

1- the learning parity with noise problem

Let n and τ in $(0,1)$.

The (search version of) the learning parity with noise problem is to find a in \mathbb{Z}_2^n from arbitrarily many samples of the form $(a_i, \langle a_i, s \rangle + e_i)$ in \mathbb{Z}_2^{n+1} , where the a_i 's are iid uniform and the e_i 's are from a Bernoulli distribution of parameter τ .

The decision version consists in distinguishing such samples from a uniform s , from uniform samples in \mathbb{Z}_2^{n+1} . (give the oracle based definition, give the distinguishing advantage)

Give the search to decision reduction?

Remarks.

- this is like decoding a random linear code in the low-weight error regime, but with an arbitrary code length.
Draw a matrix.
- τ is known.
- $\tau = 0 \Rightarrow$ trivially easy
- $\tau = 1/2 \Rightarrow$ trivially hard
- $\tau \leq 1$ without loss of generality

Given an s , how to check it's the correct one?

Take $b_i = \langle a_i, s \rangle$. It's either uniform, or Bernoulli. How to distinguish? Majority test.

$\Pr [N \geq m\tau(1+\delta) \text{ or } \leq m\tau(1-\delta)] \leq 2\exp(-m\tau\delta^2/3)$

To distinguish with error less than ϵ , take δ such that

$m\tau(1+\delta) \leq m^{1/2} \tau(1-\delta)$. For example, $\delta = (1/2 - \tau)/2$. And m linear in $\ln(1/\epsilon)/(\tau\delta^2)$.

2- Aleknovitch's encryption

Key Gen: $[A, A+e] =: B$
 $pk := B; sk := e \text{ or } s.$

Enc ($M \in \{0,1\}$): $\left\{ \begin{array}{l} \text{sample } b^\perp \text{ unif. in } B^\perp; e' \text{ Bernoulli of } \tau. \\ \text{if } M=0 \text{ } \left\{ \begin{array}{l} b^\perp \\ b^\perp + e' \end{array} \right. \\ \text{if } M=1: v \text{ unif} \end{array} \right.$

Dec: $\langle c, e \rangle = \left\{ \begin{array}{l} \text{unif if } M=1 \\ \langle b^\perp, e \rangle + \langle e, e' \rangle \\ = 0 \qquad \qquad \qquad 0? \end{array} \right.$

There are $\leq \tau m$ indices $i: e_i \neq 0$
 We want that for all of these $e_i = 0$
 This happens with probability $\exp(-\tau \ln(1-\tau)) \leq \exp(-\tau^2 m)$
 over the randomness of b^\perp . \rightarrow Take $\tau = o(1/\sqrt{m})$.

$0 \rightarrow 0$ with $pr \geq 1$
 $1 \rightarrow \left\{ \begin{array}{l} 0 \text{ with } pr = 1/2 \\ 1 \text{ with } pr = 1/2 \end{array} \right.$ } repeat and use an
 error correcting code to
 boost correctness.

Security: \cdot under LNV, B^\perp is ϵ -unif.
 \cdot under LNV for code B^\perp , $b^\perp + e'$ is ϵ -unif. \square

③ The BKW algorithm



- sample lots of $a_i, b_i = \langle a_i, s + e_i \rangle$.
- put them in buckets, according to their first ℓ entries: 2^ℓ buckets
- take a pair of samples $(a_i, b_i), (a_j, b_j)$ of the same bucket, and map them to $(a_i + a_j, b_i + b_j) = (a_i + a_j, \langle a_i + a_j, s + e_i + e_j \rangle + \langle e_i + e_j, s \rangle)$.

This maps L samples in dim n with param τ to $L/2$ $n-\ell$ $2\tau(1-\tau)$.

at a cost of $\leq \max(L, 2^\ell)$

($e_i = 0 \wedge e_j = 1$
or $e_i = 1 \wedge e_j = 0$)

Repeat m/ℓ times. At the last step, we ensure that $a_i + e_i = 0 \dots 0 1$ rather than $0 \dots 0$.

This gives samples of the form $0 \dots 0 1, s_m + e_i, i=1 \dots$

Noise rate at the end: $\tau_{end} = \frac{1}{2} - \frac{1}{2}(1-2\tau)^{m/\ell}$ (proof by induction)

$$\left(2 \cdot \left[\frac{1}{2} - \frac{1}{2}(1-2\tau)^2 \right] \left[1 - \frac{1}{2} + \frac{1}{2}(1-2\tau)^2 \right] \right) = \frac{1}{2} - \frac{1}{2}(1-2\tau)^4$$

We want to use a major vote to distinguish. By what we did earlier, we could do with $\leq \frac{m}{\tau_{end}} \cdot \left(\frac{1}{2} - \tau_{end}\right)^2$

$$\leq \frac{m}{(1 - (1-2\tau)^{m/\ell})} (1-2\tau)^{2m/\ell} \text{ samples}$$

To balance out the costs, we make τ_{end} not too small. \leq what matters is $(1-2\tau)^{2m/\ell}$. (else we could decrease ℓ , repeat more times)

Cost $\leq 2^\ell + (1-2\tau)^{2m/\ell}$. Take $\ell = \frac{m}{\ln(\frac{m}{\tau^2 \ln m})}$ ($\tau = o(1)$)

$$\Rightarrow \frac{O\left(\frac{m}{\ln(m/\ell)}\right)}{\ln(m/\ell)}$$

