Week 9

Database Management Systems

Concepts of Indexing

Consider a table: Faculty(Name, Phone)

Index on "Nam	ne"		Table "Faculty"	Index on "Phone"			
Name	Pointer	Rec#	Name	Phone	Pointer	Phone	
Anupam Basu	2	1Pa	rtha Pratim Das	81998	6	81664	
Pabitra Mitra	6	2An	upam Basu	82404	1	81998	
Partha Pratim Das	1	3Ra	njan Sen	84624	2	82404	
Prabir Kumar Biswas	7	4Su	deshna Sarkar	82432	4	82432	
Rajib Mall	5	5Ra	jib Mall	83668	5	83668	
Ranjan Sen	3	6Pa	bitra Mitra	81664	3	84624	
Sudeshna Sarkar	4	7Pra	abir Kumar Biswas	84772	7	84772	

- How to search on Name?
 - Get the phone number for 'Pabitra Mitra'
 - Use "Name" Index sorted on 'Name', search 'Pabitra Mitra' and navigate on pointer (rec #)
- How to search on Phone?
 - Get the name of the faculty having phone number = 84772
 - Use "Phone" Index sorted on 'Phone', search '84772' and navigate on pointer (rec #)
- We can keep the records sorted on 'Name' or on 'Phone' (called the primary index), but not on both

- Indexing mechanisms used to speed up access to desired data.
 - o For example:
 - Name in a faculty table
 - □ author catalog in library
- 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

search-key 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 time
- Insertion time
- Deletion time
- Space overhead

Ordered Indices

- In an ordered index, index entries are stored sorted on the search key value. For example, author catalog in library
- Primary index: in a sequentially ordered file, the 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 Indexing

- Dense index Index record appears for every search-key value in the file.
- For example, index on ID attribute of instructor relation

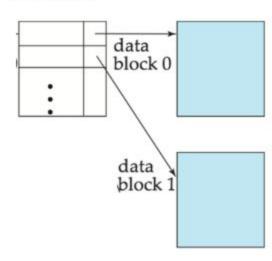
10101		10101	Srinivasan	Comp. Sci.	65000	+
12121	→	12121	Wu	Finance	90000	-
15151	→	15151	Mozart	Music	40000	-
22222 +		22222	Einstein	Physics	95000	-
32343		32343	El Said	History	60000	-
33456	→	33456	Gold	Physics	87000	-
45565	→	45565	Katz	Comp. Sci.	75000	-
58583	→	58583	Califieri	History	62000	-
76543 —	→	76543	Singh	Finance	80000	-
76766 -	→	76766	Crick	Biology	72000	-
83821 -		83821	Brandt	Comp. Sci.	92000	-
98345	→	98345	Kim	Elec. Eng.	80000	

Sparse Indexing

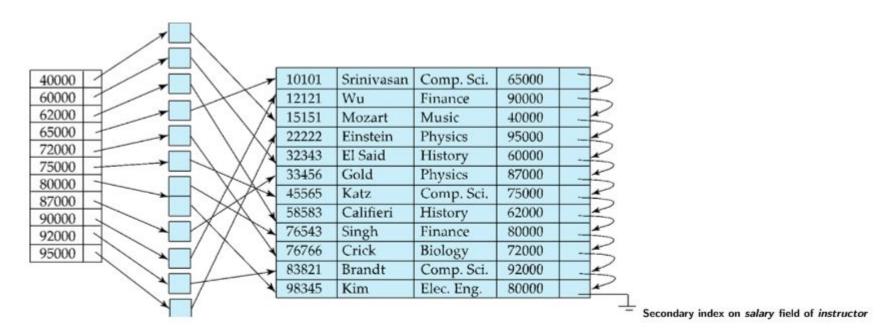
- Sparse Index: contains index records for only some search-key values.
 - Applicable when records are sequentially ordered on search-key
- To locate a record with search-key value K we:
 - Find index record with largest search-key value < K
 - Search file sequentially starting at the record to which the index record points

10101	10101	Srinivasan	Comp. Sci.	65000	-
32343	12121	Wu	Finance	90000	*
76766	15151	Mozart	Music	40000	-
	22222	Einstein	Physics	95000	-
\ \	32343	El Said	History	60000	- 4
	33456	Gold	Physics	87000	-
	45565	Katz	Comp. Sci.	75000	-
	58583	Califieri	History	62000	- 4
\	76543	Singh	Finance	80000	- 1
*	76766	Crick	Biology	72000	- 4
	83821	Brandt	Comp. Sci.	92000	- 4
	98345	Kim	Elec. Eng.	80000	×

- Compared to dense indices:
 - Less space and less maintenance overhead for insertions and deletions
 - Generally slower than dense index for locating records
- Good tradeoff: sparse index with an index entry for every block in file, corresponding to least search-key value in the block



Secondary Indices



- 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

Multilevel Indexing

- 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
 - o 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

Index Update: Deletion

 If deleted record was the only record in the file with its particular search-key value, the searchkey is deleted from the index also.

