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# Chapter 1

# Introduction

# 1.1 Security Amplification Theorems

Introduction to security amplification theorems and hardness implication statements. Example of classical results. Problems captured by weakly verifiable puzzles. Contribution of this thesis.

- 1.2 Weakly verifiable puzzles
- 1.3 Contribution of the Thesis
- 1.4 Organization of the Thesis

Overview of the content of the succeeding chapters.

# Chapter 2

# **Preliminaries**

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

#### 2.1 Notation and Definitions

(Algorithms, Bitstrings and Circuits) We define a Boolean circuit as a directed acyclic graph with input vertices and vertices implementing logical functions and, or, and not. We denote Boolean circuits using capital letters from the Greek or English alphabet. We define a probabilistic circuit as a Boolean circuit  $C_{m,n}: \{0,1\}^m \times \{0,1\}^n \to \{0,1\}^*$ . Additionally, we write  $C_{m,n}(x;r)$  which should be understood as a probabilistic circuit taking as input  $x \in \{0,1\}^m$  and auxiliary input  $r \in \{0,1\}^n$ . If a probabilistic circuit does not take any input, we abuse notation and write  $C_n(r)$ . Similarly, we use  $\{C_n\}_{n\in\mathbb{N}}$ to denote a family of probabilist circuits that takes only auxiliary input. We make sure that it is clear from the context that probabilistic circuits with only auxiliary input are not confused with circuits that do not take auxiliary input. 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). 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  $C_2$ . 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  $\delta$ .

**Define:** Distribution  $D^n$  check definition of [DIJK09]

**TODO:** Does it hold for search problems and for algorithms with not a single bit of output.

It is well known [AB09] that a probabilistic polynomial time algorithm can be represented as a circuit of polynomial size. Moreover, it can be computed in polynomial time and logarithmic space. Therefore, whenever we state a theorem about circuits it can be also generalized for the polynomial time algorithms.

We write  $poly(\alpha_1, \ldots, \alpha_n)$  to denote a polynomial on variables  $\alpha_1, \ldots, \alpha_n$ . 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 the number of steps required to evaluate A is bounded by poly(|x|), where |x| denotes the length of the input that A takes. Similarly, as for probabilistic circuits we write the randomness used by a probabilistic algorithm explicitly as a bitstring provided as an auxiliary input.

For a tuple  $x^{(l)}$  and an element  $x_{l+1}$  we use  $x^{(l)} \circ x_{l+1}$  to denote the concatenation of  $x^{(l)}$  and  $x_{(l+1)}$  which results in  $x^{(l+1)} := (x_1, \dots, x_l, x_{l+1})$ . (**Probabil**-

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

Let  $(\Omega_n, \mathcal{F}_n, \Pr)$  be a probability space and  $n \in \mathbb{N}$ . 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  $\Pr[E_n] \geq 1 - 2^{-n} poly(n)$ .

(Functions) 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

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

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

(Interactive protocols) We are often interested in situations where two probabilistic circuits interact with each other according to some protocol. We

limit ourselves to the cases where circuits interact by means of messages representable by bitstrings. Let  $\{A_n\}_{n\in\mathbb{N}}$  and  $\{B_n\}_{n\in\mathbb{N}}$  be families of circuits such that  $A_n:\{0,1\}^*\to\{0,1\}^*$  and  $B_n:\{0,1\}^*\to\{0,1\}^*$ . An interactive protocol is defined by a  $\{A_n\}_{n\in\mathbb{N}}$  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 k-th round we have  $m_k:=A_n(\rho_A,m_1,m_2,\cdots,m_{k-1})$  and in the k+1-th round  $B_n(\rho_B,m_1,m_2,\cdots,m_{k-1},m_k)$ . A protocol execution between two probabilistic circuits A and B is denoted by  $\langle A,B\rangle_B$ . The output of A in a protocol execution is denoted by  $\langle A,B\rangle_B$ . A sequence of all messages sent by A and B in the protocol execution is called a communication transcript and is denoted by  $\langle A,B\rangle_{trans}$ .

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

**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  $poly(\log |\mathcal{D}|, \log |cR|)$  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

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

It is well know [CW77] that there exists families of functions meeting the above criteria.

### Chapter 3

# Weakly Verifiable Cryptographic Primitives

This chapter gives an overview of dynamic interactive weakly verifiable puzzles. In Section 3.1 we define dynamic interactive weakly verifiable puzzles. Next, in Section 3.2, we describe a series of cryptographic primitives that are captured by the notion of dynamic interactive weakly verifiable puzzles. The Section 3.3 is devoted to the previous research concerning weakly verifiable puzzles.

# 3.1 Dynamic Interactive Weakly Verifiable Puzzles

We consider the following definition of dynamic interactive weakly verifiable puzzles.

**Definition 3.1 (Dynamic Interactive Weakly Verifiable Puzzle.)** A dynamic interactive weakly verifiable puzzle (DIWVP) is defined by a family of probabilistic circuits  $\{P_n\}_{n\in\mathbb{N}}$ . A circuit belonging to  $\{P_n\}_{n\in\mathbb{N}}$  is called a problem poser. A solver  $C_n := (C_1, C_2)$  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  $C_n(\rho) := (C_1, C_2)(\rho)$  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.

If for a certain DIWVP the number of hint queries is greater than zero, we say that such a puzzle is *dynamic*. Conversely, if a problem poser is not allowed to ask any hint queries, then such a puzzle is called *non-dynamic*.

We say that DIWVP is *interactive* if in the first phase the number of messages exchanged between a problem poser and a problem solver is greater than one.

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

**TODO:** This is not obvious as: a) algorithms are in BPP but they are search problems so no direct use of Cook-Levin b) this theorem is for deterministic algorithms

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

# 3.2 Examples

In this section we give examples of cryptographic constructions that are dynamic weakly verifiable puzzles.

#### 3.2.1 Message Authentication Codes

**TODO:** Runtime of problem poser and problem solver.

We consider the setting in which two parties a *sender* and a *receiver* communicate over an insecure channel. The messages of the sender may be read, 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 an adversary has no access. The sender appends to every message a tag which is computed as a function of the key and the message. The receiver, using the key, has a way to check whether an appended tag is valid for a received message. The receiver accepts a message if the tag is valid, otherwise it rejects. We require that it is hard for the adversary to find a tag and a message that is accepted by the receiver with non-negligible probability. We give the following formal definition of *Message Authentication Codes* based on [Mau13] and [Gol04].

**Definition 3.2 (Message Authentication Codes)** Let  $\mathcal{M} \subseteq \{0,1\}^*$  be a set of messages,  $\mathcal{K} \subseteq \{0,1\}^n$  a set of keys and  $\mathcal{T} \subseteq \{0,1\}^*$  a set of tags where  $n \in \mathbb{N}$ . We define the message authentication code (MAC) as am 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  $k \stackrel{\$}{\leftarrow} \mathcal{K}$  and  $\Gamma_H : \mathcal{M} \to \mathcal{T}$  be a polynomial size circuit computing a tag for a message where a key is fixed to k. We say that MAC is secure if there is no probabilistic polynomial time algorithm with oracle access to  $\Gamma_H$  that with nonnegligible probability outputs a message  $m \in \mathcal{M}$ , as well as a corresponding tag  $t \in \mathcal{T}$  such that f(m,k) = t, and  $\Gamma_H$  has not been queried for a tag of message m.

