

PROPERTIES OF A PROGRAM'S RUNTIME STATE SPACE

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OUTLINE, PROGRESS

1. ● Programs
2. Models in First Order Logic (. . . and programs. . .)
3. But . . . Aren't all nontrivial questions about real programs **undecidable?**
4. Subrecursive programming languages (life without CONS)
5. One-sided program analyses (approximating reachable states)
6. Termination analysis (approximating state transitions)
7. Conclusions

This talk is **not encyclopedic**. Main focus: areas close to my own research.

PROGRAMS

A program is a **syntactic object**. We typically call it p , q , ...

Its **purpose** is to realise some **computational intent**.

Semantics $\llbracket p \rrbracket$ of program p : the computation(s) specified by p .

What are computations? Many variants...

HOW computations can take place:

- ▶ deterministic versus nondeterministic;
- ▶ finite versus continuing/infinite;
- ▶ local versus global;
- ▶ concurrent; synchronous or asynchronous;
- ▶ quantum, ...

WHAT the purpose of running p may be: very many possibilities.

For this talk, program p 's purpose is to compute an **an input-output function** (possibly **partial**: p may not terminate on all inputs)

PROGRAM ANALYSIS OVERVIEW

What can we say about a program's **behavior** based on its **syntax** alone?

Practical needs: dangerous applications, reliability, performance, ...

▶ The **practice** of compilers:

- Compiler construction, including **code optimisation phases**
- Compiler correctness: what is it, how to define?

Relevant theory:

- ▶ **Theory of computability:** Turing machines, Turing completeness, the **Halting Problem**, Rice's theorem, the Rogers Isomorphism theorem, ...
- ▶ Programming language **semantics**
- ▶ **Abstract interpretation**, or program flow analysis
- ▶ Computational Complexity theory
 - **Effect of programming style** (functional, imperative, logic programming, process algebra ...) on what can be expressed by programs
(a **Whorfian hypothesis?**)

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MODELS IN FIRST ORDER LOGIC

- ▶ Near the beginning: Tarski 1936

The concept of truth in formalized languages

- ▶ Heinrich Scholz 1952:

asked for a characterization of **spectra**, i.e., sets of natural numbers that are **the cardinalities of finite models of first order sentences**.

- ▶ Scholz **probably expected** something like:

Spectra are the smallest class of sets of natural numbers that contain ... and are closed under the operations ...

- ▶ Surprise!

The first solution was a characterisation by **complexity classes** (Jones & Selman 1972).

- ▶ **Finite model theory** was then developed **much** further, e.g., by Fagin, Gurevich, Immerman, ...)

ANALOGY: FIRST ORDER LOGIC AND PROGRAMS

- ▶ A first order formula ϕ is a bit like a program p
- ▶ A **model** \mathcal{M} of FOL formula ϕ is analogous to the **runtime state space** of a program p .
- ▶ Viewed as a “program”, ϕ has only boolean logic (\wedge, \vee, \neg) and simple iteration (\forall, \exists) with no accumulator
- ▶ Dimension of interest: the **cardinality** of model \mathcal{M} .

The **Spectrum** of a FOL formula ϕ is

$$\text{SPECTRUM}(\phi) = \{n \in \mathbb{N} \mid \phi \text{ has a model } \mathcal{M} \text{ of cardinality } n\}$$

Example:

- ▶ If $\phi =$ the axioms for Boolean Algebra
- ▶ then

$$\text{SPECTRUM}(\phi) = \{2^m \mid m \geq 0\}$$

by Stone’s representation theorem

An answer to Scholz' problem:

$I \subseteq \mathbb{N}$ IS A SPECTRUM

IF AND ONLY IF

$I \in \text{NEXPTIME}$

(Viewpoint: regard I as a set of numbers written as bit strings)

Proof both ways by programming:

▶ \Rightarrow : given ϕ , find a **nondeterministic program** p to answer the question

“is $n \in \text{SPECTRUM}(\phi)$?”

(also: show that p runs in exponential time)

▶ \Leftarrow : show that

- **for any nondeterministic program** p that runs in exponential time

- there exists a **first order formula** ϕ that **simulates** the running of p

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BUT ...AREN'T ALL NONTRIVIAL QUESTIONS ABOUT REAL PROGRAMS UNDECIDABLE?

