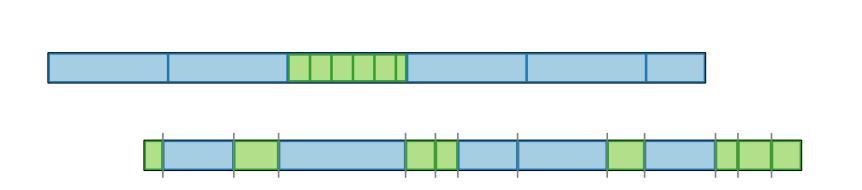


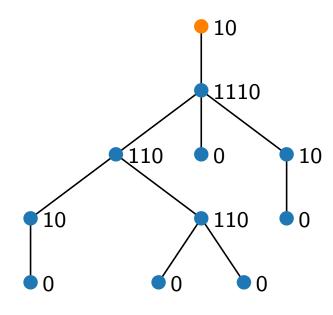
# Advanced Algorithms

#### Succinct Data Structures

Indexable Dictionaries and Trees

Johannes Zink · WS23/24





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- store,
- organize, and
- **manage** data.

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- What do we represent?
- How much space is required?
- Dynamic or static?
- Which operations are defined?
- How fast are they?

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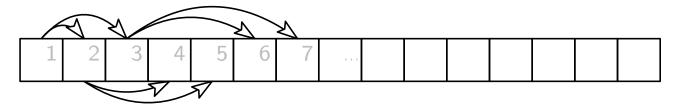
Examples!

- arrays to represent lists
  - but why not linked lists?

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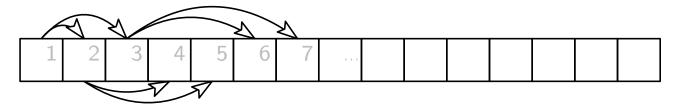
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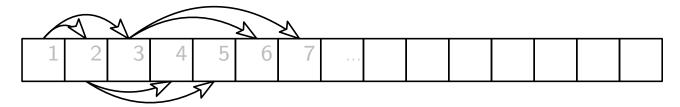
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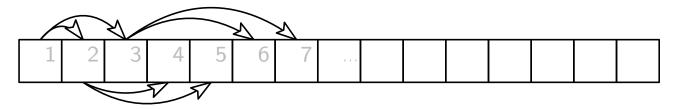
$${\tt leftChild}(i) = 2i \\ {\tt rightChild}(i) = {\tt parent}(i) = {\tt rightChild}(i)$$

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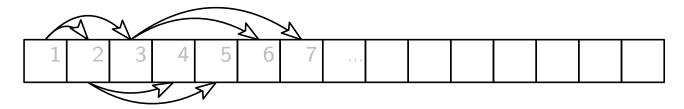
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$$leftChild(i) = 2i$$
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And unbalanced trees?

Represent a subset  $S \subseteq \{1, 2, ..., n\}$  and support the following operations in O(1) time:

- lacksquare member(i) returns if  $i \in S$
- ightharpoonup rank(i) = number of elements in S that are less or equal to i
- $\blacksquare$  select(j) = j-th element in S
- predecessor(i)
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How many bits of space do we need to distinguish them?

$$\log 2^n = n$$
 bits

Represent S with a bit vector b of length n where

$$b[i] = \begin{cases} 1 & \text{if } i \in S \\ 0 & \text{otherwise} \end{cases}$$

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$$select(5) = 9$$

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 $rank(9) =$ 

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$$select(5) = 9$$

$$rank(9) = 5$$

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$$select(5) = 9$$

$$rank(9) = 5 = rank(12)$$

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- $ightharpoonup {
  m rank}(i) = \# {
  m 1s} {
  m at} {
  m or before position} {
  m is} {
  m number of} {
  m of}$
- $\blacksquare$  select(j) = position of j-th 1 bit

$$S = \{3, 4, 6, 8, 9, 14\}$$
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 $\mathtt{member}(i)$  can trivially be answered in O(1) time (assuming that we can access any entry in constant time)

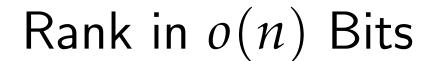
Exercise: Use these methods to  $\Rightarrow$  answer predecessor(i) and successor(i) in O(1) time.

$$select(5) = 9$$
 $rank(9) = 5 = rank(12)$ 
 $rank(15) = 6$ 

Rank in o(n) Bits

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1	
10	
$\nu$	
,,	
•	

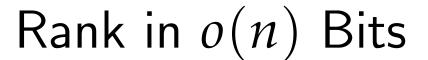
 $\log^2 n \qquad \qquad \log^2 n = (\log n)^2$ 



b

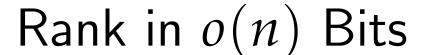
1. Split into  $(\log^2 n)$ -bit **chunks** and store cumulative rank: each needs  $\leq \log n$  bits

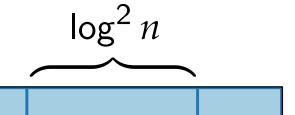
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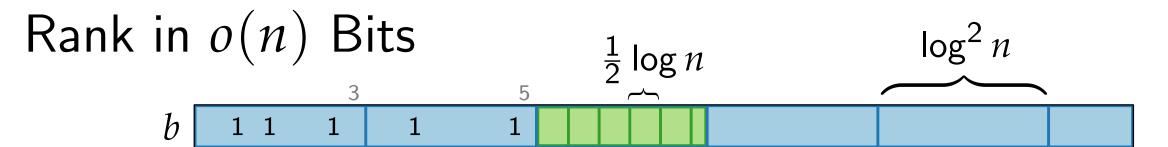


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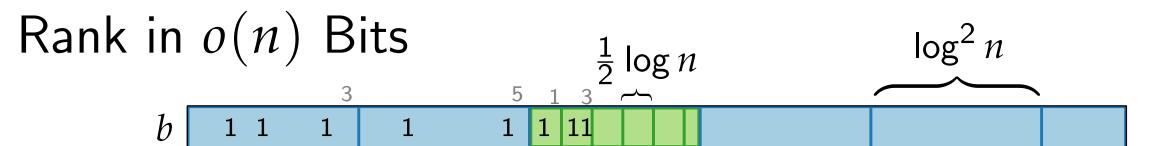
$$\Rightarrow O(\frac{n}{\log^2 n} \log n) = O(\frac{n}{\log n}) \subseteq o(n) \text{ bits}$$
# chunks rank



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2. Split **chunks** into  $(\frac{1}{2} \log n)$ -bit **subchunks** and store cumulative rank within **chunk**:

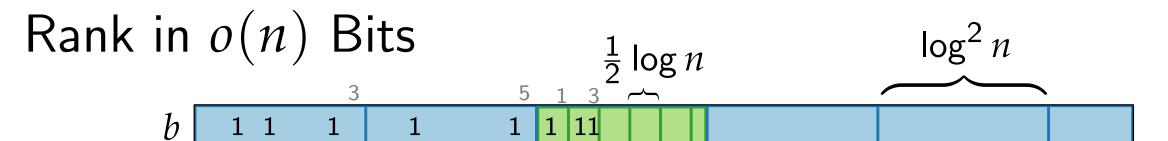


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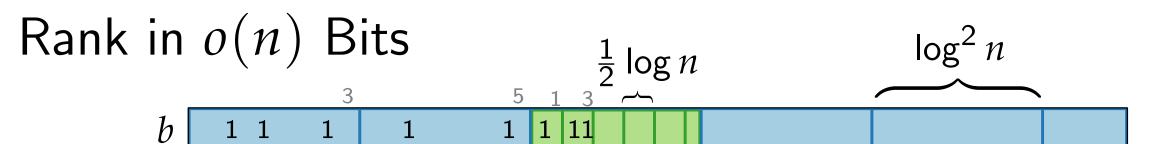


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2. Split chunks into  $(\frac{1}{2} \log n)$ -bit subchunks and store cumulative rank within chunk: each needs  $\leq \log \log^2 n = 2 \log \log n$  bits



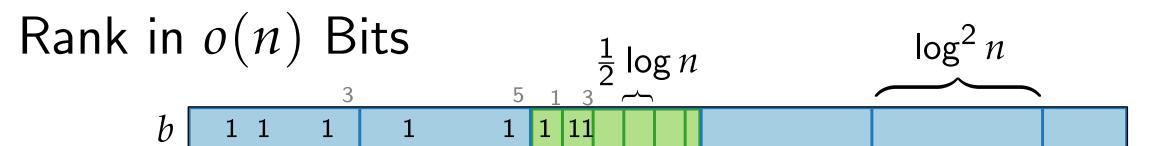
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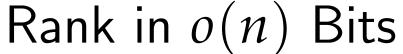


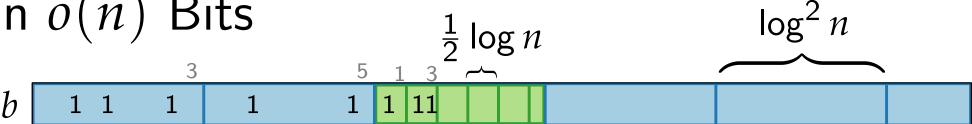
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- 2. Split **chunks** into  $(\frac{1}{2} \log n)$ -bit **subchunks** and store cumulative rank within **chunk**: each needs  $\leq \log \log^2 n = 2 \log \log n$  bits  $\Rightarrow O(\frac{n}{\log n} \log \log n) \subseteq o(n)$  bits
- 3. Use **lookup table** for bitstrings of length  $(\frac{1}{2} \log n)$ :





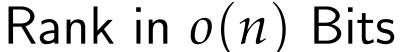
- 1. Split **Example:**  $n = 64 \Rightarrow \frac{1}{2} \log n = 3$ 
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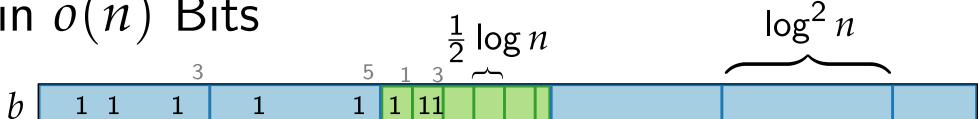
 $s \leq \log n$  bits

