Complexity-theoretic barriers to validated solution of initial value problems

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May 23, 2008

The problem

The initial value problem:

given
$$g:[0,1]\times\mathbf{R}\to\mathbf{R}$$
, find $h:[0,1]\to\mathbf{R}$ such that $h(0)=0, \qquad h'(t)=g\bigl(t,h(t)\bigr).$

Question: How computationally complex is *h*?

To discuss computational complexity of real functions, we use the framework of Computable Analysis:

- computability (Grzegorczyk 1955);
- polynomial-time computability (Ko–Friedman 1981).
 (henceforth PTIME)

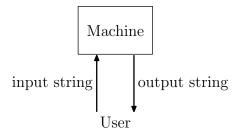
- Complexity of real functions: How do we define (PTIME) computability of real functions?
- 2. Complexity of IVPs: How complex can *h* be when *g* is PTIME computable?
 - ► Lipschitz IVP is PSPACE complete
 - ► Analytic IVP is PTIME computable
- 3. Discussion

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The "digital" computer

Computers (= Turing machines) manipulate bits.

Objects (integers, graphs, ...) must be encoded by bit strings.



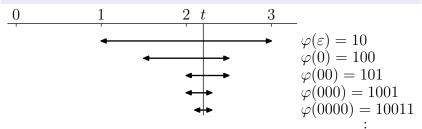
But real numbers cannot be named by bit strings, because there are uncountably many.

Representing real numbers

Represent real numbers by string-to-string functions.

Definition

Function $\varphi : \{0,1\}^* \to \{0,1\}^*$ represents $t \in \mathbf{R}$ if for each $n \in \mathbf{N}$, the value $\varphi(0^n)$ is (the binary notation of) either $\lfloor t \cdot 2^n \rfloor$ or $\lceil t \cdot 2^n \rceil$. We also say φ is a name of t.



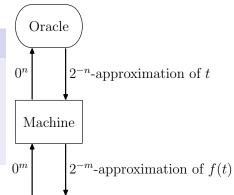
Thus, each $\varphi(0^n)$ gives a 2^{-n} -approximation of t.

Computing real functions

Use Oracle Turing Machines.

Definition (Grzegorczyk 1955)

A machine computes function $f:[0,1] \rightarrow \mathbf{R}$ if, given any name of any $t \in [0,1]$ as oracle, it computes some name of f(t).



In other words:

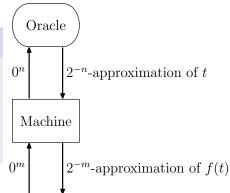
- ► The machine Turing-reduces f(t) to t.
- ► The machine produces approximations of f(t) of any precision the user desires, assuming that the user supplies the machine with approximations of t of any precision it desires.

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PTIME means that the machine halts after polynomially many steps in m.

▶ A polynomial p and a PTIME algorithm \hat{h} such that

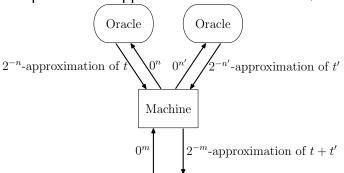
$$u \in \{ \lfloor t \cdot 2^{p(m)} \rfloor, \lceil t \cdot 2^{p(m)} \rceil \}$$

$$\implies \hat{h}(0^m, u) \in \{ \lfloor h(t) \cdot 2^m \rfloor, \lceil h(t) \cdot 2^m \rceil \}.$$

Example 1

 $Addition + : [0,1] \times [0,1] \rightarrow \textbf{R} \text{ is PTIME computable.}$

Given (functions representing) t, t' as oracles, compute a 2^{-m} -approximation of the sum t + t'.



Get 2^{-m-2} -approximations of t and t', compute their sum (a rational number), and output the closest 2^{-m} -approximation.

Example 2

The function exp : $[0,1] \rightarrow \mathbf{R}$ is PTIME computable.

To 2^{-m} -approximate

$$\exp t = \frac{1}{0!} + \frac{t}{1!} + \frac{t^2}{2!} + \frac{t^3}{3!} + \cdots,$$

it suffices to $2^{-m}/2$ -approximate the sum of the first m + 2 terms

(the remaining terms add up to $< 2^{-m}/2$).

Example 3

Comparison \leq : $[0,1]^2 \rightarrow R$ is not computable:

$$\leq (t, t') = \begin{cases} 1 & \text{if } t \leq t' \\ 0 & \text{if } t > t' \end{cases}$$

Computable functions are continuous

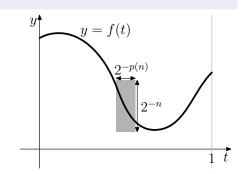
Theorem

- Every computable function is continuous.
- Every PTIME computable function has polynomial modulus of continuity.

Modulus of continuity *p*:

$$|t - t'| < 2^{-\rho(n)}$$

 $\implies |f(t) - f(t')| < 2^{-n}$



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Summary

Let $g : [0,1] \times [-1,1] \to \mathbf{R}$ and $h : [0,1] \to \mathbf{R}$ satisfy $h(0) = 0, \qquad h'(t) = g(t,h(t)).$

and suppose that g is PTIME computable. Then

- ▶ there may be infinitely many solutions h, none of which is computable (Pour-El 1979, Ko 1983);
- if we simply assume that h is the unique solution, then it is computable but can take arbitrarily long time (Miller 1970, Ko 1983);
- if we further assume that g is Lipschitz continuous, then the (unique) solution h is PSPACE computable and can be PSPACE complete;
- ▶ if we further assume that g is analytic, then the (unique) solution h is PTIME computable.

