

Data Structures II

- Partial Sums
- 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 $A[1] + A[2] + \dots + A[i]$
 - $UPDATE(i, \Delta)$: set $A[i] = A[i] + \Delta$

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

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.
- **Challenge.** How can solve the problem with current techniques?

Partial Sums

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

- **Slow sum and ultra fast updates.** 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.

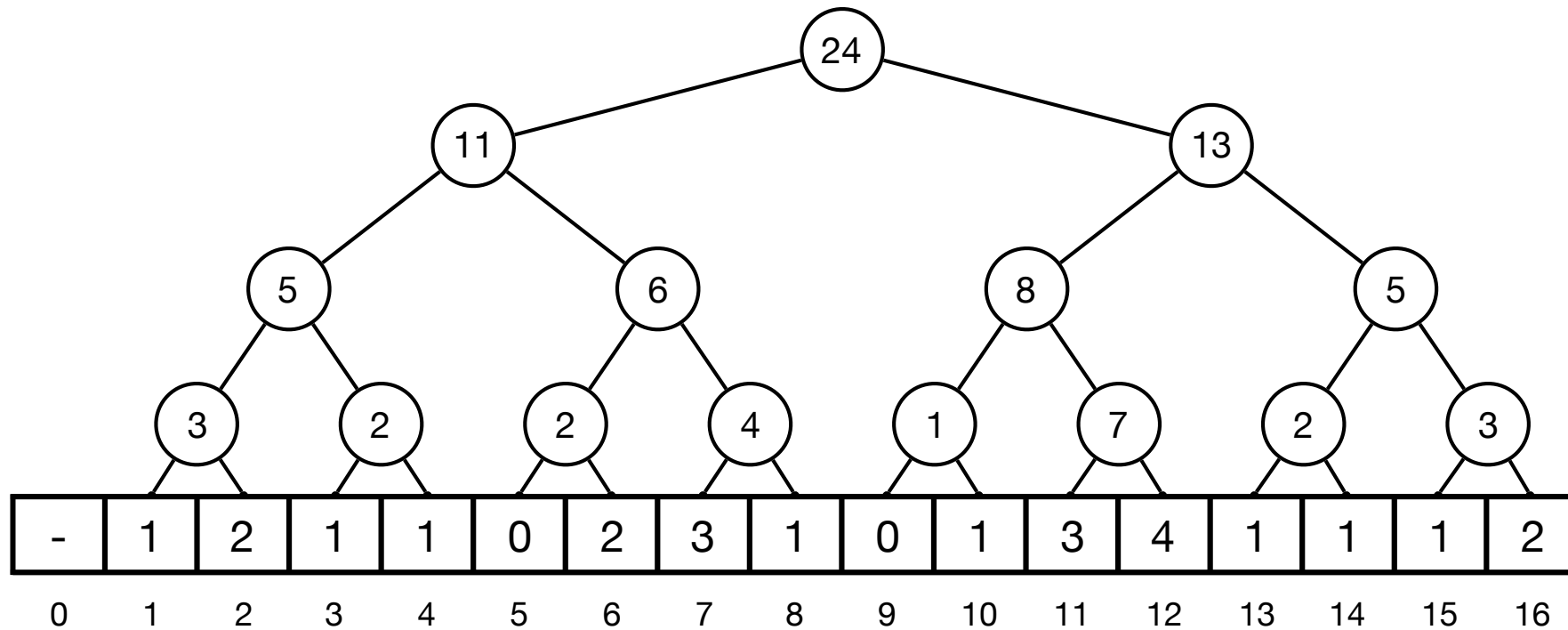
