

# The Representation Technique

## Cryptanalysis for Dlog, SubsetSum, Decoding

Alexander May

Ruhr-University Bochum

Summer School Kaliningrad, July 2019

# Discrete Logarithms

## DLP: Discrete Logarithm Problem

**Given:** Generator  $g$  for  $G = \langle g \rangle$  with  $2^{n-1} \leq |G| < 2^n$ ,  $\beta = g^x$

**Find:**  $x = \text{dlog}_g \beta \in \mathbb{Z}_{|G|}$

### Examples:

- $G = (\mathbb{Z}, +) = \langle 1 \rangle$ ,  $x = \text{dlog}_1 \beta = \beta$
- $G = (E(\mathbb{F}_p), +)$ , best algorithm  $\tilde{O}(\sqrt{|G|}) = \tilde{O}(2^{\frac{n}{2}})$ .
- $G = (\mathbb{Z}_p^*, \cdot)$ , best algorithm sub-exponential
- $G$  generic:  $\Omega(\sqrt{|G|})$

**Variants:** small  $x$ , small Hamming weight  $x$ , faulty  $x$ , many  $x$

# DLP Enumeration

## Algorithm Brute-Force DLP

**Input:**  $g, \beta$

- 1  $x = 0$ .
- 2 **While**  $(g^x \neq \beta)$  **do**  $x = x + 1$ ;

**Output:**  $x = \text{dlog}_g \beta$

### Runtime:

- Need  $x$  iterations of while-loop, each costs one group operation.
- $\mathcal{O}(x) = \mathcal{O}(|G|) = \mathcal{O}(2^n)$  group operations.
- Each group operation usually costs  $\mathcal{O}(\log^c n)$  bit operations.
- **Notice:** Brute-Force not so bad for small  $x$ .

# Reaching Square Root Complexity

## Idea:

- Write  $x = x_1 + x_2 2^{n/2}$  with  $0 \leq x_1, x_2 < 2^{n/2}$ .
- Use identity  $g^{x_1} = \beta \cdot (g^{-2^{n/2}})^{x_2}$ .

## Algorithm Meet-in-the-Middle DLP

**Input:**  $g, \beta$

- 1 **For**  $0 \leq i < 2^{n/2}$  **do** store  $(i, g^i)$  in list  $L$ .
- 2 Sort list  $L$  according to second entry.
- 3 **For**  $0 \leq j < 2^{n/2}$  **do** if  $\exists (i, \beta \cdot (g^{-2^{n/2}})^j) \in L$ , output  $x = i + j 2^{n/2}$ .

**Output:**  $x = \text{dlog}_g \beta$

**Correctness:** MitM terminates iff  $(i, j) = (x_1, x_2)$ .

**Run time:**  $\tilde{O}(2^{n/2}) = \tilde{O}(\sqrt{|G|})$ . But also memory  $\tilde{\Theta}(\sqrt{|G|})$ .

**Exercise:** Modify MitM such that it has runtime  $\tilde{O}(\sqrt{x})$ .

# Multiple Discrete Logarithms

## Multiple DLP

**Given:** Generator  $g$  for  $G = \langle g \rangle$  with  $2^{n-1} \leq |G| < 2^n$ ,  
 $\beta_1 = g^{x_1}, \dots, \beta_k = g^{x_k}$

**Find:**  $x_1, \dots, x_k$

**Easy:**  $\tilde{O}(k \cdot \sqrt{|G|})$ .

**Exercise:** Show that Multiple DLP can be solved in  $\tilde{O}(\sqrt{k \cdot |G|})$ .

# Small Weight Discrete Logarithms

## Small weight DLP

**Given:** Generator  $g$  for  $G = \langle g \rangle$  with  $2^{n-1} \leq |G| < 2^n$ ,  
 $\beta = g^x$  with known Hamming weight  $\text{wt}(x) = \alpha n$ ,  $\alpha \in [0, 1]$

**Find:**  $x$

## Algorithm Brute-Force Small weight DLP

**Input:**  $g, \beta, \alpha$

① **For all**  $x$  with  $\text{wt}(x) = \alpha n$  **do** if  $(g^x = \beta)$  output  $x$ ;

**Output:**  $x = \text{dlog}_g \beta$

**Run time:**  $\tilde{O}\left(\binom{n}{\alpha n}\right)$ . How good is that?

# Bounding Binomial Coefficients

## Theorem Binomials

We have  $\binom{n}{\alpha n} = \tilde{\Theta}(2^{H(\alpha)n})$  with  $H(\alpha) = -\alpha \log(\alpha) - (1 - \alpha) \log(1 - \alpha)$ .

