

Partial Sums and Dynamic Arrays

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- Dynamic Arrays

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Partial Sums

- **Partial sums.** Maintain array $A[0,1,\dots, n]$ of integers support the following operations.
 - SUM(i): return $A[1] + A[2] + \dots + A[i]$
 - UPDATE(i, Δ): set $A[i] = A[i] + \Delta$

-	1	2	1	1	0	2	3	1	0	1	3	4	1	1	1	2
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16

Partial Sums

- Applications.

- Dynamic lists and arrays (random access into changing lists)
- Arithmetic coding.
- Succinct data structures.
- Lower bounds and cell probe complexity.
- Basic component in many data structures.

Partial Sums

- **Goal.** Partial sums data structure with SUM and UPDATE in **optimal** $O(\log n)$ time and no extra space(!)
- **Solution in 4 steps.**
 - **Explicit array:** slow sum and ultra fast update.
 - **Explicit partial sum:** ultra fast sum and slow update.
 - **Balanced binary tree:** fast sum and fast update.
 - **Fenwick tree:** fast sum and fast update and no extra space.

Solution 1: Explicit Array

-	1	2	1	1	0	2	3	1	0	1	3	4	1	1	1	2
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16

- **Slow sum and ultra fast update.** Maintain A explicitly.
 - SUM(i): compute $A[0] + \dots + A[i]$.
 - UPDATE(i, Δ): set $A[i] = A[i] + \Delta$
- **Time.**
 - $O(i) = O(n)$ for SUM, $O(1)$ for UPDATE.

Solution 2: Explicit Partial Sum

-	1	3	4	5	5	7	10	11	11	12	15	19	20	21	22	24
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16

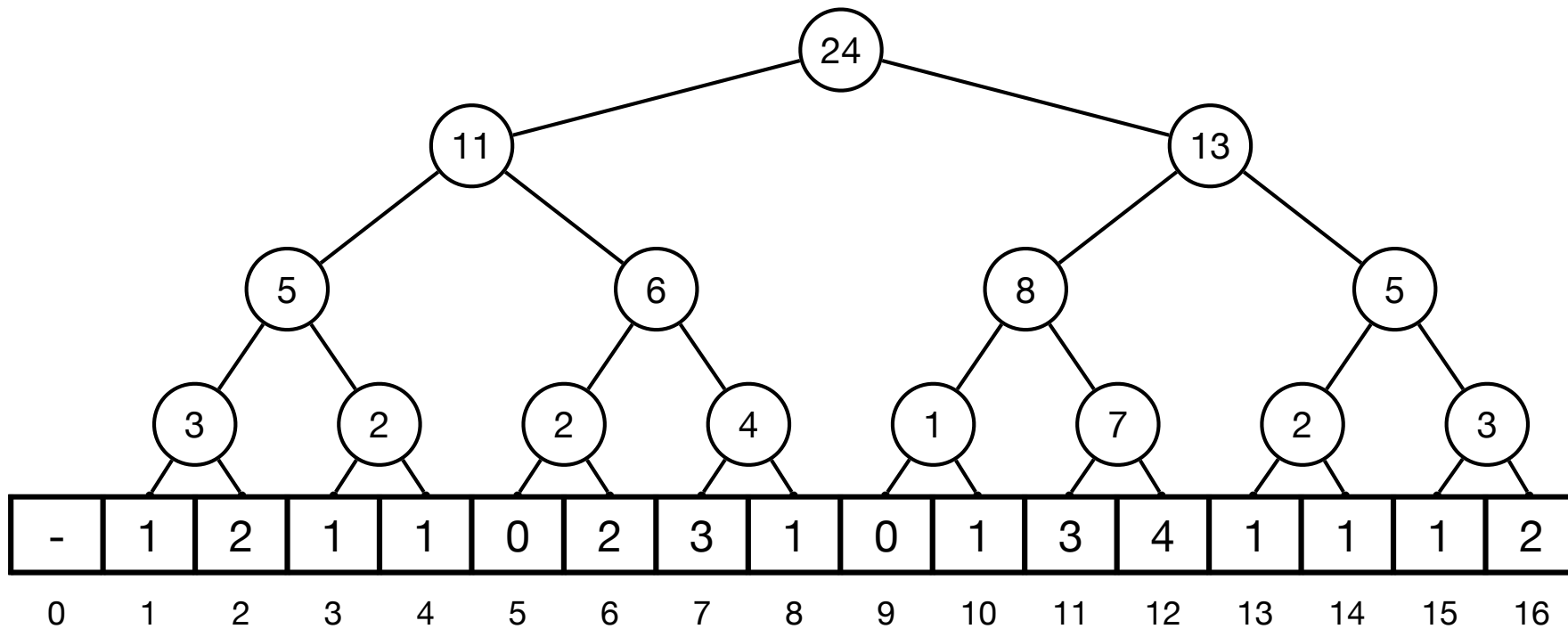
-	1	2	1	1	0	2	3	1	0	1	3	4	1	1	1	2
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16

- **Ultra fast sum and slow update.** Maintain **partial sum** P of A .
 - $SUM(i)$: return $P[i]$.
 - $UPDATE(i, \Delta)$: add Δ to $P[i], P[i+1], \dots, P[n]$.
- **Time.**
 - $O(1)$ for SUM , $O(n - i + 1) = O(n)$ for $UPDATE$.

