Optimization of Bootstrapping in Circuits

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Abstract

In 2009, Gentry proposed the first Fully Homomorphic Encryption (FHE) scheme, an extremely powerful cryptographic primitive that enables to perform computations, i.e., to evaluate circuits, on encrypted data without decrypting them first. This has many applications, particularly in cloud computing.

In all currently known FHE schemes, encryptions are associated with some (non-negative integer) noise level. At each evaluation of an AND gate, this noise level increases. This increase is problematic because decryption succeeds only if the noise level stays below some maximum level L at every gate of the circuit. To ensure that property, it is possible to perform an operation called bootstrapping to reduce the noise level. Though critical, boostrapping is a time-consuming operation. This expense motivates a new problem in discrete optimization: minimizing the number of bootstrappings in a circuit while still controlling the noise level.

In this paper, we (1) formally define the bootstrap problem, (2) design a polynomial-time L-approximation algorithm using a novel method of rounding of a linear program, and (3) show a matching hardness result: $(L - \varepsilon)$ -inapproximability for any $\varepsilon > 0$.

Keywords. DAG, circuit, approximation algorithms, inapproximability, linear programming, rounding, fully homomorphic encryption (FHE)

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1 Introduction

Imagine evaluating a circuit with noise: at each gate, the noise level may increase due to the computation. Now, imagine that you can occasionally perform a computationally expensive operation (called *bootstrapping*) on the output of a gate to reduce the noise level. Given a circuit, at which gates should you apply the bootstrapping operation to the output of the gate to keep the maximum noise level within a certain tolerance level and minimize the number of bootstrappings?

For example, if the noise level at an input gate equals 0 and the noise level at a gate with two direct predecessors u and v equals $\max(\text{noiselevel}(u), \text{noiselevel}(v)) + 1$, then in the circuit in Fig. 1, it is possible to maintain a maximum noise level of at most L=3 by doing 2 bootstrappings; that is optimal.

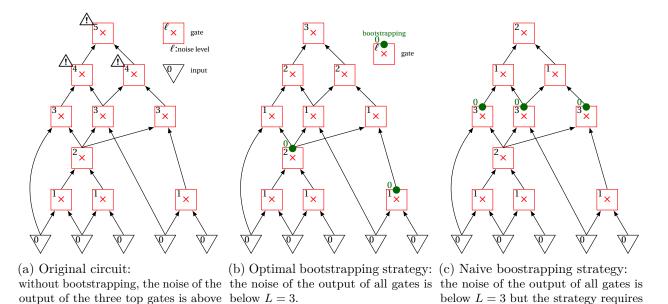


Figure 1: Example of circuit and of two bootstrapping strategies

3 bootstrappings (non-optimal).

1.1 Motivation

L=3 (warning sign).

This problem arises in cryptography in the context of fully homomorphic encryption (FHE) [2, 4-6, 8, 10-15, 17, 18, 26, 33]. A fully homomorphic encryption scheme enables one to encrypt bits and keep them confidential, while allowing anyone who is given an encryption E(a) of a bit a and an encryption E(b) of a bit a to publicly compute a (not a), a (a xor a), and a (a and a). Such a scheme permits secure computation of any binary circuit over encrypted bits. This primitive has tremendous potential for applications, including the canonical example of outsourcing computation to a remote server without compromising one's privacy. Concrete application examples include biometric identification, statistics over encrypted data [3, 27], machine learning [20], and private genomic analyses [24].

All existing instantiations of FHE follow the same blueprint [13]: ciphertexts (i.e., encryptions of bits) contain some "noise" that grows during the circuit evaluation. To ensure correctness at

decryption time, one has to regularly perform *bootstrapping* operations on the ciphertexts in an effort to lessen the noise.¹ Unfortunately, since such operations are very expensive in practice (see, e.g., [8, 12, 16, 22, 28, 31]), we ask the question:

What is a minimum set of ciphertexts that must be bootstrapped in order to correctly evaluate the circuit?

This is called the *bootstrap problem*.

In all efficient implementations of FHE schemes [4,5,9,12,17,21,22,26,31], non-linear gates (and) introduce much more additional noise than linear gates (not and xor), hence we use a simplified model where evaluation of linear gates do not increase the noise; see, e.g., [1,7,21,25,30].

Formally, each ciphertext has an associated "noise level" $\ell \in \mathbb{Z}_{\geq 0}$, and:

- Evaluating a not gate over a ciphertext does not change its noise level.
- Evaluating an xor gate yields a ciphertext whose noise level $\max(\cdot, \cdot)$ is the maximum noise level of its inputs.
- Evaluating a non-linear gates (and) yields a ciphertext with increased noise level $\max(\cdot,\cdot) + 1$.