By Stirling's formula  $n! \sim \sqrt{2\pi n} \cdot \left(\frac{n}{e}\right)^n$  we have

$$\begin{aligned}\binom{n}{\alpha n} &= \frac{n!}{(\alpha n)!((1 - \alpha)n)!} = \tilde{\Theta}\left(\frac{\left(\frac{n}{e}\right)^n}{\left(\frac{\alpha n}{e}\right)^{\alpha n} \left(\frac{(1 - \alpha)n}{e}\right)^{(1 - \alpha)n}}\right) \\ &= \tilde{\Theta}\left(2^{(-\alpha \log \alpha - (1 - \alpha) \log(1 - \alpha))n}\right) = \tilde{\Theta}(2^{H(\alpha)n})\end{aligned}$$

## Corollary

For  $0 \leq \alpha \leq \beta \leq 1$ :  $\binom{\beta n}{\alpha n} = \binom{\beta n}{\alpha \frac{1}{\beta} \beta n} = \tilde{\Theta}(2^{H(\frac{\alpha}{\beta}) \cdot \beta n})$ .

# Small weight Discrete Logarithms

Brute-Force Small Weight DLP:  $\tilde{O}\left(\binom{n}{\alpha n}\right) = \tilde{O}(2^{H(\alpha)n})$ ,  $\alpha = \frac{1}{2} : \tilde{O}(2^n)$ .

**Exercise 1:** Assume that we get the promise  $x = x_1 + x_2 2^{n/2}$  with

$$0 \leq x_1, x_2 < 2^{n/2} \text{ and } \text{wt}(x_1) = \text{wt}(x_2) = \alpha \cdot \frac{n}{2}.$$

Devise a MitM algorithm with run time  $\tilde{O}(2^{\frac{H(\alpha)}{2}n})$ .

**Exercise 2:** Do Exercise 1 without promise.



# Faulty Discrete Logarithms

## Faulty DLP

**Given:** Generator  $g$  for  $G = \langle g \rangle$  with  $2^{n-1} \leq |G| < 2^n$ ,  
 $\beta = g^x$ , faulty  $\tilde{x}$  with  $\alpha n$ ,  $\alpha \in [0, 1]$  many  $1 \rightarrow 0$ -flips of  $x$

**Find:**  $x$

**Mini Exercise:** Show how Faulty DLP relates to Small weight DLP.

# Finding a function collision

## Collision finding

**Given:** function  $f : \{0, 1\}^n \rightarrow \{0, 1\}^n$  (with random properties)

**Find:**  $x_1 \neq x_2$  with  $f(x_1) = f(x_2)$

- $\Pr_{x_1 \neq x_2}(f(x_1) = f(x_2)) = \frac{1}{2^n}$
- Brute Force: Sample  $2^n$  many pairs  $(x_1, x_2)$ .

# Birthday Paradox – Meet in the Middle

## Algorithm List Collision Finding

**Input:**  $f : \{0, 1\}^n \rightarrow \{0, 1\}^n$

- 1 Compute list  $L$  with entries  $(x_i, f(x_i))$  for  $i = 1, \dots, 2^{n/2} + 1$ .
- 2 Search for  $(x_i, y), (x_j, y) \in L$  with  $i \neq j$ .

**Output:** Collision or  $\perp$

### Run time & Success probability:

- Run time  $\tilde{O}(2^{n/2})$  (but also the same memory).
- $L$  does not contain a collision with probability

$$\prod_{i=0}^{2^{n/2}} \left(1 - \frac{i}{2^n}\right) \leq \prod_{i=1}^{2^{n/2}} e^{-\frac{i}{2^n}} = e^{-\sum_{i=1}^{2^{n/2}} \frac{i}{2^n}} = e^{-\frac{2^{n/2}(2^{n/2}+1)}{2 \cdot 2^n}} \leq e^{-\frac{1}{2}} \approx 0.6.$$

- Thus, we succeed with probability  $\approx 0.4$ .

## Iterating a function

- Consider sequence:  $x, f(x), f(f(x)), f(f(f(x))), \dots$
- Let us use notation  $f^i(x)$  for  $i$  applications.
- Let  $\gamma, \lambda > 0$  be minimal with  $f^\gamma(x) = f^{\gamma+\lambda}(x)$ . Then

$$f^{\gamma+1}(x) = f^{\gamma+\lambda+1}(x), f^{\gamma+2}(x) = f^{\gamma+\lambda+2}(x), \dots$$

- By the argumentation before we expect that  $\gamma + \lambda \approx 2^{\frac{n}{2}}$ .

