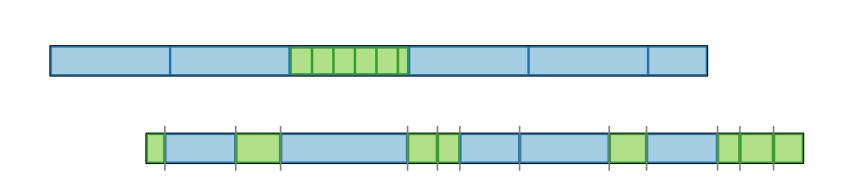


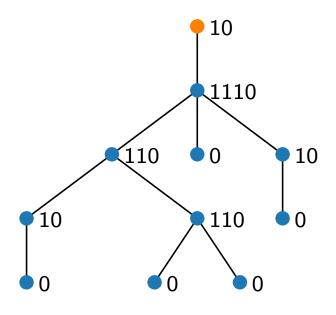
Advanced Algorithms

Succinct Data Structures

Indexable Dictionaries and Trees

Jonathan Klawitter · WS20





Data structures

A data structure is a concept to

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

As such, it is a collection of

- data values,
- their relations, and
- the operations that can applied to the data.

Data structures

A data structure is a concept to

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

As such, it is a collection of

- data values,
- their relations, and
- the operations that can applied to the data.

Remarks.

- We look at data structures as a designer/implementer (and not necessarily as a user).
- To define a data structure and to implement it are two different tasks.

Data structures

A data structure is a concept to

- store,
- organize, and
- manage data.

As such, it is a collection of

- data values,
- their relations, and
- the operations that can applied to the data.

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

Remarks.

- We look at data structures as a designer/implementer (and not necessarily as a user).
- To define a data structure and to implement it are two different tasks.

Goal.

- Use space "close" to information-theoretical minimum,
- but still support time-efficient operations.

Goal.

- Use space "close" to information-theoretical minimum,
- but still support time-efficient operations.

Let *L* be the information-theoretical lower bound to represent a class of objects.

Then a data structure, which still supports time-efficient operations, is called

Implicit, if it takes L + O(1) bits of space;

Goal.

- Use space "close" to information-theoretical minimum,
- but still support time-efficient operations.

Let L be the information-theoretical lower bound to represent a class of objects.

Then a data structure, which still supports time-efficient operations, is called

- implicit, if it takes L + O(1) bits of space;
- **succinct**, if it takes L + o(L) bits of space;

Goal.

- Use space "close" to information-theoretical minimum,
- but still support time-efficient operations.

Let L be the information-theoretical lower bound to represent a class of objects.

Then a data structure, which still supports time-efficient operations, is called

- implicit, if it takes L + O(1) bits of space;
- **succinct**, if it takes L + o(L) bits of space;
- **compact**, if it takes O(L) bits of space.

Goal.

- Use space "close" to information-theoretical minimum,
- but still support time-efficient operations.

Let L be the information-theoretical lower bound to represent a class of objects.

Then a data structure, which still supports time-efficient operations, is called

- implicit, if it takes L + O(1) bits of space;
- **succinct**, if it takes L + o(L) bits of space;
- **compact**, if it takes O(L) bits of space.

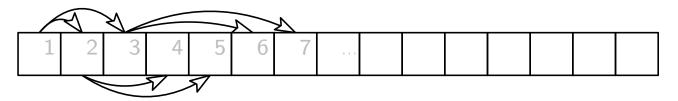
Examples!

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

- arrays to represent lists
 - but why not linked lists?
- 1-dim arrays to represent multi-dimensional arrays

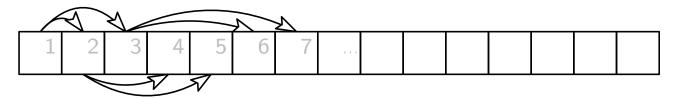
- arrays to represent lists
 - but why not linked lists?
- 1-dim arrays to represent multi-dimensional arrays
- sorted arrays to represent sorted lists
 - but why not binary search trees?

- arrays to represent lists
 - but why not linked lists?
- 1-dim arrays to represent multi-dimensional arrays
- sorted arrays to represent sorted lists
 - but why not binary search trees?
- arrays to represent complete binary trees and heaps



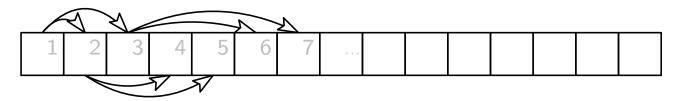
$${\tt leftChild}(i) = \\ {\tt rightChild}(i) = \\ {\tt parent}(i) = \\ {\tt rightChild}(i) = {\tt parent}(i) = {\tt parent}($$

- arrays to represent lists
 - but why not linked lists?
- 1-dim arrays to represent multi-dimensional arrays
- sorted arrays to represent sorted lists
 - but why not binary search trees?
- arrays to represent complete binary trees and heaps



$$\begin{array}{l} \texttt{leftChild}(i) = 2i \\ \texttt{rightChild}(i) = 2i + 1 \end{array} \quad \texttt{parent}(i) = \lfloor \frac{i}{2} \rfloor \end{array}$$

- arrays to represent lists
 - but why not linked lists?
- 1-dim arrays to represent multi-dimensional arrays
- sorted arrays to represent sorted lists
 - but why not binary search trees?
- arrays to represent complete binary trees and heaps



$$leftChild(i) = 2i$$
 $rightChild(i) = 2i + 1$

$$\mathtt{parent}(i) = \lfloor rac{i}{2}
floor$$

And unbalanced trees?

Represent a subset $S \subset [n]$ and support O(1)-time operations:

- lacktriangle member(i) returns if $i \in S$
- ightharpoonup rank(i) = # 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit
- predecessor and successor can be answered using rank
 and select

Represent a subset $S \subset [n]$ and support O(1)-time operations:

- lacksquare member(i) returns if $i \in S$
- ightharpoonup rank(i) = # 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit
- predecessor and successor can be answered using rank
 and select

How many different subsets of [n] are there?

How many bits of space do we need to distinguish them?

Represent a subset $S \subset [n]$ and support O(1)-time operations:

- lacksquare member(i) returns if $i \in S$
- ightharpoonup rank(i) = # 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit
- predecessor and successor can be answered using rank and select

How many different subsets of [n] are there? 2^n

How many bits of space do we need to distinguish them?

