# B-trees (Ubiquitous and otherwise)

Williams College :: CSCI 333 Spring 2020

## Last Class(es)

## **Hashing and Filters**

- Bloom filters
- Quotient filters

## Support "approximate membership" queries

- No false negatives
- Tunable false positive rate

## Cache efficiency matters

Quotient >= Bloom

#### **API** matters

Deletes? Merges? Resizing?

## Last Class(es)

## **Aging and Locality**

- Aging is a path not a point
- But we can examine any point in that path
  - Layout score: statically analyze blocks of a file to quantify locality
  - Metadata, inter-file, intra-file, and free space fragmentation
    - Can statically analyze
    - Can experimentally verify performance impacts

## Why Important

- To evaluate our hypotheses, we need well designed experiments
  - Meaningful workloads under meaningful conditions
  - ▶ Need to be reproducible

## This Class

## Review and apply the DAM model

How to analyze external memory algorithms

#### **B-trees**

- Operations
- Variants
- Discussion

# How do you keep data organized?

## An Analogy from [Comer 79 CSUR]

# Filing Cabinet: folders of records, alpha-sorted by last name

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

## Indexes (yes, colloquially pluralized that way)

## Indexes organize data

- Random 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?

## What operations does an index support?

## **Operations**

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

## In short, indexes support the *dictionary* interface.

Used when data is too big for memory.



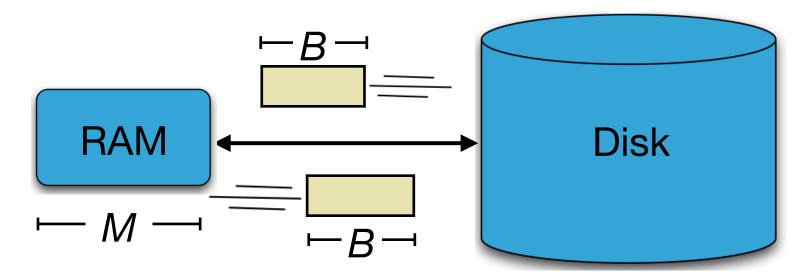
## How to we quantify index performance?

#### **DAM** model:

- Useful when data is too big for memory
  - Data is transferred in blocks between RAM and disk.
- The number of block transfers dominates the running time.
  - Searching through a given block is "free" (once 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 in B-tree Analysis

## Analyze worst-case costs by counting I/Os

- **B**: unit of transfer
  - ▶ B is the B-tree's node size, so we essentially count node transfers
- 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 the dimensions of our tree
- We will think about the tree shape (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)
  - ▶ Have a well-defined ordering

#### **Terms and Conditions**

#### Rules for our B-trees

- B-ary tree
  - Internal nodes have between d and 2d keys called pivots
    - Must 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, we must rebalance to restore the invariants!
- 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

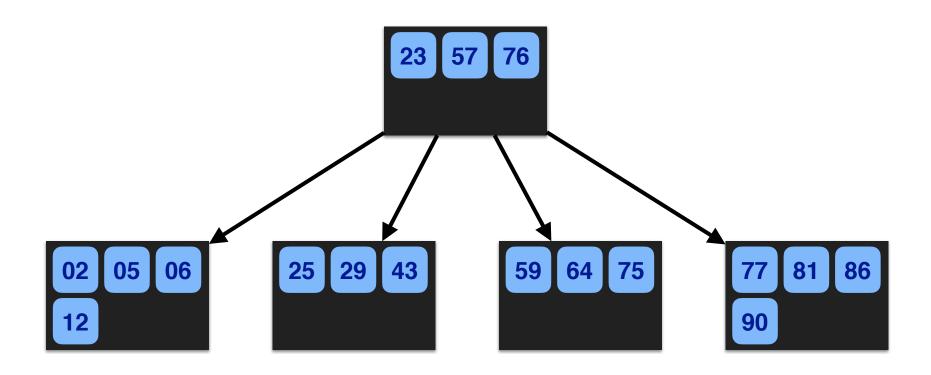
#### Terms and Conditions

#### **Several B-tree variants**

 We will describe a "B?!+--tree" here, noting features of specific variants as they come up

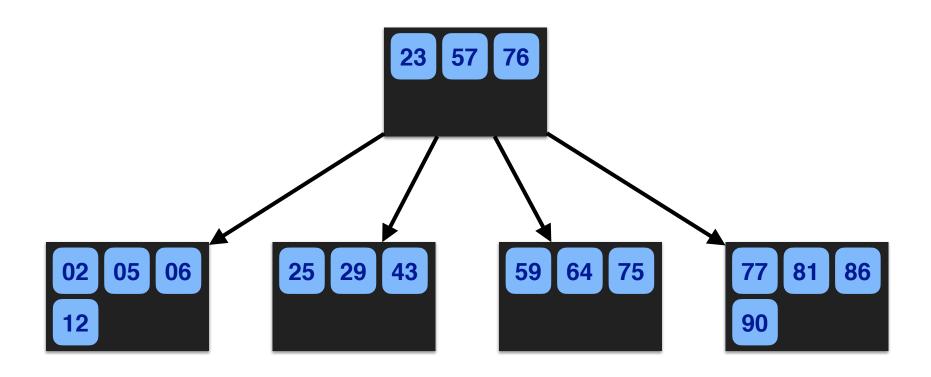
## **Popular Variants of B-trees**

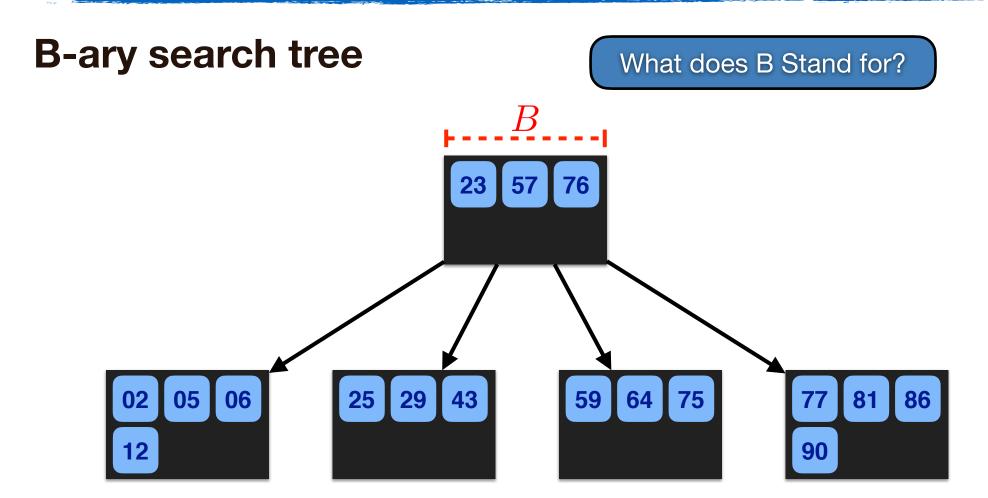
- B-tree: more-or-less what we'll describe here
- B+-tree: (B plus) B-tree where leaves form a linked list
- B\*-tree: (B star) B-tree where nodes are always 2/3 full