# Cycle Finding

## Algorithm Cycle Finding

**Input:**  $f : \{0, 1\}^n \rightarrow \{0, 1\}^n$

- 1 **Repeat** Choose start point  $x \in \{0, 1\}$  **until**  $x \neq f(x)$ .
- 2 Set  $i = 1$ ,  $k_i = f(x)$ ,  $k_{2i} = f(f(x))$ .
- 3 **While**  $k_i \neq k_{2i}$  **do**
  - 1  $k_{i+1} = f(k_i)$ ,  $k_{2(i+1)} = f(f(k_{2i}))$ . Set  $i = i + 1$ .
- 4 Set  $\ell = 0$ ,  $k_\ell = x$ .
- 5 **While**  $f(k_\ell) \neq f(k_{\ell+i})$  **do**  $k_{\ell+1} = f(k_\ell)$ ,  $k_{\ell+i+1} = f(k_{\ell+i})$ ,  $\ell = \ell + 1$ .

**Output:**  $x_1 = k_\ell$ ,  $x_2 = k_{\ell+i}$  with  $f(x_1) = f(x_2)$  and  $x_1 \neq x_2$

# Cycle Finding

## Correctness

- After the first while-loop we have  $k_i = k_{2i}$ .
- We already know that  $k_j = k_{j+c\lambda}$ ,  $c \in \mathbb{N}$  for all  $j \geq \gamma$ .
- We conclude that  $i = k\lambda$ .
- In the second loop we find the minimum  $\gamma$  for which

$$k_\gamma = k_{\gamma+k\lambda}.$$

- At termination we have  $f(k_{\gamma-1}) = f(k_{\gamma+k\lambda-1})$  which implies

$$f(x_1) = f(k_{\gamma-1}) = k_\gamma = k_{\gamma+k\lambda} = f(k_{\gamma+k\lambda-1}) = f(x_2).$$

- Furthermore,  $x_1 \neq x_2$  by minimality of  $\gamma$ .

## Complexity

- Memory consumption  $\tilde{O}(1)$ .
- After  $\gamma + \lambda \approx 2^{\frac{n}{2}}$  we cycle. The cycle length is  $\lambda$ . (**While**-loop in 3)
- In total, we need  $2(\gamma + \lambda) \approx 2^{\frac{n}{2}+1}$  iterations until termination.

## Two functions

### Theorem Rho Method

In a function  $f : \{0, 1\}^n \rightarrow \{0, 1\}^n$  we find a collision in time  $\tilde{O}(2^{\frac{n}{2}})$  with space  $\tilde{O}(1)$ .

### Two function collision finding

**Given:** functions  $f_1 : \{0, 1\}^n \rightarrow \{0, 1\}^n$ ,  $f_2 : \{0, 1\}^n \rightarrow \{0, 1\}^n$   
(with random properties)

**Find:**  $x_1, x_2$  with  $f_1(x_1) = f_2(x_2)$

### Theorem Rho Method

In two functions  $f_1, f_2 : \{0, 1\}^n \rightarrow \{0, 1\}^n$  we find a collision in time  $\tilde{O}(2^{\frac{n}{2}})$  with space  $\tilde{O}(1)$ .

**Exercise:** Adapt the previous method for two functions.

# Rho Method for DLP

**Representation of DLP:** Define

$$f_1 : \mathbb{Z}_{|G|} \rightarrow G, x \mapsto g^{x_1} \text{ and } f_2 : \mathbb{Z}_{|G|} \rightarrow G, x_2 \mapsto \beta \cdot g^{-x_2}.$$

- Any collision  $(x_1, x_2)$  satisfies  $g^{x_1} = \beta \cdot g^{-x_2}$ .
- Thus,  $x = x_1 + x_2 \pmod{|G|}$  solves DLP.
- There exist  $\mathbb{Z}_{|G|}$  many representations, respectively collisions.
- For solving DLP it suffices to find a *single* representation  $(x_1, x_2)$ .

## Definition Representation

Let  $x = x_1 + x_2$ . Then  $(x_1, x_2)$  is called a *representation* of  $x$ .

## Theorem Rho Method for DLP

DLP can be solved in any group  $G$  in time  $\tilde{O}(\sqrt{|G|})$  and memory  $\tilde{O}(1)$ .

