## **Carnegie Mellon University**

# Tree Indexes





Andrew Crotty Computer Science Carnegie Mellon University

#### ADMINISTRIVIA

**Project #1** is due Sunday, Sept 26<sup>th</sup> @11:59pm

Homework #2 is due Sunday, Oct 3<sup>rd</sup> @11:59pm



#### DATA STRUCTURES

Internal Meta-data Core Data Storage Temporary Data Structures Table Indexes



#### DATA STRUCTURES

Internal Meta-data

Core Data Storage

**Temporary Data Structures** 

Table Indexes



#### TABLE INDEXES

A **<u>table index</u>** is a replica of a subset of a table's attributes that are organized and/or sorted for efficient access using those attributes.

The DBMS ensures that the contents of the table and the index are logically synchronized.



#### TABLE INDEXES

It is the DBMS's job to figure out the best index(es) to use to execute each query.

There is a trade-off regarding the number of indexes to create per database.

- $\rightarrow$  Storage Overhead
- $\rightarrow$  Maintenance Overhead



#### TODAY'S AGENDA

B+Tree Overview Use in a DBMS Design Choices Optimizations



#### **B-TREE FAMILY**

There is a specific data structure called a **<u>B-Tree</u>**.

People also use the term to generally refer to a class of balanced tree data structures:

- $\rightarrow$  **B-Tree** (1971)
- → **B+Tree** (1973)
- → **B\*Tree** (1977?)
- $\rightarrow B^{link}$ -Tree (1981)



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SCMUDB 15-445/645 (Fall 2021)  $\rightarrow$  **B**<sup>link</sup>-**Tree** (1981)

#### Efficient Locking for Concurrent Operations

PHILIP L. LEHMAN Carnegie-Mellon University and S. BING YAO Purdue University

The B-tree and its variants have been found to be highly useful (both theoretically and in practice) The D-tree and us variance new open round to us many users toost incorrections and as partners, for storing large amounts of information, especially on secondary storage devices. We examine the for scoung sage smouths or automation, especially on secondary bounge or vices. The standard way problem of overcoming the inherent difficulty of concurrent operations on such structures, using a process os overcoming une amereten tunicanty os constartens operations on such turacultes, emag a practical storage model. A single additional "link" pointer in each node allows a process to easily presentes storage modifications performed by other concurrent processes. Our solution compares recover nom use monuncations performed by other tonoursent processes out sension comparison favorably with earlier solutions in that the locking scheme is simpler (no read-locks are used) and involancy with earlier solutions in that the needing externs is support the requirements are used, and only a (small) constant number of nodes are locked by any update process at any given time. An

Key Words and Phrases: database, data structures, B-tree, index organizations, concurrent algorithms, concurrency controls, locking protocols, correctness, consistency, multiway search trees CR Categories: 3.73, 3.74, 4.32, 4.33, 4.34, 5.24

#### 1. INTRODUCTION

The B-tree [2] and its variants have been widely used in recent years as a data structure for storing large files of information, especially on secondary storage devices [7]. The guaranteed small (average) search, insertion, and deletion time for these structures makes them quite appealing for database applications.

A topic of current interest in database design is the construction of databases that can be manipulated concurrently and correctly by several processes. In this paper, we consider a simple variant of the B-tree (actually of the B\*-tree, proposed by Wedekind [15]) especially well suited for use in a concurrent database

Methods for concurrent operations on B\*-trees have been discussed by Bayer and Schkolnick [3] and others [6, 12, 13]. The solution given in the current paper

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Authors' present addresses: P. L. Lehman, Department of Computer Science, Carnegie Mellon Automa presents doutremest. F. & Letonman, Department of Computer Science, Outremestations, University, Pittaburgh, PA 15213; S. B. Yao, Department of Computer Science and College of Business and Management, University of Maryland, College Park, MD 20742. © 1981 ACM 0362-5915/81/1200-0650 \$00.75 ACM Transactions on Database Systems, Vol. 6, No. 4, December 1981, Pages 650-670.

#### **B-TREE FAMILY**

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### B+TREE

A **B**+**Tree** is a self-balancing tree data structure that keeps data sorted and allows searches, sequential access, insertions, and deletions in O(log n).

 $\rightarrow$  Generalization of a binary search tree, since a node can have more than two children.

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 $\rightarrow$  Optimized for systems that read and write large blocks of data.

#### The Ubiguitous B-Tree

DOUGLAS COMER

Computer Science Department, Purdue University, West Lafayette, Indiana 47907

B-trees have become, de facto, a standard for file organization. File indexes of users, dedicated database systems, and general-purpose access methods have all been proposed and implemented using B-trees This paper reviews B-trees and shows why they have been so successful It discusses the major variations of the B-tree, especially the B+-tree, contrasting the relative merits and costs of each implementation. It illustrates a general purpose access method which uses a B-tree.

Keywords and Phrases: B-tree, B\*-tree, B\*-tree, file organization, index

CR Categories: 3.73 3.74 4.33 4 34

#### INTRODUCTION

The secondary storage facilities available on large computer systems allow users to store, update, and recall data from large collections of information called files. A computer must retrieve an item and place it in main memory before it can be processed. In order to make good use of the computer resources, one must organize files intelligently, making the retrieval process efficient.

The choice of a good file organization depends on the kinds of retrieval to be performed. There are two broad classes of retrieval commands which can be illustrated by the following examples:

Sequential: "From our employee file, prepare a list of all employees' names and addresses," and Random:

"From our employee file, ex- to the employee file, where the topmost tract the information about employee J. Smith".

We can imagine a filing cabinet with three folders. drawers of folders, one folder for each em-

index consists of labels on drawers, and the next level of index consists of labels on Natural hierarchies, like the one formed ployee. The drawers might be labeled "A- by considering last names as index entries, G." "H-R." and "S-Z," while the folders do not always produce the best perform-

might be labeled with the employees' last

names. A sequential request requires the

searcher to examine the entire file, one folder at a time. On the other hand, a

random request implies that the searcher,

guided by the labels on the drawers and

Associated with a large, randomly ac-

cessed file in a computer system is an index

which, like the labels on the drawers and

folders of the file cabinet, speeds retrieval

by directing the searcher to the small part

of the file containing the desired item. Fig-

ure 1 depicts a file and its index. An index

may be physically integrated with the file,

like the labels on employee folders, or phys-

ically separate, like the labels on the drawers. Usually the index itself is a file. If the

index file is large, another index may be

built on top of it to speed retrieval further,

and so on. The resulting hierarchy is similar

folders, need only extract one folder.

