

Wavelet Trees

CSCI 7000 - Advanced Data Structures
Final Project

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Navarro, G. (2014). Wavelet trees for all. Journal of Discrete Algorithms, 25, 2-20.

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Motivation

- Let's look at some problems on a sequence.
 - Counting occurrences of an element till position i ?
 - $O(n)$
 - Finding the position of the i -th occurrence of an element?
 - $O(n)$
- These problems occur at a lot of places.
- Can we do better?

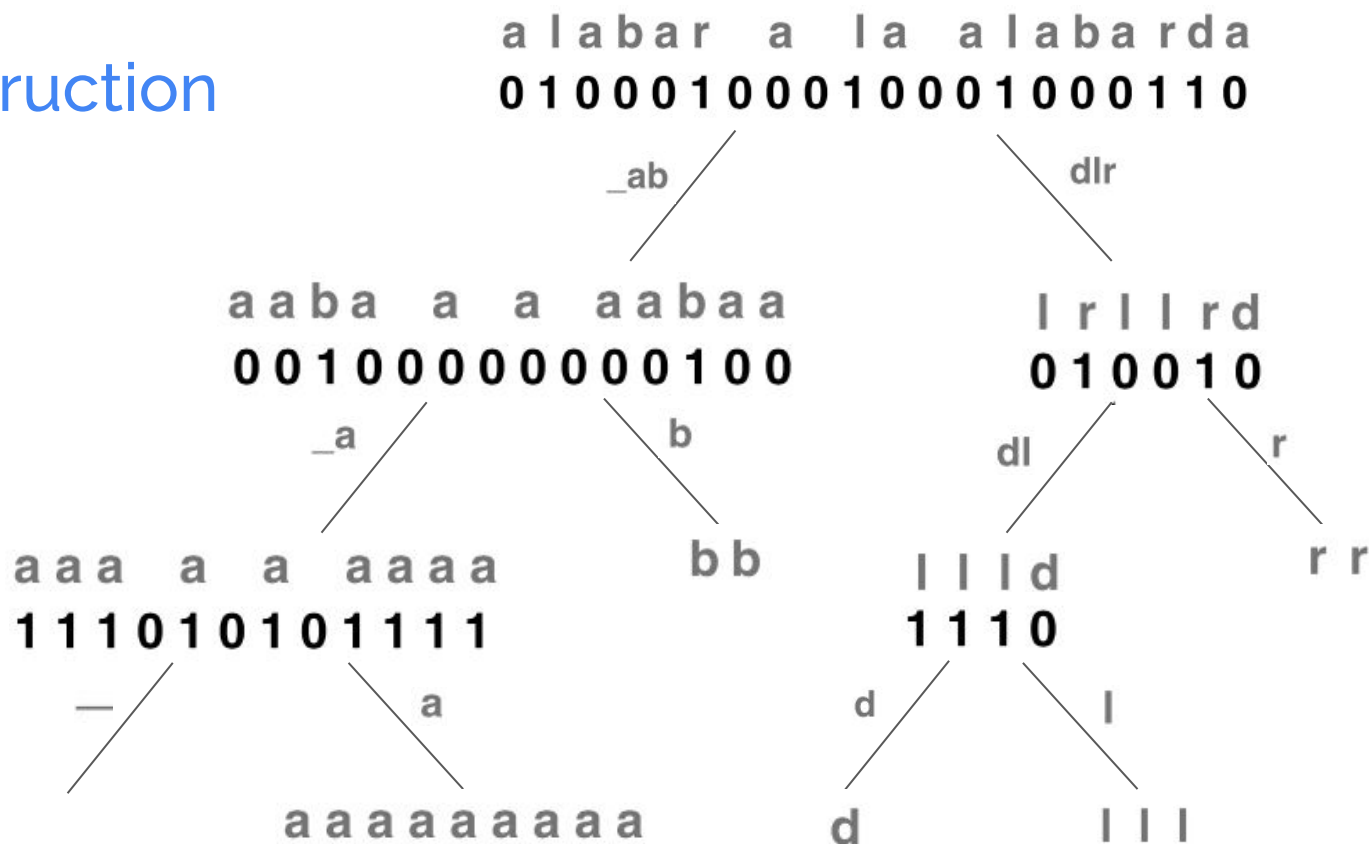
Wavelet Trees

- Wavelet tree is as a space-efficient data structure to represent a sequence and answer some queries on it.
- The name originates from "[wavelet packet decomposition](#)" in signal processing.
- The "[high](#)" and "[low](#)" symbol values of the sequence are separated and the resulting [subsequences](#) are [recursively](#) subdivided.

