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 δ . 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 Γ_V and a hint circuit Γ_H . The algorithm $S_1(\rho)$ produces no output. The circuit Γ_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 Γ_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 Γ_V and Γ_H . The execution of S_2 with the input x and the randomness fixed to ρ is denoted by $S_2(x,\rho)$. The queries of S_2 to Γ_V are called verification queries, and to Γ_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 Γ^i_V, Γ^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 π , 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 π , ρ . 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 ε, δ, 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 Γ_V Γ_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 $\pi^{(k)}$, ρ 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 γ , asking at most h hint queries and v verification queries. There exists a probabilistic algorithm **FindHash** that takes as input: parameters γ , 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 $\pi^{(k)}$, ρ 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 ρ 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, γ, n)

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

Input: Parameters h, v, γ, 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 $hash \in \mathcal{H}$

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

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

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

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

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

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 get

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

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

We define the following circuit C_2 :

$$\textbf{Circuit}\ \, \widetilde{C}_{2}^{\Gamma_{V}^{(g)},\Gamma_{H}^{(g)},C_{2},hash}(x,\rho)$$

Oracle: $\Gamma_V^{(g)}, \Gamma_H^{(k)}, hash, C_2$ Input: A transcript x, a bitstring ρ . **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

```
\begin{array}{c} \textbf{if} \ q \in P_{hash} \ \textbf{then} \\ \qquad \qquad \textbf{return} \ \bot \\ \textbf{else} \\ \qquad \qquad \text{answer the query using } \Gamma_H^{(k)}(q) \\ \\ \textbf{if} \ C_2 \ \text{asks a verification query } (q,y_1,\ldots,y_k) \ \textbf{then} \\ \qquad \qquad \textbf{if} \ q \in P_{hash} \ \textbf{then} \\ \qquad \qquad \text{ask a verification query } (q,y_1,\ldots,y_k) \\ \textbf{else} \\ \qquad \qquad \qquad \text{answer the verification query with } 0 \\ \\ \textbf{return} \ \bot \end{array}
```

We define a new solver circuit $\widetilde{C} = (C_1, \widetilde{C}_2)$ that in the first phase uses the circuit C_1 and in the second phase the circuit \widetilde{C}_2 .

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

$$\Pr_{\pi^{(k)},\rho}[Canonical Success^{P^{(g)},C,hash}(\pi^{(k)},\rho)=1] \leq \Pr_{\pi^{(k)},\rho}[Canonical Success^{P^{(g)},\widetilde{C},hash}(\pi^{(k)},\rho)=1]$$

Proof. We fix $\pi^{(k)}$, ρ . If C succeeds canonically then also \widetilde{C} succeeds canonically. Using this observation, we conclude that

$$\begin{split} \Pr_{\pi^{(k)},\rho} \left[Canonical Success^{P^{(g)},C,hash}(\pi^{(k)},\rho) = 1 \right] \\ &= \underset{\pi^{(k)},\rho}{\mathbb{E}} \left[Canonical Success^{P^{(g)},C,hash}(\pi^{(k)},\rho) = 1 \right] \\ &\leq \Pr_{\pi^{(k)},\rho} \left[Canonical Success^{P^{(g)},\widetilde{C},hash}(\pi^{(k)},\rho) = 1 \right] \end{split}$$

From a circuit C we can build a circuit \widetilde{C} that asks at most one verification query (q, y_1, \ldots, y_k) such that $q \in P_{hash}$, and every hint query on q is such that $q \notin P_{hash}$. Furthermore, we write $(q, y_1, \ldots, y_k) := \widetilde{C}_2(x, \rho)$ to denote the verification query (q, y_1, \ldots, y_k) asked by \widetilde{C}_2 . If \widetilde{C}_2 does not ask a verification query we write $\bot := \widetilde{C}_2(x, \rho)$.