The 1930s golden age: Turing/Church/Post/Kleene/...:

- ▶ They devised many formulations of **the class of all computable functions**. Superficially quite different.
- ▶ **All turned out to be equivalent** (any one could simulate any other)
- ▶ Even stronger: Hartley Rogers showed that **all programming languages are isomorphic**.

Alan Turing: **halting problem is undecidable** (by mechanical computation).

WHAT: Given, a program p , and an input d to run it on

TO DECIDE: Will p eventually terminate its computation on d ?

And yet more:

Rice's Theorem: **any nontrivial question** about program behavior is undecidable.

(shown by reduction from the halting problem)

HOW TO AVOID THIS DILEMMA ?

(or, is computer science hopeless?)

Programs are ubiquitous – we can't live without them! **What to do?**

One way: **Sacrifice Turing completeness** and use **subrecursive languages**, e.g.,

- ▶ finite models of first-order or temporal logic or
- ▶ ICC (implicit computational complexity, e.g., Dal Lago and Martini here)
- ▶ Strongly normalising programming languages, e.g., constructive type theory, System F, . . .

Another way: Do “**one-sided**” analyses of a Turing-complete language

- ▶ **Older**: practical compiler work since the 1950s. **Program flow analysis** is widely used in optimising compilers
(Fortran, Algol, C, Java, Haskell, . . .)
- ▶ More semantically based program analyses: **abstract interpretation** (Cousot, Nielson, Hankin, Jones, Muchnick, . . .)

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SUBRECURSIVE PROGRAMMING LANGUAGES

Approach:

study function classes that are smaller than the class of **all** computable partial functions.

Early works in this direction (1930s – 1958):

- ▶ **Primitive recursive functions** (Kurt Gödel, Rózsa Péter)
- ▶ **Gödel's System T** (much larger, still total and computable)
- ▶ **The Grzegorzcyk hierarchy** inside the primitive recursive functions

$$\varepsilon^0 \subsetneq \varepsilon^1 \subsetneq \varepsilon^2 \subsetneq \varepsilon^3 \subsetneq \dots \subseteq \textit{PrimRec}$$

Alas, even the small class $\varepsilon^3 = \textit{Elementary}$ is too large to be of practical relevance to computer scientists! (And the bigger classes are **much** bigger...)

PROBLEM CLASSES SMALL ENOUGH TO BE OF PRACTICAL RELEVANCE TO COMPUTER SCIENTISTS

- ▶ P , or PTIME: decision problems solvable by programs in time bounded by a polynomial function (of the length of the input)
- ▶ NP , or NPTIME: ditto, but a program/algorithm may be **nondeterministic**, i.e., it may “guess”
- ▶ LOGSPACE: a class smaller than P
- ▶ EXPTIME: a larger class, getting near to infeasibility
- ▶ NEXPTIME: still larger, a nondeterministic version of EXPTIME

Complexity classes:

$$\text{LOGSPACE} \subseteq \text{PTIME} \subseteq \text{NPTIME} \subseteq \text{PSPACE} \subseteq \text{EXPTIME} \subseteq \dots$$

(all are proper subsets of Grzegorzczuk’s “elementary” class ϵ^3)

A SMALL SUBRECURSIVE LANGUAGE: “READ-ONLY” PROGRAMS

Data Structures: Booleans, lists of booleans (and functions, for higher-order programs). **0-order = booleans or lists of Booleans.**

Read-only programs: “Life without CONS”

- ▶ no constructors or memory allocation or $x + 1$ allowed;
- ▶ only $x - 1$ or $hd(x), \dots$

Data types: 0-order, 1-order, 2-order, ... [finite orders only]

Control Structures: some choices

- ▶ **Primitive** recursion (FOR-loops only, called “folds” in functional programming), eg $f(x) = \text{for } y := 1 \text{ to } x \text{ do } \{x := x+y\}; \text{ return } x$
- ▶ **Tail** recursion (WHILE-loops; they seem a bit more powerful),
- ▶ **General** recursion: nested function calls, eg $f(x,y) = g(x,h(y))$
- ▶ Calls to **higher-order functions**, eg $\text{double}(f,x) = f(f(x))$

TYPICAL RESULTS

- ▶ Gurevich: **0-order primitive** recursive read-only programs can decide **all and only the problems in LOGSPACE**.
- ▶ Jones: **0-order general** recursive read-only programs can decide **all and only the problems in PTIME**.
- ▶ Gurevich: **1-order primitive** recursive read-only programs can decide **all and only the problems in PTIME**.