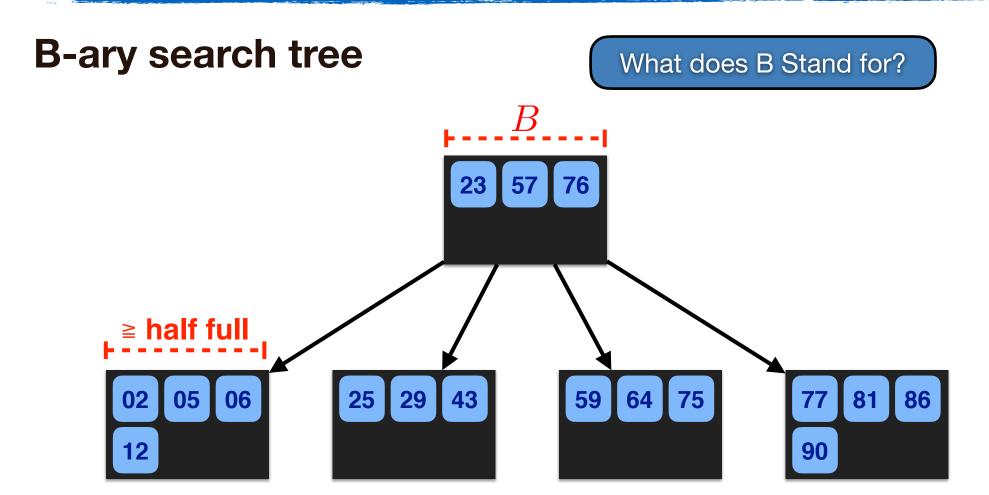


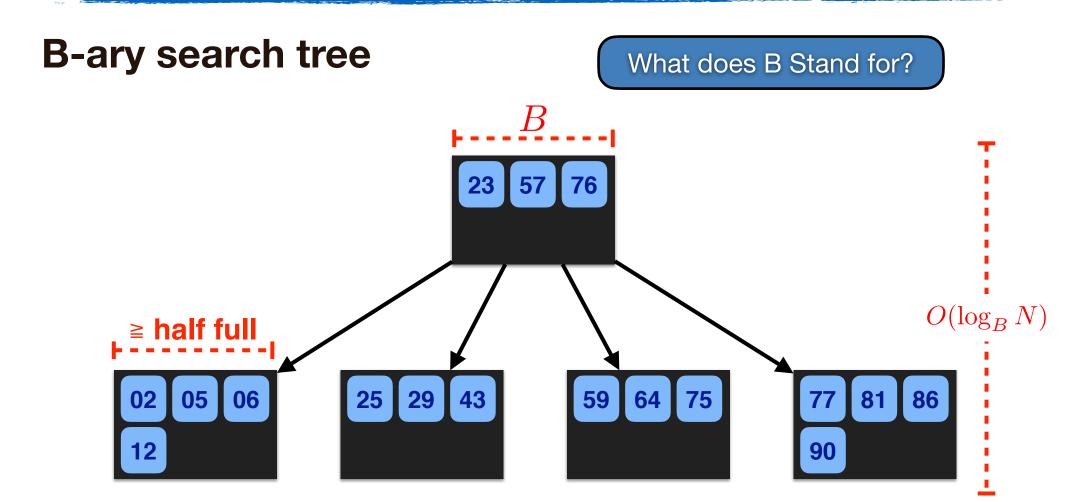
## **B-ary search tree**

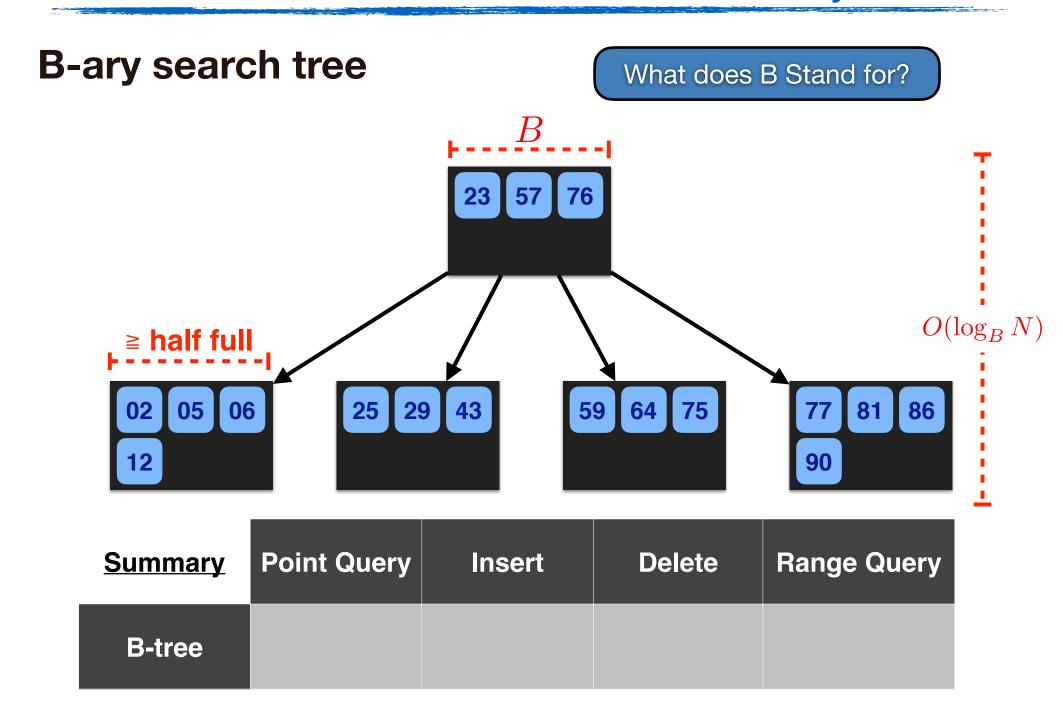
What does B Stand for?

