

Algorithms for Bichromatic Closest Pairs Problem and application to Code-based Cryptography

M. Hamdad

April, 2025

The Bichromatic Closest Pairs Problem and application to cryptanalysis

- Boolean context

The Bichromatic Closest Pairs Problem and application to cryptanalysis

- Boolean context
- Syndrome Decoding Problem (SDP) :
Find x such that $Hx = s$ with $wt(x) \leq w$ (NP-hard)

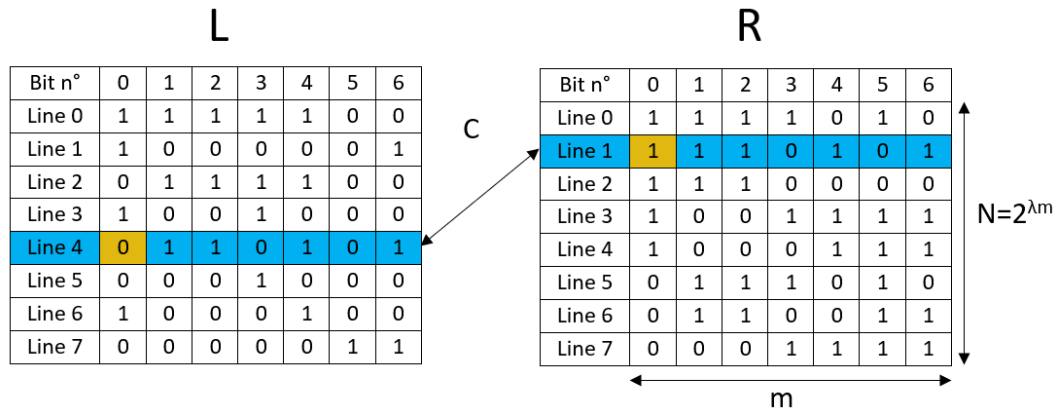
The Bichromatic Closest Pairs Problem and application to cryptanalysis

- Boolean context
- Syndrome Decoding Problem (SDP) :
Find x such that $Hx = s$ with $wt(x) \leq w$ (NP-hard)
- Solve SDP faster
 \implies improve McEliece cryptosystem's cryptanalysis

The Bichromatic Closest Pairs Problem and application to cryptanalysis

- Boolean context
- Syndrome Decoding Problem (SDP) :
Find x such that $Hx = s$ with $wt(x) \leq w$ (NP-hard)
- Solve SDP faster
 \implies improve McEliece cryptosystem's cryptanalysis
- The best-known algorithms use in a crucial way a subroutine that solves $BCPP$

Bichromatic Closest Pairs Problem in \mathbb{F}_2^m



Goal: Find all $C = (x, y) \in L \times R$ such that $wt(x + y) \leq \gamma m$
 Here $\gamma m = 1$

Bichromatic Closest Pairs Problem in \mathbb{F}_2^m

- The model : $\mathcal{M}_{Alea}(\mathbb{F}_2, 2^{\lambda m})$
- The expected number of solutions : $E = \Theta\left(m^{-\frac{1}{2}} 2^{m(2\lambda + H(\gamma) - 1)}\right)^1$
- Asymptotically, the best algorithms : Carrier [Car20], Esser et. al [EKZ21] and May-Ozerov [MO15]

¹H : Binary entropy function

Contribution [BDH25]

The May-Ozerov algorithm is galactic.

- In practice : Syndrome Decoding Estimator [EB21] \implies The projection method
- **The Crossover Point:**
The smallest code length n such that $\text{Decode}(\text{MO})$ can be faster than $\text{Decode}(\text{Projection})$.

R	D	n	MO $\log_2 N$	$\log_2 T$
0.5	0.11	533502	8121	29566
0.8	0.03	1874400	22282	63487

Table: Instance characteristics at the crossover point.

The projection method

Probability that 2 bit strings we search coincide on k columns: $p = \frac{\binom{(1-\gamma)m}{k}}{\binom{m}{k}}$

The projection method

Probability that 2 bit strings we search coincide on k columns: $p = \frac{\binom{(1-\gamma)m}{k}}{\binom{m}{k}}$

The algorithm

- Pick k columns randomly
- Sort the 2 lists in **lexicographical order** according to the selected columns
- Compare all pairs of bit strings that coincide on the k columns
- Repeat $\simeq \frac{1}{p}$ times

$k = 2$, drawn column numbers = $\{0, 2\}$

L sorted

Bit n°	2	0	1	3	4	5	6
Line 0	0	0	0	1	0	0	0
Line 1	0	0	0	0	0	1	1
Line 2	0	1	0	0	0	0	1
Line 3	0	1	0	1	0	0	0
Line 4	0	1	0	0	1	0	0
Line 5	1	0	1	1	1	0	0
Line 6	1	0	1	0	1	0	1
Line 7	1	1	1	1	1	0	0