Partial Sums

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

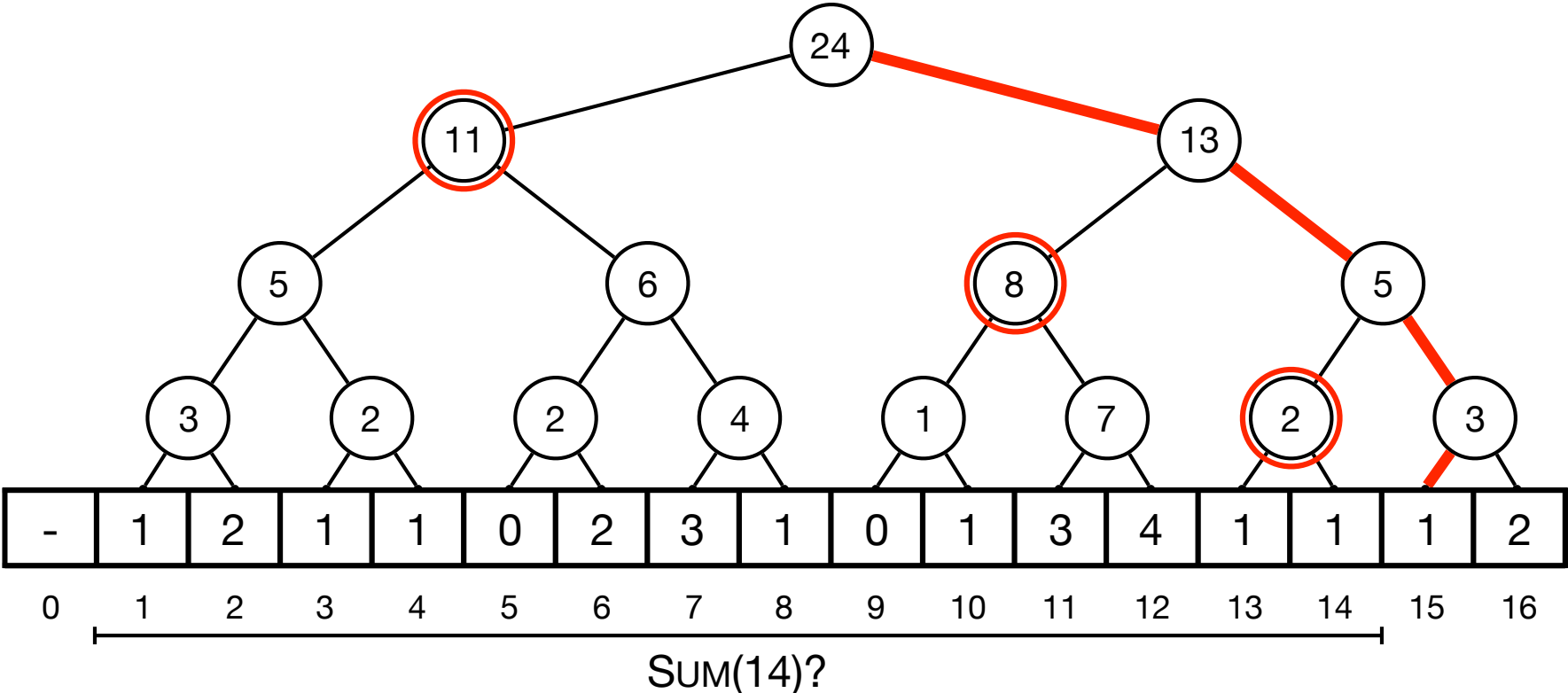
- **Ultra fast sum and slow updates.** Maintain **partial sum** P of A .
 - $SUM(i)$: return $P[i]$.
 - $UPDATE(i, \Delta)$: add Δ to $P[i], P[i+1], \dots, P[n-1]$.
- **Time.**
 - $O(1)$ for SUM , $O(n - i + 1) = O(n)$ for $UPDATE$.

Partial Sums



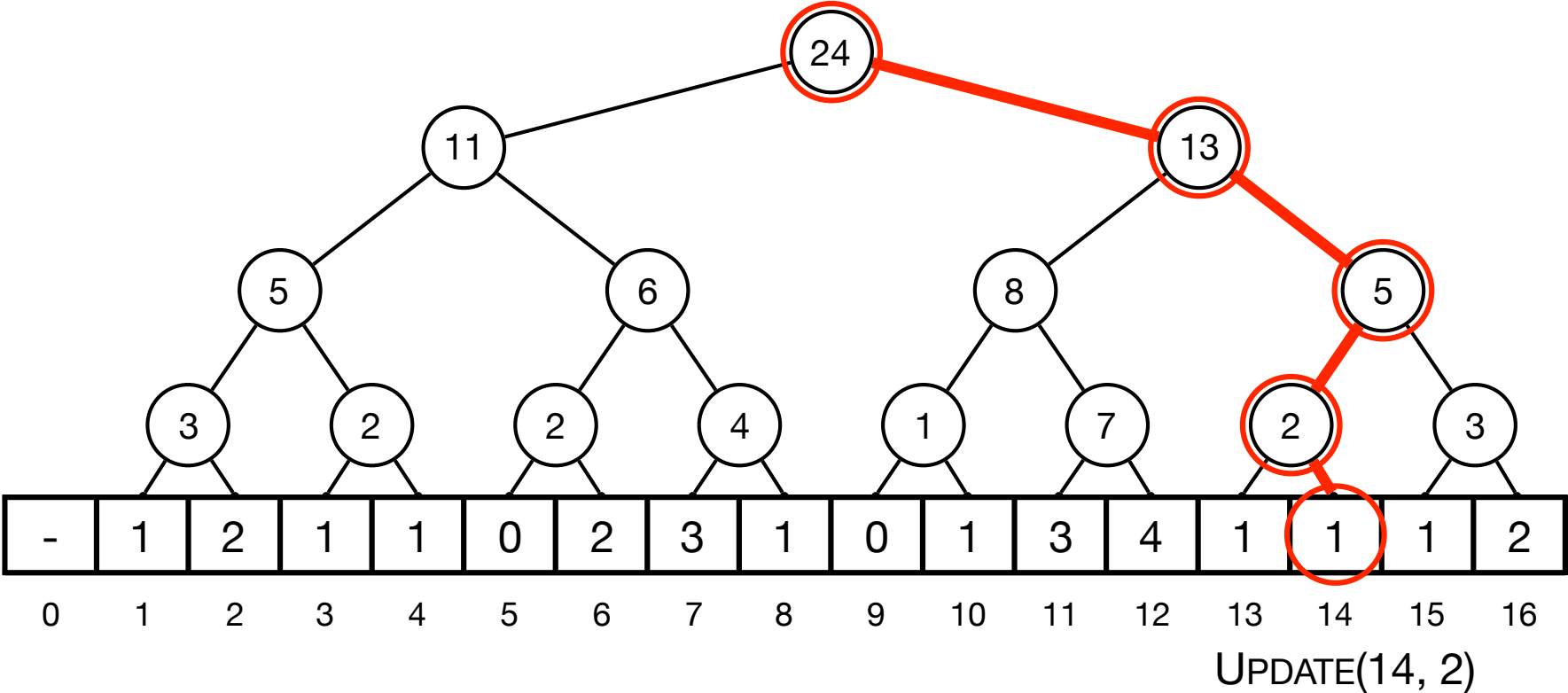
- **Fast sum and fast updates.** Maintain balanced binary tree T on A . Each node stores the sum of elements in subtree.