Wavelet Trees

- Let $S[1,n] = \{s_1, s_2, \dots, s_n\}$ be a sequence of symbols $s_i \in \Sigma$, where Σ is a finite alphabet of size σ
- Example
 - Sequence, $S = \text{alabar a la alabarda}$
 - Alphabet, $\Sigma = \{_, a, b, d, l, r\}$
 - Alphabet size, $\sigma = 6$

Construction



Binary Operations

- Rank
 - $\text{rank}_0(i)$ returns the count of unset bits till position i
 - $\text{rank}_1(i)$ returns the count of set bits till position i
- Select
 - $\text{select}_0(i)$ returns the position of the i -th unset bit
 - $\text{select}_1(i)$ returns the position of the i -th set bit

Binary Operations

bitmap = 1001

- $\text{rank}_0(4)$ = count of unset bits till position 4 = 2
- $\text{rank}_1(2)$ = count of set bits till position 2 = 1

- $\text{select}_0(1)$ = position of the 1st unset bit = 2
- $\text{select}_1(2)$ = position of the 2nd set bit = 4

Wavelet Tree Operations

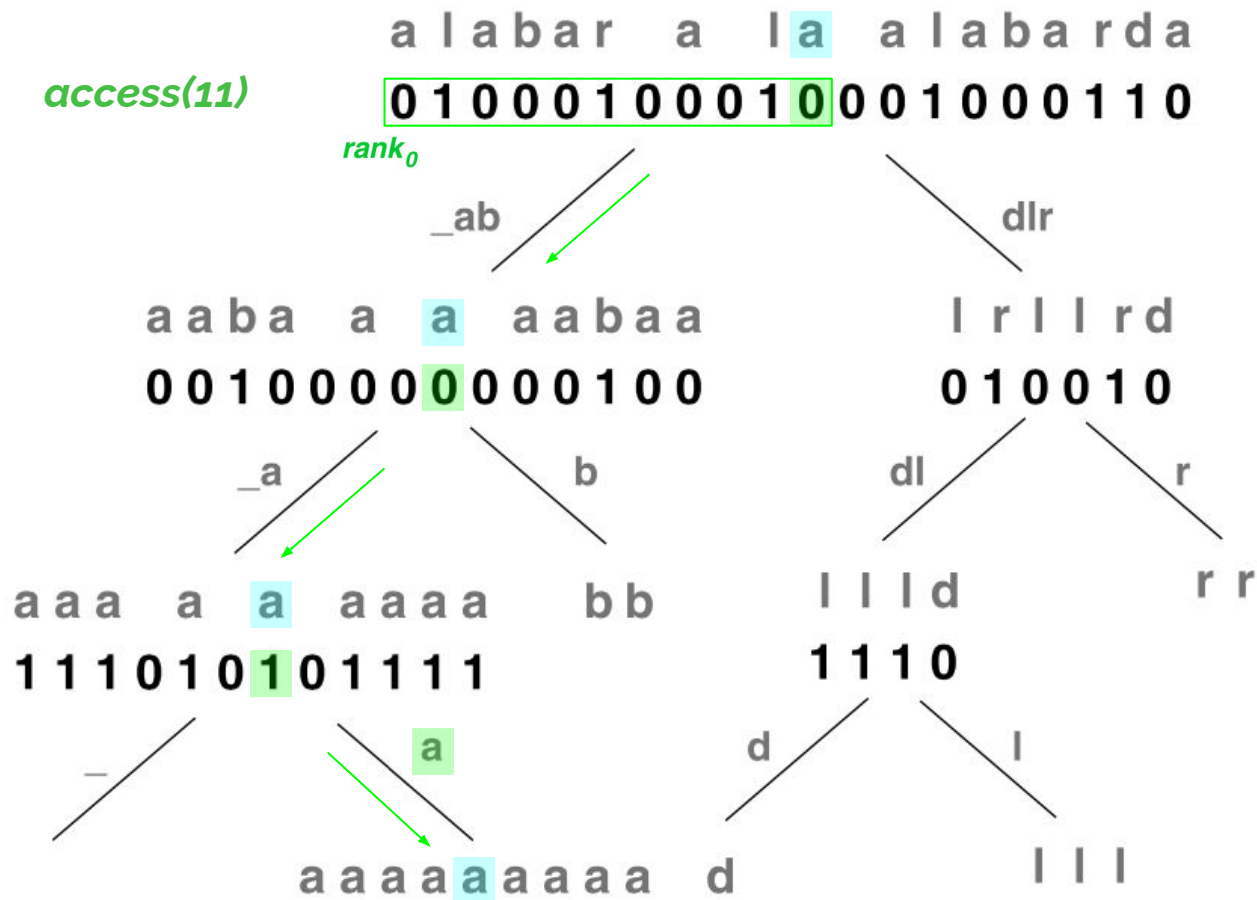
- Access
 - `access(i)` returns $S[i]$
- Rank
 - `rankc(i)` returns the count of occurrences of element c till position i
- Select
 - `selectc(i)` returns the position of the i -th occurrence of element c

