

Exam Logistics

Take-home kept in the registrar's office

- I'll create repositories like I did for the midterm, including
 - ▶ Instruction page
 - ▶ Blank question template
 - ▶ Not required to typeset (registrar gives out “blue books”, or use your own paper or write on exam, or combination of the above)
 - ▶ But must print out and submit your exam inside registrar's envelope
- **Review session:**
 - ▶ Monday evening from 7:30-8:30pm in TCL 206
- **Practice questions:**
 - ▶ Best tools are homework and questions found in slides
 - ▶ Rule of thumb for exam questions: will be representative of things you've done on homework or in-class activities (POGIL sessions in Ward Lab)

30-minute Question Review Sessions

- **Will be sign-up slots for all day Monday and Tuesday**
 - ▶ Happy to do 1-off meetings if those don't work for you
 - ▶ Not at all required; don't need to do 4 questions; use as you'd like
 - ▶ Email me which questions you wish to go over so I can have them ready

Office & TA hours

Many TA hours will happen

- Most TAs were happy to host normal hours over reading period
 - ▶ I'll update the course schedule once I get their finalized hours
- 30-minute slots can be used however you'd like
- I'm happy to add additional office hours throughout the week, on-demand!
 - ▶ Let me know in advance and I can advertise a time to the mailing list

B-trees (Ubiquitous and otherwise)

Lecture Outline

Indexing Overview

- General task & properties

DAM model

- How to analyze external memory algorithms

B-trees

- Operations
- Variants
- Discussion

Not on the exam; goal is to show you the flexibility and power of what you've learned

Why B-trees, why now when the semester is over?

- We've studied many algorithms and data structures; here is an example of how we can change our underlying algorithmic model and then apply the same analysis tools we've built throughout

Big Picture Question:
How do you keep data
organized?

An Analogy from [Comer 79 CSUR]

Filing cabinet with folders of employee records, alpha-sorted by employee last name

- We often think in terms of keys and values. Here?
 - ▶ Keys are the employees' last names
 - ▶ Values are the employee file (held in a folder, one per employee)
- A filing cabinet effectively supports two types of searches
 - ▶ Sequential
 - ▶ read through every folder in every drawer, in order
 - ▶ Random (targeted)
 - ▶ use the labels on the drawers & folders to find the single record of interest

In this scenario, the filing cabinet is an index of employee records

Indexes (yes, colloquially pluralized that way)

Indexes organize data

- Random (targeted) searches utilize an *index* to:
 - ▶ Direct our search towards a small part of the total data
 - ▶ (Hopefully) speed up our search

Questions

- ▶ What operations does an index support?
- ▶ How do we quantify index performance?
- ▶ Is the data part of the index, or does the index “sit on top of” the data?
 - ▶ Filing cabinet storing employee folders (data part of index) vs. library card catalog storing book call numbers (pointer to data stored in index)

Q1: What operations does an index support?

Operations

- $\text{Insert}(k,v)$: inserts key-value pair (k,v)
- $\text{Delete}(k)$: deletes any pair $(k,*)$
- $\text{PointQuery}(k)$: returns all pairs $(k,*)$
- $\text{RangeQuery}(k_1,k_2)$: returns all pairs $(k,*)$, $k_1 \leq k \leq k_2$

In short, indexes support the *dictionary* interface.

- Often used for very large data sets.



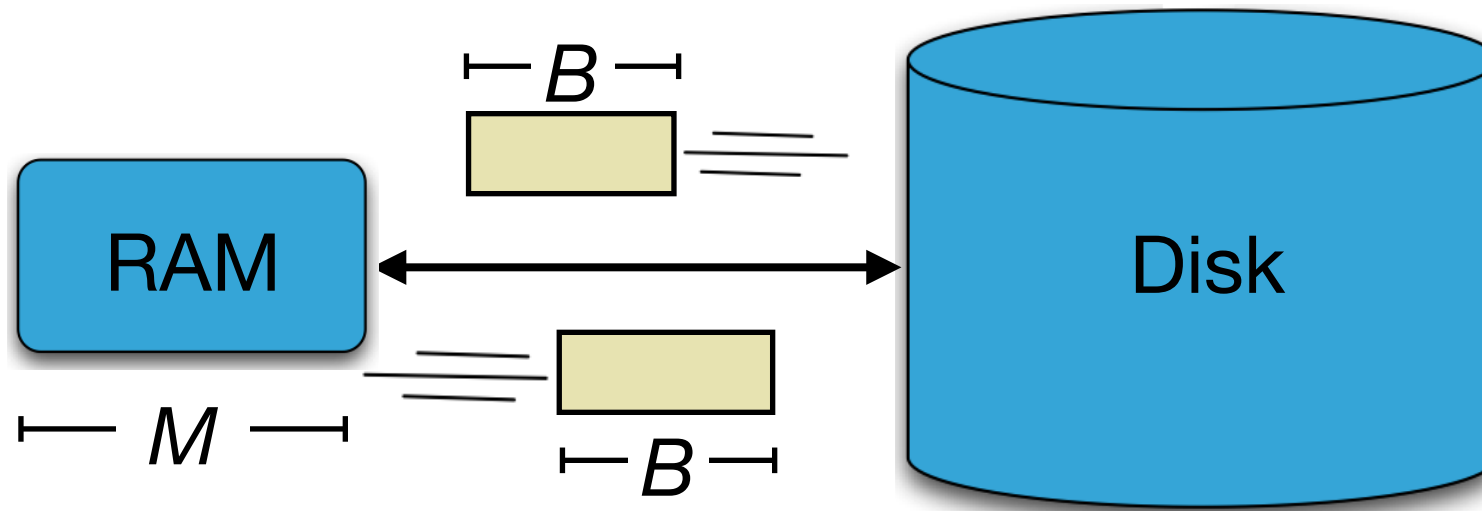
Q2: How to we quantify index performance?

Algorithmically, we can use the DAM model:

- Useful model in scenarios when data is too big for memory
 - ▶ Data is transferred in **blocks** between RAM and disk.
- Premise: the number of block transfers dominates the running time.
 - ▶ Searching through a given block is “free” (once block is in-memory)

Goal: Minimize # of I/Os

- Performance bounds are parameterized by block size B , memory size M , data size N .



[Aggarwal+Vitter '88]

DAM Model an B-tree Analysis

Analyze worst-case costs by counting I/Os

- **B**: unit of transfer
 - ▶ B-tree node size
- **M**: amount of main memory
 - ▶ We can cache M/B nodes in memory at once
- **N**: size of our data
 - ▶ We're not worried about disk space, we use **N** to describe our tree
- We will think about the tree shape (node size, height, fanout), then describe each operation's cost in terms of the DAM model

The B-tree

Terms and Conditions

B-trees store records

- Records are key-value pairs
- We assume that keys are
 - ▶ Unique (to simplify analysis)
 - ▶ Ordered

Terms and Conditions

Rules for our B-trees

- B-ary tree
 - ▶ Internal nodes have between d and $2d$ keys called *pivots*
 - ▶ This means nodes must always be half full!
 - ▶ At least $d+1$ pointers to children (one more pointer than pivot key)
- If an operation would cause a violation of one of these invariants, must rebalance the tree!
- Note: our B-tree's internal nodes **do not** store records
 - ▶ Option 1: Store (key, value) pairs in leaves
 - ▶ Option 2: Store (key, pointer to value) in leaves

Terms and Conditions

Several B-tree variants

- We will describe a “B⁺-tree” here, noting features of specific variants as they come up

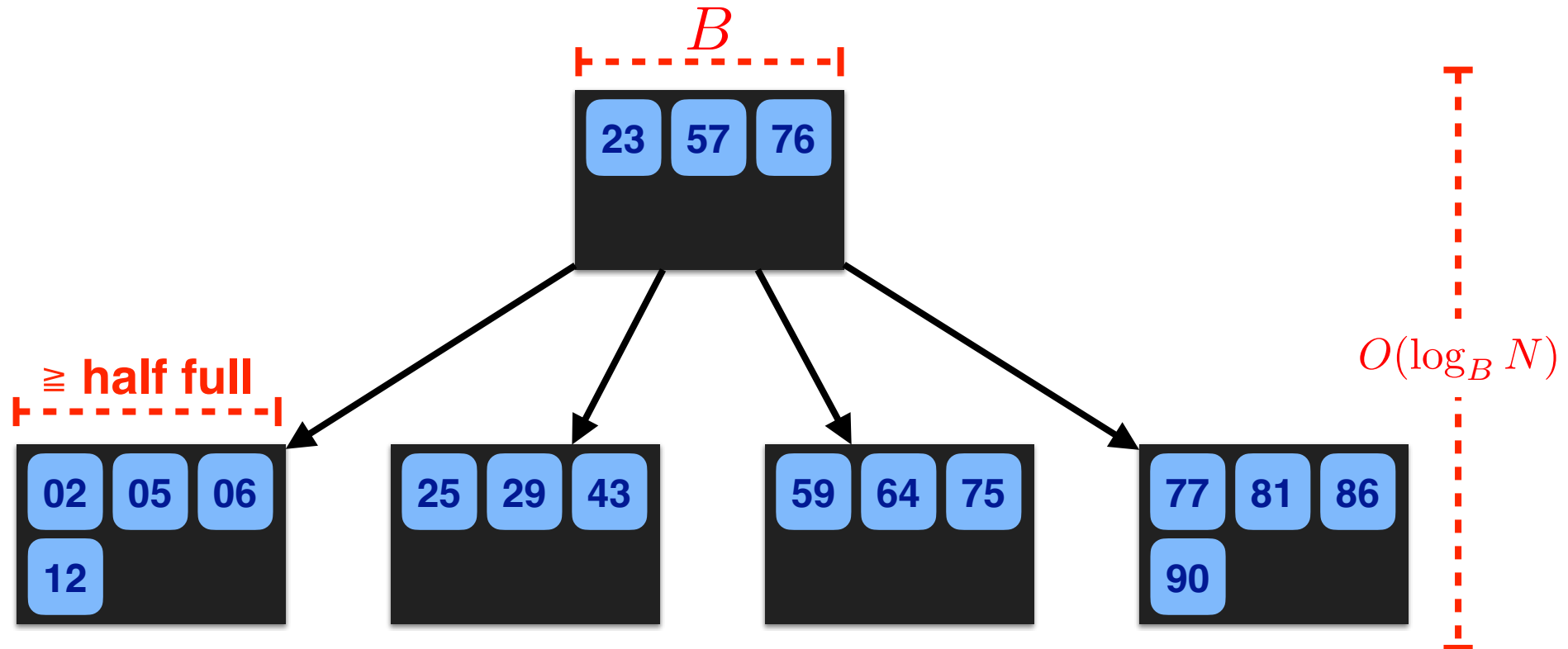
Popular Variants of B-trees

- B-tree: more-or-less what we'll describe here
- B⁺-tree: B-tree where leaves form a linked list
- B^{*}-tree: B-tree where nodes always 2/3 full

B-tree: standard DAM dictionary

B-ary search tree

What does B Stand for?



<u>Summary</u>	Point Query	Insert	Delete	Range Query
B-tree	$O(\log_B N)$	$O(\log_B N)$	$O(\log_B N)$	$O\left(\log_B N + \frac{K}{B}\right)$

B-tree Point Queries

B-tree Point Queries

Steps

- Starting at the root, find the first ***pivot*** key that is larger than your *search key*, and follow the pointer on its left
 - ▶ If there are no pivot keys larger than your search key, follow the last pointer
- Repeat until you arrive at a leaf node
- Search the leaf node (ordered list) for your target key
- Return the key-value pair (if found), or *NONE*

This work is done during an insert (need to find place where new key-value pair belongs), so we will walk through this then.

