# **Code-Based Cryptography**

Message Attacks (ISD)

Nicolas Sendrier



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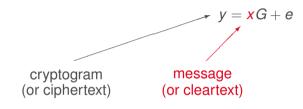
- 1. Error-Correcting Codes and Cryptography
- 2. McEliece Cryptosystem
- 3. Message Attacks (ISD)
- 4. Key Attacks
- 5. Other Cryptographic Constructions Relying on Coding Theory

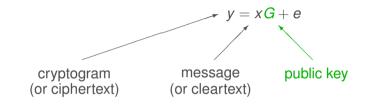
# 3. Message Attack (ISD)

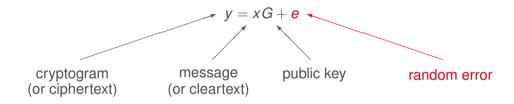
- 1. From Generic Decoding to Syndrome Decoding
- 2. Combinatorial Solutions: Exhaustive Search and Birthday Decoding
- 3. Information Set Decoding: the Power of Linear Algebra
- 4. Complexity Analysis
- 5. Lee and Brickell Algorithm
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$$y = xG + e$$

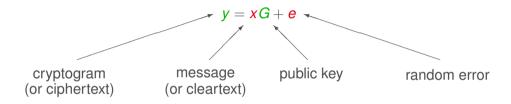






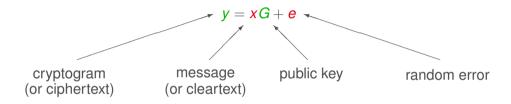


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The adversary knows the cryptogram and the public key and wishes to recover the message (or equivalently the error)

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Only an arbitrary generator matrix is known

 $\rightarrow$  generic decoding problem

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 $G \in \mathbf{F}_q^{k imes n}$  a generator matrix  $\mathcal{C} = \langle G \rangle = \{ xG \mid x \in \mathbf{F}_q^k \}$   $H \in \mathbf{F}_q^{(n-k) \times n}$  a parity check matrix  $\mathcal{C} = \langle H \rangle^{\perp} = \{ \boldsymbol{c} \in \mathbf{F}_q^n \mid \boldsymbol{c} H^T = 0 \}$ 

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$$m{H} \in \mathbf{F}_q^{(n-k) imes n}$$
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#### Generic Decoder:

$$\Phi: \ \mathbf{F}_q^n \times \mathbf{F}_q^{k \times n} \ \rightarrow \ \mathbf{F}_q^k$$
$$(y, G) \ \mapsto \ x$$

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Generic Decoder:

 $\Phi: \mathbf{F}_{q}^{n} \times \mathbf{F}_{q}^{k \times n} \to \mathbf{F}_{q}^{k}$  $\Phi(xG + e, G) = x \text{ if } e \text{ is "small"}$ 

"small" = of small Hamming weight

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 $\begin{array}{rcl} \textbf{Generic Syndrome Decoder:} \\ \Psi: & \textbf{F}_q^{n-k} \times \textbf{F}_q^{(n-k) \times n} & \rightarrow & \textbf{F}_q^n \\ & (s, H) & \mapsto & e \end{array}$ 

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Those two kinds of decoders are equivalent

 $\rightarrow$  we will consider only syndrome decoding

### The Syndrome Decoding Problem

#### Syndrome Decoding Problem

Instance: 
$$H \in \{0, 1\}^{(n-k) \times n}$$
,  $s \in \{0, 1\}^{n-k}$ , an integer  $w > 0$   
Answer:  $e \in \{0, 1\}^n$  such that  $eH^T = s$  and  $wt(e) \le w$ 

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Find *w* columns of *H* adding to *s* (modulo 2)

$$H = \begin{bmatrix} h_1 & h_2 & \cdots & h_n \\ & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\$$

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Find  $1 \le j_1 < j_2 < \cdots < j_w \le n$  such that  $h_{j_1} + h_{j_2} + \cdots + h_{j_w} = s$ 

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$$\xrightarrow{\binom{n}{W}} \frac{\binom{n}{W}}{2^{n-k}}$$
 solutions on average

0

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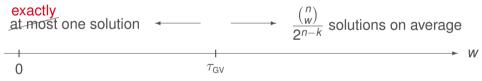
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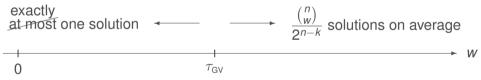
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In cryptanalysis, we only consider situations where  $CSD(H, s, w) \neq \emptyset$ 

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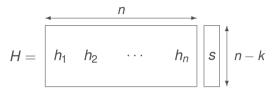
Problem: find *w* columns of *H* adding to *s* (modulo 2)

$$H = \begin{bmatrix} h_1 & h_2 & \cdots & h_n \end{bmatrix} \begin{pmatrix} n & -k & s = \end{bmatrix}$$

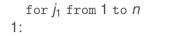
Answer: enumerate all *w*-tuples  $(j_1, j_2, \dots, j_w)$  such that  $1 \le j_1 < j_2 < \dots < j_w \le n$ and check whether  $s + h_{j_1} + h_{j_2} \cdots + h_{j_w} = 0$ 

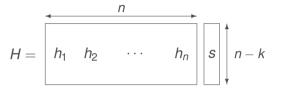
How to enumerate nicely

Enumerate 
$$\{s + eH^T \mid wt(e) = w\} = \{s + h_{j_1} + \dots + h_{j_w} \mid 1 \le j_1 < \dots < j_w \le n\}$$



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for 
$$j_1$$
 from 1 to  $n$   
1:  
for  $j_2$  from  $j_1 + 1$  to  $n$   
2:  
 $H = \begin{bmatrix} n \\ h_1 \\ h_2 \\ \dots \\ h_n \end{bmatrix} \begin{bmatrix} s \\ s \\ n - k \end{bmatrix}$ 

Enumerate  $\{s + eH^T \mid wt(e) = w\} = \{s + h_{j_1} + \dots + h_{j_w} \mid 1 \le j_1 < \dots < j_w \le n\}$ 

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for  $j_W$  from  $j_{W-1} + 1$  to  $n$   
 $H = \begin{bmatrix} n \\ h_1 \\ h_2 \\ \dots \\ h_n \end{bmatrix} \begin{bmatrix} s \\ s \\ n - k \end{bmatrix} \begin{bmatrix} n - k \\ n - k \end{bmatrix}$ 

W:

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for  $j_w$  from  $j_{w-1} + 1$  to  $n$   
 $W$ :  
 $S_w \leftarrow s + h_{j_1} + h_{j_2} + \dots + h_{j_w}$ 

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w:  
 $s_w \leftarrow s + h_{j_1} + h_{j_2} + \dots + h_{j_w}$   
[if  $s_w = 0$  then return  $(j_1, j_2, \dots, j_w)$ ] or [store $(s_w, (j_1, j_2, \dots, j_w))$ ]

Enumerate {
$$s + eH^T$$
 | wt( $e$ ) = w} = { $s + h_{j_1} + \dots + h_{j_w}$  |  $1 \le j_1 < \dots < j_w \le n$ }  
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Total cost is at most  $w \binom{n}{w}$  column additions and  $\binom{n}{w}$  tests

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for  $j_1$  from 1 to  $n$   
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 $\therefore$   
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Total cost is about  $w \binom{n}{W}$  column operations

