#### Abstract

Weakly verifiable puzzles can be seen as problems where a solution cannot be efficiently verified by the solver. They are introduced by Canetti et al. (Theory of Cryptography, 2005), who give a first hardness amplification result. Dodis et al. (Theory of cryptography, 2009) define dynamic weakly verifiable puzzles that generalize, for example, MACs and signature schemes. They prove a hardness amplification result for the case where the solver must solve some fraction of puzzles. Holenstein and Schönebeck (Theory of Cryptography, 2011) introduce interactive weakly verifiable puzzles that generalize, for example, commitment schemes. They study hardness amplification for these puzzles and give the right bound even for the setting where a monotone function applied to the output of the checking circuit of the puzzle has to evaluate to one.

In this thesis we introduce dynamic interactive weakly verifiable puzzles by combining the definitions of Dodis et al. and Holenstein et al. We combine their proofs to obtain a hardness amplification result for monotone functions in this case.

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## Chapter 1

# Introduction

## 1.1 Weakly Verifiable Puzzles

The notion of a *weakly verifiable puzzles* was introduced by Cannetti et al. [CHS05]. Intuitively, it is a problem generated by a party called a poser and solved by a solver. One does not require the solver to have an efficient way to verify a solution. On the other hand, the poser has access to some secret information which makes the task of verifying a solution easy.

An example of a weakly verifiable puzzle is a CAPTCHA (an automated Turing test) defined by Ahn et al. [VABHL03]. It is easily solvable by humans but is at least mildly harder to solve for computer programs. The poser generates an instance of a CAPTCHA together with the unique solution. Therefore, it can trivially verify solutions. The solution space is often relatively small, thus it must be hard for the solver being a computer program to verify a solution.

# 1.2 Hardness Amplification

An important task in cryptography is to turn a certain problem that is only mildly hard into one that is substantially harder. An example is the well known hardness amplification lemma for one-way function by Yao [Yao82]. This result implies that it is possible to build a strong one-way function from a function that is only weakly one-way.

A similar hardness amplification statement can be studied for weakly verifiable puzzles. Given a puzzle that is solved with substantial probability, we want to construct a puzzle that is considerable harder to solve.

There are two approaches to amplify hardness for weakly verifiable puzzles. Both base on combining several weakly-hard puzzles into a construction that is strongly-hard. First, one can use *sequential repetition* where puzzles are solved in rounds that start one after another. Ahn et al. [VABHL03] observed that

sequential repetition amplifies hardness for weakly verifiable puzzle. However, this approach may be inefficient as it increases the number of communication rounds. Often a better solution from the practical reasons is to amplify hardness by *parallel repetition* where several independent puzzles are sent in one round. We note that Bellare et al. [BIN97] show that parallel repetition may fail to amplify hardness in certain settings.

To prove hardness amplification one has to show that the following implication holds

$$A \implies B$$

where A is a statement that a problem P is hard, and B denotes that a problem Q is hard. It turns out that it is often easier to consider the following logically equivalent implication

$$\neg B \implies \neg A$$
.

This approach is used in this thesis. More precisely, we assume existence of an algorithm that successfully solves the parallel repetition of weakly verifiable puzzles with substantial probability. Under this assumption we construct an algorithm that success probability in solving a single puzzle is substantial.

#### 1.3 Previous works

The proof of Yao for amplifying hardness of one—way functions relies to a great extent on the fact that it is possible for an adversary to easily verify correctness of a solution. Therefore, to show hardness amplification for weakly verifiable puzzles a different approach has to be developed.

Cannetti et al. [CHS05] define weakly verifiable puzzles and give the hardness amplification proof for these puzzles.

Hardness amplification for weakly verifiable puzzles in a situation where the solver can make several mistakes but still successful solve a parallel repetition of puzzles is studied by Impagliazzo et al. [IJK07]. They introduce a threshold function that determines the minimal fraction of puzzles that must be successfully solved.

Holenstein and Schoenebeck [HS11] give a more general proof for hardness amplification of weakly verifiable puzzles. They use an arbitrary binary monotone function in the place of a threshold function. Additionally, they introduce interactive weakly verifiable puzzles where a puzzle is defined by an interactive protocol between the poser and the solver. These puzzles generalize, for example, the binding property of commitment protocols.

Dodis et al. [DIJK09] extend the proof of hardness amplification of Imaglizazzo et al. Not only, they consider threshold functions but also introduce *dynamic* 

interactive weakly verifiable puzzles where a puzzle is defined together with a set of indices  $\mathcal{Q}$ . The solver must give a correct answer on any  $q \in \mathcal{Q}$ . Furthermore, it can verify correctness of a limited number of solutions and obtain several hints understood as correct solutions on some  $q \in \mathcal{Q}$ . The solver must solve a puzzle on q for which it did not ask a hint before. A dynamic weakly verifiable puzzle can be seen as, for example, a game of breaking security of MAC. In this setting  $\mathcal{Q}$  is a set of messages. The task of the solver is to provide a valid tag for  $q \in \mathcal{Q}$ . The solver can obtain tags of messages of his choice, but must provide a correct tag for a message for which it did not obtain a hint before.

## 1.4 Contribution of the Thesis

In this thesis we introduce the notion of dynamic interactive weakly verifiable puzzles that generalize dynamic and interactive puzzles. We give a proof of hardness amplification for this type of puzzles by combining techniques used by Dodis et al. [DIJK09] and Holenstein et al. [HS11]. As a result we show that it is possible to amplify hardness of dynamic interactive weakly verifiable puzzles where to create an instance of a puzzle the solver and the poser engage in an interactive protocol. Furthermore, the solver can verify a limited number of solutions and obtain correct solutions on some  $q \in \mathcal{Q}$ . Finally, the proof applies also to a setting where a monotone binary function is used to decide which puzzles of the parallel repetition of puzzles have to be successfully solved.

# 1.5 Organization of the Thesis

In Chapter 2 we set up the notation and terminology used in the thesis. We also define a monotone binary function and a pairwise independent family of hash functions.

Next, in Chapter 3 we introduce dynamic interactive weakly verifiable puzzles and state the hardness amplification theorem for this type of puzzles. Proving this theorem is the main focus of this thesis.

Chapter 4 is devoted to the different types of cryptographic primitives that can be seen as weakly verifiable. We briefly describe MACs, signature schemes, commitment protocols, and CAPTCHAs and explain why these constructions are generalized by dynamic interactive weakly verifiable puzzles.

Then, in Chapter 5 we give an outline of the earlier studies of weakly verifiable puzzles and compare them with the techniques used in this thesis.

Finally, in Chapter 6 we give the proof of hardness amplification for dynamic interactive weakly verifiable puzzles and discuss this result.

## Chapter 2

# Notation and Definitions

In this chapter we set up the notation and introduce definitions used in the thesis.

**Bitstrings and tuples.** We denote a tuple  $(x_1, \ldots, x_n)$  by  $x^{(n)}$ . The *i*-th element of  $x^{(n)}$  is denote by  $x_i^{(n)}$  Furthermore, for tuples  $x^{(n)}$ ,  $y^{(k)}$  we use  $x^{(n)} \circ y^{(k)}$  to denote the concatenation of  $x^{(n)}$  and  $y^{(k)}$  which results in a tuple  $(x_1, \ldots, x_l, y_1, \ldots, y_k)$ . We denote a bitstring  $(b_1, \ldots, b_n) \in \{0, 1\}^n$  as a n-tuple of bits.

**Functions.** We write  $p(\alpha_1, \ldots, \alpha_n)$  to denote a polynomial on variables  $(\alpha_1, \ldots, \alpha_n) \in \mathbb{R}^n$ . We call a function  $f: \mathbb{N} \to \mathbb{R}$  negligible if for every p(n) there exists  $n_0 \in \mathbb{N}$  such that for all  $n \in \mathbb{N} : n > n_0$  we have

$$f(n) < \frac{1}{p(n)}.$$

On the other hand, we say that a function  $f: \mathbb{N} \to \mathbb{R}$  is non-negligible if there exists p(n) such that for some  $n_0 \in \mathbb{N}$  and for all  $n \in \mathbb{N}$ :  $n > n_0$  we have

$$f(n) \ge \frac{1}{p(n)}.$$

We note that it is possible that a function  $f: \mathbb{N} \to \mathbb{R}$  is neither negligible nor non-negligible.

We say that a function f is efficiently computable if there exists a polynomial time algorithm computing f.

For a function  $f(n): \mathbb{R} \to \mathbb{R}$  we write f(n) = poly(n) to denote that there exists a polynomial p(n) that bounds f(n) for n big enough.

**Definition 2.1** (Binary monotone function). For  $n \in \mathbb{N}$  we say that  $g : \{0,1\}^n \to \{0,1\}$  is a binary monotone function if for any two bitstrings

$$(b_1, \ldots, b_n) \in \{0, 1\}^n$$
 and  $(b'_1, \ldots, b'_n) \in \{0, 1\}^n$  and sets  $\mathcal{I} := \{i \in \{1, \ldots, n\} : b_i = 1\}$  and  $\mathcal{I}' := \{i \in \{1, \ldots, n\} : b'_i = 1\}$  it holds that if  $\mathcal{I} \subseteq \mathcal{I}'$  and  $g(b_1, \ldots, b_n) = 1$  then  $g(b'_1, \ldots, b'_n) = 1$ .

Algorithms and Circuits. We define a probabilistic circuit as a Boolean circuit  $C_{m,n}: \{0,1\}^m \times \{0,1\}^n \to \{0,1\}^*$  and write  $C_{m,n}(x;\rho)$  to denote a probabilistic circuit that takes as input  $x \in \{0,1\}^m$  and the randomness  $\rho \in \{0,1\}^n$ . If a probabilistic circuit takes as input only the randomness, we slightly abuse the notation and write  $C_n(\rho)$ . We make sure that it is clear from the context that probabilistic circuits that take as input only the randomness are not confused with deterministic Boolean circuits. We use  $\{C_n\}_{n\in\mathbb{N}}$  to denote a family of probabilistic circuits. For a (probabilistic) circuit C we write Size(C) to denote the total number of vertices of C. W define a family of (probabilistic) polynomial size circuits as a family of (probabilistic) circuits where the size of a circuit is polynomial in the number of input vertices.

Sometimes we talk about interactive protocols consisting of two phases where two probabilistic circuits interact with each other. In this settings we define a two-phase circuit  $C := (C_1, C_2)$  as a circuit where in the first phase a circuit  $C_1$  is used and in the second phase a circuit  $C_2$ . We write  $C(\delta) := (C_1, C_2)(\delta)$  to denote that in both phases  $C_1$  and  $C_2$  use the same randomness  $\delta \in \{0, 1\}^*$ .

We write Time(A) to denote the number of steps it takes to execute an algorithm A as a function of the length of input A takes. We say that A runs in the polynomial time if there exists p(n) such that Time(A) is bounded by p(n) where  $n \in \mathbb{N}$  denotes the length of the input that A takes.

Similarly as for a probabilistic circuit we often write the randomness used by a probabilistic algorithm explicitly.

**Probabilities and distributions.** For a finite set  $\mathcal{R}$  we write  $r \stackrel{\$}{\leftarrow} \mathcal{R}$  to denote that r is chosen from  $\mathcal{R}$  uniformly at random. For  $\delta \in \mathbb{R} : 0 \le \delta \le 1$  we write  $\mu_{\delta}$  to denote the Bernoulli distribution where outcome 1 occurs with probability  $\delta$  and 0 with probability  $1 - \delta$ . Moreover, we use  $\mu_{\delta}^k$  to denote the probability distribution over k-tuples where each element of a k-tuple is drawn independently according to  $\mu_{\delta}$ . Finally, let  $u \leftarrow \mu_{\delta}^k$  denote that a k-tuple u is chosen according to  $\mu_{\delta}^k$ .

Let  $(\Omega_n, \mathcal{F}_n, \Pr)$  be a probability space and  $n \in \mathbb{N}$ . Furthermore, let  $E_n \in \mathcal{F}_n$  denote an event whose probability depends on n. We say that  $E_n$  happens almost surely or with high probability if there exists  $n_0 \in \mathbb{N}$  such that for all  $n \in \mathbb{N} : n > n_0$  there exists p(n) such that  $\Pr[E_n] \geq 1 - 2^{-n}p(n)$ .

Interactive protocols. We are often interested in situations where two probabilistic circuits interact with each other according to some protocol by means of messages representable by bitstrings. Let  $\{A_n\}_{n\in\mathbb{N}}$  and  $\{B_n\}_{n\in\mathbb{N}}$  be families

of circuits such that  $A_n: \{0,1\}^* \to \{0,1\}^*$  and  $B_n: \{0,1\}^* \to \{0,1\}^{*1}$  An interactive protocol is defined by  $\{A_n\}_{n\in\mathbb{N}}$  and  $\{B_n\}_{n\in\mathbb{N}}$  where for random bitstrings  $\rho_A \in \{0,1\}^n$ ,  $\rho_B \in \{0,1\}^n$  in the first round a message  $m_0:=A_n(\rho_A)$  is sent and in the second round a message  $m_1:=B_n(\rho_B,m_0)$ . In general in the (2k-1)-th round we have  $m_{2k-2}:=A_n(\rho_A,m_1,\cdots,m_{2k-3})$  and in the 2k-th round  $m_{2k-1}:=B_n(\rho_B,m_1,\cdots,m_{2k-2})$ . The number of rounds of interaction is naturally bounded by the size of circuits  $A_n$  and  $B_n$ . The protocol execution between two probabilistic circuits A and B is denoted by  $\langle A,B\rangle_A$ . The output of A in the protocol execution is denoted by  $\langle A,B\rangle_A$  and of B by  $\langle A,B\rangle_B$ . The sequence of all messages sent by A and B in the protocol execution is called a communication transcript and is denoted by  $\langle A,B\rangle_{trans}$ . The time complexity of the protocol depends on the size of  $A_n$  and  $B_n$ . For example, we say that a protocol runs in the polynomial time if the size of  $A_n$  and  $B_n$  are bounded by some p(n).

**Oracle algorithms.** We use the notion of *oracle circuits* following the standard definition in the literature [Gol04]. If a circuit A has oracle access to a circuit B, we write  $A^B$ . If additionally B has oracle access to a circuit C, we write  $A^{B^C}$ . However, to shorten the notation, we often write  $A^B$  instead and make sure that it is clear from the context which oracle is accessed by B.

In many situations when studying the time complexity of algorithms with oracle access we count each oracle call as a single step. We emphasize this by writing that an algorithm has a certain time complexity with oracle calls. On the other hand, in some settings we are interested in giving a more rigorous bound on the running time of an algorithm. In these situations we compute the running time explicitly with regard to the time needed for accessing the oracle

**Definition 2.2** (Pairwise independent family of efficient hash functions). Let  $\mathcal{D}$  and  $\mathcal{R}$  be finite sets and  $\mathcal{H}$  be a family of functions mapping values from  $\mathcal{D}$  to values in  $\mathcal{R}$ . We say that  $\mathcal{H}$  is a family of pairwise independent efficient hash functions if  $\mathcal{H}$  has the following properties.

**Pairwise independent.** For all  $x, y \in \mathcal{D}, x \neq y$  and for all  $\alpha, \beta \in \mathcal{R}$ , it holds that

$$\Pr_{hash \leftarrow \mathcal{H}}[hash(x) = \alpha \mid hash(y) = \beta] = \frac{1}{|\mathcal{R}|}.$$

**Polynomial time samplable.** For every  $hash \in \mathcal{H}$  the function hash is samplable in time  $poly(\log |\mathcal{D}|, \log |\mathcal{R}|)$ .

**Efficiently computable.** For every  $hash \in \mathcal{H}$  there exists an algorithm running in time  $poly(\log |\mathcal{D}|, \log |\mathcal{R}|)$  which on input  $x \in \mathcal{D}$  outputs  $y \in \mathcal{R}$  such that y = hash(x).

<sup>&</sup>lt;sup>1</sup>We represent the input of  $A_n$  as some tuple that is encoded into a bitstring. The same approach is used for the input of  $B_n$ .

## 2. Notation and Definitions

We note that the pairwise independence property is equivalent to

$$\Pr_{hash \,\leftarrow \, \mathcal{H}}[hash(x) = \alpha \wedge hash(y) = \beta] = \frac{1}{|\mathcal{R}|^2}.$$

It is well known [CW77] that there exists families of functions meeting the criteria stated in Definition 2.2.

## Chapter 3

# Dynamic Interactive Weakly Verifiable Puzzles

In the previous chapter we set up the notation and terminology used in this thesis. The focus of this chapter is on *dynamic interactive weakly verifiable puzzles*, which we introduce in Section 3.1. Then, in Section 3.2 we define the *direct product of puzzles* and state the hardness amplification theorem, which proving is the main goal of this thesis.

#### 3.1 The Definition

Let us combine definitions of puzzles introduced in [DIJK09] and [HS11] and define a dynamic interactive weakly verifiable puzzle as follows.

**Definition 3.1** (Dynamic interactive weakly verifiable puzzle). A dynamic interactive weakly verifiable puzzle (DIWVP) is defined by a family of probabilistic circuits  $\{P_n\}_{n\in\mathbb{N}}$ . A circuit belonging to  $\{P_n\}_{n\in\mathbb{N}}$  is called a problem poser. A solver  $C:=(C_1,C_2)$  is a probabilistic two-phase circuit. We write  $P_n(\pi)$  to denote the execution of  $P_n$  with the randomness fixed to  $\pi \in \{0,1\}^n$  and  $C(\rho):=(C_1,C_2)(\rho)$  to denote the execution of both  $C_1$  and  $C_2$  with the randomness fixed to  $\rho \in \{0,1\}^*$ .

In the first phase, the problem poser  $P_n(\pi)$  and the solver  $C_1(\rho)$  interact. As the result of the interaction  $P_n(\pi)$  outputs a verification circuit  $\Gamma_V$  and a probabilistic hint circuit  $\Gamma_H$ . The circuit  $C_1(\rho)$  produces no output. The circuit  $\Gamma_V$  takes as input  $q \in \mathcal{Q}$  (for some set  $\mathcal{Q}$  of indices), an answer  $y \in \{0,1\}^*$  and outputs a bit. We say that an answer (q,y) is a correct solution if and only if  $\Gamma_V(q,y) = 1$ . The circuit  $\Gamma_H$  on input  $q \in \mathcal{Q}$  outputs a hint such that  $\Gamma_V(q,\Gamma_H(q)) = 1$ .