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, with oracle access to: $P^{(1)}$, a monotone function $g: \{0,1\}^{(k)} \to \{0,1\}$, a solver circuit C for $P^{(g)}$ and a function hash: $Q \to \{0,\ldots,2(h+v)-1\}$. Additionally, Gen takes as input parameters ε,δ,n , the number of verification queries v and hint queries h asked by C, the number of puzzles to solve k, and outputs a solver circuit D for $P^{(1)}$ as in Definition 1.1 such that the following holds: If C is such that

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

then D satisfies almost surely

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

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

Proof. First we define the following procedure that returns an estimate for 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 $b \in \{0,1\}$, parameters k, ε

Output: An estimate $\widetilde{g} \in [0, 1]$.

For
$$i := 1$$
 to $\frac{16k^2}{\varepsilon^2} \log(n)$ do: $(b_2, \dots, b_k) \leftarrow \mu_{\delta}^{(k-1)}$ $g_i := g(b, b_2, \dots, b_k)$ 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 the 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 $|\widetilde{g} - \Pr_{(u_2,\dots,u_k) \leftarrow \mu_{\delta}^k}[g(b,u_2,\dots,u_k) = 1]| \leq \frac{\varepsilon}{4k}$ almost surely.

Proof. We define binary random variable K_i for the event that $g_i = 1$. By Chernoff bound we get

$$\Pr\left[\left|\frac{\varepsilon^2 \log(n)}{16k^2} \sum_{i=1}^{\frac{16k^2}{\varepsilon^2} \log(n)} \widetilde{g}_i - \mathbb{E}[K_i]\right| \ge \frac{\varepsilon}{4k}\right] \le 2e^{\log(n)/3}.$$

Next we define a procedure **EvalutePuzzles**($\pi^{(k)}, \rho$) that outputs a tuple indicating puzzles are solved successfully in the random experiment $CanonicalSuccess^{P^{(g)}, \tilde{C}, hash}(\pi^{(k)}, \rho)$.

```
\mathbf{EvaluatePuzzles}^{P^{(1)},\widetilde{C},hash}(\pi^{(k)},\rho)
```

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

Input: Bitstrings $\pi^{(k)}$, ρ .

return (c_1,\ldots,c_k)

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

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

$$(q, y^{(k)}) := \widetilde{C}_2^{\Gamma_V^{(g)}, \Gamma_H^{(k)}, C, hash}(x, \rho)$$

$$&\mathbf{for}\ i := 1\ \text{to}\ k\ \mathbf{do:} \qquad //\text{simulate}\ k\ \text{rounds of sequential interaction}$$

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

$$&\mathbf{for}\ i := 1\ \text{to}\ k\ \mathbf{do:}$$

$$&\quad c_i := \Gamma_v^i(q, y_i) \end{aligned}$$

The procedure **EstimateSurplus** 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.

$\mathbf{EstimateSurplus}^{P^{(1)},C,hash}(\pi^*,b)$

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

Input: A bistring π^* , a bit b, an integer k.