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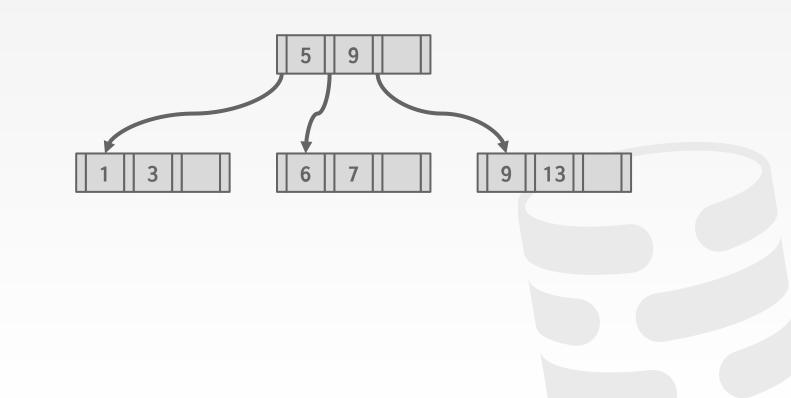
Computing Surveys, Vol. 11, No. 2, June 1979

#### **B+TREE PROPERTIES**

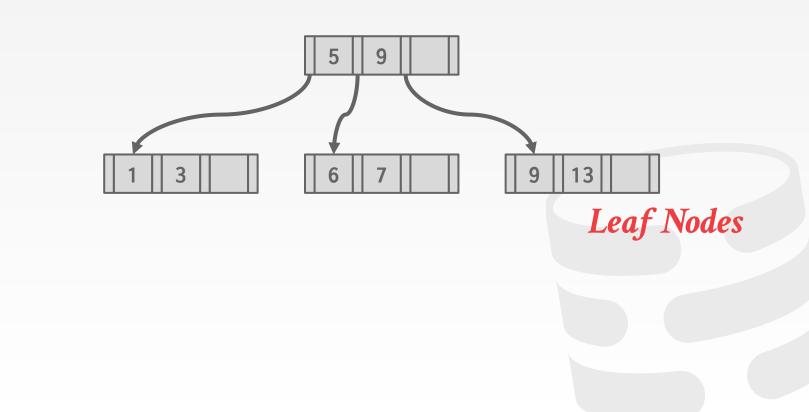
A B+Tree is an *M*-way search tree with the

- following properties:
- → It is perfectly balanced (i.e., every leaf node is at the same depth in the tree)
- → Every node other than the root is at least half-full
  M/2-1 ≤ #keys ≤ M-1
- $\rightarrow$  Every inner node with **k** keys has **k+1** non-null children

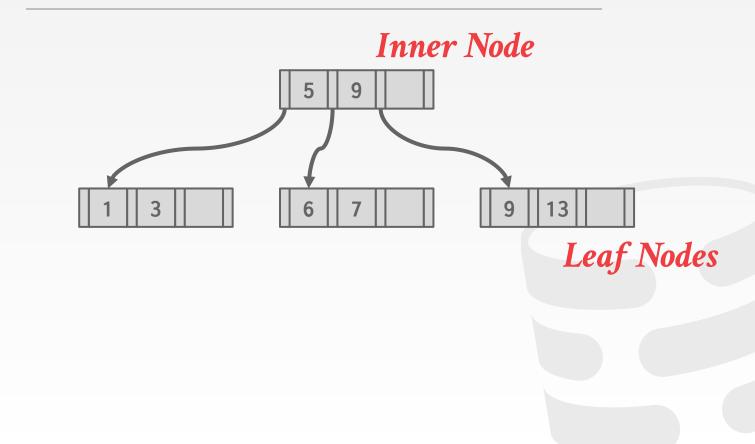




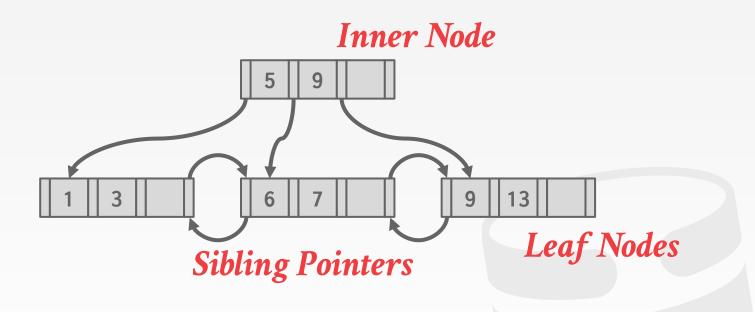




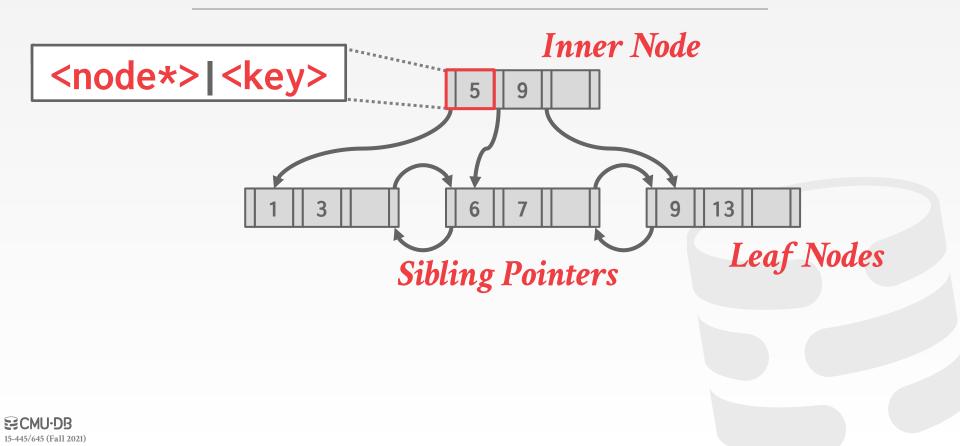
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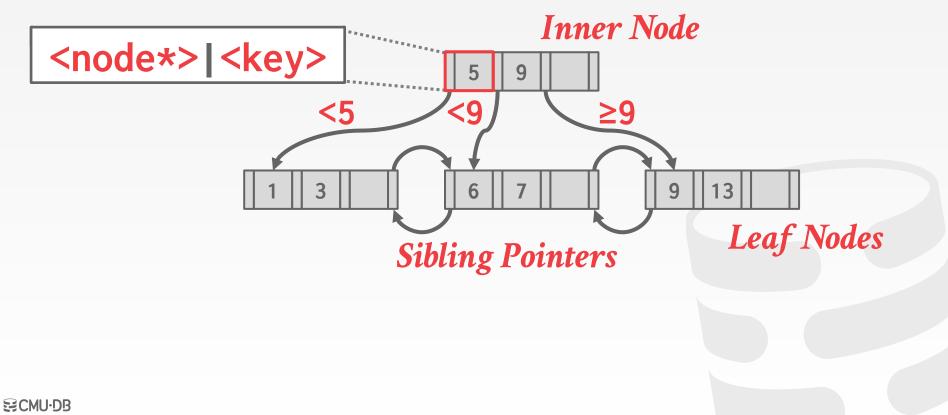


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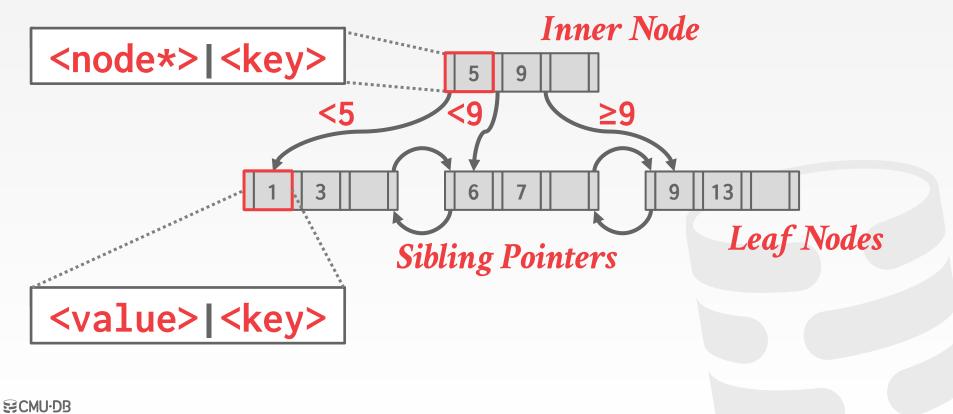








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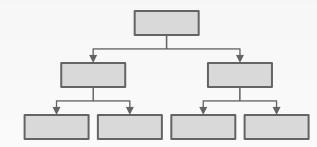
### NODES

Every B+Tree node is comprised of an array of key/value pairs.

