

Effective Web Graph Representations

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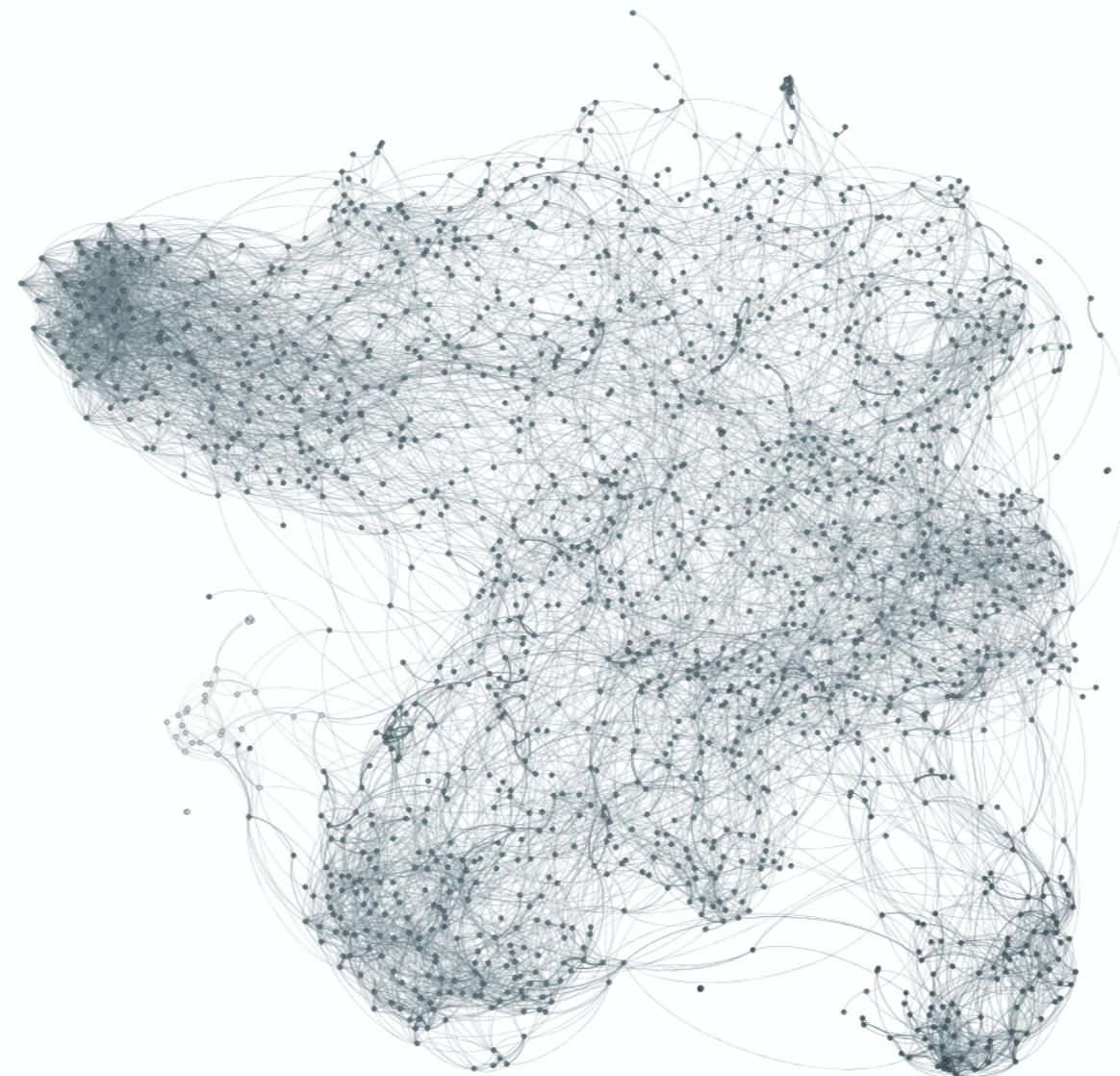
giulio.pibiri@di.unipi.it

Pisa, 29/10/2018

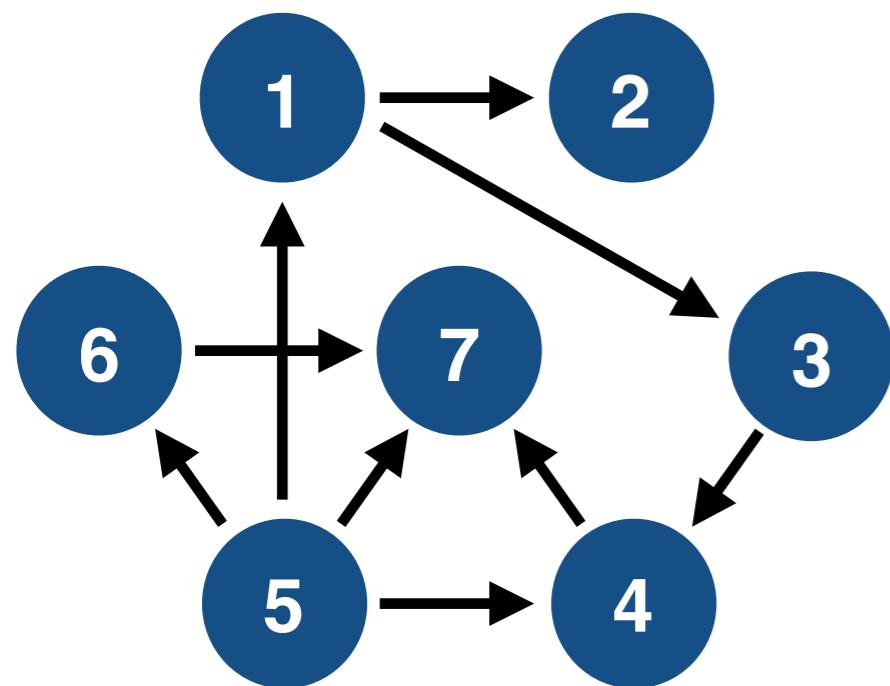
Context - Web Graphs

Web graphs are directed graphs of pages pointing to other pages on the Web.

We focus on compression effectiveness on **large real-world Web graphs**.



Context - Web Graphs



Conceptual graph

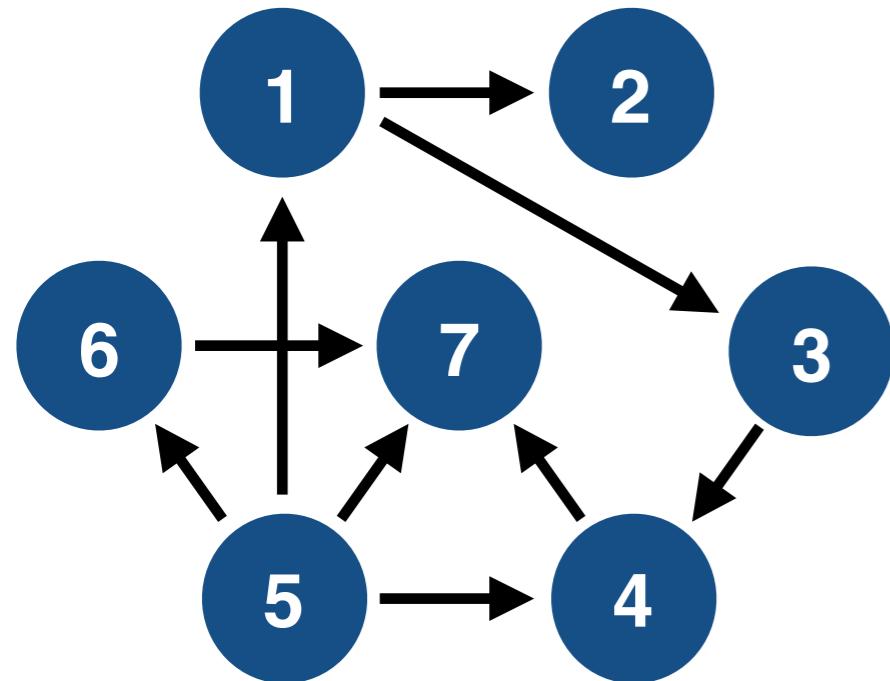
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0	0	0	0	0	0	0
0	0	0	1	0	0	0
0	0	0	0	0	0	1
1	0	0	1	0	1	1
0	0	0	0	0	0	1
0	0	0	0	0	0	0

Adjacency **matrix**

1: 2,3
2: -
3: 4
4: 7
5: 1,4,6,7
6: 7
7: -

Adjacency **lists**

Context - Web Graphs



Conceptual graph

0	1	1	0	0	0	0
0	0	0	0	0	0	0
0	0	0	1	0	0	0
0	0	0	0	0	0	1
1	0	0	1	0	1	1
0	0	0	0	0	0	1
0	0	0	0	0	0	0

Adjacency **matrix**

1:	2,3
2:	-
3:	4
4:	7
5:	1,4,6,7
6:	7
7:	-

Adjacency **lists**

Many results are known for
compressing
integer sequences.

Outline

- 1. The WebGraph framework**
- 2. k^2 -trees**
- 3. Block-trees**
- 4. 2D-Block-trees**

Outline

1. The WebGraph framework

2004

2. k^2 -trees

2009

3. Block-trees

2014

4. 2D-Block-trees

2018



The WebGraph Framework

Java/C++ framework consisting in algorithms and compression codes for managing large Web Graphs.

<http://webgraph.di.unimi.it/>

The WebGraph Framework I: Compression Techniques, Boldi-Vigna, WWW 2004

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Use *d-gap* compression.

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Locality - pages links to pages whose URL is lexicographically similar. URLs share long common prefixes.

Use *d-gap* compression.

Similarity - pages that are close together in lexicographic order, tend to have many common successors.

Use *reference* compression.

The WebGraph Framework

Exploiting **locality**.

If we have: $x: [y_1, \dots, y_k]$, then we represent
 $[y_1 - x, y_2 - y_1 - 1, y_3 - y_2 - 1, \dots, y_k - y_{k-1} - 1]$

First gap $d = y_1 - x$ is represented as $2d$ if $d \geq 0$ or $2|d|-1$ if $d < 0$

Node	Outdegree	Successors
...
15	11	13, 15, 16, 17, 18, 19, 23, 24, 203, 315, 1034
16	10	15, 16, 17, 22, 23, 24, 315, 316, 317, 3041
17	0	
18	5	13, 15, 16, 17, 50
...

Adjacency lists

Node	Outdegree	Successors
...
15	11	3, 1, 0, 0, 0, 3, 0, 178, 111, 718
16	10	1, 0, 0, 4, 0, 0, 290, 0, 0, 2723
17	0	
18	5	9, 1, 0, 0, 32
...

d -gapped adjacency lists

The WebGraph Framework

Exploiting **similarity**.

Idea: use reference compression, i.e., represent a list with respect to another one called its **reference list**.

	Node	Outdegree	Successors

1	15	11	13, 15, 16, 17, 18, 19, 23, 24, 203, 315, 1034
2	16	10	15, 16, 17, 22, 23, 24, 315, 316, 317, 3041
3	17	0	
4	18	5	13, 15, 16, 17, 50

Adjacency lists

	Node	Outd.	Ref.	Copy list	Extra nodes

	15	11	0		13, 15, 16, 17, 18, 19, 23, 24, 203, 315, 1034
	16	10	1	01110011010	22, 316, 317, 3041
	17	0			
	18	5	3	11110000000	50

Copy lists

The WebGraph Framework

Node	Outd.	Ref.	Copy list	Extra nodes
...
15	11	0	01110011010	13, 15, 16, 17, 18, 19, 23, 24, 203, 315, 1034
16	10	1		22, 316, 317, 3041
17	0			
18	5	3	11110000000	50
...

Copy lists

Node	Outd.	Ref.	# blocks	Copy blocks	Extra nodes
...
15	11	0			13, 15, 16, 17, 18, 19, 23, 24, 203, 315, 1034
16	10	1	7	0, 0, 2, 1, 1, 0, 0	22, 316, 317, 3041
17	0				
18	5	3	1	4	50
...

Copy blocks

The WebGraph Framework

Node	Outd.	Ref.	Copy list	Extra nodes
...
15	11	0	01110011010	13, 15, 16, 17, 18, 19, 23, 24, 203, 315, 1034
16	10	1		22, 316, 317, 3041
17	0			
18	5	3	11110000000	50
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Copy lists

Node	Outd.	Ref.	# blocks	Copy blocks	Extra nodes
...
15	11	0			13, 15, 16, 17, 18, 19, 23, 24, 203, 315, 1034
16	10	1	7	0, 0, 2, 1, 1, 0, 0	22, 316, 317, 3041
17	0				
18	5	3	1	4	50
...

Copy blocks

Node	Outd.	Ref.	# blocks	Copy blocks	# intervals	Left extremes	Length	Residuals
...
15	11	0			2	0, 2	3, 0	5, 189, 111, 718
16	10	1	7	0, 0, 2, 1, 1, 0, 0	1	600	0	12, 3018
17	0							
18	5	3	1	4	0			50
...

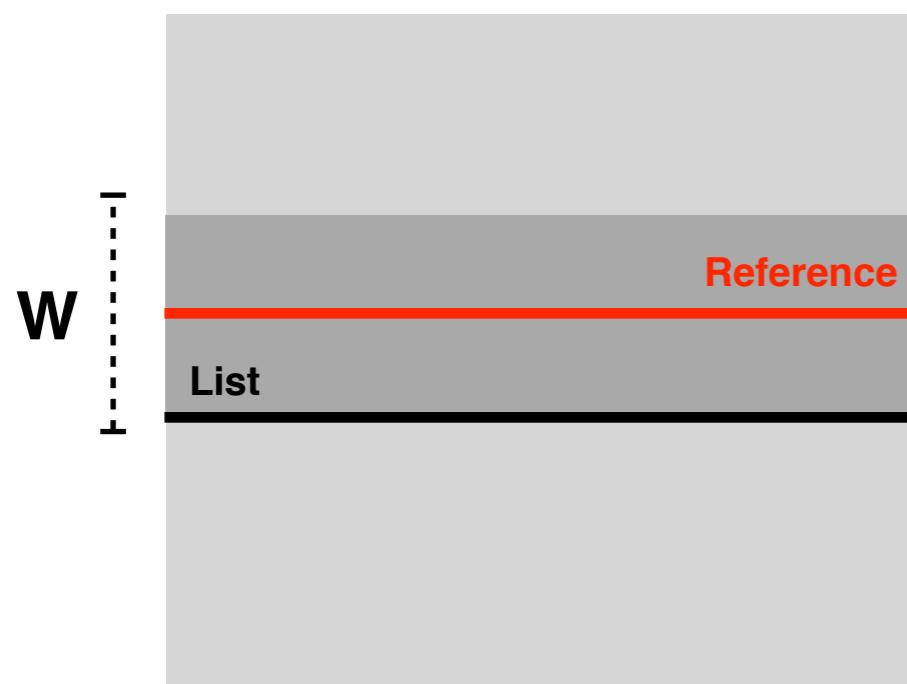
Intervals

An interval is a run, of size $\geq L$, of consecutive integers.

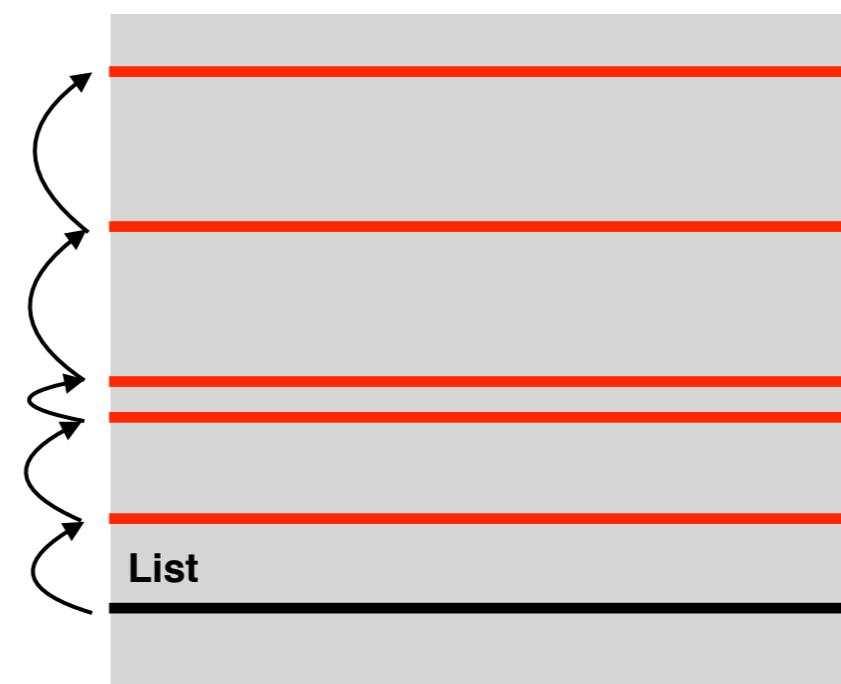
Example for $L = 2$.

The WebGraph Framework

W - window size



R - maximum reference chain



Tradeoff between compression and decoding time.

The WebGraph Framework

.uk

18.5 million pages
300 million links

WebBase

118 million pages
1 billion links

R	Bits/link		
	$W = 1$	$W = 3$	$W = 7$
∞	2.75	2.38	2.22
3	3.87	3.25	3.00
1	5.05	3.91	3.46

R	Bits/link		
	$W = 1$	$W = 3$	$W = 7$
∞	3.59	3.22	3.08
3	4.46	3.92	3.74
1	5.40	4.49	4.17

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∞	2.75	2.75	2.75		3.22	3.08	
3	3.87	3.87	3.87		3.92	3.74	
1	5.05	3.91	3.46	1	5.40	4.49	4.17

Compression rates down to
approximately 3 bits per link.

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	$W = 1$	$W = 3$	$W = 7$		$W = 1$	$W = 3$	$W = 7$
∞	2.75	2.75	2.75		3.22	3.08	
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1	5.05	3.91	3.46	1	5.40	4.49	4.17

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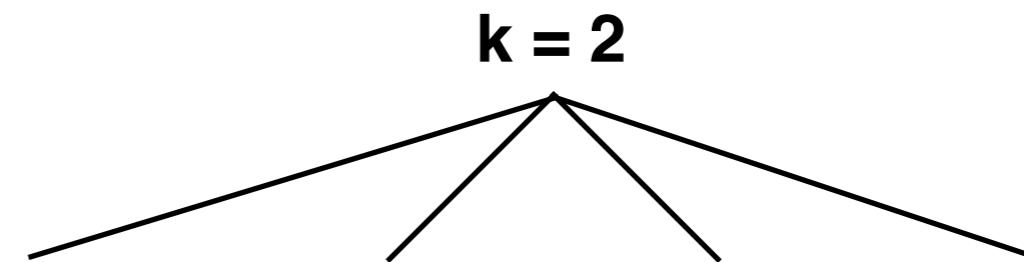
R	Graph size (MiB)	Link access time (ns)			
		seq.	$J = 1$	$J = 2$	$J = 4$
∞	79.0	198	31 237	35 752	43 699
3	106.6	206	611	753	886
1	122.9	233	442	491	605

k²-trees

k²-ary tree
representation of the
adjacency matrix.

0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	0
1	1	0	0	0	0	0	0
0	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	0	0	0	0	0

k²-trees for Compact Web Graph Representation,
Brisaboa-Ladra-Navarro, SPIRE 2009

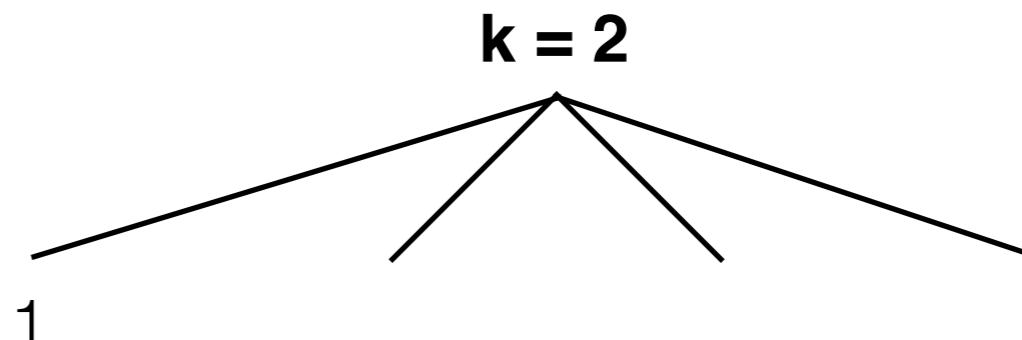


k^2 -trees

k^2 -ary tree
representation of the
adjacency matrix.

0	0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	0	0
1	1	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0

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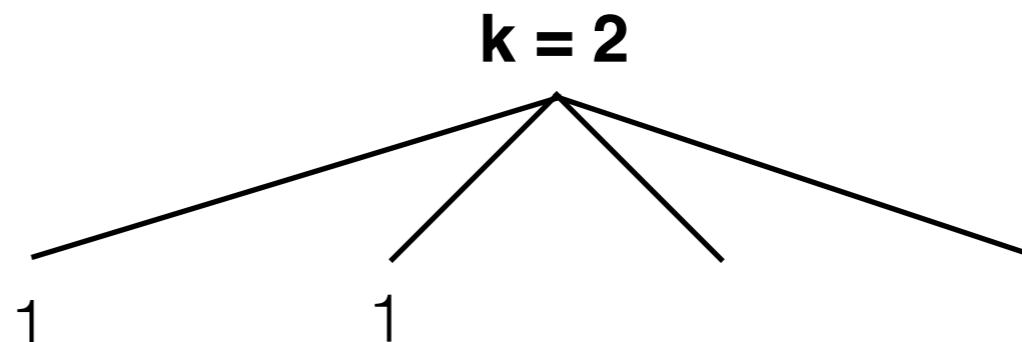


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adjacency matrix.

0	0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	0	0
1	1	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0

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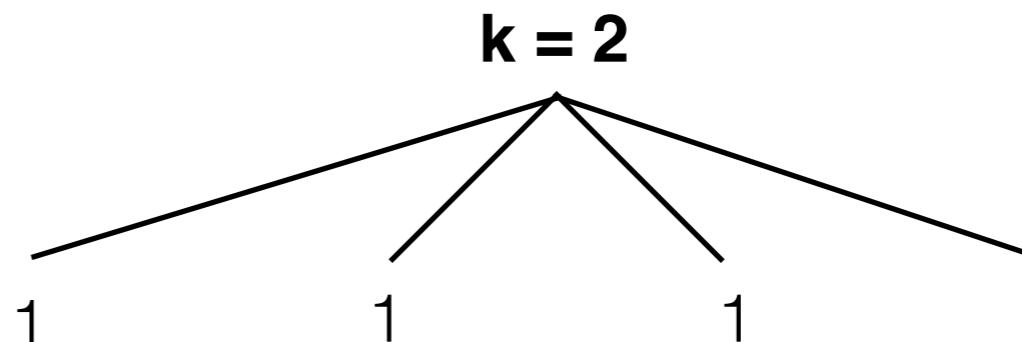


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0	0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	0	0
1	1	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0

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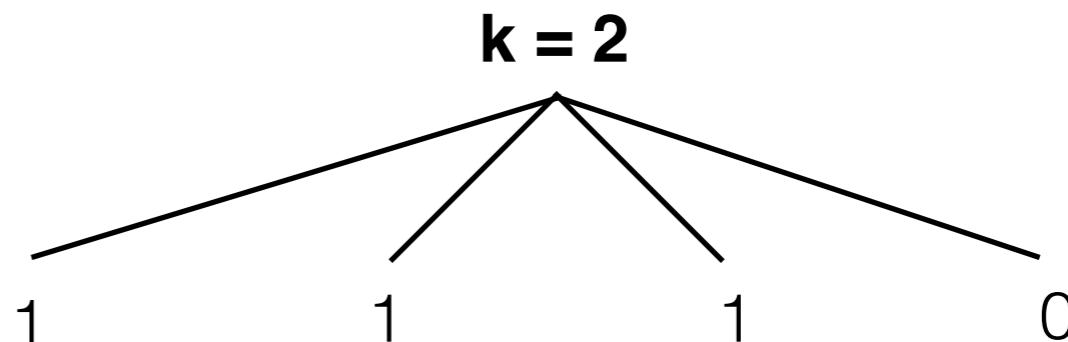


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0	0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	0	0
1	1	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0

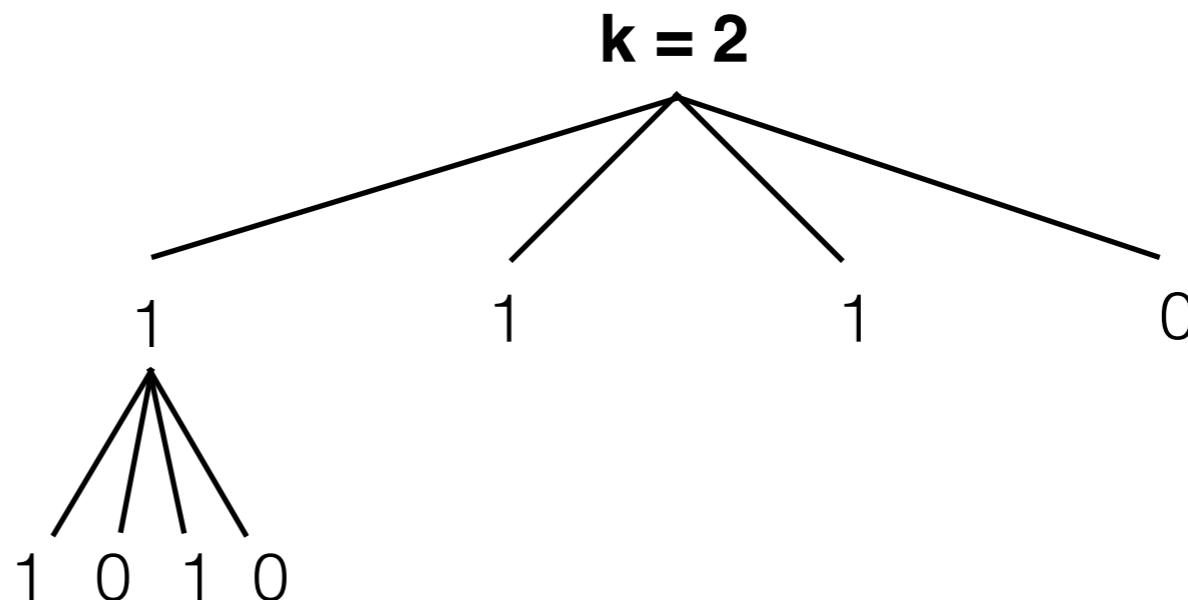
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0	0	0	0	0	0	0	0	0
1	0	1	0	0	0	1	0	0
0	1	1	1	0	0	0	0	0
1	1	1	0	0	0	0	0	0
0	1	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0
1	0	0	1	0	0	0	0	0
0	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0

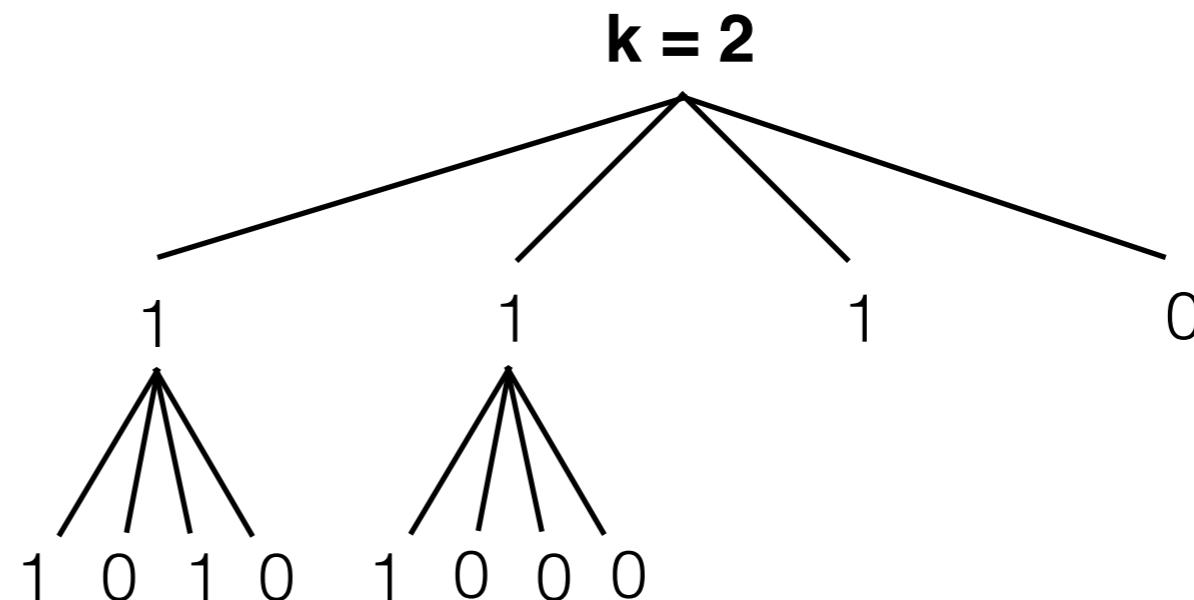


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0	0	0	0	0	0	0	0	0
1	0	0	0	0	1	0	0	0
0	1	0	0	0	0	1	0	0
1	1	1	0	0	0	0	1	0
1	1	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0
1	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0

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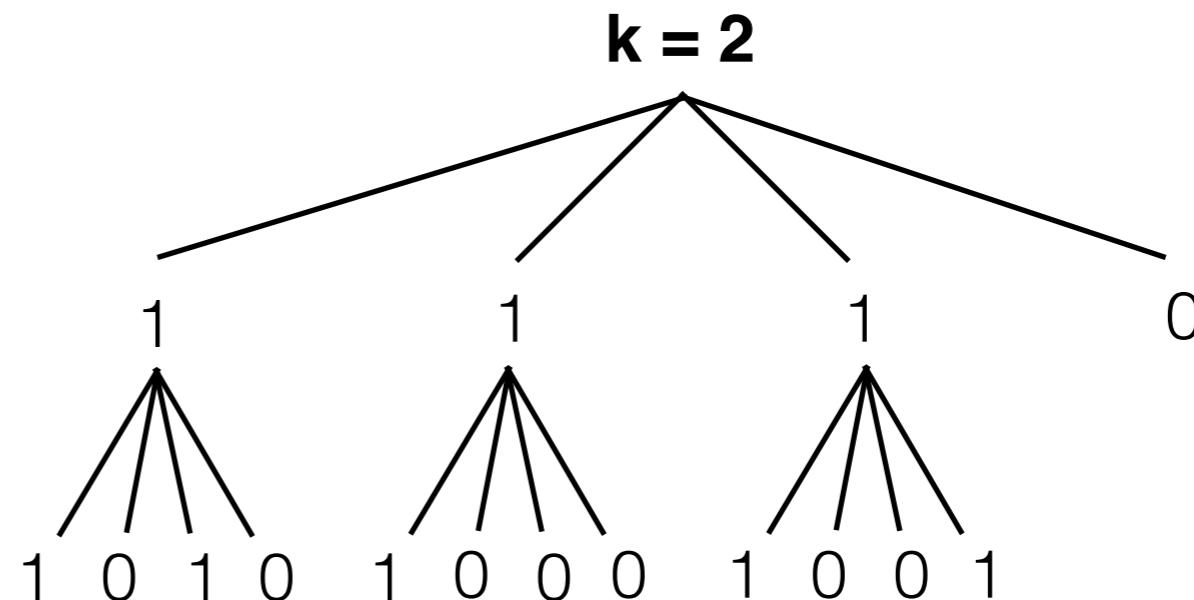


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0	0	0	0	0	0	0	0
1	0	0	0	0	1	0	0
0	1	0	0	0	0	1	0
1	1	1	0	0	0	0	0
1	1	0	0	0	0	0	0
0	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0
0	0	0	0	0	0	0	0

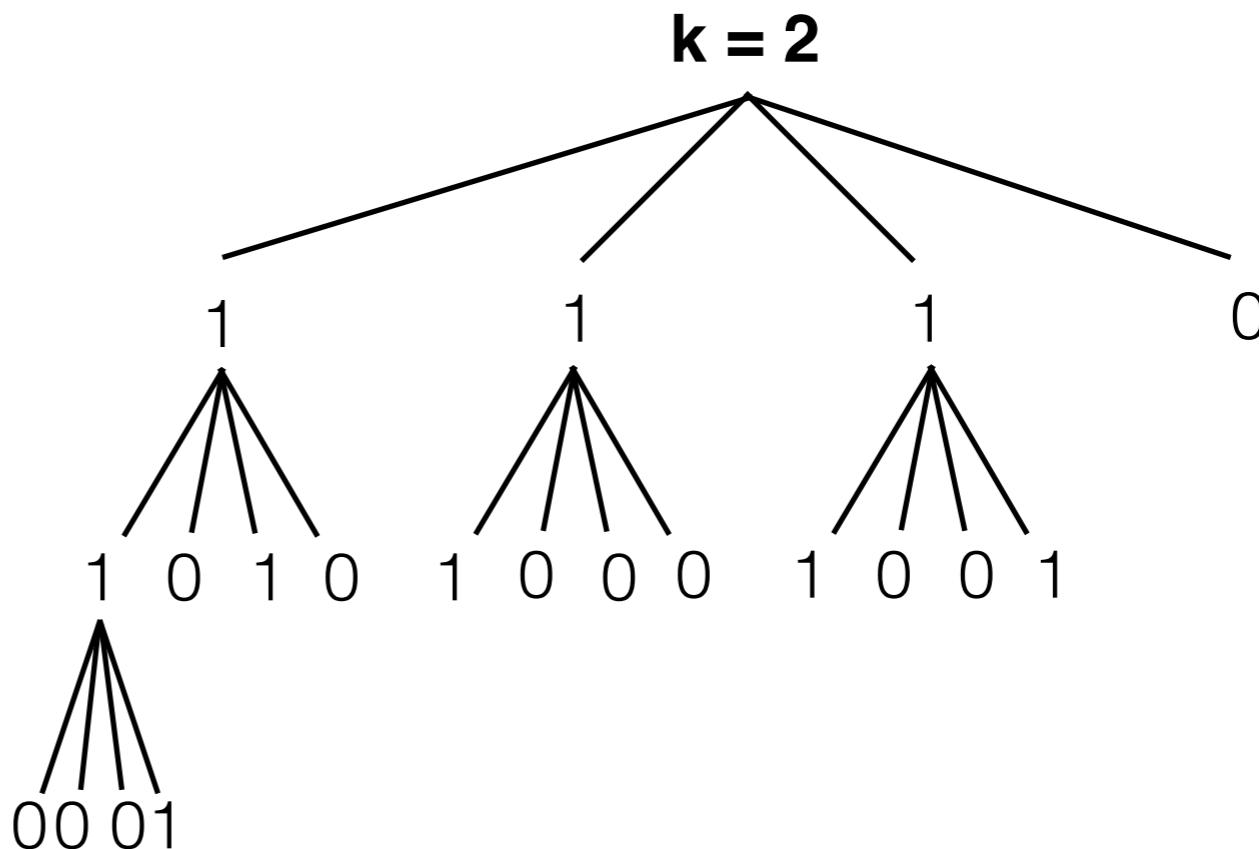
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k²-trees

k^2 -ary tree
representation of the
adjacency matrix.

