

Lehrstuhl für INFORMATIK I Effiziente Algorithmen und wissensbasierte Systeme



Advanced Algorithms

Winter term 2019/20

Lecture 9. Succinct data structures

(Based on lectures from Simon Gog and from Erik Demaine)

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Chair for Computer Science I

Goal

- use space "close" to information-theoretical minimum
- but still support time-efficient operations

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- succinct, if it takes L + o(L) bits of space;

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

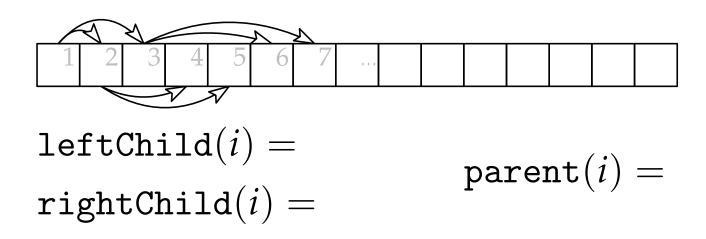
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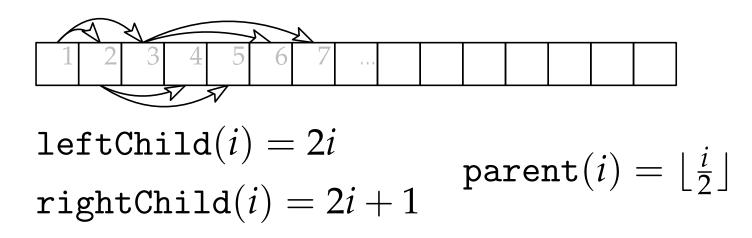
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- 1-dim array to represent multi-dimensional array
- sorted array to represent sorted list; but why not binary search tree?

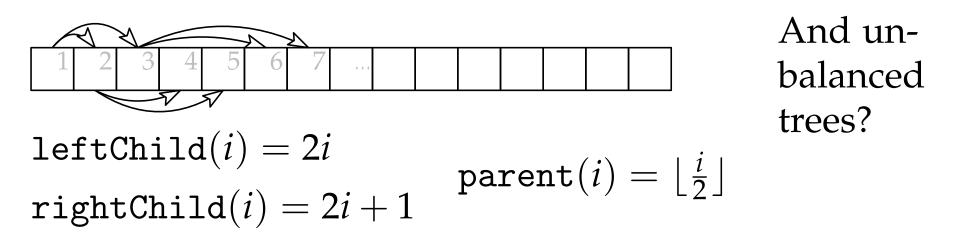
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- Represent a subset $S \subset [n]$ and support O(1) operations:
 - **member**(i) returns if $i \in S$
 - rank(i) = # 1's at or before position i
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How many bits of space do we need to distinguish them?

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 $\log 2^n = n$ bits

Represent *S* with a bit vector *b* of length *n* where

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

$$select(5) = 9$$

$$b 0 0 1 1 0 1 0 1 0 1 1 0 0 0 0 1 0$$

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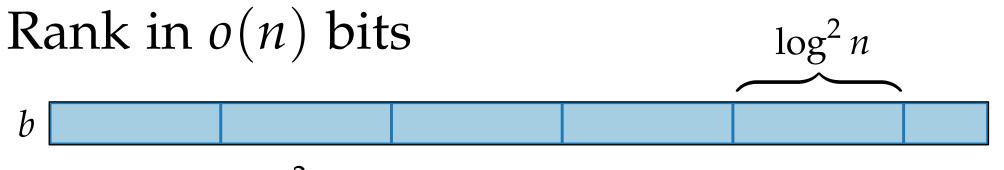
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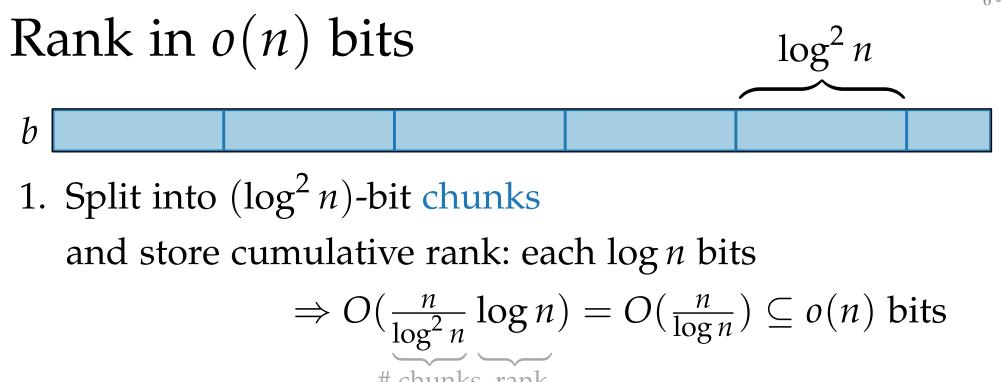
Rank in o(n) bits



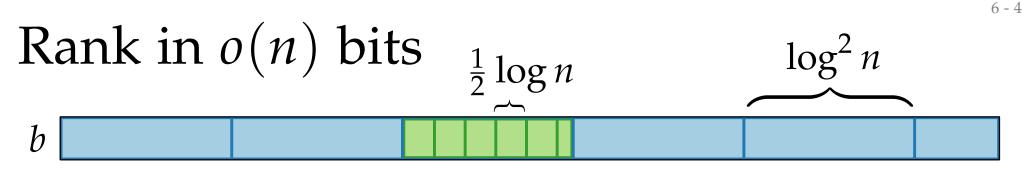


1. Split into $(\log^2 n)$ -bit chunks

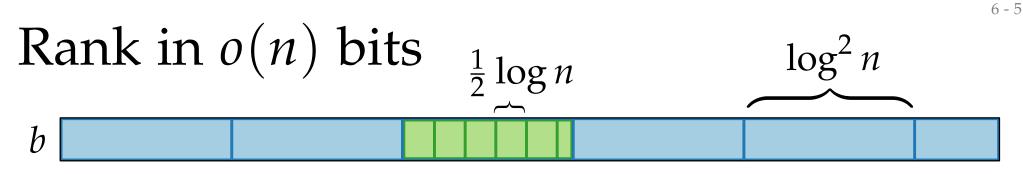
and store cumulative rank: each log *n* bits



