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Chapter 1

# Introduction

## 1.1 Security Amplification Theorems

## 1.2 Weakly verifiable puzzles

## 1.3 Contribution of the Thesis

## 1.4 Organization of the Thesis

Chapter 2

# Preliminaries

In this chapter we set up notation and terminology used in the Thesis.

## 2.1 Notation and Definitions

**(Functions)** We write $poly(\alpha_1, \ldots, \alpha_n)$ 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 polynomial $poly(n)$ there exists $n_0 \in \mathbb{N}$ such that for all $n \in \mathbb{N} : n > n_0$ the following holds

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

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

$$f(n) \geq \frac{1}{poly(n)}.$$

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 $f$.

**(Algorithms, Bitstrings, and Circuits)** We denote Boolean circuits using capital letters from the Greek or English alphabet. We define a *probabilistic circuit* as a Boolean circuit $C_{m,n} : \{0,1\}^m \times \{0,1\}^n \to \{0,1\}^*$ and write $C_{m,n}(x; r)$ to denote a probabilistic circuit taking as input $x \in \{0,1\}^m$ and as auxiliary input $r \in \{0,1\}^n$. If a probabilistic circuit does not take any input, we slightly abuse notation and write $C_n(r)$. Similarly, we use $\{C_n\}_{n \in \mathbb{N}}$ to denote a family of probabilistic circuits that takes only auxiliary input. We

make sure that it is clear from the context that probabilistic circuits (families of probabilistic circuits) with only auxiliary input are not confused with deterministic Boolean circuits (families of Boolean circuits respectively). For a (probabilistic) circuit $C$ we write $Size(C)$ to denote the total number of vertices of $C$. A *(probabilistic) polynomial size circuit* is a (probabilistic) circuit of size polynomial in the number of input vertices (including auxiliary input vertices if the circuit is probabilistic). 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 $C_2$.[1] If $C_1$ and $C_2$ are probabilistic circuits we write $C(\delta) := (C_1, C_2)(\delta)$ to denote that in both phases $C_1$ and $C_2$ take as auxiliary input the same bitstring $\delta$.

For an algorithm $A$ we write $Time(A)$ to denote the number of steps it takes to execute $A$. We say that $A$ runs in *polynomial time* if $Time(A)$ is bounded by some $poly(|x|)$ where $|x|$ denotes the length of the input that $A$ takes.

Similarly as for probabilistic circuits we often write randomness used by a probabilistic algorithm explicitly as a bitstring provided as an auxiliary input.

We denote a tuple $(x_1, \ldots, x_l)$ by $x^{(l)}$. The $i$-th element of $x^{(l)}$ is denote by $x_i^{(l)}$ Furthermore, for tuples $x^{(l)}$, $y^{(k)}$ we use $x^{(l)} \circ y^{(k)}$ to denote the concatenation of $x^{(l)}$ and $y^{(k)}$ which results in a tuple $(x_1, \ldots, x_l, y_1, \ldots, y_k)$.

**(Probabilities and distributions)** For a finite set $\mathcal{R}$ we write $r \xleftarrow{\$} \mathcal{R}$ to denote that $r$ is chosen from $\mathcal{R}$ uniformly at random. For $\delta \in \mathbb{R} : 0 \leq \delta \leq 1$ 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 that probability depends on $n$. We say that $E_n$ happens *almost surely* or with *high probability* if $\Pr[E_n] \geq 1 - 2^{-n} poly(n)$.

**(Interactive protocols)** We are often interested in situations where two probabilistic circuits interact with each other according to some protocol. We limit ourselves to cases where messages are 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 $B_n : \{0,1\}^* \to \{0,1\}^*$. 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 $m_0 := A_n(\rho_A)$ and in the second round $m_1 := B_n(\rho_B, m_1)$. 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 protocol execution between two probabilistic circuits $A$ and $B$ is denoted by $\langle A, B\rangle$. The output

[1] Where what a *phase* means should be clear from the context in which $C$ is used.

of $A$ in the protocol execution is denoted by $\langle A, B\rangle_A$ and of $B$ by $\langle A, B\rangle_B$. A sequence of all messages send 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 number of rounds needed to execute the protocol and the number of steps required to evaluate $A_n$ and $B_n$. Exemplary, the protocol runs in polynomial time if the number of rounds and time needed to evaluate $A_n$ and $B_n$ is bounded by some $poly(n)$.

**(Oracle algorithms)** We use notion of *oracle circuits* following the standard definition included in the literature [Gol04]. If a circuit $A$ gains oracle access to a circuit $B$, we write $A^B$. If additionally $B$ gains oracle access to a circuit $C$, we write $A^{B^C}$. However, to shorten 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 time complexity of algorithms with oracle access we count 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 bounds on running time of an algorithm. In these situations we compute running time explicitly with regard to time needed for accessing oracle.

**Definition 2.1 (Polynomial time sampleable distribution)** *We say that a distribution is* polynomial time sampleable *if it can be approximated by an algorithm running in time* $poly(\log|\mathcal{D}|, \log|\mathcal{R}|)$ *up to an exponential factor.*

**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 $\forall x \neq y \in \mathcal{D}$ and $\forall \alpha, \beta \in \mathcal{R}$, it holds*

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

***(Polynomial time sampleable)*** *For every $hash \in \mathcal{H}$ the function $hash$ is sampleable 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 $y = hash(x)$.*

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 know [CW77] that there exists families of functions meeting the criteria stated in Definition 2.2.

Chapter 3

# Weakly Verifiable Cryptographic Primitives

This chapter gives an overview of weakly verifiable cryptograph primitives. We start by formulating a definition of a *dynamic interactive weakly verifiable puzzle* in Section 3.1. To provide the Reader more intuition in Section 3.2 we describe a series of well known cryptographic primitives that are weakly verifiable. Section 3.3 is devoted to the previous research concerning different types of weakly verifiable puzzles.

## 3.1 Dynamic Interactive Weakly Verifiable Puzzle

We 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 the problem poser. The solver* $C := (C_1, C_2)$ *for* $P_n$ *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 hint circuit* $\Gamma_H$. *The circuit* $C_1(\rho)$ *produces no output. The circuit* $\Gamma_V$ *takes as input* $q \in Q$, *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* $q \in Q$ *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 randomness 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* $q$*.*

We note that the above definition is very general and does not pose any constrains on the size of circuits or the time complexity of the interactive protocol.

We use the term *weakly verifiable* to emphasize that there is no easy way for the solver to check correctness of a solution except 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.

Definition 3.1 generalizes and combines previous approaches that studies *weakly verifiable puzzles* [CHS04], *dynamic weakly verifiable puzzles* [DIJK09], and *interactive weakly verifiable puzzles* [HS10].

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 a case where the problem poser and the solver are probabilistic polynomial time algorithms. We use the well know fact [Hol13b] that a probabilistic polynomial time algorithm can be transformed into an equivalent family of probabilistic Boolean circuits of the polynomial size[1].

## 3.2 Examples

In this section we give examples of cryptographic constructions that motivate studies of different types of weakly verifiable puzzles.

### 3.2.1 Message Authentication Codes

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

Loosely speaking, the message authentication codes may be explained as follows. Let sender, receiver, and adversary be polynomial time algorithms and messages be represented as bitstrings. Furthermore, we assume that the sender and the receive share a secrete key to which the adversary has no access. The

[1] Theorem 6.10 from [Hol13b] is stated for the probabilistic oracle programs with single bit of output, but it can be adopted to a case when an output is longer than a single bit.

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 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, not sent before, that is accepted by the receiver with non-negligible probability. We give the following formal definition of a *message authentication code* based on [Mau13] and [Gol04].

**Definition 3.2 (Message Authentication Code)** *Let $\mathcal{M}$ be a set of messages, $\mathcal{K}$ a set of keys and $\mathcal{T}$ a set of tags. We define the* message authentication code (MAC) *as an efficiently computable function $\mathcal{M} \times \mathcal{K} \to \mathcal{T}$. Furthermore, we say that MAC is secure if it satisfies the following condition:*

*Let $k \xleftarrow{\$} \mathcal{K}$ be fixed and $H$ be a polynomial size circuit 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 non-negligible 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 $\Gamma_H$ has not been queried for a tag of $m$.*

We show how MAC is captured by notion of a dynamic weakly verifiable puzzles where at most one verification query is asked. For fixed $f$ and $n \in \mathbb{N}$ the sender corresponds to the problem poser, the adversary to the problem solver, and the key is a bitstring $\pi \in \{0, 1\}^n$ taken as auxiliary input by the problem poser. In the first phase, which is non–interactive, the problem poser outputs a hint circuit $\Gamma_H$ and a verification circuit $\Gamma_V$ where both circuits have hard-coded $\pi$. The circuit $\Gamma_H$ takes as input a message and outputs a tag for this message and corresponds to the circuit $H$ from Definition 3.2. The circuit $\Gamma_V$ that as input $m \in \mathcal{M}$ and $t \in \mathcal{T}$ and outputs a bit 1 if and only if $f(m, \pi) = t$. In the second phase the problem solver takes no input ($x^*$ is empty string) and is given oracle access to $\Gamma_H$ and $\Gamma_V$. We consider a case where at most on verification query is asked. Thus, a task of finding by an adversary a valid tag $t \in \mathcal{T}$ for a message $m \in \mathcal{M}$ such that a hint for $m$ has not been asked before corresponds to asking a successful verification query by a problem poser to $\Gamma_V$.

### 3.2.2 Public Key Signature Scheme

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

**Definition 3.3 (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:*

- *$G(1^n)$ outputs a pair of bitstrings $k_{priv} \in \{0,1\}^n$ and $k_{pub} \in \{0,1\}^n$ where $n \in \mathbb{N}$ is a security parameter. We call $k_{priv}$ a private key and $k_{pub}$ a public key.*

- *The signing algorithm $S$ takes as input $k_{priv} \in \{0,1\}^n$, $q \in \mathcal{Q}$ and outputs a signature $s \in \mathcal{S}$.*

- *The verification algorithm $V$ takes as input $k_{pub} \in \{0,1\}^n$, $q \in \mathcal{Q}$, and $s \in \mathcal{S}$ and outputs a bit $b \in \{0,1\}$.*

- *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 random coins of $V$ and $S$.*

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

**Definition 3.4** ***(Security of public key signature scheme with respect to a chosen message attack)*** *Let an* adversary *$A$ be a probabilistic polynomial time algorithm that takes as input $k_{pub}$ and has oracle access to $S$. 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 $S$ 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.*

We will show now that a public key signature scheme defined as above can be represented as a dynamic weakly verifiable puzzle. Let the problem poser correspond to an entity that generates $k_{pub}$, $k_{priv}$ and the solver to the adversary. In the first phase, the problem poser uses algorithm $G(1^n)$ to obtain $k_{pub}$, $k_{priv}$ and sends to the adversary the public key $k_{pub}$. Then the problem poser generates a hint circuit $\Gamma_H$ and a verification circuit $\Gamma_V$. The hint circuit $\Gamma_H$ takes as input $q \in \mathcal{Q}$ and outputs a valid signature for $q$. The verification circuit $\Gamma_V$ takes as input $s \in \mathcal{S}$ and $q \in \mathcal{Q}$ and outputs 1 if and only if $s \in \mathcal{S}$ is a valid signature for $q \in \mathcal{Q}$. In the second phase, the problem solver takes as input a transcript of the messages from the first round which consists solely of $k_{pub}$. Additionally, it is given oracle access to $\Gamma_V$ and $\Gamma_H$. The adversary can ask a number of hint and verification queries. It is clear that if the adversary asks a successful verification query $(q, s)$, then it finds also a valid signature for $q$.