10101	10101	Srinivasan	Comp. Sci.	65000	-
32343	12121	Wu	Finance	90000	
76766	15151	Mozart	Music	40000	-
1 /	22222	Einstein	Physics	95000	
/	32343	El Said	History	60000	
	33456	Gold	Physics	87000	
	45565	Katz	Comp. Sci.	75000	-
	58583	Califieri	History	62000	-
	76543	Singh	Finance	80000	
,	76766	Crick	Biology	72000	
	83821	Brandt	Comp. Sci.	92000	-
	98345	Kim	Elec. Eng.	80000	

Single-level index entry deletion:

- 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, it is deleted by replacing 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, the entry is deleted instead of being replaced

Index Update: Insertion

Single-level index 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

Balanced Binary Search

Trees:

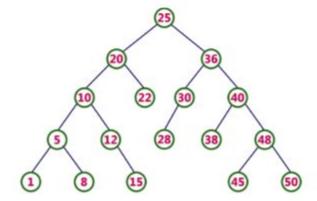
- How to search a key in a list of n data items?
 - Linear Search: O(n): Find 28 \Rightarrow 16 comparisons
 - □ Unordered items in an array search sequentially
 - ▶ Unordered / Ordered items in a list search sequentially

22	50	20	36	40	15	08	01	45	48	30	10	38	12	25	28	05	END
				1.7	5.		7.5				1000				7.7		

- Binary Search: $O(\lg n)$: Find $28 \Rightarrow 4$ comparisons -25, 36, 30, 28
 - ▷ Ordered items in an array search by divide-and-conquer

01 (05	80	10	12	15	20	22	25	28	30	36	38	40	45	48	50	END
------	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	-----

Binary Search Tree − recursively on left / right



• Worst case time (n data items in the data structure):

Data Structure	Search	Insert	Delete	Remarks			
Unordered Array	O(n)	O(1)	O(1)	The time to Insert /			
Ordered Array	O(log n)	O(n)	O(n)	Delete an item is the time after the location			
Unordered List	O(n)	O(1)	O(1)	of the item has been			
Ordered List	O(n)	O(1)	O(1)	ascertained by			
Binary Search Tree	O(h)	O(1)	O(1)	Search.			

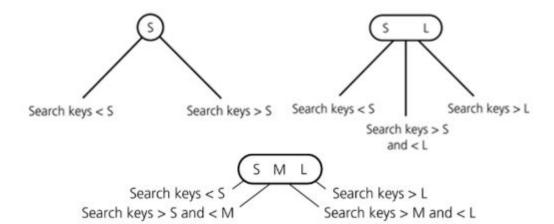
- Between an array and a list, there is a trade-off between search and insert/delete complexity
- For a BST of n nodes, $\lg n \le h < n$, where h is the height of the tree
- A BST is balanced if $h \sim O(\lg n)$: this what we desire

- In the worst case, searching a key in a BST is O(h), where h is the height of the key
- Bad Tree: $h \sim O(n)$
 - The BST is a skewed binary search tree (all the nodes except the leaf would have only one child)
 - This can happen if keys are inserted in sorted order
 - Height (h) of the BST having n elements becomes n-1
 - \circ Time complexity of search in BST becomes O(n)
- Good Tree: $h \sim O(\lg n)$
 - The BST is a balanced binary search tree
 - This is possible if
 - ▷ If keys are inserted in purely randomized order, Or
 - ▶ If the tree is explicitly balanced after every insertion
 - Height (h) of the binary search tree becomes lg n
 - \circ Time complexity of search in BST becomes $O(\lg n)$

- These data structures have optimal complexity for the required operations:
- \circ Search: $O(\lg n)$
 - o Insert: Search + O(1): $O(\lg n)$
 - Delete: Search + O(1): $O(\lg n)$
- And they are:
 - o Good for in-memory operations
 - o Work well for small volume of data
 - Has complex rotation and / or similar operations
 - Do not scale for external data structures

