

# Proving Computation Integrity

CS 276: Introduction to Cryptography

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- 1 NIZKs from Trapdoor Permutations
- 2 NIZK for NP: Graph Hamiltonicity

## Definition 1 (NIZK Proof System)

A NIZK for language  $L$  is efficient algorithms  $(K, P, V)$ :

- 1 **Key generation:**  $\sigma \leftarrow K(1^k)$  generates the CRS.
- 2 **Prover:**  $\pi \leftarrow P(\sigma, x, \omega)$  produces a proof (given witness  $\omega$ ).
- 3 **Verifier:**  $V(\sigma, x, \pi) \in \{0, 1\}$  accepts or rejects.

## Completeness

$\forall x \in L, \forall \omega \in R_L(x)$ :

$$\Pr[\sigma \leftarrow K(1^k), \pi \leftarrow P(\sigma, x, \omega) : V(\sigma, x, \pi) = 1] = 1.$$

# NIZK: Properties

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## Non-adaptive soundness

$\forall x \notin L:$

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## Adaptive soundness (stronger)

$$\Pr[\sigma \leftarrow K(1^k) : \exists (x, \pi) \text{ s.t. } x \notin L \text{ and } V(\sigma, x, \pi) = 1] = \text{negl}(k).$$

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# NIZK: Zero-Knowledge Property

## Non-adaptive zero-knowledge

$\exists$  PPT simulator  $S$  such that  $\forall x \in L, \omega \in R_L(x)$ , the following are computationally indistinguishable:

### Real:

- 1  $\sigma \leftarrow K(1^k)$
- 2  $\pi \leftarrow P(\sigma, x, \omega)$
- 3 Output  $(\sigma, \pi)$

### Simulated:

- 1  $(\sigma, \pi) \leftarrow S(1^k, x)$
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## Key point

The simulator produces **both**  $\sigma$  and  $\pi$  together. If  $S$  could not generate  $\sigma$ , the definition would be trivial (verifier could run  $S$  itself instead of interacting with  $P$ ).

## Adaptive zero-knowledge

$\exists$  PPT simulator  $(S_1, S_2)$  such that  $\forall$  PPT  $\mathcal{A}$ , the following are computationally indistinguishable:

### Real:

- 1  $\sigma \leftarrow K(1^k)$
- 2  $(x, \omega) \leftarrow \mathcal{A}(\sigma)$  with  $(x, \omega) \in R_L$
- 3  $\pi \leftarrow P(\sigma, x, \omega)$
- 4 Output  $(\sigma, x, \pi)$

### Simulated:

- 1  $(\sigma, \tau) \leftarrow S_1(1^k)$
- 2  $(x, \omega) \leftarrow \mathcal{A}(\sigma)$
- 3  $\pi \leftarrow S_2(\sigma, x, \tau)$
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Here  $\tau$  is local state (“trapdoor”) passed from  $S_1$  to  $S_2$ : the simulator sets up  $\sigma$  in a way that lets it later produce fake proofs for any  $x$  chosen by  $\mathcal{A}$ .

# Trapdoor One-Way Permutations

## Definition 2 (Trapdoor OWP)

A collection  $\{f_i : D_i \rightarrow D_i\}_{i \in I}$  where  $D_i \subset \{0, 1\}^{|i|}$  with:

- 1  $\exists$  PPT  $G$ :  $G(1^k)$  outputs  $(i, t_i)$  where  $i \in I \cap \{0, 1\}^k$ .
- 2 Easy to sample from  $D_i$  given  $i$ .
- 3  $f_i$  is easy to compute but hard to invert.
- 4  $f_i$  is a permutation.
- 5  $\exists$  PPT  $A$ :  $A(i, y, t_i) \in f_i^{-1}(y)$  (efficient inversion with trapdoor  $t_i$ ).