C

R sorted

Bit n°	2	0	1	3	4	5	6
Line 0	0	0	0	1	1	1	1
Line 1	0	1	0	1	1	1	1
Line 2	0	1	0	0	1	1	1
Line 3	1	0	1	1	0	1	0
Line 4	1	0	1	0	0	1	1
Line 5	1	1	1	1	0	1	0
Line 6	1	1	1	0	1	0	1
Line 7	1	1	1	0	0	0	0

$k = 2$, drawn column numbers = $\{1, 4\}$

L sorted

Bit n°	4	1	0	2	3	5	6
Line 0	0	0	1	0	0	0	1
Line 1	0	0	1	0	1	0	0
Line 2	0	0	0	0	0	1	1
Line 3	0	0	0	0	1	0	0
Line 4	1	0	1	0	0	0	0
Line 5	1	1	1	1	1	0	0
Line 6	1	1	0	1	1	0	0
Line 7	1	1	0	1	0	0	1

C

R sorted

Bit n°	4	1	0	2	3	5	6
Line 0	0	1	1	1	1	1	0
Line 1	0	1	1	1	0	0	1
Line 2	0	1	0	1	1	1	0
Line 3	0	1	0	1	0	1	1
Line 4	1	0	1	0	1	1	1
Line 5	1	0	1	0	0	1	1
Line 6	1	0	0	0	1	1	1
Line 7	1	1	1	1	0	0	1

The expected number of solutions : E

Complexity

$$C_{Proj} = O \left(\left(N + \frac{N^2}{2^k} \right) \frac{1}{p} \right)$$

The expected number of solutions : E

Complexity

$$C_{Proj} = O \left(\left(N + \frac{N^2}{2^k} \right) \frac{1}{p} \right)$$

If $\lambda \leq \bar{\gamma}$, we take $k = \lambda m$ else $k = (1 - 2\gamma)m$ then :

$$C_{Proj} = \begin{cases} O(2^{m(\lambda + h(\lambda, \gamma))}) & \text{for } \lambda < \bar{\gamma} \\ O(\sqrt{m}E) & \text{for } \lambda \geq \bar{\gamma} \end{cases}$$

with

$$\bar{\gamma} = (1 - 2\gamma) \text{ and } h(\lambda, \gamma) = H(\lambda) - (1 - \gamma)H\left(\frac{\lambda}{1 - \gamma}\right)$$

Carrier [Car20], Esser et. al [EKZ21] and May-Ozerov [MO15] are similar.

A high-level description

- m columns are partitioned into t strips.
- The vectors of both lists are filtered **strip after strip**.
- Filtered across their m coordinates, the lists are **small enough** for **an exhaustive search** to find solution pairs

The May-Ozerov algorithm

Three core phases

- 1 Computing the parameters of the algorithm.
- 2 Double rerandomization.
- 3 Recursive search for solutions in a tree of depth t .

Algorithm 1 The MO algorithm

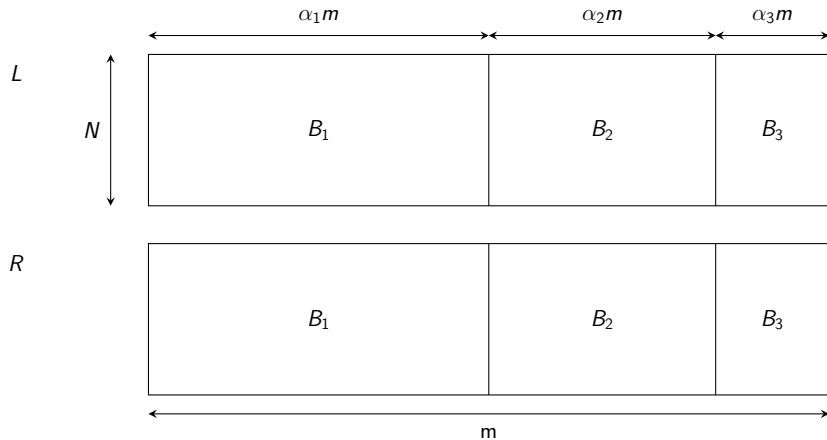
```
1: function MO(  $L, R, \lambda, \gamma, m, t, \epsilon$  )
2:    $y(\gamma, \lambda) \leftarrow (1 - \gamma) \left( 1 - H \left( \frac{H^{-1}(1-\lambda) - \gamma/2}{1-\gamma} \right) \right)$ 
3:    $\alpha_1 \leftarrow (y(\gamma, \lambda) - \lambda + \epsilon/2) / y(\gamma, \lambda)$ 
4:   for  $2 \leq j \leq t$  do
5:      $\alpha_j \leftarrow \frac{\lambda}{y(\gamma, \lambda)} \alpha_{j-1}$  ▷ Divide the lists into  $t$  strips of  $\alpha_j m$  indices
6:   for  $f_1(m)$  uniformly random permutation  $\pi$  of  $\{1, \dots, m\}$  do
7:     for  $f_2(m)$  times do
8:        $r = (r_1, \dots, r_t) \leftarrow (\text{RANDOM}(\mathbb{F}_2^{\alpha_j m}))_{j=1}^t$  s.t.  $\text{wt}(r_j) = \alpha_j \frac{m}{2}$ 
9:        $\bar{L} \leftarrow \pi(L) + r$ 
10:       $\bar{R} \leftarrow \pi(R) + r$ 
11:      Remove from  $\bar{L}$  and  $\bar{R}$  all vectors that are not of weight  $\alpha_j \frac{m}{2}$  on the  $j$ -th strip
12:       $C \leftarrow \text{RECURSIVEMO}(\bar{L}, \bar{R}, m, t, \epsilon, \gamma, \lambda, (\alpha)_{j=1}^t, 1)$ 
13:      if  $C \neq \perp$  then return  $C$ 
14: return  $\perp$ 
```