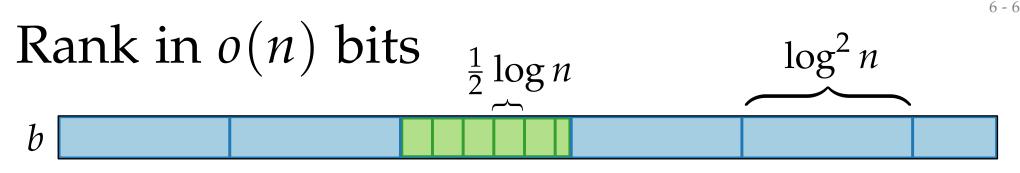
chunks rank



- 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 cummulative rank within chunk:



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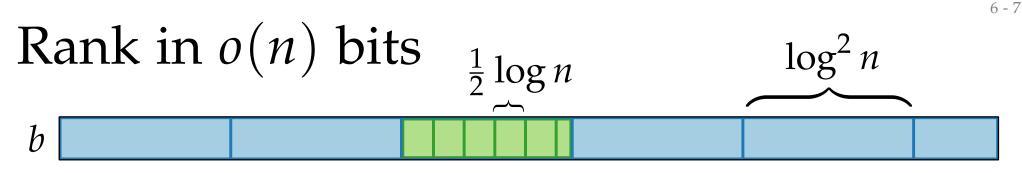


- Split into (log² n)-bit chunks
 and store cumulative rank: each log n bits
 ⇒ O(ⁿ/_{log² n} log n) = O(ⁿ/_{log n}) ⊆ o(n) bits

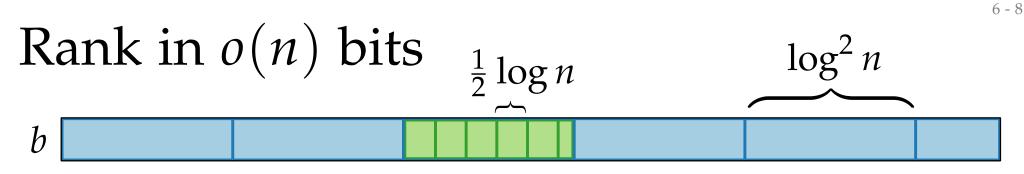
 Split chunks into (¹/₂ log n)-bit subchunks
- 2. Split chunks into $(\frac{1}{2} \log n)$ -bit subchunks and store cumulative rank within chunk: $2 \log \log n$ bits

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

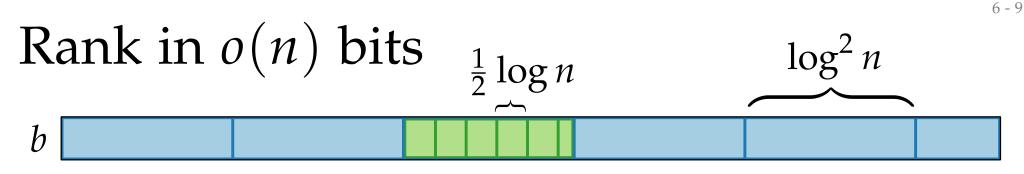
subch. rel. rank



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- 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



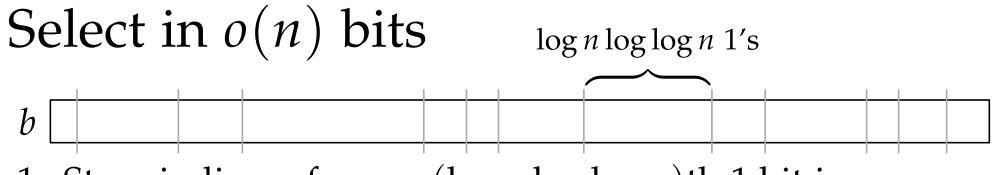
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- 4. rank = rank of chunk
 + relative rank of subchunk within chunk
 + relative rank of element within subchunk



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- 4. rank = rank of chunk $\Rightarrow O(1)$ time + relative rank of subchunk within chunk + relative rank of element within subchunk