Partial Sums



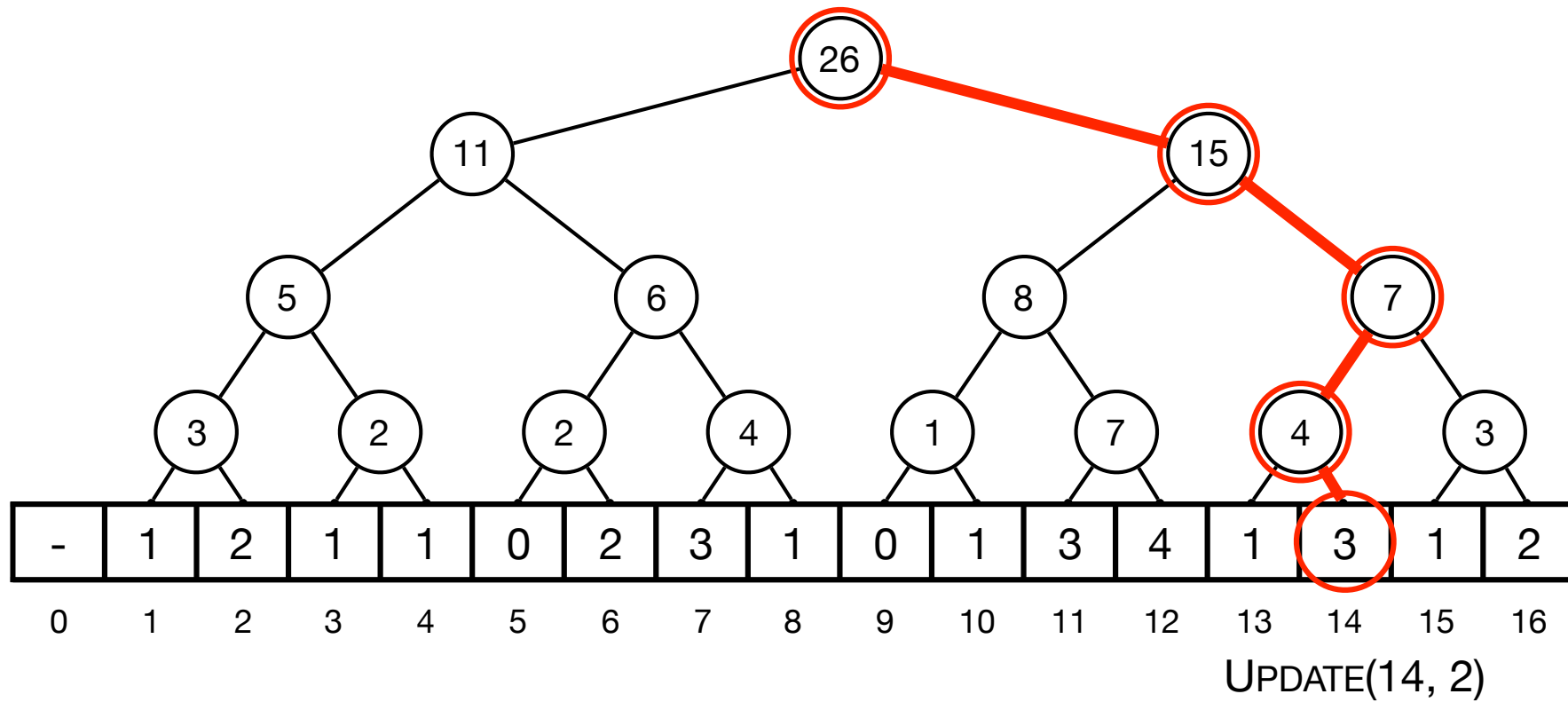
- SUM.
 - SUM(i): traverse path to $i + 1$ and sum up all **off-path** nodes.
- Time. $O(\log n)$

Partial Sums



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

Partial Sums



- **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	$\Omega(\log n)$	$\Omega(\log n)$	