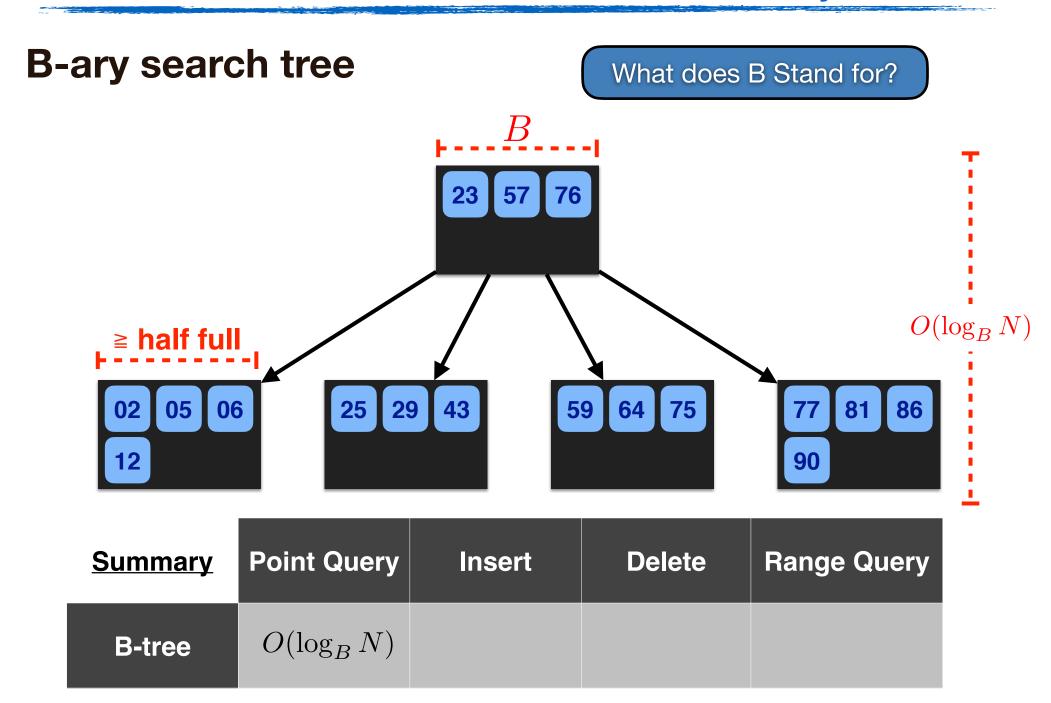


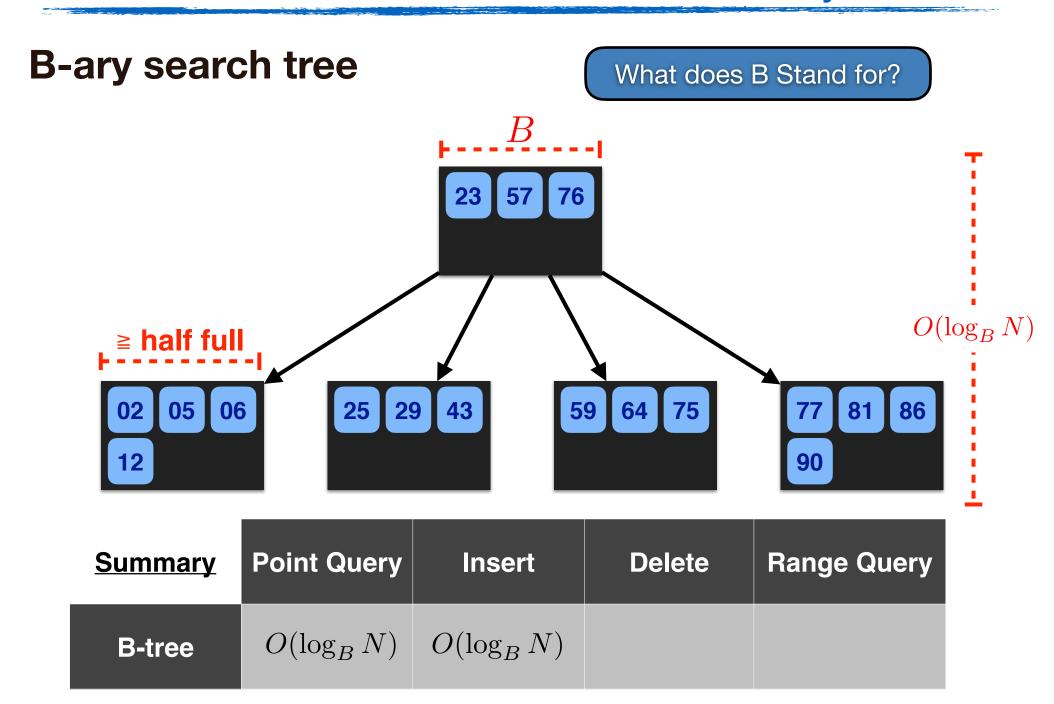


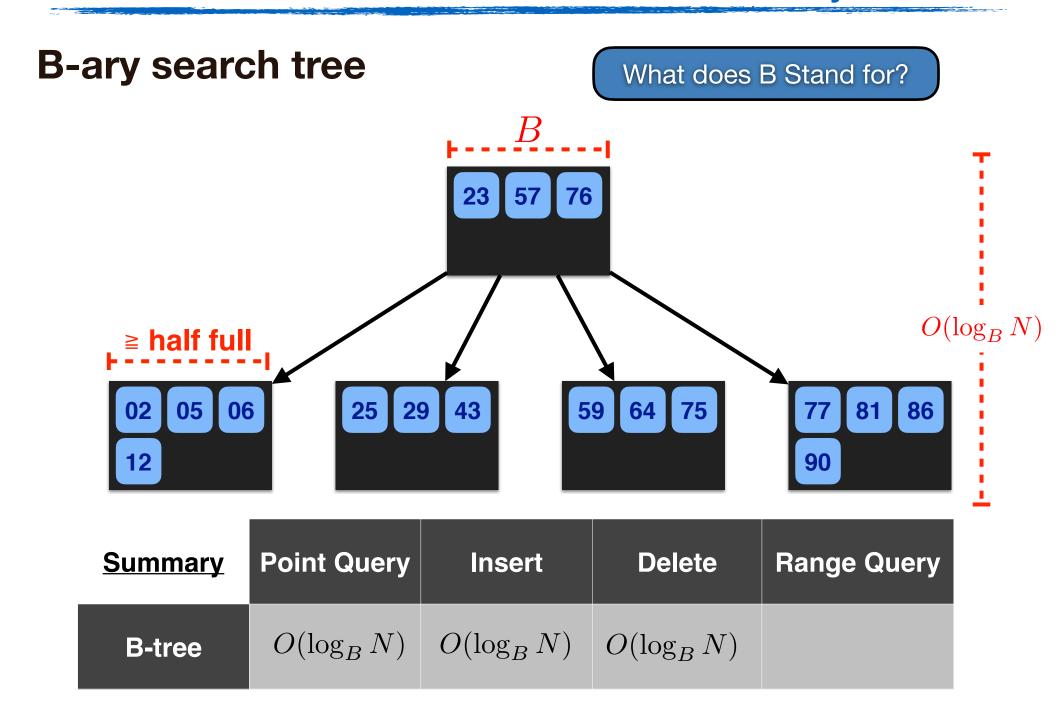


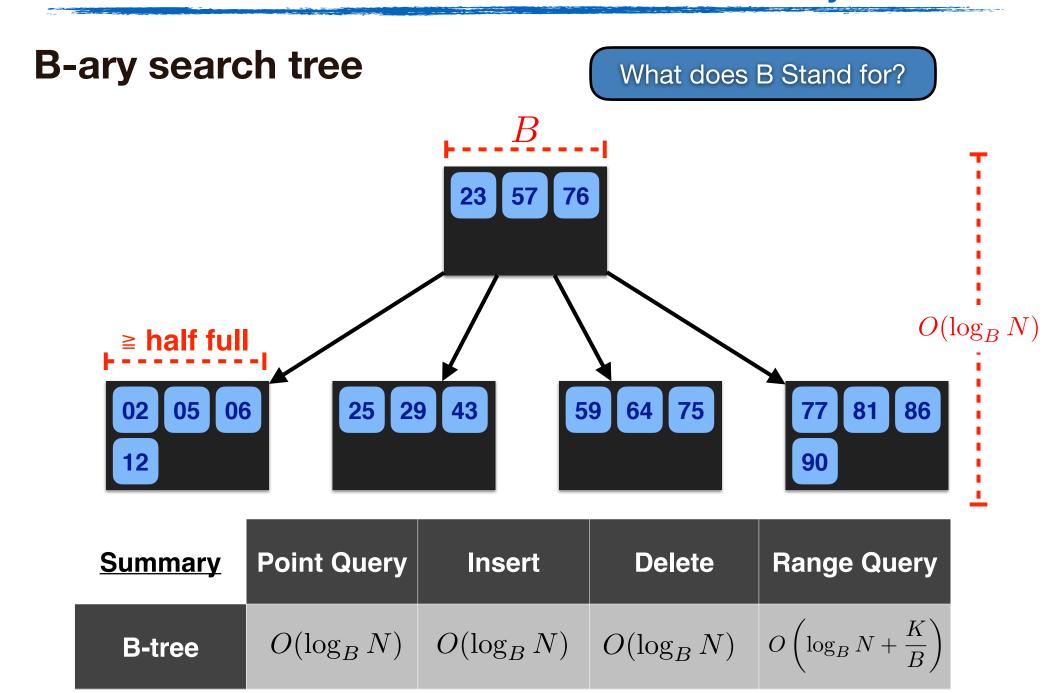












## **Steps**

- Starting at the root, find the first pivot key that is larger than your search key, and follow the pointer to 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

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

#### Cost

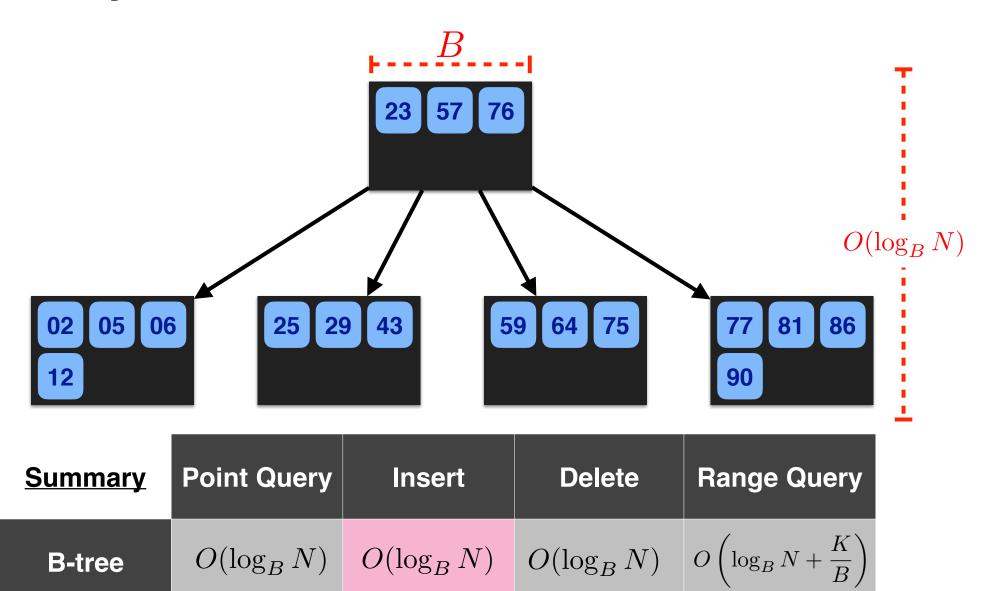
- How many nodes must be read/written in a search?
  - ▶ We read the root node to search the pivot keys
  - We recurse on the subtree
- Total cost of a search: O(h)
  - ightharpoonupRecall  $h = O(log_BN)$
  - ► So point query cost is O(log<sub>B</sub>N)