B-tree Point Queries

Cost

- How many nodes must be read/written in a search?
 - ▶ We read the root node to search the pivot keys
 - ▶ We recurse on the subtree
- Total cost of a search: $O(h)$
 - ▶ Recall $h = O(\log_B N)$

B-tree Insertions

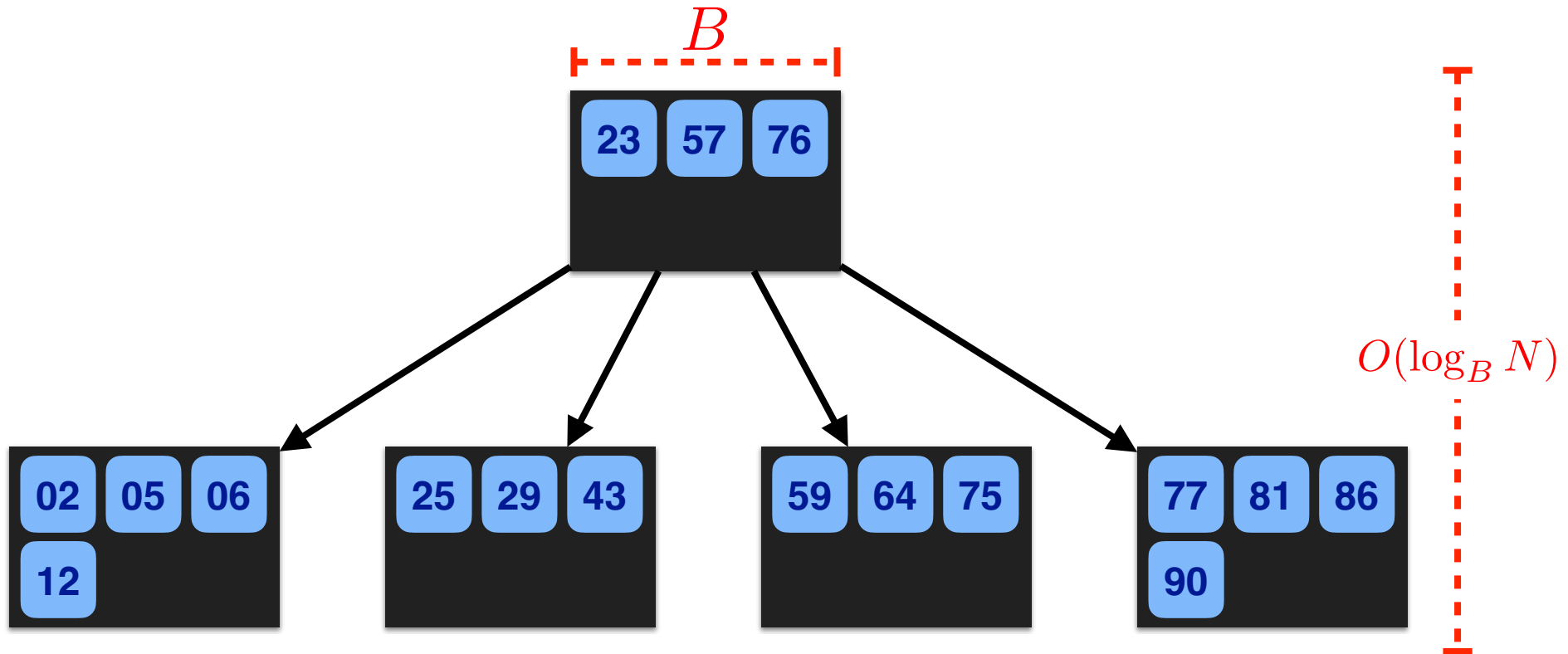
B-tree Insert

Steps

- Find the leaf node where your key-value pair belongs (point query)
- Insert your key-value pair into that leaf

