

Chapter 11: Indexing and Hashing

Database System Concepts, 6th Ed.

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Basic Concepts

- Indexing mechanisms used to speed up access to desired data
 - e.g., author catalog in library, term index at the end of a book
- **Search key** attribute to set of attributes used to look up records in a file
- An index file consists of records (called index entries) of the form

pointer

- Index files are typically much smaller than the original file
- Two basic kinds of indices:
 - Ordered indices: search keys are stored in sorted order
 - Hash indices: search keys are distributed uniformly across "buckets" using a "hash function"



Index Evaluation Metrics

- Access types supported efficiently
 - Point query: records with a specified value in the attribute
 - Range query: records with an attribute value falling in a specified range of values
- Access time
- Insertion time
- Deletion time
- Space overhead



Ordered Indices

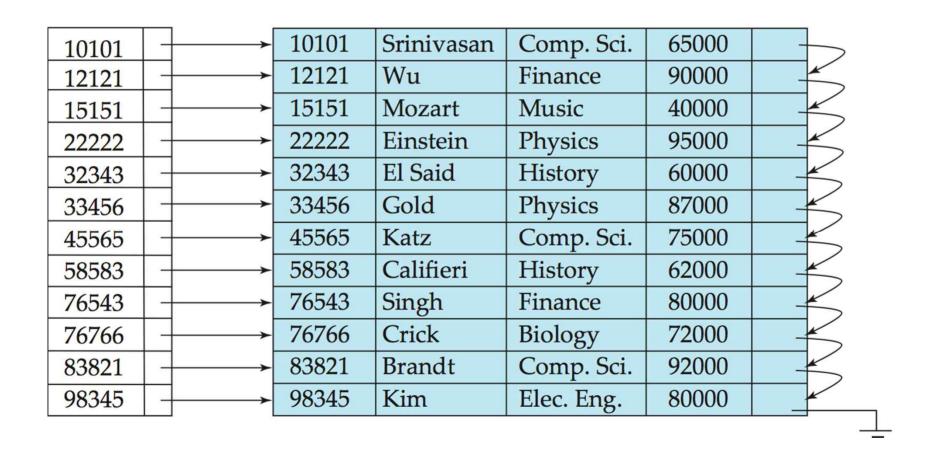
- Ordered index: index entries are stored sorted on the search key value
 - e.g., author catalog in library
- Primary index: an index whose search key specifies the sequential order of the file
 - Also called clustering index
 - The search key of a primary index is usually (but not necessarily) the primary key
- Secondary index: an index whose search key specifies an order different from the sequential order of the file
 - Also called non-clustering index
- Index-sequential file: ordered sequential file with a primary index



Dense Index Files

Dense index

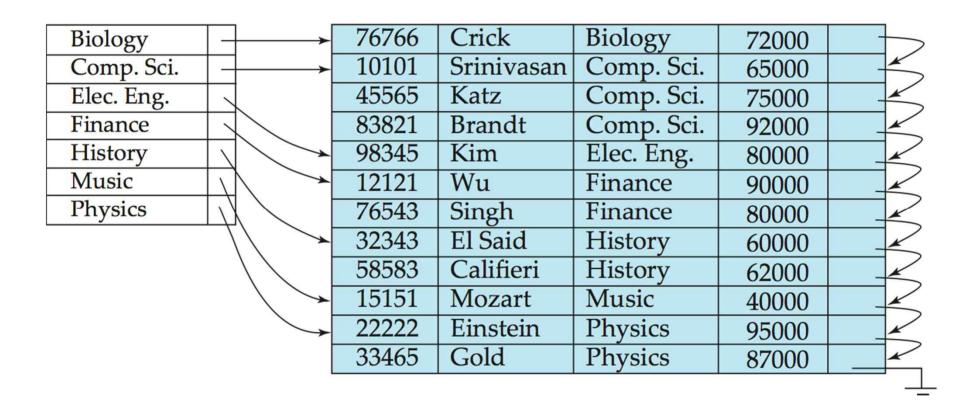
Index record appears for every search-key value in the file





Primary Indices Example

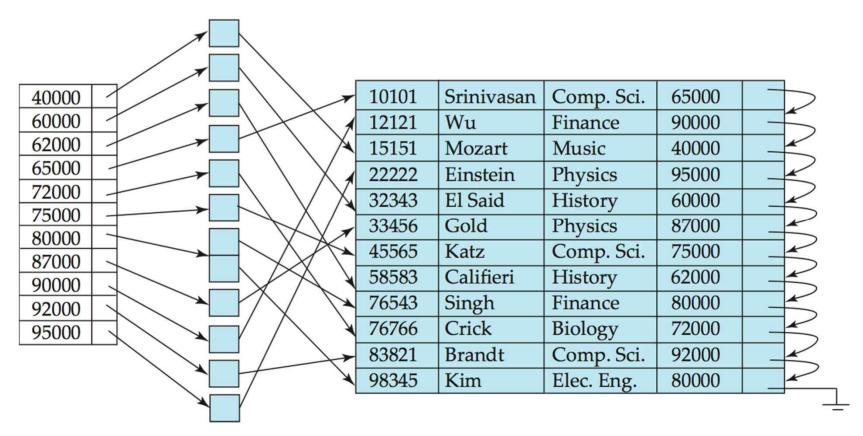
Example: dense index on dept_name, with instructor file sorted on dept_name





