

Interval-Adjoint Significance Analysis: A Case Study

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2nd Workshop on Approximate Computing, Prague, January 2016

Outline

SCoRPIo

Interval-Adjoint Significance Analysis

Overview

Fundamentals of Interval-Adjoint Significance Analysis

Extension of the Interval-Adjoint Significance Analysis

Case study: 1-D heat equation

Summary of Results and Future Work



SCoRPiO

Significance-Based Computing for
Reliability and Power Optimization



Motivation

- ▶ Compute the significance of each arithmetic operation for a given source code
- ▶ Significance identifies less important operations with regards to the quality of the output
- ▶ Insignificant operations can be approximated



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Approximation

Save energy by

- ▶ performing computations on low-power but unreliable hardware
- ▶ using lower precision on less power consuming hardware
- ▶ replacing computations with representative expressions or constants

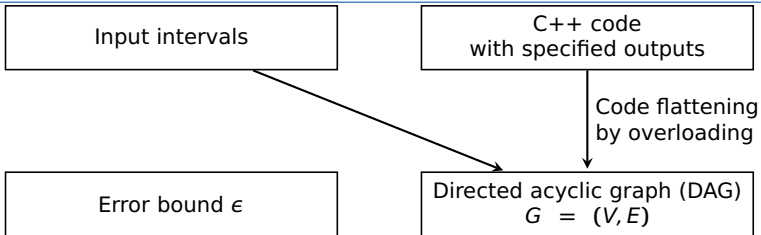
Interval-Adjoint Significance Analysis

Input intervals

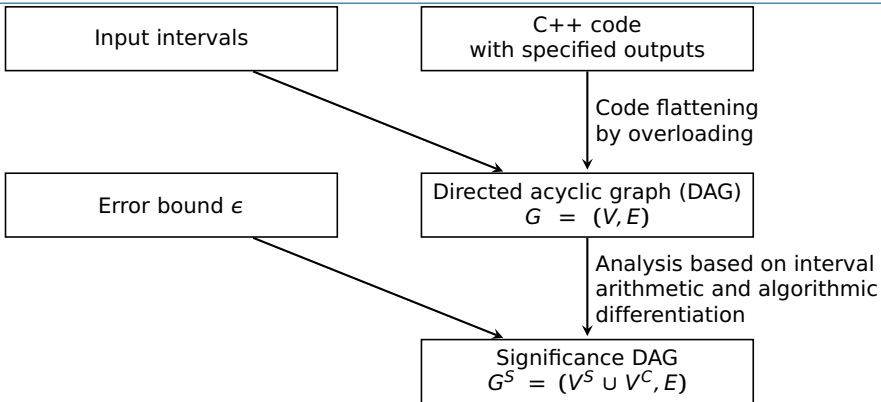
C++ code
with specified outputs

Error bound ϵ

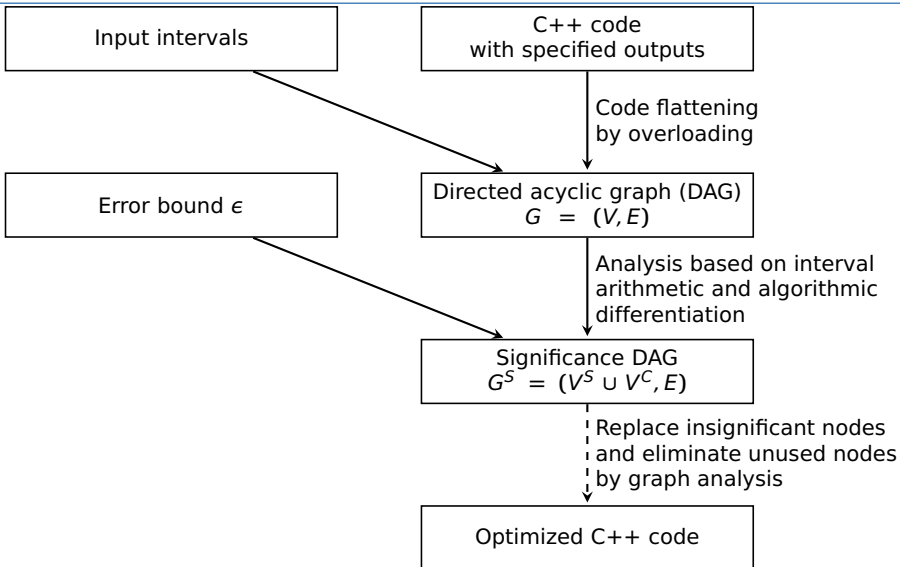
Interval-Adjoint Significance Analysis



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Fundamentals of Interval-Adjoint Significance Analysis