- $\rightarrow$  The keys are derived from the attribute(s) that the index is based on.
- $\rightarrow$  The values will differ based on whether the node is classified as an <u>inner node</u> or a <u>leaf node</u>.

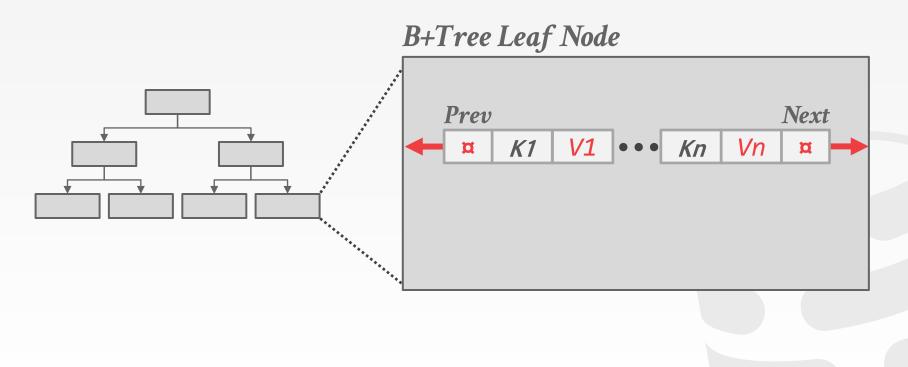
The arrays are (usually) kept in sorted key order.



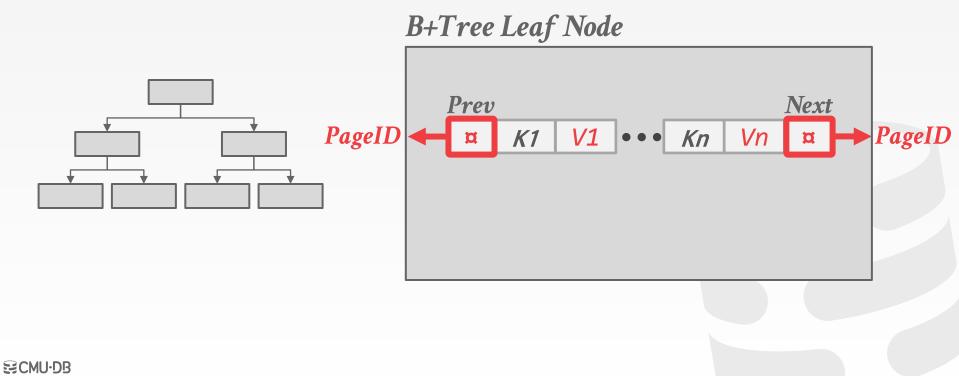




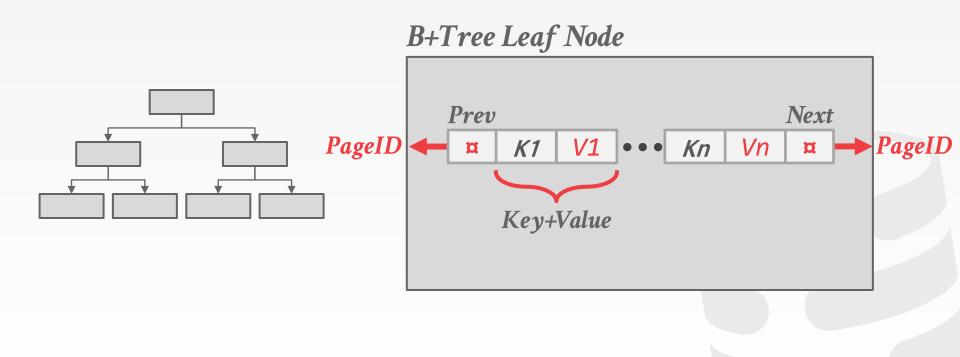
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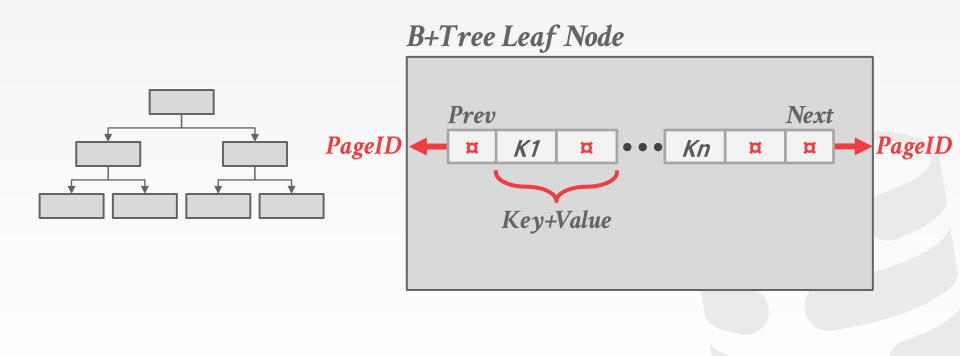




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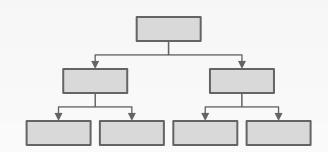


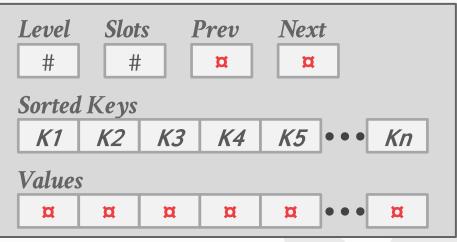






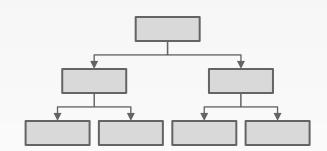
#### B+Tree Leaf Node

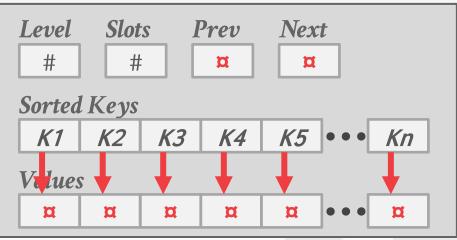






#### B+Tree Leaf Node







#### LEAF NODE VALUES

#### Approach #1: Record IDs

 $\rightarrow$  A pointer to the location of the tuple to which the index entry corresponds.

#### Approach #2: Tuple Data

- $\rightarrow$  The leaf nodes store the actual contents of the tuple.
- $\rightarrow$  Secondary indexes must store the Record ID as their values.





### LEAF NODE VALUES

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**Musql** 

PostgreSQL PostgreSQL

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#### **B-TREE VS. B+TREE**

The original **B-Tree** from 1972 stored keys and values in all nodes in the tree.

 $\rightarrow$  More space-efficient, since each key only appears once in the tree.

A **B+Tree** only stores values in leaf nodes. Inner nodes only guide the search process.



The DBMS can use a B+Tree index if the query provides any of the attributes of the search key.

Example: Index on <a,b,c>

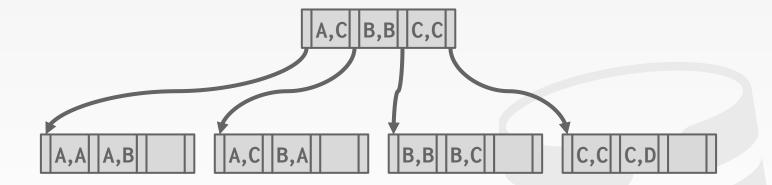
 $\rightarrow$  Supported: (a=5 AND b=3)

 $\rightarrow$  Supported: (b=3)

Not all DBMSs support this.