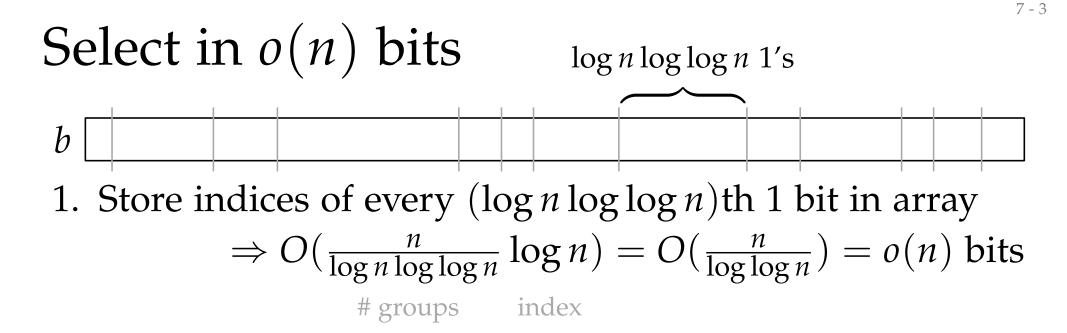
Select in o(n) bits

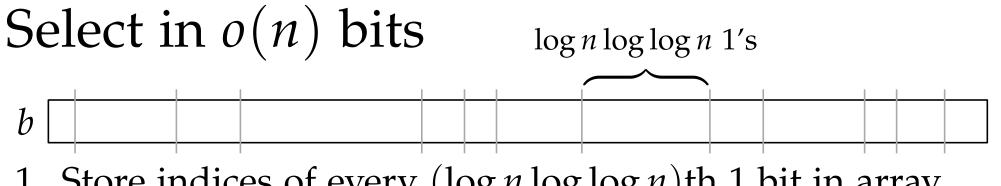




7 - 2

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

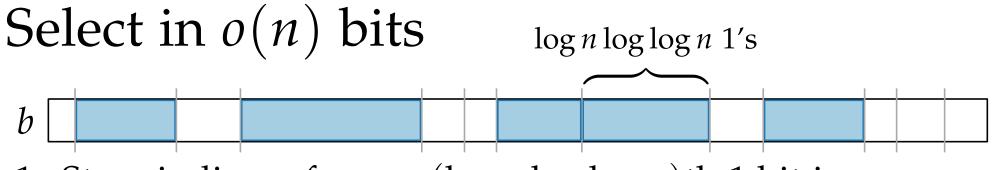




7 - 4

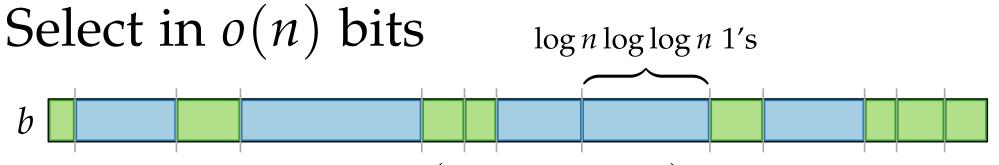
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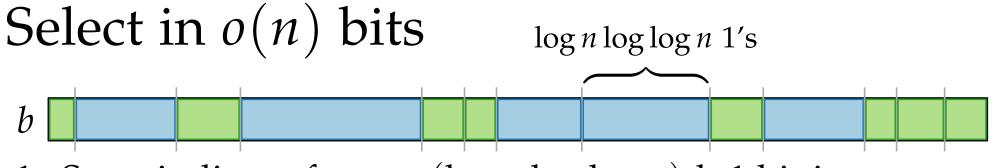
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groups # 1 bits index

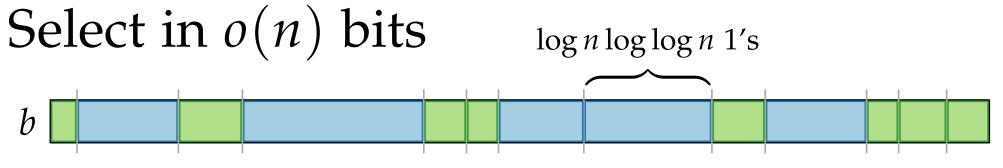


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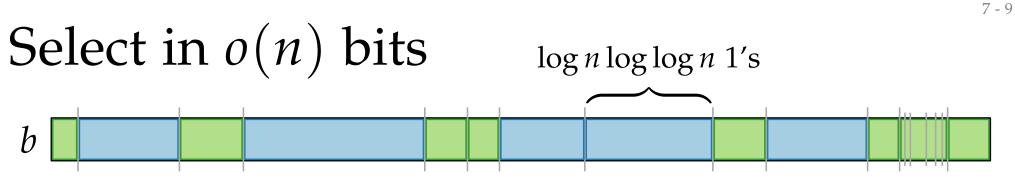
else reduced to bitstrings of length $r < (\log n \log \log n)^2$



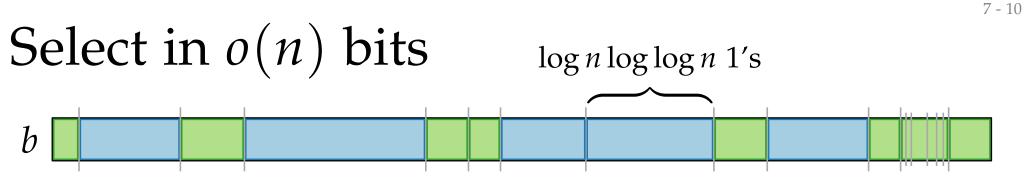
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- 3. Repeat 1. and 2. on reduced bitstrings



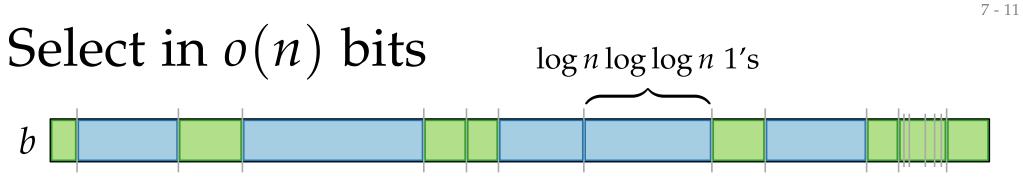
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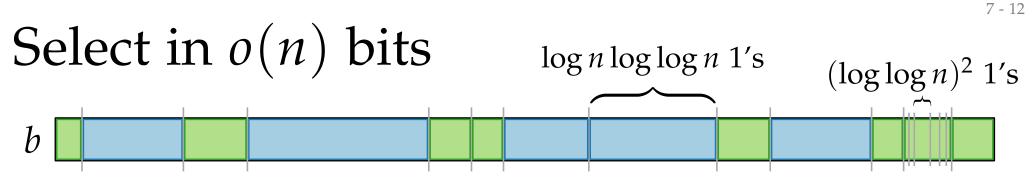
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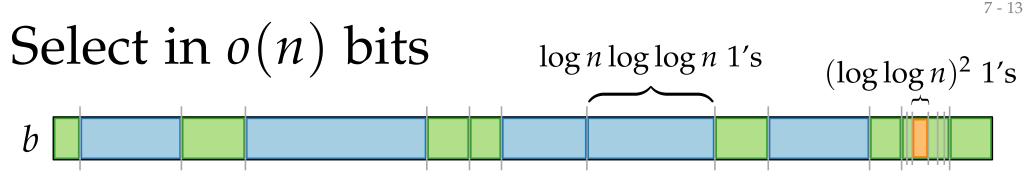
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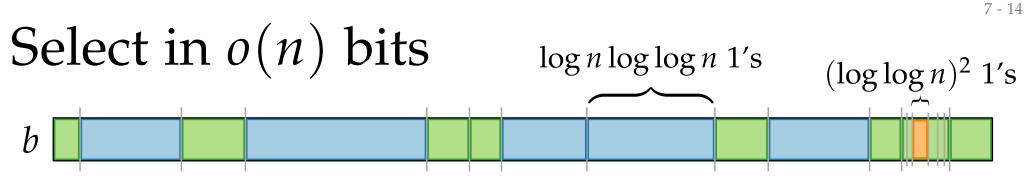
subgroups rel. index