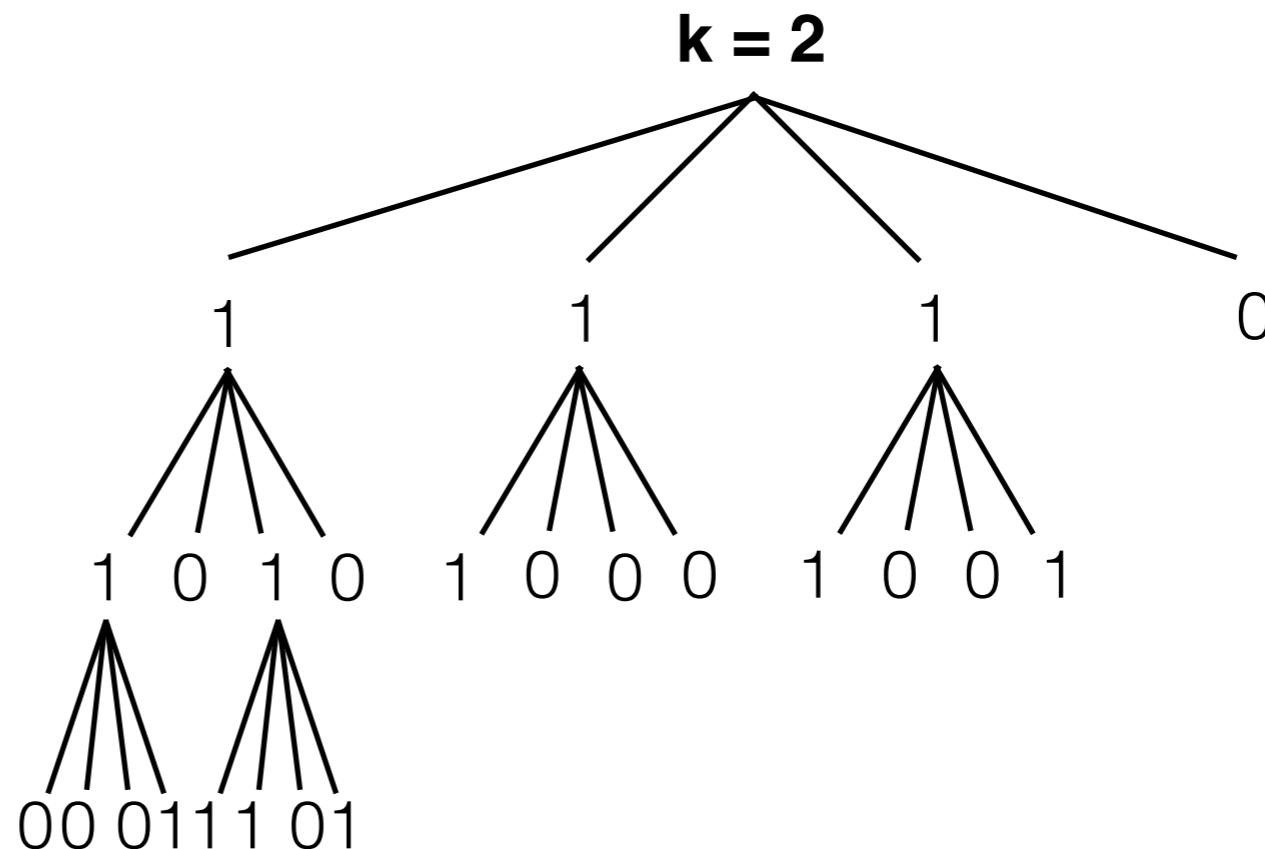
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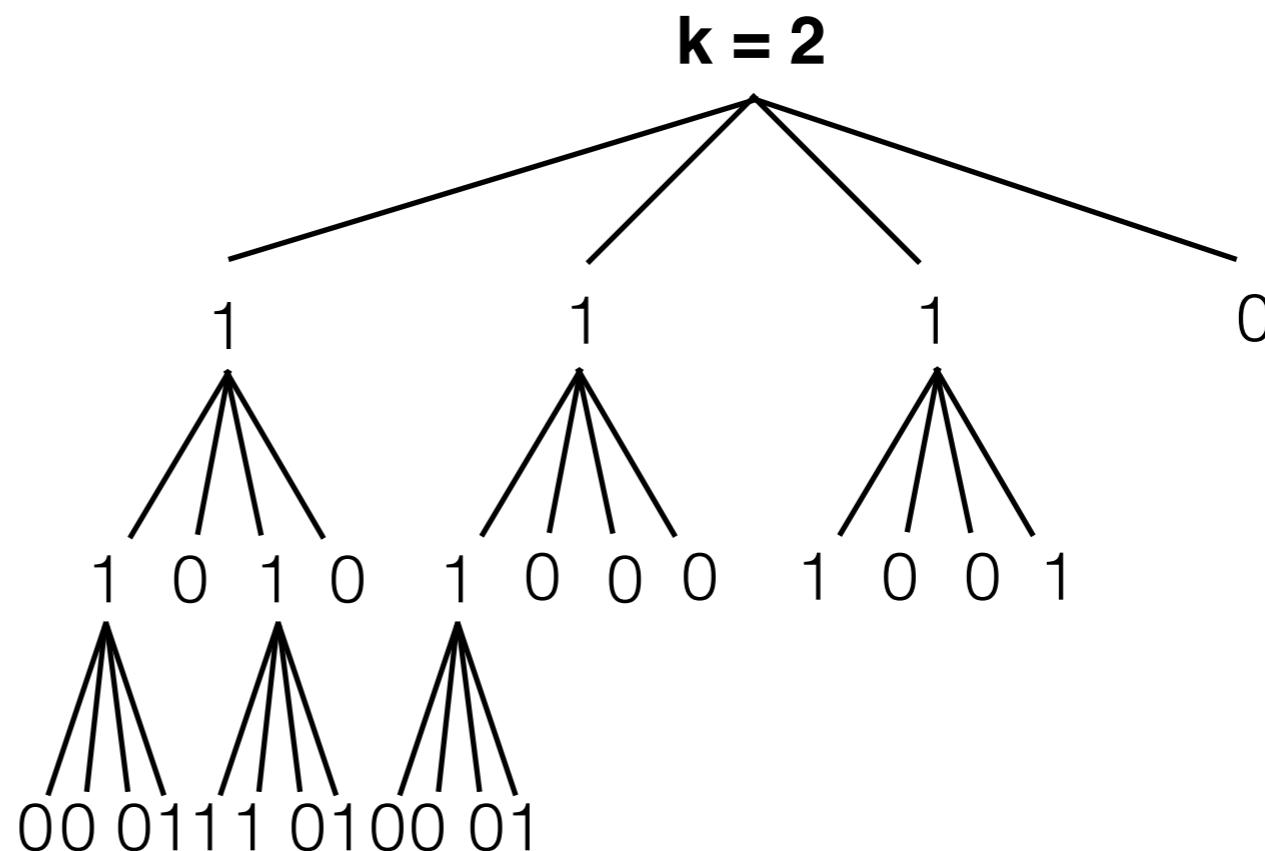
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1	1	1	0	0	0	0	0
1	1	0	0	0	0	0	0
0	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
0	0	0	1	1	0	0	0
0	0	0	0	0	0	0	0



k^2 -trees

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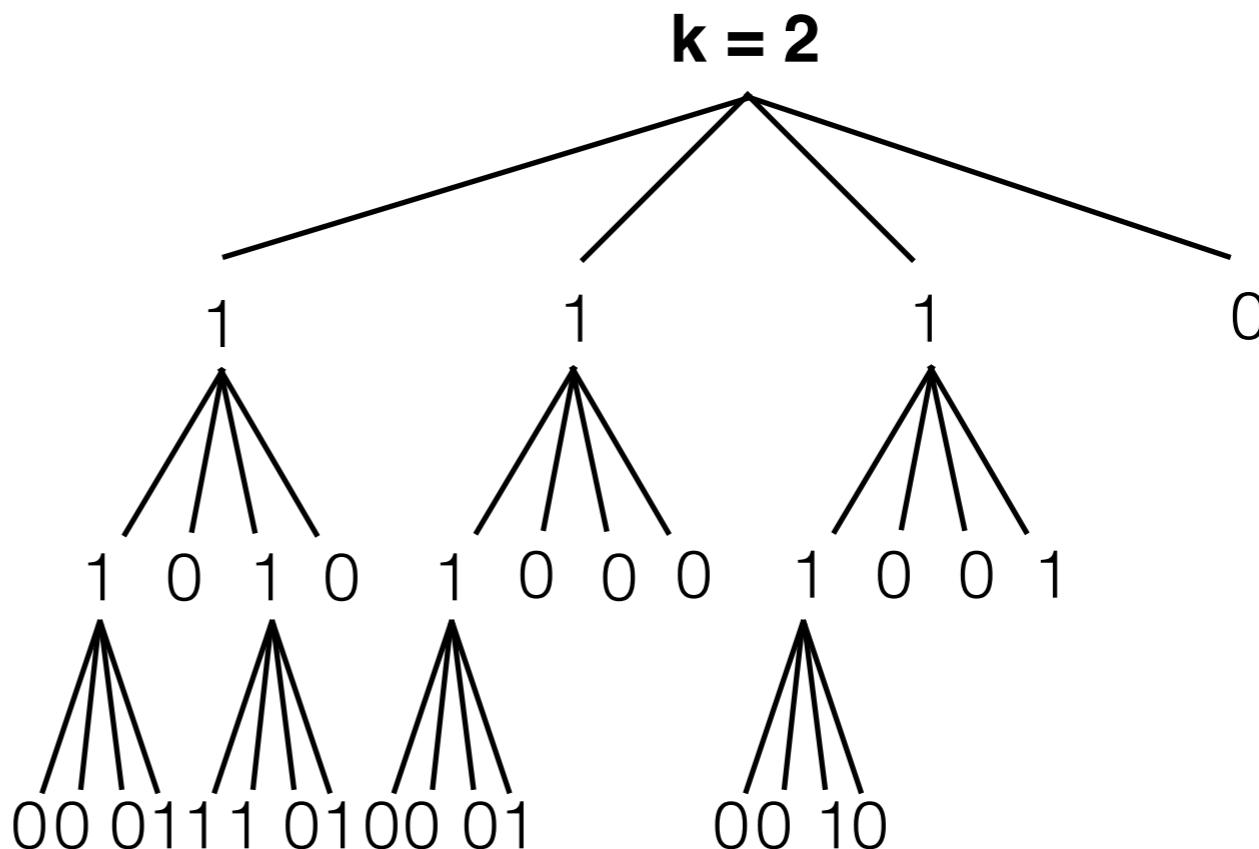
0	0	0	0	0	0	0	0
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1	1	1	0	0	0	0	0
1	1	0	0	0	0	0	0
0	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
0	0	0	1	1	0	0	0
0	0	0	0	0	0	0	0



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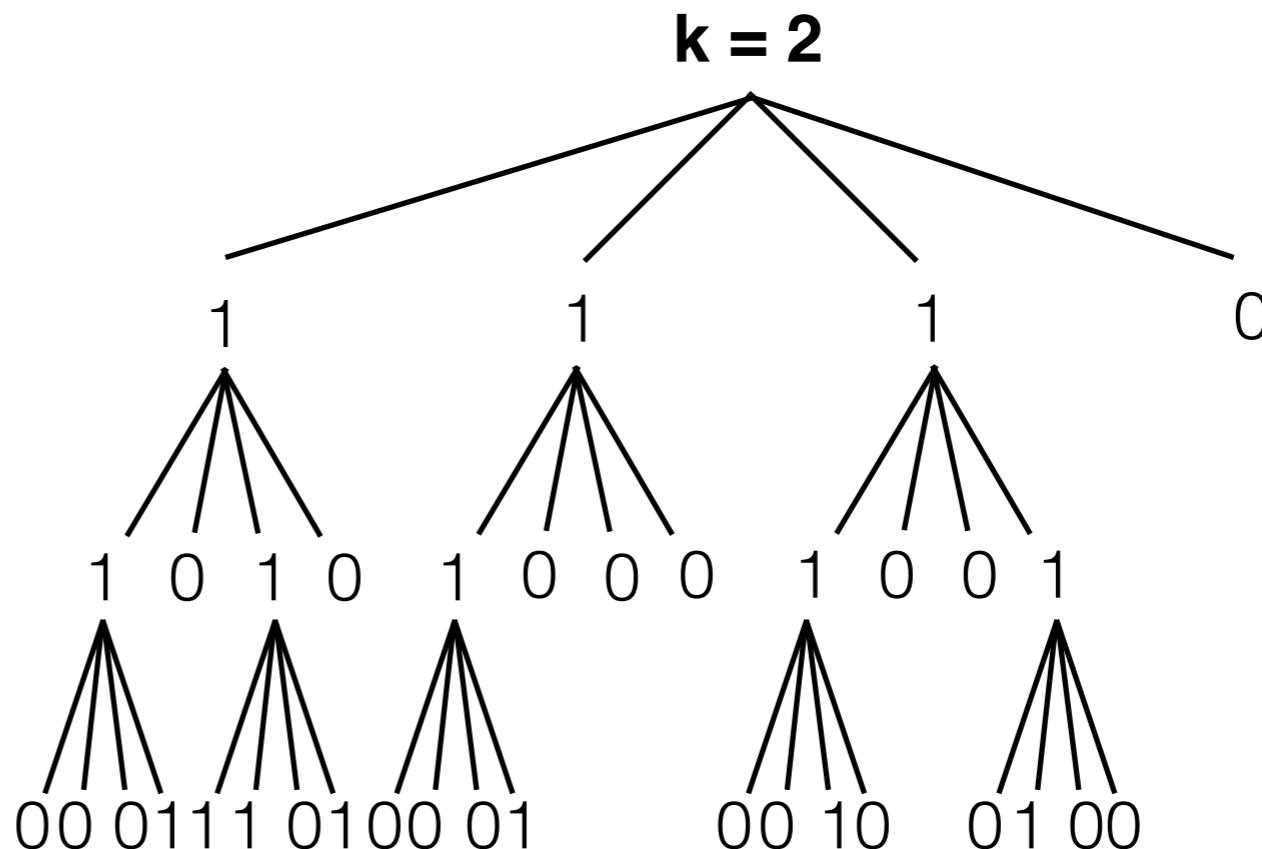
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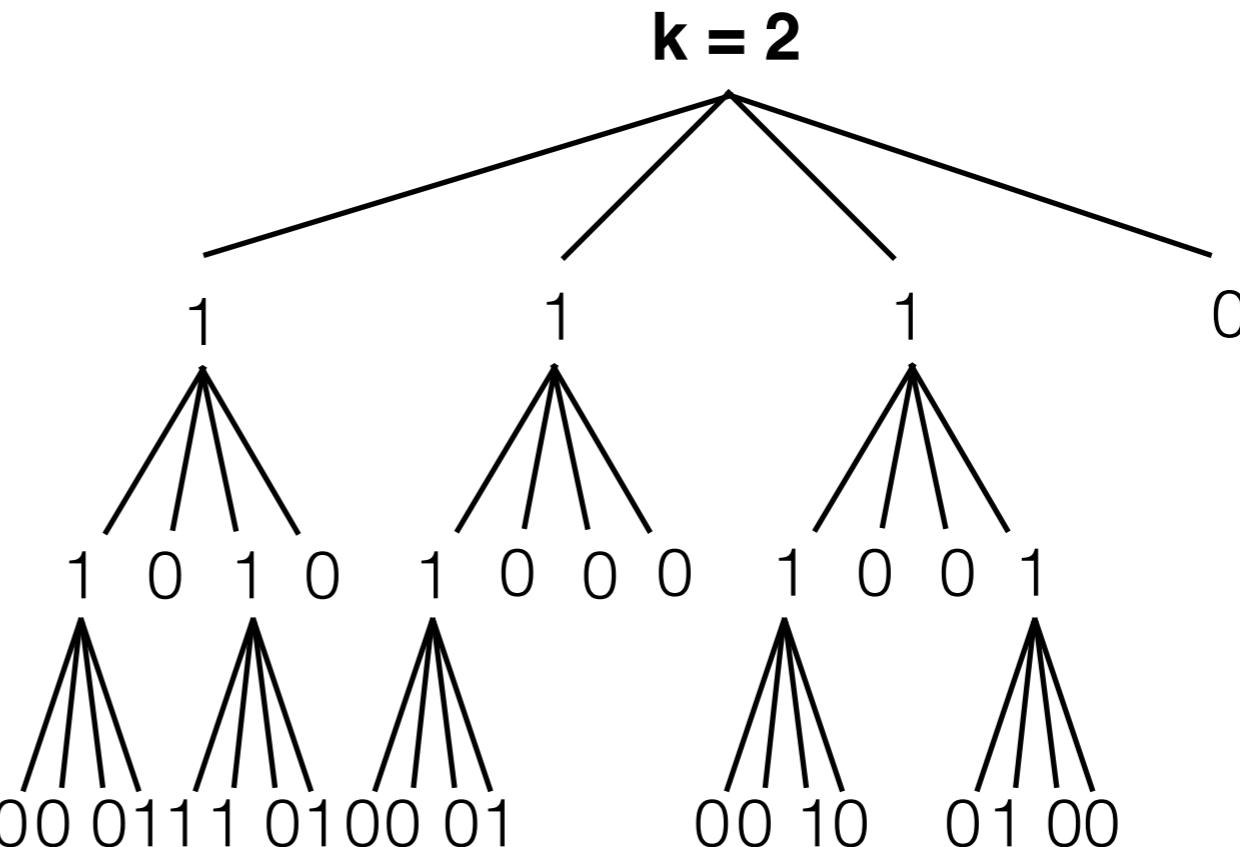
0	0	0	0	0	0	0	0
1	0	0	0	0	1	0	0
0	1	0	0	0	0	1	0
1	1	1	0	0	0	0	0
1	1	0	0	0	0	0	0
0	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
0	0	0	1	1	0	0	0
0	0	0	0	0	0	0	0



k^2 -trees

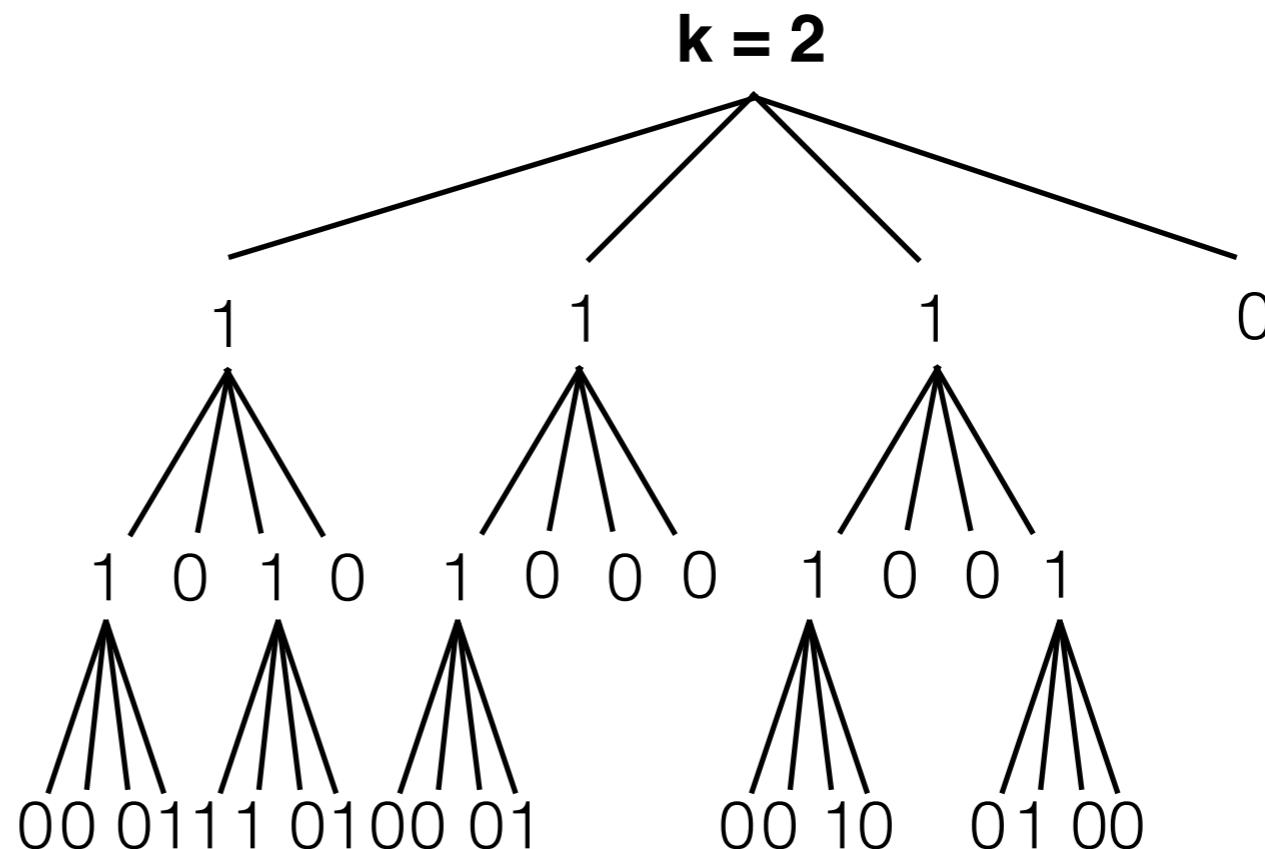
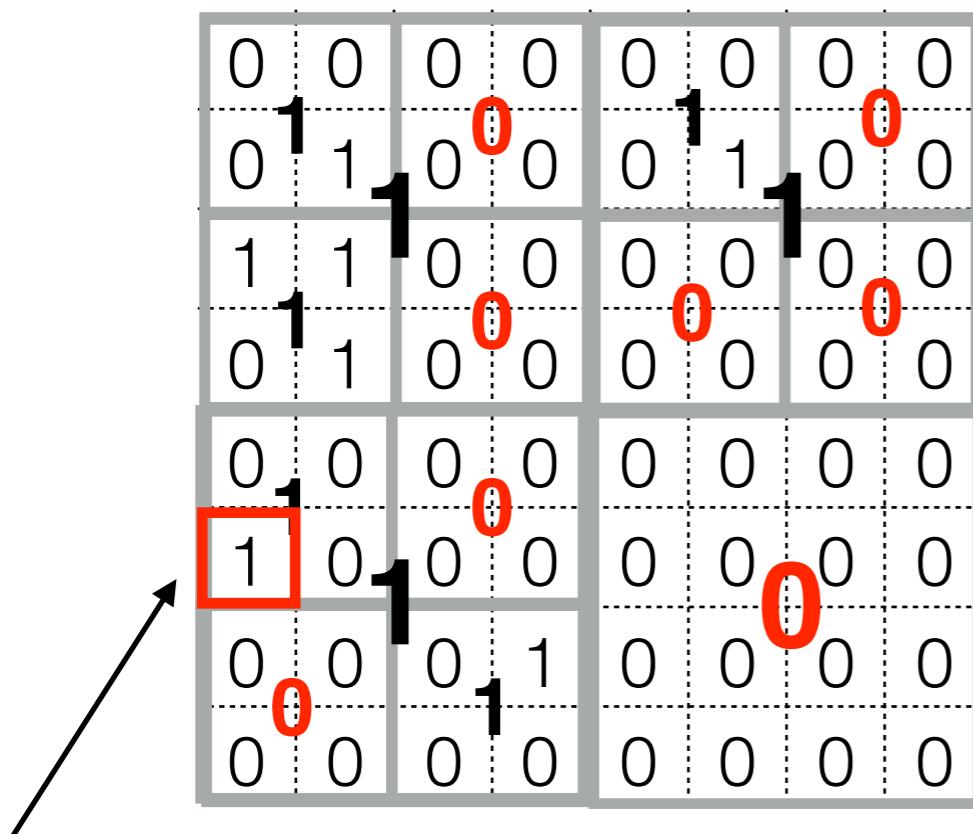
k^2 -ary tree
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0	0	0	0	0	0	0	0	0
1	0	0	0	0	1	0	0	0
0	1	0	0	0	0	1	0	0
1	1	1	0	0	0	0	1	0
1	1	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0
1	0	0	1	0	0	0	0	0
0	0	0	1	0	0	0	0	0



k^2 -trees

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Access

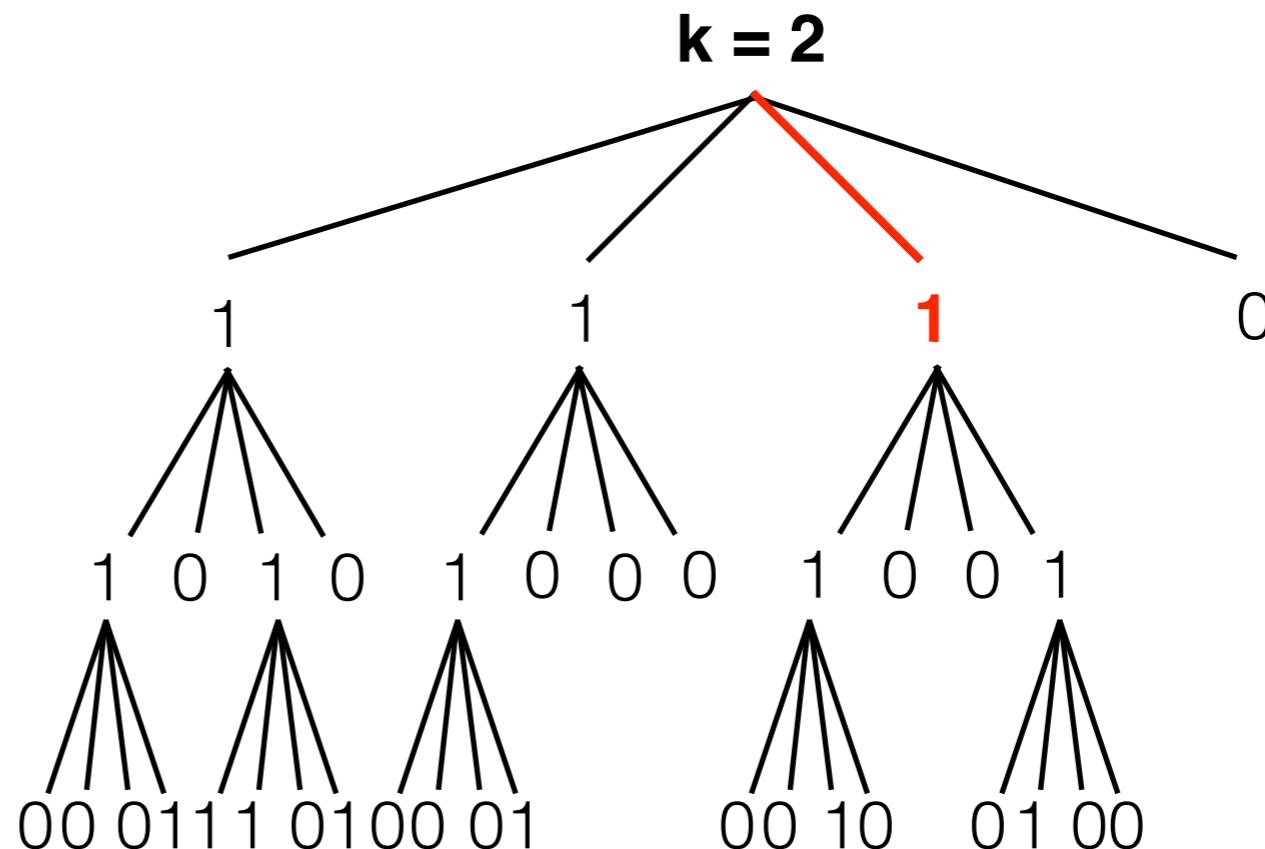
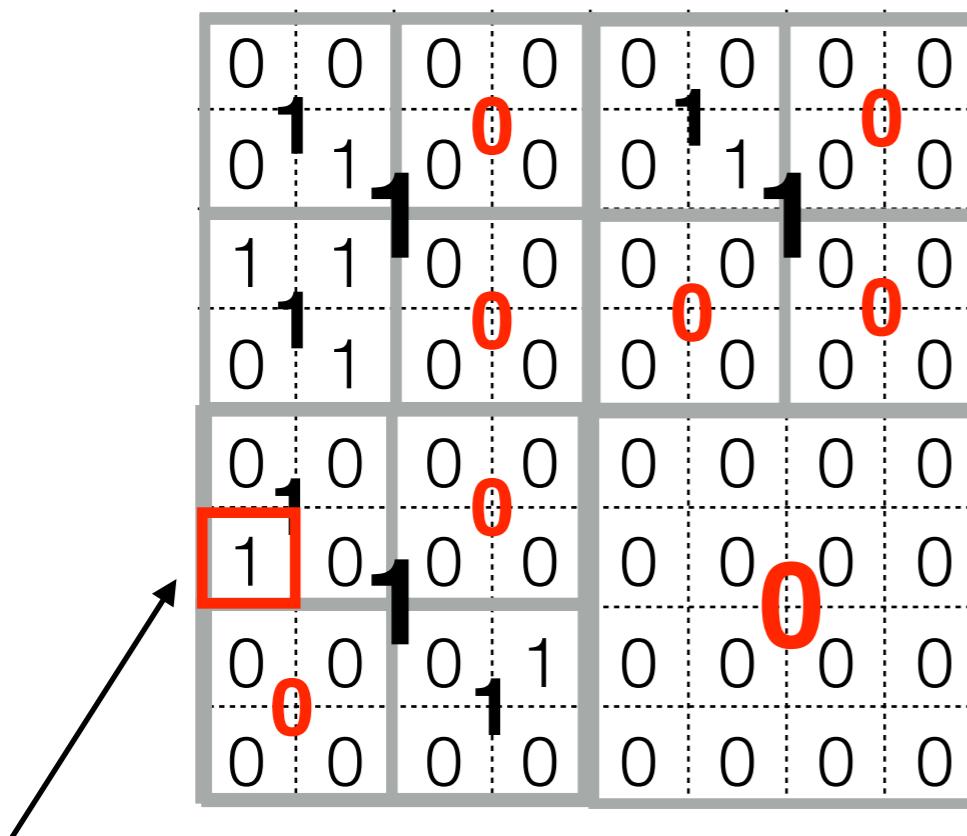
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T

L

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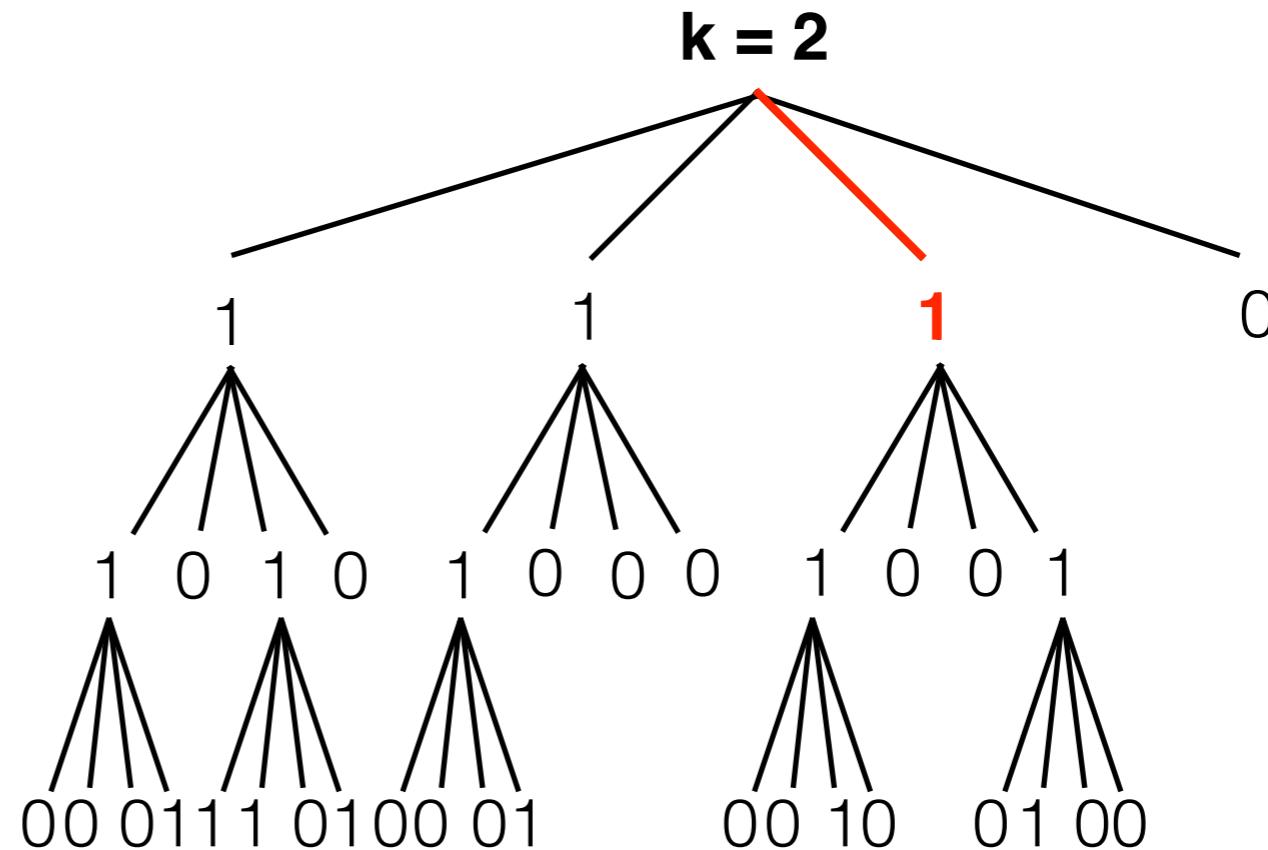
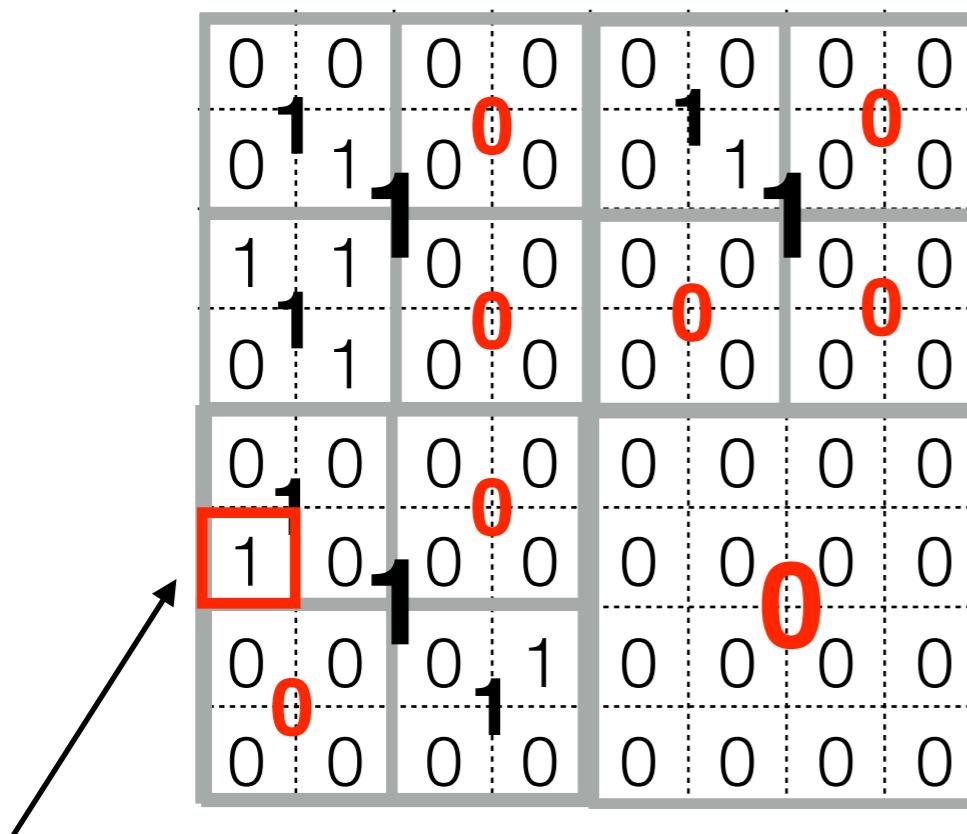
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T