Access

- `access(i)` returns $S[i]$
- Check for the bit at i and calculate $\text{rank}_{0/1}$
- Update i and traverse down to the leaf based on i .
- Return i -th leaf element.

Access

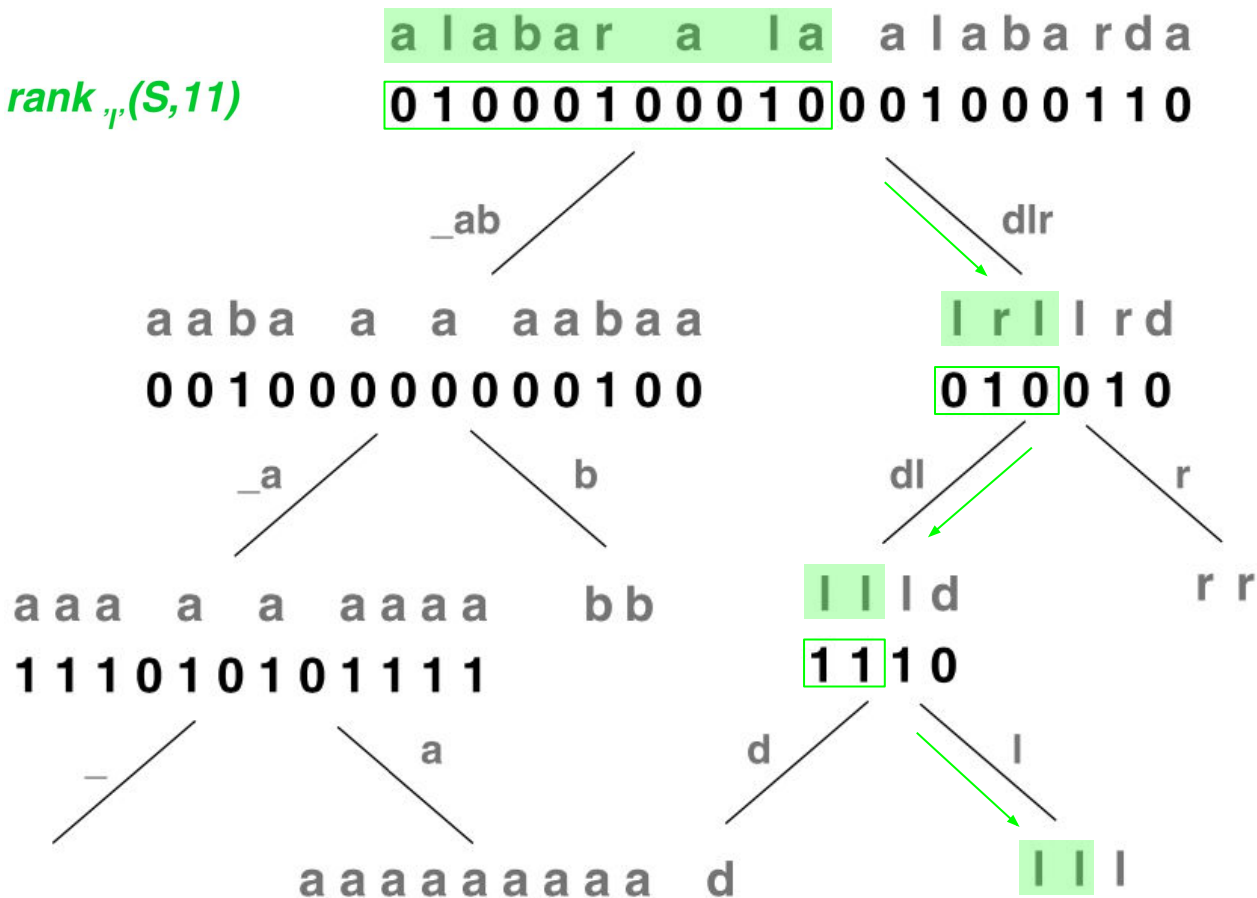
access(11)



