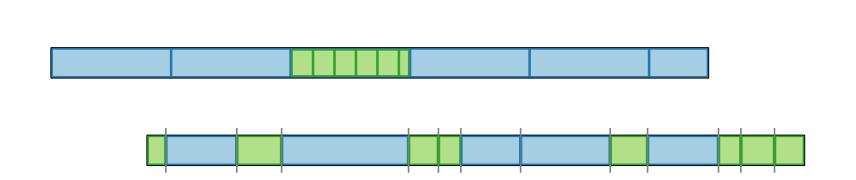


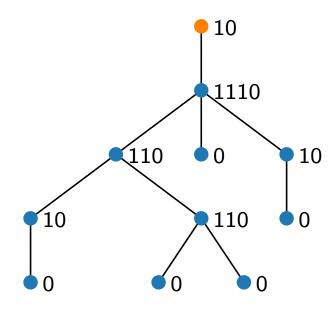
Advanced Algorithms

Succinct Data Structures

Indexable Dictionaries and Trees

Johannes Zink · WS22





A data structure is a concept to

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

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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.

A data structure is a concept to

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As such, it is a collection of

- data values,
- their relations, and
- the operations that be 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?

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- We look at data structures as a designer/implementer (and not necessarily as a user).
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Goal.

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

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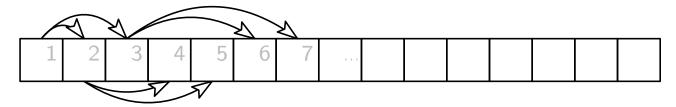
Examples!

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

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- 1-dim arrays to represent multi-dimensional arrays

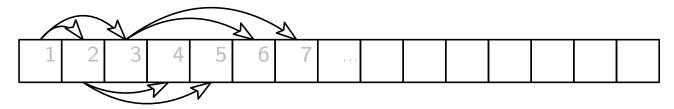
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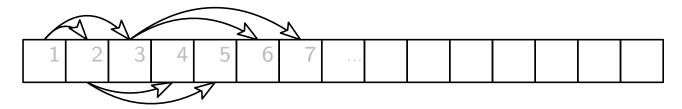
$$ext{leftChild}(i) = ext{parent}(i) = ext{rightChild}(i)$$

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

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

$$parent(i) = \lfloor \frac{i}{2} \rfloor$$

And unbalanced trees?

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

- lacktriangle 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 different subsets of $\{1, 2, ..., n\}$ are there?

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

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$$\log 2^n = n$$
 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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$$S = \{3, 4, 6, 8, 9, 14\}$$
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$$member(i)$$
 can trivially be answered in $O(1)$ time (assuming that we can access any entry in constant time)

select(5) = 9

Represent S with a bit vector b of length n where

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

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- ightharpoonup rank(i)=# 1s at or before position i
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$$select(5) = 9$$

$$rank(9) = 5$$

Represent S with a bit vector b of length n where

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$$S = \{3, 4, 6, 8, 9, 14\}$$
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$$select(5) = 9$$

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

Represent S with a bit vector b of length n where

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$$select(5) = 9$$
 $rank(9) = 5 = rank(12)$
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member(i) can trivially be answered in O(1) time (assuming that we can access any entry in constant time)

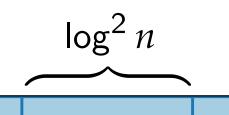
Exercise: Use these methods to answer predecessor(i) and successor(i) in O(1) time.

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

Rank in o(n) Bits

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1	
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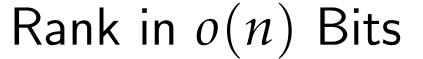
Rank in o(n) Bits

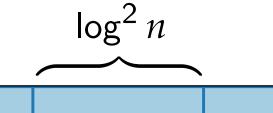


b

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

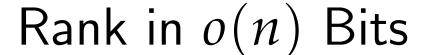
 $\log^2 n = (\log n)^2$

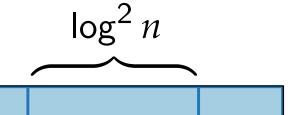






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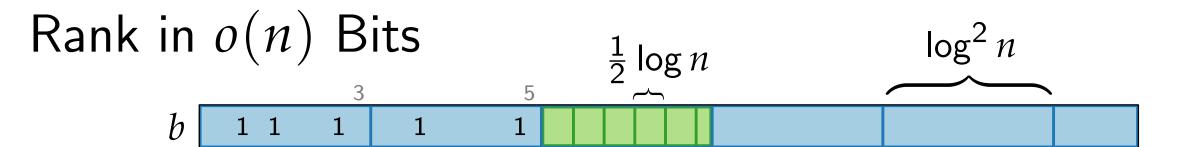


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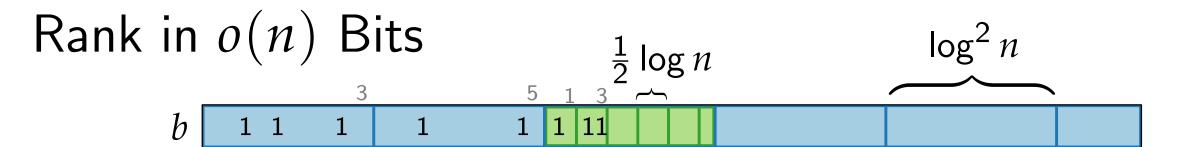


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2. Split **chunks** into $(\frac{1}{2} \log n)$ -bit **subchunks** and store cumulative rank within **chunk**:

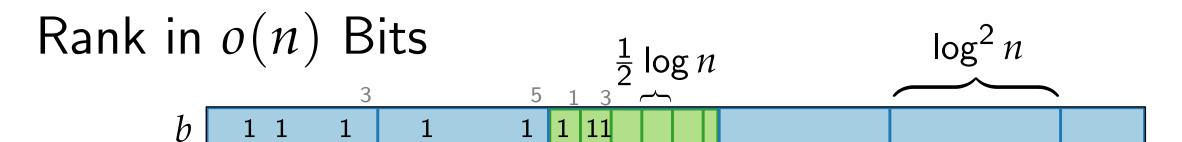


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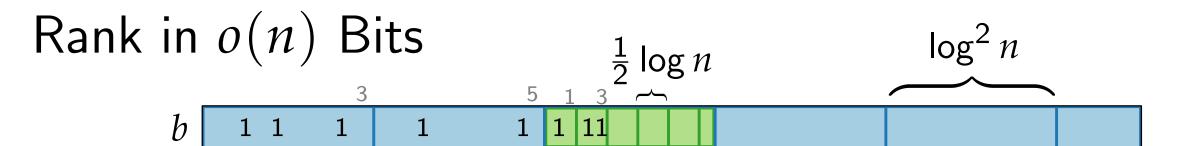
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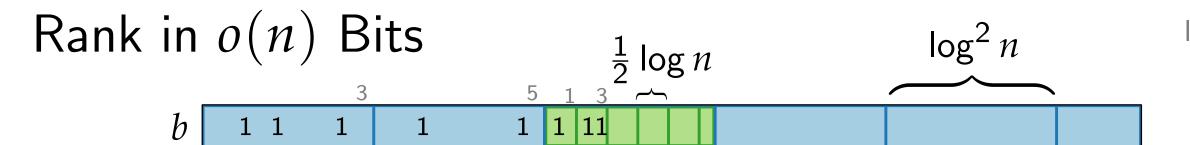
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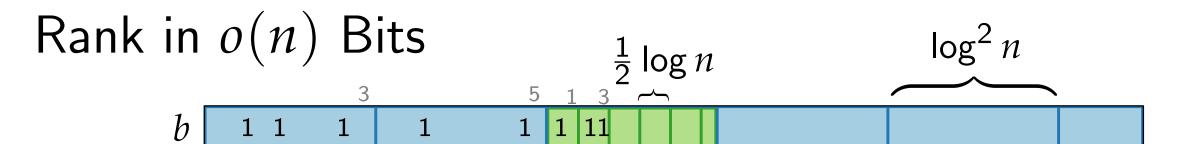


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- 3. Use lookup table for bitstrings of length $(\frac{1}{2} \log n)$:

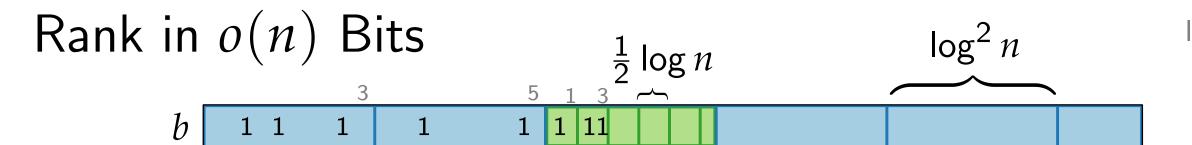


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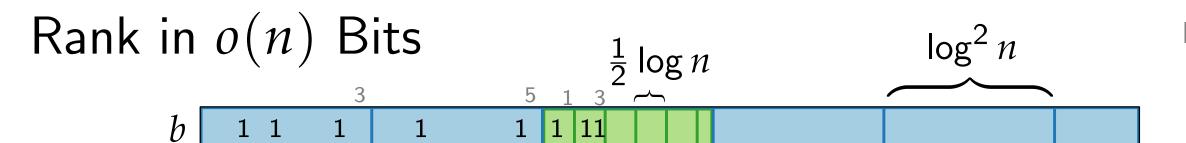


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- 3. Use lookup table for bitstrings of length $(\frac{1}{2} \log n)$: $2^{\frac{1}{2} \log n} = \sqrt{n}$ entries $\Rightarrow O(\sqrt{n} \log n \log \log n) \subseteq o(n)$ bits # bitstrings query i answer



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

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 - + relative rank of subchunk within chunk
 - + relative rank of element i within subchunk



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 bits

 $\log^2 n$

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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+ relative rank of subchunk within chunk

+ relative rank of element i within subchunk

 $\Rightarrow O(1)$ time

(assume read & write 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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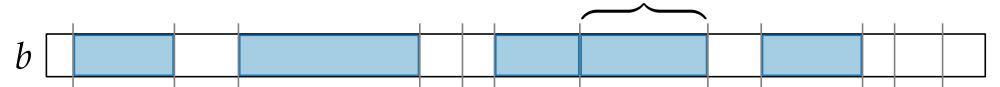
$$\Rightarrow O(\underbrace{\frac{n}{\log n \log \log n}}_{\text{ # groups}} \underbrace{\log n}) = O(\underbrace{\frac{n}{\log \log n}}) \subseteq o(n) \text{ bits}$$

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



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



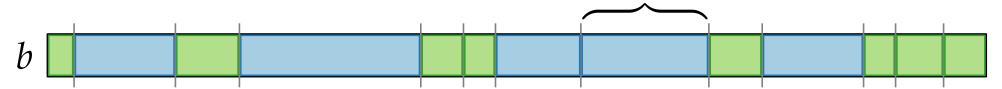
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if
$$r \ge (\log n \log \log n)^2$$

then store indices of 1 bits in group in array

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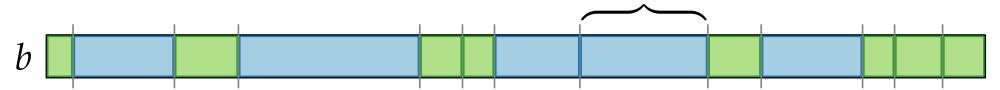
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$$\Rightarrow O(\frac{n}{(\log n \log \log n)^2}(\log n \log \log n) \log n) \subseteq O(\frac{n}{\log \log n})$$
 bits