2. Split and

and

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and

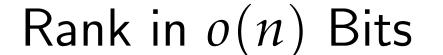
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1.	Split
	and

**Example:**  $n = 64 \Rightarrow \frac{1}{2} \log n = 3$ position

/	4	_	)	
00	0	0	0	
01	0	0	1	
10	0	1	1	

2. Split and

bitstring

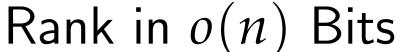
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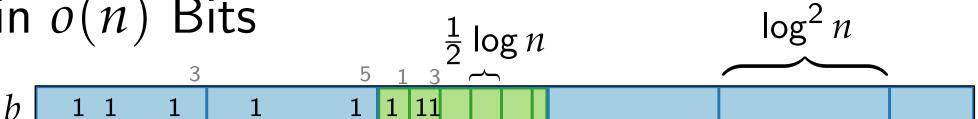
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1. Split and

2. Split

and

position  $\rightarrow$ oitstring

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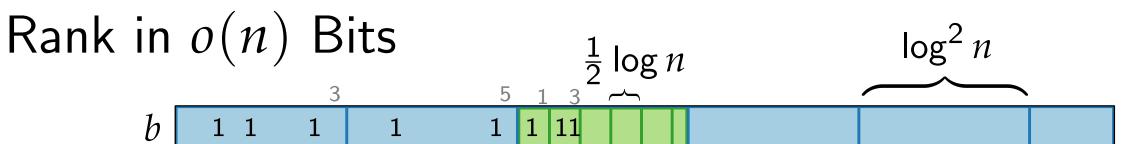
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# rows # columns rel. rank



1. Split into  $(\log^2 n)$ -bit chunks

and store cumulative rank: each needs  $\leq \log n$  bits

$$\Rightarrow O(\frac{n}{\log^2 n} \log n) = O(\frac{n}{\log n}) \subseteq o(n)$$
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- 2. Split **chunks** into  $(\frac{1}{2} \log n)$ -bit **subchunks** and store cumulative rank within **chunk**: each needs  $\leq \log \log^2 n = 2 \log \log n$  bits  $\Rightarrow O(\frac{n}{\log n} \log \log n) \subseteq o(n)$  bits
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- 4. rank(i) = rank of chunk
  - + relative rank of subchunk within chunk
  - + relative rank of element i within subchunk





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- 3. Use lookup table for bitstrings of length  $(\frac{1}{2} \log n)$ :  $2^{\frac{1}{2} \log n} = \sqrt{n}$  distinct bitstrings  $\Rightarrow O(\sqrt{n} \log n \log \log n) \subseteq o(n)$  bits
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+ relative rank of element i within subchunk

 $\Rightarrow O(1)$  time

(assume read/write numbers in O(1) time)

b

 $\log n \log \log n$  1s



1. Store indices of every  $(\log n \log \log n)$ -th 1 bit in array

 $\log n \log \log n$  1s



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$$\Rightarrow O(\frac{n}{\log n \log \log n} \log n) = O(\frac{n}{\log \log n}) \subseteq o(n) \text{ bits}$$
# groups index

 $\log n \log \log n$  1s



1. Store indices of every  $(\log n \log \log n)$ -th 1 bit in array

$$\Rightarrow O(\frac{n}{\log n \log \log n} \log n) = O(\frac{n}{\log \log n}) \subseteq o(n)$$
 bits

2. Within group of  $(\log n \log \log n)$  1 bits of length r bits:

 $\log n \log \log n$  1s



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$$\Rightarrow O(\frac{n}{\log n \log \log n} \log n) = O(\frac{n}{\log \log n}) \subseteq o(n)$$
 bits

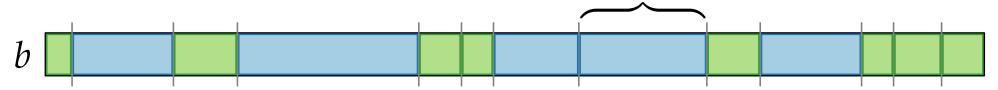
2. Within group of  $(\log n \log \log n)$  1 bits of length r bits:

if 
$$r \ge (\log n \log \log n)^2$$

then store indices of 1 bits in group in array

$$\Rightarrow O(\frac{n}{(\log n \log \log n)^2}(\log n \log \log n) \log n) \subseteq O(\frac{n}{\log \log n}) \text{ bits}$$
# groups # 1 bits index

 $\log n \log \log n$  1s



1. Store indices of every  $(\log n \log \log n)$ -th 1 bit in array

$$\Rightarrow O(\frac{n}{\log n \log \log n} \log n) = O(\frac{n}{\log \log n}) \subseteq o(n)$$
 bits

2. Within group of  $(\log n \log \log n)$  1 bits of length r bits:

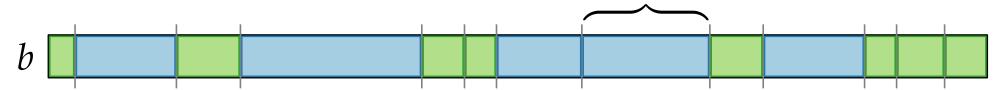
if 
$$r \ge (\log n \log \log n)^2$$

then store indices of 1 bits in group in array

$$\Rightarrow O(\frac{n}{(\log n \log \log n)^2}(\log n \log \log n) \log n) \subseteq O(\frac{n}{\log \log n})$$
 bits

else problem is reduced to bitstrings of length  $r < (\log n \log \log n)^2$ 

 $\log n \log \log n$  1s



1. Store indices of every  $(\log n \log \log n)$ -th 1 bit in array

$$\Rightarrow O(\frac{n}{\log n \log \log n} \log n) = O(\frac{n}{\log \log n}) \subseteq o(n)$$
 bits

2. Within group of  $(\log n \log \log n)$  1 bits of length r bits:

if 
$$r \ge (\log n \log \log n)^2$$

then store indices of 1 bits in group in array

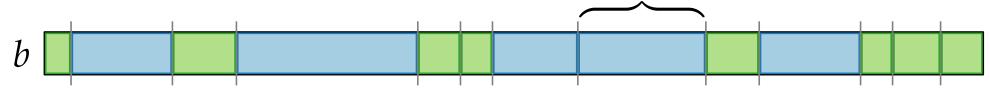
$$\Rightarrow O(\frac{n}{(\log n \log \log n)^2}(\log n \log \log n) \log n) \subseteq O(\frac{n}{\log \log n})$$
 bits

else problem is reduced to bitstrings of length  $r < (\log n \log \log n)^2$ 

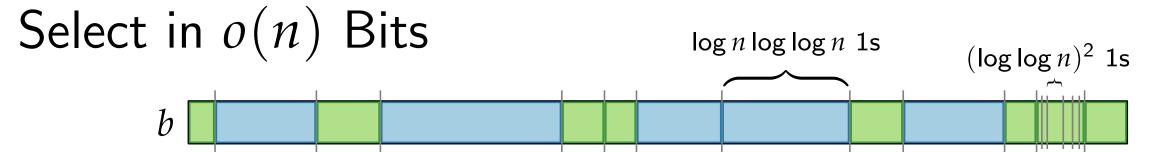
3. Repeat 1. and 2. on reduced bitstrings



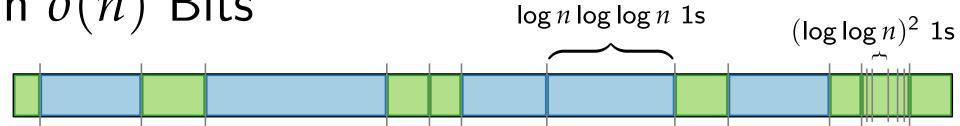
 $\log n \log \log n$  1s



3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :



- 3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :
  - 1' Store relative indices of every  $(\log \log n)^2$ -th 1 bit in array



- 3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :
  - 1' Store relative indices of every  $(\log \log n)^2$ -th 1 bit in array

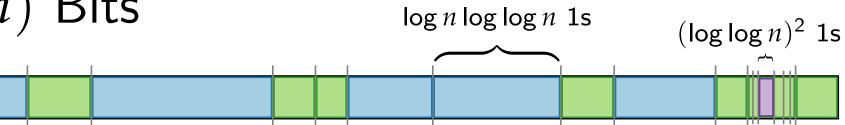
$$\Rightarrow O(\underbrace{\frac{n}{(\log\log n)^2}\log\log n}) = O(\frac{n}{\log\log n}) \text{ bits}$$
# subgroups rel. index



- 3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :
  - 1' Store relative indices of every  $(\log \log n)^2$ -th 1 bit in array

$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 bits

2' Within group of  $(\log \log n)^2$  1 bits of length r' bits:



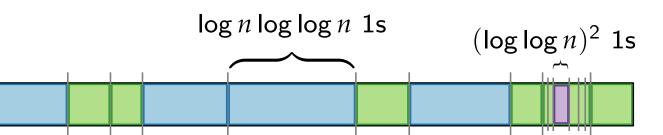
- 3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :
  - 1' Store relative indices of every  $(\log \log n)^2$ -th 1 bit in array

$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 bits

2' Within group of  $(\log \log n)^2$  1 bits of length r' bits:

if 
$$r' \ge (\log \log n)^4$$

then store relative indices of 1 bits in subgroup in array



- 3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :
  - 1' Store relative indices of every  $(\log \log n)^2$ -th 1 bit in array

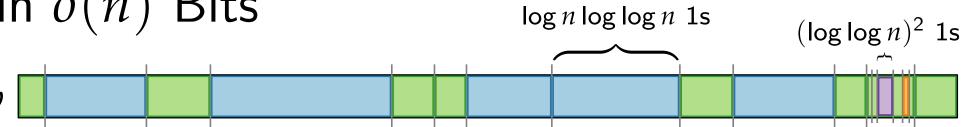
$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 bits