B-tree: standard DAM dictionary

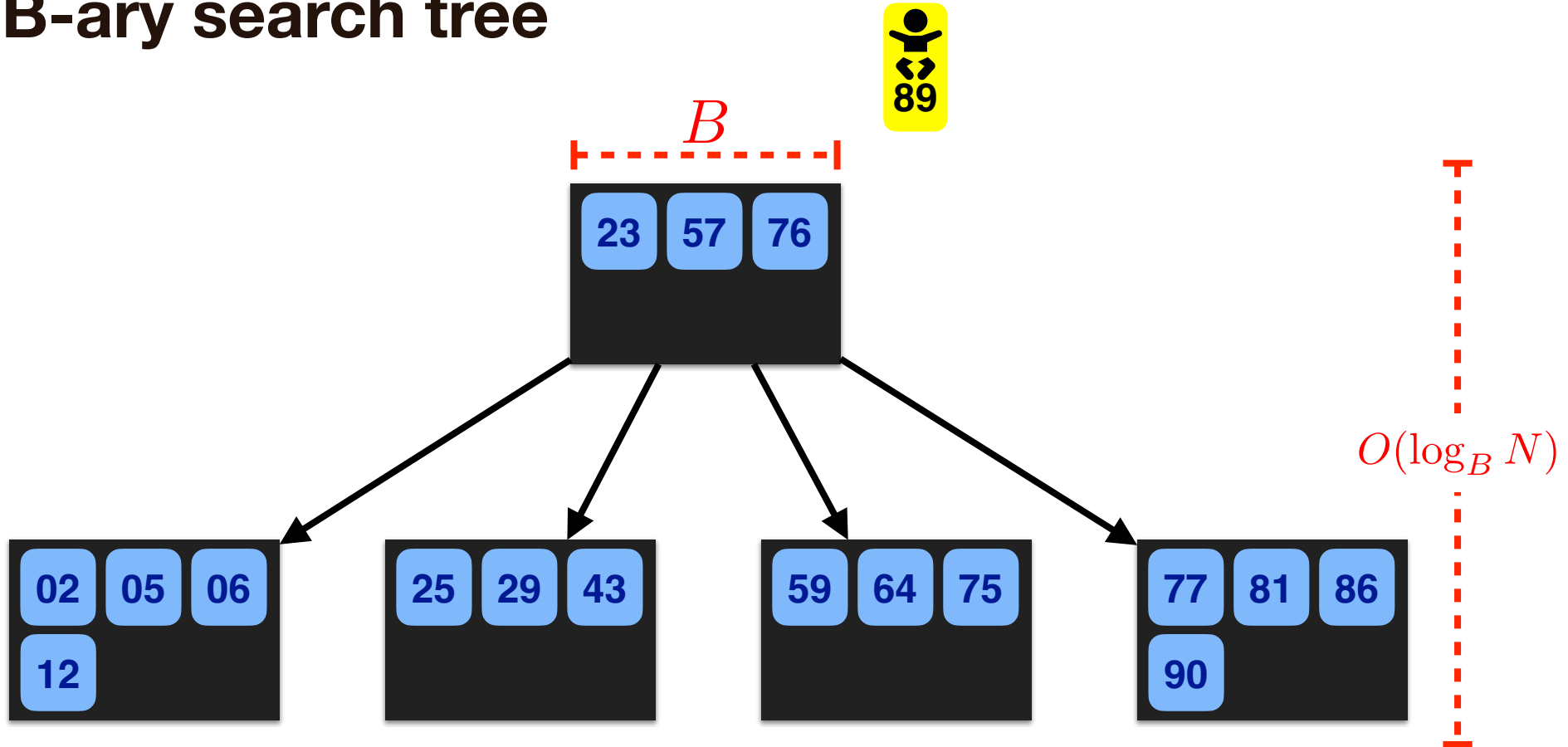
B-ary search tree



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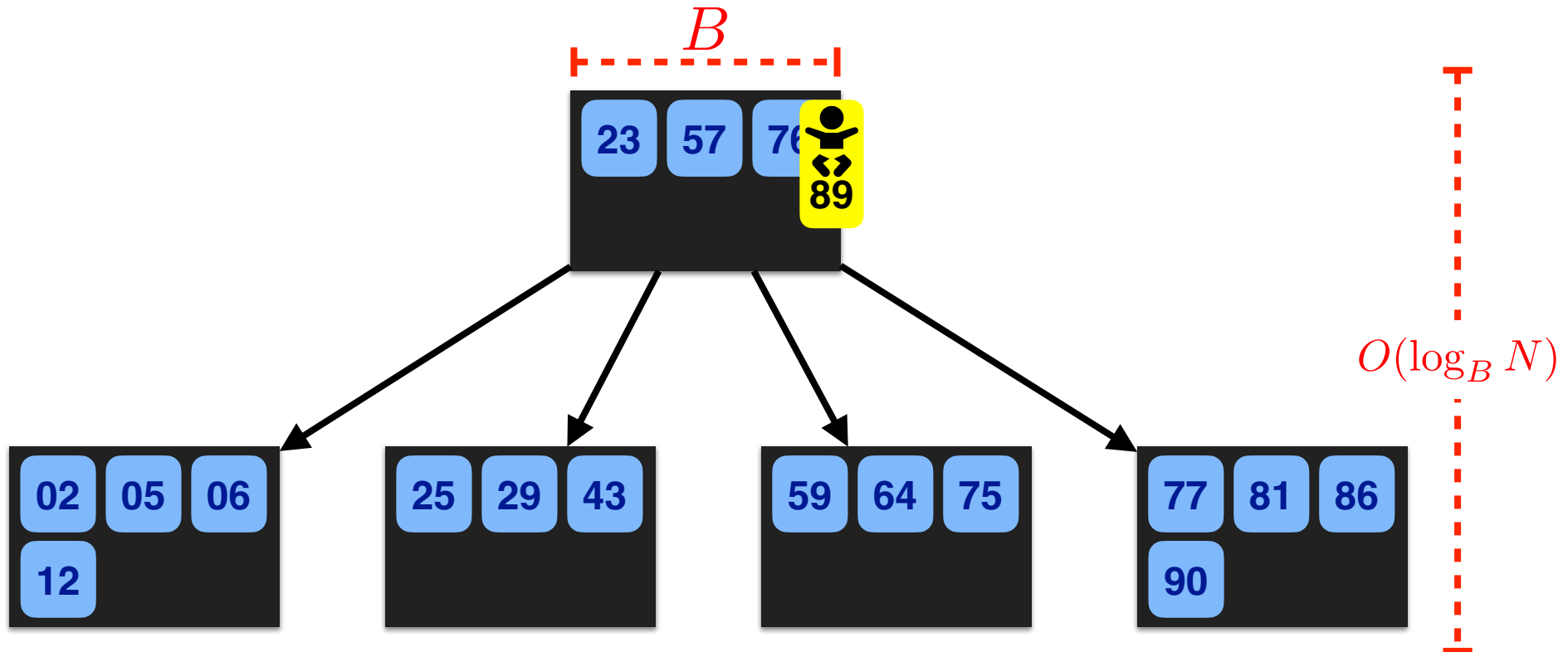
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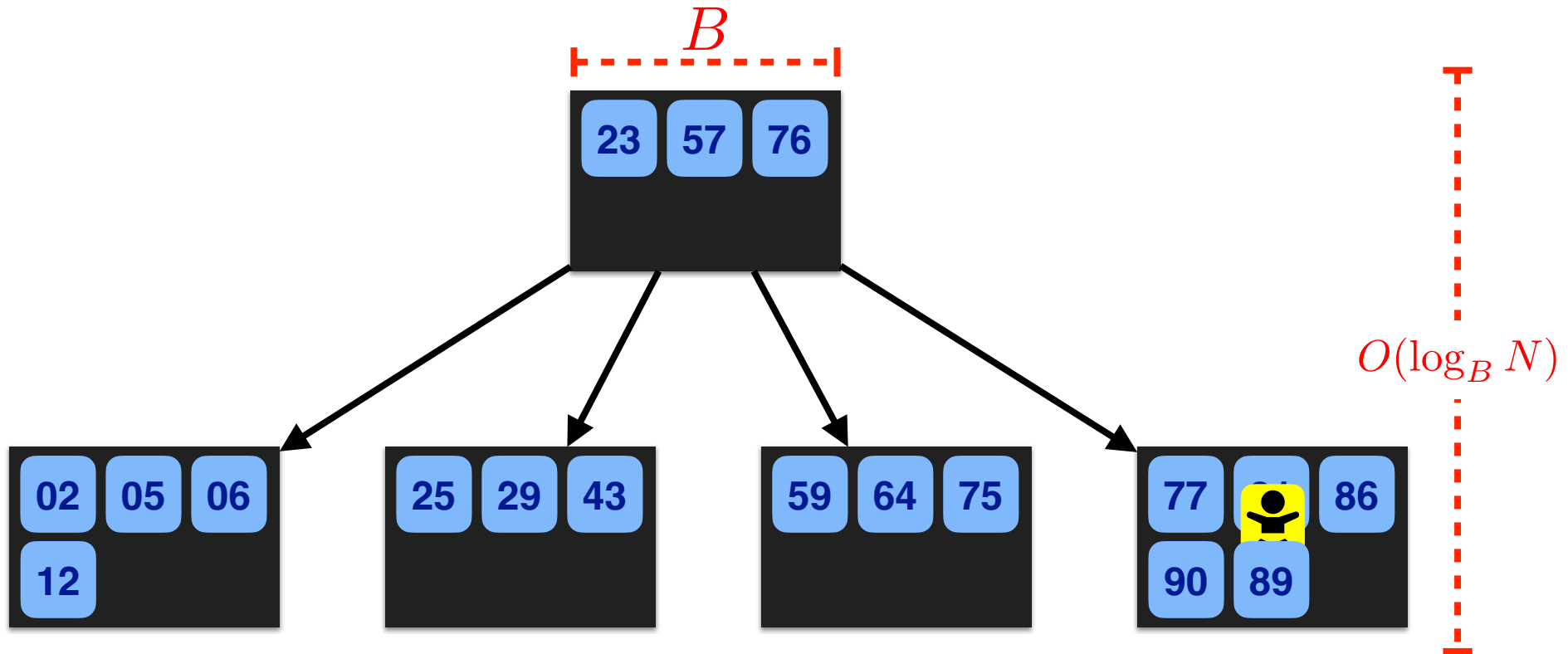
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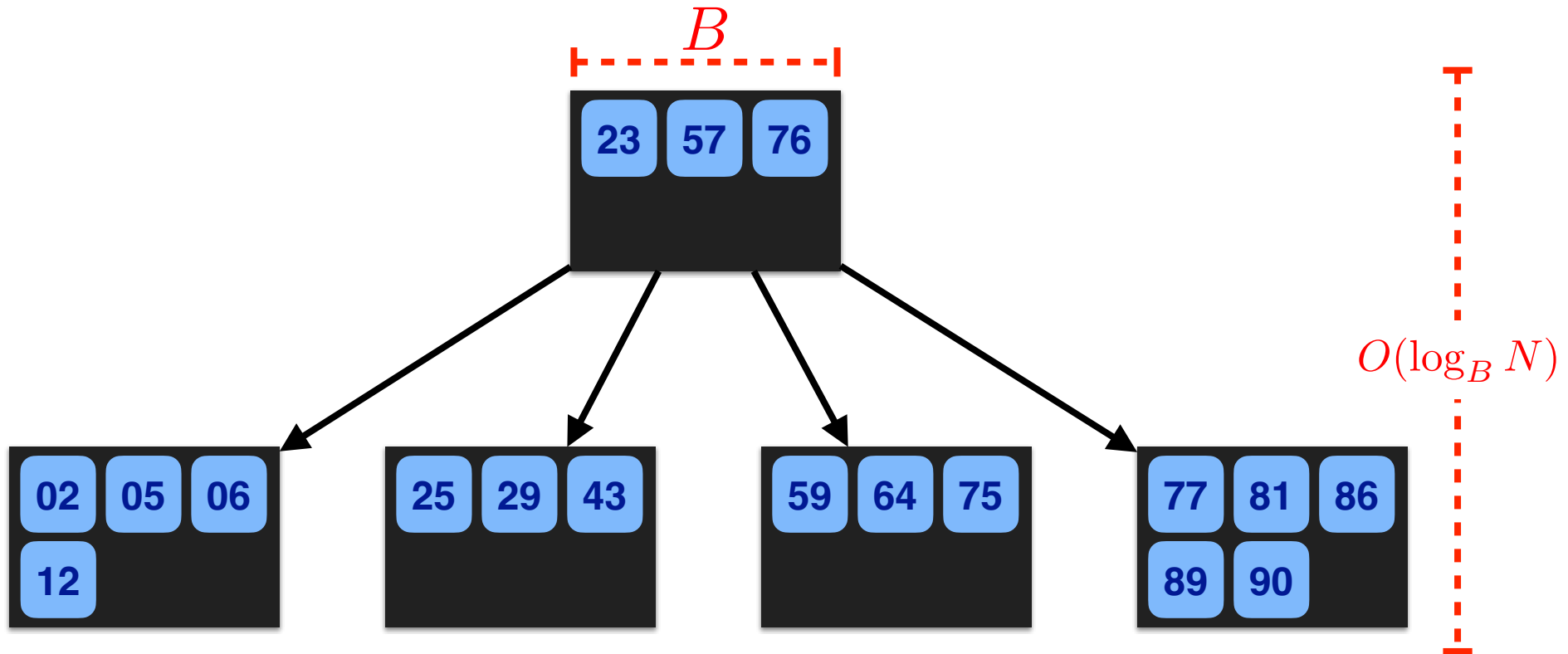
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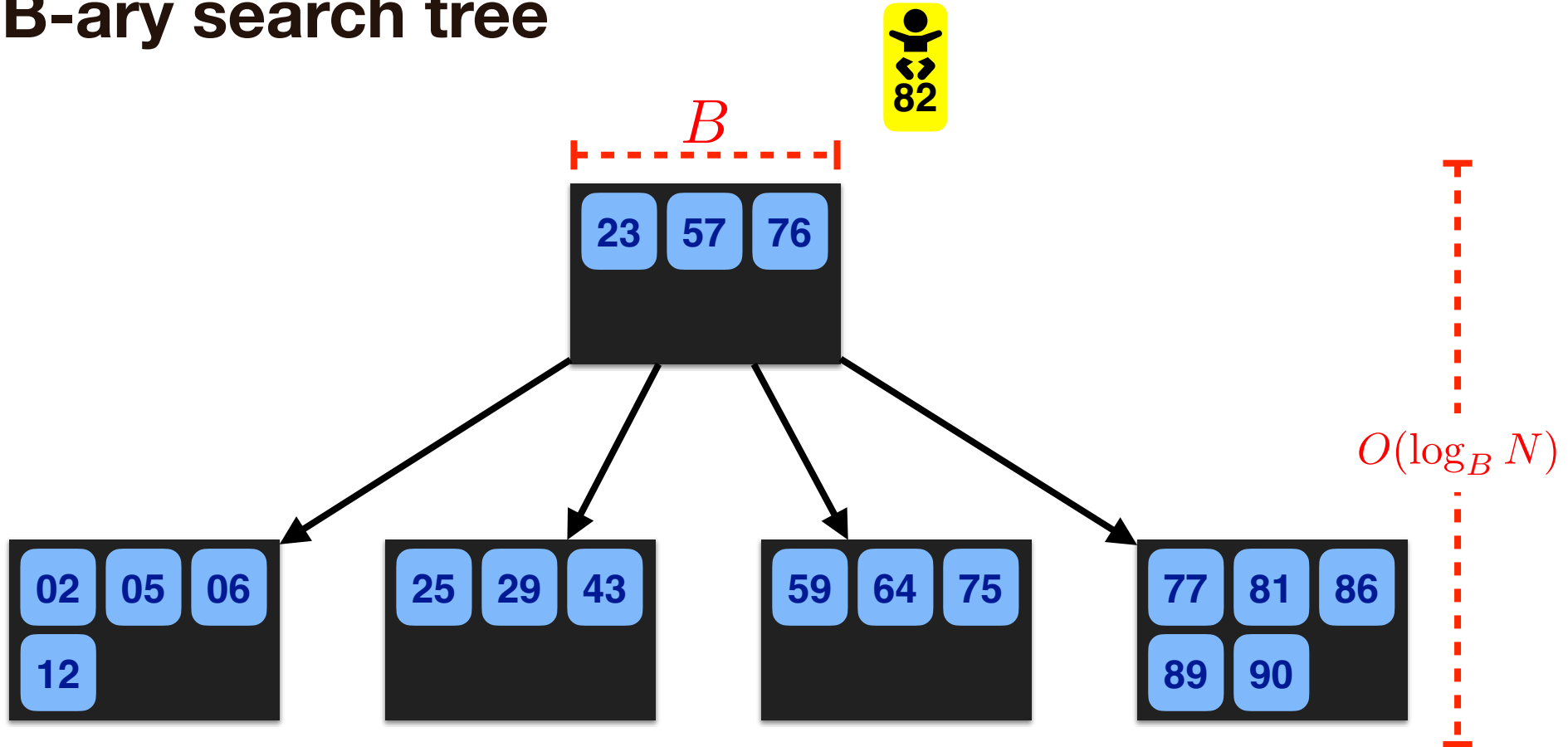
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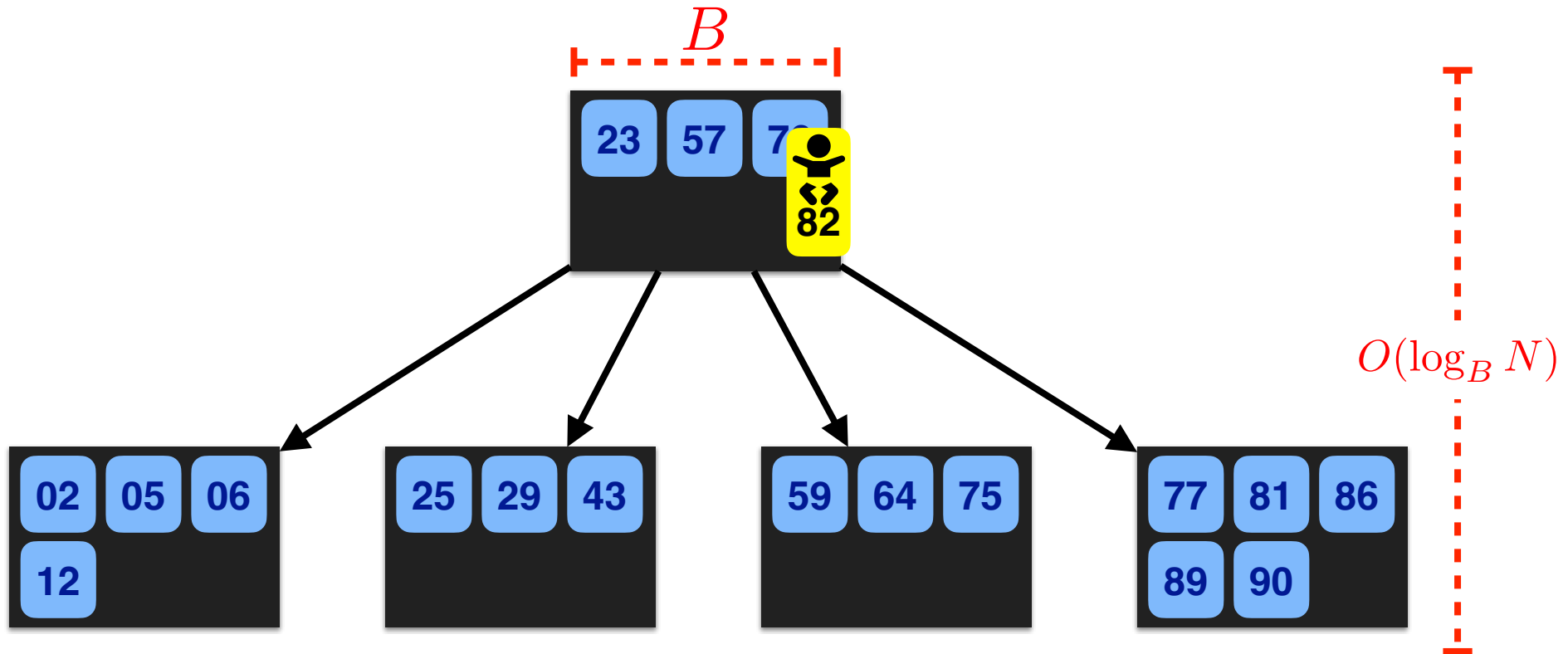
B-ary search tree