Rank

- $\text{rank}_c(i)$ returns the count of occurrences of element c till position i
- Update i using $\text{rank}_{0/1}$ based on the child node.
- Traverse depending on the bit of the element in the current node.
- Return i .

Rank

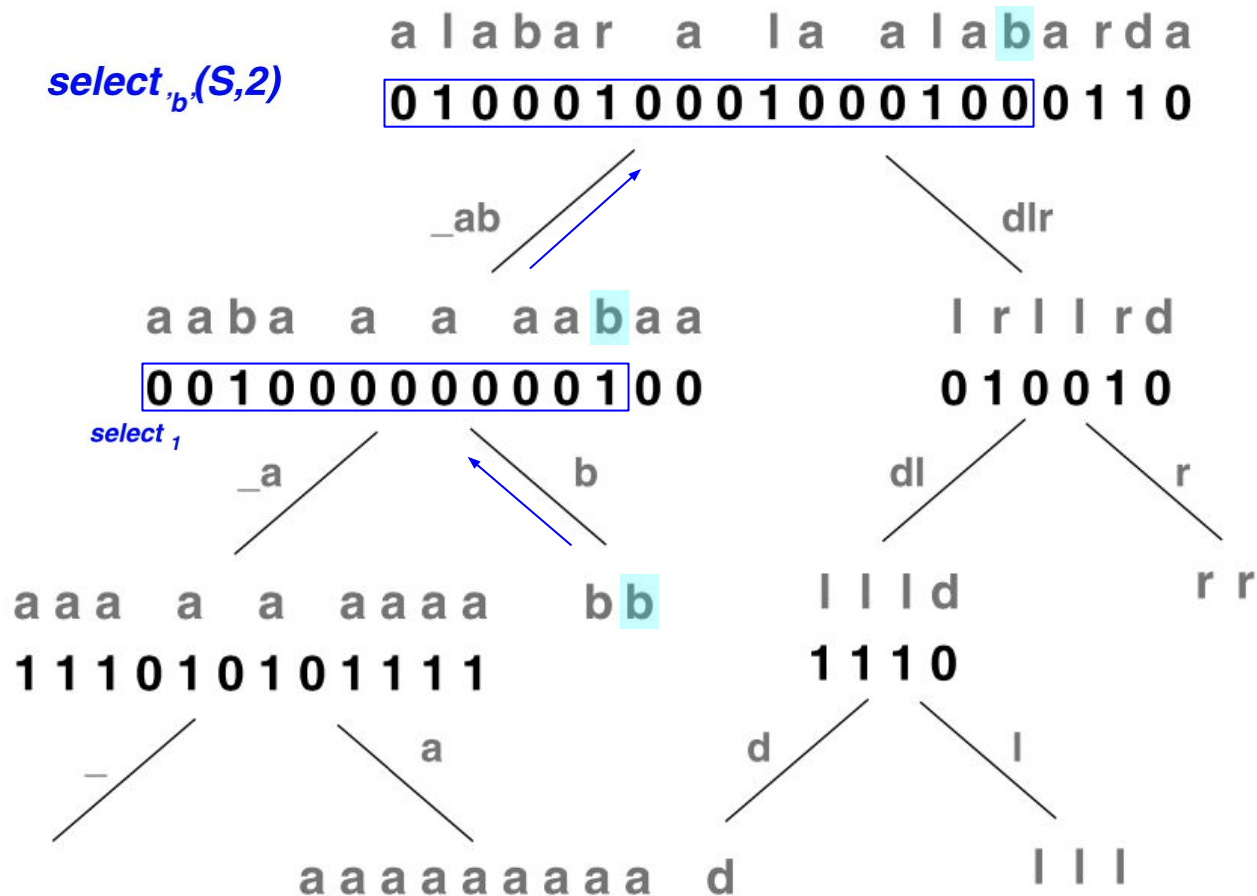
 $rank_{,,}(S, 11)$ 

Select

- $\text{select}_c(i)$ returns the position of the i -th occurrence of element c
- Start from the i -th position of the element in the corresponding leaf node.
- Traverse up to the root while finding the position of 0/1 using $\text{select}_{0/1}$ depending on if node is left/right child, up to i -th occurrence.
- Return the final position.