Secondary Indices Example

- Index record points to a bucket that contains pointers to all the actual records with that particular search-key value.
- Secondary indices have to be dense

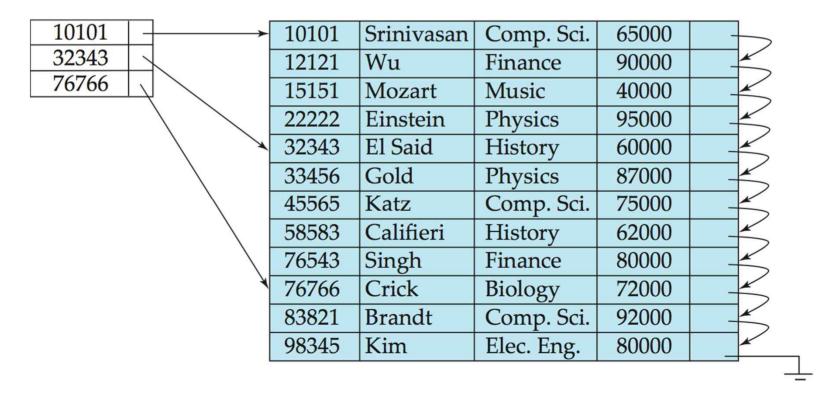


Secondary index on salary field of instructor



Sparse Index Files

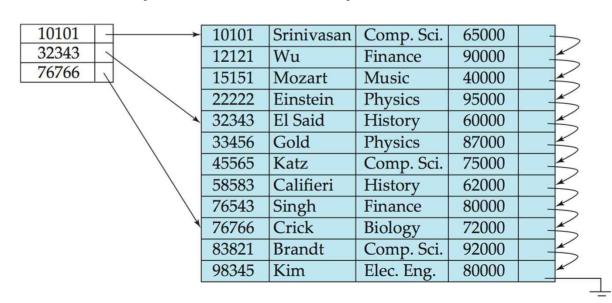
- Sparse Index: contains index records for only some search-key values.
 - Applicable when records are sequentially ordered on search-key
- Compared to dense indices:
 - Less space and less maintenance overhead for insertions and deletions
 - Generally slower than dense index for locating records





Index Update: Deletion

- If deleted record was the only record in the file with its particular search-key value, the search-key is deleted from the index also.
- **Dense indices** deletion of search-key is similar to file record deletion
- Sparse indices
 - If an entry for the search key exists in the index,
 - Replace the entry in the index with the next search-key value in the file (in search-key order)
 - If the next search-key value already has an index entry
 - Delete the entry





Index Update: Insertion

Perform a lookup using the search-key value appearing in the record to be inserted

Dense indices

If the search-key value does not appear in the index, insert it.

Sparse indices

- If index stores an entry for each block of the file
 - No change needs to be made to the index, unless a new block is created
 - If a new block is created, the first search-key value appearing in the new block is inserted into the index



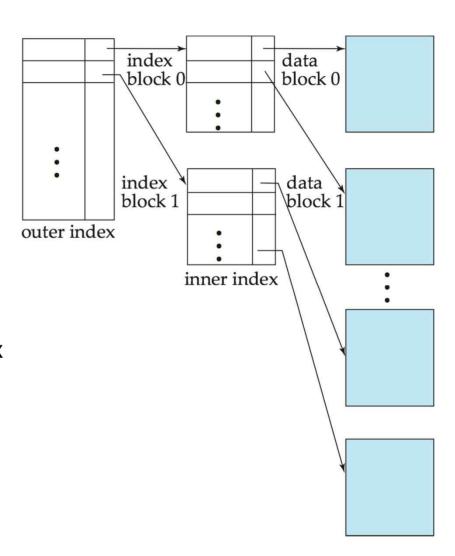
Primary and Secondary Indices

- Indices offer substantial benefits when searching for records
- When a file is modified, every index on the file must be updated
 - Updating indices imposes overhead on database modification
- Sequential scan using primary index is efficient, but a sequential scan using a secondary index is expensive
 - Each record access may fetch a new block from disk
 - Block fetch requires about 5 to 10 milliseconds, versus about 100 nanoseconds for memory access



Multilevel Index

- If primary index does not fit in memory, access becomes expensive
- Solution: treat primary index kept on disk as a sequential file and construct a sparse index on it
 - outer index a sparse index of primary index
 - inner index the primary index file
- If even outer index is too large to fit in main memory, yet another level of index can be created, and so on
- Indices at all levels must be updated on insertion or deletion from the file