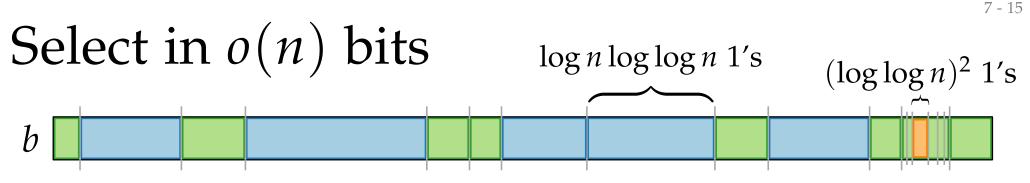
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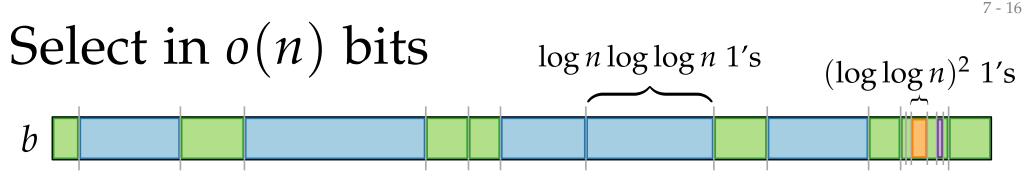
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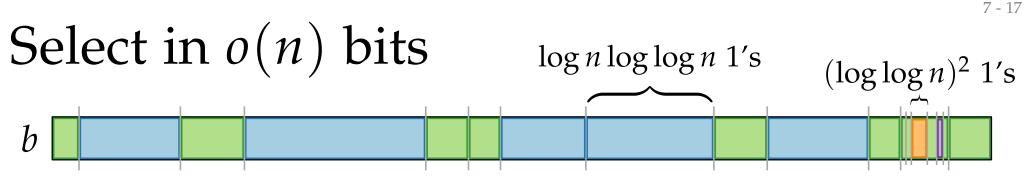


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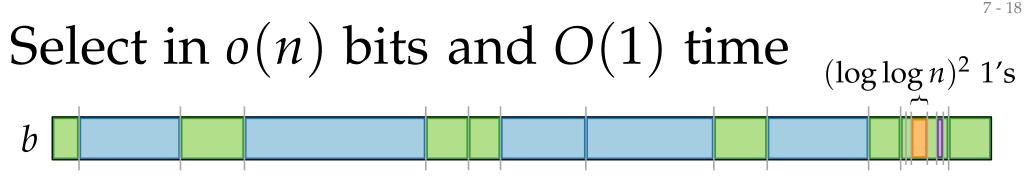
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4. Use lookup table for bitstrings of length $r' \le \frac{1}{2} \log n$ $\Rightarrow O(\sqrt{n} \log n \log \log n) = o(n)$ bits bitstring query *j* answer



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

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

Operations we want to support: parent(v), leftChild(v), rightChild(v)