- ▶ Interval Arithmetics (IA)
 - ▶ Forward propagation of intervals
 - ▶ Computes interval enclosures of all intermediates and outputs for given inputs
 - ▶ Issues: relational operators, overestimation (e.g. wrapping effect)

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$$S_y(v_i) = w[v_i] \cdot \max |\nabla_{[v_i]}[y]|$$

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- ▶ $w[v_i]$: influence of the input domain on variable v_i
- ▶ $\max |\nabla_{[v_i]}[y]|$: influence of variable v_i on the output
- ▶ Variable v_i is insignificant if

$$S_y(v_i) \leq \epsilon$$

Interval Splitting

- ▶ Control flow splitting:
 - ▶ Enables computation of piecewise differentiable functions

	$v < c$ with $v \in [\underline{v}_i, \overline{v}_i]$	
$\overline{v}_i < c$	true	
$\underline{v}_i \leq c \leq \overline{v}_i$	true	if $v \in [\underline{v}_i, c)$
$c < \underline{v}_i$	false	if $v \in [c, \overline{v}_i]$

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- ▶ Ranking of scenario k for example with number of insignificant variables $|V_k^C|$ and input domain size $\prod_{i=1}^n w(D_k^i)$ by

$$r_k = |V_k^C| \cdot \prod_{i=1}^n w(D_k^i)$$

Case study: 1-D heat equation

Given C++ code for solving the 1-D heat equation

▶ Input:

rod length	L	=	2
spatial discretization	n	=	41
temporal discretization	m	=	800
final time	t_f	=	32
heat coefficient	c	=	0.01
initial rod temperature	$T(0, x)$	=	$[280, 300], \quad 0 \leq x \leq 2$
temperature at the boundary	$T(t, 0)$	=	$[280, 300], \quad t \in [0, t_f]$
temperature at candle	$T(t, L)$	=	$[1650, 1700], \quad t \in [0, t_f]$

▶ Output:

temperature at the middle $T(t_f, L/2)$

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Result of IASA with error bound $\epsilon = 0$

- ▶ 2,702,061 nodes
- ▶ 1,292,973 nodes (48%) marked as insignificant
 → IASA exposed automatically: code is inefficient

Case study: 1-D heat equation

Model

Heat distribution:

$$\frac{\partial T}{\partial t} = c \cdot \frac{\partial^2 T}{\partial x^2}, \quad T = T(t, x, c),$$

The initial condition:

$$T(0, x, c) = T_0 \quad \forall x \in (0, 2)$$

Boundary conditions:

$$T(t, 0, c) = T_{x1}, T(t, 2, c) = T_{x2} \quad \forall t \in [0, t_f]$$

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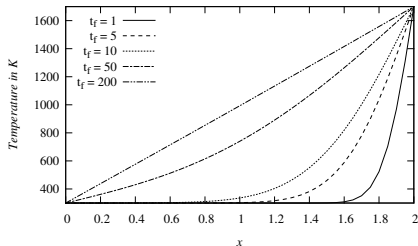
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Case study: 1-D heat equation

Discretized System

- ▶ Number of spatial discretization points n
- ▶ Number of temporal discretization points m
- ▶ Linear system

$$\begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ -a & 1+2a & -a & \ddots & \vdots \\ 0 & \ddots & \ddots & \ddots & 0 \\ \vdots & \ddots & -a & 1+2a & -a \\ 0 & \dots & 0 & 0 & 1 \end{pmatrix} T^{k+1} = T^k, \quad k = 1, \dots, m$$