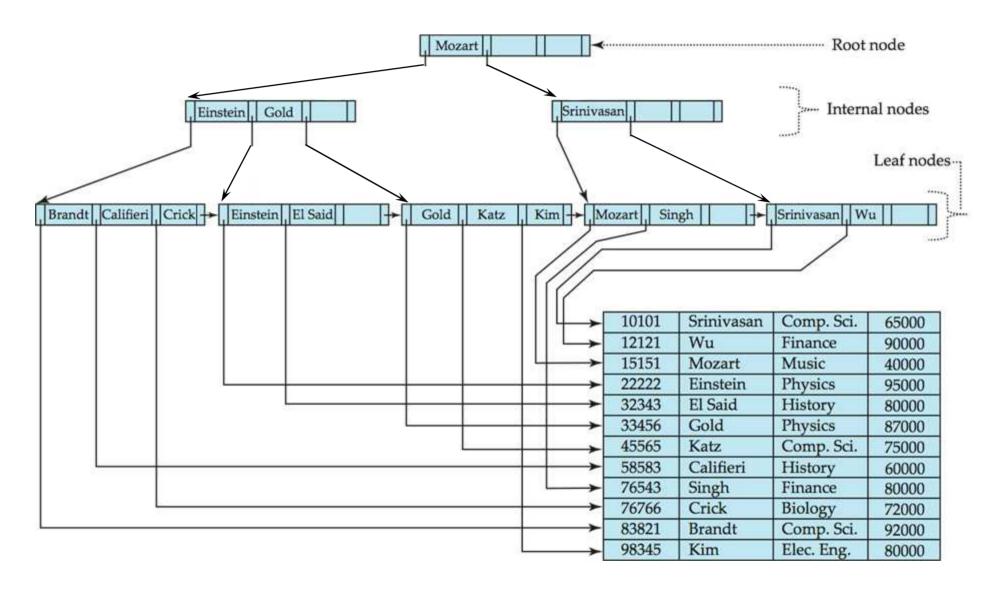


B⁺-Tree Index Files

- Advantage of B+-tree index files:
 - Automatically reorganizes itself with small local changes, in the face of insertions and deletions
 - Reorganization of entire file is not required to maintain performance
- (Minor) disadvantage of B+-trees:
 - Extra insertion and deletion overhead, space overhead
- B+-trees are used extensively
 - Advantages of B⁺-trees outweigh disadvantages



Example of B*-Tree





B⁺-Tree Index Files (Cont.)

A B⁺-tree is a rooted tree satisfying the following properties:

- All paths from root to leaf are of the same length
- Each node that is not a root or a leaf has between $\lceil n/2 \rceil$ and n children
- A leaf node has between $\lceil (n-1)/2 \rceil$ and n-1 values
- Special cases: root node
 - If the root is not a leaf, it has at least 2 children
 - If the root is a leaf (that is, there are no other nodes in the tree), it can have between 0 and (n-1) values



B⁺-Tree Node Structure

Typical node (non-leaf node)



- K_i are the search-key values
- P_i are pointers to children (for non-leaf nodes) or pointers to records (for leaf nodes)
- The search-keys in a node are ordered

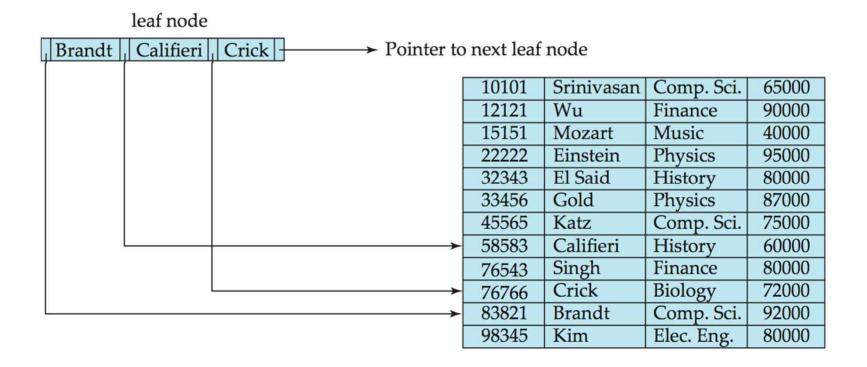
$$K_1 < K_2 < K_3 < \ldots < K_{n-1}$$

We assume no duplicate keys



Leaf Nodes in B*-Trees

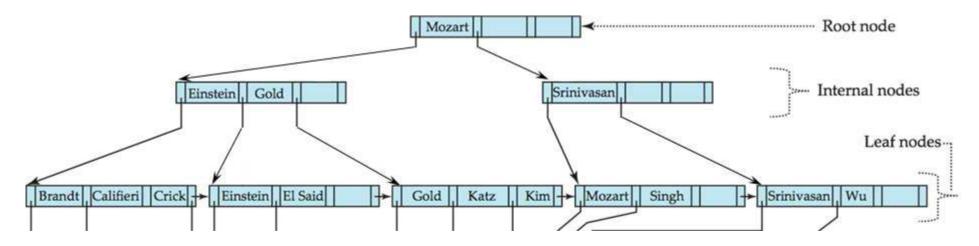
- For i = 1, 2, ..., n-1, pointer P_i points to a file record with search-key value K_i
- \blacksquare P_n points to next leaf node in search-key order
- If L_i , L_j are leaf nodes and i < j, L_i 's search-key values are less than or equal to L_i 's search-key values