Conclusion from the first two: (for 0-order data)

general recursion is stronger than primitive recursion
if and only if
PTIME properly includes LOGSPACE

Alas, **this doesn't answer** the expressibility question

“is general recursion stronger than primitive recursion ?”.

BUT: it shows it equivalent to another very hard question!

A question in family with: is $P = NP$?

DECISION POWER, BY SEVERAL PROGRAM CONTROLS AND SEVERAL DATA ORDERS

| <u>Programs</u> | <u>Data Order 0</u> | <u>Data Order 1</u> | <u>Data Order 2</u> | ... | <u>LIMIT</u> |
|------------------------------------|-------------------------|-------------------------|-------------------------|-----|---------------------------|
| Read-write | REC. ENUM | REC. ENUM | REC. ENUM | ... | REC. ENUMERABLE |
| Primitive recursive (foldr) | PRIM.REC. | PRIM ¹ REC. | PRIM ² REC. | ... | System T |
| General rec. (RO) | PTIME | EXPTIME | EXP ² TIME | ... | ELEMENTARY = ϵ^3 |
| Tail recursive RO | LOGSPACE | PSPACE | EXPSPACE | ... | ELEMENTARY = ϵ^3 |
| Primitive rec. RO | LOGSPACE | PTIME | PSPACE | ... | ELEMENTARY = ϵ^3 |

Top half notation:

- ▶ RECURSIVELY ENUMERABLE = all problems solvable by a Turing machine (that only halts on “yes” answers)
- ▶ PRIMITIVE RECURSIVE = usual Gödel-style, including successor $x + 1$
- ▶ **System T** = PRIMITIVE RECURSIVE of any finite order. **Huge!**

DECISION POWER, BY SEVERAL PROGRAM KINDS AND SEVERAL DATA ORDERS

| <u>Programs</u> | <u>Data Order 0</u> | <u>Data Order 1</u> | <u>Data Order 2</u> | ... | <u>LIMIT</u> |
|---|-------------------------|-------------------------------------|-------------------------------------|-----|------------------------------------|
| Read-write Primitive recursive (foldr) | REC. ENUM PRIM.REC. | REC. ENUM PRIM ¹ REC. | REC. ENUM PRIM ² REC. | ... | REC. ENUMERABLE System T |
| General rec. (RO) | PTIME | EXPTIME | EXP ² TIME | ... | ELEMENTARY = ϵ^3 |
| Tail recursive RO | LOGSPACE | PSPACE | EXSPACE | ... | ELEMENTARY = ϵ^3 |
| Primitive rec. RO | LOGSPACE | PTIME | PSPACE | ... | ELEMENTARY = ϵ^3 |

Bottom half notation:

Read-only (RO) programs that have no successor $x + 1$.

▶ **Complexity classes:**

$$\text{LOGSPACE} \subseteq \text{PTIME} \subseteq \text{PSPACE} \subseteq \text{EXPTIME} \subseteq \dots$$

▶ **Grzegorzczuk's class** $\text{ELEMENTARY} = \epsilon^3 = \bigcup_{k=0}^{\infty} \text{EXP}^k\text{TIME}$. Pretty big!

READ-ONLY PROGRAMS, DIFFERENT DATA ORDERS

| <u>Programs</u> | <u>Data Order 0</u> | <u>Data Order 1</u> | <u>Data Order 2</u> | ... | <u>LIMIT</u> |
|--------------------------|---------------------|---------------------|-----------------------|-----|---------------------------|
| General rec. (RO) | PTIME | EXPTIME | EXP ² TIME | ... | ELEMENTARY = ϵ^3 |
| Tail recursive RO | LOGSPACE | PSPACE | EXPSPACE | ... | ELEMENTARY = ϵ^3 |
| Primitive rec. RO | LOGSPACE | PTIME | PSPACE | ... | ELEMENTARY = ϵ^3 |

(proven by: Gödel, Gurevich, Goerdt, Seidl, Jones.)

- ▶ For read-only programs, data order 0: Is general recursion more powerful than tail recursion or primitive rec. ? **Equivalent by table:**

Is PTIME \supsetneq LOGSPACE ?