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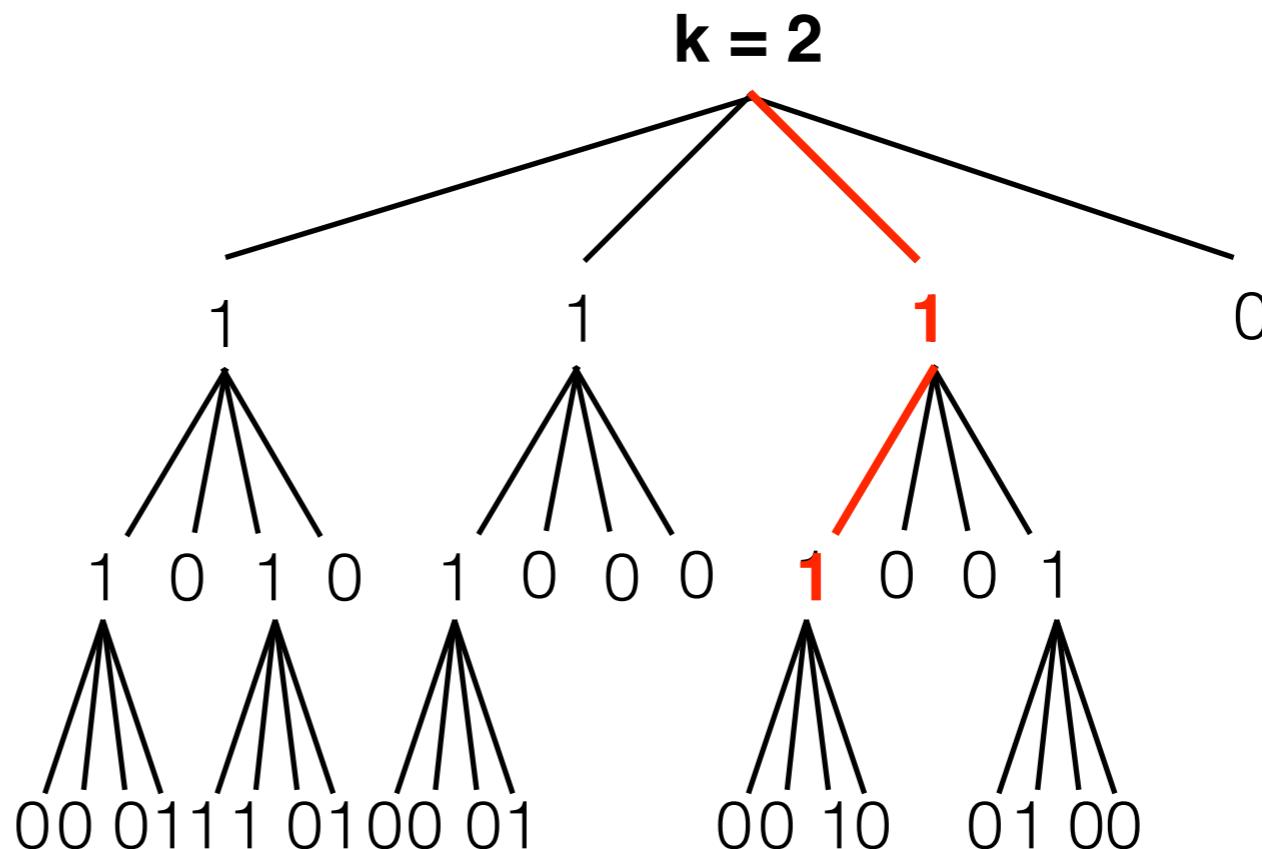
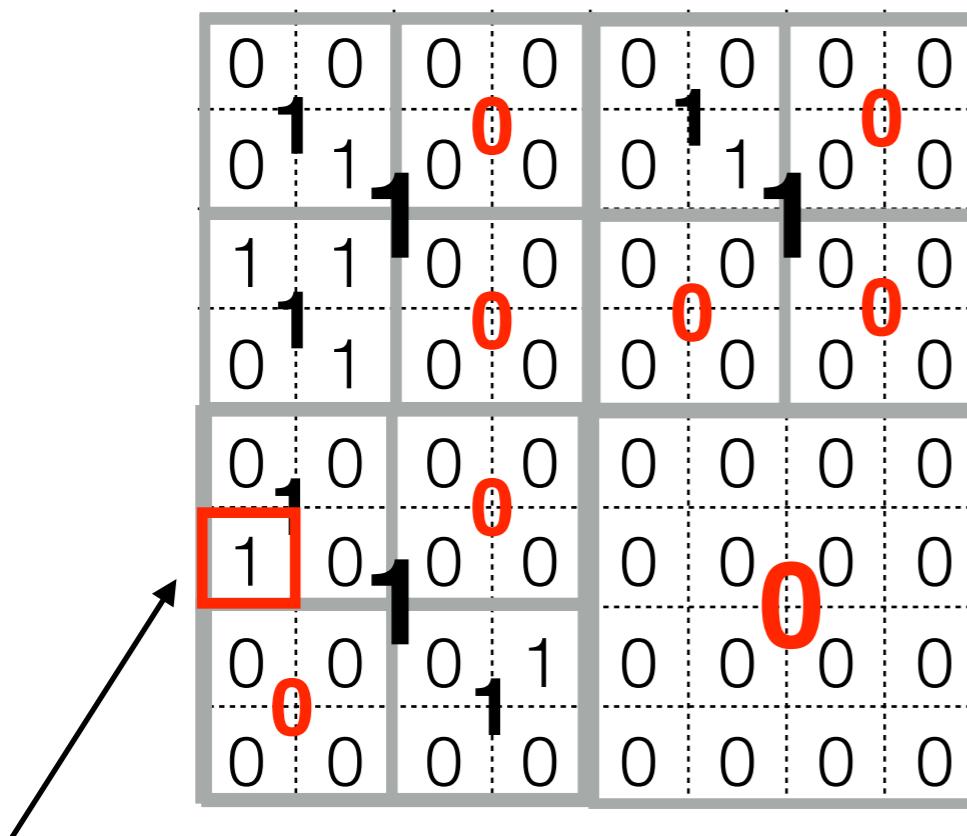


Access

1110101010001001 00011101000100100
T L

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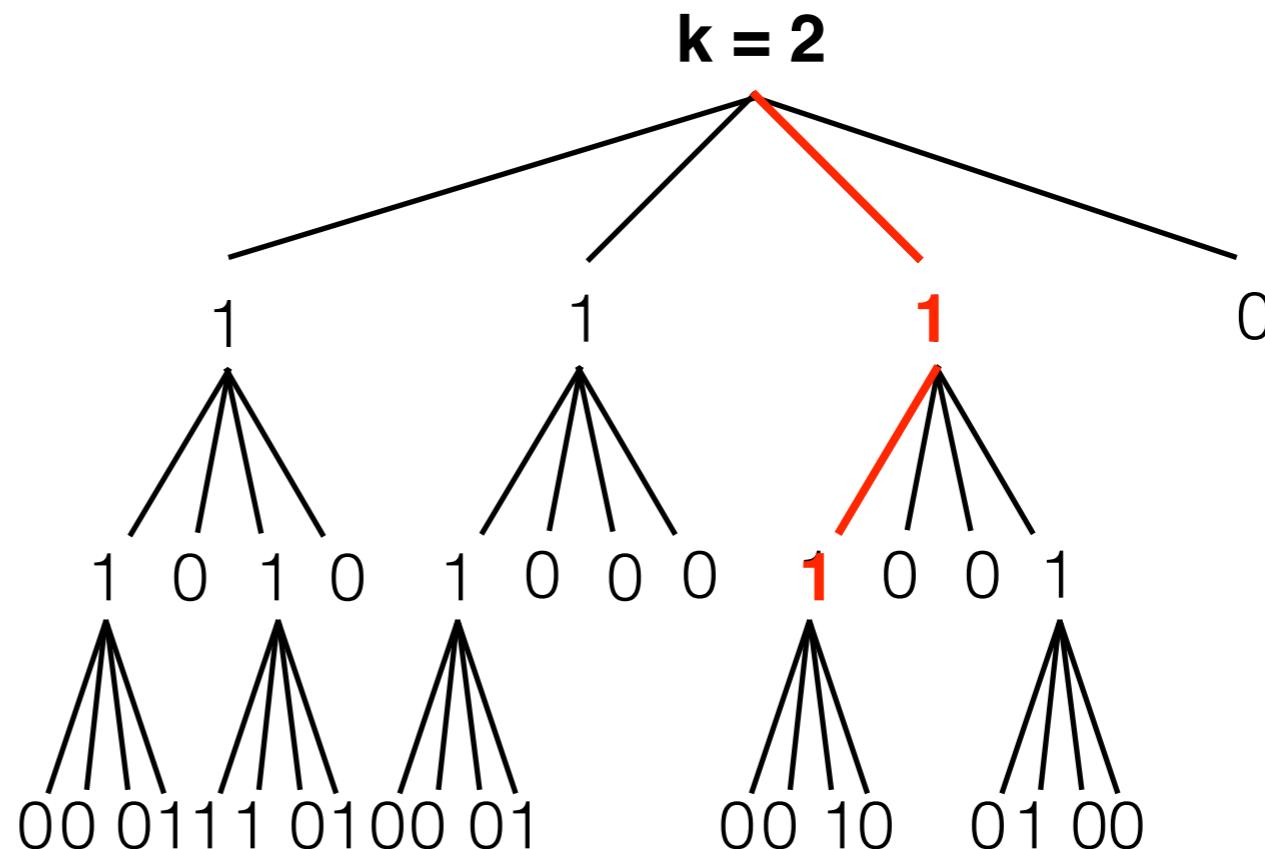
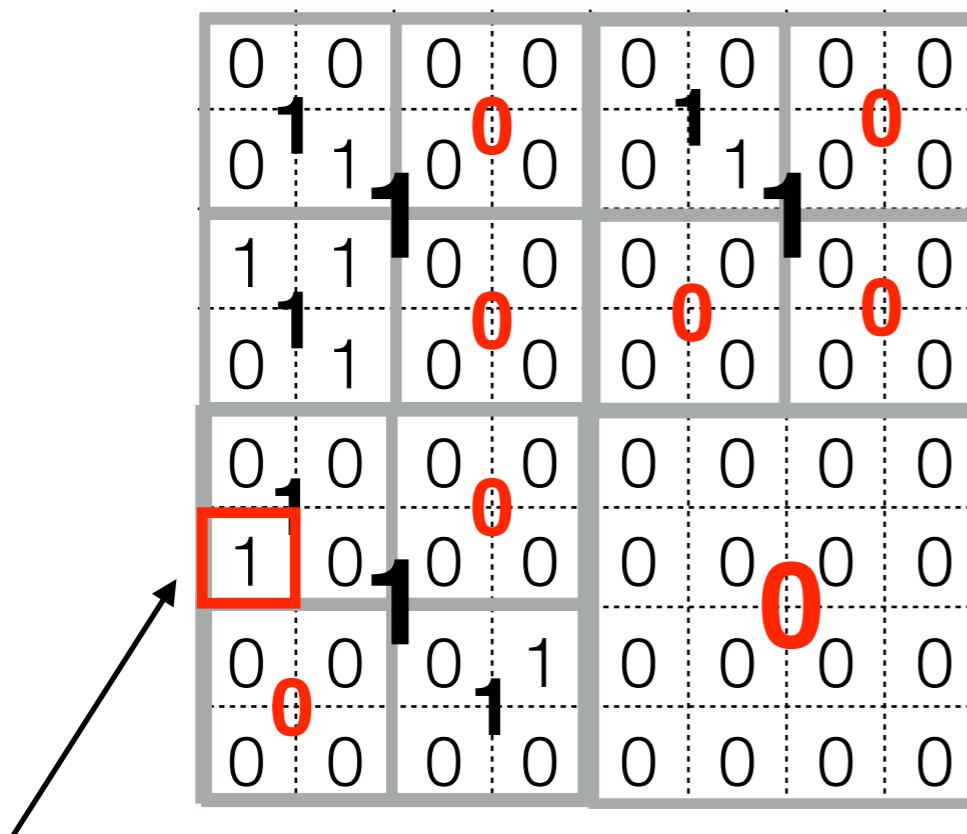


Access

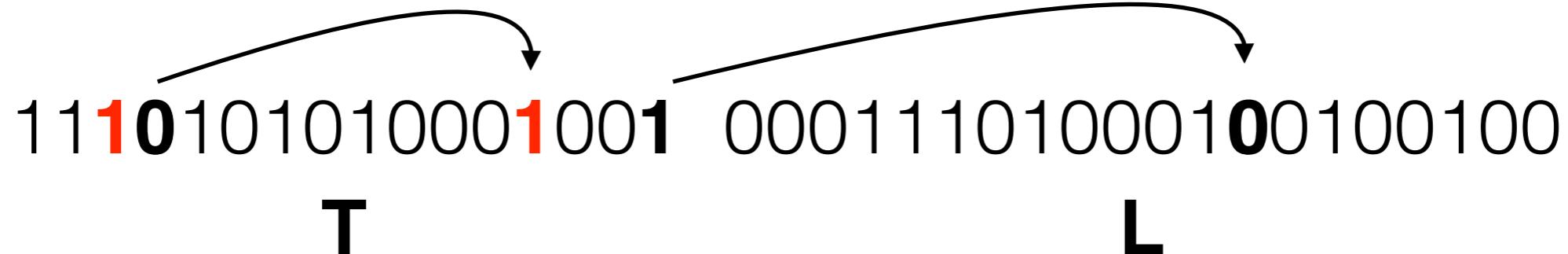
1110101010001001 00011101000100100
T L

k^2 -trees

k^2 -ary tree
representation of the
adjacency matrix.

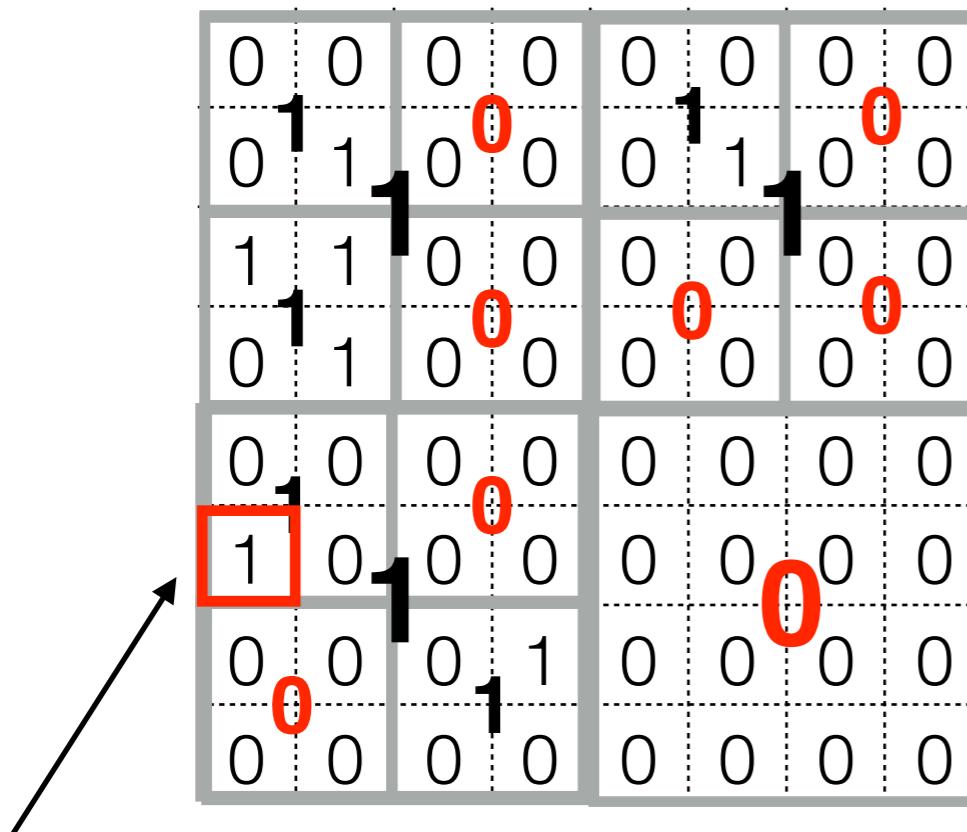


Access

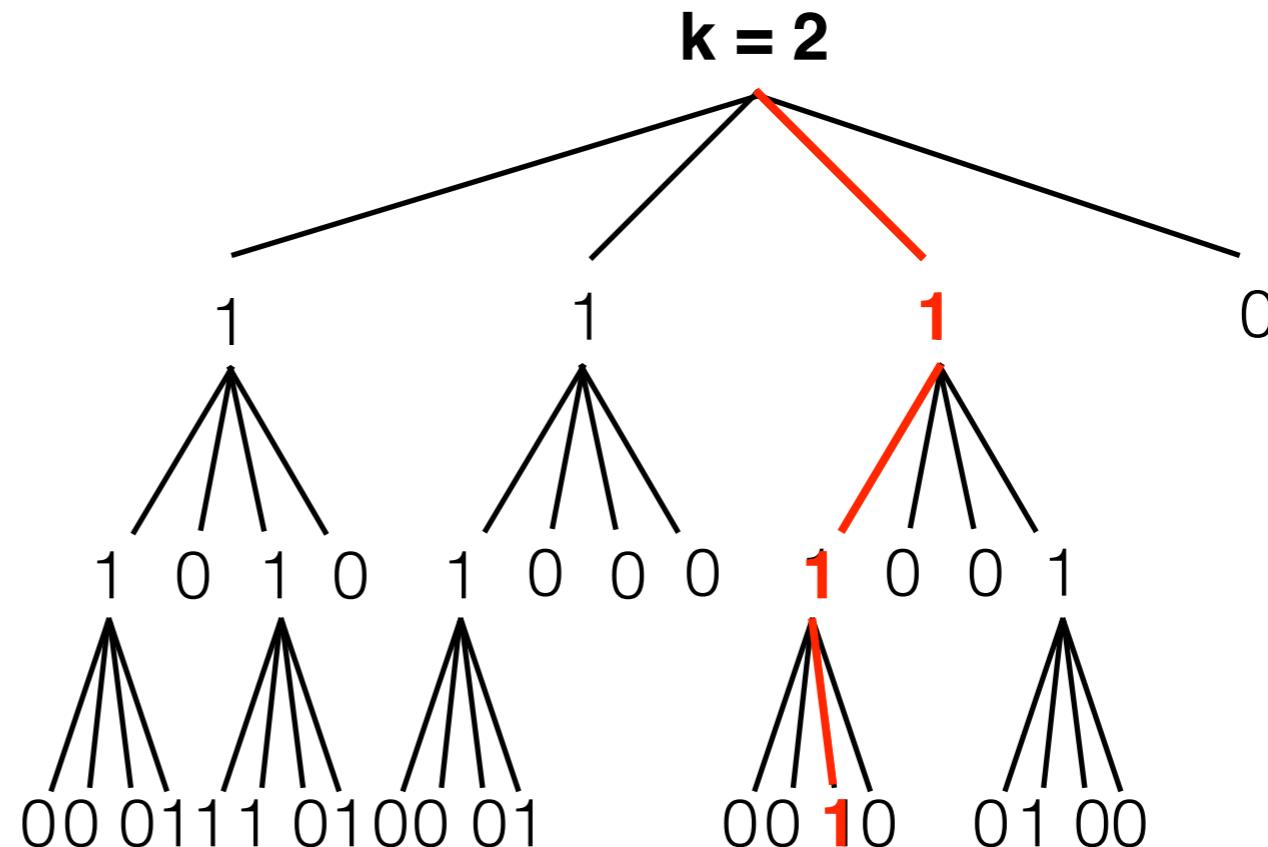


k^2 -trees

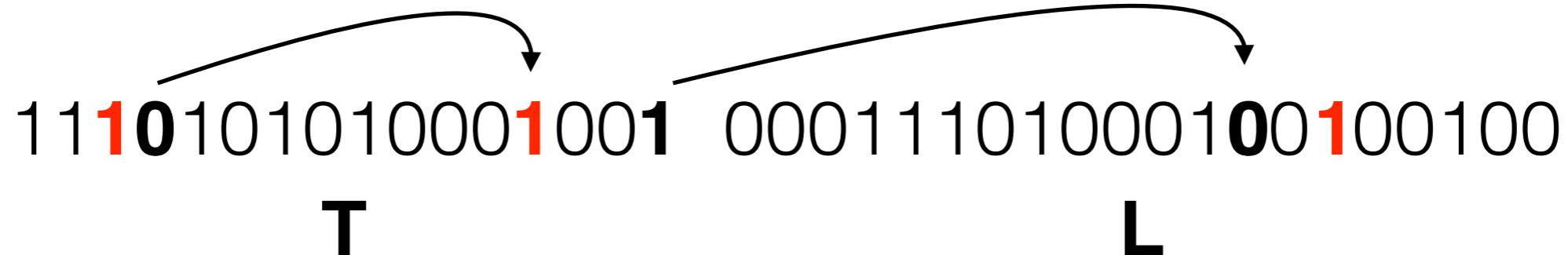
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k^2 -trees for Compact Web Graph Representation,
Brisaboa-Ladra-Navarro, SPIRE 2009

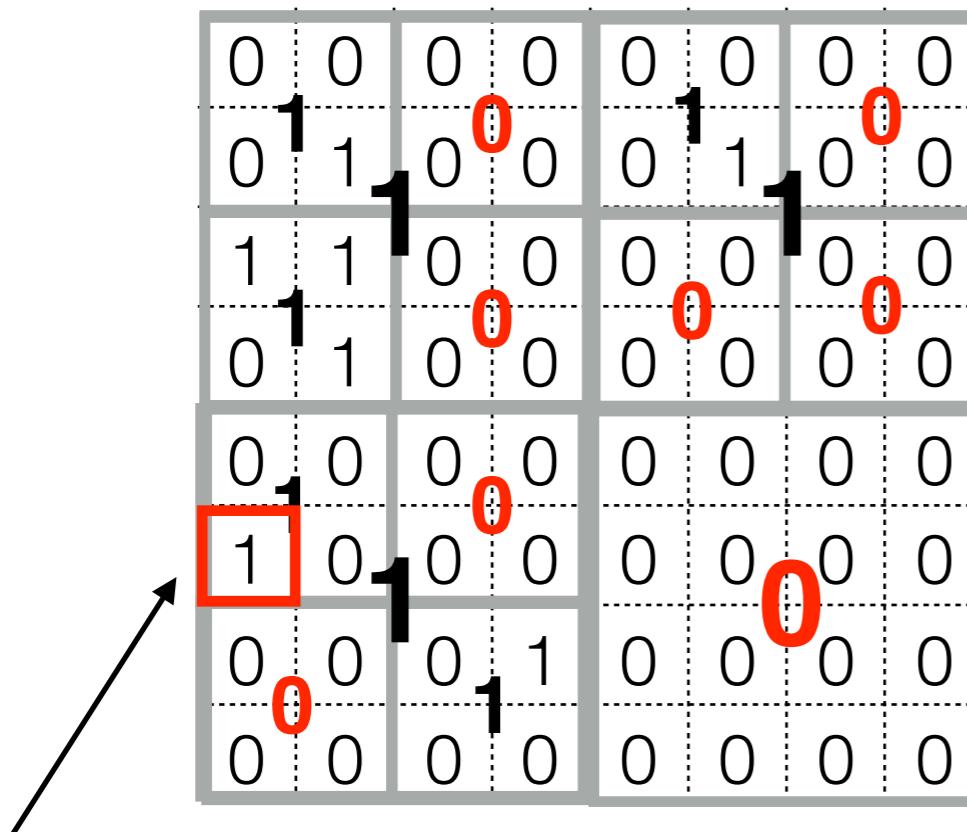


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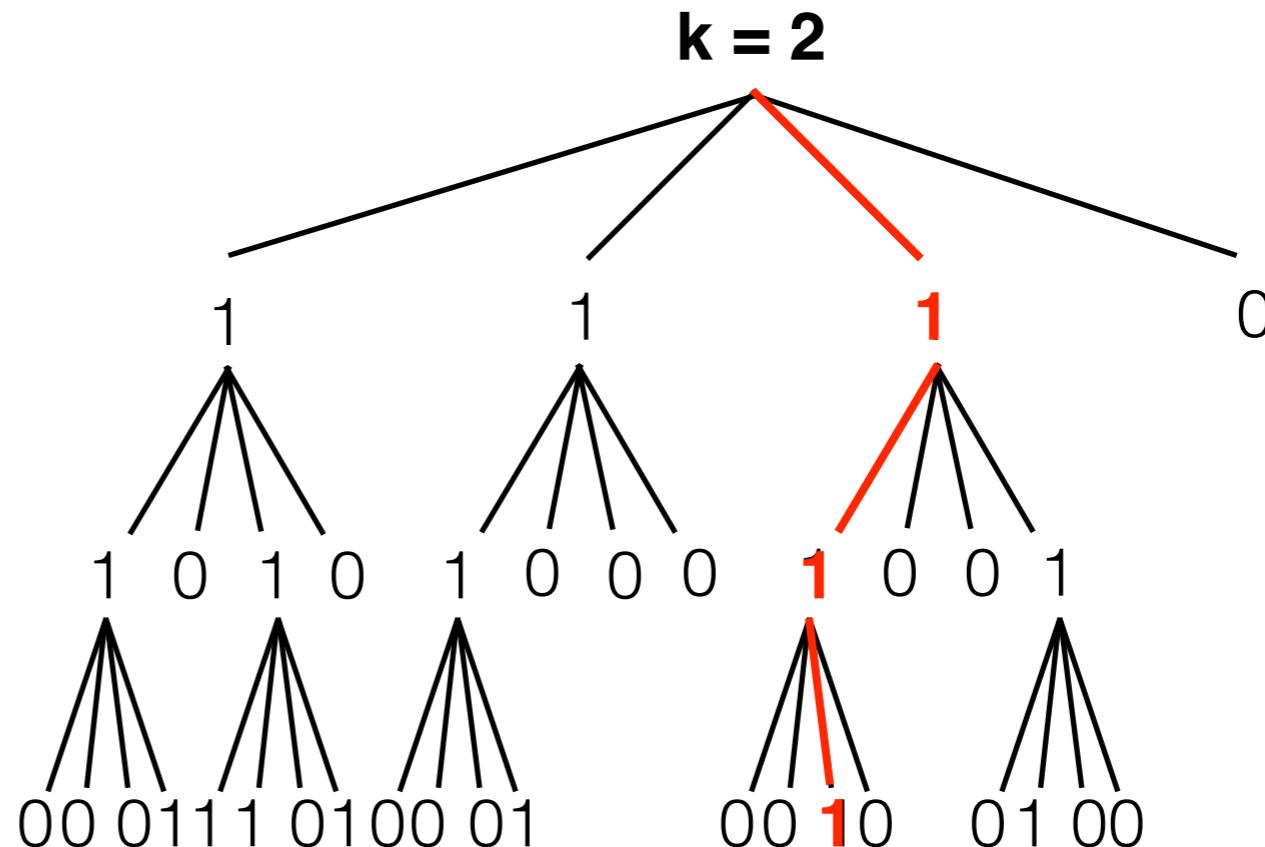


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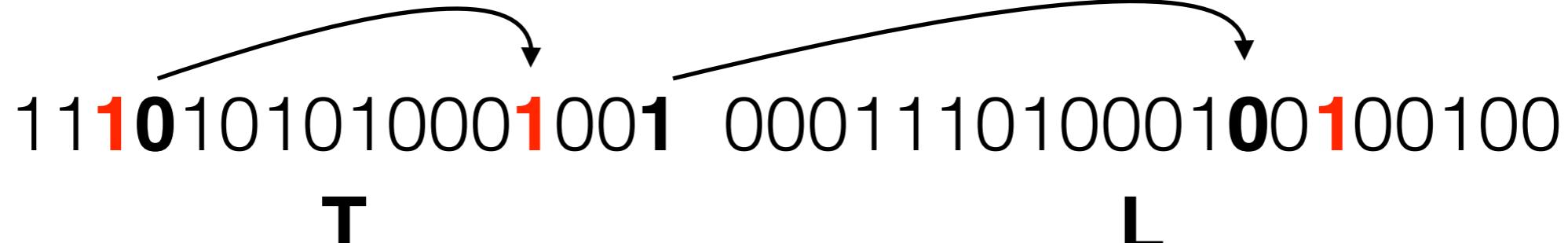
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Access



$\text{rank}_1 \times k^2$ on **T** in $O(1)$

k²-trees

n pages m links

$$h = \lceil \log_{k^2} n^2 \rceil$$

k²-trees

n pages m links

$h = \lceil \log_{k^2} n^2 \rceil$ So the total space is $mk^2 \lceil \log_{k^2} n^2 \rceil$ bits

k²-trees

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$$k^2 \sum_{i=0}^{\lfloor \log_{k^2} m \rfloor - 1} k^{2i} + mk^2(\lceil \log_{k^2} n \rceil - \lfloor \log_{k^2} m \rfloor)$$

$$= k^2 \frac{m-1}{k^2-1} + mk^2 \log_{k^2} \frac{n^2}{m}$$

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Information theoretic
lower bound

$$\log \binom{n^2}{m} \approx m \log \frac{n^2}{m} + O(m)$$

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Information theoretic
lower bound

$$k = 2$$

$$2m \log \frac{n^2}{m} + O(m)$$

$$\log \binom{n^2}{m} \approx m \log \frac{n^2}{m} + O(m)$$

2X away

k^2 -trees

File	Pages (n)	Links (m)	Links/page
EU (2005)	862,664	19,235,140	22.30
Indochina (2002)	7,414,866	194,109,311	26.18
UK (2002)	18,520,486	298,113,762	16.10

Crawl	k^2 -tree	WebGraph (dir+rev)
EU	3.22	5.62
Indochina	1.23	2.04
UK	2.04	3.29

Space results in bits x link.

Block-trees

A block-tree divides a *string* into fixed-size blocks and those appearing earlier are represented with pointers.