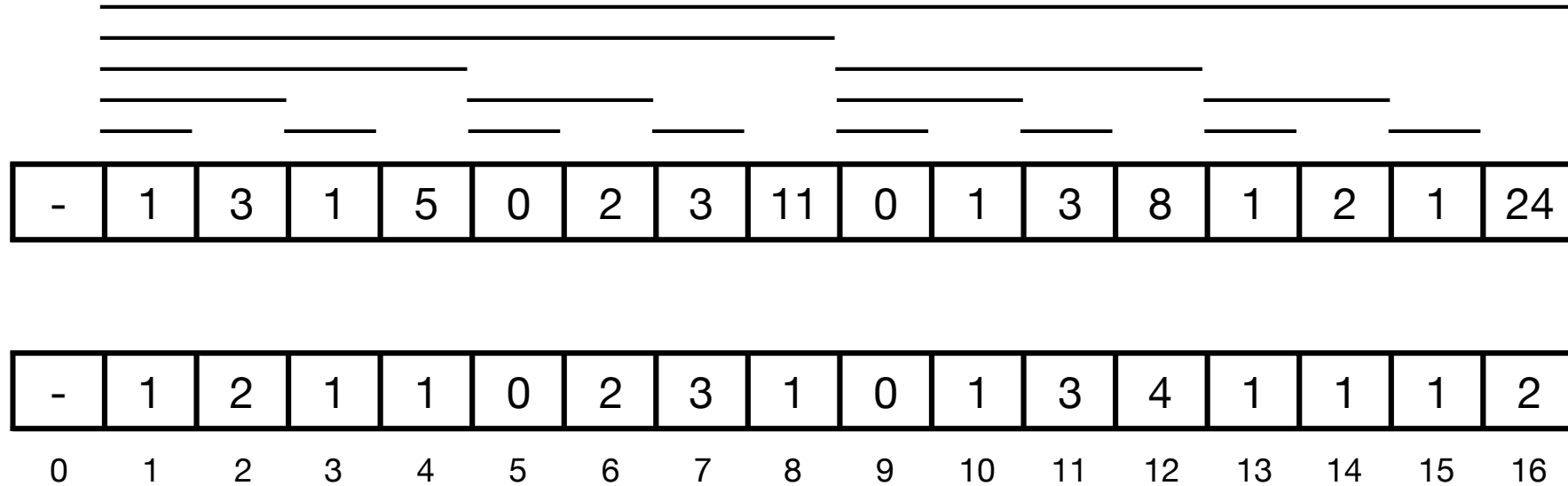
- **Challenge.** How can we improve?
- **In-place data structure.**
 - Replace input array A with data structure D of exactly same size.
 - Use only $O(1)$ space in addition to D .

Partial Sums

-	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

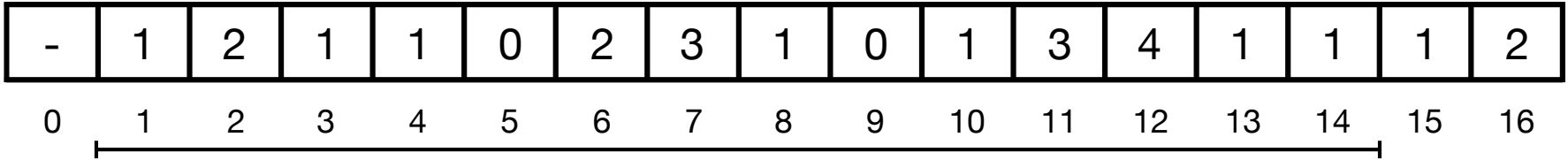
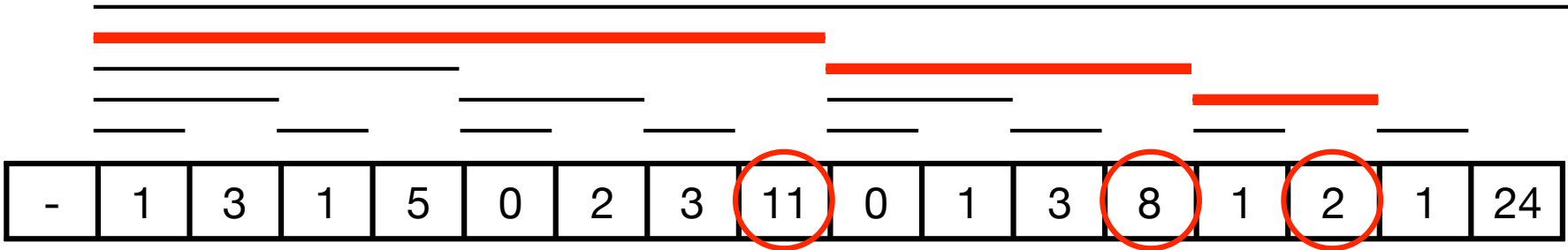
- **Fenwick tree.** Replace A by another array F.
 - Replace all even entries $A[2i]$ by $A[2i - 1] + A[2i]$.
 - Recurse on the entries $A[2, 4, \dots, n]$ until we are left with a single element.

Partial Sums



- **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.