**Exercise:** Show an  $\tilde{O}(x^{\frac{3}{2}})$ -algorithm for small  $x$ -DLP with memory  $\tilde{O}(1)$ .



## Small Weight DPL with Low Memory

**Promise:**  $x = x_1 + x_2 2^{n/2}$  with  $0 \leq x_i < 2^{n/2}$  and  $\text{wt}(x_i) = \alpha \cdot \frac{n}{2}$ .

- Search space  $\mathcal{S} = \{x_i \in \mathbb{Z}_{2^{n/2}} \mid \text{wt}(x_i) = \alpha \cdot \frac{n}{2}\}$ .
- Therefore  $|\mathcal{S}| = \binom{n/2}{\alpha \cdot n/2} = \tilde{\Theta}(2^{H(\alpha)n/2})$ .
- Let  $h : G \rightarrow \mathcal{S}$ . Define  $f_i : \mathcal{S} \rightarrow \mathcal{S}$  with

$$x_1 \mapsto h(g^{x_1}) \text{ and } x_2 \mapsto h(\beta \cdot g^{-x_2 2^{n/2}}).$$

### Algorithm Folklore Low Weight DPL with Low Memory

**Input:**  $f_1, f_2, h$

① **Repeat**

① Find a random collision  $(x_1, x_2)$  in  $f_1, f_2$

② **Until**  $g^{x_1} = \beta \cdot g^{-x_2 2^{n/2}}$

**Output:**  $x = x_1 + x_2 2^{n/2}$

## 0.75 Algorithm

### Run Time:

- Every iteration costs  $\tilde{O}(\sqrt{|\mathcal{S}|})$ .
- Since  $f_i : \mathcal{S} \rightarrow \mathcal{S}$ , we expect  $|\mathcal{S}|$  collisions.
- $x$  has a unique representation as  $x = x_1 + x_2 2^{\frac{n}{2}}$ .
- Therefore only a single collisions  $(x_1, x_2)$  satisfies  $g^{x_1} = \beta \cdot g^{-x_2 2^{n/2}}$ .
- The probability that an iteration succeeds is thus

$$p = \Pr[(x_1, x_2) \text{ satisfies } g^{x_1} = \beta \cdot g^{-x_2}] = \frac{1}{|\mathcal{S}|}.$$

- We obtain expected run time

$$p^{-1} \tilde{O}(\sqrt{|\mathcal{S}|}) = \tilde{O}(|\mathcal{S}|^{\frac{3}{2}}) = \tilde{O}(2^{\frac{3}{4}H(\alpha)n})$$

- For  $\alpha = \frac{1}{2}$  this is time  $2^{\frac{3}{4}n}$  as opposed to  $2^{\frac{1}{2}n}$  for Rho.

## Improving a bit

**Idea:** Take the representation  $x = x_1 + x_2$  with  $x_1, x_2 \in \mathbb{Z}_{|G|}$  as in Rho.

- We choose  $\text{wt}(x_1) = \text{wt}(x_2) = \frac{\alpha}{2}n$ .
- Search space  $\mathcal{S} = \{x_i \in \mathbb{Z}_{|G|} \mid \text{wt}(x_i) = \frac{\alpha}{2} \cdot n\}$ .
- Therefore  $|\mathcal{S}| = \binom{n}{\alpha/2 \cdot n} = \tilde{\Theta}(2^{H(\alpha/2)n})$ .
- Let  $h : G \rightarrow \mathcal{S}$ . Define  $f_i : \mathcal{S} \rightarrow \mathcal{S}$  with

$$x_1 \mapsto h(g^{x_1}) \text{ and } x_2 \mapsto h(\beta \cdot g^{-x_2}).$$

### Algorithm Improved Low Weight DPL with Low Memory

**Input:**  $f_1, f_2, h$

① **Repeat**

① Find a random collision  $(x_1, x_2)$  in  $f_1, f_2$

② **Until**  $g^{x_1} = \beta \cdot g^{-x_2}$

**Output:**  $x = x_1 + x_2 2^{n/2}$

## 0.72 Algorithm

### Run Time:

- Every iteration cost  $\tilde{O}(\sqrt{|\mathcal{S}|})$ .
- Since  $f_i : \mathcal{S} \rightarrow \mathcal{S}$ , we expect  $|\mathcal{S}|$  collisions.
- $x$  has  $\binom{\alpha n}{\frac{\alpha}{2}n} = \tilde{\Theta}(2^{\alpha n})$  many representation as  $x = x_1 + x_2$ .
- All representations  $(x_1, x_2)$  satisfy  $g^{x_1} = \beta \cdot g^{-x_2}$ .
- The probability that an iteration succeeds is thus

$$p = \Pr[(x_1, x_2) \text{ satisfies } g^{x_1} = \beta \cdot g^{-x_2}] = \frac{\tilde{\Theta}(2^{\alpha n})}{|\mathcal{S}|}.$$