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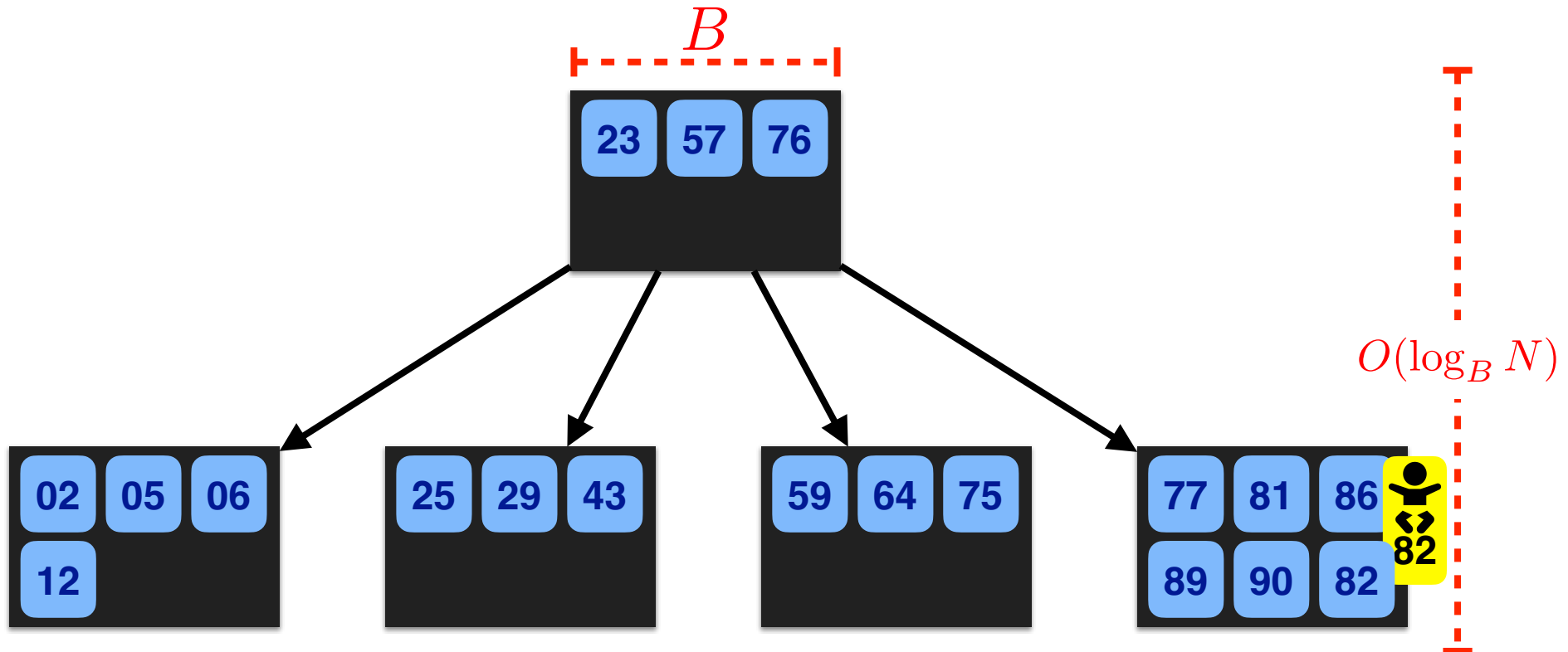
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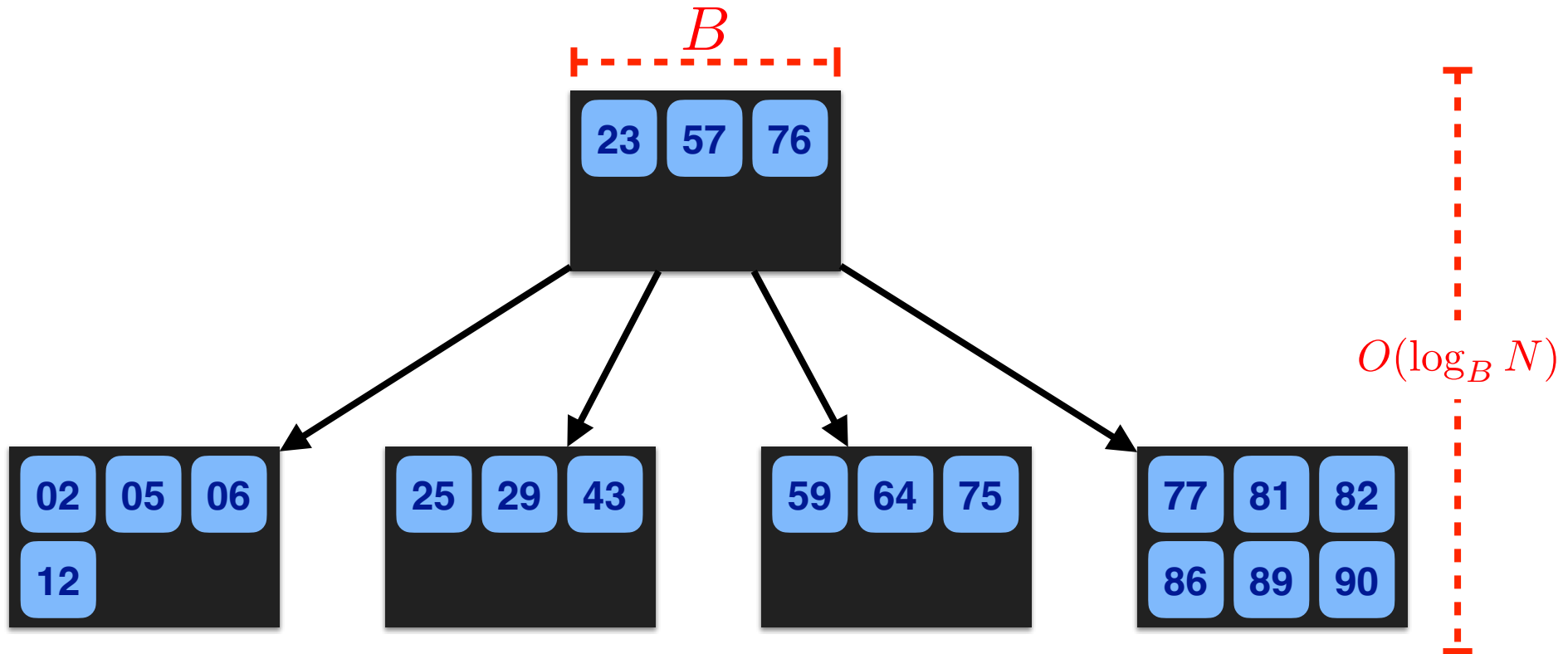
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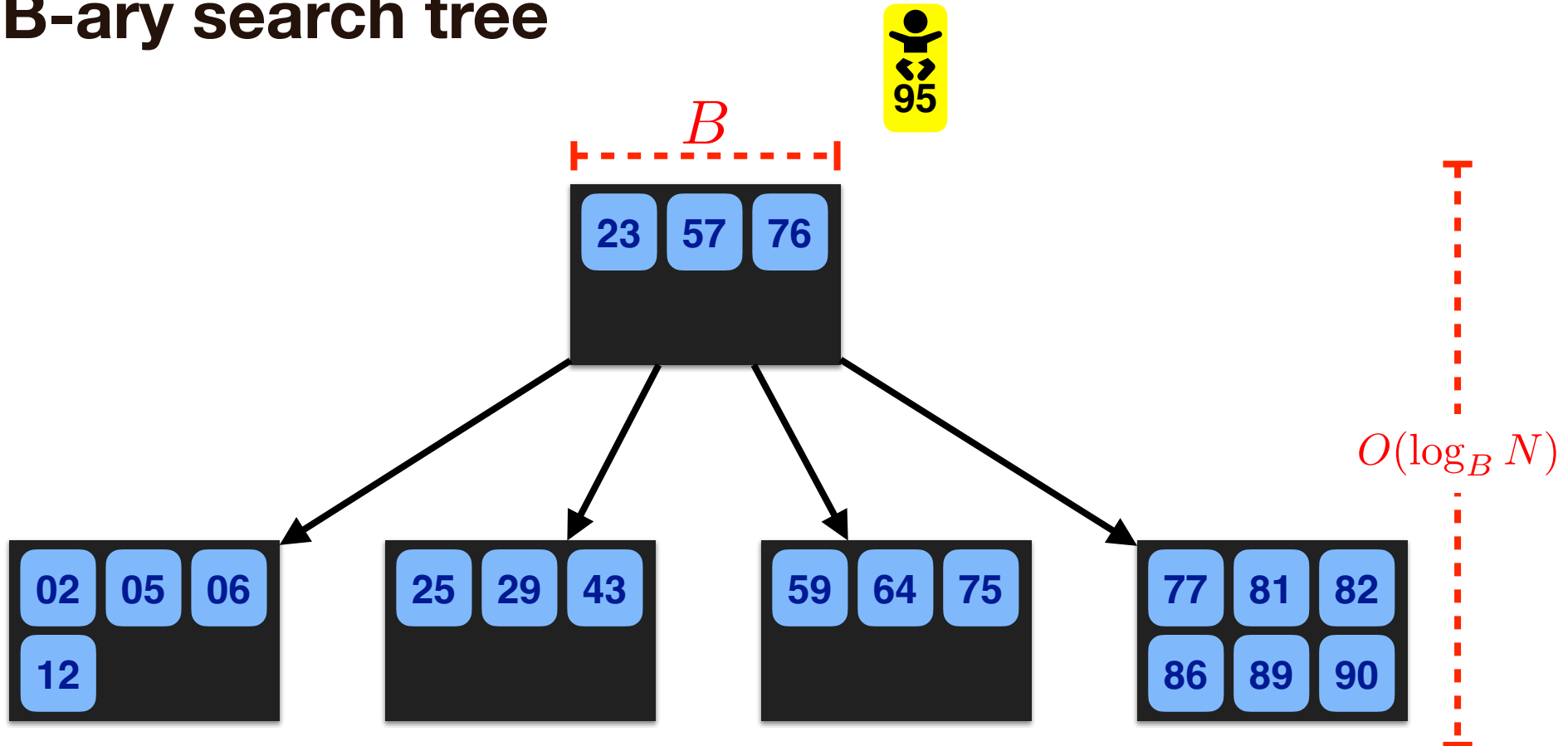
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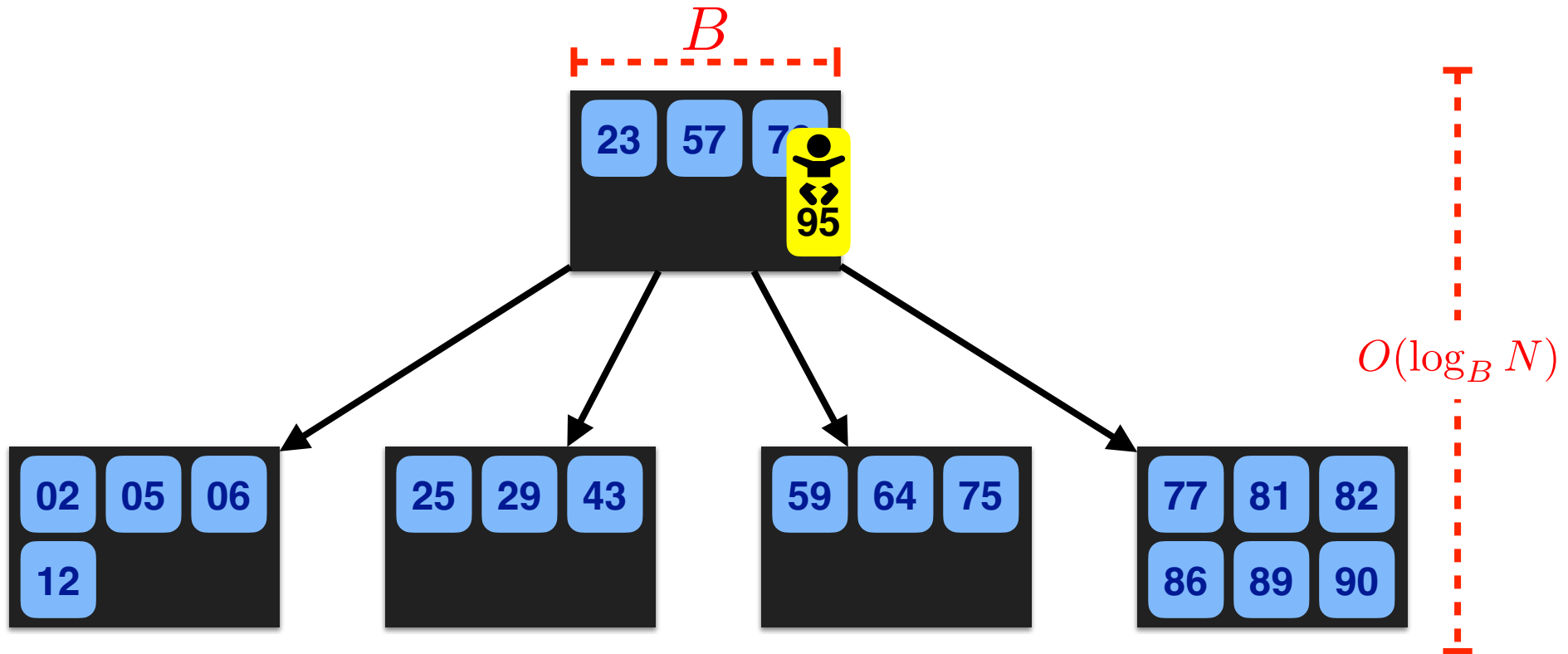
