

We write  $u \leftarrow \mu_\delta^k$  to denote a tuple  $u$  of length  $k$  which each element is independently drawn from the Bernoulli distribution with parameter  $\delta$ . The protocol execution between probabilistic algorithms  $A$  and  $B$  is denoted by  $\langle A, B \rangle$ . Additionally, the output of  $A$  in such protocol execution is denoted by  $\langle A, B \rangle_A$ , and the transcript of the 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  $S$  with the randomness fixed to  $\rho \in \{0, 1\}^*$ . The poser  $P(\pi)$  and the solver  $S_1(\rho)$  interact. We denote by  $x$  the transcript of the interaction. 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  $x$ , 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  $g : \{0, 1\}^k \rightarrow \{0, 1\}$  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  $\pi^{(k)} := (\pi_1, \dots, \pi_k)$ . 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  $\Gamma_V^i, \Gamma_H^i$ . Finally,  $P^{(g)}$  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  $C := (C_1, C_2)$  to denote that  $C$  in the first phase uses  $C_1$ , and in the second phase  $C_2$ . 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  $q \in Q$ , for which a hint query has been asked before.

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

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

**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  $C_2^{\Gamma_V, \Gamma_H}(x, \rho)$

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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

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

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

**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 the 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  $g : \{0, 1\}^k \rightarrow \{0, 1\}$ , 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*

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

*then  $D$  satisfies almost surely*

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

*Additionally,  $Gen$  and  $D$  require oracle access to  $g$ ,  $P^{(1)}, C$ . Furthermore,  $D$  requires also oracle access to  $\Gamma_V$  and  $\Gamma_H$ , and asks at most  $h$  hint queries and  $v$  verification queries. 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.

Let  $hash : Q \rightarrow \{0, 1, \dots, 2(h + v) - 1\}$ , 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 *CanonicalSuccess* 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  $C^{(\cdot, \cdot)}$  for  $P$ .

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

**Input:** Bitstrings:  $\pi, \rho$ .

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

Run  $\langle P(\pi), C_1(\rho) \rangle$

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

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

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

Let  $(q_j, y_j)$  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

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

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: a set  $\mathcal{H}$  of pairwise independent hash functions  $Q \rightarrow \{0, 1, \dots, 2(h+v)-1\}$ , 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  $Q \rightarrow \{0, 1, \dots, 2(h+v)-1\}$ . 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 *CanonicalSuccess* 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  $q_j$  we have  $hash(q_i) \neq 0$ . Conditioned on the event  $hash(q_i) = 0$ , we get

$$\begin{aligned} \Pr_{hash \leftarrow \mathcal{H}}[X = 1] &= \Pr_{hash \leftarrow \mathcal{H}}[hash(q_j) = 0 \wedge (\forall i < j : hash(q_i) \neq 0)] \\ &= \Pr_{hash \leftarrow \mathcal{H}}[\forall i < j : hash(q_i) \neq 0 \mid hash(q_j) = 0] \Pr_{hash \leftarrow \mathcal{H}}[hash(q_j) = 0]. \end{aligned}$$

Now we use (0.0.3) twice and obtain

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

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

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

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

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

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**Algorithm: FindHash**( $\mathcal{H}, h, v, \gamma, n$ )

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

**Input:** A set  $\mathcal{H}$ , parameters  $h, v, \gamma, n$

**Output:** A function  $hash \in \mathcal{H}$ .

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for  $i := 1$  to  $32(h+v)^2/\gamma^2$ 
   $hash \xleftarrow{\$} \mathcal{H}$ 
   $count := 0$ 
  for  $j := 1$  to  $32(h+v)^2/\gamma^2$ 
     $\pi^{(k)} \xleftarrow{\$} \{0, 1\}^{kn}$ 
    if  $CanonicalSuccess^{P^{(g)}, C(\cdot, \cdot), hash}(\pi^{(k)}) = 1$  then
       $count := count + 1$ 
  if  $\frac{\gamma^2}{32(h+v)^2} count \geq \frac{\gamma}{6(h+v)}$ 
    return  $hash$ 
return  $\perp$ 

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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  $hash \in \mathcal{H}$  for which

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

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  $X_1, \dots, X_N$  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 get

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

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

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

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  $\Pr_{hash \leftarrow \mathcal{H}}[Y_i = 1] = \mathbb{E}[Y_i] \geq \frac{\gamma}{4(h+v)}$ , almost surely. Thus, we get

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

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  $\tilde{C}$ .