Queries on LZ-Bounded Encodings,
Belazzougui *et al.*, DCC 2014

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input = AABAABBAABAAACBBA

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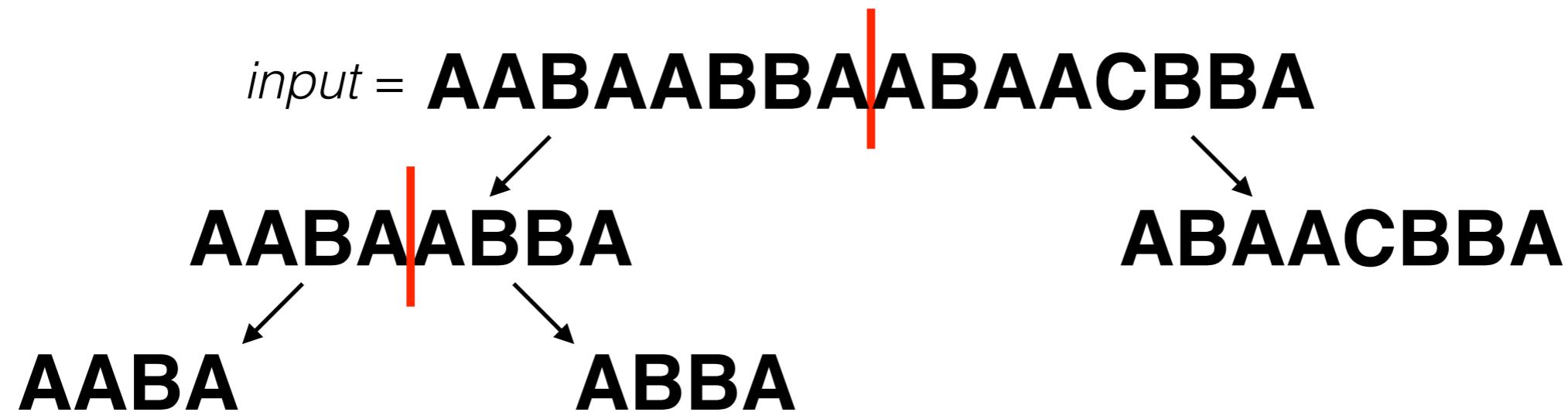
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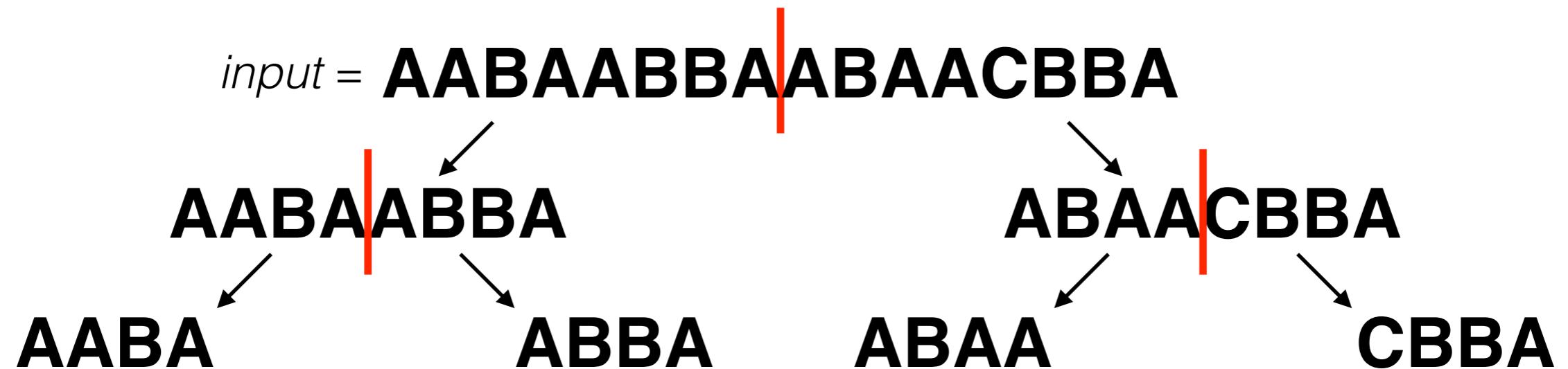
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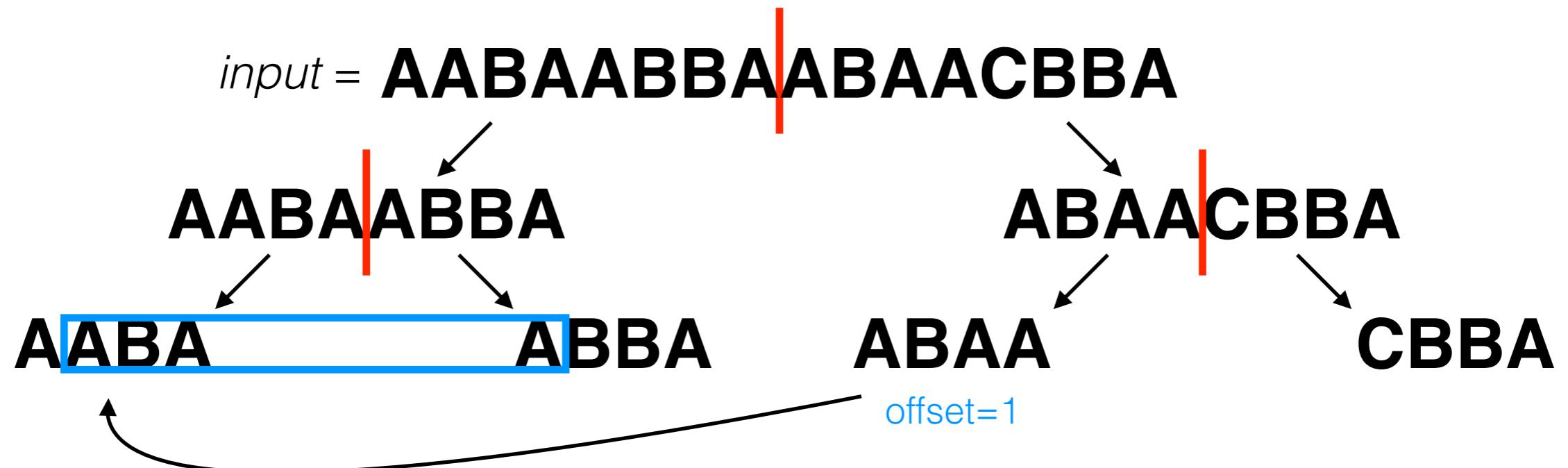
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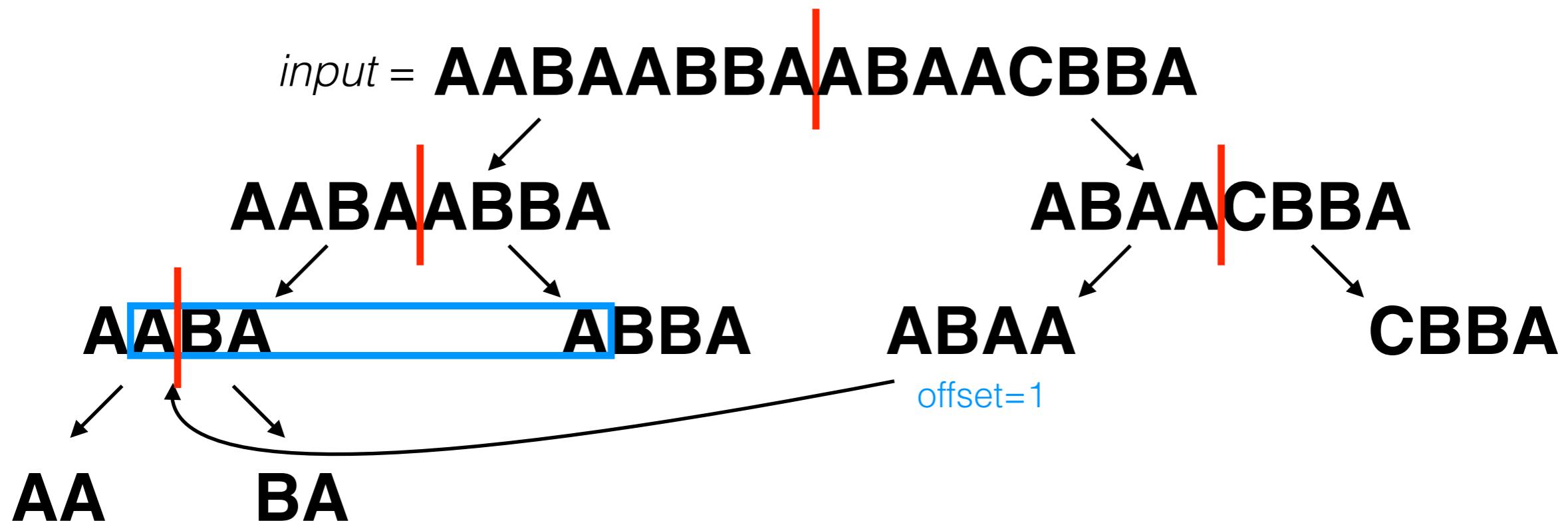
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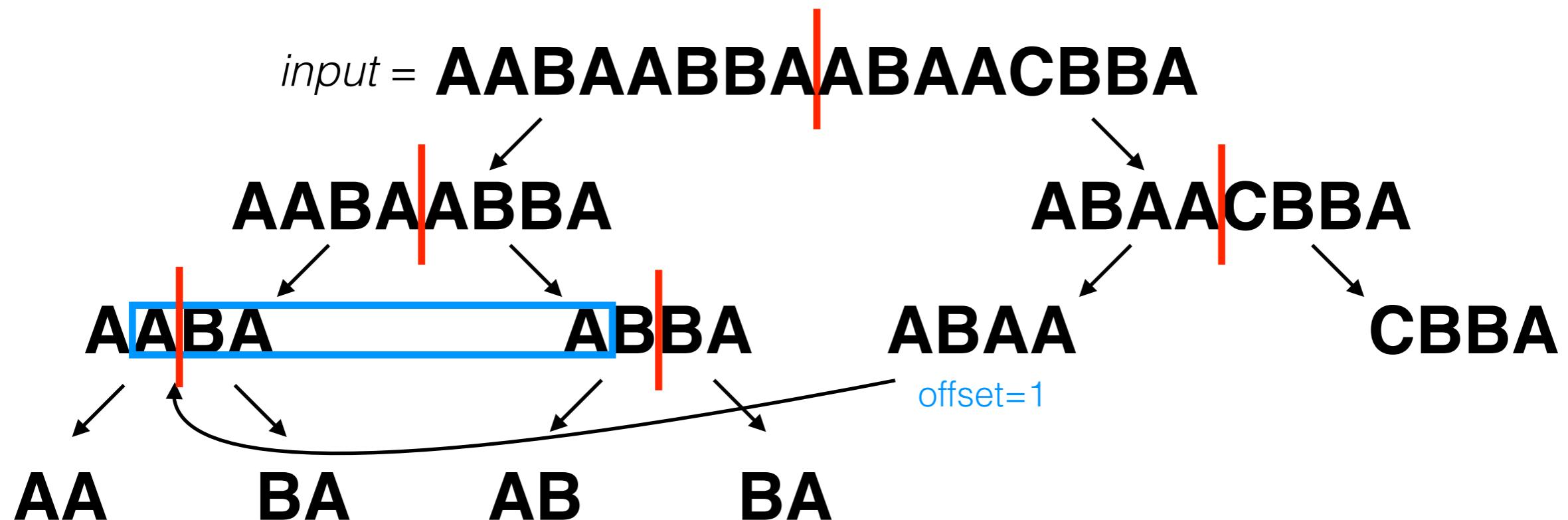
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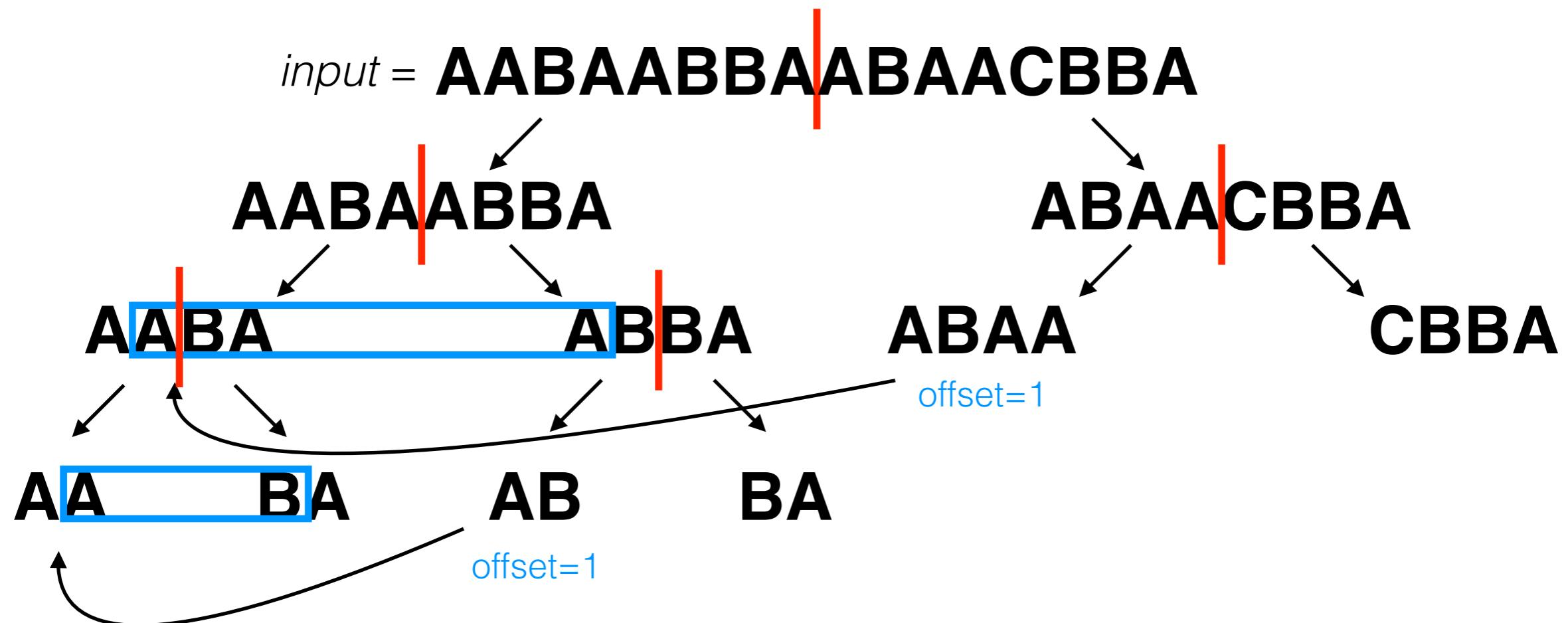
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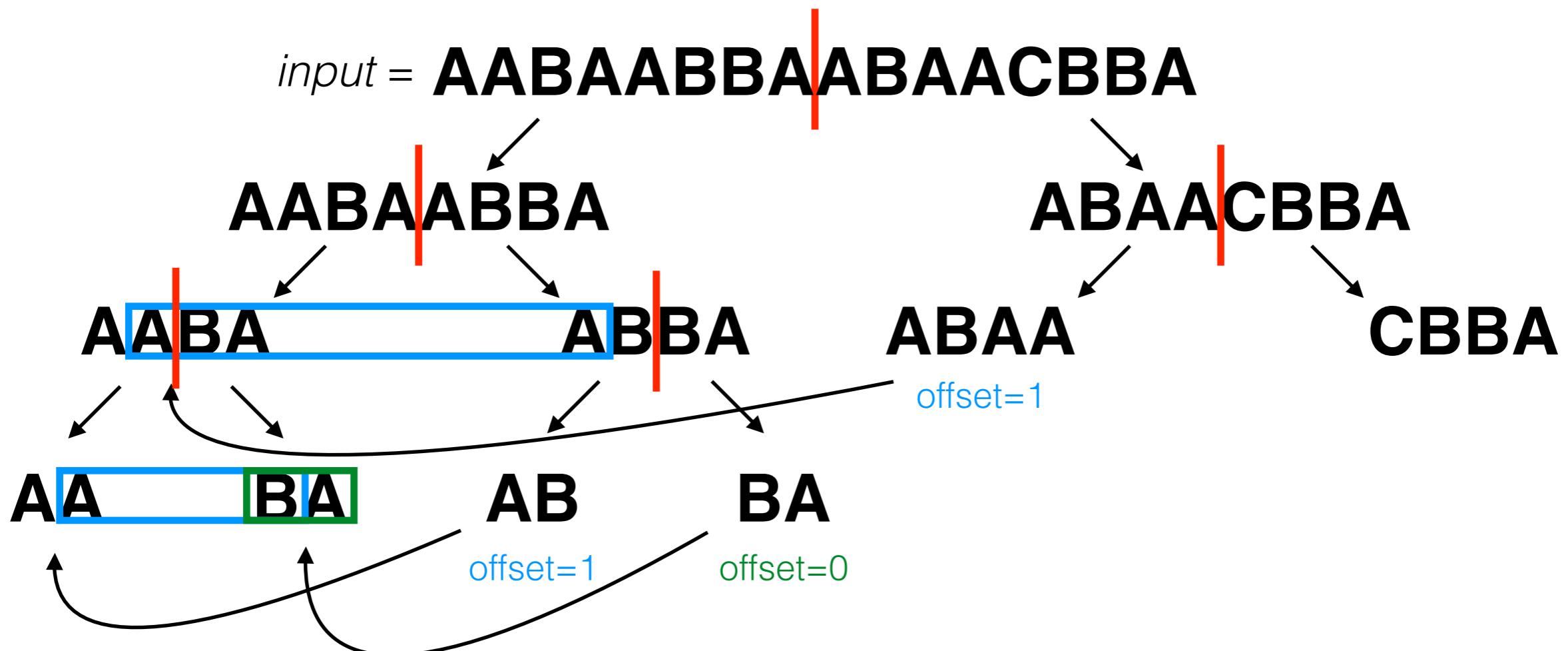
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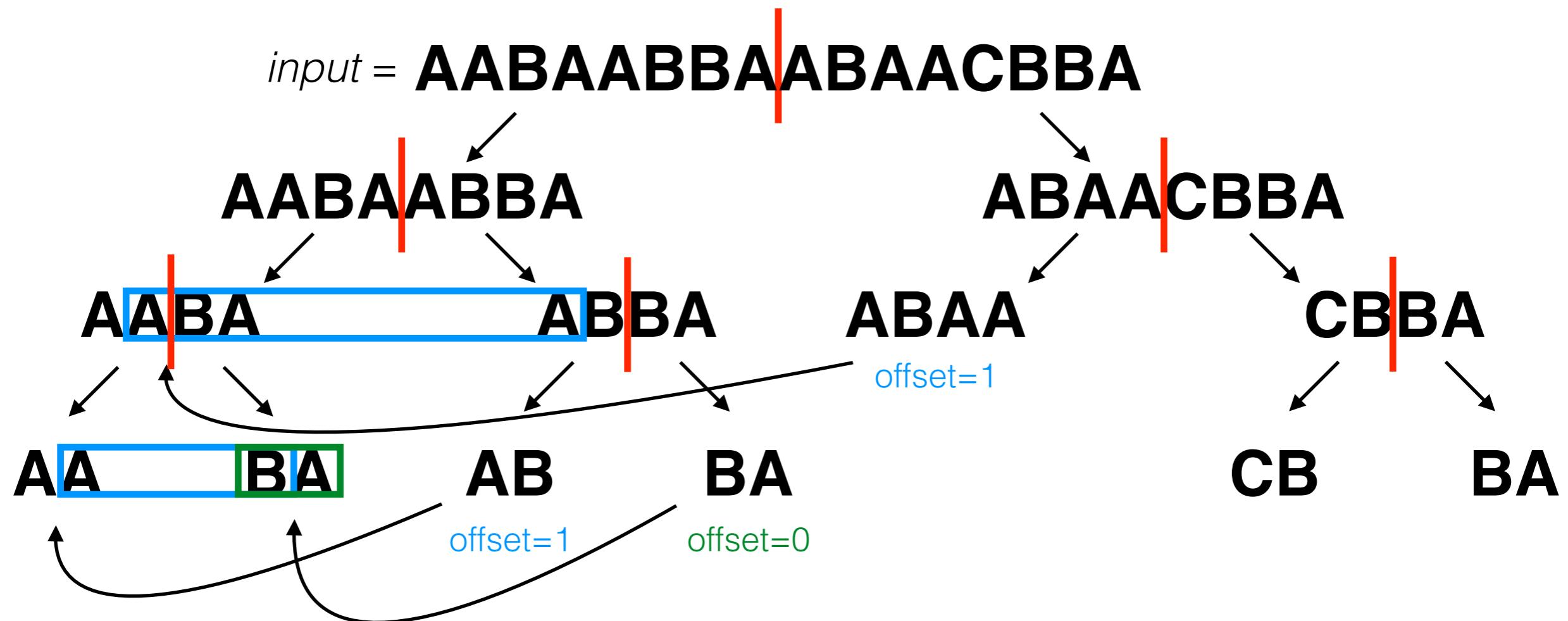
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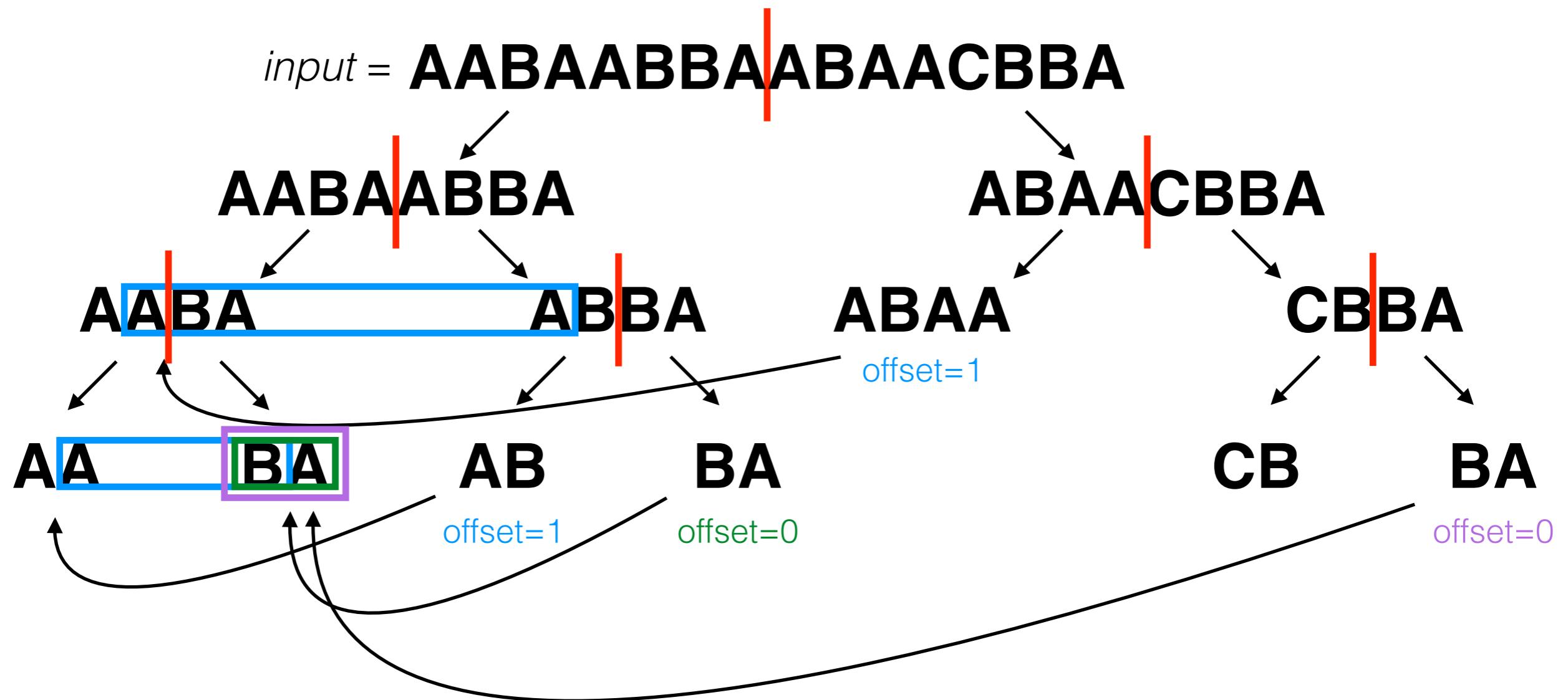
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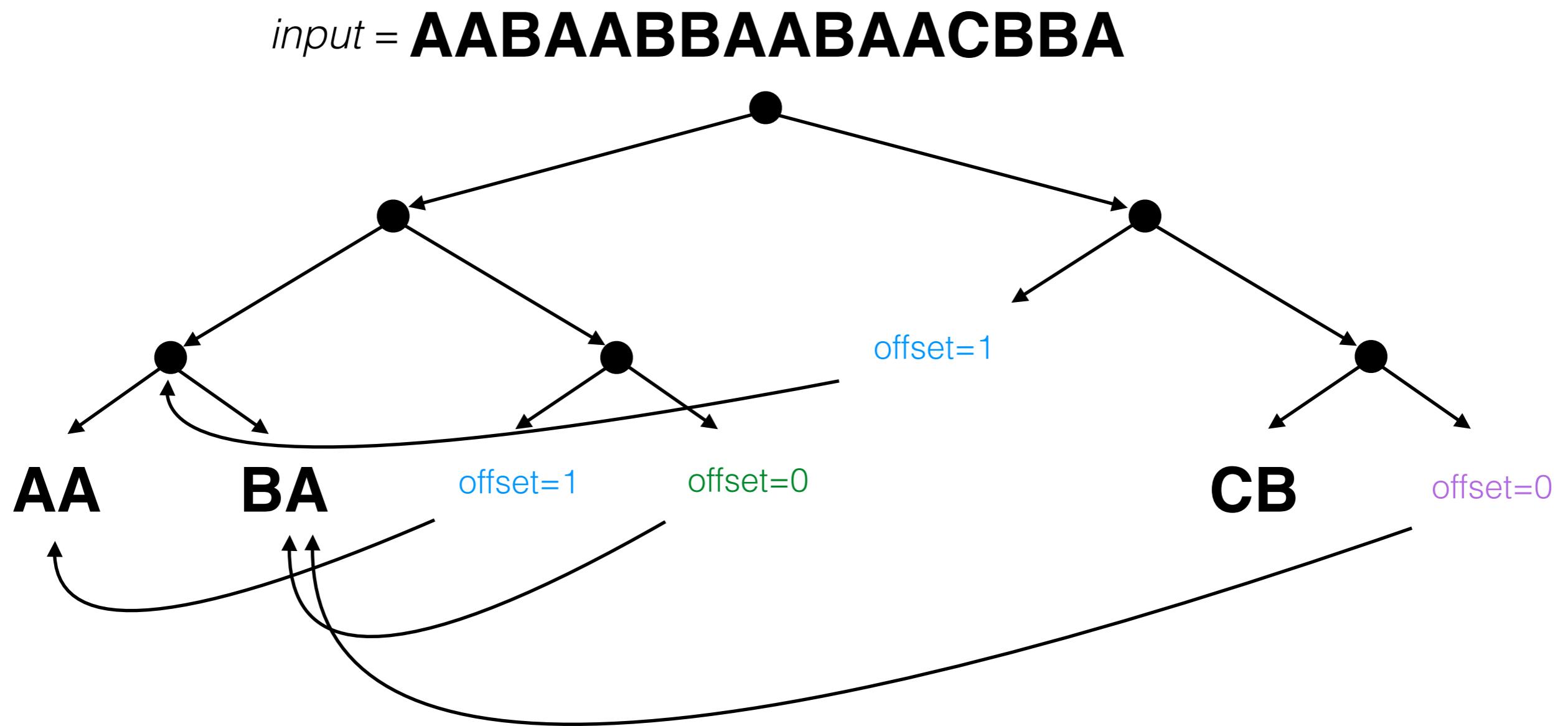
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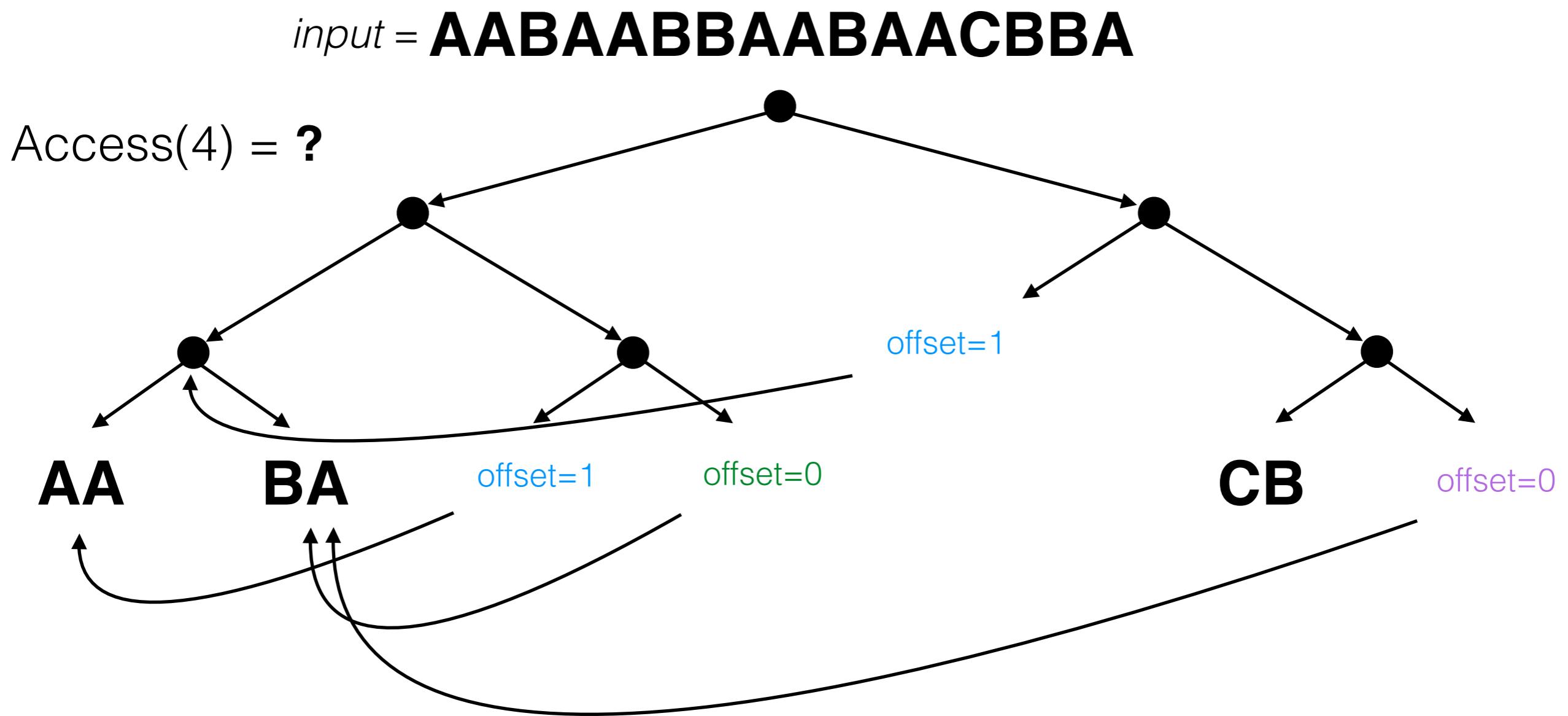
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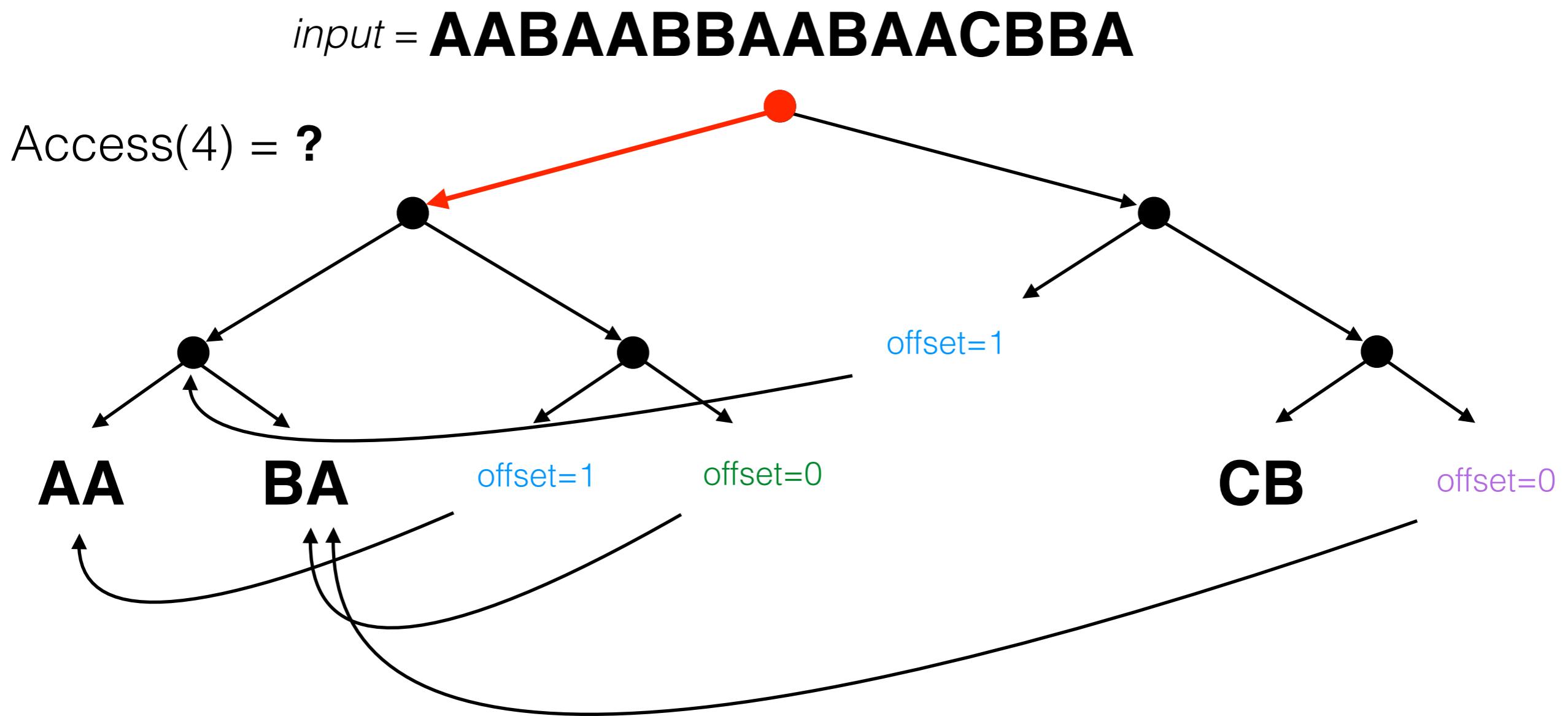
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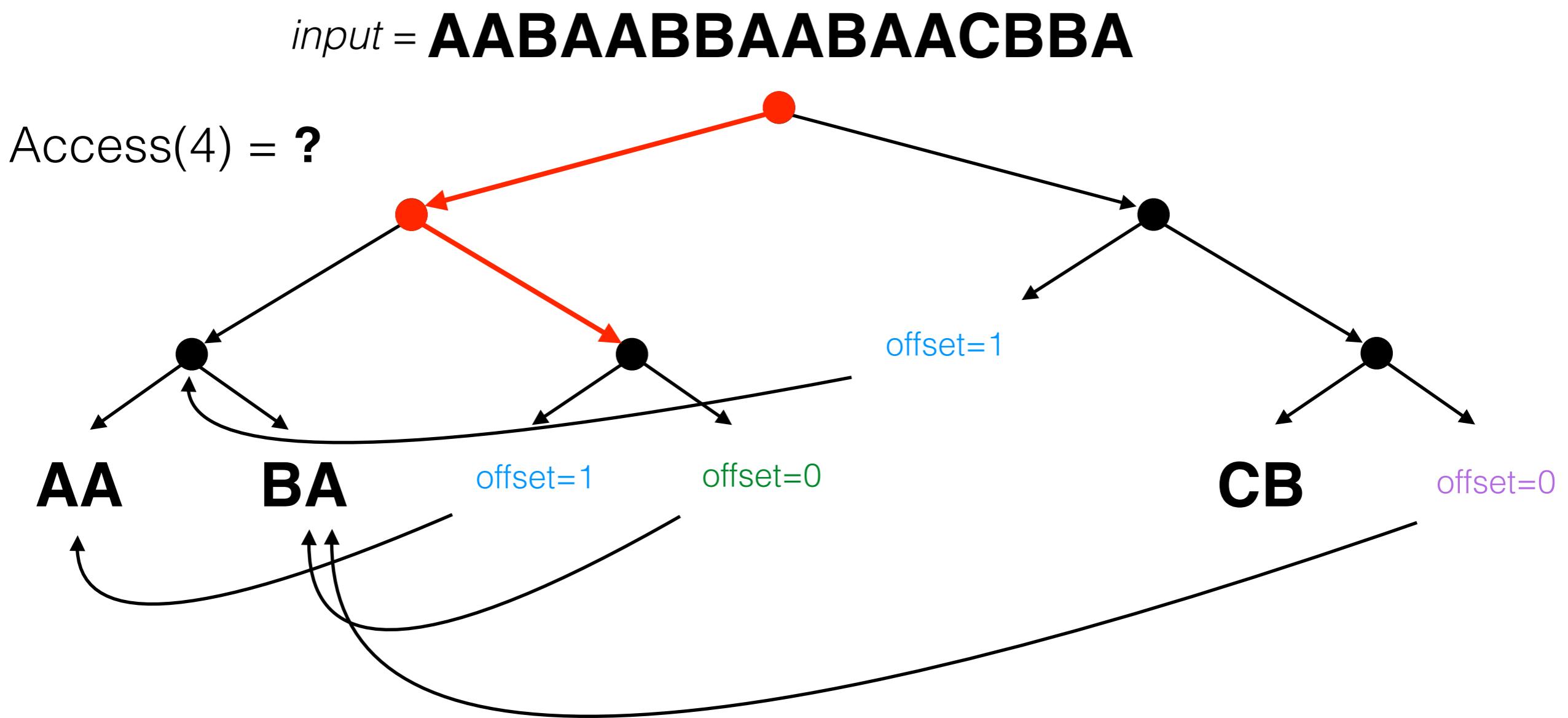
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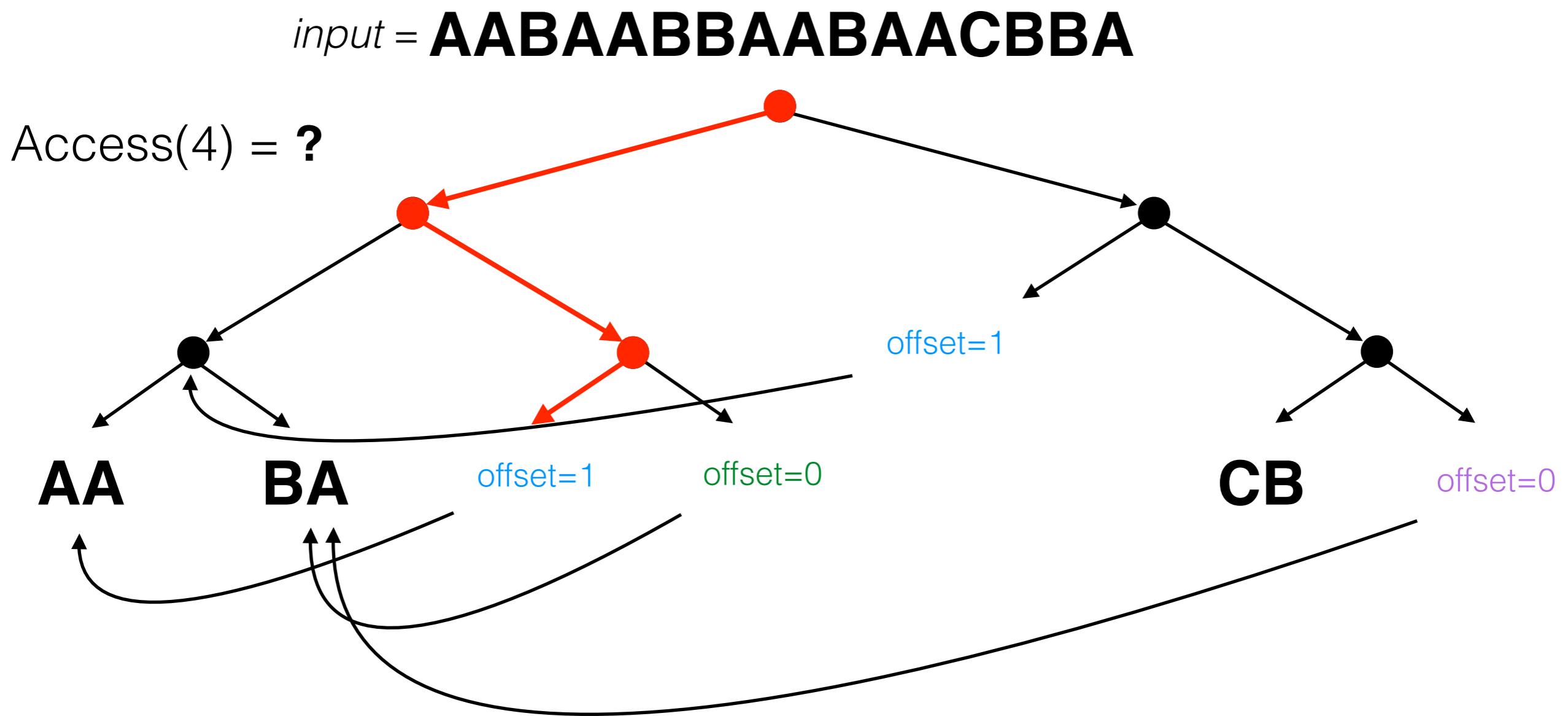
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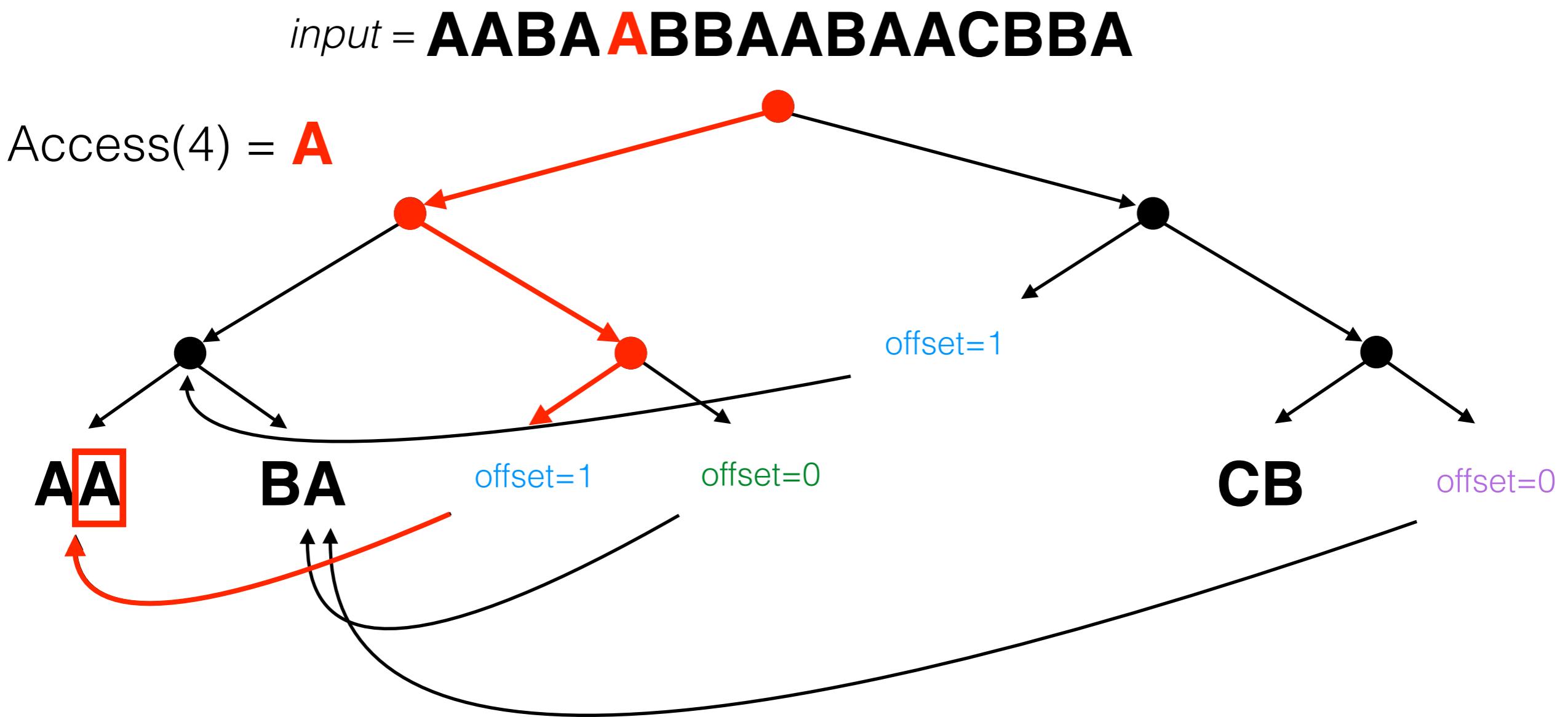
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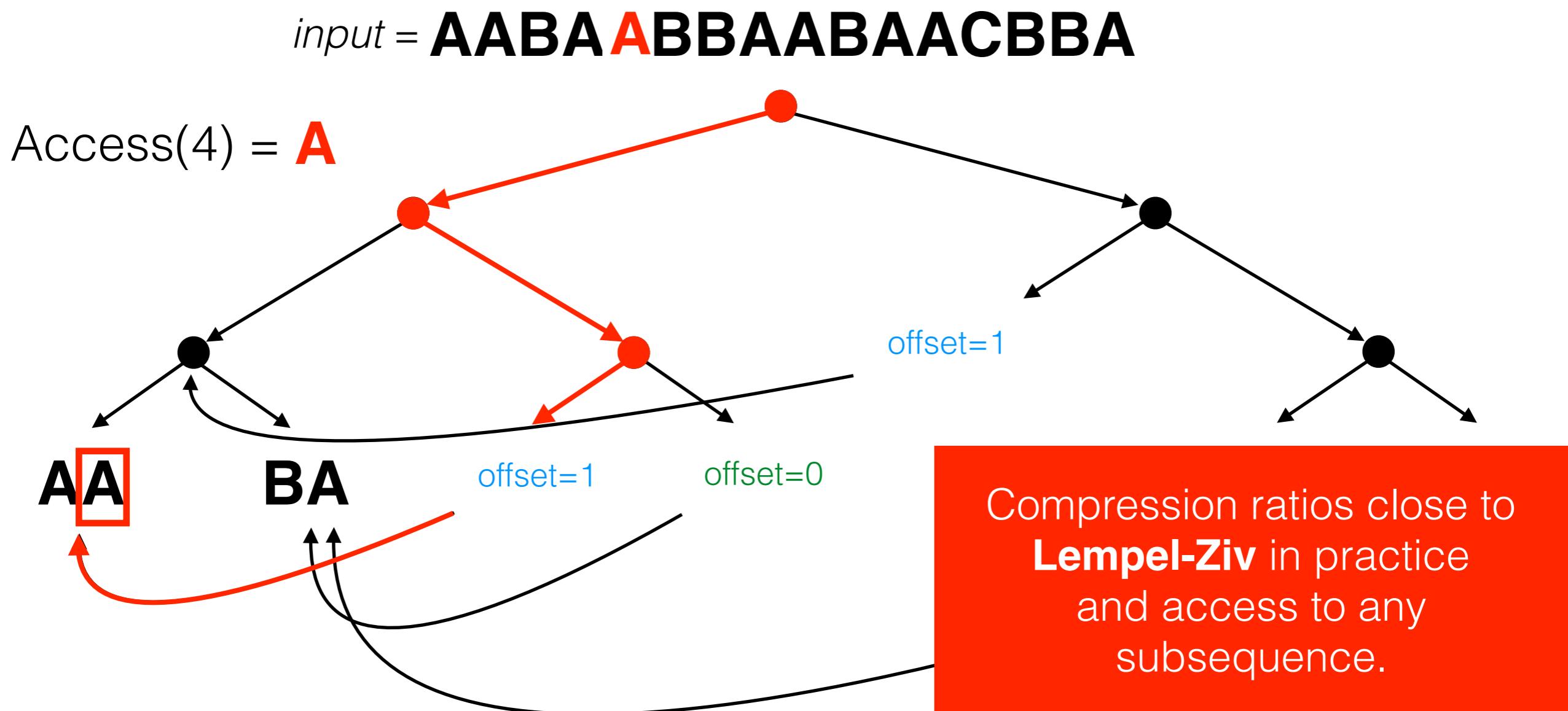
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Block-trees