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red: new results

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PSPACE completeness

Theorem

There is a PTIME computable, Lipschitz continuous g such that the solution h is PSPACE complete.

I.e., h is the "hardest" among PSPACE solvable problems: for any PSPACE predicate L, there are a polynomial p and PTIME computable functions r, s such that if $s(u) \in \{\lfloor t \cdot 2^{p(n)} \rfloor, \lceil t \cdot 2^{p(n)} \rceil\}$ and $v \in \{\lfloor h(t) \cdot 2^n \rfloor, \lceil h(t) \cdot 2^n \rceil\}$, then r(v) = L(u).



Building blocks

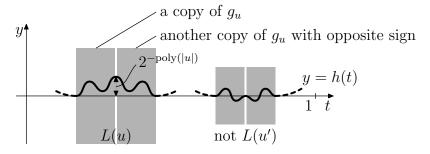
It suffices to construct $g_u : [0,1] \times [-1,1] \to \mathbf{R}$ and $h_u : [0,1] \to \mathbf{R}$ for each string u (with g_u uniformly Lipschitz and PTIME computable from u) such that

$$h_u(0) = 0, \qquad h'_u(t) = g_u(t, h_u(t))$$

and

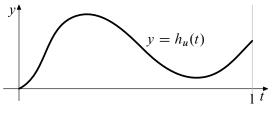
$$h_u(1) = egin{cases} 2^{-\mathsf{poly}(|u|)} & \text{if } L(u), \\ 0 & \text{otherwise}. \end{cases}$$

Once this is done, g_u 's can be put into one function g:



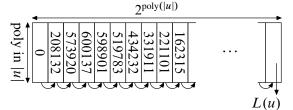
An attempt: discrete IVP

We want h_u such that:



- its slope is PTIME computed using the current value;
- $h_u(1)$ encodes the answer L(u).

Every PSPACE computation has the following description:

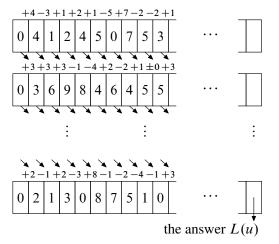


- each cell is PTIME computed from the previous cell;
- ▶ the last cell has the answer L(u).

An attempt: use values of h_u to encode this table. But this does not give Lipschitz continuous g.

Discrete Lipschitz IVP

A discrete problem simulable by Lipschitz IVP:



Each increment is PTIME computed from (*u* and) the upper left number.

By a more elaborate reduction, any PSPACE predicate can be simulated by this table also.

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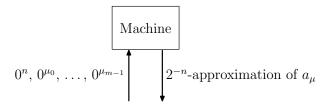
PTIME computable Taylor coefficients

Lemma

Let f be a real analytic function on a compact subset of \mathbf{R}^m around the origin:

$$f(x) = \sum_{\mu \in \mathbf{N}^m} a_{\mu} x_0^{\mu_0} \cdots x_{m-1}^{\mu_{m-1}}.$$

Then f is PTIME computable if and only if the sequence $(a_{\mu})_{\mu \in \mathbf{N}^m}$ is.



Solution by Taylor series

Substitute

$$g(t,y) = \sum_{i \in \mathbf{N}} \sum_{j \in \mathbf{N}} a_{ij} t^i y^j, \qquad h(t) = \sum_{i \in \mathbf{N}} b_i t^i$$

into h'(t) = g(t, h(t)) and compare coefficients.

Theorem

If $(a_{ij})_{(i,j)\in\mathbb{N}^2}$ is PTIME computable, so is $(b_i)_{i\in\mathbb{N}}$.

Corollary

If an analytic function g is PTIME computable, so is h.

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Algorithms and complexity

- The Church—Turing Thesis:
 - effective
 - = computed by a Turing machine
 - feasible
 - = computed by a Turing machine in polynomial time
- ► In discrete problems, hardness results sometimes give useful information for algorithm design.
- What about numerical problems?

Results (again)

Let $g:[0,1]\times[-1,1]\to\mathbf{R}$ and $h:[0,1]\to\mathbf{R}$ satisfy $h(0)=0, \qquad h'(t)=g\bigl(t,h(t)\bigr).$

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What do the positive results imply?

- ▶ unique solution ⇒ computable
- ▶ Lipschitz ⇒ PSPACE computable
- ▶ analytic ⇒ PTIME computable

What do these mean for numerical solution of IVPs? —Not much: Just because easy g gives rise to easy h does not mean that it is easy to compute $g \mapsto h$.

The first result (computability) can be strengthened nicely to a computability result of the operator $g \mapsto h$.

Open question:

How should we formulate PTIME computable operators?

What do the negative results imply?

- non-computable solution
- arbitrarily long time, even if unique
- ► PSPACE-complete, even if Lipschitz

Though we have no definition of "easy operators", they certainly should not take easy functions to hard ones.

The last result (say) implies: Unless PTIME = PSPACE, any efficient algorithm for Lipschitz IVP must fail on some g (and in particular on the one we constructed).

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