We write  $u \leftarrow \mu_{\delta}^k$  to denote a tuple u of length k which each element is an independent Bernoullidistributed random variable with the parameter  $\delta$ . The protocol execution between probabilistic algorithms A and B is denoted by  $\langle A, B \rangle_A$ . Additionally, the output of A in such a protocol execution is denoted by  $\langle A, B \rangle_A$ , and a transcript of communication by  $\langle A, B \rangle_{\text{trans}}$ .

**Definition 1.1 (Dynamic weakly verifiable puzzle.)** A dynamic weakly verifiable puzzle (DWVP) is defined by a probabilistic algorithm P called a problem poser. A problem solver  $S := (S_1, S_2)$  for P is a probabilistic two phase algorithm. We write  $P(\pi)$  to denote the execution of P with the randomness fixed to  $\pi \in \{0,1\}^n$ , and  $(S_1, S_2)(\rho)$  to denote the execution of both  $S_1$  and  $S_2$  with the randomness fixed to  $\rho \in \{0,1\}^*$ .

The poser  $P(\pi)$  and the solver  $S_1(\rho)$  interact. As the result of the interaction  $P(\pi)$  outputs a verification circuit  $\Gamma_V$  and a hint circuit  $\Gamma_H$ . The algorithm  $S_1(\rho)$  produces no output. The circuit  $\Gamma_V$  takes as input  $q \in Q$ , an answer  $y \in \{0,1\}^*$ , and outputs a bit. 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  $S_2$  takes as input  $x := \langle P(\pi), S_1(\rho) \rangle_{trans}$ , and has oracle access to  $\Gamma_V$  and  $\Gamma_H$ . The execution of  $S_2$  with the input x and the randomness fixed to  $\rho$  is denoted by  $S_2(x,\rho)$ . The queries of  $S_2$  to  $\Gamma_V$  are called verification queries, and to  $\Gamma_H$  hint queries. The algorithm  $S_2$  can ask at most h hint queries, v verification queries, and 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.

**Definition 1.2 (k-wise direct-product of DWVPs.)** Let  $g:\{0,1\}^k \to \{0,1\}$  be a monotone function and  $P^{(1)}$  a problem poser as in Definition 1.1. The k-wise direct product of  $P^{(1)}$  is a DWVP defined by a probabilistic algorithm  $P^{(g)}$ . We write  $P^{(g)}(\pi^{(k)})$  to denote the execution of  $P^{(g)}$  with the randomness fixed to  $\pi^{(k)} := (\pi_1, \ldots, \pi_k)$ . Let  $(S_1, S_2)(\rho)$  be a solver for  $P^{(g)}$  as in Definition 1.1. The algorithm  $S_1(\rho)$  sequentially interacts in k rounds with  $P^{(g)}(\pi^{(k)})$ . In the i-th round  $S_1(\rho)$  interacts with  $P^{(1)}(\pi_i)$ , and as the result  $P^{(g)}(\pi^{(k)})$  generates circuits  $\Gamma^i_V, \Gamma^i_H$ . Finally, after k rounds  $P^{(g)}$  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

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

Let C be a random circuit that corresponds to a solver S as in Definition 1.1. Similarly as for two phase algorithm, we write  $C(\pi) := (C_1, C_2)(\pi)$  to denote that the randomness used by C is fixed to  $\pi$ , and  $C(\pi)$  in the first phase uses  $C_1(\pi)$  and in the second phase  $C_2(\pi)$ . A verification query (q, y) of C for which a hint query on this q has been asked before can not be a successfully verification query. Therefore, without loss of generality, we make an assumption that C does not ask verification queries on q, for which a hint query has been asked before.

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

**Oracle:** A problem poser P, a solver circuit  $C^{(\cdot,\cdot)}$ .

Input: Bitstrings  $\pi$ ,  $\rho$ . Output: A bit  $b \in \{0, 1\}$ .

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

```
Run C_2^{\Gamma_V,\Gamma_H}(x,\rho)

if C_2^{\Gamma_V,\Gamma_H} asks a verification query (q,y) such that \Gamma_V(q,y)=1 then return 1

return 0
```

The success probability of C in solving a puzzle defined by P in the experiment Success is

$$\Pr_{\pi,\rho}[Success^{P,C^{(\cdot,\cdot)}}(\pi,\rho)=1]. \tag{0.0.1}$$

Theorem 1.3 (Security amplification for a dynamic weakly verifiable puzzle.) Let  $P^{(1)}$  be a fixed problem poser as in Definition 1.1, and  $P^{(g)}$  be a poser for the k-wise direct product of  $P^{(1)}$ . There exists a probabilistic algorithm  $Gen(C, g, \varepsilon, \delta, n, v, h)$  which takes as input: a solver circuit C for  $P^{(g)}$ , a monotone function  $g: \{0,1\}^k \to \{0,1\}$ , parameters  $\varepsilon, \delta, n$ , the number of verification queries v, and hint queries h asked by C, and outputs a random circuit D such that the following holds:

If C is such that

$$\Pr_{\pi^{(k)},\rho} \left[ Success^{P^{(g)},C}(\pi^{(k)},\rho) = 1 \right] \ge 8(h+v) \left( \Pr_{u \leftarrow \mu_{\delta}^k} \left[ g(u) = 1 \right] + \varepsilon \right)$$

then D satisfies almost surely

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

Additionally, Gen and D require oracle access to g,  $P^{(1)}$ , C. Furthermore, D requires also oracle access to  $\Gamma_V$   $\Gamma_H$ , and asks at most  $\frac{6k}{\varepsilon}\log\left(\frac{6k}{\varepsilon}\right)h$  hint queries and one verification query. Finally,  $Size(D) \leq Size(C) \cdot \frac{6k}{\varepsilon}$  and  $Time(Gen) = poly(k, \frac{1}{\varepsilon}, n, v, h)$ .

The Theorem 1.3 implies that if there is no good solver for a puzzle defined by  $P^{(1)}$ , then a good solver for a k-wise direct product of  $P^{(1)}$  does not exist.

The idea of the algorithm Gen is to output a circuit D that solves the input puzzle often. We know that C has good success probability for a k-wise 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 1.3 are true for a k-1-wise direct product. If it is 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 k-wise 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. These 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 Q.