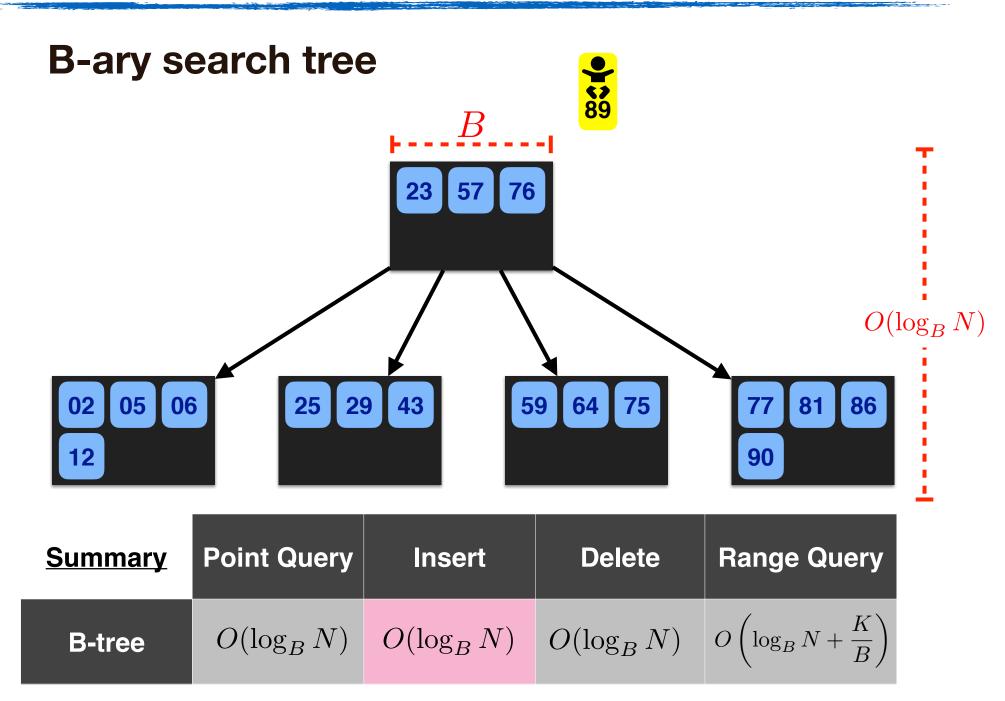
## **B-tree Insertions**

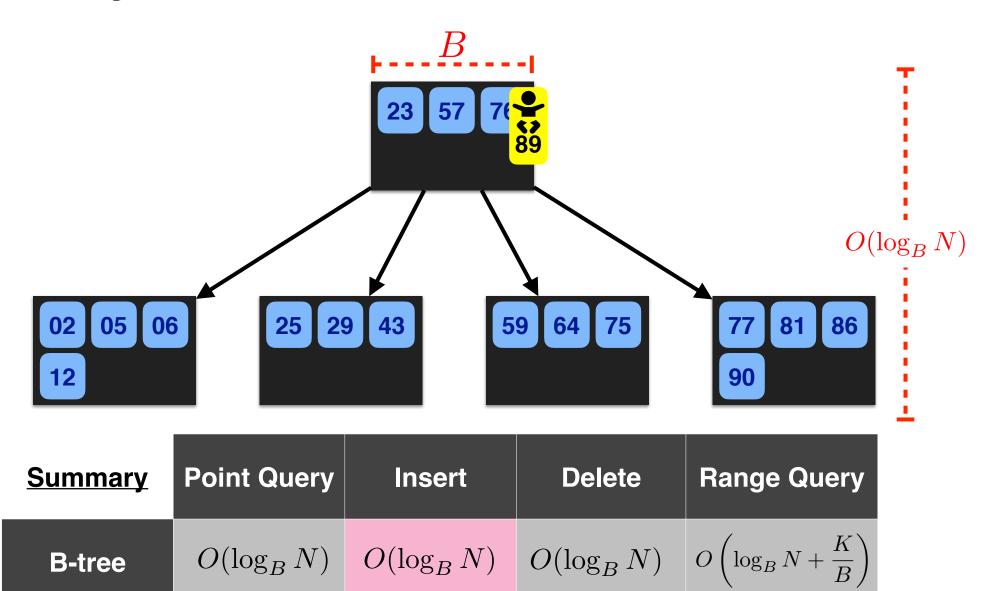
#### B-tree Insert

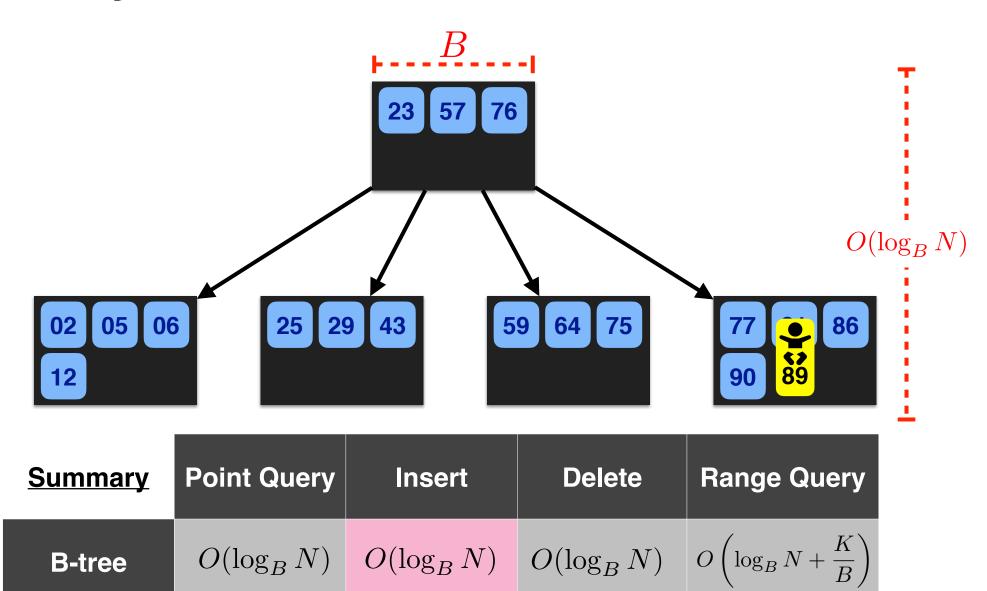
#### **Steps**

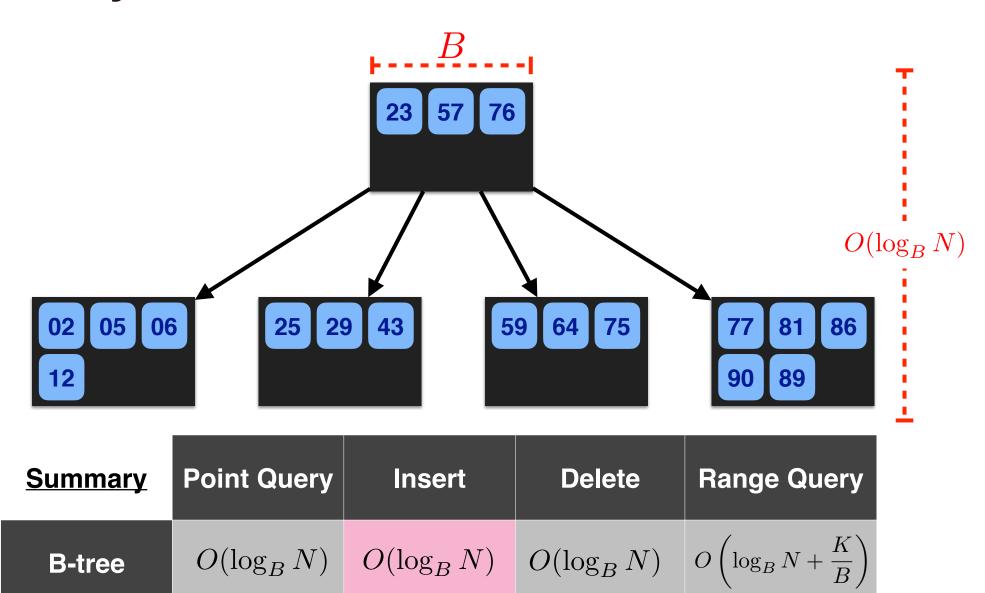
- Find the leaf node where your key-value pair belongs (same steps as a point query)
- Insert your key-value pair into that leaf