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

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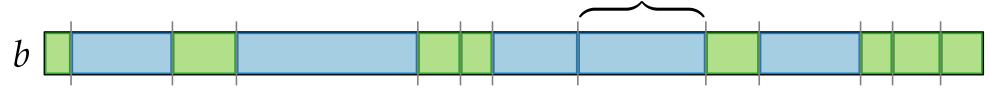
$$\Rightarrow O(\frac{n}{(\log n \log \log n)^2}(\log n \log \log n) \log n) \subseteq O(\frac{n}{\log \log n})$$
 bits

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



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



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 - 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:



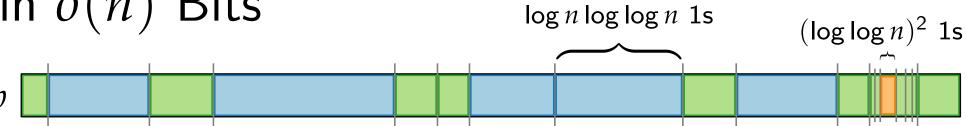
- 3. Repeat 1. and 2. on reduced bitstrings $(r < (\log n \log \log n)^2)$:
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$$\Rightarrow O(\frac{n}{(\log \log n)^2} \log \log n) = O(\frac{n}{\log \log n})$$
 bits

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



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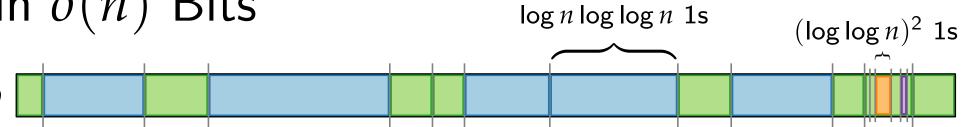
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subgroups # 1 bits rel. index



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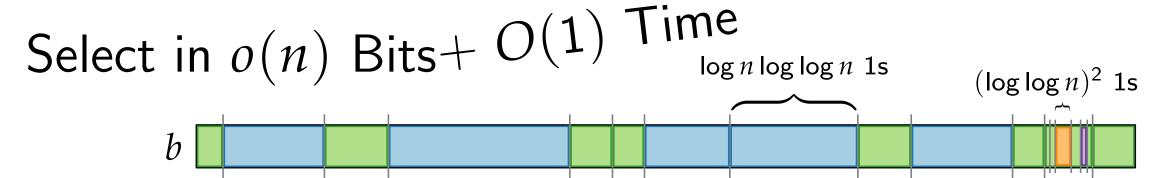
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4. Use lookup table for bitstrings of length $r' \leq (\log \log n)^4 \leq \frac{1}{2} \log n$ $\Rightarrow O(\sqrt{n} \log n \log \log n) = o(n) \text{ bitstrings query } j \text{ answer}$



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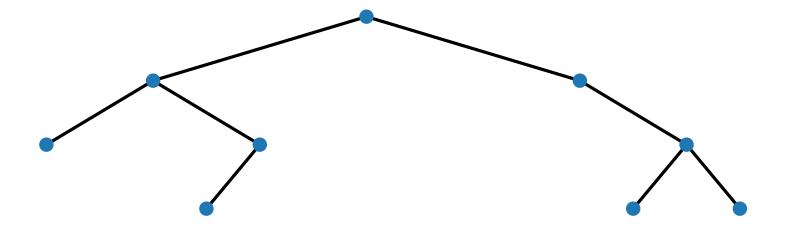
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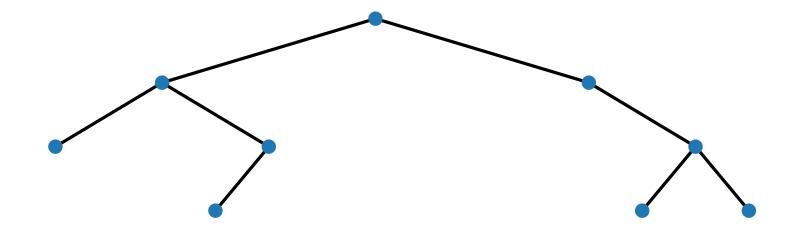
$$\Rightarrow O(\frac{n}{(\log \log n)^4}(\log \log n)^2 \log \log n) = O(\frac{n}{\log \log n})$$
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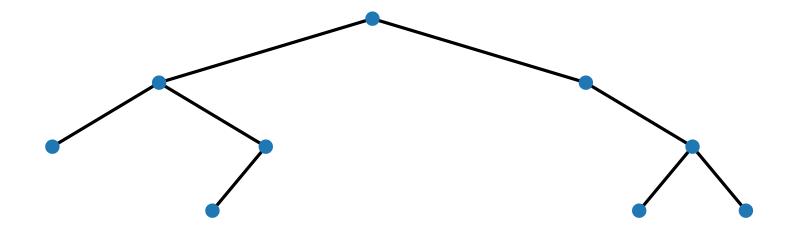


Number of binary trees on n vertices:



- C_n is the n-th $extit{Catalan number}$ and $C_0=1$

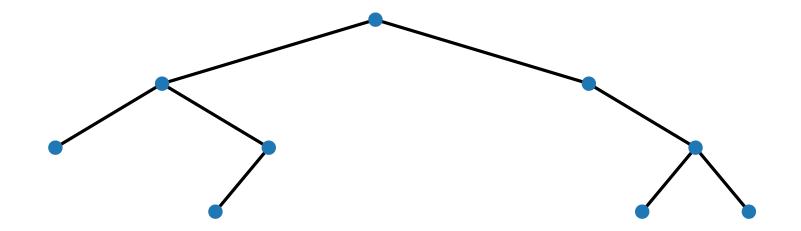
Number of binary trees on n vertices: $C_n = \sum_{i=0}^{n-1} C_i \cdot C_{n-1-i} = \frac{(2n)!}{(n+1)! \, n!}$