2' Within group of  $(\log \log n)^2$  1 bits of length r' bits:

if 
$$r' \ge (\log \log n)^4$$

then store relative indices of 1 bits in subgroup in array

$$\Rightarrow O(\frac{n}{(\log\log n)^4}(\log\log n)^2\log\log n) = O(\frac{n}{\log\log n}) \text{ bits}$$
# subgroups # 1 bits rel. index



- 3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :
  - 1' Store relative indices of every  $(\log \log n)^2$ -th 1 bit in array

$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 bits

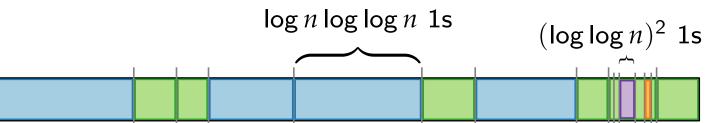
2' Within group of  $(\log \log n)^2$  1 bits of length r' bits:

if 
$$r' \ge (\log \log n)^4$$

then store relative indices of 1 bits in subgroup in array

$$\Rightarrow O(\frac{n}{(\log \log n)^4}(\log \log n)^2 \log \log n) = O(\frac{n}{\log \log n})$$
 bits

else problem is reduced to bitstrings of length  $r' < (\log \log n)^4$ 



- 3. Repeat 1. and 2. on reduced bitstrings  $(r < (\log n \log \log n)^2)$ :
  - 1' Store relative indices of every  $(\log \log n)^2$ -th 1 bit in array

$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 bits

2' Within group of  $(\log \log n)^2$  1 bits of length r' bits:

if 
$$r' \ge (\log \log n)^4$$

then store relative indices of 1 bits in subgroup in array

$$\Rightarrow O(\frac{n}{(\log \log n)^4}(\log \log n)^2 \log \log n) = O(\frac{n}{\log \log n})$$
 bits

else problem is reduced to bitstrings of length  $r' < (\log \log n)^4$ 

4. Use lookup table for bitstrings of length  $r' \leq (\log \log n)^4$ :

 $\log n \log \log n \text{ 1s} \qquad (\log \log n)^2 \text{ 1s}$ 

3. Repea

1' Store

2' With if r'

then

<b>Example:</b> $n = 10 \Rightarrow (\log \log n)^2 \approx 3$	$\frac{1}{1} \log \log n$ ) <sup>2</sup> ).
<b>Example:</b> $n = 10 \Rightarrow (\log \log n)^2 \approx 3$ $\Rightarrow r' < (\log \log n)^4 \approx 9$	$n \log \log n$

1 bit in array

$$\log \log n = O(\frac{n}{\log \log n})$$
 bits

r' bits:

up in array

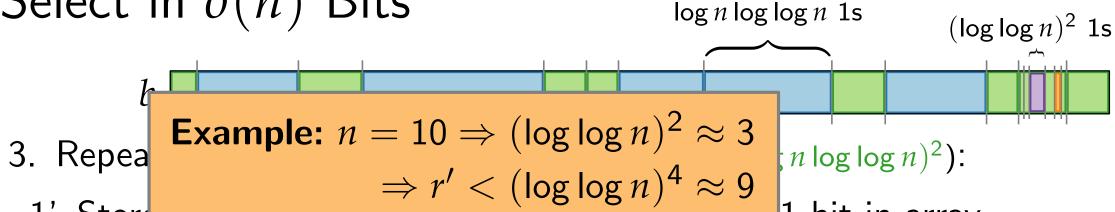
$$(n)^2 \log \log n = O(\frac{n}{\log \log n})$$
 bits

else problem is reduced to bitstrings of length  $r' < (\log \log n)^4$ 

4. Use lookup 'table for bitstrings of length  $r' \leq (\log \log n)^4$ :

select

oitstring



1' Store

2' With

then

	select  o	1	2	3
ing	00000111 00001011 00001101	6 5 5	7 7 6	8000
bitstring	: 11001000 11010000 11100000	1 1 1	2 2 2	543

1 bit in array

 $\frac{1}{(n)^2}\log\log n) = O(\frac{n}{\log\log n})$  bits

r' bits:

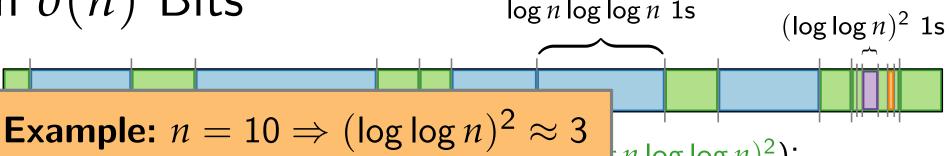
up in array

 $(n)^2 \log \log n) = O(\frac{n}{\log \log n})$  bits

else problem/is reduced to bitstrings of length  $r' < (\log \log n)^4$ 

4. Use lookup 'table for bitstrings of length  $r' \leq (\log \log n)^4$ :

$$\underbrace{2^{(\log\log n)^4}}_{\text{# rows}} \in O(2^{\frac{1}{2}\log n}) = O(\sqrt{n}); \quad \underbrace{(\log\log n)^2}_{\text{# columns}} \in O(\log n)$$



- 3. Repea
  - 1' Store

2' With if r'

then

$-10 \rightarrow (\log \log n) \sim 3$	$\frac{1}{2}$
$\Rightarrow r' < (\log \log n)^4 \approx 9$	$n \log \log n$
$\rightarrow r < (\log \log n) \sim 9$	

1 bit in array

 $\frac{1}{(n)^2} \log \log n = O(\frac{n}{\log \log n})$  bits

r' bits:

up in array

 $(n)^2 \log \log n = O(\frac{n}{\log \log n})$  bits

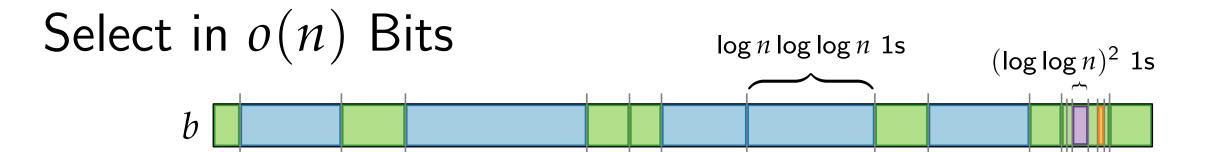
else problem is reduced to bitstrings of length  $r' < (\log \log n)^4$ 

4. Use lookup table for bitstrings of length  $r' \leq (\log \log n)^4$ :

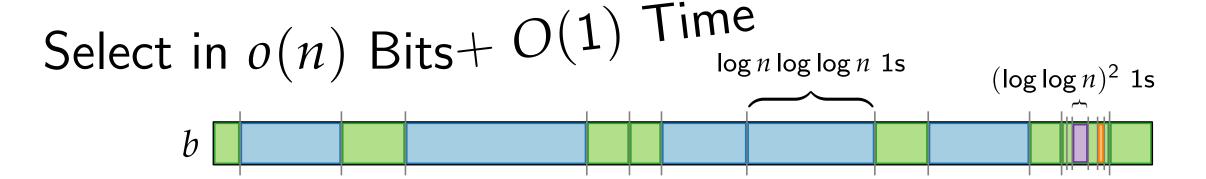
select

bitstring

$$2^{(\log\log n)^4} \in O(2^{\frac{1}{2}\log n}) = O(\sqrt{n}); \quad (\log\log n)^2 \in O(\log n) \Rightarrow O(\sqrt{n}\log \log\log n) = o(n) \text{ bits}$$
# rows # columns rel. index

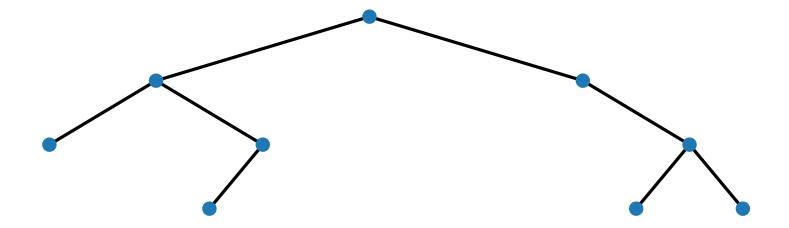


- 4. select(j) = select J-th group where  $J = \lfloor j/(\log n \log \log n) \rfloor$ 
  - + directly select (j-J)-th 1 bit or select J'-th subgroup where  $J' = \lfloor (j-J)/(\log\log n)^2 \rfloor$
  - + directly select  $(j-J-J^\prime)$ -th 1 bit or select it in the lookup table

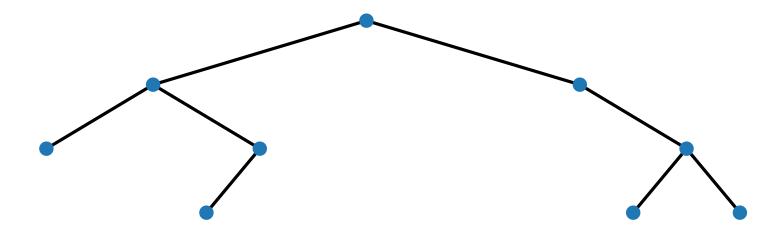


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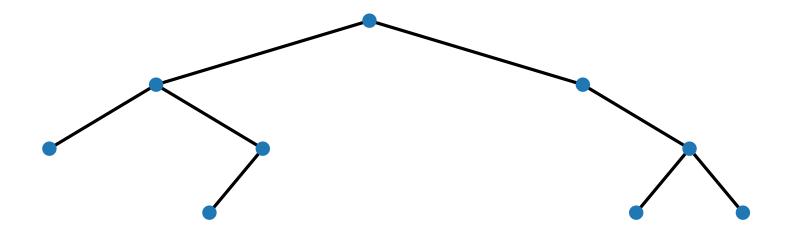
#### Succinct Representation of Binary Trees



