Probabilistic Program Analysis

Matthew B. Dwyer, Antonio Filieri, Jaco Geldenhuys, Mitchell Gerrard, Corina Păsăreanu, and Willem Visser

> Department of Computer Science and Engineering University of Nebraska - Lincoln Lincoln, Nebraska USA

> > August 2015



Your view of "Program Analysis"

When you hear the term program analysis what comes to mind?



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When you hear the term program analysis what comes to mind?

What applications?

What kinds of information are computed?

What's the relationship between that information and program semantics?

How are the analysis results computed?



Program analysis to me

I worked at a compiler company in the 1980s; responsible for the common optimizer in a family of C compilers for embedded systems

My Ph.D. work applied data flow analysis to check safety properties of concurrent programs

Co-directed a large project on software model checking (e.g., Bandera, Bogor)

Developed symbolic execution for program equivalence checking

Optimization of runtime monitoring for complex properties (e.g., state machines)

Verification of high-performance computing codes (e.g., MPI, OpenMP, CUDA, pthreads)





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 enrich your favorite language with the ability to draw from a probability distribution, e.g., normal (0,1)



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 further enrich language with the ability to condition the output, e.g., observe(x>0)



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The modern view is being driven by applications of machine learning

 Infering an input distribution from (enough) observed outputs amounts to a backward data flow analysis (Claret et al, FSE'13)



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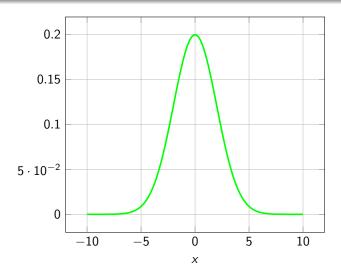
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Imagine an input distributed according to N(0,2)



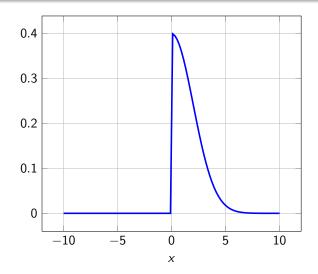


A trivial program

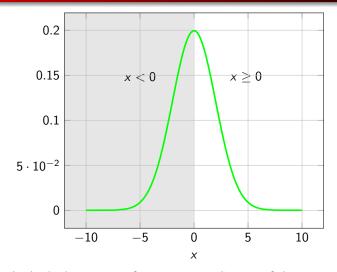
```
double abs(double x) {
  if (x<0)
    return -x;
  else
    return x;
}</pre>
```



Here is the output distribution

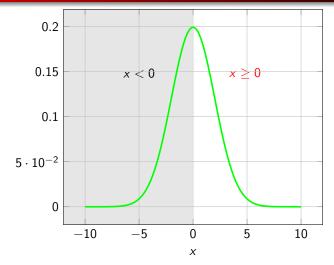






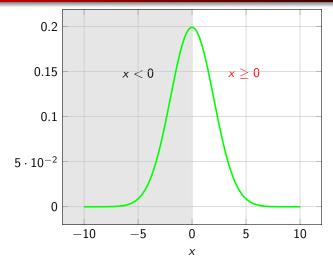
Let's think in terms of a very coarse division of the input





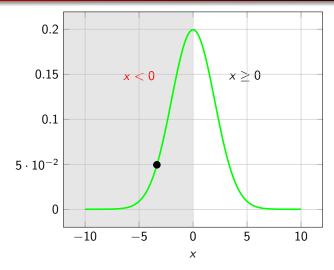
Input values $x \ge 0$ appear on the output unchanged.





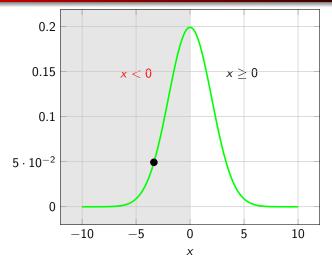
Input values $x \ge 0$ appear on the output unchanged. Their mass in the input distribution propagates to the output.



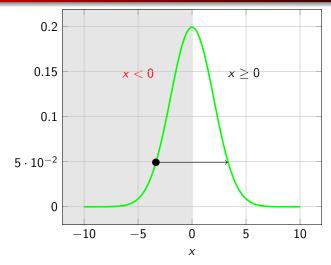


Input values x < 0 are transformed.



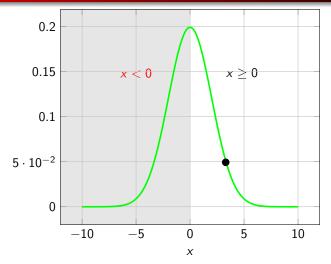


Input values x<0 are transformed. Their mass in the input distribution is shifted to -x



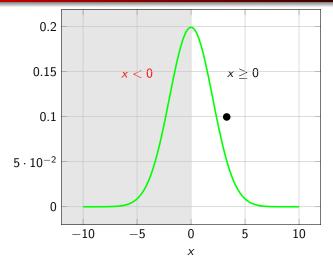
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Input values x<0 are transformed. Their mass in the input distribution is shifted to -x and accumulates in the output distribution

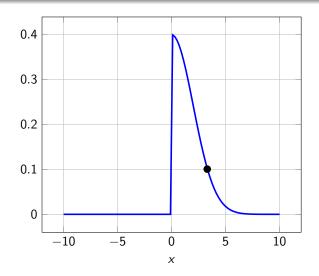




Input values x<0 are transformed. Their mass in the input distribution is shifted to -x and accumulates in the output distribution



Here is the output distribution





This briefing is about

Probabilistic (Program Analysis)

as opposed to

(Probabilistic Program) Analysis







Modeling of common duckweed and aphid system





Modeling of common duckweed and aphid system

Assuring the safety of UAV operations in close proximity agriculture







Modeling of common duckweed and aphid system

Assuring the safety of UAV operations in close proximity agriculture



These systems involve inherent uncertainty

• in the input values, sensor error, state transitions, ...

Unlike in classic program analyses, this uncertainty can be characterized and should be exploited



What kinds of questions can you answer?

How reliable is the program under an input distribution?

How frequently is this block executed?

Are these programs equivalent under an input distribution?

How sensitive is a test oracle to input distribution?

How much of the input space is covered?

How important is this dependence? (quantified info. flow)

If you can't prove it, how close did you get?



What's next ...

- Provide an overview of the key concepts in data flow analysis and symbolic execution
- Describe how researchers have enriched those frameworks with probabilistic reasoning (of varying sorts)
- Break down the literature across 3 orthogonal dimensions
- Talk in more detail about probabilistic symbolic execution
- Give you a list of 5+ PhD topics in this area