Partial Sums

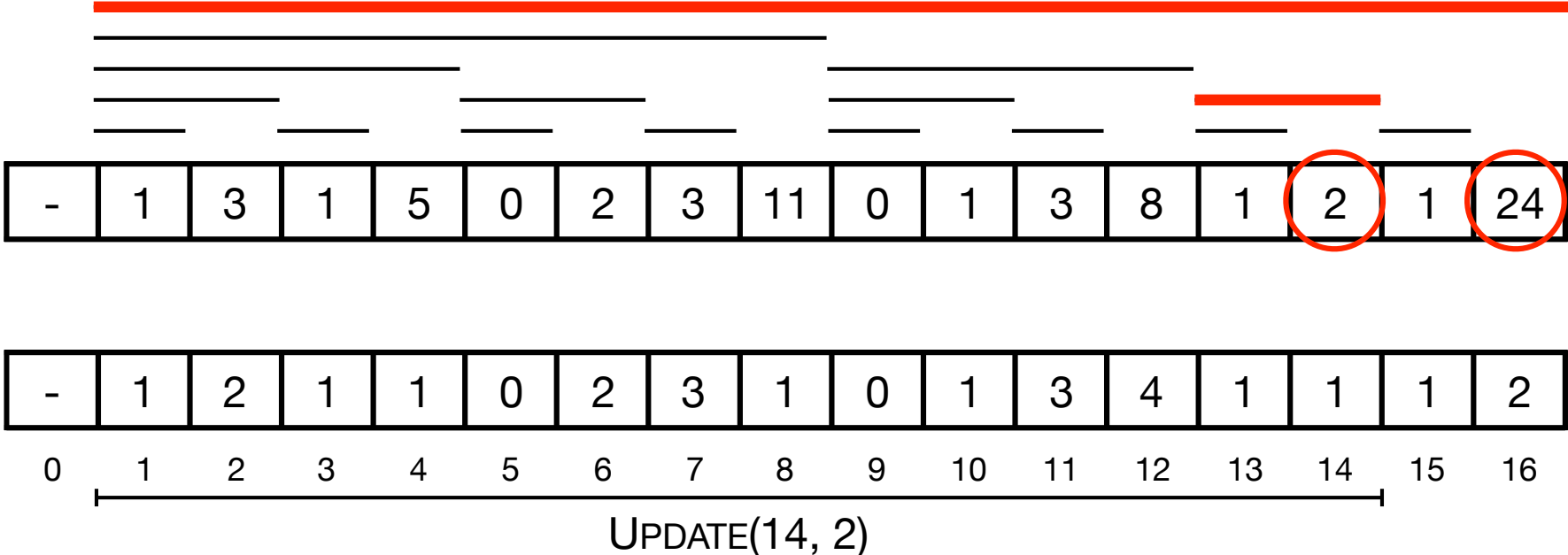


SUM(14)?

$14 = 1110_2$
 $12 = 1100_2$
 $8 = 1000_2$
 $0 = 0000_2$

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

Partial Sums



- **UPDATE.**

$14 = 1110_2$
 $16 = 10000_2$

- `UPDATE(i, Δ)`: add Δ to partial sums covering i .
- Indexes $i_0, i_1, ..$ in F given by $i_0 = i$ and $i_{j+1} = i_j + \text{rmb}(i_j)$. Stop when we get n .

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

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

Data Structures II

- Partial Sums
- Dynamic Arrays

Dynamic Arrays

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

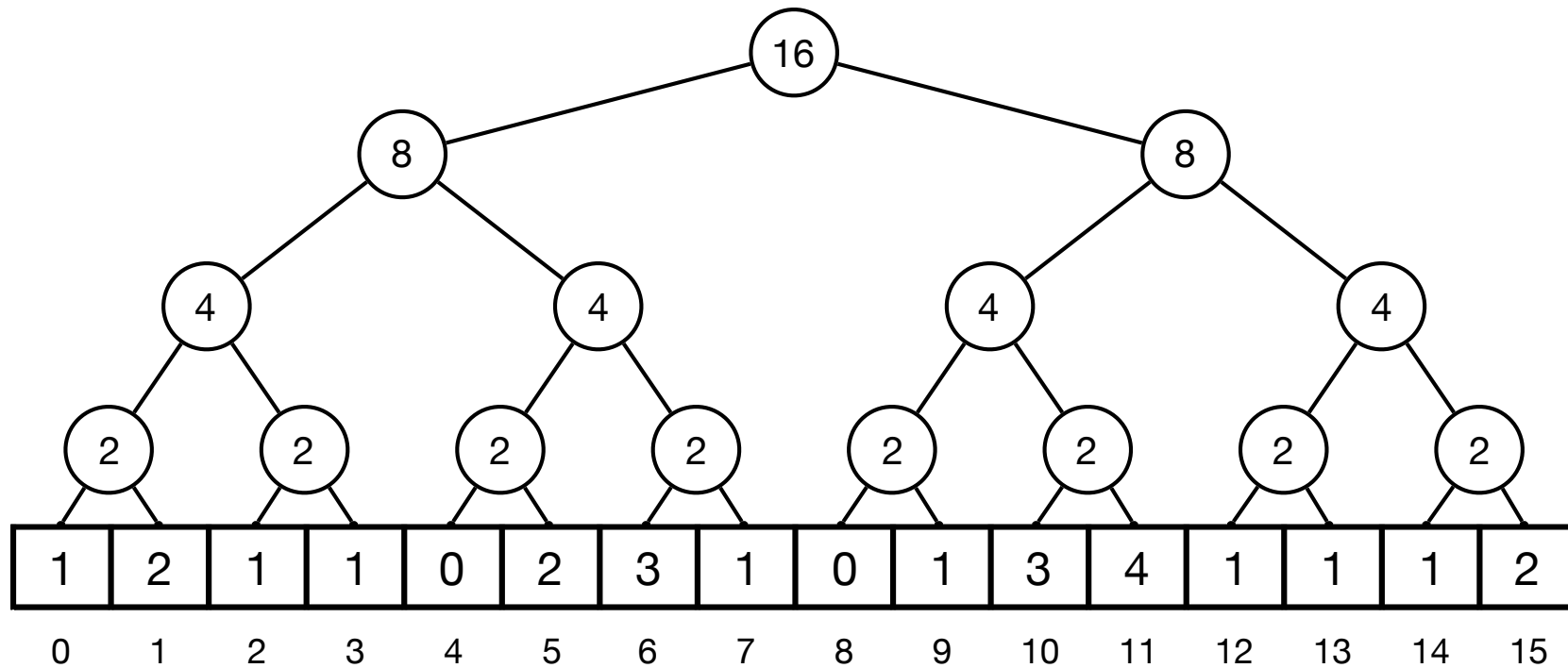
- **Applications.**
 - Dynamic lists and arrays (random access into changing lists)
 - Basic component in many data structures.
- **Challenge.** How can solve the problem with current techniques?