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

### 3.2.3 Bit Commitments

Let us 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 $y \in \{0,1\}^*$ is an information that helps the receiver to verify that the sender committed to the value $b' \in \{0,1\}$. A desirable property of a bit commitment protocol is that in the reveal phase it should be hard for the sender to find two bitstrings $y_0$ and $y_1$ such that for the receiver both $(0, y_0)$ and $(1, y_1)$ are valid decommitments.

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

**Definition 3.5 (Bit Commitment Protocol)** *A* bit commitment protocol *is defined by a pair $(S_n, R)$ where $S = (S_1, S_2)$ is a two phase probabilistic circuit, $R$ is a probabilistic circuit, and $n \in \mathbb{N}$ is a security parameter. We call $S$ the sender and $R_n$ the 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 $\rho_S \in \{0,1\}^n$ is the randomness used by the algorithm $S$. The receiver $R_n$ takes as auxiliary input a bitstring $\rho_R \in \{0,1\}^*$ that is the randomness used by $R_n$. The protocol consists of two phases. In the* commit phase*, circuits $S_1$ and $R_n$ engage in the protocol execution. As the result $S_1$ commits to $b$ and $R_n$ generates a circuit $V$. The circuit $V$ takes as input a bit $b' \in \{0,1\}$ 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 bitstring $\rho_s$ 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*

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

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

**(Hiding)** *Probability over random coins of $S$ and $R_n$ that any polynomial size circuit can guess bit $b$ 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 circuit $S := (S_1, S_2)$ such that $S_1$ and $S_2$ are of size*

*poly*$(n)$ *we have*

$$\Pr_{\substack{\rho_S \in \{0,1\}^n, \rho_R \in \{0,1\}^* \\ V := \langle S_1(b,\rho_S), R(\rho_R) \rangle_R \\ ((0,y_0),(1,y_1)) := S_2(\rho_S)}} [V(0,y_0) = 1 \wedge V(1,y_1) = 1] \leq \varepsilon(k),$$

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

Breaking the binding property of a bit commitment protocols can be generalized as an interactive weakly verifiable puzzle. The number of hint queries amounts to zero, and the number of the verification queries is at most one which means that the puzzle is non–dynamic. The receiver is the problem poser. The solver corresponds to the sender trying to break the binding property.

In the first phase the solver parses the auxiliary input such that the first bit $b$ is the bit to which it commits. The remaining part of the auxiliary input is the randomness used by the solver. The solver interacts with the problem poser. After, the first phase the problem poser generates a circuit $\Gamma_V$ that takes as input $y_0$, $y_1$ and outputs 1 if and only if $V(0,y_0) = 1 \wedge V(1,y_1) = 1$. For the bit commitment protocols $|Q| = 1$, thus we do not write on which $q \in Q$ 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 black box access to $\Gamma_V$. Therefore, it has no efficient way to check whether the solution is successful except asking a verification query.

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

### 3.2.4 Automated Turing Tests

The goal of *Automated Turing Tests* is to distinguish humans from computers. An example of such a test is *CAPTCHA* formally defined in [VABHL03]. Loosely speaking, CAPTCHA is a test for which it is hard to write a computer program that success probability is comparable or higher to the one achieved by most of humans. An example is an image depicting a distorted text. We note that the definition of hardness is defined by and bases on the opinions of AI community [VABHL03].

CAPTCHAs based on guessing the distorted text can be modeled as weakly verifiable puzzles. In the first round the problem poser and the solver engage in the execution of an interactive protocol such that after its execution the problem poser sends to the solver an image containing the distorted text. Furthermore, the problem poser generates a circuit $\Gamma_V$ that takes as input a

bitstring $y$ and outputs 1 if and only if $y$ correctly encodes the text depicted on the distorted image.

The problem poser in the second phase takes as input a distorted image, has oracle access to $\Gamma_V$ and can ask at most a single verification query. In general, checking whether a solution is correct is as hard as finding a correct solution. Thus, CAPTCHAs are weakly verifiable.

Standard CAPTCHAs are non-dynamic, as the problem solver does not gain access to the hint oracle.

It is not know how good an algorithm for recognizing CAPTCHA can be. Thus, it is likely that a gap between human performance and a performance of a computer program may be small. Therefore, it is of interest to find a way to amplify this gap. Actually, in [HS10] it has been shown that it is possible for parallel repetition. The solver is given $n$ independent puzzles. The verifier accepts when the solver succeeds on at least $\delta n$ fraction of puzzles. It turns out that in this way we can construct weakly verifiable puzzles for which success probability of a computer program is at most $\varepsilon$ and which is still easy for humans that succeeds with at least probability $1 - \varepsilon$ where $\varepsilon \in (0, 1)$.

In Chapter 4 we give a more general proof than the one in [HS10] that applies also to dynamic context.

## 3.3 Previous results

Different types of weakly verifiable cryptographic primitives have been studied in a series of works [CHS04, DIJK09, HS10]. This section is intended to give a short overview of techniques used in these works and aims to provide some intuition and insight into the problem of hardness amplification of dynamic interactive weakly verifiable puzzles.

### 3.3.1 Results of R.Canetti, S.Halevi, and M.Steiner

The notion of a *weakly verifiable puzzle* has been coined by R.Canetti, S.Halevi, and M.Steiner in the work *Hardness amplification of weakly verifiable puzzles* [CHS04]. In comparison to Definition 3.1 the puzzles considered in [CHS04] are neither dynamic nor interactive. Moreover, the number of verification queries is limited to one. This constitutes a simplified case to the one considered in this Thesis. In this section we provide the definition of weakly verifiable puzzles (WVP) that closely follows the one contained in [CHS04] and state the theorem of hardness amplification of weakly verifiable puzzles in a similar vein as in [CHS04]. Finally, we give an intuition behind the proof of this theorem. It is noteworthy that the main proof of this Thesis, contained in Chapter 4, uses many ideas from this work of R.Canetti, S.Halevi, and M.Steiner [CHS04].

**Definition 3.6 (Weakly Verifiable Puzzles)** *A* weakly verifiable puzzle *is defined by a pair of polynomial time algorithms: a probabilistic puzzle–generation algorithm $G$ and a deterministic verification algorithm $V$. We write $G(1^k;\rho)$ to denote that $G$ takes as input a bitstring $1^k$, where $k$ is a security parameter, and as auxiliary input a bitstring $\rho \in \{0,1\}^*$ which is the randomness used by $G$. The algorithm $G$ outputs a bitstring $p \in \{0,1\}^*$ and a check information $c \in \{0,1\}^*$. The* verifier *$V$ is a deterministic algorithm that takes as input $p$, $c$, an answer $a \in \{0,1\}^*$ and outputs $b \in \{0,1\}$.*

*A* solver *$S$ for $G$ is a polynomial time probabilistic 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 $P$ as*

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

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

We compare the above definition with Definition 3.1. First, we note that in Definition 3.1 we use the language of probabilistic circuits. This is more general approach as probabilistic polynomial time algorithms can be converted into families of polynomial size circuits. Next, we see that in Definition 3.6 the algorithm $G$ is parameterized by a bitstring $1^k$ meaning that the length of a random bitstring taken by $G$ is bounded by $poly(k)$. For a fixed $k$, without loss of generality, we can model the algorithm $G(1^k;\rho)$ as a polynomial size probabilistic circuit that does not take as input $1^k$, but just a bitstring $\rho$ of length $poly(k)$. The security parameters from Definition 3.6 and Definition 3.1 are not equivalent, as in the later definition the security parameter limits the length of the random bistring. Moreover, in Definition 3.6 a verification algorithm 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. Hence, the algorithm $V$ corresponds to $\Gamma_V$. The puzzles considered in Definition 3.6 are non–dynamic. Thus, there is no corresponding element for a hint circuit $\Gamma_H$ from Definition 3.1. Finally, the puzzles described in Definition 3.6 are non–interactive.

We will formulate now the definition of $n$-fold repetition of Weakly Verifiable Puzzles along the lines of [CHS04]. Later in this Thesis, for other types of weakly verifiable puzzles, we use the notion of a $k$-wise direct product of puzzles instead [2].

[2] We note that the terminology used to name the parallel repetition of $n$ weakly verifiable puzzles is different in different works. In [CHS04] the notion *$n$-fold direct product of puzzles* is used whereas in [DIJK09] a similar construction that captures dynamic puzzles is named a $k$-wise repetition of puzzles. In this Thesis we tend to use the latter formulation.

**Definition 3.7 ($n$-fold repetition of Weakly Verifiable Puzzles)** *Let $n \in \mathbb{N}$ and a weakly verifiable puzzle $P = (G, V)$ be fixed. We define the $n$-fold repetition of $P$ as a weakly verifiable puzzle where the puzzle–generation algorithm $G^{(n)}$ takes as input $1^k$, as an auxiliary input a bitstring $\rho \in \{0,1\}^*$ 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 $1 \leq i \leq n$ a pair $(p_i, c_i)$ is an independent instance of a weakly verifiable puzzle defined by $G$ and $V$ with 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 $1 \leq i \leq n$ we have $V(p_i, c_i, a_i) = 1$. We write $P^{(n)} := (G^{(n)}, V^{(n)})$ to denote the $n$–fold repetition of $P$.*

Let us lay down some terminology. For $P^{(n)} := (G^{(n)}, V^{(n)})$ when writing a *puzzle on the $i$-th coordinate* we refer to the $i$-th puzzle of the $n$–fold repetition of WVP (this puzzle that corresponds to the one generated by $G_i^{(n)}$, $V_i^{(xn)}$). The $n$-fold repetition of weakly verifiable puzzles is solved successfully if and only if all $n$ puzzles are solved successfully. In contrast, in Chapter 4 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 $n$-fold repetition of puzzles have to be solved correctly. A precise definition is given in Section 4.1. Clearly, we can assume that $g$ is such that all coordinates have to be solved successfully which matches the case considered in this section.

