

# On the Complexity of Numerical Analysis

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## Abstract

*We study two quite different approaches to understanding the complexity of fundamental problems in numerical analysis:*

- *The Blum-Shub-Smale model of computation over the reals.*
- *A problem we call the “Generic Task of Numerical Computation,” which captures an aspect of doing numerical computation in floating point, similar to the “long exponent model” that has been studied in the numerical computing community.*

*We show that both of these approaches hinge on the question of understanding the complexity of the following problem, which we call PosSLP: Given a division-free straight-line program producing an integer  $N$ , decide whether  $N > 0$ .*

- *In the Blum-Shub-Smale model, polynomial time computation over the reals (on discrete inputs) is polynomial-time equivalent to PosSLP, when there are only algebraic constants. We conjecture that using transcendental constants provides no additional power, beyond nonuniform reductions to PosSLP, and we present some preliminary results supporting this conjecture.*
- *The Generic Task of Numerical Computation is also polynomial-time equivalent to PosSLP.*

*We prove that PosSLP lies in the counting hierarchy. Combining this with work of Tiwari, we obtain that the Euclidean Traveling Salesman Problem lies in the counting hierarchy – the previous best upper bound for this important problem (in terms of classical complexity classes) being PSPACE.*

*In the course of developing the context for our results on arithmetic circuits, we present some new observations on the complexity of ACIT: the Arithmetic Circuit Identity Testing problem. In particular, we show that if  $n!$  is not ultimately easy, then ACIT has subexponential complexity.*

## 1 Introduction

The original motivation for this paper comes from a desire to understand the complexity of computation over the reals in the Blum-Shub-Smale model. In Section 1.1 we give a brief introduction to this model and we introduce the problem PosSLP and explain its importance in understanding the Blum-Shub-Smale model.

In Section 1.2 we present yet another reason to be interested in PosSLP. We isolate a computational problem that lies at the root of the task of designing numerically stable algorithms. We show that this task is computationally equivalent to PosSLP. The material in Sections 1.1 and 1.2 provides motivation for studying PosSLP and for attempting to place it within the framework of traditional complexity classes.

In Section 1.3 we discuss our main technical contributions: proving upper and lower bounds on the complexity of PosSLP. In Section 1.4 we present applications of our main result with respect to the Euclidean Traveling Salesman Problem and the Sum-of-Square-Roots problem.

## 1.1 Polynomial Time Over the Reals

The Blum-Shub-Smale model of computation over the reals provides a very well-studied complexity-theoretic setting in which to study the computational problems of numerical analysis. We refer the reader to Blum, Cucker, Shub and Smale [12] for detailed definitions and background material related to this model; here, we will recall only a few salient facts. In the Blum-Shub-Smale model, each machine computing over the reals has associated with it a finite set  $S$  of real *machine constants*. The inputs to a machine are elements of  $\bigcup_n \mathbb{R}^n = \mathbb{R}^\infty$ , and thus each polynomial-time machine over  $\mathbb{R}$  accepts a “decision problem”  $L \subseteq \mathbb{R}^\infty$ . The set of decision problems accepted by polynomial-time machines over  $\mathbb{R}$  using only constants from  $S \cup \{0, 1\}$  is denoted  $P_{\mathbb{R}}^S$ . The union of the classes  $P_{\mathbb{R}}^S$  over all  $S$  is called *polynomial time over  $\mathbb{R}$*  and is denoted  $P_{\mathbb{R}}$ . The subclass  $P_{\mathbb{R}}^\emptyset$  of “constant-free polynomial time” is commonly denoted by  $P_{\mathbb{R}}^0$ , cf. Bürgisser and Cucker [22].

There has been considerable interest in relating computation over  $\mathbb{R}$  to the classical Boolean complexity classes such as P, NP, PSPACE, etc. This is accomplished by considering the *Boolean part* of decision problems over the reals. That is, given a problem  $L \subseteq \mathbb{R}^\infty$ , the Boolean part of  $L$  is defined as  $BP(L) := L \cap \{0, 1\}^\infty$ . (Here, we follow the notation of [12];  $\{0, 1\}^\infty = \bigcup_n \{0, 1\}^n$ , which is identical to  $\{0, 1\}^*$ .) The Boolean part of  $P_{\mathbb{R}}$ , denoted  $BP(P_{\mathbb{R}})$ , is defined as  $\{BP(L) \mid L \in P_{\mathbb{R}}\}$ .

By encoding the advice function in a single real constant as in Koiran [40], one can show that  $P/\text{poly} \subseteq BP(P_{\mathbb{R}})$ . The best upper bound on the complexity of problems in  $BP(P_{\mathbb{R}})$  that is currently known was obtained by Cucker and Grigoriev [25]:

$$BP(P_{\mathbb{R}}) \subseteq PSPACE/\text{poly}. \tag{1}$$

There has been *no* work pointing to lower bounds on the complexity of  $BP(P_{\mathbb{R}})$ ; nobody has presented any compelling evidence that  $BP(P_{\mathbb{R}})$  is not equal to  $P/\text{poly}$ .

There has also been some suggestion that perhaps  $BP(P_{\mathbb{R}})$  is equal to  $PSPACE/\text{poly}$ . For instance, certain variants of the RAM model that provide for unit-cost arithmetic can simulate all of PSPACE in polynomial time [9, 35]. Since the Blum-Shub-Smale model also provides for unit-time multiplication on “large” numbers, Cucker and Grigoriev [25] mention that researchers have raised the possibility that similar arguments might show that polynomial-time computation over  $\mathbb{R}$  might be able to simulate PSPACE. Cucker and Grigoriev also observe that certain naïve approaches to provide such a simulation must fail.

One of our goals is to provide evidence that  $BP(P_{\mathbb{R}})$  lies properly between  $P/\text{poly}$  and  $PSPACE/\text{poly}$ . Towards this goal, it is crucial to understand a certain decision problem PosSLP: *The problem of deciding, for a given division-free straight-line program, whether it represents a positive integer*. More generally, for a fixed finite subset  $S \subset \mathbb{R}$ ,  $\text{PosSLP}(S)$  is the problem of deciding for a given division-free straight-line program using constants from  $S \cup \{0, 1\}$ , whether the real number represented by it is positive. (For precise definitions, see the next section.)

The immediate relationship between the Blum-Shub-Smale model and the problems  $\text{PosSLP}(S)$  is given by the proposition below.

**Proposition 1.1** *We have  $P^{\text{PosSLP}(S)} = BP(P_{\mathbb{R}}^S)$  for all finite subsets  $S \subset \mathbb{R}$ . In particular,  $P^{\text{PosSLP}} = BP(P_{\mathbb{R}}^0)$ .*

*Proof.* It is clear that  $\text{PosSLP}(S)$  is in  $BP(P_{\mathbb{R}}^S)$ , since we can implement a standard SLP interpreter in the real Turing machine framework and evaluate the result in linear time using unit cost instructions. The result is then obtained by one sign test. To show the other direction, assume we have a polynomial time machine over  $\mathbb{R}$  using only the constants in  $S \cup \{0, 1\}$ . By a usual argument (separate computation of numerator and denominator), we may assume without loss of generality that the machine does not use divisions. Given a bit string as input, we simulate the computation by storing the straight-line program representation of the intermediate results instead of their values. Branch instructions can be simulated by using the oracle  $\text{PosSLP}(S)$  to determine if the contents of a given register (represented by a straight-line program) is greater than zero. □

It was shown by Chapuis and Koiran [23] that algebraic constants do not help. More specifically,  $P_{\mathbb{R}}^0$  is equal to the class of decision problems over the reals decided by polynomial time Blum-Shub-Smale machines using real algebraic numbers as constants.