- We obtain expected run time

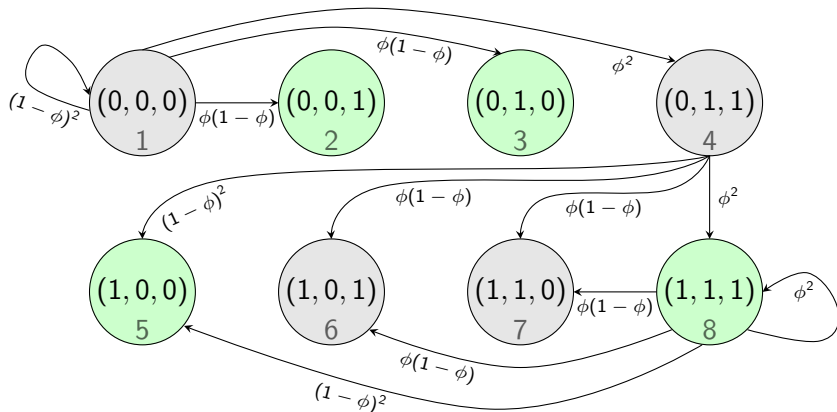
$$p^{-1} \tilde{O}(\sqrt{|\mathcal{S}|}) = \tilde{O}\left(\frac{|\mathcal{S}|^{\frac{3}{2}}}{2^{\alpha n}}\right) = \tilde{O}(2^{(\frac{3}{2}H(\alpha/2) - \alpha)n})$$

- For  $\alpha = \frac{1}{2}$  this is time  $2^{0.72n}$  as opposed to  $2^{\frac{1}{2}n}$  for Rho.

## Improving a bit more via carries

**Idea:** Take  $\text{wt}(x_1) = \text{wt}(x_2) = \phi n \geq \frac{\alpha}{2}n$  such that  $\text{wt}(x_1 + x_2) = \alpha n$ .

- Search space  $\mathcal{S} = \{x_i \in \mathbb{Z}_{|G|} \mid \text{wt}(x_i) = \phi n\}$ .
- Therefore  $|\mathcal{S}| = \binom{n}{\phi n} = \tilde{\Theta}(2^{H(\phi)n})$ .
- Analysis: Take each 1-coordinate in  $x_1, x_2$  with probability  $\phi$ .



# Analysis

- Define matrix  $M$  for Markov process.
- Process has a stationary distribution  $\pi = (\pi_1, \dots, \pi_8)$  with  $\pi = M\pi$ .
- We solve the system of linear equations

$$\pi = M\pi, \quad \pi_1 + \dots + \pi_8 = 1, \quad \pi_2 + \pi_3 + \pi_5 + \pi_8 = \alpha.$$

- Obtain  $\alpha = 4\phi^4 - 4\phi^3 - \phi^2 + 2\phi$ . Check:  $\phi = \frac{1}{2} \Rightarrow \alpha = \frac{1}{2}$ .
- Number of representations  $(x_1, x_2)$ : heuristically  $\frac{|\mathcal{S}|^2}{\binom{n}{\alpha n}}$ .
- This implies  $p = \frac{|\mathcal{S}|}{\binom{n}{\alpha n}}$  and run time

$$p^{-1}|\mathcal{S}|^{\frac{1}{2}} = \frac{\binom{n}{\alpha n}}{|\mathcal{S}|^{\frac{1}{2}}} = 2^{(H(\alpha) - \frac{1}{2}H(\phi))n}.$$

- $\alpha = \frac{1}{2}$ : Complexity  $2^{\frac{1}{2}n}$ .

# Parallel Collision Search

## Theorem PCS

Given functions  $f_0, \dots, f_k : \{0, 1\}^n \rightarrow \{0, 1\}^n$ . We find a collision between  $f_0$  and all other  $f_1, \dots, f_k$  in time  $\tilde{O}(\sqrt{k}2^{\frac{n}{2}})$  with space  $\tilde{O}(k)$ .

## Multiple Dlog

- Let  $\beta_1 = g^{x_1}, \dots, \beta_k = g^{x_k}$ . Define functions  $\mathbb{Z}_{|G|} \rightarrow G$  with

$$f_0 : x_0 \mapsto g^{x_0} \text{ and } f_i : x_i \mapsto \beta_i \cdot g^{-x_i} \text{ for } i = 1, \dots, k.$$

- Collision  $(x_0, x_i)$  solves  $i^{\text{th}}$  dlog instance.
- Run time is  $\tilde{O}(\sqrt{k|G|})$  with space only  $\tilde{O}(k)$ .