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Instead, we may keep track of partial sums

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2:  $s_2 \leftarrow s_1 + h_{j_2}$  (=  $s + h_{j_1} + h_{j_2}$ )  
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for  $j_W$  from  $j_{W-1} + 1$  to  $n$   
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for  $j_W$  from  $j_{W-1} + 1$  to  $n$   
W:  $s_W \leftarrow s_{W-1} + h_{j_W}$  (=  $s + h_{j_1} + h_{j_2} + \dots + h_{j_W}$ )  
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Line  $i$  is executed about  $\binom{n}{i}$  times

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Line  $i$  is executed about  $\binom{n}{i}$  times  
 $\rightarrow$  total of about  $\binom{n}{1} + \binom{n}{2} + \dots + \binom{n}{w}$  column additions

Enumerate { $s + eH^T$  | wt(e) = w} = { $s + h_{j_1} + \dots + h_{j_w}$  |  $1 \le j_1 < \dots < j_w \le n$ } for  $j_1$  from 1 to n1:  $s_1 \leftarrow s + h_{j_1}$ for  $j_2$  from  $j_1 + 1$  to n2:  $s_2 \leftarrow s_1 + h_{j_2}$  $H = \begin{bmatrix} h_1 & h_2 & \cdots & h_n \end{bmatrix} \begin{bmatrix} s \\ s \end{bmatrix} \begin{bmatrix} n - k \end{bmatrix}$ 

#### **Exhaustive Search**

Problem: find *w* columns of *H* adding to *s* (modulo 2)

Answer: enumerate all *w*-tuples  $(j_1, j_2, \dots, j_w)$  such that  $1 \le j_1 < j_2 < \dots < j_w \le n$ and check whether  $s + h_{j_1} + h_{j_2} \cdots + h_{j_w} = 0$ 

Requires *about*  $\binom{n}{w}$  column operations

#### **Exhaustive Search**

Problem: find w columns of H adding to s (modulo 2)

$$H = \begin{bmatrix} h_1 & h_2 & \cdots & h_n \end{bmatrix} \begin{pmatrix} n & -k & s = \end{bmatrix}$$

Answer: enumerate all *w*-tuples  $(j_1, j_2, \dots, j_w)$  such that  $1 \le j_1 < j_2 < \dots < j_w \le n$ and check whether  $s + h_{j_1} + h_{j_2} \cdots + h_{j_w} = 0$ 

How to enumerate nicely

Requires *about*  $\binom{n}{w}$  column operations

Note that we obtain all solutions

## **Birthday Decoding**

Problem: find w columns of H adding to s (modulo 2)

$$H = \begin{bmatrix} n \\ H_1 \\ H_2 \\ H_2 \\$$

#### **Birthday Decoding**

Problem: find *w* columns of *H* adding to *s* (modulo 2)

Answer: Split H into two equal parts and enumerate the two following sets

$$\mathcal{L}_1 = \left\{ e_1 H_1^T \mid \mathsf{wt}(e_1) = \frac{w}{2} \right\} \text{ and } \mathcal{L}_2 = \left\{ s + e_2 H_2^T \mid \mathsf{wt}(e_2) = \frac{w}{2} \right\}$$
$$\cap \mathcal{L}_2 \neq \emptyset, \text{ we have solution(s): } s + e_1 H_1^T + e_2 H_2^T = 0$$

Algorithm

If  $\mathcal{L}_1$ 

Compute  $\mathcal{L}_1 \cap \mathcal{L}_2 = \{e_1 H_1^T \mid wt(e_1) = \frac{w}{2}\} \cap \{s + e_2 H_2^T \mid wt(e_2) = \frac{w}{2}\}$ 

$$H = \begin{bmatrix} n \\ H_1 \\ H_2 \end{bmatrix} \begin{bmatrix} s \\ s \\ t \\ n-k \end{bmatrix}$$

Compute  $\mathcal{L}_1 \cap \mathcal{L}_2 = \left\{ e_1 H_1^T \mid \mathsf{wt}(e_1) = \frac{w}{2} \right\} \cap \left\{ s + e_2 H_2^T \mid \mathsf{wt}(e_2) = \frac{w}{2} \right\}$ 

for all  $e_1$  of weight w/2 $x \leftarrow e_1 H_1^T$ ;  $T[x] \leftarrow T[x] \cup \{e_1\}$ 

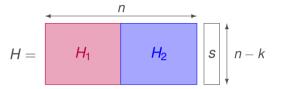
$$H = \begin{bmatrix} n \\ H_1 \\ H_2 \end{bmatrix} \begin{bmatrix} s \\ s \\ t \\ n-k \end{bmatrix}$$

Total cost:  $\binom{n/2}{w/2}$ 

 $|\mathcal{L}_1|$ 

Compute  $\mathcal{L}_1 \cap \mathcal{L}_2 = \left\{ e_1 H_1^T \mid \mathsf{wt}(e_1) = \frac{w}{2} \right\} \cap \left\{ s + e_2 H_2^T \mid \mathsf{wt}(e_2) = \frac{w}{2} \right\}$ 

for all  $e_1$  of weight w/2  $x \leftarrow e_1 H_1^T$ ;  $T[x] \leftarrow T[x] \cup \{e_1\}$ for all  $e_2$  of weight w/2 $x \leftarrow s + e_2 H_2^T$ 



Total cost: 
$$\binom{n/2}{w/2} + \binom{n/2}{w/2}$$
 $|\mathcal{L}_1| \quad |\mathcal{L}_2|$ 

Compute  $\mathcal{L}_1 \cap \mathcal{L}_2 = \left\{ e_1 H_1^T \mid \mathsf{wt}(e_1) = \frac{w}{2} \right\} \cap \left\{ s + e_2 H_2^T \mid \mathsf{wt}(e_2) = \frac{w}{2} \right\}$ 

for all 
$$e_1$$
 of weight  $w/2$   
 $x \leftarrow e_1 H_1^T$ ;  $T[x] \leftarrow T[x] \cup \{e_1\}$   
for all  $e_2$  of weight  $w/2$   
 $x \leftarrow s + e_2 H_2^T$   
for all  $e_1 \in T[x]$   
 $\mathcal{I} \leftarrow \mathcal{I} \cup \{(e_1, e_2)\}$ 

Total cost: 
$$\binom{n/2}{w/2} + \binom{n/2}{w/2} + \frac{\binom{n/2}{w/2}}{2^{n-k}}$$
$$|\mathcal{L}_1| \qquad |\mathcal{L}_2| \qquad \frac{|\mathcal{L}_1| \cdot |\mathcal{L}_2|}{2^{n-k}}$$

Compute  $\mathcal{L}_1 \cap \mathcal{L}_2 = \left\{ e_1 H_1^T \mid \mathsf{wt}(e_1) = \frac{w}{2} \right\} \cap \left\{ s + e_2 H_2^T \mid \mathsf{wt}(e_2) = \frac{w}{2} \right\}$ 