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

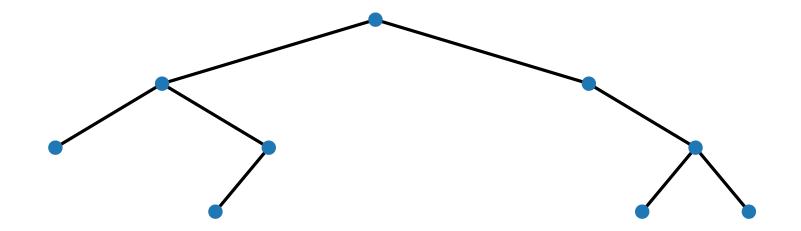


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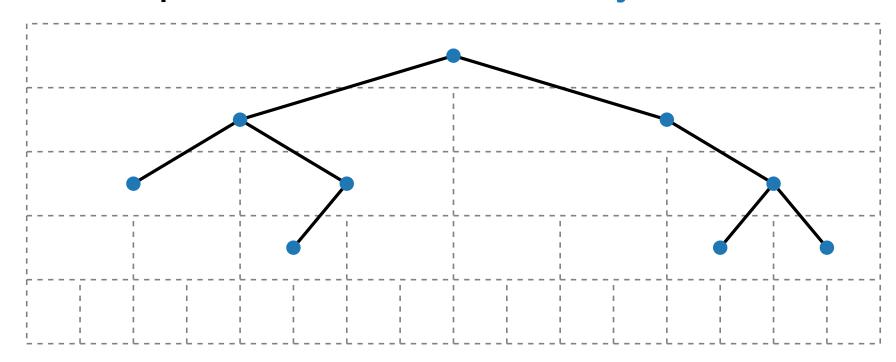
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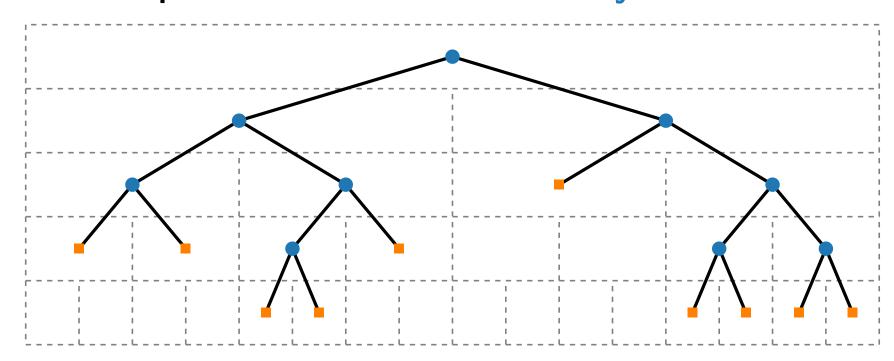
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Difficulty is when a binary tree is not full.

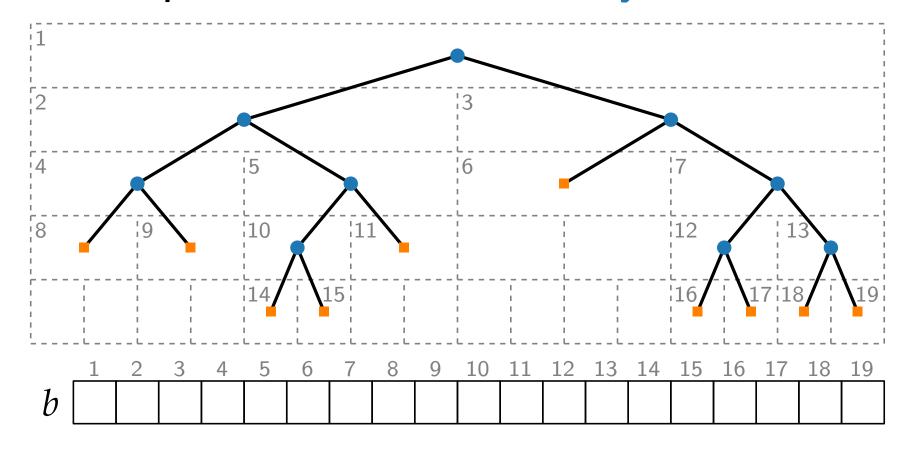


Idea.



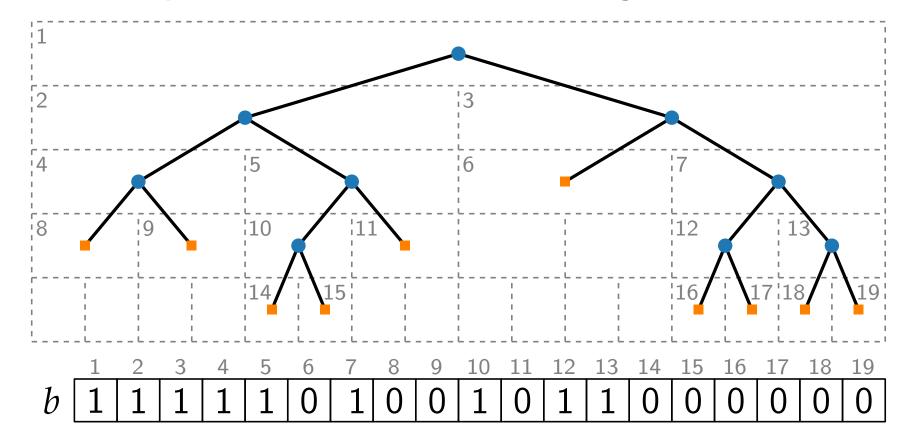
Idea.