- All leaves are at the same depth (the bottom level).
 - Height, h, of all leaf nodes are same
 - $\triangleright h \sim O(\lg n)$
- \triangleright Complexity of search, insert and delete: $O(h) \sim O(\lg n)$
- All data is kept in sorted order
- Every node (leaf or internal) is a 2-node, 3-node or a 4-node (based on the number of links or children), and holds one, two, or three data elements, respectively
- Generalizes easily to larger nodes
- Extends to external data structures

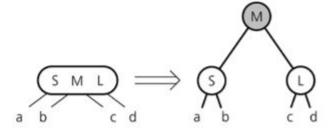
- Uses 3 kinds of nodes satisfying key relationships as shown below:
 - A 2-node must contain a single data item (S) and two links
 - o A 3-node must contain two data items (S, L) and three links
 - A 4-node must contain three data items (S, M, L) and four links
 - A leaf may contain either one, two, or three data items



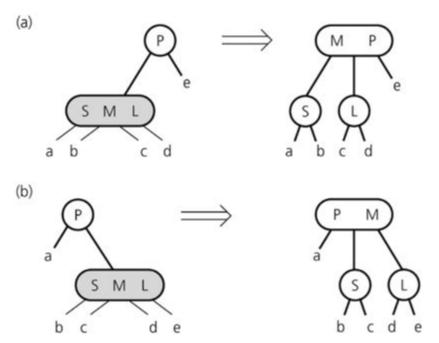
- Insert
 - Search to find expected location
 - ▶ If it is a 2 node, change to 3 node and insert
 - ▷ If it is a 3 node, change to 4 node and insert
 - ▷ If it is a 4 node, split the node by moving the middle item to parent node, then insert
 - Node Splitting
 - A 4-node is split as soon as it is encountered during a search from the root to a
 leaf
 - ▶ The 4-node that is split will
 - Be the root, or
 - Have a 2-node parent, or
 - Have a 3-node parent

Insertion in the Tree

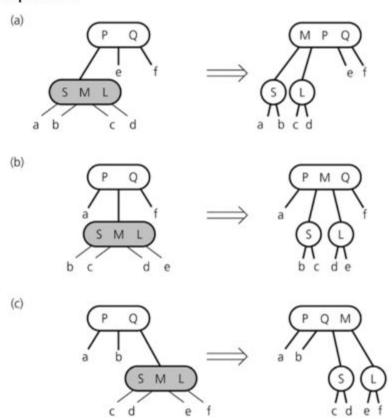
Splitting at Root



• Splitting with 2 Node parent

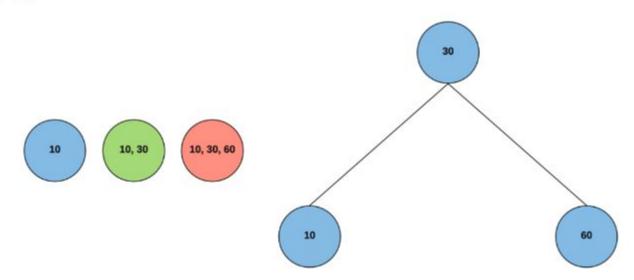


• Splitting with 3 Node parent



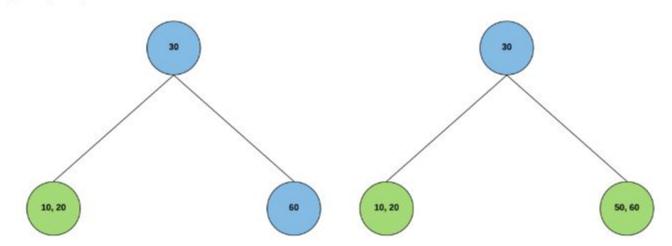
- Node Splitting: There are two strategies:
- Early: Split a 4-node as soon as you cross on in traversal. It ensures that the tree does not have a path with multiple 4-nodes at any point
 - \circ Late: Split a 4-node only when you need to insert an item in it. This might lead to cases where for one insert we may need to perform O(h) splits going till up to the root
- Both are valid and has the same complexity O(h). However, they lead to different results. Different texts and sites follow different strategies.
- Here we are following early strategy

- Insert 10, 30, 60, 20, 50, 40, 70, 80, 15, 90, 100
- 10
- 10, 30
- 10, 30, 60
- Split for 20