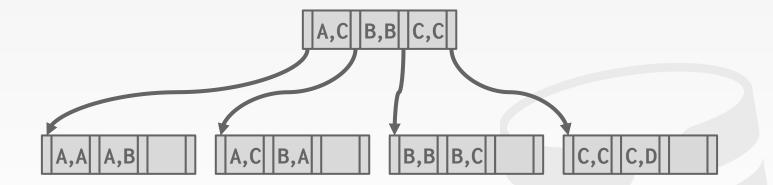
For a hash index, we must have all attributes in search key.



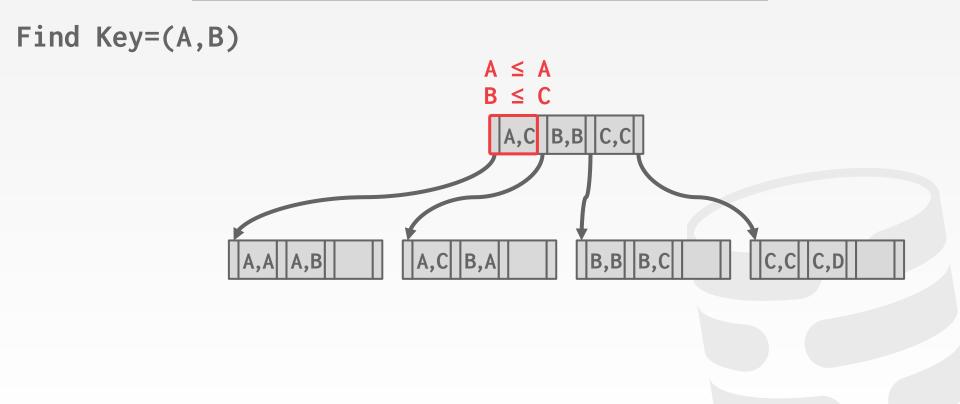




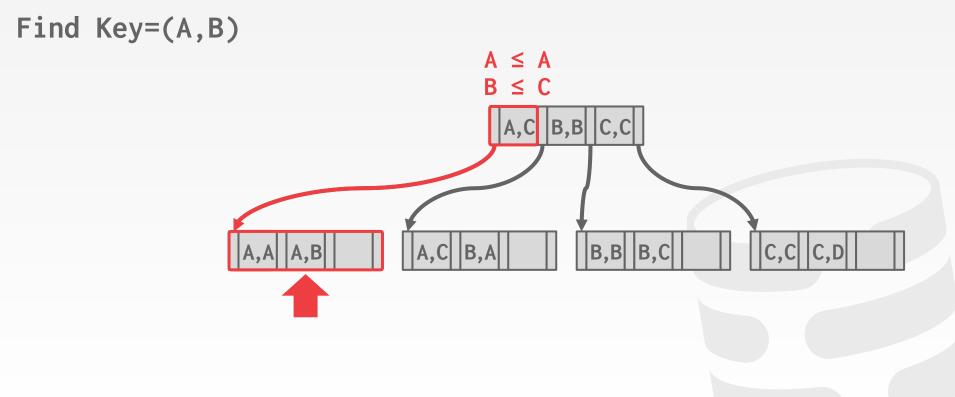
#### Find Key=(A,B)





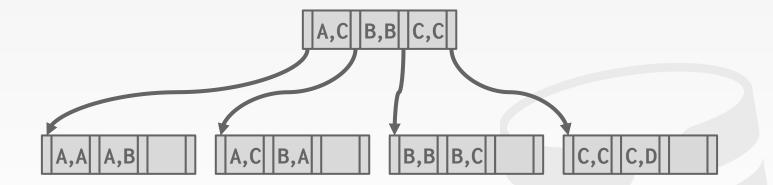






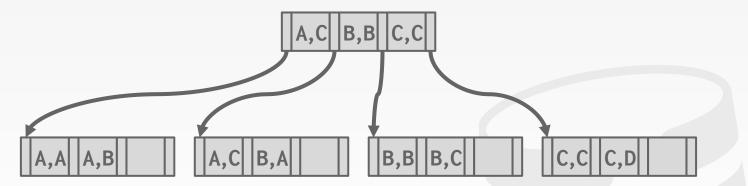


#### Find Key=(A,B)

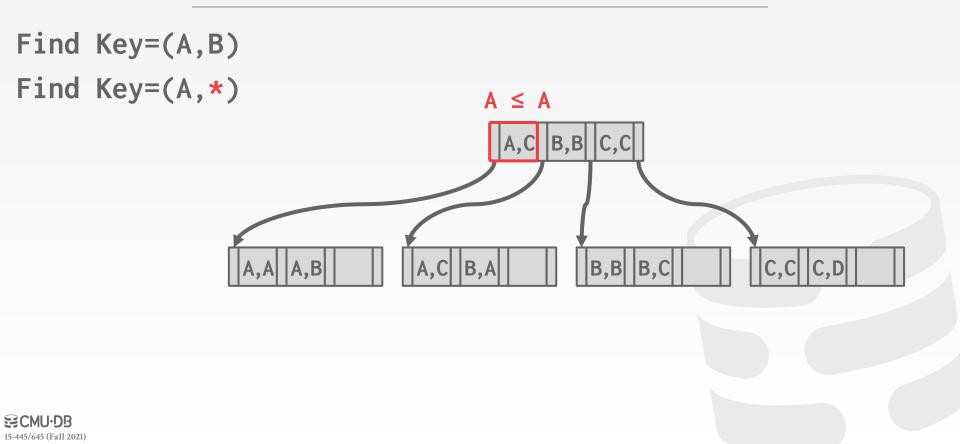


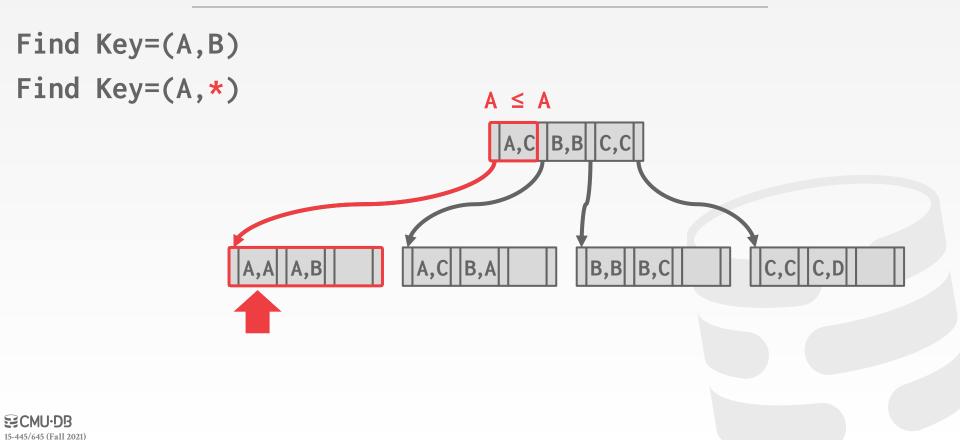


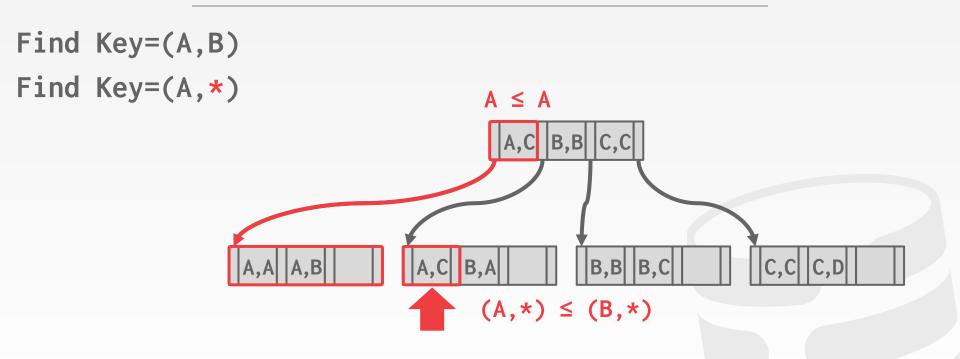
Find Key=(A,B)
Find Key=(A,\*)













## B+TREE - INSERT

Find correct leaf node L.