Let  $hash: Q \to \{0, 1, \ldots, 2(h+v)-1\}$ , then a set  $P_{hash} \subseteq Q$ , defined with respect to hash, is the set of preimages of 0 for hash. The idea is that  $P_{hash}$  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  $q \in P_{hash}$ . Therefore, if C makes a verification query (q, y) such that  $q \in P_{hash}$ , then we know that no hint query is ever asked on this q. In the experiment CanonicalSuccess a circuit C succeeds if and only if it asks a successful verification query (q, y) such that  $q \in P_{hash}$ , and no hint query is asked on  $q \in P_{hash}$ .

In the following experiment Canonical Success 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.

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

Oracle: A problem poser P, a solver circuit C^{(\cdot,\cdot)}.

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

Input: Bitstrings: \pi, \rho.

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_{\text{trans}}

Run C^{\Gamma_V, \Gamma_H}(x, \rho)

(q_j, y_j) be the first verification query such that C_2^{\Gamma_V, \Gamma_H}(q_j, y_j) = 1, or an arbitrary verification query if C_2 does not succeed.

If (\forall i < j : q_i \notin P_{hash}) and q_j \in P_{hash} and \Gamma_V(q_j, y_j) = 1 then return 1 else return 0
```

Similarly as for the experiment Success, we define the success probability of a solver C for P with respect to a function hash in the experiment CanonicalSuccess as

$$\Pr_{\pi,\rho}[CanonicalSuccess^{P,C^{(\cdot,\cdot)},hash}(\pi,\rho)=1]. \tag{0.0.2}$$

For fixed hash and  $P^{(g)}$  a canonical success of C for  $\pi^{(k)}$ ,  $\rho$  is a situation when  $CanonicalSuccess^{P^{(g)},C^{(\cdot,\cdot)},hash}(\pi^{(k)},\rho)=1.$ 

We show that if for a fixed  $P^{(1)}$  a solver circuit C often succeeds in the experiment Success for  $P^{(g)}$ , then it also often successful in the experiment CanonicalSuccess for  $P^{(g)}$ .

Lemma 1.4 (Success probability in solving a k-wise direct product of  $P^{(1)}$  with respect to a function hash.) For fixed  $P^{(g)}$  let C succeed in the experiment Success for  $P^{(g)}$  with probability  $\gamma$ , asking at most h hint queries and v verification queries. There exists a probabilistic algorithm **FindHash** that takes as input: parameters  $\gamma$ , n, the number of verification queries v and hint queries h, and has oracle access to C and  $P^{(g)}$ . Furthermore, **FindHash** runs in time  $O((h+v)^4/\gamma^4)$ , and with high probability outputs a function hash  $\in \mathcal{H}$  such that success probability of C with respect to  $P_{hash}$  in the experiment CanonicalSuccess is at least  $\frac{\gamma}{8(h+v)}$ .

**Proof.** We fix  $P^{(g)}$  and a solver C for  $P^{(h)}$  in the whole proof of Lemma 1.4. Let  $\mathcal{H}$  be a family of pairwise independent hash functions  $Q \to \{0, 1, \dots, 2(h+v)-1\}$ . For all  $i \neq j \in \{1, \dots, (h+v)\}$ and  $k, l \in \{0, 1, \dots, 2(h+v) - 1\}$  by pairwise independence property of  $\mathcal{H}$ , we have

$$\forall q_i, q_j \in Q : \Pr_{hash \leftarrow \mathcal{H}}[hash(q_i) = k \mid hash(q_j) = l] = \Pr_{hash \leftarrow \mathcal{H}}[hash(q_i) = k] = \frac{1}{2(h+v)}. \quad (0.0.3)$$

Let  $\pi^{(k)}$ ,  $\rho$  be fixed. We consider the experiment Canonical Success for  $P^{(g)}$  and C in which we define a binary random variable X for the event that  $hash(q_j) = 0$ , and for every query  $q_i$ asked before  $q_i$  we have  $hash(q_i) \neq 0$ . Conditioned on the event  $hash(q_i) = 0$ , we get

$$\Pr_{hash \leftarrow \mathcal{H}}[X = 1] = \Pr_{hash \leftarrow \mathcal{H}}[hash(q_j) = 0 \land (\forall i < j : hash(q_i) \neq 0)]$$

$$= \Pr_{hash \leftarrow \mathcal{H}}[\forall i < j : hash(q_i) \neq 0 \mid hash(q_j) = 0] \Pr_{hash \leftarrow \mathcal{H}}[hash(q_j) = 0].$$

Now we use (0.0.3) twice and obtain

$$\begin{split} \Pr_{hash \leftarrow \mathcal{H}}[X=1] &= \frac{1}{2(h+v)} \left( 1 - \Pr_{hash \leftarrow \mathcal{H}}[\exists i < j : hash(q_i) = 0 \mid hash(q_j) = 0] \right) \\ &= \frac{1}{2(h+v)} \left( 1 - \Pr_{hash \leftarrow \mathcal{H}}[\exists i < j : hash(q_i) = 0] \right). \end{split}$$

Finally, we use union bound and the fact that  $j \leq (h + v)$  to get

$$\Pr_{hash \leftarrow \mathcal{H}}[X=1] \ge \frac{1}{2(h+v)} \left( 1 - \sum_{i < j} \Pr_{hash \leftarrow \mathcal{H}}[hash(q_i) = 0] \right) \ge \frac{1}{4(h+v)}.$$

Let  $\mathcal{P}_{Success}$  be the set of all  $(\pi^{(k)}, \rho)$  for which C succeeds in the random experiment Success for  $P^{(g)}$ . Furthermore, we denote the set of those  $(\pi^{(k)}, \rho)$  for which  $CanonicalSuccess^{P^{(g)}, C(\cdot, \cdot), hash}(\pi^{(k)}) =$ 1 by  $\mathcal{P}_{Canonical}$ . For fixed  $\pi^{(k)}$  and  $\rho$  if C succeeds canonically, then it also succeeds in the experiment Success for  $P^{(g)}$ . Hence,  $\mathcal{P}_{Canonical} \subseteq \mathcal{P}_{Success}$ , and we conclude

$$\Pr_{\substack{hash \leftarrow \mathcal{H} \\ \pi^{(k)}, \rho}} \left[ Canonical Success^{P^{(g)}, C^{(\cdot, \cdot)}, hash}(\pi^{(k)}, \rho) = 1 \right] = \mathbb{E}_{(\pi^{(k)}, \rho) \in \mathcal{P}_{Success}} \left[ \Pr_{hash \leftarrow \mathcal{H}}[X = 1] \right] \\
\geq \frac{\gamma}{4(h+v)}. \tag{0.0.4}$$

Algorithm: FindHash $(h, v, \gamma, n)$ 

**Oracle:** A solver circuit  $C^{(\cdot,\cdot)}$  for the k-wise direct product of  $P^{(1)}$ .

**Input:** Parameters  $h, v, \gamma, n$ 

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

Let  $\mathcal{H}$  be a family of pairwise independent hash functions  $Q \to \{0, 1, \dots, 2(h+v) - 1\}$ for i = 1 to  $32(h+v)^2/\gamma^2$  do:

 $hash \stackrel{\$}{\leftarrow} \mathcal{H}$ 

count := 0

for j := 1 to  $32(h+v)^2/\gamma^2$  do:  $\pi^{(k)} \stackrel{\$}{\leftarrow} \{0,1\}^{kn}$ 

if  $CanonicalSuccess^{P^{(g)},C^{(\cdot,\cdot)},hash}(\pi^{(k)})=1$  then

 $\begin{array}{c} count := count + 1 \\ \textbf{if} \ \ \frac{\gamma^2}{32(h+v)^2} count \geq \frac{\gamma}{6(h+v)} \ \textbf{then} \end{array}$ 

### return $\perp$

We show that **FindHash** chooses hash such that the canonical success probability of C with respect to  $P_{hash}$  is at least  $\frac{\gamma}{4(h+v)}$  almost surely. Let  $\mathcal{H}_{Good}$  denote a family of functions  $hash \in \mathcal{H}$ 

$$\Pr_{\pi^{(k)},\rho}\left[CanonicalSuccess^{P^{(g)},C^{(\cdot,\cdot)},hash}(\pi^{(k)},\rho)=1\right] \geq \frac{\gamma}{4(h+v)}$$

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

$$\Pr_{\pi^{(k)},\rho}\left[CanonicalSuccess^{P^{(g)},C^{(\cdot,\cdot)},hash}(\pi^{(k)},\rho)=1\right] \leq \frac{\gamma}{8(h+v)}.$$

Additionally, for a fixed hash, we define binary random variables  $X_1, \ldots, X_N$  such that

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

We first show that it is unlikely that **FindHash** returns  $hash \in \mathcal{H}_{Bad}$ . For  $hash \in \mathcal{H}_{Bad}$  we have  $\mathbb{E}_{\pi^{(k)},\rho}[X_i] < \frac{\gamma}{8(h+v)}$ . Therefore, for any fixed  $hash \in \mathcal{H}_{Bad}$  using the Chernoff bound we