In the second phase,  $C_2$  takes as input  $x := \langle P_n(\pi), C_1(\rho) \rangle_{trans}$  and has oracle access to  $\Gamma_V$  and  $\Gamma_H$ . The execution of  $C_2$  with the input x and the ran-

domness fixed to  $\rho$  is denoted by  $C_2(x;\rho)$ . The queries of  $C_2$  to  $\Gamma_V$  and  $\Gamma_H$  are called *verification queries* and *hint queries*, respectively. We say that the circuit  $C_2$  succeeds if and only if it makes a verification query (q,y) such that  $\Gamma_V(q,y) = 1$ , and it has not previously asked for a hint query on this q.  $\diamond$ 

If it is clear from the context, we omit the subscript n and write P instead of  $P_n$ .

A verification query (q, y) of a solver C for which a hint query on this q has been asked before cannot be a verification query for which C succeeds. Therefore, without loss of generality throughout this chapter, we make the assumption that C does not ask verification queries on q for which a hint query has been asked before. Moreover, we assume that once C asked a verification query that succeeds, it does not ask any further hint or verification queries.

There is no loss of generality in assuming that the problem poser and the solver are defined by probabilistic circuits. Definition 3.1 embraces also the case where the problem poser and the solver are probabilistic polynomial time algorithms. We use the well known fact [Hol13b] that a probabilistic polynomial time algorithm can be transformed into an equivalent family of probabilistic Boolean circuits of the polynomial size<sup>1</sup>.

We use the term weakly verifiable to emphasize that there is no easy way for the solver to check the correctness of a solution except for asking a verification query. We call a weakly verifiable puzzle dynamic if the number of hint queries is greater than zero. Furthermore, we say that a weakly verifiable puzzle is interactive if in the first phase the number of messages exchanged between the problem poser and the solver is greater than one. Finally, we say that a weakly verifiable puzzle is non-dynamic if |Q|=1 and non-interactive if the number of messages sent in the first phase is at most one.

Definition 3.1 generalizes and combines the previous approaches that study weakly verifiable puzzles [CHS05], dynamic weakly verifiable puzzles [DIJK09], and interactive weakly verifiable puzzles [HS11].

# 3.2 The Hardness Amplification Theorem

In this section we define the k-wise direct product of puzzles and state the hardness amplification theorem for dynamic interactive weakly verifiable puzzles.

**Definition 3.2** (k-wise direct-product of DIWVPs). Let  $g: \{0,1\}^k \to \{0,1\}$  be a binary monotone function and  $P_n^{(1)}$  a problem poser as in Definition 3.1.

<sup>&</sup>lt;sup>1</sup>Theorem 6.10 from [Hol13b] is stated for probabilistic, polynomial time, oracle algorithms with a single bit of output, but it can be adopted to the case where an output is longer than a single bit.

The k-wise direct product of  $P_n^{(1)}$  is a DIWVP defined by a circuit  $P_{kn}^{(g)}$ . We write  $P_{kn}^{(g)}(\pi^{(k)})$  to denote the execution of  $P_{kn}^{(g)}$  with the randomness fixed to  $\pi^{(k)} := (\pi_1, \ldots, \pi_k)$  where  $\pi_i \in \{0, 1\}^n$  for all  $1 \le i \le n$ . Let  $(C_1, C_2)(\rho)$  be a probabilistic two-phase circuit called a solver. In the first phase, the problem poser  $P_{kn}^{(g)}(\pi^{(k)})$  sequentially interacts in k rounds with the circuit  $C_1(\rho)$ . In the i-th round  $C_1(\rho)$  interacts with  $P_n^{(1)}(\pi_i)$ , and as the result  $P_n^{(1)}(\pi_i)$  generates circuits  $\Gamma_V^i, \Gamma_H^i$ . Finally, after k rounds  $P_{kn}^{(g)}(\pi^{(k)})$  outputs a verification circuit

$$\Gamma_V^{(g)}(q, y_1, \dots, y_k) := g(\Gamma_V^1(q, y_1), \dots, \Gamma_V^k(q, y_k))$$

and a hint circuit

$$\Gamma_H^{(k)}(q) := (\Gamma_H^1(q), \dots, \Gamma_H^k(q)).$$
  $\diamond$ 

For the k-wise direct product of puzzles we require the solver to make a verification query on a single index  $q \in \mathcal{Q}$ . Otherwise, if verification queries of the form  $((q_1, y_1), \ldots, (q_k, y_k))$  were allowed, solving the k-wise direct product of puzzles would be trivial. One could fix some  $\bar{q} \in \mathcal{Q}$  and ask k hint queries such that the i-th hint query is of the form  $(q_1, \ldots, q_{i-1}, \bar{q}, q_{i+1}, \ldots, q_k)$  where for all  $1 \leq j \leq k$  such that  $i \neq j$  it holds  $q_j \in \mathcal{Q} \setminus \{\bar{q}\}$ . The solver obtains correct solutions for  $\bar{q}$  to the puzzles on all k positions and can trivially make a successful verification query.

On the following code listing we define an experiment Success such that it outputs 1 if and only if the solver C asks a successful verification query.

```
Experiment Success^{P,C}(\pi, \rho)

Oracles: A problem poser P, a solver C = (C_1, C_2) for P.

Input: Bitstrings \pi \in \{0, 1\}^n, \rho \in \{0, 1\}^*.

Output: A bit b \in \{0, 1\}.

run \langle P(\pi), C_1(\rho) \rangle
(\Gamma_V, \Gamma_H) := \langle P(\pi), C_1(\rho) \rangle_P
x := \langle P(\pi), C_1(\rho) \rangle_{trans}

run C_2^{\Gamma_V, \Gamma_H}(x; \rho)
if C_2^{\Gamma_V, \Gamma_H}(x; \rho) asks a verification query (q, y) s.t. \Gamma_V(q, y) = 1 then return 1

return 0
```

We define the success probability of C in solving a puzzle defined by P as

$$\Pr_{\pi,\rho}[Success^{P,C}(\pi,\rho)=1]. \tag{3.1}$$

Furthermore, for fixed P we say that C succeeds for  $\pi$ ,  $\rho$  if  $Success^{P,C}(\pi,\rho) = 1$ .

We now state our main theorem. Loosely speaking, we claim that it is possible to reduce a solver for the k-wise direct product of P to a solver for a single puzzle defined by P. This implies that if there is no good solver for P, then also a good solver for the k-wise direct product of P does not exist.

**Theorem 3.3** (Hardness amplification for dynamic interactive weakly verifiable puzzles). Let  $P_n^{(1)}$  be a fixed problem poser as in Definition 3.1 and  $P_{kn}^{(g)}$  a problem poser for the k-wise direct product of  $P_n^{(1)}$  as in Definition 3.2. Additionally, let C be a solver for  $P_{kn}^{(g)}$  asking at most h hint queries and v verification queries. There exists a probabilistic algorithm  $\widetilde{Gen}$  with oracle access to C, a binary monotone function  $g: \{0,1\}^k \to \{0,1\}$ , and a problem poser  $P_n^{(1)}$ . The algorithm  $\widetilde{Gen}$  takes as input parameters  $\varepsilon$ ,  $\delta$ , n, k, h, v, and outputs a solver circuit  $\widetilde{D}$  for  $P_n^{(1)}$  such that:

$$\Pr_{\substack{\pi^{(k)} \in \{0,1\}^{kn} \\ \rho \in \{0,1\}^*}} \left[ Success^{P_{kn}^{(g)},C}(\pi^{(k)},\rho) = 1 \right] \ge 16(h+v) \left( \Pr_{u \leftarrow \mu_{\delta}^k} \left[ g(u) = 1 \right] + \varepsilon \right), \tag{3.2}$$

then  $\widetilde{D}$  is a two-phase probabilistic circuit that almost surely over the randomness of  $\widetilde{Gen}$  satisfies

$$\Pr_{\substack{\pi \in \{0,1\}^n \\ \rho \in \{0,1\}^*}} \left[ Success^{P_n^{(1)}, \widetilde{D}}(\pi, \rho) = 1 \right] \ge \delta + \frac{\varepsilon}{6k}. \tag{3.3}$$

Additionally,  $\widetilde{D}$  requires oracle access to g,  $P_n^{(1)}$ , C, hint and verification circuits generated by the problem poser  $P_n^{(1)}$  after the first phase and asks at most  $\frac{6k}{\varepsilon}\log\left(\frac{6k}{\xi}\right)h$  hint queries and one verification query. Finally,  $Time(\widetilde{Gen}) = poly(k, \frac{1}{\varepsilon}, n, v, h)$  with oracle calls.

The above theorem is very general in the sense that it does not pose any constrains on the size of the circuits or the time complexity of the interactive protocol. Additionally, we count each oracle call as a single step of the algorithm.

Let us consider, as an example, the case where the solver C and the problem poser  $P_{kn}^{(g)}$  are polynomial time probabilistic algorithms. Furthermore, we assume that the solver C is such that it satisfies (3.2) and  $\frac{1}{\varepsilon}$  and k are bounded by some polynomial p(n). Clearly the solver and the poser can be represented as polynomial size families of probabilistic circuits. The running time of  $\widetilde{Gen}$  is polynomial, therefore the size of  $\widetilde{D}$  must also be polynomial. Finally, the circuit  $\widetilde{D}$  can be executed by a polynomial time algorithm. Thereby, we obtain a typical polynomial time reduction as described in the literature [AB09, Mau13].

The Theorem 3.3 holds with high probability over the randomness of  $\widetilde{\text{Gen}}$ . More precisely, the circuit output by  $\widetilde{\text{Gen}}$  satisfies the condition of the theorem with probability at least  $1-p(k,n,\frac{1}{\varepsilon})\cdot 2^n$ . Therefore, Theorem 3.3 is meaningful if there exists a polynomial p(n) that bounds k and  $\frac{1}{\varepsilon}$ .

We emphasize that the number of hint queries asked by D is greater than the number of hint queries asked by C whereas the number of verification queries is limited to at most one. For many cryptographic constructions making such an assumption about the number of hint and verification queries is reasonable. In particular, we cannot assume that a solver for a single puzzle may ask more verification queries than a solver for the k-wise direct product of puzzles.

In Chapter 2 we defined a binary monotone function. We note that the monotone restriction on g in Theorem 3.3 is essential. For g(b) := 1 - b a circuit that deliberately gives incorrect answers satisfies g with probability 1 whereas a circuit that solves a puzzle successfully with probability  $\gamma > 0$  succeeds only with probability  $1 - \gamma$ .

We postpone the proof of Theorem 3.3 until Chapter 6.

## Chapter 4

# Examples of Weakly Verifiable Cryptographic Primitives

In the previous chapter we formally defined dynamic interactive weakly verifiable puzzles, the k-wise direct product of puzzles, and stated the hardness amplification theorem for this type of puzzles. In this chapter we give an overview of well known cryptographic primitives that can be seen as various types of weakly verifiable puzzles. First, in Section 4.1 we describe message authentication codes which game based security definition can be seen being a dynamic non-interactive weakly verifiable puzzle. In Section 4.2 we describe public key signature schemes and give the chosen-message attack security definition for public key signature schemes which is yet another example of a dynamic non-interactive weakly verifiable puzzle. Then, in Section 4.3 we focus on bit commitment protocols that binding property can be seen as a non-dynamic interactive weakly verifiable puzzle. Finally, in Section 4.4 we describe a problem of solving CAPTCHAs that can be considered to be a non-dynamic non-interactive weakly verifiable puzzle.

# 4.1 Message Authentication Codes

In this section we describe message authentication codes (MAC) and conclude that the game based definition of MAC security can be seen being a dynamic non-interactive weakly verifiable puzzle.

Let us consider the setting in which two parties, a *sender* and a *receiver*, communicate over an *insecure channel* where messages of the sender may be intercepted, modified, and deleted by a third party called an *adversary*. The receiver needs a way to ensure that the received messages were indeed sent by the sender and were not modified by the adversary. The solution is to use *message authentication codes*.

Loosely speaking, message authentication codes can be explained as follows. Let the sender, receiver, and adversary be polynomial time algorithms and messages be represented as bitstrings. Furthermore, we assume that the sender and the receiver share a secret key to which the adversary has no access. The sender appends to every message a tag which is computed as a function of the key and the message. The receiver, using the key, has a way to efficiently check whether an appended tag is valid for a received message. The receiver accepts a message if the tag is valid, otherwise it rejects. We require that it is hard for the adversary to find a tag and a message that was not sent before and with the non-negligible probability is accepted by the receiver.

Below we give the formal definition of a message authentication code based on [Mau13, Gol04].

**Definition 4.1** (Message Authentication Code). Let  $\mathcal{M}$  be a set of messages,  $\mathcal{K}$  a set of keys and  $\mathcal{T}$  a set of tags. A message authentication code (MAC) is defined by an efficiently computable function  $f: \mathcal{M} \times \mathcal{K} \to \mathcal{T}$ . We say that MAC is secure if f satisfies the following condition:

Let k be chosen uniformly at random from  $\mathcal{K}$  and H be a polynomial size circuit with hard-coded k that takes as input a message  $m \in \mathcal{M}$  and outputs a tag  $t \in \mathcal{T}$  such that f(m,k) = t. We say that MAC is secure if there is no probabilistic polynomial time algorithm with oracle access to H that with nonnegligible probability outputs a message  $m \in \mathcal{M}$  as well as a corresponding tag  $t \in \mathcal{T}$  such that f(m,k) = t, and H has not been queried on m.  $\diamondsuit$ 

We describe how a task of breaking the security of MAC can be seen as solving a dynamic weakly verifiable puzzle where at most one verification query is asked. For fixed f and  $n \in \mathbb{N}$  the sender corresponds to the problem poser and the adversary to the solver. Furthermore, the key  $k_{\pi}$  depends on the randomness  $\pi \in \{0,1\}^n$  used by the problem poser. The set  $\mathcal{Q}$  is the set of messages  $\mathcal{M}$ .

In the first phase, which is non-interactive, neither the problem poser nor the solver send any messages. The problem poser outputs a hint circuit  $\Gamma_H$  and a verification circuit  $\Gamma_V$  where both circuits have hard-coded  $\pi$ . The circuit  $\Gamma_H$  corresponds to the circuit H from Definition 4.1 and takes as input a message m and outputs a tag t such that  $f(m, k_{\pi}) = t$ . The circuit  $\Gamma_V$  takes as input  $m \in \mathcal{M}$  and  $t \in \mathcal{T}$  and outputs 1 if and only if  $f(m, k_{\pi}) = t$ . In the second phase, the solver takes no input (x is empty string), is given oracle access to  $\Gamma_H$  and  $\Gamma_V$  and can ask at most one verification query.

Thus, the task of the adversary to find a valid tag  $t \in \mathcal{T}$  for some message  $m \in \mathcal{M}$  such that the hint query for m has not been asked before corresponds to making a successful verification query by a problem solver to  $\Gamma_V$ .

## 4.2 Public Key Signature Schemes

In this section we describe *public key signature schemes* (SIGs). The game base security definition of SIGs for the chosen message attack can be seen as a dynamic, non-interactive weakly verifiable puzzle.

First, we give a definition of public key encryption schemes, and what it means for such schemes to be secure. The definitions are based on [Gol04].

**Definition 4.2** (Public key signature scheme). Let  $\mathcal{Q}$  be the set of messages. A *public key signature scheme* is defined by a triple of probabilistic polynomial time algorithms: G – the key generation algorithm, V – the verification algorithm, S – the signing algorithm, such that the following conditions are satisfied:

- $G(1^n; \rho)$  outputs a pair of bitstrings  $k_{priv} \in \{0, 1\}^*$  and  $k_{pub} \in \{0, 1\}^*$  where  $n \in \mathbb{N}$  is a security parameter and  $\rho$  the randomness used by G. We call  $k_{priv}$  a private key and  $k_{pub}$  a public key.
- The signing algorithm S takes as input  $k_{priv}$ ,  $q \in \mathcal{Q}$  and outputs a signature  $s \in \mathcal{S}$ .
- The verification algorithm V takes as input  $k_{pub}$ ,  $q \in \mathcal{Q}$ , and  $s \in \mathcal{S}$  and outputs  $b \in \{0, 1\}$ .
- For every  $k_{priv}$ ,  $k_{pub}$  output by G and every  $q \in \mathcal{Q}$  it holds

$$\Pr[V(k_{pub}, q, S(k_{priv}, q))] = 1$$

where the probability is over the randomness S.

 $\Diamond$ 

We say that  $s \in \mathcal{S}$  is a valid signature for  $q \in \mathcal{Q}$  if and only if  $V(k_{pub}, q, s) = 1$ .

It seems that a game of breaking security of the public key signature scheme is not weakly verifiable as the adversary can use V to efficiently check the correctness of the solution. However, we it easy to see that this situation still can be modeled as a dynamic interactive weakly verifiable puzzle where the solver has access to the verification circuit and can ask a polynomial number of verification queries.

**Definition 4.3** (Security of a public key signature scheme with respect to a chosen message attack). Let H be a polynomial size circuit that has hard-coded  $k_{priv}$  and takes as input q and outputs  $S(k_{priv}, q)$ . An adversary A is a probabilistic polynomial time algorithm that takes as input  $k_{pub}$  and has oracle access to H. We say that A succeeds if it finds a valid signature  $s \in \mathcal{S}$  for a message  $q \in \mathcal{Q}$ , and the oracle H has not been queried for a signature of q. The public key encryption scheme is secure if there is no polynomial time adversary that succeeds with non-negligible probability.  $\diamondsuit$ 

We will describe how a task of breaking the security of the public key signature scheme can be seen as a dynamic non-interactive weakly verifiable puzzle.

The problem poser correspond to G and the solver to the adversary. The set Q is a set of messages. In the first phase, the problem poser obtains  $k_{pub}$ ,  $k_{priv}$  and sends  $k_{pub}$  to the solver. Then, the problem poser generates a hint circuit  $\Gamma_H$  and a verification circuit  $\Gamma_V$ . The hint circuit  $\Gamma_H$  corresponds to the circuit H from Definition 4.3 and takes as input  $q \in Q$  and outputs a valid signature for q. The verification circuit  $\Gamma_V$  takes as input  $s \in S$  and  $s \in Q$  and outputs 1 if and only if  $s \in S$  is a valid signature for  $s \in Q$ . In the second phase, the solver takes as input a transcript of the messages from the first phase which consists solely of  $s_{pub}$ . Additionally, the solver is given oracle access to  $s_{quad}$  and  $s_{quad}$  and

Thus, a task of finding a valid signature by the adversary in a public key signature scheme can be seen as a weakly verifiable puzzle that is dynamic but non-interactive as in the first phase only a single message is sent.