Noise behaviors of other FHE schemes are discussed later.

To ensure that the circuit evaluation is correct, the FHE scheme has a parameter $L \ge 1$, which is independent of the circuit size, and requires that all ciphertexts must have their noise levels less than or equal to L at all gates of the circuit under evaluation. This requires performing a bootstrapping operation on the output of some gates of the circuit to reduce the noise level of the ciphertext to 0.

The first instantiations of FHE were for L=1 [8,10,11,16]. Most of them merely perform a bootstrapping operation right after (see, e.g., [16]) or right before (see, e.g., [17]) each and gate evaluation. However, this can be computationally wasteful, since fewer bootstrappings may be sufficient to evaluate the whole circuit when positioned more carefully [25,30]; see also Fig. 1 (in this figure, square gates correspond to and gates).

Lepoint and Paillier [25] modeled the problem of constructively computing the exact minimum number of bootstrappings for any $L \ge 1$ based on Boolean satisfiability. They associated a Boolean to each ciphertext during the circuit evaluation. The Boolean is equal to true when the ciphertext should be bootstrapped. Using the logic circuit and the noise level constraints, the authors constructed a Boolean monotone predicate ϕ which captures the correctness of the circuit evaluation. They then described a heuristic method to recover the smallest prime implicant of ϕ , which directly yields the minimum number of bootstrappings and the ciphertexts to be bootstrapped. However, [25] claims neither a complexity analysis nor a hardness result.²

Later, Paindavoine and Vialla [30] showed that, for L=1, the bootstrap problem can be solved in polynomial time by a reduction to (s,t)-min-cut; and that, for $L \ge 2$, the bootstrap problem is NP-hard by a reduction from the vertex cover problem.³ They also provided experimental results on real-world circuits—integer addition, integer multiplication, and some cryptographic primitives—based on mixed integer linear programming.⁴

¹An upper bound on the admissible noise is given as part of the parameters of the FHE scheme.

²Lepoint and Paillier only indicated that for Boolean monotone predicates, finding the size of the smallest prime implicant is known to be #P-complete [19, Section 6].

³We note that in the terminology of [30], $l_{max} = L + 1$ since the minimum noise level in their model is 1.

⁴Their linear programming relaxation is different from the one in this paper.

1.2 Problem Formulation

In graph theory terms, the bootstrap problem can be formulated as follows: the input is a positive integer L and a directed acyclic graph (DAG) G = (V, E) whose vertices all have indegree 0 or 2, with colors on the vertices (vertices of indegree 0 are white and vertices of indegree 2 are either blue or red). G may have parallel edges. A feasible solution is a subset $S \subseteq V$ of marked vertices such that $\max_{u \in V} \ell(u) \leq L$, where the function $\ell(\cdot)$ is computed recursively as follows:⁵

$$\ell(v) = \begin{cases} 0 & \text{if } v \text{ is white,} \\ \max_{(u,v) \in E} \ell(u) \cdot \mathbbm{1}_{V \setminus S}(u) & \text{if } v \text{ is blue,} \\ \max_{(u,v) \in E} \ell(u) \cdot \mathbbm{1}_{V \setminus S}(u) + 1 & \text{if } v \text{ is red.} \end{cases}$$

The goal of the bootstrap problem is to find a feasible solution S of minimum cardinality.

If S = V, then $\ell(v) \leq 1$ for every vertex v, so this solution is always feasible. If $S = \emptyset$, then $\ell(v)$ is the maximum number of red vertices on any path ending at v, so this solution is feasible if and only if there does not exist a path in G containing L + 1 red vertices.

In terms of the problem we have been discussing, the DAG is a binary or an arithmetic circuit, the white vertices are the input variables, the blue vertices are the xor (or addition) gates, the red vertices are the and (or multiplication) gates, S is the set of ciphertexts that are bootstrapped during the computation, $\ell(\cdot)$ is the noise level, and L is the maximum allowed noise level.⁶

A feasible solution $S' \subseteq V$ is an α -approximate solution (for $\alpha \ge 1$) if $|S'| \le \alpha \cdot \text{OPT}$, where OPT denotes the minimum cardinality of a feasible solution.

1.3 Results

We characterize the complexity of the bootstrap problem by providing a polynomial-time L-approximation algorithm (Theorem 1) and (assuming the Unique Games Conjecture) showing that L is the best achievable approximation factor (Theorem 2).

Theorem 1 (approximation algorithm). Let $L \ge 1$ be an integer parameter. There is a deterministic polynomial-time approximation algorithm for the bootstrap problem within approximation factor L.