As already mentioned, by encoding the advice function in a single real constant, one can show that  $P/\text{poly} \subseteq \text{BP}(P_{\mathbb{R}})$ . The proof in fact shows even  $P^{\text{PosSLP}}/\text{poly} \subseteq \text{BP}(P_{\mathbb{R}})$ . The real constant encoding the advice function, will, of course, in general be transcendental. Thus, there is a strong relationship between non-uniformity in the classical model of computation and the use of transcendental constants in the Blum-Shub-Smale model. We conjecture that this relationship can be further strengthened:

**Conjecture 1.2**  $P^{\text{PosSLP}}/\text{poly} = \text{BP}(P_{\mathbb{R}})$ .

In Section 3 we present some preliminary results toward proving this conjecture. For instance we prove that  $\text{BP}(P_{\mathbb{R}}^{\{\alpha\}}) \subseteq P^{\text{PosSLP}}/\text{poly}$  for almost all  $\alpha \in \mathbb{R}$ , in the sense of Lebesgue measure. We also show that  $\text{BP}(P_{\mathbb{R}}^{\{\alpha\}}) \subseteq P^{\text{PosSLP}}/1$  (one bit of advice) if  $\alpha$  is the value of an elementary function on a rational number. For instance this is the case for the well-known transcendental numbers  $e$  or  $\pi$ .

## 1.2 The Task of a Numerical Analyst

The Blum-Shub-Smale model is a very elegant one, but it does not take into account the fact that actual numerical computations have to deal with *finitely* represented values. We next observe that even if we take this into account, the PosSLP problem still captures the complexity of numerical computation.

Let  $u \neq 0$  be a dyadic rational number. The *floating point* representation of  $u$  is obtained by writing  $u = v2^m$  where  $m$  is an integer and  $\frac{1}{2} \leq |v| < 1$ . The floating point representation is then given by the sign of  $v$ , and the usual binary representations of the numbers  $|v|$  and  $m$ . The floating point representation of 0 is the string 0 itself. We shall abuse notation and identify the floating point representation of a number with the number itself, using the term “floating point number” for the number as well as its representation.

Let  $u \neq 0$  be a real number. We may write  $u$  as  $u = u'2^m$  where  $\frac{1}{2} \leq |u'| < 1$  and  $m$  is an integer. Then, we define a *floating point approximation of  $u$  with  $k$  significant bits* to be a floating point number  $v2^m$  so that  $|v - u'| \leq 2^{-(k+1)}$ .

We will focus on one part of the job that is done by numerical analysts: the design of numerically-stable algorithms. In our scenario, the numerical analyst starts out with a known function  $f$ , and the task is to design a “good” algorithm for it. When we say that the function  $f$  is “known”, we mean that the analyst starts out with some method of computing (or at least approximating)  $f$ ; we restrict attention to the “easy” case where the method for computing  $f$  uses only the arithmetic operations  $+$ ,  $-$ ,  $*$ ,  $\div$ , and thus the description of  $f$  that the analyst is given can be presented as an arithmetic circuit with operations  $+$ ,  $-$ ,  $*$ ,  $\div$ . Usually, the analyst also has to worry about the problems that are caused by the fact that the inputs to  $f$  are not known precisely, but are only given as floating point numbers that are approximations to the “true” inputs – but again we will focus on the “easy” case where the analyst will merely try to compute a good approximation for  $f(x_1, \dots, x_n)$  on the exact floating point numbers  $x_1, \dots, x_n$  that are presented as input:

**The generic task of numerical computation (GTNC):** *Given an integer  $k$  in unary and a straight-line program (with  $\div$ ) taking as inputs floating point numbers, with a promise that it neither evaluates to zero nor does division by zero, compute a floating point approximation of the value of the output with  $k$  significant bits.*

The traditional approach that numerical analysts have followed in trying to solve problems of this sort is to study the numerical stability of the algorithm represented by the circuit, and in case of instability, to attempt to devise an equivalent computation that is numerically stable. Although stable algorithms have been found for a great many important functions, the task of devising such algorithms frequently involves some highly nontrivial mathematics and algorithmic ingenuity. There seems to be no expectation that there will ever be a purely automatic way to solve this problem, and indeed there seems to be no expectation that a numerically stable algorithm will exist in general. To summarize, there is substantial empirical evidence that the generic task of numerical computation is intractable. It would be of significant practical interest if, contrary to expectation, it should turn out to be very easy to solve (say, solvable in linear time).

We show that the generic task of numerical computation is equivalent in power to PosSLP.

**Proposition 1.3** *The generic task of numerical computation (GTNC) is polynomial time Turing equivalent to PosSLP.*

*Proof.* We first reduce PosSLP to the generic task of numerical computation. Given a division-free straight-line program representing the number  $N$ , we construct a straight-line program computing the value  $v = 2N - 1$ . The only inputs 0, 1 of

this program can be considered to be floating point numbers and this circuit clearly satisfies the promise of the generic task of numerical computation. Then  $N > 0$  if  $v \geq 1$  and  $N \leq 0$  if  $v \leq -1$ . Determining an approximation of  $v$  to one significant bit is enough to distinguish between these cases.

Conversely, suppose we have an oracle solving PosSLP. Given a straight-line program with inputs being floating point numbers, we first convert it to a straight-line program having only input 1; it is easy to see that this can be done in polynomial time. By standard techniques we move all  $\div$  gates to the top, so that the program computes a value  $v = v_1/v_2$ , where  $v_1, v_2$  are given by division-free straight-line programs. We can use the oracle to determine the signs of  $v_1$  and  $v_2$ . Without loss of generality assume that  $v$  is positive. Next we use the oracle to determine if  $v_1 \geq v_2$ . Suppose this is indeed the case (the opposite case is handled similarly).

We then find the least  $r$ , so that  $2^{r-1} \leq v < 2^r$ , by first comparing  $v_1$  with  $v_2 2^{2^i}$  for  $i = 0, 1, 2, 3, \dots$ , using the oracle, thus finding the minimum  $i$  so that  $v < 2^{2^i}$  and afterwards doing a binary search, again using the oracle to compare  $v_1$  to  $v_2 2^r$  for various values of  $r$ . This takes polynomial time.

The desired output is a floating point number  $u = u'2^r$ , where  $|v - u'| \leq 2^{-(k+1)}$ . To obtain  $u'$  we first want to find the integer  $w$  between  $2^k$  and  $2^{k+1} - 1$  so that  $w/2^{k+1} \leq v/2^r < (w+1)/2^{k+1}$ . Since  $w/2^{k+1} \leq v/2^r < (w+1)/2^{k+1}$  iff  $w2^r v_2 \leq v_1 2^{k+1} < (w+1)2^r v_2$ , we can determine this by another binary search, using  $O(k)$  calls to the oracle. We then output the sign of  $v$ , the binary representation of the rational  $w/2^{k+1}$ , and the binary representation of  $r$ , together forming the desired floating point approximation of  $v$ .  $\square$

The reader may wonder how GTNC fits into the numerical analysis literature. The Long Exponent Model (LEM) of Demmel [28, 29] offers the closest parallel. Demmel considers the classic problem of computation of the determinant, and he identifies three ways of modeling the problem, which he calls the Traditional Model, the Short Exponent Model (SEM), and the Long Exponent Model (LEM). Computing determinants is easy in the SEM, while in the LEM the problem is equivalent to a special case of GTNC. Namely, it is equivalent to instances of GTNC where the circuit  $C$  that is provided as input is the polynomial-size SLP for determinants given by Berkowitz [8].

Demmel goes so far as to conjecture that, in the LEM, the problem of deciding if the determinant is zero is NP-hard [28]. Since this problem is actually a special case of EquSLP and thus lies in BPP, Demmel's conjecture is almost certainly false. However, we agree with his underlying intuition, in that we believe that the problem of deciding if the determinant is *positive* in the LEM very likely is intractable (even if we see no evidence that it is NP-hard). That is, this special case of PosSLP is recognized as a difficult problem by the numerical analysis community.

### 1.3 The Complexity of PosSLP

We consider Proposition 1.3 to be evidence for the computational intractability of PosSLP. If PosSLP is in P/poly then there is a polynomial-sized "cookbook" that can be used in place of the creative task of devising numerically stable computations. This seems unlikely.

We wish to emphasize that the generic task of numerical computation models the *discrete* computational problem that underlies an important class of computational problems. Thus it differs quite fundamentally from the approach taken in the Blum-Shub-Smale model.

We also wish to emphasize that, in defining the generic task of numerical computation, we are *not* engaging in the debate over which real functions are "efficiently computable". There is by now a large literature comparing and contrasting the relative merits of the Blum-Shub-Smale model with the so-called "bit model" of computing, and there are various competing approaches to defining what it means for a real-valued function to be feasible to compute; see [10, 15, 16, 62, 63] among others. Our concerns here are orthogonal to that debate. We are not trying to determine which real-valued functions are feasible; we are studying a discrete computational problem that is relevant to numerical analysis, with the goal of proving upper and lower bounds on its complexity.