We show how MAC is captured by the dynamic interactive weakly verifiable puzzles. For fixed n the sender corresponds to a problem poser, the adversary to a problem solver, and the k key is a bitstring  $\pi \in \{0,1\}^n$  taken as auxiliary input by a problem poser. In the first phase, which is non interactive, the problem poser outputs a hint circuit  $\Gamma_H : \mathcal{M} \to \mathcal{T}$  that given a message computes a tag and a verification circuit  $\Gamma_V : \mathcal{M} \times \mathcal{T} \to \{0,1\}$  that on input  $m \in \mathcal{M}$  and  $t \in \mathcal{T}$  outputs one if and only if f(m,k) = t.

In the second phase the adversary takes no input  $(x^*)$  is empty string), as the first phase was non-interactive, and is given oracle access to  $\Gamma_H$  and  $\Gamma_V$ . 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 no 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 public key encryption scheme, and what it means for such a scheme to be secure. These definitions are based on [Gol04].

**TODO:** Length of public and private key

**Definition 3.3 (Public key signature scheme)** Let 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 is a security parameter. We call  $k_{priv}$  a private key and  $k_{pub}$  a public key.
- The signing algorithm S takes as input  $k_{priv}$ ,  $q \in \mathcal{Q}$  and outputs a signature  $s \in S$ .
- The verification algorithm takes as input  $k_{pub}$ ,  $q \in \mathcal{Q}$ , and  $s \in S$  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 the random coins of V and S.

We say that  $s \in S$  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 be a probabilistic polynomial time algorithm that takes as input  $k_{pub}$  and has oracle access to S. We say that the adversary succeeds if it finds a signature  $s \in S$  for a message  $q \in Q$  such that  $V(k_{pub}, q, s) = 1$ , 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 algorithm that succeeds with non negligible probability.

We show now that the public key signature schemes defined as above can be represented as a dynamic interactive weakly verifiable puzzle. 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  $k_{pub}$ . The problem poser generates a hint circuit  $\Gamma_H$  and a verification circuit  $\Gamma_V$ . The hint circuit takes as input  $q \in Q$  and outputs a signature for q. The verification circuit takes as input  $s \in S$  and  $s \in Q$  and checks whether  $s \in S$  is a valid signature for  $s \in Q$ . In the second phase the problem solver takes as input a transcript of message from the first round which consists of a single message  $s_{pub}$ . Additionally, it gain oracle access to  $s_V$  and  $s_H$ . It is clear that if the adversary asks a successful verification query  $s_V$ , then it also breaks the security of a public key signature scheme.

Public key signature schemes are types of puzzles that are dynamic but are not 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  $b \in \{0,1\}$ . In the reveal phase the sender opens the commitment by sending to the receiver (y,b') where  $y \in \{0,1\}^*$  and  $b' \in \{0,1\}$ . We require that after the commit phase it is hard for the receiver to correctly guess b. Additionally, in the reveal phase it should be hard for the sender to persuade the receiver that it was committed to the value  $\neg b$ .

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

**Definition 3.5 (Bit Commitment Protocol)** For a security parameter  $n \in \mathbb{N}$  a bit commitment protocol is defined by a pair  $(S_n, R_n)$  where  $S_n = (S_1, S_2)$  is a two phase probabilistic circuit, and  $R_n$  is a probabilistic circuit. The circuit  $S_1$ , used in the commit phase, takes as input a tuple  $(b, \rho_S)$  where  $b \in \{0,1\}$  is interpreted as a bit to which  $S_n$  commits and  $\rho_S \in \{0,1\}^n$  is the randomness used by the algorithm  $S_n$ . The receiver  $R_n$  takes only auxiliary input  $\rho_R \in \{0,1\}^n$  that is the randomness used by  $R_n$ . The protocol consists of two phases. In the commit phase circuits  $S_1$ ,  $S_n$  engage in the protocol execution. As the result  $S_1$  commits to  $S_n$  and generates a circuit  $S_n$  in the reveal phase the circuit  $S_n$  returns  $S_n$ . For fixed  $S_n$  and  $S_n$  we require the bit commit protocol to have the following properties:

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

$$\Pr_{\substack{\Gamma_{V} := \langle S_{1}, R \rangle_{R} \\ (b,y) := S_{2}(k_{pub}, \rho_{S})}} \left[ V(b,y) = 1 \right] \geq 1 - \varepsilon(n),$$

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

(Hiding)

**TODO:** Describe it using equations, define somehow the guess of R? Maybe as a last message in the first phase of communication

Probability over random coins of  $S_n$  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 polynomial size circuit  $S_n$  we have

For every posynomial size circuit 
$$S_n$$
 we have 
$$\Pr_{\substack{\Gamma_V := \langle S_1, R \rangle_R \\ (b,y) := S_2(k_{pub}, \rho_S)}} [\Gamma_V(0, y_0) = 1 \land V(1, y_1) = 1] \le \varepsilon(k),$$

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

The bit commitment protocols can be generalized as dynamic interactive weakly verifiable puzzles for a case where the number of hint queries amounts to zero and the number of the verification queries is at most one. The sender corresponds to a problem solver, and the receiver is a problem poser. Additionally, we require the problem solver to ask a verification query only on  $q := \neg b$  where b is a bit to which the problem solver is committed after the first phase. The first phase corresponds to the commit phase. In the reveal phase the problem poser tries to find a bitstring y such that  $V(\neg b, y) = 1$ .

#### 3.2.4 Automated Turing Tests

The goal of Automated Turing Tests is to distinguish humans from computers which is frequently used to prevent computer programs from accessing resources for humans. An example is CAPTCHA defined first in [VABHL03]. Loossly speaking, CAPTCHA is a test that human can solve with probability close to 1, but it is hard to write a computer program that has a success probability comparable to the one achieved by humans. An example of CAPTACHA is an image depicting a distorted text. Most humans guess the text which is displayed on the image correctly, but it might be hard to write a program for which it would also be easy. We note that the definition of hardness has not been particular well defined, and bases on opinions AI community opinions that distinguish between hard and easy AI problems [VABHL03].

CAPTCHAs based on guessing the distorted text are weakly verifiable puzzles. In the first round the problem poser and problem solver engage in interactive protocol, such that after the execution of the protocol the problem poser has a way to verify the solution. The problem poser in the second round takes as input a distorted image, and try to guess the text that was used to generated it. The standard CAPTCHAs are non-dynamic, as the problem poser does not gain access to the hint oracle and asks only a single verification query.

Our definition captures also the above type of problems, additionally it is also applicable in the broader context for a different AI problems.

As it is not know how good the possible algorithm can be to recognize CAPTCHA it is likely that the gap between human performance and a performance of computer programs may be small. Therefore, it is of interest to find a way to amplify this gap. It turns out that it is indeed possiThe first ble which for not dynamic puzzles was proved in [HS10]. The proof presented in Chapter 4 applies also to the dynamic context.