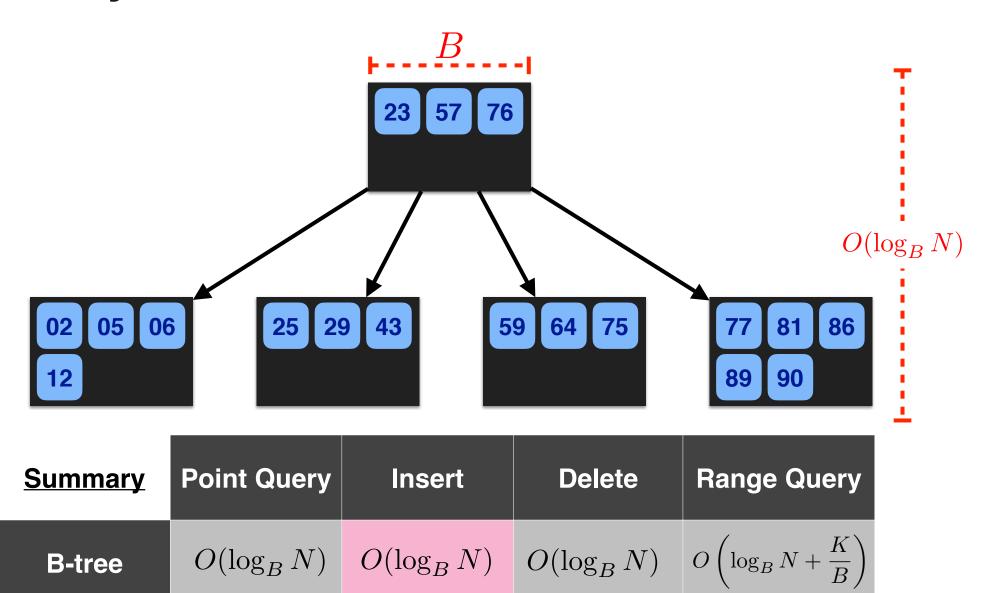


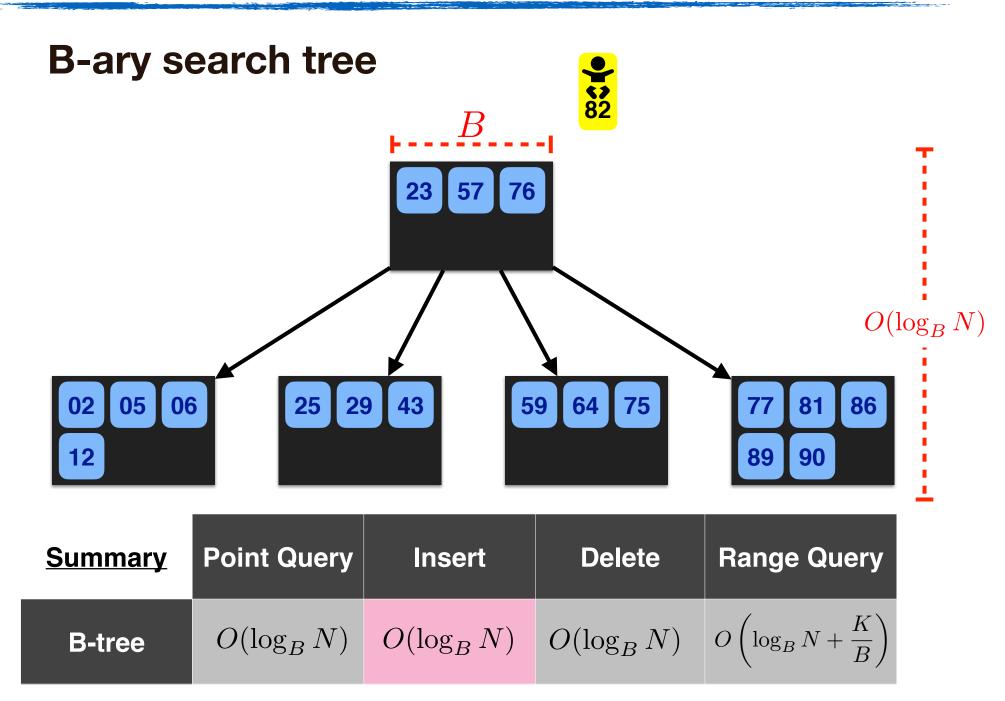


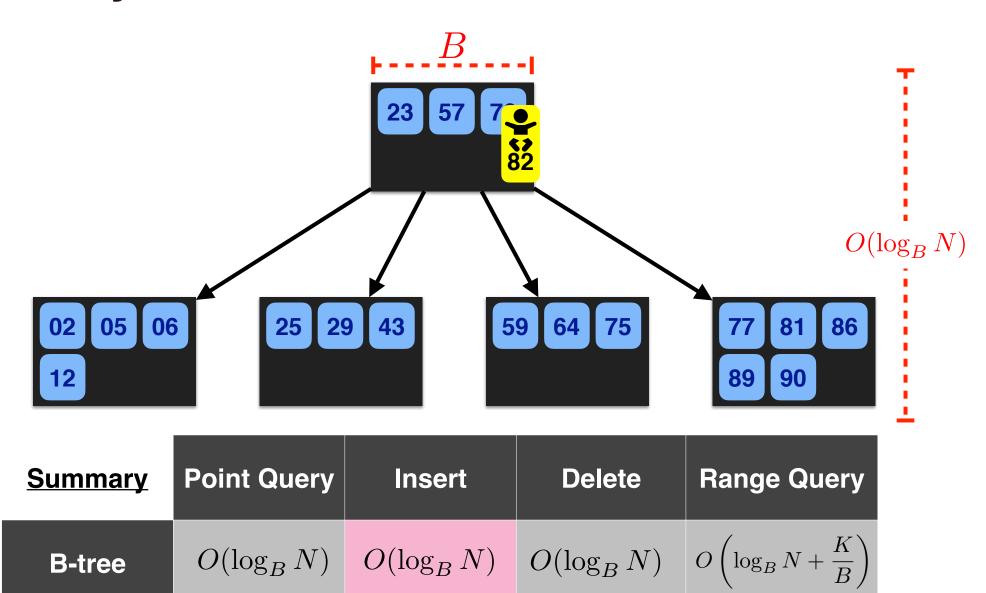


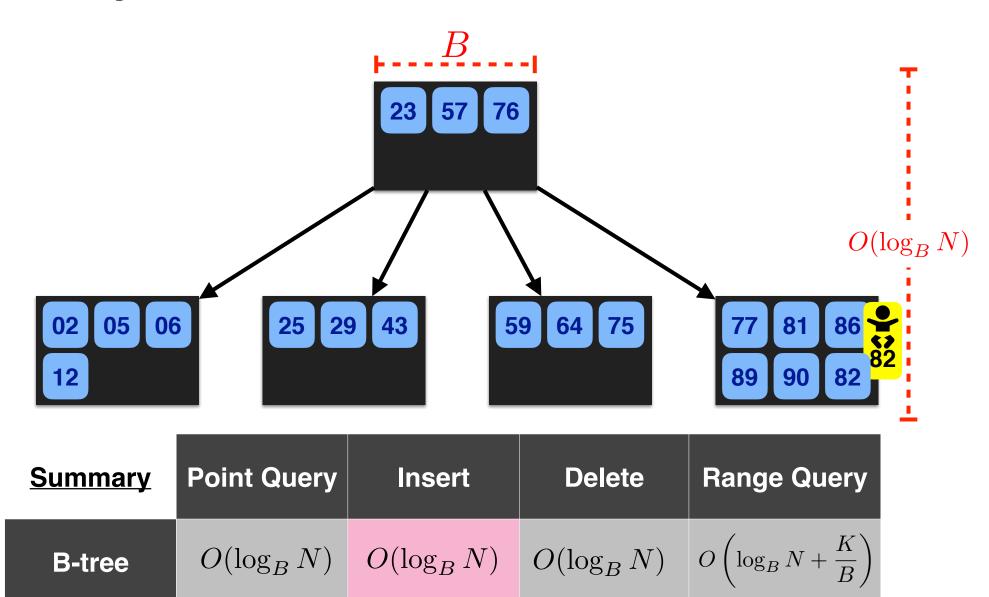


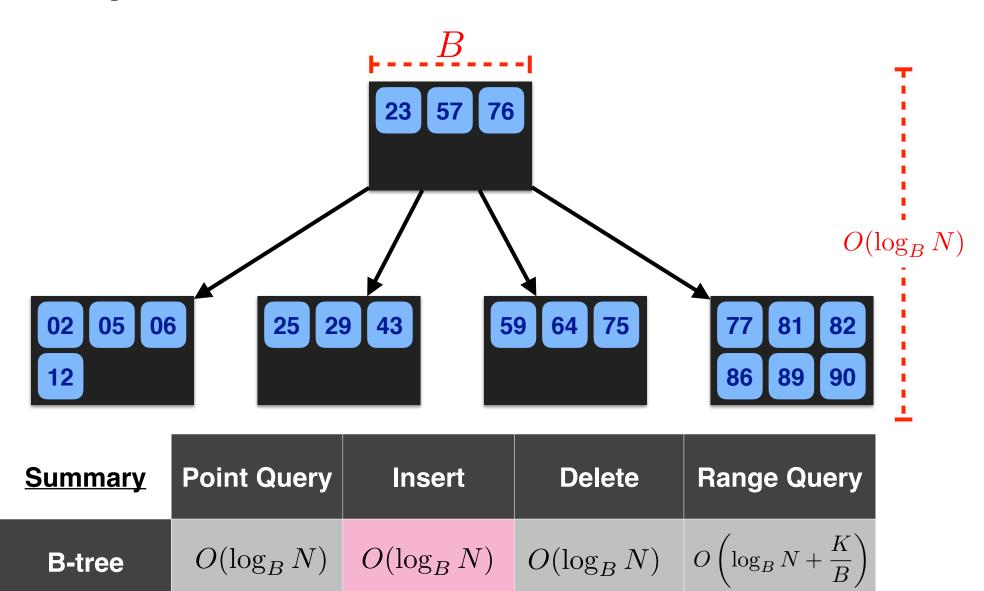


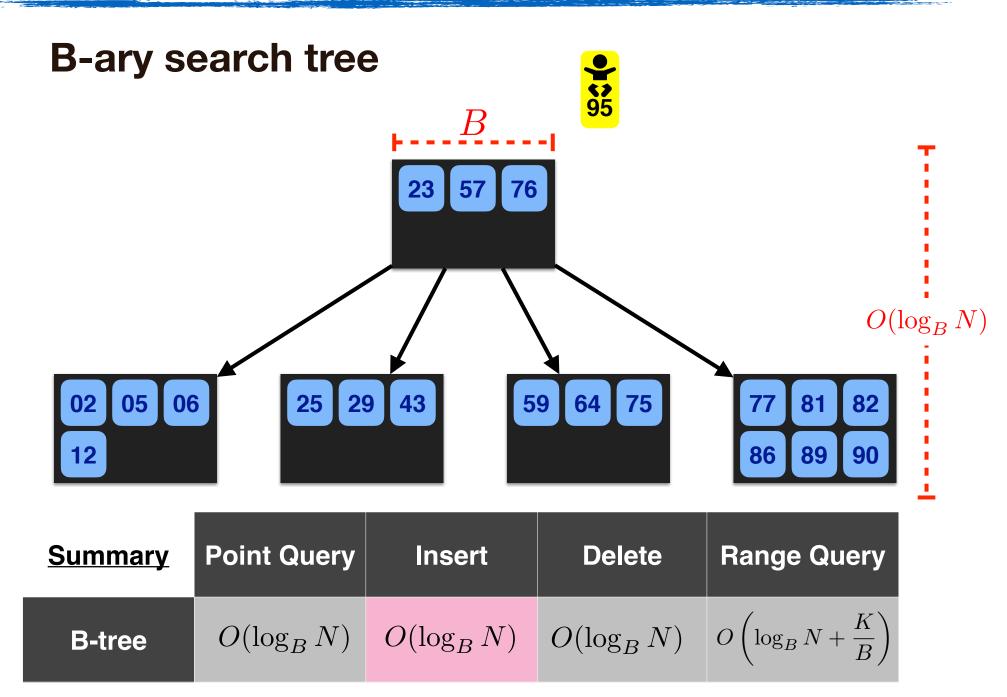


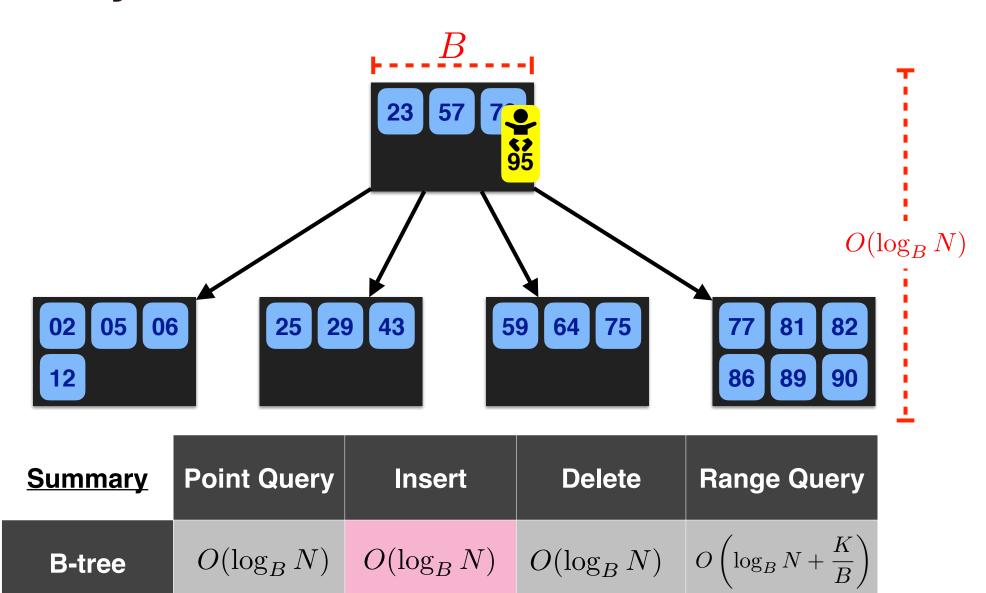