Partial Sums

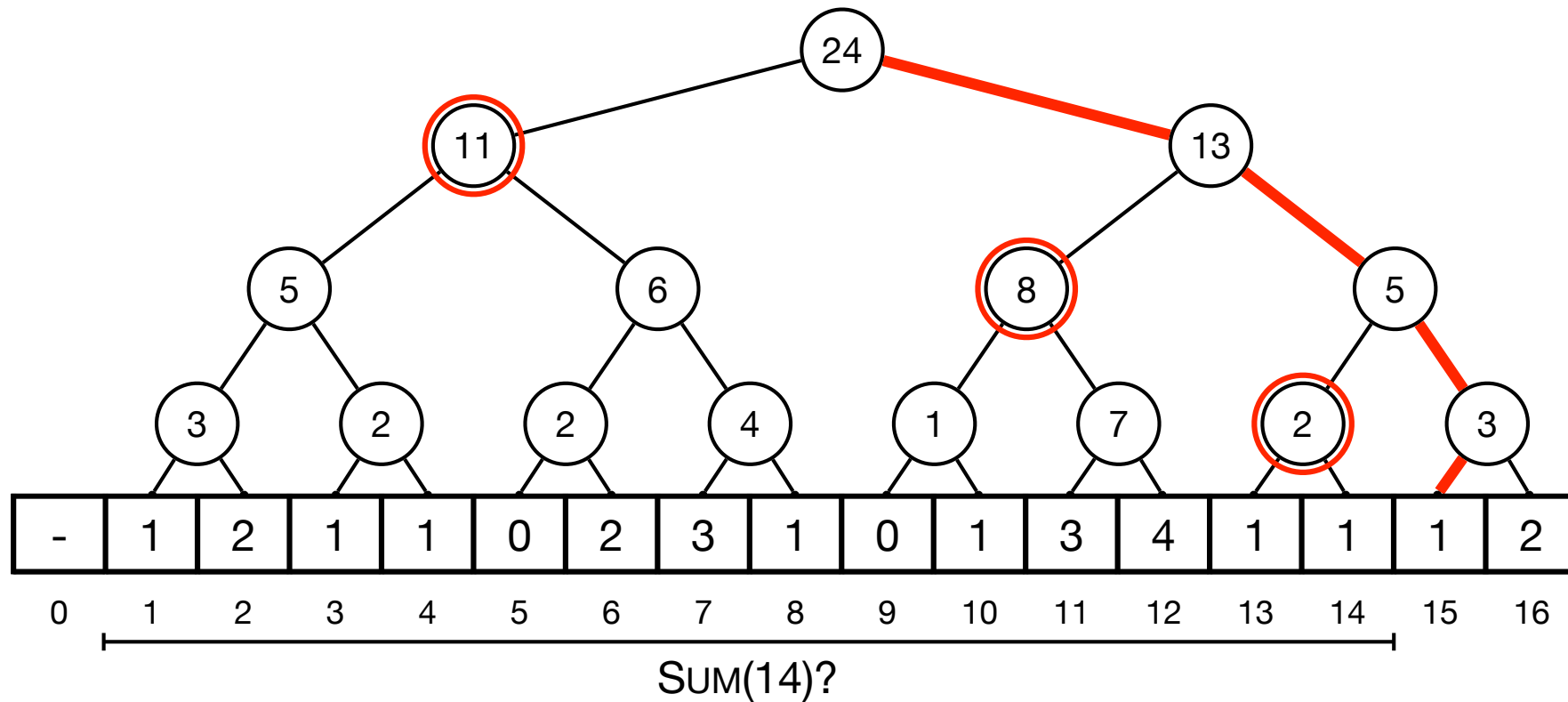
Data structure	SUM	UPDATE	Space
explicit array	$O(n)$	$O(1)$	$O(n)$
explicit partial sum	$O(1)$	$O(n)$	$O(n)$

Solution 3: Balanced Binary Tree



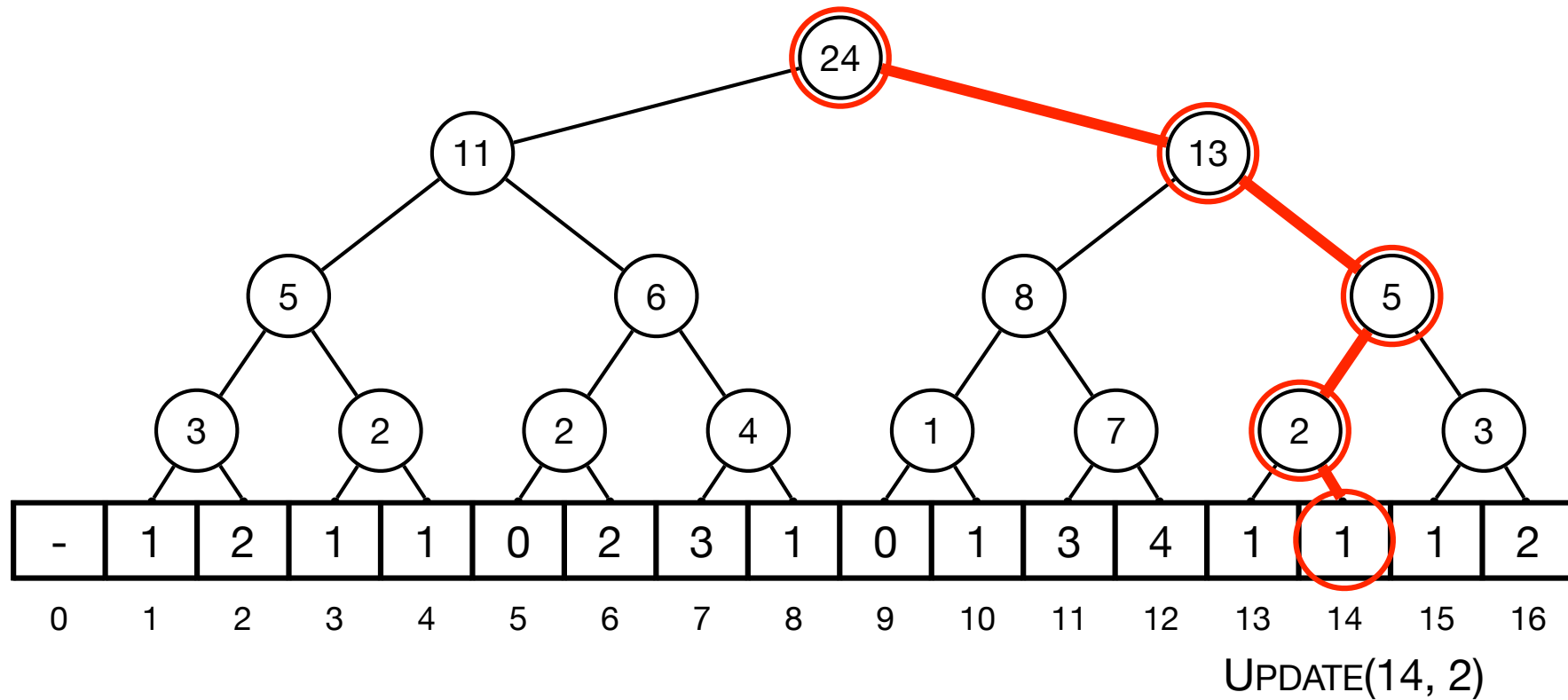
- **Fast sum and fast update.** Maintain balanced binary tree T on A . Each node stores the sum of elements in subtree.