**TODO:** Give an optimization problem for gap amplification

### 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 not only to give a short overview of techniques used in these works, but aims also to provide some intuition and insight into the problem of hardness amplification of dynamic interactive weakly verifiable puzzles.

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

The notion of the weakly verifiable puzzles has been coined by R.Canetti, S.Halevi and M.Steiner in the paper 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 in comparison 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 included in [CHS04] about hardness amplification of weakly verifiable puzzles. 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 the paper 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  $p \in \{0,1\}^*$  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

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

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

**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  $1^k$  and outputs tuples  $p^{(n)} := (p_1, \ldots, p_n) \in \{0, 1\}^*$  and  $c^{(n)} := (c_1, \ldots, c_k) \in \{0, 1\}^*$ , where for each  $1 \le i \le n$  pair  $(p_i, c_i)$  is an independent instance of weakly verifiable puzzles defined by G with security parameter k and V. Finally, the verification algorithm  $V^{(n)}$  takes as inputp<sup>(n)</sup>,  $c^{(n)}$ , an answer  $a^{(n)}$ , and outputs  $b \in \{0, 1\}$  such that b = 1 if and only if for all  $1 \le i \le n$  we have  $V(a_i, c_i, p_i) = 1$ . We write  $P^{(n)}$  to denote the n-fold repetition of P.

We compare the above definition with Definition 3.1. First, we note that 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 circuit that does not take as input the bistring  $1^k$ , but just a bitstrong  $\rho$  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. Next, 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 care hard-coded in the circuit  $\Gamma_V$  from Definition 3.1. Hence, the algorithm V corresponds to  $\Gamma_V$ . Moreover, the n-fold repetition of weakly verifiable puzzles is solved successfully if and only if all n puzzles are solved successfully. In this Thesis we are interested in a more general situation where whether a monotone function  $g:\{0,1\}^n \to \{0,1\}$  is used to decide which coordinates have to be solved correctly. A precise definition is included in Section 4.1. Clearly, we can assume that q is such that all coordinates have to be solved successfully which matches the case considered in Definition 3.7.

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  $n: \mathbb{N} \to \mathbb{N}$ ,  $q \in \mathbb{N}$ , and  $\delta \to (0,1)$  be efficiently computable functions, and P a weakly verifiable puzzle as in Definition 3.6. Furthermore, let running times of the generation algorithm G for P be denote 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 success probability  $\delta(1-\frac{1}{q})$  and running time  $O\left(\frac{nq^3}{\delta^{2n-1}}(T+nT_G+nT_V)\right)$ .

The following algorithm is used in [CHS04] to prove Theorem ??. It uses a solver  $S^{(n)}$  for the *n*-fold repetition of P that succeeds with probability at least  $\delta^n$  to solve a single puzzle with probability at least  $\delta(1-\frac{1}{q})$ . Where q is a slackness parameter, as it is not possible to achieve a perfect hardness

amplification. We note that we take into account time needed for oracle calls to  $S^{(n)}$ , V, G in the analysis of running time of CHS-solver.

**TODO:** Explicit probabilities in G

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

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

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

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

**Algorithm:** OnlinePhase  $S^{(n)}, V, G$  (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 (n-i)-tuple of bitstrings prefix, a bitstring  $p \in \{0,1\}^*$ , parameters  $v, n, k, q, \delta$ .

```
\begin{array}{l} \textbf{Repeat} \ K_i := \frac{6q \ln(6q)}{\delta^{n-v+1}} \\ \quad ((p_{v+1}, \dots, p_n), (c_{v+1}, \dots, c_n)) := G^{(n-v)}(1^k) \\ \quad a^{(n)} := S^{(n)}(prefix, p, p_{v+1}, \dots, p_n) \\ \quad \textbf{if} \ \forall_{v+1 \leq i \leq n} V(p_i, c_i, a_i) = 1 \ \textbf{then return} \ a_v \\ \textbf{return} \ \bot \end{array}
```

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

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

**Input:** A (n-i)-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 do:  $(p^*,c^*):=G(1^k)$ 

```
ar{\mu}_i := EstimateResSuccProb^{G,V}(prefix \circ p^*, i, n, k, q, \delta)
if ar{\mu}_i \geq \delta^{n-i} then return p^*
return ot
```

```
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 (n-i)-tuple of puzzles *prefix*, parameters  $i, n, k, q, \delta$ .