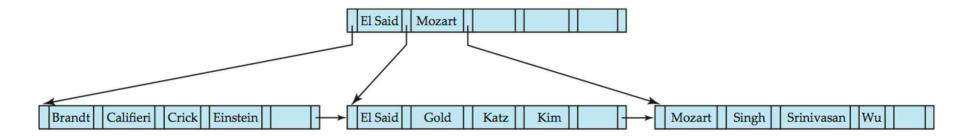
Non-Leaf Nodes in B⁺-Trees

- Non leaf nodes form a multi-level sparse index on the leaf nodes
- For a non-leaf node with m pointers $(m \le n)$:
 - All the search-keys in the subtree to which P₁ points are less than K₁
 - For $2 \le i \le m-1$, all the search-keys in the subtree to which P_i points have values greater than or equal to K_{i-1} and less than K_i
 - All the search-keys in the subtree to which P_m points have values greater than or equal to K_{m-1}





Example of B*-tree



B⁺-tree for *instructor* file (n = 6)

- Leaf nodes must have between 3 and 5 values
 - $\lceil (n-1)/2 \rceil$ and n-1, with n=6
- Non-leaf nodes other than root must have between 3 and 6 children
 - $\lceil (n/2) \rceil$ and n, with n = 6
- Root must have at least 2 children



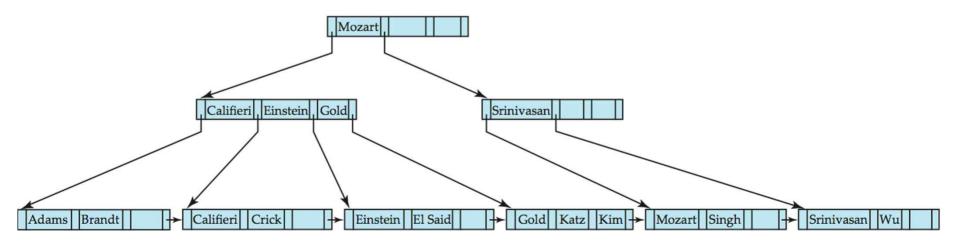
Observations about B+-trees

- Since the inter-node connections are done by pointers, "logically" close blocks need not be "physically" close
- The non-leaf levels of the B+-tree form a hierarchy of sparse indices
- The B⁺-tree contains a relatively small number of levels
 - ▶ Level below root has at least 2 * \[n/2 \] values
 - ▶ Next level has at least 2 * \[n/2 \] * \[n/2 \] values
 - .. etc.
 - If there are K search-key values in the file, the tree height is no more than $\lceil \log_{\lceil n/2 \rceil}(K) \rceil$
 - Thus searches can be conducted efficiently
- Insertions and deletions to the main file can be handled efficiently, as the index can be restructured in logarithmic time (as we shall see).



Queries on B*-Trees

- Find record with search-key value V
 - 1. C = root
 - 2. While C is not a leaf node {
 - 1. Let *i* be least value s.t. $V \le K_i$.
 - 2. If no such exists, set C = last non-null pointer in C
 - 3. Else { if $(V = K_i)$ set $C = P_{i+1}$ else set $C = P_i$ }
 - 3. Let *i* be least value s.t. $K_i = V$
 - 4. If there is such a value i, follow pointer P_i to the desired record
 - 5. Else no record with search-key value *k* exists





Queries on B⁺-Trees (Cont.)

- If there are K search-key values in the file, the height of the tree is no more than $\lceil \log_{\lceil n/2 \rceil}(K) \rceil$
- A node is generally the same size as a disk block, typically 4 kilobytes
 - n is typically around 100 (40 bytes per index entry)
- With 1 million search key values and n = 100
 - At most $log_{50}(1,000,000) = 4$ nodes are accessed in a lookup
- Contrast this with a balanced binary tree with 1 million search key values
 - Around 20 nodes are accessed in a lookup
 - Difference is significant since every node access may need a disk I/O, costing around 20 milliseconds



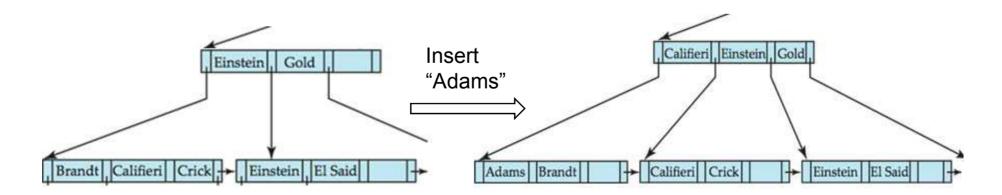
Updates on B*-Trees: Insertion

- 1. Find the leaf node in which the search-key value would appear
- 2. Add the record to the file
- 3. If there is room in the leaf node, insert (key-value, pointer) pair in the leaf node
- 4. Otherwise, split the node (along with the new (key-value, pointer) entry)



Leaf Node Split in B⁺-Trees

- Splitting a leaf node:
 - Take the n (search-key value, pointer) pairs (including the one being inserted) in sorted order
 - ▶ Place the first $\lceil n/2 \rceil$ in the original node, and the rest in a new node
 - Make the original node point to the new node
 - Let the new node be p, and let k be the least key value in p
 - ▶ Insert (*k*, *p*) in the parent of the node being split
 - If the parent is full, split it and **propagate** the split further up
- Splitting of nodes proceeds upwards till a node that is not full is found