n size of the string

σ alphabet size

Stop recursion when

$$B \lceil \log \sigma \rceil > \lceil \log n \rceil$$

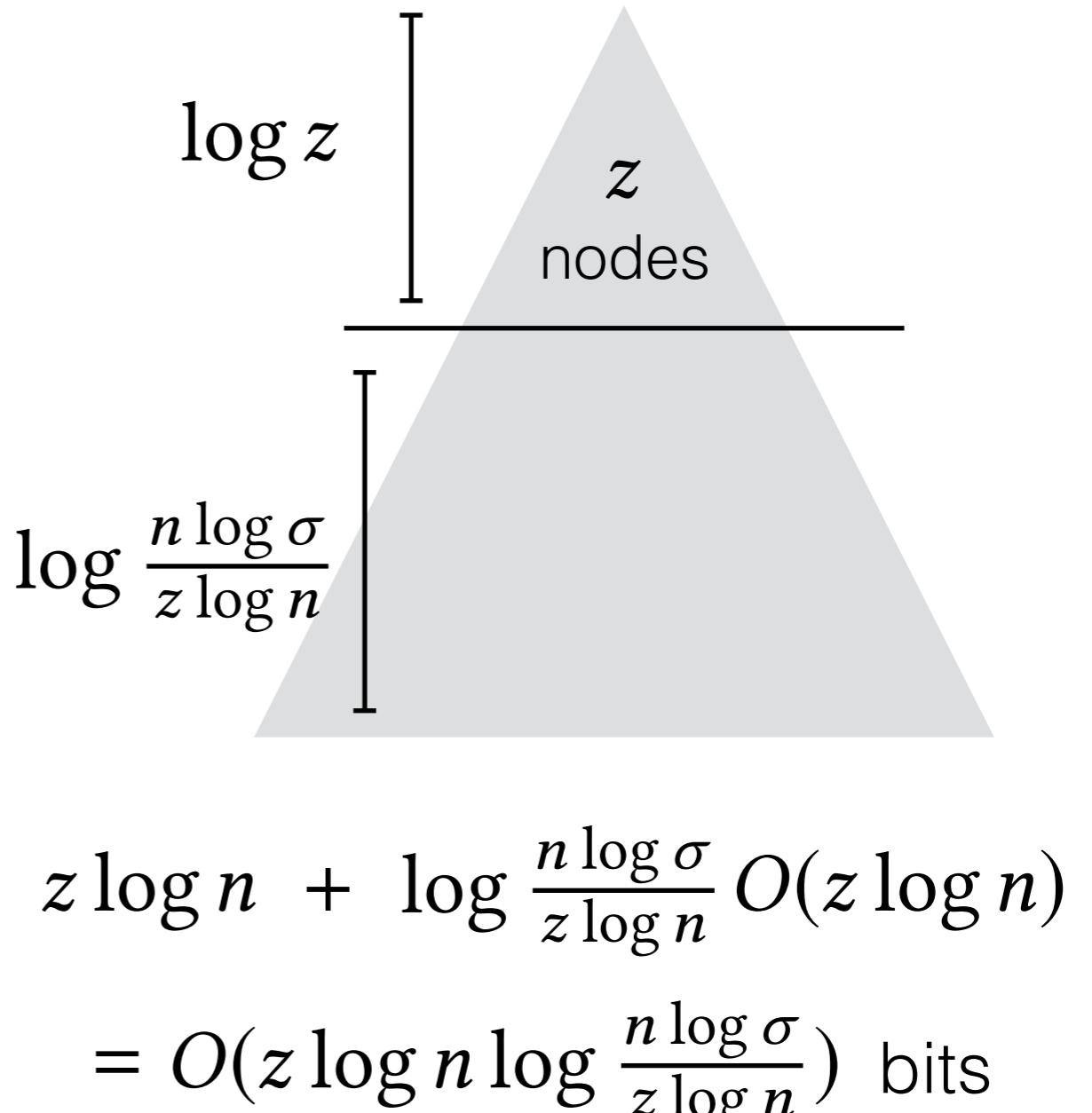
$$B = O(\log n / \log \sigma)$$

$$h = \log \frac{n \log \sigma}{\log n}$$

At level i

t_i nodes take

$$O(t_i \log n) \text{ bits}$$

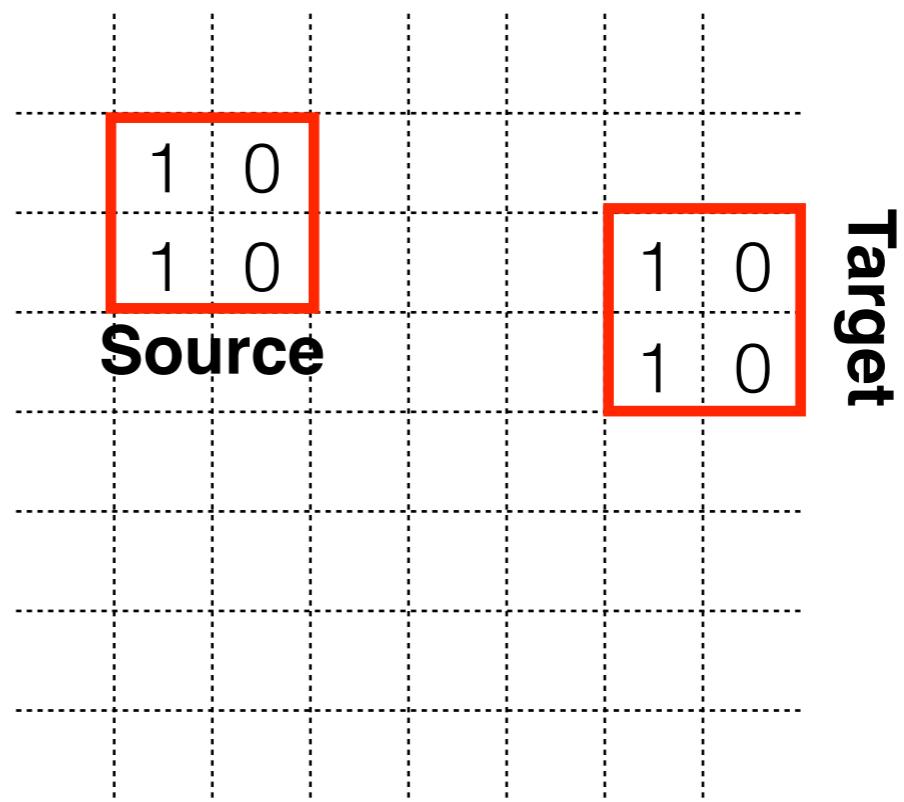


2D Block-trees

A *hybrid* between the k^2 -tree to exploit the clustering of the 0s and the block tree to exploit the repetitiveness of the adjacency matrix.

Two-Dimensional Block Trees,
Brisaboa, Gagie, Gómez-Brandón, Navarro,
DCC 2018

A **source** block may overlap up to **4** adjacent blocks.

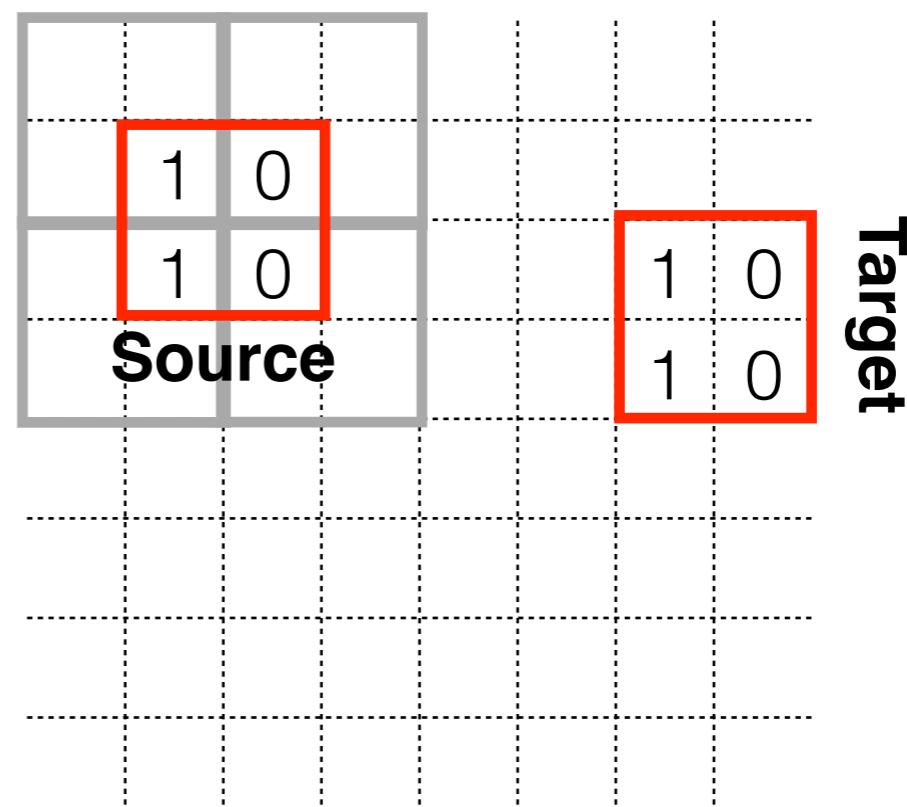


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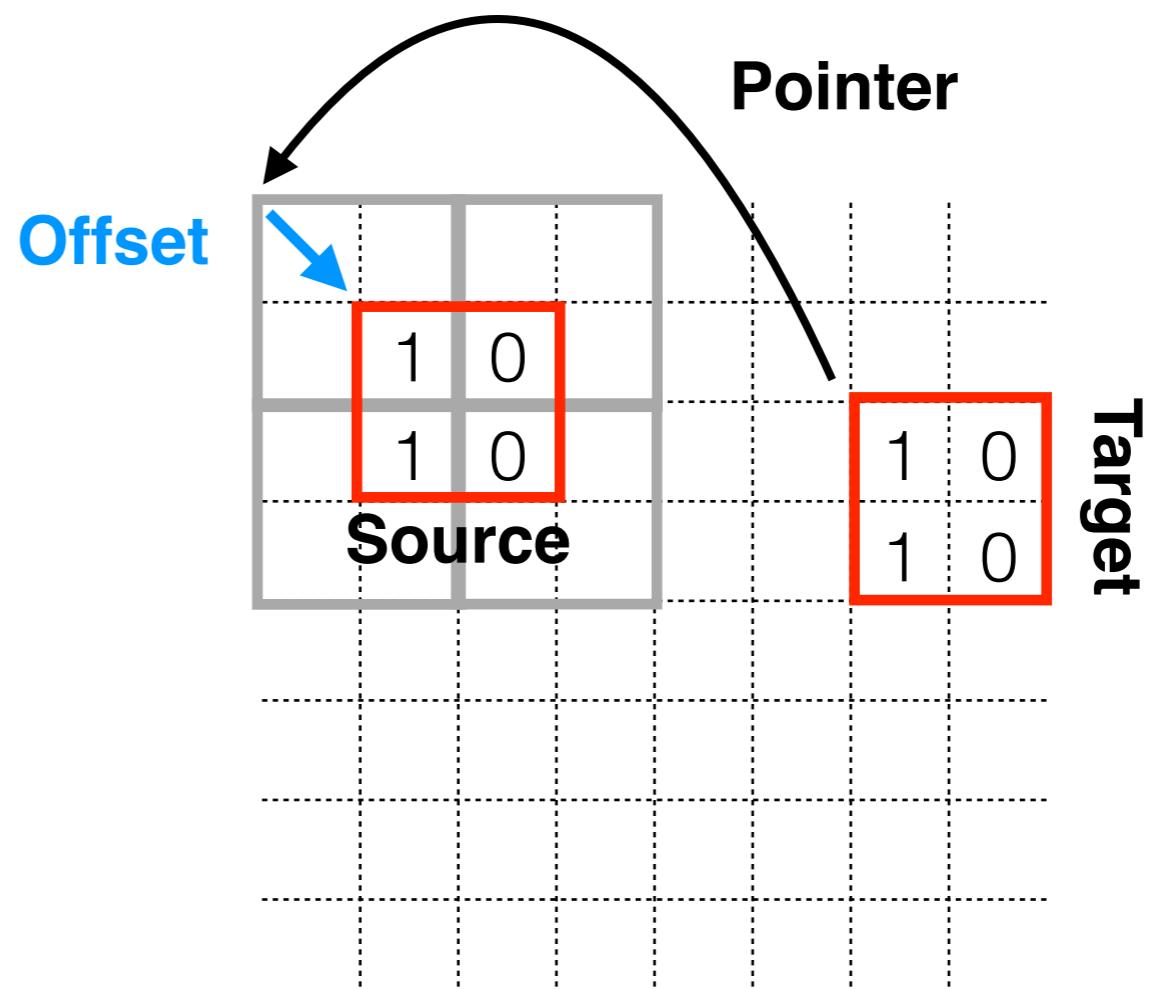
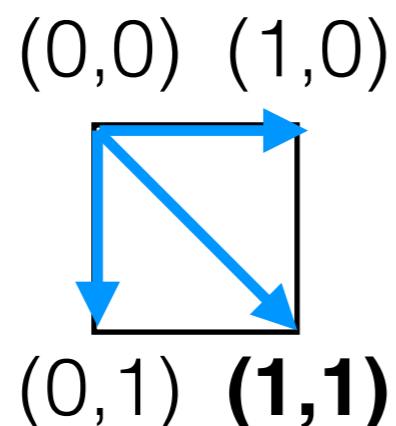


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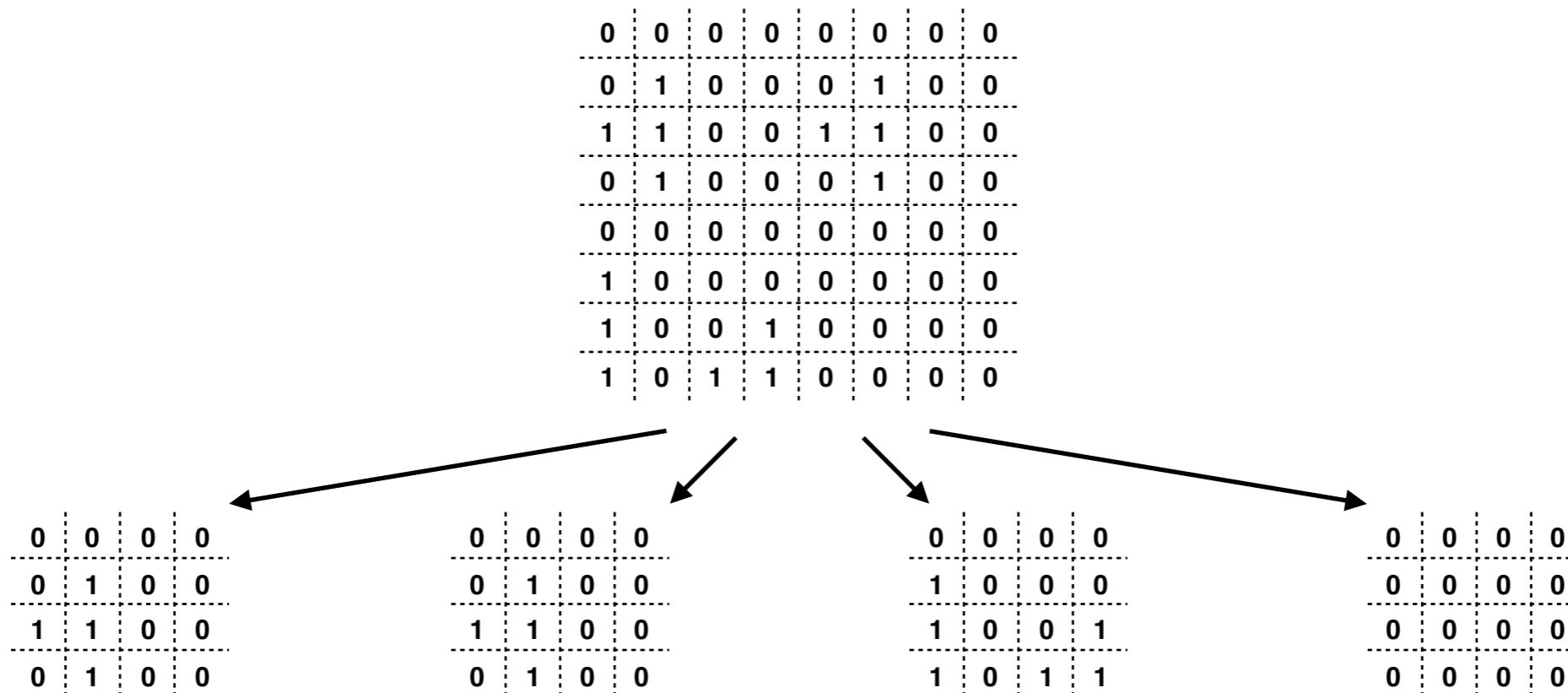
A **source** block may overlap up to **4** adjacent blocks.



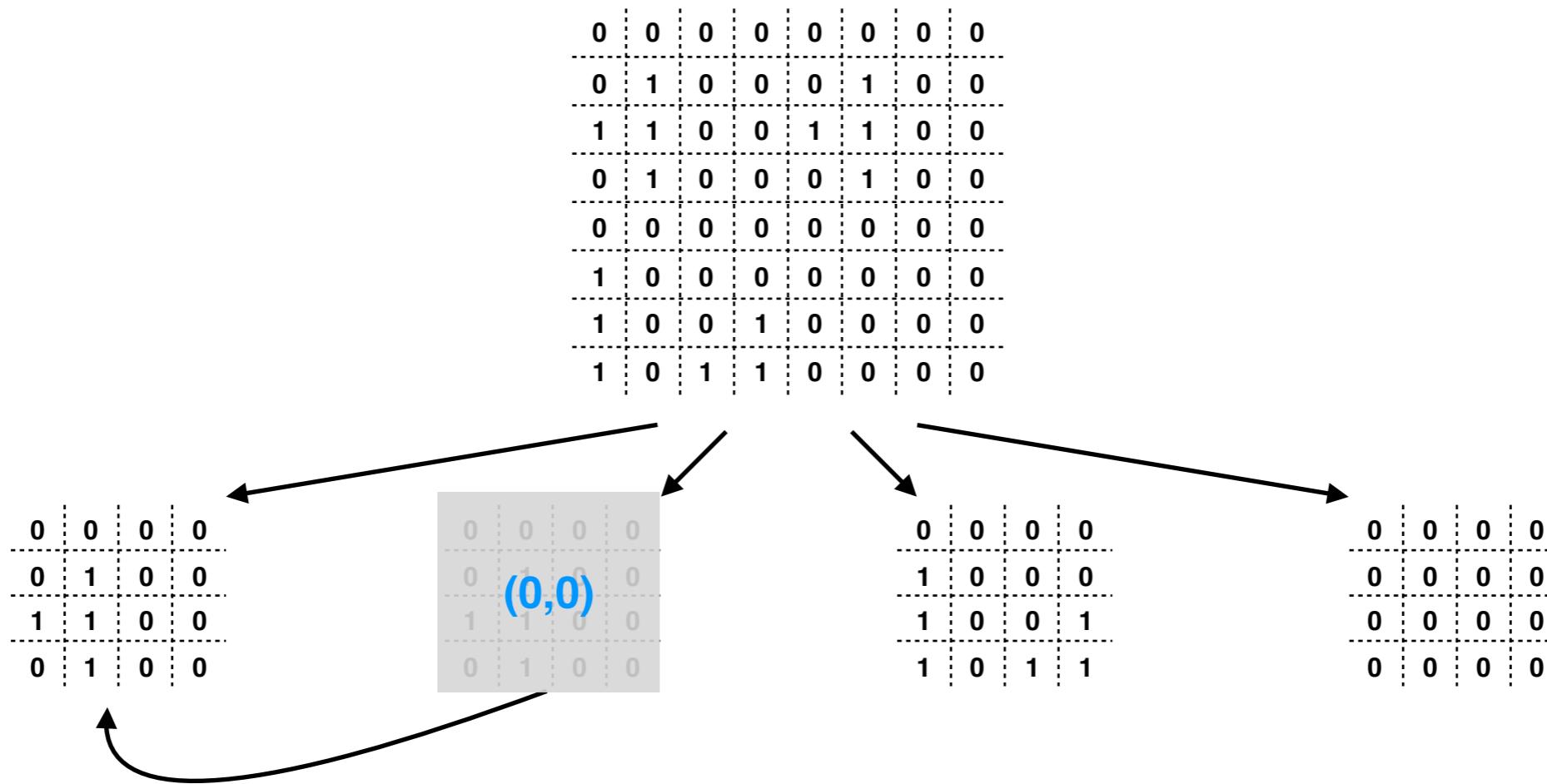
2D Block-trees

0	0	0	0	0	0	0	0
0	1	0	0	0	1	0	0
1	1	0	0	1	1	0	0
0	1	0	0	0	1	0	0
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
1	0	0	1	0	0	0	0
1	0	1	1	0	0	0	0