```
\begin{aligned} successes &:= 0 \\ \mathbf{Repeat} \ M_i &:= \tfrac{84q^2}{\delta^{n-i}} \ln \left( \tfrac{18qn \cdot N_i}{\delta} \right) \text{ times} \\ & \left( (p_{i+1}, \cdots, p_n), (c_{i+1}, \cdots, c_n) \right) := G^{(n-i)}(1^k) \\ & a^{(n)} &:= A(prefix, p_{i+1}, \cdots, p_n) \\ & \text{if } \forall_{1 \leq i \leq n} : V(p_i, c_i, a_i) = 1 \text{ then } successes := successes + 1 \\ & \text{return } successes/M_i \end{aligned}
```

A detail proof of Theorem 3.8 is presented in [CHS04]. We limit ourselves to providing an intuition why the above algorithm is used in the proof of this theorem.

Let us consider the *n*-fold repetition of P, and a deterministic solver  $S^{(n)}$  for  $P^{(n)}$ . We define a matrix M as follows. The columns of M are labeled with all possible bitstrings  $p_1$ , whereas the rows corresponds to all possible tuples  $(p_2, \ldots, p_n)$ , where  $G^{(n)}$  is 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  $a^{(n)} := S(p^{(n)})$  and  $p^{(n)}$  is a tuple of bitstring inferred by a column and row of the cell. We make the following observation, with a simplifying assumption that the problem solver is deterministic.

**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  $1^k$  given that the cell is of the form that last (n-1) bits equal 1 is at least  $\delta$ .

We show how the algorithm *CHS-solver* uses Observation 3.9 to solve WVP with probability at least  $\delta(1-\frac{1}{q})$ . The algorithm starts with the first position and tries to fix a puzzle 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 this  $p^*$  on this position and repeat the whole procedure again in the consecutive iteration for the next position. If  $S^{(n)}$  fails to find such a bitstring  $p^*$ , then we may assume that there is no

column of M that contains  $\delta^{(n-1)}$  fraction of cell that are all of the form  $1^n$ . We use Observation 3.9 and hope 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 p (which denotes the input puzzle) on this position and note that all remaining puzzles are generated by CHS - solver. Thus, it is possible to efficiently verify whether they 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, it may also happen that no such column exists, but the algorithm fails to find a cell such that last (n-1) bits are one. 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, we intuitively see that the algorithm CHS-solver solves the puzzle successfully with probability at least  $\delta$  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 situation when a solver successfully solves the n-fold repetition  $P^{(n)}$  although it solved successfully only on a subset  $S \subset \{1, 2, \ldots, n\}$  of puzzles P.

It turns out that it is possible to use a similar technique of fixing puzzles P on consecutive positions of  $P^{(n)}$  to show the hardness amplification in the more general case.

#### 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 the *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 to  $x \in \{0,1\}^*$ . Furthermore, we consider a set  $\mathcal{Q}$  of indices and a probabilistic polynomialtime 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 function.

 $<sup>^{1}</sup>$ Actually, in Chapter 4 we define the k-wise repetition of puzzles which in case of WVP is equivalent to the n-fold repetition of puzzles.

A solver S takes as input x and can ask hint queries on  $q \in Q$  which are answered using  $H(\alpha, q)$  and verification queries (q, r) checking whether  $R(\alpha, q, r) = 1$ . We say that S succeeds if and only if it makes a verification query on (q, r) such that 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.

The definition stated in Section 3.1 generalizes for dynamic weakly verifiable puzzles. Instead of considering a distribution on pairs  $(x, \alpha)$  we use a problem poser. Furthermore, the problem poser may interact in the first phase with the problem solver. Thus, in our case the distribution of puzzles is defined by both the problem poser and the problem solver. The hint and verification relations corresponds to the hint and verification circuits generated by the problem poser.

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  $V^n(q, x_1, \ldots, x_n)$  evaluates to 1 if and only if for  $1 \le i \le n$  at least  $n - (1 - \gamma)\delta n$  is such that  $V(q, x_i, \alpha_i) = 1$  where  $0 \le \gamma, \delta \le 1$ . We write  $P^{(n)} := (D^{(n)}, H^{(n)}, V^{(n)})$  to denote the n-wise direct product of P := (D, H, V).

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

**Theorem 3.12** 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.

We define the  $success\ probability$  of a solver S for DWVP as

**TODO:** Define success probability

It worth seeing why the approach presented in the previous section can not be applied in the setting of dynamic weakly verifiable puzzles (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 that would make CHS-solver successfully solves the input puzzle. Thus, we can not dismiss the situation where the success probability of the solver  $S^{(n)}$  decrease in after some iterations.

The solution proposed in [DIJK09] is to partition the set Q into a set of attacking queries  $Q_{attack}$  and a set of advice queries  $Q_{adv}$ . The idea is to allow a solver for the n-wise direct product to ask a hint queries only on  $q \in Q_{adv}$ , and to halt the execution whenever a hint query is asked on  $q \in Q_{attack}$ .

It is possible, for a solver S that asks at most h hint queries and v verification queries, to find a function  $Q \to \{0, 1, \dots 2(h+v)\}$  such that the success probability of S with respect to  $Q_{attack}$  and  $Q_{adv}$  is multiplied by  $O(\frac{1}{h+v})$ . If h and v are not to big then the success probability of S can still be substantial.

Additionally, we define a canonical success probability with respect to a hash function as

**TODO:** Define success probability and canonical success probability

More formally, in [DIJK09] the following lemma is proved.

**Lemma 3.13** Let S be a solver for DWVP which success probability is at least  $\varepsilon$ , 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  $poly(h, v, \frac{1}{\varepsilon}, t)$  that outputs a function  $hash: Q \to \{0, 1, \ldots, 2(h+v)\}$  such that the canonical success probability of S with respect to hash is at least  $\frac{\varepsilon}{8(h+v)}$ .

A function hash can be found by using a natural sampling technique. We follow the same approach of domain partitioning 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  $H(\alpha, q)$ .

```
Algorithm: DWVP-solver^{S^{(n)},hash,H^{(n)}_{\alpha},R^{(n)}_{\alpha}}(x)

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\}^*.

Repeat at most O(\frac{h+v}{\varepsilon} \cdot \log(\frac{1}{\gamma\delta})) times

Choose uniformly at random position i \in \{1,\ldots,n\} 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 the fresh randomness. run S^{(n)}(x_1,\ldots,x_{i-1},x,x_{i+1},\ldots,x_n)
```