## 4.3 Bit Commitment Protocols

In this section we describe bit commitment protocols. A task of breaking the binding property of a bit commitment protocol can be seen as an interactive non-dynamic weakly verifiable puzzle where  $|\mathcal{Q}| = 1$ .

Loosely speaking, we consider the following bit commitment protocol that involves two parties, a sender and a receiver. We suppose that the sender and the receiver are polynomial time probabilistic algorithms. The protocol consists of a commit phase and a reveal phase. In the commit phase the sender and the receiver interact, as the result the sender commits to a value  $b \in \{0, 1\}$ . We require that after the commit phase it is hard for the receiver to correctly guess b. In the reveal phase the sender opens the commitment by sending to the receiver a pair (b', y) where  $y \in \{0, 1\}^*$  should convince the receiver that the sender committed in the commit phase to the value  $b' \in \{0, 1\}$ . A desirable property of a bit commitment protocol is that in the reveal phase it is hard for the sender to find two bitstrings  $y_0$  and  $y_1$  such that the receiver recognizes both  $(0, y_0)$  and  $(1, y_1)$  to be valid decommitments.

We base the following formal definition of a bit commitment protocol on [Hol13a].

**Definition 4.4** (Bit Commitment Protocol). A bit commitment protocol is defined by families of circuits  $\{S_n\}_{n\in\mathbb{N}}$  and  $\{R_n\}_{n\in\mathbb{N}}$  where  $S_n=(S_1,S_2)$  is a two-phase probabilistic circuit,  $R_n$  is a probabilistic circuit, and  $n\in\mathbb{N}$  is

a security parameter. We call  $S_n$  a sender and  $R_n$  a receiver. The circuit  $S_1$  takes as input a pair  $(b, \rho_S)$  where  $b \in \{0, 1\}$  is interpreted as a bit to which S commits, and  $\rho_S \in \{0, 1\}^*$  is the randomness used by S. The receiver  $R_n$  uses the randomness  $\rho_R \in \{0, 1\}^*$ . The protocol consists of two phases. In the commit phase,  $S_1$  and  $R_n$  engage in the protocol execution. As the result,  $S_n$  commits to b and  $R_n$  generates a circuit V. The circuit V takes as input  $b' \in \{0, 1\}$  and a bitstring  $y \in \{0, 1\}^*$  and outputs a bit. In the reveal phase, the circuit  $S_2$  takes as input a communication transcript from the commitment phase  $\langle S_1, R_n \rangle_{trans}$ , the randomness  $\rho_s$  and returns (b', y). We require a bit commitment protocol to have the following properties:

Correctness. For a fixed  $b \in \{0, 1\}$ , we have

$$\Pr_{\substack{\rho_S \in \{0,1\}^*, \rho_R \in \{0,1\}^* \\ V := \langle S_1(b, \rho_S), R_n(\rho_R) \rangle_{R_n} \\ (b', y) := S_2(\langle S_1(b, \rho_s), R_n(\rho_R) \rangle_{trans}, \rho_S)}} \left[ V(b', y) = 1 \right] \ge 1 - \varepsilon(n),$$

where  $\varepsilon(n)$  is a negligible function of n.

**Hiding.** For every  $b \in \{0, 1\}$  the probability over the random coins of  $S_n$  and  $R_n$  that any polynomial size circuit can guess b correctly after the commit phase is at most  $\frac{1}{2} + \varepsilon(n)$  where  $\varepsilon(n)$  is a negligible function of n.

**Binding.** For every probabilistic polynomial size two-phase circuit  $S_n^*(\rho_S) := (S_1^*, S_2^*)(\rho_S)$  we have

$$\Pr_{\substack{\rho_S \in \{0,1\}^*, \rho_R \in \{0,1\}^* \\ V := \langle S_1^*(b,\rho_S), R_n(\rho_R) \rangle_{R_n} \\ ((0,y_0),(1,y_1)) := S_2^*(\rho_S)}} [V(0,y_0) = 1 \land V(1,y_1) = 1] \le \varepsilon(n),$$

where  $\varepsilon(n)$  is a negligible function in n.

 $\Diamond$ 

Breaking the binding property of a bit commitment protocol can be seen as a problem of solving an interactive weakly verifiable puzzle. The receiver corresponds the problem poser and the solver to the sender that tries to break the binding property of the bit commitment protocol. The number of hint queries is zero, and the number of the verification queries is at most one. Thus, we conclude that the puzzle is non-dynamic.

In the first phase, the solver uses its randomness to choose a bit to which it commits. Then, it engage in interaction with the problem poser.

When the interaction in the commit phase is completed the problem poser generates circuits  $\Gamma_V$ ,  $\Gamma_H$ . The circuit  $\Gamma_V$  takes as input  $y_0 \in \{0,1\}^*$ ,  $y_1 \in \{0,1\}^*$  and outputs 1 if and only if  $V(0,y_0) = 1 \wedge V(1,y_1) = 1$ . The hint circuit  $\Gamma_H$  outputs  $\perp$  for any query. Thus, without loss of generality, we

assume that the solver does not ask any verification queries. Furthermore, for the bit commitment protocols  $|\mathcal{Q}| = 1$ , thus we do not write explicitly on which  $q \in \mathcal{Q}$  the solver asks a verification query.

In the second phase, the solver is given oracle access to  $\Gamma_V$  and is allowed to ask at most one verification query. We emphasize that the solver has only oracle access to  $\Gamma_V$ . Therefore, it has no efficient way to check whether the solution is successful except asking a verification query.

Finally, we conclude that for the problem poser and the solver defined as above a task of making a successful verification query corresponds to breaking the binding property of the bit commitment protocol.

## 4.4 Automated Turing Tests

In this section we describe CAPTCHAs which can be seen as non-dynamic and non-interactive weakly verifiable puzzles.

The goal of Automated Turing tests is to distinguish humans from computer programs. An example of such a test is a CAPTCHA, which is formally defined in [VABHL03]. Loosely speaking, a CAPTCHA is a test for which it is hard to write a computer program whose success probability is comparable to or higher than the one achieved by humans. An example is an image depicting a distorted text where the goal is to guess the text used to generate the image.

A task of solving a CAPTCHA based on guessing the distorted text can be seen as a non-dynamic, non-interactive weakly verifiable puzzle. In the first round the problem poser sends the solver an image containing the distorted text. Then, the problem poser generates circuits  $\Gamma_V$  and  $\Gamma_H$ . The circuit  $\Gamma_V$  takes as input a bitstring y and outputs 1 if and only if y correctly encodes the text depicted on the distorted image. A hint circuit  $\Gamma_H$  outputs  $\bot$  in response to every query. Therefore, without loss of generality, we assume that the solver makes no hint queries.

In the second phase, the solver takes as input the image with the distorted text, has oracle access to  $\Gamma_V$  and can ask at most a single verification query. In general, for CAPTCHAs checking the correctness of a solution is comparably hard to finding a correct solution. Thus we conclude that a task of correctly guessing a text depicted on the image can be seen as a weakly verifiable non-interactive and non-dynamic puzzle.

It is not known how good an algorithm for solving CAPTCHAs can be. Thus, it is likely that a gap between the human performance and a performance of the best computer programs may be small. We are interested in finding a way to amplify this gap. In the paper [?] it was shown that this is possible by means of the parallel repetition of non-dynamic interactive weakly verifiable puzzles.

## Chapter 5

# **Previous Results**

In the last chapter we gave an overview of different types of cryptographic primitives that motivated studies of weakly verifiable puzzles. The focus of this chapter is on giving the outline of the previous research. We give a short overview of techniques used in the series of papers [CHS05, DIJK09, HS11] and provide some intuition and insight into the problem of hardness amplification for weakly verifiable puzzles. First, we describe weakly verifiable puzzles studied in [CHS05] that are neither interactive nor dynamic. Then, in Section 5.2 we bring our focus on the dynamic non-interactive puzzles studied in [DIJK09]. Finally, in Section 5.3 we give an overview of the results of Holenstein and Schoenebeck [HS11] where non-dynamic but interactive weakly verifiable puzzles are studied.

# 5.1 Weakly Verifiable Puzzles

The notion of weakly verifiable puzzles has been coined by Canetti, Halevi, and Steiner in the paper Hardness amplification of weakly verifiable puzzles [CHS05]. The puzzles considered there are non-dynamic and non-interactive. Moreover, the number of verification queries is limited to one. This constitutes a special case of Definition 3.1. In this section we provide the definition of weakly verifiable puzzles (WVPs) and state the theorem of hardness amplification for WVPs in a similar vein as in [CHS05]. Finally, we give the intuition behind the proof of this theorem. It is noteworthy that the main proof of this thesis, presented in Chapter 6, uses many ideas of the work of Canetti, Halevi, and Steiner [CHS05].

#### 5.1.1 The definition

We give the definition of WVP from [CHS05]. However, we use the notation and terminology defined in this thesis.

**Definition 5.1** (Weakly Verifiable Puzzle, [CHS05]). A weakly verifiable puzzle is defined by a pair of polynomial time algorithms: a probabilistic puzzle-generation algorithm G and a deterministic verification algorithm V. For a security parameter  $k \in \mathbb{N}$  we write  $G(1^k; \rho)$  to denote that G takes as input  $1^k$  and uses the randomness  $\rho \in \{0, 1\}^*$ . The algorithm G outputs  $p \in \{0, 1\}^*$  and a check information  $c \in \{0, 1\}^*$ . The verifier V takes as input p, c, an answer  $a \in \{0, 1\}^*$  and outputs  $b \in \{0, 1\}$ .

A solver S for G is a probabilistic polynomial time algorithm that takes as input p and outputs a. We denote the randomness used by S as  $\pi \in \{0, 1\}^*$  and define the success probability of S in solving a puzzle defined by (V, G) as

$$\Pr_{\substack{\rho \in \{0,1\}^*, \pi \in \{0,1\}^* \\ (p,c) := G(1^k; \rho) \\ a := S(p,\pi)}} \left[ V(p,c,a) = 1 \right].$$

We write P := (G, V) to denote a weakly verifiable puzzle P defined by algorithms G and V.  $\diamondsuit$ 

Let us argue that Definition 5.1 is a special case of Definition 3.1. First, we note that for G that takes as input  $1^k$  the length of the randomness used is bounded by some polynomial p(k). For a fixed k, without loss of generality, we can represent  $G(1^k; \rho)$  as a probabilistic circuit of polynomial size that takes as input only the randomness  $\rho$ . In Definition 5.1 a verification algorithm V takes as input p, c, a. Again, without loss of generality, we can assume that bitstrings p and c are hard-coded in the circuit  $\Gamma_V$  from Definition 3.1. Furthermore, for weakly verifiable puzzles we have that |Q| = 1 and thus, it is not necessary to pass q as the parameter to  $\Gamma_V$ . Hence, for fixed p and c the algorithm V corresponds to  $\Gamma_V$ . The puzzles considered in Definition 5.1 are non-dynamic. Thus, there is no element corresponding to the hint circuit  $\Gamma_H$ , and without loss of generality, we can assume that  $\Gamma_H$  outputs  $\bot$  for every query. Finally, we note that the puzzles described in Definition 5.1 are non-interactive.

#### 5.1.2 The hardness amplification

We give the definition of the n-wise direct product of weakly verifiable puzzles.

**Definition 5.2** (n-wise direct product of weakly verifiable puzzles, [CHS05]). Let  $n \in \mathbb{N}$  and a weakly verifiable puzzle P = (G, V) be fixed. We define the n-wise direct product of P as a weakly verifiable puzzle where the puzzle-generation algorithm  $G^{(n)}$  takes as input  $1^{k \cdot n}$ , uses the randomness  $\rho \in \{0, 1\}^*$  and outputs tuples  $p^{(n)} := (p_1, \ldots, p_n) \in \{0, 1\}^*$  and  $c^{(n)} := (c_1, \ldots, c_k) \in \{0, 1\}^*$  where for each  $1 \le i \le n$  a pair  $(p_i, c_i)$  is an independent instance of a weakly verifiable puzzle defined by G and V with the security parameter k.

Finally, the verification algorithm  $V^{(n)}$  takes as input  $p^{(n)}$ ,  $c^{(n)}$ , an answer  $a^{(n)}$ , and outputs  $b \in \{0,1\}$  such that b=1 if and only if for all  $1 \le i \le n$  it holds  $V(p_i,c_i,a_i)=1$ .

Let us give the additional notation and terminology. We write  $P^{(n)} := (G^{(n)}, V^{(n)})$  to denote the *n*-wise direct product of P. For  $P^{(n)} := (G^{(n)}, V^{(n)})$  we say puzzle on the *i*-th coordinate to refer to the *i*-th puzzle of the *n*-wise direct product of P (this puzzle corresponds to the one generated by  $G_i^{(n)}, V_i^{(n)}$ ).

The *n*-wise direct product of WVP is solved successfully if and only if all *n* puzzles are solved successfully. In contrast, in Definition 3.2 we are interested in a more general situation where a monotone function  $g: \{0,1\}^n \to \{0,1\}$  is used to decide which coordinates of the *n*-wise direct product of puzzles have to be solved successfully. Clearly, we can assume that g is such that the puzzles on all coordinates have to be solved successfully which matches the case considered in Definition 5.2.

The main theorem proved in [CHS05] states that it is possible to turn a good solver for  $P^{(n)}$  into a good solver for P.

**Theorem 5.3** (Hardness amplification for weakly verifiable puzzles, [CHS05]). Let  $n, q \in \mathbb{N}$  and  $\delta : \mathbb{N} \to (0,1)$  be an efficiently computable function. Moreover, let P = (G, V) be a weakly verifiable puzzle. We denote the running time of the puzzle-generation algorithm G by  $T_G$  and of the verification algorithm V by  $T_V$ . If  $S^{(n)}$  is a solver for the n-wise direct product of P that success probability is at least  $\delta^n$  and the running time is T, then there exists a solver S for P with oracle access to  $S^{(n)}$  that success probability is at least  $\delta(1 - \frac{1}{q})$  and the running time is  $O\left(\frac{nq^3}{\delta^{2n-1}}(T + nT_G + nT_V)\right)$ .

The parameter q is introduced as it is not possible to achieve the perfect hardness amplification. We note that in the analysis of the running time of the CHS-solver we explicitly take into account the time needed for the oracle calls to  $S^{(n)}$ , V, G.

The following algorithm CHS-solver has oracle access to  $S^{(n)}$  and solves a puzzle defined by P with probability at least  $\delta(1-\frac{1}{q})$ . We denote by  $p \in \{0,1\}^*$  the output of G, which is also the input taken by the CHS-solver. To make the notation shorter in the following code excerpts we do not write the randomness used by G explicitly.

**Algorithm:** CHS-solver  $S^{(n)}, V, G(p, n, k, q, \delta)$ 

**Oracle:** A solver  $S^{(n)}$  for  $P^{(n)}$ , a verification algorithm V for P, a puzzlegeneration algorithm G for P.

**Input:** A bistring  $p \in \{0,1\}^*$ , parameters  $n, k, q, \delta$ .

```
\begin{array}{l} \textit{prefix} := \emptyset \\ \textbf{for} \ i := 1 \ \textbf{to} \ n - 1 \ \textbf{do:} \\ p^* := ExtendPrefix^{S^{(n)},V,G}(\textit{prefix},i,n,k,q,\delta) \\ \textbf{if} \ p^* = \bot \ \textbf{then} \ \textbf{return} \ \textit{OnlinePhase}^{S^{(n)},V,G}(\textit{prefix},p,i,n,k,q,\delta) \\ \textbf{else} \ \textit{prefix} := \textit{prefix} \circ p^* \\ a^{(n)} := S^{(n)}(\textit{prefix} \circ p) \\ \textbf{return} \ a_n \end{array}
```

**Algorithm:** OnlinePhase  $S^{(n)}, V, G$  (prefix,  $p, i, n, k, q, \delta$ )

**Oracle:** A solver algorithm  $S^{(n)}$  for  $P^{(n)}$ , a puzzle-generation algorithm G for P, a verification algorithm V for P.

**Input:** A (v-1)-tuple of bitstrings *prefix*, a bitstring  $p \in \{0,1\}^*$ , parameters  $v, n, k, q, \delta$ .

```
 \begin{split} \mathbf{Repeat} & \left\lceil \frac{6q \ln(6q)}{\delta^{n-i+1}} \right\rceil \text{ times} \\ & \left( (p_{i+1}, \dots, p_n), (c_{i+1}, \dots, c_n) \right) := G^{(n-i-1)}(1^k) \\ & a^{(n)} := S^{(n)}(prefix, p, p_{i+1}, \dots, p_n) \\ & \text{ if } \forall_{i+1 \leq j \leq n} V(p_j, c_j, a_j) = 1 \text{ then return } a_i \\ & \text{return } \bot \end{split}
```

**Algorithm:** ExtendPrefix $S^{(n)}, V, G$  (prefix,  $i, n, k, q, \delta$ )

**Oracle:** A solver algorithm  $S^{(n)}$  for  $P^{(n)}$ , a puzzle-generation algorithm G for P, a verification algorithm V for P.

Input: A (i-1)-tuple of puzzles prefix, parameters  $i, n, k, q, \delta$ .

```
\begin{aligned} \mathbf{Repeat} & \left\lceil \frac{6q}{\delta^{n-v+1}} \ln(\frac{18qn}{\delta}) \right\rceil \text{ times} \\ & (p^*,c^*) := G(1^k) \\ & \bar{\nu}_i := EstimateResSuccProb^{S^{(n)},G,V}(prefix \circ p^*,i,n,k,q,\delta) \\ & \text{if } \bar{\nu}_i \geq \delta^{n-i} \text{ then return } p^* \\ & \text{return } \bot \end{aligned}
```

**Algorithm:** EstimateResSuccProb<sup> $S^{(n)}$ </sup>,V,G(prefix,  $i, n, k, q, \delta$ )

**Oracle:** A solver algorithm for  $P^{(n)}$ , a verification algorithm V for P, a puzzle-generation algorithm G for P

**Input:** A *i*-tuple of puzzles *prefix*, parameters  $i, n, k, q, \delta$ .

```
successes := 0
\mathbf{Repeat} \ M := \left\lceil \frac{84q^2}{\delta^{n-i}} \ln \left( \frac{18qn \cdot N_i}{\delta} \right) \right\rceil \text{ times}
\left( (p_{i+1}, \dots, p_n), (c_{i+1}, \dots, c_n) \right) := G^{(n-i)}(1^k)
a^{(n)} := A(prefix, p_{i+1}, \dots, p_n)
\mathbf{if} \ \forall_{i+1 \leq j \leq n} : V(p_j, c_j, a_j) = 1 \ \mathbf{then} \ successes := successes + 1
\mathbf{return} \ successes / M
```

A full proof of Theorem 5.3 is presented in [CHS05]. We limit ourselves to providing intuition why the CHS-solver transforms a good solver for the n-wise direct product of P into a good solver for a single puzzle defined by P.

