

Pseudorandom functions in almost constant depth from low-noise LPN

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Outline

- Introduction to LPN
 - Decisional and Computational LPN
 - Asymptotic hardness of LPN
 - Related work
 - (randomized) PRFs and PRGs
- The road map
 - Overview of the LPN-based randomized PRG in $AC_0(\text{mod } 2)$
 - Bernoulli noise extractor in $AC_0(\text{mod } 2)$
 - Bernoulli-like noise sampler in AC_0
 - randomized PRG \rightarrow randomized PRF
- Conclusion and open problems

Learning Parity with Noise (LPN)

Challenger: $a \xleftarrow{\$} \{0,1\}^{q \times n}$, $x \xleftarrow{\$} \{0,1\}^n$, $e \leftarrow \text{Ber}_\mu^q$, $y := Ax + e$

$$\begin{bmatrix} 1 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 1 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} \pmod{2}$$

Ber_μ : Bernoulli distribution

of noise rate $0 < \mu < \frac{1}{2}$

$\Pr[\text{Ber}_\mu = 1] = \mu$

$\Pr[\text{Ber}_\mu = 0] = 1 - \mu$

Ber_μ^q : q-fold of Ber_μ

(mod 2)

Search LPN: given a and y , find out s

Decisional LPN: distinguish (a, y) from (a, U_q)

[Blum et al.94, Katz&Shin06]: the two versions are (poly) equivalent

In fact: can use $x \leftarrow \text{Ber}_\mu^n$ instead of $x \xleftarrow{\$} \{0,1\}^n$

Hardness of LPN

- worst-case hardness

LPN (decoding random linear code) is NP-hard.

- average-case hardness

- constant noise $\mu = O(1)$

BKW (Blum, Kalai, Wasserman): $t = q = 2^{O(n/\log n)}$

Lyubashevsky's tradeoff: $t = 2^{O(n/\log\log n)}$, $q = n^{1+\epsilon}$

- low noise $\mu = n^{-c}$ (for constant $0 < c < 1$)

$t = 2^{O(n^{1-c})}$, $q = n + O(1)$

- quantum resistance

Related Work

- public-key cryptography from LPN
 - CPA PKE from **low-noise** LPN [Alekhnovich 03]
 - CCA PKE from **low-noise** LPN [Dottling et al.12, Kiltz et al. 14]
 - CCA PKE from **constant-noise** LPN [Yu & J. Zhang C16]
- symmetric cryptography from LPN
 - Pseudorandom generators [Blum et al.93, Applebaum et al.09]
 - Authentication schemes [Hopper&Blum 01, Juels et al. 05,...]
[Kiltz et al.11, Dodis et al.12, Lyu & Masny13, Cash et al.16]
 - Perfectly binding string commitment scheme [Jain et al. 12]
 - Pseudorandom functions from (**low-noise**) LPN?

This work

Main results

- Low-noise LPN implies
 - Polynomial-stretch pseudorandom generators (PRGs) in $AC_0(\text{mod } 2)$
 $AC_0(\text{mod } 2)$: polynomial-size, $O(1)$ -depth circuits with unbounded fan-in \wedge, \vee, \oplus .
 - Pseudorandom functions (PRFs) in $\widetilde{AC}_0(\text{mod } 2)$
 $\widetilde{AC}_0(\text{mod } 2)$: polynomial-size, $\omega(1)$ -depth circuits with unbounded fan-in \wedge, \vee, \oplus
- [Razborov & Rudich 94]: good PRFs do **NOT exist** in $AC_0(\text{mod } 2)$
- More about the PRGs/PRFs:
 - weak seed/key of **sublinear** entropy & security \approx LPN on **linear** size secret
 - uniform seed/key of size λ & security up to $2^{O(\lambda/\log\lambda)}$
- Technical tools:
 - Bernoulli noise extractor in $AC_0(\text{mod } 2)$
Rényi entropy source \rightarrow Bernoulli distribution
 - Bernoulli-like noise sampler in AC_0
Uniform randomness \rightarrow Bernoulli-like distribution
 - Security-preserving and depth-preserving domain extender for PRFs