The generic task of numerical computation is one way of formulating the notion of what is feasible to compute in a world where *arbitrary precision* arithmetic is available for free. In contrast, the Blum-Shub-Smale model can be interpreted as formulating the notion of feasibility in a world where *infinite precision* arithmetic is available for free. According to Proposition 1.3, both of these approaches are *equivalent* (and captured by  $P^{\text{PosSLP}}$ ) when only algebraic constants are allowed in the Blum-Shub-Smale model. Conjecture 1.2 claims that this is also true when allowing arbitrary real constants.

As another demonstration of the computational power of PosSLP, we show in §2 that the problem of determining the total degree of a multivariate polynomial over the integers given as a straight-line program reduces to PosSLP.

The above discussion suggests that PosSLP is not an easy problem. Can more formal evidence of this be given? Although it would be preferable to show that PosSLP is hard for some well-studied complexity class, the best that we can do is observe

that a somewhat stronger problem (BitSLP) is hard for  $\#P$ . This will be done in §2.

The above discussion also suggests that non-trivial upper bounds for PosSLP are of great interest. Prior to this paper, the best upper bound was PSPACE. Our main technical result is an improved upper bound: We show, based on results on the uniform circuit complexity of integer division and the relationship between constant depth circuits and subclasses of PSPACE [5, 37], that PosSLP lies in the counting hierarchy CH, a well-studied subclass of PSPACE that bears more or less the same relationship to  $\#P$  as the polynomial hierarchy bears to NP [59, 61].

**Theorem 1.4** PosSLP is in  $P^{PP^{PP^{PP}}}$ .

Another interesting upper bound for PosSLP was recently discovered by Tarasov and Vyalyi [56], who give a reduction from PosSLP to the *Semidefinite Feasibility Problem* (SFDP), i.e. the feasibility version of the optimization problem *Semidefinite Programming*. Their result can be seen as a lower bound for SFDP. SFDP is known to reduce to its complement and to lie in  $NP_{\mathbb{R}}$  [50]; also it is easy to see that SFDP reduces to the existential theory of the reals (for instance, see the discussion in [50]), and thus SFDP  $\in$  PSPACE.

We suspect that PosSLP lies at an even lower level of CH. We leave as major open problems the question of providing better upper bounds for PosSLP and the question of providing any sort of hardness theorem, reducing a supposedly intractable problem to PosSLP.

We also believe that it would be very interesting to verify Conjecture 1.2, as this would give a characterization of  $BP(P_{\mathbb{R}})$  in terms of classical complexity classes. But in fact, it would be equally interesting to refute it under some plausible complexity theoretic assumption, as this would give evidence that the power of using transcendental constants in the Blum-Shub-Smale model goes beyond the power of non-uniformity in classical computation.

## 1.4 Applications

The *Sum-of-square-roots problem* is a well-known problem with many applications to computational geometry and elsewhere. The input to the problem is a list of integers  $(d_1, \dots, d_n)$  and an integer  $k$ , and the problem is to decide if  $\sum_i \sqrt{d_i} \geq k$ . The complexity of this problem is posed as an open question by Garey, Graham and Johnson [34] in connection with the Euclidean traveling salesman problem, which is not known to be in NP, but which is easily seen to be solvable in NP relative to the Sum-of-square-roots problem. See also O’Rourke [48, 49] and Etessami and Yannakakis [32] for additional information. Although it has been conjectured [47] that the problem lies in P, it seems that no classical complexity class smaller than PSPACE has been known to contain this problem. On the other hand, Tiwari [57] showed that the problem can be decided in polynomial time on an “algebraic random-access machine”. In fact, it is easy to see that the set of decision problems decided by such machines in polynomial time is exactly  $BP(P_{\mathbb{R}}^0)$ . Thus by Proposition 1.1 we see that the Sum-of-square-roots problem reduces to PosSLP. Theorem 1.4 thus yields the following corollary.

**Corollary 1.5** *The Sum-of-square-roots problem and the Euclidean Traveling Salesman Problem are in CH.*

## 2 Preliminaries

Our definitions of arithmetic circuits and straight-line programs are standard. An *arithmetic circuit* is a directed acyclic graph with input nodes labeled with the constants 0, 1 or with indeterminates  $X_1, \dots, X_k$  for some  $k$ . Internal nodes are labeled with one of the operations  $+$ ,  $-$ ,  $*$ ,  $\div$ . A *straight-line program* is a sequence of instructions corresponding to a sequential evaluation of an arithmetic circuit. If it contains no  $\div$  operation it is said to be *division free*. Unless otherwise stated, all the straight-line programs considered will be division-free. Thus straight-line programs can be seen as a very compact representation of a polynomial over the integers. In many cases, we will be interested in division-free straight-line programs using no indeterminates, which thus represent an integer.

By the  $n$ -bit binary representation of an integer  $N$  such that  $|N| < 2^n$  we understand a bit string of length  $n + 1$  consisting of a *sign bit* followed by  $n$  bits encoding  $|N|$  (padded with leading zeroes, if needed).

We consider the following problems:

EquSLP Given a straight-line program representing an integer  $N$ , decide whether  $N = 0$ .

ACIT Given a straight-line program representing a polynomial  $f \in \mathbb{Z}[X_1, \dots, X_k]$ , decide whether  $f = 0$ .

DegSLP: Given a straight-line program representing a polynomial  $f \in \mathbb{Z}[X_1, \dots, X_k]$ , and given a natural number  $d$  in binary, decide whether  $\deg f \leq d$ .

PosSLP Given a straight-line program representing  $N \in \mathbb{Z}$ , decide whether  $N > 0$ .

BitSLP Given a straight-line program representing  $N$ , and given  $n, i \in \mathbb{N}$  in binary, decide whether the  $i$ th bit of the  $n$ -bit binary representation of  $N$  is 1.

It is not clear that any of these problems is in P, since straight-line program representations of integers can be exponentially smaller than ordinary binary representation.

There is an immediate relationship between the Blum-Shub-Smale model over the complex numbers  $\mathbb{C}$  and the problem EquSLP. Let  $P_{\mathbb{C}}^0$  denote the class of decision problems over  $\mathbb{C}$  decided by polynomial time Blum-Shub-Smale machines using only the constants 0, 1. Similarly as for Proposition 1.1 one can show that  $P^{\text{EquSLP}} = \text{BP}(P_{\mathbb{C}}^0)$ . On the other hand, it is known that constants can be eliminated in this setting [11, 41], hence  $\text{BP}(P_{\mathbb{C}}) = \text{BP}(P_{\mathbb{C}}^0)$ . We therefore have

**Proposition 2.1**  $P^{\text{EquSLP}} = \text{BP}(P_{\mathbb{C}})$ .

Clearly, EquSLP is a special case of ACIT. Schönhage [53] showed that EquSLP is in coRP, using computation modulo a randomly chosen prime. Ibarra and Moran [38], building on DeMillo and Lipton [27], Schwartz [54] and Zippel [64], extended this to show that ACIT lies in coRP. In the spirit of Adleman’s observation [1], Heintz and Schnorr [36] established the existence of nonuniform polynomial time algorithms for an algebraic variant of the ACIT problem (allowing any field elements as constants). The problem ACIT has recently attracted much attention due to the work of Kabanets and Impagliazzo [39] who showed that a deterministic algorithm for ACIT would yield circuit lower bounds. (See [43] for some progress on finding deterministic algorithms for certain versions of the problem.) As far as we know, it has not been pointed out before that ACIT is actually polynomial time equivalent to EquSLP. In other words, disallowing indeterminates in the straight-line program given as input does not make ACIT easier. Or more optimistically: It is enough to find a deterministic algorithm for this special case in order to have circuit lower bounds.

**Proposition 2.2** ACIT is polynomial-time equivalent to EquSLP.

*Proof.* We are given a straight-line program of size  $n$  with  $m$  indeterminates  $X_1, \dots, X_m$ , computing the polynomial  $p(X_1, \dots, X_m)$ . Define  $B_{n,i} = 2^{2^{in^2}}$ . Straight-line-programs computing these numbers using iterated squaring can easily be constructed in polynomial time, so given a straight-line-program for  $p$ , we can easily construct a straight-line program for  $p(B_{n,1}, \dots, B_{n,m})$ . We shall show that for  $n \geq 3$ ,  $p$  is identically zero iff  $p(B_{n,1}, \dots, B_{n,m})$  evaluates to zero.