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## Example: RSA

$G(1^k) = ((N, e), d)$  with  $N = pq$  (primes),  $\gcd(e, \phi(N)) = 1$ ,  $d = e^{-1} \bmod \phi(N)$ .

$$F_{N,e}(x) = x^e \bmod N, \quad A((N, e), y, d) = y^d \bmod N.$$

Correctness:  $(x^e)^d = x \pmod{N}$ .

## Definition 3 (Hidden-Bit Model (HBM))

A NIZK in the HBM for statement  $x$  (witness  $\omega$ ) is  $(K_H, P_H, V_H)$ :

- 1  $r \leftarrow K_H(1^k)$ : generates a hidden random string ( $\ell$  bits).
- 2  $(I, \phi) \leftarrow P_H(r, x, \omega)$ : prover reads all of  $r$ , selects indices  $I \subseteq [\ell]$  to reveal, and produces proof  $\phi$ .
- 3  $V_H(I, \{r_i\}_{i \in I}, x, \phi)$ : verifier sees only revealed bits, accepts or rejects.

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## Intuition

The prover can selectively reveal parts of a random string — like clouds obscuring a random sky, where the prover chooses which clouds to clear.

## Theorem 4

*Given a NIZK  $(P_H, V_H)$  in the HBM, we can construct a NIZK  $(P, V)$  in the CRS model using trapdoor one-way permutations.*

# HBM $\Rightarrow$ Standard Model via Trapdoor OWPs

## Theorem 4

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## Setup

Let  $\mathcal{F}$  be a family of trapdoor OWPs with hard-core bit  $B(\cdot)$ . CRS:  $\sigma = \sigma_1 \cdots \sigma_\ell$  (each  $\sigma_i \in \{0, 1\}^k$ ). Amplify HBM soundness to  $\leq 2^{-2k}$  by repetition.

## HBM $\Rightarrow$ CRS: Construction

### Prover $P(\sigma, x, \omega)$

- 1 Sample trapdoor OWP:  $(f, f^{-1}) \leftarrow \mathcal{F}(1^k)$ .
- 2 Compute  $\alpha_i = f^{-1}(\sigma_i)$  for all  $i \in [\ell]$ .
- 3 Compute hidden bits  $r_i = B(\alpha_i)$  for all  $i \in [\ell]$ . Let  $r := r_1 \cdots r_\ell$ .
- 4 Run the HBM prover:  $(I, \phi) \leftarrow P_H(r, x, \omega)$ .
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## Verifier $V(\sigma, x, f, I, \{\alpha_i\}_{i \in I}, \phi)$

- 1 Confirm  $f \in \mathcal{F}$  and  $f(\alpha_i) = \sigma_i$  for all  $i \in I$ .
- 2 Compute revealed bits  $r_i = B(\alpha_i)$  for all  $i \in I$ .
- 3 Output  $V_H(I, \{r_i\}_{i \in I}, x, \phi)$ .

## HBM $\Rightarrow$ CRS: Completeness and Soundness

### Completeness

Honest prover:  $\alpha_i = f^{-1}(\sigma_i)$  uniformly random (since  $f$  is a permutation), so  $r_i = B(\alpha_i)$  is unbiased. Reduces exactly to HBM distributions.

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## Soundness

For a fixed  $f = f_0$ :  $r_i$ 's are uniformly random, so by HBM soundness:

$$\Pr_{\sigma}[P^* \text{ cheats using } f_0] \leq 2^{-2k}.$$

A cheating  $P^*$  may choose  $f$  adversarially, but  $|\mathcal{F}| = 2^k$ , so by union bound:

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**Important:**  $V$  must verify  $f \in \mathcal{F}$ . If  $f$  is not a permutation,  $f^{-1}(\sigma_i)$  can be multi-valued, letting the prover “explain”  $\sigma_i$  with different  $\alpha_i$ 's having different hard-core bits.