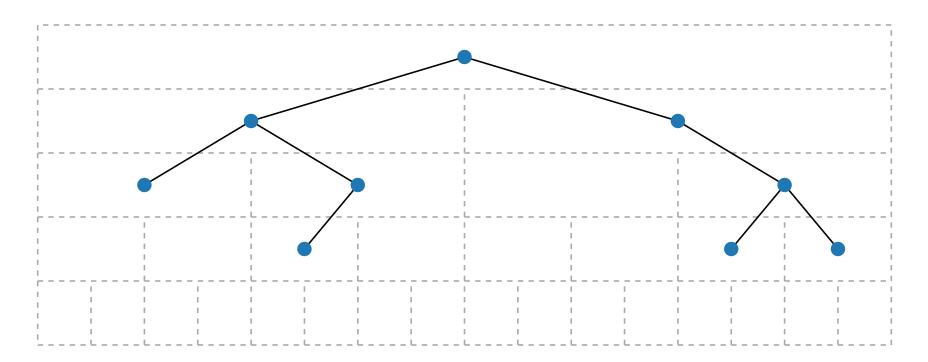
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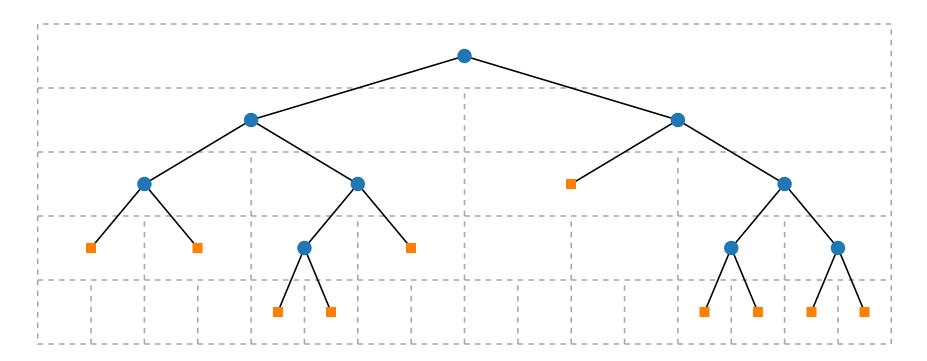
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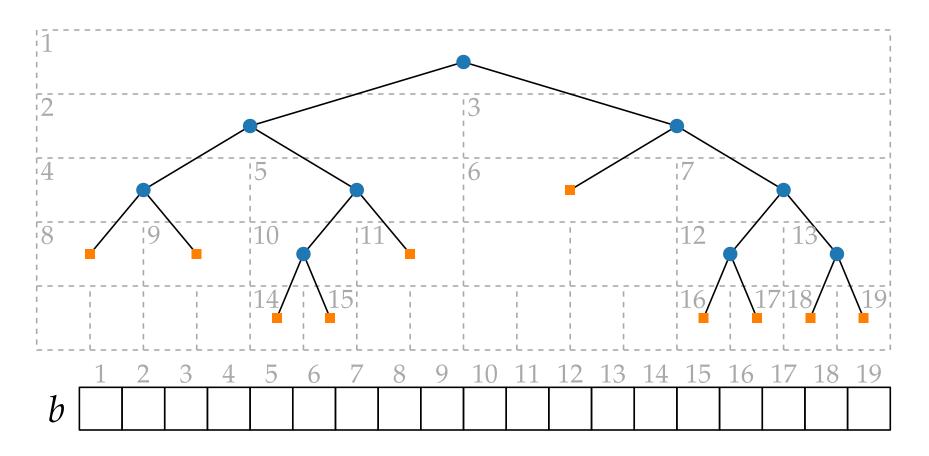
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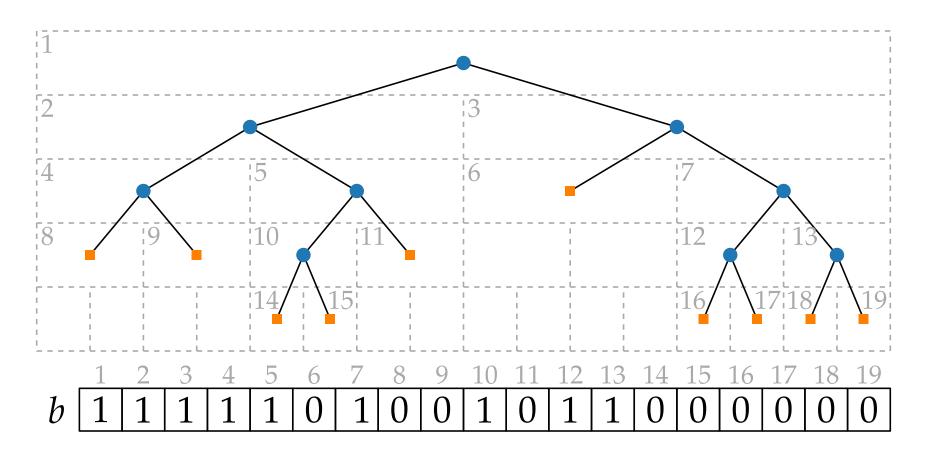
Idea:

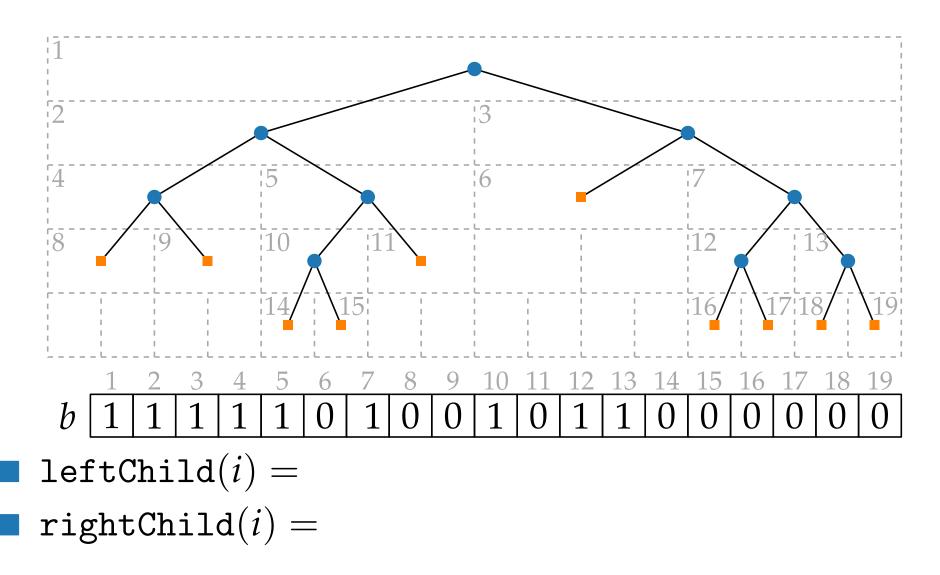
- add external nodes
- read internal nodes as 1
- read external nodes as 0
- use rank and select