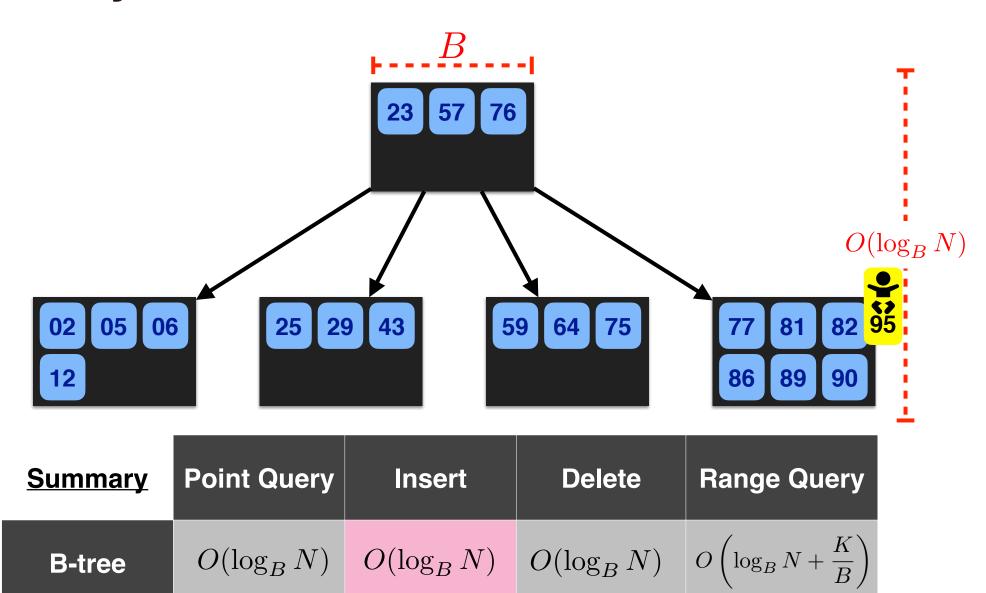


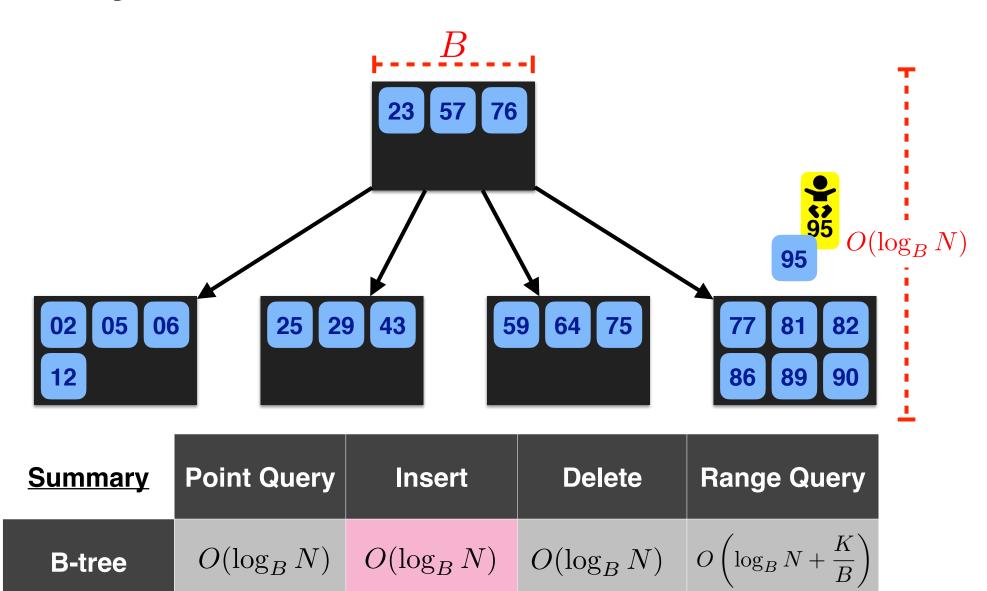


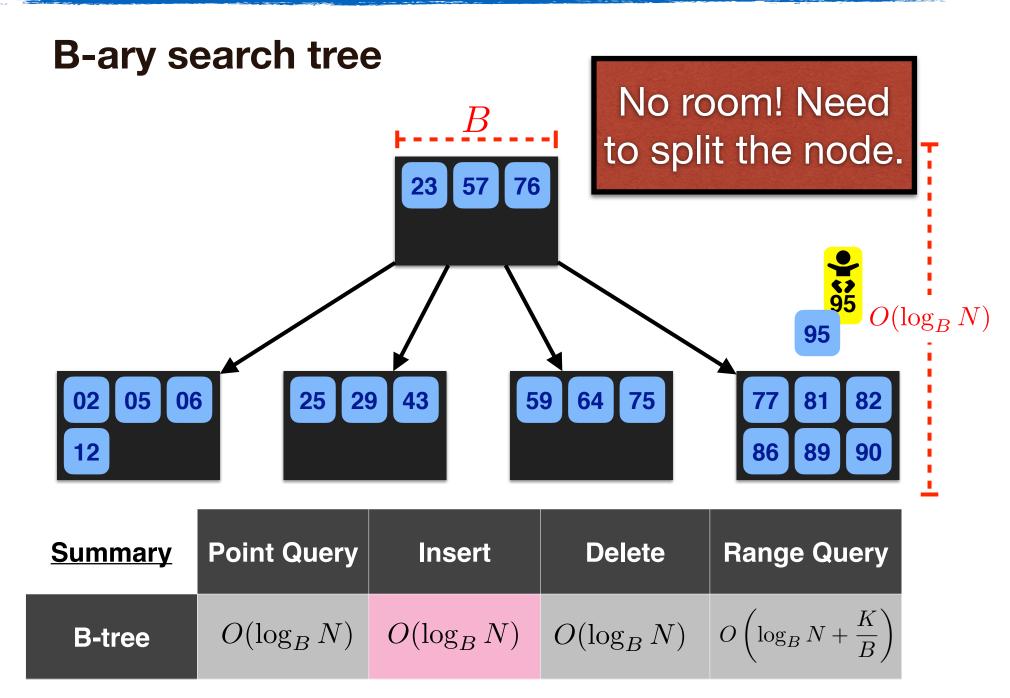








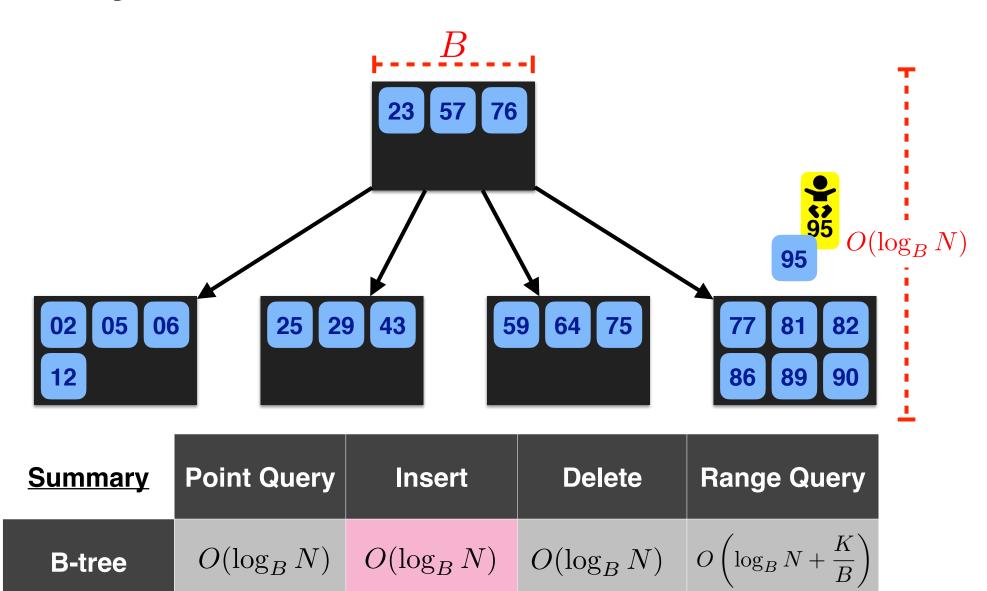




## Splitting a B-tree node

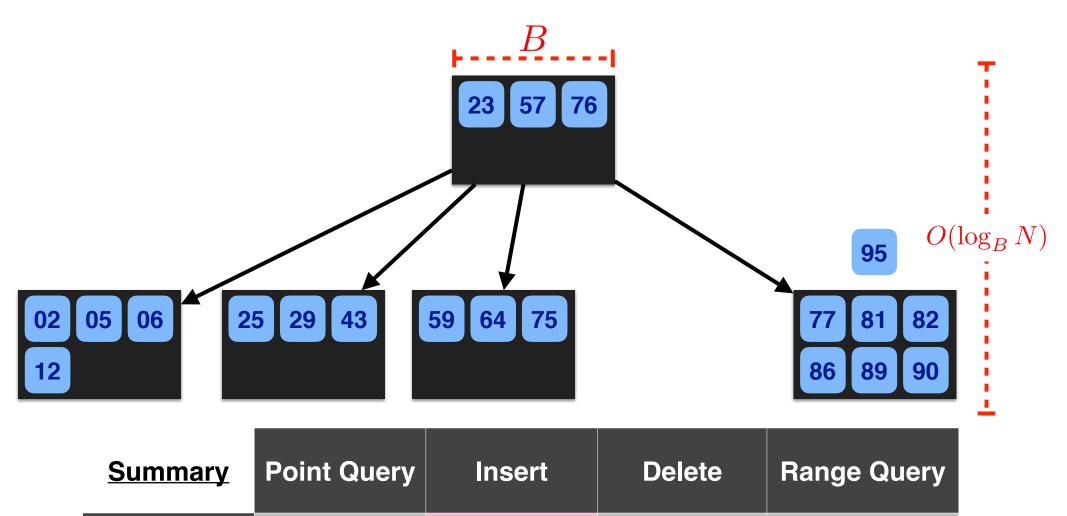
#### **Steps**

- 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)