The main theorem proved in [CHS04] states that it is possible to turn a good solver for $P^{(n)}$ to a good solver for $P$.

**Theorem 3.8 (Hardness amplification of Weakly Verifiable Puzzles)** *Let $n : \mathbb{N} \to \mathbb{N}$ and $\delta : \mathbb{N} \to (0,1)$ be efficiently computable functions and $q \in \mathbb{N}$ a slackness parameter. Moreover, let $P = (G, V)$ be a weakly verifiable puzzle. We denote the running time of the puzzle–generation algorithm $G$ for $P$ by $T_G$ and of the verification algorithm $V$ for $P$ by $T_V$. If $S^{(n)}$ is a solver for the $n$–fold repetition of $P$ that success probability is at least $\delta^n$ and running time is $T$, then there exists a solver $S$ for $G$ with oracle access to $S^{(n)}$ that success probability is at least $\delta(1 - \frac{1}{q})$ and running time $O\Big(\frac{nq^3}{\delta^{2n-1}}(T + nT_G + nT_V)\Big)$.*

The following algorithm is used by R.Cannetti, S.Halevi, and M.Steiner in the proof of Theorem 3.8. It transforms $S^{(n)}$ for $P^{(n)}$ with success probability at least $\delta^n$ to a solver for a single puzzle $P$ with probability at least $\delta(1 - \frac{1}{q})$. The slackness parameter $q$ is introduced as it is not possible to achieve the perfect hardness amplification. We note that in the analysis of running time of *CHS–solver* we explicitly take into account time needed for oracle calls to $S^{(n)}, V, G$.

We denote by $p \in \{0,1\}^*$ a bitstring output by $G$ and taken as input by *CHS–solver*. To make notation shorter in the following code excerpts we do not write randomness used by $G$ explicitly.

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

**Oracle:** A solver $S^{(n)}$ for $P^{(n)}$, a verification algorithm $V$ for $P$, a puzzle–generation algorithm $G$ for $P$.
**Input:** A bistring $p \in \{0,1\}^*$, parameters $n, k, q, \delta$.

$\mathit{prefix} := \emptyset$
**for** $i = 1$ **to** $n-1$ **do:**
  $p^* := \mathit{ExtendPrefix}^{S^{(n)},V,G}(\mathit{prefix}, i, n, k, q, \delta)$
  **if** $p^* = \perp$ **then return** $\mathit{OnlinePhase}^{S^{(n)},V,G}(\mathit{prefix}, p, i, n, k, q, \delta)$
  **else** $\mathit{prefix} := \mathit{prefix} \circ p^*$
$a^{(n)} := S^{(n)}(\mathit{prefix} \circ p)$
**return** $a_n$

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

**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 $v, n, k, q, \delta$.

**Repeat** $\left\lceil \frac{6q \ln(6q)}{\delta^{n-v+1}} \right\rceil$ times
  $((p_{v+1}, \ldots, p_n), (c_{v+1}, \ldots, c_n)) := G^{(n-v-1)}(1^k)$
  $a^{(n)} := S^{(n)}(\mathit{prefix}, p, p_{v+1}, \ldots, p_n)$
  **if** $\forall_{v+1 \le i \le n} V(p_i, c_i, a_i) = 1$ **then return** $a_v$
**return** $\perp$

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

**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 $i, n, k, q, \delta$.

**Repeat** $\left\lceil \frac{6q}{\delta^{n-v+1}} \ln(\frac{18qn}{\delta}) \right\rceil$ times
  $(p^*, c^*) := G(1^k)$
  $\bar{\nu}_i := \mathit{EstimateResSuccProb}^{G,V}(\mathit{prefix} \circ p^*, i, n, k, q, \delta)$
  **if** $\bar{\nu}_i \ge \delta^{n-i}$ **then return** $p^*$
**return** $\perp$

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

**Oracle:** A solver algorithm for $P^{(n)}$, a verification algorithm $V$ for $P$, a generation algorithm $G$ for $P$
**Input:** A $i$–tuple of puzzles *prefix*, parameters $i, n, k, q, \delta$.

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

A detail proof of Theorem 3.8 is presented in [CHS04]. We limit ourselves to providing an intuition why the above algorithm transforms a good solver for the $n$–wise direct product of $P$ to a good solver for $P$.

Let us consider the $n$-fold repetition of $P$, and for simplicity a deterministic solver $S^{(n)}$ for $P^{(n)} := (G^{(n)}, V^{(n)})$. Furthermore, 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 algorithm $G^{(n)}$ 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 bitstring inferred by a column and a row of the cell. We make the following observation.

**Observation 3.9** *For a deterministic polynomial time algorithm $S^{(n)}$ that successfully solves the $n$–fold repetition of $P$ with probability at least $\delta^n$ the matrix $M$ defined as above has either a column with $\delta^{(n-1)}$ fractions of cells that are all one vectors, or a conditional probability that a cell is of the form $1^n$ given that the last $(n-1)$ bits of the cell are equal 1 is at least $\delta$.*

We show, at least intuitively using Observation 3.9, how the algorithm *CHS–solver* can be used to solve $P$ with substantial probability given oracle access to $S^{(n)}$ for $P^{(n)}$. The algorithm 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 *CHS–solver* fails to find a bitstring $p^*$, then we assume that there is no column of $M$ that contains $\delta^{(n-1)}$ fraction of cells that are of the form $1^n$. We use Observation 3.9 to assume 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 all remaining puzzles are generated by *CHS–solver*. Thus, it is possible to efficiently verify whether these puzzles are successful solved by $S^{(n)}$.

Obviously, the algorithm *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 an estimate returned by *EstimateResSuccProb* is incorrect.

It is possible to show that all these events happen with negligible probability. Therefore, at least intuitively we see that the algorithm *CHS–solver* solves a single WVP puzzle successfully with probability at least $\delta(1-\frac{1}{q})$ almost surely.

In Chapter 4 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 $n$-fold repetition of puzzles[3] although it successful only on some coordinates of the $n$-fold repetition of $P$. It turns out that it is possible to use a similar technique of fixing puzzles on consecutive positions of the $n$-fold repetition of puzzles to prove the hardness amplification in this more general setting.

### 3.3.2 Results of Y.Dodis, R.Impagliazzo, R.Jaiswal, V.Kabanets

Some of the cryptographic constructions presented in Section 3.2 are not only weakly verifiable but also dynamic (MAC and SIG). This type of puzzles are defined and studied in [DIJK09]. We give a short overview of this work, state the definition of a *dynamic weakly verifiable puzzle* that closely follows the one included in [DIJK09]. Finally, we provide intuition for the proof of the hardness amplification of DWVP included in [DIJK09].

**Definition 3.10 (Dynamic Weakly Verifiable Puzzle.)** *A* dynamic weakly verifiable puzzle *(DWVP) is defined by a distribution $\mathcal{D}$ on pairs $(x, \alpha)$ where $\alpha \in \{0,1\}^*$ is an advice used to generate and evaluate responses and $x \in \{0,1\}^*$ is a bitstring taken as input by the solver. Furthermore, we consider a set $\mathcal{Q}$ and a probabilistic polynomial time computable relation $R$ such that $R(\alpha, q, r) = 1$ if and only if $r$ is a correct answer to $q \in \mathcal{Q}$ on the set of puzzle determined by $\alpha$. Finally, let $H(\alpha, q)$ be a probabilistic polynomial time computable* hint *relation.*

*A solver $S$ takes as input $x$ and can ask hint queries on $q \in \mathcal{Q}$ which are answered using $H(\alpha, q)$ and verification queries of the form $(q, r)$ answered by means of $R(\alpha, q, r)$. We say that $S$ succeeds if and only if it makes a*

[3] Actually, in Chapter 4 we define the $k$-wise repetition of puzzles following the terminology used in [DIJK09] which is equivalent to the $n$-fold repetition of puzzles.

*verification query on $(q, r)$ such that $R(\alpha, q, r) = 1$ and it has not previously asked for a hint query on this $q$. We write $P := (\mathcal{D}, R, H)$ to denote a DWVP with a distribution $\mathcal{D}$ of pairs $(x, \alpha)$, and $R$, $H$ being a verification and hint relations respectively.*

We show now how the above definition is generalized by Definition 3.1. First, instead of considering a distribution on pairs $(x, \alpha)$ in Definition 3.1 we use a probabilistic problem poser that outputs circuits $\Gamma_H$ and $\Gamma_V$ that corresponds to hint and verification relations respectively. Furthermore, the problem poser may interact in the first phase with the problem solver. In particular, the problem poser can send a bitstring $x$ as a message in the first phase. Thus, Definition 3.1 captures a more general case where the distribution of puzzles is defined by both the problem poser and the problem solver.

We define the *$n$-wise direct product of DWVPs* which is conceptually similar to the $n$-fold repetition of WVPs.

**Definition 3.11 ($n$-wise direct product of DWVPs)** *For a dynamic weakly verifiable puzzle $P := (\mathcal{D}, R, H)$ we define the $n$-wise direct product of $P$ as a DWVP with a distribution $\mathcal{D}^{(n)}$ on tuples $(x_1, \alpha_1), \ldots, (x_n, \alpha_n)$. 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 $R^{(n)}(\alpha_1, \ldots, \alpha_n, r_1, \ldots, r_n, q)$ evaluates to 1 if and only if for $1 \leq i \leq n$ at least $n - (1 - \gamma)\delta n$ is such that $R(\alpha_i, q, r_i) = 1$ where $0 \leq \gamma, \delta \leq 1$. We write $P^{(n)} := (D^{(n)}, H^{(n)}, R^{(n)})$ to denote the $n$-wise direct product of $P := (D, H, R)$.*

In contrast to the $n$-fold repetition of puzzles defined in previous section, here we require a solver to succeed only on a fraction of puzzles.

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

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

With no loss of generality we make an assumption that the solver does not make hint queries on the successful verification queries. We define the *success probability* of a solver $S$ for $P := (G, V)$ as

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

**Theorem 3.12** ***(Hardness amplification for dynamic weakly verifiable puzzle).*** *Let $S^{(n)}$ be a probabilistic algorithm for the $P^{(n)}$ 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. Then there exists a probabilistic algorithm $S$ that succeeds in solving $P$ with probability at least $1-\delta$ making $O(h(h+v)/\varepsilon) \cdot \log(1/\gamma\delta)$ hint queries and at most one verification query. Furthermore, the running time $poly(h, v, \frac{1}{\varepsilon}, t, \omega, \log(1/\gamma\delta))$ 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 $n$-fold repetition of WVP cannot be applied for the $k$-wise direct product of DWVP (moving aside for a moment the issue of solving only a fraction of puzzle successfully). 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 runs $S^{(n)}$ asks a hint query on $q$ for which in one of the later runs a verification query $(q, r)$ is asked for which algorithm would return an answear for the input puzzle (in other words the condition $\forall_{v+1 \leq i \leq n} V(p_i, c_i, a_i) = 1$ is satisfied). However, the fact that a hint query on this $q$ has been asked makes it impossible to ask a successful verification query on this $q$. Thus, we can not 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 $n$-wise 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 $q \in \mathcal{Q}_{attack}$.