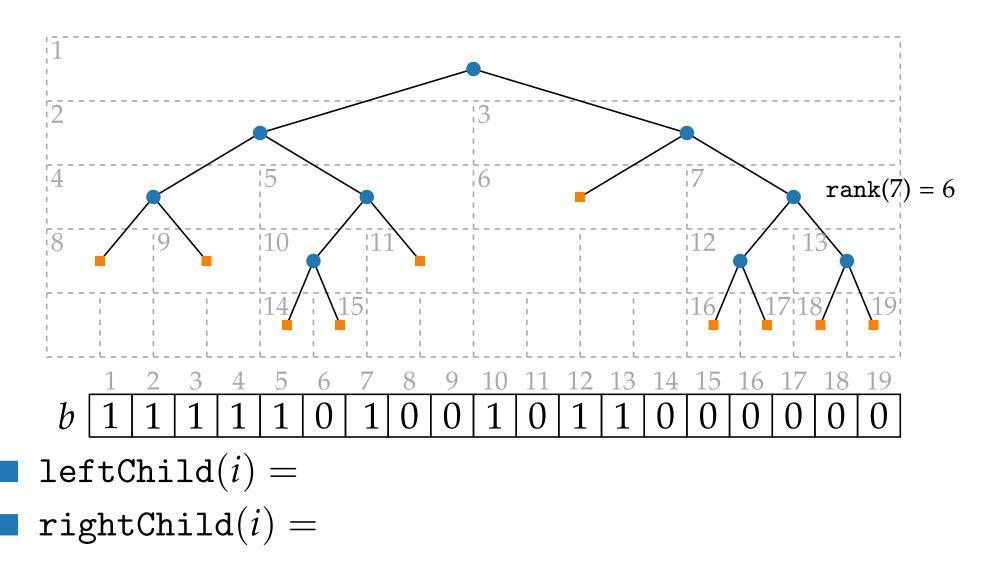


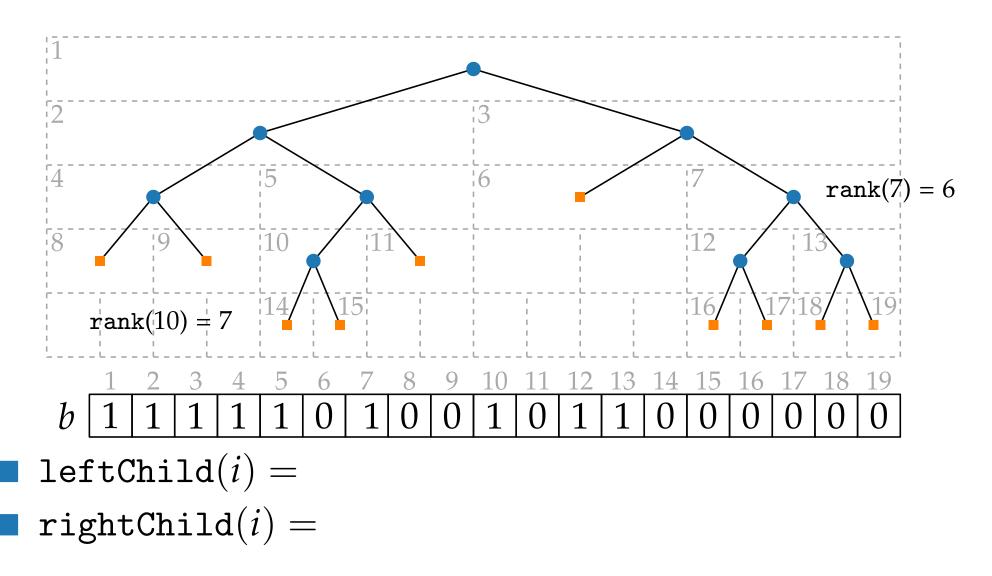


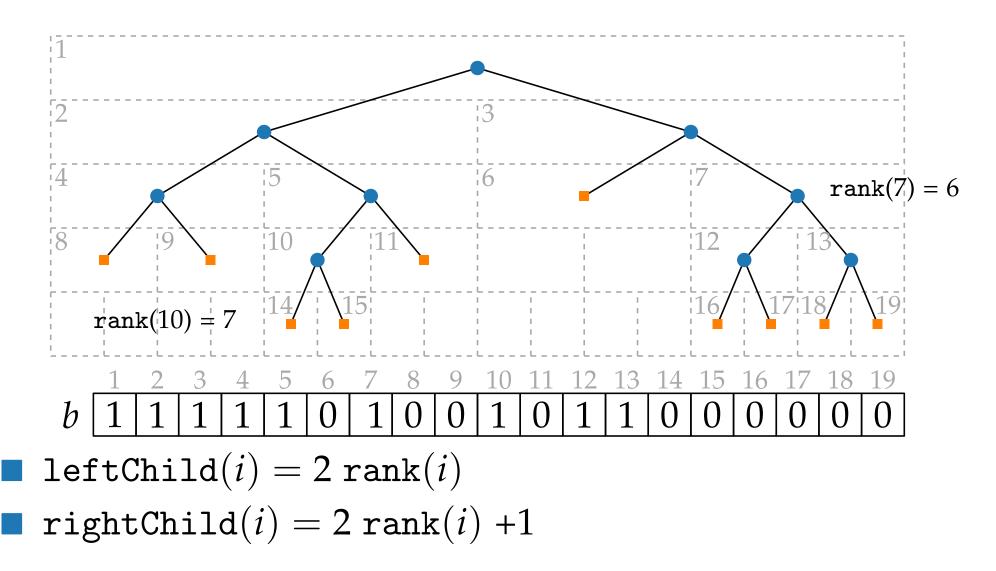


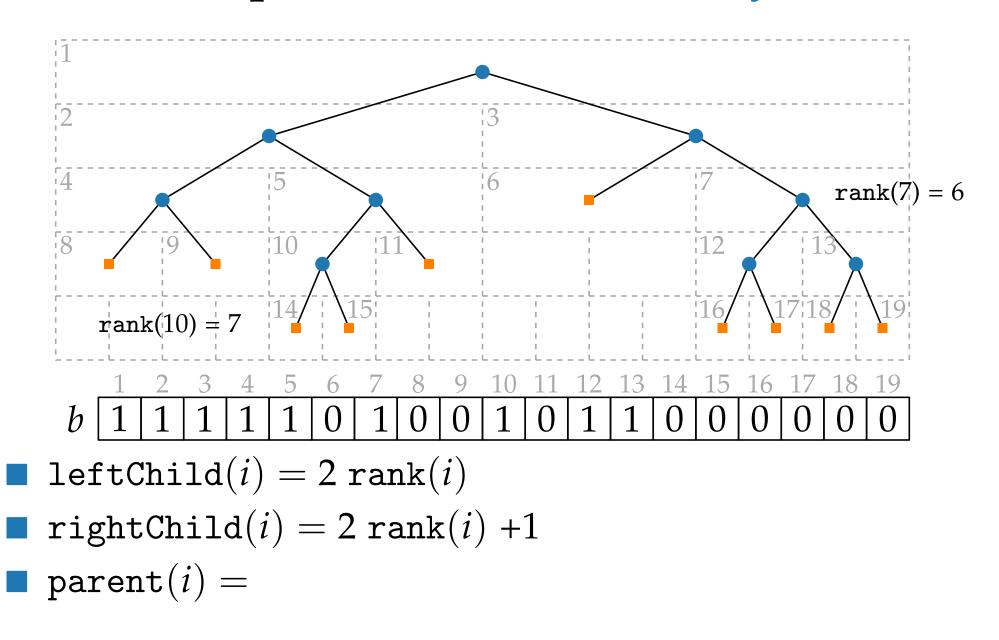


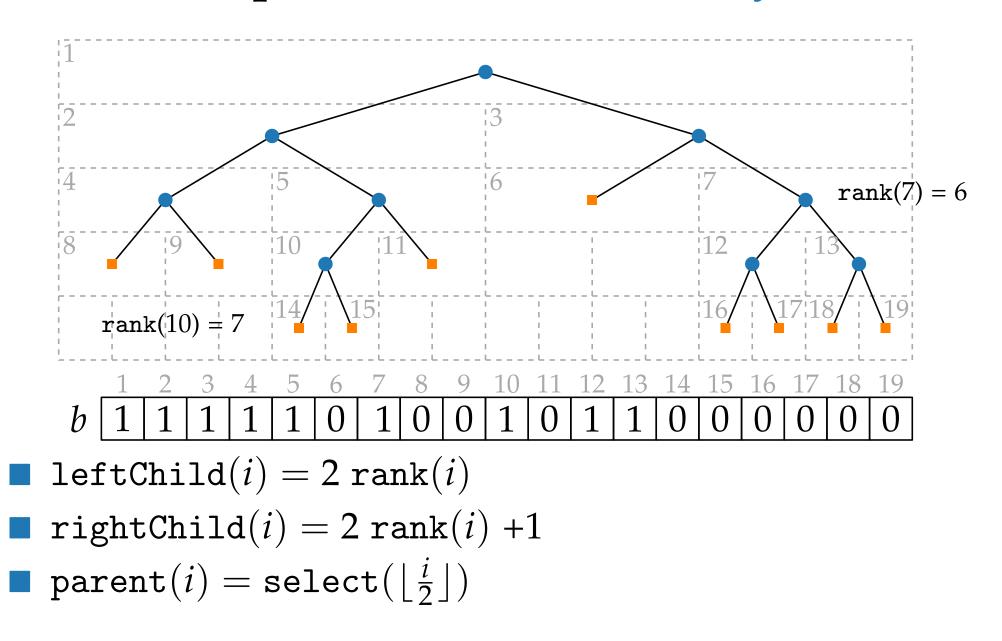


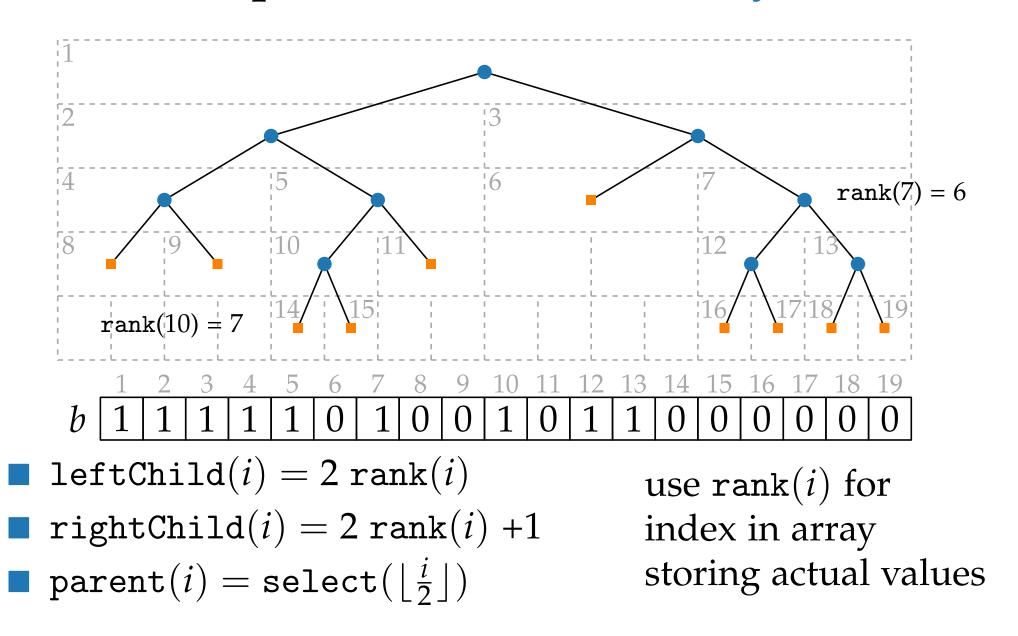


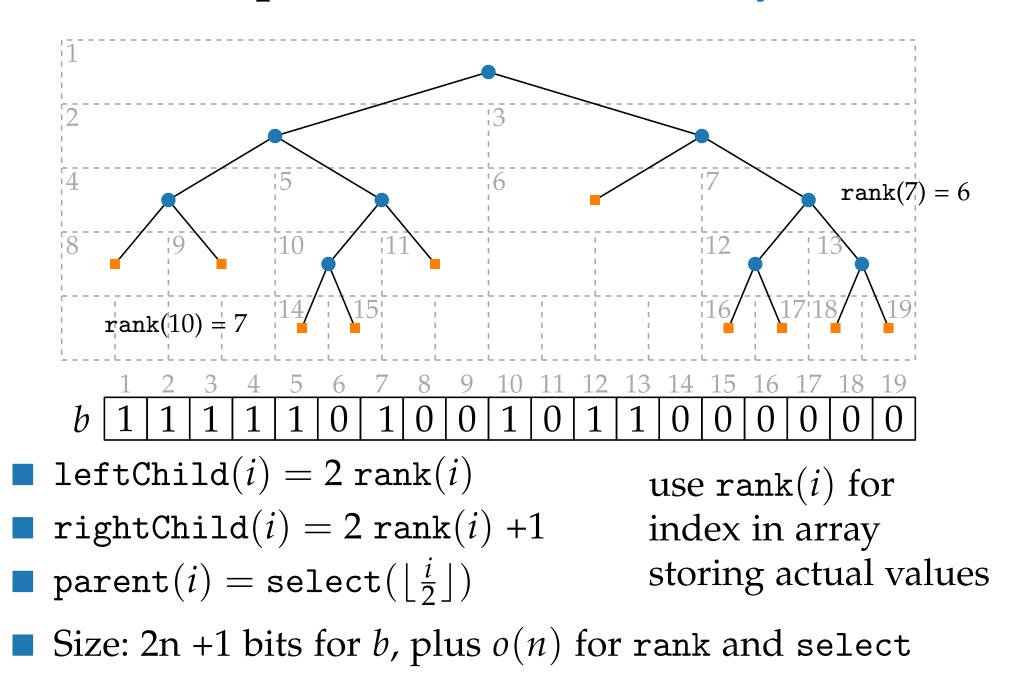




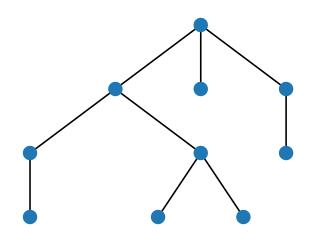




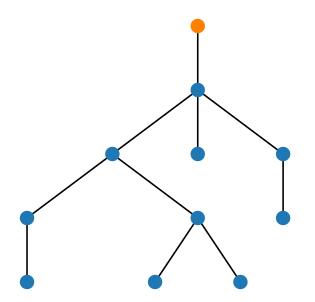




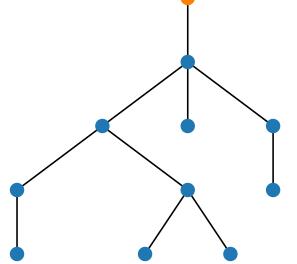
Succinct representation of trees - LOUDS Level order unary degree sequence



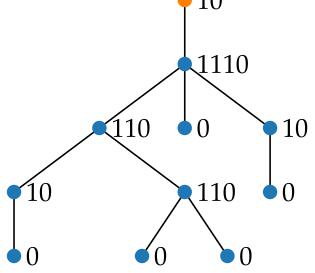
Succinct representation of trees - LOUDS Level order unary degree sequence



Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree



Succinct representation of trees - LOUDS Level order unary degree sequence ¹⁰ unary decoding of outdegree



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Succinct representation of trees - LOUDS Level order unary degree sequence

- unary decoding of outdegree
 gives LOUDS sequence
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each node represented twice use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits

10 - 7 Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence 1110 110 10 0 0 110 each node represented twice 10use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$

10 - 8 Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence 1110 110 10 0 0 1 0 $\mathbf{0}$ except 110 each node represented twice root 10use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$

 $\texttt{firstChild}(\textbf{8}) = \texttt{select}_0(\texttt{rank}_1(\textbf{8})) + 1$