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

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

**Input:** A transcript  $x$ , a bitstring  $\rho$ .

**Output:** A tuple  $(q, y_1, \dots, y_k)$  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**  $\perp$

**else**

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

**if**  $C_2$  asks a verification query  $(q, y_1, \dots, y_k)$  **then**

**if**  $q \in P_{hash}$  **then**

**return**  $(q, y_1, \dots, y_k)$

**else**

      answer the verification query with 0

**return**  $\perp$

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

$$\begin{aligned} & \Pr_{\pi^{(k)}, \rho} [CanonicalSuccess^{P^{(g)}, C, hash}(\pi^{(k)}, \rho) = 1] \\ & \leq \Pr_{\pi^{(k)}, \rho} \left[ \Gamma_V^{(g)}(\tilde{C}^{\Gamma_V^{(g)}, \Gamma_H^{(k)}, hash}(x, \rho)) = 1 \right]. \\ & \quad (\Gamma_V^{(g)}, \Gamma_H^{(k)}) := \langle P^{(g)}(\pi^{(k)}), S(\rho) \rangle_{P^{(g)}} \\ & \quad x := \langle P^{(g)}(\pi^{(k)}), S(\rho) \rangle_{trans} \end{aligned}$$

**Proof.** We fix  $\pi^{(k)}, \rho$ . Then 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

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

Using this observation, we conclude that

$$\begin{aligned} & \Pr_{\pi^{(k)}, \rho} \left[ \text{CanonicalSuccess}^{P^{(g)}, C, \text{hash}}(\pi^{(k)}, \rho) = 1 \right] \\ &= \mathbb{E}_{\pi^{(k)}, \rho} \left[ \Pr \left[ \text{CanonicalSuccess}^{P^{(g)}, C, \text{hash}}(\pi^{(k)}, \rho) = 1 \right] \right] \\ &= \Pr_{\pi^{(k)}, \rho} \left[ \Gamma_V^{(g)}(\widetilde{C}_2^{\Gamma_V^{(g)}, \Gamma_H^{(g)}, \text{hash}}(x, \rho)) = 1 \right]. \\ & \quad \left( \Gamma_V^{(g)}, \Gamma_H^{(k)} \right) := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{P^{(g)}} \\ & \quad x := \langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle_{\text{trans}} \end{aligned}$$

□

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

**Lemma 1.6 (Security amplification of a dynamic weakly verifiable puzzle with respect to  $P_{\text{hash}}$ .)** For fixed  $P^{(1)}$  there exists an algorithm  $\text{Gen}$ , which takes as input a solver circuit  $C$  for  $P^{(g)}$ , a monotone function  $g : \{0, 1\}^{(k)} \rightarrow \{0, 1\}$ , a function  $\text{hash} : Q \rightarrow \{0, \dots, 2(h + v) - 1\}$ , 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

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

then  $D$  satisfies almost surely

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

Additionally,  $\text{Gen}$  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  $\frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon}) h$  hint queries and no verification queries. Finally,  $\text{Size}(D) \leq \text{Size}(C) \frac{6k}{\varepsilon}$  and  $\text{Time}(\text{Gen}) = \text{poly}(k, \frac{1}{\varepsilon}, n, v, h)$ .

**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:**

$(b_2, \dots, b_k) \leftarrow \mu_\delta^{(k-1)}$

$\tilde{g}_i := g(b, b_2, \dots, b_k)$ <b>then</b> <b>return</b> $\frac{\varepsilon^2}{16k^2 \log(n)} \sum_{i=1}^{\frac{16k^2}{\varepsilon^2} \log(n)} g_i$
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**Lemma 1.7 (Estimate of function  $g$ .)** *The procedure **EvaluateFunctionProbability** outputs an estimate  $\tilde{g}$  for the function  $g : \{0, 1\}^n \rightarrow \{0, 1\}$  with the first bit fixed to  $b \in \{0, 1\}$  such that  $|\tilde{g} - \Pr_{(u_2, \dots, u_k) \leftarrow \mu_\delta^k} [g(b, u_2, \dots, u_k) = 1]| \leq \frac{\varepsilon}{4k}$  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

$$\Pr \left[ \left| \frac{1}{N} \sum_{i=1}^N \tilde{g}_i - \mathbb{E}[X] \right| \geq \frac{\varepsilon}{4k} \right] \leq 2e^{\log(n)/3}. \quad \square$$

We define helper procedures **EvaluatePuzzles** and **EvaluateSurplus**.