(randomized) PRGs, PRFs and LPN

- $G_a: \{0,1\}^n \times \{0,1\}^m \rightarrow \{0,1\}^l$ ($n < l$) is **randomized PRG** if

$$(G_a(U_n), a) \sim_c (U_l, a)$$

- $F_{k,a}: \{0,1\}^n \times \{0,1\}^n \times \{0,1\}^m \rightarrow \{0,1\}^l$ is **randomized PRF** if for every PPT A
| $\Pr[A^{F_{k,a}}(a) = 1] - \Pr[A^R(a) = 1] | = \text{negl}(n)$

where $R: \{0,1\}^m \rightarrow \{0,1\}^l$ is a random function.

- Can we obtain (randomized) PRGs and weak PRFs from LPN ?

try eliminating the noise (like LWR from LWE)

$$\underbrace{\langle a_1, x \rangle, \dots, \langle a_i, x \rangle}_{L(\cdot)}, \dots, \underbrace{\langle a_{q-i+1}, x \rangle, \dots, \langle a_q, x \rangle}_{L(\cdot)}$$

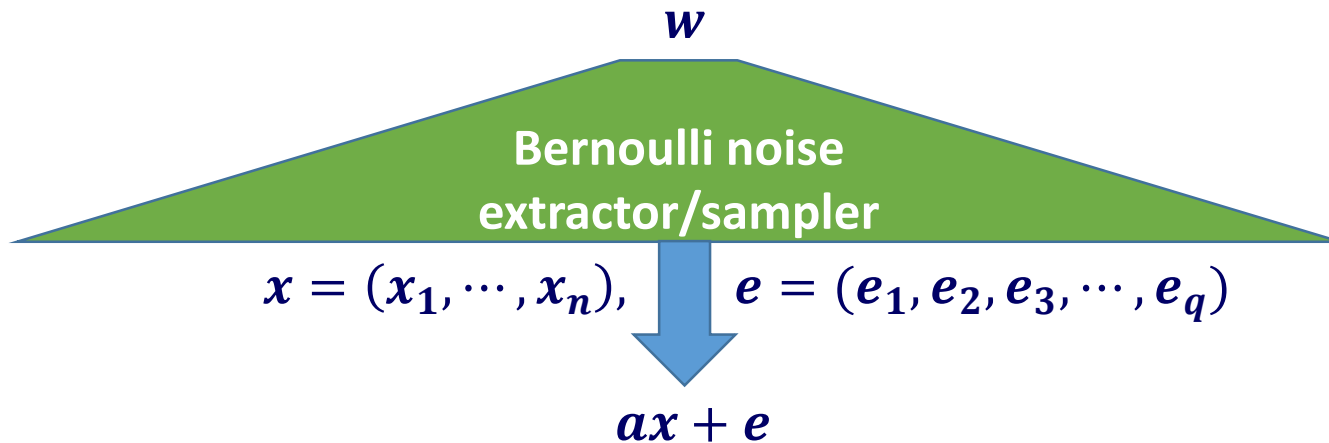
where $L(\cdot)$ is deterministic, $G_a(x) = L(a \cdot x)$, $F_x(a) = L(a \cdot x)$

[Akavia et al.14]: **may not work!**

our approach: convert entropy source w into Bernoulli noise

Overview: LPN-based randomized PRG

- Input: (possibly weak) seed w and public coin a
- Noise sampling: convert (almost all entropy of) w into Bernoulli-like noise (x, e)
- Output: $G_a(w) = ax + e$



- **Theorem:** Assume that the decisional LPN is $(q = (1 + O(1))n, t, \varepsilon)$ -hard on secret of size n and noise rate $\mu = n^{-c}$ ($0 < c < 1$), then G_a is a $(t - \text{poly}(n), O(\varepsilon))$ -hard randomized PRG in $AC_0(\text{mod } 2)$ on
 - weak seed w of entropy $O(n^{1-c} \cdot \log n)$
 - uniform seed w of size $O(n^{1-c} \cdot \log n)$

