# Restricted Non-interactive Zero Knowledge Proofs 

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## Outline

(1) Basic complexity background
(2) Zero knowledge proofs
(3) Known results

4 New results

## Log-space computation

## Definition

Figure 1: Log-space transducer

Read-only input tape


Write-only output tape


Read/write work tape


Has only logarithmic work space, but still relatively powerful

- Closed under function composition
- Closed under complement
- Can run for (at most) polynomial time

L is the class of problems computable by some log-space transducer

## Circuit complexity

## Example

Figure 2: Output node is 1 iff input has at least one 0 and one 1


Circuit (family) classes:

- $\mathrm{NC}^{0}$ : Fan-in 2, constant depth, poly-size
- $\mathrm{AC}^{0}$ : Unbounded fan-in, constant depth, poly-size
- Projection: no gates, only wires from input to output (possibly negated)


## (NISZK) Non-interactive, statistical zero knowledge proof

 Intuition: prover wants to prove input $x$ has some property to the verifier without revealing additional information
## Definition

An NISZK proof system consists of four main parts:

- Prover: very powerful machine, outputs some distribution of proofs given $(x, \sigma)$.
- Verifier: randomized, limited machine which almost always accepts if correct proof is provided for $(x, \sigma)$ (completeness), and almost always rejects on a fake or insufficient proof (soundness)
- Simulator: used to guarantee zero knowledge. Should output a distribution of proofs statistically similar to the prover's distribution
- Reference string $\sigma$ : uniformly random string provided to verifier and prover to use during the proof - "shared randomness"


## Known results

## NISZK $_{L}$

NISZK, but with log-space verifier and simulator. Introduced in 2020 paper by Allender and REU students [All+21].

Was shown to have two complete problems - $S D U_{N C^{0}}$ and $E A_{N C^{0}}$ - which are modified versions of complete problems for NISZK.

## Definition

Promise-EA $\mathrm{NC}^{0}$ : a promise problem over pairs $(C, k)$, where $C$ is an $\mathrm{NC}_{4}^{0}$ circuit and $k$ an integer.

$$
\begin{aligned}
& (C, k) \in E_{Y E S} \text { if the Shannon entropy of } C \text { is } \geq k+1 \\
& (C, k) \in E_{N O} \text { if the Shannon entropy of } C \text { is } \leq k-1 .
\end{aligned}
$$

Not much else was known about this class outside of this paper.

## Perfect randomized encodings

Introduced in [AIK06] to show that $\mathrm{NC}^{0}$ had OWFs iff more powerful classes like NL did.

## Definition

Adapted from [All+21, Definition 27] (perfect randomized encoding):
Let $f:\{0,1\}^{n} \rightarrow\{0,1\}^{(n)}$ be a function. We say that
$\hat{f}:\{0,1\}^{n} \times\{0,1\}^{m} \rightarrow\{0,1\}^{s}$ is a perfect uniform randomized encoding of $f$ with blowup $b$ if it is:

- Input Independent: for every $x, x^{\prime} \in\{0,1\}^{n}$ such that $f(x)=f\left(x^{\prime}\right)$, $\hat{f}\left(x, U_{m}\right)$ and $\hat{f}\left(x^{\prime}, U_{m}\right)$ are identically distributed.
- Output Disjoint: for every $x, x^{\prime} \in\{0,1\}^{n}$ such that $f(x) \neq f\left(x^{\prime}\right)$, supp $\hat{f}\left(x, U_{m}\right) \cap \operatorname{supp} \hat{f}\left(x^{\prime}, U_{m}\right)=\emptyset$.
- Uniform: for every $x \in\{0,1\}^{n}$ the random variable $\hat{f}\left(x, U_{m}\right)$ is uniform over supp $\hat{f}\left(x, U_{m}\right)$.
- Balanced: for every $x, x^{\prime} \in\{0,1\}^{n}$

$$
\left|\operatorname{supp} \hat{f}\left(x, U_{m}\right)\right|=\left|\operatorname{supp} \hat{f}\left(x^{\prime}, U_{m}\right)\right|=b .
$$