Parameter setup:

- $y(\gamma, \lambda)$
- Divide the lists into t strips B_1, \dots, B_t containing $\alpha_1 m, \dots, \alpha_t m$ indices respectively, such that:

$$\sum_{j=1}^t \alpha_j m = m$$

$t = 3$



Algorithm 1 The MO algorithm

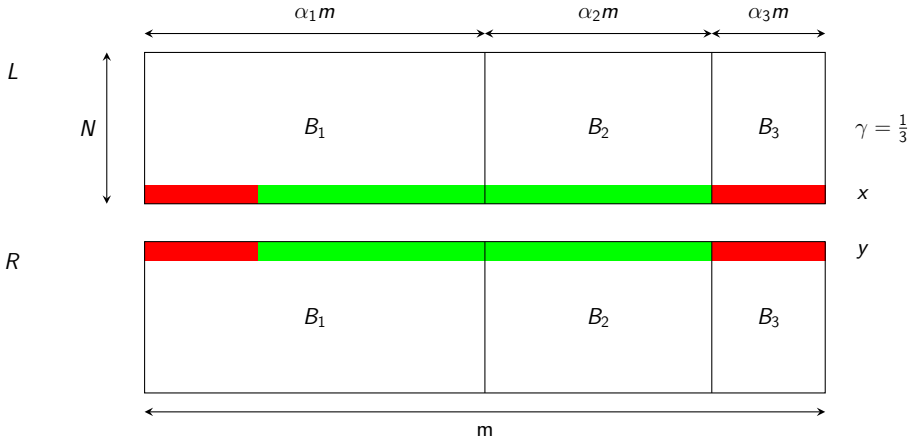
```
1: function MO(  $L, R, \lambda, \gamma, m, t, \epsilon$  )
2:    $y(\gamma, \lambda) \leftarrow (1 - \gamma) \left( 1 - H \left( \frac{H^{-1}(1-\lambda) - \gamma/2}{1-\gamma} \right) \right)$ 
3:    $\alpha_1 \leftarrow (y(\gamma, \lambda) - \lambda + \epsilon/2) / y(\gamma, \lambda)$ 
4:   for  $2 \leq j \leq t$  do
5:      $\alpha_j \leftarrow \frac{\lambda}{y(\gamma, \lambda)} \alpha_{j-1}$  ▷ Divide the lists into  $t$  strips of  $\alpha_j m$  indices
6:   for  $f_1(m)$  uniformly random permutation  $\pi$  of  $\{1, \dots, m\}$  do
7:     for  $f_2(m)$  times do
8:        $r = (r_1, \dots, r_t) \leftarrow (\text{RANDOM}(\mathbb{F}_2^{\alpha_j m}))_{j=1}^t$  s.t.  $\text{wt}(r_j) = \alpha_j \frac{m}{2}$ 
9:        $\bar{L} \leftarrow \pi(L) + r$ 
10:       $\bar{R} \leftarrow \pi(R) + r$ 
11:      Remove from  $\bar{L}$  and  $\bar{R}$  all vectors that are not of weight  $\alpha_j \frac{m}{2}$  on the  $j$ -th strip
12:       $C \leftarrow \text{RECURSIVEMO}(\bar{L}, \bar{R}, m, t, \epsilon, \gamma, \lambda, (\alpha)_{j=1}^t, 1)$ 
13:      if  $C \neq \perp$  then return  $C$ 
14:   return  $\perp$ 
```