Add external nodes to have out-degree 2



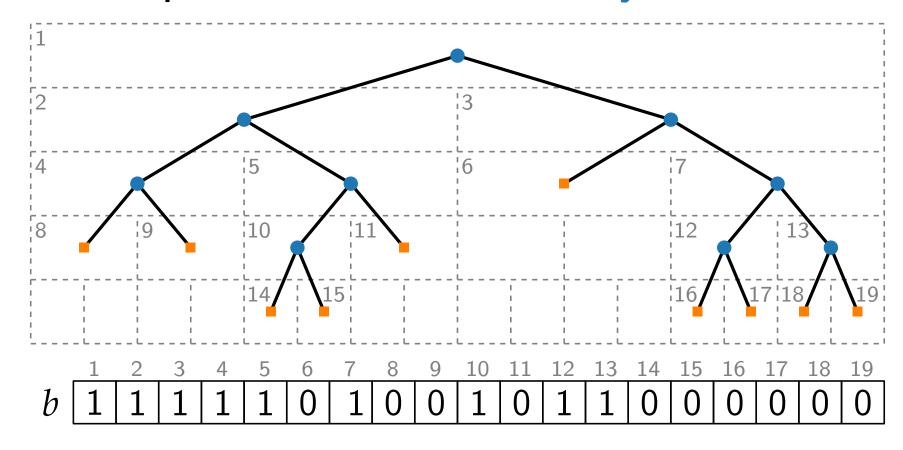
Idea.

■ Add external nodes to have out-degree 2



Idea.

- Add external nodes to have out-degree 2
- Read internal nodes as 1
- Read external nodes as 0

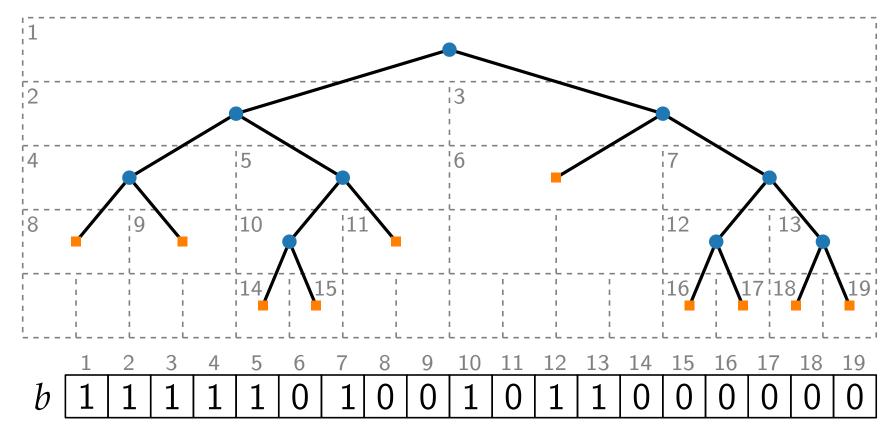


Size.

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

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Size.

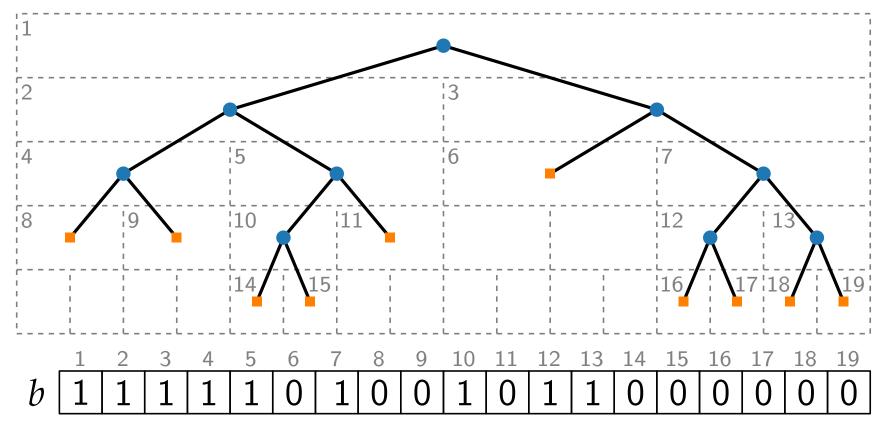
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Idea.

Add external nodes to have out-degree 2

- Read internal nodes as 1
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- \blacksquare parent(i) = ?
- lacksquare leftChild(i)= ?
- \blacksquare rightChild(i) = ?



Size.

