

3. Message Attack (ISD)

1. From Generic Decoding to Syndrome Decoding
2. Combinatorial Solutions: Exhaustive Search and Birthday Decoding
3. Information Set Decoding: the Power of Linear Algebra
4. **Complexity Analysis**
5. Lee and Brickell Algorithm
6. Stern/Dumer Algorithm
7. May, Meurer, and Thomae Algorithm
8. Becker, Joux, May, and Meurer Algorithm
9. Generalized Birthday Algorithm for Decoding
10. Decoding One Out of Many

ISD – Complexity Analysis

We will refer to Information Set Decoding (ISD) to designate is a family of algorithms similar to Prange algorithm

All variants of Information Set Decoding repeat a (large) number of times an independent iteration which has

- a constant (expected) cost \mathcal{K}
- a success probability \mathcal{P}
→ an expected number of iteration $\mathcal{N} = 1/\mathcal{P}$

The workfactor is $\mathcal{N} \cdot \mathcal{K}$

ISD – One Solution or All Solutions?

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1. If $\binom{n}{w} < 2^{n-k}$ (i.e. $w < \tau_{\text{GV}}$) there is **exactly** one solution
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First case (the most common) → no difference

Second case → finding only one solution should be easier

(intuitively by a factor $\binom{n}{w}/2^{n-k}$)

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If $N = |\text{CSD}(H, s, w)|$, we have

$$\mathcal{P}_1 = 1 - (1 - \mathcal{P}_\infty)^N \approx \min(1, N\mathcal{P}_\infty) \text{ up to a small constant factor}$$

or simply $\mathcal{P}_1 = N\mathcal{P}_\infty$ if N is not too large (which corresponds to intuition)

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- “ $w < \tau_{\text{GV}}$ ” or “ $w > \tau_{\text{GV}}$ and we want all solutions”
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→ we expect to execute $\mathcal{N}_\infty = 1/\mathcal{P}_\infty$ iterations
- “ $w > \tau_{\text{GV}}$ and we want only one solution”

→ we expect to execute $\mathcal{N}_1 = \mathcal{N}_\infty / N = \frac{2^{n-k}}{\binom{n}{w} \mathcal{P}_\infty}$ iterations

Prange Algorithm – Complexity Analysis

An error pattern is found if it has the following form $e =$

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Total workfactor is

- for all solutions $\text{WF}_{\text{Prange}} = n(n-k) \frac{\binom{n}{w}}{\binom{n-k}{w}}$

- for one solution $n(n-k) \frac{\min(2^{n-k}, \binom{n}{w})}{\binom{n-k}{w}}$

indeed the values are identical when $\binom{n}{w} < 2^{n-k}$

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