u_j : the $|B_j|$ -bit vector of u 's entries indexed by B_j

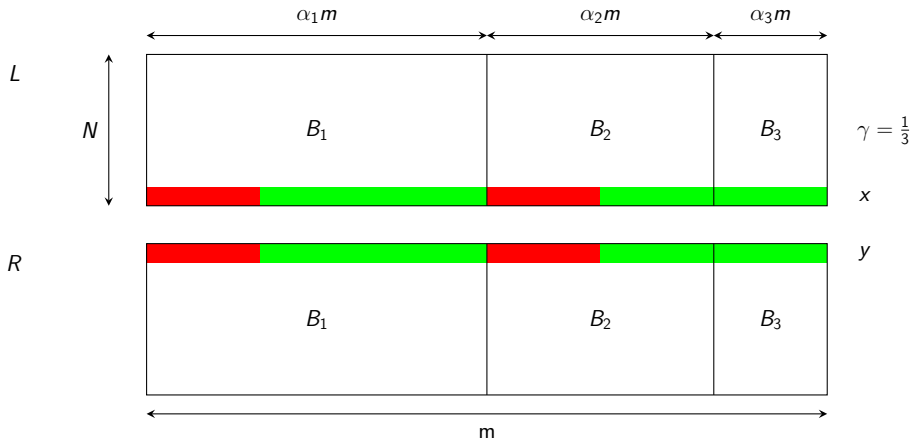
Double rerandomisation

The recursive search requires the particular solution (x, y) to satisfy the following conditions.
For all $1 \leq j \leq t$:

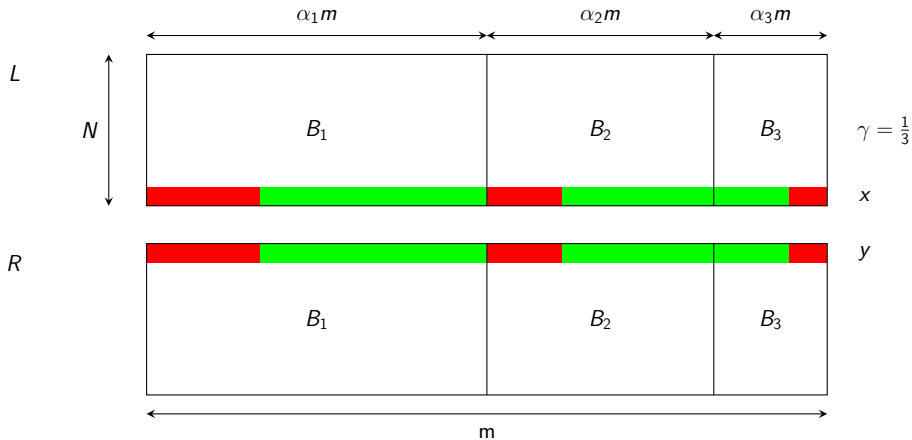
- i) $d(x_j, y_j) = \gamma |B_j| = \gamma \alpha_j m$;
- ii) $wt(x_j) = wt(y_j) = \frac{\alpha_j m}{2}$.