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 of weight  $w/2$   
 $x \leftarrow e_1 H_1^T$ ;  $T[x] \leftarrow T[x] \cup \{e_1\}$   
for all  $e_2$  of weight  $w/2$   
 $x \leftarrow s + e_2 H_2^T$   
for all  $e_1 \in T[x]$   
 $\mathcal{I} \leftarrow \mathcal{I} \cup \{(e_1, e_2)\}$   
return  $\mathcal{I}$   
Total cost:  $\binom{n/2}{w/2} + \binom{n/2}{w/2} + \frac{\binom{n/2}{w/2}^2}{2^{n-k}}$   
 $|\mathcal{L}_1| \quad |\mathcal{L}_2| \quad \frac{|\mathcal{L}_1| \cdot |\mathcal{L}_2|}{2^{n-k}}$ 

$$H = \begin{bmatrix} H_1 & H_2 \\ H_1 & H_2 \end{bmatrix} \begin{bmatrix} s \\ s \\ t \\ n - k \end{bmatrix}$$

## **Birthday Decoding**

Problem: find *w* columns of *H* adding to *s* (modulo 2)

Answer: Split H into two equal parts and enumerate the two following sets

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If  $\mathcal{L}_1 \cap \mathcal{L}_2 \neq \emptyset$ , we have solution(s):  $s + e_1 H_1^T + e_2 H_2^T = 0$ 

Algorithm

Requires about 
$$2\binom{n/2}{w/2} + \frac{\binom{n/2}{w/2}^2}{2^{n-k}}$$
 column operations

Can also be written  $2L + L^2/2^{n-k}$  where  $L = |\mathcal{L}_1| = |\mathcal{L}_2|$ 

Problem: find *w* columns of *H* adding to *s* (modulo 2)

$$H = \begin{array}{c} n \\ H_1 \\ H_2 \\ H_2 \\ H_2$$

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One particular error of Hamming weight *w* splits evenly with probability  $\mathcal{P} = \frac{\binom{n/2}{w/2}^2}{\binom{n}{w}}$ 

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We may have to repeat with H divided in several different ways



or more generally by picking the two halves randomly

Problem: find w columns of H adding to s (modulo 2)

$$H = \begin{array}{c} n \\ H_1 \\ H_2 \\ H_2 \\ H_2$$

To obtain all solutions:



Problem: find w columns of H adding to s (modulo 2)

$$H = \begin{array}{c} n \\ H_1 \\ H_2 \\ H_2 \\ H_2$$

To obtain all most solutions: repeat with  $\approx \frac{1}{P}$  different splitting: { 1. compute  $\mathcal{L}_1$  and  $\mathcal{L}_2$ 2. compute  $\mathcal{L}_1 \cap \mathcal{L}_2$ 

$$\mathcal{P} = \frac{\binom{n/2}{w/2}^2}{\binom{n}{w}}$$

Problem: find *w* columns of *H* adding to *s* (modulo 2)

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To obtain all most solutions: repeat with  $\approx \frac{1}{\mathcal{P}}$  different splitting:  $\begin{cases} 1. \text{ compute } \mathcal{L}_1 \text{ and } \mathcal{L}_2 \\ 2. \text{ compute } \mathcal{L}_1 \cap \mathcal{L}_2 \end{cases}$ Total cost  $\frac{2\binom{n/2}{w/2} + \binom{n/2}{w/2}^2/2^{n-k}}{\mathcal{P}} = \frac{2\binom{n}{w}}{\binom{n/2}{w/2}} + \frac{\binom{n}{w}}{2^{n-k}}$  operations

Problem: find *w* columns of *H* adding to *s* (modulo 2)

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\end{cases}$ Total cost  $\frac{2\binom{n/2}{w/2} + \binom{n/2}{w/2}^2/2^{n-k}}{\mathcal{P}} = \frac{2\binom{n}{w}}{\binom{n/2}{w/2}} + \frac{\binom{n}{w}}{2^{n-k}}$  operations  $\approx \sqrt[4]{8\pi w} \sqrt{\binom{n}{w}} + \# \text{Solutions}$   $\mathcal{P} = \frac{\binom{n/2}{w/2}^2}{\binom{n}{w}}$ 

Problem: find w columns of H adding to s (modulo 2)

$$H = \underbrace{H_{1}}_{n/2 + \varepsilon} \underbrace{H_{2}}_{n/2 + \varepsilon} \left( n - k \quad S = \right)$$

$$f = \underbrace{H_{1}}_{n/2 + \varepsilon} \underbrace{H_{2}}_{n/2 + \varepsilon} \left( n - k \quad S = \right)$$

$$\mathcal{P} = \frac{\binom{n/2 + \varepsilon}{w/2}}{\binom{n}{w}}$$

To obtain all most solutions: repeat with  $\approx \frac{1}{P}$  different splitting: { 1. compute  $\mathcal{L}_1$  and  $\mathcal{L}_2$ 2. compute  $\mathcal{L}_1 \cap \mathcal{L}_2$ 

Relaxation: allow overlapping  $\rightarrow$   $H_1$  and  $H_2$  are wider by  $\varepsilon$ 

Problem: find w columns of H adding to s (modulo 2)

$$H = \underbrace{H_1}_{n/2 + \varepsilon} \underbrace{H_2}_{n/2 + \varepsilon} n - k \quad s = \begin{bmatrix} n \\ n/2 + \varepsilon \\ w/2 \end{bmatrix}$$

To obtain all most solutions: repeat with  $\approx \frac{1}{\mathcal{D}}$  different splitting:  $\begin{cases}
1. compute <math>\mathcal{L}_1 \text{ and } \mathcal{L}_2 \\
2. compute <math>\mathcal{L}_1 \cap \mathcal{L}_2
\end{cases}$ 

 $\approx 1$  $\binom{n}{w}$ 

Relaxation: allow overlapping  $\rightarrow H_1$  and  $H_2$  are wider by  $\varepsilon$ We choose  $\varepsilon$  such that  $\binom{n/2+\varepsilon}{w/2} \approx \sqrt{\binom{n}{w}} \rightarrow$  single repetition

Problem: find w columns of H adding to s (modulo 2)

$$H = \underbrace{H_1}_{n/2 + \varepsilon} \underbrace{H_2}_{n/2 + \varepsilon} \left( n - k \quad s = \right)^2$$

To obtain all most solutions: repeat with  $\approx \frac{1}{\mathcal{P}}$  different splitting:  $\begin{cases} 1. \text{ compute } \mathcal{L}_1 \text{ and } \mathcal{L}_2 \\ 2. \text{ compute } \mathcal{L}_1 \cap \mathcal{L}_2 \end{cases}$ 

$$\mathcal{P} = \frac{\binom{n/2+\varepsilon}{w/2}^2}{\binom{n}{w}} \approx 1$$

Relaxation: allow overlapping  $\rightarrow H_1$  and  $H_2$  are wider by  $\varepsilon$ We choose  $\varepsilon$  such that  $\binom{n/2+\varepsilon}{w/2} \approx \sqrt{\binom{n}{w}} \rightarrow$  single repetition

Total cost:  $2\sqrt{\binom{n}{w}} + \binom{n}{w}/2^{n-k} = 2L + L^2/2^{n-k}$  with  $L = \sqrt{\binom{n}{w}}$ (up to a small constant factor)

# 3. Message Attack (ISD)

- 1. From Generic Decoding to Syndrome Decoding
- 2. Combinatorial Solutions: Exhaustive Search and Birthday Decoding
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- 10. Decoding One Out of Many