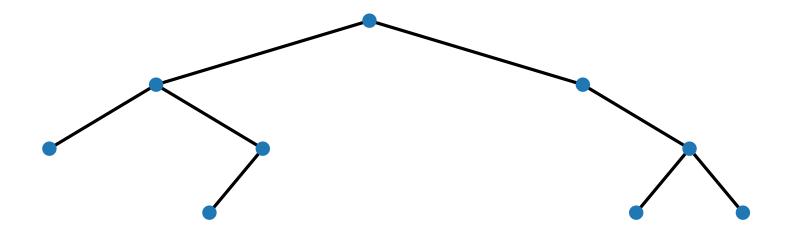
Number of binary trees on n vertices:



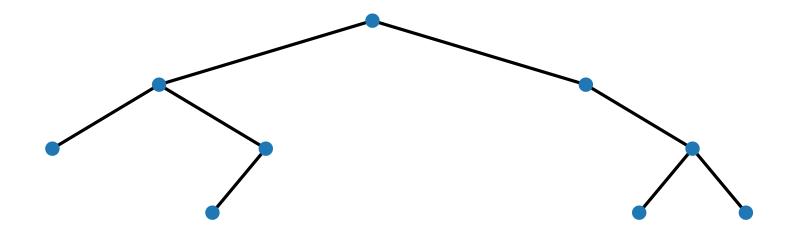
```
n=0: "empty tree" 1 possibility
```



```
n=0: "empty tree" 1 \text{ possibility} n=1: \bullet 1 \text{ possibility}
```



```
n=0: "empty tree" n=2: \bullet start with root n=1: \bullet n=1: \bullet 1 possibility
```

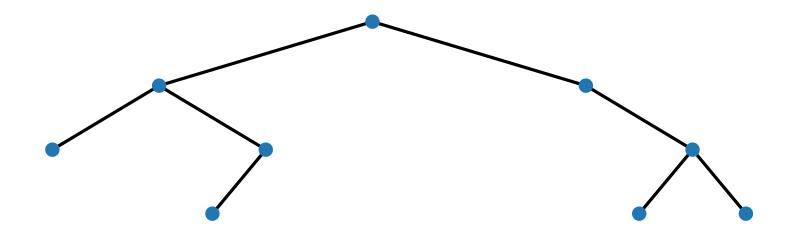


```
n=0: "empty tree" n=2: start with root

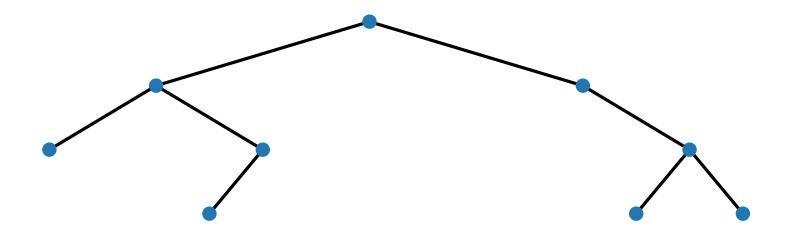
1 possibility append no child left

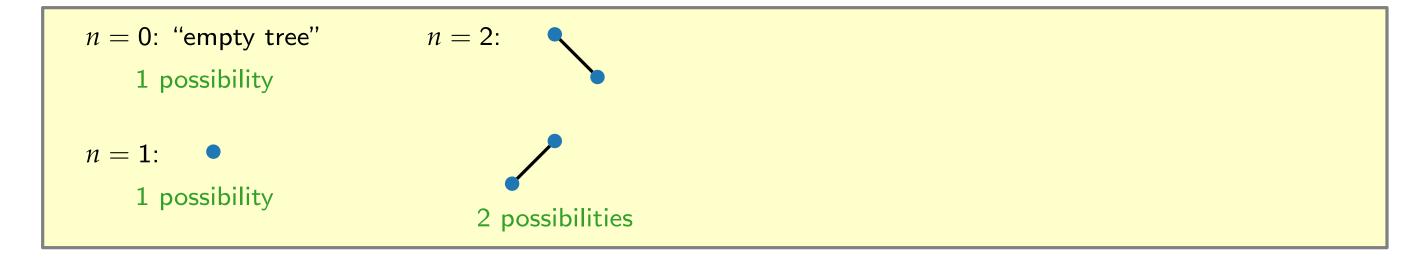
n=1: append 1 child right

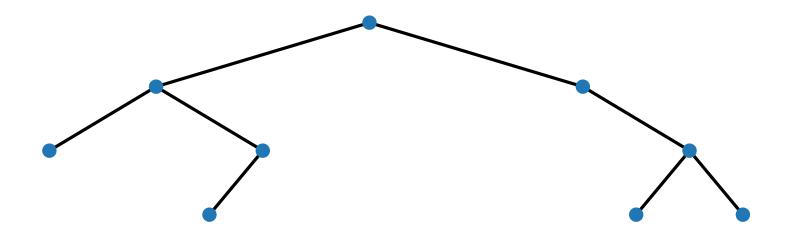
1 possibility
```

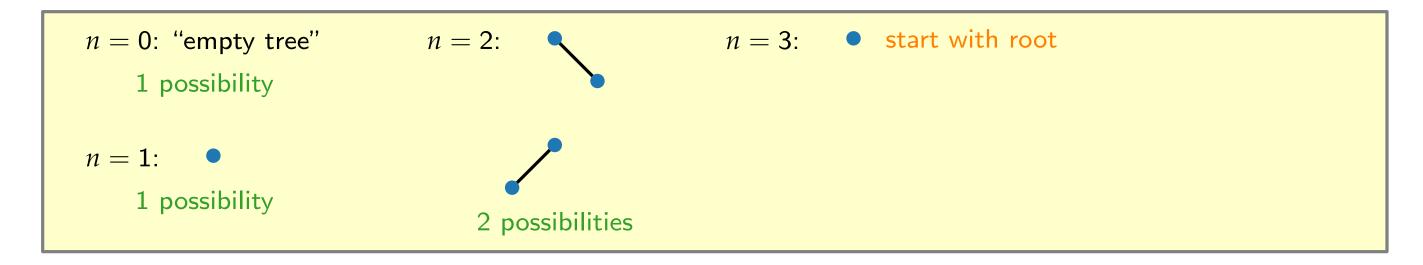


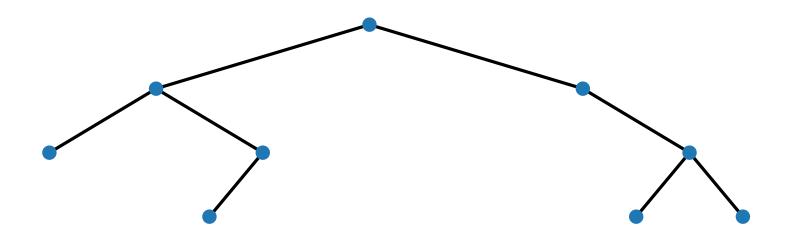






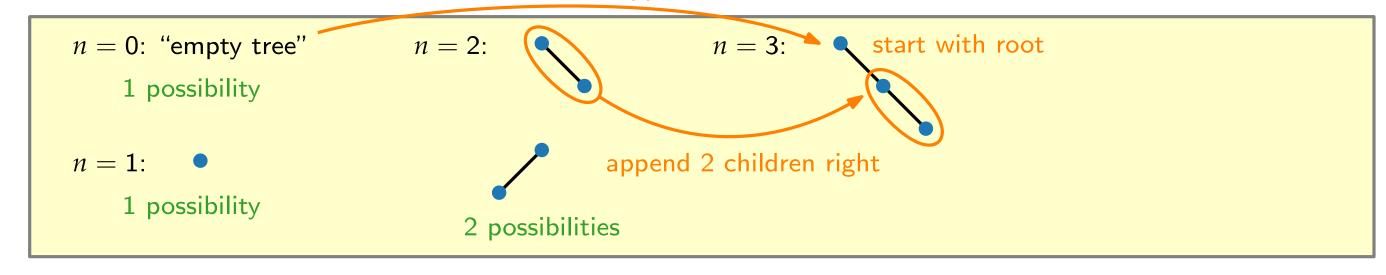


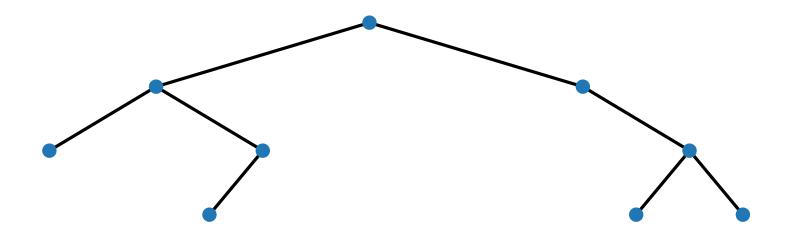




Number of binary trees on n vertices:

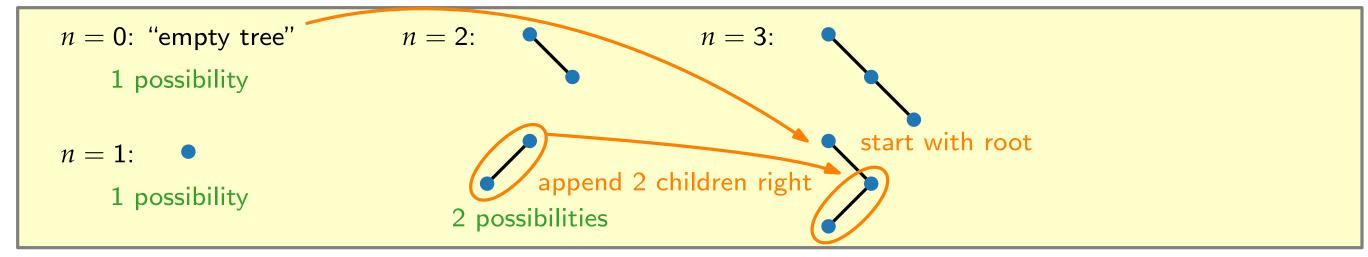
append no child left

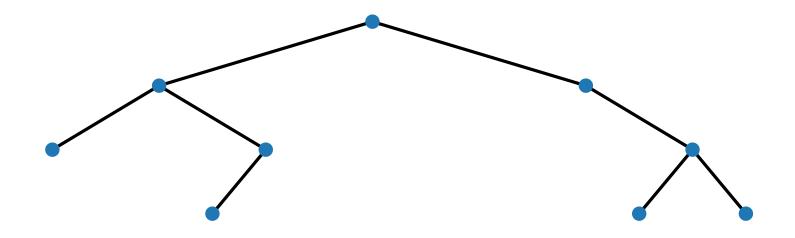


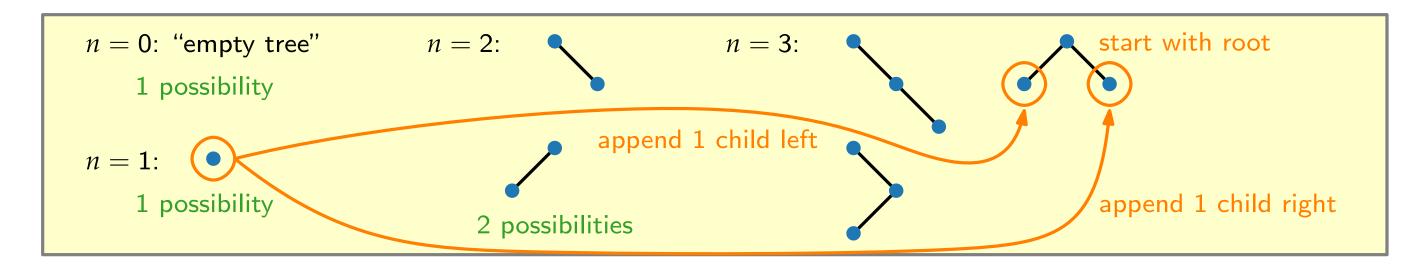


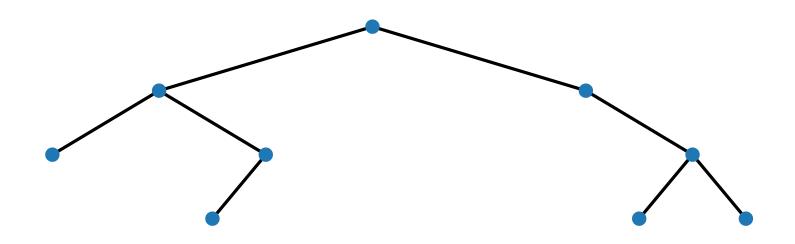
#### Number of binary trees on n vertices:

append no child left





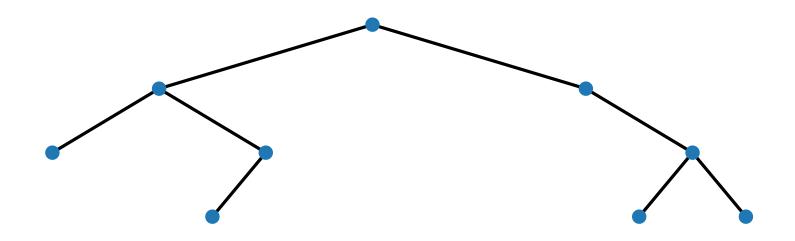




Number of binary trees on n vertices:

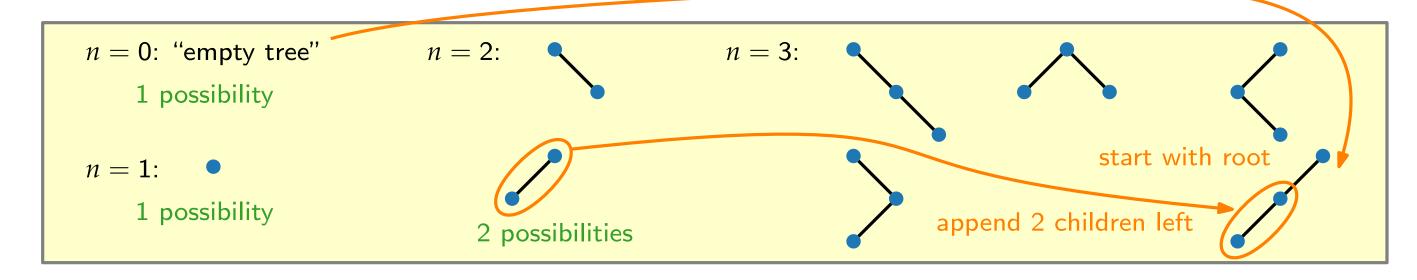
append no child right

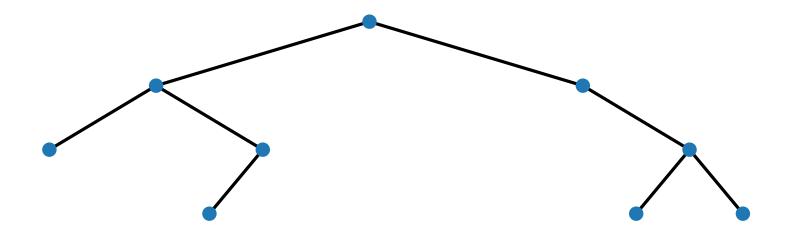
n=0: "empty tree" n=2: n=3: n=1: append 2 children left 2 possibilities

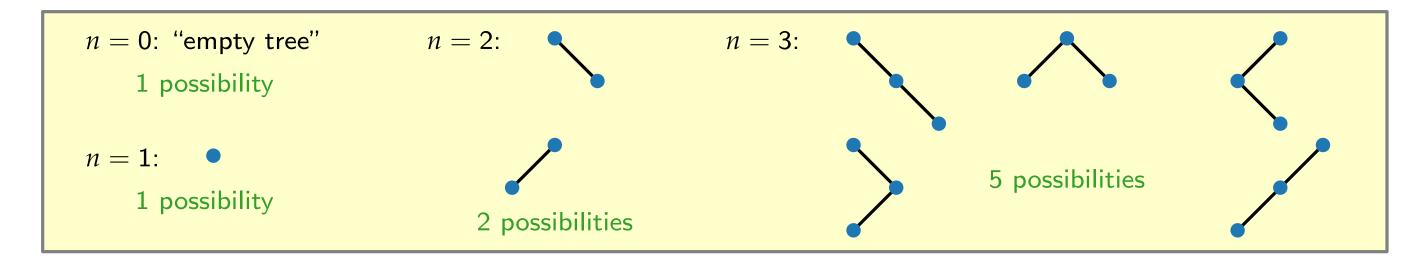


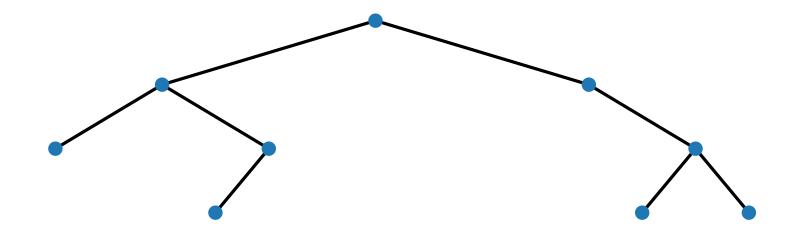
Number of binary trees on n vertices:

append no child right



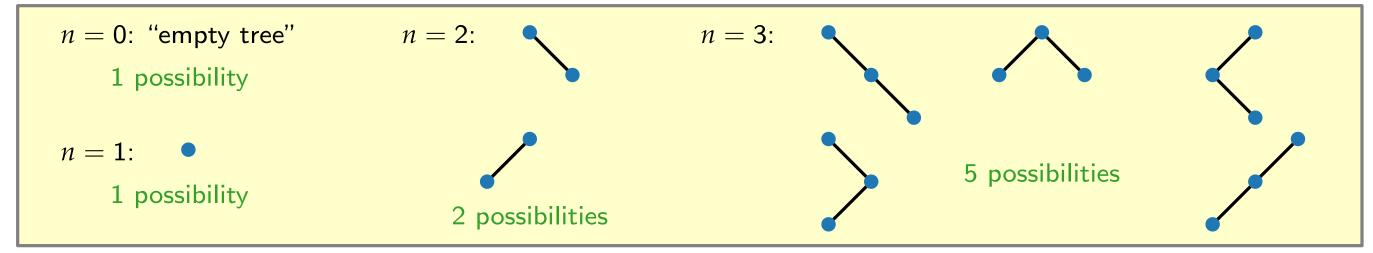


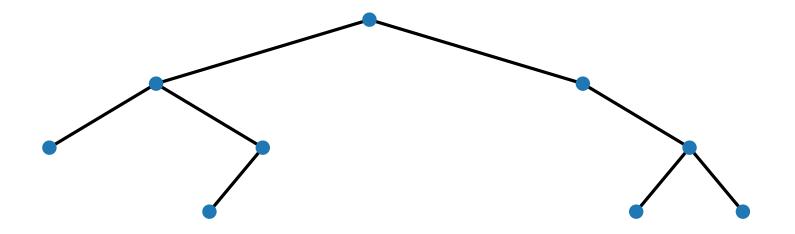




 $C_n$  is the n-th extstyle Catalan number and  $C_0=1$ 

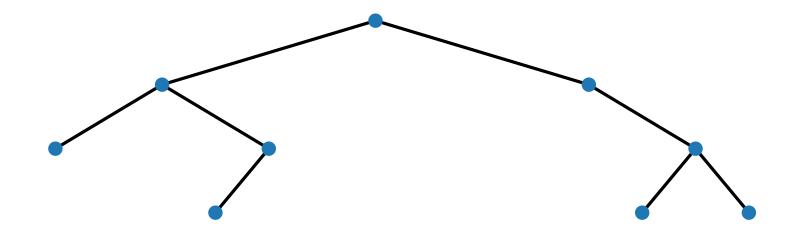
Number of binary trees on n vertices:  $C_n = \sum_{i=0}^{n-1} C_i \cdot C_{n-1-i} = \frac{(2n)!}{(n+1)! \, n!}$ 





Number of binary trees on n vertices:  $C_n = \sum_{i=0}^{n-1} C_i \cdot C_{n-1-i} = \frac{(2n)!}{(n+1)! \, n!}$ 

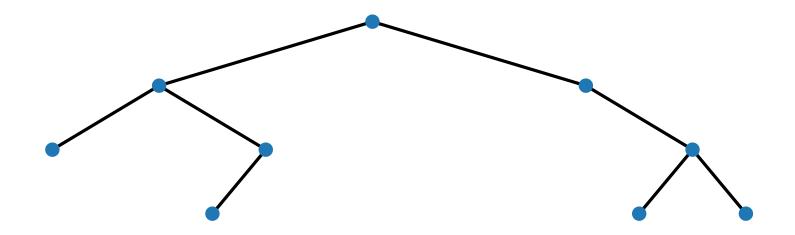
 $\log C_n = 2n + o(n)$  (by Stirling's approximation)



Number of binary trees on n vertices:  $C_n = \sum_{i=0}^{n-1} C_i \cdot C_{n-1-i} = \frac{(2n)!}{(n+1)! \, n!}$ 

 $\log C_n = 2n + o(n)$  (by Stirling's approximation)

 $\Rightarrow$  We can use 2n + o(n) bits to represent binary trees.

