The level-ancestor problem
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Data: A tree

Query:
- Given a vertex $v$ and a number $k$, find the ancestor of $v$ that is $k$ steps higher in the tree
- Equivalently: Given a vertex $v$ and a number $d$ find the ancestor whose depth (number of steps from root) is $d$

We could just follow parent links up the tree in time $O(k)$ but we want small query time even when $k$ is large
Inefficient solution

Each node $v$ stores $O(\log n)$ ancestors, the ones $k$ steps higher for $k = 1, 2, 4, 8, \ldots 2^i, \ldots$

To find the ancestor $k$ steps higher when $k$ is not a power of two:
- Decompose $k$ into a sum of powers of two (its binary representation)
- Use stored ancestor pointers to jump up by each power

Space $O(n \log n)$, query time $O(\log k)$

We can reduce space but first we need to speed up queries
An easy special case

If the tree is a path, rooted at left end:

- Store an array $A$ of the vertices in the path
- Each node records its position in the array
- If $V$ is in $A[i]$, its $k$-step ancestor is in $A[i-k]$

Linear space, constant query time
Decomposition into long paths

Each non-leaf node selects one child, the one leading to the deepest leaf.

The selected edges form a system of paths covering all non-leaf nodes and some of the leaves.

Remaining leaves form one-vertex paths.

But if we use the path solution for these paths, how do we find ancestors on different paths?
Extended paths

Instead of disjoint paths, extend each path to be twice as long (or all the way to the root if there is no ancestor twice as high)

Store each extended path in an array, and for each tree vertex store both which array it belongs to and its position in the array

Total length of all arrays $\leq 2n \implies$ linear space

Can answer some but not all queries: when $v$ is $h$ steps above a leaf, we can find ancestors $h$ steps above $v$
Combined data structure

Store both power-of-2 ancestors and extended paths

To find the ancestor $k$ steps higher from $v$:

- Make one power-of-2 jump, the biggest one that would still be below the ancestor
  Jump amount $= 2^{\lfloor \log_2 k \rfloor}$

- You are now high enough above a leaf to use extended paths

Constant query time but space is still $O(n \log n)$
(The paths use linear space but the jump tables are too big)

[Bender and Farach-Colton 2004]
Level ancestors and Euler tours

To find the ancestor of $v$ at height $d$:

- Look in the Euler tour of the tree, starting from the last copy of $v$ in the tour
- The next vertex with height $d$ is the ancestor
Log-shaving

Structure:
- Break Euler tour into blocks of length $b = \frac{1}{2} \log_2 n$
- Within block, depths differ by $\pm 1$; label each block by its pattern of $\pm$ choices, a $(b - 1)$-bit binary number
- Precompute tables for queries with answer in same block
- Store power-of-two tables only at the last vertex of each block

Query:
- Use table to find answer in $v$’s block (if it is there)
- If not, vertex $u$ at block end has same level-$d$ ancestor as $v$
- Use the jump table stored at $u$ to find an ancestor $w$ high enough that we can use the extended paths for $w$

[Berkman and Vishkin 1994]
The list ordering problem
List ordering vs tree ancestors

Simplification of common ancestor problem:
Test whether one tree vertex is an ancestor of another, or not

Data structure:
Number in preorder
Number in postorder

Ancestor = earlier in preorder and later in postorder
Dynamic order comparison

Maintain a collection of elements ordered into a list

Operations:
- Insert $x$ at the start or end of the list
- Insert $x$ immediately before or after another element $y$
- Find the element immediately before or after $x$
- Remove $x$
- Test whether element $x$ is earlier than or later than element $y$

Most operations can easily be done in constant time, for example by using doubly linked lists

The only missing one: testing relative ordering
Typical properties of numbers of buildings in US streets:

- Ordered: number tells you relative position along street
- Usually small integers
- Not necessarily consecutive: there may be gaps in the numbering
- Renumbering is expensive, so don’t do it very often

Intuition: apply similar scheme to list ordering by numbering list elements and using numbers to test relative position
Partial history

- Dietz [1982]: logarithmic update, $O(1)$ order-comparison
- Tsakalidis [1984]: constant amortized update and comparison
- Dietz and Sleator [1987]: maintain ordered numbering, all numbers polynomially large, constant amortized update, complicated
- Bender et al. [2002]: simplification of same results
- Bender et al. [2017]: worst case rather than amortized
- Devanny et al. [2017]: few relabelings per element

We will follow Bender et al. [2002]
Application of house numbering: Dynamic arrays, revisited
Dynamic arrays with insertion and deletion

Suppose we want to maintain a sequence of values with the following operations:

- Look up the value at position \( i \) in the sequence
- Change the value at position \( i \) in the sequence
- Add a new value at position \( i \) in the sequence, shifting all later values to higher positions
- Remove the value at position \( i \) in the sequence, shifting all later values to earlier positions

Dynamic arrays allow only the first two operations, and add/remove at the end of the array; adding and removing fast lookup of the element at position \( i \), and fast insertion or deletion at the end of the array.