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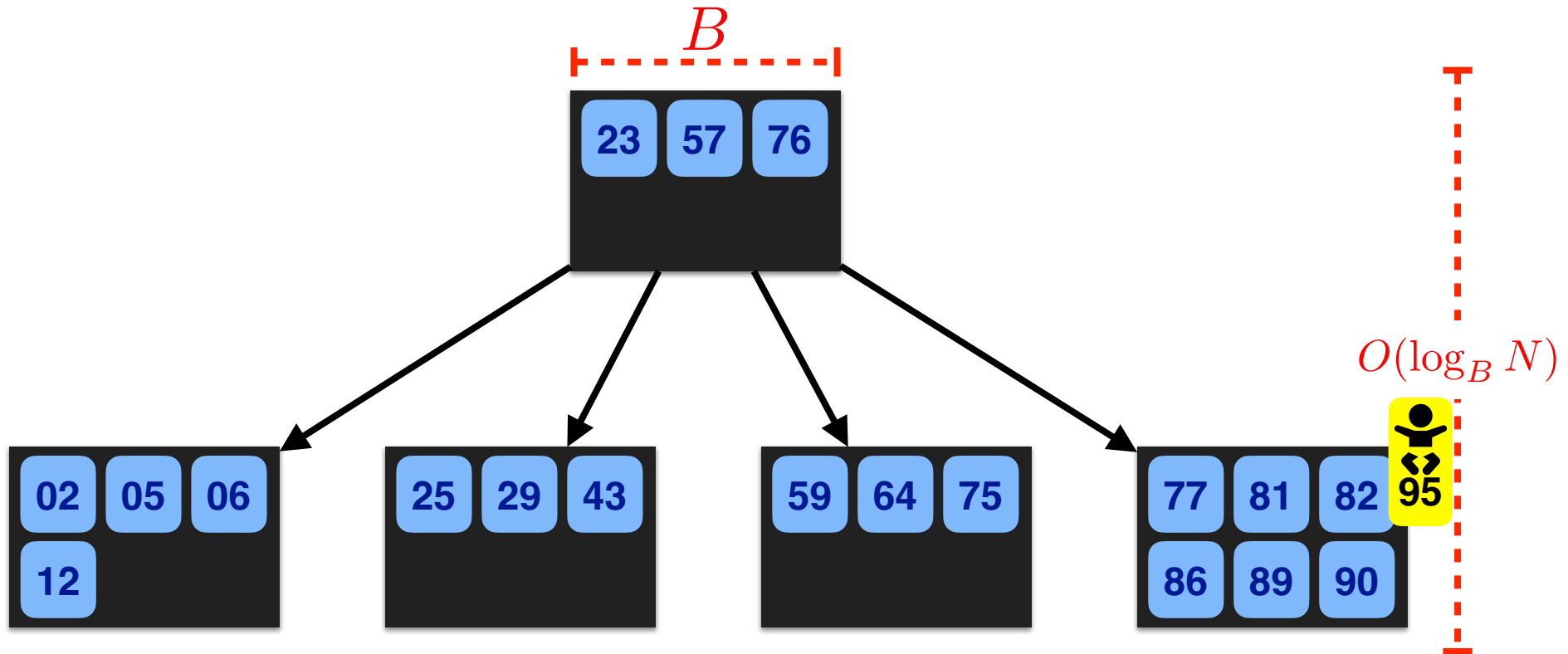
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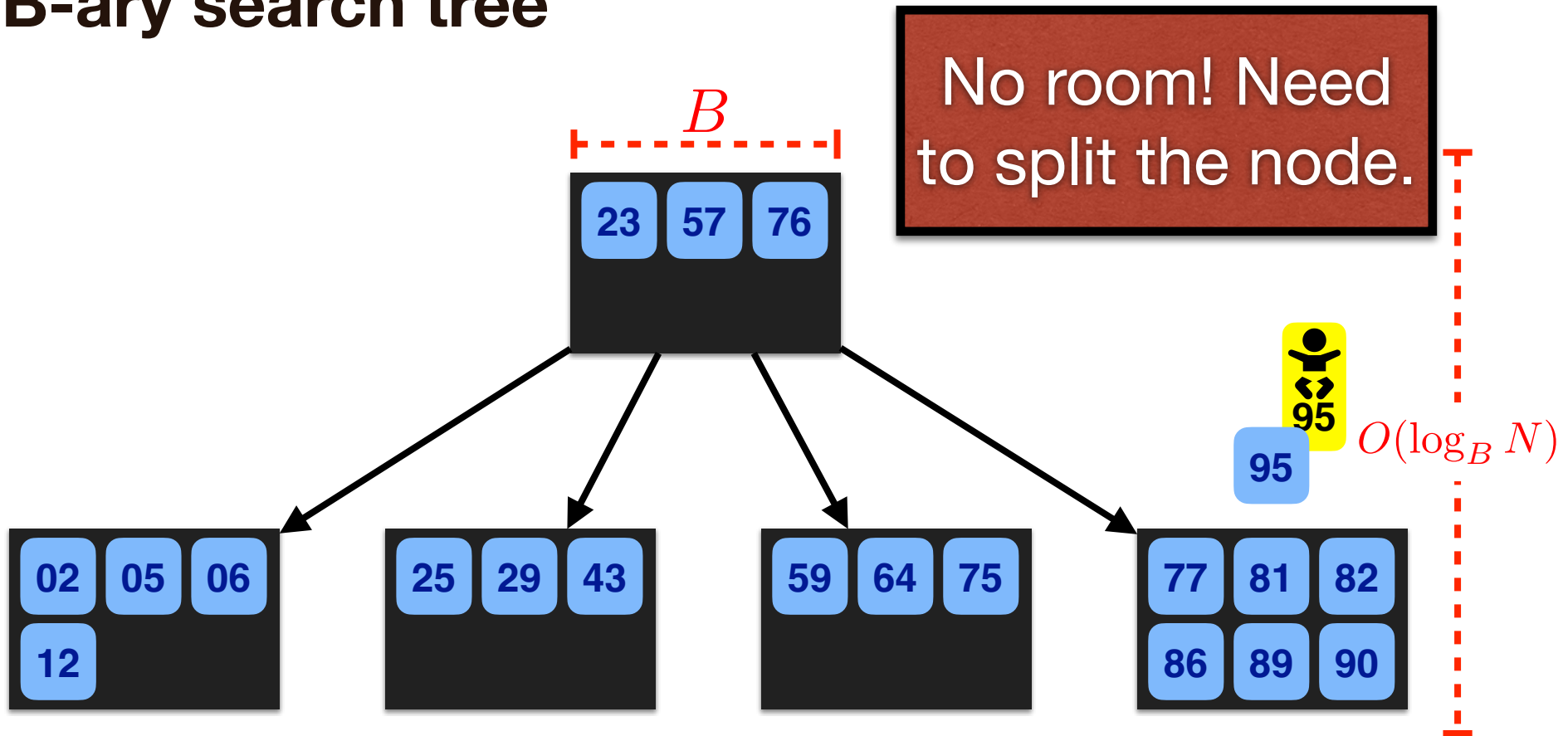
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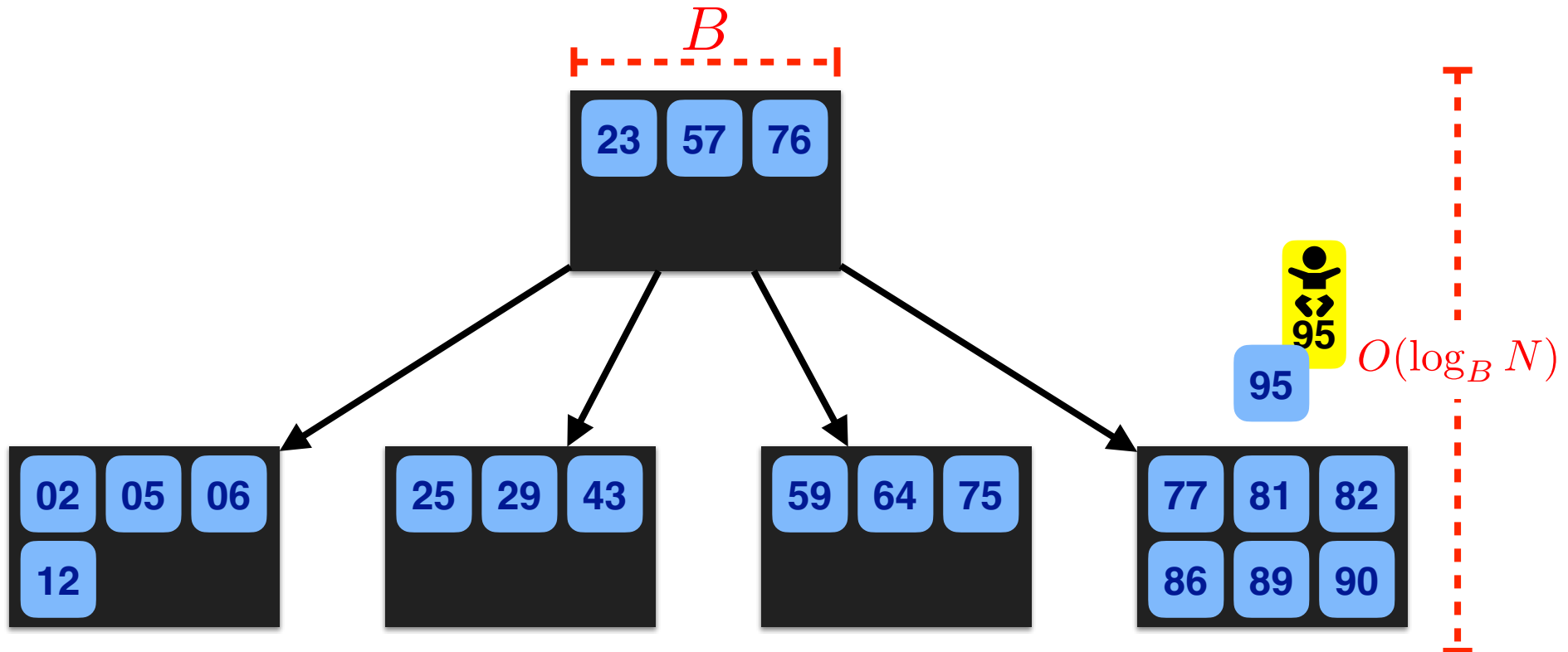
Splitting a B-tree node