Output: A circuit D.

```
\widetilde{g}_b := \mathbf{EvaluteFunctionProbability}^g(b, \varepsilon, \delta)
For i := 1 to \frac{16k^2}{\varepsilon^2} \log(n) do:
           (\pi_{m+1},\ldots,\pi_k) \stackrel{\$}{\leftarrow} \{0,1\}^{(k-m-1)n}
           \rho \stackrel{\$}{\leftarrow} \{0,1\}^*
           (c_1,\ldots,c_k) := \mathbf{EvalutePuzzles}^{P^{(1)},C,hash}(\pi_1,\ldots,\pi_m,\pi^*,\ldots,\pi_k,\rho)
           \widetilde{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)}\widetilde{S}^i_{\pi^*,b}-\widetilde{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 π_1, \ldots, π_{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 Γ_V^*, Γ_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 Γ_V^* and Γ_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:

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

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) := \widetilde{C}^{\Gamma_V^{(g)}, \Gamma_H^{(k)}, C, hash}((x_1, \dots, x_{m-1}, x^*, \dots, x_k), \rho)

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

return (q, y^*)
```

```
Algorithm Gen^{C,P^{(1)},g,hash}(\varepsilon,\delta,n,v,h,k)
Oracle: P^{(1)}, C, g, hash
Input: \varepsilon, \delta, n, v, h, k
Output: D
for i := 1 to \frac{6k}{\epsilon} \log(n) do:
      \pi^* \stackrel{\$}{\leftarrow} \{0,1\}^n
      \widetilde{S}_{\pi^*,0} := \mathbf{EstimateSurplus}^{P^{(1)},C,hash}(\pi^*,0)
      \widetilde{S}_{\pi^*,1} := \mathbf{EstimateSurplus}^{P^{(1)},C,hash}(\pi^*,1)
      if \exists b \in \{0,1\} : \widetilde{S}_{\pi^*,b} \geq (1-\frac{3}{4k})\varepsilon then
             Let C'_1 simulate first round of interaction between C_1 and P^{(g)} using P^{(1)}(\pi^*),
             then use C_1 for remaining k-1 rounds.
             C' := (C_1, ', C_2)
             g'(b_2,\ldots,b_k) := g(b,b_2,\ldots,b_k)
             return Gen^{C',P^{(1)},g',hash}(\varepsilon,\delta,n,v,h,k-1)
// all estimates are lower than (1-\frac{3}{4k})\varepsilon
return D^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 simply uses the circuit \widetilde{C} . In case g is a constant function the statement is vacuously true.

For fixed $\pi^{(k)}$, ρ 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, let (Γ_V^i, Γ_H^i) be the verification and hint circuits generated in the *i*-th round of the interaction between $P^{(g)}(\pi^{(k)})$ and $C_1(\rho)$. Finally, for $(q, y_1, \dots, y_k) := \widetilde{C}_2(x^{(k)}, \rho)$ we denote $c_i := \Gamma_V^i(q, y_i)$. We define the surplus:

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

The surplus $S_{\pi^*,b}$ tells us how good \widetilde{C} performs when the bitstring π_1 is fixed to π^* , and the fact whether \widetilde{C} succeeds in solving the first puzzle defined by $P^{(1)}(\pi_1)$ is disregarded. Instead, the bit b is used as the input on the first position of the function g.

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

Proof. We use union bound and similar argument as in Lemma 1.7 which yields 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.

In case Gen manages to find an estimate that satisfies $\widetilde{S}_{\pi^*,b} \geq (1 - \frac{3}{4k})\varepsilon$ we define a monotone function $g'(b_2, \ldots, b_k) := g(b, b_2, \ldots, b_k)$, and a circuit $\widetilde{C}' = (C'_1, \widetilde{C}'_2)$, where C'_1 first simulates the interaction between C_1 and $P^{(1)}(\pi^*)$, and then interacts with $P^{(1)}$.

The circuit \widetilde{C} satisfies the conditions of Lemma 1.6 for the remaining k-1 puzzles and we recurse using g' and \widetilde{C}' .

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 δ . 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 δ . We now show that this intuition is indeed correct. For a fixed π^* 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 $g(1, u_2, \ldots, u_k) = 1 \land g(0, u_2, \ldots, u_k) = 0$, and correspondingly $G_{\pi^{(k)}} := g(1, c_2, \ldots, c_k) = 1) \land (g(0, c_2, \ldots, c_k) = 0$. From (0.0.7) for $\pi = \pi^*$ fixed we obtain

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

$$(0.0.8)$$

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_{π} 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

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

$$\ge \frac{\varepsilon + \delta \Pr_{\mu_{\delta}^{(k)}}[G_{\mu}] - (1 - \frac{1}{6k})\varepsilon}{\Pr_{\mu_{\delta}^{(k)}}[G_{\mu}]} \ge \delta + \frac{\varepsilon}{6k} \qquad (0.0.21)$$

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

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