To see this, first note that the “only if” part is trivial, so we only have to show the “if” part. Thus, assume that  $p(X_1, \dots, X_m)$  is not the zero-polynomial. Let  $q(X_1, \dots, X_m)$  be the largest monomial occurring in  $p$  with respect to inverse lexicographic order<sup>1</sup> and let  $k$  be the number of monomials. We can write  $p = \alpha q + \sum_{i=1}^{k-1} \alpha_i q_i$ , where  $(q_i)_{i=1, \dots, k-1}$  are the remaining monomials. An easy induction in the size of the straight line program shows that  $|\alpha_i| \leq 2^{2^{2n}}$ ,  $k \leq 2^{2^n}$  and that the degree of any variable in any  $q_i$  is at most  $2^n$ .

Now, our claim is that the absolute value  $|\alpha q(B_{n,1}, \dots, B_{n,m})|$  is strictly bigger than  $|\sum_{i=1}^{k-1} \alpha_i q_i(B_{n,1}, \dots, B_{n,m})|$ , and thus we cannot have that  $p(B_{n,1}, \dots, B_{n,m}) = 0$ .

Indeed, since the monomial  $q$  was the biggest in the inverse lexicographic ordering, we have that for any other monomial  $q_i$  there is an index  $j$  so that

$$\frac{q(B_{n,1}, \dots, B_{n,m})}{q_i(B_{n,1}, \dots, B_{n,m})} \geq \frac{2^{2^{jn^2}}}{\prod_{l=1}^{j-1} 2^{2^{ln^2} \cdot 2^n}} > 2^{2^{n^2-1}},$$

so we can bound

$$\begin{aligned} \left| \sum_{i=1}^{k-1} \alpha_i q_i(B_{n,1}, \dots, B_{n,m}) \right| &\leq 2^{2^n} 2^{2^{2n}} \left| \max_{i=1}^{k-1} q_i(B_{n,1}, \dots, B_{n,m}) \right| \\ &\leq 2^{2^n} 2^{2^{2n}} 2^{-2^{n^2-1}} |q(B_{n,1}, \dots, B_{n,m})| < q(B_{n,1}, \dots, B_{n,m}) \leq |\alpha q(B_{n,1}, \dots, B_{n,m})|, \end{aligned}$$

which proves the claim.  $\square$

<sup>1</sup> $X_1^{\alpha_1} \dots X_m^{\alpha_m}$  is greater than  $X_1^{\beta_1} \dots X_m^{\beta_m}$  in this order iff the right-most nonzero component of  $\alpha - \beta$  is positive, cf. Cox, Little and O’Shea [24, p. 59].

We believe that Proposition 2.2 could be a useful tool for devising deterministic algorithms for ACIT. Indeed, in Section 5, we use it to devise a new subexponential algorithm for ACIT based on the assumption that a conjecture of Shub and Smale is correct.

The problem DegSLP is not known to lie in BPP, even for the special case of univariate polynomials. Here, we show that it reduces to PosSLP.

**Proposition 2.3** *DegSLP polynomial time many-one reduces to PosSLP.*

*Proof.* We first show the reduction for the case of univariate polynomials (i.e., straight-line-programs with a single indeterminate) and afterwards we reduce the multivariate case to the univariate case.

Let  $f \in \mathbb{Z}[X]$  be given by a straight-line program of length  $n$ . To avoid having to deal with the zero polynomial of degree  $-\infty$  and to ensure that the image of the polynomial is a subset of the non-negative integers, we first change the straight-line program computing  $f$  into a straight-line program computing  $f_1(X) = (Xf(X) + 1)^2$  by adding a few extra lines. We can check if the degree of  $f$  is at most  $d$  by checking if the degree of  $f_1$  is at most  $D = 2(d + 1)$  (except for  $d = -\infty$  in which case we check if the degree of  $f_1$  is at most  $D = 0$ ).

Let  $B_n$  be the integer  $2^{2^{n^2}}$ . As in the proof of Proposition 2.2, we can easily construct a straight-line program computing  $B_n$  and from this a straight-line program computing  $f_1(B_n)$ .

Now, suppose that  $\deg f_1 \leq D$ . Using the same bounds on sizes of the coefficients as in the proof of Proposition 2.2 and assuming without loss of generality that  $n \geq 3$ , we then have

$$f_1(B_n) \leq \sum_{i=0}^D 2^{2^{2n}} B_n^i < (2^n + 1)2^{2^{2n}} B_n^D \leq (2^{2^n} + 1)2^{2^{2n} - 2^{n^2}} B_n^{D+1} < B_n^{D+1}/2.$$

On the other hand suppose that  $\deg f_1 \geq D + 1$ . Then we have

$$f_1(B_n) \geq (B_n)^{D+1} - \sum_{i=0}^D 2^{2^{2n}} B_n^i \geq B_n^{D+1} - 2^{2^n} 2^{2^{2n}} 2^{-2^{n^2}} B_n^{D+1} > B_n^{D+1}/2.$$

Thus, to check whether  $\deg f_1 \leq D$ , we just need to construct a straight-line-program for  $2f_1(B_n) - B_n^{D+1}$  and check whether it computes a positive integer. This completes the reduction for the univariate case.

We next reduce the multivariate case to the univariate case. Thus, let  $f \in \mathbb{Z}[X_1, \dots, X_m]$  be given by a straight-line program of length  $n$ . Let  $f^* \in \mathbb{Z}[X_1, \dots, X_m, Y]$  be defined by  $f^*(X_1, \dots, X_m, Y) = f(X_1Y, \dots, X_mY)$ . We claim that if we let  $B_{n,i} = 2^{2^{in^2}}$  as in the proof of Proposition 2.2, then, for  $n \geq 3$ , the degree of the univariate polynomial  $f^*(B_{n,1}, \dots, B_{n,m}, Y)$  is equal to the total degree of  $f$ . Indeed, we can write  $f^*$  as a polynomial in  $Y$  with coefficients in  $\mathbb{Z}[X_1, \dots, X_m]$ :

$$f^*(X_1, \dots, X_m, Y) = \sum_{j=0}^{d^*} g_j(X_1, \dots, X_m) Y^j$$

where  $d^*$  is the degree of variable  $Y$  in the polynomial  $f^*$ . Note that this is also the total degree of the polynomial  $f$ . Now, the same argument as used in the proof of Proposition 2.2 shows that since  $g_{d^*}$  is not the zero-polynomial,  $g_{d^*}(B_{n,1}, B_{n,2}, \dots, B_{n,m})$  is different from 0.  $\square$

As PosSLP easily reduces to BitSLP, we obtain the chain of reductions

$$\text{EquSLP} \equiv \text{ACIT} \leq_m^p \text{DegSLP} \leq_m^p \text{PosSLP} \leq_m^p \text{BitSLP}.$$

In §4 we will show that all the above problems in fact lie in the counting hierarchy CH.

The complexity of BitSLP contrasts sharply with that of EquSLP.

**Proposition 2.4** *BitSLP is hard for #P.*

*Proof.* A similar result is stated without proof in [28]. The proof that we present is quite similar to that of Bürgisser [20, Prop. 5.3], which in turn is based on ideas of Valiant [60]. We show that computing the permanent of matrices with entries from  $\{0,1\}$  is reducible to BitSLP.

Given a matrix  $X$  with entries  $x_{i,j} \in \{0, 1\}$ , consider the univariate polynomial

$$f_n = \sum_i f_{n,i} Y^i = \prod_{i=1}^n \left( \sum_{j=1}^n x_{i,j} Y^{2^{j-1}} \right)$$

which can be represented by a straight-line program of size  $O(n^2)$ . Then  $f_{n,2^{n-1}}$  equals the permanent of  $X$ . Let  $N$  be the number that is represented by the straight-line program that results by replacing the indeterminate  $Y$  with  $2^{n^3}$ . It is easy to see that the binary representation of  $f_{n,2^{n-1}}$  appears as a sequence of consecutive bits in the binary representation of  $N$ .  $\square$

### 3 Transcendental Constants

We present here some first results toward establishing our Conjecture 1.2.