Select

select_b(S,2)



Space Analysis

- Balanced Wavelet Tree (on the alphabet)
 - Height of tree = $\lceil \log_2 \sigma \rceil$
 - Exactly n bits at each level. At most n bits at the last level.
 - Total space = $n \lceil \log_2 \sigma \rceil$ bits (upper bound)
-
- What about space taken by pointers?
 - $\sigma - 1$ internal nodes, each pointer taking `wordsize` bits of space
 - Total space = $O(\sigma \cdot \text{wordsize})$ bits

Time Analysis - Access

Looking for symbol at position i in the Tree. Starting from the root, at each level,

- Find child: $O(1)$ using the bitmap at each level
- Find position of i in the child: $O(\text{time for rank})$ using $\text{rank}_{0/1}$ on the bitmaps

There are $\lceil \log_2 \sigma \rceil$ levels.

Thus, total time =

- $O(n \log_2 \sigma)$ - naive rank using linear scan
- $O(\log_2 \sigma)$ - precomputed rank

Time Analysis - Rank

Finding rank of symbol s , up to given index i in the Wavelet Tree. Starting from the root, at each level,

- Find child: $O(1)$ using the bitmap at each level
- Find corresponding index for i in the child: $O(\text{time for rank})$ using $\text{rank}_{0/1}$ on the bitmaps

There are $\lceil \log_2 \sigma \rceil$ levels.

Thus, total time =

- $O(n \log_2 \sigma)$ - naive rank using linear scan
- $O(\log_2 \sigma)$ - precomputed rank

Time Analysis - Select

Finding index of the i -th occurrence of symbol s in the original sequence.
Traversing upwards from the leaf, at each level,

- Find child bit from which we traversed up: $O(1)$.
- Find index of i -th occurrence of child bit: $O(\text{time for select})$ using $\text{select}_{0/1}$ on the bitmaps.

There are $\lceil \log_2 \sigma \rceil$ levels.

Thus, total time =

- $O(n \log_2 \sigma)$ - naive select using linear scan
- $O(\log_2 \sigma)$ - precomputed select

Binary Rank & Select Optimizations

- **POPCNT** Instruction - $O(1)$ space, $O(n/\text{wordsize})$ time.
 - Native CPU instruction for **population count** (number of set bits in a machine word).
- Naive Precomputation - $O(n \log_2 n)$ space, $O(1)$ time.
 - At each node, precompute ranks for each bit in the bitmap.
 - Store it in an array of size = $\text{len}(\text{bitmap})$

Binary Rank & Select Optimizations

- Smart Precomputation - $O(n)$ space, $O(1)$ time.
 - Essentially, divide the bitmap into equal-length **samples**.
 - Store rank value for each **sample**.
 - Further divide each **sample** into equal-length **subsamples**.
 - Store **offset** rank values (from previous **sample**) for each **subsample**.
 - Use **popcount** within the **subsample** to get the final rank value.

Succinctness

- Succinct data structure: Uses amount of space almost as much as the information theoretical lower bound, e.g. Heap
- Succinctness in Wavelet Trees: Compressed bitmap representation
- The compression that can be achieved for a sequence S is given by its Empirical Zero-Order Entropy, $H_0(S)$.
- What is entropy?
 - Essentially, it's the number of bits required to represent each member of a set.
 - For a set of size n , worst-case entropy, $H_{wc}(S) = \log_2 n$

Empirical Zero-Order Entropy

- Compression of a sequence S is achieved by using the difference in frequencies of individual elements in S .
- Empirical zero-order entropy gives the amount of compression that can be achieved.

$$H_0(S) = \sum_{c \in \Sigma} (n_c/n) \log_2(n/n_c)$$

where n_c is the number of occurrences of c in S

Empirical Zero-Order Entropy

- If bitmaps of a wavelet tree are compressed to H_0 , then overall space occupied by the tree = $n \cdot H_0(S)$
- Thus, any zero-order entropy coded bitmap representation with $O(1)$ rank and select can be used to get a succinct wavelet tree.
- Example: 'succinct indexable dictionary' by Raman, Raman, and Rao.