<b>EvaluatePuzzles</b> <sup><math>P^{(1)}, C, hash</math></sup> ( $\pi^{(k)}, \rho$ )
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**Oracle:** A circuit  $C$ , an algorithm  $P^{(1)}$ , a function  $hash$ .

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

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

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**Run**  $\langle P^{(g)}(\pi^{(k)}), C_1(\rho) \rangle$

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

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

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

**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**  $(c_1, \dots, c_k)$

<b>EvaluateSurplus</b> <sup><math>P^{(1)}, C, hash</math></sup> ( $\pi^*, b$ )
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**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  $D$ .

---

$\tilde{g}_b := \text{EvaluateFunctionProbability}^g(b, \varepsilon, \delta)$

**For**  $i := 1$  **to**  $\frac{16k^2}{\varepsilon^2} \log(n)$  **do:**

$(\pi_{m+1}, \dots, \pi_k) \xleftarrow{\$} \{0, 1\}^{(k-m-1)n}$

$\rho \xleftarrow{\$} \{0, 1\}^*$

$(c_1, \dots, c_k) := \text{EvaluatePuzzles}^{P^{(1)}, C, hash}(\pi_1, \dots, \pi_m, \pi^*, \dots, \pi_k, \rho)$

$\tilde{s}_{\pi^*, b}^i := g(b, c_{m+1}, \dots, c_k)$

---

**return**  $\frac{\varepsilon^2 \log(n)}{16k^2} \sum_{i=1}^{\frac{16k^2}{\varepsilon^2} \log(n)} \tilde{s}_{\pi^*, b}^i - \tilde{g}_b$

---

**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  $\pi_1, \dots, \pi_{m-1}$ .

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

---

**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  $x_1, \dots, x_{k-1}$  be the same as in the **Phase I**.

**for** at most  $\frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon})$  iterations **do:**

$\pi^{(k)} \leftarrow$  read  $k \cdot n$  bits from  $\sigma$

**for**  $i := 1$  **to**  $m - 1$  **do:**                      // finish remaining simulation of puzzles

        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  $\Gamma_V^{(g)} := g(\Gamma_V^1, \dots, \Gamma_V^{m-1}, \Gamma_V^*, \Gamma_V^{m+1}, \dots, \Gamma_V^k)$

    Let  $\Gamma_H^{(k)} := (\Gamma_H^1, \dots, \Gamma_H^{m-1}, \Gamma_H^*, \Gamma_H^{m+1}, \dots, \Gamma_H^k)$

$(q, y_1, \dots, y_{m-1}, y^*, \dots, y_k) := \tilde{C}_{\Gamma_V^{(g)}, \Gamma_H^{(k)}, C, \text{hash}}((x_1, \dots, x_{m-1}, x^*, \dots, x_k), \rho)$

**if**  $g(1, c_{m+1}, \dots, c_k) = 1 \wedge g(0, c_{m+1}, \dots, c_k) = 0$  **then**

**return**  $(q, y^*)$

**return**  $\perp$

---

Let  $m$  be the global variable with the initial value 0.



**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  $Gen$ .

**for**  $i := 1$  to  $\frac{6k}{\varepsilon} \log(n)$

$\pi^* \leftarrow \{0, 1\}^n$

$\tilde{S}_{\pi^*, 0} := \mathbf{EvaluateSurplus}^{P^{(1)}, C, hash}(\pi^*, 0)$

$\tilde{S}_{\pi^*, 1} := \mathbf{EvaluateSurplus}^{P^{(1)}, C, hash}(\pi^*, 1)$

**if**  $\exists b \in \{0, 1\} : \tilde{S}_{\pi^*, b} \geq (1 - \frac{3}{4k})\varepsilon$

Fix  $\pi_m := \pi^*$

$g'(b_2, \dots, b_k) := g(b, b_2, \dots, b_k)$

$m := m + 1$

**return**  $Gen(\tilde{C}, g', \varepsilon, \delta, n, v, h, hash)$

// all estimates are lower than  $(1 - \frac{3}{4k})\varepsilon$

Hard code  $\pi_1, \dots, \pi_{m-1}$  into the circuit  $D$ .