The proof of Theorem 1 is in Section 2.

Theorem 2 (hardness of approximation). Let $L \ge 2$ be an integer parameter. For any $\epsilon > 0$, it is NP-hard to approximate the bootstrap problem within a factor of $L - \epsilon$, assuming the Unique Games Conjecture.

The proof of Theorem 2 is in Section 3.

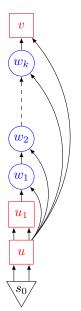
To design the approximation algorithm used in Theorem 2, we first observe that a set of marked vertices is a feasible solution if and only if, for every path $p = v_1 \dots v_k$ that starts and ends at red vertices and that traverses L + 1 red vertices (including endpoints), at least one vertex among v_1, \dots, v_{k-1} is marked. Such path p is called an *interesting path*.

⁵The indicator function $\mathbb{1}_{V\setminus S}$ has value 1 on $V\setminus S$ and 0 on S.

⁶Without loss of generality, we assume there are no **not** gates, since they do not influence the noise level.

Based on this observation, our algorithm starts by solving a linear program relaxation with one constraint for each interesting path and obtains, for every $v \in V$, a value $x_v \in [0, 1]$ that indicates whether the vertex v should be bootstrapped. The challenging part of the algorithm is the rounding.

In a naive attempt to do the rounding, we define $\delta(u,v)$ as the u-to-v distance in the metric induced by $\{x_v\}$. To ensure that every interesting path from u to v contains a marked vertex, we choose a value $t \in [0, \delta(u, v)]$ (randomly or according to some rules such as in the region growing technique [34]), and then mark a vertex $w \in V$ if and only if $\delta(u, w) \leq t \leq \delta(u, w) + x_w$. This approach does not yield a good approximation because $\delta(u, v)$ might be very small or even zero (see Fig. 2), as there might exist short non-interesting u-to-v paths. Hence, the major difference between the bootstrap problem and classical cut problems (e.g., min-cut, multi-cut, multi-terminal cut) is that in the bootstrap problem, for each pair of vertices (u, v), we only want to "cut" the interesting u-to-v paths, but the non-interesting u-to-v paths may remain.



Graph with 4 + k vertices. L = 2.

The only interesting path is: $uu_1w_1w_2...w_kv$.

Circle vertices are blue, square vertices are red, and the triangle vertex is white.

Suppose that the fractional solution obtained from the linear program is the following:

$$\begin{cases} x_{w_i} = \frac{1}{k} & \text{for } i = 1, \dots, k; \\ x_u = x_{u_1} = x_v = 0. \end{cases}$$

Then $\delta(u, v) = 0$.

Figure 2: Example of circuit for which naive rounding does not work

Another attempt is to apply *iterative rounding* [23,34]. However, this does not seem to help for the bootstrap problem, mainly because the family of interesting paths is not closed under union, intersection, and difference.

Instead, our approach to rounding separates paths according to the number of red vertices which they traverse and defines a function f_i with respect to all paths that traverses exactly i red vertices. This is a main idea of the algorithm. We perform a rounding for each function separately and obtain L sets of marked vertices. By taking the union of these sets, we obtain a feasible solution of cardinality at most $L \cdot \text{OPT}$.

Remark. When L=1, the output of algorithm in Theorem 1 is optimal.

To prove the lower bound of Theorem 2, we first look at a related problem called the *DAG Vertex Deletion (DVD)* problem [29,32]. In the DVD problem, we are given a directed acyclic graph H and an integer $L \ge 2$ and we want to delete the minimum number of vertices so that the resulting

graph has no path containing L vertices. Svensson [32] showed that if we assume the Unique Games (UG) Conjecture, approximating DVD within an $L - \epsilon$ factor is NP-hard.

To show the UG-hardness of approximating the bootstrap problem, we provide an approximation-preserving reduction from the DVD problem to the bootstrap problem.

1.4 Discussion of the Model

Necessity of bootstrappings. To date, the bootstrapping paradigm is the only known way of obtaining an unbounded FHE scheme, i.e., one that can homomorphically evaluate any efficient function using constant-size keys and ciphertexts. Therefore, to exploit the full potential of fully homomorphic encryption, one must resort to bootstrapping.