$$\Pr_{\pi^{(k)}, \rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \ge \frac{\gamma}{6(h+v)} \right] \le \Pr_{\pi^{(k)}, \rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \ge (1 + \frac{1}{3}) \mathbb{E}[X_i] \right] \le e^{-\frac{\gamma}{4(h+v)}N/27}.$$

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

$$\Pr_{\pi^{(k)}, \rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \le \frac{\gamma}{6(h+v)} \right] \le \Pr_{\pi^{(k)}, \rho} \left[ \frac{1}{N} \sum_{i=1}^{N} X_i \le (1 - \frac{1}{3}) \mathbb{E}[X_i] \right] \le e^{-\frac{\gamma}{4(h+v)}N/27}.$$

Finally, we show that **FindHash** picks in one of its iteration  $hash \in \mathcal{H}_{Good}$  almost surely. Let  $Y_i$  be a binary random variable such that

$$Y_i = \begin{cases} 1 & \text{if in the } i\text{-th iteration } hash \in \mathcal{H}_{Good} \text{ is picked} \\ 0 & \text{otherwise} \end{cases}$$

From equation (0.0.4) we know that  $\Pr_{hash \leftarrow \mathcal{H}}[Y_i = 1] = \mathbb{E}[Y_i] \ge \frac{\gamma}{4(h+v)}$ , almost surely. Thus, we

$$\Pr_{hash \leftarrow \mathcal{H}} \left[ \sum_{i=1}^{K} Y_i = 0 \right] \le \left( 1 - \frac{\gamma}{4(h+v)} \right)^K \le e^{-\frac{\gamma}{4(h+v)}K}.$$

The bound stated in the Lemma 1.4 is achieved for  $K = N = 32(h+v)^2/\gamma^2$ 

We define the following solver circuit  $\hat{C}$ :

Circuit  $\widetilde{C}^{\Gamma_V^{(g)},\Gamma_H^{(g)},C,hash}(x,\rho)$ 

Oracle:  $\Gamma_V^{(g)}, \Gamma_H^{(k)}, hash, C$ 

**Input:** A transcript x, a bitstring  $\rho$ . **Output:** A tuple  $(q, y_1, \ldots, y_k)$  or  $\perp$ .

Run  $C_2^{\Gamma_V^{(g)},\Gamma_H^{(k)}}(x,\rho)$  if  $C_2$  asks a hint query on q then

if  $q \in P_{hash}$  then

return 
$$\bot$$
else
answer the query using  $\Gamma_H^{(k)}(q)$ 

if  $C_2$  asks a verification query  $(q, y_1, \ldots, y_k)$  then
if  $q \in P_{hash}$  then
return  $(q, y_1, \ldots, y_k)$ 
else
answer the verification query with 0

**Lemma 1.5** For fixed  $P^{(g)}$ , C and hash the following statement is true

$$\begin{split} \Pr_{\pi^{(k)},\rho}[Canonical Success^{P^{(g)},C,hash}(\pi^{(k)},\rho) &= 1] \\ &\leq \Pr_{\pi^{(k)},\rho} \left[ \Gamma_V^{(g)}(\widetilde{C}^{\Gamma_V^{(g)},\Gamma_H^{(k)},hash}(x,\rho)) &= 1 \right]. \\ & \qquad \qquad (\Gamma_V^{(g)},\Gamma_H^{(k)}) := & \langle P^{(g)}(\pi^{(k)}),S(\rho) \rangle_{P^{(g)}} \\ & \qquad \qquad x := & \langle P^{(g)}(\pi^{(k)}),S(\rho) \rangle_{trans} \end{split}$$

**Proof.** We fix  $\pi^{(k)}$ ,  $\rho$ . For  $(\Gamma_V^{(g)}, \Gamma_H^{(g)}) := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{P^{(g)}}$  and  $x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{\text{trans}}$ ,  $C(\rho)$  succeeds canonically, if and only if

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

Using this observation, we conclude that

$$\begin{split} \Pr_{\pi^{(k)},\rho} \left[ Canonical Success^{P^{(g)},C,hash}(\pi^{(k)},\rho) &= 1 \right] \\ &= \underset{\pi^{(k)},\rho}{\mathbb{E}} \left[ \Pr\left[ Canonical Success^{P^{(g)},C,hash}(\pi^{(k)},\rho) &= 1 \right] \right] \\ &= \underset{\pi^{(k)},\rho}{\Pr} \left[ \Gamma_{V}^{(g)}(\widetilde{C}^{\Gamma_{V}^{(g)},\Gamma_{H}^{(k)},hash}(x,\rho)) &= 1 \right]. \\ & \left( \Gamma_{V}^{(g)},\Gamma_{H}^{(k)} \right) &:= \langle P^{(g)}(\pi^{(k)}),C_{1}(\rho) \rangle_{P^{(g)}} \\ & x := \langle P^{(g)}(\pi^{(k)}),C_{1}(\rho) \rangle_{trans} \end{split}$$

Therefore, from a circuit C we can build a circuit  $\widetilde{C}$  that outputs  $\bot$  or  $(q, y_1, \ldots, y_k)$  such that  $q \in P_{hash}$ . Furthermore, the circuit  $\widetilde{C}$  asks no verification queries, and every hint query on q is such that  $q \notin P_{hash}$ .

Lemma 1.6 (Security amplification of a dynamic weakly verifiable puzzle with respect to  $P_{hash}$ .) For fixed  $P^{(1)}$  there exists an algorithm Gen, which takes as input a solver circuit C for  $P^{(g)}$ , a monotone function  $g: \{0,1\}^{(k)} \to \{0,1\}$ , a function hash  $: Q \to \{0,\ldots,2(h+v)-1\}$ , parameters  $\varepsilon,\delta,n$ , number of verification queries v and hint queries h asked by C, and outputs a circuit D such that the following holds: If C is such that

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

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then D satisfies almost surely

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

Additionally, Gen and D requires oracle access to g,  $P^{(1)}$  and C. Furthermore, D requires also oracle access to  $\Gamma_V$  and  $\Gamma_H$ , and asks at most  $\frac{6k}{\varepsilon}\log\left(\frac{6k}{\varepsilon}\right)h$  hint queries and no verification queries. Finally,  $Size(D) \leq Size(C)\frac{6k}{\varepsilon}$  and  $Time(Gen) = poly(k, \frac{1}{\varepsilon}, n, v, h)$ .

**Proof.** The following procedure estimates the function g with the first bit set to  $b \in \{0, 1\}$ .

### EvaluateFunctionProbability $^g(b, \varepsilon, \delta)$

Oracle: A function g.

**Input:** A bit  $b \in \{0,1\}$ , an integer  $k, \varepsilon$ 

**Output:** A number  $\widetilde{g} \in [0, 1]$ .

For 
$$i := 1$$
 to  $\frac{16k^2}{\varepsilon^2} \log(n)$  do:  
 $(b_2, \dots, b_k) \leftarrow \mu_{\delta}^{(k-1)}$   
 $\widetilde{g}_i := g(b, b_2, \dots, b_k)$  then  
return  $\frac{\varepsilon^2}{16k^2 \log(n)} \sum_{i=1}^{\frac{16k^2}{\varepsilon^2} \log(n)} g_i$ 

**Lemma 1.7 (Estimate of function g.)** The procedure **EvaluteFunctionProbability** outputs an estimate  $\widetilde{g}$  for the function  $g: \{0,1\}^n \to \{0,1\}$  with the first bit fixed to  $b \in \{0,1\}$  such that  $|\widetilde{g} - \Pr_{(u_2,\ldots,u_k) \leftarrow \mu_{\delta}^k}[g(b,u_2,\ldots,u_k) = 1]| \leq \frac{\varepsilon}{4k}$  almost surely.

**Proof.** We define binary random variable  $K_i$  for the event that  $g_i = 1$  and set  $N = \frac{\varepsilon}{4k} \log(n)$ . By Chernoff bound we get

$$\Pr\left[\left|\frac{1}{N}\sum_{i=1}^{N}\widetilde{g}_{i} - \mathbb{E}[X]\right| \ge \frac{\varepsilon}{4k}\right] \le 2e^{\log(n)/3}.$$

We define helper procedures EvalutePuzzles and EvaluateSurplus.

# EvaluatePuzzles $^{P^{(1)},C,hash}(\pi^{(k)},\rho)$

**Oracle:** A circuit C, an algorithm  $P^{(1)}$ , a function hash.

**Input:** Bitstrings  $\pi^{(k)}$ ,  $\rho$ .

**Output**: A tuple  $(c_1, \ldots, c_k)$ .

Run 
$$\langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle$$
  
 $(\Gamma_V^{(g)}, \Gamma_H^{(g)}) := \langle P(\pi^{(k)}), C_1(\rho) \rangle_{P^{(g)}}$   
 $x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{\text{trans}}$ 

$$(q,y^{(k)}):=\widetilde{C}^{\Gamma_V^{(g)},\Gamma_H^{(k)},C,hash}(x,\rho)$$

for i := 1 to k do: //simulate k rounds of sequential interaction

 $(\Gamma_V^i, \Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{P^{(1)}}$ 

for i := 1 to k do:

$$c_i := \Gamma_v^i(q, y_i)$$
  
return  $(c_1, \dots, c_k)$ 

## EvaluateSurplus $^{P^{(1)},C,hash}(\pi^*,b)$

**Oracle:** An algorithm  $P^{(1)}$ , a circuit C, a function hash, a function g.

**Input:** A bistring  $\pi^*$ , a bit b, an integer k.

Output: A circuit D.