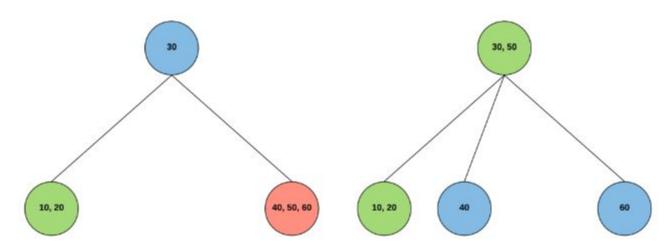


• 10, 30, 60, 20

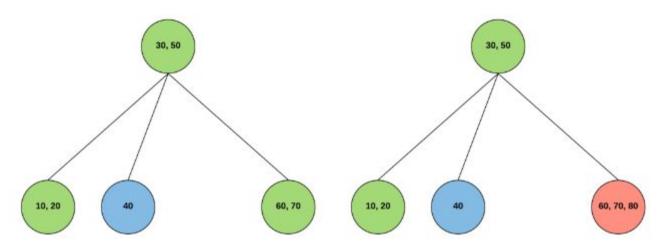
• 10, 30, 60, 20, 50



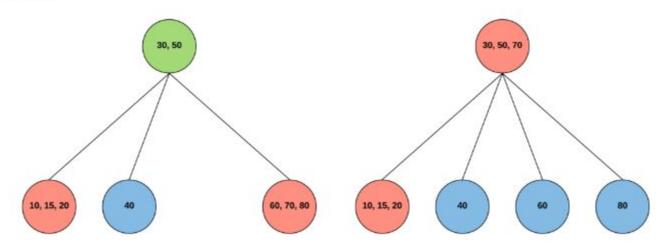
- 10, 30, 60, 20, 50, 40
- Split for 70



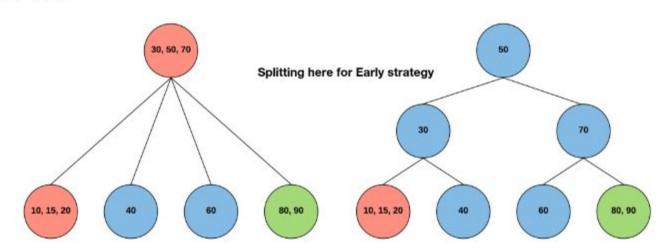
- 10, 30, 60, 20, 50, 40, 70
- 10, 30, 60, 20, 50, 40, 70, 80



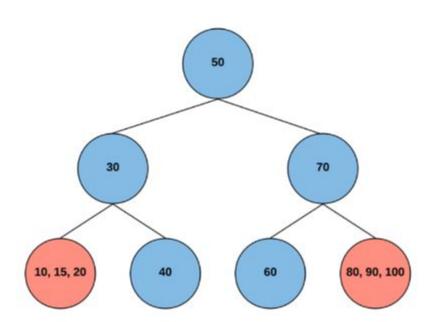
- 10, 30, 60, 20, 50, 40, 70, 80, 15
- Split for 90



- 10, 30, 60, 20, 50, 40, 70, 80, 15, 90
- Split for 100



• 10, 30, 60, 20, 50, 40, 70, 80, 15, 90, 100



Delete

- Locate the node n that contains the item theltem
- Find theltem's inorder successor and swap it with theltem (deletion will always be at a leaf)
- o If that leaf is a 3-node or a 4-node, remove theltem
- o To ensure that theltem does not occur in a 2-node
 - ▶ Transform each 2-node encountered into a 3-node or a 4-node
 - Reverse different cases illustrated for splitting

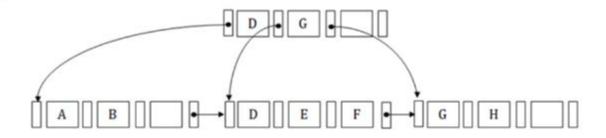
- Advantages
- o All leaves are at the same depth (the bottom level): Height, $h \sim O(\lg n)$
 - o Complexity of search, insert and delete: $O(h) \sim O(\lg n)$
 - All data is kept in sorted order
 Generalizes easily to larger nodes
 - Extends to external data structures
- Disadvantages
- Uses variety of node types need to destruct and construct multiple nodes for converting a 2 Node to 3 Node, a 3 Node to 4 Node, for splitting etc.