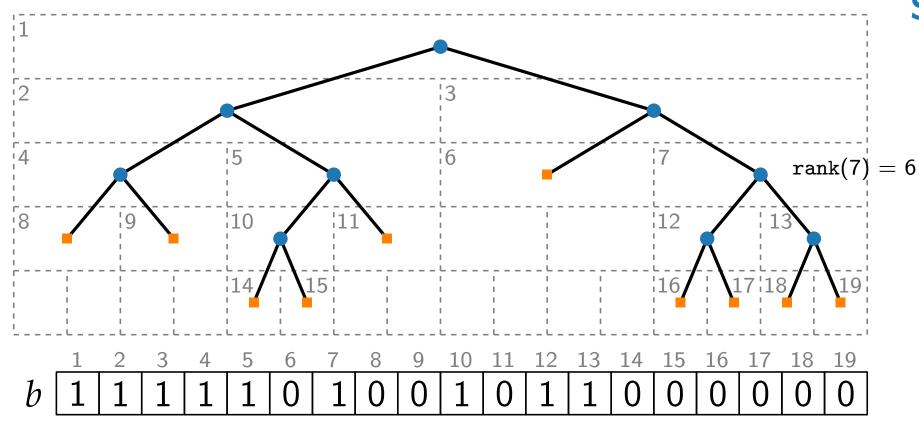
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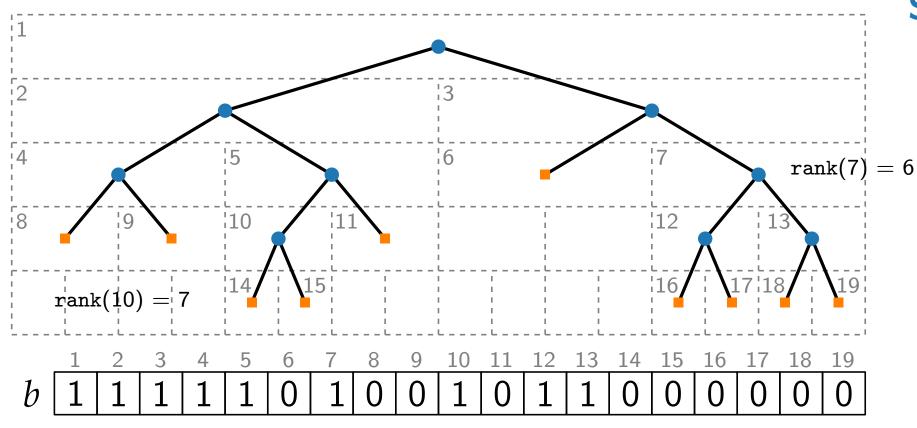
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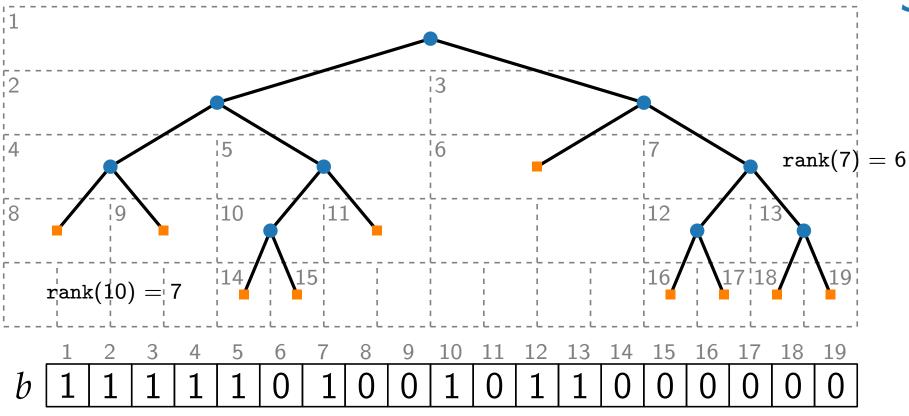
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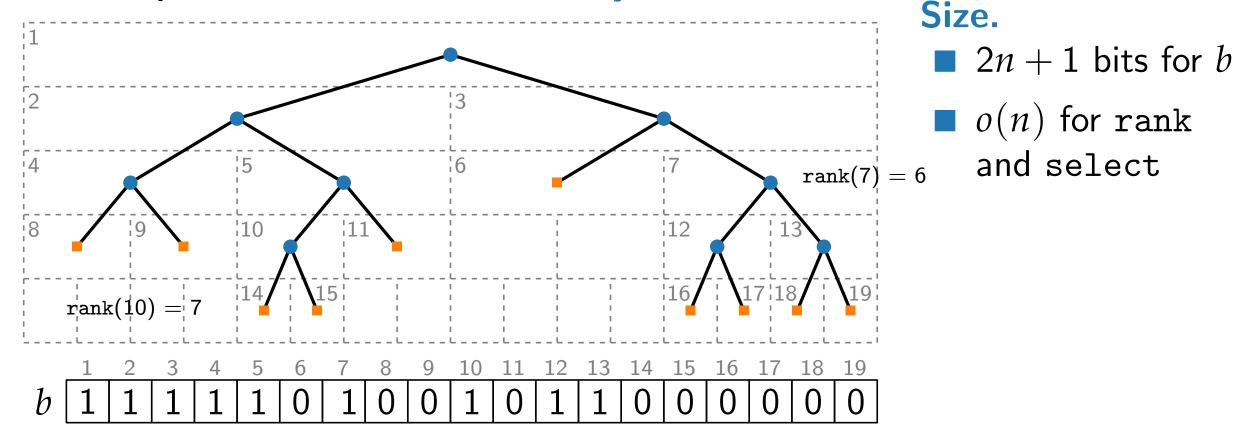
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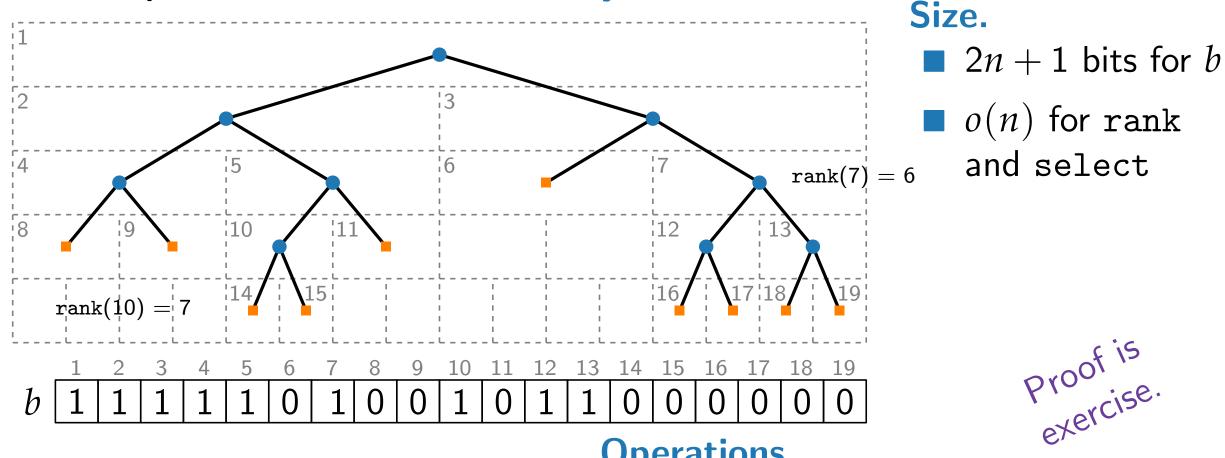
- \blacksquare parent(i) = ?
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- rightChild(i) = 2 rank(i) + 1