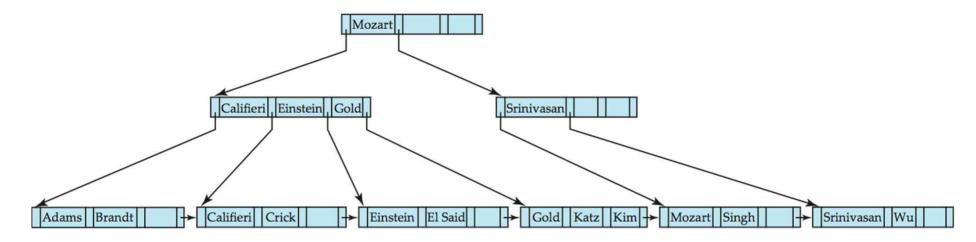


Non-Leaf Node Split in B⁺-Trees

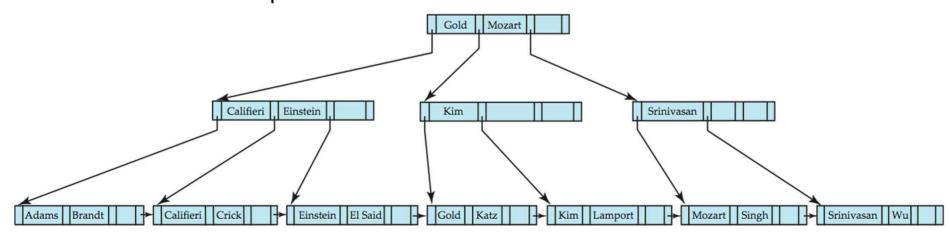
- Splitting a non-leaf node: when inserting (k,p) into an already full internal node N
 - Copy N to an in-memory area M that can hold N and (k,p)
 - Insert (k,p) into M
 - Copy P₁,K₁, ..., K_{⌈n/2⌉-1},P_{⌈n/2⌉} from M back into node N
 - Copy P_{[n/2]+1},K_{[n/2]+1},...,K_{n-1},P_n from M into newly allocated node N'
 - Insert (K_{∫n/2}¬,N') into parent N



B⁺-Tree Non-Leaf Node Split Example



After insertion of "Lamport"





Updates on B*-Trees: Deletion

- 1. Find the record to be deleted, and remove it from the main file
- 2. Remove (search-key value, pointer) from the leaf node
- If the node has too few entries due to the removal, merge siblings or redistributed pointers (next slide)
- 4. The node deletions may cascade upwards till a node which has $\lceil n/2 \rceil$ or more pointers is found
- If the root node has only one pointer after deletion, it is deleted and the sole child becomes the root

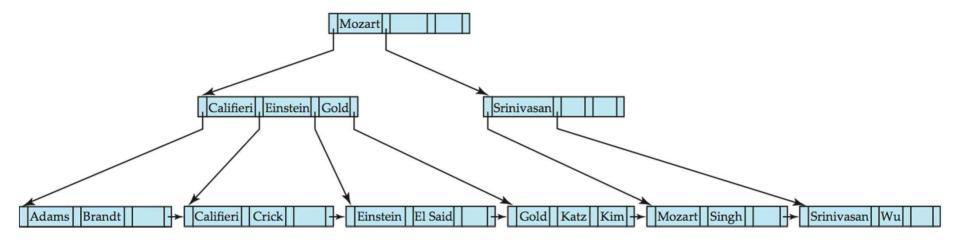


Updates on B*-Trees: Deletion (Cont.)

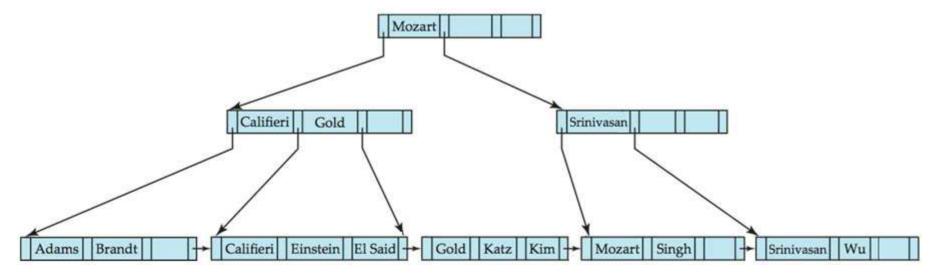
- 3. If the node has too few entries due to the removal,
 - 1. The entries in the node and a sibling fit into a single node *merge siblings*:
 - Insert all the search-key values in the two nodes into a single node (the one on the left), and delete the other node
 - ▶ Delete the pair (K_{i-1}, P_i) , where P_i is the pointer to the deleted node, from its parent, recursively using the above procedure
 - The entries in the node and a sibling do not fit into a single node redistribute pointers:
 - Redistribute the pointers between the node and a sibling such that both have more than the minimum number of entries
 - Update the corresponding search-key value in the parent of the node
- Order to check an adjacent sibling
 - Right sibling first
 - If not possible, then, left sibling