**return**  $D^{\tilde{C}}$

For  $k = 1$  the function  $g : \{0, 1\} \rightarrow \{0, 1\}$  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 at 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  $(q, y_1, \dots, y_k) := \tilde{C}(x^{(k)}, \rho)$  let  $c_i := \Gamma_V^i(q, y_i)$ . We define the surplus:

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

The surplus  $S_{\pi^*, b}$  tells us how good  $\tilde{C}$  performs when the first randomness  $\pi_1$  of the poser is fixed, and the fact whether  $\tilde{C}$  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  $\tilde{S}_{\pi^*, b}$  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  $\tilde{S}_{\pi^*, b}$  returned by *EvaluateEstimate* 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  $\tilde{S}_{\pi^*, b}^i = 1$ . Then using union bound and similar argument as in Lemma 1.7 we obtain that  $\frac{\varepsilon^2 \log(n)}{16k^2} \sum_{i=1}^{\frac{16k^2}{\varepsilon^2} \log(n)} \tilde{S}_{\pi^*, b}^i$  differs from  $\mathbb{E}[g(b, c_2, \dots, c_k)]$  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.  $\square$

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

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

$$\begin{aligned} & \Pr_{u \leftarrow \mu_\delta^k} [g(1, u_2, \dots, u_k) = 1] - \Pr_{u \leftarrow \mu_\delta^k} [g(0, u_2, \dots, u_k) = 1] = \\ & \Pr_{\pi^{(k)}, \rho} [g(1, c_2, \dots, c_k) = 1 \mid \pi_1 = \pi^*] - \Pr_{\pi^{(k)}, \rho} [g(0, c_2, \dots, c_k) = 1 \mid \pi_1 = \pi^*] - (S_{\pi^*,1} - S_{\pi^*,0}). \end{aligned} \quad (0.0.6)$$

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

$$\begin{aligned} & \Pr_{u \leftarrow \mu_\delta^k} [g(1, u_2, \dots, u_k) = 1 \wedge g(0, u_2, \dots, u_k) = 0] = \\ & \Pr_{\pi^{(k)}, \rho} [g(1, c_2, \dots, c_k) = 1 \wedge g(0, c_2, \dots, c_k) = 0 \mid \pi_1 = \pi^*] - (S_{\pi^*,1} - S_{\pi^*,0}). \end{aligned} \quad (0.0.7)$$

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

$$\begin{aligned} & \Pr_{\rho} [\Gamma_V(D_2(x^*, \rho)) = 1] = \frac{\Pr_{\rho} [\Gamma_V(D_2(x^*, \rho)) = 1 \mid \pi_1 = \pi^*] \Pr_{\rho} [G_{\pi} \mid \pi_1 = \pi^*]}{\Pr_{u \leftarrow \mu_\delta^k} [G_{\mu}]} \\ & (\Gamma_V, \Gamma_H) := \langle P^{(1)}(\pi^*), D_1(\rho) \rangle_{P^{(1)}} \\ & x^* := \langle P^{(1)}(\pi^*), D_1(\rho) \rangle_{\text{trans}} \\ & - \frac{\Pr_{\rho} [\Gamma_V(D_2(x^*, r)) = 1 \mid \pi_1 = \pi^*] (S_{\pi^*,1} - S_{\pi^*,0})}{\Pr_{u \leftarrow \mu_\delta^k} [G_{\mu}]} \end{aligned} \quad (0.0.8)$$

We can write the first summand of (0.0.8) as

$$\begin{aligned} & \Pr_{\rho} [\Gamma_V(D_2(x^*, \rho)) = 1 \mid \pi_1 = \pi^*] \Pr_{\pi^{(k)}} [G_{\pi} \mid \pi_1 = \pi^*] = \\ & \Pr_{\rho} [D_2(x^*, \rho) \neq \perp \mid \pi_1 = \pi^*] \Pr_{\pi^{(k)}} [c_1 = 1 \mid G_{\pi}, \pi_1 = \pi^*] \Pr_{\pi^{(k)}} [G_{\pi} \mid \pi_1 = \pi^*] \end{aligned} \quad (0.0.9)$$