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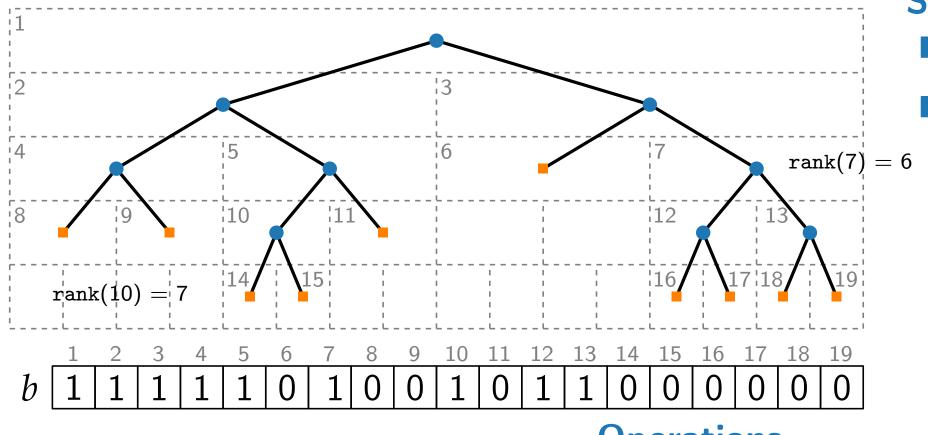
- lacksquare parent $(i) = \operatorname{select}(\lfloor \frac{1}{2} \rfloor)$
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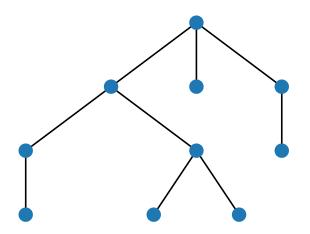
Proof is exercise

Idea.

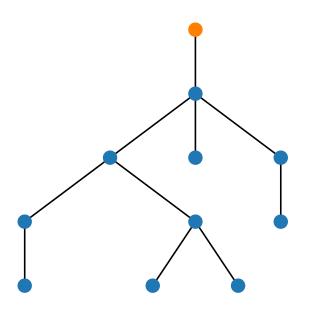
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- lacksquare leftChild $(i)=2 \ \mathrm{rank}(i)$
- lacksquare rightChild $(i)=2 \ \mathrm{rank}(i)+1$
- \blacksquare rank(i) is index for extra storing array

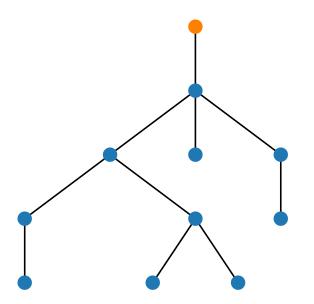
LOUDS = Level Order Unary Degree Sequence



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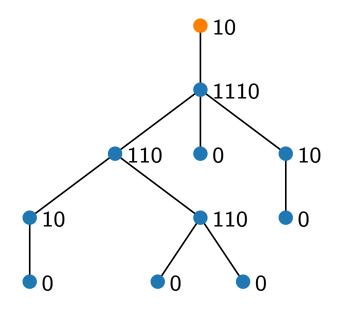


LOUDS = Level Order Unary Degree Sequence



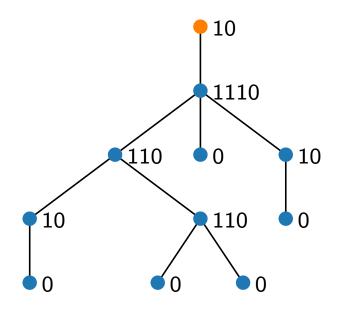
unary decoding of outdegree

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unary decoding of outdegree

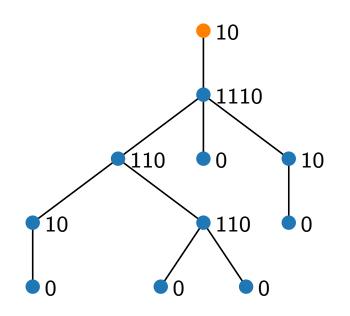
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- unary decoding of outdegree
- gives LOUDS sequence

_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

LOUDS = Level Order Unary Degree Sequence



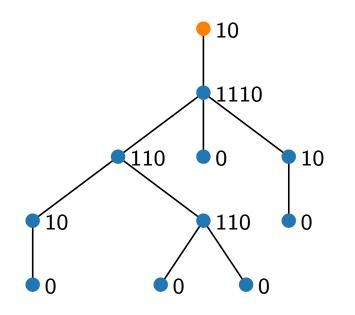
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	1	0	1	1	1	0	1	1	0	0	1	0	1	0	1	1	0	0	0	0	0

Size.

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

LOUDS = Level Order Unary Degree Sequence



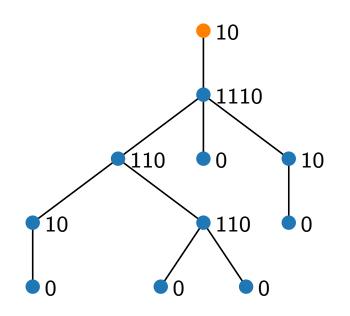
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Size.

- 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

LOUDS = Level Order Unary Degree Sequence

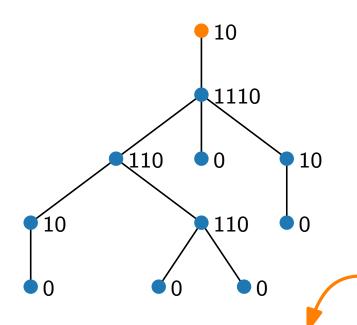


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- Let i be index of 1 in LOUDS sequence.
- ightharpoonup rank(i) is index for array storing vertex objects/values.