Represent a subset $S \subset [n]$ and support O(1)-time operations:

- lacksquare member(i) returns if $i \in S$
- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit
- predecessor and successor can be answered using rank and select

How many different subsets of [n] are there? 2^n

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

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

Represent S with a bit vector b of length n where

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

$$S = \{3, 4, 6, 8, 9, 14\}$$
 where $n = 15$

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

$$S = \{3, 4, 6, 8, 9, 14\}$$
 where $n = 15$

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

$$select(5) =$$

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

$$select(5) = 9$$

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

$$select(5) = 9$$

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

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

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

Represent S with a bit vector b of length n where

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

- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

$$select(5) = 9$$

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

$$rank(15) = 6$$

Represent S with a bit vector b of length n where

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

plus o(n)-space structures to answer in O(1) time

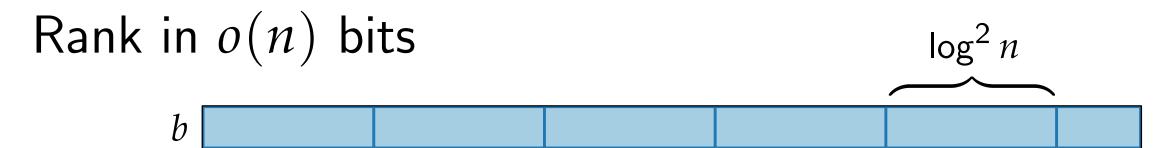
- ightharpoonup rank(i)=# 1's at or before position i
- \blacksquare select(j) = position of jth 1 bit

⇒ Exercise: Use them to answer predecessor and successor.

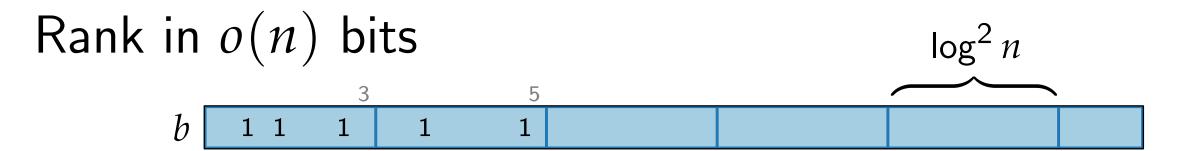
$$select(5) = 9$$
 $rank(9) = 5 = rank(12)$
 $rank(15) = 6$

Rank in o(n) bits

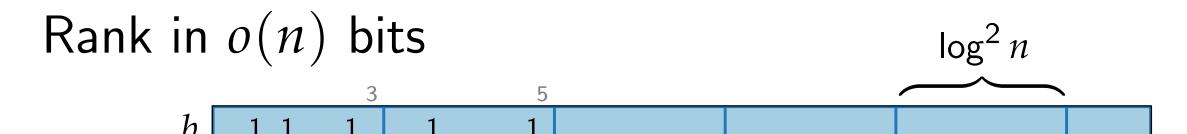
b



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



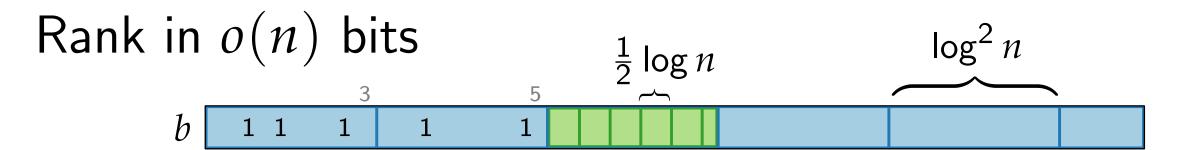
1. Split into $(\log^2 n)$ -bit **chunks** and store cumulative rank: each $\log n$ bits



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

and store cumulative rank: each log n bits

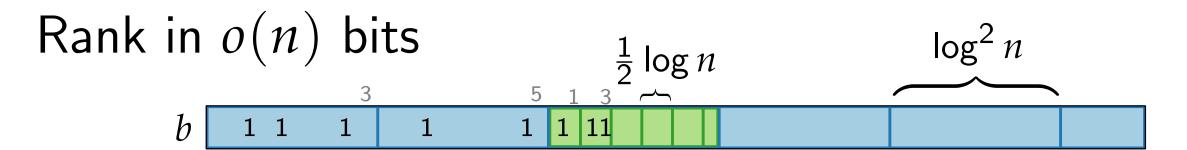
$$\Rightarrow O(\frac{n}{\log^2 n} \log n) = O(\frac{n}{\log n}) \subseteq o(n) \text{ bits}$$
chunks rank



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

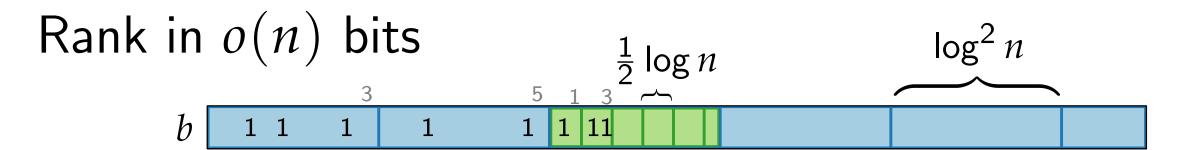
$$\Rightarrow O(\frac{n}{\log^2 n} \log n) = O(\frac{n}{\log n}) \subseteq o(n)$$
 bits

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



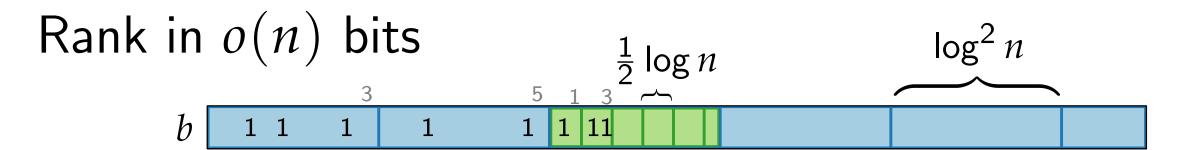
$$\Rightarrow O(\frac{n}{\log^2 n} \log n) = O(\frac{n}{\log n}) \subseteq o(n)$$
 bits

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



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

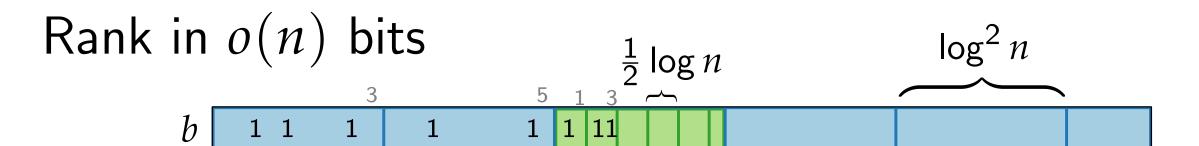
2. Split **chunks** into $(\frac{1}{2} \log n)$ -bit **subchunks** and store cumulative rank within **chunk**: $2 \log \log n$ bits



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

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

$$\Rightarrow O(\frac{n}{\log n} \log \log n) \subseteq o(n) \text{ bits}$$
subch. rel. rank