$t = 3$ 

$t = 3$



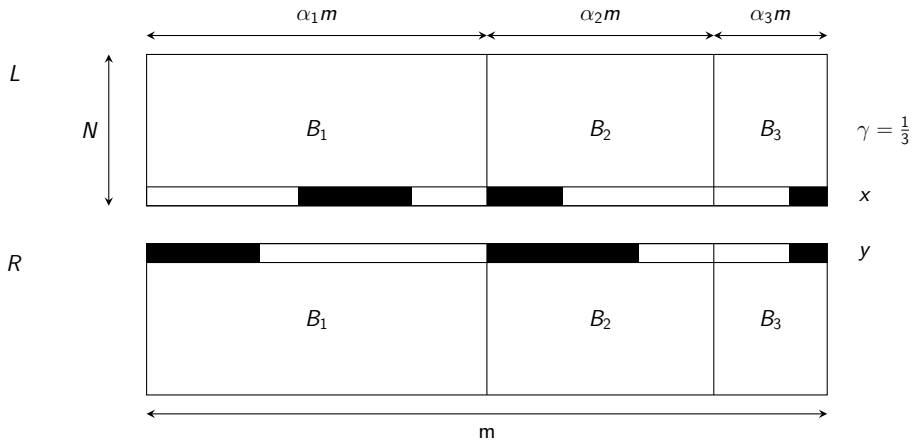
$t = 3$



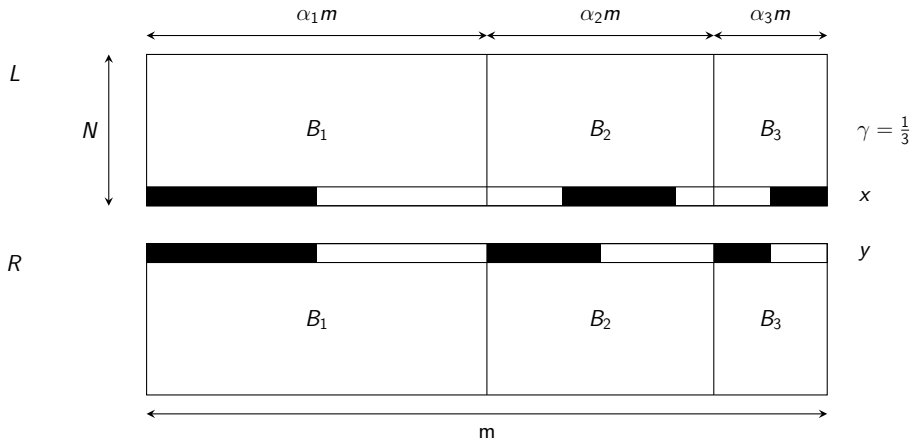
Algorithm 1 The MO algorithm

```
1: function MO(  $L, R, \lambda, \gamma, m, t, \epsilon$  )
2:    $y(\gamma, \lambda) \leftarrow (1 - \gamma) \left( 1 - H \left( \frac{H^{-1}(1-\lambda) - \gamma/2}{1-\gamma} \right) \right)$ 
3:    $\alpha_1 \leftarrow (y(\gamma, \lambda) - \lambda + \epsilon/2) / y(\gamma, \lambda)$ 
4:   for  $2 \leq j \leq t$  do
5:      $\alpha_j \leftarrow \frac{\lambda}{y(\gamma, \lambda)} \alpha_{j-1}$  ▷ Divide the lists into  $t$  strips of  $\alpha_j m$  indices
6:   for  $f_1(m)$  uniformly random permutation  $\pi$  of  $\{1, \dots, m\}$  do
7:     for  $f_2(m)$  times do
8:        $r = (r_1, \dots, r_t) \leftarrow (\text{RANDOM}(\mathbb{F}_2^{\alpha_j m}))_{j=1}^t$  s.t.  $\text{wt}(r_j) = \alpha_j \frac{m}{2}$ 
9:        $\bar{L} \leftarrow \pi(L) + r$ 
10:       $\bar{R} \leftarrow \pi(R) + r$ 
11:      Remove from  $\bar{L}$  and  $\bar{R}$  all vectors that are not of weight  $\alpha_j \frac{m}{2}$  on the  $j$ -th strip
12:       $C \leftarrow \text{RECURSIVEMO}(\bar{L}, \bar{R}, m, t, \epsilon, \gamma, \lambda, (\alpha)_{j=1}^t, 1)$ 
13:      if  $C \neq \perp$  then return  $C$ 
14: return  $\perp$ 
```

$t = 3$



$t = 3$



Algorithm 1 The MO algorithm

```
1: function MO(  $L, R, \lambda, \gamma, m, t, \epsilon$  )
2:    $y(\gamma, \lambda) \leftarrow (1 - \gamma) \left( 1 - H \left( \frac{H^{-1}(1-\lambda) - \gamma/2}{1-\gamma} \right) \right)$ 
3:    $\alpha_1 \leftarrow (y(\gamma, \lambda) - \lambda + \epsilon/2) / y(\gamma, \lambda)$ 
4:   for  $2 \leq j \leq t$  do
5:      $\alpha_j \leftarrow \frac{\lambda}{y(\gamma, \lambda)} \alpha_{j-1}$  ▷ Divide the lists into  $t$  strips of  $\alpha_j m$  indices
6:   for  $f_1(m)$  uniformly random permutation  $\pi$  of  $\{1, \dots, m\}$  do
7:     for  $f_2(m)$  times do
8:        $r = (r_1, \dots, r_t) \leftarrow (\text{RANDOM}(\mathbb{F}_2^{\alpha_j m}))_{j=1}^t$  s.t.  $wt(r_j) = \alpha_j \frac{m}{2}$ 
9:        $\bar{L} \leftarrow \pi(L) + r$ 
10:       $\bar{R} \leftarrow \pi(R) + r$ 
11:      Remove from  $\bar{L}$  and  $\bar{R}$  all vectors that are not of weight  $\alpha_j \frac{m}{2}$  on the  $j$ -th strip
12:       $C \leftarrow \text{RECURSIVEMO}(\bar{L}, \bar{R}, m, t, \epsilon, \gamma, \lambda, (\alpha)_{j=1}^t, 1)$ 
13:      if  $C \neq \perp$  then return  $C$ 
14:   return  $\perp$ 
```

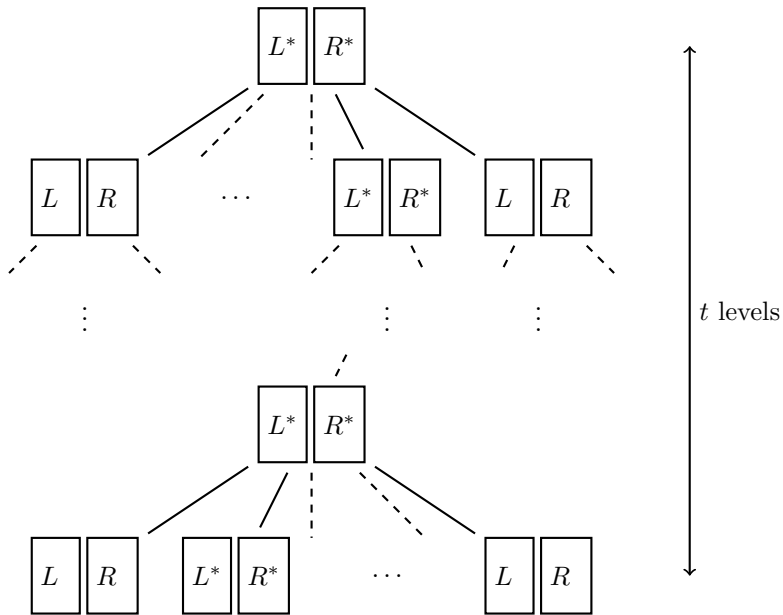
Lemma 1 [BDH25]