Noise levels. The bootstrap problem is a simplification of the behavior of FHE schemes, since the noise grows in practice in a more complex manner. Indeed, all encryption procedures in FHE schemes consist of adding a short noise (a bit, or more generally, an integer) to the encoding of the message. Since the noise is added to the encoding, and computing an xor gate homomorphically essentially corresponds to adding the ciphertexts and thus adding the corresponding noises, on a logarithmic scale the amount of noise remains approximately as large as the maximum input noise (up to one bit). On the other hand, computing an and gate requires a multiplication of the ciphertexts, and makes the noise growth noticeably larger [21]. This is why the cryptographic community introduced the simplified model of Section 1.1 and started building circuits for which the noise does not increase too much in this model [1,7].

In this paper, we say that the noise level of a ciphertext is in [0, L], where L is the parameter of the FHE scheme. In previous works, the level was either in $[1, \ell_{\text{max}}]$ [25, 30], or in $[9, \ell_{\text{max}}]$ [21], where $\ell_{\text{max}} = L + 1$. This is equivalent: $\ell = 0$ should not be interpreted to mean that a ciphertext is noise-free, but that the amount of noise contained in the ciphertext results from a bootstrapping operation.

Other noise behaviors. Though the noise model in Section 1.1 corresponds to the family of FHE schemes that are the most efficient in practice, there exist other families of FHE schemes.

One family corresponds to the first implementations that were proposed [8,11,16]. Therein, non-linear gates behave as $\cdot + \cdot$ for the noise levels. Hence, the noise growth is exponential with the multiplicative depth of the circuit and these schemes will never be used in practice. In addition, to get reasonable parameters, the proof-of-concept implementations set L=1 (compared to, e.g., L=41 in the HElib implementation [22]). In this particular case, the noise model is actually equivalent to the one we are considering (when L=1).

Another family is the one of the GSW scheme [6, 18]. Variants of this scheme have been implemented [9, 12]. These implementations have a faster wall-clock time for bootstrapping than [8, 22], but support neither large message spaces nor vector messages and thus have larger amortized per-bit timing. The noise behavior is slightly different there: it is asymmetric (i.e., the order of the inputs matters). Modeling the noise behavior for these schemes, and extending our results within this new model, are left as interesting open problems.

⁷The 9 comes from the fact that a "fresh" ciphertext (i.e., an unprocessed encryption of a bit) is said to have noise level 1, and that after a bootstrapping, the resulting ciphertext has a noise level 9.

⁸For this model, the lowest noise level must be set to 1 instead of 0, and the maximum allowed noise is L+1 instead of L.

Computing model. In the bootstrap problem, we minimize the total number of bootstrappings (i.e., marked vertices), thus accounting for classical sequential complexity. We could also consider a parallel computing model in which any number of bootstrappings performed in parallel will cost the same as a single bootstrapping. This might be relevant in some Cloud-based scenarios where the user encrypting the data has an unbounded amount of money and only want to minimize the time to get the result of the circuit evaluation over the encrypted data. However, the financial cost would basically be proportional to the total number of bootstrappings. Furthermore, we remark that this parallel version of the bootstrap problem has a trivial solution: topologically sorting the DAG and greedily marking the vertices with noise level greater than L.

1.5 Other Related Work

The DVD problem (see Section 1.3) was introduced by Paik, Reddy, and Sahni in [29] in the context of certain VLSI design and communication problems. Svensson showed that the DVD problem is UG-hard [32]. His work was mainly motivated by the classical *Discrete Time-Cost Tradeoff Problem* in the completely different setting of *Project Scheduling*.

2 Approximation Algorithm

To prove Theorem 1, we give a randomized algorithm (Algorithm 1) in Section 2.1, analyze it in Section 2.2, and derandomize it in Section 2.3.

2.1 Algorithm

For a path $p = v_1 \dots v_k$, the vertex v_k is called the *final vertex* of p and the vertices v_1, \dots, v_{k-1} are called the *non-final vertices* of p.

The following fact is used throughout the paper.

Fact 3. A set of marked vertices is a feasible solution if and only if every path that starts and ends at a red vertex and that contains exactly L+1 red vertices (including endpoints) has a non-final vertex that is marked.

Proof. We first observe that a set S of marked vertices is a feasible solution if and only if, for all red vertices $u \in V$, the noise level $\ell(u)$ with respect to S is at most L. This is because a white vertex has noise level 0 and a blue vertex has noise level not exceeding those of its direct predecessors. We further observe that, for each red vertex $u \in V$, the level $\ell(u)$ is the maximum number of red vertices on $any\ v$ -to-u path (for some red vertex v) that does not contain any marked vertices (except u if appropriate). Thus, $\ell(u) \leqslant L$ if and only if every path that starts at a red vertex, ends at u, and contains exactly L+1 red vertices (including endpoints) has a non-final vertex that is marked. Therefore, all red vertices $u \in V$ are such that $\ell(u) \leqslant L$ if and only if every path that starts and ends at a red vertex and that contains exactly L+1 red vertices (including endpoints) has a non-final vertex that is marked.