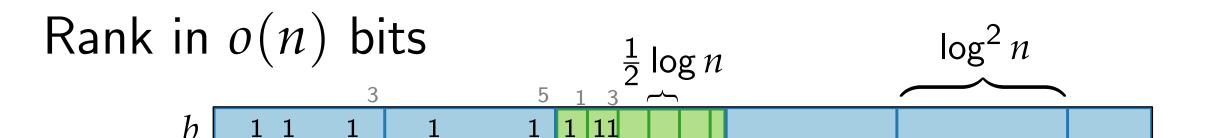
$$\Rightarrow O(\frac{n}{\log^2 n} \log n) = O(\frac{n}{\log n}) \subseteq o(n)$$
 bits

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

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

3. Use lookup table for bitstrings of length $(\frac{1}{2} \log n)$

$$\Rightarrow O(\sqrt{n} \log n \log \log n) \subseteq o(n)$$
 bits bitstring query i answer



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

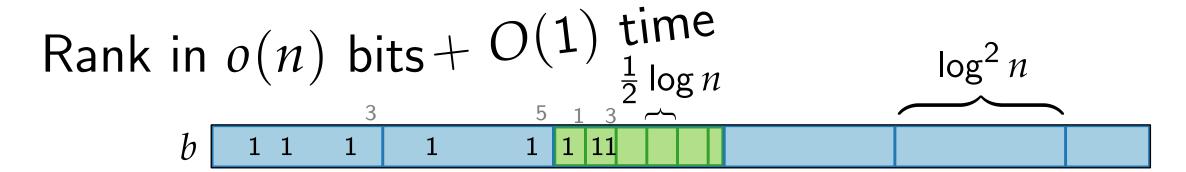
2. Split **chunks** into $(\frac{1}{2} \log n)$ -bit **subchunks** and store cumulative rank within **chunk**: $2 \log \log n$ bits

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

3. Use lookup table for bitstrings of length $(\frac{1}{2} \log n)$

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

- 4. rank = rank of chunk
 - + relative rank of **subchunk** within **chunk**
 - + relative rank of element within **subchunk**



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

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

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

 $\Rightarrow O(1)$ time

3. Use **lookup table** for bitstrings of length $(\frac{1}{2} \log n)$

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

+ relative rank of subchunk within chunk

4. rank = rank of chunk

+ relative rank of element within subchunk

b

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



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

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



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

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

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

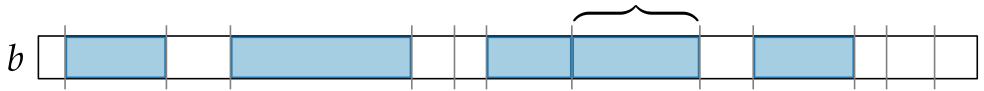


1. Store indices of every $(\log n \log \log n)$ th 1 bit in array

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

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

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



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

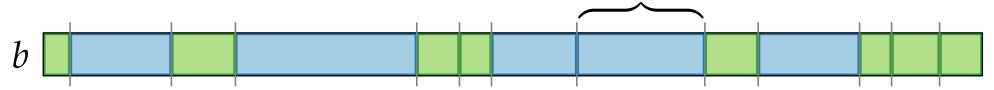
2. Within group of $(\log n \log \log n)$ 1 bits of length r bits:

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

then store indices of 1 bits in group in array

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

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



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:

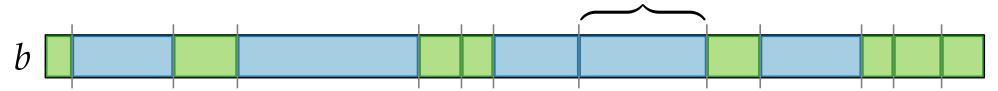
if
$$r \ge (\log n \log \log n)^2$$

then store indices of 1 bits in group in array

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

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

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



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

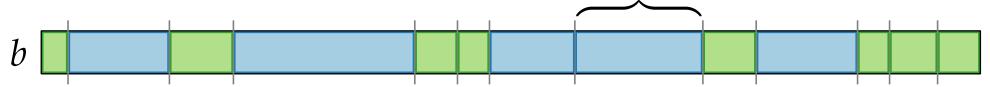
$$\Rightarrow O(\frac{n}{(\log n \log \log n)^2}(\log n \log \log n) \log n) \subseteq O(\frac{n}{\log \log n})$$

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

3. Repeat 1. and 2. on reduced bitstrings



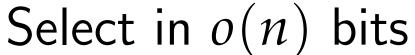
 $\log n \log \log n$ 1's

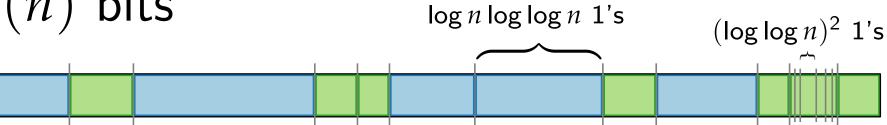


3. Repeat 1. and 2. on reduced bitstrings $(r < (\log n \log \log n)^2)$:



- 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

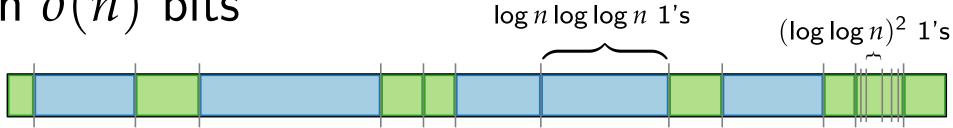




- 3. Repeat 1. and 2. on reduced bitstrings $(r < (\log n \log \log n)^2)$:
 - 1' Store relative indices of every $(\log \log n)^2$ th 1 bit in array

$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 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 $(\log \log n)^2$ th 1 bits of length r' bits:



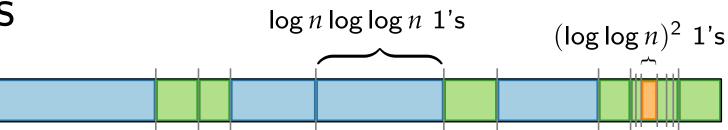
- 3. Repeat 1. and 2. on reduced bitstrings $(r < (\log n \log \log n)^2)$:
 - 1' Store relative indices of every $(\log \log n)^2$ th 1 bit in array

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

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

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

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

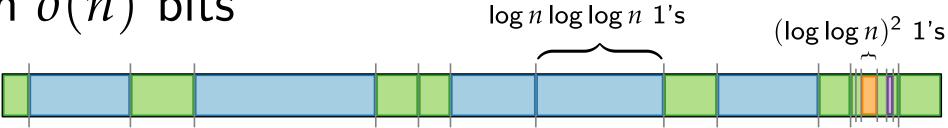
$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 bits