# Subset Sum

## Problem Subset Sum

**Given:**  $\mathbf{a} = (a_1, \dots, a_n) \in \mathbb{Z}_{2^n}^n$ ,  $t \in \mathbb{Z}_{2^n}$

**Find:**  $\mathbf{e} = (e_1, \dots, e_n) \in \{0, 1\}^n$  with  $\sum_{i=1}^n e_i a_i = t \pmod{2^n}$

### Cryptanalysis basics:

- Brute Force:  $\tilde{O}(2^n)$
- Meet-in-the-Middle:  $\tilde{O}(2^{\frac{n}{2}})$
- **Questions:** Low Memory Algorithms? Faster?

### Exercise:

Express DLP as a subset sum problem in  $(G, \cdot)$  instead of  $(\mathbb{Z}_{2^n}, +)$ .



# Schroepel-Shamir algorithm (1979)

**Idea:** Write  $\sum_{i=1}^{\frac{n}{4}} e_i a_i + \sum_{i=\frac{n}{4}+1}^{\frac{1}{2}n} e_i a_i = t - \sum_{i=\frac{3}{4}n}^{\frac{1}{2}n} e_i a_i - \sum_{i=\frac{3}{4}n+1}^n e_i a_i$ .

## Algorithm 4-List algorithm

**Input:**

- 1 Generate lists  $L_1, \dots, L_4$  with

$$L_1 = \left\{ \sum_{i=1}^{\frac{n}{4}} e_i a_i \mid (e_1, \dots, e_{\frac{n}{4}}) \in \{0, 1\}^{\frac{n}{4}} \right\}, \text{ etc.}$$

- 2 Repeat

- 1 Choose  $r \in_R \mathbb{Z}_{2^{\frac{n}{4}}}$ .

- 2 Compute  $L_{12} = L_1 \bowtie_{\frac{n}{4}} L_2 := \left\{ \sum_{i=1}^{\frac{n}{2}} e_i a_i \mid \sum_{i=1}^{\frac{n}{2}} e_i a_i = r \pmod{2^{\frac{n}{4}}} \right\}$  and  $L_{34} = L_3 \bowtie_{\frac{n}{4}} L_4 := \left\{ t - \sum_{i=\frac{n}{2}+1}^n e_i a_i \mid t - \sum_{i=\frac{n}{2}+1}^n e_i a_i = r \pmod{2^{\frac{n}{4}}} \right\}$ .

- 3 Compute  $L = L_{12} \bowtie_n L_{34} := \left\{ \sum_{i=1}^n e_i a_i \mid \sum_{i=1}^n e_i a_i = t \pmod{2^n} \right\}$

- 3 Until  $|L| \neq \emptyset$

**Output:** e from L

# Analysis Shamir-Shroepel

## Correctness (Termination):

- Let  $\mathbf{e}$  be a subset sum solution. Let  $r = \sum_{i=1}^{n/2} e_i a_i \bmod 2^{n/4}$ .
- Assume that we choose  $r$  in Step 2.2.
- Then our algorithm terminates with output  $\mathbf{e}$ .

## Run time:

- Each iteration costs on expectation  $\tilde{O}(2^{n/4})$  time/memory.
- On expectation, it takes  $2^{n/4}$  iterations for finding  $r$ .

**Question:** Is there a  $\tilde{O}(1)$  memory algorithm faster than brute-force?

## 0.75 Subset Sum

**Idea:** Use collision finding in  $f_1, f_2 : \{0, 1\}^{\frac{n}{2}} \rightarrow \mathbb{Z}_{2^{\frac{n}{2}}}$  with

$$f_1 : (e_1, \dots, e_{n/2}) \mapsto \sum_{i=1}^{n/2} e_i a_i \pmod{2^{n/2}} \text{ and}$$

$$f_2 : (e_{n/2+1}, \dots, e_n) \mapsto t - \sum_{i=n/2+1}^n e_i a_i \pmod{2^{n/2}}.$$

### Algorithm Subset Sum with Low Memory

**Input:**  $f_1, f_2$

① **Repeat**

① Find a random collision  $(x_1, x_2)$  in  $f_1, f_2$

② **Until**  $\sum_{i=1}^{n/2} e_i a_i = t - \sum_{i=n/2+1}^n e_i a_i$

**Output:**  $e$

# Analysis

## Run time:

- We have  $f_i : \{0, 1\}^{\frac{n}{2}} \rightarrow \mathbb{Z}_{2^{\frac{n}{2}}}$  with search space size  $|\mathcal{S}| = 2^{\frac{n}{2}}$ .
- We expect that  $f_1, f_2$  have  $|\mathcal{S}| = 2^{\frac{n}{2}}$  many collisions.
- Since we uniquely represent  $\mathbf{e}$ , only a single collision is good.
- Need on expectation  $\tilde{O}(2^{\frac{n}{2}})$  iteration with cost  $\tilde{O}(2^{\frac{n}{4}})$  each.