```
\begin{split} \widetilde{g}_b &:= \mathbf{EvaluteFunctionProbability}^g(b, \varepsilon, \delta) \\ \mathbf{For} \ i &:= 1 \ \mathrm{to} \ \frac{16k^2}{\varepsilon^2} \log(n) \ \mathbf{do:} \\ & (\pi_{m+1}, \dots, \pi_k) \overset{\$}{\leftarrow} \{0, 1\}^{(k-m-1)n} \\ & \rho \overset{\$}{\leftarrow} \{0, 1\}^* \\ & (c_1, \dots, c_k) := \mathbf{EvalutePuzzles}^{P^{(1)}, C, hash}(\pi_1, \dots, \pi_m, \pi^*, \dots, \pi_k, \rho) \\ & \widetilde{s}^i_{\pi^*, b} := g(b, c_{m+1}, \dots, c_k) \\ & \mathbf{return} \ \frac{\varepsilon^2 \log(n)}{16k^2} \sum_{i=1}^{\frac{16k^2}{\varepsilon^2}} \log(n) \ \widetilde{s}^i_{\pi^*, b} - \widetilde{g}_b \end{split}
```

Circuit  $D = (D_1, D_2)(\sigma)$ 

Phase I  $D_1^{P^{(1)},C}(\sigma)$ 

**Oracle:** A poser  $P^{(1)}$ , a circuit C, a function hash.

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

**Hard coded:** Bitstrings  $\pi_1, \ldots, \pi_{m-1}$ .

**Output**: Transcripts  $x_1, \ldots, x_{m-1}, x^*$ .

for i := 1 to m-1 do: Simulate  $\langle P^{(1)}(\pi_i), C_1(\rho) \rangle$ Let  $x_i := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{\text{trans}}$ Let  $(\Gamma_V^i, \Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{P^{(1)}}$ 

Interact with the problem poser using  $C_1(\rho)$ .

Let  $x^*$  be the transcript of the interaction

Let  $\Gamma_V^*, \Gamma_H^*$  be the verification and hint oracles output by the problem poser.

Let 
$$\Gamma_V^{(m-1)} := (\Gamma_V^1, \dots, \Gamma_V^{m-1})$$
  
Let  $\Gamma_H^{(m-1)} := (\Gamma_H^1, \dots, \Gamma_H^{m-1})$   
Let  $x^{(m-1)} := (x_1, \dots, x_{m-1})$ 

Phase II  $D_2^{P^{(1)},C}(x^*,\sigma)$ 

**Oracle:** A poser  $P^{(1)}$ , a circuit C, a function hash, circuits  $\Gamma_V^*$  and  $\Gamma_H^*$ .

**Input:** A transcript  $x^*$ , a bitstring  $\sigma \in \{0, 1\}^*$ .

Output: A circuit D.