Put data entry into L in sorted order.

If L has enough space, done!

Otherwise, split L keys into L and a new node L2

 $\rightarrow$  Redistribute entries evenly, copy up middle key.

 $\rightarrow$  Insert index entry pointing to L2 into parent of L.

To split inner node, redistribute entries evenly, but push up middle key.

Source: Chris Re CMU-DB

# B+TREE – DELETE

Start at root, find leaf L where entry belongs. Remove the entry.

If L is at least half-full, done!

If L has only M/2-1 entries,

 $\rightarrow$  Try to re-distribute, borrowing from sibling (adjacent node with same parent as L).

 $\rightarrow$  If re-distribution fails, merge L and sibling.

If merge occurred, must delete entry (pointing to L or sibling) from parent of L.

Source: Chris Re CMU-DB

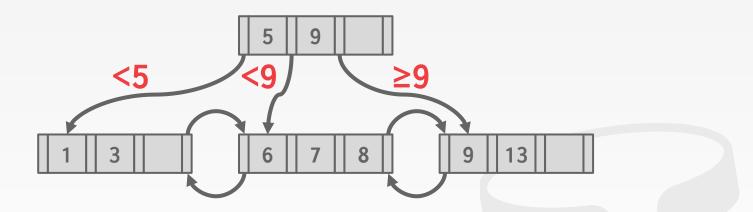
# B+TREE – DUPLICATE KEYS

#### Approach #1: Append Record ID

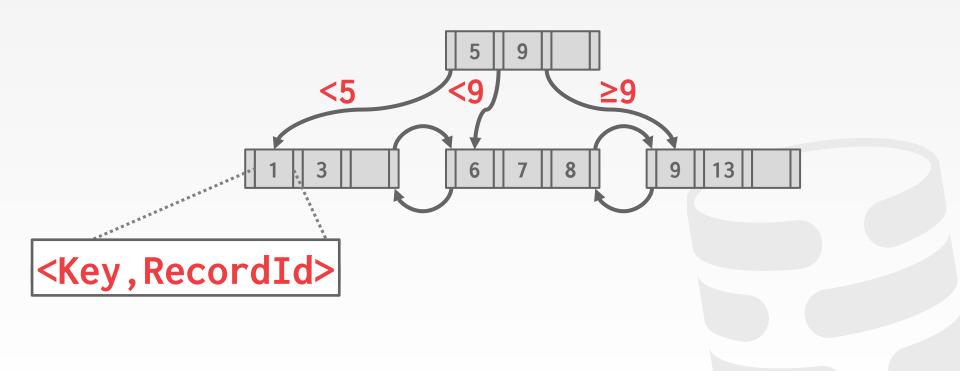
- $\rightarrow$  Add the tuple's unique Record ID as part of the key to ensure that all keys are unique.
- $\rightarrow$  The DBMS can still use partial keys to find tuples.

#### Approach #2: Overflow Leaf Nodes

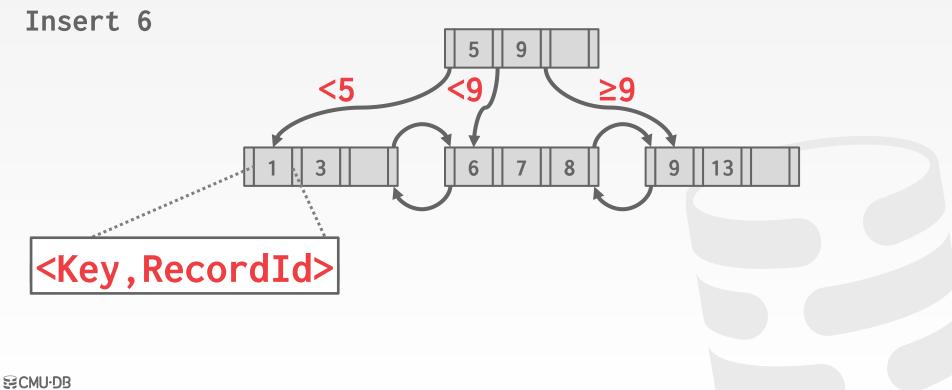
- $\rightarrow$  Allow leaf nodes to spill into overflow nodes that contain the duplicate keys.
- $\rightarrow$  This is more complex to maintain and modify.