Let  $S$  denote a fixed finite subset of  $\mathbb{R}$ . By an *SLP over  $S$*  we shall understand a division-free straight-line program using constants from  $S \cup \{0, 1\}$ . Recall the following problem:

PosSLP( $S$ ) Given an SLP over  $S$ , decide whether the real number represented by it is positive.

**Remark 3.1** We could have defined a variant of PosSLP( $S$ ) by allowing divisions in the straight-line programs. However, this variant is easily seen to be polynomial time equivalent to PosSLP( $S$ ). Indeed, by computing separately with numerators and denominators we can transform an SLP representing  $\alpha$  into two division-free SLPs representing numbers  $A, B$  such that  $\alpha = A/B$ . Hereby, the length of the SLPs increases at most by a factor of four. Now  $\alpha$  is positive iff  $AB$  is positive.

A result by Chapuis and Koiran [23] implies that algebraic constants can be eliminated. It can be stated as follows:

**Proposition 3.2** Let  $S \subseteq \mathbb{R}$  be finite and  $\alpha \in \mathbb{R}$  be algebraic over the field  $\mathbb{Q}(S)$ . Then  $\mathsf{P}^{\text{PosSLP}(S \cup \{\alpha\})} = \mathsf{P}^{\text{PosSLP}(S)}$ .

Our first goal is to prove that almost all transcendental constants can be eliminated.

**Theorem 3.3** For all  $(\alpha_1, \alpha_2, \dots, \alpha_k) \in \mathbb{R}^k$  except in a subset of Lebesgue measure zero we have  $\mathsf{P}^{\text{PosSLP}(\{\alpha_1, \dots, \alpha_k\})} / \text{poly} = \mathsf{P}^{\text{PosSLP}} / \text{poly}$ .

The proof will require some lemmas. The idea is to eliminate one by one the elements of such sets  $S$ , replacing each element with appropriate advice of polynomial size.

We denote by  $R_n^S \subset \mathbb{R}$  the set of all real numbers that occur as a root of some nonzero univariate polynomial that is computed by a division-free straight-line program of size  $n$  that uses constants in  $S$ . Note that  $\mathbb{R} \setminus R_n^S$  consists of a collection of open intervals. Clearly, any univariate polynomial computed from  $S$  by an SLP of size  $n$  has constant sign on each of these intervals. For  $\alpha \in \mathbb{R} \setminus R_n^S$ , we denote by  $I_n^S(\alpha)$  the unique interval containing  $\alpha$ .

**Remark 3.4** A real number  $\alpha$  is transcendental over  $\mathbb{Q}(S)$  iff  $\alpha \notin R_n^S$  for all  $n$  (or equivalently, for infinitely many  $n$ ).

**Definition 3.5** We call a real number  $\alpha$  *approximable with respect to  $S$*  if either  $\alpha$  is algebraic over  $\mathbb{Q}(S)$  or else if  $\alpha$  is transcendental over  $\mathbb{Q}(S)$  and satisfies the following condition: there exists a polynomial  $p$  such that for all sufficiently large  $n \in \mathbb{N}$  the interval  $I_n^S(\alpha)$  contains an element  $x_n$  that can be represented by an SLP over  $S$  of size  $p(n)$ , possibly using divisions. (Note that this interval is well-defined as  $\alpha \notin R_n^S$ , cf. Remark 3.4.) We say that  $\alpha$  is approximable iff it is approximable with respect to the empty set.

**Lemma 3.6** If  $\alpha \in \mathbb{R}$  is approximable with respect to  $S$ , then  $\mathsf{P}^{\text{PosSLP}(S \cup \{\alpha\})} / \text{poly} = \mathsf{P}^{\text{PosSLP}(S)} / \text{poly}$ .

*Proof.* Suppose  $\alpha \in \mathbb{R}$  is approximable with respect to  $S$ . By Proposition 3.2 we may assume that  $\alpha$  is transcendental over  $\mathbb{Q}(S)$ . Then, for all sufficiently large  $n$ , there exist  $x_n \in I_n^S(\alpha)$  computed by an SLP  $\Gamma_n$  over  $S$  (using divisions) of size polynomial in  $n$ .

It is sufficient to show that PosSLP( $S \cup \{\alpha\}$ ) is contained in  $\mathsf{P}^{\text{PosSLP}(S)} / \text{poly}$ . Let  $C$  be an SLP (of size  $n$ ) over  $S \cup \{\alpha\}$  computing  $v \in \mathbb{R}$ . We want to decide whether  $v$  is positive. If we replace the constant  $\alpha$  by the variable  $X$ , then this SLP

computes a polynomial  $f(X)$  and we have  $v = f(\alpha)$ . Since the sign of  $f$  is constant on the interval  $I_{p(n)}^S(\alpha)$ ,  $v$  has the same sign as  $f(x_{p(n)})$ .

We interpret the SLP  $\Gamma_n$  over  $S$  as an advice of polynomial size. By concatenating  $\Gamma_n$  with the SLP for  $f$ , we obtain an SLP over  $S$  that computes  $f(x_{p(n)})$ . We eliminate the divisions in the concatenated SLP according to Remark 3.1. Then the sign of this number is obtained by one oracle call to  $\text{PosSLP}(S)$ .  $\square$

**Lemma 3.7** *We have:*

1.  $|R_n^S| \leq (6(n + |S|))^n$ .
2. *The minimal distance between two different elements of  $R_n^\emptyset$  is at least  $2^{-2^{N_n}}$  with  $N_n = O(n \log n)$ .*

*Proof.* Let  $F_n$  be the product of all nonzero univariate polynomials  $f$  that can be computed from the variable  $X$  by an SLP over  $S$  of size  $n$ . Note that such  $f$  have degree at most  $2^n$ . Then  $R_n^S$  is the set of roots of  $F_n$ . There are at most  $\prod_{i=1}^n 3(|S| + i - 1)^2 \leq (3(|S| + n)^2)^n$  many SLPs over  $S$ . Therefore,  $\deg F_n \leq (6(|S| + n)^2)^n$ , which shows the first assertion.

Before showing the second assertion we introduce a notation: let  $\|g\|_1$  denote the sum of the absolute values of the coefficients of a univariate polynomial  $g$ . It is easy to see that  $\|g \cdot h\|_1 \leq \|g\|_1 \cdot \|h\|_1$ .

Suppose now  $S = \emptyset$ . If  $f(X)$  is computed by an SLP of size  $n$  over  $\emptyset$  from the variable  $X$ , then one can show that  $\log \|f\|_1 \leq (n + 1)2^n$ , see e.g. [19, Lemma 4.16]. By the submultiplicativity of  $\|\cdot\|_1$  we conclude

$$\log \|F_n\|_1 \leq (3n^2)^n (n + 1)2^n \leq 2^{O(n \log n)}.$$

Rump [51] has shown that the distance between any two distinct real roots in a univariate polynomial  $P$  with integer coefficients and degree  $d$  is at least  $2\sqrt{2}(d^{\frac{d}{2}+1}(\|P\|_1 + 1)^d)^{-1}$ . The second assertion follows by applying this bound to the polynomial  $F_n$ .  $\square$

**Lemma 3.8** *For any finite  $S \subset \mathbb{R}$ , the set of real numbers that are not approximable with respect to  $S$  has Lebesgue measure zero.*

*Proof.* Let  $\alpha \in \mathbb{R}$  and  $x_n$  be the binary approximation of  $\alpha$  with a precision of  $n^2$  digits, i.e.,  $|\alpha - x_n| < 2^{-n^2}$ . Clearly, there is an SLP over  $\{1/2\}$  of size  $O(n^2)$  representing  $x_n$ . Furthermore, suppose that  $\alpha$  has distance at least  $2^{-n^2}$  from  $R_n^S$  for all sufficiently large  $n$ , say for  $n \geq m$ . Then  $x_n$  is contained in the interval  $I_n^S(\alpha)$  for  $n \geq m$ . Hence, by definition,  $\alpha$  is approximable with respect to  $S$ .