```
Let \Gamma_V^{(m-1)}, \Gamma_H^{(m-1)} and x_1, \ldots, x_{k-1} be the same as in the Phase I.

for at most \frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon}) iterations do:
\pi^{(k)} \leftarrow \operatorname{read} k \cdot n \text{ bits from } \sigma
\text{for } i := 1 \text{ to } m - 1 \text{ do:} \qquad // \text{ finish remaining simulation of puzzles}
\operatorname{Simulate} \langle P^{(1)}(\pi_i), C_1(\rho) \rangle
\operatorname{Let} x_i := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle_{\operatorname{trans}}
\operatorname{Let} \Gamma_V^{(g)} := g(\Gamma_V^1, \ldots, \Gamma_V^{m-1}, \Gamma_V^*, \Gamma_V^{m+1}, \ldots, \Gamma_V^k)
\operatorname{Let} \Gamma_H^{(k)} := (\Gamma_H^1, \ldots, \Gamma_H^{m-1}, \Gamma_H^*, \Gamma_H^{m+1}, \ldots, \Gamma_H^k)
(q, y_1, \ldots, y_{m-1}, y^*, \ldots, y_k) := \widetilde{C}^{\Gamma_V^{(g)}, \Gamma_H^{(k)}, C, hash}((x_1, \ldots, x_{m-1}, x^*, \ldots, x_k), \rho)
\operatorname{if} g(1, c_{m+1}, \ldots, c_k) = 1 \wedge g(0, c_{m+1}, \ldots, c_k) = 0 \text{ then}
\operatorname{return} (q, y^*)
```