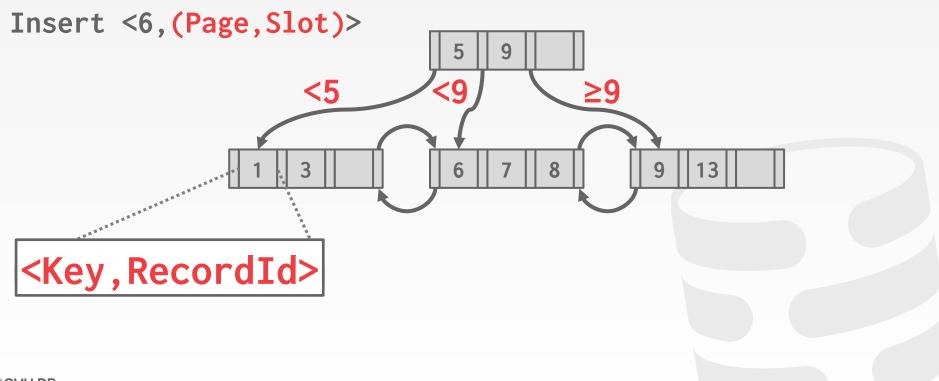




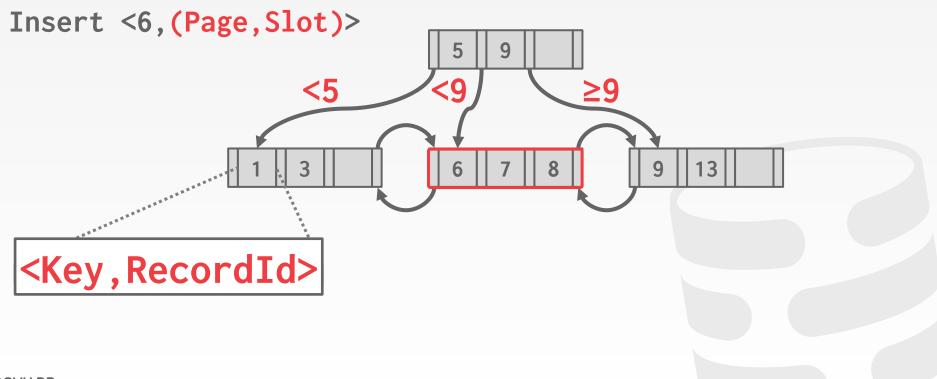




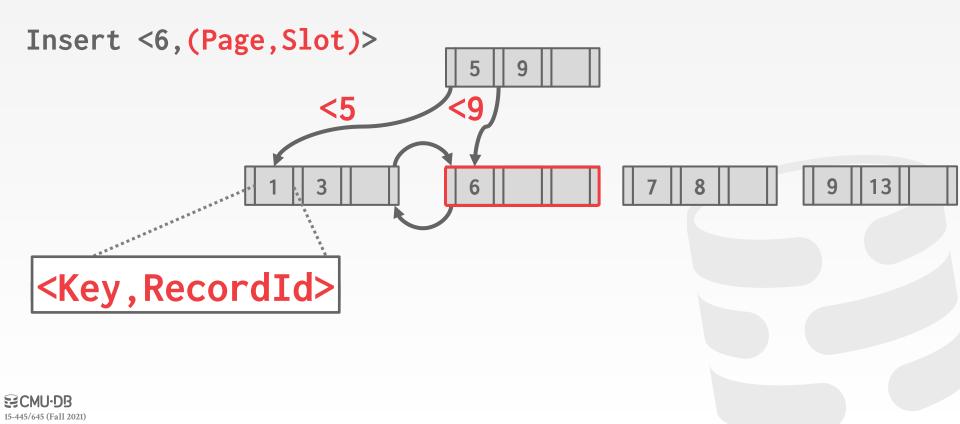
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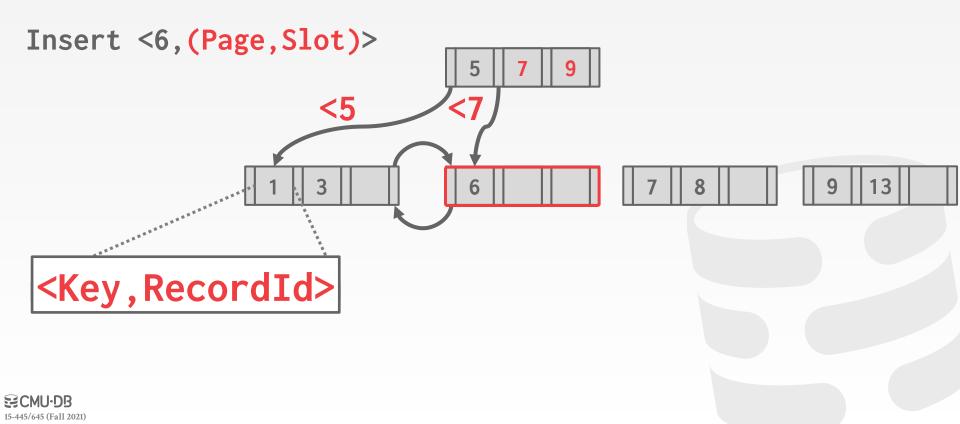


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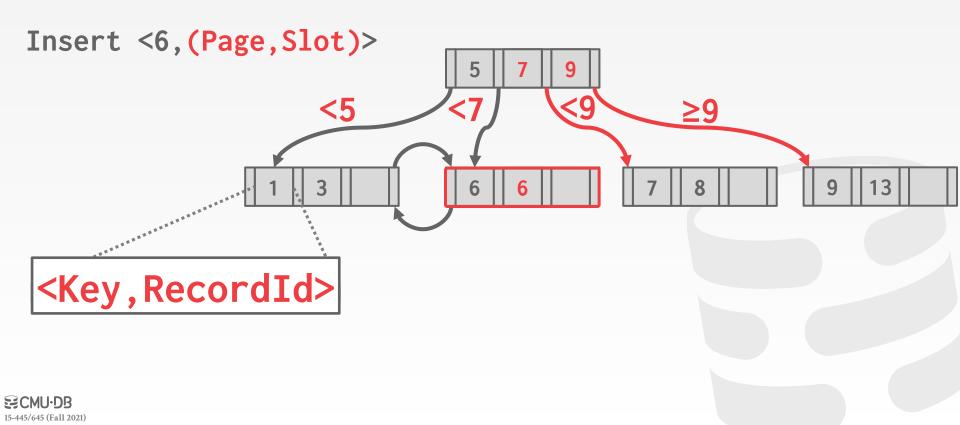


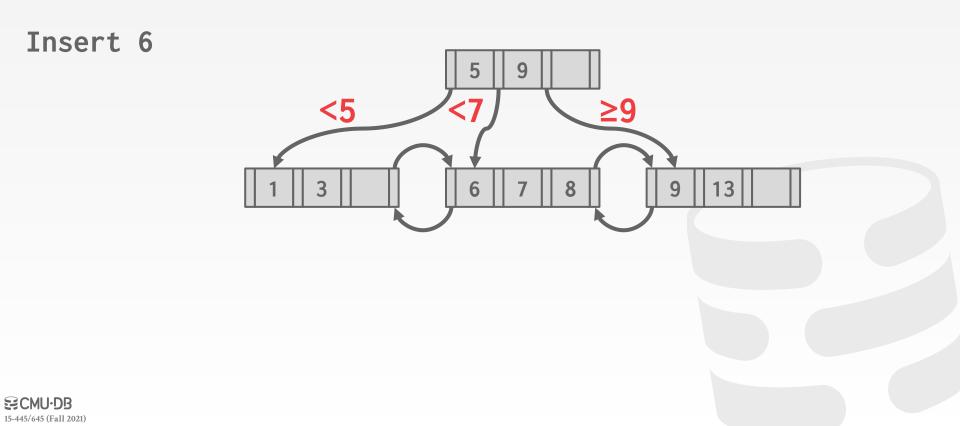
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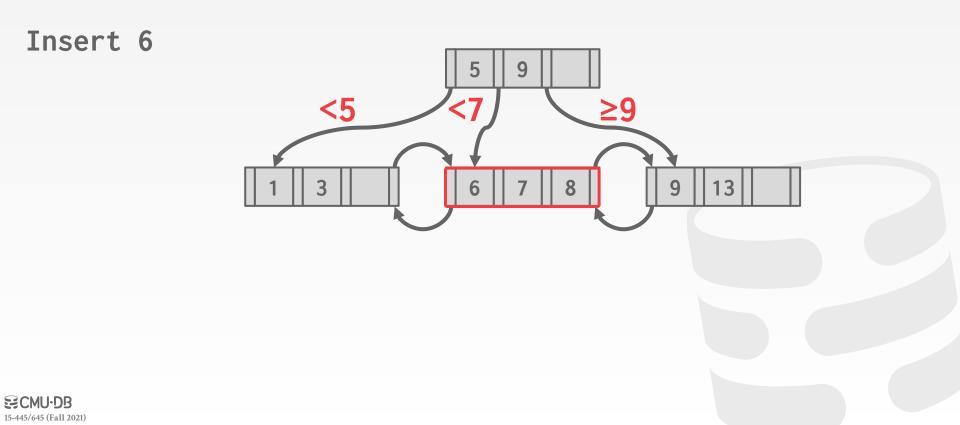