What if we want to extend arrays to allow insertion or deletion at other positions?
Store the sequence in a binary search tree augmented for ranking and unranking

(The sequence order is the left-to-right tree order; we don’t need to store keys with the
tree nodes, only the associated array values, so house numbering not needed.)

To find or change the value at position $i$: use unrank to find its tree node

To add or remove a value: standard binary search tree insertion/deletion operations
Dynamic arrays from house numbering

Maintain:

- House-numbering solution for sequence of elements
- Dictionary mapping house numbers to linked list nodes
- Structure for ranking and unranking house numbers with $O(\log n / \log \log n)$ time per operation (mentioned briefly last week)

To find or change the value at position $i$: use unrank to find its house number and then use that number as a dictionary key.

To add or remove a value, update the house numbers and propagate any changes in numbering to the ranking/unranking structure.
House numbering solution
Terminology

We will store a doubly linked list, whose elements are called keys.

Because it’s stored as a linked list, we can quickly find adjacent keys. The two adjacent keys are the predecessor and successor.

We wish to assign numbers to the keys to allow fast comparisons of their positions; these numbers are called tags.

We want to maintain a correspondence keys \( \rightarrow \) tags so that the numerical ordering of tags = the list ordering of keys.
Main idea

Delete a key ⇒ do not renumber other tags

If we insert key \( x \), and there is any available tag \( i \) between tags of its predecessor and successor, set \( \text{tag}(x) = i \)

Remaining case: Partition possible tag values recursively into ranges of tags with power-of-two sizes

Find the smallest range of tags (size \( 2^k \)) surrounding new element location that is used by few keys: fewer than \( c^k \)

Renumber the keys evenly within this range (including \( x \))
Main idea implementation details

The parameter $c$ can be any fixed number in range $1 < c < 2$ (smaller $c \Rightarrow$ bigger tags; larger $c \Rightarrow$ more renumberings)

To find a range of tags that is used by few keys: scan left and right from $x$ in the sequence, finding increasingly large ranges in hierarchical partition of tags, until finding a range with few keys

Let $k = \log_c n$, and let $\alpha = \log_c 2$. If max tag $> n^\alpha$, then range of all tags is bigger than $2^k$ and holds only $n = c^k$ keys, so $\exists$ range with few keys and search terminates

When search terminates, time it took to find the range and renumber its elements are both proportional to $\#$ keys in it
Main idea analysis

When we renumber a range of size $2^i$, left or right half-range is full.

(If both half-ranges had few elements, we would have renumbered one of them before getting to the larger range.)

Therefore, when we renumber a range of size $2^k$, we renumber between $c^{k-1}$ and $c^k$ keys and it takes total time $\Theta(c^k)$

After renumber, each half-range has $\leq \frac{c}{2}c^{k-1}$ keys, below full by a factor of $c/2 \Rightarrow$ cannot fill up again before we do another $\Omega(c^k)$ insertions $\Rightarrow$ amortized time for ranges of size $2^i$ is $O(1)$

The same analysis holds separately for each choice of $i$, but there are $O(\log n)$ choices $\Rightarrow$ total amortized time per update is $O(\log n)$
Log-shaving

To achieve constant instead of logarithmic amortized time per update, again use a blocking strategy:

- Group keys into dynamic blocks of logarithmic length
- Split block when it gets too long; merge pairs of consecutive blocks when total length small enough
- Use main idea to number blocks
- Allocate polynomially many tags within each block
- New key in a block gets average of predecessor and successor
- Renumber all keys in a block when block structure changes or when a new element has no tag; this happens only $O(1)$ times per $O(\log n)$ insertions
Summary
Representation of trees and binary trees with $2n$ bits
Blocking and table lookup strategy for saving logarithmic factors in the space bounds for many data structures
Common ancestor problem and its applications to shortest paths and bandwidth maximization
Equivalence between common ancestors and range minima
Common ancestors in $O(n)$ space and $O(1)$ query time
Maintaining order in a list in $O(1)$ amortized time
Level ancestors in $O(n)$ space and $O(1)$ query time
References and image credits, 1


References and image credits, II