```
Algorithm Gen(C, g, \varepsilon, \delta, n, v, h, hash)
Oracle: P^{(1)}, C, g, hash
Input: \varepsilon, \delta, n, v, h
Output: A circuit D
Let m be the recursion depth of Gen.
 \begin{aligned} \textbf{for} \ i := 1 \ \text{to} \ \frac{6k}{\varepsilon} \log(n) \\ \pi^* \leftarrow \{0,1\}^n \end{aligned} 
       \widetilde{S}_{\pi^*,0} := \mathbf{EvaluateSurplus}^{P^{(1)},C,hash}(\pi^*,0)
       \widetilde{S}_{\pi^*,1} := \mathbf{EvaluateSurplus}^{P^{(1)},C,hash}(\pi^*,1)
       if \exists b \in \{0,1\} : \widetilde{S}_{\pi^*,b} \ge (1 - \frac{3}{4k})\varepsilon
               Let \widetilde{C}' be a circuit that first simulates interaction with P^{(1)}, C_1
               Fix \pi_1 := \pi^*
               g'(b_2,\ldots,b_k):=g(b,b_2,\ldots,b_k)
               m := m + 1
               return Gen(\widetilde{C}, g', \varepsilon, \delta, n, v, h, hash)
// all estimates are lower than (1-\frac{3}{4k})\varepsilon
Hard code \pi_1, \ldots, \pi_{m-1} into the circuit D.
return D^{\tilde{C}}
```

For 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 success probability of C in the random experiment CanonicalSuccess is as least  $\delta + \varepsilon$ , and D runs circuit C to obtain (q,y). In case g is a constant function the statement is vacuously true.

For fixed  $\pi^{(k)}$ ,  $\rho$  let  $(\Gamma_V^{(g)}, \Gamma_H^{(k)}) := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{P^{(g)}}$  and  $x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{\text{trans}}$ . Additionally, for the *i*-th round of interaction between  $P^{(g)}(\pi^{(k)})$  and  $C_1(\rho)$  let us denote  $(\Gamma_V^i, \Gamma_H^i) := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle$ , and  $x_i := \langle P^{(1)}(\pi_i), C_1(\rho) \rangle$ . Finally, for  $(q, y_1, \dots, y_k) := \widetilde{C}(x^{(k)}, \rho)$  let  $c_i := \Gamma_V^i(q, y_i)$ . We define the surplus:

$$S_{\pi^*,b} = \Pr_{\pi^{(k)}} \left[ g(b, c_2, \dots, c_k) = 1 \right] - \Pr_{\mu^{(k)}} \left[ g(b, u_2, \dots, u_k) = 1 \right]$$
 (0.0.5)

The surplus  $S_{\pi^*,b}$  tells us how good  $\widetilde{C}$  performs when the first randomness  $\pi_1$  of the poser is fixed, and the fact whether  $\widetilde{C}$  succeeds in solving the puzzle defined by  $P^{(1)}(\pi_1)$  is disregarded. Instead, the bit b is used as the first input to g.

The procedure **EvaluateSurplus** returns the estimate  $\widetilde{S}_{\pi^*,b}$  for  $S_{\pi^*,b}$ . All puzzles used during obtaining the estimate are generated internally. Therefore, it is possible to answer all hint and verification queries, without calls to the verification oracles.

**Lemma 1.8** The estimate  $\widetilde{S}_{\pi^*,b}$  returned by EvaluteEstimate differs from  $S_{\pi^*,b}$  by at most  $\frac{\varepsilon}{2k}$  almost surely.

**Proof.** First we define a random variable  $K_i$  for the event that  $\widetilde{s}_{\pi^*,b}^i = 1$ . Then using union bound and similar argument as in Lemma 1.7 we obtain that  $\frac{\varepsilon^2 \log(n)}{16k^2} \sum_{i=1}^{\frac{16k^2}{\varepsilon^2}} \log^{(n)} \widetilde{s}_{\pi^*,b}^i$  differs from  $\mathbb{E}[g(b,c_2,\ldots,c_k)]$  by at most  $\frac{\varepsilon}{4k}$  almost surely. Together, with Lemma 1.7 we conclude that the surplus estimate returned by **EstimateSurplus** differs from  $S_{\pi^*,b}$  by at most  $\frac{\varepsilon}{2k}$ , almost surely.

From Lemma 1.8 we conclude that if  $\widetilde{S}_{\pi^*,b} \geq (1 - \frac{3}{4k})\varepsilon$ , then  $S_{\pi^*,b} \geq (1 - \frac{1}{k})\varepsilon$  almost surely.

Let us assume that Gen manages to find an estimate that satisfies  $\widetilde{S}_{\pi^*,b} \geq (1 - \frac{3}{4k})\varepsilon$ . In this case we define a new monotone function  $g'(b_2,\ldots,b_k) := g(b,b_2,\ldots,b_k)$ , and fix the m-th randomness used by problem poser to  $\pi_m$ . The circuit  $\widetilde{C}$  satisfies the conditions of Lemma 1.6 for the remaining k-1 puzzles and we recurse using g'.

If all estimates are less than  $(1-\frac{3}{4k})\varepsilon$ , then intuitively C does not perform much better on the remaining k-1 puzzles than an algorithm that solves each puzzle independent with probability  $\delta$ . However, from the assumption we know that on all k puzzles  $\widetilde{C}$  has higher success probability. Therefore, it is likely that the first puzzle is correctly solved with the probability higher than  $\delta$ . We now show that this intuition is indeed correct. For a fixed  $\pi^*$  using (0.0.5), we get

$$\Pr_{u \leftarrow \mu_{\delta}^{k}}[g(1, u_{2}, \dots, u_{k}) = 1] - \Pr_{u \leftarrow \mu_{\delta}^{k}}[g(0, u_{2}, \dots, u_{k}) = 1] = 
\Pr_{\pi^{(k)}, \rho}[g(1, c_{2}, \dots, c_{k}) = 1 \mid \pi_{1} = \pi^{*}] - \Pr_{\pi^{(k)}, \rho}[g(0, c_{2}, \dots, c_{k}) = 1 \mid \pi_{1} = \pi^{*}] - (S_{\pi^{*}, 1} - S_{\pi^{*}, 0}).$$