- ▶ with

$$a = \frac{c\Delta t}{\Delta x^2}$$

Dense solver

- | | |
|---|---------------------------|
| ▶ dense LU decomposition of system matrix A | $\mathcal{O}(n^3)$ |
| ▶ for $t = 1, \dots, m$ | |
| ▶ dense forward substitution | $\mathcal{O}(n^2)$ |
| ▶ dense backward substitution | $\mathcal{O}(n^2)$ |
| <hr/> | |
| ▶ total complexity of the code with a dense solver: | $\mathcal{O}(mn^2 + n^3)$ |

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 - ▶ dense backward substitution $\mathcal{O}(n^2)$
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- ▶ total complexity of the code with a dense solver: $\mathcal{O}(mn^2 + n^3)$

Tridiagonal sparse solver

- ▶ sparse LU decomposition of system matrix A $\mathcal{O}(n)$
 - ▶ for $t = 1, \dots, m$
 - ▶ sparse forward substitution $\mathcal{O}(n)$
 - ▶ sparse backward substitution $\mathcal{O}(n)$
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- ▶ total complexity of the code with a sparse solver: $\mathcal{O}(mn)$

Case study: 1-D heat equation

Result of IASA with error bound $\epsilon = 0$ for the sparse code

- ▶ 161,018 nodes
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New Configuration

- ▶ Error bound $\epsilon = 1$
- ▶ Quantification mode:
 - ▶ Binary splitting
 - ▶ Only splitting of final time t_f

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Interval Splitting Results and Ranking

k	t_f	$ V_k^C $	$\frac{r_k}{1000}$	k	t_f	$ V_k^C $	$\frac{r_k}{1000}$
1	[0, 1]	95583	95583	6	[2, 4]	16412	32824
2	[1, 2]	71007	71007	7	[4, 5]	31887	31887
3	[0, 2]	30832	61664	8	[5, 6]	27068	27068
4	[2, 3]	51363	51363	9	[0, 4]	5964	23856
5	[3, 4]	39132	39132	10	[6, 7]	23594	23594

Case study: 1-D heat equation

Results of the case study

- ▶ Quantification mode yields that for small final simulation times most of the computations are insignificant
- ▶ 95,583 of 161,018 computations are marked as insignificant for $t_f = 1$

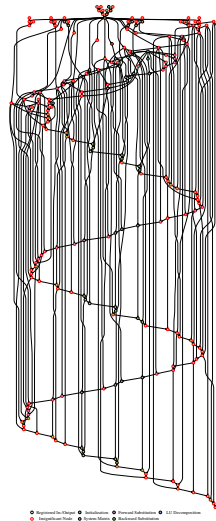
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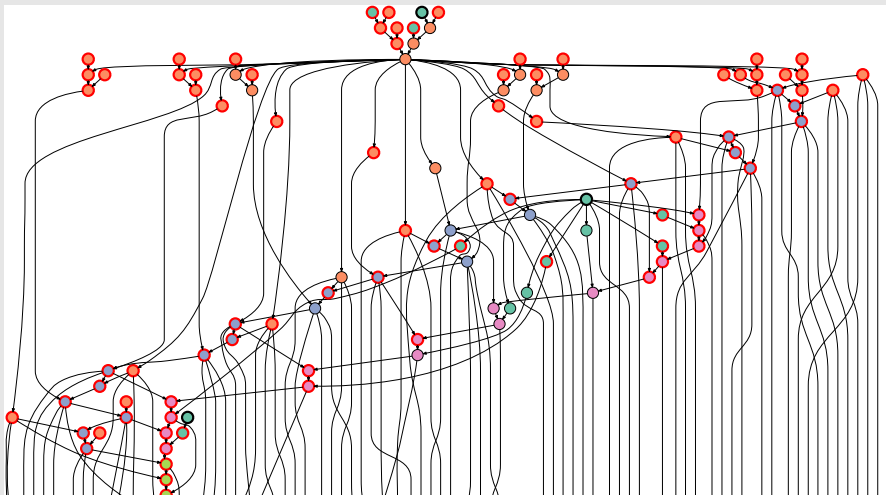
Visualization of the Significance Graph