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

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

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

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

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

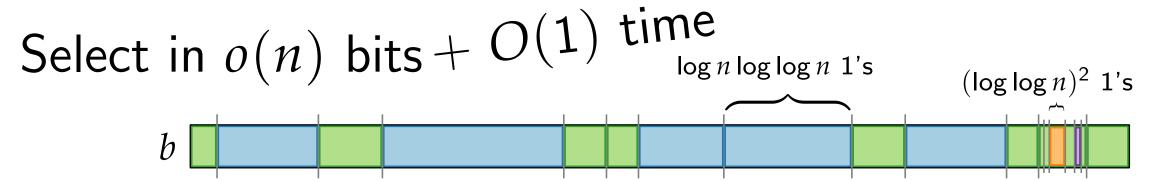
if
$$r' \ge (\log \log n)^4$$

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 $r' \leq (\log \log n)^4 \leq \frac{1}{2} \log n$ $\Rightarrow O(\sqrt{n} \log n \log \log n) = o(n) \text{ bitstring query } j \text{ answer}$



- 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 $(\log \log n)^2$ th 1 bits of length r' bits:

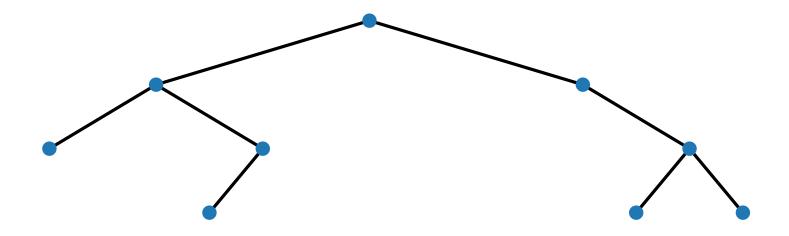
if
$$r' \geq (\log \log n)^4$$

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

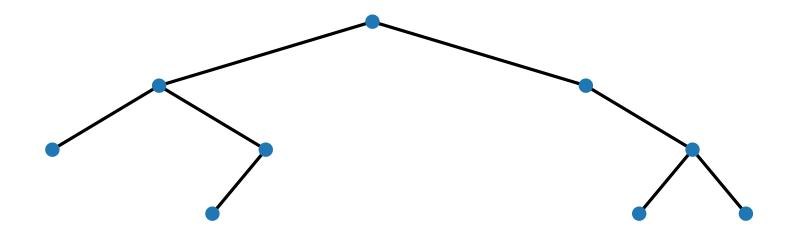
else problem is reduced to bitstrings of length $r' < (\log \log n)^4$

4. Use lookup table for bitstrings of length $r' \leq (\log \log n)^4 \leq \frac{1}{2} \log n$ $\Rightarrow O(\sqrt{n} \log n \log \log n) = o(n) \text{ bitstring query } j \text{ answer}$



Number of binary trees on n vertices: $C_n = \frac{1}{n+1} {2n \choose n}$

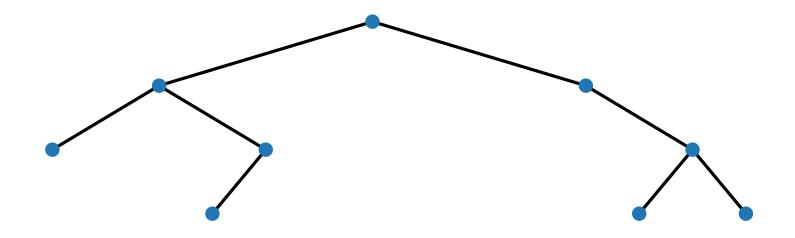
 $\log C_n = 2n + o(n)$ (by Stirling's approximation)



Number of binary trees on n vertices: $C_n = \frac{1}{n+1} {2n \choose n}$

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

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

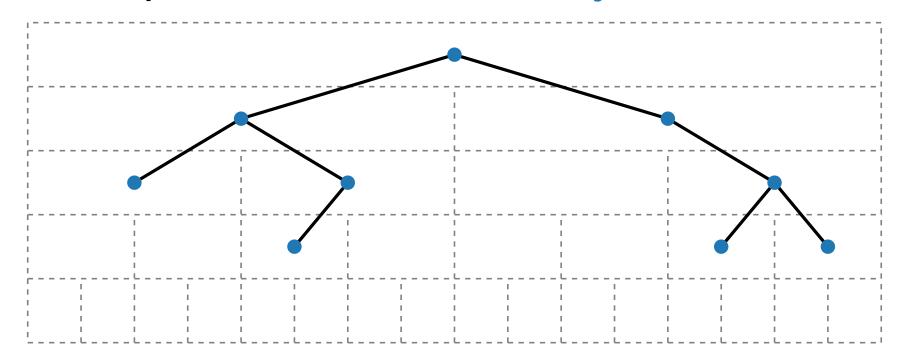


Number of binary trees on n vertices: $C_n = \frac{1}{n+1} {2n \choose n}$

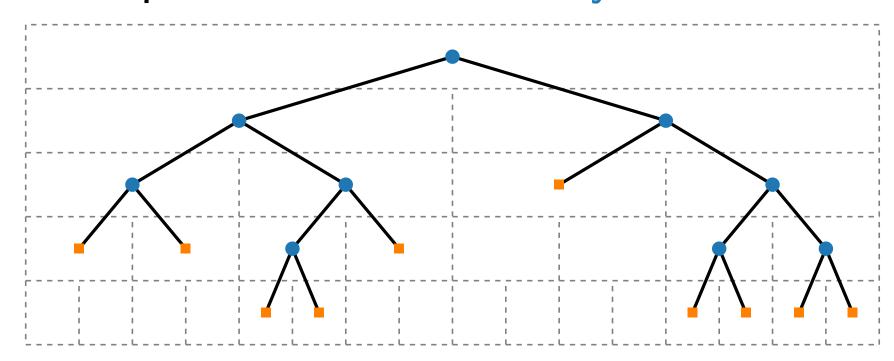
 $\log C_n = 2n + o(n)$ (by Stirling's approximation)

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

Difficulty is when binary tree is not full.

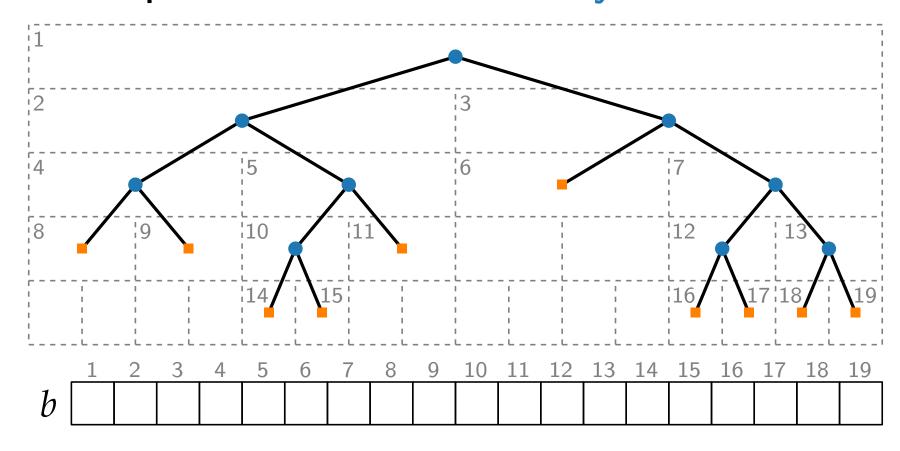


Idea.