Representations

- As a sequence of values
 - The wavelet tree on a sequence $S[1,n] = \{s_1, s_2, \dots, s_n\}$ represents the values s_i .
- As a reordering
 - The wavelet tree structure describes a stable ordering of the symbols in S .
- As a grid of points
 - We have an $n \times n$ grid with n points so that no two points share the same row or column.

Applications

- As a sequence of values
 - Positional inverted indexes, Full-text indexes, Graphs, etc.
- As a reordering
 - Generic numeric sequences, Permutations, Document retrieval indexes, etc.
- As a grid of points
 - Discrete grids, Binary relations, Colored range queries, etc.

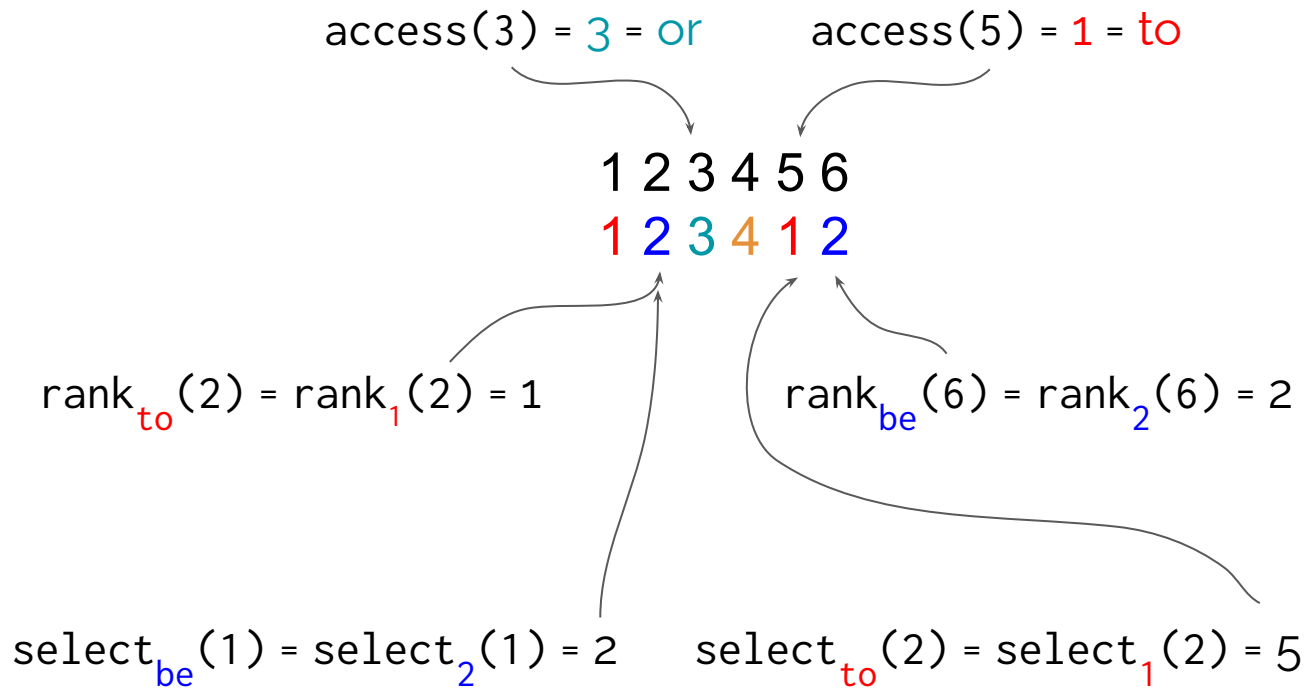
Positional inverted indexes

- A positional inverted index is a data structure that stores, for each word, the list of the positions where it appears in the collection.
- Example
 - Sequence, S = to be or not to be
 - Alphabet, Σ = to be or not
 - Alphabet size, $\sigma = 4$