# HBM $\Rightarrow$ CRS: Zero-Knowledge (Hybrid Argument)

## Hybrids (differences from previous in red)

- $H_0$  Real protocol: sample  $\sigma$ , compute  $\alpha_i = f^{-1}(\sigma_i)$ ,  $r_i = B(\alpha_i)$ , run  $P_H$ .
- $H_1$  Sample  $\alpha_i$  randomly, set  $\sigma_i = f(\alpha_i)$ . Identical distribution ( $f$  is a permutation).
- $H_2$  Sample  $r_i \leftarrow \{0, 1\}$  first, then  $\alpha_i$  with  $B(\alpha_i) = r_i$ . Identical (reordering).
- $H_3$  For  $i \notin I$ :  $\sigma_i \leftarrow \{0, 1\}^k$  uniformly (instead of  $f(\alpha_i)$ ). Indistinguishable by hard-core bit security.
- $H_4$  Replace  $P_H$  with HBM simulator  $S_H$ . Indistinguishable by HBM zero-knowledge.

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$H_4$  is the NIZK simulator: it does **not** use the witness. So  $(P, V)$  is zero-knowledge.  $\square$

## $H_2 \approx_c H_3$ : Why Unrevealed Bits Look Random

### Key claim

For a fixed known bit  $r$ :  $\{f(B^{-1}(r))\} \approx_c \{f(B^{-1}(\bar{r}))\}$ , where randomness is over sampling  $B^{-1}$ .

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### Proof

Distinguishing the above  $\equiv$  guessing the hard-core bit from  $f(\alpha)$ . Therefore:

$$\{f(B^{-1}(r))\} \approx_c \mathcal{U}_k$$

since  $\mathcal{U}_k$  can be generated by sampling random  $b$ , then outputting  $f(B^{-1}(b))$  (using that  $f$  is a permutation).

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This justifies replacing  $\sigma_i = f(\alpha_i)$  with random for  $i \notin I$  in hybrid  $H_3$ .

# Graph Hamiltonicity

## Definition 5 (Hamiltonian Graph)

A graph  $G = (V, E)$  with  $|V| = n$  is Hamiltonian if there exist distinct  $x_1, \dots, x_n \in V$  such that  $(x_i, x_{i+1}) \in E$  for all  $i$ , and  $(x_n, x_1) \in E$ .

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## Matrix representations

- **Adjacency matrix**  $M_G$ :  $M_G[i, j] = 1$  iff  $(i, j) \in E$ .
- **Permutation matrix**: boolean  $n \times n$  matrix with exactly one 1 per row and column. There are  $n!$  of these.
- **Cycle matrix**: permutation matrix corresponding to a single Hamiltonian cycle. There are  $(n - 1)!$  of these.

## NIZK for Hamiltonicity in the HBM (Ideal Case)

### Assumption

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The prover knows Hamiltonian cycle  $x_1, \dots, x_n$  in  $G$ .

- 1 Find bijection  $\phi : V \rightarrow \{1, \dots, n\}$  mapping the Hamiltonian cycle onto the cycle of  $M_c$ :  $M_c[\phi(x_i), \phi(x_{i+1})] = 1$  for all  $i$ .
- 2 For all **non-edges**  $(u, v) \notin E$ : reveal  $M_c[\phi(u), \phi(v)] = 0$ .
- 3 Send  $\phi$  and the revealed entries to the verifier.

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- 3 Send  $\phi$  and the revealed entries to the verifier.

## Verifier

Check that all revealed entries are 0. This proves: every non-edge of  $G$  maps to a non-edge of  $M_c$ , so the edges of  $M_c$  (which form a Hamiltonian cycle) must all lie within  $G$ 's edges.

# Ideal Case: Properties

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If  $P$  knows a Hamiltonian cycle, it can always find  $\phi$  mapping the cycle onto  $M_C$ .

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If  $G$  has no Hamiltonian cycle, then for any  $\phi$ , at least one edge of  $M_C$  maps to a non-edge of  $G$ . That entry is 1 but must be revealed as a non-edge  $\Rightarrow$  verifier rejects with probability 1.