Solution 3: Balanced Binary Tree



- **SUM.**
 - SUM(i): traverse path to $i + 1$ and sum up all **off-path** nodes.
- **Time.** $O(\log n)$

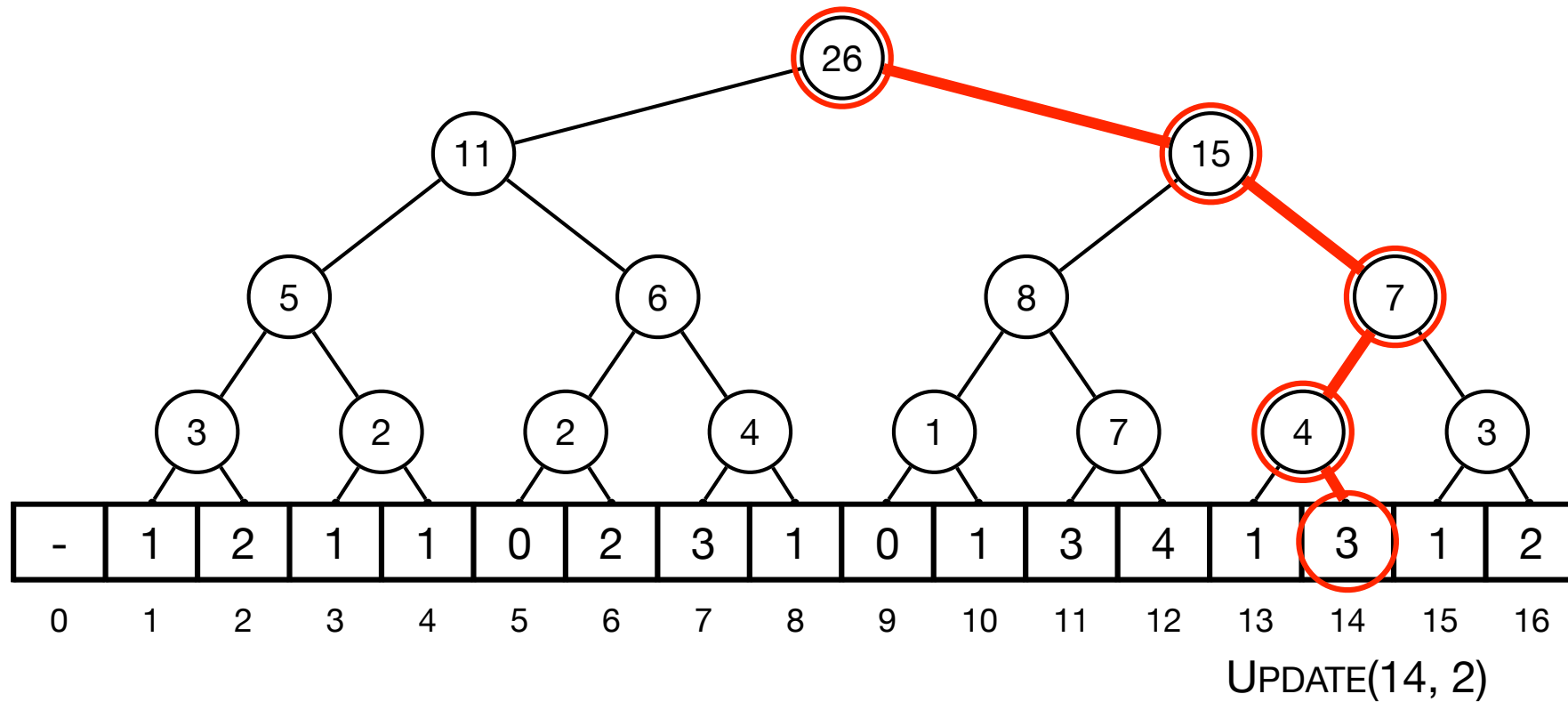
Solution 3: Balanced Binary Tree



- UPDATE.

- UPDATE(i, Δ): add Δ to nodes on path to i .

Solution 3: Balanced Binary Tree



- **UPDATE.**
 - UPDATE(i, Δ): add Δ to nodes on path to i .
- **Time.** $O(\log n)$

Partial Sums

Data structure	SUM	UPDATE	Space
explicit array	$O(n)$	$O(1)$	$O(n)$
explicit partial sum	$O(1)$	$O(n)$	$O(n)$
balanced binary tree	$O(\log n)$	$O(\log n)$	$O(n)$
lower bound	$\Omega(\log n)$	$\Omega(\log n)$	