Dynamic Arrays

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

- **Very fast access and slow updates.** Maintain A explicitly.
 - ACCESS(i): return A[i].
 - INSERT(i, x): set A[i] = x. Shift all elements to the right of entry i to the right by 1.
 - 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



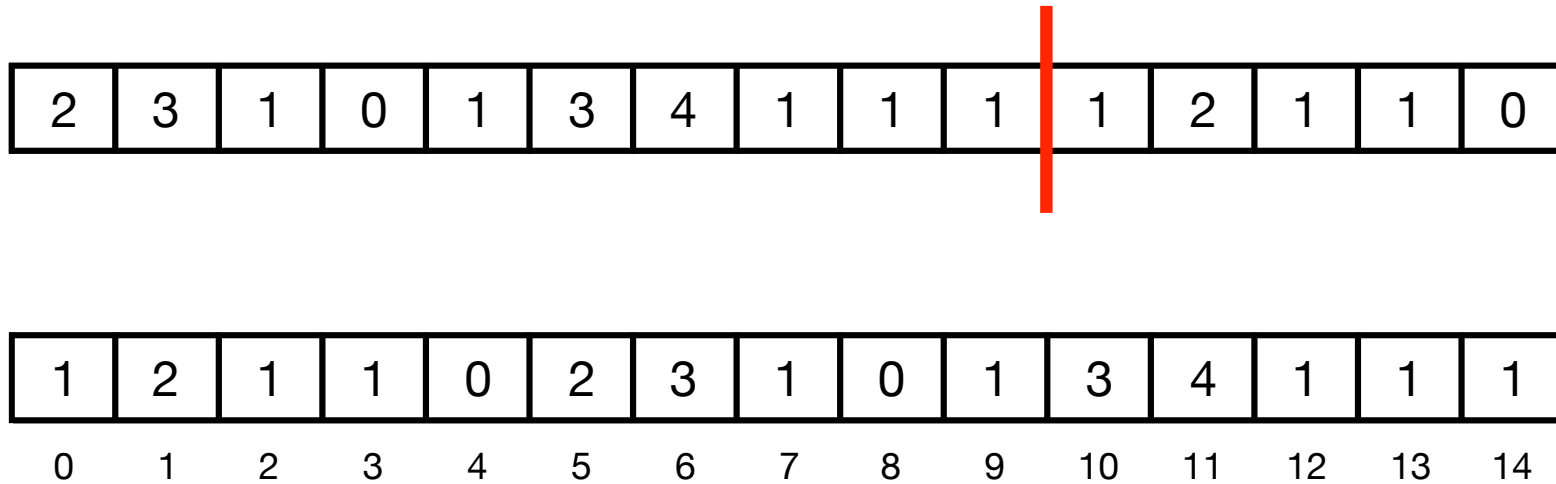
- **Fast access and fast updates.** Maintain balanced binary tree T on A . Each node stores the number of elements in subtree.
 - $\text{ACCESS}(i)$: traverse path to leaf j .
 - $\text{INSERT}(i, x)$: insert new leaf and update tree.
 - $\text{DELETE}(i)$: delete new leaf and update tree.
- **Time.**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

Dynamic Arrays

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

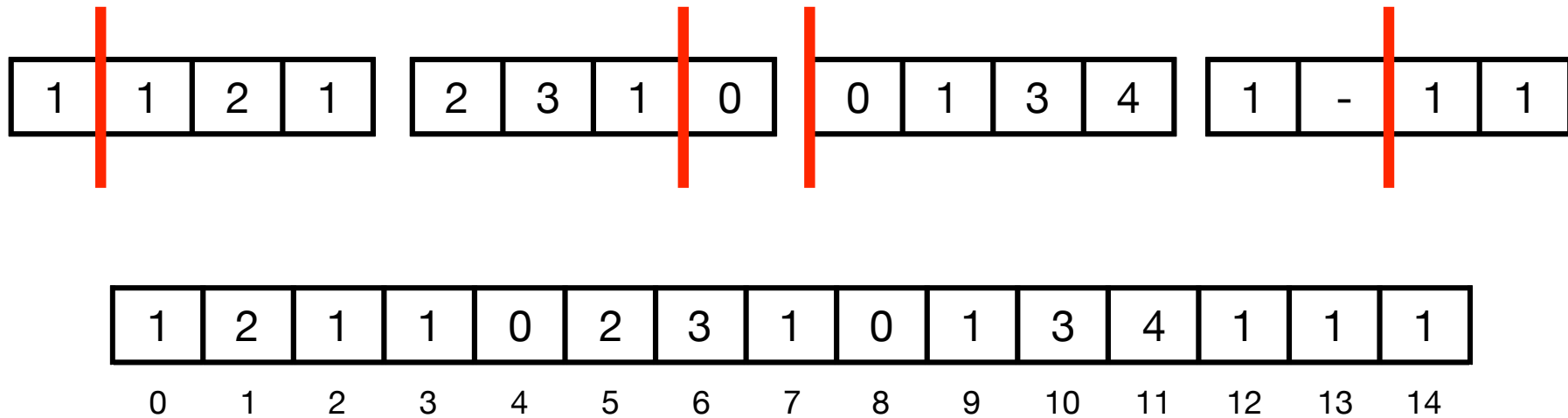
- **Challenge.** What can we get if we insist on **constant time** ACCESS?