10 - 9 Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence 1110 · 0 · 110 10 0 1 0 $\mathbf{0}$ except 110 each node represented twice root 10use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$ $firstChild(8) = select_0(rank_1(8)) + 1$ = select₀(6) + 1

10 - 10 Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence 1110 110 10 0 $\mathbf{0}$ 0 $\mathbf{0}$ except 110 each node represented twice root 10 use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$

$$\label{eq:select_0} \begin{split} \texttt{firstChild}(\mathbf{8}) &= \texttt{select}_0(\texttt{rank}_1(\mathbf{8})) + 1 \\ &= \texttt{select}_0(6) + 1 = 10 + 1 = 11 \end{split}$$

10 - 11 Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence .1110 110 10 0 1 0 0 $\mathbf{0}$ except each node represented twice root 110 10 use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$ $firstChild(8) = select_0(rank_1(8)) + 1$ = select₀(6) + 1 = 10 + 1 = 11

nextSibling(i) = i+1

10 - 12 Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence 1110 110 10 0 0 1 0 1 0 except each node represented twice root 110 10 use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$ $firstChild(8) = select_0(rank_1(8)) + 1$ = select₀(6) + 1 = 10 + 1 = 11 Exercise: child(i, j)nextSibling(i) = i + 1with validity check

10 - 13 Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence .1110110 10 0 0 0 1 0 except each node represented twice root 110 10 use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$ $firstChild(8) = select_0(rank_1(8)) + 1$ = select₀(6) + 1 = 10 + 1 = 11 Exercise: child(i, j)nextSibling(i) = i + 1with validity check $parent(i) = select_1(rank_0(i))$