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

In [DIJK09] the following lemma is proved.

**Lemma 3.13** *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 $h$ and $v$ respectively. There exists a probabilistic algorithm that runs in time $poly(h, v, \frac{1}{\delta}, t)$ that outputs a function $hash : \mathcal{Q} \to \{0, 1, \ldots, 2(h+v)-1\}$ that partitions $\mathcal{Q}$ to $\mathcal{Q}_{attack}$ 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 has been asked on $q \in \mathcal{Q}_{adv}$.*

A function *hash* can be found by using a natural sampling technique. We follow exactly the same approach of the partitioning the domain in the main

proof of this Thesis in Section 4.1.3.

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 $R_\alpha(q, r)$ to denote a polynomial time probabilistic algorithm that computes relations $R(\alpha, q, r)$ and has hard-coded bitstring $\alpha$. The following algorithm is used in [DIJK09] in the proof of lemma 3.12. It gains oracle access to $R_\alpha$ and $H_\alpha$ as well as a function *hash* from Lemma 3.13.

**Algorithm:** $DWVP\text{–}solver^{S^{(n)}, hash, H_\alpha^{(n)}, R_\alpha^{(n)}}(x)$

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**Oracle:** A solver $S^{(n)}$ for $P^{(n)}$, a function $hash : Q \to \{0, 1, \ldots, 2(h+v)\}$.
**Input:** A bistring $x \in \{0,1\}^*$.

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**Repeat** at most $O(\frac{h+v}{\varepsilon} \cdot \log(\frac{1}{\gamma\delta}))$ times
  Let $i \xleftarrow{\$} \{1, \ldots, n\}$ 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 $(q, r_1, \ldots, r_n)$ **then**
      **if** $hash(q) = 0$ **then** answer the query with 0
      Let $m := |j : V(q, r_j) = 1, j \neq i|$
      **if** $m \geq n - n(1-\gamma)\delta$ **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 the above algorithm we execute multiple times a solver $S^{(n)}$ for the $k$-wise direct product of DWVPs. In each iteration the position for $x \in \{0,1\}^*$ 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. We use a function *hash* to partition the query domain and assume that *hash* is such that the success probability of $S^{(n)}$ with respect to *hash* is at least $\frac{\delta}{8(h+v)}$. We check on which $q$ the solver $S^{(n)}$ asks hint and verification queries. If a hint query is asked on $q$ such that $hash(q) = 0$ then

the execution of $S^{(n)}$ is aborted and we go to the next iteration. This way we make sure that the algorithm 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 such a verification query with 0.

Finally, in case when $S^{(n)}$ asks a verification query using index $q$ such that $hash(q) = 0$, then we use a soft decision system 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. The detail calculations provided in [DIJK09] show that this approach yields a demanded result. We do not give a more detail description of the proof of [DIJK09] as in Chapter 4, except the domain partitioning, we use a different technique.

In case of weakly verifiable primitives like CAPTCHAs we assume that most people have at least slightly higher probability of solving these kind of puzzles than the best computer programs. Still, it may happen that humans do not solve all puzzles. This is a motivation to introduce a threshold function such that on average solutions of humans are treated as solved successfully but the ones of computer programs on average are classified as not successful. This motivates a study of the situations where only a fraction of puzzles is solved successfully.

In Chapter 4 we consider a weakly verifiable puzzles that are not only dynamic but also interactive. We use a very similar technique to partition domain $Q$ into advice and hint queries as presented in [DIJK09]. Instead of the requirement to succeed only on a fraction of puzzles we consider an arbitrary, monotone function $g : \{0,1\}^n \to \{0,1\}$ that determines on which coordinates the solver has to succeed in order to successfully solve the $n$-wise direct product of puzzles.

To show the hardness amplification for the $n$-wise direct product with the domain partitioned we use the approach similar to the one presented in Section 3.3.1. Namely, we try to find a good position for the input puzzle instead of choosing the position uniformly on random as in [DIJK09].

### 3.3.3 Results of T.Holenstein and G.Scheonebeck

The hardness amplification of interactive weakly verifiable puzzles has been studied by T.Holenstein and G.Schoenebeck in [HS10]. We will give now an overview of this work and compare it with our approach.

The following definition of an *interactive weakly verifiable puzzle* closely follow the one from [HS10].

**Definition 3.14** *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:*

$$\Pr_{\substack{\rho,\pi \\ \Gamma^{(g)}:=\langle P(\rho),S(\pi)\rangle_P}} \left[\Gamma^{(g)}(\langle P(\rho), S(\pi)\rangle_S) = 1\right].$$

We are intrested in the hardness amplification of interactive weakly verifiable puzzles. Thus, similarly as in the previous sections we define the $k$-wise direct product of puzzles.

**Definition 3.15** ***($k$-wise direct product of interactive weakly verifiable puzzles)*** *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 $(P, S)$ defined by $(P^{(g)}, S^{(g)})$ is an interactive weakly verifiable puzzles in which the sender and the receiver sequentially interact in $k$ rounds where in each round $(P, S)$ is used to generate an instance of interactive weakly verifiable puzzle. As the result circuits $\Gamma^{(1)}, \ldots, \Gamma^{(k)}$ for $P$ are generated. Finally, $P^{(g)}$ outputs the circuit $\Gamma^{(g)}(y_1, \ldots, y_k) := g(\Gamma^{(1)}(y_1), \ldots, \Gamma^{(k)}(y_k))$.*

Similarly as in Definition 3.1 the puzzles considered in [HS10] are interactive. Furthermore, a monotone binary function is used to determine whether the $k$-wise repetition has been successfully solved. Unlike, puzzles in Definition 3.1, the puzzles studied by T.Holenstein and G.Schoenebeck are non-dynamic. Thus, only a verification circuit $\Gamma$ is generated and no hint circuit is ever used.

The following hardness amplification theorem is proved in [HS10].

**Theorem 3.16** *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 $P$, 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*

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

*then, $D$ satisfies almost surely,*

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

*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 $Time(Gen) = poly(k, \frac{1}{\varepsilon}, n)$ with oracle calls to $C$.*

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 were compared with the algorithms that either solves all puzzles ([CHS04]) or allowed a fraction of puzzles to be solved incorrectly ([DIJK09]). In the above definition a more general case is considered where we use a binary monotone function $g$. More precisely we are interested functions that are binary monotonously non-decreasing.

We emphasize that the monotone restriction on $g$ is essential. For $g(b) := 1-b$ an algorithm that deliberately gives incorrect answers satisfies $g$ with probability 1 whereas an algorithm that solves a puzzle successfully with probability $\gamma > 0$ succeeds only with probability $1-\gamma$. A desirable property for the solver for the $k$-wise direct product of IWVP is that an algorithm that solves puzzles on all coordinates with the higher probability does not the $k$–wise direct product with probability lower than an algorithm that success probability on these coordinates is lower.

The proof technique used by T.Holenstein and G.Schoenebeck is similar to the one presented in Section 3.3.1. In chapter 4 we use very similar approach and fix puzzles on consecutive coordinates of the $n$-wise direct product.

## 3.4 Limitations of Security Amplification

Chapter 4

# Hardness amplification for weakly verifiable puzzles

In the previous chapter we gave an overview of the former studies of different types of weakly verifiable puzzles. We also defined a more general notion of a dynamic interactive weakly verifiable puzzle. The focus of this chapter is on a constructive proof of hardness amplification for dynamic interactive weakly verifiable puzzles. In Section 4.1 we formulate the theorem which is then proved in the succeeding sections. We begin with constructing an algorithm that finds an efficiently computable function that is used to partition the domain of hint and verification queries. Next, we give a proof of hardness amplification under the assumption that the domain is well partitioned. Finally, in Section 4.1.5 we finish the proof by combining the previous steps.

## 4.1 Main theorem

### 4.1.1 Direct product of interactive weakly verifiable puzzles

We begin by providing the definition of the $k$-wise direct product of dynamic interactive weaky verifiable puzzles.

**Definition 4.1 ($k$-wise direct-product of DIWVPs.)** *Let $g : \{0,1\}^k \to \{0,1\}$ be a monotone function and $P_n^{(1)}$ a problem poser as in Definition 3.1. 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 each $1 \leq i \leq n$. Let $(C_1, C_2)(\rho)$ be a solver for $P_{kn}^{(g)}$. In the first phase, the algorithm $C_1(\rho)$ sequentially interacts in $k$ rounds with $P_{kn}^{(g)}(\pi^{(k)})$. 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*

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

*and a hint circuit*

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

If it is clear from the context, we omit the subscript $n$ and write $P(\pi)$ instead of $P_n(\pi)$ where $\pi \in \{0,1\}^n$.

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, 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.

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

---

**Oracle:** 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

$$\Pr_{\pi,\rho}[Success^{P,C}(\pi, \rho) = 1]. \tag{4.1}$$

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

**Theorem 4.2** ***(Hardness amplification for dynamic 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 $P_n^{(1)}$. 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 Gen with oracle access to $C$, a monotone*

*function* $g : \{0,1\}^k \to \{0,1\}$*, and problem posers* $P_n^{(1)}$*,* $P_{kn}^{(g)}$*. Furthermore, Gen takes as input parameters* $\varepsilon$*,* $\delta$*,* $n$*,* $k$*,* $h$*,* $v$*, and outputs a solver circuit* $D$ *for* $P_n^{(1)}$ *such that the following holds:*
*If* $C$ *is such that*

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

*then* $D$ *is a two phase probabilistic circuit that with high probability satisfies*

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

*Additionally,* $D$ *requires oracle access to* $g$*,* $P_n^{(1)}$*,* $C$*, hint and verification circuits generated by the problem poser* $P^{(1)}$ *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,* $Time(Gen) = poly(k,\frac{1}{\varepsilon},n,v,h)$ *with oracle calls.*

We emphasize that the number of hint queries asked by $D$ is greater than the number of hint asked by $C$ but the number of verification queries is limited to at most one. In most applications making such an assumption about number of hint and verification queries is reasonable. In particular, we cannot assume that one can ask more verification queries than a solver for the $k$-wise direct product.