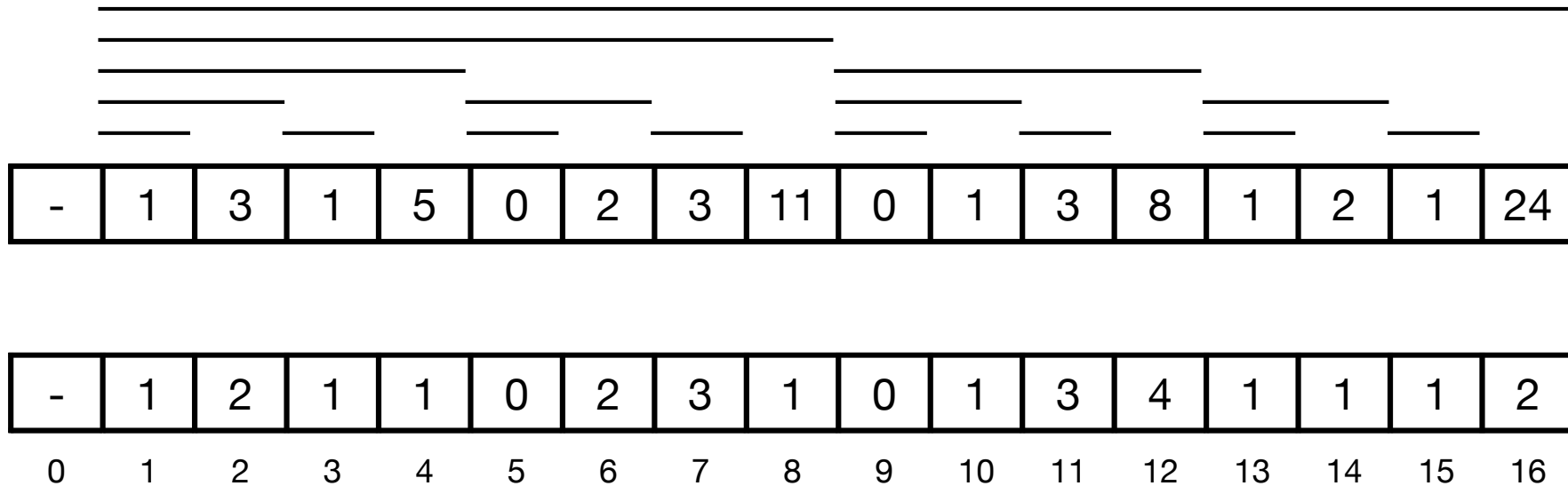
- **Challenge.** How can we improve?
- **In-place data structure.**
 - Replace input array A with data structure of exactly same size.
 - Use only $O(1)$ extra space.

Solution 4: Fenwick Tree

-	1	3	1	5	0	2	3	11	0	1	3	8	1	2	1	24
-	1	3	1	5	0	2	3	11	0	1	3	8	1	2	1	13
-	1	3	1	5	0	2	3	6	0	1	3	8	1	2	1	5
-	1	3	1	2	0	2	3	4	0	1	3	7	1	2	1	3
-	1	2	1	1	0	2	3	1	0	1	3	4	1	1	1	2
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16

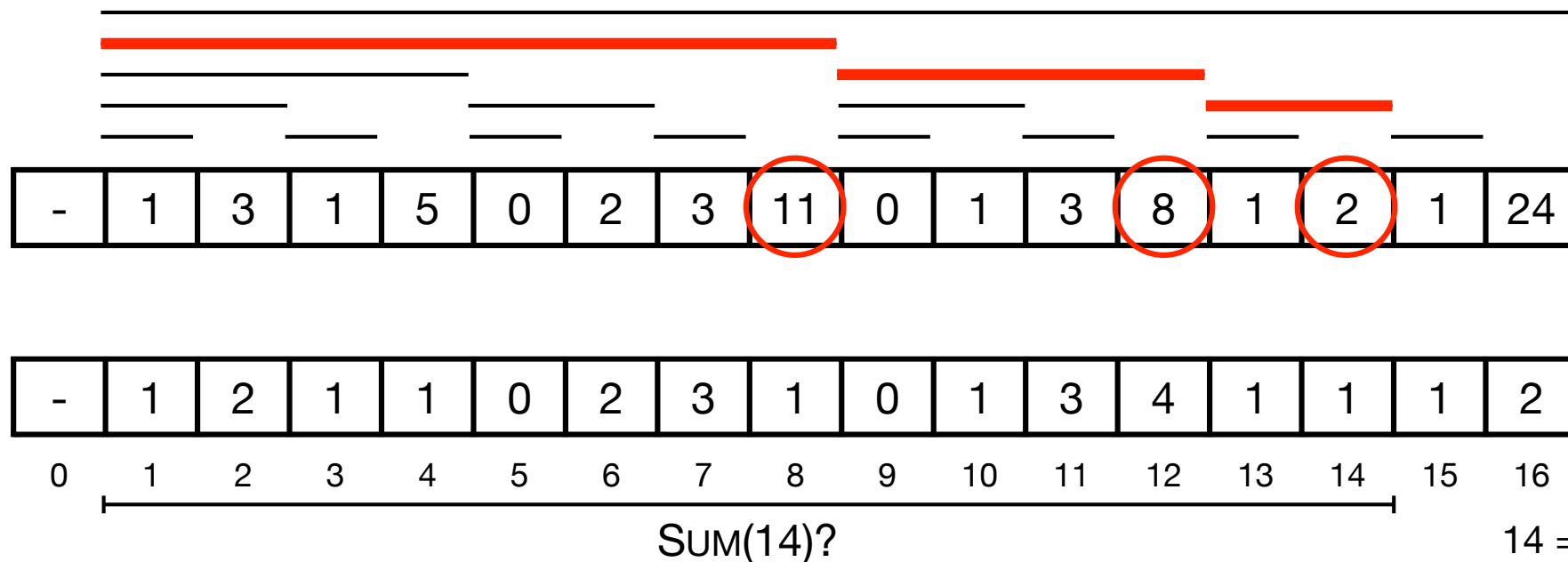
- **Fenwick tree.** Replace A by another array F.
 - Replace all even entries $A[2i]$ by $A[2i - 1] + A[2i]$.
 - Recurse on the entries $A[2, 4, \dots, n]$ until we are left with a single element.

Solution 4: Fenwick Tree



- **Fenwick tree.** Replace A by another array F.
 - Replace all even entries $A[2i]$ by $A[2i - 1] + A[2i]$.
 - Recurse on the entries $A[2, 4, \dots, n]$ until we are left with a single element.
- **Space.**
 - In-place. No extra space.