(0.0.6)

From the monotonicity of g we know that for any set of tuples  $(b_1, \ldots, b_k)$  and sets  $\mathcal{B}_0 = \{(b_1, b_2, \ldots, b_k) : g(0, b_2, \ldots, b_k) = 1\}$ ,  $\mathcal{B}_1 = \{(b_1, b_2, \ldots, b_k) : g(1, b_2, \ldots, b_k) = 1\}$  we have  $\mathcal{B}_0 \subseteq \mathcal{B}_1$ . Hence, we can write (0.0.6):

$$\Pr_{u \leftarrow \mu_{\delta}^{k}} [g(1, u_{2}, \dots, u_{k}) = 1 \land g(0, u_{2}, \dots, u_{k}) = 0] = 
\Pr_{\pi^{(k)}, \rho} [g(1, c_{2}, \dots, c_{k}) = 1 \land g(0, c_{2}, \dots, c_{k}) = 0 \mid \pi_{1} = \pi^{*}] - (S_{\pi^{*}, 1} - S_{\pi^{*}, 0}).$$
(0.0.7)

Let  $G_{u^{(k)}}$  denote the event  $g(1,u_2,\ldots,u_k)=1 \wedge g(0,u_2,\ldots,u_k)=0$ , and correspondingly  $G_{\pi^{(k)}}:=g(1,c_2,\ldots,c_k)=1) \wedge (g(0,c_2,\ldots,c_k)=0$ . From (0.0.7) for  $\pi=\pi^*$  fixed we obtain

$$\Pr_{\substack{\rho \\ (\Gamma_{V}, \Gamma_{H}) := \langle P^{(1)}(\pi^{*}), D_{1}(\rho) \rangle_{P^{(1)}} \\ x^{*} := \langle P^{(1)}(\pi^{*}), D_{1}(\rho) \rangle_{\text{trans}}}} [\Gamma_{V}(D_{2}(x^{*}, \rho)) = 1] = \frac{\Pr_{\rho}[\Gamma_{V}(D_{2}(x^{*}, \rho)) = 1 \mid \pi_{1} = \pi^{*}] \Pr_{\rho}[G_{\pi} \mid \pi_{1} = \pi^{*}]}{\Pr_{u \leftarrow \mu_{\delta}^{k}}[G_{\mu}]} \\
- \frac{\Pr_{\sigma}[\Gamma_{V}(D_{2}(x^{*}, r)) = 1 \mid \pi_{1} = \pi^{*}](S_{\pi^{*}, 1} - S_{\pi^{*}, 0})}{\Pr_{\sigma}[G_{\mu}]}$$

$$\frac{\Pr_{\sigma}[\Gamma_{V}(D_{2}(x^{*}, r)) = 1 \mid \pi_{1} = \pi^{*}](S_{\pi^{*}, 1} - S_{\pi^{*}, 0})}{\Pr_{\sigma}[G_{\mu}]}$$

$$\frac{\Pr_{\sigma}[G_{\mu}]}{(0.0.8)}$$

We can write the first summand of (0.0.8) as

$$\Pr_{\rho}[\Gamma_{V}(D_{2}(x^{*},\rho)) = 1 \mid \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[G_{\pi} \mid \pi_{1} = \pi^{*}] = 
\Pr_{\rho}[D_{2}(x^{*},\rho) \neq \bot \mid \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[c_{1} = 1 \mid G_{\pi}, \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[G_{\pi} \mid \pi_{1} = \pi^{*}]$$
(0.0.9)

where we make use of the fact that the event  $G_{\pi}$  implies  $D(x^*, r) \neq \bot$ . We consider two cases. For  $\Pr_{\pi^k}[g(1, c_2, \ldots, c_k) = 1 \land g(0, c_2, \ldots, c_k) = 0 \mid \pi_1 = \pi^*] \leq \frac{\varepsilon}{6k}$  then

$$\Pr_{\pi^{(k)}}[c_1 = 1 \mid G_{\pi}, \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[G_{\pi} \mid \pi_1 = \pi^*] \le \frac{\varepsilon}{6k}, \tag{0.0.10}$$

and when  $\Pr_{\pi^k}[g(1,c_2,\ldots,c_k)=1 \land g(0,c_2,\ldots,c_k)=0] > \frac{\varepsilon}{6k}$  then circuit D outputs  $\bot$  only if it fails in all  $\frac{6k}{\varepsilon}\log(\frac{6k}{\varepsilon})$  iterations to find  $\pi^{(k)}$  such that  $g(1,c_2,\ldots,c_k)=1 \land g(0,c_2,\ldots,c_k)=0$  which happens with probability

$$\Pr_{r}[D(x^*, r) = \bot \mid \pi_1 = \pi^*] \le (1 - \frac{\varepsilon}{6k})^{\frac{6k}{\varepsilon} \log(\frac{\varepsilon}{6k})} \le \frac{\varepsilon}{6k}.$$
 (0.0.11)

We conclude that in both cases:

$$\Pr_{r}[D(x^{*}, r) \neq \bot \mid \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[c_{1} = 1 \mid G_{\pi}, \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[G_{\pi} \mid \pi_{1} = \pi^{*}] 
\geq \Pr_{\pi^{(k)}}[c_{1} = 1 \mid G_{\pi}, \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[G_{\pi} \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k}.$$
(0.0.12)

Therefore, we have

$$\Pr_{r}[D(x^{*}, r) \neq \bot \mid \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[c_{1} = 1 \mid G_{\pi}, \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[G_{\pi} \mid \pi_{1} = \pi^{*}]$$

$$= \Pr_{\pi^{(k)}}[c_{1} = 1 \land g(1, c_{2}, \dots, c_{k}) = 1 \land g(0, c_{2}, \dots, c_{k}) = 0 \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k}$$

$$= \Pr_{\pi^{(k)}}[g(c_{1}, c_{2}, \dots, c_{k}) = 1 \land g(0, c_{2}, \dots, c_{k}) = 0 \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k}$$

$$= \Pr_{\pi^{(k)}}[g(c_{1}, c_{2}, \dots, c_{k}) = 1 \mid \pi_{1} = \pi^{*}] - \Pr_{\pi^{(k)}}[g(0, c_{2}, \dots, c_{k}) = 0 \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k},$$

and finally by (0.0.5)

$$\Pr_{r}[D(x^{*}, r) \neq \bot \mid \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[c_{1} = 1 \mid G_{\pi}, \pi_{1} = \pi^{*}] \Pr_{\pi^{(k)}}[G_{\pi} \mid \pi_{1} = \pi^{*}]$$

$$= \Pr_{\pi^{(k)}}[g(c_{1}, c_{2}, \dots, c_{k}) = 1 \mid \pi_{1} = \pi^{*}] - \Pr_{\mu_{\delta}^{(k)}}[g(0, \mu_{2}, \dots, \mu_{k}) = 0 \mid \pi_{1} = \pi^{*}] - S_{\pi^{*}, 0} - \frac{\varepsilon}{6k}.$$

$$(0.0.13)$$

Inserting this result into the equation (0.0.8) yields

$$\Pr_{r,\pi}[\Gamma_{V}(D(x,r)) = 1] = \mathbb{E}_{\pi} \left[ \Pr_{r}[D(x,r) = 1 \mid \pi_{1} = \pi^{*}] \right] \\
= \mathbb{E}_{\pi} \left[ \frac{\Pr_{\pi^{(k)}}[g(c) = 1 \mid \pi_{1} = \pi^{*}] - \Pr_{\mu_{\delta}^{(k)}}[g(0,\mu_{2},\dots,\mu_{k}) = 0 \mid \pi_{1} = \pi^{*}] - \frac{\varepsilon}{6k}}{\Pr_{\mu_{\delta}^{k}}[G_{\mu}]} \right] \\