## New results

## $\mathrm{NISZK}_{\mathrm{L}}=$ NISZK $_{\mathrm{AC}}$

Since $\mathrm{AC}_{0} \in \mathrm{~L}$, it is clear that $\mathrm{NISZK}_{\mathrm{AC}}^{0} 0$, NISZK $_{\mathrm{L}}$. To prove: NISZK $_{\mathrm{L}} \in \operatorname{NISZK}_{\mathrm{AC}_{0}}$ The Entropy Approximation Problem in $N C_{0}$, denoted $E A_{N C_{0}}$, is a complete problem for NISZK $_{\mathrm{L}}$. We will show that $E A_{N C_{0}} \in$ NISZK $_{A C_{0}}$.

## Proving $E A \in$ NISZK

Transform an instance $(C, k)$ of $E A$ of length $s$ into a distribution $Z$ by

- Taking poly(s) copies of $X$
- hashing
- repeat items 1 and 2

The Problem: $\mathrm{AC}_{0}$ circuits cannot compute hash functions. They can be computed in logspace.

## Solution

## Theorem

There exists a constant $c$ such that, for every $k(n)>n / \operatorname{poly}(\log n)$, and every $r(n) \in[\Omega(\log n), k(n) / c]$, extraction of $(1+c) \cdot r(n)$ bits that are $\epsilon(n)=\frac{1}{n^{3}}$ close to $\{0,1\}^{(1+c) \cdot r(n)}$ in total variation distance is possible in uniform $A C_{0}$ using a seed of length $r(n)$.

## Theorem

There is a polynomial time computable function that takes an instance $(X, m-a)$ of $E A_{N C_{0}}$ and a parameter $s$, and produces a distribution $Z$ on $\{0,1\}^{\prime}$ such that:
(1) If $H(X)>m-a+1$, then $Z$ has statistical difference at most $1 / \operatorname{poly}(s)$ from the uniform distribution on $\{0,1\}^{\prime}$.
(2) If $H(X)<m-a-1$, then the support of $Z$ is at most a $2^{-s}$ fraction of $\{0,1\}^{\prime}$.

## Non Interactive Proof System

(1) Let $Z$ be the distribution on $\{0,1\}^{\prime}$ obtained from $(X, m-a)$ taking $s$ to be the total description length of $(X, m-a)$ in bits. Let $\sigma_{1}, \sigma_{2}, \ldots, \sigma_{s / \log s} \in\{0,1\}^{\prime}$ be the reference strings. The verifier sends $\sigma_{1}, \sigma_{2}, \ldots, \sigma_{s / \log s}$ to the prover.
(2) The prover picks an $i$ at random from $i \in\{0,1, \ldots, s / \log (s)\}$ such that $\left|\left\{r_{i} \mid Z\left(r_{i}\right)=\sigma_{i}\right\}\right| \neq \phi$. Then, after fixing $i$,it picks a random $r_{i}$ from $\left\{r_{i} \mid Z\left(r_{i}\right)=\sigma_{i}\right\}$. It sends $r_{i}$ to the verifier.
(3) $V$ accepts if $Z\left(r_{i}\right)=\sigma_{i}$.

## Simulator

(1) Let $Z$ be obtained from $(X, m-a)$ as in the proof system.
(2) Sample an $i$ uniformly at random from $\{1,2, \ldots, s / \log s\}$.
(3) For this index $i$,sample $r_{i}$ at random, and compute $Z\left(r_{i}\right)=\sigma_{i}$.
(9) For all $j \in\{1,2, \ldots, i-1, i+1, \ldots, s / \log (s)\}$, sample $\sigma_{j}$ uniformly at random.
(6) Output $\left(r_{i}, \sigma_{1}, \ldots \sigma_{i}=Z\left(r_{i}\right), \ldots \sigma_{s / \log (s)}\right)$

## Completeness

## Claim

If $H(X)>m-a+1$, then the verifier accepts with probability $\geq 1-\frac{1}{2^{s}}$.