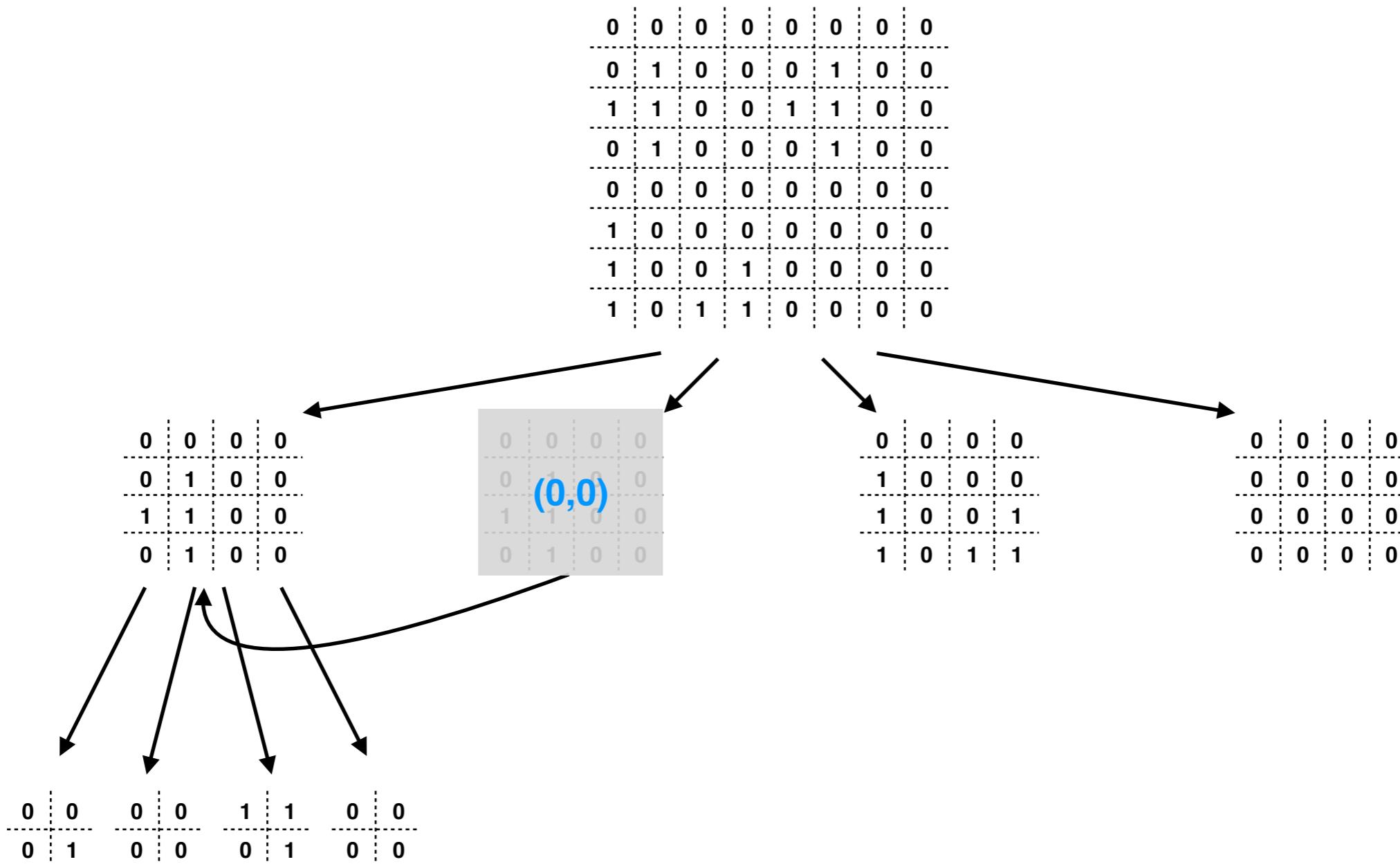
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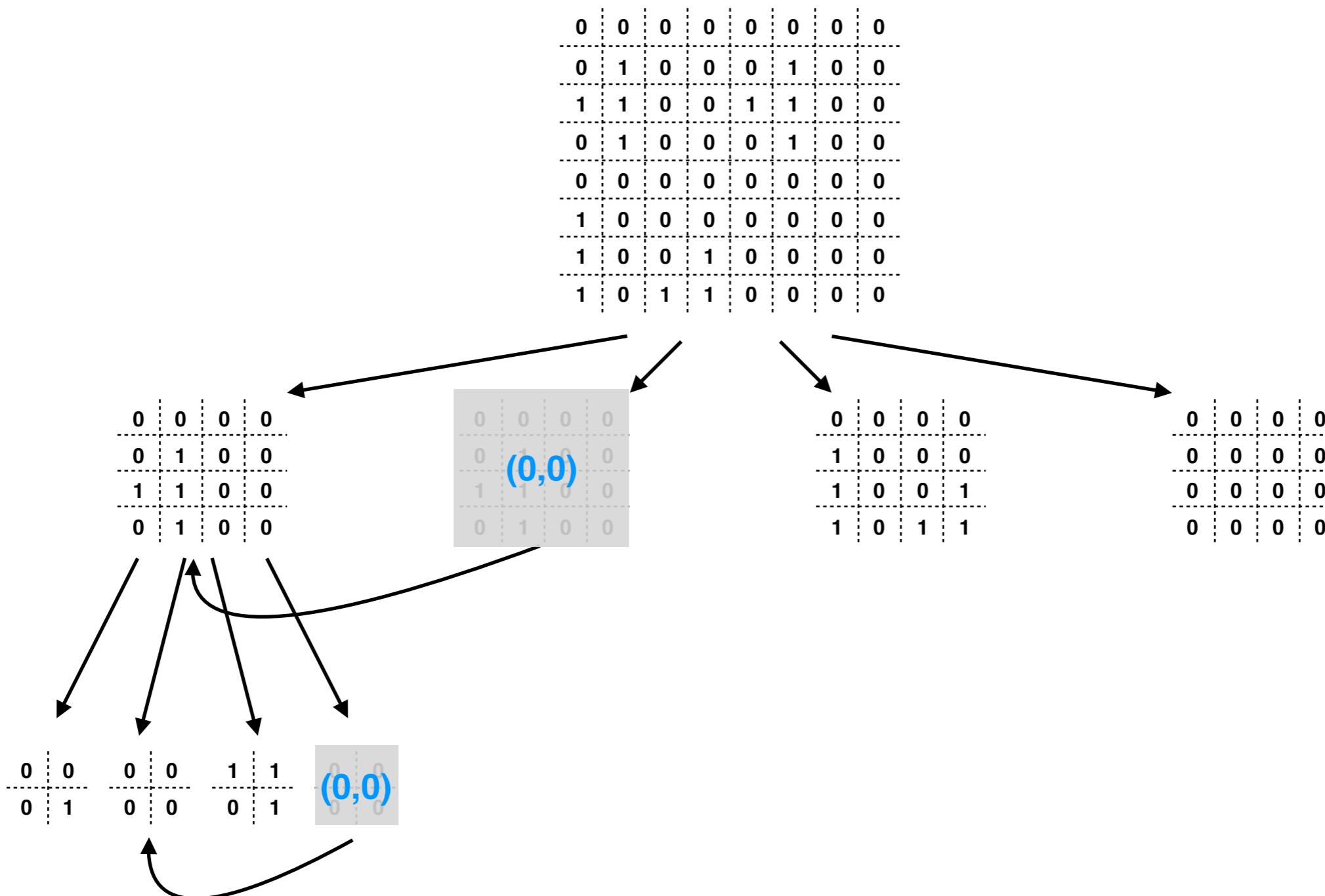
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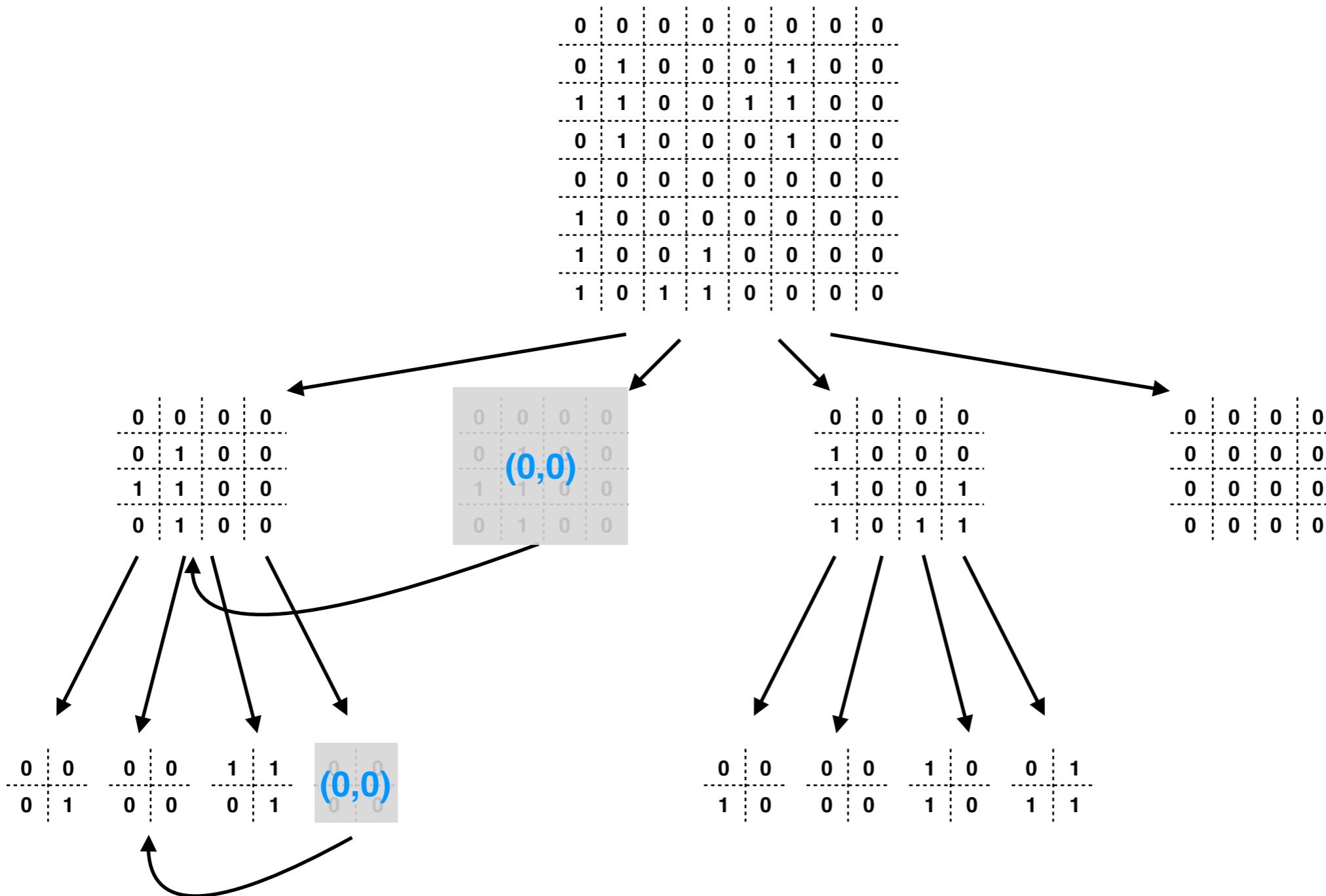
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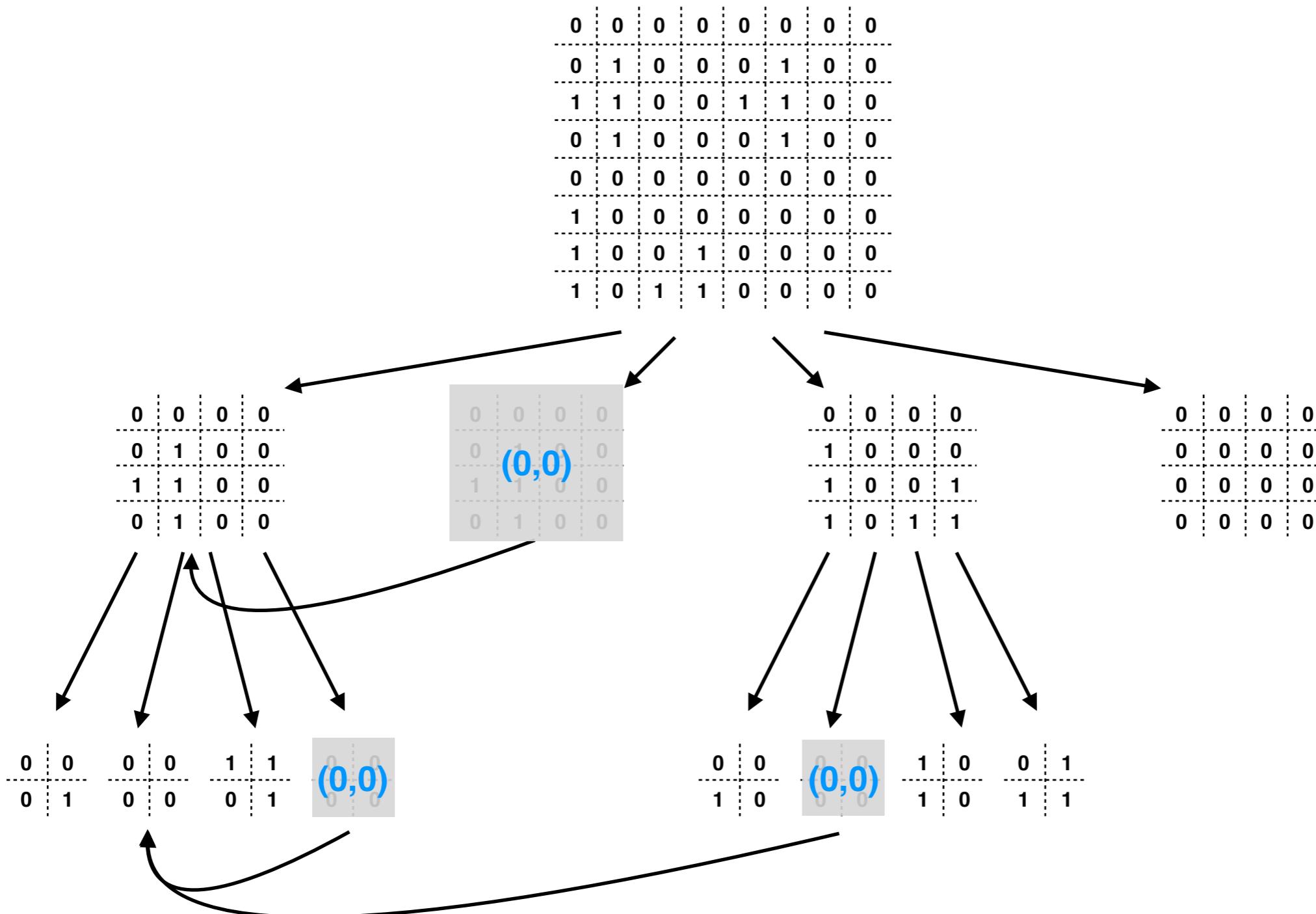
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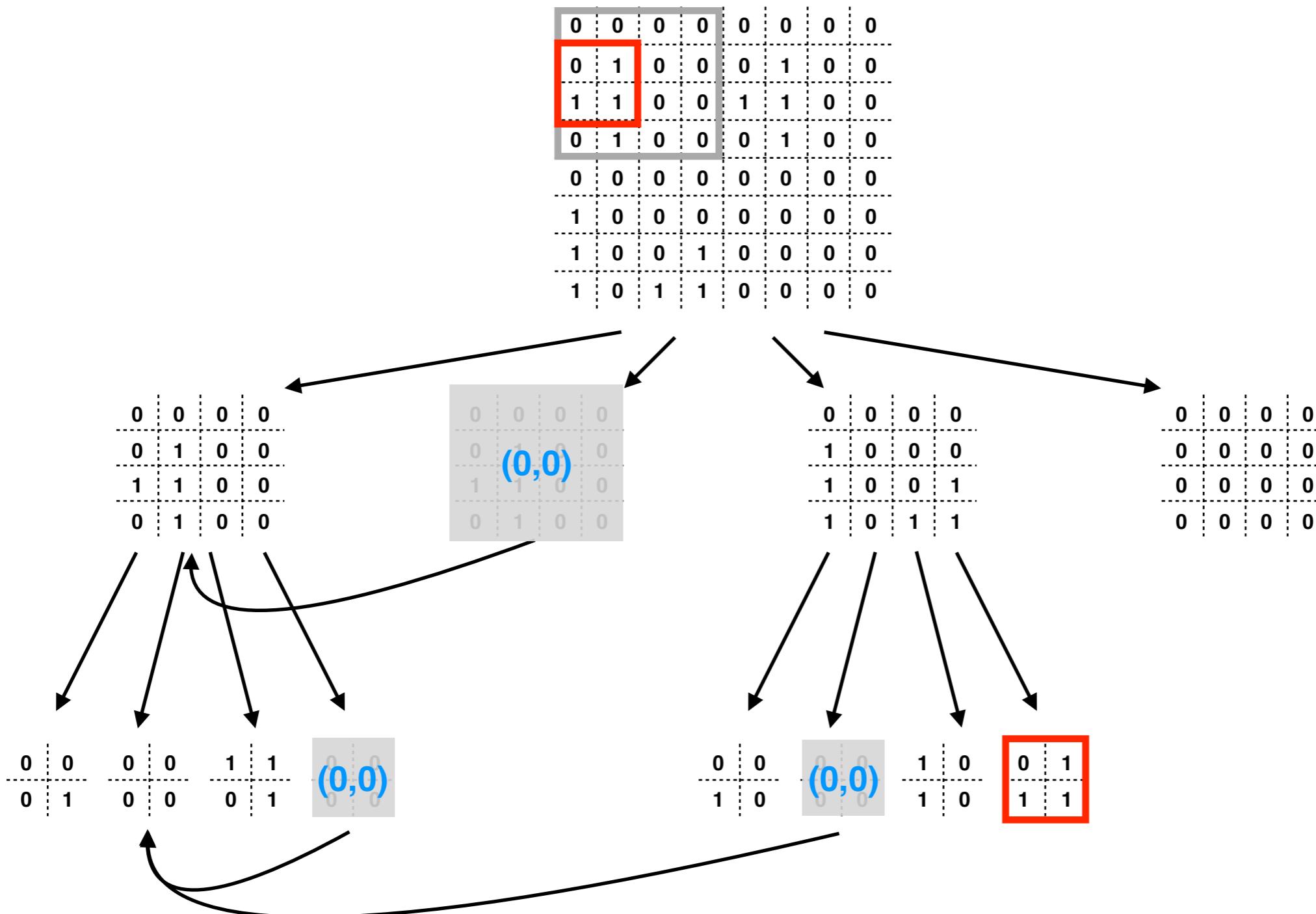
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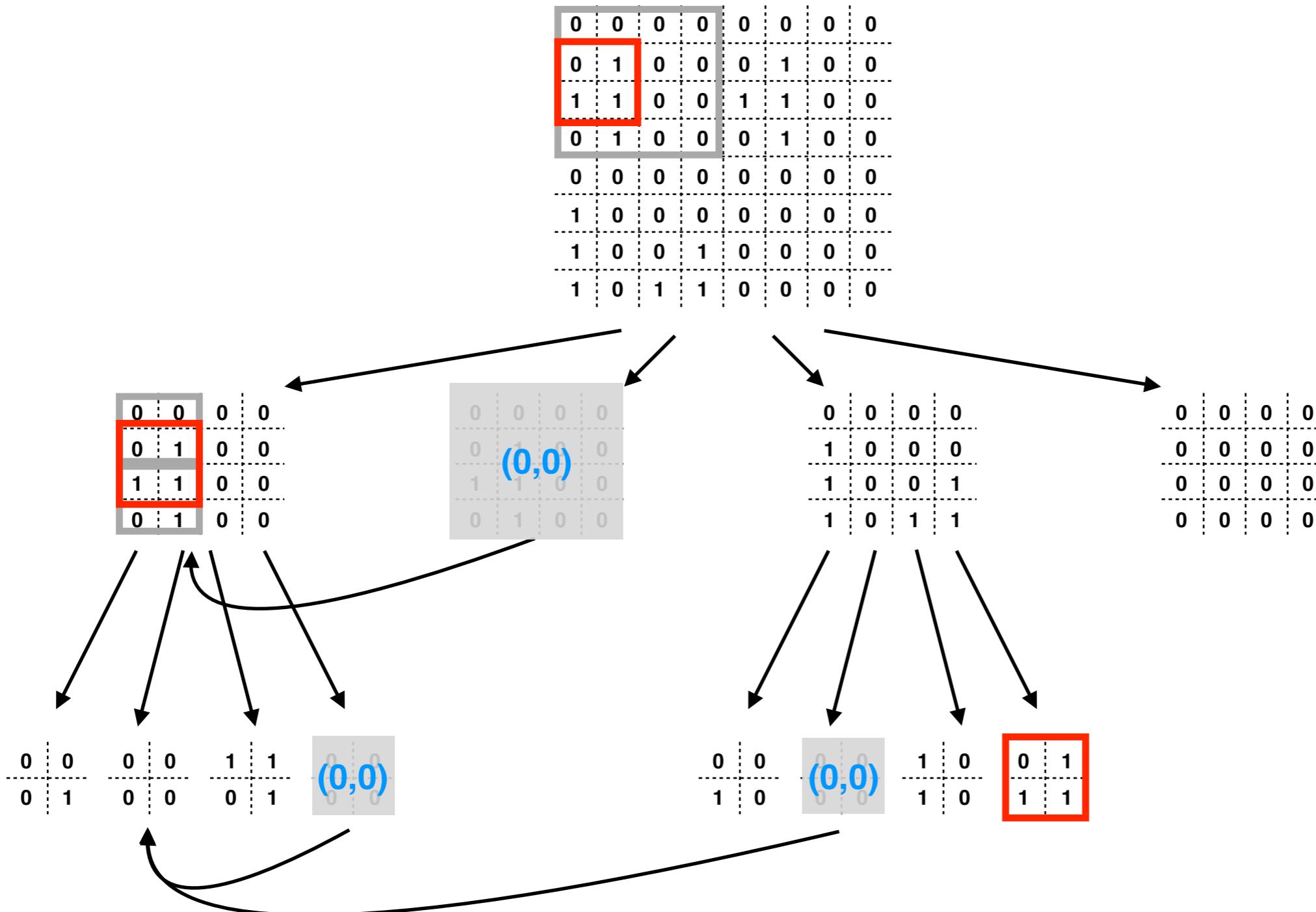
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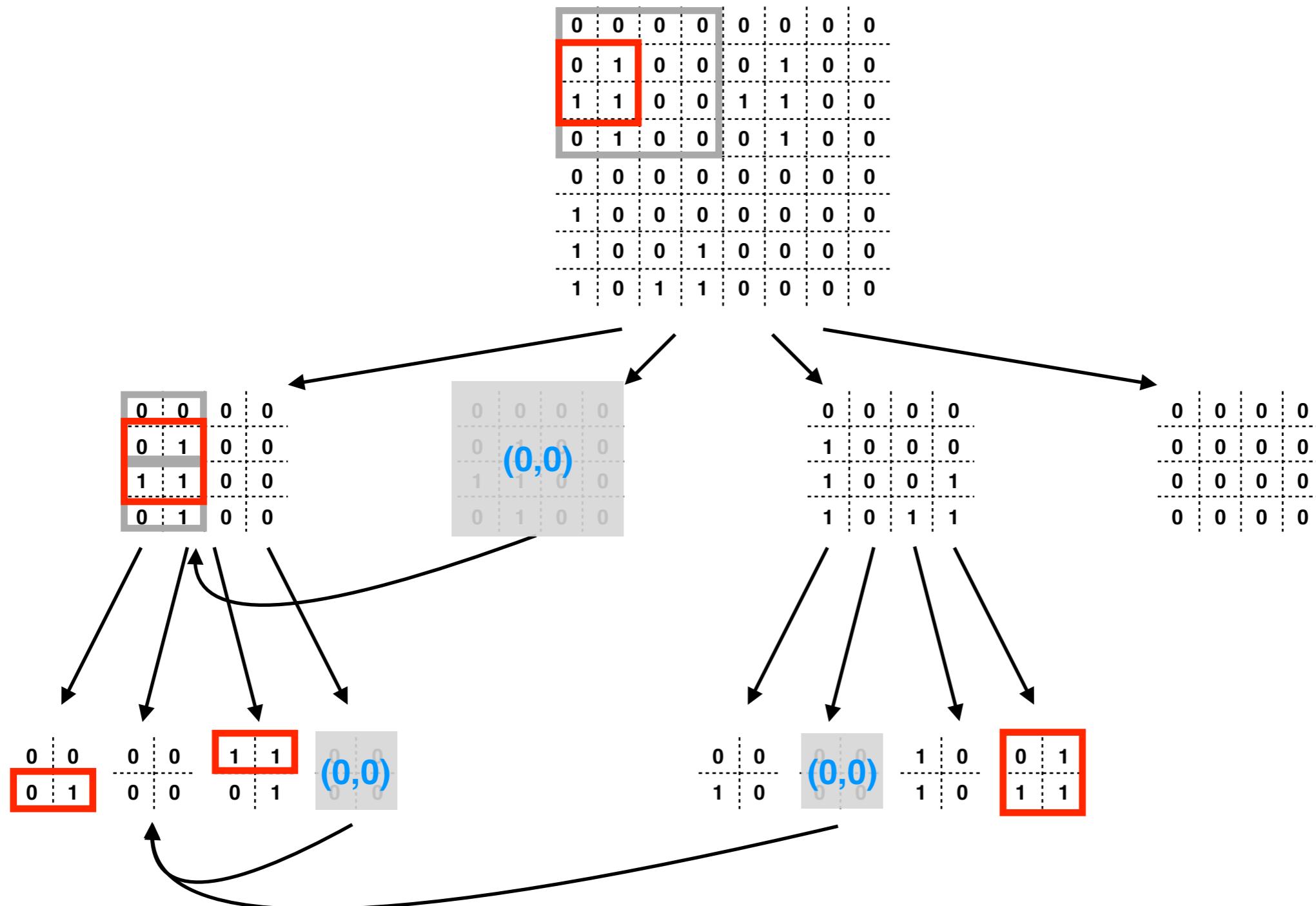
2D Block-trees



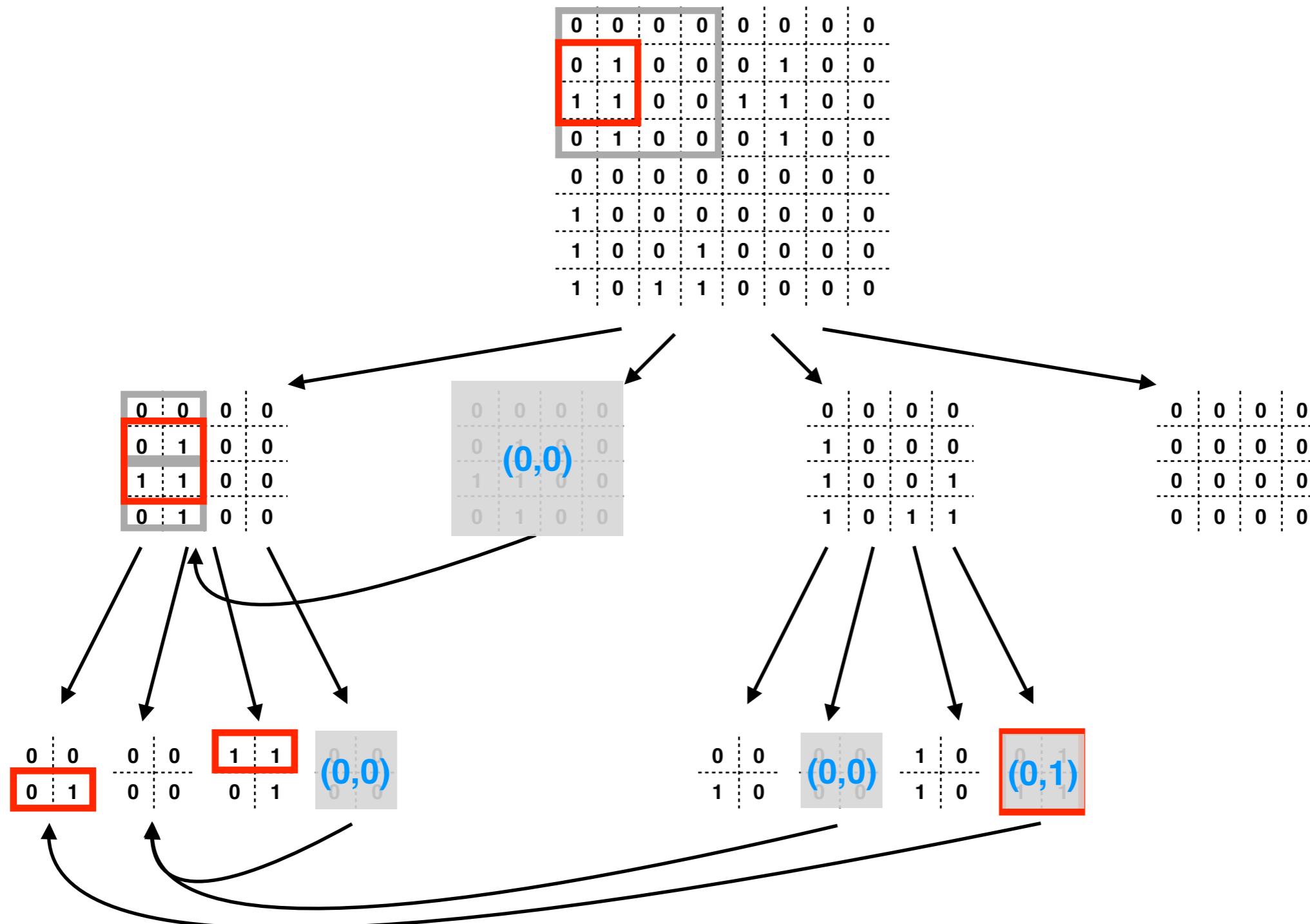
2D Block-trees



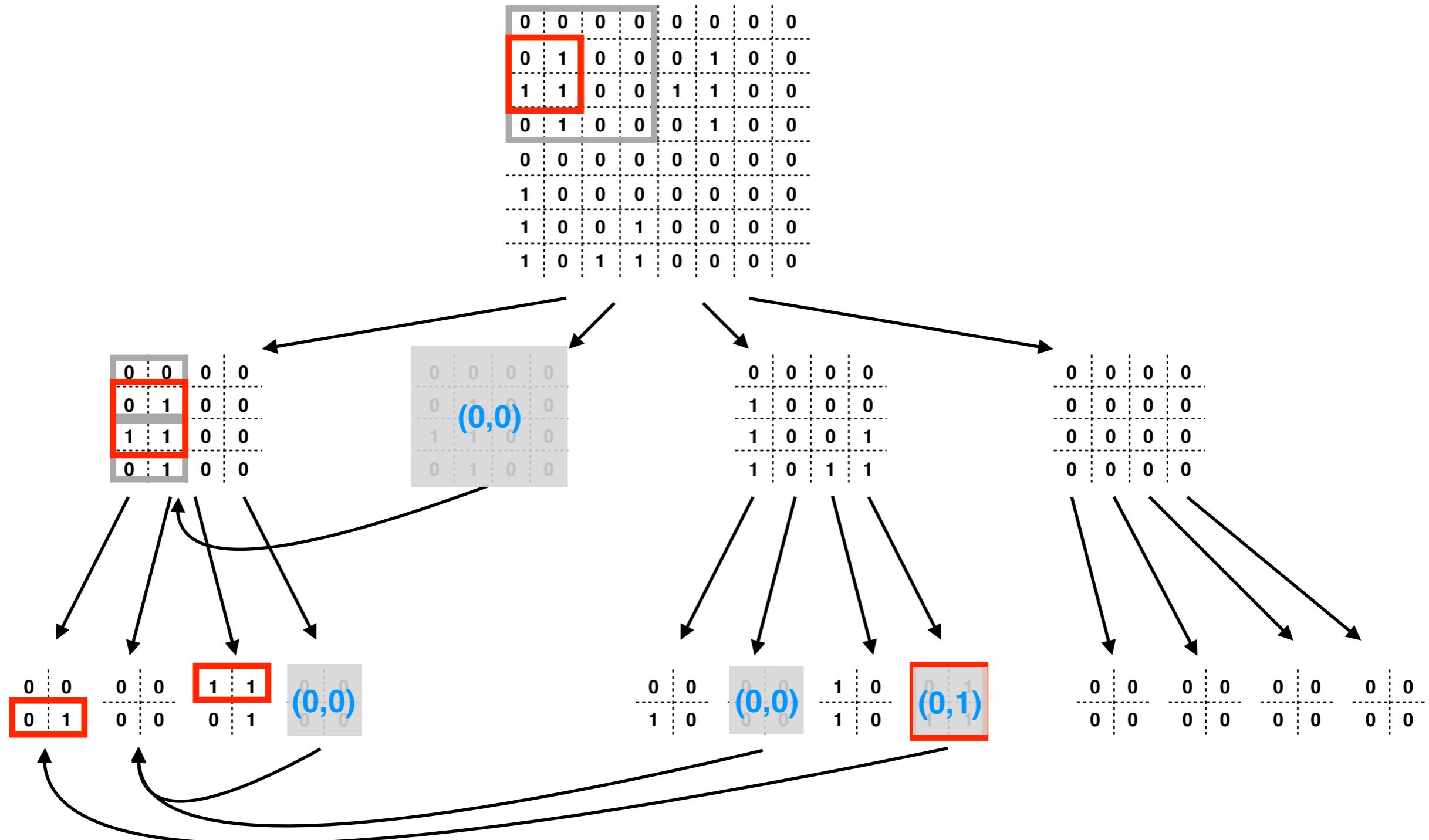
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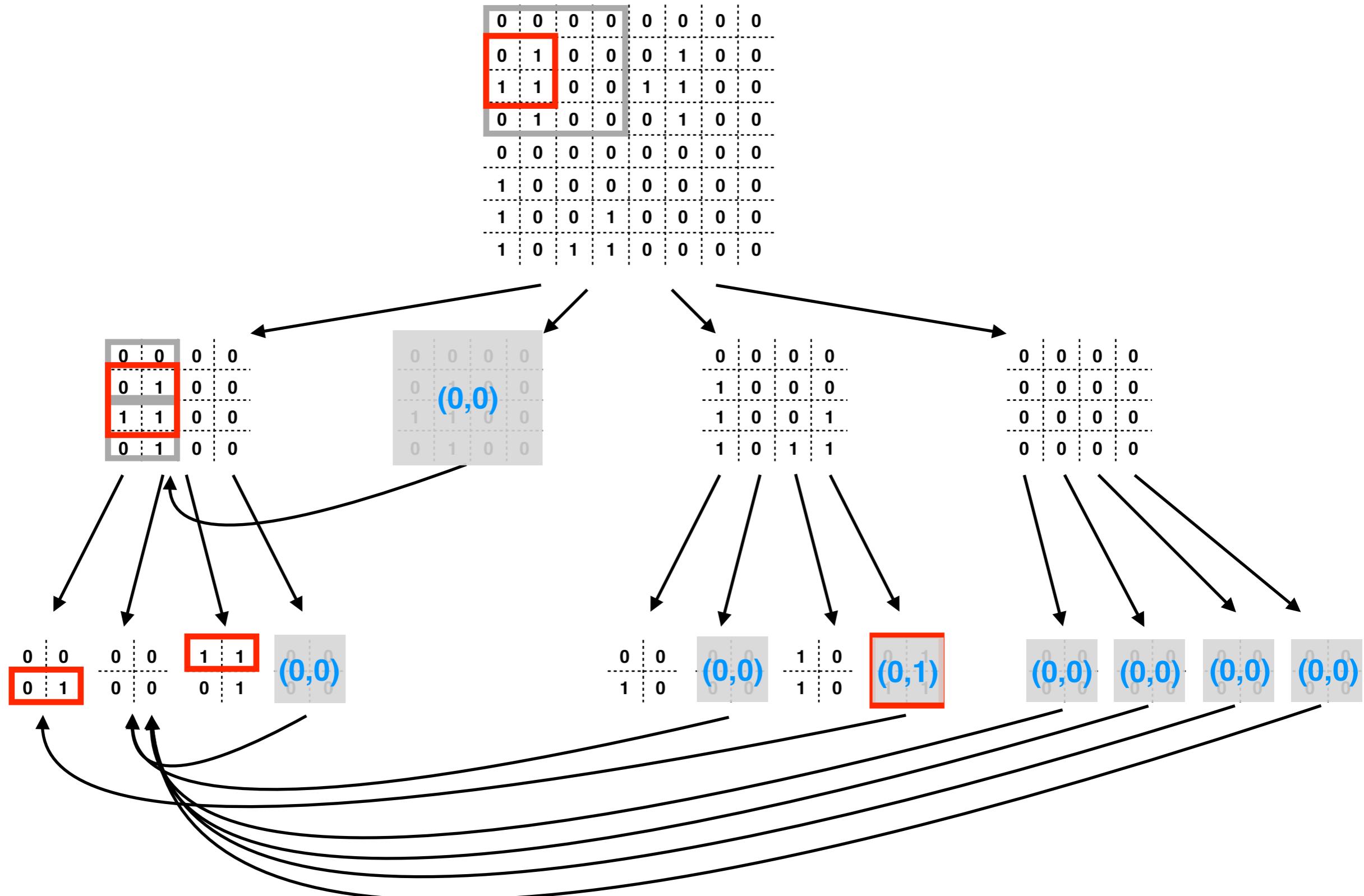
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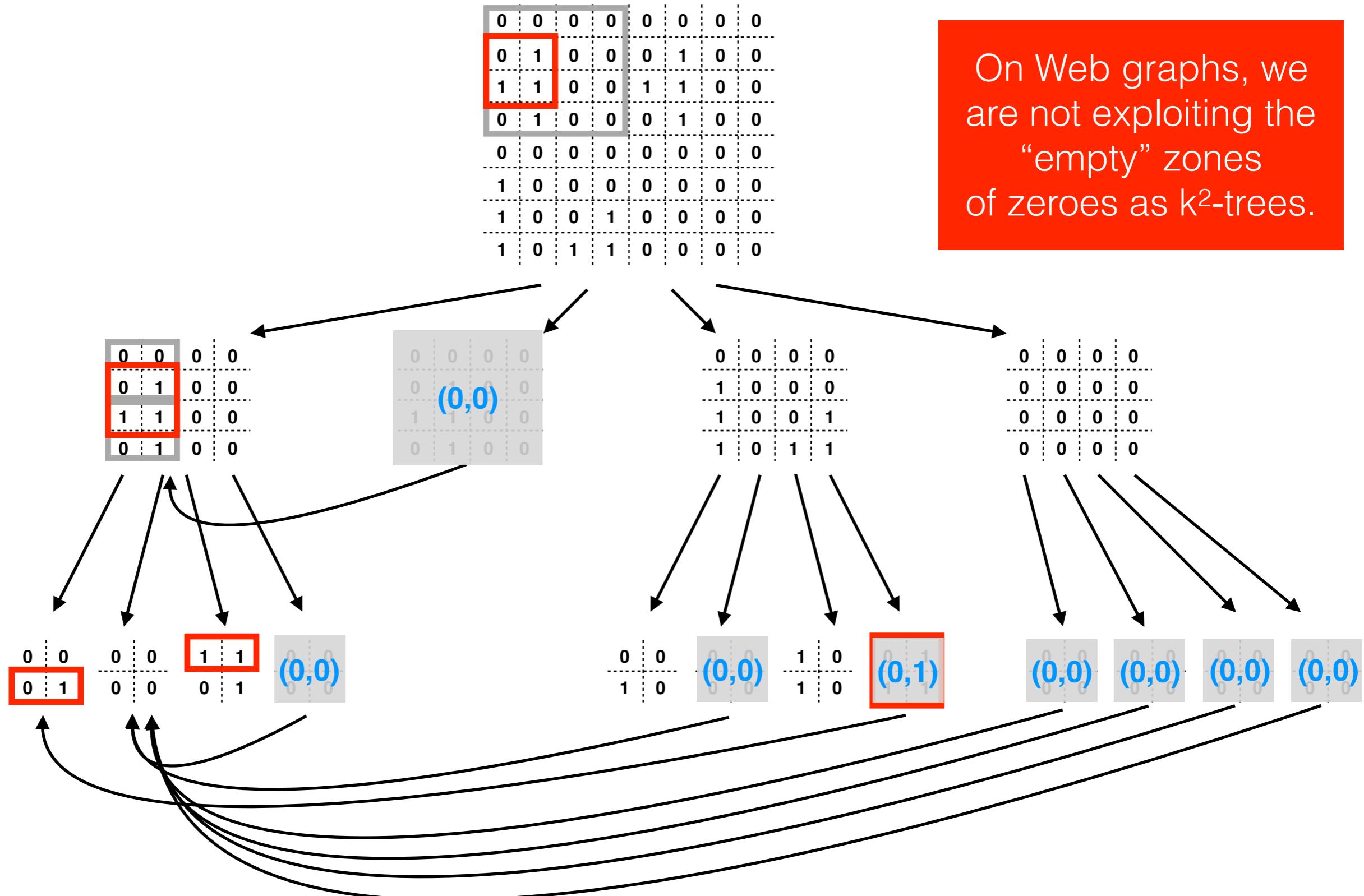
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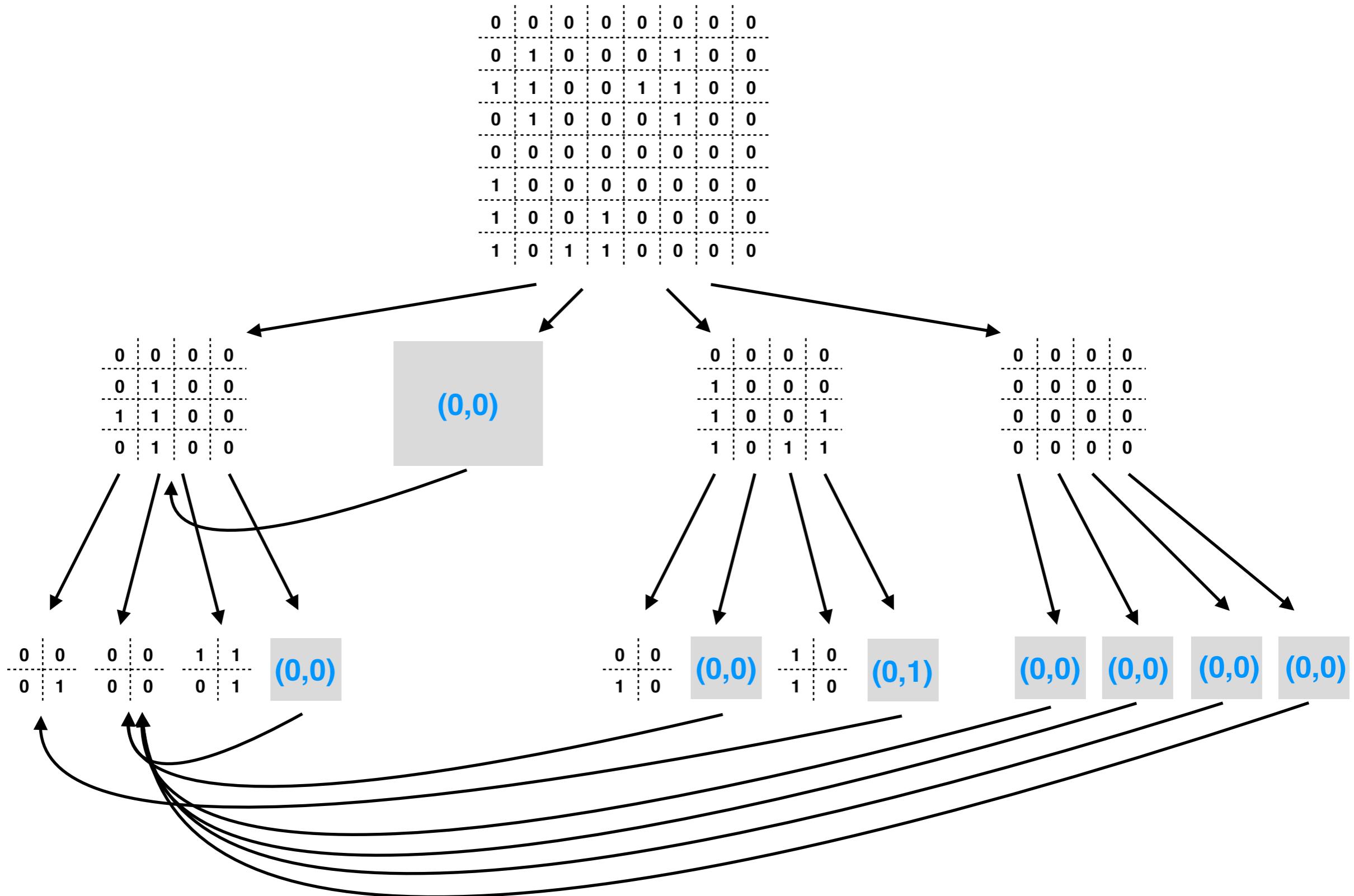
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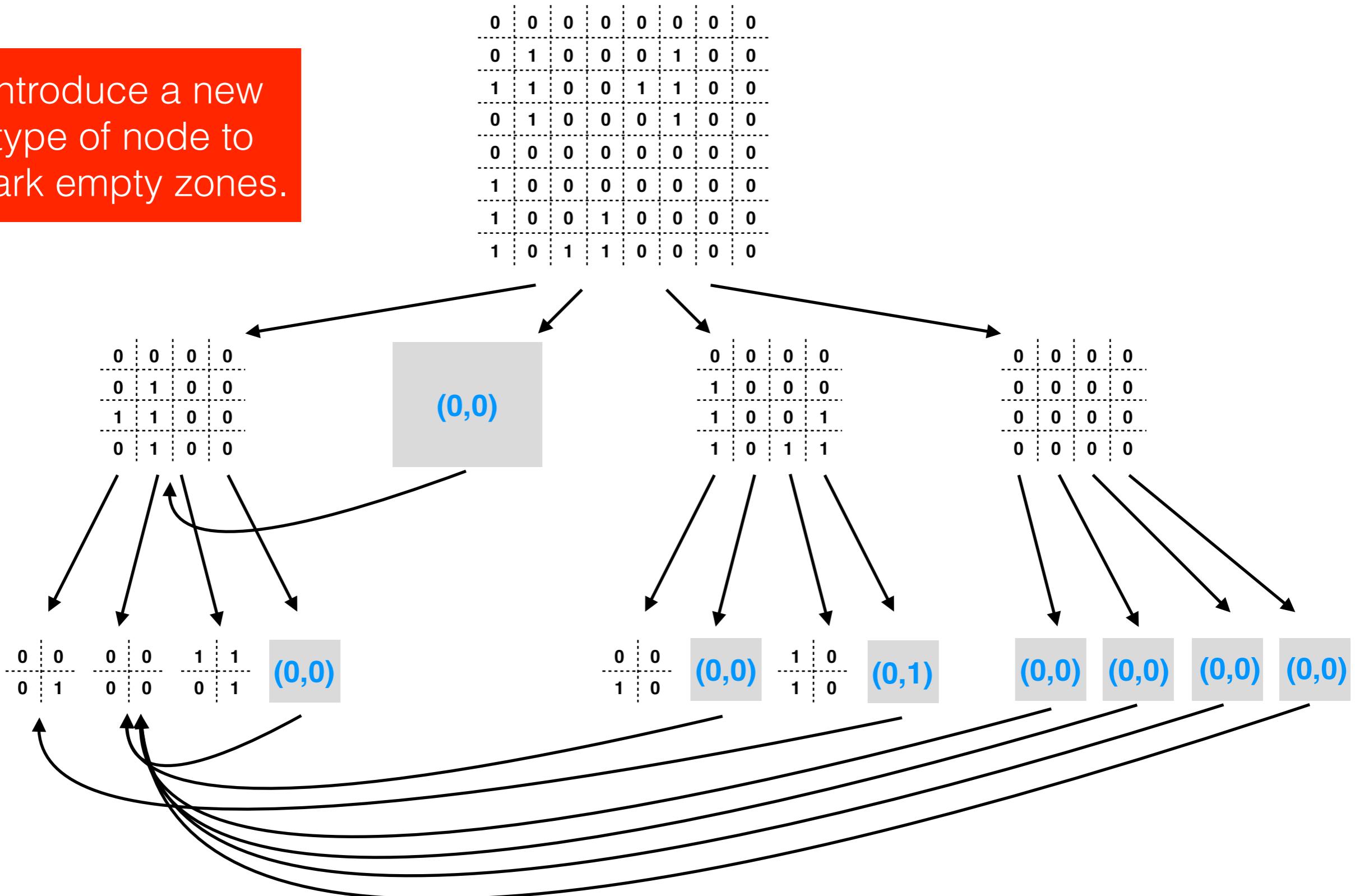