A long-standing open problem!

- ▶ For read-only programs, data order 1: general recursion **IS** more powerful than primitive recursion **since**

EXPTIME \supsetneq PTIME

A problem with deriving programs from proofs?

Reasoning: **induction proofs lead to primitive recursion.**

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ONE-SIDED PROGRAM ANALYSES (FLOW ANALYSIS, ABSTRACT INTERPRETATION)

Semantic and **logical** aspects of a computational practice.

- ▶ Practice: **decades of practical experience** in writing compilers
(though correctness is rarely addressed by compiler hackers!)
- ▶ **Data flow analysis**: informal, pragmatic, ad hoc methods from the 1950s.
- ▶ Engineering methodology: **program analysis by fix-point computations**.

Theory: **Semantics-based program analysis**

- ▶ Formally based in program semantics.
Cousot, Hankin, Jones, Muchnick, Nielson, many others.
- ▶ Research since 1970's under the name of **Abstract Interpretation**
- ▶ January 2008 conference in San Francisco:
“30 Years of Abstract Interpretation.”

TOWARDS UNDERSTANDING THE PROBLEM

Consider the **code elimination** transformation

$$[x := a]^\ell \Rightarrow [\text{skip}]^\ell$$

(It sounds trivial, but it's significant in practice!)

Semantic reasons that make it valid: (control flow and data flow)

1. ℓ is **unreachable**: \exists no control flow **from the program's start** to $[x := a]^\ell$
2. ℓ is **dead**: \exists no control flow from $[x := a]^\ell$ **to the program's end**. E.g.,
 - ▶ The program will **definitely loop** after point ℓ . Or
 - ▶ The program will **definitely abort** execution after point ℓ .
3. x is **dead** at ℓ : the value of x is never used again.
4. x is **already equal to** a (if control ever gets to ℓ)
5. a is an **uninitialised variable**: the value of x is completely undependable

ALAS, MOST OF THESE REASONS ARE AS UNDECIDABLE AS THE HALTING PROBLEM (!)

Remark: many (all!) of the above program behavior properties are **undecidable** (if you insist on exact answers).

Proof Rice's Theorem from Computability Theory.

So what do we do?

An answer: allow “one-sided” errors (in practice of **program analysis** and theory of **abstract interpretation**).

- ▶ Find **safe** descriptions of program behavior. Meaning of safety:
 - if the analysis says that a program has a certain behavior (e.g., that x is dead at point ℓ),
 - then it **definitely has that behavior** in all computations.
- ▶ Allow the analysis to be imprecise, i.e., “one-sided” :
the analysis can answer “don't know” even when the property is true
(this is the trick to gain decidability; it can be misused)

“ONE-SIDED” REASONING TO DISCOVER PROGRAM PROPERTIES

“Program-point-centric” analysis: approximate the **control flow** or **data flow** at each program point ℓ .

The flow properties at a program point ℓ are determined by

▶ the **time dimension**:

- the computational **past** (of computations that get as far as ℓ); or
- the computational **future** (of computations after ℓ)

▶ the **path modality**:

- a property of **all computation paths** from (or to) ℓ , or by
- a property of **at least one computation path** from (or to) ℓ

Does this look familiar? The practitioners’ methods in effect achieve

▶ **applied temporal logic**

▶ **on finite models**

WHAT AND HOW

What: program transformation to improve efficiency

▶ Based on **program flow analysis**

▶ Must be correct.

Semantic question: what does this mean?

Study object: the space of all run-time states

▶ **Important:** efficiency, complex hardware, human limits, etc

▶ **Semantically subtle**

How: several steps in program optimisation. First: **program analysis.**

▶ Choose a **data flow lattice** to describe program properties:

one-sided finite descriptions of run-time state sets

▶ Build a system of **data flow equations** from the program:

time dimension = future/past, modality = may/must.

▶ **Solve** the system of data flow equations

Then **transform** the program, usually to optimise it

STATE-BASED PROGRAM ANALYSIS

An example: for every program point ℓ , over-approximate

$$\text{Reach}(\ell) = \{\sigma \mid \text{some initial state } (\ell_0, \sigma_0) \text{ can reach } (\ell, \sigma)\}$$

Program analysis

- ▶ must be finitely (and feasibly!) computable
- ▶ is computed **uniformly** for **all points** ℓ in the given program.
- ▶ is a mass act: applied **automatically** to **any input program**
(in contrast to: one-program-at-a-time verification)
- ▶ **Adjacent program points** will have properties that are related, e.g., by classic **flow equations** of dataflow analysis for compiler construction.