For any invertible  $U \in \{0,1\}^{(n-k)\times(n-k)}$  and any permutation matrix  $P \in \{0,1\}^{n\times n}$ 

$$(eH^T = s) \Leftrightarrow (e'H'^T = s')$$
 where  $\begin{cases} H' \leftarrow UHP \\ s' \leftarrow sU^T \\ e' \leftarrow eP \end{cases}$ 

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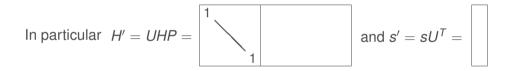
Proof: 
$$e'H'^T = (eP)(UHP)^T$$
  
=  $(eP)P^TH^TU^T$   
=  $eH^TU^T$   
=  $sU^T$   
=  $s'$ 

For any invertible  $U \in \{0,1\}^{(n-k)\times(n-k)}$  and any permutation matrix  $P \in \{0,1\}^{n\times n}$ 

 $CSD(H, s, w) \equiv CSD(UHP, sU^T, w)$ 

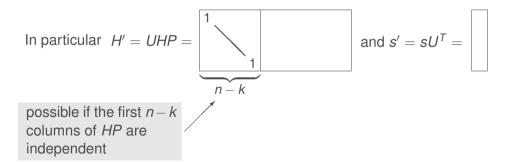
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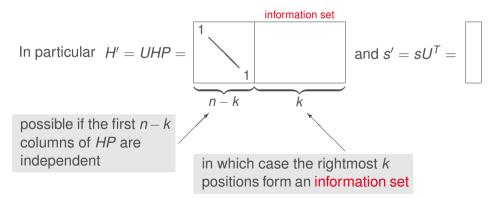
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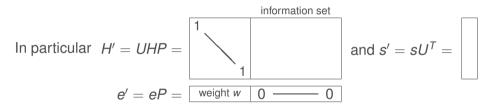
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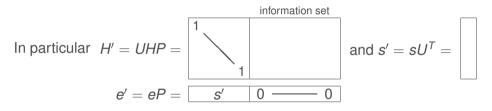


If we are lucky

- the error positions are out of the information set

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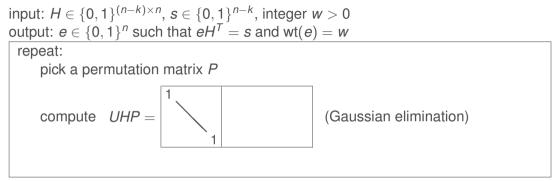
- the error positions are out of the information set
- easy to check because  $e' = (s' \mid 0)$  and wt(s') = w

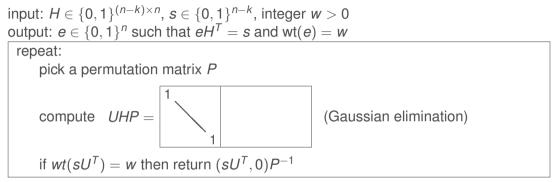
input:  $H \in \{0,1\}^{(n-k)\times n}$ ,  $s \in \{0,1\}^{n-k}$ , integer w > 0output:  $e \in \{0,1\}^n$  such that  $eH^T = s$  and wt(e) = w

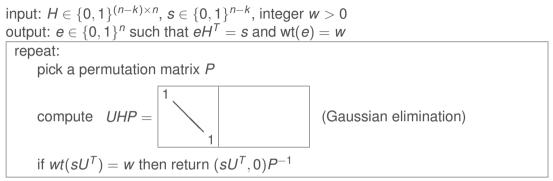
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repeat:

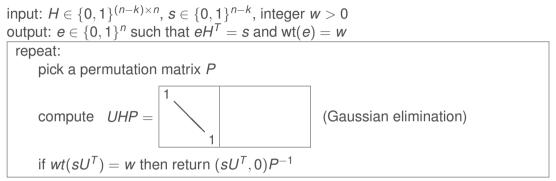
```
pick a permutation matrix P
```







Each iteration costs about n(n-k) column operations



#### Each iteration costs about n(n-k) column operations

Repeat until a solution has its non-zero coordinates "all left"

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# **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  ${\cal K}$
- a success probability  ${\cal P}$

 $\rightarrow$  an expected number of iteration  $\mathcal{N}=1/\mathcal{P}$ 

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

We consider the problem CSD(H, s, w) with  $H \in \{0, 1\}^{(n-k) \times n}$  and  $s \in \{0, 1\}^{n-k}$ 

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 $\rightarrow$  there is always at least one solution

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 $\rightarrow$  there is always at least one solution

1. If  $\binom{n}{w} < 2^{n-k}$  (*i.e.*  $w < \tau_{GV}$ ) there is exactly one solution 2. If  $\binom{n}{w} > 2^{n-k}$  (*i.e.*  $w > \tau_{GV}$ ) there are  $\binom{n}{w}/2^{n-k}$  solutions (on average)

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First case (the most common)  $\rightarrow$  no difference

Second case  $\rightarrow$  finding only one solution should be easier (intuitively by a factor  $\binom{n}{m}/2^{n-k}$ )

ISD performs many independent iterations. For one iteration, we denote

•  $\mathcal{P}_{\infty}$  the probability to find one specific element of CSD(H, s, w)

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- $\mathcal{P}_1$  the probability to find any one element of CSD(H, s, w)

If N = |CSD(H, s, w)|, we have

 $\mathcal{P}_1 = 1 - (1 - \mathcal{P}_\infty)^N \approx \min(1, N\mathcal{P}_\infty)$  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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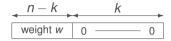
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For the complexity analysis, there are two situations

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

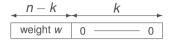
 $\rightarrow$  we expect to execute  $\mathcal{N}_1 = \mathcal{N}_\infty / N = \frac{2^{n-k}}{\binom{n}{k} \mathcal{P}_\infty}$  iterations

An error pattern is found if it has the following form e = weight w 0 - 0



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It follows that 
$$\mathcal{P}_{\infty} = rac{\binom{n-k}{w}}{\binom{n}{w}}$$
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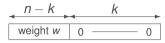
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Total workfactor is

• for all solutions 
$$WF_{Prange} = n(n-k)\frac{\binom{n}{w}}{\binom{n-k}{w}}$$
  
• for one solution  $n(n-k)\frac{\min(2^{n-k},\binom{n}{w})}{\binom{n-k}{w}}$   
ndeed the values are identical when  $\binom{n}{w} < 2^{n-k}$ 



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# 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
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Idea: relax Prange algorithm to amortize the cost of the Gaussian elimination

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Allow error patterns of the form 
$$e = \frac{n-k}{|weight w - p||weight p|}$$

At each iteration, we try the  $\binom{k}{p}$  possible values for the right hand side block

(Prange Algorithm corresponds to p = 0)

Idea: relax Prange algorithm to amortize the cost of the Gaussian elimination

input:  $H \in \{0, 1\}^{(n-k) \times n}$ ,  $s \in \{0, 1\}^{n-k}$ , integer w > 0, a parameter  $p, 0 \le p \le w$ output:  $e \in \{0, 1\}^n$  such that  $eH^T = s$  and wt(e) = w