This fact leads to the definition of *interesting* paths.

Definition 4 (interesting path). A path in G is called *interesting* if it starts and ends at red vertices, and traverses exactly L+1 red vertices (including endpoints). For a given vertex $v \in V$

Algorithm 1 Approximation algorithm for the bootstrap problem

1: Solve the following LP relaxation, where for each vertex v we have one variable x_v , which represents whether there is a bootstrapping on v, and one constraint for each interesting path p.

$$\min \sum_{v \in V} x_v$$
s. t.
$$\sum_{\substack{\text{non-final } \\ \text{vertex } v \text{ of } p}} x_v \geqslant 1 \qquad \forall \text{ interesting path } p$$

$$0 \leq x_v \leq 1 \qquad \forall v \in V$$

2: For every red vertex u and blue vertex v, compute

$$\delta(u, v) = \min \{ x_u + x_{v_2} + \dots + x_{v_{k-1}} : \text{ path } p = uv_2 \dots v_{k-1}v \text{ such that } v_2, \dots, v_{k-1} \text{ are blue} \},$$

using a classical shortest path algorithm. By convention $\delta(u,v) := \infty$ if no such path exists.

3: For every vertex v and integer $i \in \{1, \dots, L+1\}$, compute

$$f_i(v) = \min \{x_{v_1} + x_{v_2} + \dots + x_{v_{k-1}} : \text{ path } p = v_1 v_2 \dots v_{k-1} v \text{ is } (v, i) \text{-interesting} \}$$

using the side table δ and a dynamic program (see Section 2.1). By convention $f_i(v) := \infty$ if no such path exists.

4: Rounding: Pick a uniformly random value $t \in [0, 1]$; A vertex v is marked if and only if there exists $i \in \{1, ..., L\}$ s.t. $t \in [f_i(v), f_i(v) + x_v]$.

and a given level $i \in \{1, ..., L+1\}$, a path in G is called (v, i)-interesting if it starts at a red vertex, ends at v, and traverses exactly i red vertices (including endpoints, if appropriate).

We associate to each vertex $v \in V$ a non-negative weight x_v . In our algorithm, these weights come from a solution of a linear program (LP). These weights induce a metric. More formally, we define the following notion of length.

Definition 5 (length). Let p be a path in G. We define the length len(p) of p as the sum of the weights x_v of all the non-final vertices v of p. We define $f_i(v)$ as the minimum length of a (v, i)-interesting path.

We remark that, for any red vertex $v \in V$, a (v, L+1)-interesting path is an interesting path. To compute $\{f_i(v)\}_{v,i}$, we use the following dynamic program, which proceeds in phases corresponding to $i = 1, \ldots, L+1$.

• For the base case
$$i=1$$
: $f_1(v)=\begin{cases} \infty & \text{if } v \text{ is white,} \\ 0 & \text{if } v \text{ is red,} \\ \min\limits_{u \text{ red}} \delta(u,v) & \text{if } v \text{ is blue.} \end{cases}$

 $^{{}^{9}}f_{i}(v) := \infty$ if there is no (v, i)-interesting path.

• For
$$i \in \{2, \dots, L+1\}$$
: $f_i(v) = \begin{cases} \infty & \text{if } v \text{ is white,} \\ \min\limits_{(u,v) \in E} (f_{i-1}(u) + x_u) & \text{if } v \text{ is red,} \\ \min\limits_{u \text{ red}} (f_i(u) + \delta(u,v)) & \text{if } v \text{ is blue.} \end{cases}$

2.2 Analysis

We now prove that the output of Algorithm 1 is a feasible solution (correctness property) and has cardinality at most $L \cdot \text{OPT}$ (approximation factor L). We then show that Algorithm 1 runs in polynomial time.

Analysis of Correctness. Consider an interesting path $p = v_1 \dots v_k$.

Lemma 6. Let $p = v_1 \dots v_k$ be an interesting path. For every $j \in \{1, \dots, k\}$, let $i_j \in \mathbb{N}$ denote the number of red vertices on the subpath $v_1 \dots v_j$ of p. Then, for any $t \in [0,1]$, there exists $j \in \{1, \dots, k-1\}$ such that $t \in [f(v_j, i_j), f(v_j, i_j) + x_{v_j}]$.

Applying Lemma 6, and using the fact that $i_j \in \{1, ..., L\}$ for every $j \in \{1, ..., k-1\}$, we see that the algorithm marks at least one non-final vertex of p, and so by Fact 3 the output is a feasible solution, proving correctness.

