

Algebra-Based Loop Analysis

Laura Kovács TU Wien Vienna, Austria

ABSTRACT

Automating loop analysis, and in particular synthesizing loop invariants, is a central challenge in the computer-aided verification of programs with loops, with applications in compiler optimization, probabilistic programming and IT security. While this challenge is in general undecidable, several techniques have emerged to automatically summarize the functional behaviour of software loops, thus providing inductive loop invariants that may prevent programmers from introducing errors while making changes in their code. In this tutorial, we show that novel combinations of methods from computer algebra, algorithmic combinatorics and static loop analysis provide powerful workhorses to derive (all) polynomial loop invariants, synthesize affine loops from invariants, and infer quantitative properties over the value distributions of probabilistic loop variables.

CCS CONCEPTS

- Theory of computation → Invariants; Logic and verification;
- Mathematics of computing \rightarrow Combinatorics.

KEYWORDS

Program Verification, Loop Synthesis, Loop Invariants, Algebraic Recurrences, Polynomial Ideals

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OVERVIEW

One of the main challenges in computer-aided formal software verification comes with the burden to verify or synthesize recursive behavior arising from (unbounded) software loops.

Intuitively, formal verification assigns to every program statement meaning in form of a logical formula, and then tries to prove that every formula is implied by the previous ones, starting with the precondition [3–5]. If we are able to prove that eventually the postconditions holds, we established the correctness of the program with respect to the given precondition and postcondition. Within this process of deductive verification, arguably the most challenging program constructs are the ones with recursive behavior such as loops. For (unbounded) loops, it is not known how often a loop



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is iterated when statically analyzing the program, that is, there is no straightforward way to assign a logical formula to loops. Formal verification therefore requires additional program assertions for loops in the form of *inductive loop invariants*. Intuitively, a loop invariant is a formal description of the behavior of the loop, expressing loop properties that hold before and after each loop iteration. In order to automate formal verification, we need rigorous techniques for automatically inferring and analyzing loop invariants.

The present tutorial overviews algebra-based algorithms for loop analysis. The key ingredients of our work root in the combination of the following four approaches:

- (i) static analysis, to model loops as algebraic recurrence equations over the loop counter and to impose structural constraints over loops in order to guarantee that the resulting recurrences equations can be reduced to linear recurrences with constant coefficients [10];
- (ii) *symbolic summation and polynomial algebra*, to solve recurrences and eliminate loop counters/recurrence variables by applying Gröbner basis computation over closed-forms [7, 9];
- (iii) applied statistics, to compute quantitative loop invariants over higher-order statistical moments of polynomial probabilistic loops [11];
- (iv) automated reasoning, to simplify loop constraints and generate invariants towards ensuring functional correctness of loops [6, 8].

Illustrative Examples

We motivate (applications of) our work with the loops of Figure 1. Within *invariant synthesis*, we compute the polynomial relation $c=n^3$ as a *polynomial loop invariant* of the polynomial loop of Figure 1. On the other hand, within *loop synthesis*, we strength reduce the polynomial loop of Figure 1 to the linear loop of Figure 1, by replacing loop variable multiplications with additions among program variables; doing so, we ensure that the linear loop of Figure 1 satisfies the polynomial loop invariant $c=n^3$ of the polynomial loop of Figure 1. In other words, the polynomial and linear loops of Figure 1 both satisfy the invariant $c=n^3$, and hence these loops are equivalent modulo the invariant $c=n^3$.

While synthesizing loop invariants from loops is crucial for formal software verification, synthesizing loops from loop invariants is in particular useful within program optimization and repair [2]. The combination of the algebraic techniques presented in this tutorial therefore provide efficient workhorses both for software verification and compiler optimization. The strength of this combined framework stems from the fact that we compute the set of *all polynomial equality invariants* for a given (polynomial) loop, that is, we precisely capture the behavior of the loop. Furthermore, the algebraic problem constructed by our synthesis procedure precisely

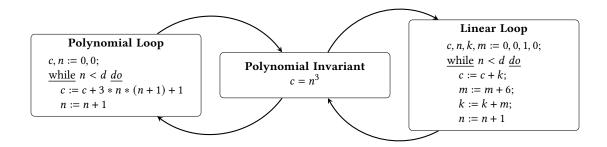


Figure 1: Algebra-Based Loop Analysis - Synthesizing Invariants and Loops

characterizes the set of *all* (*linear*) *loops*, modulo additional structural constraints, that satisfy the polynomial invariants given by the invariant generation procedure.

The central aspect of our work is that we represent numeric loops as systems of recurrence equations and employ properties of those recurrence systems to either generate invariants for a given loop or synthesize loops satisfying a given invariant. Our framework naturally extends to arbitrary loops whose functional summaries can be described by or reduced to linear recurrences with constant coefficients over loop counters [1], allowing for example the quantitative analysis of probabilistic loops by solving recurrences over expected values of loop variables [11].

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