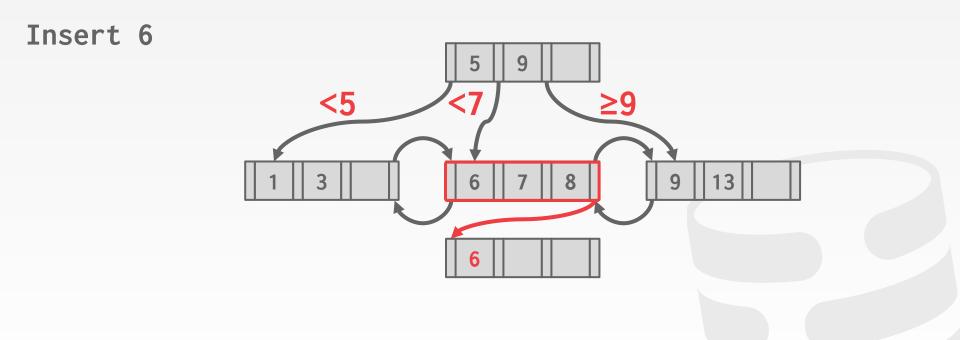


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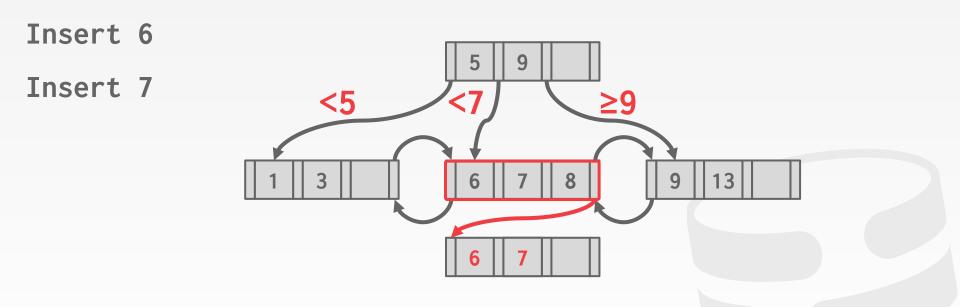




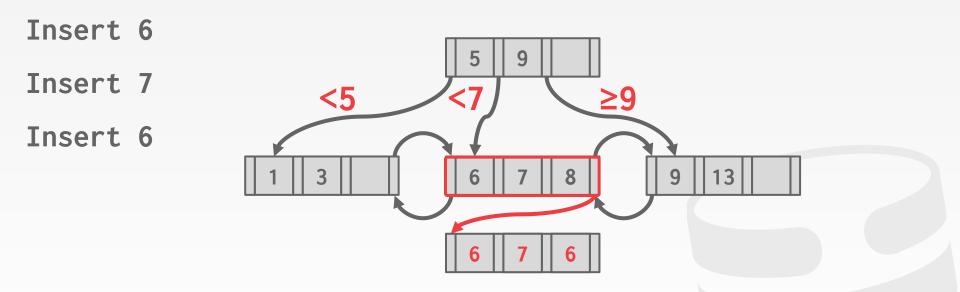














## **CLUSTERED INDEXES**

The table is stored in the sort order specified by the primary key.  $\rightarrow$  Can be either heap- or index-organized storage.

Some DBMSs always use a clustered index.
 → If a table does not contain a primary key, the DBMS will automatically make a hidden primary key.

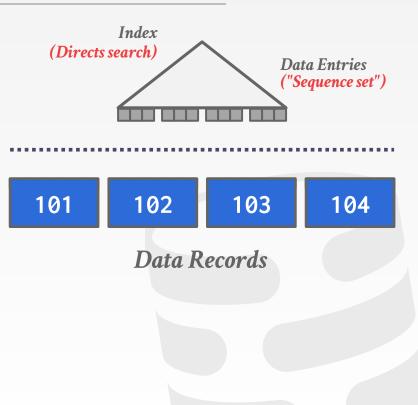
Other DBMSs cannot use them at all.



## **CLUSTERED B+TREE**

Traverse to the left-most leaf page and then retrieve tuples from all leaf pages.

This will always be better than external sorting.

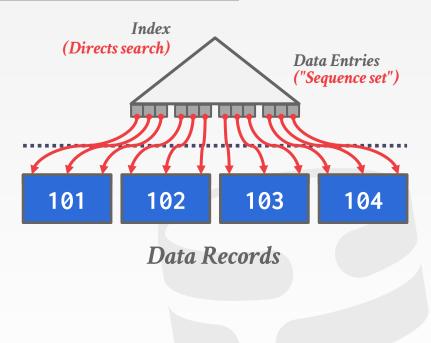




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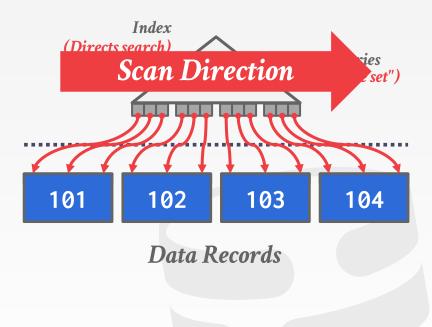




## **CLUSTERED B+TREE**

Traverse to the left-most leaf page and then retrieve tuples from all leaf pages.

This will always be better than external sorting.





Retrieving tuples in the order they appear in a non-clustered index can be very inefficient.



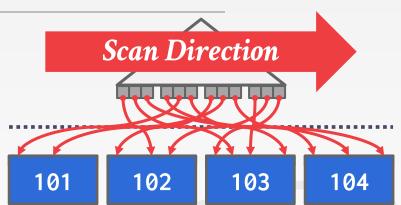


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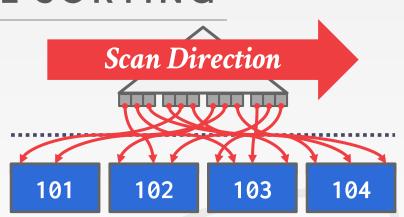
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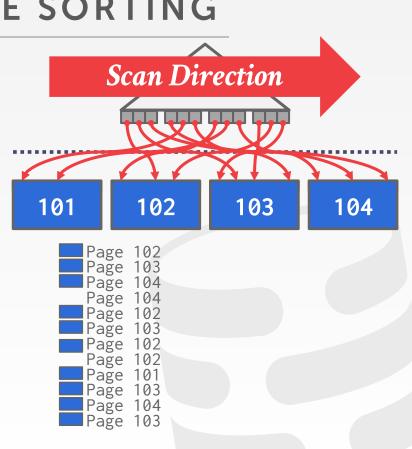
Retrieving tuples in the order they appear in a non-clustered index can be very inefficient.

The DBMS can first figure out all the tuples that it needs and then sort them based on their Page ID.



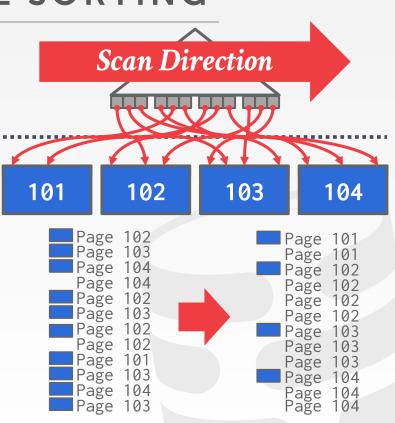
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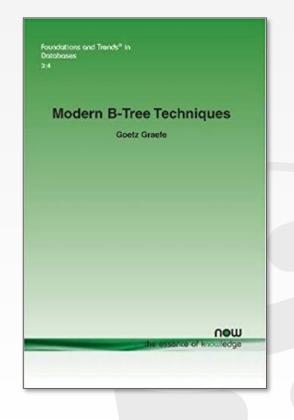
Retrieving tuples in the order they appear in a non-clustered index can be very inefficient.