## Proof.

If $H(X)>m-a+1$, then $T V\left(Z, U_{\{0,1\}^{\prime}}\right) \leq \frac{1}{\text { poly }(s)}$. Thus, for a given $i$

$$
\begin{aligned}
P\left(\exists r_{i} \mid Z\left(r_{i}\right)=\sigma_{i}\right) & \geq 1-P\left(\nexists r_{i} \mid Z\left(r_{i}\right)=\sigma_{i}\right) \\
& \geq 1-\prod_{i=1}^{s / \log (S)} \frac{1}{\operatorname{poly}(s)} \\
& \geq 1-\frac{1}{\operatorname{poly}(s)^{s / \log s}} \\
& \geq 1-\frac{1}{2^{s}}
\end{aligned}
$$

## Soundness

## Claim

If $H(X)<m-a-1$, then the verifier accepts with probability $\leq \frac{1}{2^{s}}$.

## Proof.

If $H(X)<m-a-1$, then, by Lemma 0.2 , the support of $Z$ is at most a $2^{-s}$ fraction of $\{0,1\}^{\prime}$. Thus,

$$
\begin{aligned}
P(\text { verifier accepts }) & =P\left(\exists i \mid Z\left(r_{i}\right)=\sigma_{i}\right) \\
& \leq \sum_{i=1}^{s / \log s} \frac{1}{2^{s}} \\
& \leq \frac{s}{\log s} \cdot \frac{1}{2^{s}} \\
& \approx \frac{1}{2^{s}}
\end{aligned}
$$

## Construction of Distribution $Z$ by $A C_{0}$ Circuits

We let $s \approx m$ be the length of the description of an instance of $(X, k)$ of EA.
Let the threshold for the EA problem be $k=m-a$, where $a$ is a small constant $\in(0,1)$.

## STEP 1: Many copies of distribution $X$

Let $m$ (resp. $n$ ) be the number input (resp. output) gates to $X$. We take $q=4 s m^{2}$ independent copies of $X$ to create distribution $X^{\prime}$. Observe that $H\left(X^{\prime}\right)=q \cdot H(X)$. For every $x, P(X=x) \geq \frac{1}{2^{m}}$. So the flattening lemma implies that $X^{\prime}$ is $\delta=\sqrt{q} \cdot m=2 \sqrt{s} \cdot m^{2}$ flat.
Thus,
(1) if $H(X)>k+1$, then $H\left(X^{\prime}\right)>q \cdot k+q>q k$.
(2) If $H(X)<k-1$, then $H\left(X^{\prime}\right)<q \cdot k-q$.

## STEP 2: Using $A C_{0}$ Randomness Extractor on $X^{\prime}$

Randomness source: $r \in\{0,1\}^{q k / c}$, where $c$ is the constant mentioned in theorem 1. Consider distribution $Y$ on $E\left(x^{\prime}, r\right):\{0,1\}^{q m} \times\{0,1\}^{q k / c} \rightarrow\{0,1\}^{q k+q k / c}$.
(1) If $H(X)>k+1$, then the statistical difference of $Y$ from the uniform distribution over $\{0,1\}^{q k+q k / c}$ is at most $1 /(q m)^{3}$.
(2) If $H(X)<q k-1$, then $H(Y)<q \cdot k-q+q k / c$.

## STEP 3: Many copies of distribution $Y$

(1) If $H(X)>k+1$, then $Y^{\prime}$ has statistical difference at most $q^{\prime} \cdot \frac{1}{(q m)^{3}}=\left(4 s(q m)^{2}\right) \cdot \frac{1}{(q m)^{3}}=\frac{4 s}{q m}=\frac{1}{m^{3}}=\mathcal{O} \frac{1}{\text { poly }(s)}$.
(2) If $H(X)<k-1$, then $H\left(Y^{\prime}\right)<q^{\prime} \cdot H(Y)<q^{\prime} \cdot q \cdot k \cdot\left(\frac{c+1}{c}\right)-q^{\prime} \cdot q$.

