

Remarks on Bias Correctors

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Overview

- 1 Bias Correctors
- 2 Some Problems
- 3 Conclusion

- Bias correctors decrease input bias, at the price of compression; fundamental for TRNGs!
- 1-bit output example: celebrated XOR [Dav02]
- Multiple-bit output: non-linear maps [Dic07] or resilient linear codes [Lac08]
- How about optimality of assumptions and bounds?

What is exactly bias?

Fix a distribution $X \in \{0, 1\}^n$ and a candidate $f : \{0, 1\}^n \rightarrow \{0, 1\}^m$ for a bias corrector.

We study one of the following metrics for $Z = f(X)$

- $\text{MAX-BIAS}(Z) = \max_y |\Pr[Z = y] - 2^{-m}|$
- $\text{TOTAL-BIAS}(Z) = \frac{1}{2} \sum_y |\Pr[Z = y] - 2^{-m}|$

For 1-bit Z simplifies to $\text{BIAS}(Z) = |\Pr[Z = 1] - \Pr[Z = 0]| = |\mathbb{E}(-1)^Z|$.

Bias by Fourier Analysis

- Consider linear spaces $\mathbf{F}_2^n \equiv \{0, 1\}^n$ and $\mathbf{F}_2^m \equiv \{0, 1\}^m$
- Compute bias for *all linear combinations of the output f*

$$f \cdot u(x) = \sum f(x)_i \cdot u_i = \bigoplus_{i: u_i \neq 0} f(x)_i$$

$$\Delta(u) = \mathbb{E}_{x \sim X} (-1)^{f \cdot u(x)}$$

- One-dimensional biases $\Delta(u)$ are connected to the original bias [Lac08, Gol95]
- Bias for a single-valued $g : \mathbf{F}_2^n \rightarrow \mathbf{F}_2$ also computed by the Fourier expansion

$$(-1)^g = \sum_I \hat{g}_I \prod_{i \in I} (-1)^{x_i}$$

- Works very well under the *independent bits model*

Bias by Fourier Analysis

The following result shows how to connect single- and multidimensional biases.

Theorem (Multidimensional Output Bias / XOR Lemma)

Let Δ be as before, then

- $\text{MAX-BIAS}(f(X)) \leq \max_u \|\Delta(u)\|_\infty$ [Lac08, Gol95]
- $\text{TOTAL-BIAS}(f(X)) \leq \frac{1}{2} \cdot 2^{m/2} \max_u \|\Delta(u)\|_\infty$ [Gol95]

Compute Bias with Fourier Analysis - Examples

Example (single-bit output)

If X has independent bits each with bias ϵ , then $f(x) = \bigoplus_i x_i$ has bias of $\frac{1}{2} \cdot (2\epsilon)^n$.

Example (multi-bit output from linear codes)

If X has independent bits each with bias ϵ , and $f : \{0, 1\}^n \rightarrow \{0, 1\}^m$ is a linear code with distance d then $|\Delta(u)| \leq \frac{1}{2} \cdot (2\epsilon)^d$ for each u .

Example (resilient codes)

A linear (n, m, d) code is $t = d - 1$ resilient because with $n - t$ unbiased bits the output is unbiased.

Better Bias Analysis

Using total bias and sharper bounds on fourier transforms, one gets better bounds than Lacharme.

Consider inputs with bias $\epsilon = \frac{1}{4}$ and a (n, m, t) -resilient linear code. Then

- MAX-BIAS = 2^{-t} , equivalently min-entropy is $m - \log(1 + 2^{m-t})$
- TOTAL-BIAS = $2^{\frac{m}{2}-t}$, closeness to the uniform distribution (smooth min-entropy)

Comparison

- In both cases resilience t large enough compared to m .
- But the second one preferable for indistinguishability applications (e.g. ciphers) from the theoretical perspective.

Removing Independence

Under the Markov model one can work with the *conditional input bias* defined as

$$\text{BIAS}(X) = \max_{x_{<n}} |\mathbb{E}(-1)^{X_n} | X_{<n} = x_{<n}|$$

and then previous results hold true with $\epsilon = \text{BIAS}(X)$.

Beating XOR Extractors

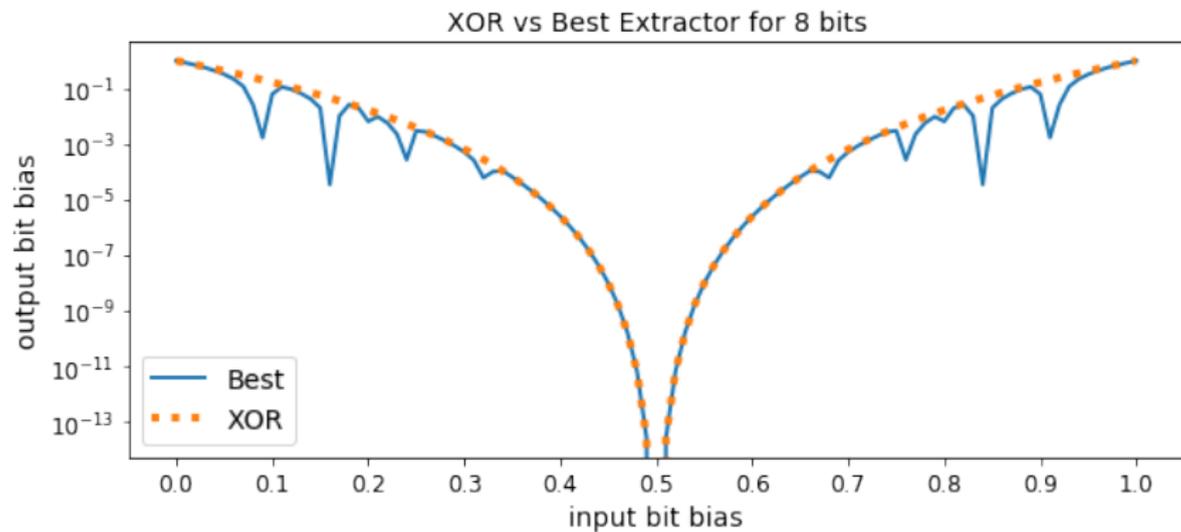
For a given class of distributions (e.g. bias prior region in the IID model), how to build an optimal (min-max) corrector?

Some observations for 1-bit correctors ¹

- XOR is optimal for small bias; for some bias values one can do better!
- Dimensionality reduction: under IID bits and with sufficiently many bias possibilities, the corrector depends on the hamming weight. Search space shrinks from 2^{2^n} to 2^n .
- Dimensionality reduction: under IID bits and symmetric bias prior, the corrector is symmetric w.r.t. the hamming weight. Search space shrinks to $2^{n/2}$
- ...

¹Unpublished work

Beating XOR Extractors



Summary

- Other stochastic models for discrete sources?
- Trade resilience for entropy (to get condensers)?
- Solve min-max for multidimensional outputs?
- ...

References I

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