Let us consider the *n*-wise direct product of P, and for simplicity a deterministic solver  $S^{(n)}$  for  $P^{(n)} := (G^{(n)}, V^{(n)})$ . We write  $p^{(n)}, c^{(n)}$  to denote the output of  $G^{(n)}$ . We define a matrix M as follows. The columns of M are labeled with all possible bitstrings  $p_1$  whereas the rows are labeled with all possible tuples  $(p_2, \ldots, p_n)$  output by  $G^{(n)}$  when executed with different randomness. A cell of M contains a binary n-tuple such that the i-th bit equals 1 if and only if  $V_i(p_i, c_i, a_i) = 1$  where  $a^{(n)} := S^{(n)}(p^{(n)})$  and  $p^{(n)}$  is a tuple of bitstrings inferred by a column and a row of the cell. We make the following observation.

**Observation 5.4.** Let  $S^{(n)}$  be a deterministic polynomial time solver for  $P^{(n)}$  that success probability is at least  $\delta^n$ . Then, the matrix M defined as above has either a column with a  $\delta^{(n-1)}$  fraction of cells that are  $1^n$  vectors, or a conditional probability that a cell is of the form  $1^n$  given that the last (n-1) bits of this cell are equal 1 is at least  $\delta$ .

Let us explain, at least intuitively using Observation 5.4, how the CHS-solver can be used to solve a puzzle defined by P with substantial probability given oracle access to  $S^{(n)}$  for  $P^{(n)}$ . The CHS-solver starts with the first position and tries to fix a bitstring  $p^*$  on this position such that the success probability of  $S^{(n)}$  on the remaining (n-1) position is at least  $\delta^{(n-1)}$ . If it is possible to find  $p^*$  such that this condition is satisfied, then we fix  $p^*$  on this position and repeat the whole procedure again in the consecutive iteration for the next position. If the CHS-solver fails to find a bitstring  $p^*$ , then we assume that there is no column of M that contains a  $\delta^{(n-1)}$  fraction of cells that are of the form  $1^n$ . We use Observation 5.4 to conclude that the conditional probability

of solving the first puzzle given that all puzzles on the remaining position are solved successfully is at least  $\delta$ . We place the input puzzle p on this position and note that the remaining puzzles are generated by the CHS-solver. Thus, it is possible to efficiently verify whether these puzzles are successful solved by  $S^{(n)}$ .

Obviously, the CHS-solver can still fail. First, it may happen that it does not find a column with a high fraction of puzzles that are solved successfully, although such a column exists. Secondly, we cannot exclude a situation where no such column exists, but the algorithm fails to find a cell such that last (n-1) bits are 1. Finally, it is also possible that the estimate returned by EstimateResSuccProb is incorrect.

It is possible to show that all these events happen with small probability such that the success probability of the CHS-solver is at least  $\delta(1-\frac{1}{q})$  almost surely.

In Chapter 6 we study a more general class of puzzles that are not only weakly verifiable but also dynamic and interactive. Furthermore, we allow a more general situation where a solver successfully solves the n-wise direct product of puzzles, although it succeeds only on a subset of coordinates of the n-wise direct product of P. It turns out that it is possible to use a similar technique of fixing puzzles on the consecutive positions of the n-wise direct product of puzzles to prove hardness amplification in this more general setting.

# 5.2 Dynamic Weakly Verifiable Puzzles

Some of the cryptographic primitives presented in Chapter 4 are not only weakly verifiable but also dynamic (MAC and SIG). This type of puzzles is defined and studied in [DIJK09]. We give a short overview of this work and state the definition of a *dynamic weakly verifiable puzzle* (DWVP) that closely follows the one included in [DIJK09]. Finally, we describe important parts of the proof of hardness amplification for DWVPs.

#### 5.2.1 The definition

**Definition 5.5** (Dynamic Weakly Verifiable Puzzle, [HS11]). A dynamic weakly verifiable puzzle is defined by a distribution  $\mathcal{D}$  on pairs  $(x, \alpha)$  where  $\alpha$  is a secret information and x is a bitstring. Furthermore, we consider a set  $\mathcal{Q}$  (for some set of indices  $\mathcal{Q}$ ) and a probabilistic polynomial time computable verification relation V such that  $V(\alpha, q, r) = 1$  if and only if  $r \in \{0, 1\}^*$  is a correct answer to  $q \in \mathcal{Q}$  on the set determined by  $\alpha$ . Finally, let H be a probabilistic polynomial time computable hint function that on input  $\alpha$ , q outputs a bitstring  $\{0, 1\}^*$ .

A solver S takes as input x and can make hint queries on  $q \in \mathcal{Q}$  which are answered using  $H(\alpha, q)$  and verification queries (q, r) answered by means of

 $V(\alpha, q, r)$ . We say that S succeeds if and only if it makes a verification query on (q, r) such that  $V(\alpha, q, r) = 1$  and it has not previously asked for a hint query on this q. We write  $P := (\mathcal{D}, V, H)$  to denote a DWVP with the distribution  $\mathcal{D}$  and V, H being a verification and hint relations, respectively.

The above definition is a special case of Definition 3.1. To see this we observe that instead of considering a distribution  $\mathcal{D}$  on pairs  $(x, \alpha)$  we can use a probabilistic problem poser that outputs circuits  $\Gamma_H$  and  $\Gamma_V$  with hard-coded  $\alpha$  that correspond to H and V, respectively. It is clear that the solver S from Definition 5.5 can be turned into a family of probabilistic polynomial size circuits with oracle access to  $\Gamma_H$  and  $\Gamma_V$ . Furthermore, in the first phase, which is non-interactive, a bitstring x is sent by the problem poser to the solver.

#### 5.2.2 The hardness amplification theorem

**Definition 5.6** (*n*-wise direct product of DWVPs, [DIJK09]). For a dynamic weakly verifiable puzzle  $P := (\mathcal{D}, V, H)$  we define the *n*-wise direct product of P as a DWVP with a distribution  $\mathcal{D}^{(n)}$  on tuples  $(x_1, \alpha_1), \ldots, (x_n, \alpha_n)$  where each  $(x_i, \alpha_i)$  is drawn from the distribution  $\mathcal{D}$ . Furthermore, the hint relation is defined by  $H^{(n)}(q, \alpha_1, \ldots, \alpha_n) := (H(\alpha_1, q), \ldots, H(\alpha_n, q))$  and the verification relation  $V^{(n)}(\alpha_1, \ldots, \alpha_n, r_1, \ldots, r_n, q)$  evaluates to 1 if and only if for at least  $n - (1 - \gamma)\delta n$  of bitstrings  $r_1, \ldots, r_n$  it holds  $V(\alpha_i, q, r_i) = 1$  where  $0 \leq \gamma, \delta \leq 1$ .

We write  $P^{(n)} := (D^{(n)}, H^{(n)}, V^{(n)})$  to denote the *n*-wise direct product of P.

The above definition is more general than defined in the previous section definition of the n-wise direct product of WVP as in Definition 5.2 it is enough if the solver succeeds only on a fraction of coordinates.

Dynamic weakly verifiable puzzles generalize games of breaking security of message authentication codes and public key signature schemes. In the case of MAC x is an empty bitstring. For the public key signature schemes x is a public key.

We write  $(\mathcal{H}_{hint}, \mathcal{V}_{verif}) \leftarrow S(x; \delta)$  to denote the execution of S with the input  $x \in \{0, 1\}^*$  and using the randomness  $\delta \in \{0, 1\}^*$  where  $\mathcal{H}_{hint}$  is the set of all hint queries asked by S, and  $\mathcal{V}_{verif}$  is the set of pairs (q, r) of all verification queries asked in the execution of S.

With no loss of generality, we make the assumption that once S made a successful verification query it does not ask any further hint and verification queries.

We define the success probability of a solver S as

$$\Pr_{\substack{\delta \in \{0,1\}^* \\ (x,\alpha) \leftarrow \mathcal{D} \\ (\mathcal{H}_{hint}, \mathcal{V}_{verif}) \leftarrow S(x;\delta))}} \left[ \exists (q,r) \in \mathcal{V}_{verif} : q \notin \mathcal{H}_{hint} \land V(\alpha, q, r) = 1 \right]$$

**Theorem 5.7** (Hardness amplification for dynamic weakly verifiable puzzles [DIJK09]). Let  $S^{(n)}$  be a probabilistic algorithm for  $P^{(n)}$  that succeeds with probability at least  $\varepsilon$ , where  $\varepsilon \geq (800/\gamma\delta) \cdot (h+v) \cdot e^{-\gamma^2\delta n/40}$ , and h and v denote the number of hint and verification queries asked by  $S^{(n)}$ , respectively. There exists a probabilistic algorithm S that solves a puzzle defined by P with probability at least  $1-\delta$ . Furthermore, S makes  $O(h(h+v)/\varepsilon) \cdot \log(1/\gamma\delta)$  hint queries and at most one verification query. The running time of S is polynomial in  $h, v, \frac{1}{\varepsilon}, t, \omega, \log(1/\gamma\delta)$  where  $\omega$  is time needed to ask a single hint query.

It is worth seeing why the approach presented in the previous section that works well for the direct product of WVPs cannot be applied for the direct product of DWVPs (moving aside for a moment the issue of solving only a fraction of puzzles successfully). Loosely speaking, for DWVP the algorithm CHS-solver breaks in the *OnlinePhase* where the solver  $S^{(n)}$  can be called multiple times. It is possible that in one of these calls  $S^{(n)}$  asks a hint query on q which prevents in one of the later runs a verification query (q, r) from succeeding. The fact that a hint query on q has been asked before makes it impossible to make a successful verification query on this q. Thus, we cannot dismiss a situation where the success probability of  $S^{(n)}$  decreases with the number of iterations.

The solution proposed in [DIJK09] is to partition the set  $\mathcal{Q}$  into a set of attacking queries  $\mathcal{Q}_{attack}$  and a set of advice queries  $\mathcal{Q}_{adv}$ . The idea is to allow a solver for the direct product to ask hint queries only on  $q \in \mathcal{Q}_{adv}$ , and to halt the execution whenever a hint query is asked on  $q \in \mathcal{Q}_{attack}$ . More formally, for a function  $hash: \mathcal{Q} \to \{0, 1, \dots, 2(h+v)-1\}$  we define  $\mathcal{Q}_{attack} := \{q \in \mathcal{Q}: hash(q) = 0\}$  and  $\mathcal{Q}_{adv} := \{q \in \mathcal{Q}: hash(q) \neq 0\}$ .

It is possible, for a fixed solver S that asks at most h hint queries and v verification queries, to find a function  $hash: \mathcal{Q} \to \{0, 1, \dots 2(h+v)-1\}$  such that the success probability of S with respect to  $\mathcal{Q}_{attack}$  and  $\mathcal{Q}_{adv}$  partitioned using hash is multiplied by  $\frac{1}{8(h+v)}$ . If h and v are not too big, then the success probability of S can be still substantial. More formally, the following lemma is proved in [DIJK09].

**Lemma 5.8** (Success probability in solving a dynamic weakly verifiable puzzle with respect to a function hash, [DIJK09]). Let S be a solver for DWVP which success probability is at least  $\delta$ , the running time is at most t, and the number of hint and verification queries is at most t and t, respectively. There exists a

probabilistic algorithm that runs in time polynomial in  $h, v, \frac{1}{\delta}$ , t that outputs a function hash:  $\mathcal{Q} \to \{0, 1, \dots, 2(h+v)-1\}$  that partitions  $\mathcal{Q}$  into  $\mathcal{Q}_{attack}$  and  $\mathcal{Q}_{adv}$  such that with probability at least  $\frac{\delta}{8(h+v)}$  the first successful verification query (q', a) asked by S is such that  $q' \in \mathcal{Q}_{attack}$  and all previous hint and verification queries have been asked on  $q \in \mathcal{Q}_{adv}$ .

A function hash can be found by means of a natural sampling technique. We follow exactly the same approach of partitioning Q in Section 6.1.3.

Let  $H_{\alpha}(q)$  denote a polynomial time probabilistic algorithm that takes as input q, has hard-coded  $\alpha$  and outputs  $H(\alpha,q)$ . Similarly, we use  $V_{\alpha}(q,r)$  to denote a polynomial time probabilistic algorithm that computes  $V(\alpha,q,r)$  and has hard-coded bitstring  $\alpha$ . The algorithm DWVP-solver has oracle oracle access to  $S^{(n)}$ ,  $V_{\alpha}$  and  $H_{\alpha}$  as well as a function hash as in Lemma 5.8.

```
Algorithm: DWVP-solver<sup>S^{(n)}</sup>, hash, H^{(n)}_{\alpha}, V^{(n)}_{\alpha} (x)
Oracle: A solver S^{(n)} for P^{(n)}, algorithms V_{\alpha} and H_{\alpha}, a function hash:
Q \to \{0, 1, \dots, 2(h+v) - 1\}.
Input: A bistring x \in \{0,1\}^*.
Repeat at most O(\frac{h+v}{\varepsilon} \cdot \log(\frac{1}{\gamma\delta})) times
      Let i \stackrel{\$}{\leftarrow} \{1, \dots, n\} be a position for x.
      Generate (x_1, \alpha_1), \ldots, (x_{i-1}, \alpha_{i-1}), (x_{i+1}, \alpha_{i+1}), \ldots, (x_n, \alpha_n)
      using (n-1) calls to P each time with fresh randomness.
      run S^{(n)}(x_1,\ldots,x_{i-1},x,x_{i+1},\ldots,x_n)
            if S^{(n)} asks a hint query on q then
                  if hash(q) \neq 0 then abort current run of S^{(n)}
                  Ask a verification query r := H(q)
                  Let (r_1, \ldots, r_{i-1}, r_{i+1}, \ldots, r_n) be hints for query q for puzzle
                  sets (x_1, \ldots, x_{i-1}, x_{i+1}, x_n)
                  Answer the hint query of S^{(n)} using (r_1, \ldots, r_{i-1}, r, r_{i+1}, r_n)
            if S^{(n)} asks a verification query (q, r_1, \dots, r_n) then
                  if hash(q) = 0 then answer the query with 0
                  Let m := |j: V(q, r_i) = 1, j \neq i|
                  if m \ge n - n(1 - \gamma)\delta then
                        make a verification query (q, r_i) and halt.
                  else with probability \rho^{m-n(1-\gamma)\delta} ask a verification query
                        (q, r_i) and halt.
                  Halt the current run of S^{(n)} and go to the next iteration.
return \perp
```

In each iteration of the DWVP-solver the position for the input puzzle is

chosen uniformly at random. The remaining (n-1) puzzles are generated by the algorithm, thus it is possible to answer all hint and verification queries for these puzzles. The DWVP-solver calls  $S^{(n)}$  multiple times, but the function hash is used to partition the query domain into  $Q_{attack}$  and  $Q_{adv}$  such that if a hint query is asked on q such that hash(q) = 0 then the current execution of  $S^{(n)}$  is aborted, and the DWVP-solver goes to the next iteration. In this way, we ensure that the DWVP-solver never asks a hint query that could prevent a verification query from succeeding. If a verification query is asked on q such that  $hash(q) \neq 0$  we answer this query with 0.

Finally, in the case where  $S^{(n)}$  asks a verification query on q such that hash(q) = 0, a  $soft\ decision\ system$  is used to decide whether to ask a verification query. The idea is that if there are many puzzles among the ones generated by the algorithm that are solved successfully, then it is likely that also the input puzzle is solved successfully. We discount  $\gamma \delta n$  to take into account that not all puzzles have to be solved successfully. Detail calculations provided in [DIJK09] show that this approach yields a demanded result.

In Chapter 6 we consider a weakly verifiable puzzles that are not only dynamic but also interactive. We use a very similar technique to partition domain  $\mathcal{Q}$  into advice and attacking queries. Instead of the requirement to succeed on at least a fraction of puzzles we consider a more general approach where an arbitrary monotone function  $g: \{0,1\}^n \to \{0,1\}$  is used to determine on which coordinates the solver has to succeed in order to successfully solve the k-wise direct product of puzzles.

# 5.3 Interactive Weakly Verifiable Puzzles

Hardness amplification for interactive but non-dynamic weakly verifiable puzzles has been studied by Holenstein and Schoenebeck in [HS11]. We will give now an overview of this work and compare it with our approach.

#### 5.3.1 The definition

**Definition 5.9** (Interactive Weakly Verifiable Puzzle, [HS11]). An interactive weakly verifiable puzzle is defined by a protocol given by two probabilistic algorithms P and S. The algorithm P is called the problem poser and produces as output a verification circuit  $\Gamma$ . The algorithm S called the problem solver produces no output. Furthermore, the success probability of the algorithm  $S^*$  in solving an interactive weakly verifiable puzzle defined by (P, S) is:

$$\Pr_{\substack{\rho,\pi\\\Gamma:=\langle P(\rho),S^*(\pi)\rangle_P}} \Big[\Gamma(\langle P(\rho),S^*(\pi)\rangle_{S^*})=1\Big].$$



It is not hard to see that Definition 5.9 is a special case of Definition 3.1. The puzzles in Definition 5.9 are non-dynamic, thus |Q| = 1. Furthermore, we can assume that  $\Gamma$  corresponds to the verification circuit from Definition 3.1 and the circuit  $\Gamma_H$  always outputs  $\perp$ .

Similarly as in the previous sections, we define the k-wise direct product of puzzles.

**Definition 5.10** (k-wise direct product of interactive weakly verifiable puzzles, [HS11]). Let  $g: \{0,1\}^k \to \{0,1\}$  be a monotone function and (P,S) be a fixed interactive weakly verifiable puzzle. The k-wise direct product of (P,S) denoted by  $(P^{(g)}, S^{(g)})$  is an interactive weakly verifiable puzzle where the problem poser and the solver sequentially interact in k rounds. In each round (P,S) is used to produce an instance of the interactive weakly verifiable puzzle. As the result circuits  $\Gamma^{(1)}, \ldots, \Gamma^{(k)}$  for P are generated. Finally,  $P^{(g)}$  outputs the circuit  $\Gamma^{(g)}(y_1, \ldots, y_k) := g(\Gamma^{(1)}(y_1), \ldots, \Gamma^{(k)}(y_k))$ .