**Exercise:** Generalize to solutions  $\mathbf{e}$  with  $\text{wt}(\mathbf{e}) = \alpha$ .

## 0.72 Algorithm (Becker, Coron, Joux 2011)

### Idea:

- Represent  $\mathbf{e} = \mathbf{e}_1 + \mathbf{e}_2$  with  $\mathbf{e}_1, \mathbf{e}_2 \in \{0, 1\}^n$ ,  $\text{wt}(\mathbf{e}_i) = \frac{n}{4}$ .
- Let  $\mathcal{S} := \{\mathbf{e}' \in \{0, 1\}^n \mid \text{wt}(\mathbf{e}') = \frac{n}{4}\}$  with  $|\mathcal{S}| = \binom{n}{n/4} \approx 2^{0.811n}$ .
- Use collision finding in  $f_1, f_2 : \mathcal{S} \rightarrow \mathbb{Z}_{|\mathcal{S}|}$  with

$$f_1 : (\mathbf{e}_1, \dots, \mathbf{e}_n) \mapsto \sum_{i=1}^n e_i a_i \bmod 2^{0.811n} \text{ and}$$

$$f_2 : (\mathbf{e}_1, \dots, \mathbf{e}_n) \mapsto t - \sum_{i=1}^n e_i a_i \bmod 2^{0.811n}.$$

### Algorithm Subset Sum with Low Memory

**Input:**  $f_1, f_2$

**1 Repeat**

**1** Find a random collision  $(x_1, x_2)$  in  $f_1, f_2$

**2 Until**  $\sum_{i=1}^{n/2} e_i a_i = t - \sum_{i=n/2+1}^n e_i a_i \bmod 2^{0.811n}$

**Output:**  $\mathbf{e}$

# Analysis

## Run Time:

- There are  $\binom{n/2}{n/4} = \tilde{\Theta}(2^{\frac{n}{2}})$  representations  $\mathbf{e}$ .
- Overall run time is

$$\tilde{O}(|\mathcal{S}|) \cdot \frac{\binom{n/2}{n/4}}{|\mathcal{S}|} = \frac{|\mathcal{S}|^{\frac{3}{2}}}{\binom{n/2}{n/4}} = \tilde{O}(2^{0.72n}).$$

## Remarks:

- Hash function  $h$  from DLP is now simply the ring homomorphism

$$\mathbb{Z}_{2^n} \rightarrow \mathbb{Z}_{2^{0.811n}}, \quad x \bmod 2^n \mapsto x \bmod 2^{0.811n}.$$

- Hence subset sum allows more (subgroup) structure than DLP.
- Especially we can do a nested collision finding on the whole  $\mathbb{Z}_{2^n}$ .

## Theorem Esser, May (2019)

Subset Sum can be solved in time  $2^{0.65n}$  and space  $\tilde{O}(1)$ .

# Howgrave-Graham Joux (2010)

## Idea:

- Represent  $\mathbf{e} = \mathbf{e}_1 + \mathbf{e}_2$  with  $\mathbf{e}_1, \mathbf{e}_2 \in \{0, 1\}^n$ ,  $\text{wt}(\mathbf{e}_i) = \frac{n}{4}$ .
- Let  $\mathcal{S}_1 := \{\mathbf{e}' \in \{0, 1\}^n \mid \text{wt}(\mathbf{e}') = \frac{n}{4}\}$  with  $|\mathcal{S}_1| = \binom{n}{n/4} \approx 2^{0.811n}$ .
- We have  $R_1 = \binom{n/2}{n/4}$  representations of  $\mathbf{e}$ . Then  $\log R_1 \approx \frac{n}{2}$ . Define

$$L_1 = \left\{ \sum_{i=1}^n e_i a_i \mid \mathbf{e} \in \mathcal{S}, \sum_{i=1}^n e_i a_i = 0 \pmod{2^{\frac{n}{2}}} \right\},$$

$$L_2 = \left\{ t - \sum_{i=1}^n e_i a_i \mid \mathbf{e} \in \mathcal{S}, \sum_{i=1}^n t - e_i a_i = 0 \pmod{2^{\frac{n}{2}}} \right\}.$$

- Then (on expectation) there exists a representation in  $L_1 \times L_2$ .
- We have  $|L_1| = |L_2| = 2^{0.311n}$ . Thus, we require at least time  $2^{0.311n}$ .
- **Observe:** Constructing  $L_1, L_2$  is again a subset sum problem.