```
 \begin{tabular}{ll} \textbf{if } S^{(n)} \ asks a hint query on $q$ then \\ & \textbf{if } hash(q) \neq 0 \textbf{ 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 $(r_1,\ldots,r_{i-1},r,r_{i+1},r_n)$ \\ & \textbf{if } S^{(n)}$ asks a verification query $(q,r_1,\ldots,r_n)$ then \\ & \textbf{if } hash(q) = 0 \textbf{ then} \ answer the query with 0 \\ & Let $m:=|j:V(q,r_j)=1,j\neq i|$ \\ & \textbf{if } m\geq n-n(1-\gamma)\delta \textbf{ then} $$ \\ & \text{make a verification query } (q,r_i) \ and \ halt. \\ & \textbf{else } \ \text{with probability } \rho^{m-n(1-\gamma)\delta} \ \text{ask a verification query} $(q,r_i)$ and halt. \\ & \text{Halt the current run of } S^{(n)} \ \text{and go to the next iteration.} \\ & \textbf{return } \bot \\ \end{tabular}
```

The above algorithm substantially differ from the one used in [CHS04].

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. We 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 goto to the next iteration. This way me make sure that the algorithm never asks a hint query that could prevent a verification query to succeed.

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 is solved correctly then it is likely that also the input puzzle is solved correctly. We also have to discount  $\gamma \delta n$  to take into account that we require not all puzzles to be solved successfully. The detail calculations provided in [DIJK09] shows that this approach yields a demanded result. More details can be found in [DIJK09, IJK07].

**TODO:** Say why it makes sense to consider a threshold function

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

In Chapter 4 we consider a weakly verifiable puzzles that are interactive and dynamic. We use similar technique to partition domain Q into advice and hint queries as presented in [DIJK09]. Instead of the requirement to succeed only on the fraction of puzzles we consider an arbitrary, monotone function  $g:\{0,1\}^n \to \{0,1\}$  that determine 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 this position on random as in [DIJK09].

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

# 3.4 Limitations of Security Amplification

# Chapter 4

# Security amplification for dynamic weakly verifiable puzzles

In this chapter we show that it is possible to amplify security of dynamic weakly verifiable puzzles. In section 4.1 we state the theorem, which is next proved in three steps. First, in Section 4.1.3, we show how to use to partition the domain on which hint and verification queries are asked, next we give a prove of security amplification under the simplifying assumption that there is no collisions of hint and verification queries. Finally, in Section 4.1.5 we combine both former steps which yields the desirable result.

#### 4.1 Main theorem

We start by giving the definition of the k-wise direct product of weaky verifiable puzzles.

#### 4.1.1 The k-wise direct product of weakly verifiable puzzle

**Definition 4.1 (k-wise direct-product of DWVPs.)** 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 DWVP 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 for each  $1 \le i \le n : \pi_i \in \{0,1\}^n$ . Let  $(C_1, C_2)(\rho)$  be a solver for  $P_{kn}^{(g)}$  as in Definition 3.1. 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

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

and a hint circuit

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

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 (Security 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 problem 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 a solver circuit 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

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

then D is a two phase probabilistic circuit and satisfies almost surely

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

Additionally, D requires oracle access to g,  $P_n^{(1)}$ , C, hint and verification circuits 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 access to C.

#### 4.1.2 Intuition

**TODO:** add intuition for hashing

The idea of the algorithm Gen is to output a circuit D that solves the input puzzle often. We know that C has high success probability in solving the k-wise direct product of  $P^{(1)}$ . The algorithm Gen tries to find a puzzle such that when C runs with this puzzle fixed on the first position and disregards whether this puzzle is correctly solved then the assumptions of Theorem 4.2 are true for the (k-1)-wise direct product. If it was possible to find such a puzzle, then Gen could recurse and solve a smaller problem. In the optimistic case we can reach k=1, which means that we found a good circuit for solving a single puzzle by just fixing the initial puzzles of C.

Otherwise, when the first position is disregarded then the success probability of C is not substantially better. This is remarkable, as we know that C performs good for the k-wise direct product. It means that the first position is important, in the sense that C solves the puzzle on that position unusually often. Therefore, it is reasonable to construct the circuit D using C by placing the input puzzle of D on that position, and then finding remaining k-1 puzzles. The (k-1) remaining puzzles are generated by the circuit D, hence it is possible to check whether they are correctly solved by the circuit C. We know that circuit C has good success probability, and the puzzle on the first position is important. Therefore, if we are able to find a (k-1) puzzles such that the fact whether the k-wise direct product is correctly solved depends on whether the puzzle on the first position is correctly solved then we can assume that C is often correct on this first position.

There are some problems with this approach, first we have to ensure that we can make a decision when the algorithm *Gen* should recurse and when not

correctly with high probability. Then, we have to show that it is possible to find a puzzles such that C is often correct on the first position. Finally, we also have to be sure that we do not ask a hint query, on the final verification query to the oracle. To satisfy the last requirement we split the set Q.

#### 4.1.3 Domain partitioning

Let  $hash: Q \to \{0, 1, \dots, 2(h+v)-1\}$ , the idea is to partition Q such that the set of preimages of 0 for hash contains  $q \in Q$  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. We define the following experiment CanonicalSuccess in which the set 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  $q_i$  is 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: Q \to \{0, \dots, 2(h+v)-1\}.

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

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

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

run C_2^{\Gamma_V, \Gamma_H}(x, \rho) 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 Canonical Success  $P,C,hash(\pi,\rho)=1$ .

We show 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.

Lemma 4.3 (Success probability in solving DWVP with respect to a function hash.) For fixed  $P_n$  let C be a solver for  $P_n$  with success probability at least  $\gamma$ , asking at most h hint queries and v verification queries. Let  $\mathcal{H}$  be an efficient family of pairwise independent hash functions  $Q \to \{0, 1, \dots, 2(h + v) - 1\}$ . There exists a probabilistic algorithm FindHash that takes as input parameters  $\gamma$ , n, h, v, and has oracle access to C and  $P_n$ . Furthermore, FindHash runs in time 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 (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

$$\forall q_k \neq q_l \in Q : \Pr_{hash \leftarrow \mathcal{H}}[hash(q_k) = \alpha \mid hash(q_l) = \beta] = \Pr_{hash \leftarrow \mathcal{H}}[hash(q_k) = \alpha]$$
$$= \frac{1}{2(h+v)}. \tag{4.3}$$

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 function hash of the event  $CanonicalSuccess^{P,C,hash}(\pi^*, \rho^*) = 1$ . Let

 $(q_j, y_j)$  denote the first query such that  $\Gamma_V(q_j, y_j) = 1$ . We have

$$\Pr_{hash \leftarrow \mathcal{H}} \left[ Canonical Success^{P,C,hash}(\pi^*, \rho^*) = 1 \right] \\
= \Pr_{hash \leftarrow \mathcal{H}} [hash(q_j) = 0 \land (\forall i < j : hash(q_i) \neq 0)] \\
= \Pr_{hash \leftarrow \mathcal{H}} [\forall i < j : hash(q_i) \neq 0 \mid hash(q_j) = 0] \Pr_{hash \leftarrow \mathcal{H}} [hash(q_j) = 0] \\
\stackrel{(4.3)}{=} \frac{1}{2(h+v)} \left( 1 - \Pr_{hash \leftarrow \mathcal{H}} [\exists i < j : hash(q_i) = 0 \mid hash(q_j) = 0] \right) \\
\stackrel{(*)}{\geq} \frac{1}{2(h+v)} \left( 1 - \sum_{i < j} \Pr_{hash \leftarrow \mathcal{H}} [hash(q_i) = 0 \mid hash(q_j) = 0] \right) \\
\stackrel{(4.3)}{=} \frac{1}{2(h+v)} \left( 1 - \sum_{i < j} \Pr_{hash \leftarrow \mathcal{H}} [hash(q_i) = 0] \right) \\
\stackrel{(4.3)}{=} \frac{1}{2(h+v)} \left( 1 - \sum_{i < j} \Pr_{hash \leftarrow \mathcal{H}} [hash(q_i) = 0] \right) \\
\stackrel{(4.3)}{=} \frac{1}{2(h+v)} \left( 1 - \sum_{i < j} \Pr_{hash \leftarrow \mathcal{H}} [hash(q_i) = 0] \right) \\
\stackrel{(4.3)}{\geq} \frac{1}{4(h+v)}, \tag{4.4}$$

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

$$\Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \right] \\
= \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \in \mathcal{P}_{Success} \right] \Pr_{\pi, \rho} [(\pi, \rho) \in \mathcal{P}_{Success}] \\
+ \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \notin \mathcal{P}_{Success} \right] \Pr_{\pi, \rho} [(\pi, \rho) \notin \mathcal{P}_{Success}] \\
= \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \in \mathcal{P}_{Success} \right] \Pr_{\pi, \rho} [(\pi, \rho) \in \mathcal{P}_{Success}] \\
\geq \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho}} \left[ CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \mid (\pi, \rho) \in \mathcal{P}_{Success} \right] \cdot \gamma \\
= \Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi, \rho \in \mathcal{P}_{Success}}} \left[ \Pr_{\substack{hash \leftarrow \mathcal{H}}} [CanonicalSuccess^{P,C,hash}(\pi, \rho) = 1 \right] \cdot \gamma \\
\geq \frac{\gamma}{4(h + v)}. \tag{4.5}$$

# **Algorithm** 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  $hash: Q \rightarrow \{0, 1, \dots, 2(h+v) - 1\}.$ 