Proof. (Proof of Lemma 6) By definition of interesting paths, the sequence $\{i_j\}_j$ is non-decreasing, $i_1 = 1$, $i_{k-1} = L$, and $i_k = L + 1$. It is sufficient to show that the interval [0,1] is contained in the union of the intervals $[f(v_j, i_j), f(v_j, i_j) + x_{v_j}]$ over all $j \in \{1, \ldots, k-1\}$, which is a direct consequence of the three following properties (see Fig. 3):

- 1. $f(v_1, i_1) = 0$;
- 2. for every $j \in \{1, \dots, k-1\}$, $f(v_{j+1}, i_{j+1}) \leq f(v_j, i_j) + x_{v_j}$;
- 3. $f(v_k, i_k) \ge 1$.

The first property follows directly from the definition of f since v_1 is red and $i_1 = 1$.

To show the second property, for any $j \in \{1, \ldots, k-1\}$, consider a (v_j, i_j) -interesting path p' that achieves the length $f(v_j, i_j)$. We observe that the concatenation of p' and v_{j+1} is a (v_{j+1}, i_{j+1}) -interesting path and it has length $f(v_j, i_j) + x_{v_j}$. From the definition of $f(v_{j+1}, i_{j+1})$, we have $f(v_{j+1}, i_{j+1}) \leq f(v_j, i_j) + x_{v_j}$.

To show the third property, consider a (v_k, i_k) -interesting path p' that achieves the length $f(v_k, i_k)$. Then p' is an interesting path since v_k is red and $i_k = L + 1$. (p' may differ from p though.) Therefore, the constraint on p' in the LP implies that $len(p') \ge 1$. Hence $f(v_k, i_k) = len(p') \ge 1$.

This concludes the proof. \Box

Analysis of Quality of Approximation. The expected value of the output is the expected number of marked vertices, $\sum_{v \in V} \Pr(v \text{ marked})$. Let $v \in V$. By the algorithm and a union bound:

$$\Pr(v \text{ marked}) = \Pr(\exists i \in \{1, \dots, L\} : t \in [f_i(v), f_i(v) + x_v]) \leqslant \sum_i \Pr(t \in [f_i(v), f_i(v) + x_v]).$$

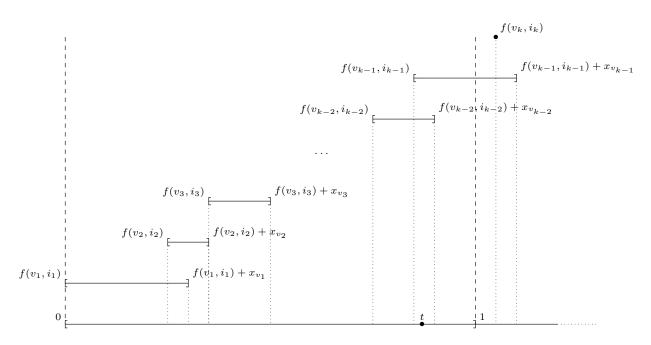


Figure 3: Illustration for the proof of Lemma 6

For t uniformly random in [0,1], the probability that $t \in [f_i(v), f_i(v) + x_v]$ is at most x_v . Thus $\Pr(v \text{ marked}) \leq Lx_v$ and the expected value of the output is at most $L(\sum_{v \in V} x_v)$. Since the linear program is a relaxation of the problem, $\sum_v x_v$ is less than or equal to the optimum value of the bootstrap problem, proving that the output is an L-approximation.

Analysis of Running Time. Clearly, computing $\{f_i(v)\}_{v,i}$ takes polynomial time. Next, we show that the LP in Step 1 of the algorithm can be solved in polynomial time (regardless of an exponential number of constraints). To that end, it is well known (see, e.g., [34]) that a polynomial-time separation oracle¹⁰ for this LP suffices.

To check whether an oracle input $\{x_v\}_v$ is a feasible solution to the LP, we compute $\{f_i(v)\}_{v,i}$ with respect to $\{x_v\}_v$ using the same dynamic program as before, in polynomial time. From the definition of f, we have that $\{x_v\}_v$ is a feasible solution if and only if $f_{L+1}(v) \ge 1$ for every red vertex $v \in V$. Suppose there is some red vertex $v \in V$ with $f_{L+1}(v) < 1$. Then, there must be an interesting path p with final vertex v such that the constraint on p is violated. It is easy to enrich the dynamic program in a standard manner, so that we obtain the entire path p. Thus, we complete the description of the polynomial-time separation oracle.

Therefore, the overall running time of Algorithm 1 is polynomial.