2D Block-trees on Web graphs

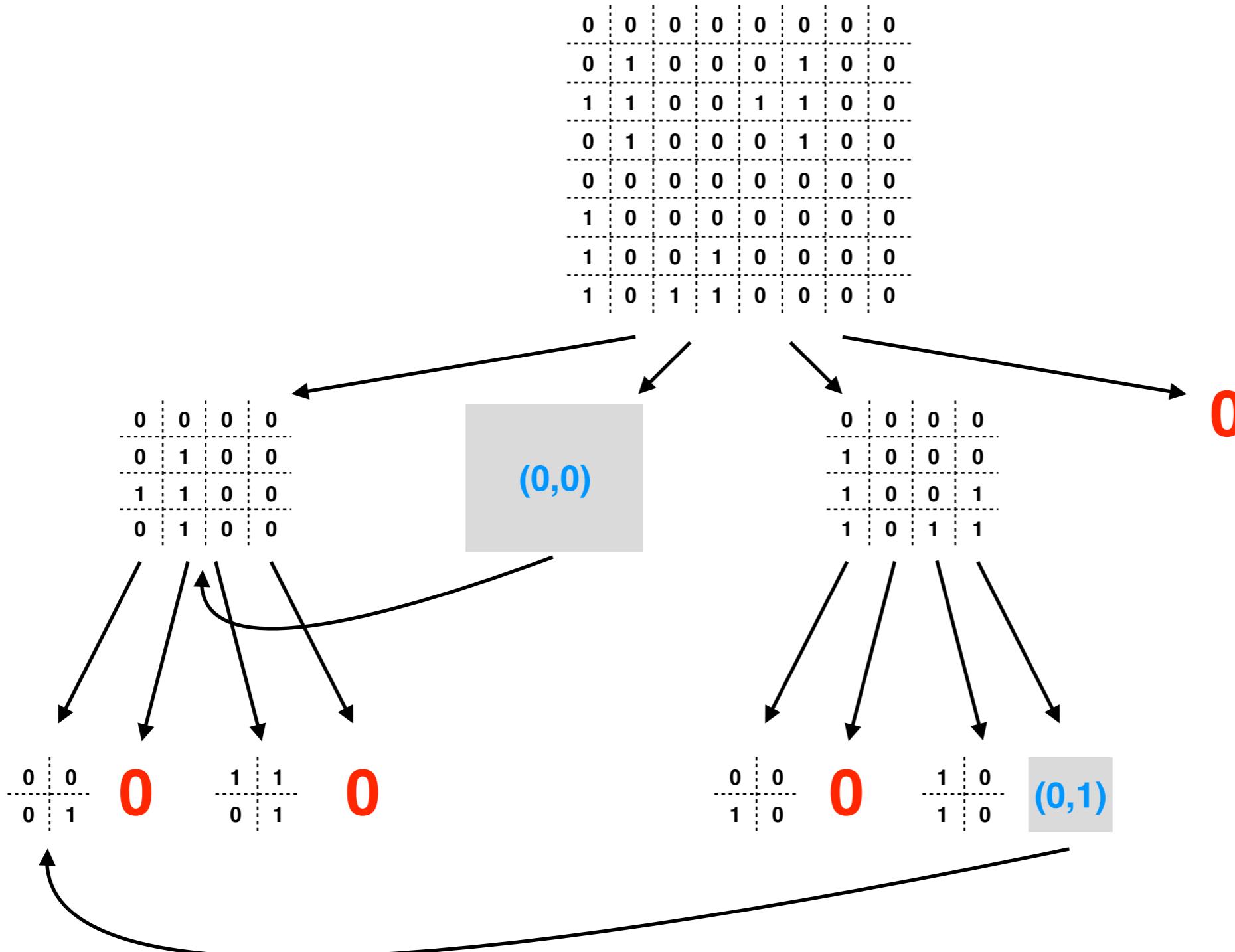


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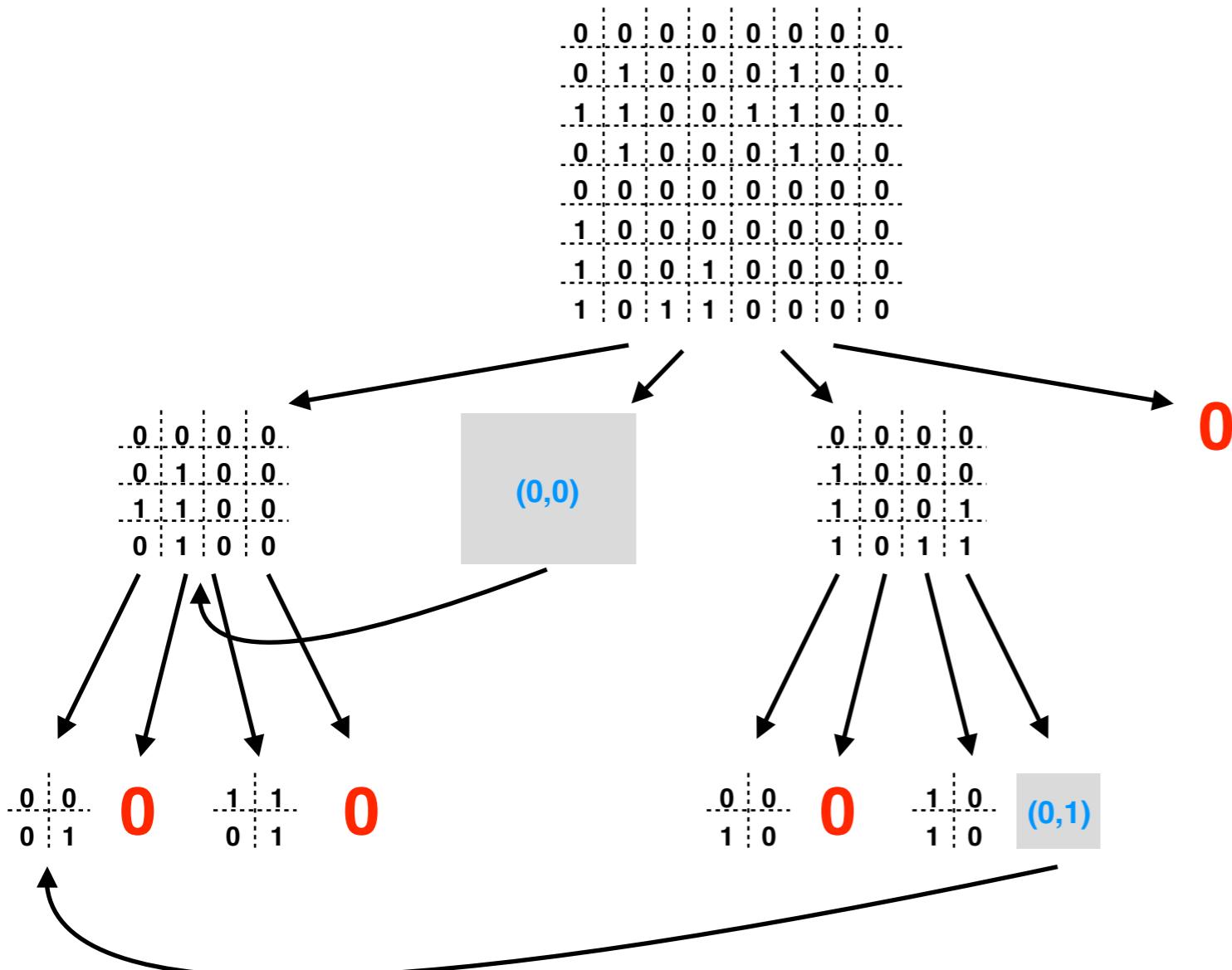
Introduce a new type of node to mark empty zones.



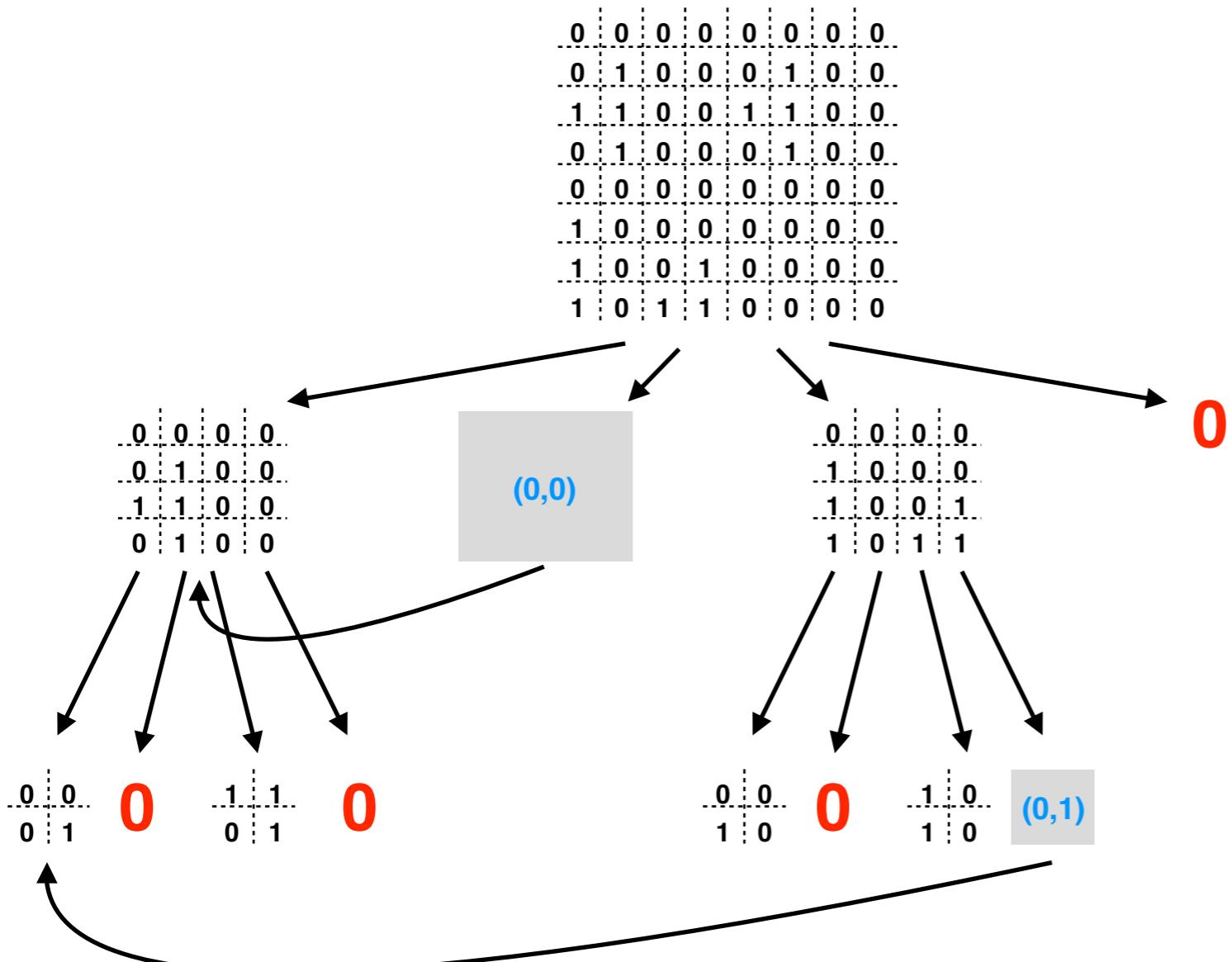
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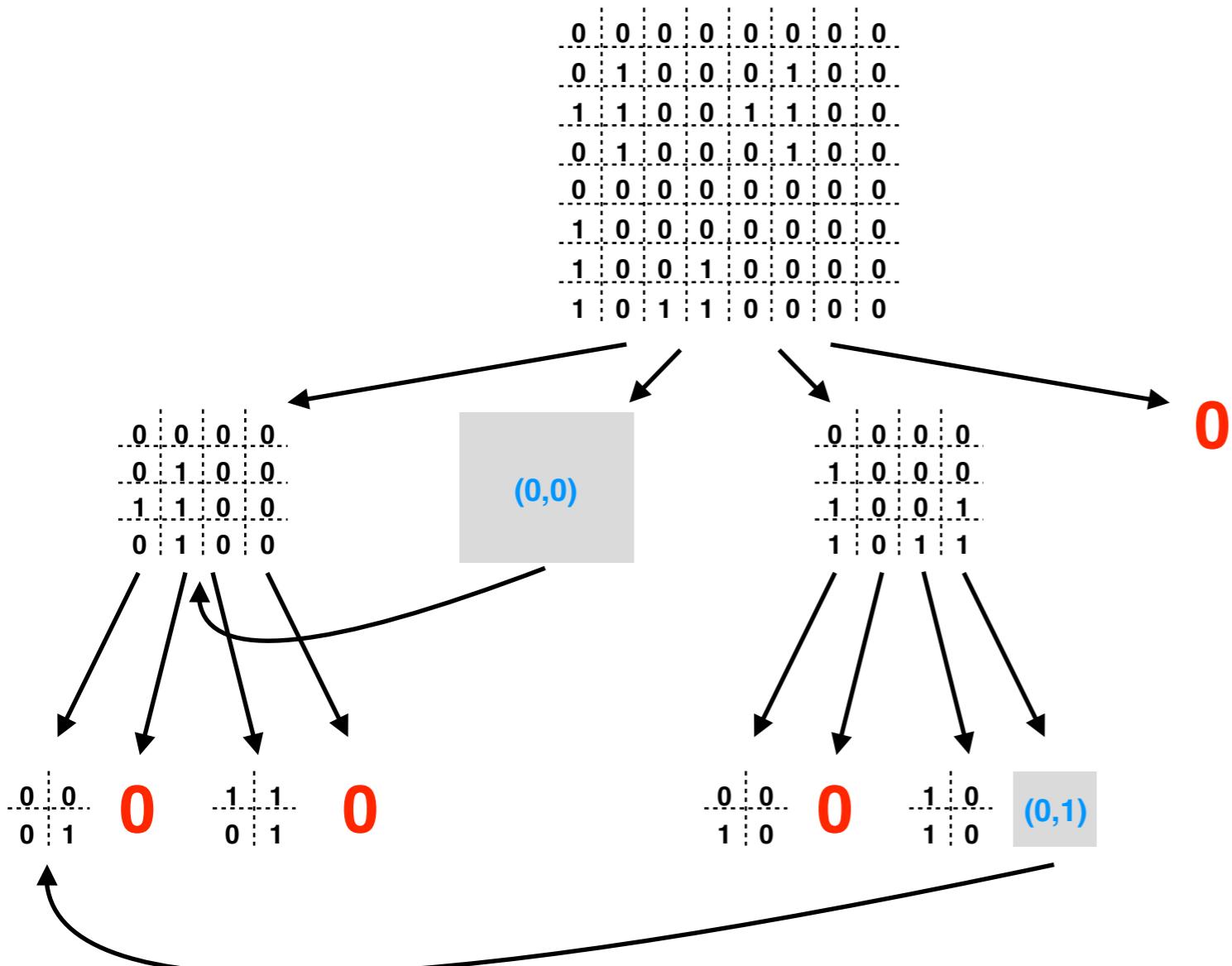
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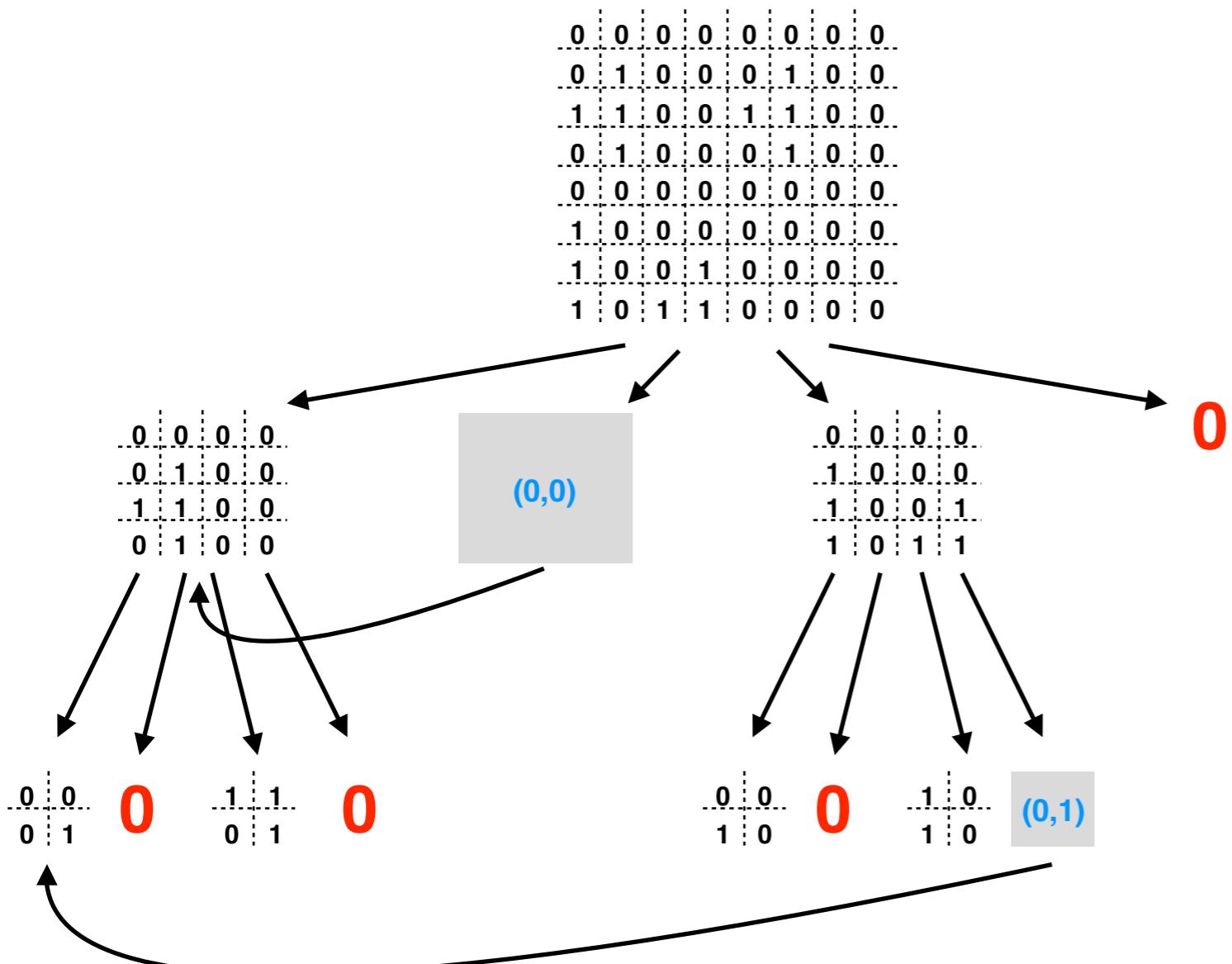


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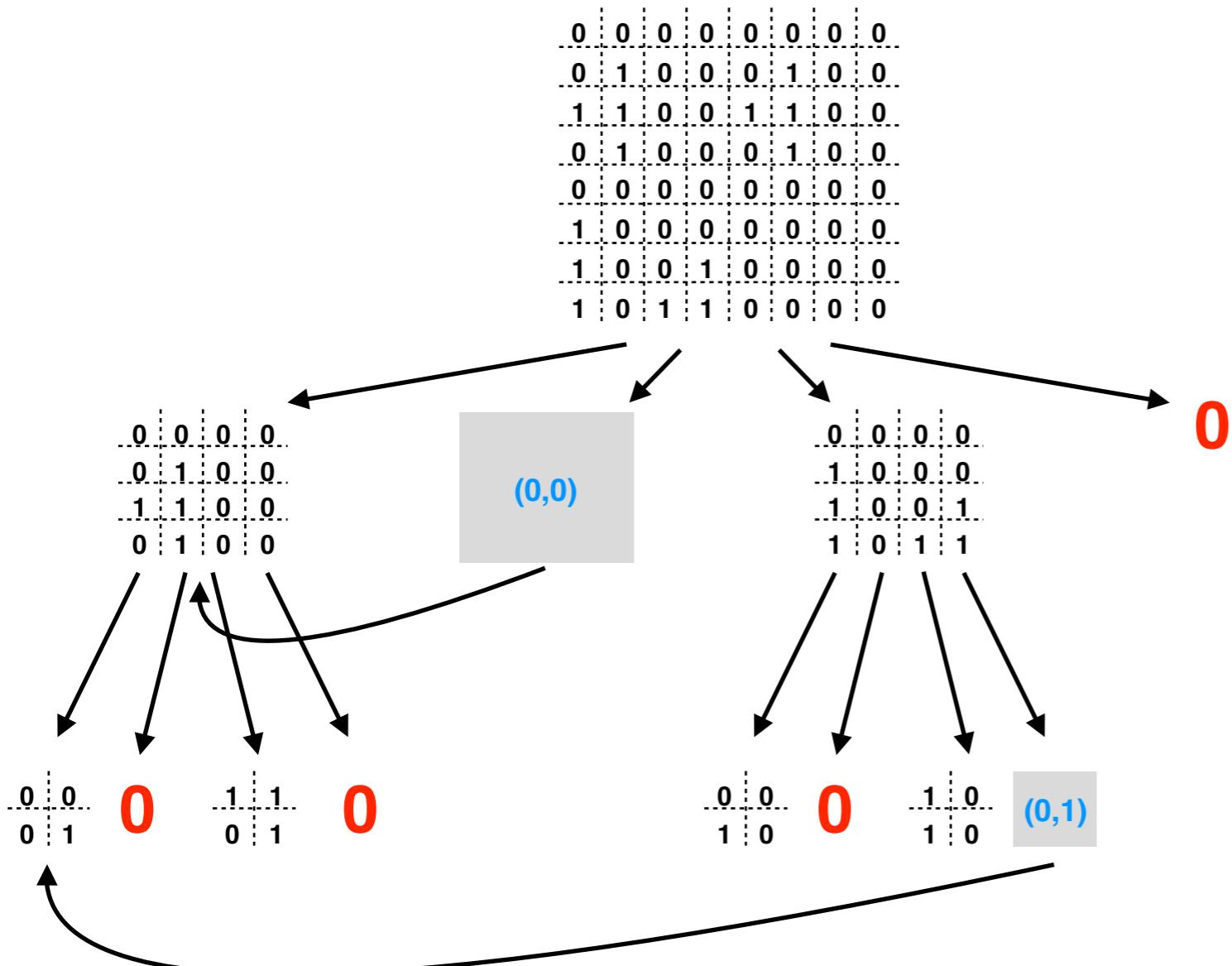
T 1010 1010 1010
L 0001 1101 0010 1010
N 100001

2D Block-trees on Web graphs



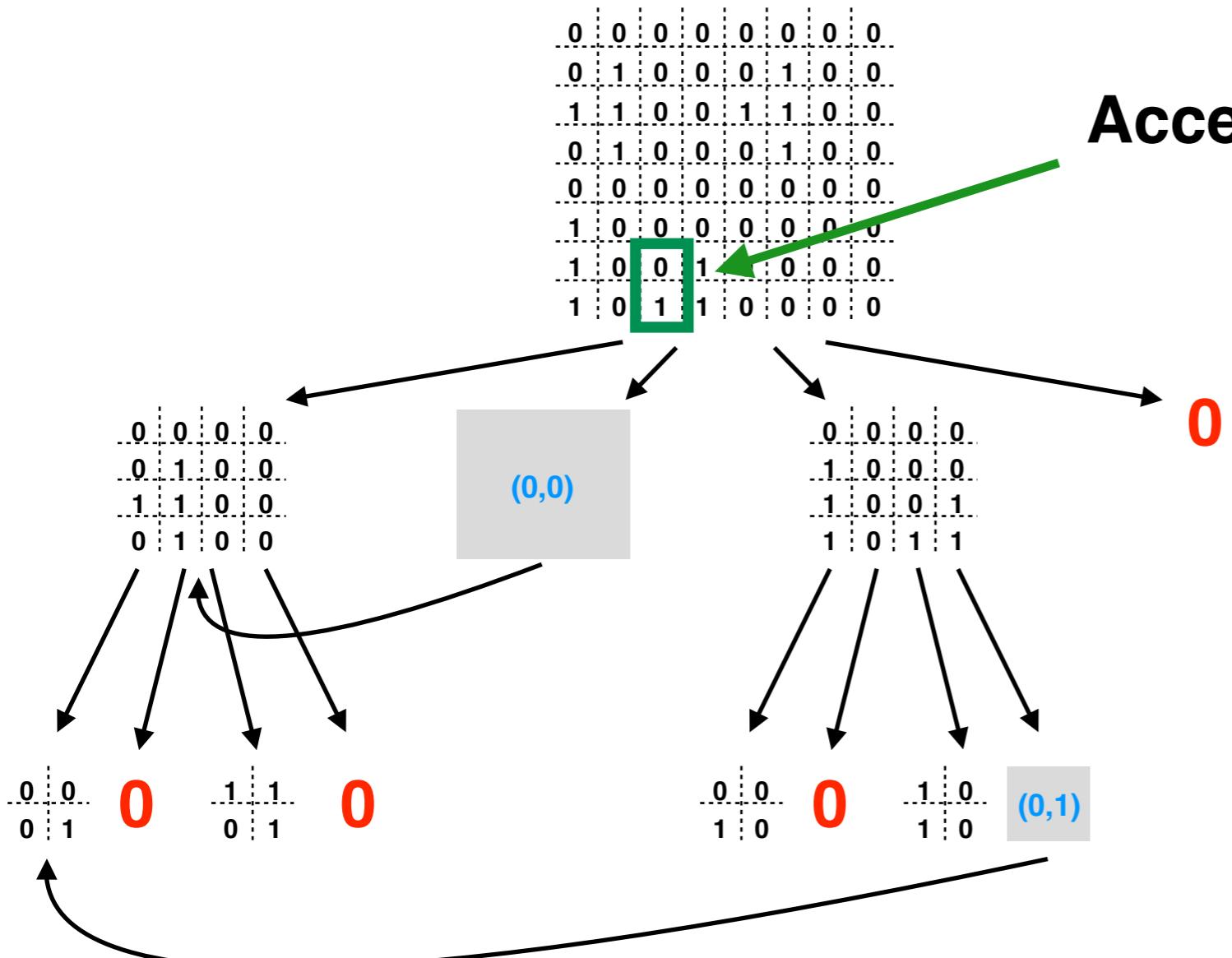
T 1010 1010 1010
L 0001 1101 0010 1010
N 100001
P₁ 1 **O₁** 00

2D Block-trees on Web graphs



T 1010 1010 1010
L 0001 1101 0010 1010
N 100001
P₁ 1 **O₁** 00
P₂ 7 **O₂** 01

2D Block-trees on Web graphs



Access

T 1010 1010 1010

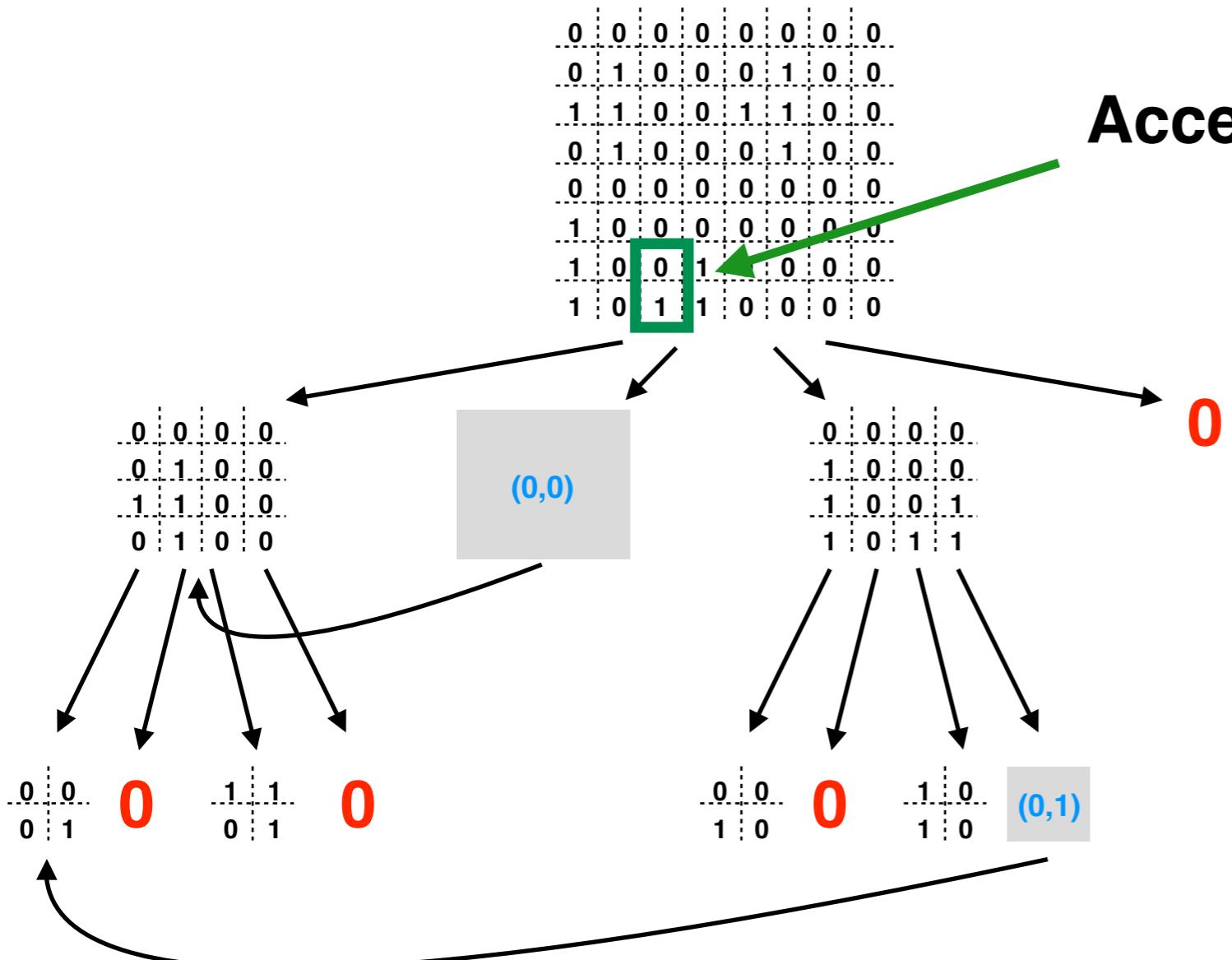
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N 100001

P₁ 1 O₁ 00

P₂ 7 O₂ 01

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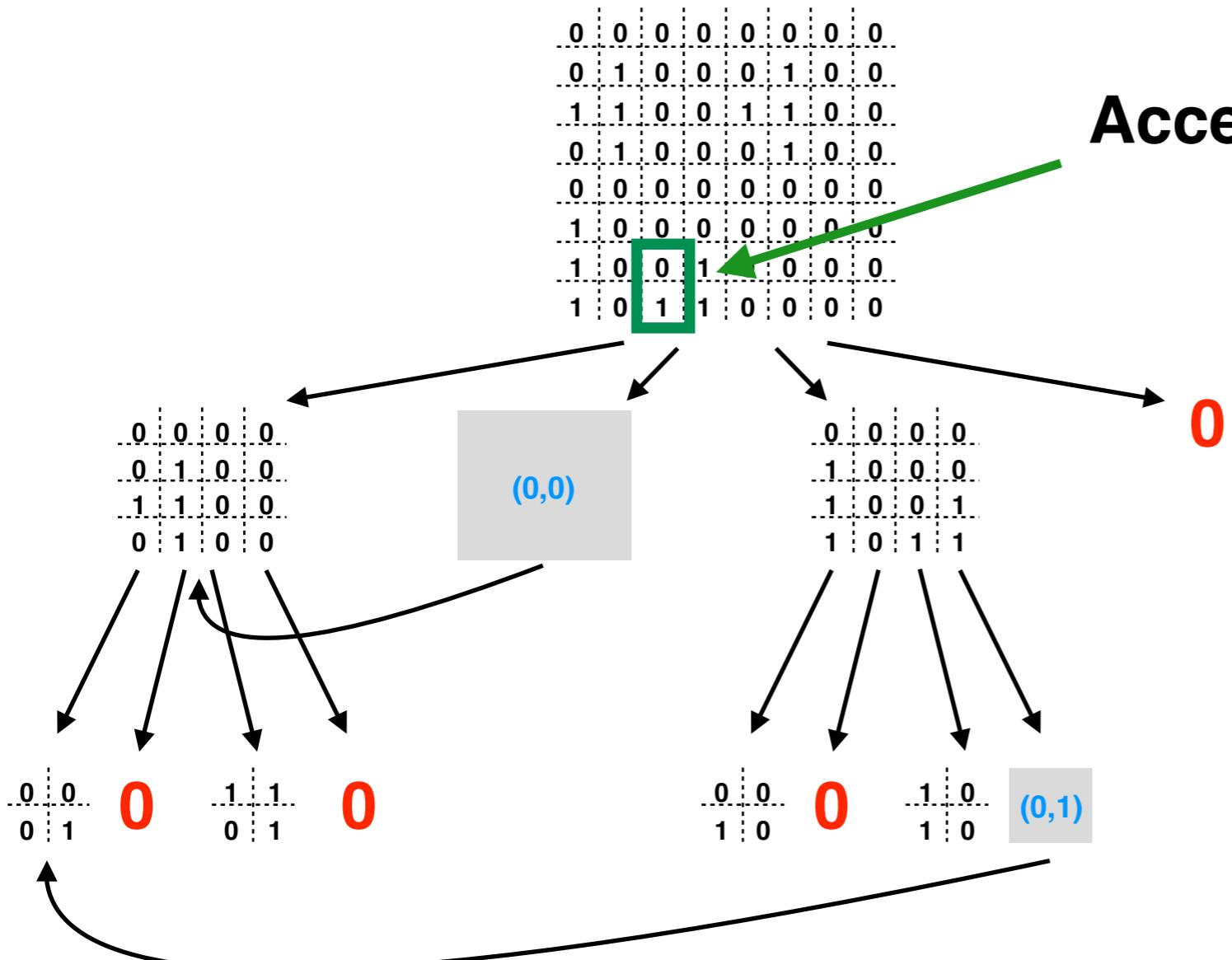
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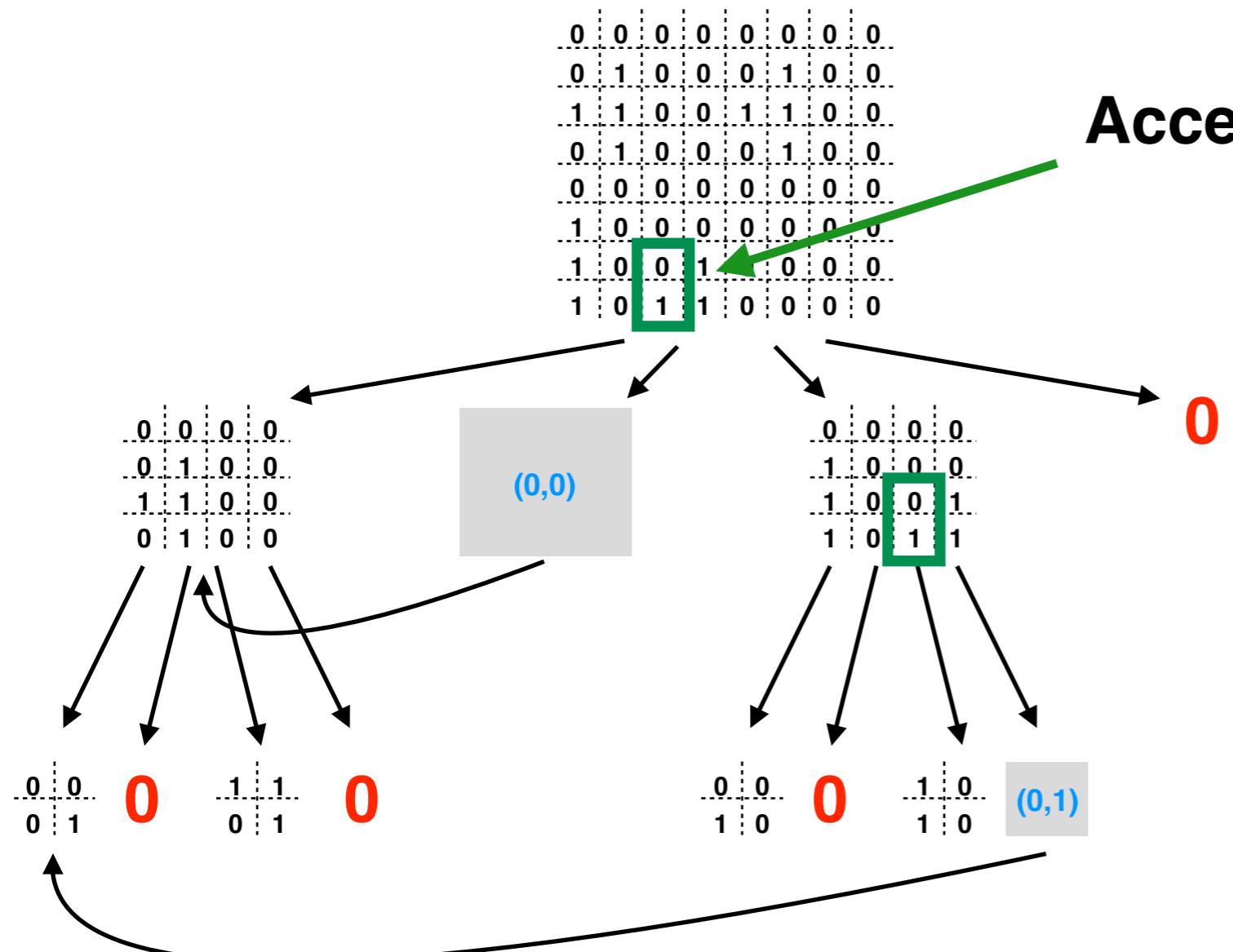
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P₂ 7 O₂ 01