An analogy: heat flow equations as in Physics.

(ALTHOUGH... heat flows 2-ways, but program flows are asymmetric.)

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TERMINATION ANALYSIS

Aim: prove that program p will **terminate on all possible runs**.

This is

- ▶ **not** a property of the **states** (ℓ, σ) reachable at program point ℓ , but
- ▶ rather a property of the **state transitions** from a program point ℓ to its successors

Definition p terminates iff there does not exist an infinite transition sequence

$$(\ell_0, \sigma_0) \rightarrow (\ell_1, \sigma_1) \rightarrow \dots \rightarrow (\ell_i, \sigma_i) \rightarrow \dots$$

Flow analysis approach: over-approximate, for each program point ℓ, ℓ' , the set

$$(\ell, \sigma) \rightarrow (\ell', \sigma')$$

What is new here?

Focus not on “one-state-at-a-time” (as for flow analysis). Focus is now on **relations between states**, between one state and its successor.

SIZE-CHANGE TERMINATION ANALYSIS

Size-Change Termination criterion:

- ▶ **Methods, viewpoint:** from automata theory and graph theory
- ▶ Size-change Termination can serve as a **simple common core** of many termination analyses
- ▶ **Challenge:** reduce need for human creativity, e.g., designing and finding lexicographic orders, polynomial interpretations, etc.
- ▶ Early motivation from partial evaluation: a **binding-time analysis** sufficient to guarantee that program specialization will stop

A one-sided analysis:

- ▶ **Any size-change terminating program is terminating**
- ▶ **Some terminating programs may not be size-change terminating**

... so we're not solving the halting problem!

OVERVIEW OF METHODS, RESULTS

1. Assume:

▶ Set *Value* has a well-founded order $>$

▶ **Size changes** known from program operations, e.g.,

$$\text{cons}(X, Y) > X, \quad \text{cons}(X, Y) > Y, \quad X > \text{head}(X)$$

2. For each transition ℓ in program p , obtain a “safe” size-change graph G_ℓ

3. Program p is **size-change terminating** if ...

▶ (a property of the graphs G_ℓ that implies)

▶ \exists **no infinite transition sequence** in any computation.

4. Algorithms to test the graph property:

(a) Compute **closure of a set of graphs** or, an alternative:

(b) Operations on **Büchi automata** (test \subseteq);

5. Upper and lower complexity bounds: both PSPACE.

APPROACH TO TERMINATION ANALYSIS

1. Consider all traces (finite or infinite transition sequences) that might occur in actual computations
2. Identify as **Bad** traces: all those of infinite length
3. Identify as **Dead** traces: all those that are impossible because they would cause infinite descent

The **Size-change principle**: program p terminates if

Every infinite transition sequence would cause an infinite descent in at least one value

In other words: every “bad” trace is a “dead” trace.

Decide using regular (finite-state) approximations to **Bad**, **Dead**.

SIZE-CHANGE GRAPHS

Consider a program transition

$$\begin{aligned} f(x_1, \dots, x_i, \dots, x_n) &= \dots \boxed{\ell} : g(e_1, \dots, e_j, \dots, e_m) \dots \\ g(y_1, \dots, y_j, \dots, y_m) &= \dots \end{aligned}$$

Size-change graph G_ℓ for the transition $\ell : f \rightarrow g$ has a labeled arc

▶ **from** f parameter x_i **to** g parameter y_j
if x_i is used to compute e_j .