Succinct representation of trees - LOUDS Level order unary degree sequence unary decoding of outdegree 10 gives LOUDS sequence **1**110 Í**1**0 10 0 0 0 0 except each node represented twice root 110 10 use index of its corresponding 1 $\Rightarrow 2n + o(n)$ bits $firstChild(i) = select_0(rank_1(i)) + 1$ $firstChild(8) = select_0(rank_1(8)) + 1$ = select₀(6) + 1 = 10 + 1 = 11 Exercise: child(*i*, *j*) nextSibling(i) = i + 1with validity check $parent(i) = select_1(rank_0(i))$ $parent(8) = select_1(rank_0(8)) = select_1(2) = 3$

Discussion

- Succinct data structures are
 space efficient
 support fast operations but
 - are mostly static (dynamic at extra cost),
 - number of operations are limited,
 - complex \rightarrow harder to implement

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- Succinct data structures are
 space efficient
 support fast operations but
 - are mostly static (dynamic at extra cost),
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 - complex \rightarrow harder to implement
- Rank and select form basis for many succinct representations

References

- Lecture 17 of Advanced Data Structures (MIT, Fall'17) by Erik Demaine
- see also Lecture 18 on compact & succinct suffix arrays & trees

References

- Lecture 17 of Advanced Data Structures (MIT, Fall'17) by Erik Demaine
- see also Lecture 18 on compact & succinct suffix arrays & trees
- Guy Jacobson "Space efficient Static Trees and Graphs", FOCS'89
- also contains how to store planar graphs in linear space