Example: B*-Tree Deletion – Merge Siblings



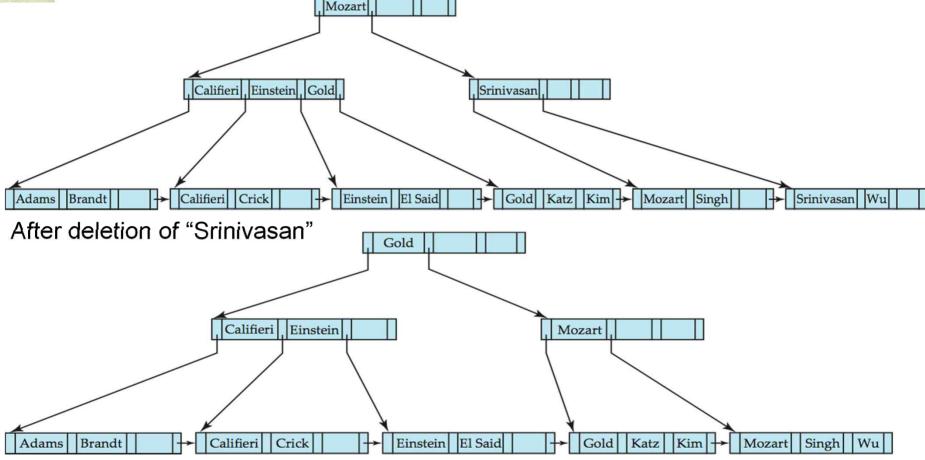
After deletion of "Crick"



Deleting "Crick" causes merging of under-full leaves



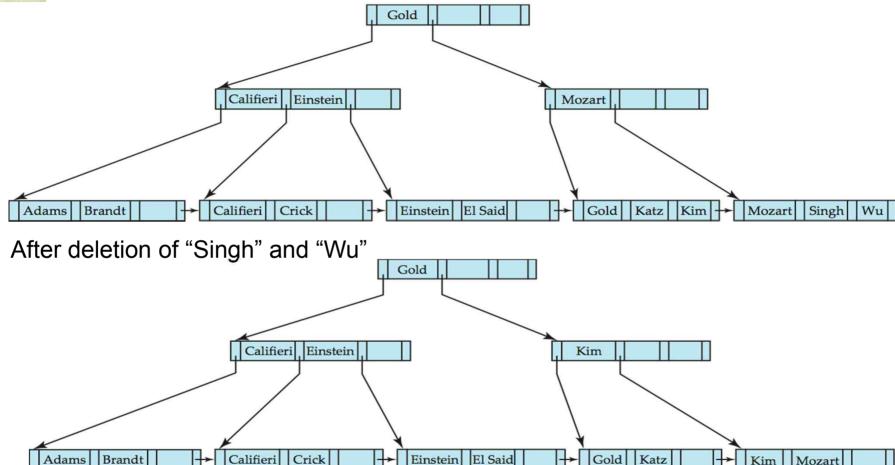
Example: B*-Tree Deletion – Merge Siblings (*)



- Deleting "Srinivasan" causes merging of under-full leaves
- Parent node becomes underfull, and is merged with its sibling
 - Value separating two nodes (at the parent) is pulled down when merging
- Root node then has only one child, and is deleted, but re-created after merging



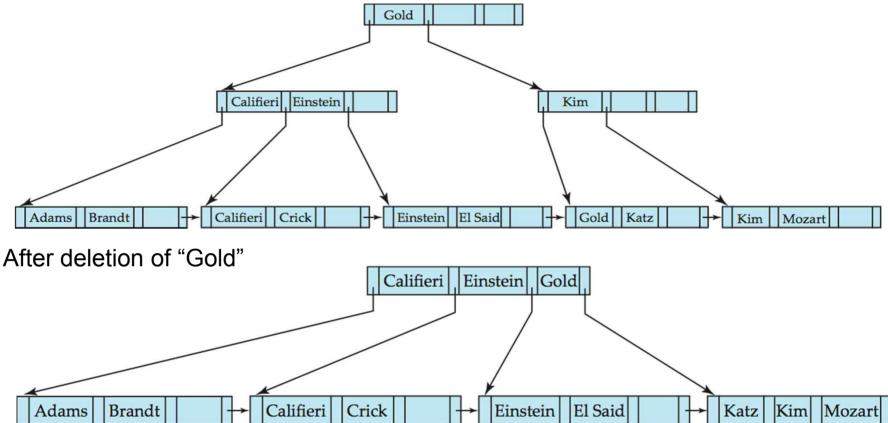
Example: B*-tree Deletion – Redistribute Pointers



- Leaf containing Singh and Wu became underfull, and borrowed a value Kim from its left sibling
- Search-key value in the parent changes as a result



Example: B*-tree Deletion – Root Node Deletion



- Node with Gold and Katz became underfull, and was merged with its sibling
- Parent node becomes underfull, and is merged with its sibling
 - Value separating two nodes (at the parent) is pulled down when merging
- Root node then has only one child, and is deleted



Handling Non-Unique Search Keys

- Buckets on separate block (bad idea)
- List of tuple pointers with each key
 - Extra code to handle long lists
 - Deletion of a tuple can be expensive if there are many duplicates on search key
 - Low space overhead, no extra cost for queries
- Make search key unique by adding a record-identifier
 - Extra storage overhead for keys
 - Simpler code for insertion/deletion
 - Widely used