Let N_{it} the number of iterations of the double rerandomization. The particular solution (x, y) satisfies Conditions i) and ii) in at least one of the iterations with probability greater than $1/4$ only if

$$N_{it} = f_1(m)f_2(m) \geq \frac{1}{8\sqrt{2}} \left(\frac{\pi^{\frac{3}{2}}}{2} \right)^t m^{t-\frac{1}{2}} (\gamma(1-\gamma))^{t-\frac{1}{2}} \left(\frac{y(\gamma, \lambda) - \lambda + \frac{\epsilon}{2}}{y(\gamma, \lambda)} \right)^t \left(\frac{\lambda}{y(\gamma, \lambda)} \right)^{\frac{t(t-1)}{2}}.$$

Algorithm 1 The MO algorithm

```
1: function MO(  $L, R, \lambda, \gamma, m, t, \epsilon$  )
2:    $y(\gamma, \lambda) \leftarrow (1 - \gamma) \left( 1 - H \left( \frac{H^{-1}(1-\lambda) - \gamma/2}{1-\gamma} \right) \right)$ 
3:    $\alpha_1 \leftarrow (y(\gamma, \lambda) - \lambda + \epsilon/2) / y(\gamma, \lambda)$ 
4:   for  $2 \leq j \leq t$  do
5:      $\alpha_j \leftarrow \frac{\lambda}{y(\gamma, \lambda)} \alpha_{j-1}$  ▷ Divide the lists into  $t$  strips of  $\alpha_j m$  indices
6:   for  $f_1(m)$  uniformly random permutation  $\pi$  of  $\{1, \dots, m\}$  do
7:     for  $f_2(m)$  times do
8:        $r = (r_1, \dots, r_t) \leftarrow (\text{RANDOM}(\mathbb{F}_2^{\alpha_j m}))_{j=1}^t$  s.t.  $\text{wt}(r_j) = \alpha_j \frac{m}{2}$ 
9:        $\bar{L} \leftarrow \pi(L) + r$ 
10:       $\bar{R} \leftarrow \pi(R) + r$ 
11:      Remove from  $\bar{L}$  and  $\bar{R}$  all vectors that are not of weight  $\alpha_j \frac{m}{2}$  on the  $j$ -th strip
12:       $C \leftarrow \text{RECURSIVEMO}(\bar{L}, \bar{R}, m, t, \epsilon, \gamma, \lambda, (\alpha)_{j=1}^t, 1)$ 
13:      if  $C \neq \perp$  then return  $C$ 
14: return  $\perp$ 
```



At a node at depth j , filtering is performed on B_j

Filtering on B_j

- Pick a random subset A of $\frac{\alpha_j m}{2}$ indices inside strip j
- $L' \leftarrow \{u \in L \text{ s.t. } wt(u_A) = H^{-1}(1 - \lambda) \frac{\alpha_j m}{2}\}$
- $R' \leftarrow \{v \in R \text{ s.t. } wt(v_A) = H^{-1}(1 - \lambda) \frac{\alpha_j m}{2}\}$

$$\mathbb{P}[(x, y) \in L' \times R' \mid (x, y) \in L \times R] = \frac{1}{s_j} = \tilde{O}\left(2^{-\alpha_j y(\gamma, \lambda)m}\right)$$

$$\mathbb{P}[u \in L' \mid u \in L] = p_j = \tilde{O}\left(2^{-\lambda \alpha_j m}\right)$$

$$\mathbb{P}[(x, y) \in L' \times R' \mid (x, y) \in L \times R] = \frac{1}{s_j} = \tilde{O}\left(2^{-\alpha_j \gamma(\gamma, \lambda)m}\right)$$

$$\mathbb{P}[u \in L' \mid u \in L] = p_j = \tilde{O}\left(2^{-\lambda \alpha_j m}\right)$$

- Filtering ms_j times on strip B_j
 \implies the particular solution (x, y) survives with **overwhelming probability**.

$$\mathbb{P}[(x, y) \in L' \times R' \mid (x, y) \in L \times R] = \frac{1}{s_j} = \tilde{O}\left(2^{-\alpha_j \gamma(\gamma, \lambda) m}\right)$$

$$\mathbb{P}[u \in L' \mid u \in L] = p_j = \tilde{O}\left(2^{-\lambda \alpha_j m}\right)$$