Steps

- Sort all $2d+1$ keys ($2d$ + new key that causes overflow)
- Make new node with first d keys
- Make new node with last d keys
- Move middle key as a pivot of the parent
- Add pointers to new children
- Recurse up the tree if necessary (rare)

B-tree: standard DAM dictionary

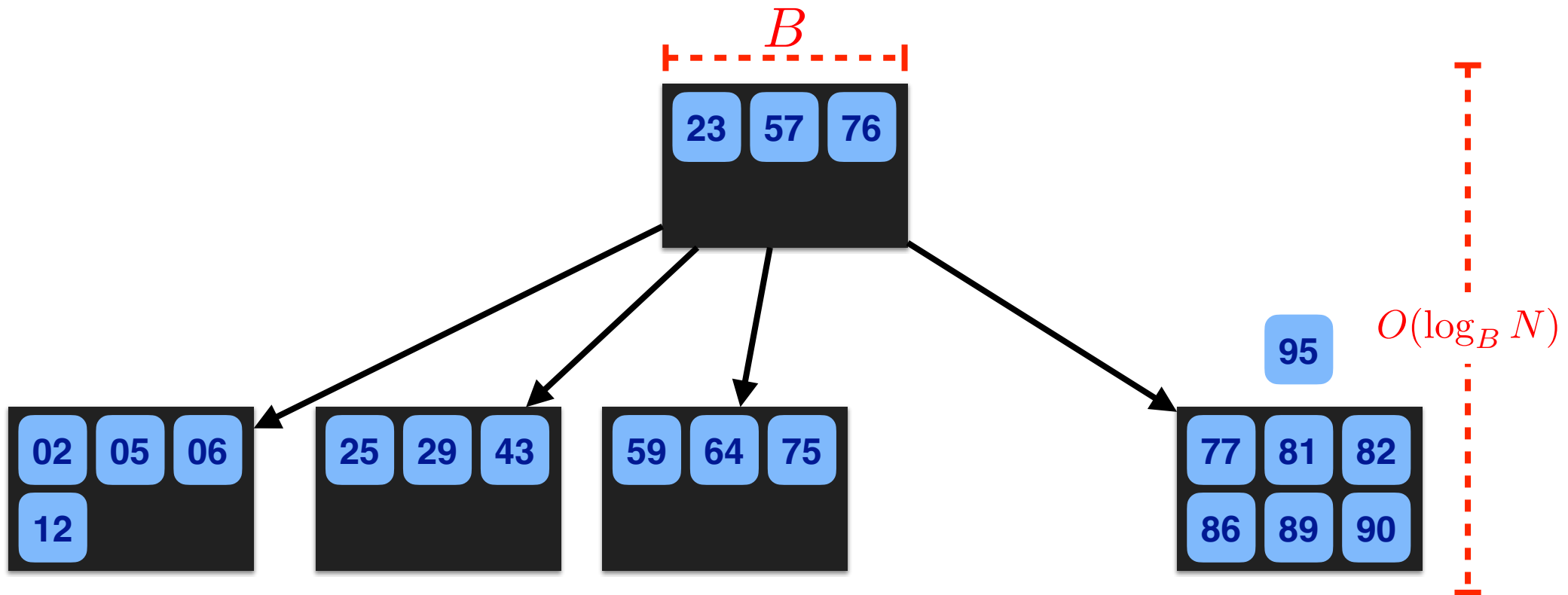
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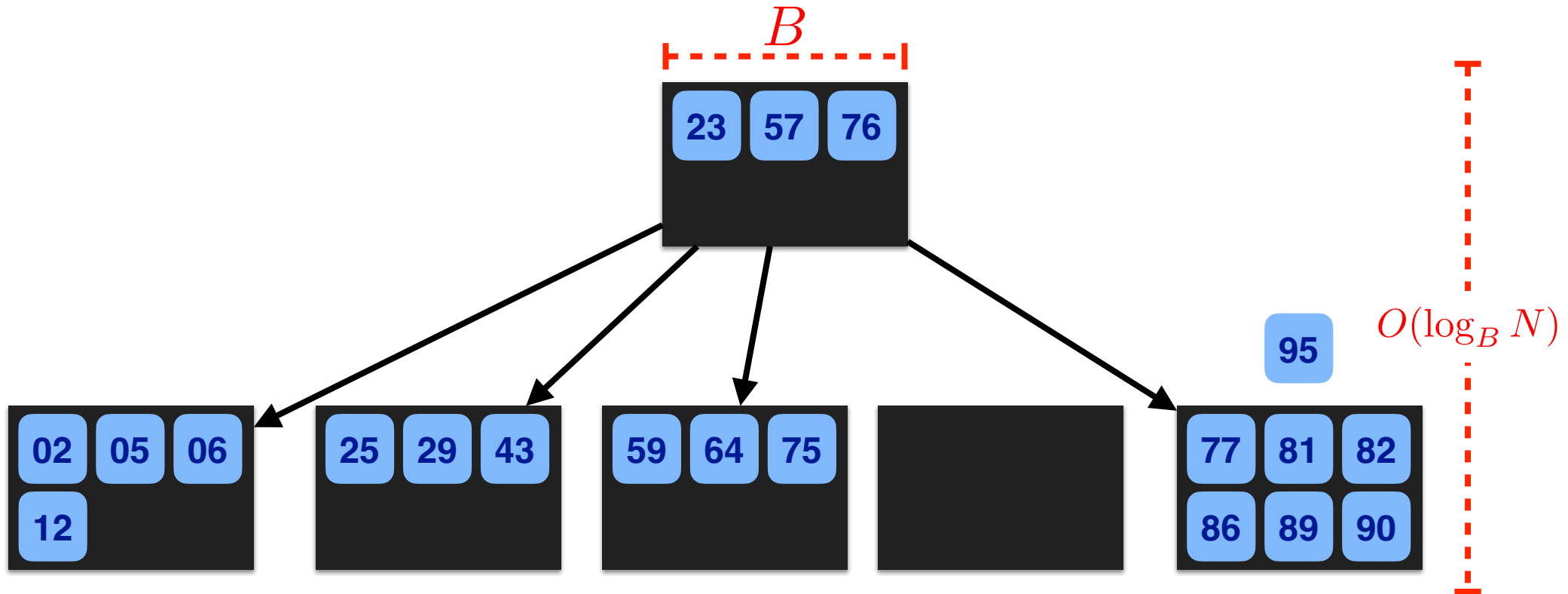
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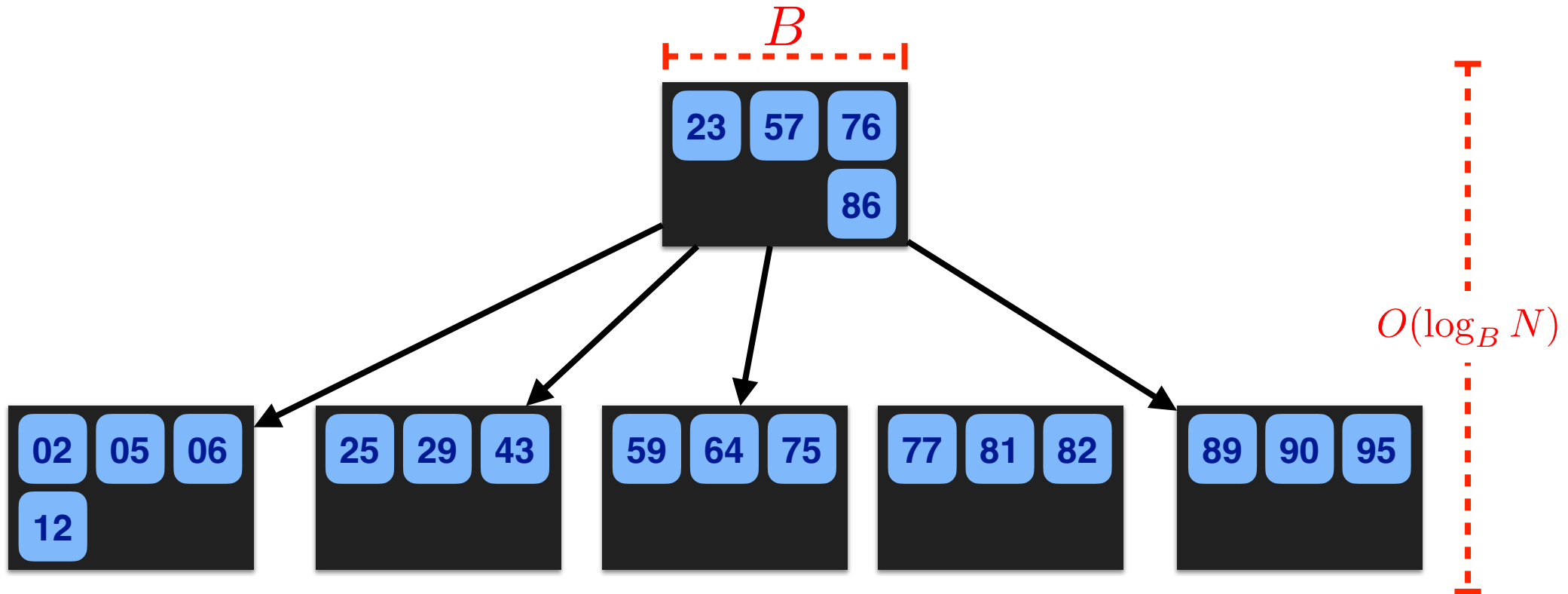
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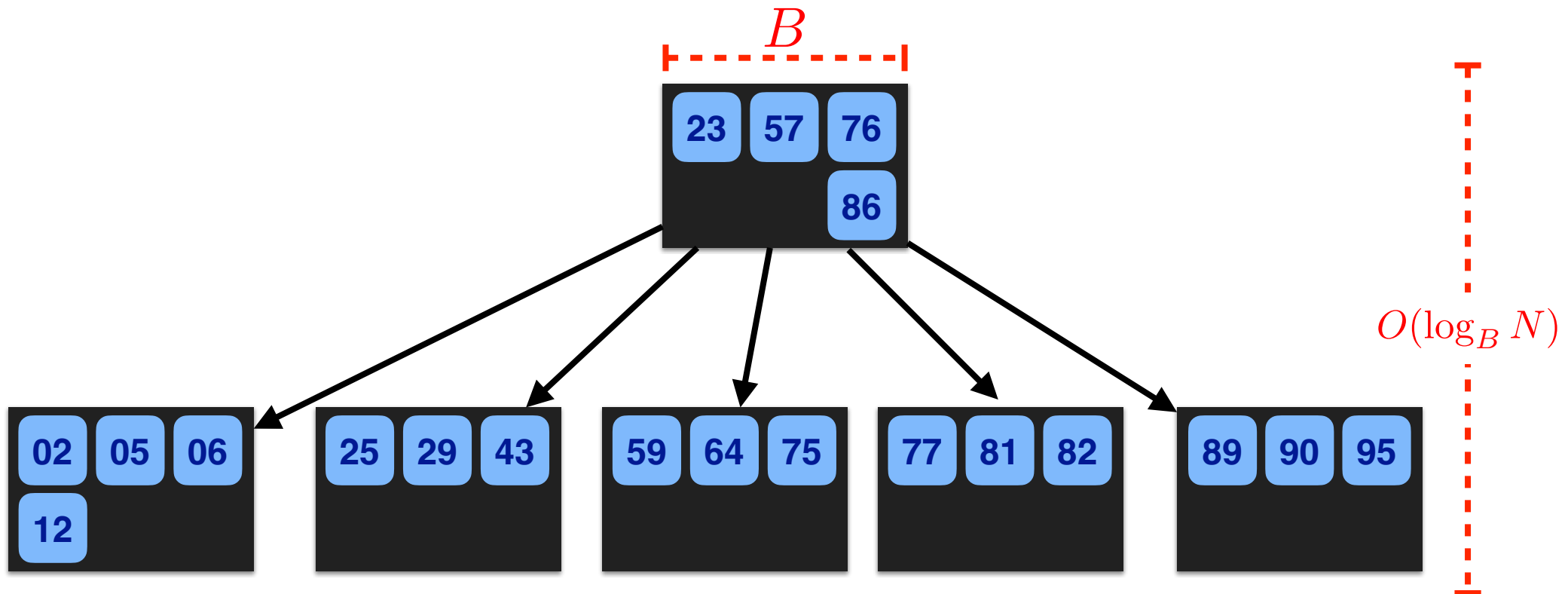
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Splitting a B-tree node

Cost

- How many nodes must be read/written in a local split?
 - ▶ We read the node being split
 - ▶ We write the old node and the new node (first **d** keys, last **d** keys)
 - ▶ We read/write the parent node
- What if we overflow the parent?
 - ▶ If we recurse, we already read the parent, so we repeat the same steps one level above
- Total cost of an insert: $O(h)$
 - ▶ Reads: $O(h)$
 - ▶ Writes: $O(2h)$

B-tree Range Queries

B-tree Range Query

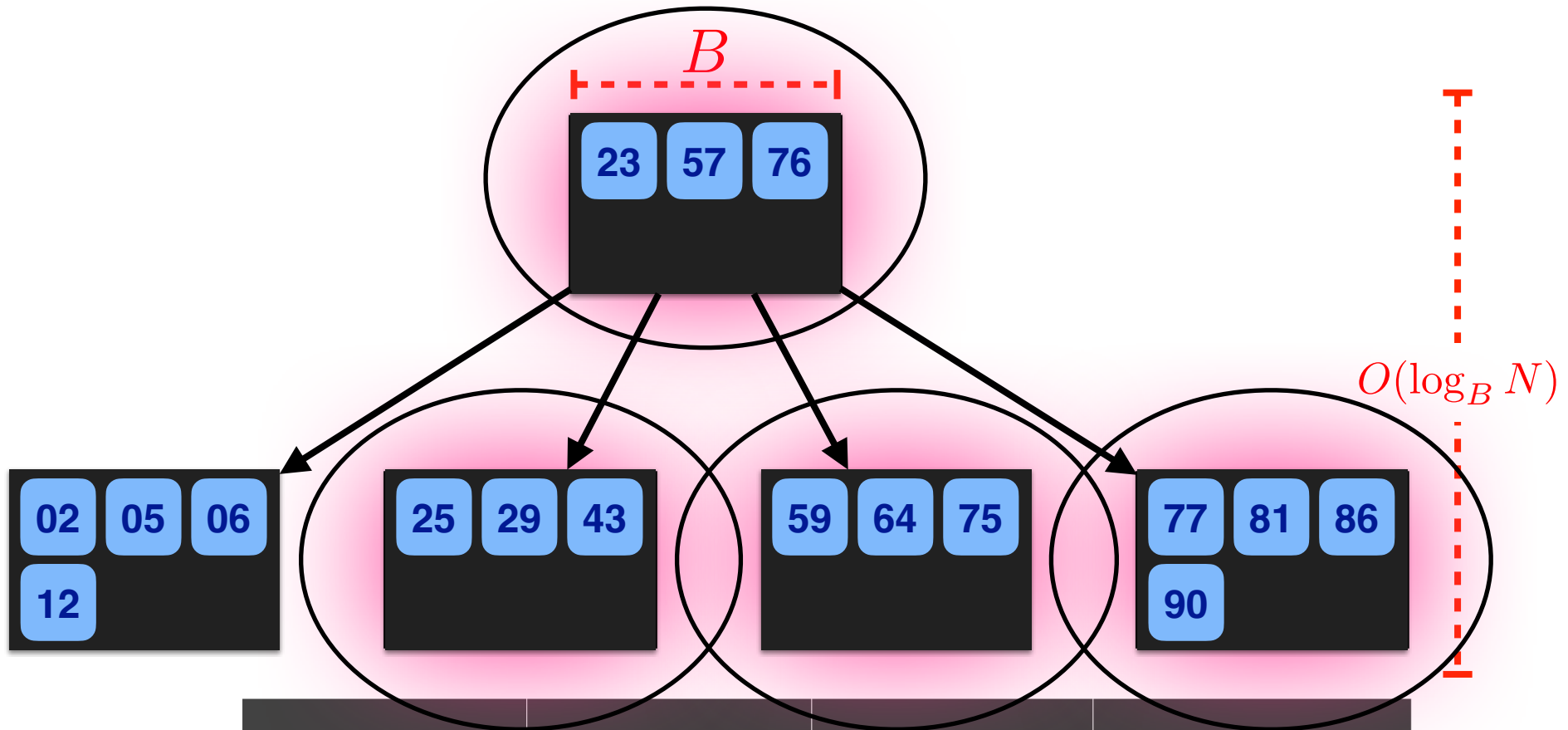
(Range query: point query + successor^k)