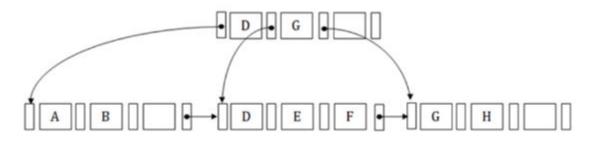
- Consider only one node type with space for 3 items and 4 links
- Internal node (non-root) has 2 to 4 children (links)
- Leaf node has 1 to 3 items Wastes some space, but has several advantages for external data structure
- Generalizes easily to larger nodes
- All paths from root to leaf are of the same length
 - \circ Each node that is not a root or a leaf has between $\lceil \frac{n}{2} \rceil$ and *n* children. \circ A leaf node has between $\left\lceil \frac{(n-1)}{2} \right\rceil$ and n-1 values
 - Special cases:
 - ▶ If the root is not a leaf, it has at least 2 children.
 - \triangleright If the root is a leaf, it can have between 0 and (n-1) values.
- Extends to external data structures
- B-Tree
 - o 2-3-4 Tree is a B-Tree where n=4

B+ Trees

The B⁺ Tree

- Is a balanced binary search tree
 - Follows a multi-level index format like 2-3-4 Tree
- Has the leaf nodes denoting actual data pointers
- Ensures that all leaf nodes remain at the same height (like 2-3-4 Tree)
- Has the leaf nodes are linked using a link list
 - Can support random access as well as sequential access
- Example:

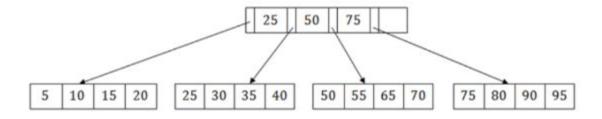




- Internal node contains
 - \circ At least $\frac{n}{2}$ child pointers, except the root node
 - At most n pointers
- Leaf node contains
 - At least $\frac{n}{2}$ record pointers and $\frac{n}{2}$ key values
 - At most n record pointer and n key values
 - One block pointer P to point to next leaf node

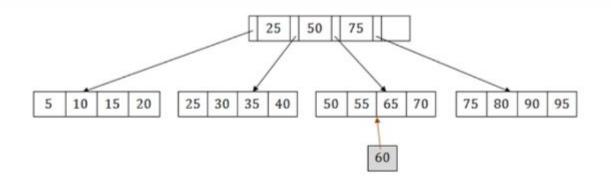
Note: These are approximate values, we will discuss more precise values later in this lecture.

- Suppose we have to search 55 in the B⁺ tree below
 - First, we will fetch for the intermediary node which will direct to the leaf node that can contain a record for 55
- So, in the intermediary node, we will find a branch between 50 and 75 nodes
 - o Then at the end, we will be redirected to the third leaf node
 - Here DBMS will perform a sequential search to find 55

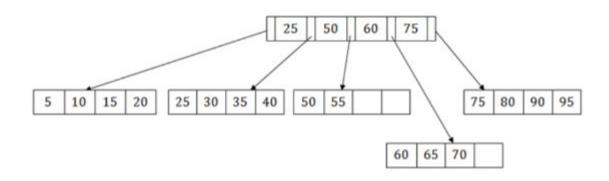


Source: B+ Tree

B+ Tree Insertion

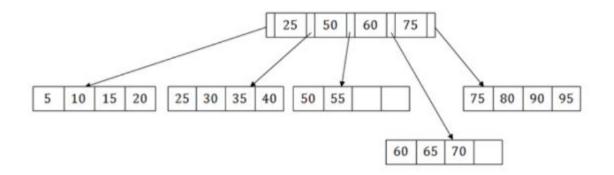


- Suppose we want to insert a record 60 that goes to 3rd leaf node after 55
- The leaf node of this tree is already full, so we cannot insert 60 there
- So we have to split the leaf node, so that it can be inserted into tree without affecting the fill factor, balance and order
- The 3rd leaf node has the values (50, 55, 60, 65, 70) and its current root node is 50
- We will split the leaf node of the tree in the middle so that its balance is not altered

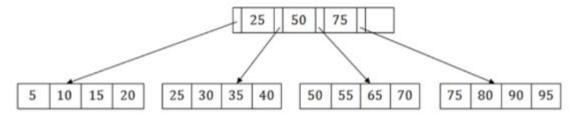


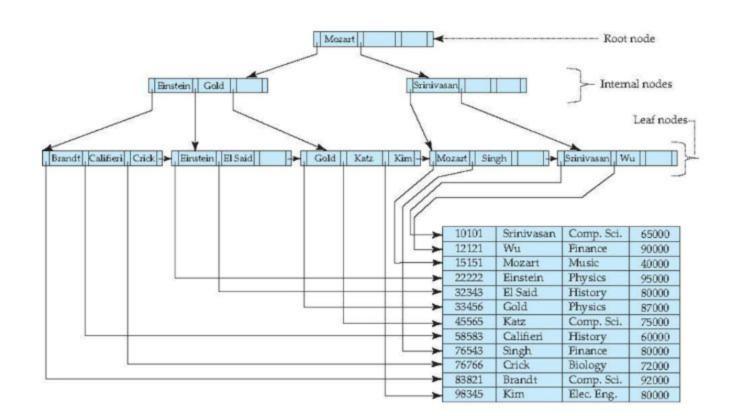
- So we can group (50, 55) and (60, 65, 70) into 2 leaf nodes
- If these two has to be leaf nodes, the intermediate node cannot branch from 50
- It should have 60 added to it, and then we can have pointers to a new leaf node