#### 5.3.2 The hardness amplification theorem

The following hardness amplification theorem is proved in [HS11].

**Theorem 5.11** (Hardness amplification for interactive weakly verifiable puzzles, [HS11]). There exists an algorithm  $Gen(C, g, \varepsilon, \delta, n)$  which takes as input a solver circuit C for the k-wise direct product of (P, S), a monotone function  $g: \{0,1\}^* \to \{0,1\}$ , and parameters  $\varepsilon, \delta, n$ . The algorithm Gen outputs a solver circuit D for P such that the following holds. If C is such that

$$\Pr\left[\Gamma^{(g)}(\langle P^{(g)}, C \rangle_C) = 1\right] \ge \Pr_{u \leftarrow \mu_{\varepsilon}^{(k)}} \left[g(u) = 1\right] + \varepsilon,$$

where the probability is over the randomness of  $P^{(g)}$  and C, then D satisfies almost surely

$$\Pr\left[\Gamma(\langle P, D\rangle_D) = 1\right] \ge \delta + \frac{\varepsilon}{6k},$$

where the probability is over the randomness of P and D. Additionally, Gen and D only require oracle access to g and C. Furthermore,  $Size(D) \leq Size(C) \cdot \frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon})$ , and the running time of Gen is polynomial in  $k, \frac{1}{\varepsilon}, n$  with oracle calls.

First, we notice that the above definition does not impose any restrictions on the time complexity of the poser and the solver. We consider a general approach where Gen is used to define a polynomial time reduction between a solver for the k-wise direct product of puzzles to a solver for a single puzzle. Furthermore, in the previous sections we considered solvers for the k-wise direct product that are compared with the algorithms that either solve all puzzles

#### 5. Previous Results

[CHS05] or allow a fraction of puzzles to be solved incorrectly [DIJK09]. In the above definition a more general approach is considered where a binary monotone function g is used.

In Chapter 6 we use very similar approach as Holenstein and Schoenebeck. The difference is that proven by us hardness amplification theorem captures also puzzles that are both dynamic and interactive.

### Chapter 6

### Hardness Amplification for Dynamic Interactive Weakly Verifiable Puzzles

In the previous chapter we gave an overview of the former studies of different types of weakly verifiable puzzles. The focus of this chapter is on giving a constructive proof of hardness amplification for dynamic interactive weakly verifiable puzzles. First, in Section 6.1.1 we remind the hardness amplification theorem for dynamic interactive weakly verifiable puzzles stated in Section 3.2. We give an intuition behind the proof in Section 6.1.2. Next, in Section 6.1.3 we construct an algorithm that finds an efficiently computable function that is used to partition the domain of hint and verification queries. Then, in Section 6.1.4 we show that is it possible to amplify hardness for dynamic interactive weakly verifiable puzzles under the assumption that there are no hint queries that prevents verification queries from succeeding. Finally, in Section 6.1.5 we complete the proof by combining the previous steps.

### 6.1 Proof of the Hardness Amplification

#### 6.1.1 The hardness amplification theorem

First, let us remind the experiment Success defined in Chapter 3. The problem poser for an interactive dynamic weakly verifiable puzzles is denoted by P and the solver by C. Additionally,  $\pi$  and  $\rho$  denote the randomness of the problem poser and the solver, respectively.

Experiment  $Success^{P,C}(\pi, \rho)$ 

**Oracles:** A problem poser P, a solver  $C = (C_1, C_2)$  for P.

```
Input: Bitstrings \pi \in \{0,1\}^n, \rho \in \{0,1\}^*.

Output: A bit b \in \{0,1\}.

 run \langle P(\pi), C_1(\rho) \rangle 
 (\Gamma_V, \Gamma_H) := \langle P(\pi), C_1(\rho) \rangle_P 
 x := \langle P(\pi), C_1(\rho) \rangle_{trans} 
 run \ C_2^{\Gamma_V, \Gamma_H}(x; \rho) 
 if \ C_2^{\Gamma_V, \Gamma_H}(x; \rho) \text{ asks a verification query } (q, y) \text{ s.t. } \Gamma_V(q, y) = 1 \text{ then } 
 return \ 1 
 return \ 0
```

Similarly, as in Chapter 3 without loss of generality throughout this chapter, we make the assumption that C does not ask verification queries on q for which a hint query has been asked before. Furthermore, we assume that once C asked a verification query that succeeds, it does not ask any further hint or verification queries.

**Theorem 3.3** (Hardness amplification for dynamic interactive weakly verifiable puzzles). Let  $P_n^{(1)}$  be a fixed problem poser as in Definition 3.1 and  $P_{kn}^{(g)}$  a problem poser for the k-wise direct product of  $P_n^{(1)}$  as in Definition 3.2. Additionally, let C be a solver for  $P_{kn}^{(g)}$  asking at most h hint queries and v verification queries. There exists a probabilistic algorithm  $\widetilde{Gen}$  with oracle access to C, a binary monotone function  $g:\{0,1\}^k \to \{0,1\}$ , and a problem poser  $P_n^{(1)}$ . The algorithm  $\widetilde{Gen}$  takes as input parameters  $\varepsilon$ ,  $\delta$ , n, k, h, v, and outputs a solver circuit  $\widetilde{D}$  for  $P_n^{(1)}$  such that: If C satisfies

$$\Pr_{\substack{\pi^{(k)} \in \{0,1\}^{kn} \\ \rho \in \{0,1\}^*}} \left[ Success^{P_{kn}^{(g)},C}(\pi^{(k)},\rho) = 1 \right] \ge 16(h+v) \left( \Pr_{u \leftarrow \mu_{\delta}^k} \left[ g(u) = 1 \right] + \varepsilon \right), \tag{3.2}$$

then  $\widetilde{D}$  is a two-phase probabilistic circuit that almost surely over the randomness of  $\widetilde{Gen}$  satisfies

$$\Pr_{\substack{\pi \in \{0,1\}^n \\ \rho \in \{0,1\}^*}} \left[ Success^{P_n^{(1)}, \widetilde{D}}(\pi, \rho) = 1 \right] \ge \delta + \frac{\varepsilon}{6k}. \tag{3.3}$$

Additionally,  $\widetilde{D}$  requires oracle access to g,  $P_n^{(1)}$ , C, hint and verification circuits generated by the problem poser  $P_n^{(1)}$  after the first phase and asks at most  $\frac{6k}{\varepsilon}\log\left(\frac{6k}{\varepsilon}\right)h$  hint queries and one verification query. Finally,  $Time(\widetilde{Gen}) = poly(k, \frac{\varepsilon}{\varepsilon}, n, v, h)$  with oracle calls.

#### 6.1.2 The intuition behind the proof of hardness amplification

In the proof of Theorem 3.3 we combine techniques used in [DIJK09] [HS11]. First, we give the intuition behind the part of the proof that bases on [HS11]. Next, we describe ideas from [DIJK09] that let us extend to the case of dynamic puzzles. Finally, we provide intuition how to combine these techniques to prove Theorem 3.3.

Let us fix the notation as in Theorem 3.3. We refer to a puzzle solved by  $\widetilde{D}$  as an *input puzzle*.

**Technique of [CHS05, HS11]** The idea is to use the circuit C to solve k puzzles among which is the input puzzle and the remaining k-1 puzzles are generated. If a position for the input puzzle is carefully chosen as well as the randomness used to generate the remaining puzzles, then the input puzzle can be solved with substantial probability almost surely.

More precisely, we try to fix the randomness used to generate a puzzle on the first position such that the circuit C satisfies the assumptions of Theorem 3.3 for the (k-1)-wise direct product of puzzles. The success probability on the remaining puzzles, where the first puzzles is fixed, can be estimated using a natural sampling technique.

When it is possible to find this randomness, then we recursively solve the smaller sub-problem for the (k-1)-wise direct product of puzzles. If the recursion reaches k = 1, then we place the input puzzle on the last position.

We cannot exclude the situation where in one of the recursive calls we cannot find the randomness for the first position such that the success probability of C on the remaining k-1 coordinates is substantial. Nevertheless, we know that C has the substantial success probability when all positions are considered. This let us conclude that the first coordinate is somehow important in the sense that C solves a puzzle on this coordinate unusually often. Therefore, it makes sense to place the input puzzle on this position. What is left is to find the randomness used to generate puzzles on the remaining positions such that when the input puzzle is placed on the first position, then it is solved correctly often.

In Section 5.1 we described the same approach for the special case where non-interactive puzzles are considered and the function g is such that all puzzles have to be correctly solved. There we use Observation 5.4 to conclude that the remaining puzzles should be chosen such that they are all correctly solved by C.

Here, we generalize this approach. Namely, we search for the randomness used to generate the remaining k-1 puzzles such that when a puzzle on the first position was solved successfully, then the k-wise direct product would be solved

successfully, and if a puzzle on the first position was unsuccessfully solved, then the k-wise direct product would be also solved unsuccessfully. For a function g which requires all puzzles to be solved correctly, this approach corresponds exactly to the one presented in Section 5.1 for weakly verifiable puzzles.

All puzzles except for the input puzzle are generated by  $\widehat{G}$ en or  $\widehat{D}$ . Therefore, it is possible to answer all hint and verification queries concerning these puzzles. For the input puzzle to answer hint and verification queries we have to use the hint and verification oracle respectively.

The above described approach cannot be directly used in the context of dynamic puzzles. Not only, we have to find a good position for the input puzzles, but also the good randomness from which the remaining puzzles are generated. This requires running C several times. In one of this runs C can ask a hint query which prevents a later verification query from succeeding. This decreases the success probability of C in the consecutive calls.

**Technique of [DIJK09]** More precisely, we would like to ensure that no hint query ever asked that could prevent a verification query from being successful. To satisfy this requirement we used technique described in [DIJK09]. Intuitively, we partition Q into two sets. The set of advice queries contains these  $q \in Q$  on which C is allowed to ask hint queries. The set of attacking queries is the set on which C can ask successful verification queries. It turns out that it is possible to use a natural sampling technique to find a function that partitions Q such that the success probability of C with respect to the new partitioned domain Q is still substantial.

#### Putting it together

**TODO:** describe our contribution

### 6.1.3 Domain partitioning

Let  $hash: \mathcal{Q} \to \{0, 1, \dots, 2(h+v)-1\}$ , the idea is to partition  $\mathcal{Q}$  such that the set of preimages of 0 for hash contains  $q \in \mathcal{Q}$  on which C is not allowed to ask hint queries, and the first successful verification query (q, y) of C is such that hash(q) = 0. Therefore, if C makes a verification query (q, y) such that hash(q) = 0, then we know that no hint query is ever asked on this q.

We denote the *i*-th query of C by  $q_i$  if it is a hint query and by  $(q_i, y_i)$  if it is a verification query. Let us define the following experiment CanonicalSuccess in which Q is partitioned using a function hash. We say that a solver circuit succeeds in the experiment CanonicalSuccess if it asks a successful verification query  $(q_j, y_j)$  such that  $hash(q_j) = 0$ , and no hint query  $q_i$  has been asked before  $(q_i, y_i)$  such that  $hash(q_i) = 0$ .

```
Experiment CanonicalSuccess^{P,C,hash}(\pi,\rho)

Oracle: A problem poser P, a solver circuit C = (C_1, C_2) for P, a function hash: \mathcal{Q} \to \{0,1,\dots,2(h+v)-1\}.

Input: Bitstrings \pi \in \{0,1\}^n, \rho \in \{0,1\}^*.

Output: A bit b \in \{0,1\}.

run \langle P(\pi), C_1(\rho) \rangle
(\Gamma_V, \Gamma_H) := \langle P(\pi), C_1(\rho) \rangle_P
x := \langle P(\pi), C_1(\rho) \rangle_{trans}

run C_2^{\Gamma_V, \Gamma_H}(x, \rho)
if C_2^{\Gamma_V, \Gamma_H}(x, \rho) does not succeed for any verification query then return 0
Let (q_j, y_j) be the first verification query such that \Gamma_V(q_j, y_j) = 1.

if (\forall i < j : hash(q_i) \neq 0) and (hash(q_j) = 0) then return 1 else return 0
```

We define the *canonical success probability* of a solver circuit C for P with respect to a function hash as

$$\Pr_{\pi,\rho}[CanonicalSuccess^{P,C,hash}(\pi,\rho)=1]. \tag{6.1}$$

For fixed hash and P a canonical success of C for bistrings  $\pi$ ,  $\rho$  is a situation where Canonical Success  $P,C,hash(\pi,\rho)=1$ .

We show now that if a solver circuit C for P often succeeds in the experiment Success, then there exists a function hash such that C also often succeeds in the experiment CanonicalSuccess provided that the number of hint and verification queries is not too large.

**Lemma 6.1** (Success probability in solving a dynamic interactive weakly verifiable puzzle with respect to a function hash). For a fixed problem poser  $P_n$  let C be a solver for  $P_n$  with the success probability at least  $\gamma$ , asking at most h hint queries and v verification queries. Moreover, let  $\mathcal{H}$  be a family of pairwise independent efficient hash functions  $Q \to \{0, 1, \dots, 2(h+v)-1\}$ . There exists a probabilistic algorithm FindHash that takes as input parameters  $\gamma$ , n, h, v, and has oracle access to C and  $P_n$ . Furthermore, FindHash runs in time polynomial in  $(h, v, \frac{1}{\gamma}, n)$  and with high probability outputs a function  $hash \in \mathcal{H}$  such that the canonical success probability of C with respect to hash is at least  $\frac{\gamma}{16(h+v)}$ .

Proof of Lemma 6.1. We fix a problem poser P and a solver C for P in the whole proof of Lemma 6.1. For  $k, l \in \{1, \ldots, (h+v)\}$  and  $\alpha, \beta \in \{0, 1, \ldots, 2(h+v)-1\}$  by the pairwise independence property, we have that for every  $q_k, q_l \in \mathcal{Q}$  such that  $q_k \neq q_l$  the following holds

$$\Pr_{hash \leftarrow \mathcal{H}}[hash(q_k) = \alpha \mid hash(q_l) = \beta] = \Pr_{hash \leftarrow \mathcal{H}}[hash(q_k) = \alpha]$$

$$= \frac{1}{2(h+v)}.$$
(6.2)

We write  $\mathcal{P}_{Success}$  to denote a set containing all  $(\pi, \rho)$  for which  $Success^{P,C}(\pi, \rho) = 1$ . Let us fix  $(\pi^*, \rho^*) \in \mathcal{P}_{Success}$ . We are interested in the probability over a choice of hash of the event  $CanonicalSuccess^{P,C,hash}(\pi^*, \rho^*) = 1$ . Let  $(q_j, y_j)$  denote the first verification query such that  $\Gamma_V(q_j, y_j) = 1$ , we have

$$\Pr_{hash \leftarrow \mathcal{H}} \left[ CanonicalSuccess^{P,C,hash}(\pi^*, \rho^*) = 1 \right] \\
= \Pr_{hash \leftarrow \mathcal{H}} \left[ hash(q_j) = 0 \land (\forall i < j : hash(q_i) \neq 0) \right] \\
= \Pr_{hash \leftarrow \mathcal{H}} \left[ \forall i < j : hash(q_i) \neq 0 \mid hash(q_j) = 0 \right] \Pr_{hash \leftarrow \mathcal{H}} \left[ hash(q_j) = 0 \right] \\
\stackrel{(6.2)}{=} \frac{1}{2(h+v)} \left( 1 - \Pr_{hash \leftarrow \mathcal{H}} \left[ \exists i < j : hash(q_i) = 0 \mid hash(q_j) = 0 \right] \right) \\
\stackrel{(*)}{\geq} \frac{1}{2(h+v)} \left( 1 - \sum_{i < j} \Pr_{hash \leftarrow \mathcal{H}} \left[ hash(q_i) = 0 \mid hash(q_j) = 0 \right] \right) \\
\stackrel{(6.2)}{=} \frac{1}{2(h+v)} \left( 1 - \sum_{i < j} \Pr_{hash \leftarrow \mathcal{H}} \left[ hash(q_i) = 0 \right] \right) \\
\stackrel{(6.2)}{\geq} \frac{1}{4(h+v)}, \qquad (6.3)$$

where in (\*) we used the union bound. Let us denote the set of those  $(\pi, \rho)$  for which  $CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1$  by  $\mathcal{P}_{Canonical}$ . If for  $\pi$ ,  $\rho$  the circuit C succeeds canonically, then for the same  $\pi$ ,  $\rho$  we also have  $Success^{P,C}(\pi, \rho) = 1$ . Hence,  $\mathcal{P}_{Canonical} \subseteq \mathcal{P}_{Success}$ , and we conclude

$$\begin{aligned} &\Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \right] \\ &= \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \in \mathcal{P}_{Success} \right] \Pr_{\pi, \rho} [(\pi, \rho) \in \mathcal{P}_{Success}] \\ &+ \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \notin \mathcal{P}_{Success} \right] \Pr_{\pi, \rho} [(\pi, \rho) \notin \mathcal{P}_{Success}] \\ &= \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \in \mathcal{P}_{Success} \right] \Pr_{\pi, \rho} [(\pi, \rho) \in \mathcal{P}_{Success}] \end{aligned}$$

$$\geq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \in \mathcal{P}_{Success} \right] \cdot \gamma$$

$$= \mathbb{E}_{(\pi, \rho) \in \mathcal{P}_{Success}} \left[ \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}} [CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1] \right] \cdot \gamma$$

$$\geq \frac{\gamma}{4(h+v)}. \tag{6.4}$$

```
Algorithm FindHash<sup>P,C</sup>(\gamma, n, h, v)
```

**Oracle:** A problem poser P, a solver circuit C for P.

**Input:** Parameters  $\gamma$ , n. The number of hint queries h and of verification queries v.

**Output:** A function  $hash : \mathcal{Q} \to \{0, 1, \dots, 2(h+v) - 1\}.$ 

We show that FindHash chooses  $hash \in \mathcal{H}$  such that the canonical success probability of C with respect to hash is at least  $\frac{\gamma}{16(h+v)}$  almost surely. Let  $\mathcal{H}_{Good}$  denote a family of functions  $hash \in \mathcal{H}$  for which

$$\Pr_{\pi,\rho} \left[ Canonical Success^{P,C,hash}(\pi,\rho) = 1 \right] \ge \frac{\gamma}{8(h+v)}, \tag{6.5}$$

and  $\mathcal{H}_{Bad}$  be a family of functions  $hash \in \mathcal{H}$  such that

$$\Pr_{\pi,\rho} \left[ CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1 \right] \le \frac{\gamma}{16(h+v)}. \tag{6.6}$$