Dynamic Arrays



- Rotated array.
 - Circular shift of array by an **offset**.
- Idea.
 - By moving offset we can delete and insert at endpoints in $O(1)$ time.
 - Lead to **underflow** or **overflow**.

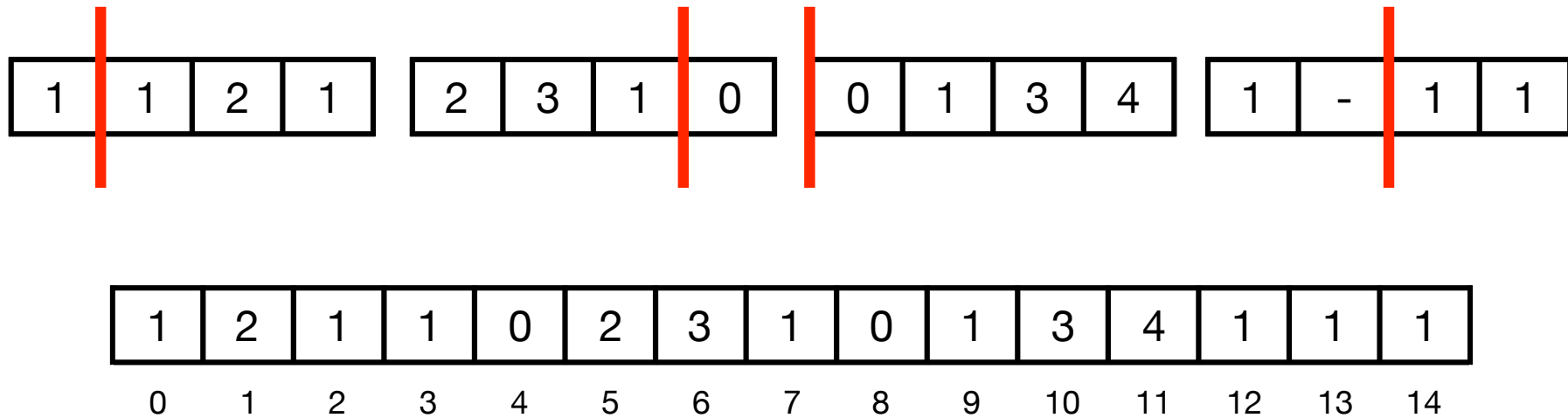
Dynamic Arrays



- 2-level rotated arrays.

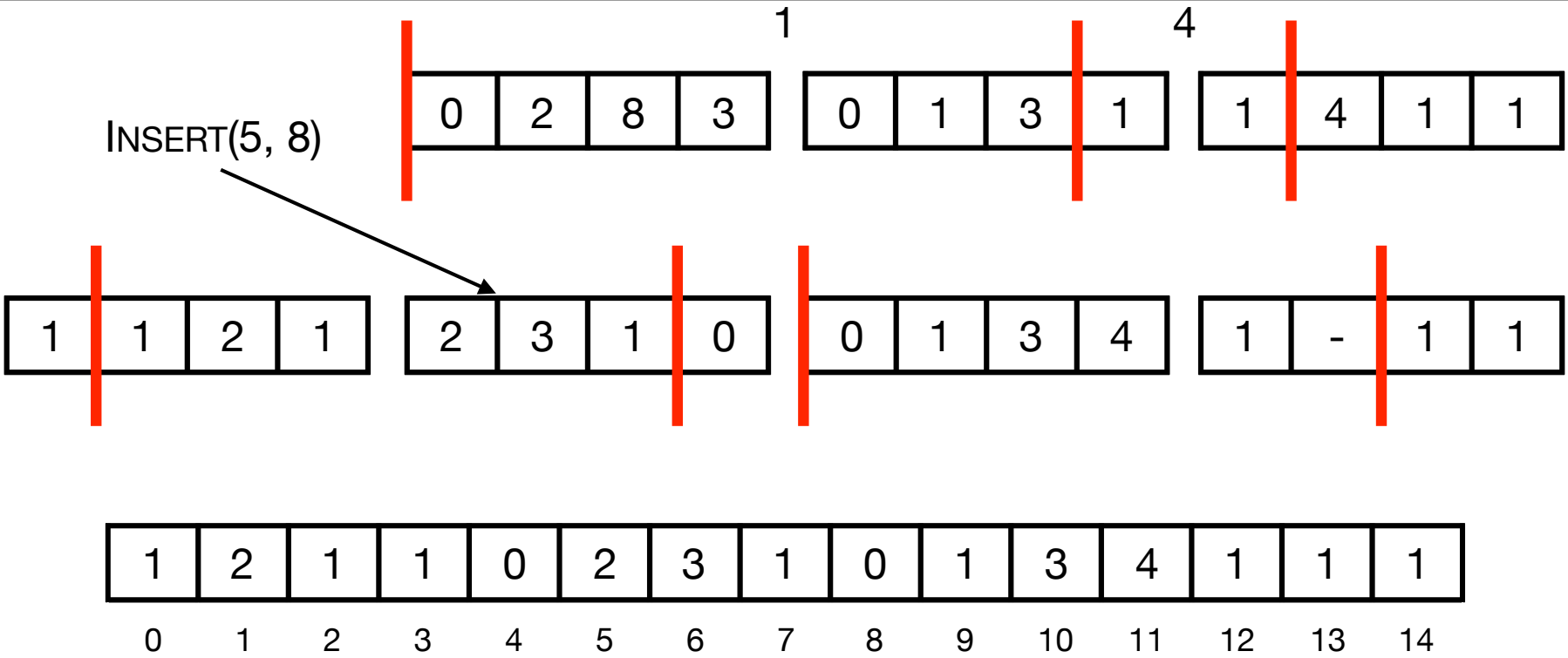
- Store \sqrt{n} rotated arrays $R_0, \dots, R_{\sqrt{n}-1}$ with **capacity** \sqrt{n} (last may have smaller capacity).