2.3 Derandomization

Algorithm 1 can be easily derandomized: $\{f_i(v)\}_{v,i} \cup \{f_i(v) + x_v\}_{v,i}$ contains at most $2|V| \cdot L$ different values, so they separate the [0,1] interval into at most $2|V| \cdot L + 1$ sub-intervals. We can

¹⁰A separation oracle takes as input a supposedly feasible solution to the linear program, and either verifies that it is indeed a feasible solution to the linear program or, if it is infeasible, produces a violated constraint.

enumerate one value t for each sub-interval, compute a feasible solution with respect to each value t, and finally return the best solution among them.

3 Hardness of Approximation

In this section, we prove Theorem 2. First, we recall the definition of the DAG Vertex Deletion (DVD) problem (see Section 1.3). In Lemma 8, we reduce the DVD problem to the bootstrap problem. The hardness of the bootstrap problem then follows from the hardness of the DVD problem [32].

Lemma 7 (Adapted from [32, Theorem 1.1]). Let $L \ge 2$ be an integer parameter. For any $\epsilon > 0$, it is NP-hard to approximate the DVD problem within a factor of $L - \epsilon$, assuming the Unique Games Conjecture.

Lemma 8. There is an approximation-preserving reduction from the DVD problem to the bootstrap problem.

Theorem 2 follows immediately from Lemmas 7 and 8. In the rest of the section, we prove Lemma 8. The proof is elementary, but delicate: in the bootstrap problem, vertices have indegree at most 2; in the DVD problem, vertices may have arbitrary indegree.

Consider a DVD instance with the DAG $H = (V_H, E_H)$ and the integer parameter $L \ge 2$. As a warm-up, let us first suppose that all vertices in V_H have indegree at most 2. We construct a DAG G = (V, E) for the bootstrap problem:

- We create a new vertex set V'_H (which can be viewed as a clone of V_H): for every $v \in V_H$, V'_H contains a new vertex v'. We also create a new vertex s_0 . The vertex set V of G is defined as $V := V_H \cup V'_H \cup \{s_0\}$. The vertices in $V_H \cup V'_H$ are red, and the vertex s_0 is white.
- The edge set E of G consists of all edges of E_H and the following edges: for every vertex $v \in V_H$, 2 copies of the edge (v, v') and (2 indegree(v)) copies of the edge (s_0, v) .

The following lemma implies that there is an approximation-preserving reduction.

Lemma 9. A feasible solution to the DVD problem for the instance (L, H) can be transformed into a feasible bootstrap solution for the instance (L, G) with at most the same cardinality, and vice versa.

Proof. Let S be a feasible solution to the DVD problem, i.e., every path of L vertices in H contains a vertex in S. We show that S is a feasible bootstrap solution for the instance (L, G). Using Fact 3, we only need to show that every interesting path in G contains a non-final vertex that is in S. Let $p = v_1 \dots v_k$ be an interesting path in G. We observe that v_1, \dots, v_k are all red vertices: v_1 is red by the definition of an interesting path, and v_j (for any $1 \le i \le k$) is red since it has positive indegree (and thus cannot be $1 \le i \le k$). By the definition of an interesting path, we know that $1 \le i \le k$ contains $1 \le i \le k$. It is in $1 \le i \le k$ positive outdegree (and thus cannot be in $1 \le i \le k$). Thus, $1 \le i \le k$ contains $1 \le i \le k$. Thus, $1 \le i \le k$ contains a non-final vertex that is in $1 \le i \le k$. Thus, $1 \le i \le k$ contains a non-final vertex that is in $1 \le i \le k$.

Conversely, let S be a feasible bootstrap solution. We show that $S \cap V_H$ is a feasible solution to the DVD problem. Let $p = v_1, \ldots, v_L$ be a path of L vertices in H. We only need to show that p

contains a vertex in $S \cap V_H$. We construct a path p' in G that is the concatenation of p and the vertex v'_L . We remark that p' starts and ends on red vertices and contains exactly L+1 red vertices. Therefore, it is an interesting path. From Fact 3, S contains a vertex u that is a non-final vertex of p', i.e., $u \in \{v_1, \ldots, v_L\}$, hence u is on p. Since $\{v_1, \ldots, v_L\} \subseteq V_H$, $u \in S \cap V_H$. Thus p contains a vertex in $S \cap V_H$.