### 4.1.2 Intuition

We refer to the puzzle solved by a circuit $D$ as an *input puzzle*. The idea is to use a solver $C$ for the $k$-wise direct product of $P^{(1)}$ with the input puzzle placed on one of the $k$ coordinates. The puzzles on the remaining $k-1$ coordinates are generated by Gen and $D$ and chosen in such a way that the input puzzle is solved with substantial probability. The approach used in this Thesis is similar to the one in [CHS04, HS10] and briefly described in Section 3.3.

Loosely speaking, we fix randomness used to generate puzzles on the consecutive coordinates of the $k$-wise direct product of $P^{(1)}$ as long as the success probability of $C$ on the remaining puzzles is still substantial. More precisely, the success probability of $C$, with the first puzzle fixed, for the remaining $k-1$ puzzles should satisfy the assumptions of Theorem 4.2 for the $(k-1)$-wise direct product of $P^{(1)}$.

If it is possible to find such randomness for the puzzle on the first positions, then Gen recursively solves the problem for the $(k-1)$–wise direct product of DIWVP. If Gen reaches $k=1$, then from the assumptions of Theorem 4.2 for the 1-wise direct product of puzzles the solver $C$ can be easily used to successfully solve the input puzzle placed on the last position.

Unfortunately, we cannot exclude a situation where Gen does not find randomness that could be used to generate a puzzle on the first coordinate such that the success probability of $C$ on the remaining coordinates is substantial. However, we make the following important observation which is very much in the same vein as Observation 3.9. If Gen fails to find a good randomness for the puzzle on the first position, then we assume that the success probability of $C$ on the remaining coordinates is no longer substantial. Furthermore, we also know that when all coordinates are considered, then $C$ has substantial success probability. Intuitively, this means that the first coordinate is somehow important in the sense that $C$ solves a puzzle on the first coordinate unusually often. Therefore, we place the input puzzle on the first position.

What is left is to find randomness used to generate puzzles on the remaining $(k-1)$ coordinates such that the puzzle on the first position is solved often. In Section 3.3.1 where we made Observation 3.9, we have seen that in the context of weakly verifiable puzzles it makes sense to consider a situation where all the remaining $k-1$ puzzles are correctly solved. Here, we generalize this approach. We fix randomness for the remaining $k$–1 puzzles such that when the puzzle on the first position was solved successfully, then the $k$-wise direct product would be solved successfully, and if the puzzle on the first position was unsuccessfully solved, then the $k$-wise direct product would be also solved unsuccessfully. It turns out that in case of a function $g$ which requires all puzzles to be solved correctly, the above described approach corresponds exactly to the one presented in Section 3.3.1. Later in the proof of Theorem 4.2 we show that this approach is indeed correct.

All puzzles except the input puzzle are generated by Gen and $C$. 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.

In Section 3.3.2 we have already described that the approach of fixing the puzzles on the consecutive coordinates may fail for dynamic weakly verifiable puzzles. The very similar arguments are also valid for the approach described above. More precisely, we would like to be sure that we never ask a hint query that prevents one of the (later) verification query to succeed. To satisfy the last requirement we split the set $Q$ into a set of advice queries and a set of attacking queries in the similar way as it is described in Section 3.3.2 and [DIJK09].

### 4.1.3 Domain partitioning

Let $hash : \mathcal{Q} \to \{0, 1, \ldots, 2(h+v)-1\}$, 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 $\mathcal{Q}$ 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 $(q_j, y_j)$ 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 $hash : \mathcal{Q} \to \{0, 1, \ldots, 2(h + v) - 1\}$.
**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

$$\Pr_{\pi,\rho}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1]. \tag{4.2}$$

For fixed *hash* and $P$ a *canonical success* of $C$ for bistrings $\pi$, $\rho$ is a situation where $CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1$.

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 4.3** ***(Success probability in solving 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 $\mathcal{Q} \to \{0, 1, \ldots, 2(h+v)-1\}$. Then 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 poly$(h, v, \frac{1}{\gamma}, n)$ 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 4.3).** We fix a problem poser $P$ and a solver $C$ for $P$ in the whole proof of Lemma 4.3. 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

$$\begin{aligned}\Pr_{hash \leftarrow \mathcal{H}}[hash(q_k) = \alpha \mid hash(q_l) = \beta] &= \Pr_{hash \leftarrow \mathcal{H}}[hash(q_k) = \alpha] \\ &= \frac{1}{2(h+v)}. \end{aligned} \tag{4.3}$$

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

$$\begin{aligned}
&\Pr_{hash \leftarrow \mathcal{H}}\Big[CanonicalSuccess^{P,C,hash}(\pi^*, \rho^*) = 1\Big] \\
&\quad = \Pr_{hash \leftarrow \mathcal{H}}\big[\, hash(q_j) = 0 \wedge (\forall i < j : hash(q_i) \neq 0)\big] \\
&\quad = \Pr_{hash \leftarrow \mathcal{H}}\big[\forall i < j : hash(q_i) \neq 0 \mid hash(q_j) = 0\big] \Pr_{hash \leftarrow \mathcal{H}}[hash(q_j) = 0] \\
&\quad \overset{(4.3)}{=} \frac{1}{2(h+v)}\left(1 - \Pr_{hash \leftarrow \mathcal{H}}[\exists i < j : hash(q_i) = 0 \mid hash(q_j) = 0]\right) \\
&\quad \overset{(*)}{\geq} \frac{1}{2(h+v)}\left(1 - \sum_{i<j} \Pr_{hash \leftarrow \mathcal{H}}[hash(q_i) = 0 | \, hash(q_j) = 0]\right) \\
&\quad \overset{(4.3)}{=} \frac{1}{2(h+v)}\left(1 - \sum_{i<j} \Pr_{hash \leftarrow \mathcal{H}}[hash(q_i) = 0]\right) \\
&\quad \overset{(4.3)}{\geq} \frac{1}{4(h+v)},
\end{aligned} \tag{4.4}$$

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

$$\begin{aligned}
&\Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi,\rho}}\Big[CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1\Big] \\
&\quad = \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi,\rho}}\Big[CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1 \mid (\pi,\rho) \in \mathcal{P}_{Success}\Big] \Pr_{\pi,\rho}[(\pi,\rho) \in \mathcal{P}_{Success}] \\
&\qquad + \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi,\rho}}\Big[CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1 \mid (\pi,\rho) \notin \mathcal{P}_{Success}\Big] \Pr_{\pi,\rho}[(\pi,\rho) \notin \mathcal{P}_{Success}] \\
&\quad = \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi,\rho}}\Big[CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1 \mid (\pi,\rho) \in \mathcal{P}_{Success}\Big] \Pr_{\pi,\rho}[(\pi,\rho) \in \mathcal{P}_{Success}] \\
&\quad \geq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi,\rho}}\Big[CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1 \mid (\pi,\rho) \in \mathcal{P}_{Success}\Big] \cdot \gamma \\
&\quad = \mathop{\mathbb{E}}_{(\pi,\rho) \in \mathcal{P}_{Success}}\Big[\Pr_{hash \leftarrow \mathcal{H}}[CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1]\Big] \cdot \gamma \\
&\quad \overset{(4.4)}{\geq} \frac{\gamma}{4(h+v)}.
\end{aligned} \tag{4.5}$$

**Algorithm** $\text{FindHash}^{P,C}(\gamma, n, h, v)$

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

```
for i := 1 to 32n(h+v)^2/γ^2 do:
    hash ← H
    count := 0
    for j := 1 to 32n(h+v)^2/γ^2 do:
        π ← {0,1}^n
        ρ ← {0,1}^*
        if CanonicalSuccess^{P,C,hash}(π,ρ) = 1 then
            count := count + 1
    if count ≥ γ/(12(h+v)) · 32(h+v)^2/γ^2 n then
        return hash
return ⊥
```

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

$$\Pr_{\pi,\rho}\Big[CanonicalSuccess^{P,C,hash}(\pi,\rho) = 1\Big] \geq \frac{\gamma}{8(h+v)}, \tag{4.6}$$

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

$$\Pr_{\pi,\rho}\Big[CanonicalSuccess^{P,C,hash}(\pi,\rho)=1\Big] \leq \frac{\gamma}{16(h+v)}. \tag{4.7}$$

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 (4.7) it follows that $\mathbb{E}_{\pi,\rho}[X_i] \leq \frac{\gamma}{16(h+v)}$. In inequalities (4.8) and (4.9) in steps denoted with $(*)$ we used the trivial facts that $h+v \geq 1$ and $\gamma \leq 1$. For any fixed $hash \in \mathcal{H}_{Bad}$ using the Chernoff bound we get

$$\begin{aligned}\Pr_{\pi,\rho}\left[\frac{1}{N}\sum_{i=1}^{N} X_i \geq \frac{\gamma}{12(h+v)}\right] &\leq \Pr_{\pi,\rho}\left[\frac{1}{N}\sum_{i=1}^{N} X_i \geq \big(1+\frac{1}{3}\big)\mathbb{E}[X_i]\right] \\ &\leq e^{-\frac{\gamma}{16(h+v)}N/27} \leq e^{-\frac{2}{27}\frac{(h+v)}{\gamma}n} \overset{(*)}{\leq} e^{-\frac{2}{27}n}\end{aligned} \tag{4.8}$$

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

$$\begin{aligned}\Pr_{\pi,\rho}\left[\frac{1}{N}\sum_{i=1}^{N} X_i \leq \frac{\gamma}{12(h+v)}\right] &\leq \Pr_{\pi,\rho}\left[\frac{1}{N}\sum_{i=1}^{N} X_i \leq \big(1-\frac{1}{3}\big)\mathbb{E}[X_i]\right] \\ &\leq e^{-\frac{\gamma}{8(h+v)}N/18} \leq e^{-\frac{2}{9}\frac{(h+v)}{\gamma}n} \overset{(*)}{\leq} e^{-\frac{2}{9}n}\end{aligned} \tag{4.9}$$

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

$$\Pr_{hash \leftarrow \mathcal{H}}[hash \in \mathcal{H}_{Good}] < \frac{\gamma}{8(h+v)}. \tag{4.10}$$