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

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

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

$$\Pr_r [D(x^*, r) = \perp \mid \pi_1 = \pi^*] \leq (1 - \frac{\varepsilon}{6k})^{\frac{6k}{\varepsilon} \log(\frac{6k}{\varepsilon})} \leq \frac{\varepsilon}{6k}. \quad (0.0.11)$$

We conclude that in both cases:

$$\begin{aligned} & \Pr_r[D(x^*, r) \neq \perp \mid \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[c_1 = 1 \mid G_\pi, \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[G_\pi \mid \pi_1 = \pi^*] \\ & \geq \Pr_{\pi^{(k)}}[c_1 = 1 \mid G_\pi, \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[G_\pi \mid \pi_1 = \pi^*] - \frac{\varepsilon}{6k}. \end{aligned} \quad (0.0.12)$$

Therefore, we have

$$\begin{aligned} & \Pr_r[D(x^*, r) \neq \perp \mid \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[c_1 = 1 \mid G_\pi, \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[G_\pi \mid \pi_1 = \pi^*] \\ & = \Pr_{\pi^{(k)}}[c_1 = 1 \wedge g(1, c_2, \dots, c_k) = 1 \wedge g(0, c_2, \dots, c_k) = 0 \mid \pi_1 = \pi^*] - \frac{\varepsilon}{6k} \\ & = \Pr_{\pi^{(k)}}[g(c_1, c_2, \dots, c_k) = 1 \wedge g(0, c_2, \dots, c_k) = 0 \mid \pi_1 = \pi^*] - \frac{\varepsilon}{6k} \\ & = \Pr_{\pi^{(k)}}[g(c_1, c_2, \dots, c_k) = 1 \mid \pi_1 = \pi^*] - \Pr_{\pi^{(k)}}[g(0, c_2, \dots, c_k) = 0 \mid \pi_1 = \pi^*] - \frac{\varepsilon}{6k}, \end{aligned}$$

and finally by (0.0.5)

$$\begin{aligned} & \Pr_r[D(x^*, r) \neq \perp \mid \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[c_1 = 1 \mid G_\pi, \pi_1 = \pi^*] \Pr_{\pi^{(k)}}[G_\pi \mid \pi_1 = \pi^*] \\ & = \Pr_{\pi^{(k)}}[g(c_1, c_2, \dots, c_k) = 1 \mid \pi_1 = \pi^*] - \Pr_{\mu_\delta^{(k)}}[g(0, \mu_2, \dots, \mu_k) = 0 \mid \pi_1 = \pi^*] - S_{\pi^*, 0} - \frac{\varepsilon}{6k}. \end{aligned} \quad (0.0.13)$$

Inserting this result into the equation (0.0.8) yields

$$\begin{aligned} & \Pr_{r, \pi}[\Gamma_V(D(x, r)) = 1] = \mathbb{E}_\pi \left[ \Pr_r[D(x, r) = 1 \mid \pi_1 = \pi^*] \right] \\ & = \mathbb{E}_\pi \left[ \frac{\Pr_{\pi^{(k)}}[g(c) = 1 \mid \pi_1 = \pi^*] - \Pr_{\mu_\delta^{(k)}}[g(0, \mu_2, \dots, \mu_k) = 0 \mid \pi_1 = \pi^*] - \frac{\varepsilon}{6k}}{\Pr_{\mu_\delta^k}[G_\mu]} \right] \\ & \quad - \mathbb{E}_\pi \left[ \frac{S_{\pi^*, 0} + \Pr_r[\Gamma_V^{(g)}(D(x^*, r)) = 1 \mid \pi_1 = \pi^*](S_{\pi^*, 1} - S_{\pi^*, 0})}{\Pr_{\mu_\delta^k}[G_\mu]} \right] \end{aligned} \quad (0.0.14)$$

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

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

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

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

Let us define a set

$$\mathcal{W} = \left\{ \pi : \left( S_{\pi, 0} \leq (1 - \frac{1}{2k})\varepsilon \right) \wedge \left( S_{\pi, 1} \leq (1 - \frac{1}{2k})\varepsilon \right) \right\} \quad (0.0.17)$$