```
\begin{array}{l} \mathbf{for}\ i := 1\ \mathbf{to}\ 32n(h+v)^2/\gamma^2\ \mathbf{do:}\\ hash \leftarrow \mathcal{H}\\ count := 0\\ \mathbf{for}\ j := 1\ \mathbf{to}\ 32n(h+v)^2/\gamma^2\ \mathbf{do:}\\ \pi \leftarrow \{0,1\}^n\\ \rho \leftarrow \{0,1\}^*\\ \mathbf{if}\ CanonicalSuccess}^{P,C,hash}(\pi,\rho) = 1\ \mathbf{then}\\ count := count + 1\\ \mathbf{if}\ count \geq \frac{\gamma}{12(h+v)} \frac{32(h+v)^2}{\gamma^2} n\ \mathbf{then}\\ \mathbf{return}\ hash \\ \mathbf{return}\ \bot \end{array}
```

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

$$\Pr_{\pi,\rho} \left[ Canonical Success^{P,C,hash}(\pi,\rho) = 1 \right] \ge \frac{\gamma}{8(h+v)}, \tag{4.6}$$

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

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

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. We define independent, 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 now turn to the case when  $hash \in \mathcal{H}_{Bad}$  and show that it is unlikely that hash is returned by FindHash. From (4.7) it follows that  $\mathbb{E}_{\pi,\rho}[X_i] \leq \frac{\gamma}{16(h+v)}$ . Therefore, for any fixed  $hash \in \mathcal{H}_{Bad}$  using the Chernoff bound we get

$$\Pr_{\pi,\rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \ge \frac{\gamma}{12(h+v)} \right] \le \Pr_{\pi,\rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \ge \left(1 + \frac{1}{3}\right) \mathbb{E}[X_i] \right]$$

$$\le e^{-\frac{\gamma}{16(h+v)} N/27} \le e^{-\frac{2}{27} \frac{(h+v)}{\gamma} n} \stackrel{(*)}{\le} e^{-\frac{2}{27} n},$$

where in (\*) we used the trivial facts that  $h+v \ge 1$  and  $\gamma \le 1$ . The probability that  $hash \in \mathcal{H}_{Good}$ , when picked, is not returned amounts

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

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

$$\Pr_{hash \leftarrow \mathcal{H}}[hash \in \mathcal{H}_{Good}] < \frac{\gamma}{8(h+v)}.$$
 (4.8)

From this it follows that we can bound probability of canonical success as follows

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

which contradicts (4.5). Therefore, we conclude that the probability of choosing a  $hash \in \mathcal{H}_{Good}$  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 be the number of iterations of the outer loop of FindHash and  $Y_i$  be a random variable for the event that in the i-th iteration of the outer loop  $hash \notin \mathcal{H}_{Good}$  is picked. We use  $\Pr_{hash \leftarrow \mathcal{H}}[hash \in \mathcal{H}_{Good}] \geq \frac{\gamma}{8(h+v)}$  and  $K \leq \frac{32(h+v)^2}{\gamma^2}n$ , and conclude

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

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

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

#### 4.1.4 Amplification proof for partitioned domain

Let  $C = (C_1, C_2)$  be a solver circuit for a dynamic 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  $C_2$  ask hint or verification queries, then 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 : Q \to \{0, 1, \dots, 2(h+v) - 1\}.
Input: Bitstrings x \in \{0, 1\}^*, \rho \in \{0, 1\}^*.
Output: A pair (q, y).
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)
                  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  $\bot$ .

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

$$\Pr_{\pi,\rho}[Canonical Success^{P,C,hash}(\pi,\rho)=1] \leq \Pr_{\pi,\rho}[\Gamma_V(\widetilde{C}_2^{\Gamma_H,C_2,hash}(x,\rho))=1] \\ \underset{(\Gamma_V,\Gamma_H):=\langle P(\pi),C_1(\rho)\rangle_{P}}{\underset{r:=\langle P(\pi),C_1(\rho)\rangle$$

**Proof.** If for fixed  $\pi$ ,  $\rho$ , and hash the circuit C succeeds canonically, then for the same  $\pi$ ,  $\rho$ , and hash also  $\widetilde{C}$  succeeds. Using this observation, we conclude that

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

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

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

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

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

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

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

Lemma 4.5 (Security amplification for dynamic weakly verifiable puzzles with respect to hash.) Let  $g:\{0,1\}^k \to \{0,1\}$  be a monotone function and  $P_n^{(1)}$  a fixed problem poser as in Definition 3.1 and  $\widetilde{C}:=(C_1,\widetilde{C}_2)$  a circuit with oracle access to a function hash:  $Q \to \{0,1,\ldots,2(h+v-1)\}$  and a solver circuit  $C:=(C_1,C_2)$  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

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

then D satisfies almost surely

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

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 define additional algorithms. First, 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 EstimateFunctionProbability.

**Algorithm** EstimateFunctionProbability<sup>g</sup> $(b, k, \varepsilon, \delta, n)$ 

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

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

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

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

**Lemma 4.6** The algorithm EstimateFunctionProbability<sup>g</sup> $(b, k, \varepsilon, \delta, n)$  outputs an estimate  $\widetilde{g}_b$  such that  $|\widetilde{g}_b - \Pr_{u \leftarrow \mu_k^k}[g(b, u_2, \dots, u_k) = 1]| \leq \frac{\varepsilon}{8k}$  almost surely.

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

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

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

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

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.

To make the notation easier 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<sup>P(1)</sup>,\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: Q \to \{0, 1, \dots, 2(h+v)-1\}.

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

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

for i := 1 to k do: //simulate k rounds of interaction (\Gamma^i_{i-1}, \Gamma^i_{i-1}) := \langle P^{(1)}(\pi_i), C_1(\rho_i) \rangle^i_{i-1}.
```

```
(\Gamma_V^i, \Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{P^{(1)}}^i
x_i := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{trans}^i
x := (x_1, \dots, x_k)
\Gamma_H^{(k)} := (\Gamma_H^1, \dots, \Gamma_H^k)
(q, y_1, \dots, y_k) := \widetilde{C}_2^{\Gamma_H^{(k)}, hash}(x, \rho)
if (q, y_1, \dots, y_k) = \bot then
\mathbf{return} \ (0, \dots, 0)
(c_1, \dots, c_k) := (\Gamma_V^1(q, y_1), \dots, \Gamma_V^k(q, y_k))
\mathbf{return} \ (c_1, \dots, c_k)
```

All puzzles used by EvalutePuzzles are generated internally thus the algorithm has access to hint circuit, and can answer itself all queries of  $\widetilde{C}_2$ .

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

$$S_{\pi^*,b} = \Pr_{\pi^{(k)},\rho} [g(b, c_2, \dots, c_k) = 1 \mid \pi_1 = \pi^*] - \Pr_{u \leftarrow \mu_{\delta}^k} [g(b, u_2, \dots, u_k) = 1],$$
(4.9)