Using (4.10) we conclude that it is possible to bound the probability of

the canonical success as follows

$$\begin{aligned}
&\Pr_{\substack{hash\leftarrow\mathcal{H}\\ \pi,\rho}}[CanonicalSuccess^{P,C,hash}(\pi,\rho)=1]\\
&\quad= \Pr_{\substack{hash\leftarrow\mathcal{H}\\ \pi,\rho}}[CanonicalSuccess^{P,C,hash}(\pi,\rho)=1 \mid hash\in\mathcal{H}_{Good}]\Pr_{hash\leftarrow\mathcal{H}}[hash\in\mathcal{H}_{Good}]\\
&\qquad+\Pr_{\substack{hash\leftarrow\mathcal{H}\\ \pi,\rho}}[CanonicalSuccess^{P,C,hash}(\pi,\rho)=1 \mid hash\notin\mathcal{H}_{Good}]\Pr_{hash\leftarrow\mathcal{H}}[hash\notin\mathcal{H}_{Good}]\\
&\quad\le \Pr_{hash\leftarrow\mathcal{H}}[hash\in\mathcal{H}_{Good}]+\Pr_{\substack{hash\leftarrow\mathcal{H}\\ \pi,\rho}}[CanonicalSuccess^{P,C,hash}(\pi,\rho)=1 \mid hash\notin\mathcal{H}_{Good}]\\
&\quad\overset{\substack{(4.6)\\(4.10)}}{<} \frac{\gamma}{8(h+v)}+\frac{\gamma}{8(h+v)}=\frac{\gamma}{4(h+v)},
\end{aligned}$$

which contradicts (4.5). Therefore, 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 $\Pr_{hash\leftarrow\mathcal{H}}[hash\in\mathcal{H}_{Good}]\ge\frac{\gamma}{8(h+v)}$ and $K\le\frac{32(h+v)^2}{\gamma^2}n$ to conclude

$$\begin{aligned}
\Pr_{hash\leftarrow\mathcal{H}}\Big[\bigcap_{1\le i\le K}Y_i\Big]\le\left(1-\frac{\gamma}{8(h+v)}\right)^{\frac{32(h+v)^2}{\gamma^2}n} &\le e^{-\frac{\gamma}{8(h+v)}\frac{32(h+v)^2}{\gamma^2}n}\\
&\le e^{-\frac{4(h+v)}{\gamma}n}\le e^{-n}.
\end{aligned}$$

It is clear that running time of FindHash is $poly(n,h,v,\gamma)$ with oracle calls. This finishes the proof of Lemma 4.3. □

### 4.1.4 Hardness amplification proof with partitioned query domain

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 $\widetilde{C}_2$.

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

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

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 $\perp$.

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 4.4** *For fixed $P$, $C$, and hash the following statement is true*

$$\Pr_{\pi,\rho}[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1] \leq \Pr_{\substack{\pi,\rho \\ x:=\langle P(\pi),C_1(\rho)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P(\pi),C_1(\rho)\rangle_P}}[\Gamma_V(\widetilde{C}_2^{\Gamma_H, C_2, hash}(x, \rho)) = 1].$$

**Proof.** If for fixed $\pi$, $\rho$, 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

$$\begin{aligned}
&\Pr_{\pi,\rho}\Big[CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1\Big] \\
&\qquad \leq \mathop{\mathbb{E}}_{\substack{\pi,\rho \\ x:=\langle P(\pi),C_1(\rho)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P(\pi),C_1(\rho)\rangle_P}}\big[\Gamma_V(\widetilde{C}_2^{\Gamma_H, C_2, hash}(x, \rho)) = 1\big] \\
&\qquad = \Pr_{\substack{\pi,\rho \\ x:=\langle P(\pi),C_1(\rho)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P(\pi),C_1(\rho)\rangle_P}}\big[\Gamma_V(\widetilde{C}_2^{\Gamma_H, C_2, hash}(x, \rho)) = 1\big]
\end{aligned}$$

□

**Lemma 4.5** ***(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 $: \mathcal{Q} \to \{0, 1, \ldots, 2(h+v-1)\}$ 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 $D := (D_1, D_2)$ such that the following holds:*
*If $\widetilde{C}$ is such that*

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

*then $D$ satisfies almost surely*

$$\Pr_{\substack{\pi \in \{0,1\}^n, \rho \in \{0,1\}^* \\ x := \langle P^{(1)}(\pi), D_1^{\widetilde{C}}(\rho) \rangle_{trans} \\ (\Gamma_H, \Gamma_V) := \langle P^{(1)}(\pi), D_1^{\widetilde{C}}(\rho) \rangle_{P^{(1)}}}} [\Gamma_V(D_2^{P^{(1)}, \widetilde{C}, hash, g, \Gamma_H}(x, \rho)) = 1] \geq \delta + \frac{\varepsilon}{6k}.$$

*Furthermore, $D$ asks at most $\frac{6k}{\varepsilon} \log\left(\frac{6k}{\varepsilon}\right) h$ hint queries and no verification queries. Finally, $Time(Gen) = poly(k, \frac{1}{\varepsilon}, n)$ with oracle access to $\widetilde{C}$.*

Before we give the proof of Lemma 4.5 we have to define additional algorithms. First, in the following code excerpt the algorithm Gen from Lemma 4.5 is defined. The procedures used by Gen are defined on the succeeding code excerpts.

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

---

**Oracle:** A poser $P^{(1)}$, a circuit $\widetilde{C}$ for $P^{(g)}$, functions $g : \{0,1\}^k \to \{0,1\}$, $hash : \mathcal{Q} \to \{0, 1, \ldots, 2(h+v) - 1\}$.
**Input:** Parameters $\varepsilon$, $\delta$, $n$, $k$.
**Output:** A circuit $D$.

---

**for** $i := 1$ **to** $\frac{6k}{\varepsilon} n$ **do:**
  $\pi^* \xleftarrow{\$} \{0,1\}^n$
  $\widetilde{S}_{\pi^*,0} := \text{EstimateSurplus}^{P^{(1)}, \widetilde{C}, g, hash}(\pi^*, 0, k, \varepsilon, \delta, n)$
  $\widetilde{S}_{\pi^*,1} := \text{EstimateSurplus}^{P^{(1)}, \widetilde{C}, g, hash}(\pi^*, 1, k, \varepsilon, \delta, n)$
  **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 $\pi^*$.

$\widetilde{C}' := (C_1', \widetilde{C}_2')$
$g'(b_2, \ldots, b_k) := g(b, b_2, \ldots, b_k)$
**return** $Gen^{P^{(1)}, \widetilde{C}', g', hash}(\varepsilon, \delta, n, k-1)$
// *all estimates are lower than* $(1 - \frac{3}{4k})\varepsilon$
**return** $D^{P^{(1)}, \widetilde{C}, hash, g}$

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 $\Pr_{u \leftarrow \mu_\delta^k}[g(b, u_2, \ldots, u_k) = 1]$.

**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 4.6** *The algorithm EstimateFunctionProbability$^g(b, k, \varepsilon, \delta, n)$ outputs an estimate $\widetilde{g}_b$ such that $|\widetilde{g}_b - \Pr_{u \leftarrow \mu_\delta^k}[g(b, u_2, \ldots, u_k) = 1]| \le \frac{\varepsilon}{8k}$ 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 $1 \le i \le N$ the random variable $K_i$ takes value $g_i$. We use the Chernoff bound to obtain

$$\Pr\Big[\Big|\widetilde{g}_b - \Pr_{u \leftarrow \mu_\delta^k}[g(b, u_2, \ldots, u_k) = 1]\Big| \ge \frac{\varepsilon}{8k}\Big]$$
$$= \Pr\Big[\Big|\Big(\frac{1}{N}\sum_{i=1}^N K_i\Big) - \mathbb{E}_{u \leftarrow \mu_\delta^k}[g(b, u_2, \ldots, u_k)]\Big| \ge \frac{\varepsilon}{8k}\Big] \le 2 \cdot e^{-n/3}. \square$$

The algorithm EvalutePuzzles$^{P^{(1)},\widetilde{C},hash}(\pi^{(k)},\rho,n,k)$ evaluates which of the $k$ puzzles of the $k$-wise direct product of $P^{(1)}$ are solved successfully by $\widetilde{C}(\rho) := (C_1,\widetilde{C}_2)(\rho)$. 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 $\widetilde{C}$. Therefore, the algorithm EvalutePuzzles runs $k$ times $P^{(1)}(\pi_u)$ 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$^{P^{(1)},\widetilde{C},hash}(\pi^{(k)},\rho,n,k)$

---

**Oracle:** A problem poser $P^{(1)}$, 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:** Bitstrings $\pi^{(k)} \in \{0,1\}^{kn}$, $\rho \in \{0,1\}^*$, parameters $n$, $k$.
**Output**: A tuple $(c_1,\ldots,c_k) \in \{0,1\}^k$.

---

**for** $i := 1$ **to** $k$ **do:** *//simulate $k$ rounds of interaction*
  $(\Gamma_V^i,\Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho)\rangle^i_{P^{(1)}}$
  $x_i := \langle P^{(1)}(\pi_i), C_1(\rho)\rangle^i_{trans}$
$x := (x_1,\ldots,x_k)$
$\Gamma_H^{(k)} := (\Gamma_H^1,\ldots,\Gamma_H^k)$
$(q,y_1,\ldots,y_k) := \widetilde{C}_2^{\Gamma_H^{(k)},hash}(x,\rho)$
**if** $(q,y_1,\ldots,y_k) = \perp$ **then**
  **return** $(0,\ldots,0)$
$(c_1,\ldots,c_k) := (\Gamma_V^1(q,y_1),\ldots,\Gamma_V^k(q,y_k))$
**return** $(c_1,\ldots,c_k)$

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 first puzzle defined by $P^{(1)}(\pi_1)$ is neglected, and instead a bit $b$ is used. More formally, we define the surplus $S_{\pi^*,b}$ as

$$S_{\pi^*,b} = \Pr_{\pi^{(k)},\rho}\left[g(b, c_2, \ldots, c_k) = 1 \mid \pi_1 = \pi^*\right] - \Pr_{u \leftarrow \mu_\delta^k}\left[g(b, u_2, \ldots, u_k) = 1\right], \tag{4.11}$$

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

The algorithm EstimateSurplus 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 $S_{\pi^*,b}$.

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

**Lemma 4.7** *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 4.6 which yields that $\frac{1}{N}\sum_{i=1}^N \widetilde{s}^i_{\pi^*,b}$ differs from $\mathbb{E}[g(b, c_2, \ldots, c_k)]$ by at most $\frac{\varepsilon}{8k}$ almost surely. Together, with Lemma 4.6 we conclude that the surplus estimate returned by EstimateSurplus differs from $S_{\pi^*,b}$ by at most $\frac{\varepsilon}{4k}$ almost surely.□

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

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

**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.

**Circuit** $\widetilde{C}_2^{\prime\Gamma_H^{(k-1)},\widetilde{C},hash}(x^{(k-1)},\rho)$

---

**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 $\langle P^{(1)}(\pi^*), C_1(\rho)\rangle^1$
$(\Gamma_H^*,\Gamma_V^*) := \langle P^{(1)}(\pi^*), C_1(\rho)\rangle^1_{P^{(1)}}$
$x^* := \langle P^{(1)}(\pi^*), C_1(\rho)\rangle^1_{trans}$
$\Gamma_H^{(k)} := (\Gamma_H^*,\Gamma_H^2,\ldots,\Gamma_H^k)$
$x^{(k)} := (x^*,x_2,\ldots,x_k)$
$(q,y_1,\ldots,y_k) := \widetilde{C}_2^{\Gamma_H^{(k)},hash}(x^{(k)},\rho)$
**return** $(q,y_2,\ldots,y_k)$

We are ready to define the solver circuit $D = (D_1,D_2)$ for $P^{(1)}$ output by Gen.