B+ Tree Deletion



- To delete 60, we have to remove 60 from intermediate node as well as 4th leaf node
- If we remove it from the intermediate node, then the tree will not remain a B+ tree
- So with deleting 60 we re-arranging the nodes:





- 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 $\lceil \frac{n}{2} \rceil$ and n children
- A leaf node has between an $\lceil \frac{n-1}{2} \rceil$ and n-1 values

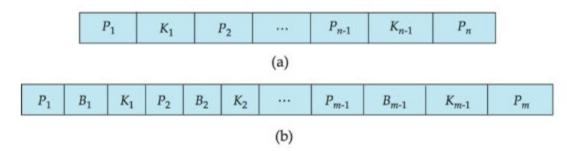
Special cases:

- 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.

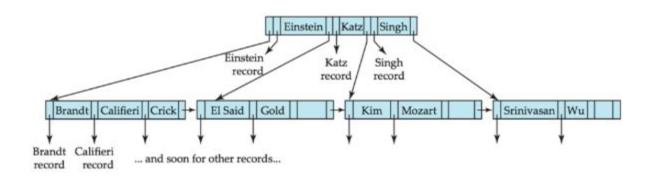
- Since the inter-node connections are done by pointers, logically close blocks need not be physically close
- The non-leaf levels of the B⁺ tree form a hierarchy of sparse indices
- The B+ tree contains a relatively small number of levels
- Level below root has at least $2 * \left\lceil \frac{n}{2} \right\rceil$ values
 - Next level has at least $2 * \lceil \frac{n}{2} \rceil * \lceil \frac{n}{2} \rceil$ values
 - o ... etc.
 - o 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

B Trees:

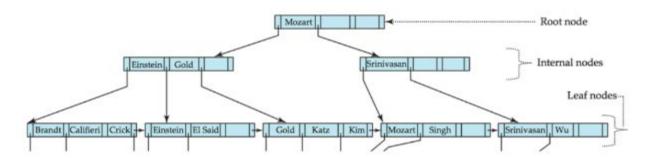
- Similar to B⁺ tree, but B-tree allows search-key values to appear only once; eliminates redundant storage of search keys
- Search keys in non-leaf nodes appear nowhere else in the B-tree; an additional pointer field for each search key in a non-leaf node must be included
- Generalized B-tree leaf node



Non-leaf node - pointers Bi are the bucket or file record pointers



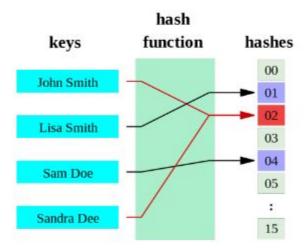
B-tree (above) and B+ tree (below) on same data



Hashing:

Static Hashing

- A hash function h maps data of arbitrary size (from domain D) to fixed-size values (say, integers from 0 to N > 0 h : D → [0..N]
- Given key k, h(k) is called hash values, hash codes, digests, or simply hashes
- If for two keys $k_1 \neq k_2$, we have $h(k_1) = h(k_2)$, we say a collision has occurred
- A hash function should be Collision Free and Fast



- A **bucket** is a unit of storage containing one or more records (a bucket is typically a disk block)
- In a hash file organization we obtain the bucket of a record directly from its search-key value using a hash function
- Hash function h is a function from the set of all search-key values K to the set of all bucket addresses B
- Hash function 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

Hash file organization of instructor file, using dept_name as key

- There are 10 buckets
- The binary representation of the i^{th} character is assumed to be the integer i
- The binary representation of the 1th character is assumed to be the integer 1
 The hash function returns the sum of the binary representations of the characters
- modulo 10

 o For example

$$h(Music) = 1$$
 $h(History) = 2$
 $h(Physics) = 3$ $h(Elec. Eng.) = 3$

bucket	0		
bucket	1		
15151	Mozart	Music	40000
bucket	2		
32343	El Said	History	80000
58583	Califieri	History	60000
bucket	3		
	Einstein	Physics	95000
33456		Physics	87000
98345	Kim	Elec. Eng.	80000