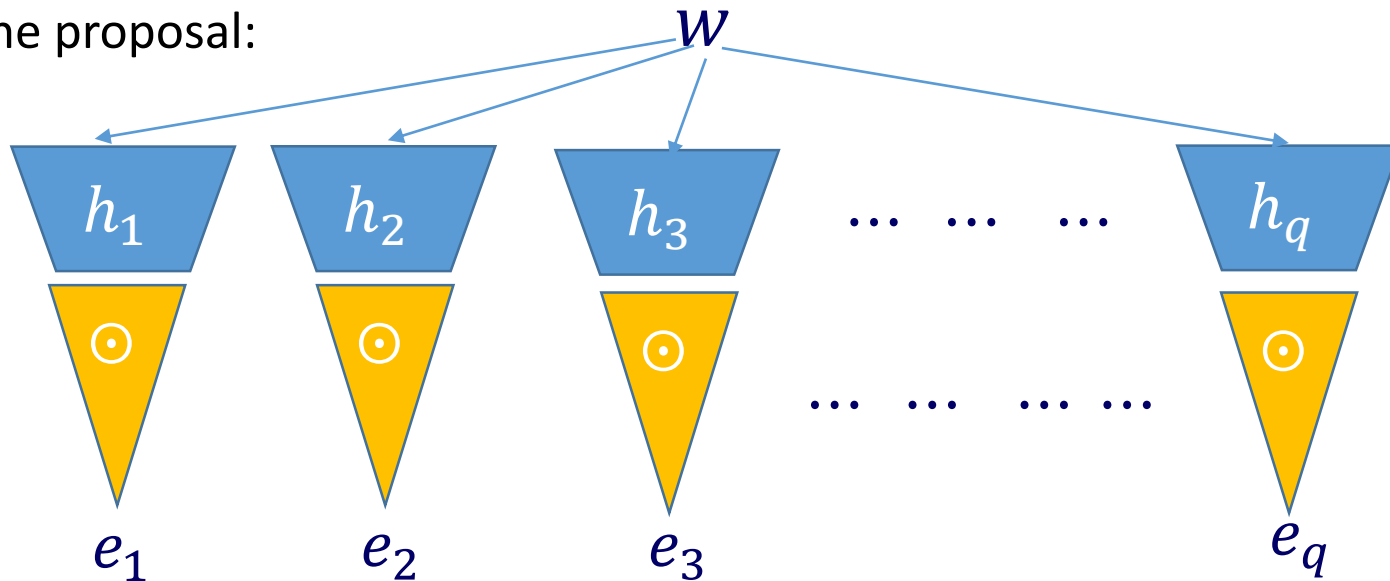
Bernoulli Noise Extractor

- Sample Ber_μ ($\mu = 2^{-i}$): output $\odot(w_1, \dots, w_i) = w_1 w_2 \cdots w_i$
- For $\mu = n^{-c}$ ($i = c \log n$), Shannon entropy $\mathbf{H}(\text{Ber}_\mu) \approx \mu \log(1/\mu)$
 λ random bits $\rightarrow (\lambda/i) = O(\lambda/\log n)$ Bernoulli bits

in theory: λ random bits $\rightarrow \lambda/\mathbf{H}(\text{Ber}_\mu) \approx O(\lambda n^c / \log n)$ Bernoulli bits

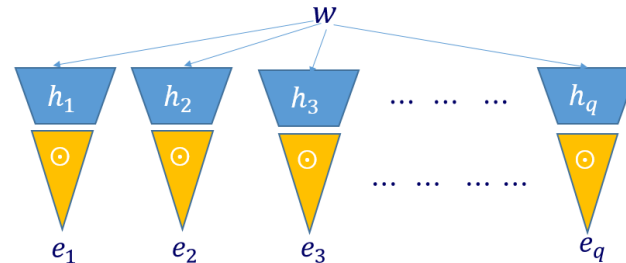
[Applebaum et al.09]: w remains a lot of entropy given the noise sampled

- The proposal:



h_1, h_2, \dots, h_q : 2-wise independent hash functions (randomized by a)

Bernoulli Noise Extractor (cont'd)