**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 $x^* := \langle P^{(1)}, C_1(\rho)\rangle^1_{trans}$.

**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 $hash : \mathcal{Q} \to \{0,1,\ldots,2(h+v)-1\}$, $g : \{0,1\}^k \to \{0,1\}$,
a hint circuit $\Gamma_H$ for $P^{(1)}$.
**Input:** A communiation 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:**
  $(\pi_2, \ldots, \pi_k) \leftarrow$ read next $(k-1)\cdot n$ bits from $\sigma$
  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:**
    **run** $\langle P^{(1)}(\pi_i), C_1(\rho)\rangle^i$
      $(\Gamma_V^i, \Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho)\rangle^i_{P^{(1)}}$
      $x_i := \langle P^{(1)}(\pi_i), C_1(\rho)\rangle^i_{trans}$
  $\Gamma_H^{(k)}(q) := (\Gamma_H(q), \Gamma_H^2(q), \ldots, \Gamma_H^k(q))$
  $(q, y^*, y_2, \ldots, y_k) := \widetilde{C}_2^{\Gamma_H^{(k)}, hash}((x^*, x_2, \ldots, x_k), \rho)$
  $(c_2, \ldots, c_k) := (\Gamma_V^2(q, y_2), \ldots, \Gamma_V^k(q, y_k))$
  **if** $g(1, c_2, \ldots, c_k) = 1$ **and** $g(0, c_2, \ldots, c_k) = 0$ **then**
    **return** $(q, y^*)$
**return** $\perp$

**Proof (Lemma 4.5).** 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. 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 4.5 it follows that $\widetilde{C}$ succeeds with probability at least $\delta + \varepsilon$. Hence, $D$ trivially satisfies the statement of Lemma 4.5. In the latter case $g$ is a constant function, and the statement is vacuously true.

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

In the former case we define 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 $\widetilde{C} := (C_1, \widetilde{C}_2)$. By Lemma 4.7 it follows that $S_{\pi^*,b} \geq \widetilde{S}_{\pi^*,b} - \frac{\varepsilon}{4k} \geq (1 - \frac{1}{k})\varepsilon$ almost surely. Therefore, the circuit $\widetilde{C}'$ succeeds in solving the $(k-1)$-wise direct product of puzzles with probability at least $\Pr_{u \leftarrow \mu_\delta^{(k-1)}}[g'(u_1, \ldots, u_{k-1})] + (1 - \frac{1}{k})\varepsilon$. In this case $\widetilde{C}'$ satisfies the conditions of Lemma 4.5 for the $(k-1)$-wise direct product of puzzles. Therefore, the recursive call to Gen with access to $g'$ and $\widetilde{C}$ returns a circuit $D = (D_1, D_2)$ that satisfies with high probability

$$\Pr_{\substack{\pi,\rho \\ x := \langle P^{(1)}(\pi), D_1^{\widetilde{C}}(\rho)\rangle_{trans} \\ (\Gamma_H, \Gamma_V) := \langle P^{(1)}(\pi), D_1^{\widetilde{C}}(\rho)\rangle_{P^{(1)}}}}[\Gamma_V(D_2^{P^{(1)}, \widetilde{C}, hash, g, \Gamma_H}(x, \rho)) = 1] \geq \delta + \left(1 - \frac{1}{k}\right)\frac{\varepsilon}{6(k-1)}$$

$$= \delta + \frac{\varepsilon}{6k}. \tag{4.12}$$

We consider now a case when none of the estimates is greater than $(1 - \frac{3}{4k})\varepsilon$. Intuitively this means that $\widetilde{C}$ does not succeed on the remaining $k-1$ puzzles with much higher probability than an algorithm that successfully solves each puzzle with probability $\delta$. From the assumptions of Lemma 4.5 we know that on all $k$ puzzles the success probability of $\widetilde{C}$ is higher. Therefore, it is likely that the first puzzle is correctly solved unusually often. It remains to prove that this intuition is indeed correct.

We fix notation as in the code excerpt of $D_2$. Let us consider a single iteration of the outer loop of $D_2$, in which values $\pi_2, \ldots, \pi_k$ are fixed. We write $\pi_1$ to denote randomness used by the problem poser in generation of 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 $\mathcal{G}_b := \{(b_1, b_2, \ldots, b_k) : g(b, b_2, \ldots, b_k) = 1\}$. Using this notation we express

$$\begin{aligned} \Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_b] &= \Pr_{u \leftarrow \mu_\delta^k}[g(b, u_2, \ldots, u_k) = 1] \\ \Pr_{\pi^{(k)}, \rho}[c \in \mathcal{G}_b] &= \Pr_{\pi^{(k)}, \rho}[g(b, c_2, \ldots, c_k) = 1]. \end{aligned} \tag{4.13}$$

Let us fix randomness $\pi_1$ of the problem poser $P^{(1)}$ to $\pi^*$. We use (4.11) and (4.13) to obtain

$$\begin{aligned} &\Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_1] - \Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_0] \\ &= \Pr_{\pi^{(k)}, \rho}[c \in \mathcal{G}_1 \mid \pi_1 = \pi^*] - \Pr_{\pi^{(k)}, \rho}[c \in \mathcal{G}_0 \mid \pi_1 = \pi^*] - (S_{\pi^*,1} - S_{\pi^*,0}) \end{aligned} \tag{4.14}$$

By monotonicity of $g$ it holds $\mathcal{G}_0 \subseteq \mathcal{G}_1$, and we write (4.14) as

$$\Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0] = \Pr_{\pi^{(k)}, \rho}[c \in \mathcal{G}_1 \setminus \mathcal{G}_0 \mid \pi_1 = \pi^*] - (S_{\pi^*,1} - S_{\pi^*,0}). \tag{4.15}$$

We multiply both sides of (4.15) by

$$\Pr_{\substack{r \\ x^* := \langle P^{(1)}(\pi^*), D_1(r) \rangle_{trans} \\ (\Gamma_V, \Gamma_H) := \langle P^{(1)}(\pi^*), D_1(r) \rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*, r)) = 1] / \Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0],$$

which yields

$$\begin{aligned}
&\Pr_{\substack{r\\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}\\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*,r))=1]\\
&\quad= \Pr_{\substack{r\\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}\\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*,r))=1]\Pr_{\pi^{(k)},\rho}[c\in\mathcal{G}_1\setminus\mathcal{G}_0\mid \pi_1=\pi^*]\frac{1}{\Pr_{u\leftarrow\mu_\delta^k}[u\in\mathcal{G}_1\setminus\mathcal{G}_0]}\\
&\qquad- \Pr_{\substack{r\\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}\\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*,r))=1](S_{\pi^*,1}-S_{\pi^*,0})\frac{1}{\Pr_{u\leftarrow\mu_\delta^k}[u\in\mathcal{G}_1\setminus\mathcal{G}_0]}.
\end{aligned}\tag{4.16}$$

Let us study the first summand of (4.16). First, we have

$$\begin{aligned}
&\Pr_{\substack{r\\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}\\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*,r))=1]\\
&\quad= \Pr_{\substack{r\\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}\\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*,r))=1|D_2(x^*,r)\neq\perp]\Pr_{\substack{r\\ x^*=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}}}[D_2(x^*,r)\neq\perp]\\
&\quad\overset{(*)}{=} \Pr_{\pi^{(k)},\rho}[c_1=1\mid c\in\mathcal{G}_1\setminus\mathcal{G}_0,\pi_1=\pi^*]\Pr_{\substack{r\\ x^*=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}}}[D_2(x^*,r)\neq\perp]
\end{aligned}\tag{4.17}$$

where in $(*)$ we use the observation that $D_2(x^*,r)\neq\perp$ 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 $\Gamma_V(D_2(x^*,r))=1$ occurs if and only if $c_1=1$. Inserting (4.17) to the numerator of the first summand of (4.16) yields

$$\begin{aligned}
&\Pr_{\substack{r\\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}\\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*,r))=1]\Pr_{\pi^{(k)},\rho}[c\in\mathcal{G}_1\setminus\mathcal{G}_0\mid\pi_1=\pi^*]\\
&\quad= \Pr_{\substack{r\\ x^*=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}}}[D_2(x^*,r)\neq\perp]\Pr_{\pi^{(k)},\rho}[c_1=1\mid c\in\mathcal{G}_1\setminus\mathcal{G}_0,\pi_1=\pi^*]\Pr_{\pi^{(k)},\rho}[c\in\mathcal{G}_1\setminus\mathcal{G}_0\mid\pi_1=\pi^*].
\end{aligned}\tag{4.18}$$

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

$$\Pr_{\pi^{(k)},\rho}[c_1=1\mid c\in\mathcal{G}_1\setminus\mathcal{G}_0,\pi_1=\pi^*]\Pr_{\pi^{(k)},\rho}[c\in\mathcal{G}_1\setminus\mathcal{G}_0\mid\pi_1=\pi^*]\leq\frac{\varepsilon}{6k}.\tag{4.19}$$

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 $\perp$ 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

$$\Pr_{\substack{r \\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}}}[D_2(x^*,r)=\perp] \leq \left(1-\frac{\varepsilon}{6k}\right)^{\frac{6k}{\varepsilon}\log(\frac{6k}{\varepsilon})} \leq \frac{\varepsilon}{6k}. \tag{4.20}$$