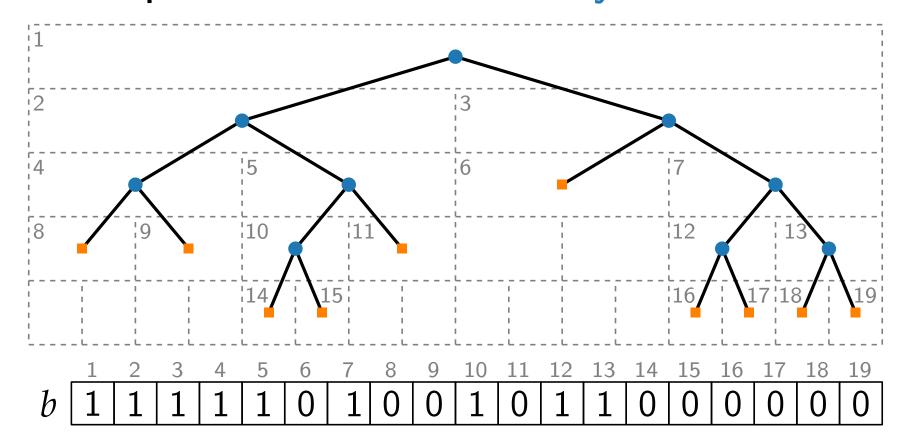
Idea.

Add external nodes



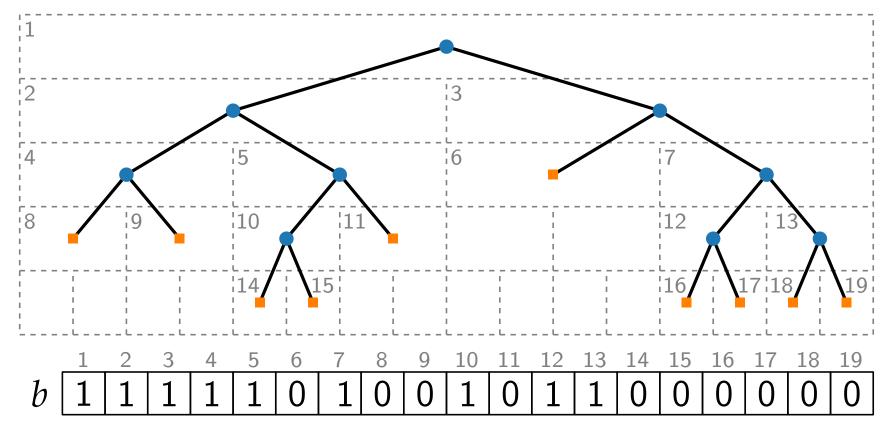
Idea.

Add external nodes



Idea.

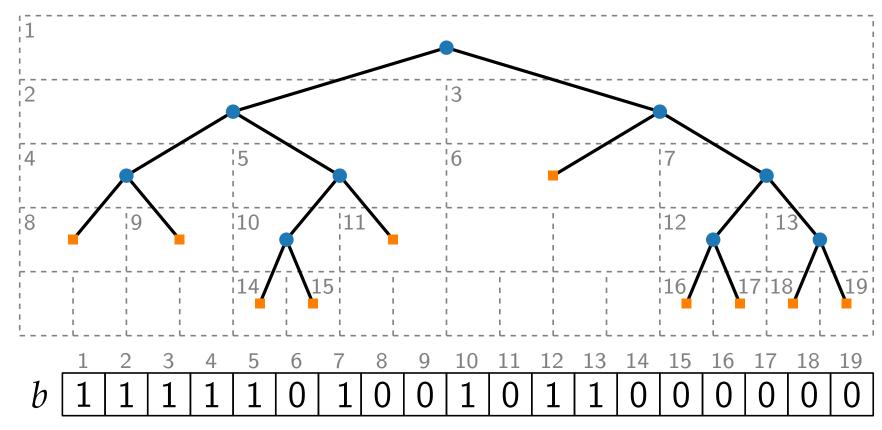
- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0



Idea.

- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0

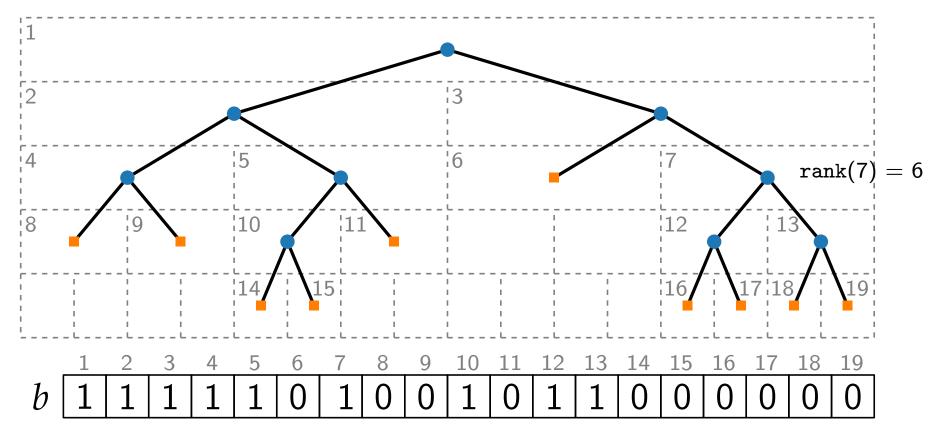
- \blacksquare parent(i) = ?
- \blacksquare leftChild(i) = ?
- \blacksquare rightChild(i) = ?



Idea.

- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

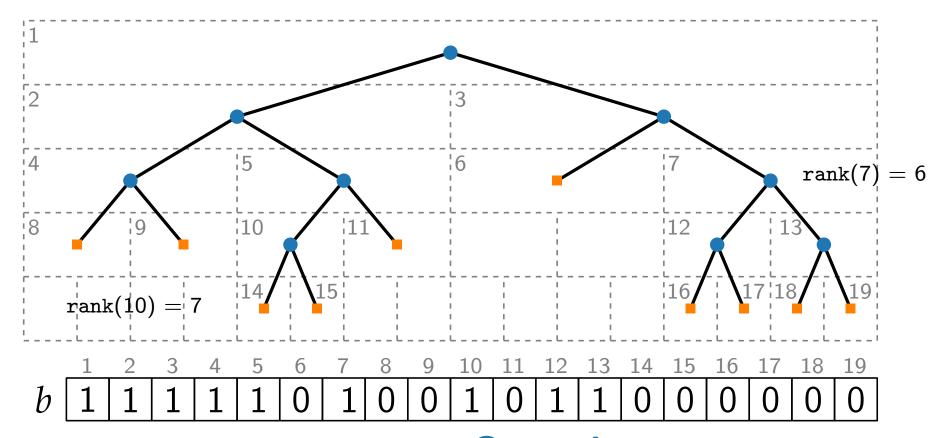
- \blacksquare parent(i) = ?
- \blacksquare leftChild(i) = ?
- \blacksquare rightChild(i) = ?