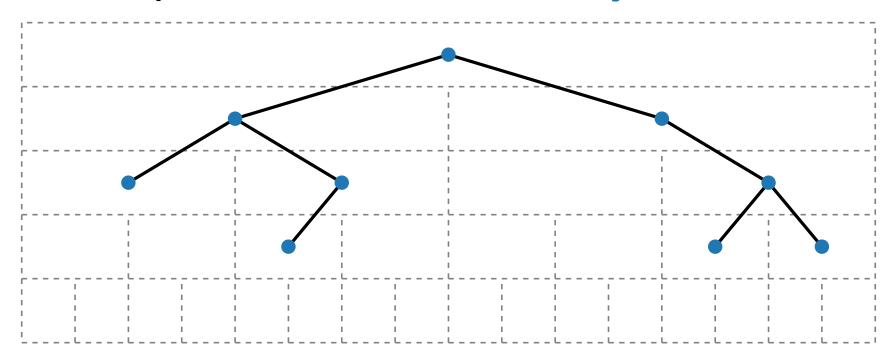


Number of binary trees on n vertices:  $C_n = \sum_{i=0}^{n-1} C_i \cdot C_{n-1-i} = \frac{(2n)!}{(n+1)! \, n!}$ 

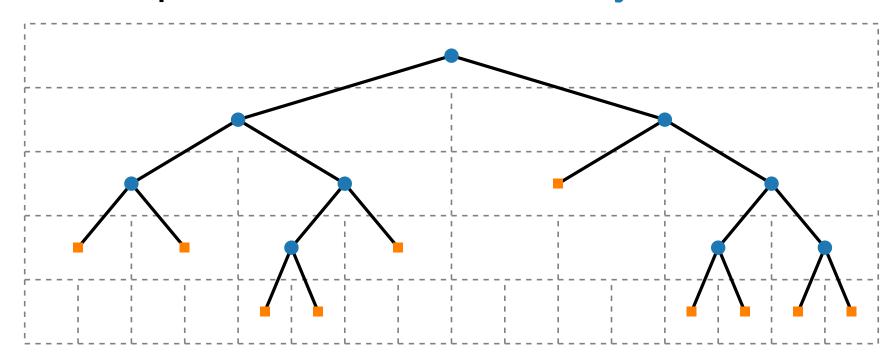
 $\log C_n = 2n + o(n)$  (by Stirling's approximation)

 $\Rightarrow$  We can use 2n + o(n) bits to represent binary trees.

**Difficulty** is when a binary tree is not full.

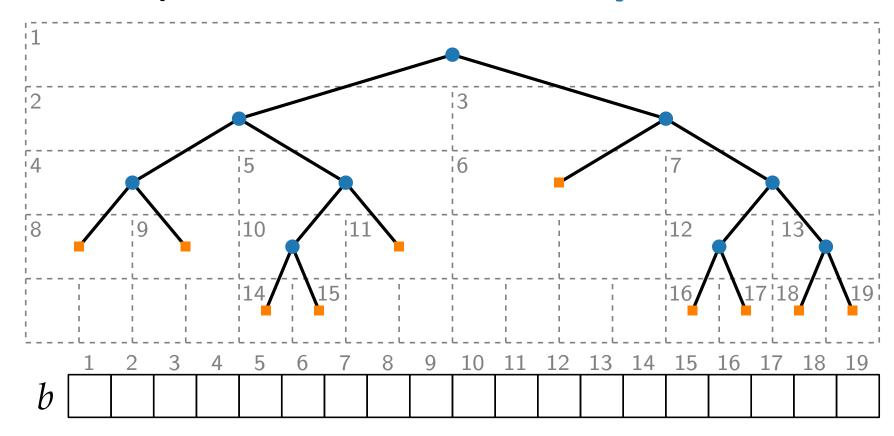


Idea.



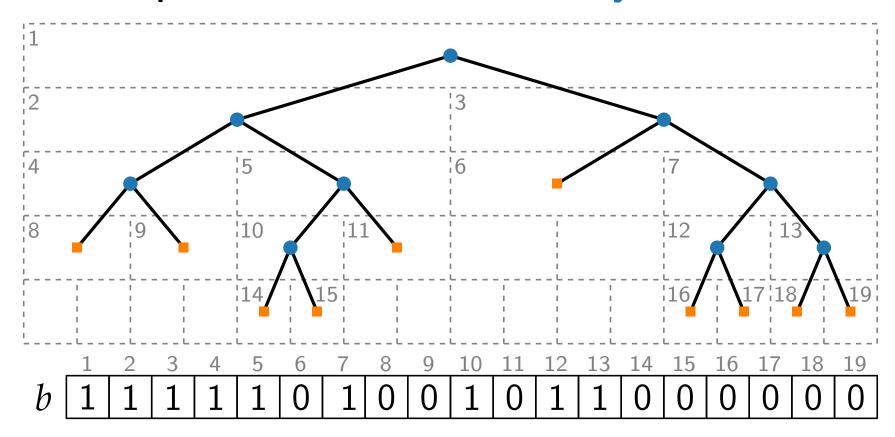
#### Idea.

Add external nodes to have out-degree 2 or 0 at every node



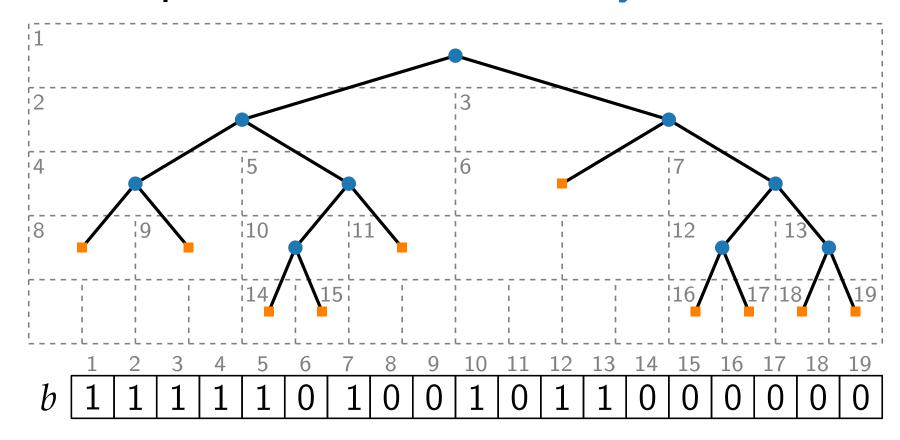
#### Idea.

Add external nodes to have out-degree 2 or 0 at every node



#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0

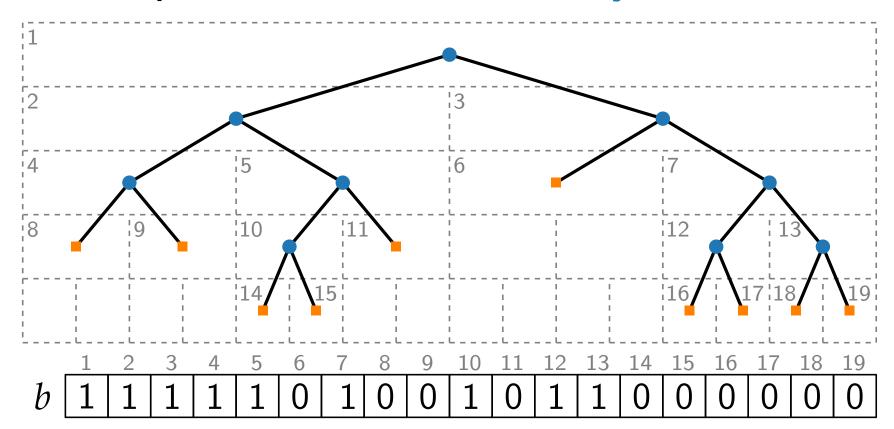


#### Size.

- $\blacksquare$  2*n* + 1 bits for *b*
- o(n) for rank and select

#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
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- Read external nodes as 0



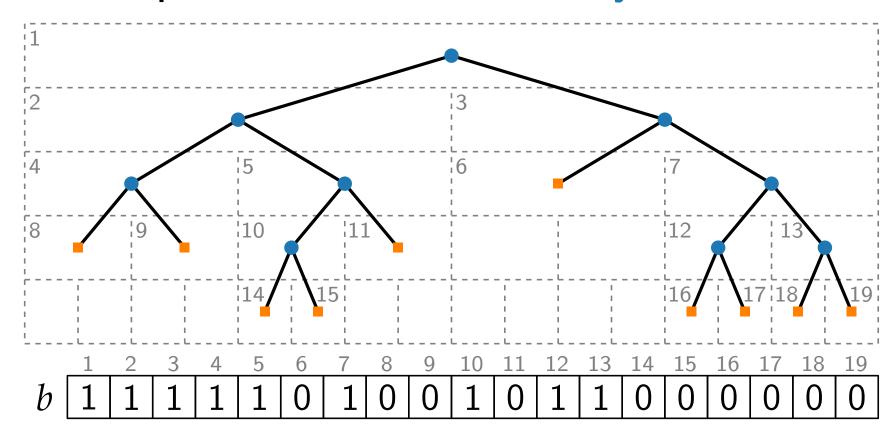
#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0

#### Operations.

- $\blacksquare$  parent(i) = ?
- $\blacksquare$  leftChild(i) = ?
- $\blacksquare$  rightChild(i) = ?