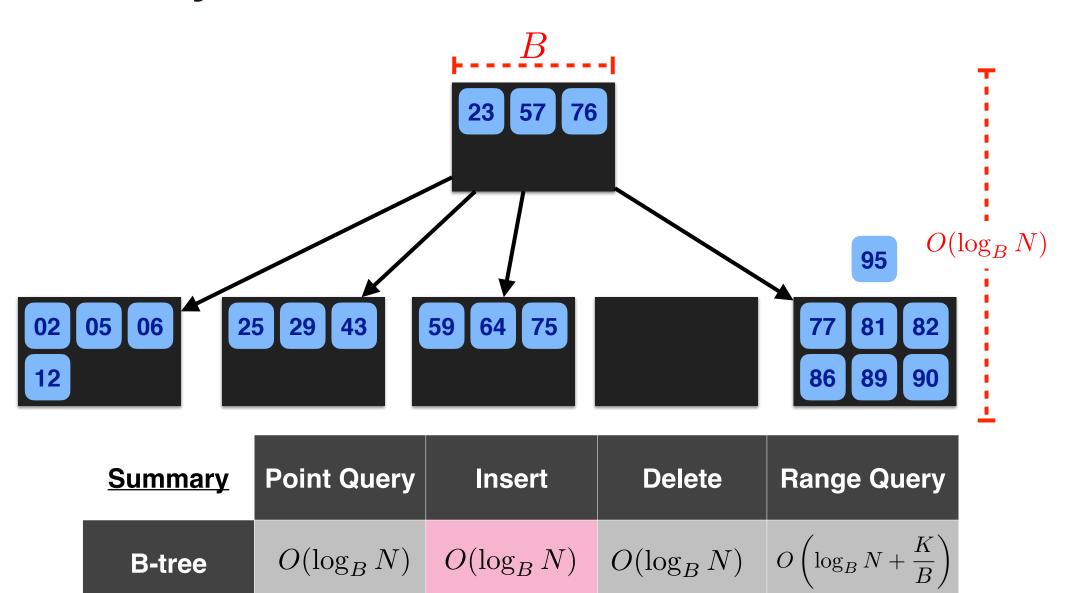


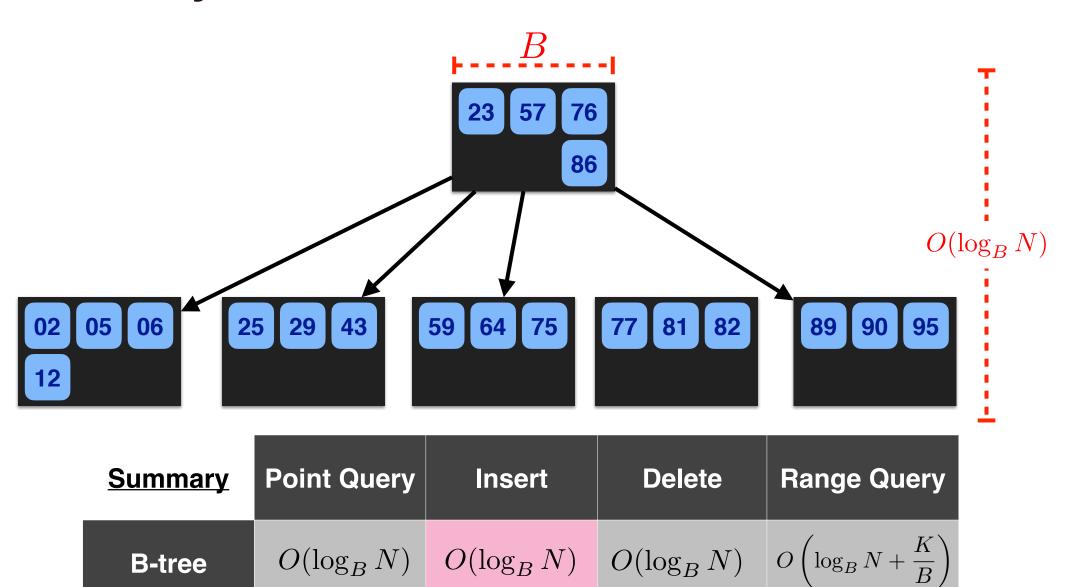
#### **B-ary search tree**

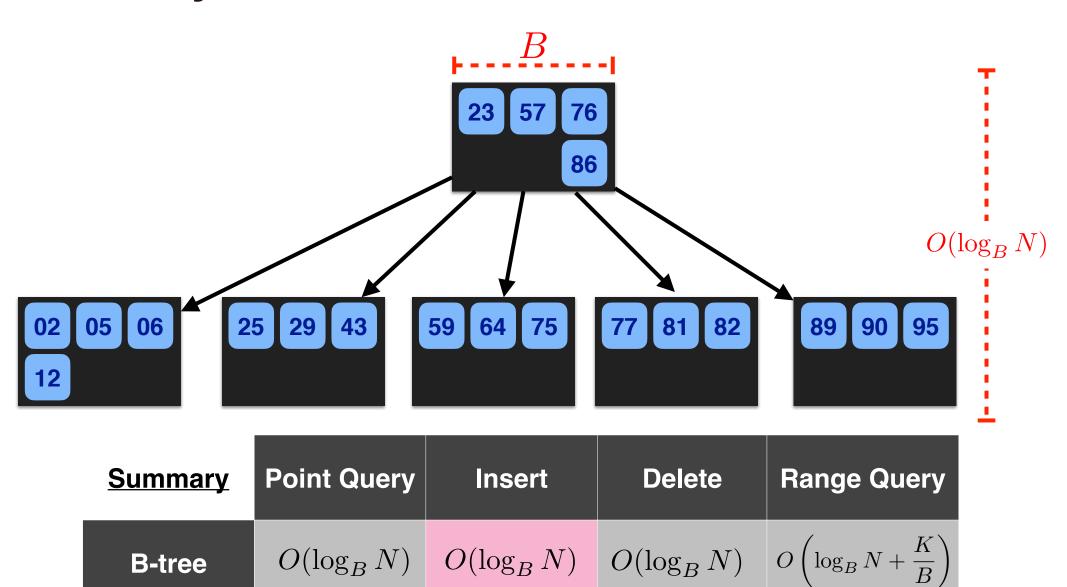
**B-tree** 



 $O(\log_B N)$   $O(\log_B N)$   $O(\log_B N)$   $O(\log_B N + \frac{K}{B})$ 







## 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)

## Splitting a B-tree node

#### Cost

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- 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)
  - $\blacktriangleright \text{Recall } \mathbf{h} = \mathbf{O}(\log_{\mathbf{B}}\mathbf{N})$
  - So insert cost is O(log<sub>B</sub>N)