- Filtering ms_j times on strip B_j
 \implies the particular solution (x, y) survives with **overwhelming probability**.
- $\mathbb{E}[\#R'] = \mathbb{E}[\#L'] = N \prod_{i=1}^j p_i = \tilde{O}\left(2^{\lambda m(1 - \sum_{i=1}^j \alpha_j)}\right)$

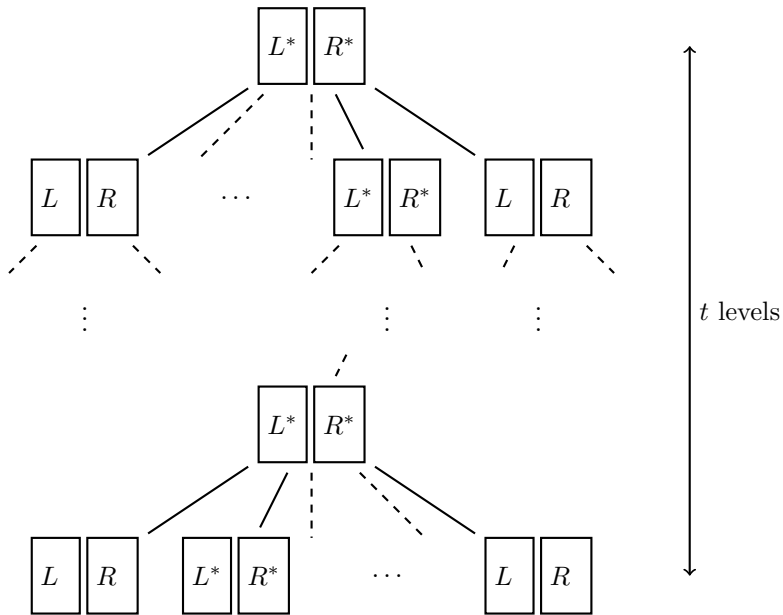
$$\mathbb{P}[(x, y) \in L' \times R' \mid (x, y) \in L \times R] = \frac{1}{s_j} = \tilde{O}\left(2^{-\alpha_j \gamma(\gamma, \lambda)m}\right)$$

$$\mathbb{P}[u \in L' \mid u \in L] = p_j = \tilde{O}\left(2^{-\lambda \alpha_j m}\right)$$

- Filtering ms_j times on strip B_j
 \implies the particular solution (x, y) survives with **overwhelming probability**.
- $\mathbb{E}[\#R'] = \mathbb{E}[\#L'] = N \prod_{i=1}^j p_i = \tilde{O}\left(2^{\lambda m(1 - \sum_{i=1}^j \alpha_j)}\right)$
- Let $X = \#R'$ (Tchebitchev inequality)
 $\implies \mathbb{P}\left[X - \mathbb{E}[X] \geq 2^{\frac{\epsilon}{2}m} \mathbb{E}[X]\right] \leq 2^{-\epsilon m}$

Algorithm 2 RECURSIVEMO

```
1: function RECURSIVEMO( $L, R, m, t, \epsilon, \lambda, \gamma, (\alpha)_1^t, j$ )
2:   if  $j = t + 1$  then
3:     return QUADRATICNN( $L, R, \gamma m$ )
4:    $C \leftarrow \perp$ 
5:    $s_j \leftarrow \frac{\binom{\alpha_j m}{\frac{1}{2}\alpha_j m}}{\binom{(1-\gamma)\frac{\alpha_j m}{2}}{(1-h-\frac{\gamma}{2})\frac{\alpha_j m}{2}} \binom{(1-\gamma)\frac{\alpha_j m}{2}}{(h-\frac{\gamma}{2})\frac{\alpha_j m}{2}} \left(\gamma\frac{\alpha_j m}{2}\right)^2}$ 
6:   for  $ms_j$  times do
7:     Pick a random subset  $A$  of  $\frac{\alpha_j m}{2}$  indices inside strip  $j$ 
8:      $L' \leftarrow \{u \in L \text{ s.t. } wt(u_A) = H^{-1}(1-\lambda)\frac{\alpha_j m}{2}\}$ 
9:      $R' \leftarrow \{v \in R \text{ s.t. } wt(v_A) = H^{-1}(1-\lambda)\frac{\alpha_j m}{2}\}$ 
10:    if  $|L'|$  and  $|R'|$  are not too big then
11:       $x \leftarrow \text{RECURSIVEMO}(L', R', m, t, \epsilon, \lambda, \gamma, (\alpha)_1^t, j+1)$ 
12:      if  $x \neq \perp$  then
13:        return  $x$ 
14:  return  $C$ 
```



For all $2 \leq j \leq t$

Total complexity at depth j

$$T_j = \tilde{O} \left((2^m)^{\sum_{i=1}^{j-1} \alpha_i y(\gamma, \lambda) + \alpha_j y(\gamma, \lambda) + \lambda \left(1 - \sum_{i=1}^{j-1} \alpha_i \right) + \frac{\epsilon}{2}} \right)$$