Hash file organization of instructor file, using dept_name as key

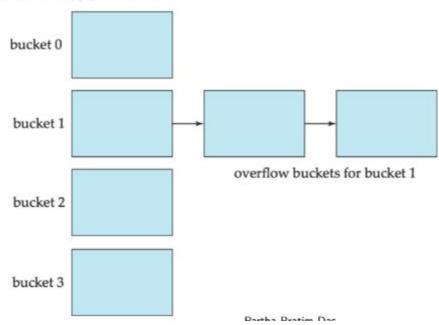
- Worst hash function maps all search-key values to the same bucket; this makes access time proportional to the number of search-key values in the file
- An ideal hash function is uniform, i.e., each bucket is assigned the same number of search-key values from the set of all possible values
- Ideal hash function is random, so 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
- For example, for a string search-key, the binary representations of all the characters in the string could be added and the sum modulo the number of buckets could be returned

- · Bucket overflow can occur because of
 - Insufficient buckets
 - o 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
- o it is handled by using overflow buckets

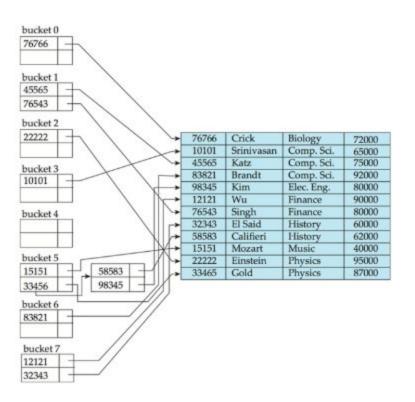
- Overflow chaining the overflow buckets of a given bucket are chained together in a linked list
- Above scheme is called closed hashing

Database Management Systems

 An alternative, called open hashing, which does not use overflow buckets, is not suitable for database applications



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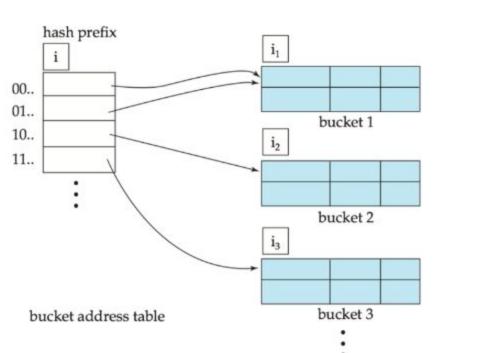
- Hash index on instructor, on attribute ID
- Computed by adding the digits modulo 8

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).
 - o 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:

- Good for database that grows and shrinks in size
- · Allows the hash function to be modified dynamically
- Extendable hashing one form of dynamic hashing
- Hash function generates values over a large range typically b-bit integers, with b = 32
 - At any time use only a prefix of the hash function to index into a table of bucket addresses
 - Let the length of the prefix be *i* bits, $0 \le i \le 32$
 - \triangleright Bucket address table size = 2^i . Initially i = 0
 - Value of i grows and shrinks as the size of the database grows and shrinks
 - Multiple entries in the bucket address table may point to a bucket (why?)
 Thus, actual number of buckets is < 2ⁱ
 - ▷ The number of buckets also changes dynamically due to coalescing and splitting of buckets

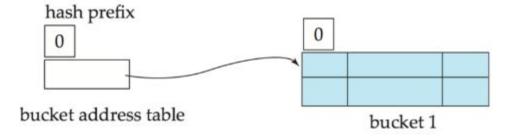


- Each bucket j stores a value i;
- \circ All the entries that point to the same bucket have the same values on the first i_j bits
- To locate the bucket containing search-key K_i
- Compute $h(K_i) = X$ Use the first i high order bits of X as a displacement into bucket address table, and follow the pointer to appropriate bucket
- To insert a record with search-key value K_i
- Follow same procedure as look-up and locate the bucket, say j If there is room in the bucket j insert record in the bucket
 - Else the bucket must be split and insertion re-attempted (next slide)
 - Overflow buckets used instead in some cases (will see shortly)

Example:

dept_name	h(dept_name)			
Biology	0010 1101 1111 1011 0010 1100 0011 0000			
Comp. Sci.	1111 0001 0010 0100 1001 0011 0110 1101			
Elec. Eng.	0100 0011 1010 1100 1100 0110 1101 1111			
Finance	1010 0011 1010 0000 1100 0110 1001 1111			
History	1100 0111 1110 1101 1011 1111 0011 1010			
Music	0011 0101 1010 0110 1100 1001 1110 1011			
Physics	1001 1000 0011 1111 1001 1100 0000 0001			