Solution 4: Fenwick Tree



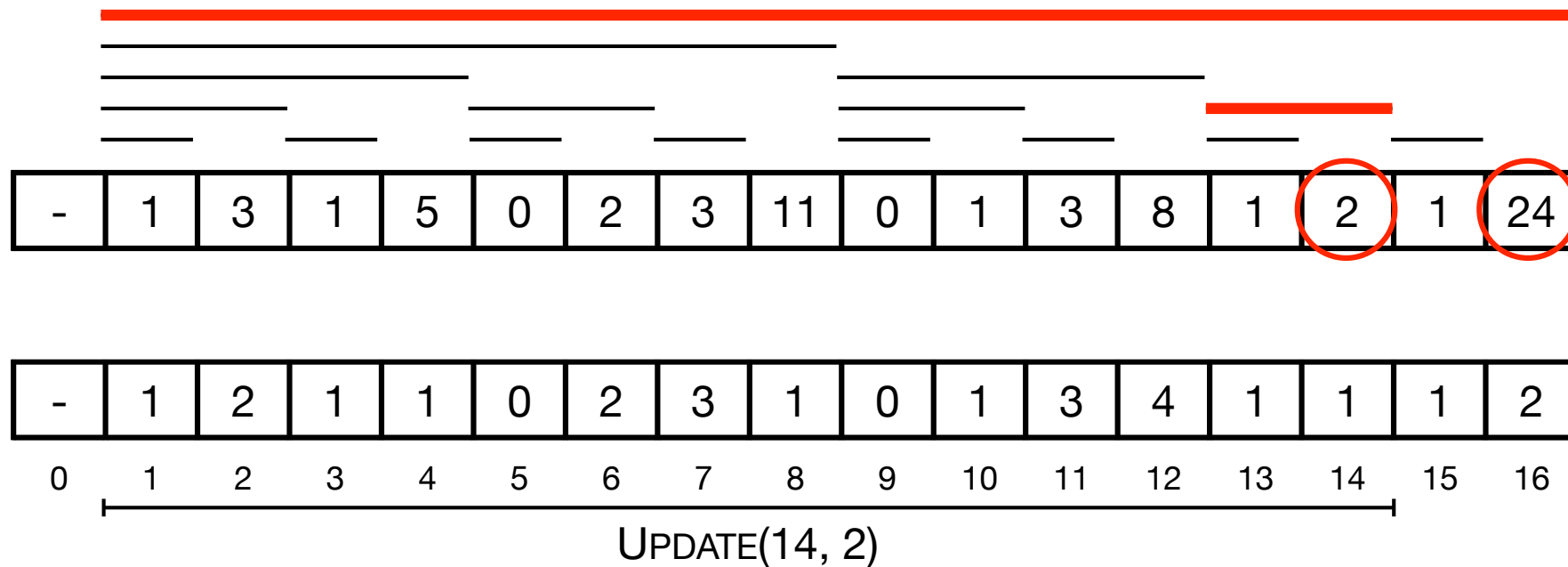
- **SUM.**

- SUM(i): add largest partial sums covering $[1, \dots, i]$.
- Indexes i_0, i_1, \dots in F given by $i_0 = i$ and $i_{j+1} = i_j - \text{rmb}(i_j)$, where $\text{rmb}(i_j)$ is the integer corresponding to the rightmost 1-bit in i . Stop when we get 0.

$$\begin{aligned}
 14 &= 1110_2 \\
 12 &= 1100_2 \\
 8 &= 1000_2 \\
 0 &= 0000_2
 \end{aligned}$$

- **Time.** $O(\log n)$

Solution 4: Fenwick Tree



- **UPDATE.**

- UPDATE(i, Δ): add Δ to partial sums covering i .

- Indexes i_0, i_1, \dots in F given by $i_0 = i$ and $i_{j+1} = i_j + \text{rmb}(i_j)$. Stop when we get n .

- **Time.** $O(\log n)$

$$14 = 1110_2$$

$$16 = 10000_2$$

Partial Sums

Data structure	SUM	UPDATE	Space
explicit array	$O(n)$	$O(1)$	$O(n)$
explicit partial sum	$O(1)$	$O(n)$	$O(n)$
balanced binary tree	$O(\log n)$	$O(\log n)$	$O(n)$
lower bound	$\Omega(\log n)$	$\Omega(\log n)$	
Fenwick tree	$O(\log n)$	$O(\log n)$	in-place

- [Practical?](#) Fenwick trees for competitive programming.

Partial Sums and Dynamic Arrays

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- Dynamic Arrays

Dynamic Arrays

- **Dynamic arrays.** Maintain array $A[0, \dots, n-1]$ of integers and support the following operations.
 - **ACCESS(i):** return $A[i]$.
 - **INSERT(i, x):** insert a new entry with value x immediately to the left of entry i .
 - **DELETE(i):** Remove entry i .

1	2	1	1	0	2	3	1	0	1	3	4	1	1	1	2
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

