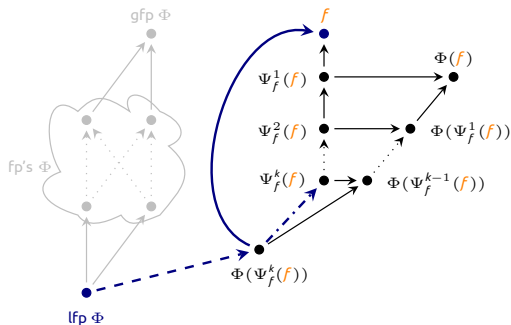


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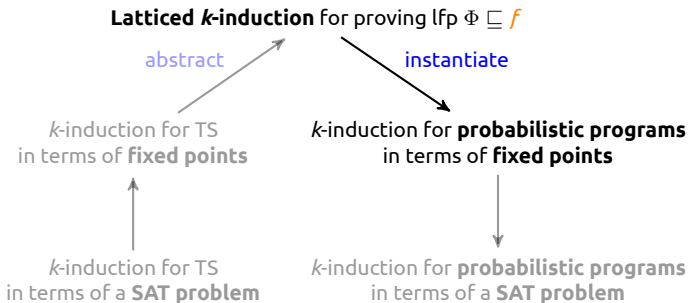
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f is a k -inductive invariant $\iff \Psi_f^{k-1}(f)$ is an inductive invariant *stronger than* f .

Idea Sketch



Weakest Preexpectation Transformer

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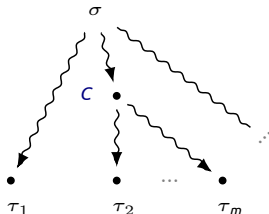
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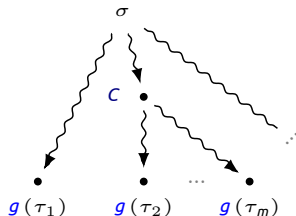
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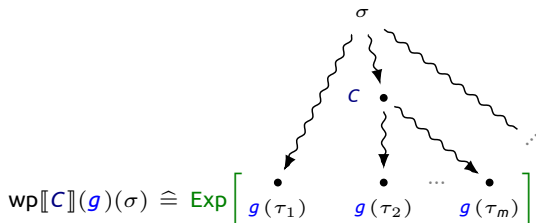
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k -Induction for Probabilistic Programs

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

Goal : Prove $\text{wp}\llbracket C \rrbracket(g) \leq f$.

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Here

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

Tool Support

kipro2: k -Induction for PRObabilistic PROgrams

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For *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

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

Example — Geometric Distribution

For C given by

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

the property

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

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

Example — Uniform Sampling by Fair-Coin Flips [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;  
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    }  
    }{  
        skip  
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    #on termination, determine correct index  
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For *arbitrary* array of fixed size
 $n = \{2, 3, 4, 5\}$, we verify

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

Empirical Results (Partial)

	postexpectation	variant	result	k	#formulae	formulae_t	sat_t	total_t
geo	x	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_san	$[c = i]$	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
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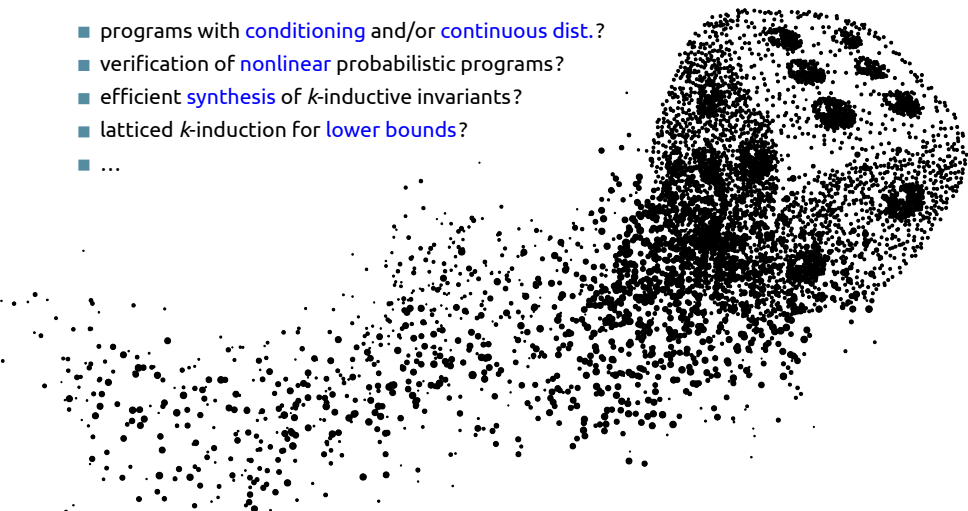
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Future Directions

- programs with **conditioning** and/or **continuous dist.**?
- verification of **nonlinear** probabilistic programs?
- efficient **synthesis** of k -inductive invariants?
- latticed k -induction for **lower bounds**?
- ...



In Combination with Latticed BMC

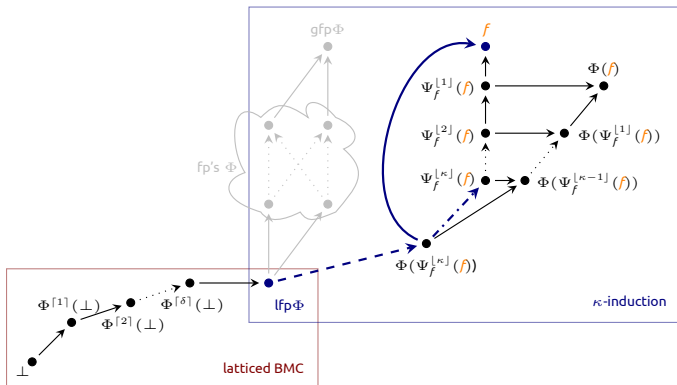


Figure – κ -induction and latticed BMC in case that $\text{lfp } \Phi \sqsubseteq f$.