Steps

- Find the leaf node where the first key-value pair belongs (point query)
- Read all key-value pairs from that node that are part of your range
- Consult your parent to find its next child pointer
- Read all key-value pairs from that node that are part of your range
- Loop

B-tree: standard DAM dictionary

B-ary search tree



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B-tree Deletes

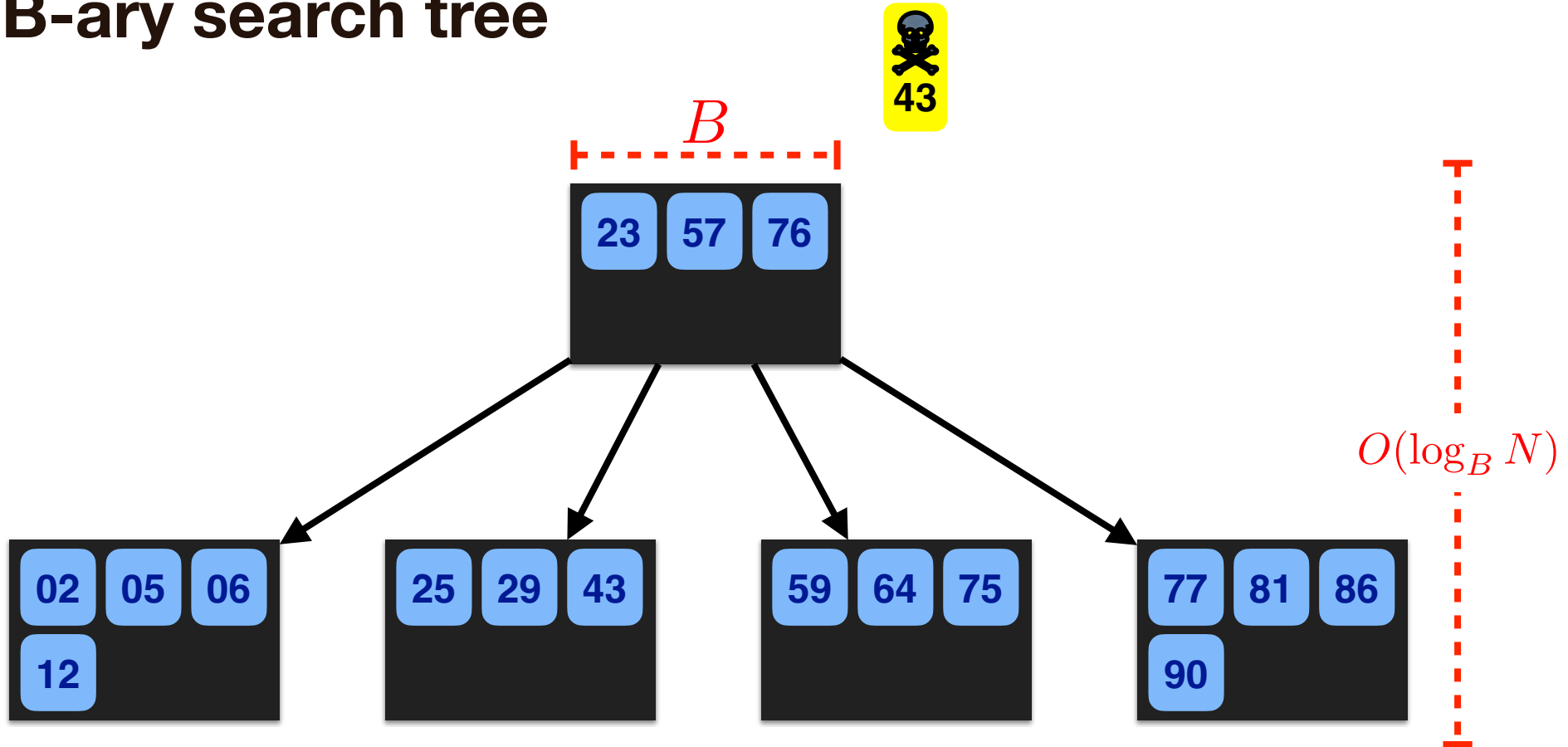
B-tree Deletions

Steps

- Search for the leaf containing the target key-value pair (point query)
- Remove the element from the leaf (if present)
- If the size of the node drops below **d**, merge with a neighbor
 - ▶ Remove extra pivot key and pointer from parent (the pointer to the node that is being deleted as part of the merge)
 - ▶ Merge contents of nodes
 - ▶ Write parent and merged node
 - ▶ If the parent size dropped below **d**, recurse upwards