## STEP 4: Using $A C_{0}$ randomness extractor on $Y^{\prime}$.

Let $Z$ be the resulting distribution.
$Z\left(r^{\prime}\right)=\left(Y^{\prime}\left(r^{\prime}\right), E\left(r^{\prime}, r\right)\right)$
If $H(X)>k+1$, the statistical distance of $Z$ from uniform is $\approx 1 / \operatorname{poly}(s)$.
If $H(X)<k-1$,
S1: $\left\{\left(Y^{\prime}\left(r^{\prime}\right), E\left(r^{\prime}, r\right)\right) \mid P\left(Y^{\prime}\left(r^{\prime}\right)=y^{\prime}\right) \leq 2^{-N-2 M}\right\}$
S2: $\left\{\left(Y^{\prime}\left(r^{\prime}\right), E\left(r^{\prime}, r\right)\right) \mid 2^{-N-2 M} \leq P\left(Y^{\prime}\left(r^{\prime}\right)=y^{\prime}\right) \leq 2^{-N}\right\}$
S3: $\left\{\left(Y^{\prime}\left(r^{\prime}\right), E\left(r^{\prime}, r\right)\right) \mid P\left(Y^{\prime}\left(r^{\prime}\right)=y^{\prime}\right) \geq 2^{-N}\right\}$
For $i \in\{1,2,3\},\left|S_{i}\right| /|D|=\frac{1}{2^{s}}$, where $D$ is the uniform distribution.
Thus, $Z=S 1 \cup S 2 \cup S 3$
$|Z| /|D|=\frac{1}{2^{s}}$

## NISZK $_{L}=$ NISZK $_{\text {NL }}$

For an arbitary problem $\Pi \in \operatorname{NISZK}_{N L}$, let $(S, P, V)$ be its protocol. Then, we will run the following:
Algorithm 1: $M_{x}(s)$

Data: $x \in \Pi \cup \bar{\Pi}$, $s$ coin flips, $(S, V)$ the simulator and verifier $(\sigma, p) \leftarrow S(x) / /$ under coin flip $s$
if $V(\sigma, p)=1$ then
return $\sigma$;
else
return $0^{|\sigma|}$;
(Convince yourself that $M_{x}(s)$ is NL-computable). Now, on $x \in \Pi$, w.h.p. $M_{x}(s)$ is uniform over its support, and on $x \notin \Pi$ w.h.p. $M_{x}(s)$ has small support. We want to find an encoding of $M_{x}(s)$ in $\mathrm{NC}^{0}$ with similar entropy.

## $\operatorname{NISZK}_{\mathrm{L}}=$ NISZK $_{\mathrm{NL}}$

Idea: [RA97] shows that for any instance of

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\text { PATH }:=\{\langle G, s, t\rangle: s \rightsquigarrow t\}
$$

there is a random weight assignment
$W=\left\{w_{i}: i \in\left[n^{2}\right]\right\}, w_{i}: E(G) \rightarrow\left[4 n^{2}\right]$ such that with probability exponentially close to 1 : $\exists i$ minimal such that $\left(G, w_{i}\right)$ has a unique minimum path from $s$ to $t$.
This reduces any NL problem to a UL problem, where UL is a class already known to have perfect randomized encodings in $\mathrm{NC}^{0}$. This completes the high level idea of our proof.
We also proved that NISZK $_{\mathrm{L}}=$ NISZK $_{\text {PM }}$ using a similar proof, but due to time constraints it won't be explained here.

## Summary

Presented results:

- NISZK $_{A C^{0}}=$ NISZK $_{L}=$ NISZK $_{\text {NL }}$

Going forward:

- NISZK ${ }_{L} \stackrel{?}{=}$ NISZK $_{\text {DET }}$
- OWFs $\in \mathrm{NC}^{0} \stackrel{?}{\leftrightarrow}$ OWFs $\in D E T$


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