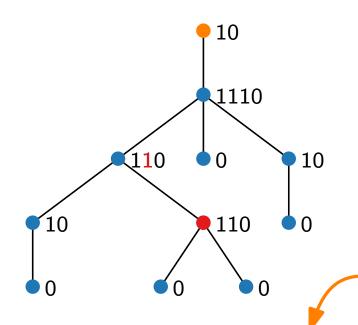
LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

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

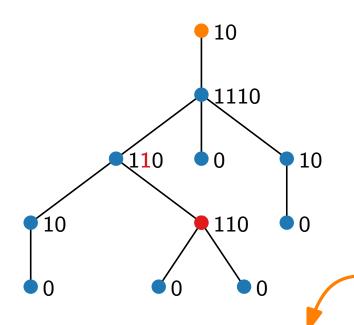
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- unary decoding of outdegree
- gives LOUDS sequence

execute select(j) on the 0s instead of the 1s (as before) firstChild(i) = select $_0(\mathrm{rank}_1(i)) + 1$ $\mathtt{firstChild}(\textcolor{red}{8}) = \mathtt{select}_0(\mathtt{rank}_1(\textcolor{red}{8})) + 1$

LOUDS = Level Order Unary Degree Sequence

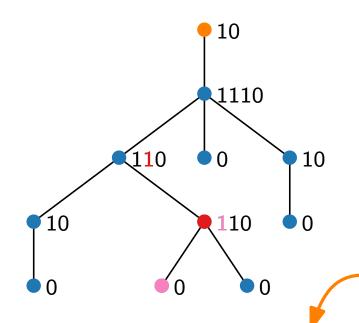


- unary decoding of outdegree
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execute select(j) on the 0s instead of the 1s (as before) firstChild(i) = select $_0(\operatorname{rank}_1(i)) + 1$

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

LOUDS = Level Order Unary Degree Sequence



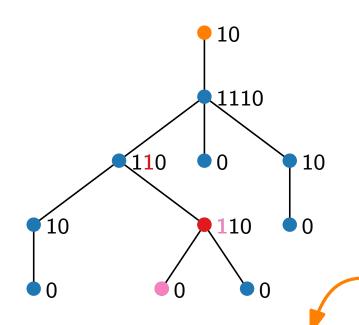
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execute select(j) on execute rank(i) on the 0s instead of the 1s (as before)

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

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

LOUDS = Level Order Unary Degree Sequence



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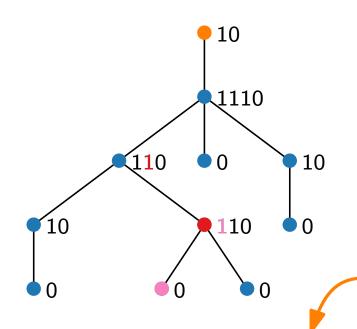
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nextSibling(i) = i + 1

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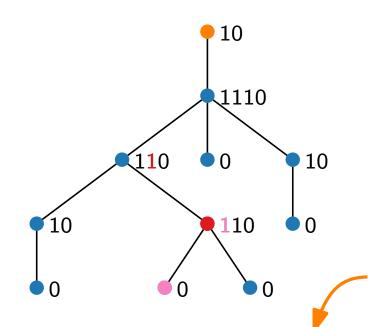
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execute select(j) on execute rank(i) on the 0s instead of the 1s (as before)

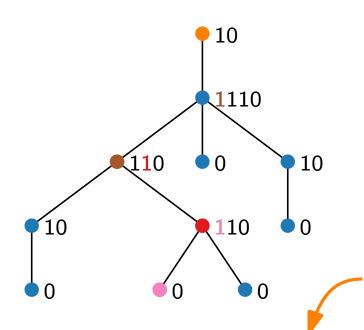
 $\texttt{firstChild}(i) = \texttt{select}_0(\texttt{rank}_1(i)) + 1 \quad / \quad \texttt{parent}(i) = \texttt{select}_1(\texttt{rank}_0(i))$

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

 $= select_0(6) + 1 = 14 + 1 = 15$

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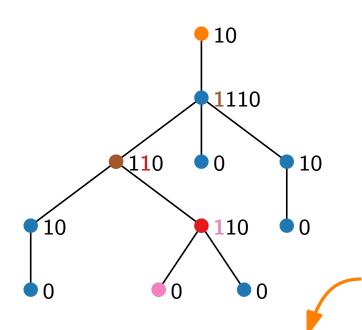
 $firstChild(8) = select_0(rank_1(8)) + 1$ $= select_0(6) + 1 = 14 + 1 = 15$

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

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

LOUDS = Level Order Unary Degree Sequence



- unary decoding of outdegree
- gives LOUDS sequence

execute select(j) on execute rank(i) of the 0s instead of the 1s (as before)

execute rank(i) on

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

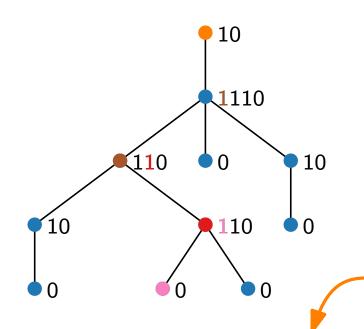
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with validity check $\mathtt{nextSibling}(i) = i+1$

 $lacktriant{\blacksquare}$ parent $(i) = \operatorname{select}_1(\operatorname{rank}_0(i))$

 $parent(8) = select_1(rank_0(8))$ = select₁(2)

LOUDS = Level Order Unary Degree Sequence



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execute select(j) on execute rank(i) on the 0s instead of the 1s (as before)

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Exercise: child(i,j)
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 $parent(8) = select_1(rank_0(8))$ = select₁(2) = 3

Discussion

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

but

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

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- are mostly static (dynamic at extra cost),
- number of operations is 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