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## Zero knowledge

Simulator: sample random bijection  $\phi$ , reveal 0's for all non-edges of  $\phi(G)$ . The verifier only sees which positions are 0 — this reveals nothing beyond the graph structure (which is public).

# From Uniform Random Bits to Cycle Matrices

## Problem

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## Construction

- 1 Use  $\ell = \lceil 3 \log_2 n \rceil \cdot n^4$  random bits. View as  $n^4$  blocks of  $\lceil 3 \log_2 n \rceil$  bits.
- 2 Define  $r'_i = 1$  iff all bits in block  $i$  are 1. Then  $\Pr[r'_i = 1] \approx 1/n^3$ .
- 3 Arrange  $r'$  as an  $n^2 \times n^2$  matrix  $M$ .
- 4 If  $M$  has exactly  $n$  ones, all in distinct rows and columns, extract the  $n \times n$  submatrix  $M_c$ .
- 5 If  $M_c$  is a cycle matrix, use it. Otherwise, reveal all bits and retry.

## Probability Analysis (I)

Pr[exactly  $n$  ones in  $M$ ]

Expected number of ones:  $n^4/n^3 = n$ . By Chebyshev:  $\Pr[|x - n| \geq n] < 1/n$ . Since  $\Pr[x = k]$  is maximized at  $k = n$ :

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Pr[no row/column collision |  $x = n$ ]

After  $k$  ones placed, probability the next avoids all occupied rows and columns is  $(1 - k/n^2)^2$ .

$$\prod_{k=1}^{n-1} \left(1 - \frac{k}{n^2}\right)^2 > 1 - \frac{n-1}{n} = \frac{1}{n}.$$

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Amplification

Repeat  $n^4$  times:

$$\Pr[\text{at least one success}] > 1 - \left(1 - \frac{1}{3n^3}\right)^{n^4} \approx 1 - e^{-n/3} = 1 - \text{negl}(n).$$

# Full HBM Proof System for Hamiltonicity

## Protocol (repeated $n^4$ times)

For each iteration with its portion of the random string  $r$ :

- 1 Attempt cycle matrix extraction from  $r$ .
- 2 **If extraction fails:** reveal all bits in this portion. Verifier checks the extraction indeed fails.
- 3 **If extraction succeeds:** reveal non-used entries of  $M$  (confirming they are 0), then apply the Hamiltonian NIZK using the cycle matrix  $M_c$ .

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## Properties

- **Completeness:** prover succeeds whenever a cycle matrix is found.
- **Soundness:** if any iteration produces a cycle matrix, a cheating prover is caught (probability  $1 - \text{negl}(n)$ ).
- **Zero knowledge:** failed iterations reveal only random bits; successful iterations use the ZK protocol for cycle matrices.

# NIZK for All of NP

## Theorem 6

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## Proof

Hamiltonicity is NP-complete:  $\exists$  poly-time  $f$  with  $x \in L \iff f(x)$  is Hamiltonian.

Prover computes  $f(x)$  and runs the HBM NIZK for Hamiltonicity.  $\square$

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## Theorem 7

*Assuming trapdoor one-way permutations exist, for any  $L \in NP$  there is a NIZK proof in the CRS model for  $L$ .*

# NIZK for All of NP

## Theorem 6

*For any  $L \in NP$ , there is a NIZK proof in the HBM for  $L$ .*

## Proof

Hamiltonicity is NP-complete:  $\exists$  poly-time  $f$  with  $x \in L \iff f(x)$  is Hamiltonian.  
Prover computes  $f(x)$  and runs the HBM NIZK for Hamiltonicity.  $\square$

## Theorem 7

*Assuming trapdoor one-way permutations exist, for any  $L \in NP$  there is a NIZK proof in the CRS model for  $L$ .*

## Proof

Combine: (1) NIZK in HBM for all of NP (via Hamiltonicity), and (2) HBM  $\Rightarrow$  CRS transformation using trapdoor OWPs.  $\square$