Getting below  $2^{\frac{n}{2}}$ .

### Algorithm Subset Sum 1

**Input:**  $a_1, \dots, a_n, t$

- 1 Construct  $L_1, L_2$  with Schroeppe-Shamir.
- 2 Construct  
 $L = (L_1 \times L_2) \cap \{\mathbf{e}' = \mathbf{e}_1 + \mathbf{e}_2 \mid \mathbf{e}_i \in \mathcal{S}, \sum_{i=1}^n e_i a_i = t \pmod{2^n}\}$ .

**Output:**  $L \cap \{0, 1\}^n$

**Run Time:**

- Step 1 runs in time  $2^{0.406n}$ .
- We expect

$$|L| = \frac{|L_1| \cdot |L_2|}{2^{\frac{n}{2}}} = 2^{0.122n}.$$

- We can construct  $L$  in time  $\tilde{O}(\max\{|L_1|, |L_2|, |L|\}) = 2^{0.311n}$ .
- Therefore, we obtain total run time  $2^{0.406n}$ .

**Idea:** Construct  $L_1, L_2$  recursively with algorithm Subset Sum 1.



## One more iteration

- We show how to construct  $L_1$ ,  $L_2$  is analogous. Recall that  $L_1 = \left\{ \sum_{i=1}^n e_i a_i \mid \mathbf{e} \in \{0, 1\}^n, \text{wt}(\mathbf{e}) = \frac{n}{4}, \sum_{i=1}^n e_i a_i = 0 \pmod{2^{\frac{n}{2}}} \right\}$ .
- Represent  $\mathbf{e} = \mathbf{e}_1 + \mathbf{e}_2$  with  $\mathbf{e}_1, \mathbf{e}_2 \in \{0, 1\}^n, \text{wt}(\mathbf{e}_i) = \frac{n}{8}$ .
- Let  $\mathcal{S}_2 := \{\mathbf{e}' \in \{0, 1\}^n \mid \text{wt}(\mathbf{e}') = \frac{n}{8}\}$  with  $|\mathcal{S}| = \binom{n}{n/8} \approx 2^{0.5435n}$ .
- We have  $R_2 = \binom{n/4}{n/8}$  representations of  $\mathbf{e}$ . Then  $\log_2 R \approx \frac{n}{4}$ . Define

$$L_{11} = \left\{ \sum_{i=1}^n e_i a_i \mid \mathbf{e} \in \mathcal{S}_2, \sum_{i=1}^n e_i a_i = 0 \pmod{2^{\frac{n}{4}}} \right\}.$$

- Then (on expectation) there exists a representation in  $L_{11} \times L_{11}$ .
- We expect that  $|L_{11}| = 2^{0.2935n}$ .

# Getting to $2^{0.337n}$

## Algorithm Howgrave-Graham Joux (2010)

**Input:**  $a_1, \dots, a_n, t$

① Construct  $L_1, L_2$  with Algorithm Subset Sum 1.

② Construct

$$L = (L_1 \times L_2) \cap \{\mathbf{e}' = \mathbf{e}_1 + \mathbf{e}_2 \mid \mathbf{e}_i \in \mathcal{S}, \sum_{i=1}^n e_i a_i = t \pmod{2^n}\}.$$

**Output:**  $L \cap \{0, 1\}^n$

**Run Time:**

- We expect  $|L_1| = \frac{|L_{11}| \cdot |L_{11}|}{2^{\frac{n}{4}}} = 2^{2 \cdot 0.2935n - 0.25n} = 2^{0.337n}$ .

## Theorem Run Time of HGJ algorithm

HGJ solves subset sum instances in  $2^{0.337n}$ .

# The Becker-Coron-Joux algorithm (2011)

**Idea** of the BCJ algorithm:

- Represent  $\mathbf{e} = \mathbf{e}_1 + \mathbf{e}_2$  with  $\mathbf{e}_i \in \{-1, 0, 1\}^n$ .
- Advantage: Many more representations as in HGJ.
- Disadvantage: Also more sums that do not end up in  $\{0, 1\}^n$ .

## Theorem Becker-Coron-Joux (2011)

BCJ solves subset sum in  $2^{0.291n}$ .

## Theorem Esser, May (2019)

Subset Sum can be solved in  $2^{0.255n}$ .

- See <https://arxiv.org/abs/1907.04295>.
- Technique: Sampling instead of enumeration.