Bulk Loading and Bottom-Up Build

- Inserting entries one-at-a-time into a B⁺-tree requires ≥ 1 IO per entry
 - assuming leaf level does not fit in memory
 - can be very inefficient for loading a large number of entries at a time (bulk loading)
- Efficient alternative 1:
 - Sort entries first (using efficient external-memory sort algorithms discussed later in Section 12.4)
 - Insert in sorted order
 - insertion will go to existing page (or cause a split)
 - much improved IO performance, but most leaf nodes half full
- Efficient alternative 2: Bottom-up B+-tree construction
 - As before sort entries
 - And then create tree layer-by-layer, starting with leaf level
 - details as an exercise
 - Implemented as part of bulk-load utility by most database systems



Multiple-Key Access

- Use multiple indices for certain types of queries.
- Example:

```
select ID
```

from instructor

where dept_name = "Finance" and salary = 80000

- Possible strategies for processing query using indices on single attributes:
 - 1. Use index on *dept_name* to find instructors with department name Finance; test *salary* = 80000
 - 2. Use index on *salary* to find instructors with a salary of \$80000; test *dept_name* = "Finance".
 - 3. Use *dept_name* index to find pointers to all records pertaining to the "Finance" department. Similarly use index on *salary*. Take intersection of both sets of pointers obtained.



Indices on Multiple Keys

- Composite search keys are search keys containing more than one attribute
 - E.g. (dept_name, salary)
- Lexicographic ordering: $(a_1, a_2) < (b_1, b_2)$ if either
 - $a_1 < b_1$, or
 - $a_1 = b_1$ and $a_2 < b_2$
- Suppose we have an index on combined search-key (dept_name, salary).
- Can efficiently handle where dept_name = "Finance" and salary = 80000
 - Fetch only records that satisfy both conditions
- Can also efficiently handle where dept_name = "Finance" and salary < 80000</p>
- But cannot efficiently handle where dept_name < "Finance" and balance = 80000</p>
 - May fetch many records that satisfy the first but not the second condition



Static Hashing

Bucket

- A unit of storage containing one or more records
- Typically a disk block
- Hash file organization
 - The bucket of a record is directly obtained from its search-key value using a hash function
- Hash function h
 - A function from the set of all search-key values K to the set of all bucket addresses B
 - is used to locate records for access, insertion as well as deletion.
- Records with different search-key values may be mapped to the same bucket
 - Thus entire bucket has to be searched sequentially to locate a record



Example of Hash File Organization

Hash file organization of *instructor* file, using *dept_name* as key

- # of buckets = 10
- The binary representation of the *i*th character is assumed to be the integer *i*
- The hash function returns the sum of the binary representations of the characters modulo 10
 - E.g. h(Music) = 1
 h(History) = 2
 h(Physics) = 3
 h(Elec. Eng.) = 3

		иері_па	iiie as				
bucket	0			bucket	4		
				12121	Wu	Finance	90000
				76543	Singh	Finance	80000
bucket	1	dan all		bucket	5		
15151	Mozart	Music	40000	76766	Crick	Biology	72000
				-			
bucket	2			bucket	6		
32343	El Said	History	80000	10101	Srinivasan	Comp. Sci.	65000
58583	Califieri	History	60000	45565	Katz	Comp. Sci.	75000
				83821	Brandt	Comp. Sci.	_
bucket	- 2			bucket	7	48	
ALIENS AND ADDRESS OF	Einstein	Physics	95000	Ducket	,		
		-	Section 100 contracts				
33456		Physics	87000				
98345	Kim	Elec. Eng.	80000				



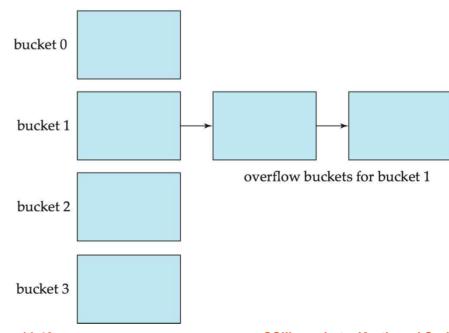
Hash Functions

- Worst hash function
 - All search-key values are mapped to the same bucket
 - Access time is proportional to the number of search-key values in the file
- Ideal hash function
 - Uniform: each bucket is assigned the same number of search-key values from the set of all possible values
 - Random: each bucket will have the same number of records assigned to it irrespective of the *actual distribution* of search-key values in the file
- Typical hash functions perform computation on the internal binary representation of the search-key



Handling of Bucket Overflows

- Bucket overflow can occur because of
 - Insufficient buckets
 - Skew in distribution of records. This can occur due to two reasons:
 - multiple records have same search-key value
 - chosen hash function produces non-uniform distribution of key values
- Although the probability of bucket overflow can be reduced, it cannot be eliminated; it is handled by using overflow buckets.
- Overflow chaining
 - The overflow buckets of a given bucket are chained together in a linked list