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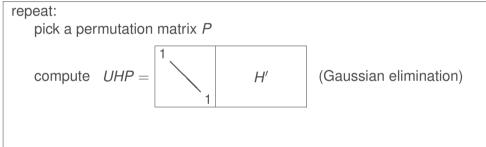
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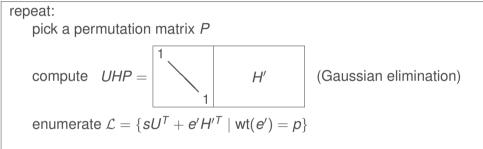
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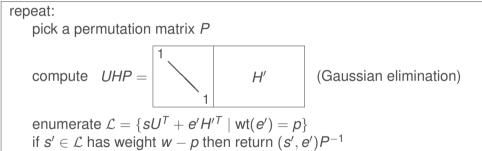
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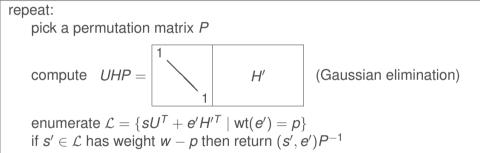
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 $\mathcal{K} = n(n-k) + \binom{k}{p}$  (Gaussian elimination + enumeration)

# Lee and Brickell Algorithm – Complexity Analysis

For an error pattern 
$$e = \frac{n-k}{weight w - p} \frac{k}{weight p}$$
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Never gains more than a polynomial factor over Prange algorithm

$$\mathsf{WF}_{\mathsf{LB}}(p) = \mathcal{N}_{\infty} \cdot \mathcal{K} = \frac{\binom{n}{w}}{\binom{n-k}{w-p}} \left(1 + \frac{n(n-k)}{\binom{k}{p}}\right) > \frac{\binom{n}{w}}{\binom{n-k}{w-p}} > \frac{\binom{n}{w}}{\binom{n-k}{w}} = \frac{1}{n(n-k)}\mathsf{WF}_{\mathsf{Prange}}$$

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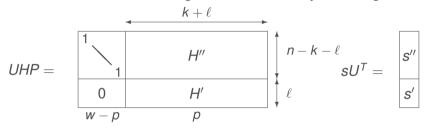
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Except for extravagant parameters, p = 2 is optimal

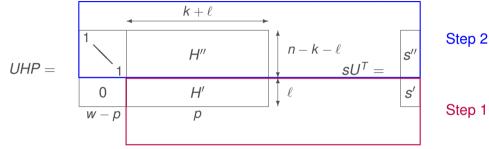
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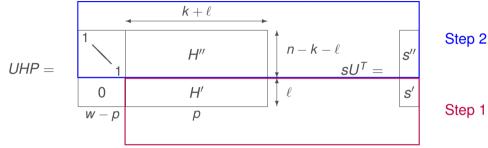


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Step 1: Find all  $e' \in CSD(H', s', p)$ Step 2: Check wt $(e'H''^T + s'') = w - p$ 

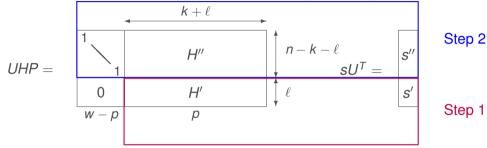
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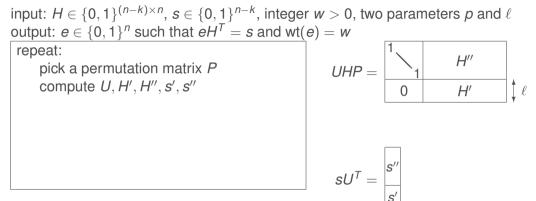
If step 1 is solved by birthday decoding  $\rightarrow$  Dumer Algorithm

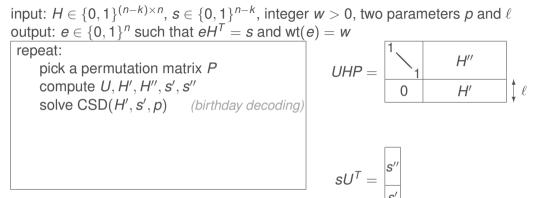
input:  $H \in \{0, 1\}^{(n-k) \times n}$ ,  $s \in \{0, 1\}^{n-k}$ , integer w > 0, two parameters p and  $\ell$  output:  $e \in \{0, 1\}^n$  such that  $eH^T = s$  and wt(e) = w

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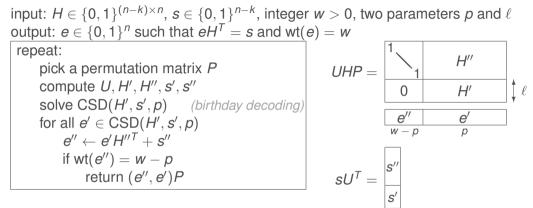
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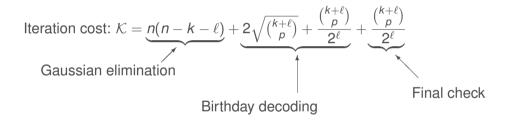


Note: Stern's algorithm (1989) was the first to use birthday decoding, Dumer's algorithm (1991) is only marginally better We will refer now to the Stern/Dumer Algorithm

Iteration cost: 
$$\mathcal{K} = n(n-k-\ell) + 2\sqrt{\binom{k+\ell}{p}} + \frac{\binom{k+\ell}{p}}{2^{\ell}} + \frac{\binom{k+\ell}{p}}{2^{\ell}}$$

Iteration cost: 
$$\mathcal{K} = \underbrace{n(n-k-\ell)}_{\checkmark} + 2\sqrt{\binom{k+\ell}{p}} + \frac{\binom{k+\ell}{p}}{2^{\ell}} + \frac{\binom{k+\ell}{p}}{2^{\ell}}$$
  
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In general, we can write

$$\mathcal{K} = \mathbf{K}_{0} \cdot \mathbf{n}(\mathbf{n} - \mathbf{k} - \ell) + \mathbf{K}_{1} \cdot \sqrt{\binom{k+\ell}{p}} + \mathbf{K}_{2} \cdot \frac{\binom{k+\ell}{p}}{2^{\ell}}$$

where  $K_0$ ,  $K_1$ , and  $K_2$  are small (implementation dependent) constants we will set  $K_0 = K_1 = K_2 = 1$  to simplify the formula

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To be minimized over p and  $\ell$  (positive integers)

The optimization parameters p and  $\ell$  grow with the problem parameters (n, k, w)

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In most situations, the above formula is minimal when the addends are equal

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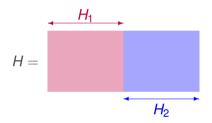
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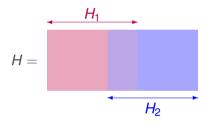
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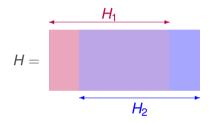
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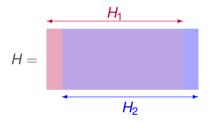
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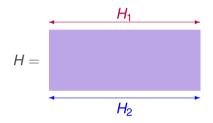
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### Improved Birthday Decoding

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For any binary vector, let  $\phi_r(x)$  denote the last *r* bits of *x*, we define

$$\mathcal{L}_1(r) = \left\{ \mathbf{e}_1 \mathbf{H}^T \mid \mathsf{wt}(\mathbf{e}_1) = \frac{w}{2}, \phi_r(\mathbf{e}_1 \mathbf{H}^T) = \mathbf{0} \right\}$$
$$\mathcal{L}_2(r) = \left\{ \mathbf{s} + \mathbf{e}_2 \mathbf{H}^T \mid \mathsf{wt}(\mathbf{e}_2) = \frac{w}{2}, \phi_r(\mathbf{s} + \mathbf{e}_2 \mathbf{H}^T) = \mathbf{0} \right\}$$