These reasonings show that for all  $m \in \mathbb{N}$ :

$$B := \{\alpha \in \mathbb{R} \mid \alpha \text{ is not approximable wrt. } S\} \subseteq \bigcup_{n \geq m} U_n,$$

where  $U_n := \{x \in \mathbb{R} \mid \exists \rho \in R_n^S \mid |x - \rho| < 2^{-n^2}\}$  denotes the  $2^{-n^2}$ -neighborhood of  $R_n^S$ . Denoting by  $\lambda(A)$  the Lebesgue measure of a set  $A \subseteq \mathbb{R}$ , we get from Lemma 3.7

$$\lambda(U_n) \leq 2|R_n^S| 2^{-n^2} \leq 2(6(n + |S|))^n 2^{-n^2} \leq 2^{-\frac{1}{2}n^2}$$

for sufficiently large  $n$ . Therefore, we conclude that for all sufficiently large  $m$

$$\lambda(B) \leq \sum_{n=m}^{\infty} 2^{-\frac{1}{2}n^2}.$$

Since the series  $\sum_n 2^{-\frac{1}{2}n^2}$  is convergent and  $m$  was arbitrary, we conclude that  $\lambda(B) = 0$ .  $\square$

*Proof of Theorem 3.3.* We consider for  $\alpha := (\alpha_1, \dots, \alpha_k) \in \mathbb{R}^k$  and  $0 \leq i \leq k$  the complexity classes  $\mathcal{C}_i(\alpha) := \text{P}^{\text{PosSLP}(\{\alpha_1, \dots, \alpha_i\})}/\text{poly}$ . Clearly,  $\mathcal{C}_k(\alpha) \neq \mathcal{C}_0(\alpha)$  implies that  $\mathcal{C}_s(\alpha) \neq \mathcal{C}_{s-1}(\alpha)$  for some index  $s$ . By applying Lemma 3.6 to the set of constants  $S = \{\alpha_1, \dots, \alpha_{s-1}\}$  we obtain

$$\{\alpha \in \mathbb{R}^k \mid \mathcal{C}_k(\alpha) \neq \mathcal{C}_0(\alpha)\} \subseteq \bigcup_{s=1}^k \{\alpha \in \mathbb{R}^k \mid \alpha_s \text{ is not approximable wrt. } \{\alpha_1, \dots, \alpha_{s-1}\}\}.$$

Lemma 3.8 says that, for fixed  $\alpha_1, \dots, \alpha_{s-1}$ , the set  $\{\alpha_s \in \mathbb{R} \mid \alpha_s \text{ is not approximable wrt. } \{\alpha_1, \dots, \alpha_{s-1}\}\}$  has Lebesgue measure zero. It follows from Fubini that the right-hand subset of  $\mathbb{R}^k$  has measure zero as well, which shows the assertion.  $\square$

We can actually prove for many specific real numbers that they are approximable. Indeed, quite surprisingly, for any elementary function  $f(X)$  there exists a sequence  $(R_n(X))$  of rational functions such that  $|R_n(x) - f(x)| < 2^{-n}$  for all  $x \in [0, 1]$ , and such that  $R_n(X)$  can be computed by a straight-line program of polylogarithmic size (using divisions) from  $X$ . The elementary functions include the algebraic functions, the natural logarithm and the exponential function. For algebraic functions, such approximating rational functions can be constructed with Newton's method, see Kung and Traub [42]. For the natural logarithm, the construction of such approximations relies on the AGM iteration going back to Gauss, Lagrange and Legendre, which, in particular, gives very good approximations of  $\pi$ . The latter algorithms were discovered by Brent [17] and Salamin [52]. The book by Borwein and Borwein [14] provides a complete and in-depth exposition of this subject.

More precisely, we shall understand by an *elementary function* a function built up from rational constants by finitely many arithmetic operations, applications of  $\exp$ ,  $\ln$ , and the operation of taking a solution of a polynomial equation. (For a formal definition see [18].)

**Theorem 3.9** *Let  $\alpha$  be the value of an elementary function at a rational number. Then:*

1.  $\alpha$  is approximable. In particular,  $e = \exp(1)$  and  $\pi$  are approximable.
2. We have  $\text{P}^{\text{PosSLP}(\{\alpha\})} \subseteq \text{P}^{\text{PosSLP}}/1$ , where  $/1$  means one bit of advice.

*Proof.* 1. By Lemma 3.7 we know that  $\epsilon_n = 2^{-2^{N_n}}$  with  $N_n = O(n \log n)$  is a lower bound on the minimum distance between two different elements of  $R_n^0$ . Note that there is an SLP over  $\{1/2\}$  of polynomial size computing  $\epsilon_n$  (repeated squaring).

Let  $\alpha$  be as in the statement of the theorem. Without loss of generality we may assume that  $\alpha$  is transcendental. According to Borwein and Borwein [13, Table 1], for each  $n$ , there is an SLP of size  $n^{O(1)}$  (using divisions) computing an approximation  $a_n$  of  $\alpha$  that satisfies  $|a_n - \alpha| < \frac{1}{2}\epsilon_n$ . By checking the proofs (cf. Borwein and Borwein [14]) one sees that these SLPs are uniform, i.e., they can be constructed in polynomial time in  $n$ .

We claim that there exist  $b_n \in \{0, 1\}$ , such that  $x_n = a_n + b_n \frac{1}{2}\epsilon_n$  lies in the interval  $I_n^0(\alpha)$ , and thus satisfies the requirement in Definition 3.5. Hence  $\alpha$  is approximable.

Indeed, let  $\ell_n$  and  $r_n$  denote the left and right endpoint of the interval  $I_n^0(\alpha)$  and denote by  $m_n := \frac{1}{2}(\ell_n + r_n)$  its midpoint. Consider first the case where  $\alpha < m_n$ . If  $\alpha \leq a_n$ , then  $a_n < \alpha + \frac{1}{2}\epsilon_n < m_n + \frac{1}{2}\epsilon_n \leq r_n$ , hence  $x_n := a_n \in I_n^0(\alpha)$ . Else if  $a_n < \alpha$ , then  $\alpha < a_n + \frac{1}{2}\epsilon_n < \alpha + \frac{1}{2}\epsilon_n \leq r_n$ , hence  $x_n := a_n + \frac{1}{2}\epsilon_n \in I_n^0(\alpha)$  does the job. In the case where  $\alpha \geq m_n$  one argues similarly.

2. We follow the proof of Lemma 3.6. However, since the SLPs computing the approximation  $a_n$  are polynomial time uniform, only one bit of advice (corresponding to  $b_n$ ) is in fact needed to emulate the computation with  $\alpha$ .  $\square$

We have not been able to find a specific number that is *provably* non-approximable. It is quite possible that there are no non-approximable numbers at all.

## 4 PosSLP lies in CH

The counting hierarchy CH was defined by Wagner [61] and was studied further by Toran [59]; see also [7, 5]. A problem lies in CH if it lies in one of the classes in the sequence  $\text{PP}, \text{PP}^{\text{PP}}, \dots$ .

**Theorem 4.1** *BitSLP is in CH.*

*Proof.* It was shown by Hesse et al. [37] that there are Dlogtime-uniform threshold circuits of polynomial size and constant depth that compute the following function:

**Input** A number  $X$  in Chinese Remainder Representation. That is, a sequence of values indexed  $(p, j)$  giving the  $j$ -th bit of  $X \bmod p$ , for each prime  $p < n^2$ , where  $0 \leq X \leq 2^n$  (thus we view  $n$  as an appropriate “size” measure of the input).

**Output** The binary representation of the unique natural number  $X < \prod_{p \text{ prime}, p < n^2} p$  whose value modulo each small prime is encoded in the input.

Let this circuit family be denoted  $\{D_n\}$ .

Now, as in the proof of [5, Lemma 5], we consider the following exponentially-big circuit family  $\{E_n\}$ , that computes BitSLP.

Given as input an encoding of a straight-line program representing integer  $W$ , we first build a new program computing the positive integer  $X = W + 2^{2^n}$ . Note that the bits of the binary representation of  $W$  (including the sign bit) can easily be obtained from the bits of  $X$ .