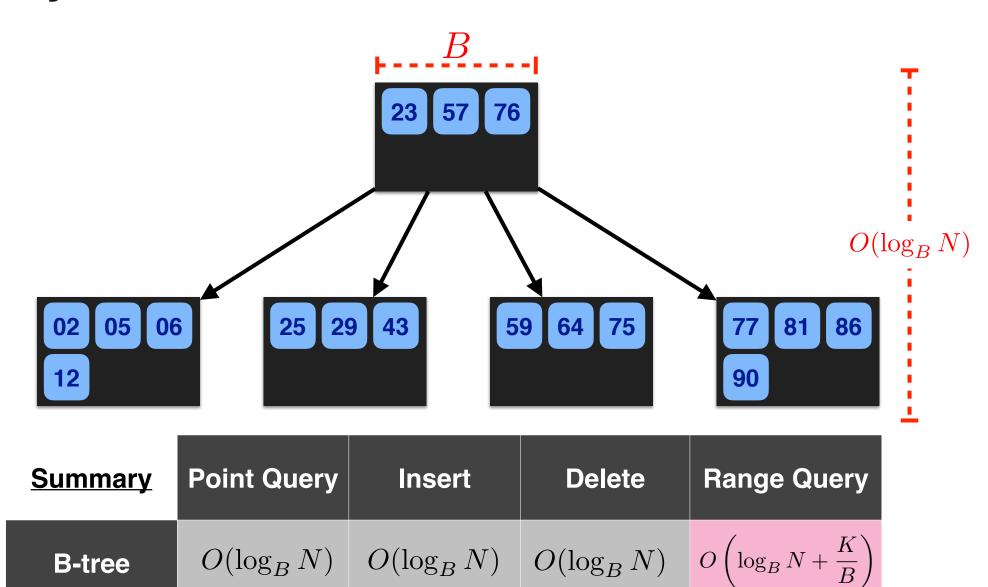
# B-tree Range Queries

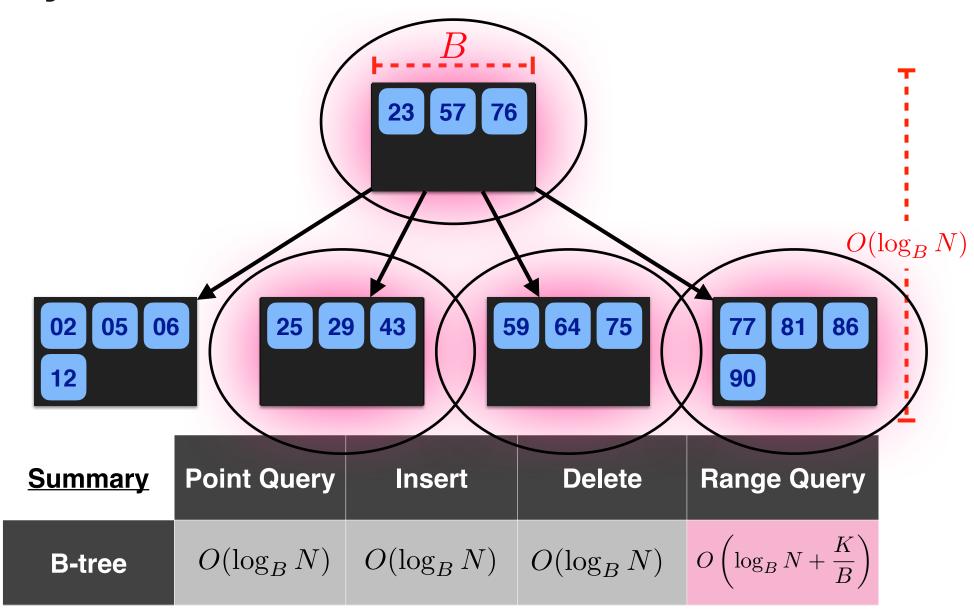
## B-tree Range Query

#### (Range query: point query + successork)

#### **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



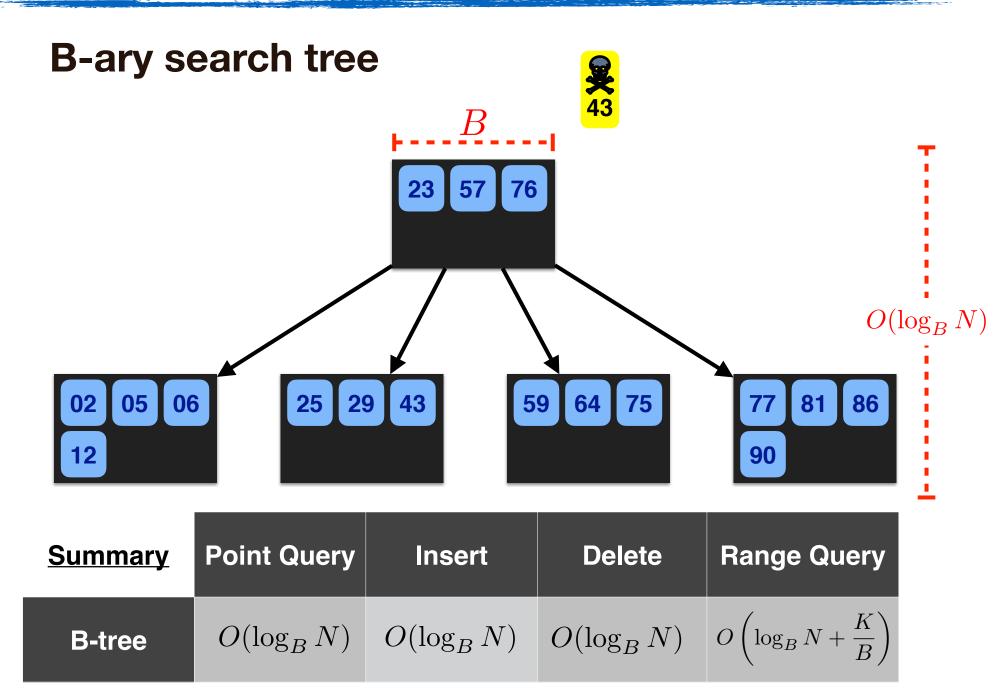


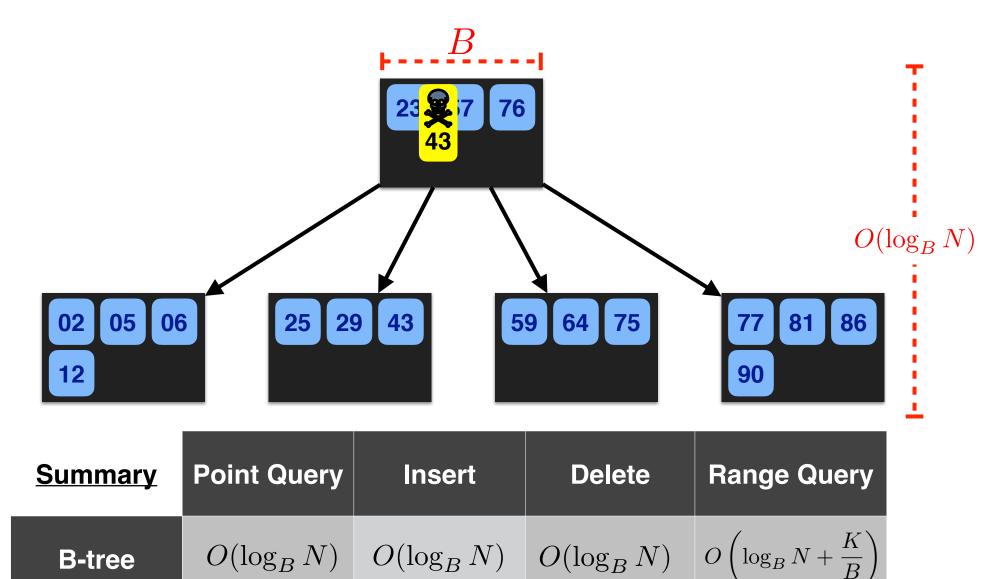
## **B-tree Deletes**

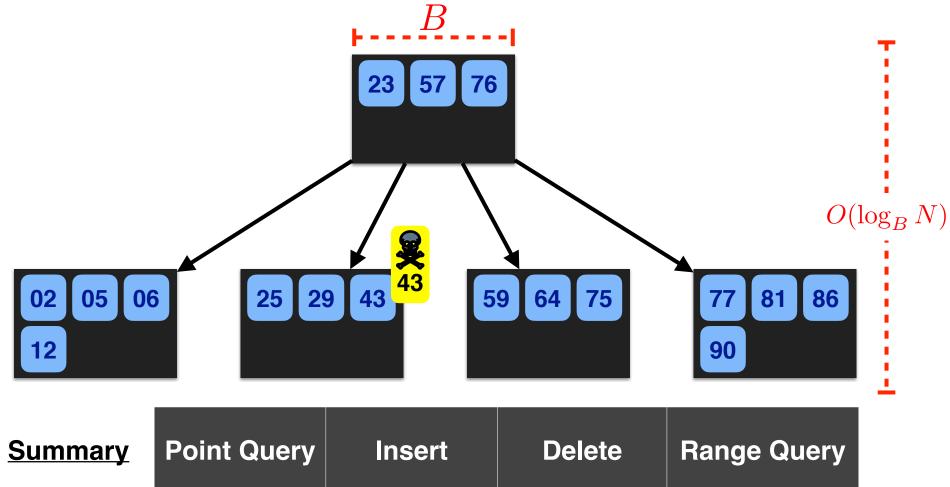
#### **B-tree Deletions**

#### **Steps**

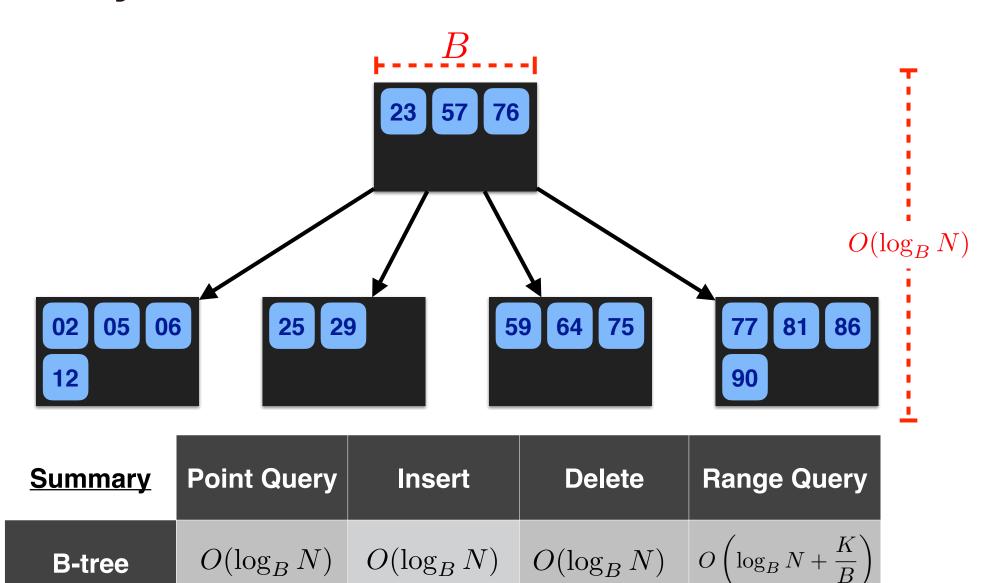
- Search for the leaf containing the target key-value pair (point query)
- Remove the element from the leaf (if present)
- If the size of the node drops below d, merge with a neighbor
  - Remove extra pivot key and pointer from parent (the pointer to the node that is being deleted as part of the merge)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

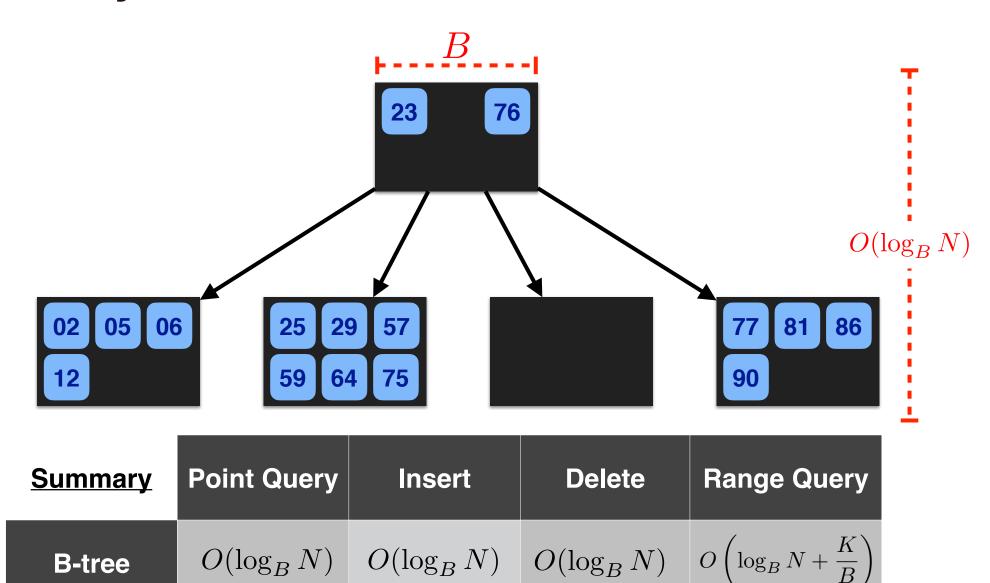


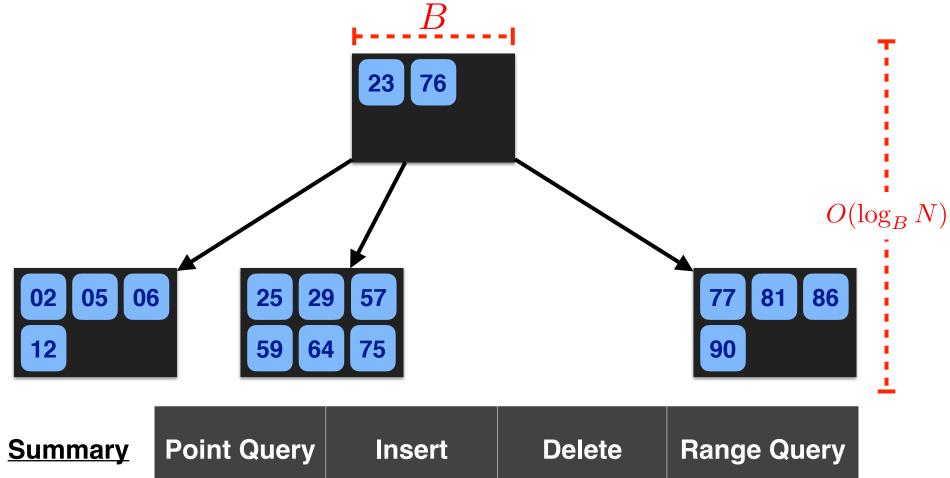




<u>Summary</u>	Point Query	Insert	Delete	Range Query
B-tree	$O(\log_B N)$	$O(\log_B N)$	$O(\log_B N)$	$O\left(\log_B N + \frac{K}{B}\right)$







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- 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
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- Concurrent access how to lock the tree?
  - ▶ Hand-over-hand locking for queries
  - Reservations or top-down splitting

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#### 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**

#### More trees

- Log structured merge trees (next class)
- Be-trees Monday

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### Write optimization

- Making our trees lazy!
  - ▶ Better I/O performance for writes
  - Not worse off for reads