### Improved Birthday Decoding

Idea: Use the "representation technique" (Howgrave-Graham and Joux, 2010) Let  $\mathcal{L}_1 = \left\{ e_1 H^T \mid wt(e_1) = \frac{w}{2} \right\}$  and  $\mathcal{L}_2 = \left\{ s + e_2 H^T \mid wt(e_2) = \frac{w}{2} \right\}$ 

Each  $e \in \text{CSD}(H, s, w)$  "represented"  $\binom{w}{w/2}$  times as  $e = e_1 + e_2$  with  $e_1H^T = s + e_2H^T \in \mathcal{L}_1 \cap \mathcal{L}_2$ 

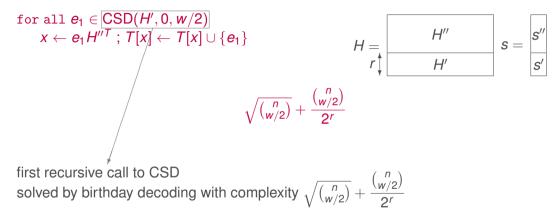
We may decimate  $\mathcal{L}_1$  and  $\mathcal{L}_2$  while keeping the solutions in  $\mathcal{L}_1\cap\mathcal{L}_2$ 

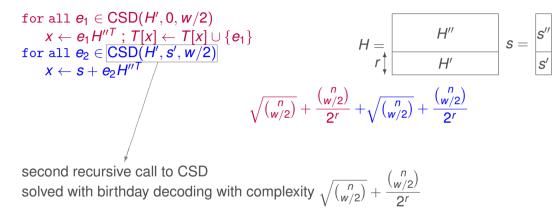
For any binary vector, let  $\phi_r(x)$  denote the last *r* bits of *x*, we define

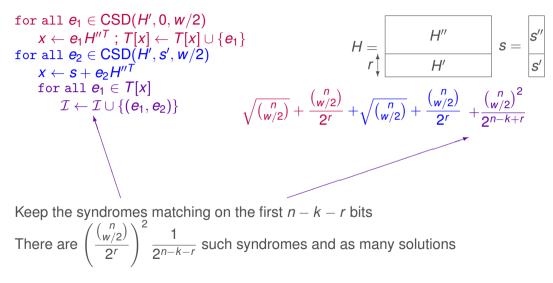
$$\mathcal{L}_1(r) = \left\{ e_1 H^T \mid \mathsf{wt}(e_1) = \frac{w}{2}, \phi_r(e_1 H^T) = 0 \right\}$$
$$\mathcal{L}_2(r) = \left\{ s + e_2 H^T \mid \mathsf{wt}(e_2) = \frac{w}{2}, \phi_r(s + e_2 H^T) = 0 \right\}$$

**Claim:** if  $2^r = \binom{w}{w/2}$  then any  $e \in \text{CSD}(H, s, w)$  is "represented in  $\mathcal{L}_1(r) \cap \mathcal{L}_2(r)$ " with probability > 1/2

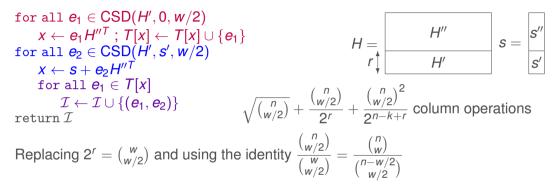




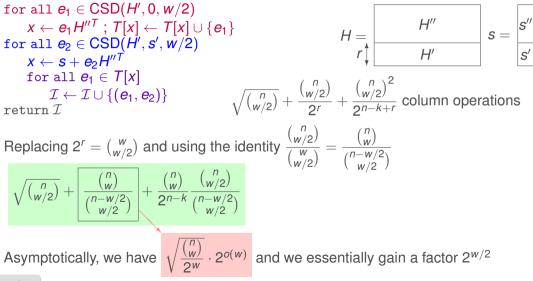




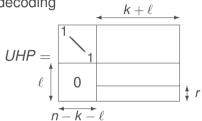
for all  $e_1 \in \operatorname{CSD}(H', 0, w/2)$   $x \leftarrow e_1 H''^T$ ;  $T[x] \leftarrow T[x] \cup \{e_1\}$ for all  $e_2 \in \operatorname{CSD}(H', s', w/2)$   $x \leftarrow s + e_2 H''^T$ for all  $e_1 \in T[x]$   $\mathcal{I} \leftarrow \mathcal{I} \cup \{(e_1, e_2)\}$ return  $\mathcal{I}$   $H = \begin{bmatrix} H'' \\ H' \end{bmatrix}$   $H = \begin{bmatrix} H'' \\ H' \end{bmatrix}$   $r \downarrow \begin{bmatrix} H'' \\ H'' \end{bmatrix}$   $r \downarrow \begin{bmatrix} H'' \\ H' \end{bmatrix}$   $r \downarrow \begin{bmatrix} H'' \\ H' \end{bmatrix}$   $r \downarrow \begin{bmatrix} H'' \\ H' \end{bmatrix}$   $r \downarrow \begin{bmatrix} H'' \\ H'' \end{bmatrix}$  $r \downarrow$ 

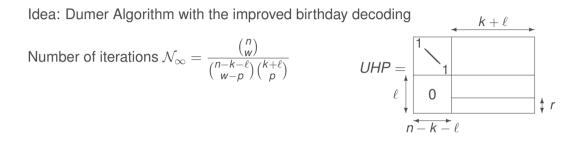


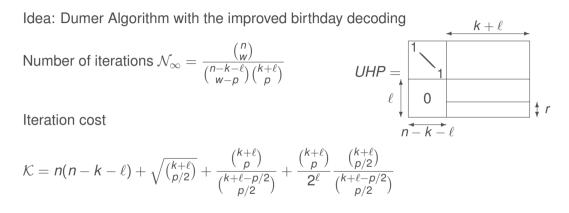
for all 
$$e_1 \in \operatorname{CSD}(H', 0, w/2)$$
  
 $x \leftarrow e_1 H''^T$ ;  $T[x] \leftarrow T[x] \cup \{e_1\}$   
for all  $e_2 \in \operatorname{CSD}(H', s', w/2)$   
 $x \leftarrow s + e_2 H''^T$   
for all  $e_1 \in T[x]$   
 $\mathcal{I} \leftarrow \mathcal{I} \cup \{(e_1, e_2)\}$   
return  $\mathcal{I}$   
 $Replacing 2^r = \binom{w}{w/2}$  and using the identity  $\frac{\binom{n}{w/2}}{\binom{m}{w/2}} = \frac{\binom{n}{w}}{\binom{n-w/2}{w/2}}$   
 $\sqrt{\binom{n}{w/2}} + \frac{\binom{n}{w}}{\binom{n-w/2}{w/2}} + \frac{\binom{n}{w}}{\binom{n-w/2}{w/2}}$ 