Idea.

- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

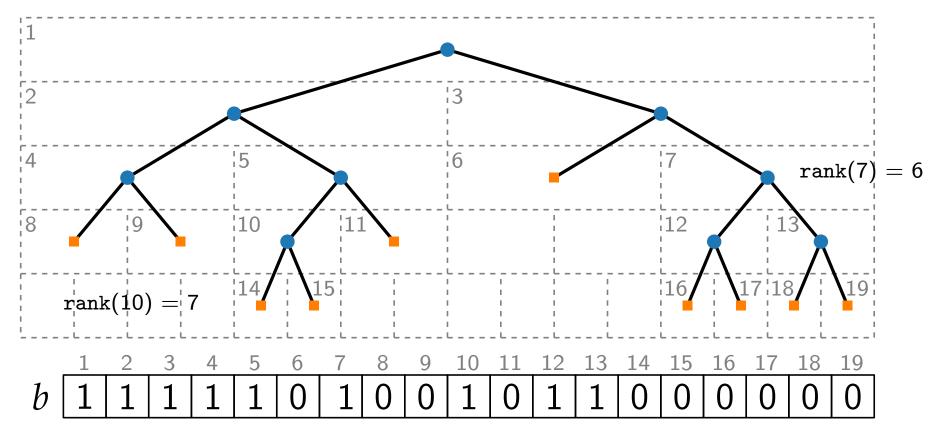
- \blacksquare parent(i) = ?
- \blacksquare leftChild(i) = ?
- lacksquare rightChild(i)= ?



Idea.

- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

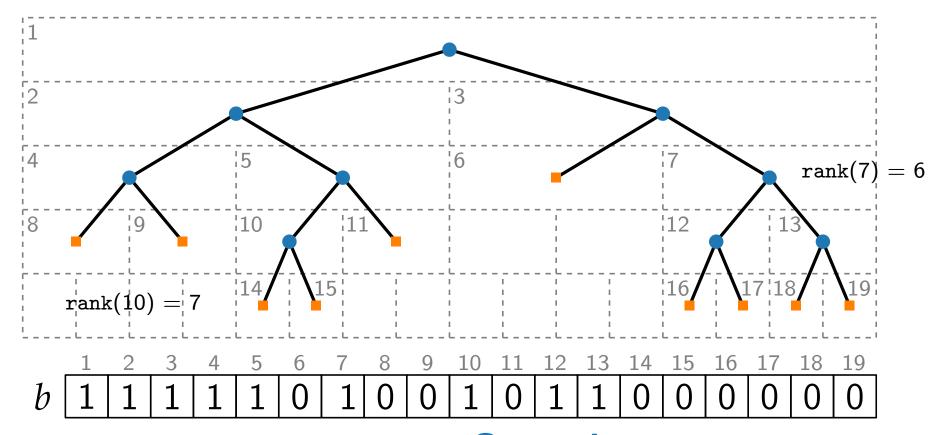
- \blacksquare parent(i) = ?
- \blacksquare leftChild(i) = ?
- \blacksquare rightChild(i) = ?



Idea.

- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

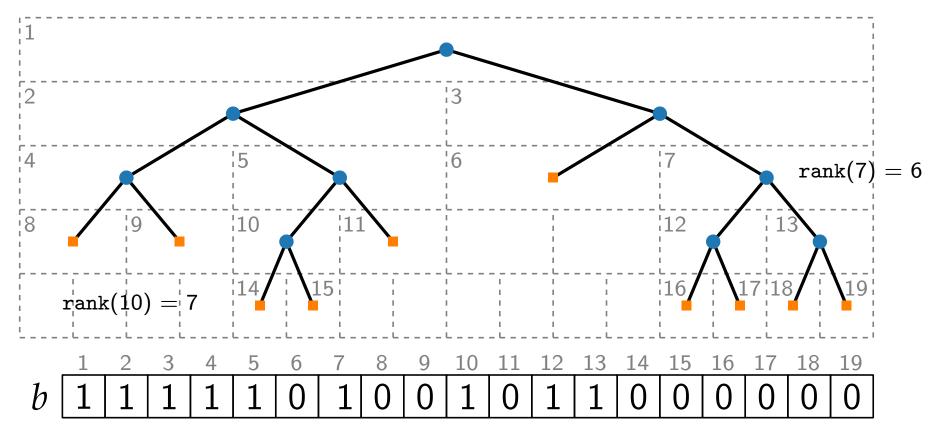
- \blacksquare parent(i) = ?
- \blacksquare leftChild $(i) = 2 \operatorname{rank}(i)$
- \blacksquare rightChild $(i) = 2 \operatorname{rank}(i) + 1$



Idea.

- Add external nodes
- 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 $(i) = 2 \operatorname{rank}(i) + 1$



Idea.

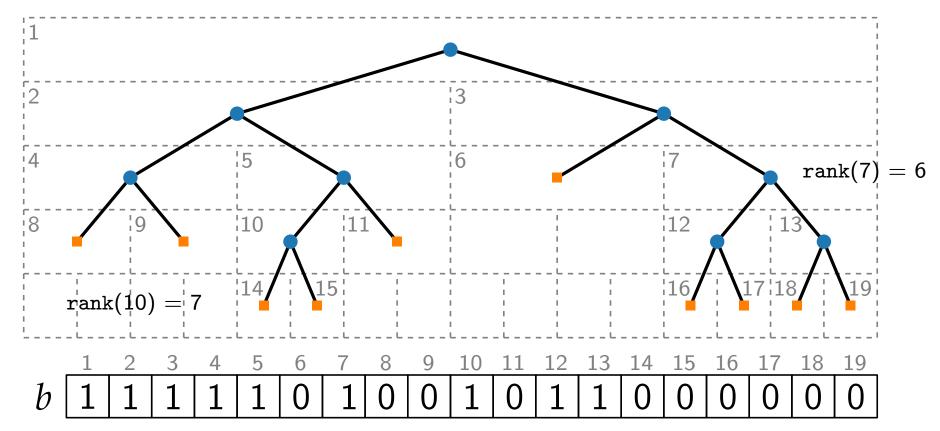
- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

Operations.

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

Proof is exercise.