- The extractor is in $AC0 \pmod{2}$

- **Theorem (informal):** Let h_1, h_2, \dots, h_q be 2-wise independent hash functions, for any source W of Renyi entropy λ , for any constant $0 < \Delta \leq 1$,

$$\text{Stat-Dist} \left((a, (e_1, \dots, e_q)), (a, \text{Ber}_{\mu}^q) \right) < 2^{\frac{(1+\Delta)H(\text{Ber}_{\mu}^q) - \lambda}{2}} + 2^{-\Delta^2 \mu q / 3}$$

- Parameters: $\mu = n^{-c}$, set $q = \Omega(n)$, $\lambda = (1 + 2\Delta)H(\text{Ber}_{\mu}^q) = \Omega(n^{1-c} \cdot \log n)$

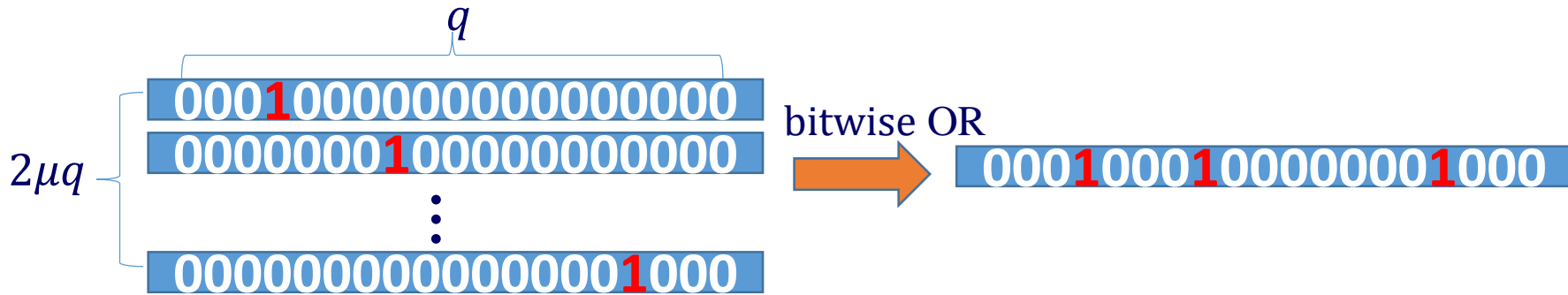
- PRG's stretch: $\frac{\text{output length}}{\text{input length}} = \frac{q-n}{\lambda} = n^{\Omega(1)}$

- Proof:

Cauchy-Schwarz + 2-wise independence like the crooked LHL [Dodis & Smith 05]	+	flattening Shannon entropy Chernoff [HILL99]
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An alternative: Bernoulli noise sampler

- Use **uniform randomness** (~~weak random source~~), and do it in AC_0 (~~$AC_q(\text{mod } 2)$~~)
- The idea: take conjunction of $2\mu q$ copies of random Hamming-weight-1 distributions



- The above distribution (denoted as ψ_μ^q) need $2\mu q(\log q)$ uniform random bits
- Asymptotically optimal: for $\mu = n^{-c}$, $q = \text{poly}(n)$, $2\mu q \log q = O(H(\text{Ber}_\mu^q))$
- PRG: $G_a(w) = ax + e$ by sampling $(x, e) \leftarrow \psi_\mu^{n+q}$ from uniform w
- Theorem: G_a is a randomized PRG of seed length $O(n^{1-c} \log n)$ with comparable security to the underlying standard LPN of secret size n .

Proof. (1) computational LPN \rightarrow computational ψ_μ^{n+q} -LPN

(2) computational ψ_μ^{n+q} LPN \rightarrow decisional ψ_μ^{n+q} -LPN

sample-preserving reduction by [Applebaum et al.07]

Randomized PRGs to PRFs

- Given **randomized PRG** $G_a: \{0,1\}^n \times \{0,1\}^m \rightarrow \{0,1\}^{n^2}$ in $AC_0(\text{mod } 2)$

how to construct a **PRF** in $\widetilde{AC}_0(\text{mod } 2)$?