- $\blacksquare$  2*n* + 1 bits for *b*
- o(n) for rank and select



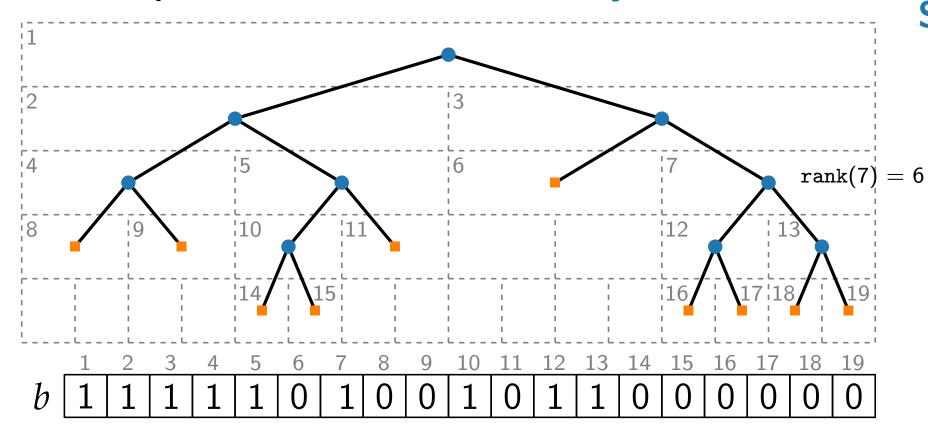
#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

#### Operations.

- $\blacksquare$  parent(i) = ?
- $\blacksquare$  leftChild(i) = ?
- $\blacksquare$  rightChild(i) = ?

- $\blacksquare$  2*n* + 1 bits for *b*
- o(n) for rank and select



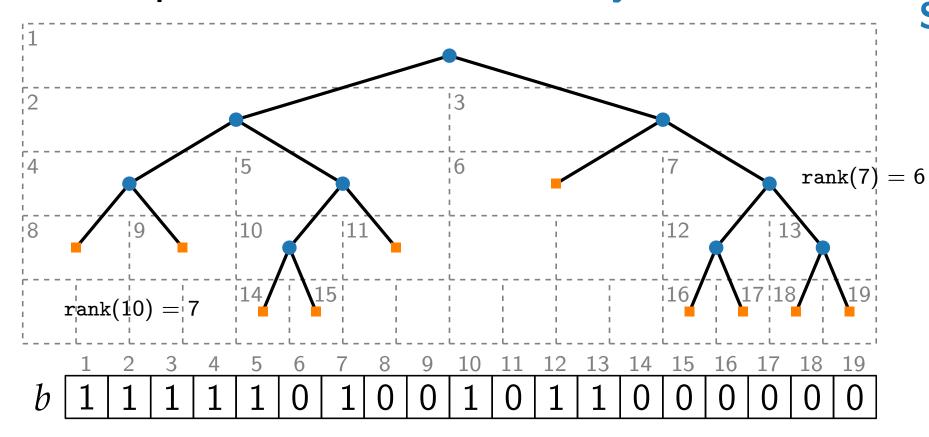
#### ldea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

#### Operations.

- $\blacksquare$  parent(i) = ?
- lacksquare leftChild(i)= ?
- $\blacksquare$  rightChild(i) = ?

- $\blacksquare$  2n+1 bits for b
- o(n) for rank and select



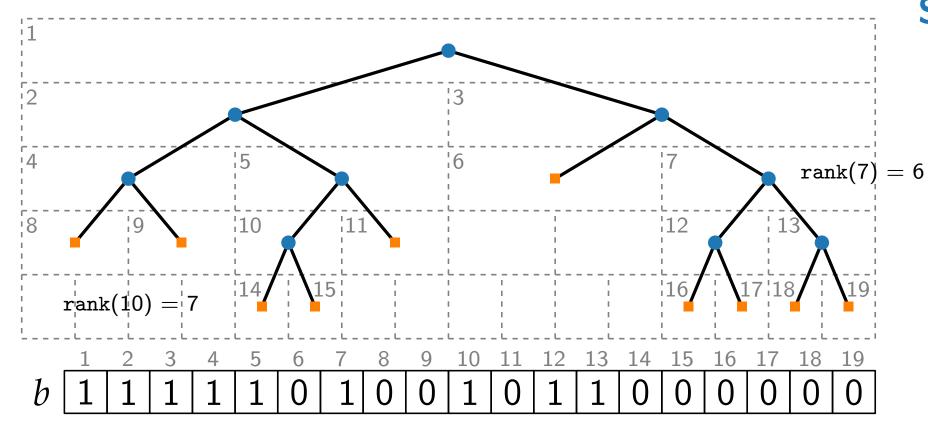
#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

#### Operations.

- $\blacksquare$  parent(i) = ?
- lacksquare leftChild(i) = ?
- $\blacksquare$  rightChild(i) = ?

- $\blacksquare$  2n+1 bits for b
- o(n) for rank and select



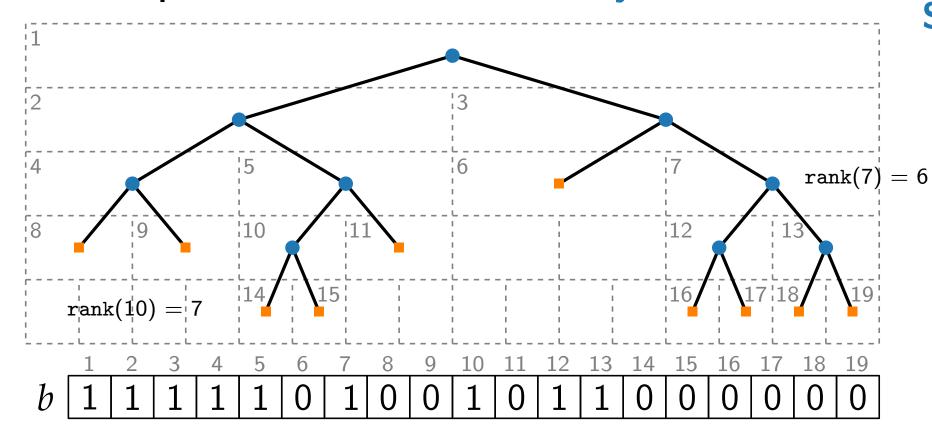
#### Size.

- $\blacksquare$  2*n* + 1 bits for *b*
- o(n) for rank and select

#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

- $\blacksquare$  parent(i) = ?
- $\blacksquare$  leftChild $(i) = 2 \operatorname{rank}(i)$
- $\blacksquare$  rightChild $(i) = 2 \operatorname{rank}(i) + 1$



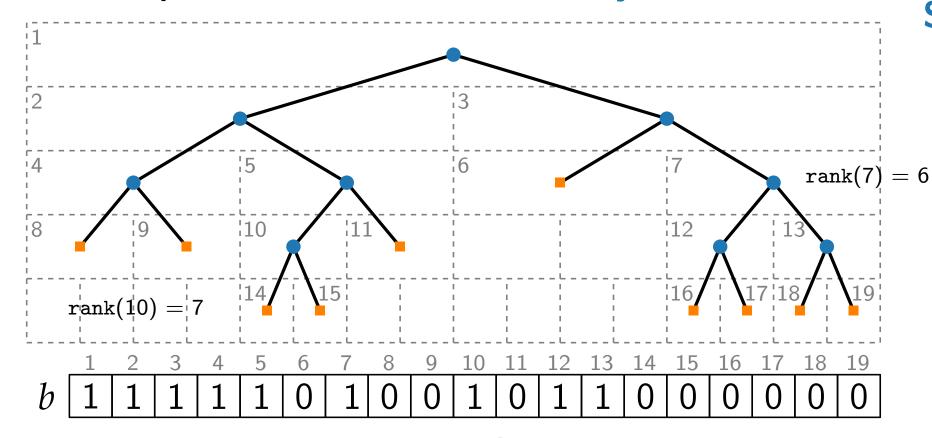
#### Size.

- $\blacksquare$  2*n* + 1 bits for *b*
- o(n) for rank and select

#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

- lacksquare parent $(i) = \operatorname{select}(\lfloor \frac{1}{2} \rfloor)$
- $\blacksquare$  leftChild $(i) = 2 \operatorname{rank}(i)$
- $\blacksquare$  rightChild $(i) = 2 \operatorname{rank}(i) + 1$



#### Size.

 $\blacksquare$  2*n* + 1 bits for *b* 

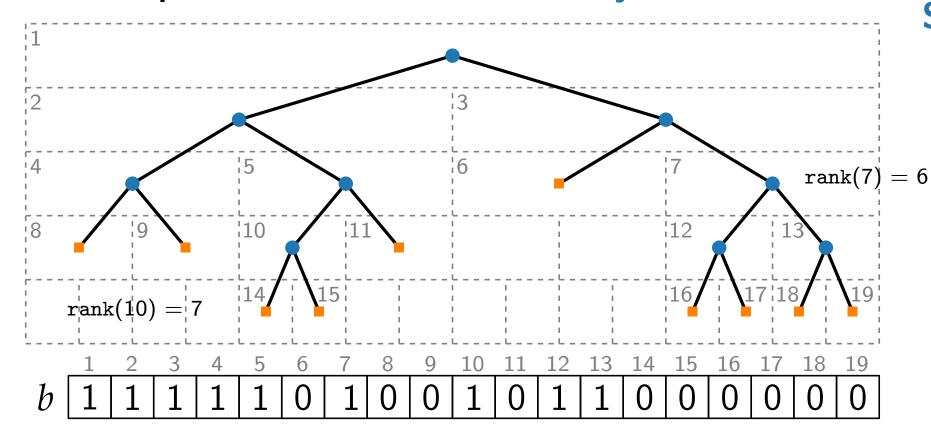
Proof is exercise.

o(n) for rank and select

#### Idea.

- Add external nodes to have out-degree 2 or 0 at every node
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

- parent $(i) = select(\lfloor \frac{1}{2} \rfloor)$
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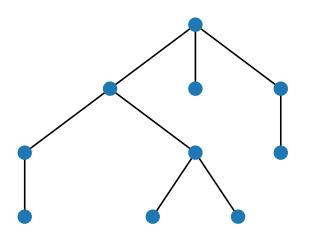
Proof is exercise.