Let N denote the number of iterations of the inner loop of FindHash. We consider a single iteration of the outer loop of FindHash in which hash is fixed. Let us define independent and identically distributed binary random variables  $X_1, \ldots, X_N$  such that

$$X_i = \begin{cases} 1 & \text{if in the } i\text{-th iteration of the inner loop } count \text{ is increased} \\ 0 & \text{otherwise.} \end{cases}$$

We turn now to the case where  $hash \in \mathcal{H}_{Bad}$  and show that it is unlikely that  $hash \in \mathcal{H}_{Bad}$  is returned by FindHash. From (6.6) it follows that  $\mathbb{E}_{\pi,\rho}[X_i] \leq \frac{\gamma}{16(h+v)}$ . In the following inequalities (6.7) and (6.8) in steps denoted with (\*) we use the trivial facts  $h + v \geq 1$  and  $\gamma \leq 1$ . For any fixed  $hash \in \mathcal{H}_{Bad}$  using the Chernoff bound we get

$$\Pr_{\pi,\rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \ge \frac{\gamma}{12(h+v)} \right] \le \Pr_{\pi,\rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \ge \left(1 + \frac{1}{3}\right) \mathbb{E}[X_i] \right] \\
\le e^{-\frac{\gamma}{16(h+v)} N/27} \le e^{-\frac{2}{27} \frac{(h+v)}{\gamma} n} \stackrel{(*)}{\le} e^{-\frac{2}{27} n}.$$
(6.7)

The probability that  $hash \in \mathcal{H}_{Good}$ , when picked, is not returned amounts

$$\Pr_{\pi,\rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \le \frac{\gamma}{12(h+v)} \right] \le \Pr_{\pi,\rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \le \left(1 - \frac{1}{3}\right) \mathbb{E}[X_i] \right] \\
\le e^{-\frac{\gamma}{8(h+v)}N/18} \le e^{-\frac{2}{9} \frac{(h+v)}{\gamma} n} \stackrel{(*)}{\le} e^{-\frac{2}{9}n} \tag{6.8}$$

where we once more used the Chernoff bound. We show now that the probability of picking  $hash \in \mathcal{H}_{Good}$  is at least  $\frac{\gamma}{8(h+v)}$ . To obtain a contradiction suppose that

$$\Pr_{hash \leftarrow \mathcal{H}}[hash \in \mathcal{H}_{Good}] < \frac{\gamma}{8(h+v)}.$$
 (6.9)

Using (6.9) we conclude that it is possible to bound the probability of the canonical success as follows

$$\begin{aligned} &\Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1] \\ &= \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid hash \in \mathcal{H}_{Good}] \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &+ \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid hash \notin \mathcal{H}_{Good}] \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \notin \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] + \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid hash \notin \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] + \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid hash \notin \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] + \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid hash \notin \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] + \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid hash \notin \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] + \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid hash \notin \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] + \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ hash \leftarrow \mathcal{H}}}[hash \in \mathcal{H}_{Good}] \\ &\leq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ ha$$

which contradicts (6.4). Hence, we conclude that the probability of choosing  $hash \in \mathcal{H}_{Good}$  amounts at least  $\frac{\gamma}{8(h+v)}$ .

We show that FindHash picks in one of its iteration  $hash \in \mathcal{H}_{Good}$  almost surely. Let K denote the random variable that takes value equal to the number

of iterations of the outer loop of FindHash and  $Y_i$  be a random variable for the event that in the *i*-th iteration of the outer loop  $hash \notin \mathcal{H}_{Good}$  is picked. We use  $\Pr_{hash \leftarrow \mathcal{H}}[hash \in \mathcal{H}_{Good}] \geq \frac{\gamma}{8(h+v)}$  and  $K \leq \frac{32(h+v)^2}{\gamma^2}n$  to conclude

$$\Pr_{hash \leftarrow \mathcal{H}} \left[ \bigcap_{1 \le i \le K} Y_i \right] \le \left( 1 - \frac{\gamma}{8(h+v)} \right)^{\frac{32(h+v)^2}{\gamma^2} n} \le e^{-\frac{\gamma}{8(h+v)} \frac{32(h+v)^2}{\gamma^2} n}$$

$$\le e^{-\frac{4(h+v)}{\gamma} n} \le e^{-n}.$$

It is clear that the running time of FindHash is  $poly(n, h, v, \gamma)$  with oracle calls. This finishes the proof of Lemma 6.1.

#### 6.1.4 The hardness amplification proof for partitioned domains

Let  $C = (C_1, C_2)$  be a solver circuit for a dynamic interactive weakly verifiable puzzle as in Definition 3.1. We write  $C_2^{(\cdot,\cdot)}$  to emphasize that  $C_2$  does not obtain direct access to hint and verification circuits. Instead, whenever  $C_2$  asks a hint or verification query, it is answered explicitly as in the following code excerpt of the circuit  $\widetilde{C}_2$ .

Given  $C = (C_1, C_2)$  we define the circuit  $\widetilde{C} = (C_1, \widetilde{C}_2)$ . Every hint query q asked by  $\widetilde{C}$  is such that  $hash(q) \neq 0$ . Furthermore,  $\widetilde{C}$  asks no verification queries. Instead, it returns (q, y) such that hash(q) = 0 or  $\bot$ .

```
Circuit \widetilde{C}_2^{\Gamma_H,C_2,hash}(x,\rho)

Oracle: A hint circuit \Gamma_H, a circuit C_2,
   a function hash: \mathcal{Q} \to \{0,1,\dots,2(h+v)-1\}.

Input: Bitstrings x \in \{0,1\}^*, \rho \in \{0,1\}^*.

Output: A pair (q,y) where q \in \mathcal{Q} and y \in \{0,1\}^*.

run C_2^{(\cdot,\cdot)}(x,\rho)

if C_2^{(\cdot,\cdot)}(x,\rho) asks a hint query on q then
   if hash(q) = 0 then
    return \bot

else
   answer the query of C_2^{(\cdot,\cdot)}(x,\rho) using \Gamma_H(q)

if C_2^{(\cdot,\cdot)}(x,\rho) asks a verification query (q,y) then
   if hash(q) = 0 then
   return (q,y)

else
   answer the verification query of C_2^{(\cdot,\cdot)}(x,\rho) with 0
```

#### return $\perp$

For fixed  $\pi$ ,  $\rho$ , and hash we say that the circuit  $\widetilde{C}$  succeeds if for  $x := \langle P(\pi), C_1(\rho) \rangle_{trans}$ ,  $(\Gamma_V, \Gamma_H) := \langle P(\pi), C_1(\rho) \rangle_P$ , we have

$$\Gamma_V(\widetilde{C}_2^{\Gamma_H,C_2,hash}(x,\rho)) = 1.$$

**Lemma 6.2.** For fixed P, C, and hash the following statement is true

$$\Pr_{\pi,\rho}[CanonicalSuccess^{P,C,hash}(\pi,\rho)=1] \leq \Pr_{\pi,\rho}[\Gamma_V(\widetilde{C}_2^{\Gamma_H,C_2,hash}(x,\rho))=1].$$

$$x:=\langle P(\pi),C_1(\rho)\rangle_{trans}$$

$$(\Gamma_V,\Gamma_H):=\langle P(\pi),C_1(\rho)\rangle_P$$

*Proof.* If for some  $\pi$ ,  $\rho$ , and hash a circuit  $C = (C_1, C_2)$  succeeds canonically, then for the same  $\pi$ ,  $\rho$ , and hash a circuit  $\widetilde{C} := (C_1, \widetilde{C}_2)$  also succeeds. Using this observation, we conclude that

$$\Pr_{\pi,\rho} \left[ CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1 \right]$$

$$\leq \mathop{\mathbb{E}}_{\pi,\rho} \left[ \Gamma_V(\widetilde{C}_2^{\Gamma_H,C_2,hash}(x,\rho)) = 1 \right]$$

$$x:=\langle P(\pi),C_1(\rho)\rangle_{trans}$$

$$(\Gamma_V,\Gamma_H):=\langle P(\pi),C_1(\rho)\rangle_P$$

$$= \mathop{\Pr}_{\pi,\rho} \left[ \Gamma_V(\widetilde{C}_2^{\Gamma_H,C_2,hash}(x,\rho)) = 1 \right]$$

$$x:=\langle P(\pi),C_1(\rho)\rangle_{trans}$$

$$(\Gamma_V,\Gamma_H):=\langle P(\pi),C_1(\rho)\rangle_P$$

Lemma 6.3. (Hardness amplification for a dynamic interactive weakly verifiable puzzle with respect to hash.) Let  $g: \{0,1\}^k \to \{0,1\}$  be a monotone function,  $P_n^{(1)}$  a fixed problem poser and  $\widetilde{C} := (C_1, \widetilde{C}_2)$  a circuit with oracle access to a function hash:  $Q \to \{0,1,\ldots,2(h+v)-1\}$  and a solver  $C := (C_1,C_2)$  for  $P_{kn}^{(g)}$  which asks at most h hint queries and v verification queries. There exists an algorithm Gen that takes as input parameters  $\varepsilon$ ,  $\delta$ , n, k, has oracle access to  $P_n^{(1)}$ ,  $\widetilde{C}$ , hash, g, and outputs a circuit  $D := (D_1,D_2)$ 

If C is such that

such that the following holds:

$$\Pr_{\substack{\pi^{(k)} \in \{0,1\}^{kn}, \rho \in \{0,1\}^{s} \\ x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{trans} \\ (\Gamma_H^{(k)}, \Gamma_V^{(g)}) := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{P(g)}}} \Pr_{\substack{x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{P(g)} \\ }} \Pr_{\substack{x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{P(g)} \\ }} [g(u) = 1] + \varepsilon,$$

then D satisfies almost surely

```
\begin{aligned} &\Pr[\Gamma_V(D_2^{P^{(1)},\widetilde{C},hash,g,\Gamma_H}(x,\rho)) = 1] \geq \delta + \frac{\varepsilon}{6k}.\\ &x{:=} \langle P^{(1)}(\pi), D_1^{\widetilde{C}}(\rho) \rangle_{trans}\\ &(\Gamma_H,\Gamma_V){:=} \langle P^{(1)}(\pi), D_1^{\widetilde{C}}(\rho) \rangle_{P^{(1)}} \end{aligned}
```

Furthermore, D asks at most  $\frac{6k}{\varepsilon} \log \left(\frac{6k}{\varepsilon}\right) h$  hint queries and no verification queries. Finally, the running time of Gen is polynomial in  $k, \frac{1}{\varepsilon}, n$  with oracle calls.

We note that the circuit D from Lemma 6.3 instead of asking verification queries outputs a pair (q, y) such that hash(q) = 0 or  $\bot$ .

Before we give the proof of Lemma 6.3 we define additional algorithms. First, in the following code listing the algorithm Gen from Lemma 6.3 is defined. The procedures used by Gen are presented on the succeeding code listings.

```
Algorithm \operatorname{Gen}^{P^{(1)},\widetilde{C},g,hash}(\varepsilon,\delta,n,k)

Oracle: A poser P^{(1)}, a solver \widetilde{C} for P^{(g)}, functions g:\{0,1\}^k \to \{0,1\}, hash: \mathcal{Q} \to \{0,1,\dots,2(h+v)-1\}.

Input: Parameters \varepsilon,\delta,n,k.

Output: A circuit D.

for i:=1 to \frac{6k}{\varepsilon}n do:
\pi^* \overset{\$}{\leftarrow} \{0,1\}^n
\widetilde{S}_{\pi^*,0} := \operatorname{EstimateSurplus}^{P^{(1)},\widetilde{C},g,hash}(\pi^*,0,k,\varepsilon,\delta,n)
\widetilde{S}_{\pi^*,1} := \operatorname{EstimateSurplus}^{P^{(1)},\widetilde{C},g,hash}(\pi^*,1,k,\varepsilon,\delta,n)
if \exists b \in \{0,1\}: \widetilde{S}_{\pi^*,b} \geq (1-\frac{3}{4k})\varepsilon then

Let C_1' have oracle access to \widetilde{C}, and have hard-coded \pi^*.
Let \widetilde{C}_2' have oracle access to \widetilde{C}, and have hard-coded \pi^*.
\widetilde{C}' := (C_1',\widetilde{C}_2')
g'(b_2,\dots,b_k) := g(b,b_2,\dots,b_k)
return \operatorname{Gen}^{P^{(1)},\widetilde{C}',g',hash}(\varepsilon,\delta,n,k-1)
// all estimates are lower than (1-\frac{3}{4k})\varepsilon
```

We are interested in the probability that for  $u \leftarrow \mu_{\delta}^k$  and a bit b we have  $g(b, u_2, \ldots, u_k) = 1$ . The estimate of this probability is calculated by the algorithm EstimateFunctionProbability.

### **Algorithm** EstimateFunctionProbability $^g(b, k, \varepsilon, \delta, n)$

**Oracle:** A function  $g : \{0, 1\}^k \to \{0, 1\}$ .

**Input:** A bit  $b \in \{0,1\}$ , parameters  $k, \varepsilon, \delta, n$ .

**Output:** An estimate  $\widetilde{g}_b$  of  $\Pr_{u \leftarrow \mu_s^k}[g(b, u_2, \dots, u_k) = 1]$ .

for 
$$i:=1$$
 to  $N:=\frac{64k^2}{\varepsilon^2}n$  do:  $u\leftarrow\mu_\delta^k$   $g_i:=g(b,u_2,\ldots,u_k)$  return  $\frac{1}{N}\sum_{i=1}^N g_i$ 

For fixed  $\pi^{(k)}$ ,  $\rho$ , and hash we say that the circuit  $\widetilde{C} := (C_1, \widetilde{C}_2)$  succeeds on the i-th coordinate if for  $x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{trans}$ ,  $(\Gamma_V^{(g)}, \Gamma_H^{(k)}) := \langle P^{(g)}(\pi), C_1(\rho) \rangle_{P^{(g)}}$  and  $(q, y^{(k)}) := \widetilde{C}_2(x, \rho)$  we have

$$\Gamma_V^i(q, y_i) = 1.$$

**Lemma 6.4.** The algorithm EstimateFunctionProbability<sup>g</sup> $(b, k, \varepsilon, \delta, n)$  outputs an estimate  $\widetilde{g}_b$  such that  $|\widetilde{g}_b - \Pr_{u \leftarrow \mu_k^k}[g(b, u_2, \dots, u_k) = 1]| \leq \frac{\varepsilon}{8k}$  almost surely.

*Proof.* We fix notation as in the code excerpt of the algorithm EstimateFunctionProbability. Let us define independent and identically distributed binary random variables  $K_1, K_2, \ldots, K_N$  such that for each  $1 \leq i \leq N$  the random variable  $K_i$  takes value  $g_i$ . We use the Chernoff bound to obtain

$$\Pr\left[\left|\widetilde{g}_{b} - \Pr_{u \leftarrow \mu_{\delta}^{k}}\left[g(b, u_{2}, \dots, u_{k}) = 1\right]\right| \geq \frac{\varepsilon}{8k}\right]$$

$$= \Pr\left[\left|\left(\frac{1}{N}\sum_{i=1}^{N}K_{i}\right) - \mathbb{E}_{u \leftarrow \mu_{\delta}^{k}}\left[g(b, u_{2}, \dots, u_{k})\right]\right| \geq \frac{\varepsilon}{8k}\right] \leq 2 \cdot e^{-n/3}.$$

The algorithm EvalutePuzzles  $P^{(1)}, \tilde{C}, hash(\pi^{(k)}, \rho, n, k)$  evaluates which of the k puzzles of the k-wise direct product of  $P^{(1)}$  are solved successfully by  $\tilde{C}(\rho) := (C_1, \tilde{C}_2)(\rho)$ . To decide whether the i-th puzzle of the k-wise direct product is solved successfully we need to gain access to the verification circuit for the puzzle generated in the i-th round of the interaction between  $P^{(g)}$  and  $\tilde{C}$ . Therefore, the algorithm EvalutePuzzles runs k times  $P^{(1)}(\pi_i)$  to simulate the interaction with  $C_1(\rho)$  where in each round of interaction a fresh random bitstring  $\pi_i \in \{0,1\}^n$  is used.

Let us introduce some additional notation. We denote by  $\langle P^{(1)}(\pi_i), C_1(\rho) \rangle^i$  the execution of the *i*-th round of the sequential interaction. We use  $\langle P^{(1)}(\pi_i), C_1(\rho) \rangle^i_{P^{(1)}}$  to denote the output of  $P^{(1)}(\pi_i)$  in the *i*-th round. Finally, we write  $\langle P^{(1)}(\pi_i), C_1(\rho) \rangle^i_{trans}$  to denote the transcript of communication in the *i*-th round. We note that the *i*-th round of the interaction between  $P^{(1)}$  and  $C_1$  is well defined only if all previous rounds have been executed before.