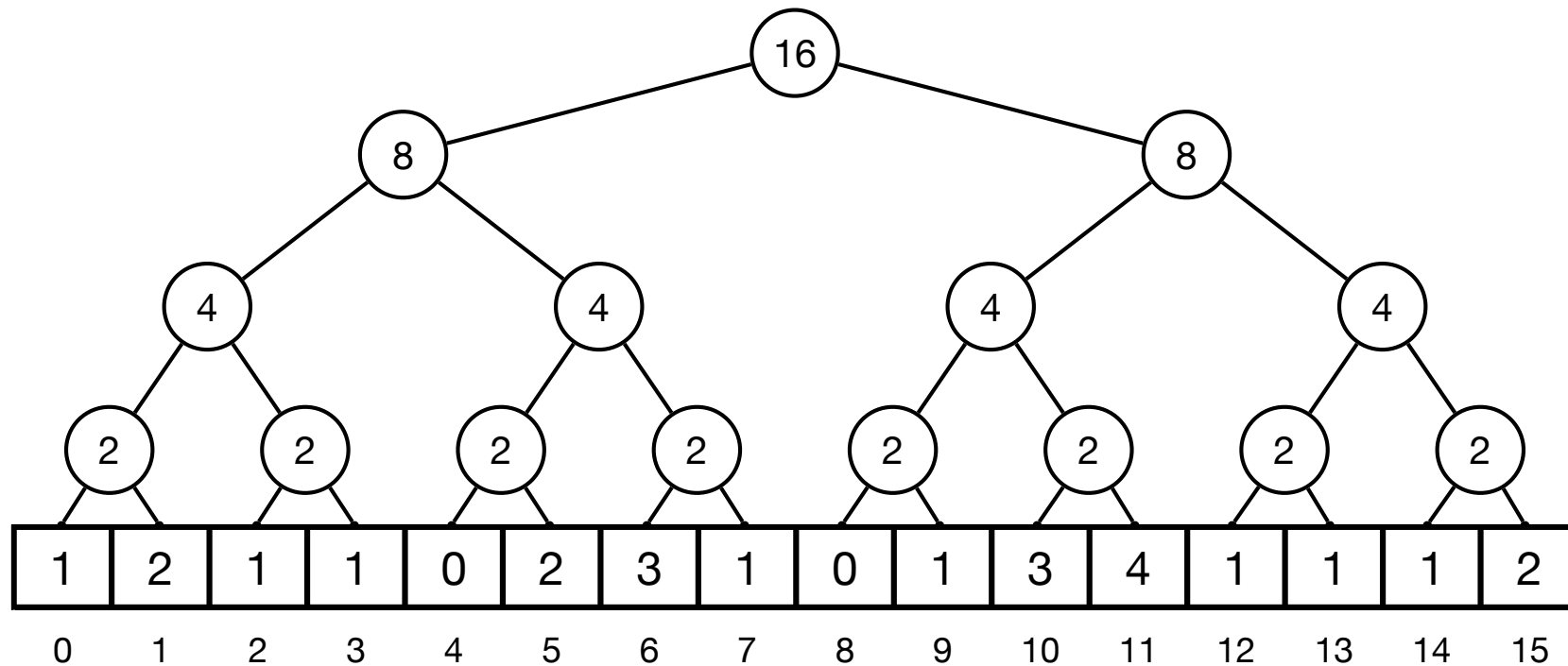
Dynamic Arrays

- Applications.
 - Dynamic lists and arrays (random access into changing lists)
 - Basic component in many data structures.

Dynamic Arrays

- **Goal.** Dynamic array data structure with linear space that supports ACCESS in $O(1)$ time and INSERT and DELETE in $O(\sqrt{n})$ time.
- **Solution in 4 steps.**
 - **Balanced binary tree:** fast access and fast update.
 - **Explicit array:** ultra fast access and slow update.
 - **Rotated array:** ultra fast access and slow update (but with ultrafast updates at endpoints).
 - **2-level rotated array:** ultra fast access and faster update.

Solution 1: Balanced Binary Tree



- **Fast access and fast update.** Maintain balanced binary tree T on A . Each node stores the number of elements in subtree.
 - $\text{ACCESS}(i)$: traverse path to leaf j .
 - $\text{INSERT}(i, x)$: insert new leaf and update tree.
 - $\text{DELETE}(i)$: delete new leaf and update tree.
- **Time.** $O(\log n)$ for ACCESS , INSERT , and DELETE .

Solution 2: Explicit Array

1	2	1	1	0	2	3	1	0	1	3	4	1	1	1	2
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

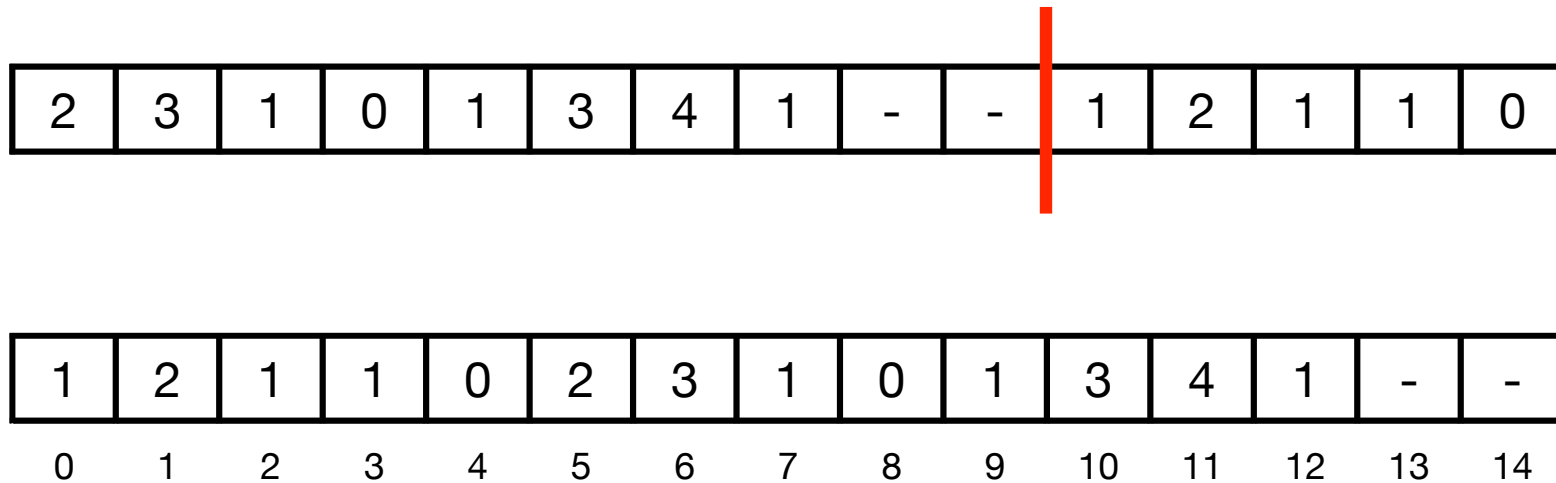
- **Ultra fast access and slow update.** Maintain A explicitly.
 - ACCESS(i): return A[i].
 - INSERT(i, x): Shift all elements from i to n by 1 to the right. Set A[i] = x.
 - DELETE(i): shift all elements to the right of entry i to the left by 1.
- **Time.**
 - O(1) for ACCESS and $O(n-i+1) = O(n)$ for INSERT and DELETE.

Dynamic Arrays

Data structure	ACCESS	INSERT	DELETE	Space
balanced binary tree	$O(\log n)$	$O(\log n)$	$O(\log n)$	$O(n)$
explicit array	$O(1)$	$O(n)$	$O(n)$	$O(n)$
lower bound	$\Omega(\log n / \log \log n)$	$\Omega(\log n / \log \log n)$	$\Omega(\log n / \log \log n)$	