Let us now deal with the general case where the vertices of H have arbitrary indegrees. First, we initialize the DAG G using the same transformation as before, except that G now contains (2 - indegree(v)) copies of an edge (s_0, v) only if the vertex v has indegree at most 2 (rather than for any vertex v). After this transformation, every red vertex in G has indegree at least 2. To transform G into a DAG for the bootstrap problem, we just need to deal with the red vertices with indegree at least 3. Let v be such a vertex. We observe that $v \in V_H$. Let v_1, \ldots, v_d be the direct successors of v in H. We remove from G the edges (v_i, v) (for each i) and add to G new blue vertices $w_1^{(v)}, \ldots, w_d^{(v)}$ and the following edges:

- 1. two copies of the edge $(v_1, w_1^{(v)})$, 2. for i = 2, ..., d, an edge $(w_{i-1}^{(v)}, w_i^{(v)})$ and an edge $(v_i, w_i^{(v)})$,
- 3. two copies of the edge $(w_d^{(v)}, v)$.

The transformation is depicted in Fig. 4. Let G = (V, E) be the final graph. We can verify that (L,G) is an instance of the bootstrap problem.

We show that Lemma 9 holds in this general setting. The transformation from a feasible DVD solution to a feasible bootstrap solution is a trivial extension from the previous proof. Let us now focus on the transformation from a feasible bootstrap solution to a feasible DVD solution. The following proposition is the key to the proof.

Proposition 10. Let S be a feasible bootstrap solution for (L,G) which contains a blue vertex $w_i^{(v)}$ for some $v \in V_H$ and some integer i. Then, $(S \setminus \{w_i^{(v)}\}) \cup \{v\}$ is also a feasible bootstrap solution for (L,G).

Proof. Let $S' = (S \setminus \{w_i^{(v)}\}) \cup \{v\}$. From Fact 3, we only need to prove that every interesting path contains a non-final vertex that is in S'. Since S is a feasible bootstrap solution, the only non-trivial part is to prove that every interesting path ending in v contains a non-final vertex that is in S'. Let $p = v_1 \dots v_k$ be such a path, and let j < k be the index of the last non-final red vertex of p (i.e., v_j is a red vertex and there is no red vertex among v_{j+1}, \ldots, v_{k-1}). Let p' be the path $v_1 \ldots v_j v_j'$. Since there are exactly L red vertices among v_1, \ldots, v_L and v_i' is red, p' is an interesting path. Thus some non-final vertex u of p' (i.e., $u \in \{v_1, \dots, v_i\}$) is in S. We observe that p' cannot contain $w_i^{(v)}$ since $w_i^{(v)} \in \{v_{i+1}, \dots, v_{k-1}\}$. Thus $u \in S'$ and is a non-final vertex of p.

Let S be a feasible bootstrap solution. We construct another feasible bootstrap solution S' with $|S'| \leq |S|$ such that S' only contains red and white vertices. As soon as S contains a blue vertex, let it be $w_i^{(v)}$ for some $v \in V_H$ and some integer i. We then replace the blue vertex $w_i^{(v)}$ in S with the red vertex v. Let S' be the final S. Then S' contains only red and white vertices and has cardinality at most |S|. From Proposition 10, S' is a feasible bootstrap solution.

We now show that $S' \cap V_H$ is a solution to the DVD problem using similar arguments as before. Let $p = v_1 \dots v_L$ be a path of L vertices in H. We only need to show that p contains a vertex in $S' \cap V_H$. We construct a (unique) path p' in G, which starts at v_1 , goes through v_2, \ldots, v_L , and ends at the red vertex v'_L . We remark that p' starts and ends on red vertices and contains exactly L+1 red vertices, namely $\{v_1,\ldots,v_L,v'_L\}$. Therefore, it is an interesting path. From Fact 3, S' contains a vertex u that is a non-final vertex of p'. Since S' contains only red and white vertices, we have $u \in \{v_1,\ldots,v_L\}$, hence u is on p. Since $\{v_1,\ldots,v_L\} \subseteq V_H$, $u \in S' \cap V_H$. Thus p contains a vertex in $S' \cap V_H$.

This concludes the proof of Lemma 8.

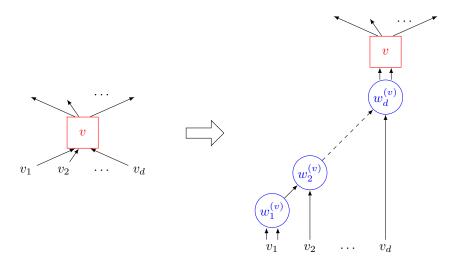


Figure 4: Reducing of the Indegree of a Vertex v (in the Proof of Lemma 8). Circle vertices are blue, square vertices are red, v_1, \ldots, v_d are the direct predecessors of $v, w_1^{(v)}, \ldots, w_d^{(v)}$ are new blue vertices.

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