For simplicity of notation in the code excerpts of circuits  $C_2$ ,  $D_2$ , and EvalutePuzzles we omit superscripts of some oracles. Exemplary, we write  $\widetilde{C}_2^{\Gamma_H^{(k)},hash}$  instead of  $\widetilde{C}_2^{\Gamma_H^{(k)},C,hash}$  where the superscript of the oracle circuit C is omitted. We make sure that it is clear from the context which oracles are used.

```
Algorithm EvaluatePuzzles P^{(1)}, \tilde{C}, hash(\pi^{(k)}, \rho, n, k)

Oracle: A problem poser P^{(1)}, a solver circuit \tilde{C} = (C_1, \tilde{C}_2) for P^{(g)}, a function hash: \mathcal{Q} \to \{0, 1, \dots, 2(h+v)-1\}.

Input: Bitstrings \pi^{(k)} \in \{0, 1\}^{kn}, \rho \in \{0, 1\}^*, parameters n, k.

Output: A tuple (c_1, \dots, c_k) \in \{0, 1\}^k.

for i := 1 to k do: /simulate \ k rounds of interaction (\Gamma_V^i, \Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{P^{(1)}}^i x_i := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{trans}^i x := (x_1, \dots, x_k) \Gamma_H^{(k)} := (\Gamma_H^1, \dots, \Gamma_H^k) (q, y_1, \dots, y_k) := \tilde{C}_2^{\Gamma_H^{(k)}, hash}(x, \rho) if (q, y_1, \dots, y_k) = \bot then return (0, \dots, 0) (c_1, \dots, c_k) := (\Gamma_V^1(q, y_1), \dots, \Gamma_V^k(q, y_k)) return (c_1, \dots, c_k)
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All puzzles used by EvalutePuzzles are generated internally. Thus, the algorithm can answer all queries of  $\widetilde{C}_2$  itself.

We are interested in the success probability of  $\widetilde{C}$  with the bitstring  $\pi_1$  fixed to  $\pi^*$  where the fact whether  $\widetilde{C}$  succeeds in solving the input puzzle defined by  $P^{(1)}(\pi_1)$  placed on the first position is neglected, and instead a bit b is used. More formally, we define the surplus  $S_{\pi^*,b}$  as

$$S_{\pi^*,b} = \Pr_{\pi^{(k)},\rho} \left[ g(b, c_2, \dots, c_k) = 1 \mid \pi_1 = \pi^* \right] - \Pr_{u \leftarrow \mu_{\delta}^k} \left[ g(b, u_2, \dots, u_k) = 1 \right],$$
(6.10)

where  $(c_2, c_3, \ldots, c_k)$  is obtained as in EvalutePuzzles.

The algorithm Estimate Surplus returns an estimate  $\widetilde{S}_{\pi^*,b}$  for  $S_{\pi^*,b}$ .

**Algorithm** EstimateSurplus  $P^{(1)}, \widetilde{C}, g, hash(\pi^*, b, k, \varepsilon, \delta, n)$ 

**Oracle:** A problem poser  $P^{(1)}$ , a circuit  $\widetilde{C}$  for  $P^{(g)}$ , functions  $g:\{0,1\}^k \to \{0,1\}$  and  $hash: \mathcal{Q} \to \{0,1,\ldots,2(h+v)-1\}$ .

**Input:** A bistring  $\pi^* \in \{0,1\}^n$ , a bit  $b \in \{0,1\}$ , parameters  $k, \varepsilon, \delta, n$ .

Output: An estimate  $\widetilde{S}_{\pi^*,b}$  for  $S_{\pi^*,b}$ .

 $\begin{aligned} & \textbf{for } i := 1 \textbf{ to } N := \frac{64k^2}{\varepsilon^2} n \textbf{ do:} \\ & (\pi_2, \dots, \pi_k) \overset{\$}{\leftarrow} \{0, 1\}^{(k-1)n} \\ & \rho \overset{\$}{\leftarrow} \{0, 1\}^* \\ & (c_1, \dots, c_k) := \text{EvalutePuzzles}^{P^{(1)}, \widetilde{C}, hash}((\pi^*, \pi_2, \dots, \pi_k), \rho, n, k) \\ & \widetilde{s}^i_{\pi^*, b} := g(b, c_2, \dots, c_k) \\ & \widetilde{g}_b := \text{EstimateFunctionProbability}^g(b, k, \varepsilon, \delta, n) \\ & \textbf{return } \left(\frac{1}{N} \sum_{i=1}^N \widetilde{s}^i_{\pi^*, b}\right) - \widetilde{g}_b \end{aligned}$ 

**Lemma 6.5.** The estimate  $\widetilde{S}_{\pi^*,b}$  returned by EstimateSurplus differs from  $S_{\pi^*,b}$  by at most  $\frac{\varepsilon}{4k}$  almost surely.

*Proof.* We use the union bound and similar argument as in Lemma 6.4 which yields that  $\frac{1}{N} \sum_{i=1}^{N} \widetilde{s}_{\pi^*,b}^i$  differs from  $\mathbb{E}[g(b,c_2,\ldots,c_k)]$  by at most  $\frac{\varepsilon}{8k}$  almost surely. Together, with Lemma 6.4 we conclude that the surplus estimate returned by EstimateSurplus differs from  $S_{\pi^*,b}$  by at most  $\frac{\varepsilon}{4k}$  with probability at least  $1-2e^{-n}$ .

We define the following solver circuit  $C' = (C'_1, C'_2)$  for the (k-1)-wise direct product of  $P^{(1)}$ .

Circuit  $C_1^{\prime \widetilde{C},P^{(1)}}(\rho)$ 

**Oracle:** A solver circuit  $\widetilde{C} = (C_1, \widetilde{C}_2)$  for  $P^{(g)}$ , a poser  $P^{(1)}$ .

**Input:** A bitstring  $\rho \in \{0, 1\}^*$ .

**Hard-coded:** A bitstring  $\pi^* \in \{0,1\}^n$ .

Simulate  $\langle P^{(1)}(\pi^*), C_1(\rho) \rangle^1$ 

Use  $C_1(\rho)$  for the remaining k-1 rounds of interaction.

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Circuit \widetilde{C}_{2}^{\Gamma_{H}^{(k-1)},\widetilde{C},hash}(x^{(k-1)},\rho)

Oracle: A hint oracle \Gamma_{H}^{(k-1)}:=(\Gamma_{H}^{2},\ldots,\Gamma_{H}^{k}),
a solver circuit \widetilde{C}=(C_{1},\widetilde{C}_{2}) for P^{(g)},
a function hash:\mathcal{Q}\to\{0,1,\ldots,2(h+v)-1\}.

Input: A transcript of k-1 rounds of interaction x^{(k-1)}:=(x_{2},\ldots,x_{k})\in\{0,1\}^{*}, a bitstring \rho\in\{0,1\}^{*}.

Hard-coded: A bitstring \pi^{*}\in\{0,1\}^{n}.

Simulate \langle P^{(1)}(\pi^{*}),C_{1}(\rho)\rangle^{1}
(\Gamma_{H}^{*},\Gamma_{V}^{*}):=\langle P^{(1)}(\pi^{*}),C_{1}(\rho)\rangle^{1}_{trans}
\Gamma_{H}^{(k)}:=(\Gamma_{H}^{*},\Gamma_{H}^{2},\ldots,\Gamma_{H}^{k})
x^{(k)}:=(x^{*},x_{2},\ldots,x_{k})
(q,y_{1},\ldots,y_{k}):=\widetilde{C}_{2}^{\Gamma_{H}^{(k)},hash}(x^{(k)},\rho)
return (q,y_{2},\ldots,y_{k})
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We are ready to define the solver circuit  $D = (D_1, D_2)$  for  $P^{(1)}$  output by Gen.

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Circuit D_1^{\widetilde{C}}(r)

Oracle: A solver circuit \widetilde{C} = (C_1, \widetilde{C}_2) for P^{(g)}.

Input: A pair r := (\rho, \sigma) where \rho \in \{0, 1\}^* and \sigma \in \{0, 1\}^*.

Interact with the problem poser \langle P^{(1)}, C_1(\rho) \rangle^1.

Let x^* := \langle P^{(1)}, C_1(\rho) \rangle^1_{trans}.
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Circuit D_2^{P^{(1)}, \widetilde{C}, hash, g, \Gamma_H}(x^*, r)

Oracle: A poser P^{(1)}, a solver circuit \widetilde{C} = (C_1, \widetilde{C}_2) for P^{(g)}, functions hash: \mathcal{Q} \to \{0, 1, \dots, 2(h+v)-1\}, g: \{0, 1\}^k \to \{0, 1\}, a hint circuit \Gamma_H for P^{(1)}.

Input: A communication transcript x^* \in \{0, 1\}^*, a bitstring r := (\rho, \sigma) where \rho \in \{0, 1\}^* and \sigma \in \{0, 1\}^*

Output: A pair (q, y^*).

for at most \frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon}) iterations do: (\pi_2, \dots, \pi_k) \leftarrow \text{read next } (k-1) \cdot n \text{ bits from } \sigma
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Use x^* to simulate the first round of interaction of C_1(\rho) with the problem poser P^{(1)}.

for i:=2 to k do:

 \operatorname{run} \langle P^{(1)}(\pi_i), C_1(\rho) \rangle^i 
 (\Gamma_V^i, \Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{P^{(1)}}^i 
 x_i := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{trans}^i 
 \Gamma_H^{(k)}(q) := (\Gamma_H(q), \Gamma_H^2(q), \dots, \Gamma_H^k(q)) 
 (q, y^*, y_2, \dots, y_k) := \widetilde{C}_2^{\Gamma_H^{(k)}, hash} ((x^*, x_2, \dots, x_k), \rho) 
 (c_2, \dots, c_k) := (\Gamma_V^2(q, y_2), \dots, \Gamma_V^k(q, y_k)) 
 \operatorname{if} g(1, c_2, \dots, c_k) = 1 \text{ and } g(0, c_2, \dots, c_k) = 0 \text{ then } 
 \operatorname{return} (q, y^*) 
 \operatorname{return} \bot
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of Lemma 6.3. First, let us consider the case where k=1. The function  $g:\{0,1\}\to\{0,1\}$  is either the identity or a constant function. In the latter case, when g is a constant function, Lemma 6.3 is vacuously true. If g is the identity function, then the circuit D returned by Gen directly uses  $\widetilde{C}$  to find a solution. From the assumptions of Lemma 6.3 it follows that  $\widetilde{C}$  succeeds with probability at least  $\delta+\varepsilon$ . Hence, D trivially satisfies Lemma 6.3.

For the general case, we consider two possibilities. Namely, either Gen in one of the iterations finds an estimate with high surplus such that  $\widetilde{S}_{\pi,b} \geq (1 - \frac{3}{4k})\varepsilon$  and recurses, or in all iterations it fails and outputs the circuit D.

If it is possible to find an estimate with high surplus, then we introduce a new monotone function  $g': \{0,1\}^{k-1} \to \{0,1\}$  such that  $g'(b_2,\ldots,b_k) := g(b,b_2,\ldots,b_k)$  and a new circuit  $\widetilde{C}' = (C'_1,\widetilde{C}'_2)$  with oracle access to  $\widetilde{C} := (C_1,\widetilde{C}_2)$ . W apply Lemma 6.5 and conclude that almost surely it holds

$$S_{\pi^*,b} \ge \widetilde{S}_{\pi^*,b} - \frac{\varepsilon}{4k} \ge \left(1 - \frac{1}{k}\right)\varepsilon.$$

It follows that  $\widetilde{C}'$  succeeds in solving the (k-1)-wise direct product of puzzles with probability at least

$$\Pr_{u \leftarrow \mu_{\delta}^{(k-1)}}[g'(u_1, \dots, u_{k-1})] + \left(1 - \frac{1}{k}\right)\varepsilon.$$

We see that in this case  $\widetilde{C}'$  satisfies the conditions of Lemma 6.3 for the (k-1)-wise direct product of puzzles. Therefore, the recursive call to Gen with access

to g' and  $\widetilde{C}$  returns  $D = (D_1, D_2)$  that with high probability satisfies

$$\Pr_{\substack{x:=\langle P^{(1)}(\pi), D_1^{\tilde{C}}(\rho)\rangle_{trans} \\ (\Gamma_H, \Gamma_V):=\langle P^{(1)}(\pi), D_1^{\tilde{C}}(\rho)\rangle_{P^{(1)}}}} = \delta + \left(1 - \frac{1}{k}\right) \frac{\varepsilon}{6(k-1)}$$

$$= \delta + \frac{\varepsilon}{6k}.$$
(6.11)

Let us bring our attention to the case where none of the estimates is greater than  $(1-\frac{3}{4k})\varepsilon$ . If all surpluses  $S_{\pi,0}$  and  $S_{\pi,1}$  were lower than  $(1-\frac{1}{k})\varepsilon$ , then it would mean that  $\widetilde{C}$  does not succeed on the remaining k-1 puzzles with much higher probability than an algorithm that solves each puzzle independently with success probability  $\delta$ . However, from the assumptions of Lemma 6.3 we know that on all k puzzles the success probability of  $\widetilde{C}$  is higher. Hence, we suspect that the first puzzle is correctly solved unusually often. It remains to show that the fact that Gen fails to find a surplus estimate that is large implies that with high probability there are only few surpluses greater than  $(1-\frac{1}{k})\varepsilon$  and their influence is can be neglected. Additionally, we have to show that the circuit D finds outputs an answer almost surely.

We fix notation as in the code listing of the circuit  $D_2$ . Let us consider a single iteration of the outer loop of  $D_2$  where values  $\pi_2, \ldots, \pi_k$  are fixed. We write  $\pi_1$  to denote randomness used by the problem poser to generate the input puzzle. Furthermore, we define  $c_1 := \Gamma_V(q, y_1)$  where  $\Gamma_V$  is the verification circuit generated by  $P^{(1)}(\pi_1)$  in the first phase when interacting with  $D_1(r)$ . We write  $c := (c_1, c_2, \ldots, c_k)$ , and for  $b \in \{0, 1\}$  we define a set

$$\mathcal{G}_b := \{(b_1, b_2, \dots, b_k) : g(b, b_2, \dots, b_k) = 1\}.$$

Using this notation we express

$$\Pr_{u \leftarrow \mu_{\delta}^{k}}[u \in \mathcal{G}_{b}] = \Pr_{u \leftarrow \mu_{\delta}^{k}}[g(b, u_{2}, \dots, u_{k}) = 1]$$

$$\Pr_{\pi^{(k)}, \rho}[c \in \mathcal{G}_{b}] = \Pr_{\pi^{(k)}, \rho}[g(b, c_{2}, \dots, c_{k}) = 1].$$
(6.12)

Let us fix randomness  $\pi_1$  used by the problem poser to generate the input puzzle to  $\pi^*$ . We use (6.10) and (6.12) to obtain

$$\Pr_{u \leftarrow \mu_{\delta}^{k}} [u \in \mathcal{G}_{1}] - \Pr_{u \leftarrow \mu_{\delta}^{k}} [u \in \mathcal{G}_{0}] 
= \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{1} \mid \pi_{1} = \pi^{*}] - \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - (S_{\pi^{*}, 1} - S_{\pi^{*}, 0}) \quad (6.13)$$

By monotonicity of g it holds  $\mathcal{G}_0 \subseteq \mathcal{G}_1$ , and we write (6.13) as

$$\Pr_{u \leftarrow \mu_{\delta}^{k}} [u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] = \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - (S_{\pi^{*}, 1} - S_{\pi^{*}, 0}).$$
 (6.14)

Let us multiply both sides of (6.14) by

$$\Pr_{\substack{x^* := \langle P^{(1)}(\pi^*), D_1(r) \rangle_{trans} \\ (\Gamma_V, \Gamma_H) := \langle P^{(1)}(\pi^*), D_1(r) \rangle_{P^{(1)}}}} / \Pr_{\substack{u \leftarrow \mu_\delta^k \\ Pr}} [u \in \mathcal{G}_1 \setminus \mathcal{G}_0],$$

which yields

$$\Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
= \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \Pr_{\pi^{(k)}, \rho} \left[ c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] \frac{1}{\Pr_{u \leftarrow \mu_{\delta}^{k}} \left[ u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right]} \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
- \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] (S_{\pi^{*}, 1} - S_{\pi^{*}, 0}) \frac{1}{\Pr_{u \leftarrow \mu_{\delta}^{k}} \left[ u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right]} \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
(6.15)$$

Let us study the first summand of (6.15). First, we have

$$\Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
= \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \middle| D_{2}(x^{*}, r) \neq \bot \right] \Pr_{r} \left[ D_{2}(x^{*}, r) \neq \bot \right] \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \qquad x^{*} = \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
\stackrel{(*)}{=} \Pr_{\pi^{(k)}, \rho} \left[ c_{1} = 1 \middle| c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}, \pi_{1} = \pi^{*} \right] \Pr_{r} \left[ D_{2}(x^{*}, r) \neq \bot \right] \quad (6.16) \\
x^{*} = \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans}$$

where in (\*) we use the observation that  $D_2(x^*, r) \neq \bot$  occurs if and only if  $D_2(x^*, r)$  finds  $\pi^{(k)}$  such that  $c \in \mathcal{G}_1 \setminus \mathcal{G}_0$ . Furthermore, conditioned on  $c \in \mathcal{G}_1 \setminus \mathcal{G}_0$  we have that  $\Gamma_V(D_2(x^*, r)) = 1$  happens if and only if  $c_1 = 1$ . Inserting (6.16) to the numerator of the first summand of (6.15) yields

$$\Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \Pr_{\pi^{(k)}, \rho} \left[ c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
= \Pr_{r} \left[ D_{2}(x^{*}, r) \neq \bot \right] \Pr_{\pi^{(k)}, \rho} \left[ c_{1} = 1 \mid c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}, \pi_{1} = \pi^{*} \right] \Pr_{\pi^{(k)}, \rho} \left[ c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] \\
x^{*} = \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \tag{6.17}$$

We consider the following two cases. First, if  $\Pr_{\pi^{(k)},\rho}[c \in \mathcal{G}_1 \setminus \mathcal{G}_0 \mid \pi_1 = \pi^*] \leq \frac{\varepsilon}{6k}$  then

$$\Pr_{\pi^{(k)}, \rho}[c_1 = 1 \mid c \in \mathcal{G}_1 \setminus \mathcal{G}_0, \pi_1 = \pi^*] \Pr_{\pi^{(k)}, \rho}[c \in \mathcal{G}_1 \setminus \mathcal{G}_0 \mid \pi_1 = \pi^*] \le \frac{\varepsilon}{6k}. \quad (6.18)$$

In case when  $\Pr_{\pi^{(k)},\rho}[c \in \mathcal{G}_1 \setminus \mathcal{G}_0 \mid \pi_1 = \pi^*] > \frac{\varepsilon}{6k}$  the circuit  $D_2$  outputs  $\bot$  if and only if it fails in all  $\frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon})$  iterations to find  $\pi^{(k)}$  such that  $c \in \mathcal{G}_1 \setminus \mathcal{G}_0$  which happens with probability

$$\Pr_{r}[D_{2}(x^{*},r) = \bot] \le \left(1 - \frac{\varepsilon}{6k}\right)^{\frac{6k}{\varepsilon}\log(\frac{6k}{\varepsilon})} \le \frac{\varepsilon}{6k}.$$

$$x^{*}:=\langle P^{(1)}(\pi^{*}), D_{1}(r)\rangle_{trans}$$
(6.19)