Dynamic Arrays



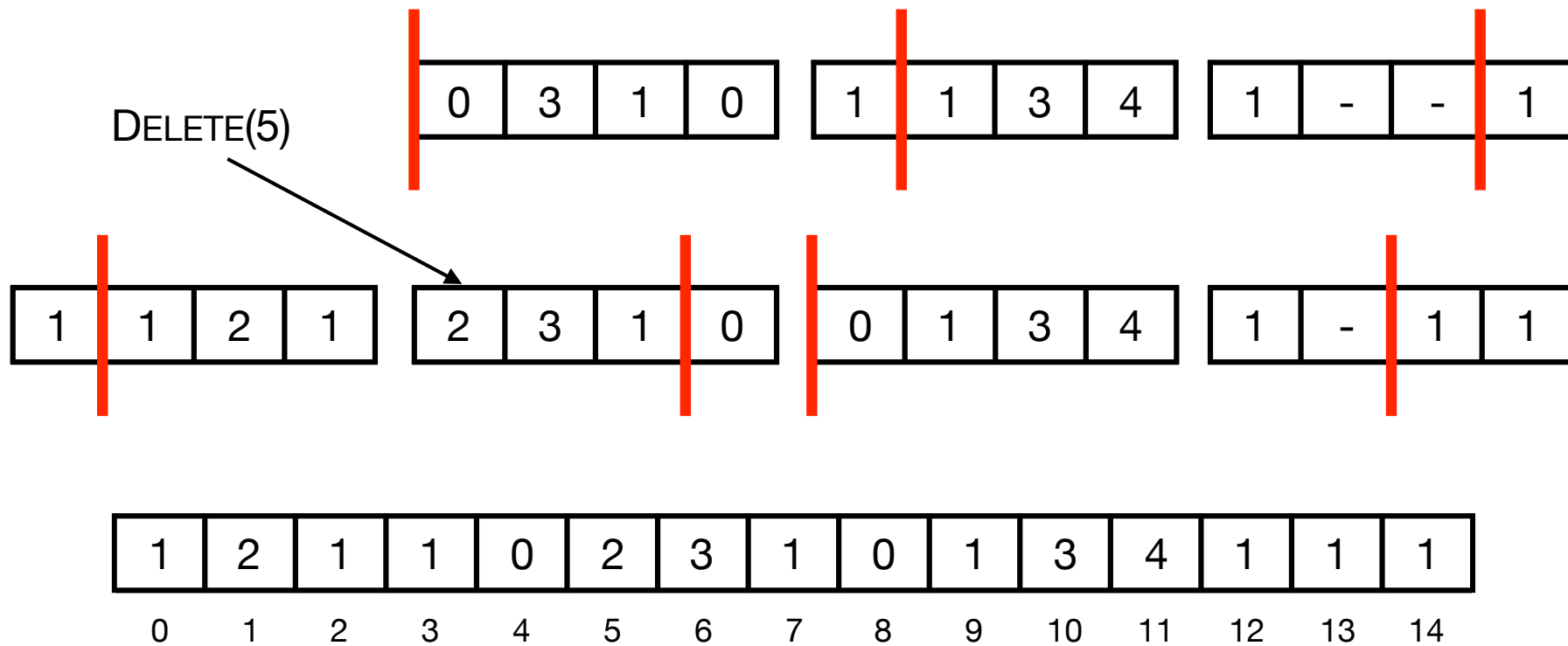
- **ACCESS.**
 - ACCESS(i): compute rotated array R_j and index k corresponding to i . Return $R_j[k]$.
- **Time.** $O(1)$

Dynamic 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. $O(\sqrt{n})$

Dynamic 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
explicit array	$O(1)$	$O(n)$	$O(n)$	$O(n)$
balanced binary tree	$O(\log n)$	$O(\log n)$	$O(\log n)$	$O(n)$
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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