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<sup>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:Q \to \{0,1,\dots,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}.</sup>
```

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

**Lemma 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  $\frac{1}{N} \sum_{i=1}^{N} \tilde{s}_{\pi^*,b}^i$  differs from  $\mathbb{E}[g(b,c_2,\ldots,c_k)]$  by at most  $\frac{\varepsilon}{8k}$  almost surely. Together, with Lemma 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  $\widetilde{C}_{2}^{/\Gamma_{H}^{(k-1)},\widetilde{C},hash}(x^{(k-1)},\rho)$ 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  $hash: Q \to \{0, 1, \dots, 2(h+v)-1\}$ . **Input:** A transcript of k-1 rounds of interaction

A transcript of k-1 rounds of interaction  $x^{(k-1)} := (x_2, \dots, x_k) \in \{0, 1\}^*$ , a bitstring  $\rho \in \{0, 1\}^*$ 

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

Simulate  $\langle P^{(1)}(\pi^*), C_1(\rho) \rangle^1$   $(\Gamma_H^*, \Gamma_V^*) := \langle P^{(1)}(\pi^*), C_1(\rho) \rangle_{P^{(1)}}^1$  $x^* := \langle P^{(1)}(\pi^*), C_1(\rho) \rangle_{trans}^1$ 

$$\begin{split} \Gamma_H^{(k)} &:= (\Gamma_H^*, \Gamma_H^2, \dots, \Gamma_H^k) \\ x^{(k)} &:= (x^*, x_2, \dots, x_k) \\ (q, y_1, \dots, y_k) &:= \tilde{C}_2^{\Gamma_H^{(k)}, hash}(x^{(k)}, \rho) \\ &\mathbf{return} \ (q, y_2, \dots, y_k) \end{split}$$

We are ready to define the solver circuit  $D = (D_1, D_2)$  for  $P^{(1)}$  and the algorithm 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: Q \to \{0, 1, \dots, 2(h+v)-1\}, g: \{0, 1\}^k \to \{0, 1\},
               a hint circuit \Gamma_H for P^{(1)}.
Input: A communication transcript x^* \in \{0,1\}^*, a bitstring r := (\rho, \sigma)
               where \rho \in \{0, 1\}^* and \sigma \in \{0, 1\}^*
Output: A pair (q, y^*).
for at most \frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon}) iterations do:
        (\pi_2, \ldots, \pi_k) \leftarrow \text{read next } (k-1) \cdot n \text{ bits from } \sigma
        Use x^* to simulate the first round of interiaction 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_{D^{(1)}}^i
                      x_i := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{trans}^i
       \Gamma_H^{(k)}(q) := (\Gamma_H(q), \Gamma_H^2(q), \dots, \Gamma_H^k(q))
       (q, y^*, y_2, \dots, y_k) := \widetilde{C}_2^{\Gamma_H^{(k)}, hash}((x^*, x_2, \dots, x_k), \rho)
(c_2, \dots, c_k) := (\Gamma_V^2(q, y_2), \dots, \Gamma_V^k(q, y_k))
        if g(1, c_2, ..., c_k) = 1 and g(0, c_2, ..., c_k) = 0 then
               return (q, y^*)
return \perp
```

```
Algorithm Gen<sup>P^{(1)},\widetilde{C},g,hash (\varepsilon, \delta, n, k)</sup>
Oracle: A poser P^{(1)}, a solver circuit \widetilde{C} for P^{(g)}, functions g:\{0,1\}^k \to \mathbb{C}
\{0,1\},\
                   hash: Q \to \{0, 1, \dots, 2(h+v) - 1\}.
Input: Parameters \varepsilon, \delta, n, k.
Output: A circuit D.
for i := 1 to \frac{6k}{\varepsilon}n do:
         \pi^* \stackrel{\$}{\leftarrow} \{0,1\}^n
         \begin{split} \widetilde{S}_{\pi^*,0} &:= \mathsf{EstimateSurplus}^{P^{(1)},\widetilde{C},g,hash}(\pi^*,0,k,\varepsilon,\delta,n) \\ \widetilde{S}_{\pi^*,1} &:= \mathsf{EstimateSurplus}^{P^{(1)},\widetilde{C},g,hash}(\pi^*,1,k,\varepsilon,\delta,n) \end{split}
         if \exists b \in \{0,1\} : \widetilde{S}_{\pi^*,b} \geq (1-\frac{3}{4k})\varepsilon then
                  Let C_1' have oracle access to \widetilde{C}, and have hard-coded \pi^*
                   Let \widetilde{C}'_2 have oracle access to \widetilde{C}, and have hard-coded \pi^*.
                   \widetilde{C}' := (C_1', \widetilde{C}_2')
                  g'(b_2, \dots, b_k) := g(b, b_2, \dots, b_k)
return 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}
```

**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. Either Gen in one of the iterations finds an estimate  $\widetilde{S}_{\pi,b} \geq (1 - \frac{3}{4k}\varepsilon)$  or it fails and returns the circuit D.

In the former case we define a new monotone function  $g'(b_2,\ldots,b_k):=g(b,b_2,\ldots,b_k)$  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 with high probability satisfies

$$\Pr_{\pi,\rho}[\Gamma_{V}(D_{2}^{P^{(1)},\widetilde{C},hash,g,\Gamma_{H}}(x,\rho)) = 1] \geq \delta + \left(1 - \frac{1}{k}\right)\frac{\varepsilon}{6(k-1)} = \delta + \frac{\varepsilon}{6k}. \quad (4.10)$$

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

The only point remaining concerns the behavior of Gen when none of the estimates is greater than  $(1 - \frac{3}{4k})\varepsilon$ . Assume that...

# TODO: Give more intuition (and the correct one with references to the equations)

Intuitively this means that  $\widetilde{C}$  does not succeed on the remaining k-1 puzzles with much higher probability than an algorithm that correctly solves each puzzle with probability  $\delta$ . However, 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 the notation used in the code excerpt of the circuit  $D_2$ . We consider a single iteration of the outer loop of  $D_2$ , in which values  $\pi_2, \ldots, \pi_k$  are fixed. Additionally, we write  $\pi_1$  to denote the randomness of the problem poser and 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  $D_1(r)$ . Finally, we introduce the additional notation  $\mathcal{G}_b := \{(b_1, b_2, \ldots, b_k) : g(b, b_2, \ldots, b_k) = 1\}$  and  $c = (c_1, c_2, \ldots, c_k)$ . Using the new notation the following holds

$$\Pr_{u \leftarrow \mu_{\delta}^{k}} [u \in \mathcal{G}_{b}] = \Pr_{u \leftarrow \mu_{\delta}^{k}} [g(b, u_{2}, \dots, u_{k}) = 1] 
\Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{b}] = \Pr_{\pi^{(k)}, \rho} [g(b, c_{2}, \dots, c_{k}) = 1].$$
(4.11)

We fix the randomness  $\pi_1$  of the problem poser  $P^{(1)}$  to  $\pi^*$  and use (4.9), (4.11) to obtain

$$\Pr_{u \leftarrow \mu_{\delta}^{k}} [u \in \mathcal{G}_{1}] - \Pr_{u \leftarrow \mu_{\delta}^{k}} [u \in \mathcal{G}_{0}] 
= \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{1} \mid \pi_{1} = \pi^{*}] - \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - (S_{\pi^{*}, 1} - S_{\pi^{*}, 0}) \quad (4.12)$$

We know that the function g is monotone, hence it holds  $\mathcal{G}_0 \subseteq \mathcal{G}_1$ , and we write (4.12) as

$$\Pr_{u \leftarrow \mu_{\delta}^{k}} [u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}] = \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - (S_{\pi^{*}, 1} - S_{\pi^{*}, 0}).$$
 (4.13)

Still having  $\pi_1 = \pi^*$  fixed we multiply both sides of (4.13) by

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

which yields

$$\Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{p(1)} \\
= \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \Pr_{\pi^{(k)}, \rho} \left[ c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*} \right] \frac{1}{\Pr_{u \leftarrow \mu_{\delta}^{k}} \left[ u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right]} \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{p(1)} \\
- \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \left( S_{\pi^{*}, 1} - S_{\pi^{*}, 0} \right) \frac{1}{\Pr_{u \leftarrow \mu_{\delta}^{k}} \left[ u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \right]} \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{p(1)} \\
(4.14)$$

