

Correction of Samplable Additive Errors

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Two Well-Studied Channel Models

- Binary Symmetric Channels (BSC)
 - Each bit is flipped with constant probability p

- Worst-Case (or Adversarial) Channels
 - Only **weight of error vectors** is restricted

Two Well-Studied Channel Models

■ Binary Symmetric Channels (BSC)

- Each bit is flipped with constant probability p
- Introduce errors by simply flipping a coin
(Low-cost computation)

■ Worst-Case (or Adversarial) Channels

- Only weight of error vectors is restricted
- Introduce errors based on the full knowledge
(High-cost computation)

Computationally-Bounded Channels

- Introduced by Lipton (STACS 1994)
- Channels = Polynomial-time bounded algorithms
(with error weight restriction)
 - Lie between BSC and Worst-Case Channels
- Related results
 - PKI setting [Micali et al. (TCC 2005)]
 - Without shared randomness or PKI [Guruswami, Smith (FOCS 2010)]

This Work

- Introduce “**Samplable Additive-Error Channels**” as another intermediate model
- Investigate the possibilities and limitations for the reliable communication over this channel

Samplable Distributions

- Distribution Z over $\{0,1\}^n$ is *samplable*
 - ↔ Exist probabilistic polynomial-time algorithm S s.t. $S(1^n)$ is distributed according to Z
- Related work on samplable distributions
 - Data compression [GS91, Wee04, TVZ05]
 - Randomness extraction [TV00, Vio11, DW12, DRV12, DPW14]

Samplable Additive-Error Channels

- Additive channel $C^Z : \{0,1\}^n \rightarrow \{0,1\}^n$ is defined by a samplable distribution Z s.t.

$$C^Z(x) = x + z, z \sim Z$$

- Error vector z does not depend on x
- Dist. Z does not depend on the code
 - Conversely, the code can depend on Z
- Weight of z is not bounded

Reasons for Introducing This Model

- Error distr. is same for every code/codeword
→ Error-correction problem is simple
- Error distribution is samplable
→ Computational constraints on error vectors can help error correction?
- Errors without weight restriction
→ High-weight errors are correctable if they have “nice structure” ?

What Z is correctable?

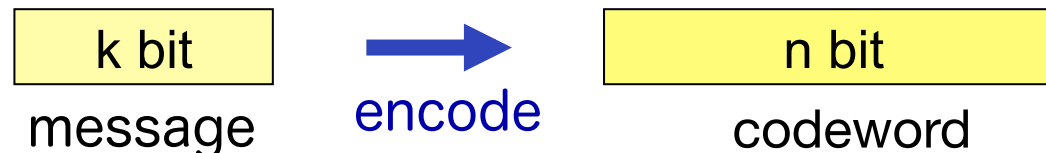
Correctability of Samplable Additive Errors

- Use **Shannon entropy** $H(Z)$ as criterion

$$H(Z) = \mathbb{E}_z \left[\log \frac{1}{p_Z(z)} \right] = \sum_{z \in \text{supp}(Z)} p_Z(z) \log \frac{1}{p_Z(z)}$$

- $H(Z) \in [0, n]$ since Z is over $\{0, 1\}^n$
- $H(Z) = 0 \rightarrow$ easily **correctable**
- $H(Z) = n \rightarrow$ **uncorrectable**

- What **coding rate** $R = k/n$ is achievable?



Observation 1: Z is BSC

- Z can simulate BSC



Theorem 1

$\exists Z$ with $H(Z) = n \cdot h(p)$ ($= \Omega(n)$)
uncorrectable for $R > 1 - h(p)$

Observation 2: Z is pseudo-random

- Z is the output of pseudo-random generator,
→ Not correctable by poly-time algorithms



Theorem 2

- $\exists Z$ with $H(Z) \leq n^\epsilon$, $0 < \epsilon < 1$,
uncorrectable by polynomial-time algorithms
- Assume OWF exists

Observation 3: Z forms a linear subspace

Theorem 3

\forall vector in linear space $Z \subseteq \{0,1\}^n$ of dim. m is **correctable** by a linear code with rate $R = 1 - m/n$

- $H(Z) = m$
- Decoding is efficient

■ Proof sketch:

- For basis $\{z_1, \dots, z_m\}$,
 \exists linear map $T : \{0,1\}^n \rightarrow \{0,1\}^m$ s.t.
 $T(z) = (a_1, \dots, a_m)$ for \forall error vector $z = \sum_i a_i z_i$
- The code = $\{ x : T(x) = 0 \}$

Observation 4: Z is flat

Theorem 4.1

\forall flat Z is **correctable** by linear code with rate $R < 1 - m/n - \Omega(\log(1/\epsilon)/n)$ and error $\epsilon > 0$

- $H(Z) = m$ ($|\text{supp}(Z)| = 2^m$)
- Code is not explicitly given

- **Proof sketch:** Equivalence between linear code ensemble and linear lossless condenser [Cheraghchi (ISIT 2009)]

Theorem 4.2

\forall flat Z is **uncorrectable** for rate $R \geq 1 - m/n + O(1/n)$ and error $\epsilon < 1/2$ ($|\text{supp}(Z)| = 2^m$)

- **Proof sketch:** Need to divide received word space $\{0,1\}^n$ into 2^{Rn} disjoint sets each of size $(1 - \epsilon)2^m$

Observation 5: Uncorrectable Z with low entropy

Theorem 5

$\forall \omega(\log n) < m < n, \exists$ samplable Z with $H(Z) = m$
uncorrectable by “efficient syndrome decoding”
for rate $R > \omega(\log n)/n$

- Assume “oracle access” to some oracle

■ Proof sketch:

- \exists samplable Z with $H(Z) = \omega(\log n)$ not efficiently compressible to length $< n - \omega(\log n)$ [Wee (CCC2004)] (Assuming “oracle access”)
- Z is compressible to length $n(1 - R)$ by linear function $\Leftrightarrow Z$ is correctable with rate R by syndrome decoding [Cair et al. 2004]

Observation 6: Z is small-biased distribution

- Sample space $S \subseteq \{0,1\}^n$ is δ -biased
 $\Leftrightarrow \forall$ non-zero $a \in \{0,1\}^n, |E_{x \sim S}[(-1)^{a \cdot x}]| \leq \delta$

Z is small-biased \Leftrightarrow Indistinguishable from uniform
by linear functions

Theorem 6

$\forall Z$ is **uncorrectable** for rate
 $R > 1 - \Omega(\log(1/\delta) / n)$ with error $\varepsilon < 1/2$
if Z is uniform over δ -biased sample space S

- **Proof sketch:** Z can work as the key of the one-time pad
if message has entropy [Dodis, Smith (TCC2005)]

Corollary 6.1

$\exists Z$ with $H(Z) = m$ **uncorrectable** for
rate $R \geq 1 - m/n + O((\log n)/n)$

Correctability of Samplable Additive Errors (Summary)

$H(Z)$	Correctability	Assump.	References
0	Efficiently correctable		Trivial
$\omega(\log n)$	Efficiently uncorrectable by syndrome decoding for $R > \omega((\log n)/n)$	Oracle access	Theorem 5
n^ε ($0 < \varepsilon < 1$)	Efficiently uncorrectable	OWF	Theorem 2
$n \cdot h(p)$ ($0 < p < 1$)	Uncorrectable for $R > 1 - h(p)$		Theorem 1
$0 \leq m \leq n$	\forall linear space Z of dim. m is correctable for $R \leq 1 - m/n$		Theorem 3
$0 \leq m \leq n$	\forall flat Z is correctable for $R \leq 1 - m/n - \Omega(\log(1/\varepsilon)/n)$		Theorem 4.1
$0 \leq m \leq n$	\forall flat Z is uncorrectable for $R > 1 - m/n + O(1/n)$		Theorem 4.2
$0 \leq m \leq n$	\forall δ -biased Z is uncorrectable for $R \geq 1 - m/n + O((\log n)/n)$		Corollary 6.1
n	Uncorrectable		Trivial

Conclusions

Our Results

- Introduce “Samplable Additive-Error Channels”
- Investigate the correctability

Future Work

- Any practical situations captured by this model (with positive results)?
- More positive results (on more restricted Z ?)
 - Log-space/constant-depth samplable Z
- Prove without assumptions (OWF, oracle access)
 - Or prove the assumption is necessary