Succinct representation of binary trees



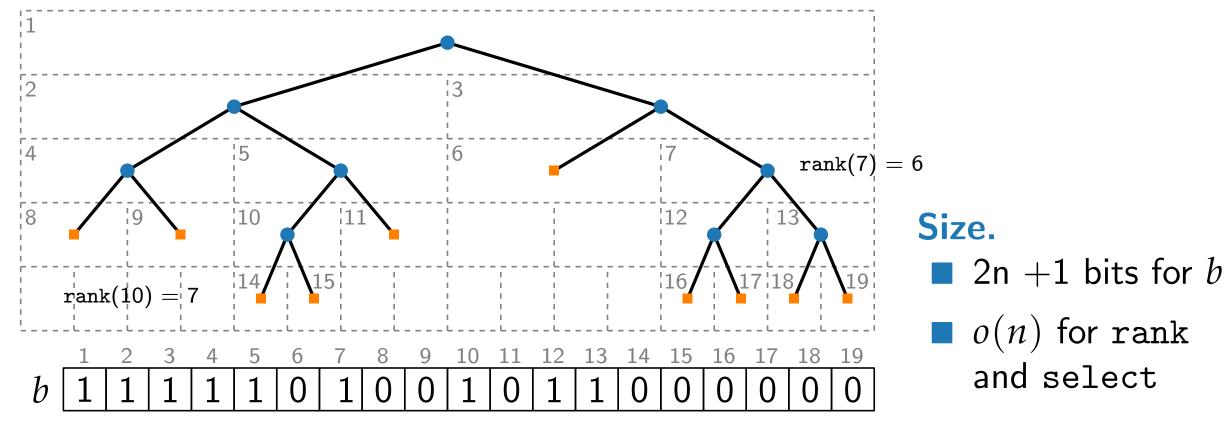
Idea.

- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

Operations.

- lacksquare parent $(i) = \text{select}(\lfloor \frac{i}{2} \rfloor)$
- \blacksquare leftChild $(i) = 2 \operatorname{rank}(i)$
- \blacksquare rightChild $(i) = 2 \operatorname{rank}(i) + 1$
- ightharpoonup rank(i) is index for array storing actual values

Succinct representation of binary trees

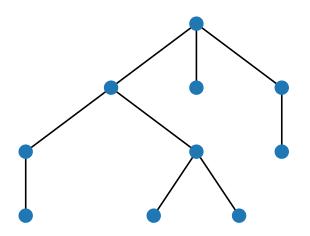


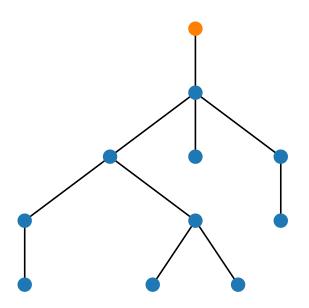
Idea.

- Add external nodes
- Read internal nodes as 1
- Read external nodes as 0
- Use rank and select

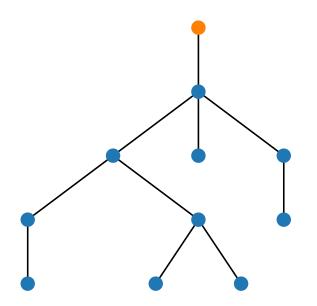
Operations.

- lacksquare parent $(i) = \operatorname{select}(\lfloor \frac{1}{2} \rfloor)$
- lacksquare leftChild $(i) = 2 \operatorname{rank}(i)$
- lacksquare rightChild $(i)=2 \; \mathrm{rank}(i)+1$
- ightharpoonup rank(i) is index for array storing actual values



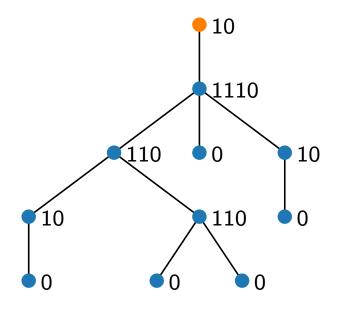


LOUDS = Level Order Unary Degree Sequence

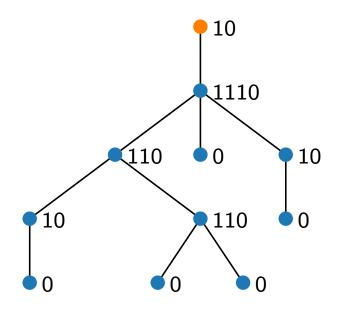


unary decoding of outdegree

LOUDS = Level Order Unary Degree Sequence



unary decoding of outdegree

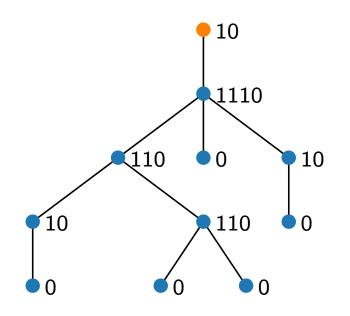


- unary decoding of outdegree
- gives LOUDS sequence

```
    1
    2
    3
    4
    5
    6
    7
    8
    9
    10
    11
    12
    13
    14
    15
    16
    17
    18
    19
    20
    21

    1
    0
    1
    1
    1
    0
    0
    1
    0
    1
    0
    1
    1
    0
    0
    0
    0
    0
```

LOUDS = Level Order Unary Degree Sequence



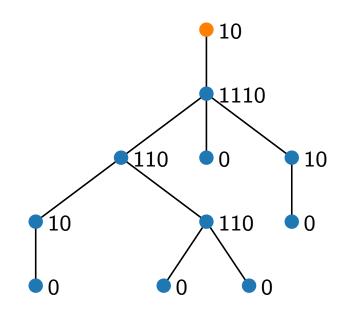
- unary decoding of outdegree
- gives LOUDS sequence



Size.

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

LOUDS = Level Order Unary Degree Sequence



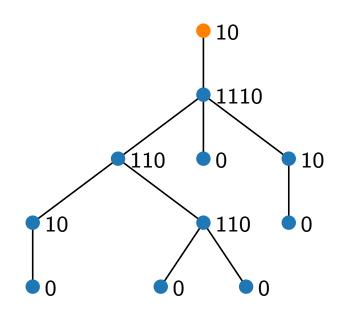
- unary decoding of outdegree
- gives LOUDS sequence

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21
1	0	1	1	1	0	1	1	0	0	1	0	1	0	1	1	0	0	0	0	0

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

LOUDS = Level Order Unary Degree Sequence



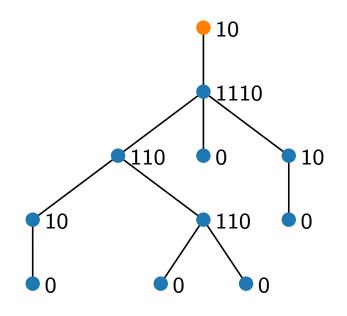
- unary decoding of outdegree
- gives LOUDS sequence

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21
1	0	1	1	1	0	1	1	0	0	1	0	1	0	1	1	0	0	0	0	0

Operations.

- \blacksquare Let i be index of 1 in louds sequence.
- ightharpoonup rank(i) is index for array storing vertex objects/values.