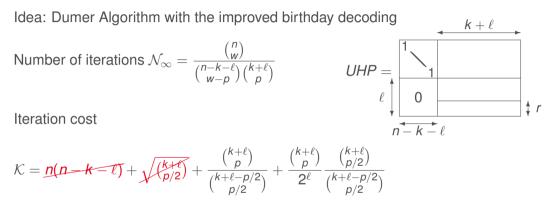


Idea: Dumer Algorithm with the improved birthday decoding

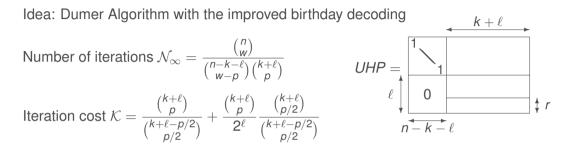


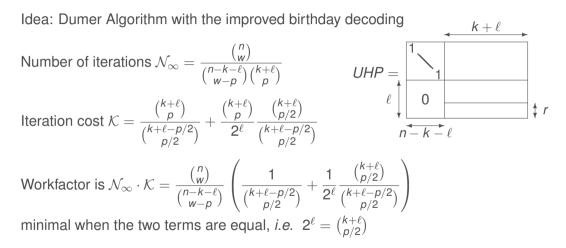


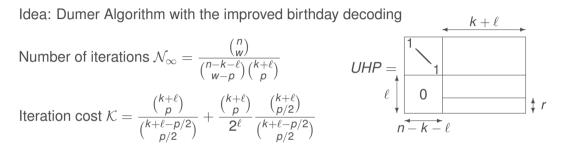




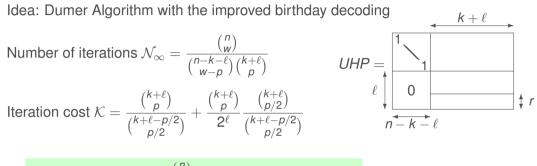
First two terms can be neglected (to be checked a posteriori)







$$\mathsf{WF}_{\mathsf{MMT}} = \min_{p} \frac{\binom{n}{w}}{\binom{n-k-\ell}{w-p}\binom{k+\ell-p/2}{p/2}} \text{ with } \ell = \log_2 \binom{k+\ell}{p/2}$$



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Asymptotic gain  $\approx 2^{p/2}$  compared with Dumer's algorithm

# 3. Message Attack (ISD)

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- 4. Complexity Analysis
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- 10. Decoding One Out of Many

 $\mathcal{L}_{1}(r,\varepsilon) = \left\{ e_{1}H^{T} \mid \mathsf{wt}(e_{1}) = \frac{w}{2} + \varepsilon, \phi_{r}(e_{1}H^{T}) = 0 \right\}$  $\mathcal{L}_{2}(r,\varepsilon) = \left\{ s + e_{2}H^{T} \mid \mathsf{wt}(e_{2}) = \frac{w}{2} + \varepsilon, \phi_{r}(s + e_{2}H^{T}) = 0 \right\}$ 

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Idea:

two words of weight  $\frac{w}{2}$  and length *n* are expected to have  $\begin{cases} \frac{w^2}{4n} \text{ non-zero positions in common} \\ \text{a sum of weight } w - \frac{w^2}{2n} \end{cases}$ 

$$\mathcal{L}_{1}(r,\varepsilon) = \left\{ e_{1}H^{T} \mid \mathsf{wt}(e_{1}) = \frac{w}{2} + \varepsilon, \phi_{r}(e_{1}H^{T}) = 0 \right\}$$
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Idea: if  $\varepsilon = \frac{(w/2+\varepsilon)^2}{n}$ , two words of weight  $\frac{w}{2} + \varepsilon$  and length *n* are expected to have  $\begin{cases} \varepsilon & \text{non-zero positions in common} \\ \text{a sum of weight } w \end{cases}$ 

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Note also that there are  $\binom{w}{w/2}\binom{n-w}{\varepsilon}$  different ways to write  $e = e_1 + e_2$  with wt(e) = w and wt( $e_1$ ) = wt( $e_2$ ) =  $\frac{w}{2} + \varepsilon$ 

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Workfactor "simplifies" to

$$\sqrt{\binom{n}{w/2+\varepsilon}} + \frac{\binom{n}{w}}{\binom{n}{w/2+\varepsilon}} + \frac{\binom{n}{w}}{2^{n-k}}$$

(up to a polynomial factor)

# Impact on MMT Algorithm Complexity

#### Instead of

$$\mathsf{WF}_{\mathsf{MMT}} = \min_{p} \frac{\binom{n}{w}}{\binom{n-k-\ell}{w-p}\binom{k+\ell-p/2}{p/2}} \text{ with } \ell = \log_2 \binom{k+\ell}{p/2}$$

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(up to a constant factor)

We set 
$$\varepsilon = \frac{(w/2+\varepsilon)^2}{n}$$
, and the workfactor reduces to  

$$WF = \min_{p} \frac{\binom{n}{w}}{\binom{n-k-\ell}{p/2+\varepsilon}} \text{ with } \ell = \log_2 \binom{k+\ell}{p/2+\varepsilon}$$

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This is the embryo of the next improvement of ISD

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## Becker, Joux, May, and Meurer Algorithm (1/2)

Idea: what happens if we let  $\varepsilon$  grows (much) beyond  $w^2/4n$ ?

 $\mathcal{L}_{1}(r,\varepsilon) = \left\{ e_{1}H^{T} \mid \mathsf{wt}(e_{1}) = \frac{w}{2} + \varepsilon, \phi_{r}(e_{1}H^{T}) = 0 \right\}$  $\mathcal{L}_{2}(r,\varepsilon) = \left\{ s + e_{2}H^{T} \mid \mathsf{wt}(e_{2}) = \frac{w}{2} + \varepsilon, \phi_{r}(s + e_{2}H^{T}) = 0 \right\}$ 

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# BJMM Algorithm (2/2)

BJMM Algorithm, key features:

- increase *ε* leading to FIBD (Further Improved Birthday Decoding)
- make an additional level of recursive call to FIBD (improved birthday decoding makes two calls to smaller CSD problems)
- embed all this into Information Set Decoding framework

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- embed all this into Information Set Decoding framework

Improves the workfactor

Algorithm and analysis are very elaborated

#### **Comparison of the Various ISD Variants**

 $\mathsf{WF} = 2^{c \cdot n(1+o(1))}$ 

*c* a constant (asymptotic exponent)

## **Comparison of the Various ISD Variants**

	$c = \lim_{n \to \infty} \frac{\log_2 WF}{n}$	
	<i>k</i> = 0.5 <i>n</i>	
	<i>w</i> = 0.11 <i>n</i>	
Enumeration	0.5	
Birthday Decoding	0.25	
Prange	0.1198	
Stern	0.1154	
Dumer	0.1151	
MMT	0.1101	
BJMM	0.1000	

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## **Comparison of the Various ISD Variants**

	$c = \lim_{n \to \infty} \frac{\log_2 WF}{n}$	
	<i>k</i> = 0.5 <i>n</i>	<i>k</i> = 0.8 <i>n</i>
	<i>w</i> = 0.11 <i>n</i>	w = 0.03 <i>n</i>
Enumeration	0.5	0.2
Birthday Decoding	0.25	0.1
Prange	0.1198	0.0724
Stern	0.1154	0.0680
Dumer	0.1151	0.0679
MMT	0.1101	0.0638
BJMM	0.1000	0.0562

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*c* a constant (asymptotic exponent)

Remark that Birthday Decoding is comparatively better when k/n grows

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## **Generalized Birthday Algorithm**