For all $2 \leq j \leq t$

Total complexity at depth j

$$T_j = \tilde{O} \left((2^m)^{\sum_{i=1}^{j-1} \alpha_i y(\gamma, \lambda) + \alpha_j y(\gamma, \lambda) + \lambda \left(1 - \sum_{i=1}^{j-1} \alpha_i \right) + \frac{\epsilon}{2}} \right)$$

For all $2 \leq j \leq t$

Total complexity at depth j

$$T_j = \tilde{O} \left(\binom{2^m}{\sum_{i=1}^{j-1} \alpha_i y(\gamma, \lambda) + \alpha_j y(\gamma, \lambda) + \lambda \left(1 - \sum_{i=1}^{j-1} \alpha_i \right) + \frac{\epsilon}{2}} \right)$$

$$\text{Case } j = 1: T_1 = \tilde{O}(2^{\lambda m + \alpha_1 y(\gamma, \lambda) m + \frac{\epsilon}{2} m})$$

Depth $t + 1 \implies$ Exhaustive search

Maximum list size: $\tilde{O} \left((2^m)^{\lambda(1 - \sum_{i=1}^t \alpha_i) + \frac{\epsilon}{2}} \right) = \tilde{O} \left(2^{\frac{\epsilon}{2}m} \right)$

Number of node: $\tilde{O} \left((2^m)^{\sum_{i=1}^t \alpha_i y(\gamma, \lambda)} \right) = \tilde{O} \left(2^{y(\gamma, \lambda)m} \right)$

Total complexity at depth $t + 1$

$$T_{t+1} = \tilde{O} \left(2^{(y(\gamma, \lambda) + \epsilon)m} \right)$$

Algorithm 3 RECURSIVEMO

```
1: function RECURSIVEMO( $L, R, m, t, \epsilon, \lambda, \gamma, (\alpha)_1^t, j$ )
2:   if  $j = t + 1$  then
3:     return QUADRATICNN( $L, R, \gamma m$ )
4:    $C \leftarrow \perp$ 
5:    $s_j \leftarrow \frac{\binom{\alpha_j m}{\frac{1}{2}\alpha_j m}}{\binom{(1-\gamma)\frac{\alpha_j m}{2}}{(1-h-\frac{\gamma}{2})\frac{\alpha_j m}{2}} \binom{(1-\gamma)\frac{\alpha_j m}{2}}{(h-\frac{\gamma}{2})\frac{\alpha_j m}{2}} \left(\gamma\frac{\alpha_j m}{2}\right)^2}$ 
6:   for  $ms_j$  times do
7:     Pick a random subset  $A$  of  $\frac{\alpha_j m}{2}$  indices inside strip  $j$ 
8:      $L' \leftarrow \{u \in L \text{ s.t. } wt(u_A) = H^{-1}(1-\lambda)\frac{\alpha_j m}{2}\}$ 
9:      $R' \leftarrow \{v \in R \text{ s.t. } wt(v_A) = H^{-1}(1-\lambda)\frac{\alpha_j m}{2}\}$ 
10:    if  $|L'|$  and  $|R'|$  are not too big then
11:       $x \leftarrow \text{RECURSIVEMO}(L', R', m, t, \epsilon, \lambda, \gamma, (\alpha)_1^t, j+1)$ 
12:      if  $x \neq \perp$  then
13:        return  $x$ 
14:  return  $C$ 
```

To conclude, choosing:

① $\alpha_{j+1} = \frac{\lambda \alpha_j}{y(\gamma, \lambda)}$ for all $j \in \{1, \dots, t-1\}$

② $t = \left\lceil \frac{\log(2(y(\gamma, \lambda) - \lambda)/\epsilon + 1)}{\log(y(\gamma, \lambda)/\lambda)} \right\rceil$

leads to

$$T_1 = \tilde{O}\left(2^{(y(\gamma, \lambda) + \epsilon)m}\right)$$

$$T_2 = \tilde{O}\left(2^{(y(\gamma, \lambda) + \epsilon)m}\right)$$

$$\dots = \tilde{O}\left(2^{(y(\gamma, \lambda) + \epsilon)m}\right)$$

$$T_t = \tilde{O}\left(2^{(y(\gamma, \lambda) + \epsilon)m}\right)$$

$$T_{t+1} = \tilde{O}\left(2^{(y(\gamma, \lambda) + \epsilon)m}\right)$$

- The model : $\mathcal{M}_{Alea}(\mathbb{F}_2, 2^{\lambda m})$
- The expected number of solutions : $E = \Theta\left(m^{-\frac{1}{2}} 2^{m(2\lambda + H(\gamma) - 1)}\right)$
- Fonctions LSH
- A lower bound in a nearby model [KL21] : $2^{\frac{\lambda}{1-\gamma}m}$

Thank you