▶ Arc label: **how is x_i value related to e_j value?**

G_ℓ must **safely approximate** size relations:

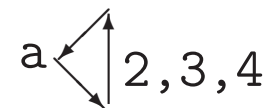
▶ $x_i \xrightarrow{\downarrow} y_j$ implies x_i 's value is greater than e_j 's new value

▶ $x_i \xrightarrow{\Downarrow} y_j$: same, but “greater than or equal”

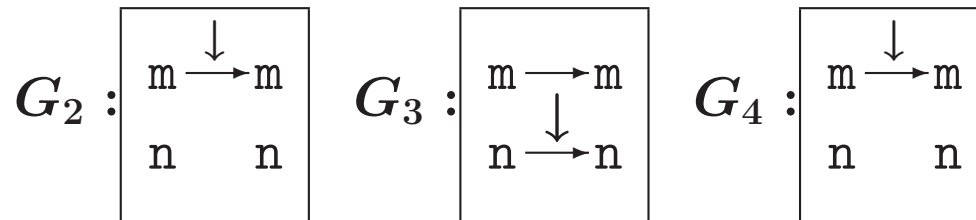
ACKERMANN: A NONLINEAR PROGRAM

$a(m,n) =$ if $m=0$ then $n+1$ else
if $n=0$ then $\boxed{2}:a(m-1,1)$ else
 $\boxed{4}:a(m-1, \boxed{3}:a(m,n-1))$

Transition graph

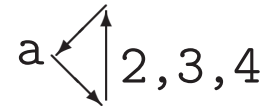


Size-change graphs

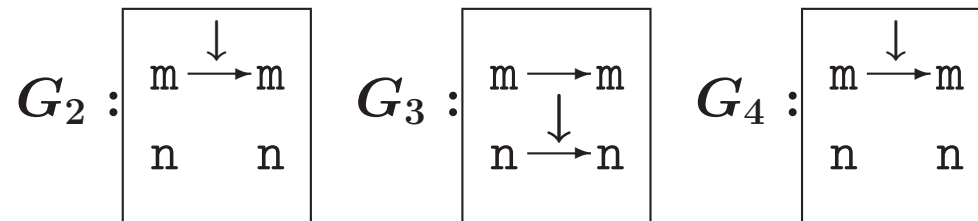


SIZE-CHANGE TERMINATION REASONING:

Transition graph



Size-change graphs



Consider any infinite transition sequence:

$$\pi \in (2 + 3 + 4)^\omega$$

- ▶ If $\pi = \dots 3^\omega$, then **n descends infinitely**.
- ▶ Otherwise π has infinitely many 2's or 4's; so **m descends infinitely**.

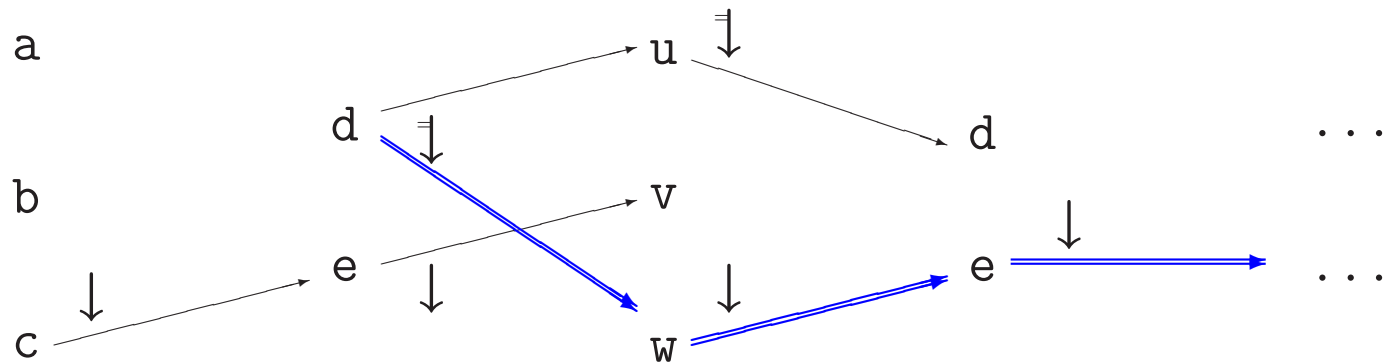
There is no obvious bound on length of a descending chain.

(No big surprise, since Ackermann's function isn't primitive recursive!)

THE MULTIPATH OF A TRANSITION SEQUENCE

Definition: The **multipath** $\mathcal{M}(\pi)$ of transition sequence $\pi = \tau_1\tau_2\dots$ is:

$$\mathcal{M}(\tau_1\tau_2\dots) = \text{concatenation of } G_{\tau_1}, G_{\tau_2}, \dots$$



$$\mathcal{M}(\tau_1\tau_2\dots) = \boxed{G_{\tau_1} : f \rightarrow g} \boxed{G_{\tau_2} : g \rightarrow h} \boxed{G_{\tau_3} : h \rightarrow g} \boxed{G_{\tau_4} : g \rightarrow k} \dots$$

Definition:

1. A **thread** in multipath $\mathcal{M}(cs)$ is a connected sequence of labeled arcs.
2. The thread is of **infinite descent** if its arc label sequence contains infinitely many \downarrow .