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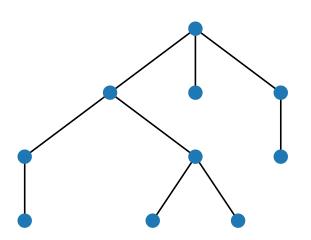
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- $\blacksquare$  rank(i) is index for array storing actual values

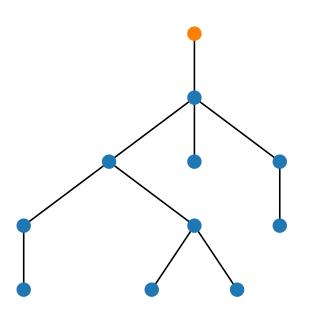


[Level Order Unary Degree Sequence]



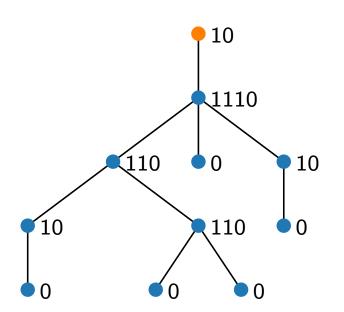
[Level Order Unary Degree Sequence]

add extra root with out-degree 1



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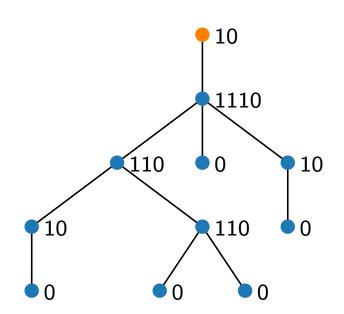


[Level Order Unary Degree Sequence]



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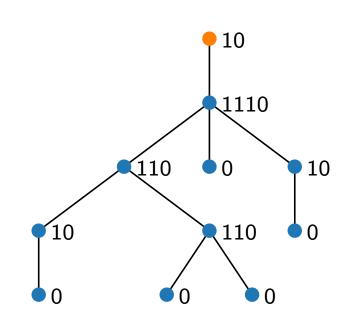


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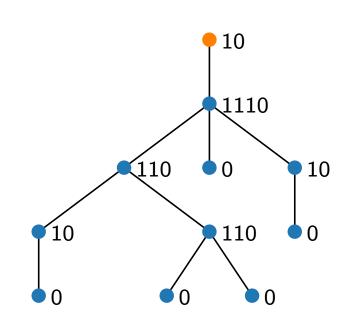


- each vertex (except root) is represented twice, namely with a 1 and with a 0
- o(n) bits for rank and select

[Level Order Unary Degree Sequence]



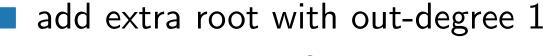
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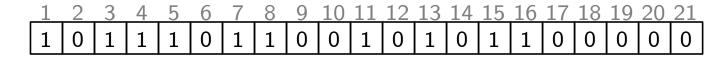
#### Size.

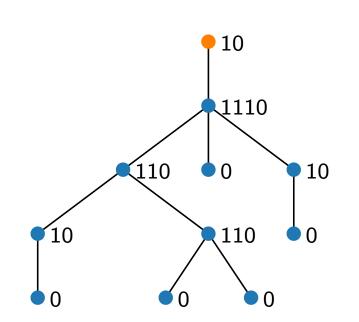
- each vertex (except root) is represented twice, namely with a 1 and with a 0  $\Rightarrow 2n + o(n)$  bits
- o(n) bits for rank and select

[Level Order Unary Degree Sequence]



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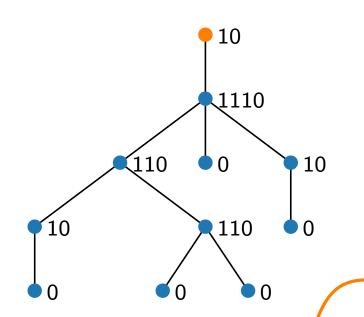


#### **Operations.**

- Let *i* be index of 1 in LOUDS sequence. This 1 represents a node (e.g. first 1 represents the root).
- ightharpoonup rank(i) is index for array storing actual values of the nodes.

[Level Order Unary Degree Sequence]

- add extra root with out-degree 1
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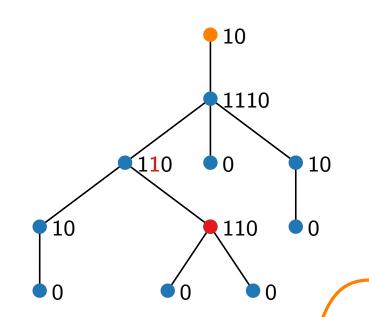


execute select(j) on the 0s instead of the 1s (as before) firstChild(i) = select $_0(\operatorname{rank}_1(i)) + 1$ 

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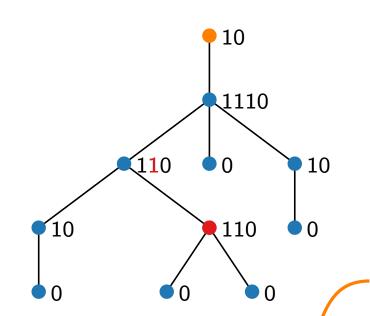


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 $\mathtt{firstChild}(\textcolor{red}{8}) = \mathtt{select}_0(\mathtt{rank}_1(\textcolor{red}{8})) + 1$ 

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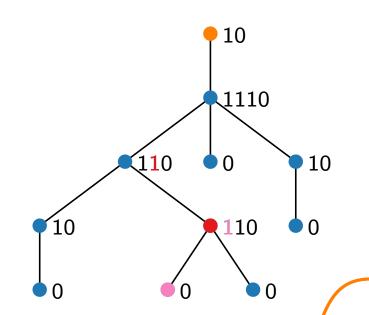


execute select(j) on the Os instead of the 1s (as before) the 1stChild(i) = select $_0(\operatorname{rank}_1(i)) + 1$ 

 $firstChild(8) = select_0(rank_1(8)) + 1$  $= select_0(6) + 1$ 

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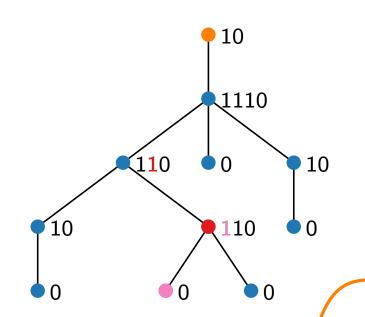


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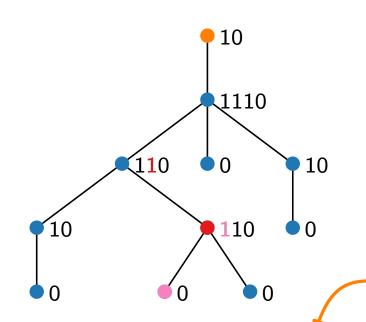
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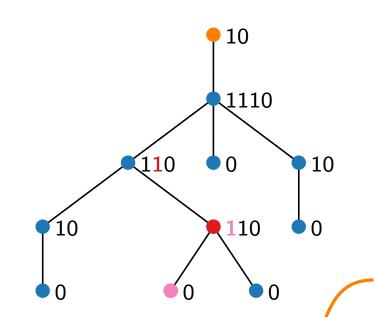
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 $\blacksquare$  parent $(i) = select_1(rank_0(i))$ 



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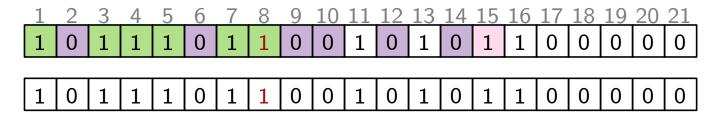
 $firstChild(8) = select_0(rank_1(8)) + 1$  $= select_0(6) + 1 = 14 + 1 = 15$ 

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Exercise: child(i, j)
with validity check

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1110

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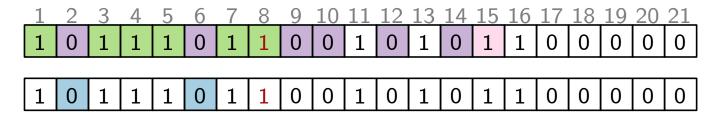
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execute rank(i) on

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1110

110

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Exercise: child(i, j)
with validity check

$$\begin{aligned} & \texttt{firstChild}(8) = \texttt{select}_0(\texttt{rank}_1(8)) + 1 & \texttt{parent}(8) = \texttt{select}_1(\texttt{rank}_0(8)) \\ &= \texttt{select}_0(6) + 1 = 14 + 1 = 15 & \texttt{select}_1(2) \end{aligned}$$

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**\1110** 

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ightharpoonup parent $(i) = select_1(rank_0(i))$ 

 $parent(8) = select_1(rank_0(8))$ = select<sub>1</sub>(2) = 3

 $\mathtt{nextSibling}(i) = i + 1$ 

Exercise: child(i, j)
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  - support fast operations

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- rank and select form the basis for many succinct representations (e.g., for specific types of trees or strings).
- There are implementations of succinct data structures being used in practice for large data sets in information retrieval, language model representation, bioinformatics, etc.

#### Literature

#### Main reference:

- Lecture 17 of Advanced Data Structures (MIT, Fall'17) by Erik Demaine
- [Jac '89] "Space efficient Static Trees and Graphs"

#### Recommendations:

■ Lecture 18 of Demaine's course on compact & succinct arrays & trees