Positional inverted indexes

to be or not to be

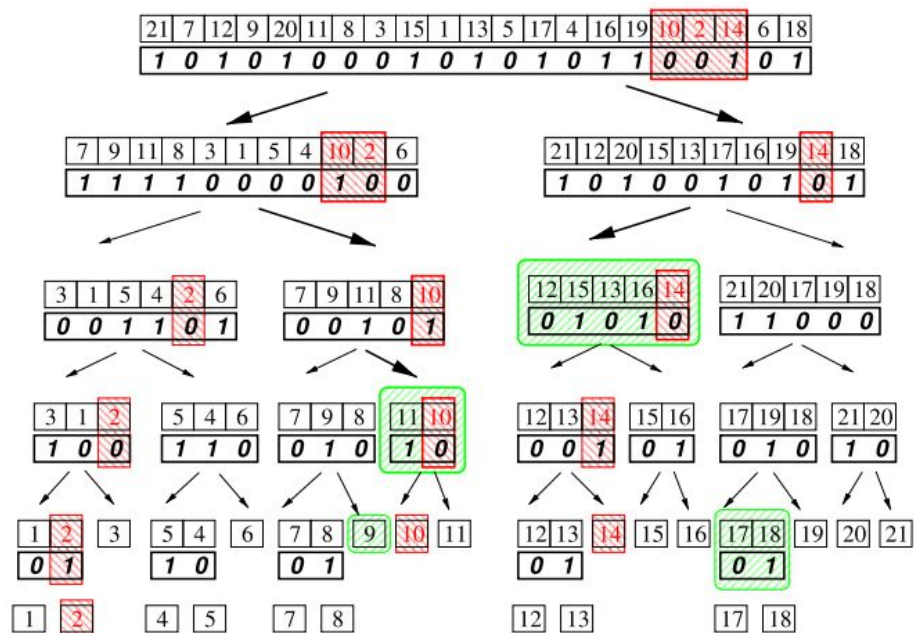
1	to	1, 5
2	be	2, 6
3	or	3
4	not	4



Range quantile query

- For a sequence of numbers $S[1,n]$ on the domain $[1..\sigma]$, given a range $[1,r]$ and a value k , we can compute the k -th smallest element in $S[1,r]$.
- It uses the **rank** operation to successively narrow down the range until we arrive at a leaf, whose label is the k -th smallest element in $S[1,r]$.
- Nontrivial compared to Positional inverted indexes

A 21x21 grid with a green shaded region (rows 9-18, columns 1-18), a red shaded region (rows 1-8, columns 19-21), and a purple cross-hatched region (rows 9-18, columns 19-21). Blue dots are scattered across the grid.



Implementation

- We implemented a wavelet tree library in C++ that represents a wavelet tree as a `sequence`, along with `precomputation`.
- Thanks to `C++ generics`, it supports creation of a wavelet tree of a sequence of any data type that satisfies the following two conditions-
 - Can be `hashed`
 - Can be compared using `<= operator`
- We implemented two applications
 - Positional inverted indexes
 - Range quantile queries

Implementation

Sequence, S : a l a b a r _ a _ l a _ a l a b a r d a

Alphabet, Σ : _ a b d l r

Alphabet size, σ : 6

traverse():

```
|—— level: 0
subsequence: a l a b a r _ a _ l a _ a l a b a r d a
bitmap: 0 1 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 1 0
rank_0: 1 1 2 3 4 4 5 6 7 7 8 9 10 10 11 12 13 13 13 14
select_0: 0 2 3 4 6 7 8 10 11 12 14 15 16 19
select_1: 1 5 9 13 17 18
```

```
|—— level: 1
subsequence: a a b a _ a _ a _ a a b a a
bitmap: 0 0 1 0 0 0 0 0 0 0 0 0 1 0 0
rank_0: 1 2 2 3 4 5 6 7 8 9 10 10 11 12
select_0: 0 1 3 4 5 6 7 8 9 10 12 13
select_1: 2 11
```

```
access(10) = a
access(15) = b
access(-3) = Index out of range!
access(50) = Index out of range!
```

```
rank(l, 11) = 2
rank(b, 16) = 2
rank(z, 3) = Invalid input!
```

```
select(b, 2) = 15
select(a, 6) = 12
select(z, 3) = Invalid input!
```


Conclusion

- Wavelet tree is a versatile **succinct** data structure with a plethora of **applications**.

Thank You!