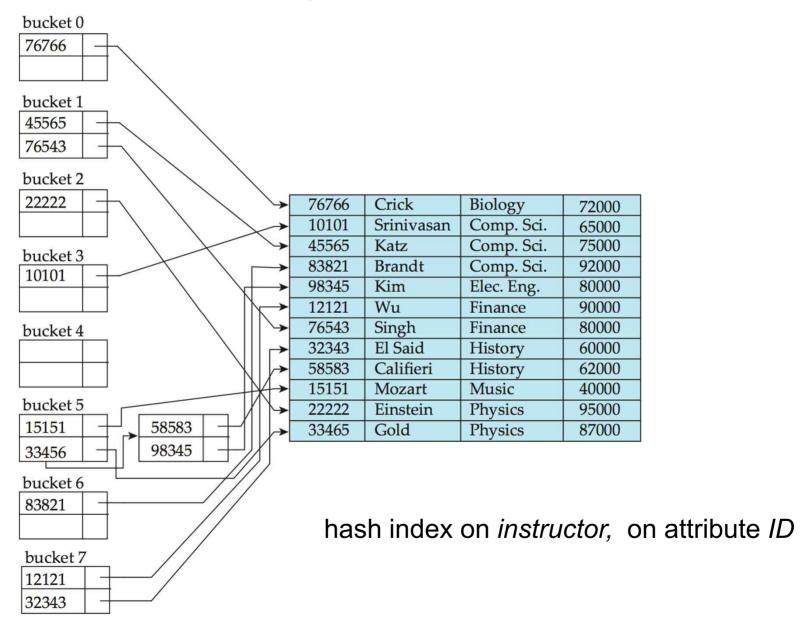


Hash Indices

- A hash index organizes the search keys, with their associated record pointers, into a hash file structure.
- Strictly speaking, hash indices are always secondary indices
 - If the file itself is organized using hashing, a separate primary hash index on it using the same search-key is unnecessary
 - However, we use the term hash index to refer to both secondary index structures and hash organized files



Example of Hash Index





Deficiencies of Static Hashing

- In static hashing, function h maps search-key values to a fixed set of B of bucket addresses
 - Databases grow or shrink with time
 - If initial number of buckets is too small, and file grows, performance will degrade due to too much overflows
 - If space is allocated for anticipated growth, a significant amount of space will be wasted initially (and buckets will be underfull)
 - If database shrinks, again space will be wasted
- One solution: periodic re-organization of the file with a new hash function
 - Expensive, disrupts normal operations
- Better solution: allow the number of buckets to be modified dynamically
 - Dynamic hashing is not covered in this class



Comparison of Ordered Indexing and Hashing

- Cost of periodic re-organization
- Relative frequency of insertions and deletions
- Is it desirable to optimize average access time at the expense of worst-case access time?
- Expected type of queries:
 - Hashing is generally better at retrieving records having a specified value of the key (point query)
 - If range queries are common, ordered indices are to be preferred

In practice:

- PostgreSQL supports hash indices, but discourages use due to poor performance
- Oracle supports static hash organization, but not hash indices
- SQLServer supports only B⁺-trees



Bitmap Indices

- Bitmap: simply an array of bits
- Bitmap index: a specialized type of index designed for efficient querying on multiple keys
- In the simplest form, a bitmap index on an attribute has a bitmap for each value of the attribute
 - Bitmap has as many bits as records
 - In a bitmap for value v, the bit for a record is 1 if the record has the value v for the attribute, and is 0 otherwise

record number	ID	gender	income_level
0	76766	m	L1
1	22222	f	L2
2	12121	f	L1
3	15151	m	L4
4	58583	f	L3

Bitmap	s for gender		itmaps for		
m	10010	17	income_level		
f	01101	L1	10100		
		L2	01000		
		L3	00001		
		L4	00010		
		L5	00000		



Bitmap Indices (Cont.)

- Applicable on attributes that take on a relatively small number of distinct values
 - E.g. gender, country, state, ...
 - E.g. income-level (income broken up into a small number of levels such as 0-9999, 10000-19999, 20000-50000, 50000- infinity)
- Bitmap indices generally very small compared with relation size
 - E.g. if record is 100 bytes, space for a single bitmap is 1/800 of space used by relation
 - If # of distinct attribute values is 8, bitmap is only 1% of relation size



Bitmap Indices (Cont.)

- Bitmap indices are useful for queries on multiple attributes
 - not particularly useful for single attribute queries
- Queries are answered using bitmap operations
 - Intersection (and): e.g., 100110 AND 110011 = 100010
 - Union (or): e.g., 100110 OR 110011 = 110111
 - Complementation (not): e.g., 100110 NOT 100110 = 011001
- Each operation takes two bitmaps of the same size and applies the operation on corresponding bits to get the result bitmap
 - E.g. males with income level L1: 10010 AND 10100 = 10000
 - Can then retrieve required tuples
 - Counting number of matching tuples is even faster



Index Definition in SQL

Create an index

E.g.: **create index** *dept_index* **on** *instructor* (*dept_name*)

- Use create unique index to indirectly specify and enforce the condition that the search key is a candidate key
 - Not really required if SQL unique integrity constraint is supported
- To drop an index

drop index <index-name>

Most database systems allow specification of type of index, and clustering



End of Chapter 11

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