## **B+TREE DESIGN CHOICES**

Node Size Merge Threshold Variable-Length Keys Intra-Node Search





## NODE SIZE

The slower the storage device, the larger the optimal node size for a B+Tree.

- $\rightarrow$  HDD: ~1MB
- $\rightarrow$  SSD: ~10KB
- $\rightarrow$  In-Memory: ~512B

Optimal sizes can vary depending on the workload  $\rightarrow$  Leaf Node Scans vs. Root-to-Leaf Traversals



### MERGE THRESHOLD

Some DBMSs do not always merge nodes when they are half full.

Delaying a merge operation may reduce the amount of reorganization.

It may also be better to just let smaller nodes exist and then periodically rebuild entire tree.



## VARIABLE-LENGTH KEYS

#### Approach #1: Pointers

 $\rightarrow$  Store the keys as pointers to the tuple's attribute.

#### Approach #2: Variable-Length Nodes

- $\rightarrow$  The size of each node in the index can vary.
- $\rightarrow$  Requires careful memory management.

#### Approach #3: Padding

 $\rightarrow$  Always pad the key to be max length of the key type.

#### Approach #4: Key Map / Indirection

 $\rightarrow$  Embed an array of pointers that map to the key + value list within the node.

## **INTRA-NODE SEARCH**

#### Approach #1: Linear

 $\rightarrow$  Scan node keys from beginning to end.

#### Approach #2: Binary

→ Jump to middle key, pivot left/right depending on comparison.

#### **Approach #3: Interpolation**

→ Approximate location of desired key based on known distribution of keys.



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Find Key=8



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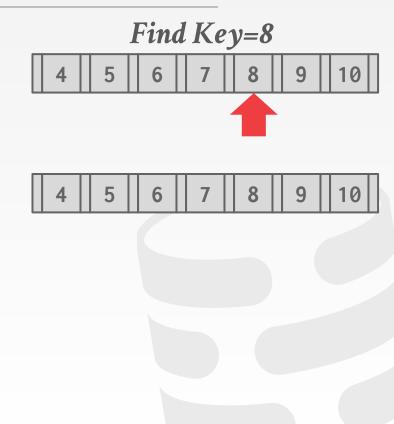
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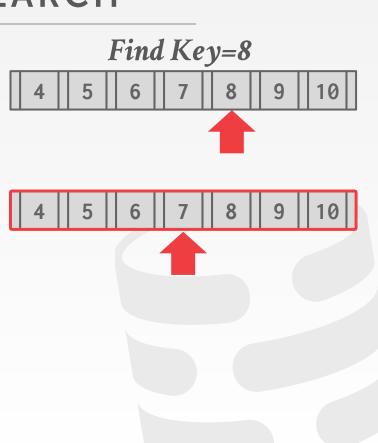
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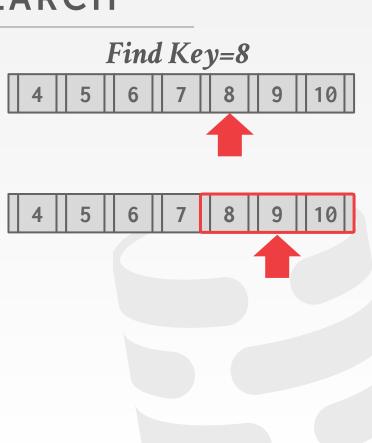
### Approach #1: Linear

 $\rightarrow$  Scan node keys from beginning to end.

### Approach #2: Binary

→ Jump to middle key, pivot left/right depending on comparison.

### **Approach #3: Interpolation**





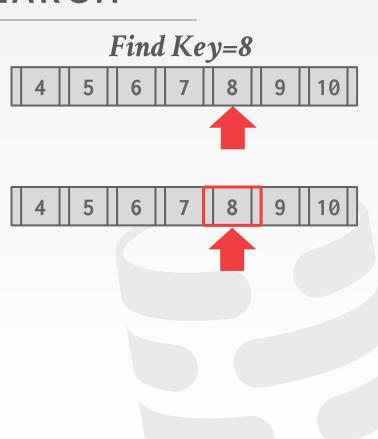
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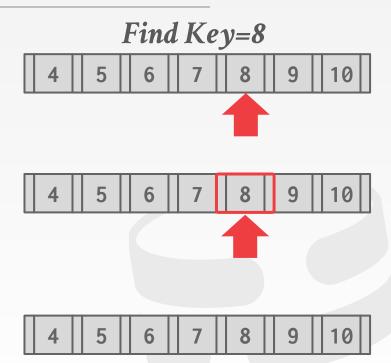
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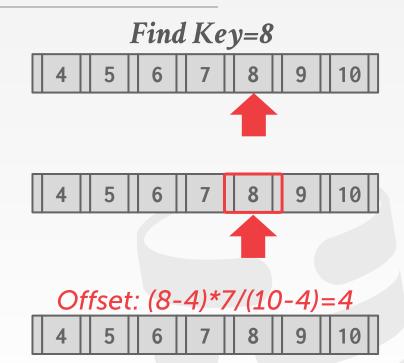
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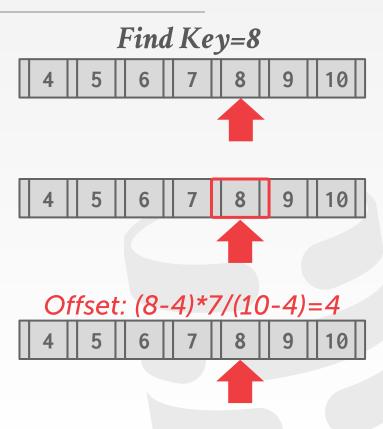
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### **OPTIMIZATIONS**

Prefix Compression Deduplication Bulk Insert Many more...



## PREFIX COMPRESSION

Sorted keys in the same leaf node are likely to have the same prefix.

Instead of storing the entire key each time, extract common prefix and store only unique suffix for each key.  $\rightarrow$  Many variations.

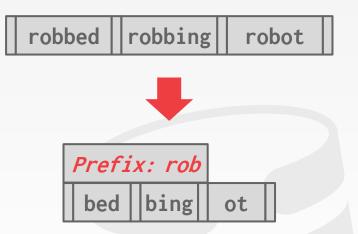




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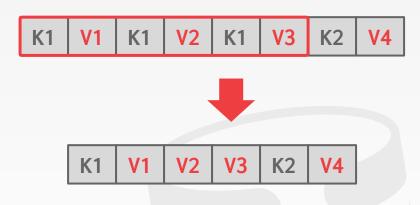




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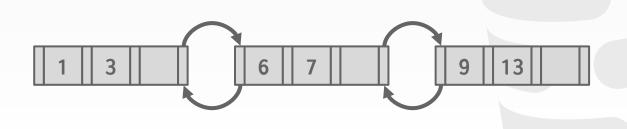
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Keys: 3, 7, 9, 13, 6, 1 Sorted Keys: 1, 3, 6, 7, 9, 13



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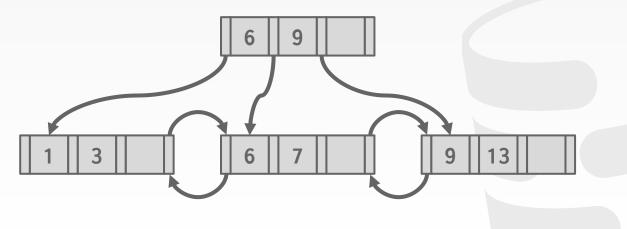
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## CONCLUSION

The venerable B+Tree is (almost) always a good choice for your DBMS.



### NEXT CLASS

#### Index Concurrency Control