We conclude that in both aforementioned cases using (4.19) and (4.20) the following holds

$$\begin{aligned}
&\Pr_{\substack{r \\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans}}}[D_2(x^*,r)\neq\perp] \Pr_{\pi^{(k)},\rho}[c_1=1 \mid c\in\mathcal{G}_1\setminus\mathcal{G}_0,\pi_1=\pi^*] \Pr_{\pi^{(k)},\rho}[c\in\mathcal{G}_1\setminus\mathcal{G}_0 \mid \pi_1=\pi^*] \\
&\geq \Pr_{\pi^{(k)},\rho}[c_1=1 \mid c\in\mathcal{G}_1\setminus\mathcal{G}_0,\pi_1=\pi^*] \Pr_{\pi^{(k)},\rho}[c\in\mathcal{G}_1\setminus\mathcal{G}_0 \mid \pi_1=\pi^*] - \frac{\varepsilon}{6k} \\
&= \Pr_{\pi^{(k)},\rho}[c_1=1 \wedge c\in\mathcal{G}_1\setminus\mathcal{G}_0 \mid \pi_1=\pi^*] - \frac{\varepsilon}{6k} \\
&= \Pr_{\pi^{(k)},\rho}[g(c)=1 \mid \pi_1=\pi^*] - \Pr_{\pi^{(k)},\rho}[c\in\mathcal{G}_0 \mid \pi_1=\pi^*] - \frac{\varepsilon}{6k} \\
&\overset{(4.11)}{=} \Pr_{\pi^{(k)},\rho}[g(c)=1 \mid \pi_1=\pi^*] - \Pr_{u\leftarrow\mu_\delta^{(k)}}[u\in\mathcal{G}_0] - S_{\pi^*,0} - \frac{\varepsilon}{6k}.
\end{aligned} \tag{4.21}$$

We insert (4.21) into (4.16) and calculate the expected value of over $\pi^*$ which yields

$$\begin{aligned}
\Pr_{\substack{\pi,r \\ x:=\langle P^{(1)}(\pi),D_1(r)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x,r))=1] &\geq \mathbb{E}_{\pi^*}\left[\frac{\Pr_{\pi^{(k)},\rho}[g(c)=1|\pi_1=\pi^*]-\Pr_{u\leftarrow\mu_\delta^{(k)}}[u\in\mathcal{G}_0]-\frac{\varepsilon}{6k}}{\Pr_{u\leftarrow\mu_\delta^{(k)}}[u\in\mathcal{G}_1\setminus\mathcal{G}_0]}\right] \\
&- \mathbb{E}_{\pi^*}\left[\left(\Pr_{\substack{r \\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x^*,r))=1](S_{\pi^*,1}-S_{\pi^*,0})+S_{\pi^*,0}\right)\frac{1}{\Pr_{u\leftarrow\mu_\delta^k}[u\in\mathcal{G}_1\setminus\mathcal{G}_0]}\right].
\end{aligned} \tag{4.22}$$

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

$$\Pr_{\pi}\left[\left(S_{\pi,0}\leq(1-\frac{1}{2k})\varepsilon\right)\wedge\left(S_{\pi,1}\leq(1-\frac{1}{2k})\varepsilon\right)\right]<1-\frac{\varepsilon}{6k}, \tag{4.23}$$

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

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

almost surely, where we used Lemma 4.7. Therefore, under the assumption that Gen does not recurse with high probability it holds

$$\Pr_{\pi,\rho}\left[\left(S_{\pi,0}\leq(1-\frac{1}{2k})\varepsilon\right)\wedge\left(S_{\pi,1}\leq(1-\frac{1}{2k})\varepsilon\right)\right]\geq 1-\frac{\varepsilon}{6k}. \tag{4.24}$$

Let us define a set

$$\mathcal{W} = \left\{ \pi : \left( S_{\pi,0} \leq (1 - \frac{1}{2k})\varepsilon \right) \wedge \left( S_{\pi,1} \leq (1 - \frac{1}{2k})\varepsilon \right) \right\}. \tag{4.25}$$

Additionally, let $\overline{\mathcal{W}}$ denote the complement of $\mathcal{W}$. We bound the numerator of the second summand in (4.22)

$$\begin{aligned}
&\mathbb{E}_{\pi^*}[S_{\pi^*,0} + \Pr_{\substack{r \\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}} [\Gamma_V(D_2(x^*,r)) = 1](S_{\pi^*,1} - S_{\pi^*,0})] \\
&\quad = \mathbb{E}_{\pi^*}\Big[S_{\pi^*,0} + \Pr_{\substack{r \\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}} [\Gamma_V(D_2(x^*,r) = 1](S_{\pi^*,1} - S_{\pi^*,0}) \mid \pi^* \in \overline{\mathcal{W}}] \Pr_{\pi^*}[\pi^* \in \overline{\mathcal{W}}]\Big] \\
&\qquad + \mathbb{E}_{\pi^*}\Big[S_{\pi^*,0} + \Pr_{\substack{r \\ x^*:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}} [\Gamma_V(D_2(x^*,r)) = 1](S_{\pi^*,1} - S_{\pi^*,0}) \mid \pi^* \in \mathcal{W}] \Pr_{\pi^*}[\pi^* \in \mathcal{W}]\Big] \\
&\quad \leq \frac{\varepsilon}{6k} + \mathbb{E}_{\pi^*}\Big[S_{\pi^*,0} + \Pr_{\substack{r \\ x:=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{trans} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(r)\rangle_{P^{(1)}}}} \big[\Gamma_V(D_2^{\widetilde{C}}(x^*,r)) = 1\big] \big((1 - \frac{1}{2k})\varepsilon - S_{\pi^*,0}\big) \mid \pi^* \in \mathcal{W}\Big] \\
&\quad \leq \frac{\varepsilon}{6k} + (1 - \frac{1}{2k})\varepsilon = (1 - \frac{1}{3k})\varepsilon.
\end{aligned} \tag{4.26}$$

We observe that

$$\begin{aligned}
\Pr_{u \leftarrow \mu_\delta^k}[g(u) = 1] &= \Pr[u \in \mathcal{G}_0 \vee (u \in \mathcal{G}_1 \setminus \mathcal{G}_0 \wedge u_1 = 1)] \\
&= \Pr[u \in \mathcal{G}_0] + \delta \Pr[u \in \mathcal{G}_1 \setminus \mathcal{G}_0].
\end{aligned} \tag{4.27}$$

Finally, we insert (4.26) into (4.22) which yields

$$\Pr_{\substack{\pi,\rho \\ x:=\langle P^{(1)}(\pi),D_1(\rho)\rangle_{\text{trans}} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi),D_1(\rho)\rangle_{P^{(1)}}}}[\Gamma_V(D_2(x,\rho)) = 1] \geq \mathop{\mathbb{E}}_{\pi^*}\left[\frac{\Pr_{\pi^{(k)},\rho}[g(c) = 1 \mid \pi_1 = \pi^*] - \Pr_{u \leftarrow \mu_\delta^k}[u \in G_0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0]}\right].$$

From the assumptions of Lemma 4.5 it follows that

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

Hence, we obtain

$$\begin{aligned}
\Pr_{\substack{\pi^*,\rho \\ x:=\langle P^{(1)}(\pi^*),D_1(\rho)\rangle_{\text{trans}} \\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi^*),D_1(\rho)\rangle_{P^{(1)}}}} [\Gamma_V(D_2(x,\rho)) = 1] &\geq \frac{\Pr_{u \leftarrow \mu_\delta^k}[g(u) = 1] + \varepsilon - \Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0]} \\
&\overset{(4.27)}{\geq} \frac{\varepsilon + \delta \Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{u \leftarrow \mu_\delta^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0]} \geq \delta + \frac{\varepsilon}{6k}
\end{aligned} \tag{4.28}$$

Clearly, the running time of Gen is $poly(k, \frac{1}{\varepsilon}, n)$. This finishes the proof of Lemma 4.5. □

### 4.1.5 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 4.5 we can build a circuit that returns a solution that is likely to be good. It remains to use these build blocks and prove Theorem 4.2.

**Proof (of Theorem 4.2).** Let $\widetilde{\text{Gen}}$ be the algorithm from Theorem 4.2, and $\widetilde{D} = (D_1, D_2)$ the corresponding solver circuit for $P$ that is output by $\widetilde{\text{Gen}}$ as in Theorem 4.2. They are defined on the following code excerpts.

**Circuit** $\widetilde{D}_2^{D,P^{(1)},hash,g,\Gamma_V,\Gamma_H}(x, \rho)$

**Oracle:** A circuit $D := (D_1, \widetilde{D}_2)$ from Lemma 4.5, a problem poser $P^{(1)}$, functions $hash : Q \to \{0, 1, \ldots, 2(h+v)-1\}$, $g : \{0,1\}^k \to \{0,1\}$ 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 $\Gamma_V$.

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

**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 4.4 with oracle access to $C$, $hash$.
$D := Gen^{P^{(1)},\widetilde{C},g,hash}(\varepsilon, \delta, n, k)$
**return** $\widetilde{D} := (D_1, \widetilde{D}_2)$

We show that Theorem 4.2 follows from Lemma 4.3 and Lemma 4.5. 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

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

First, we note that $C$ meets the requirements of Lemma 4.3. Thus, the algorithm $\widetilde{\text{Gen}}$ can call FindHash to obtain *hash* such that with high probability it holds

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

Applying Lemma 4.4 for $C$ we obtain a circuit $\widetilde{C}=(C_1,\widetilde{C}_2)$ that satisfies

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

Now, we use the algorithm Gen as in Lemma 4.5 that yields a circuit $D=(D_1,D_2)$ which with high probability satisfies

$$\Pr_{\substack{\pi,\rho\\ x:=\langle P^{(1)}(\pi),D_1^{\widetilde{C}}(\rho)\rangle_{trans}\\ (\Gamma_H,\Gamma_V):=\langle P^{(1)}(\pi),D_1^{\widetilde{C}}(\rho)\rangle_{P^{(1)}}}}[\Gamma_V(D_2^{P^{(1)},\widetilde{C},hash,g,\Gamma_H}(x,\rho))=1] \geq (\delta+\frac{\varepsilon}{6k}). \tag{4.29}$$

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

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

The running time of FindHash is $\mathit{poly}(h,v,\frac{1}{\varepsilon},n)$ with oracle calls and of Gen $\mathit{poly}(k,\frac{1}{\varepsilon},n)$ with oracle access. Thus, the overall running time of $\widetilde{\mathit{Gen}}$ is $\mathit{poly}(k,\frac{1}{\varepsilon},h,v,n,t)$ 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 $\mathit{Size}(\widetilde{D}) \leq \mathit{Size}(C)\cdot\frac{6k}{\varepsilon}$. This finishes the proof of Theorem 4.2. □

## 4.2 Discussion

Appendix A

# Appendix

## A.1 Basic Inequalities

**Lemma A.1 (Chernoff Bounds)** *For independent, identically distributed Bernoulli random variables* $X_1, \ldots, X_n$ *with* $X := \sum_{i=1}^{n} X_i$ *with* $\Pr[X_i = 1] = p_i$ *and* $\Pr[X_i = 0] = 1 - p_i$ *for all* $1 \leq i \leq n$*. we have the following inequalities for* $0 \leq \delta \leq 1$ *and* $\mathbb{E}[X] = \sum_{i=1}^{n} p_i$*:*

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

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

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

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