Proposed by D. Wagner in 2002, in a more general context

The Generalized Birthday Algorithm (GBA) of order *a* solves the following problem:

Instance:  $2^a$  lists of vectors  $\mathcal{L}_i \subset \{0,1\}^\ell$ ,  $i = 1, 2, ..., 2^a$ Answer:  $x_i \in \mathcal{L}_i$ ,  $i = 1, 2, ..., 2^a$  such that  $x_1 + x_2 + ... + x_{2^a} = 0$ 

If the lists are large enough, then GBA runs in time  $O(2^{\ell/(a+1)})$ 

(the case a = 1 corresponds to the usual birthday paradox)

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GBA can be applied to decoding

- it applies to instances of CSD with many solutions
- it aims at finding one solution only

Let  $H \in \{0, 1\}^{(n-k) \times n}$ ,  $s \in \{0, 1\}^{n-k}$ , and w > 0, we consider CSD(H, s, w) where

- there are many solutions: exact condition to be determined
- we only want one solution

$$H = H_1 H_2$$

 $s = s_1 + s_2$  arbitrarily

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We build two lists of size L

$$\mathcal{L}_i \subset \{ oldsymbol{s}_i + oldsymbol{e}_i H_i^{\mathcal{T}} \mid \mathsf{wt}(oldsymbol{e}_i) = w/2 \}, i \in \{1,2\}$$

Any element of  $\mathcal{L}_1 \cap \mathcal{L}_2$  provides a solution

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 arbitrarily

We must have 
$$|\mathcal{L}_1 \cap \mathcal{L}_2| = rac{L^2}{2^{n-k}} \geq 1$$

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Choosing  $L = 2^{(n-k)/2}$  the workfactor is  $O(2^{(n-k)/2})$ 

Let  $H \in \{0,1\}^{(n-k) \times n}$ ,  $s \in \{0,1\}^{n-k}$ , and w > 0, we consider CSD(H, s, w) where

- there are many solutions:  $\binom{n/2}{w/2}^2 \ge 2^{n-k}$
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We must have 
$$|\mathcal{L}_1 \cap \mathcal{L}_2| = rac{L^2}{2^{n-k}} \geq 1$$

Choosing  $L = 2^{(n-k)/2}$  the workfactor is  $O(2^{(n-k)/2})$ L cannot exceed  $\binom{n/2}{w/2}$ , and thus we need  $\binom{n/2}{w/2}^2 \ge 2^{n-k}$ 

$$H = H_1 H_2$$

$$s = s_1 + s_2$$
 arbitrarily

$$H = \begin{bmatrix} H_1 & H_2 & H_3 & H_4 \end{bmatrix}$$

$$s = s_1 + s_2 + s_3 + s_4$$

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Let  $\mathcal{L}_i \subset \{s_i + e_i H_i^T \mid wt(e_i) = w/4\}, i \in \{1, 2, 3, 4\}$ 

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Let  $\mathcal{L}_i \subset \{s_i + e_i H_i^T \mid wt(e_i) = w/4\}, i \in \{1, 2, 3, 4\}$  of size  $L = 2^{\ell}, \ell = (n - k)/3$ 

$H = H_1$	H <sub>2</sub>	H <sub>3</sub>	H <sub>4</sub>
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After computing  $\mathcal{L}_1, \mathcal{L}_2, \mathcal{L}_3, \mathcal{L}_4, \mathcal{L}_{1,2}, \mathcal{L}_{3,4}$  we expect to find an element in  $\mathcal{L}_{1,2} \cap \mathcal{L}_{3,4}$  from which we derive a solution to CSD(H, s, w)

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The computing effort is  $O(2^{(n-k)/3})$  possible only if  $\binom{n/4}{w/4} \ge 2^{(n-k)/3}$ 

In general the order *a* GBA decoding will have a cost  $O\left(2^{\frac{n-k}{a+1}}\right)$ 



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Finally, note that improvements of birthday decoding apply This allows to lower the complexity in some cases

# **Comparing GBA and ISD**

Information Set Decoding (all variants) and its complexity analysis can easily be adapted to the case where we seek one solution among many

In practice ISD is almost always more efficient

GBA is more efficient only when the code rate k/n is close to 1 and even then, it is only better for a limited range of values of w

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#### *N*-Syndrome Decoding

Instance:  $S \subset \{0,1\}^{n-k}$ , |S| = N,  $H \in \{0,1\}^{(n-k) \times n}$ , an integer w > 0Answer:  $e \in \{0,1\}^n$  such that  $eH^T \in S$  and  $wt(e) \le w$ 

We will denote  $CSD_N(H, S, w)$  the set of all solutions to the above problem

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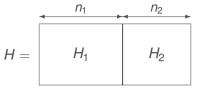
Improvement:

- we get the N solutions at the expense of a factor  $\approx \sqrt{N}$
- or we get one solution with a gain of a factor  $\approx \sqrt{N}$

## **Birthday Decoding With Multiple Instances**

Solve  $CSD_N(H, S, w)$  with birthday decoding

Let 
$$\begin{cases} \mathcal{L}_1 = \{ e_1 H_1^T \mid \mathsf{wt}(e_1) = w_1 \} \\ \mathcal{L}_2 = \{ s + e_2 H_2^T \mid s \in S, \mathsf{wt}(e_2) = w_2 \} \end{cases}$$



$$n = n_1 + n_2, w = w_1 + w_2$$

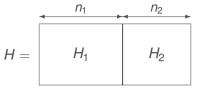
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We choose  $w_1$  and  $w_2$  such that

$$\frac{w_1}{n_1} = \frac{w_2}{n_2}$$
 and  $|\mathcal{L}_1| = \binom{n_1}{w_1} = |\mathcal{L}_2| = N\binom{n_2}{w_2}$ 



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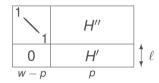
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**Claim:** If  $N \leq \binom{n}{w}$ , we obtain all solutions of  $CSD_N(H, S, w)$ for a cost  $\sqrt{N\binom{n}{w}} + \frac{N\binom{n}{w}}{2^{n-k}}$  (up to a polynomial factor)

Solve  $\text{CSD}_N(H, S, w)$  when  $S \subset \{eH^T \mid wt(e) = w\}$  with Dumer Algorithm

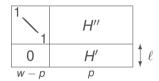
The problem has N solutions and we only want one



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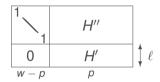
A specific solution requires 
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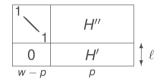


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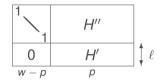


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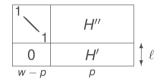
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 $\rightarrow$  gain of a factor  $\approx \sqrt{N}$  as long as  $N \leq \min\left(\mathcal{N}_{\infty}, \binom{k+\ell}{p}\right)$ 

A gain is also possible with an Order 2 GBA Decoding when  $N = |S| = \binom{n/3}{w/3}$ 

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To be compared with  $\sqrt{\binom{n}{w}}$  with the birthday decoding, gaining a factor  $\approx \sqrt{N}$ 

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# **Code-Based Cryptography**

- 1. Error-Correcting Codes and Cryptography
- 2. McEliece Cryptosystem
- 3. Message Attacks (ISD)
- 4. Key Attacks
- 5. Other Cryptographic Constructions Relying on Coding Theory