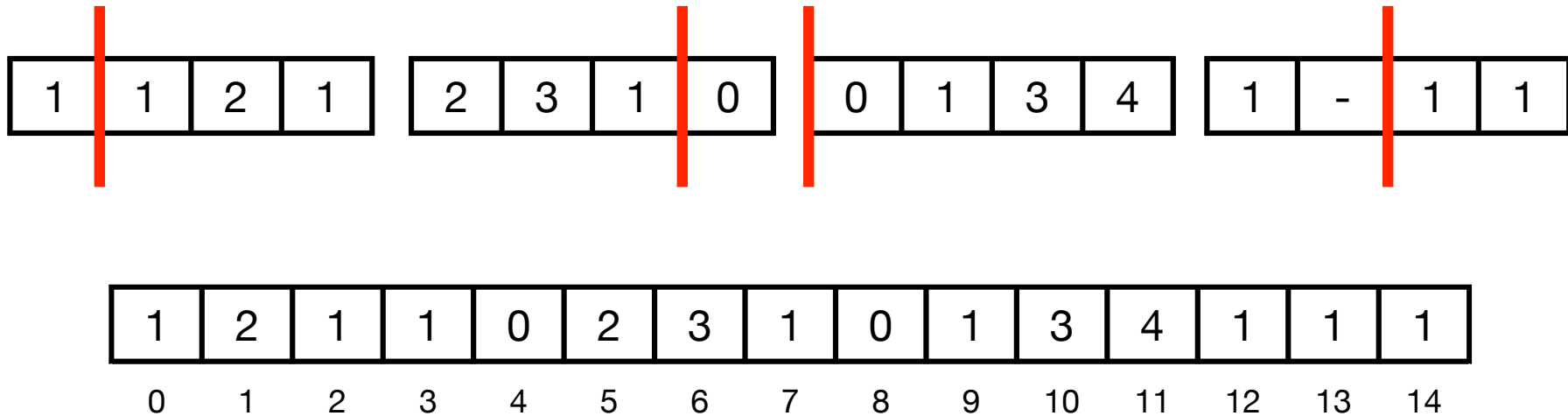
- **Challenge.** What can we get if we insist on **constant time** ACCESS?

Solution 3: Rotated array



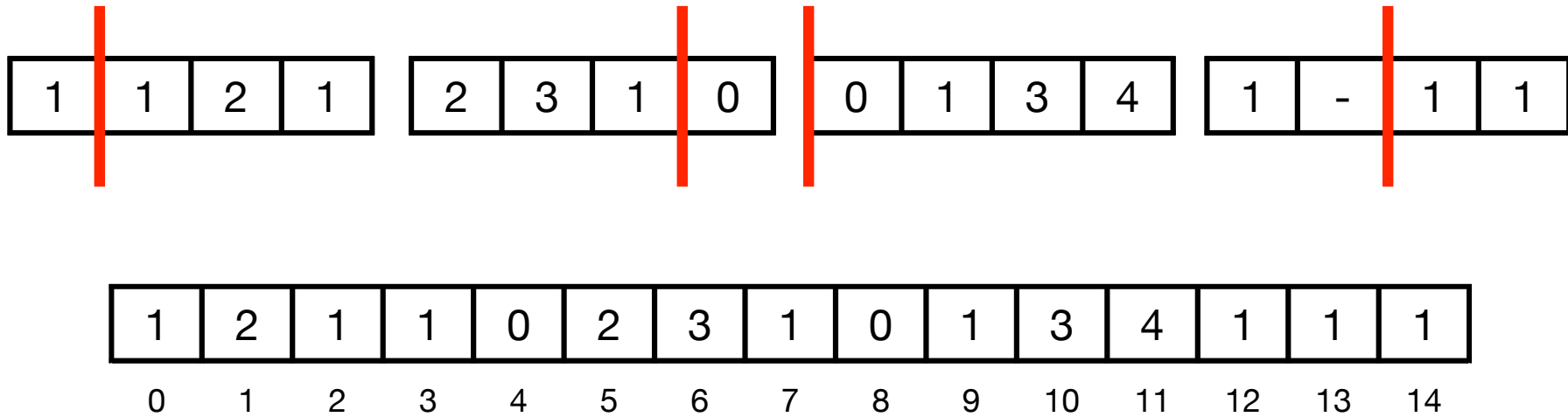
- **Rotated array with capacity N.** Maintain a circular shift of array in an array with capacity N. Store **offset** to mark start of array.
 - ACCESS, INSERT, and DELETE as in solution 1 but shifted by offset.
- **Time.** $O(1)$ for ACCESS and $O(n)$ for INSERT and DELETE.
- INSERT and DELETE at **endpoints** in $O(1)$ time.

Solution 4: 2-level Rotated Arrays



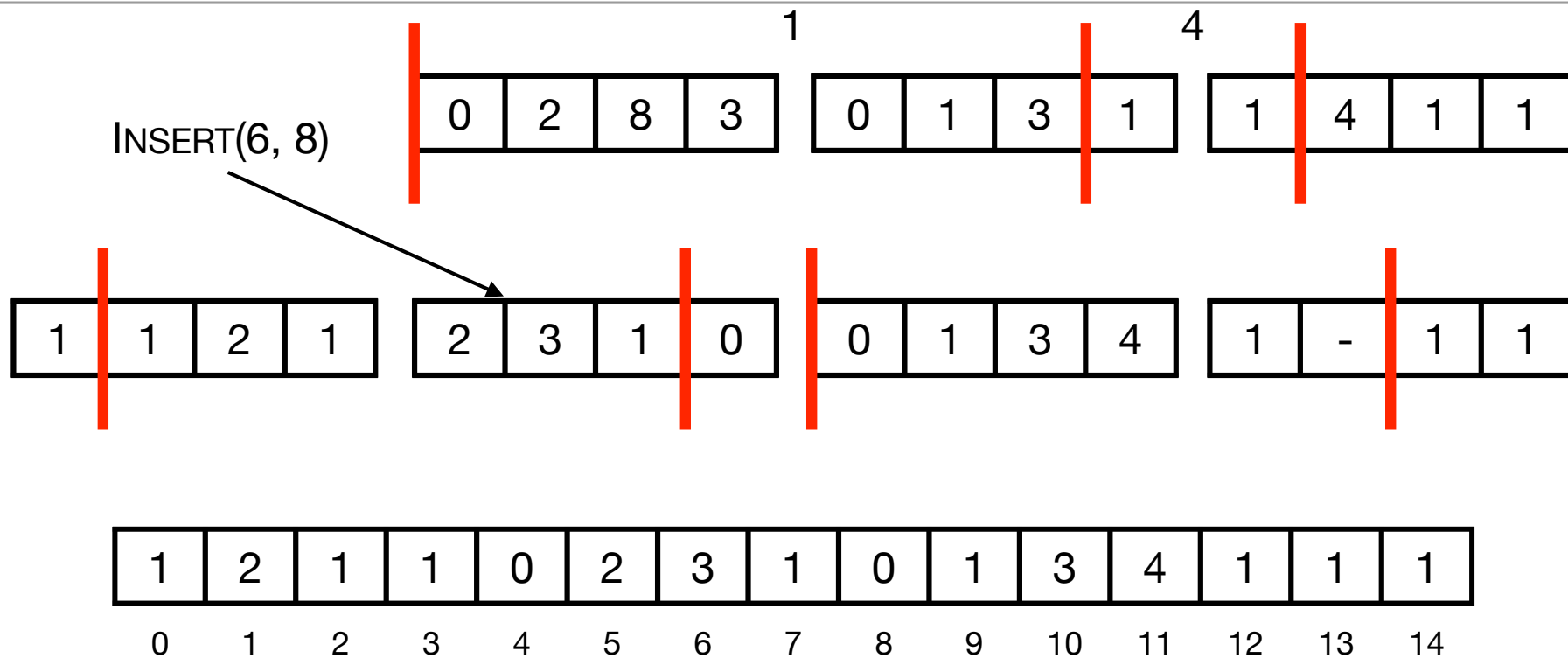
- 2-level rotated arrays.
 - Store \sqrt{n} rotated arrays $R_0, \dots, R_{\sqrt{n}-1}$ with **capacity** \sqrt{n} (last may have smaller capacity).