THE SIZE-CHANGE TERMINATION CRITERION

Recall: Program p is size-change terminating if $\text{Bad} \subseteq \text{Dead}$, where

$$\text{Bad} = \{ \pi \in C^\omega \mid \pi \text{ follows } p\text{'s flow chart} \}$$

$$\text{Dead} = \{ \pi \in C^\omega \mid \mathcal{M}(\pi) \text{ has a thread with infinite } \downarrow \}$$

Algorithms:

1. Graph calculation used in practice:

Compute and test the **closure** of the set of size-change graphs.

Theorem p is size-change terminating iff every idempotent graph in the closure has an arc $x \xrightarrow{\downarrow} x$.

2. Asymptotically better, practicality as yet unclear:

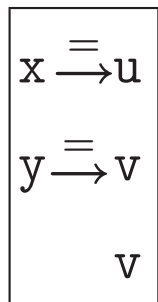
Büchi automaton algorithms (Sasha, Vardi, Fogarty).

TERMINATION BY GRAPH CALCULATION

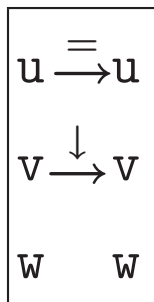
$f(x,y) = \text{if } x = 0 \text{ then } y \text{ else } \boxed{1}: g(x,y,0)$

$g(u,v,w) = \text{if } w = 0 \text{ then } \boxed{3}: f(u-1,w) \text{ else } \boxed{2}: g(u,v-1,w+2)$

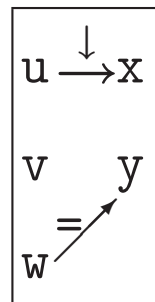
Size-change graphs \mathcal{G}



G_1

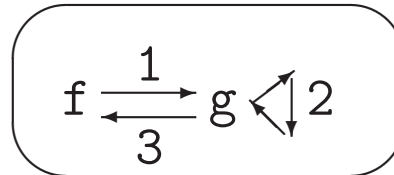


$*G_2$

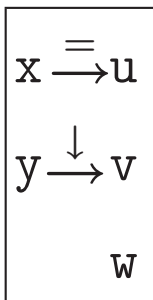


G_3

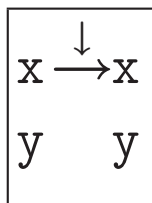
Transition graph



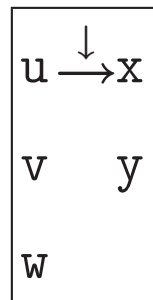
Closure: $\{G_1, G_2, G_3, G_{12}, G_{13}, G_{23}, G_{31}, G_{131}\}$ (**all reachable** data flows)



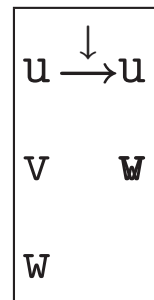
G_{12}



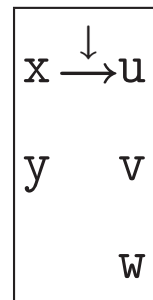
$*G_{13}$



G_{23}



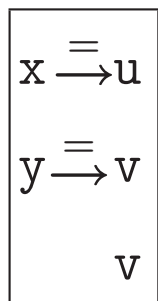
G_{31}



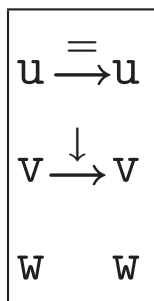
G_{131}

TERMINATION BY GRAPH CALCULATION II

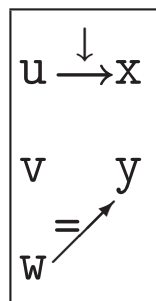
Size-change graphs \mathcal{G}



G_1

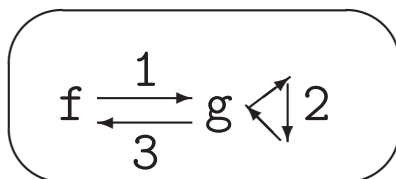


$*G_2$



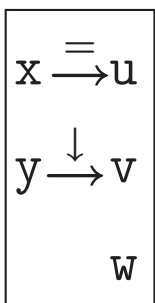
G_3

Transition graph

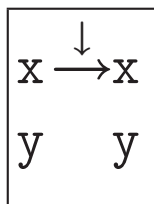


$G_2 = G_2; G_2$, so G_2 is idempotent.