B-tree: standard DAM dictionary

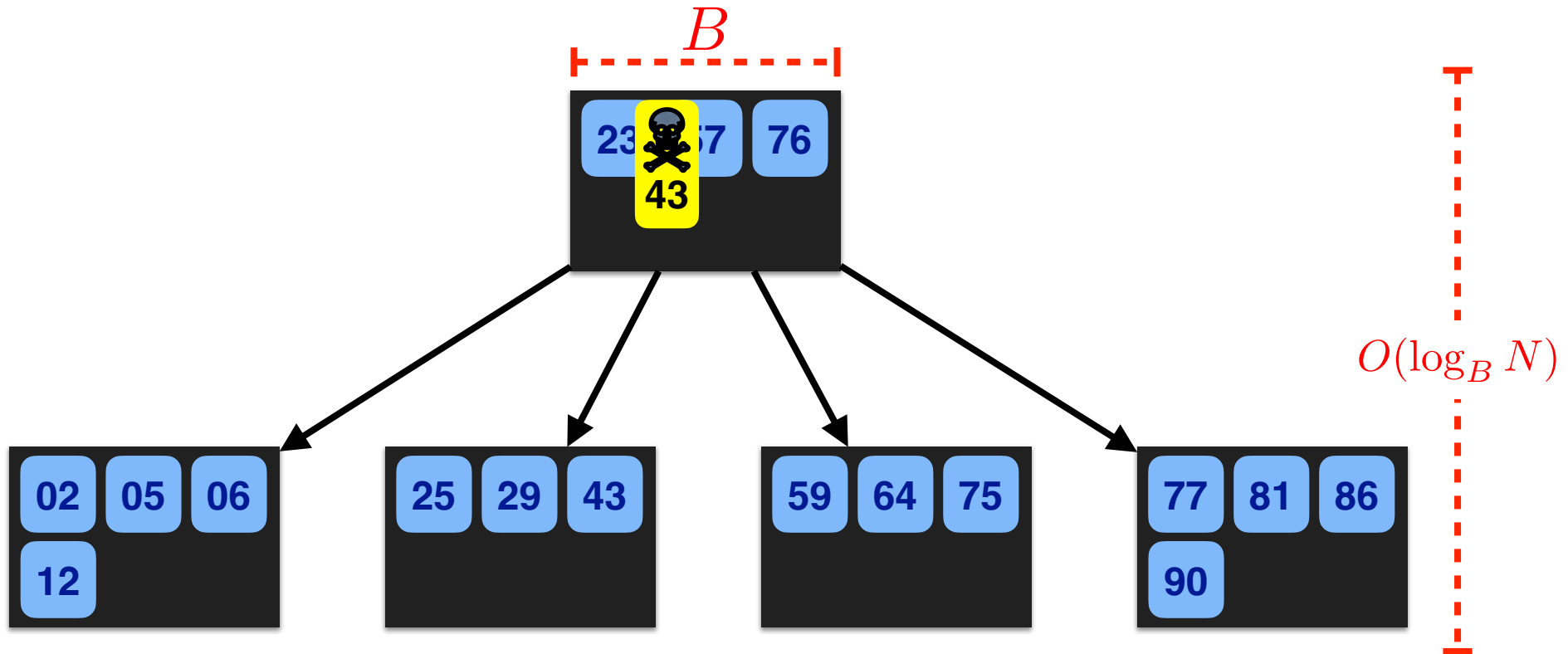
B-ary search tree



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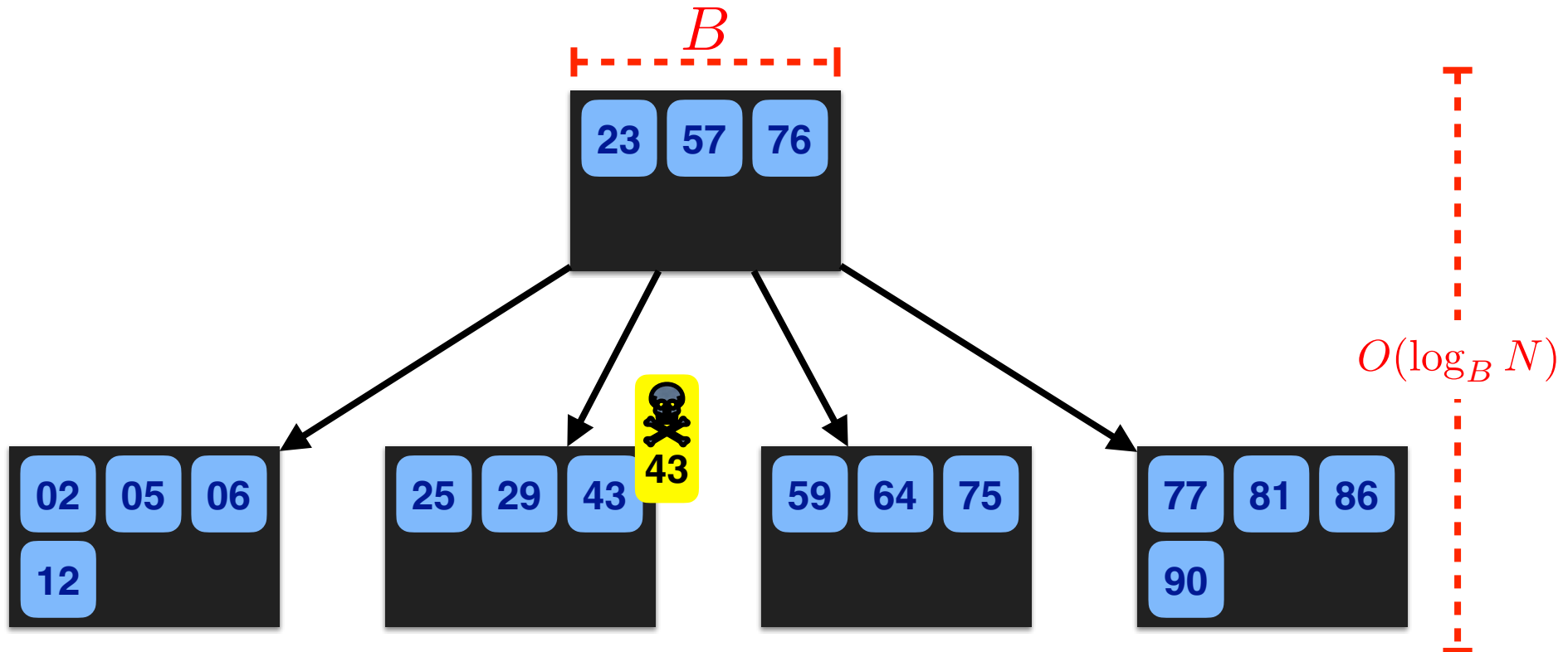
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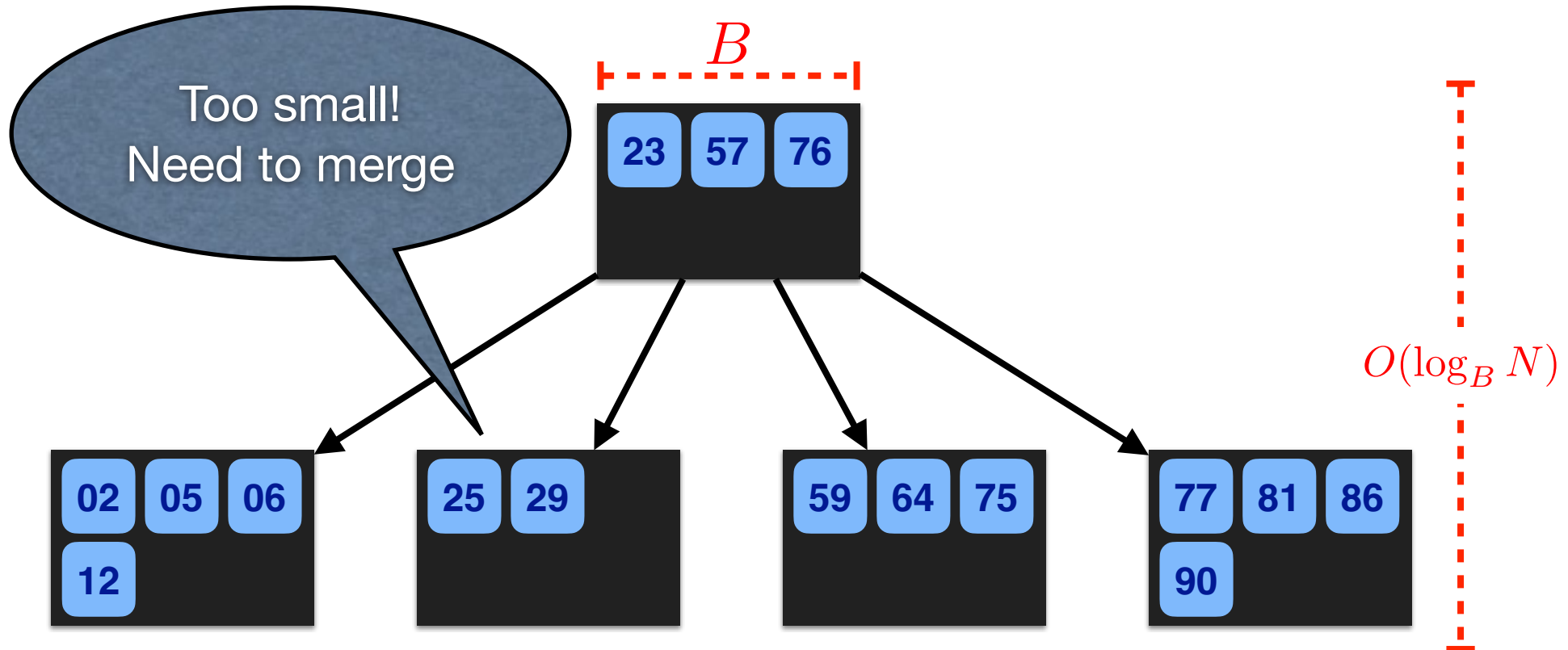
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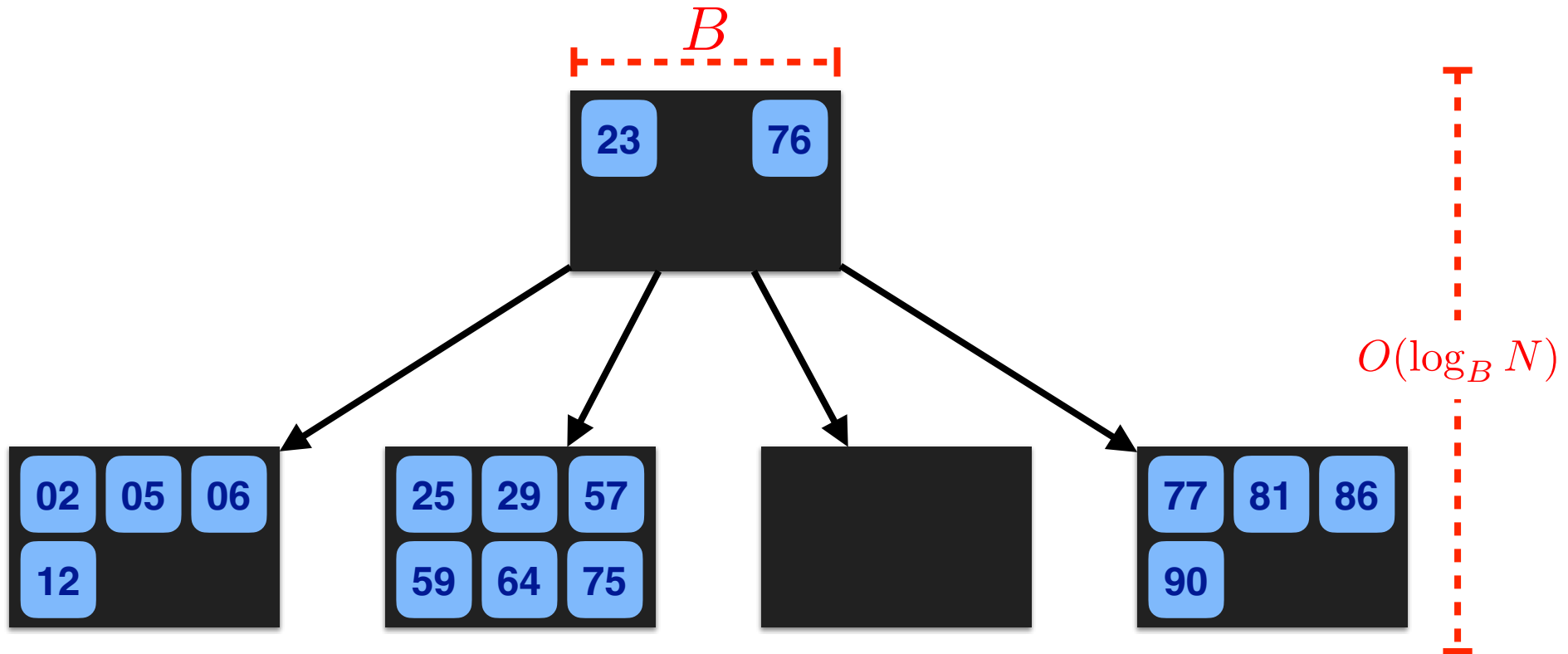
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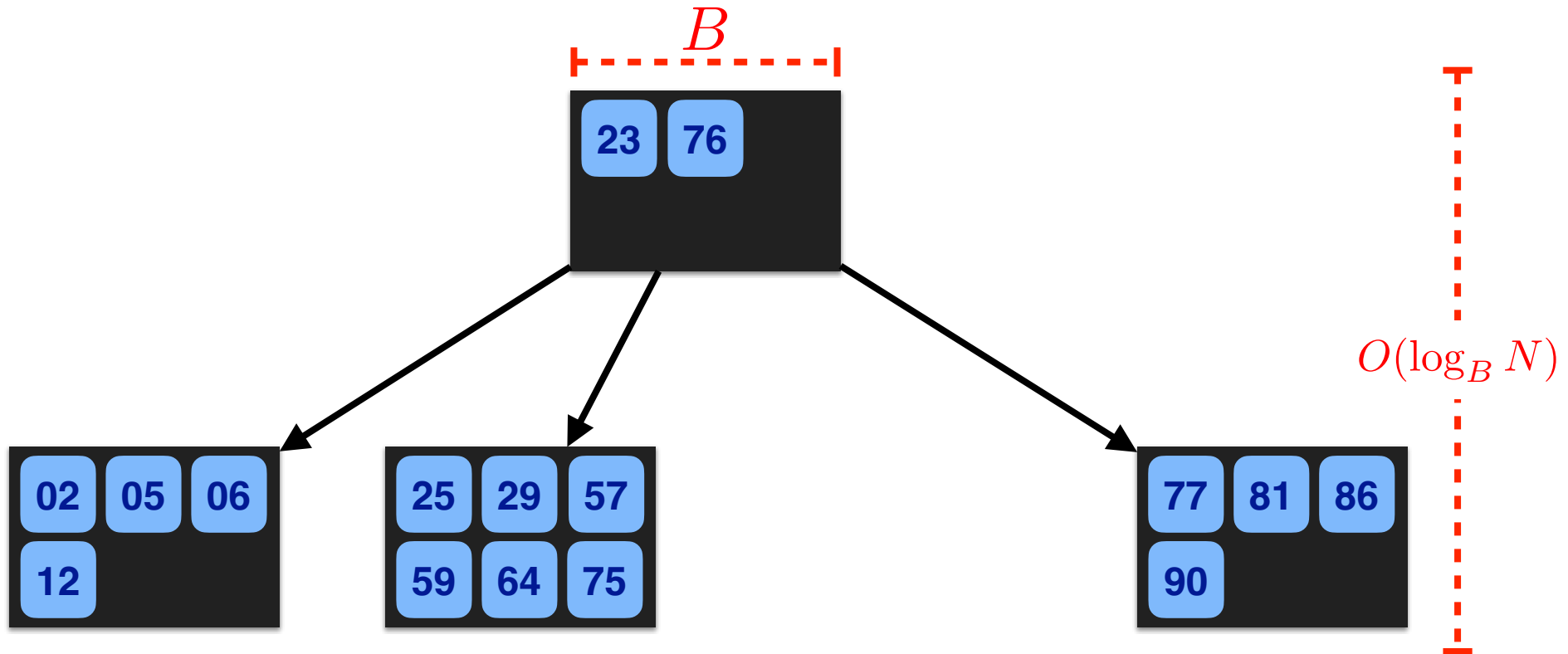
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Summary

- B-trees are the de-facto search structure for external memory applications
- Variants exist to tune utilization and range scan performance, but the idea is the same
- We can analyze performance using the DAM model

Other discussions

- **Concurrent access - how to lock the tree?**
 - ▶ Hand-over-hand locking for queries
 - ▶ Reservations or top-down splitting
- **How to choose the node size (B)?**
 - ▶ Must balance competing goals:
 - ▶ Small B minimizes write amplification (each update requires writing whole node)
 - ▶ Large B minimizes fragmentation (more data read per seek)

Looking Ahead

B-trees because they are widely used, but they also serve as a starting point to discuss more recent advances in trees

- Log structured merge trees
- B^e-trees

The above trees employ *write optimization*

- ▶ Better I/O performance for writes
- ▶ Not asymptotically worse off for reads

If you like this type of analysis (the intersection of asymptotic analysis and system optimization)

- Several electives!
 - ▶ Applied algorithms
 - ▶ Storage Systems
 - ▶ Parallel Processing // Distributed Systems

Taking Stock

We've covered a lot this semester

- My first time teaching 256, so I've learned a lot
 - ▶ I hope to learn what was most helpful for you
- Covered important topics that will help you think about, formally define, and quantify problems and their solutions
 - ▶ Asymptotic analysis
 - ▶ Graphs: traversals, algorithms, and applications
 - ▶ Greedy algorithms & proving their optimality
 - ▶ Divide and conquer // Recurrences
 - ▶ Dynamic programming
 - ▶ Network flow & problem reductions
 - ▶ Intractability: NP & NP hardness & more problem reductions
 - ▶ Randomized algorithms and analysis

What's Next?

CS256 Opens up several doors

- Prerequisite to both theory and “applications” electives
- Prerequisite to Theory of Computation
- There isn't anyone in this room I wouldn't recommend for research in CS (if that is what you actually want...)
 - ▶ Theorists in CS faculty include...
 - ▶ Sam McCauley
 - ▶ Shikha Singh
 - ▶ Aaron Williams
 - ▶ Intersection of Theory and XXXXX
 - ▶ Data structures/indexing/filters (Bill & Sam & Shikha)
 - ▶ Many PL problems (Dan & Steve)
 - ▶ Distributed Systems (Jeannie)
 - ▶ Algorithmic Game Theory (Shikha)
- Summer research, theses, and (sometimes) RAs
- REU programs are available in other places too!

CS Courses

<http://cs.williams.edu/~jannen/teaching/cs-prereqs.svg>