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

$$\begin{aligned} \mathbb{E}_\pi \left[ S_{\pi^*,0} + \Pr_r[\Gamma_V^{(g)}(D(x^*, r)) = 1 \mid \pi_1 = \pi^*](S_{\pi^*,1} - S_{\pi^*,0}) \right] \\ = \mathbb{E}_{\pi \in \mathcal{W}^c} \left[ S_{\pi^*,0} + \Pr_r[\Gamma_V^{(g)}(D(x^*, r)) = 1 \mid \pi = \pi^*](S_{\pi^*,1} - S_{\pi^*,0}) \right] \\ + \mathbb{E}_{\pi \in \mathcal{W}} \left[ S_{\pi^*,0} + \Pr_r[\Gamma_V^{(g)}(D(x^*, r)) = 1 \mid \pi = \pi^*](S_{\pi^*,1} - S_{\pi^*,0}) \right] \end{aligned} \quad (0.0.18)$$

$$\leq \frac{\varepsilon}{6k} + \mathbb{E}_{\pi \in \mathcal{W}^c} \left[ S_{\pi^*,0} + \Pr_r[\Gamma_V^{(g)}(D(x^*, r)) = 1 \mid \pi = \pi^*]\left((1 - \frac{1}{2k})\varepsilon - S_{\pi^*,0}\right) \right] \quad (0.0.19)$$

$$\leq \frac{\varepsilon}{6k} + 1 - \frac{\varepsilon}{2k} = 1 - \frac{\varepsilon}{3k} \quad (0.0.20)$$

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

$$\begin{aligned} \Pr[g(u) = 1] &= \Pr[(g(0, \mu_2, \dots, \mu_k) = 1) \vee (g(1, \mu_2, \dots, \mu_k) = 1 \wedge g(0, \mu_2, \dots, \mu_k) = 0 \wedge \mu_1 = 1)] \\ &= \Pr[g(0, \mu_2, \dots, \mu_k) = 1] + \Pr[g(1, \mu_2, \dots, \mu_k) = 1 \wedge g(0, \mu_2, \dots, \mu_k) = 0] \Pr[\mu_1 = 1] \end{aligned}$$

which yields

$$\Pr_{r,\pi}[D(x, r) = 1] \geq \mathbb{E}_\pi \left[ \frac{\Pr_{\pi^{(k)}}[g(c) = 1 \mid \pi_1 = \pi^*] - \Pr_{\mu_\delta^{(k)}}[g(0, \mu_2, \dots, \mu_k) = 0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{\mu_\delta^k}[G_\mu]} \right]$$

Using the assumptions of Lemma 1.6, we get

$$\begin{aligned} \Pr_{r,\pi}[\Gamma_V(D(x, r)) = 1] &\geq \frac{\Pr_{\mu_\delta^{(k)}}[g(\mu) = 1] + \varepsilon + \Pr_{\mu_\delta^{(k)}}[g(0, \mu_2, \dots, \mu_k) = 0] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{\mu_\delta^k}[G_\mu]} \\ &\geq \frac{\varepsilon + \delta \Pr_{\mu_\delta^{(k)}}[G_\mu] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{\mu_\delta^k}[G_\mu]} \geq \delta + \frac{\varepsilon}{6k} \end{aligned} \quad (0.0.21) \quad \square$$

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

<b>E</b>
<b>Oracle:</b> Circuit $D$ from Lemma 1.6
<b>Input:</b> A bitstring $\rho \in \{0, 1\}^*$
Run circuit $(q, y) = D(\rho)$
<b>if</b> $(q, y) \neq \perp$ <b>then</b>
make a verification query $(q, y)$

The circuit  $E$  is output by the following algorithm  $Gen$ .

<b>Gen</b>
<b>Oracle:</b> $C, P^{(1)}, g$
<b>Input:</b> $\varepsilon, \delta, n, h, v$
Let $\mathcal{H}$ be a set of pairwise independent hash functions $Q \rightarrow \{0, 1, \dots, 2(h + v) - 1\}$
$hash := \mathbf{FindHash}(\mathcal{H}, h + v)$
$D := Gen(C, g, \varepsilon, \delta, n, h, v, hash)$
<b>return</b> $D^{P^{(1)}, C, hash}(\rho)$

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  $\tilde{C}$  with respect to function  $hash$  is at least

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

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

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

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