Solution 4: 2-level Rotated Arrays



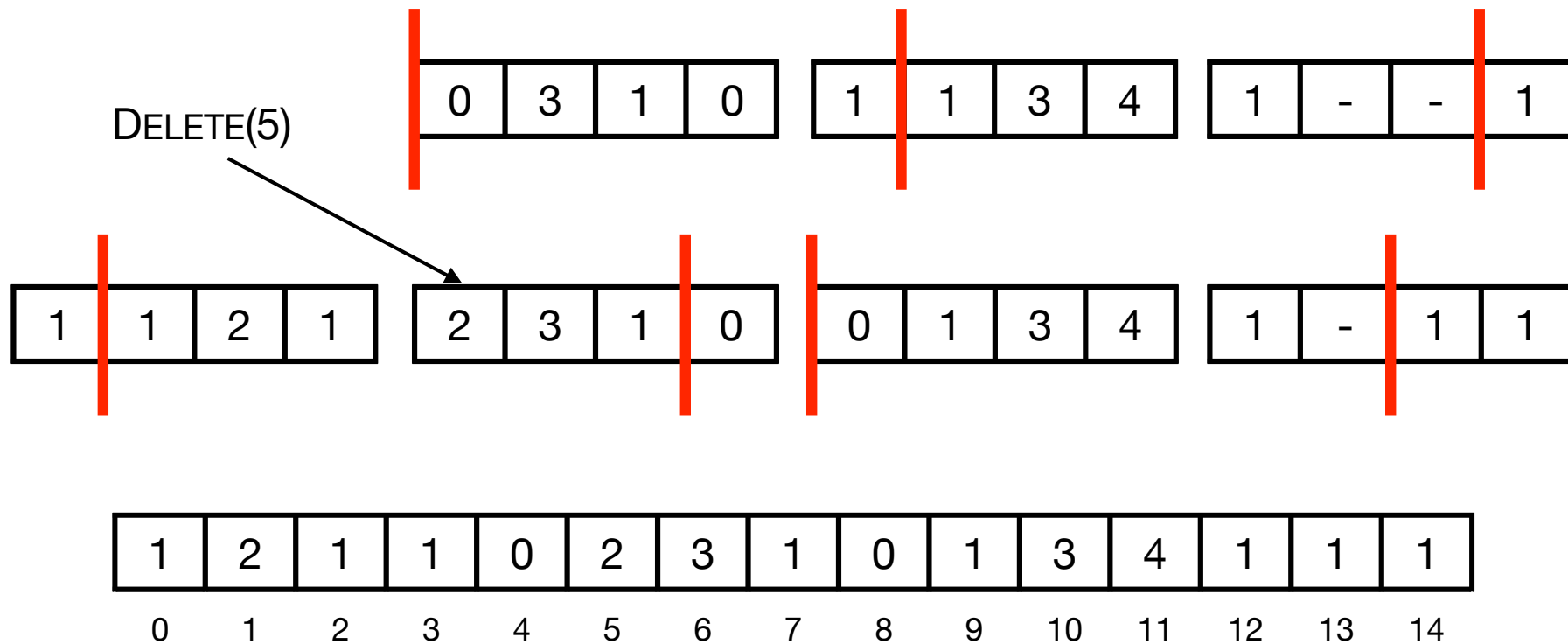
- **ACCESS.**
 - ACCESS(i): compute rotated array R_j and index k corresponding to i . Return $R_j[k]$.
- **Time.** $O(1)$

Solution 4: 2-level Rotated Arrays



- **INSERT.**
 - INSERT(i, x): find R_j and k as in ACCESS.
 - Rebuild R_j with new entry inserted.
 - Propagate **overflow** to R_{j+1} **recursively**.
- **Time.** $O(\sqrt{n})$

Solution 4: 2-level Rotated Arrays



- DELETE.
 - DELETE(i): find R_j and k as in ACCESS.
 - Rebuild R_j with entry i deleted.
 - Propagate **underflow** to R_{j+1} **recursively**.
- Time. $O(\sqrt{n})$

Dynamic Arrays

Data structure	ACCESS	INSERT	DELETE	Space
balanced binary tree	$O(\log n)$	$O(\log n)$	$O(\log n)$	$O(n)$
explicit array	$O(1)$	$O(n)$	$O(n)$	$O(n)$
lower bound	$\Omega(\log n / \log \log n)$	$\Omega(\log n / \log \log n)$	$\Omega(\log n / \log \log n)$	
2-level rotated array	$O(1)$	$O(\sqrt{n})$	$O(\sqrt{n})$	$O(n)$
$O(1)$ -level rotated array	$O(1)$	$O(n^\epsilon)$	$O(n^\epsilon)$	$O(n)$

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