- ▶ Number of time steps $m = 3$
- ▶ Number of space distributions $n = 9$
- ▶ Red framed nodes are insignificant
- ▶ Black framed nodes are Inputs or Outputs
- ▶ Different sections of the code have different colors



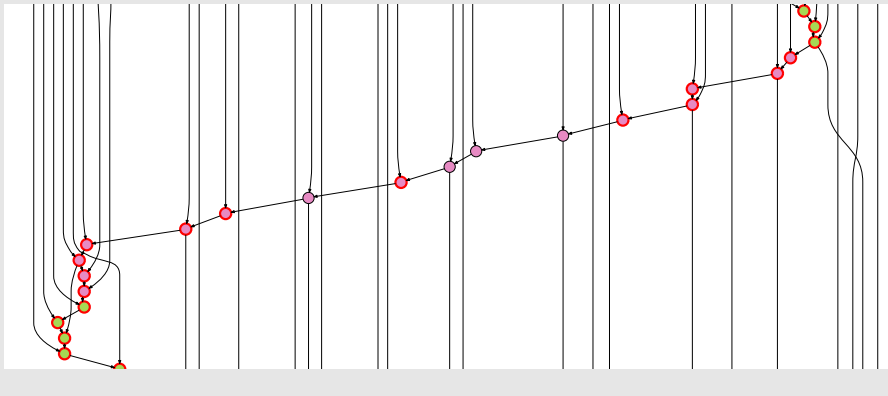
Case study: 1-D heat equation

Initialization, system matrix, first forward substitution



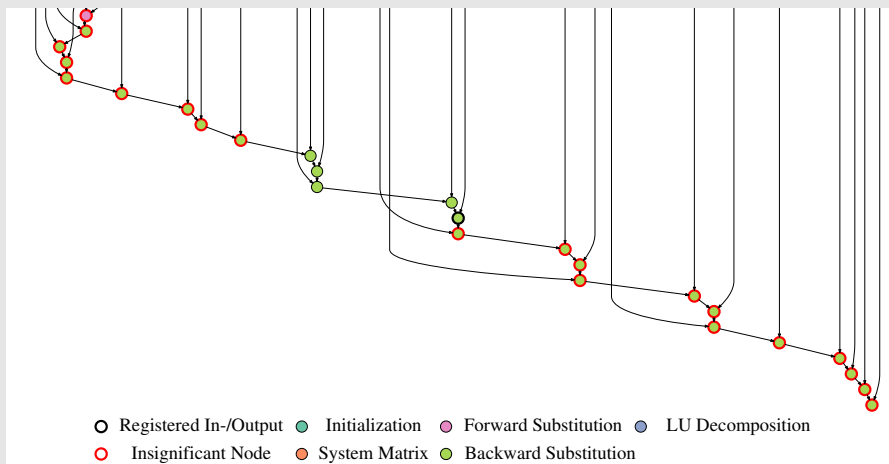
Case study: 1-D heat equation

Forward substitution



Case study: 1-D heat equation

Backward substitution



Summary of Results and Future Work

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- ▶ IASA is able to identify sparse systems
- ▶ Quantification mode was used to verify intuitive and well-known characteristics
- ▶ Results of the case study were used as a sanity check for IASA

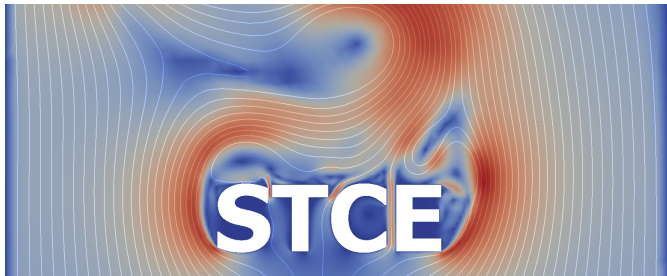
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Future Work

- ▶ Identify the quality of the optimized code for the corresponding input domain
- ▶ Implementation and testing of the exploration mode
- ▶ Calculate significances of larger applications by using IASA



Thank you very much!

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