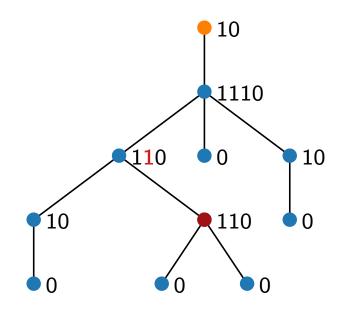
LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

firstChild $(i) = \mathtt{select}_0(\mathtt{rank}_1(i)) + 1$

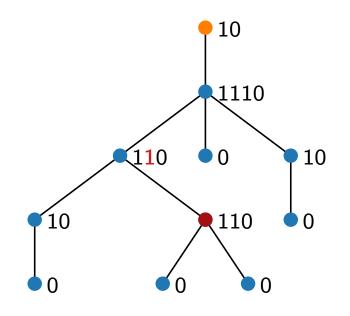
LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

firstChild(i) = select $_0(\operatorname{rank}_1(i)) + 1$ firstChild(8) = select $_0(\operatorname{rank}_1(8)) + 1$

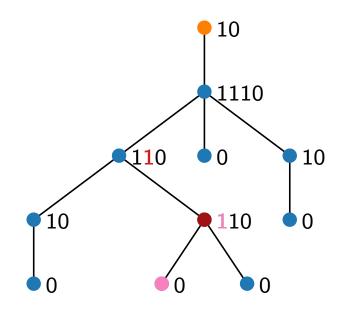
LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

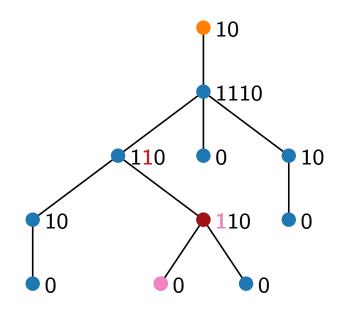
firstChild(i) = select₀(rank₁(i)) + 1 firstChild(i) = select₀(rank₁(i)) + 1 = select₀(i) + 1

LOUDS = Level Order Unary Degree Sequence



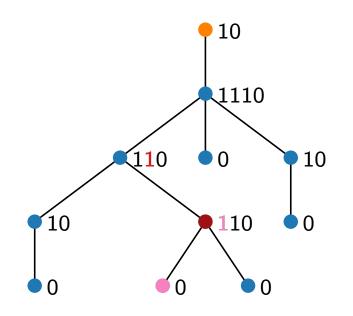
- unary decoding of outdegree
- gives LOUDS sequence

firstChild(i) = select₀(rank₁(i)) + 1 firstChild(i) = select₀(rank₁(i)) + 1 = select₀(i) + 1 = 14 + 1 = 15



- unary decoding of outdegree
- gives LOUDS sequence

- firstChild(i) = select₀(rank₁(i)) + 1 firstChild(i) = select₀(rank₁(i)) + 1 = select₀(i) + 1 = 14 + 1 = 15
- lacksquare nextSibling(i)=i+1



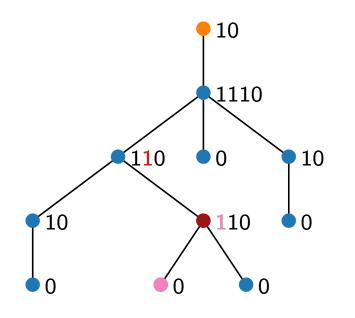
- unary decoding of outdegree
- gives LOUDS sequence

```
    1
    2
    3
    4
    5
    6
    7
    8
    9
    10
    11
    12
    13
    14
    15
    16
    17
    18
    19
    20
    21

    1
    0
    1
    1
    1
    0
    0
    1
    0
    1
    0
    1
    1
    0
    0
    0
    0
    0
```

- firstChild(i) = select₀(rank₁(i)) + 1 firstChild(i) = select₀(rank₁(i)) + 1 = select₀(i) + 1 = 14 + 1 = 15
- nextSibling(i) = i + 1Exercise: child(i, j)with validity check

LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

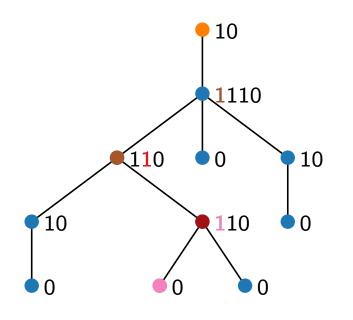
- firstChild(i) = select₀(rank₁(i)) + 1 parent(i) = select₁(rank₀(i))

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

nextSibling(i) = i + 1

Exercise: child(i, j) with validity check

LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

Exercise: child(i, j) with validity check

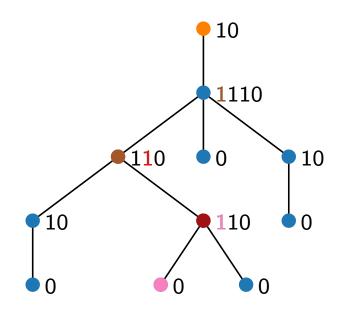
- firstChild(i) = select $_0(\operatorname{rank}_1(i)) + 1$ firstChild(8) = select $_0(\operatorname{rank}_1(8)) + 1$ = select $_0(6) + 1 = 14 + 1 = 15$
- lacksquare nextSibling(i)=i+1

parent(i) = select₁(rank₀(i))

parent(2) = select₁(rank₀(2))

 $parent(8) = select_1(rank_0(8))$

LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

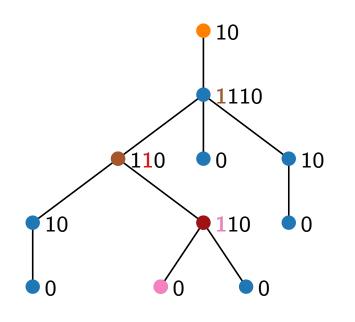
- firstChild(i) = select₀(rank₁(i)) + 1 firstChild(i) = select₀(rank₁(i)) + 1 = select₀(i) + 1 = 14 + 1 = 15
- nextSibling(i) = i + 1Exercise: child(i, j)

parent(i) = select₁(rank₀(i))

parent(8) = select₁(rank₀(8))

= select₁(2)

LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

Exercise: child(i, j) with validity check

- firstChild(i) = select₀(rank₁(i)) + 1 firstChild(i) = select₀(rank₁(i)) + 1 = select₀(i) + 1 = 14 + 1 = 15
- lacksquare nextSibling(i)=i+1

parent(i) = select₁(rank₀(i))

parent(8) = select₁(rank₀(8))

= select₁(2) = 3

Discussion

- Succinct data structures are
 - space efficient
 - support fast operations

but

- are mostly static (dynamic at extra cost),
- number of operations are limited,
- \blacksquare complex \rightarrow harder to implement

Discussion

- Succinct data structures are
 - space efficient
 - support fast operations

but

- are mostly static (dynamic at extra cost),
- number of operations are limited,
- \blacksquare complex \rightarrow harder to implement
- Rank and select form basis for many succinct representations

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