Level 1 of the circuit  $E_n$  consists of gates labeled  $(p, j)$  for each prime  $p$  such that  $p < 2^{2^n}$  and for each  $j : 1 \leq j \leq \lceil \log p \rceil$ . The output of this gate records the  $j$ th bit of  $X \bmod p$ . (Observe that there are exponentially many gates on level 1, and also note that the output of each gate  $(p, j)$  can be computed in time polynomial in the size of the binary encoding of  $p$  and the size of the given straight-line program representing  $X$ . Note also that the gates on Level 1 correspond to the gates on the input level of the circuit  $D_{2^{2^n}}$ .)

The higher levels of the circuit are simply the gates of  $D_{2^{2^n}}$ .

Now, similar to the proof of [5, Lemma 5], we claim that for each constant  $d$ , the following language is in the counting hierarchy:  $L_d = \{(F, P, b) : F \text{ is the name of a gate on level } d \text{ of } E_n \text{ and } F \text{ evaluates to } b \text{ when given straight-line program } P \text{ as input}\}$ .

We have already observed that this is true when  $d = 1$ . For the inductive step, assume that  $L_d \in \text{CH}$ . Here is an algorithm to solve  $L_{d+1}$  using oracle access to  $L_d$ . On input  $(F, P, b)$ , we need to determine if the gate  $F$  is a gate of  $E_n$ , and if so, we need to determine if it evaluates to  $b$  on input  $P$ .  $F$  is a gate of  $E_n$  iff it is connected to some gate  $G$  such that, for some  $b'$ ,  $(G, P, b') \in L_d$ . This can be determined in  $\text{NP}^{L_d} \subseteq \text{PP}^{L_d}$ , since  $D_n$  is Dlogtime-uniform. That is, we can guess a gate  $G$ , check that  $G$  is connected to  $F$  (this takes only linear time because of the uniformity condition) and then use our oracle for  $L_d$ . If  $F$  is a gate of  $E_n$ , we need to determine if the majority of the gates that feed into it evaluate to 1. (Note that all of the gates in  $D_n$  are MAJORITY gates.) That is, we need to determine if it is the case that for most bit strings  $G$  such that  $G$  is the name of a gate that is connected to  $F$ ,  $(G, P, 1) \in L_d$ . This is clearly computable in  $\text{PP}^{L_d}$ .

Thus in order to compute BitSLP, given program  $P$  and index  $i$ , compute the name  $F$  of the output bit of  $E_n$  that produces the  $i$ th bit of  $N$  (which is easy because of the uniformity of the circuits  $D_{2^{2^n}}$ ) and determine if  $(F, P, 1) \in L_d$ , where  $d$  is determined by the depth of the constant-depth family of circuits presented in [37].  $\square$

Theorem 4.1 shows that  $\text{BP}(\text{P}_{\mathbb{R}}^0)$  lies in CH. A similar argument can be applied to an analogous restriction of “digital”  $\text{NP}_{\mathbb{R}}$  (i.e., where nondeterministic machines over the reals can guess “bits” but cannot guess arbitrary real numbers). Bürgisser and Cucker [22] present some problems in PSPACE that are related to *counting* problems over  $\mathbb{R}$ . It would be interesting to know if these problems lie in CH.

Although Theorem 4.1 shows that BitSLP and PosSLP both lie in CH, some additional effort is required in order to determine the level of CH where these problems reside. We present a more detailed analysis for PosSLP, since it is our main concern in this paper. (A similar analysis can be carried out for BitSLP, showing that it lies in  $\text{PH}^{\text{PP}^{\text{PP}^{\text{PP}^{\text{PP}}}}}$  [6].)

The following result implies Theorem 1.4, since Toda’s Theorem [58] shows that  $\text{PP}^{\text{PH}^A} \subseteq \text{P}^{\text{PP}^A}$  for every oracle  $A$ .

**Theorem 4.2**  $\text{PosSLP} \in \text{PH}^{\text{PP}^{\text{PP}}}$ .

*Proof.* We will use the Chinese remaindering algorithm of [37] to obtain our upper bound on PosSLP. (Related algorithms, which do not lead directly to the bound reported here, have been used on several occasions [2, 26, 31, 44, 45].) Let us introduce some notation relating to Chinese remaindering.

For  $n \in \mathbb{N}$  let  $M_n$  be the product of all odd primes  $p$  less than  $2^{n^2}$ . By the prime number theorem,  $2^{2^n} < M_n < 2^{2^{n^2}+1}$  for  $n$  sufficiently large. For such primes  $p$  let  $h_{p,n}$  denote the inverse of  $M_n/p \bmod p$ .

Any integer  $0 \leq X < M_n$  can be represented uniquely as a list  $(x_p)$ , where  $p$  runs over the odd primes  $p < 2^{n^2}$  and  $x_p = X \bmod p$ . Moreover,  $X$  is congruent to  $\sum_p x_p h_{p,n} M_n / p$  modulo  $M_n$ . Hence  $X/M_n$  is the fractional part of  $\sum_p x_p h_{p,n} / p$ .

Define the family of approximation functions  $app_n(X)$  to be  $\sum_p B_p$ , where  $B_p = x_p h_{p,n} \sigma_{p,n}$  and  $\sigma_{p,n}$  is the result of truncating the binary expansion of  $1/p$  after  $2^{n^4}$  bits. Note that for  $n$  sufficiently large and  $X < M_n$ ,  $app_n(X)$  is within  $2^{-2^{n^3}}$  of  $X/M_n$ .

Let the input to PosSLP be a program  $P$  of size  $n$  representing the integer  $W$  and put  $Y_n = 2^{2^n}$ . Since  $|W| \leq Y_n$ , the number  $X := W + Y_n$  is nonnegative and we can easily transform  $P$  into a program of size  $2n + 2$  representing  $X$ . Clearly,  $W > 0$  iff  $X > Y_n$ . Note that if  $X > Y_n$ , then  $X/M_n$  and  $Y_n/M_n$  differ by at least  $1/M_n > 2^{-2^{n^2+1}}$ , which implies that it is enough to compare the binary expansions of  $app_n(X)$  and  $app_n(Y_n)$ . (Interestingly, this seems to be somewhat easier than computing the bits of  $X$  directly.)

We can determine if  $X > Y_n$  in PH relative to the following oracle:  $A = \{(P, j, b, 1^n) : \text{the } j\text{-th bit of the binary expansion of } app_n(X) \text{ is } b, \text{ where } X \text{ is the number represented by straight-line program } P \text{ and } j \text{ is given in binary}\}$ . Lemma 4.3 completes the proof by showing that  $A \in \text{PH}^{\text{PP}^{\text{PP}}}$ .  $\square$

**Lemma 4.3**  $A \in \text{PH}^{\text{PP}^{\text{PP}}}$ .

*Proof.* Assume for the moment that we can show that  $B \in \text{PH}^{\text{PP}}$ , where  $B := \{(P, j, b, p, 1^n) : \text{the } j\text{-th bit of the binary expansion of } B_p (= x_p h_{p,n} \sigma_{p,n}) \text{ is } b, \text{ where } p < 2^{n^2} \text{ is an odd prime, } x_p = X \bmod p, X \text{ is the number represented by the straight-line program } P, \text{ and } j \text{ is given in binary}\}$ . In order to recognize the set  $A$ , it clearly suffices to compute  $2^{n^4}$  bits of the binary representation of the sum of the numbers  $B_p$ . A uniform circuit family for iterated sum is presented by Maciel and Thérien in [46, Corollary 3.4.2] consisting of MAJORITY gates on the bottom (input) level, with three levels of AND and OR gates above. As in the proof of Theorem 4.1, the construction of Maciel and Thérien immediately yields a  $\text{PH}^{\text{PP}^B}$  algorithm for  $A$ , by simulating the MAJORITY gates by  $\text{PP}^B$  computation, simulating the OR gates above the MAJORITY gates by  $\text{NP}^{\text{PP}^B}$  computation, etc. The claim follows, since by Toda's Theorem [58]  $\text{PH}^{\text{PP}^B} \subseteq \text{PH}^{\text{PP}^{\text{PP}^B}} = \text{PH}^{\text{PP}^{\text{PP}}}$ . It remains only to show that  $B \in \text{PH}^{\text{PP}}$ .  $\square$

**Lemma 4.4**  $B \in \text{PH}^{\text{PP}}$ .

*Proof.* Observe that given  $(P, j, b, p)$  we can determine in polynomial time if  $p$  is prime [3], and we can compute  $x_p$ .