- \mathbb{E}_{\pi} \left[ \frac{S_{\pi^{*},0} + \Pr_{r}[\Gamma_{V}^{(g)}(D(x^{*},r)) = 1 \mid \pi_{1} = \pi^{*}](S_{\pi^{*},1} - S_{\pi^{*},0})}{\Pr_{\mu_{\delta}^{k}}[G_{\mu}]} \right] \tag{0.0.14}$$

For the second summand we show that if we do not recurse, then almost surely majority of estimates is low. Let assume

$$\Pr_{\pi} \left[ \left( S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon \right) \wedge \left( S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon \right) \right] < 1 - \frac{\varepsilon}{6k}, \tag{0.0.15}$$

then the algorithm recurses almost surely. Therefore, under the assumption that Gen does not recurse, we have almost surely

$$\Pr_{\pi} \left[ \left( S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon \right) \wedge \left( S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon \right) \right] \ge 1 - \frac{\varepsilon}{6k}. \tag{0.0.16}$$

Let us define a set

$$W = \left\{ \pi : \left( S_{\pi,0} \le (1 - \frac{1}{2k})\varepsilon \right) \land \left( S_{\pi,1} \le (1 - \frac{1}{2k})\varepsilon \right) \right\}$$
 (0.0.17)

and use  $\mathcal{W}^c$  to denote the complement of  $\mathcal{W}$ . We bound the second summand in (0.0.14)

$$\mathbb{E}_{\pi} \left[ S_{\pi^*,0} + \Pr_{r} [\Gamma_{V}^{(g)}(D(x^*,r)) = 1 \mid \pi_{1} = \pi^*] (S_{\pi^*,1} - S_{\pi^*,0}) \right] \\
= \mathbb{E}_{\pi \in \mathcal{W}^{c}} \left[ S_{\pi^*,0} + \Pr_{r} [\Gamma_{V}^{(g)}(D(x^*,r)) = 1 \mid \pi = \pi^*] (S_{\pi^*,1} - S_{\pi^*,0}) \right] \\
+ \mathbb{E}_{\pi \in \mathcal{W}} \left[ S_{\pi^*,0} + \Pr_{r} [\Gamma_{V}^{(g)}(D(x^*,r)) = 1 \mid \pi = \pi^*] (S_{\pi^*,1} - S_{\pi^*,0}) \right] \qquad (0.0.18) \\
\leq \frac{\varepsilon}{6k} + \mathbb{E}_{\pi \in \mathcal{W}^{c}} \left[ S_{\pi^*,0} + \Pr_{r} [\Gamma_{V}^{(g)}(D(x^*,r)) = 1 \mid \pi = \pi^*] ((1 - \frac{1}{2k})\varepsilon - S_{\pi^*,0}) \right] \qquad (0.0.19) \\
\leq \frac{\varepsilon}{6k} + 1 - \frac{\varepsilon}{2k} = 1 - \frac{\varepsilon}{3k} \qquad (0.0.20)$$

Finally, we insert this result into equation (0.0.14) and make use of the fact

$$\Pr[g(u) = 1] = \Pr[(g(0, \mu_2, \dots, \mu_k) = 1) \lor (g(1, \mu_2, \dots, \mu_k) = 1 \land g(0, \mu_2, \dots, \mu_k) = 0 \land \mu_1 = 1)]$$

$$= \Pr[g(0, \mu_2, \dots, \mu_k) = 1] + \Pr[g(1, \mu_2, \dots, \mu_k) = 1 \land g(0, \mu_2, \dots, \mu_k) = 0] \Pr[\mu_1 = 1]$$

which yields

$$\Pr_{r,\pi}[D(x,r)=1] \ge \mathbb{E}_{\pi} \left[ \frac{\Pr_{\pi^{(k)}}[g(c)=1 \mid \pi_{1}=\pi^{*}] - \Pr_{\mu_{\delta}^{(k)}}[g(0,\mu_{2},\dots,\mu_{k})=0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{\mu_{\delta}^{k}}[G_{\mu}]} \right]$$

Using the assumptions of Lemma 1.6, we get

$$\Pr_{r,\pi}[\Gamma_{V}(D(x,r)) = 1] \ge \frac{\Pr_{\mu_{\delta}^{(k)}}[g(\mu) = 1] + \varepsilon + \Pr_{\mu_{\delta}^{(k)}}[g(0,\mu_{2},\dots,\mu_{k}) = 0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{\mu_{\delta}^{(k)}}[G_{\mu}]} \\
\ge \frac{\varepsilon + \delta \Pr_{\mu_{\delta}^{(k)}}[G_{\mu}] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{\mu_{\delta}^{k}}[G_{\mu}]} \ge \delta + \frac{\varepsilon}{6k} \tag{0.0.21}$$

Now, we can show that the Theorem 1.1 follows by Lemma 1.4 and Lemma 1.6. First we define the following circuit:

 $\mathbf{E}$ 

Oracle: Circuit D from Lemma 1.6

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

Run circuit  $(q, y) = D(\rho)$ 

if  $(q, y) \neq \bot$  then

make a verification query (q, y)

The circuit E is output by the following algorithm Gen.

### Gen

Oracle:  $C, P^{(1)}, g$ Input:  $\varepsilon, \delta, n, h, v$ 

Let  $\mathcal{H}$  be a set of pairwise independent hash functions  $Q \to \{0, 1, \dots, 2(h+v)-1\}$  $hash := \mathbf{FindHash}(\mathcal{H}, h+v)$ 

$$\begin{split} D := Gen(C, g, \varepsilon, \delta, n, h, v, hash) \\ \mathbf{return} \ D^{P^1, C, hash}(\rho) \end{split}$$

From the assumptions of Theorem 1.3 we know that success probability of C is at least

$$8(h+v)\left(\Pr_{u\leftarrow\mu_{\delta}^{k}}[g(u)=1]+\varepsilon\right),\,$$

then by Lemma 1.4, the canonical success probability of  $\widetilde{C}$  with respect to function hash is at least

$$\left(\Pr_{u \leftarrow \mu_{\delta}^{k}} \left[ g(u) = 1 \right] + \varepsilon \right).$$

Then we apply Lemma 1.6 with respect to  $\widetilde{C}$  and hash which yields a circuit D that outputs (q,y) such that

$$\Pr_{\substack{\pi,\sigma\\ (\Gamma_V,\Gamma_H):=\langle P^{(1)}(\pi),D(\rho)\rangle_{P^{(1)}}\\ x:=\langle P^{(1)}(\pi),D(\rho)\rangle_{\text{trans}}}} \left[\Gamma_V(D^{P^{(1)},C,\Gamma_V,\Gamma_H,hash}(x,\sigma))=1\right] \geq (\delta+\frac{\varepsilon}{6k}).$$

Hence, the probability that the verification query made by E is successful is at least  $(\delta + \frac{\varepsilon}{6k})$ .