We analyze the first summand of (4.14). First, we have

$$\Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \right] \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
= \Pr_{r} \left[ \Gamma_{V}(D_{2}(x^{*}, r)) = 1 \middle| D_{2}(x^{*}, r) \neq \bot \right] \Pr_{r} \left[ D_{2}(x^{*}, r) \neq \bot \right] \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \qquad x^{*} = \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \\
\stackrel{(*)}{=} \Pr_{\pi^{(k)}, \rho} \left[ c_{1} = 1 \middle| c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}, \pi_{1} = \pi^{*} \right] \Pr_{r} \left[ D_{2}(x^{*}, r) \neq \bot \right], \quad (4.15) \\
x^{*} = \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans}$$

where in (\*) we use the observation that the event  $D_2(x^*,r) \neq \bot$  happens if and only if the circuit  $D_2(x^*,r)$  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 (4.15) to the numerator of the first summand of (4.14) yields

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

$$x^* = \langle P^{(1)}(\pi^*), D_1(r) \rangle_{trans}$$
(4.16)

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

$$\Pr_{\pi^{(k)},\rho}[c_1 = 1 \mid c \in \mathcal{G}_1 \setminus \mathcal{G}_0, \pi_1 = \pi^*] \Pr_{\pi^{(k)},\rho}[c \in \mathcal{G}_1 \setminus \mathcal{G}_0 \mid \pi_1 = \pi^*] \le \frac{\varepsilon}{6k}. \quad (4.17)$$

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

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

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

$$(4.18)$$

We conclude that in both cases by (4.17) and (4.18) we have

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

$$\geq \Pr_{\pi^{(k)}, \rho} [c_{1} = 1 \mid c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}, \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k}$$

$$= \Pr_{\pi^{(k)}, \rho} [c_{1} = 1 \land c \in \mathcal{G}_{1} \setminus \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k}$$

$$= \Pr_{\pi^{(k)}, \rho} [g(c) = 1 \mid \pi_{1} = \pi^{*}] - \Pr_{\pi^{(k)}, \rho} [c \in \mathcal{G}_{0} \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k}$$

$$\stackrel{(4.9)}{=} \Pr_{\pi^{(k)}, \rho} [g(c) = 1 \mid \pi_{1} = \pi^{*}] - \Pr_{u \leftarrow \mu_{\delta}^{(k)}} [u \in \mathcal{G}_{0}] - S_{\pi^{*}, 0} - \frac{\varepsilon}{6k}. \quad (4.19)$$

We insert (4.19) to (4.14) and calculate the expected value of over  $\pi^*$  which yields

$$\Pr_{\substack{x := \langle P^{(1)}(\pi), D_{1}(r) \rangle_{trans} \\ (\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi), D_{1}(r) \rangle_{P^{(1)}}}} = 1] \ge \mathbb{E}_{\pi^{*}} \left[ \frac{\Pr_{\pi^{(k)}, \rho}[g(c) = 1 | \pi_{1} = \pi^{*}] - \Pr_{u \leftarrow \mu_{\delta}^{(k)}}[u \in \mathcal{G}_{0}] - \frac{\varepsilon}{6k}}{\Pr_{u \leftarrow \mu_{\delta}^{(k)}}[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]} \right] \\
- \mathbb{E}_{\pi^{*}} \left[ \left( \Pr_{r}[\Gamma_{V}(D_{2}(x^{*}, r)) = 1](S_{\pi^{*}, 1} - S_{\pi^{*}, 0}) + S_{\pi^{*}, 0} \right) \frac{1}{\Pr_{u \leftarrow \mu_{\delta}^{k}}[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]} \right] \cdot \\
x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{trans} \\
(\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(r) \rangle_{P^{(1)}} \tag{4.20}$$

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

$$\Pr_{\pi} \left[ \left( S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon \right) \wedge \left( S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon \right) \right] < 1 - \frac{\varepsilon}{6k}, \tag{4.21}$$

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

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

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

$$\Pr_{\pi,\rho}\left[\left(S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon\right) \land \left(S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon\right)\right] \ge 1 - \frac{\varepsilon}{6k}.\tag{4.22}$$

Let us define a set

$$W = \left\{ \pi : \left( S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon \right) \land \left( S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon \right) \right\}, \tag{4.23}$$

and use  $\overline{\mathcal{W}}$  to denote the complement of  $\mathcal{W}$ . We bound the numerator of the second summand in (4.20)

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

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

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

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

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

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

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

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

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

We observe that

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

$$= \Pr[u \in \mathcal{G}_{0}] + \delta \Pr[u \in \mathcal{G}_{1} \setminus \mathcal{G}_{0}]. \tag{4.25}$$

Finally, we insert (4.24) into (4.20) which yields

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

From the assumptions of Lemma 4.5 it follows that

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

thus we get

$$\Pr_{\substack{x := \langle P^{(1)}(\pi^*), D_1(\rho) \rangle_{\text{trans}} \\ (\Gamma_V, \Gamma_H) := \langle P^{(1)}(\pi^*), D_1(\rho) \rangle_{P^{(1)}}}} \ge \frac{\Pr_{u \leftarrow \mu_{\delta}^k}[g(u) = 1] + \varepsilon - \Pr_{u \leftarrow \mu_{\delta}^k}[u \in \mathcal{G}_0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{u \leftarrow \mu_{\delta}^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0]} \\
\ge \frac{\Pr_{u \leftarrow \mu_{\delta}^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0]}{\Pr_{u \leftarrow \mu_{\delta}^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0] - (1 - \frac{1}{6k})\varepsilon} \\
\ge \frac{\varepsilon + \delta \Pr_{u \leftarrow \mu_{\delta}^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{u \leftarrow \mu_{\delta}^k}[u \in \mathcal{G}_1 \setminus \mathcal{G}_0]} \ge \delta + \frac{\varepsilon}{6k}$$
(4.26)

Clearly, the running time of Gen is  $poly(k, \frac{1}{\epsilon}, n)$ .

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

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

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

**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  $\widetilde{D} = (D_1, \widetilde{D}_2)$ .

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

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

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

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

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

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

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

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

$$\Pr_{\pi,\rho} \left[ \Gamma_V \left( D_2^{P^{(1)},\tilde{C},hash,g,\Gamma_H}(x,\rho) \right) = 1 \right] \ge \left( \delta + \frac{\varepsilon}{6k} \right). \tag{4.27}$$

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

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

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

The running time of FindHash is  $poly(h, v, \frac{1}{\varepsilon}, n)$  with oracle calls and of Gen  $poly(k, \frac{1}{\varepsilon}, n)$  with oracle access. Thus, the overall running time of  $\widetilde{Gen}$  is  $poly(k, \frac{1}{\varepsilon}, h, v, n, t)$  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  $Size(\widetilde{D}) \leq Size(C) \cdot \frac{6k}{\varepsilon}$ . 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  $1 \le i \le n$ . we have the following inequalities for  $0 \le \delta \le 1$  and  $\mathbb{E}[X] = \sum_{i=1}^n p_i$ :

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

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

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

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