In  $\text{PH} \subseteq \text{P}^{\text{PP}}$  we can find the least generator  $g_p$  of the multiplicative group of the integers mod  $p$ . The set  $C = \{(q, g_p, i, p) : p \neq q \text{ are primes and } i \text{ is the least number for which } g_p^i \equiv q \pmod{p}\}$  is easily seen to lie in PH. We can compute the discrete log base  $g_p$  of the number  $M_n/p \bmod p$  in  $\#\text{P}^C \subseteq \text{P}^{\text{PP}}$ , by the algorithm that nondeterministically guesses  $q$  and  $i$ , verifies that  $(q, g_p, i, p) \in C$ , and if so generates  $i$  accepting paths. Thus we can compute the number  $M_n/p \bmod p$  itself in  $\text{P}^{\text{PP}}$  by first computing its discrete log, and then computing  $g_p$  to that power, mod  $p$ . The inverse  $h_{p,n}$  is now easy to compute in  $\text{P}^{\text{PP}}$ , by finding the inverse of  $M_n/p \bmod p$ .

Our goal is to compute the  $j$ -th bit of the binary expansion of  $x_p h_{p,n} \sigma_{p,n}$ . We have already computed  $x_p$  and  $h_{p,n}$  in  $\text{P}^{\text{PP}}$ , so it is easy to compute  $x_p h_{p,n}$ . The  $j$ th bit of  $1/p$  is 1 iff  $2^j \bmod p$  is odd, so bits of  $\sigma_{p,n}$  are easy to compute in polynomial time. (Note that  $j$  is exponentially large.)

Thus our task is to obtain the  $j$ -th bit of the product of  $x_p h_{p,n}$  and  $\sigma_{p,n}$ , or (equivalently) adding  $\sigma_{p,n}$  to itself  $x_p h_{p,n}$  times. The problem of adding  $\log^{O(1)} n$  many  $n$ -bit numbers lies in uniform  $\text{AC}^0$  [30]. Simulating these  $\text{AC}^0$  circuits leads to the desired  $\text{PH}^{\text{PP}}$  algorithm for  $B$ .  $\square$

## 5 An Observation on Derandomizing ACIT

The connections between algebraic complexity and the counting hierarchy in the preceding section were first introduced in an earlier version of this paper [4]. Recently, these connections have led to further developments. Bürgisser shows in [21] that the counting hierarchy provides a useful tool for showing implications among several hypotheses in algebraic complexity theory that were not previously known to be related. In that same paper, he also improves a theorem of Koïran, relating the

arithmetic circuit complexity of the permanent to a frequently-studied question about the complexity of expressing  $n!$ . We have some new observations to present on this topic, and start by recalling some background and definitions.

We will follow the terminology of Shub and Smale [55], and say that  $n!$  is “easy” if there is a sequence of SLPs  $C_n$  of size  $\log^{O(1)} n$ , where  $C_n$  represents the number  $n!$ . Following the same convention, we say that  $n!$  is “ultimately easy” if there is a sequence of SLPs  $C_n$  of size  $\log^{O(1)} n$ , where  $C_n$  represents a nonzero multiple of the number  $n!$ . (It does not matter which multiple is represented.) Shub and Smale conjectured that  $n!$  is not ultimately easy, and they showed that this condition implies that  $P_C \neq NP_C$ . It is also pointed out in [12] that if factoring is sufficiently hard to compute, it implies that  $n!$  is not easy.

Note that if  $n!$  is not ultimately easy, it says merely that there are *infinitely many*  $n$  for which multiples of  $n!$  require large circuits. It may be useful also to consider the hypothesis that this condition holds for *all* large  $n$ : that is, for all  $k$  there is an  $m$  such that for all  $n > m$ , there is no SLP of size  $\log^k n$  representing a nonzero multiple of  $n!$ . Let us call this condition “ $n!$  is ultimately hard”.

The following implications are known to hold:

$$\begin{aligned} n! \text{ is ultimately hard} &\Rightarrow n! \text{ is not ultimately easy} \Rightarrow n! \text{ is not easy} \\ &\Rightarrow \text{the permanent requires arithmetic circuits of superpolynomial size} \Rightarrow \text{AFIT} \in \bigcap_{\epsilon > 0} \text{io-}[\text{DTIME}(2^{n^\epsilon})], \end{aligned}$$

where AFIT denotes Arithmetic Formula Identity Testing: a special case of ACIT. The third implication is from [21], the fourth is from [39, Theorem 7.7]. Derandomization results such as those of [39] usually come in two flavors. If one assumes that a particular function (such as the permanent) is hard on infinitely many input lengths, then one obtains only algorithms that work correctly on infinitely many input lengths. One can also obtain an algorithm that works correctly on all input lengths, if one starts with a stronger assumption, such as that the permanent requires large circuits on all input lengths.

It has not been known whether any of these hypotheses are sufficiently strong to derandomize ACIT itself, although it is known that if ACIT is in  $\bigcap_{\epsilon > 0} \text{DTIME}(2^{n^\epsilon})$  (or even in  $\bigcap_{\epsilon > 0} \text{NTIME}(2^{n^\epsilon})$ ), then either the permanent requires arithmetic circuits of superpolynomial size, or  $\text{NEXP} \not\subseteq P/\text{poly}$  [39]. We observe now that the following implication holds.

**Theorem 5.1** *We have the following:*

1. *If  $n!$  is ultimately hard, then  $\text{ACIT} \in \bigcap_{\epsilon > 0} \text{DTIME}(2^{n^\epsilon})$ .*
2. *If  $n!$  is not ultimately easy, then  $\text{ACIT} \in \bigcap_{\epsilon > 0} \text{io-}[\text{DTIME}(2^{n^\epsilon})]$ .*

*Proof.* We prove only the second claim. The first is easier, and follows by the same method.

First note that by Proposition 2.2, it is sufficient to prove the implication for EquSLP instead of ACIT. Assume that  $n!$  is not ultimately easy. Then for every  $k$ , there is an infinite set  $I(k)$  of numbers such that for all  $m \in I(k)$  no SLP of size at most  $\log^k m$  can produce a nonzero multiple of  $m!$ .

Given  $\epsilon > 0$ , pick any  $\gamma$  such that  $0 < \gamma < \epsilon$ . Choose  $k$  to be the least integer larger than  $1/\gamma$ . For any  $m \in I(k)$  put  $n = \lfloor \log^k m \rfloor$ .

Suppose we are given as input an SLP  $C$  of size  $n$ . Note that the binary encoding of  $m$  has length at most  $n^{1/k}$  – but we do not know what  $m$  is. Thus we try all numbers  $z$  having binary encoding of length at most  $n^{1/k}$  (one of which will be  $m$ ). We then compute the binary representation of  $z!$  with the obvious algorithm, which takes time at most  $z^2 \log^{O(1)} z$ , which is less than  $2^{n^\gamma}$  for sufficiently large  $n$ . Then we evaluate the SLP  $C$  modulo  $z!$ ; we accept iff the result is zero for all of the numbers  $z$ . This algorithm works correctly, since by our assumption, the SLP  $C$  cannot produce a nonzero multiple of  $m!$ . The running time is  $2^{O(n^\gamma)} 2^{O(n^\gamma)}$ , which is less than  $2^{n^\epsilon}$  for all large  $n$ .  $\square$

## 6 Closing Remarks

NP-hardness is firmly established as a useful tool for providing evidence of intractibility. We believe that PosSLP can become a useful tool for providing evidence of intractibility for problems that do not appear to be NP-hard, and for providing evidence that certain problems do not lie in NP. Indeed, results of this flavor have already started to appear: Etesami and Yannakakis have recently shown that PosSLP reduces to the problem of computing Nash equilibria for three-person games [33].

There are several directions for further research suggested by the results that we have presented. We would very much like to see a resolution of our Conjecture 1.2, and we think that it is likely that that PosSLP lies at a lower level of the counting

hierarchy than is proved in Theorem 1.4. Perhaps better upper bounds can be presented at least for the sum-of-square-roots problem. Can better evidence be presented for the intractibility of PosSLP? Can some important problems in  $\text{NP}_{\mathbb{R}}$  (such as the existential theory of the reals) be shown to lie in the counting hierarchy?

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