2D Block-trees on Web graphs

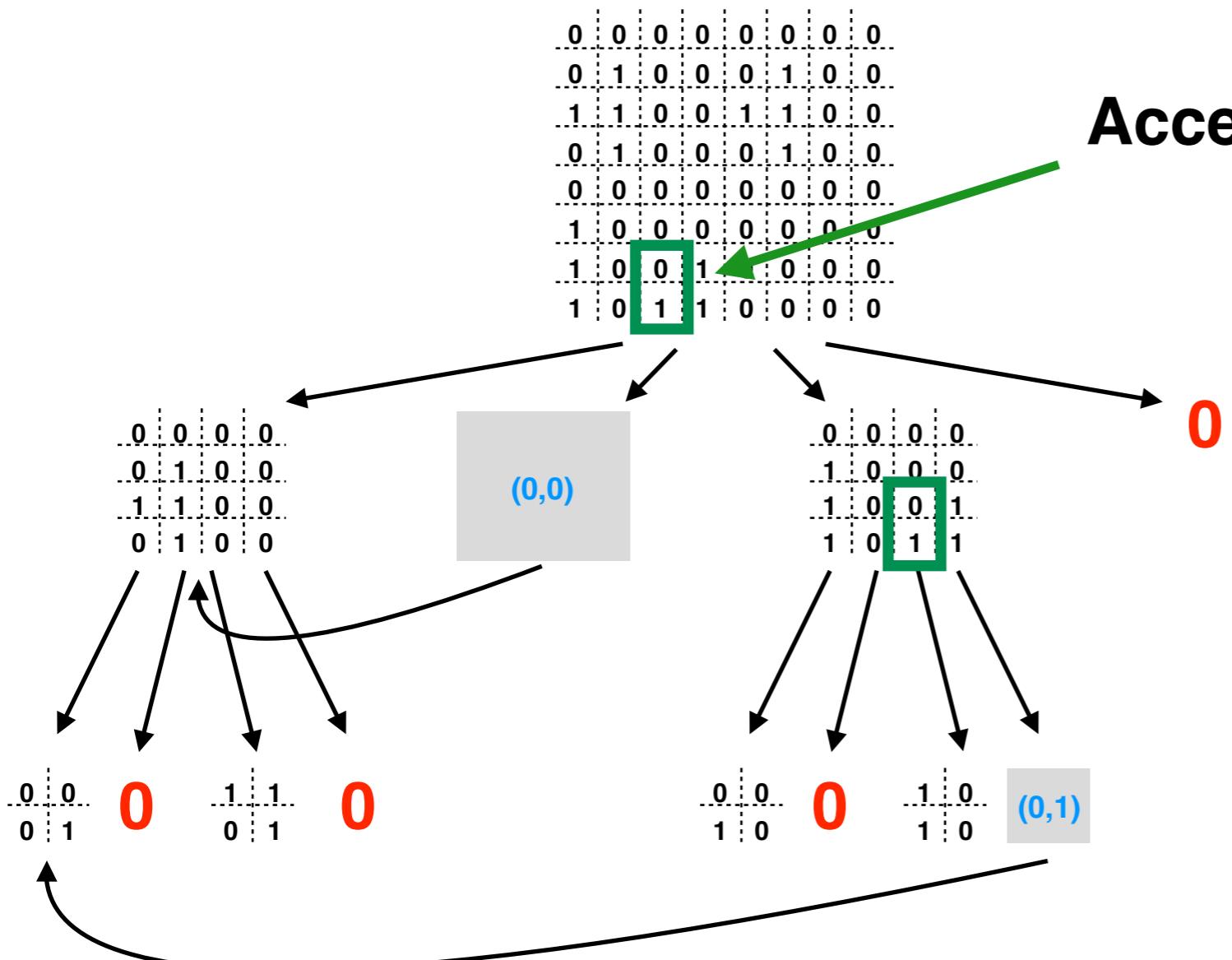


2D Block-trees on Web graphs



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2D Block-trees on Web graphs



T 1010 1010 1010

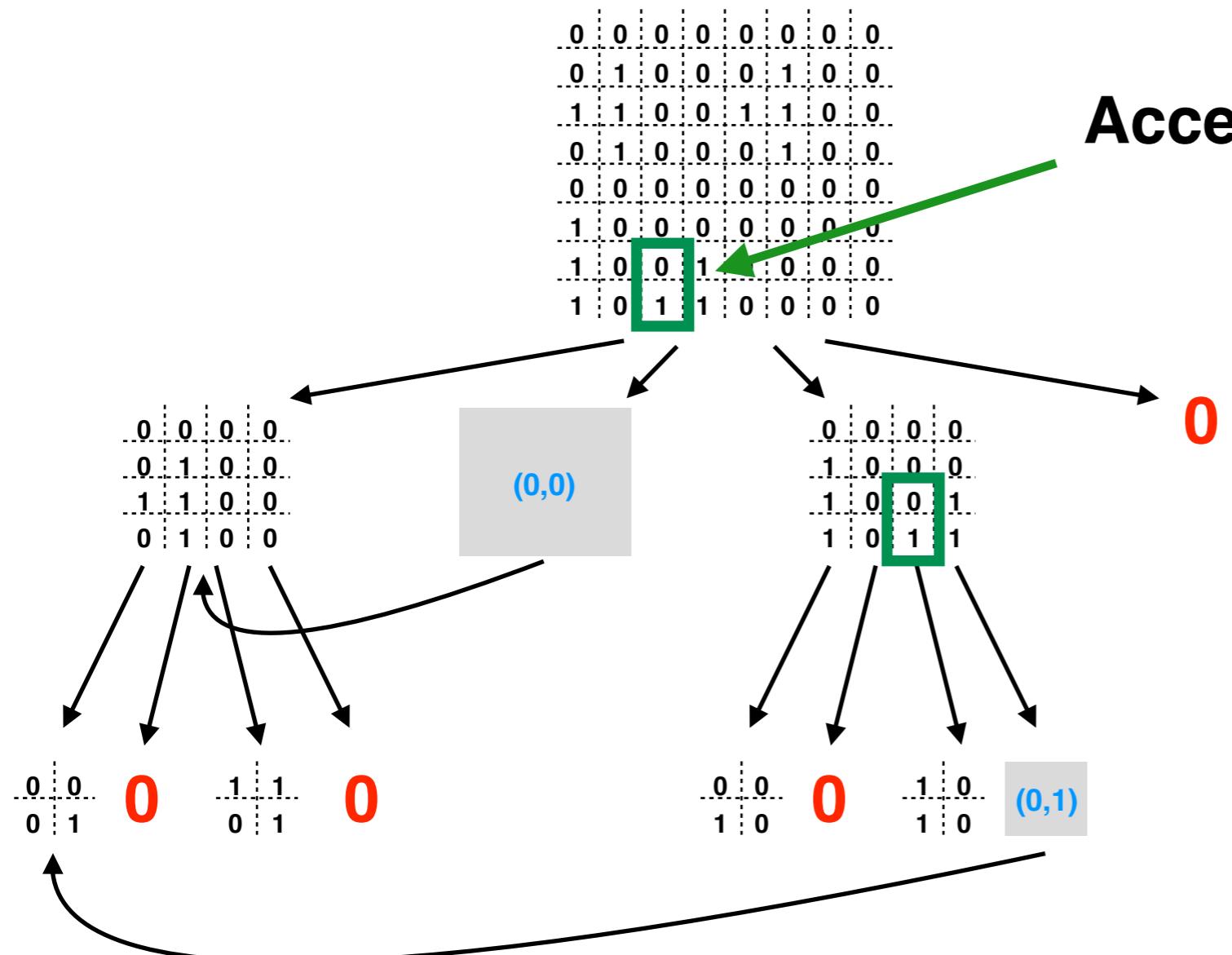
L 0001 1101 0010 1010

N 100001

P₁ 1 **O₁** 00

P₂ 7 **O₂** 01

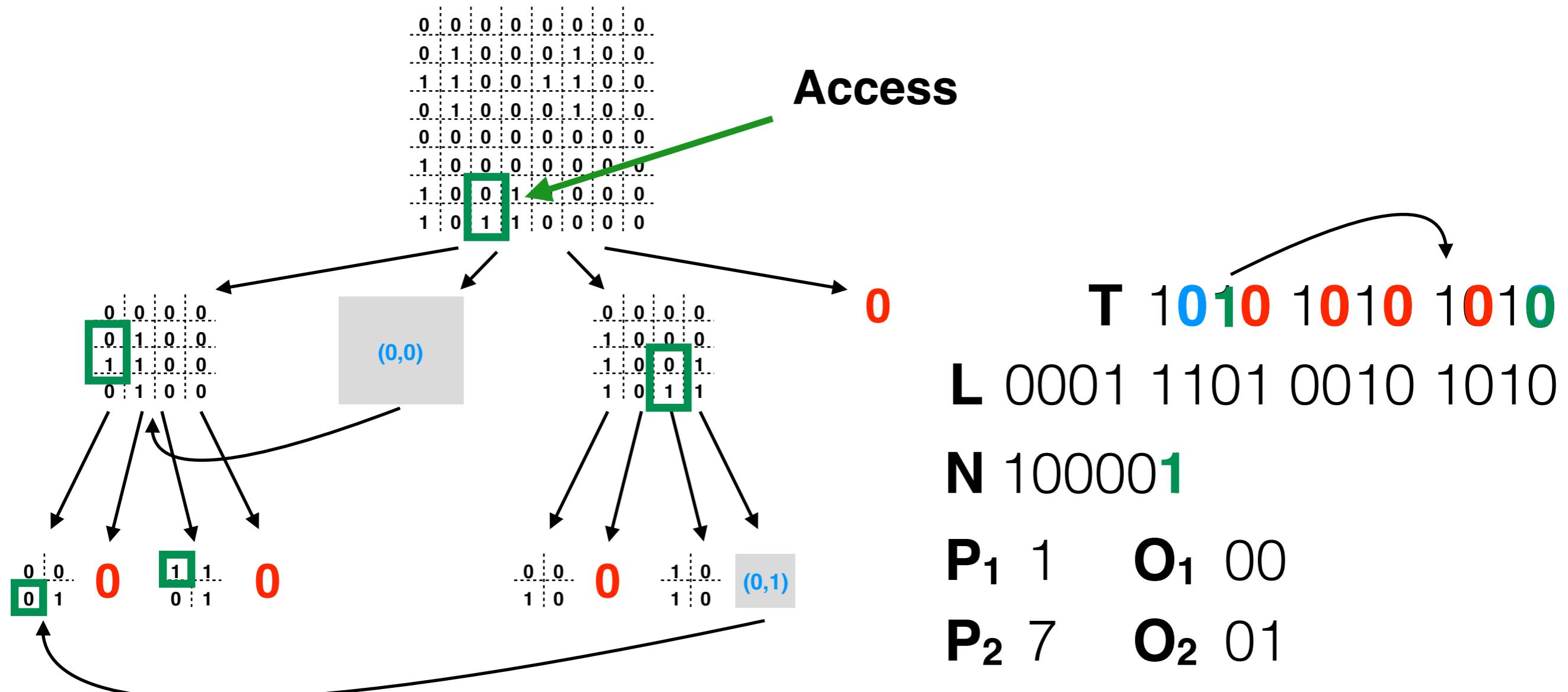
2D Block-trees on Web graphs



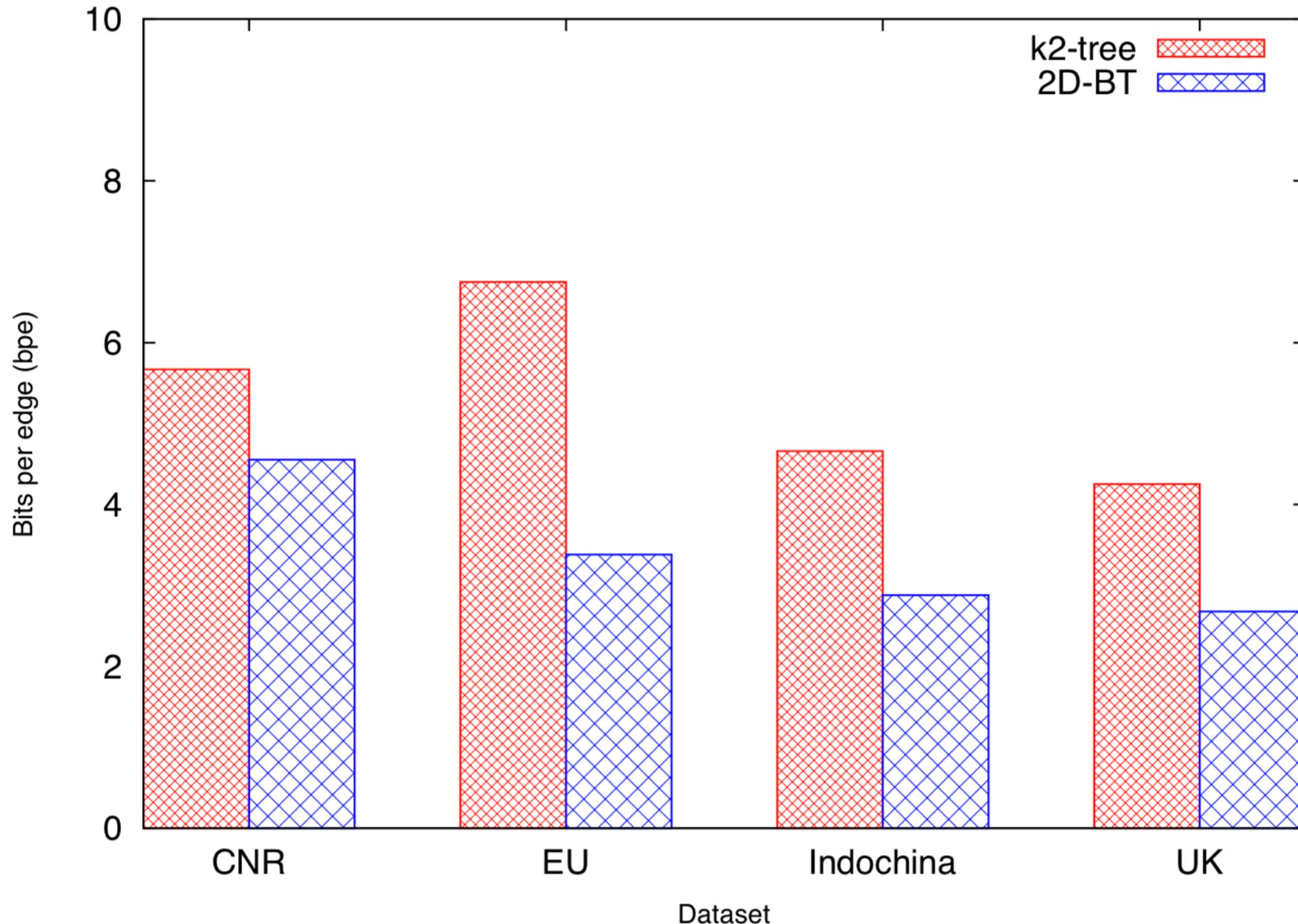
T 1010 1010 1010
L 0001 1101 0010 1010
N 100001
P₁ 1 **O₁** 00
P₂ 7 **O₂** 01



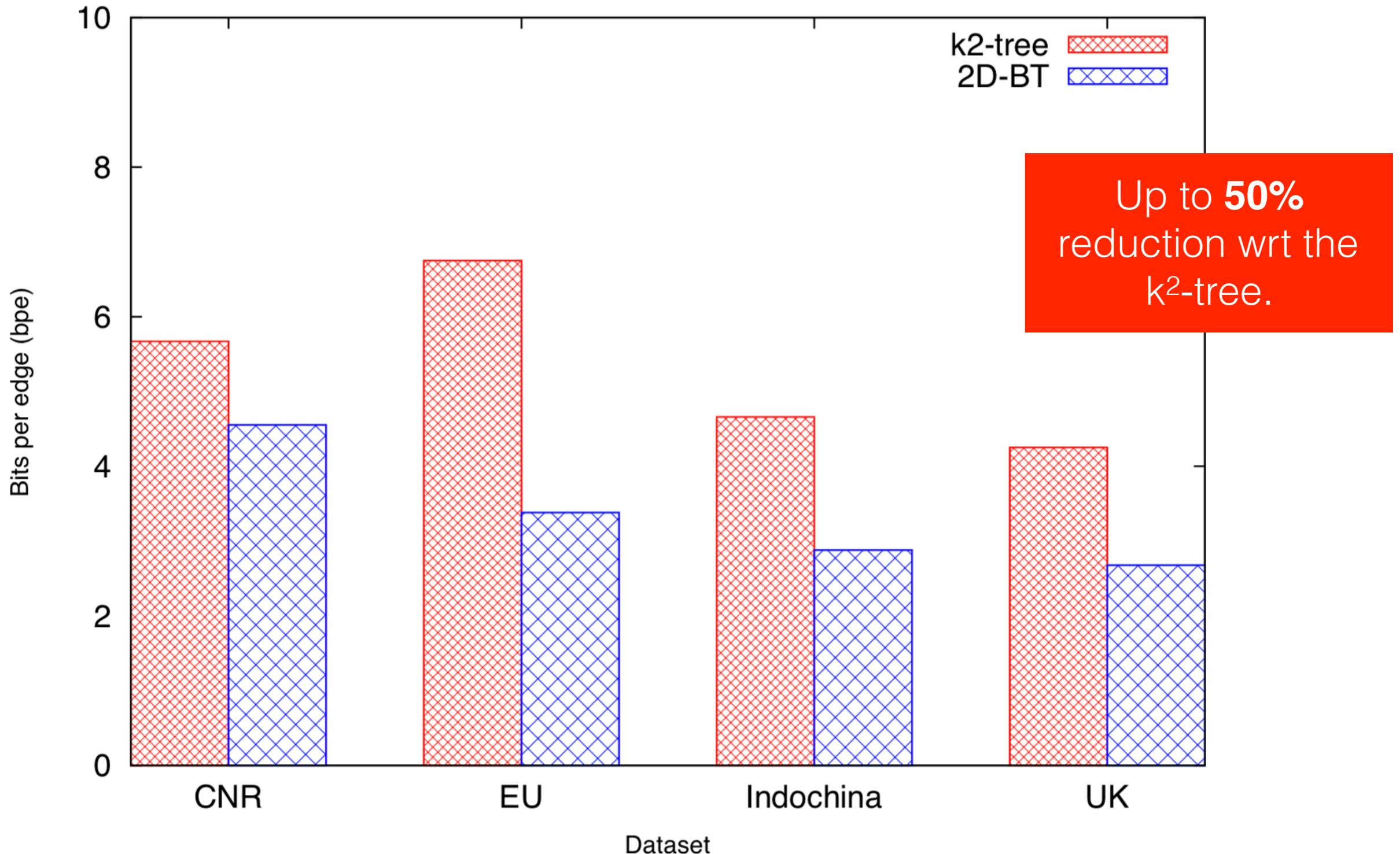
2D Block-trees on Web graphs



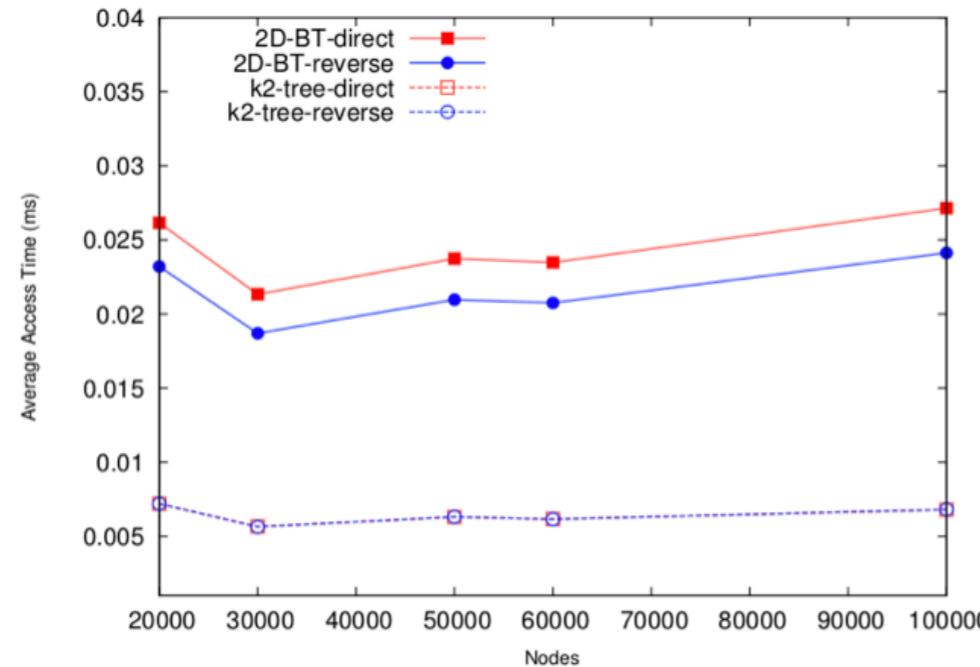
2D Block-trees on Web graphs



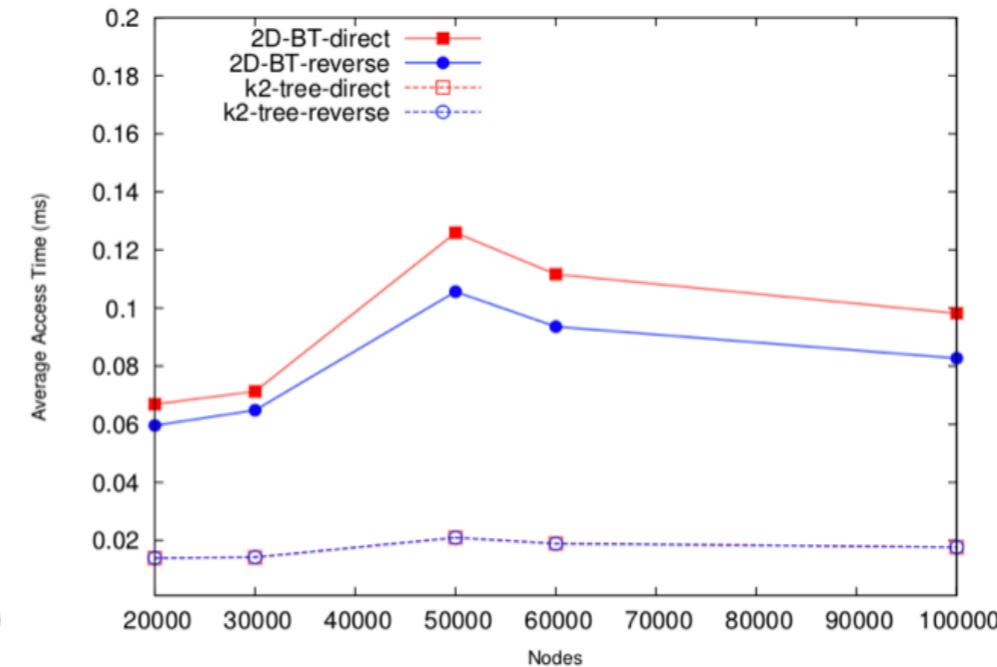
2D Block-trees on Web graphs



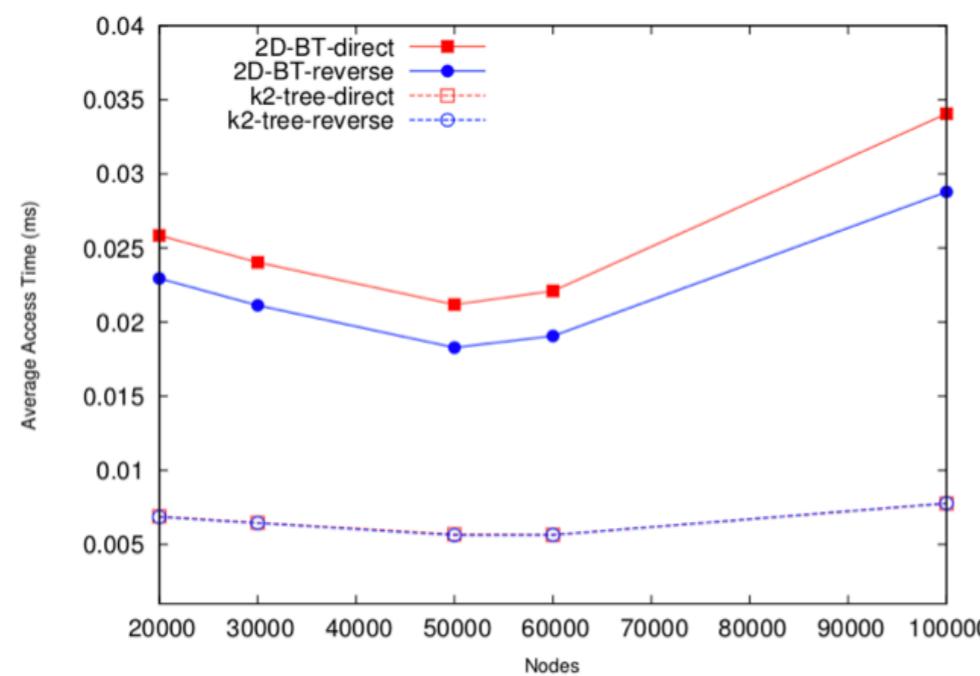
2D Block-trees on Web graphs



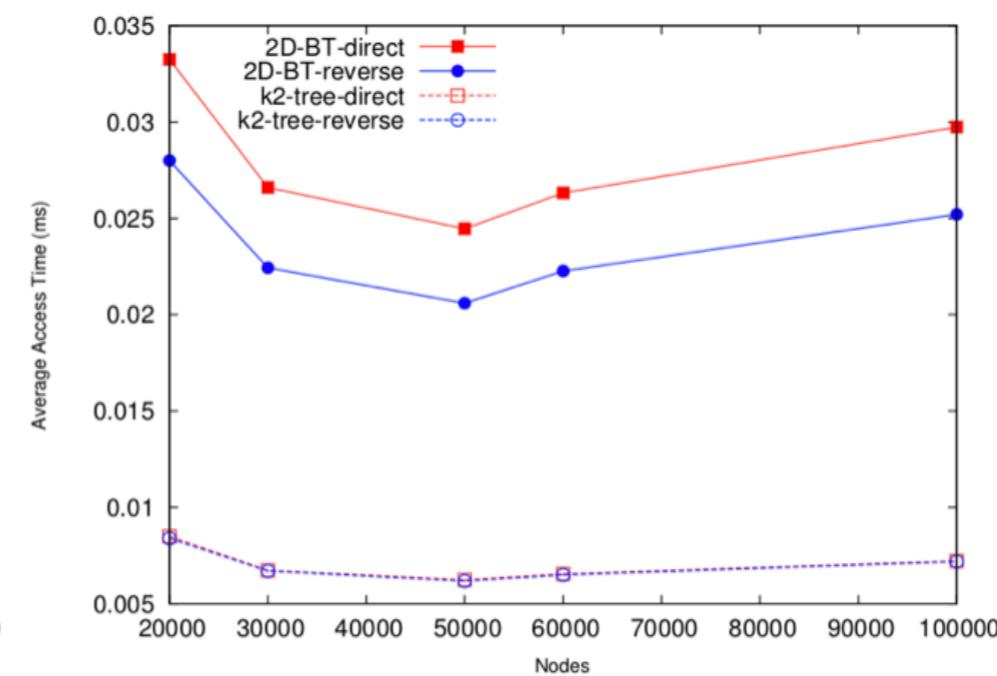
(a) Dataset CNR



(b) Dataset EU

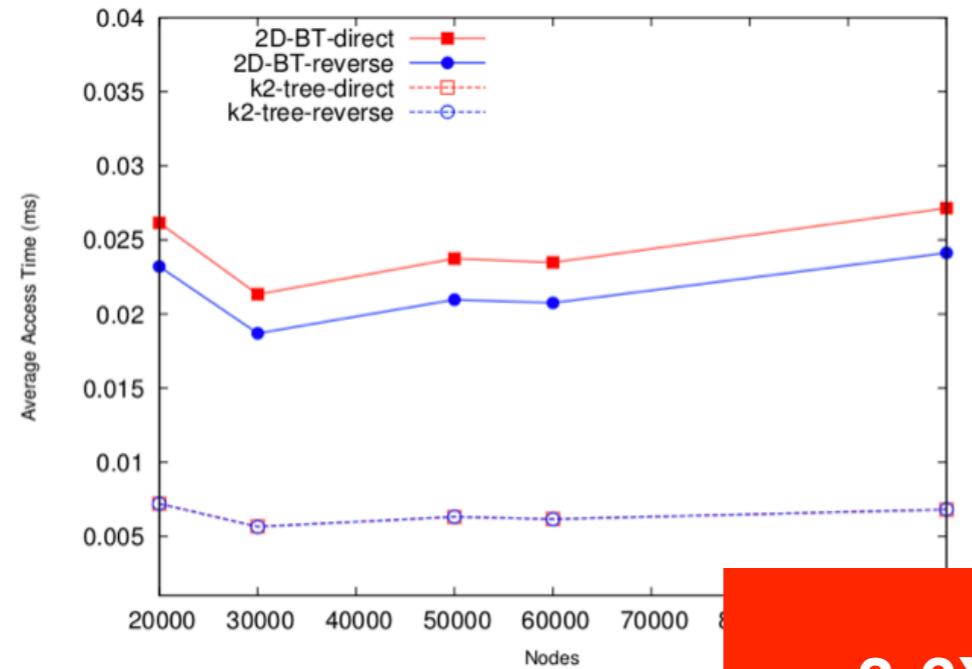


(c) Dataset Indochina

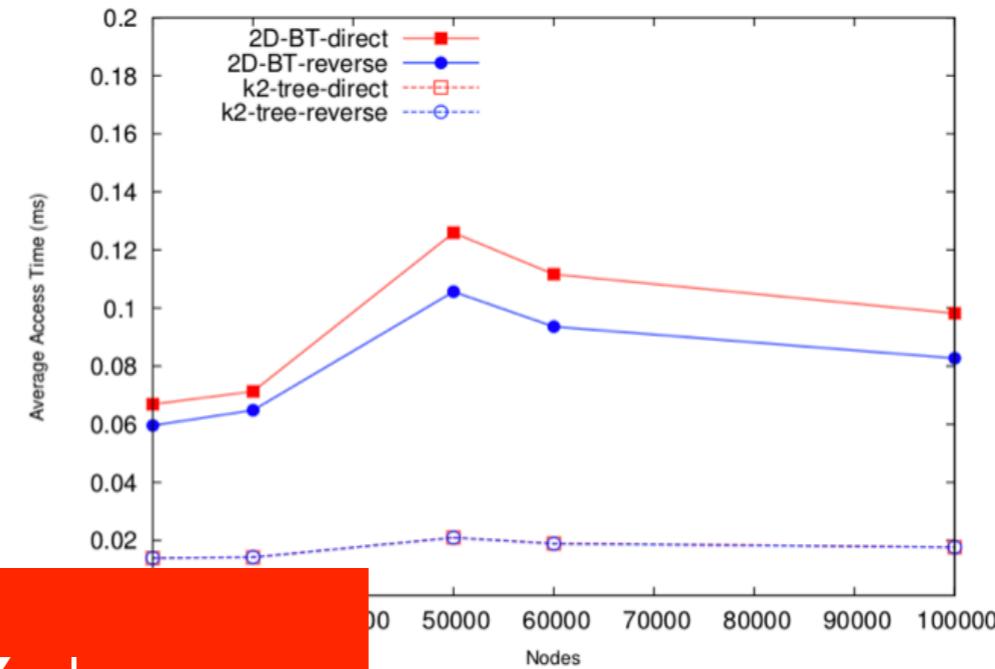


(d) Dataset UK

2D Block-trees on Web graphs

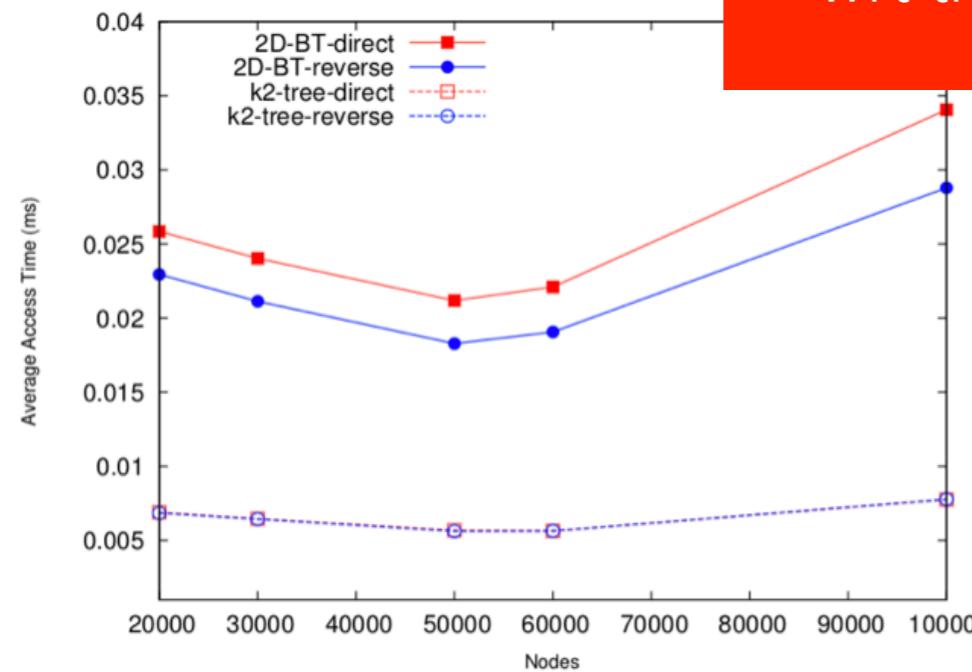


(a) Dataset CNR

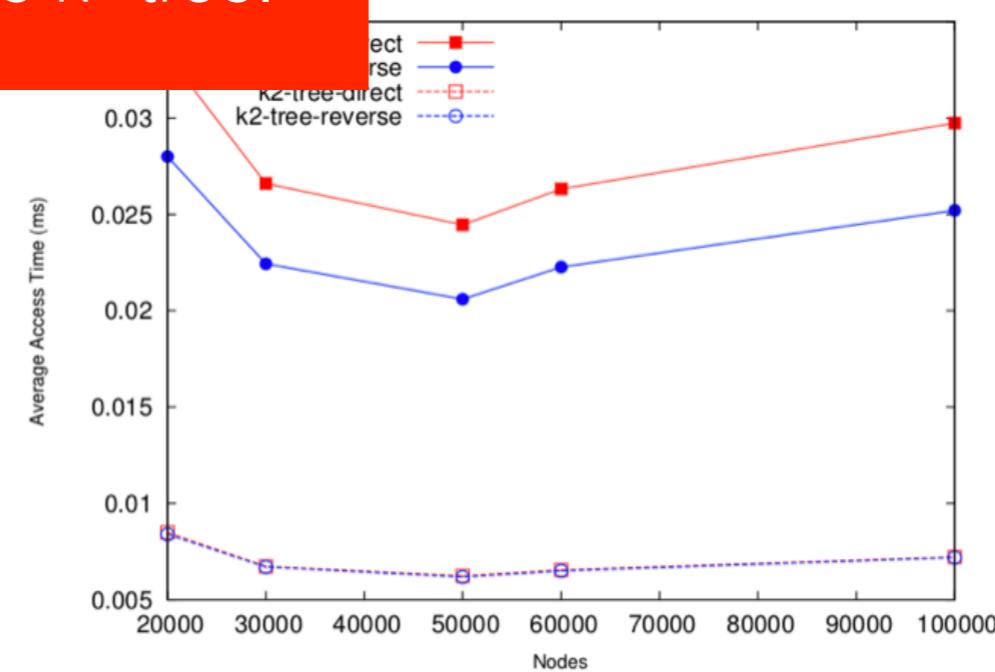


3-6X slower
wrt the k²-tree.

(b) Dataset EU



(c) Dataset Indochina



(d) Dataset UK

Summing up

WebGraph supports efficient extraction of direct neighbours in excellent compressed space.

k²-trees support direct and reverse navigation;
good trade-off between space and time;
do not exploit repetitiveness.

Block-trees compress (1D) strings to compression ratios close to LZ and support efficient random access to any substring.

2D-block-trees combines the capturing-sparseness behaviour of k²-trees with the capturing-repetitiveness of block-trees.
Up to 50% reduction in space, but 3-6X slower than k²-trees.

Take home messages

WebGraph is still the most compact representation
if only direct navigation is allowed.

k²-trees achieve a good trade-off between space and time *when* both direct and reverse navigation is needed.

2D-block-trees are even smaller than k²-trees *but* much slower.

Thanks for your attention,
time, patience!

Any questions?