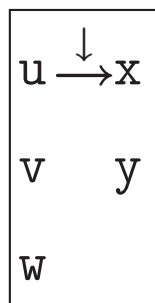
Closure: $\{G_1, G_2, G_3, G_{12}, G_{13}, G_{23}, G_{31}, G_{131}\}$ (all reachable data flows)



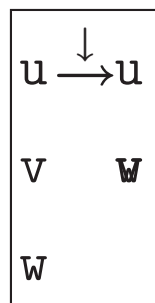
G_{12}



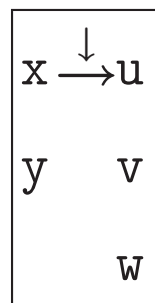
$*G_{13}$



G_{23}



G_{31}



G_{131}

G_{13} is idempotent

Idempotent graphs G_2, G_{13} have decreasing variables, so no infinite traces are possible.
Therefore program p terminates on all inputs.

MORE ABOUT TERMINATION ANALYSIS

- ▶ **Ramsey's theorem** is the key to prove correctness of the closure method.
- ▶ **Worst-case behavior:** Size-change termination is complete for PSPACE.

Related work (far from all...)

1. Early functional: Abel and Altenkirch
2. Damien Sereni's 2006 Ph.D. thesis: higher-order functions
3. Term rewriting: Giesl, Arts, many others
4. Logic programming: Codish, Lindenstrauss, Plümer, Sagiv, Taboch, ...
5. Chin Soon Lee's 2002 Ph.D. thesis:
 - ▶ Application to partial evaluation
 - ▶ Program analysis in PTIME (weaker but strong enough in practice)

OUTLINE, PROGRESS

1. Programs
2. Models in First Order Logic (. . . and programs. . .)
3. But . . . Aren't all nontrivial questions about real programs **undecidable?**
4. Subrecursive programming languages (life without CONS)
5. One-sided program analyses (flow analysis, abstract interpretation)
6. Termination analysis (approximating state transitions)
7. ● Conclusions

CONCLUSIONS

Overviewed in this talk:

- ▶ A **model in FOL** resembles a **program's runtime state space**
- ▶ Relate unsolved problems to each other, e.g.:

RECURSION $\not\supseteq$ ITERATION **if and only if** PTIME $\not\supseteq$ LOGSPACE

- ▶ Connections between practice (**compilers' flow analysis**) and theory (**abstract interpretation, model checking**)
- ▶ The **halting problem** can be dealt with, if not completely solved

More generally:

- ▶ Properties of a program's runtime state space are fascinating
- ▶ There are many connections, some unexpected, with theoretical and applied Computer Science, and Logic too (spectra, temporal logic, ...)
- ▶ Don't give up, even on undecidable problems!

5 RELEVANT PAPERS

References

- [1] Neil D. Jones. **The expressive power of higher-order types or, life without cons.** J. Funct. Program., 11(1):5–94, 2001.
- [2] Neil D. Jones and Alan L. Selman. **Turing machines and the spectra of first-order formulas with equality.** ACM Symp. Theory of Computing, pp.157–167, 1972. Journal of Symbolic Logic, 39:139–150, 1974.
- [3] Neil D. Jones and Flemming Nielson. **Abstract interpretation: a semantics-based tool for program analysis.** In Handbook of Logic in Computer Science. Oxford University Press, 1994. 527–629.
- [4] David Lacey, Neil D. Jones, Eric Van Wyk, and Carl Christian Frederiksen. **Compiler optimization correctness by temporal logic.** Higher-Order and Symbolic Computation, 17(3):173–206, 2004.
- [5] Chin Soon Lee, Neil D. Jones, Amir M. Ben-Amram. **The size-change principle for program termination.** ACM POPL, pages 81–92, 2001.