- ① a PRF of input size $\omega(\log n)$: n -ary GGM tree of depth $d = \omega(1)$

$$G_a(k) \stackrel{\text{def}}{=} \underbrace{G_a^{0 \cdots 00}(k)}_{n \text{ bits}} \underbrace{G_a^{0 \cdots 01}(k)}_{n \text{ bits}} \cdots \underbrace{G_a^{1 \cdots 11}(k)}_{n \text{ bits}}$$

n blocks

$$F_{k,a}(x_1 \cdots x_{d \log n}) \stackrel{\text{def}}{=} G_a^{x_{(d-1)\log n+1} \cdots x_{d \log n}} (\cdots G_a^{x_{\log n+1} \cdots x_{2 \log n}} (G_a^{x_1 \cdots x_{\log n}}(k)) \cdots)$$

- ② Domain extension from $\{0,1\}^{\omega(\log n)}$ to $\{0,1\}^n$ (w. security & depth preserved)

Generalized Levin's trick:

$$F'_{k,a}(x) \stackrel{\text{def}}{=} F_{k_1,a}(h_1(x)) \oplus F_{k_2,a}(h_2(x)) \oplus \cdots \oplus F_{k_l,a}(h_l(x))$$

universal hash functions $h_1, \cdots, h_l: \{0,1\}^n \rightarrow \{0,1\}^{\omega(\log n)}$, $k \stackrel{\text{def}}{=} (k_1, h_1, \cdots, k_l h_l)$

Randomized PRGs to PRFs (cont'd)

Theorem [Generalized Levin's trick]: For random functions $R_1, \dots, R_l : \{0,1\}^{\omega(\log n)} \rightarrow \{0,1\}^n$ and universal hash functions $h_1, \dots, h_l : \{0,1\}^n \rightarrow \{0,1\}^{\omega(\log n)}$, let

$$R'(x) \stackrel{\text{def}}{=} R_1(h_1(x)) \oplus R_2(h_2(x)) \oplus \dots \oplus R_l(h_l(x))$$

Then, R' is $q \left(\frac{q}{n^{\omega(1)}}\right)^l$ -indistinguishable from random function $\{0,1\}^n \rightarrow \{0,1\}^n$ for any (computationally unbounded) adversary making up to q oracle queries.

- See [Bellare et al.99] [Maurer 02][Dottling,Schröder15] [Gazi&Tessaro15]
- Our proof: using the Patarin's H-coefficient technique
- Security is preserved for $q = n^{\omega(1)}$ and $l = O(n/\log n)$

Theorem [The PRF] Assume the decisional LPN is $(q = (1 + O(1))n, t, \varepsilon)$ -hard on secret of size n and noise rate $\mu = n^{-c}$ ($0 < c < 1$), then for any $\omega(1)$ there exists $(q = n^{\omega(1)}, t - \text{poly}(q, n), O(dq\varepsilon))$ -hard randomized PRF $F'_{k,a}$ in $\widetilde{AC}_0(\text{mod } 2)$ of depth $\omega(1)$ on any weak key k of entropy $O(n^{1-c} \cdot \log n)$.

Conclusion and open problems

From low-noise LPN we construct:

- Polynomial-stretch pseudorandom generators (PRGs) in $AC_0(\text{mod } 2)$

- Pseudorandom functions (PRFs) in $\widetilde{AC}_0(\text{mod } 2)$

 - Same (actually better) t/ϵ security than the underlying LPN

 - seed/key of entropy $\lambda = n^{1-c} \log n$ with t/ϵ security up to $2^{O(n^{1-c})} = 2^{O(\lambda/\log \lambda)}$

 - Query complexity $q = n^{\omega(1)}$. $\omega(1)$: depth of the circuit.

- Open problems

 - LPN-based PRFs in constant depth

 - weak PRFs in $AC_0(\text{mod } 2)$

 - PRFs in TC_0

 - More cryptomania objects from LPN?

 - Collision Resistant Hash Function (CRHF)

 - Fully Homomorphic Encryption (FHE)

 - Etc.

Thank you!