We conclude that in both aforementioned cases using (6.17), (6.18) and (6.19) the following holds

$$\Pr_{r} \left[ D_{2}(x^{*}, r) \neq \bot \right] \Pr_{\pi^{(k)}, \rho} \left[ c_{1} = 1 \mid c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}, \pi_{1} = \pi^{*} \right] \Pr_{\pi^{(k)}, \rho} \left[ c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] \\
\times^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
\geq \Pr_{\pi^{(k)}, \rho} \left[ c_{1} = 1 \mid c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}, \pi_{1} = \pi^{*} \right] \Pr_{\pi^{(k)}, \rho} \left[ c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] - \frac{\varepsilon}{6k} \\
= \Pr_{\pi^{(k)}, \rho} \left[ c_{1} = 1 \wedge c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] - \frac{\varepsilon}{6k} \\
= \Pr_{\pi^{(k)}, \rho} \left[ g(c) = 1 \mid \pi_{1} = \pi^{*} \right] - \Pr_{\pi^{(k)}, \rho} \left[ c \in \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] - \frac{\varepsilon}{6k} \\
\stackrel{(6.10)}{=} \Pr_{\pi^{(k)}, \rho} \left[ g(c) = 1 \mid \pi_{1} = \pi^{*} \right] - \Pr_{u \leftarrow \mu_{\delta}^{(k)}} \left[ u \in \mathcal{G}_{0} \right] - S_{\pi^{*}, 0} - \frac{\varepsilon}{6k} . \tag{6.20}$$

We insert (6.20) into (6.15) and calculate the expected value over  $\pi^*$  which yields

$$\Pr_{\substack{x := \langle P^{(1)}(\pi), D_{1}(r) \rangle_{trans} \\ (\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi), D_{1}(r) \rangle_{P^{(1)}}}} \left[ \frac{\Pr_{\pi^{(k)}, \rho}[g(c) = 1 | \pi_{1} = \pi^{*}] - \Pr_{u \leftarrow \mu_{\delta}^{(k)}}[u \in \mathcal{G}_{0}] - \frac{\varepsilon}{6k}}{\Pr_{u \leftarrow \mu_{\delta}^{(k)}}[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]} \right] \\
- \mathbb{E}_{\pi^{*}} \left[ \left( \Pr_{r}[\Gamma_{V}(D_{2}(x^{*}, r)) = 1](S_{\pi^{*}, 1} - S_{\pi^{*}, 0}) + S_{\pi^{*}, 0} \right) \frac{1}{\Pr_{u \leftarrow \mu_{\delta}^{(k)}}[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]} \right] \cdot \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}}$$
(6.21)

We show now that if Gen does not recurse, then the majority of estimates is low almost surely. Let us assume that

$$\Pr_{\pi} \left[ \left( S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon \right) \wedge \left( S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon \right) \right] < 1 - \frac{\varepsilon}{6k}, \tag{6.22}$$

then Gen recurses almost surely, because the probability that Gen does not find  $\widetilde{S}_{\pi,b} \geq (1 - \frac{3}{4k})\varepsilon$  in all of the  $\frac{6k}{\varepsilon}n$  iterations is at most

$$\left(1 - \frac{\varepsilon}{6k}\right)^{\frac{6k}{\varepsilon}n} \le e^{-n}$$

almost surely, where we used Lemma 6.5. Therefore, under the assumption that Gen does not recurse with high probability it holds

$$\Pr_{\pi,\rho}\left[\left(S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon\right) \land \left(S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon\right)\right] \ge 1 - \frac{\varepsilon}{6k}.\tag{6.23}$$

Let us define a set

$$W = \left\{ \pi : \left( S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon \right) \land \left( S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon \right) \right\}. \tag{6.24}$$

Additionally, let  $\overline{W}$  denote the complement of W. We bound the numerator of the second summand in (6.21)

$$\mathbb{E}_{\pi^*} \Big[ S_{\pi^*,0} + \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^*,r)) = 1 \right] (S_{\pi^*,1} - S_{\pi^*,0}) \Big]$$

$$x^* := \langle P^{(1)}(\pi^*), D_{1}(r) \rangle_{trans}$$

$$(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^*), D_{1}(r) \rangle_{P^{(1)}}$$

$$= \mathbb{E}_{\pi^*} \Big[ S_{\pi^*,0} + \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^*,r) = 1) \right] (S_{\pi^*,1} - S_{\pi^*,0}) \mid \pi^* \in \overline{\mathcal{W}} \Big] \Pr_{\pi^*} [\pi^* \in \overline{\mathcal{W}}]$$

$$x^* := \langle P^{(1)}(\pi^*), D_{1}(r) \rangle_{trans}$$

$$(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^*), D_{1}(r) \rangle_{P^{(1)}}$$

$$+ \mathbb{E}_{\pi^*} \Big[ S_{\pi^*,0} + \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^*,r)) = 1 \right] (S_{\pi^*,1} - S_{\pi^*,0}) \mid \pi^* \in \mathcal{W} \Big] \Pr_{\pi^*} [\pi^* \in \mathcal{W}]$$

$$x^* := \langle P^{(1)}(\pi^*), D_{1}(r) \rangle_{trans}$$

$$(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^*), D_{1}(r) \rangle_{P^{(1)}}$$

$$\leq \frac{\varepsilon}{6k} + (1 - \frac{1}{2k})\varepsilon = (1 - \frac{1}{3k})\varepsilon.$$

$$(6.25)$$

Finally, we insert (6.25) into (6.21) which yields

$$\Pr_{\substack{x := \langle P^{(1)}(\pi), D_1(\rho) \rangle_{\text{trans}} \\ (\Gamma_V, \Gamma_H) := \langle P^{(1)}(\pi), D_1(\rho) \rangle_{P^{(1)}}}} \Pr_{\substack{x := \langle P^{(1)}(\pi), D_1(\rho) \rangle_{P^{(1)}}}}} \Pr_{\substack{x := \langle P^{(1)}(\pi), D_1(\rho) \rangle_{P^{(1)}}}}} \Pr_{\substack{x := \langle P^{(1)}(\pi), D_1(\rho) \rangle_{P^{(1)}}}} \Pr_{\substack{x := \langle$$

From the assumptions of Lemma 6.3 it follows that

$$\Pr_{\pi^{(k)}, \rho}[g(c) = 1] \ge \Pr_{u \leftarrow \mu_{\delta}^{(k)}}[g(u) = 1] + \varepsilon. \tag{6.26}$$

We observe that

$$\Pr_{u \leftarrow \mu_{\delta}^{k}}[g(u) = 1] = \Pr[u \in \mathcal{G}_{0} \lor (u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \land u_{1} = 1)]$$

$$= \Pr[u \in \mathcal{G}_{0}] + \delta \Pr[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]. \tag{6.27}$$

Using (6.27) and (6.26) we obtain

$$\Pr_{\substack{x := \langle P^{(1)}(\pi), D_{1}(\rho) \rangle_{\text{trans}} \\ (\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi), D_{1}(\rho) \rangle_{P^{(1)}}}} \ge \frac{\Pr_{u \leftarrow \mu_{\delta}^{k}}[g(u) = 1] + \varepsilon - \Pr_{u \leftarrow \mu_{\delta}^{k}}[u \in \mathcal{G}_{0}] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{u \leftarrow \mu_{\delta}^{k}}[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]} \\
\ge \frac{(6.27)}{2} \frac{\varepsilon + \delta \Pr_{u \leftarrow \mu_{\delta}^{k}}[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{u \leftarrow \mu_{\delta}^{k}}[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]} \ge \delta + \frac{\varepsilon}{6k}.$$
(6.28)

Clearly, the running time of Gen is bounded by some polynomial  $p(k, \frac{1}{\varepsilon}, n)$  with oracle calls. Furthermore, the algorithm Gen outputs a circuit D such that it satisfies with probability at least  $1 - p(k, \frac{1}{\varepsilon}, n) \cdot 2^n$  the statement of Lemma 6.3. This concludes the proof of Lemma 6.3.

### 6.1.5 Putting it together

In the previous sections we have shown that it is possible to partition the domain Q such that if the number of hint and verification queries is small, then success probability of a solver for the k-wise direct product is still substantial. As shown in Lemma 6.3 we can build a circuit that returns a solution that is likely to be good. It remains to use these building blocks and prove Theorem 3.3.

of Theorem 3.3. Let  $\widetilde{\text{Gen}}$  be the algorithm from Theorem 3.3 that outputs the solver circuit  $\widetilde{D}$  for P. First, in the following code listing we define  $\widetilde{\text{Gen}}$ .

Algorithm  $\widetilde{\operatorname{Gen}}^{P^{(1)},g,C}(n,\varepsilon,\delta,k,h,v)$ 

**Oracle:** A problem poser  $P^{(1)}$ , a function  $g: \{0,1\}^k \to \{0,1\}$ , a solver circuit C for  $P^{(g)}$ .

Input: Parameters  $n, \varepsilon, \delta, k, h, v$ .

**Return:** A circuit  $\widetilde{D} = (D_1, \widetilde{D}_2)$ .

 $hash := \text{FindHash}((h+v)\varepsilon, n, h, v)$ 

Let  $\widetilde{C} := (C_1, \widetilde{C}_2)$  be as in Lemma 6.2 with oracle access to C, hash.

 $D := Gen^{P^{(1)}, \widetilde{C}, g, hash}(\varepsilon, \delta, n, k)$ 

return  $\widetilde{D} := (D_1, \widetilde{D}_2)$ 

Circuit  $\widetilde{D}^{D,P^{(1)},hash,g,\Gamma_{V},\Gamma_{H}}_{2}(x,\rho)$ 

**Oracle:** A circuit  $D := (D_1, \widetilde{D}_2)$  from Lemma 6.3, a problem poser  $P^{(1)}$ ,

functions  $hash: Q \to \{0, 1, \dots, 2(h+v)-1\}, g: \{0, 1\}^k \to \{0, 1\}$  a verification oracle  $\Gamma_V$ , a hint oracle  $\Gamma_H$ .

**Input:** Bitstrings  $x \in \{0,1\}^*, \rho \in \{0,1\}^*$ .

$$(q,y) := D_2^{P^{(1)},\widetilde{C},hash,g,\Gamma_H}(x,\rho)$$
  
Ask verification query  $(q,y)$  to  $\Gamma_V$ .

We show that Theorem 3.3 follows from Lemma 6.1 and Lemma 6.3. We fix  $P^{(1)}$ , g,  $P^{(g)}$  in the whole proof and consider a solver circuit  $C = (C_1, C_2)$ , asking at most h hint queries and v verification queries such that

$$\Pr_{\pi^{(k)},\rho} \left[ Success^{P^{(g)},C}(\pi^{(k)},\rho) = 1 \right] \ge 16(h+v) \Big( \Pr_{u \leftarrow \mu_{\delta}^k} \left[ g(u) = 1 \right] + \varepsilon \Big).$$

First, we note that C meets the requirements of Lemma 6.1. Furthermore, trivially success probability of C is at least  $(h+v)\varepsilon$ . Thus, the algorithm Gen can call FindHash to obtain  $hash: \mathcal{Q} \to \{0,1,\ldots,2(h+v)-1\}$  such that with high probability it holds

$$\Pr_{\pi^{(k)},\rho} \left[ Canonical Success^{P^{(g)},C,hash}(\pi^{(k)},\rho) = 1 \right] \geq \Pr_{u \leftarrow \mu_{\delta}^{k}} \left[ g(u) = 1 \right] + \varepsilon.$$

Additionally, the running time of FindHash is  $poly(h, v, \frac{1}{\varepsilon}, n)$  with oracle calls. Applying Lemma 6.2 for  $C = (C_1, C_2)$  we obtain a circuit  $\widetilde{C} = (C_1, \widetilde{C}_2)$  that satisfies

$$\Pr_{\substack{\pi^{(k)}, \rho \\ x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{trans} \\ (\Gamma_V^{(g)}, \Gamma_H^{(k)}) := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{P^{(g)}}}} [\Gamma_V^{(g)}(\widetilde{C}_2^{\Gamma_H^{(k)}, C_2, hash}(x, \rho)) = 1] \ge \Pr_{\substack{u \leftarrow \mu_\delta^k \\ 0 < u < \mu_\delta}} [g(u) = 1] + \varepsilon.$$

Now, we use the algorithm Gen as in Lemma 6.3 that yields a circuit  $D = (D_1, D_2)$  which with high probability satisfies

$$\Pr_{\pi,\rho} \left[ \Gamma_{V} \left( D_{2}^{P^{(1)}, \tilde{C}, hash, g, \Gamma_{H}}(x, \rho) \right) = 1 \right] \ge \left( \delta + \frac{\varepsilon}{6k} \right).$$

$$x := \langle P^{(1)}(\pi), D_{1}^{\tilde{C}}(\rho) \rangle_{trans}$$

$$(\Gamma_{H}, \Gamma_{V}) := \langle P^{(1)}(\pi), D_{1}^{\tilde{C}}(\rho) \rangle_{P^{(1)}}$$
(6.29)

Finally,  $\widetilde{\text{Gen}}$  outputs  $\widetilde{D} = (D_1, \widetilde{D}_2)$  with oracle access to  $D, P^{(1)}, hash, g$  such that with high probability it holds

$$\Pr_{\pi,\rho} \left[ Success^{P^{(1)},\widetilde{D}}(\pi,\rho) = 1 \right] \ge \delta + \frac{\varepsilon}{6k}.$$

The running time of Gen is polynomial in  $k, \frac{1}{\varepsilon}, n)$  with oracle calls. Thus, the overall running time of  $\widetilde{Gen}$  is polynomial in  $k, \frac{1}{\varepsilon}, h, v, n$  with oracle access. Furthermore, the circuit  $\widetilde{D}$  asks at most  $\frac{6k}{\varepsilon}\log(\frac{6k}{\varepsilon})h$  hint queries and one verification query. Finally, we have  $Size(\widetilde{D}) \leq Size(C) \cdot \frac{6k}{\varepsilon}$ . This finishes the proof of Theorem 3.3.

### 6.2 Discussion

#### 6.2.1 Optimality of the result

**TODO:** When does p.r. fails to a.h.

**TODO:** Why fixing all puzzles is not possible

**TODO:** Why we believe that it is not possible to do any better

**TODO:** Mention the number of the verification queries

**TODO:** Try to explain somehow the number of hint queries

**TOOD:** What we tried to prove the lower bound

**TOOD:** Why it was not possible

- no easy way to handle inefficient solver without visible effect on the number of hint queries
- complicated breaker + puzzles has to depend on each other,
- has to ask many hint and verification queries in each phase on the more less the same domain
- it interacts with the solver that is computationally bounded, hence limitation on the number of hint and verification queries
- the solver may be dishonest answer the queries with different (not true response)

### Appendix A

### **Appendix**

### A.1 Concentration Bounds

**Lemma A.1** (Chernoff Bounds). For independent and identically distributed Bernoulli random variables  $X_1, \ldots, X_n$  with  $X := \sum_{i=1}^n X_i$  and  $\Pr[X_i = 1] = p_i$  for all  $1 \le i \le n$  the following inequalities hold

$$\Pr[X \ge (1+\delta)\mathbb{E}[X]] \le e^{-\mathbb{E}[X]\delta^2/3} \tag{A.1}$$

$$\Pr[X \le (1 - \delta)\mathbb{E}[X]] \le e^{-\mathbb{E}[X]\delta^2/2} \tag{A.2}$$

$$\Pr[|X - \mathbb{E}[X]| \ge \delta \mathbb{E}[X]] \le 2e^{-\mathbb{E}[X]\delta^2/3},\tag{A.3}$$

where  $0 \le \delta \le 1$ .

# A.2 The Proof of Lemma 6.3 Under the Simplifying Assumptions

We give a proof of Lemma 6.3 where for the sake of simplicity we make the following assumptions

$$\Pr_{\pi,\rho}[D_2(x,\rho) \neq \bot] = 1$$

$$x := \langle P^{(1)}(\pi), D_1(\rho) \rangle_{trans}$$
(A.4)

$$\forall \pi \in \{0, 1\}^n : S_{\pi, b} \le (1 - \frac{1}{k}). \tag{A.5}$$

Loosely speaking, in (A.4) we assume that for every  $\pi$  circuit D outputs an answer, and in (A.5) we make the assumption that all surpluses are low. In the complete proof of Lemma 6.3 this assumptions fail only slightly such that it is still possible to obtain the desired result. However, the calculations are fairly lengthy. The following simplified calculations are intended to give the Reader better intuition about the full proof. We have

$$\begin{split} \Pr_{\pi,\rho}[\Gamma_{V}(D_{2}(x,\rho)) &= 1] \overset{(A.4)}{=} \Pr_{\pi,\rho}[\Gamma_{V}(D_{2}(x^{*},\rho)) = 1 \mid D_{2}(x,\rho) \neq \bot] \\ x &:= \langle P^{(1)}(\pi), D_{1}(\rho) \rangle_{trans} \qquad x := \langle P^{(1)}(\pi), D_{1}(\rho) \rangle_{trans} \\ (\Gamma_{V}, \Gamma_{H}) &:= \langle P^{(1)}(\pi), D_{1}(\rho) \rangle_{P(1)} \end{cases} \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}[c_{1} = 1 \mid c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}[c_{1} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}[c_{1} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}[c_{1} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}[c_{1} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] \left( \Pr_{\pi^{(k-1)}}[c \in \mathcal{G}_{0} \setminus \mathcal{G}_{1}] - (S_{\pi^{*},1} - S_{\pi^{*},0})) \right) \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ \Pr_{\pi^{(k-1)}}[c_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] \left( \Pr_{\pi^{(k-1)}}[c \in \mathcal{G}_{0} \setminus \mathcal{G}_{1}] - (1 - \frac{1}{k})\varepsilon \right) \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \left( \Pr_{\pi^{(k-1)}}[c \in \mathcal{G}_{0} \setminus \mathcal{G}_{1}] - (1 - \frac{1}{k})\varepsilon \right) \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ c_{1} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] - (1 - \frac{1}{k})\varepsilon \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] - (1 - \frac{1}{k})\varepsilon \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k-1)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k-1)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] - (1 - \frac{1}{k})\varepsilon \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=} \Pr_{\pi^{(k)}}\left[ e_{1}^{*} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right] \\ &\stackrel{x}{=}$$

where in (\*) we use the facts that  $D_2(x,\rho) \neq \bot$  if and only if  $D_2$  finds  $c \in \mathcal{G}_1 \setminus \mathcal{G}_0$  and conditioned on  $D_2(x,\rho) \neq \bot$  we have that  $\Gamma_V(D_2(x,r)) = 1$  if and only if  $c_1 = 1$ .

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