• Initial Hash structure; bucket size = 2

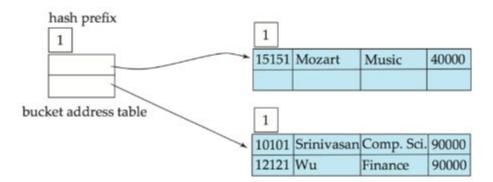


Insert "Mozart", "Srinivasan", and "Wu" records

dept_name	h(dept_name)			
Biology	0010 1101 1111 1011 0010 1100 0011 0000			
Comp. Sci.	1111 0001 0010 0100 1001 0011 0110 1101			
Elec. Eng.	0100 0011 1010 1100 1100 0110 1101 1111			
Finance	1010 0011 1010 0000 1100 0110 1001 1111			
History	1100 0111 1110 1101 1011 1111 0011 1010			
Music	0011 0101 1010 0110 1100 1001 1110 1011			
Physics	1001 1000 0011 1111 1001 1100 0000 0001			

76766	Crick	Biology	72000
10101	Srinivasan	Comp. Sci.	65000
45565	Katz	Comp. Sci.	75000
83821	Brandt	Comp. Sci.	92000
98345	Kim	Elec. Eng.	80000
12121	Wu	Finance	90000
76543	Singh	Finance	80000
32343	El Said	History	60000
58583	Califieri	History	62000
15151	Mozart	Music	40000
22222	Einstein	Physics	95000
33465	Gold	Physics	87000

Hash structure after insertion of "Mozart", "Srinivasan", and "Wu" records

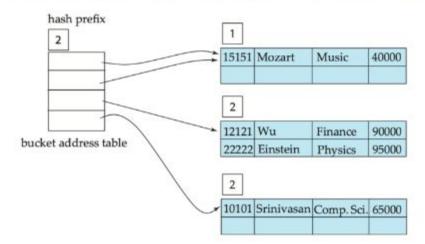


Insert Einstein record

dept_name	h(dept_name)
Biology	0010 1101 1111 1011 0010 1100 0011 0000
Comp. Sci.	1111 0001 0010 0100 1001 0011 0110 1101
Elec. Eng.	0100 0011 1010 1100 1100 0110 1101 1111
Finance	1010 0011 1010 0000 1100 0110 1001 1111
History	1100 0111 1110 1101 1011 1111 0011 1010
Music	0011 0101 1010 0110 1100 1001 1110 1011
Physics	1001 1000 0011 1111 1001 1100 0000 0001

ł	76766	Crick	Biology	72000
I	10101	Srinivasan	Comp. Sci.	65000
ĺ	45565	Katz	Comp. Sci.	75000
ł	83821	Brandt	Comp. Sci.	92000
ł	98345	Kim	Elec. Eng.	80000
ł	12121	Wu	Finance	90000
ł	76543	Singh	Finance	80000
ł	32343	El Said	History	60000
ł	58583	Califieri	History	62000
ł	15151	Mozart	Music	40000
ł	22222	Einstein	Physics	95000
I	33465	Gold	Physics	87000

Hash structure after insertion of "Einstein" record

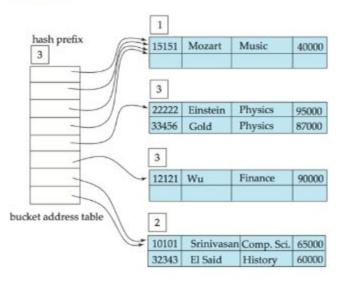


· Insert "Gold" and "El Said" records

dept_name	h(dept_name)
Biology	0010 1101 1111 1011 0010 1100 0011 0000
Comp. Sci.	1111 0001 0010 0100 1001 0011 0110 1101
Elec. Eng.	0100 0011 1010 1100 1100 0110 1101 1111
Finance	1010 0011 1010 0000 1100 0110 1001 1111
History	1100 0111 1110 1101 1011 1111 0011 1010
Music	0011 0101 1010 0110 1100 1001 1110 1011
Physics	1001 1000 0011 1111 1001 1100 0000 0001

Crick	Biology	72000
Srinivasan	Comp. Sci.	65000
Katz	Comp. Sci.	75000
Brandt	Comp. Sci.	92000
Kim	Elec. Eng.	80000
Wu	Finance	90000
Singh	Finance	80000
El Said	History	60000
Califieri	History	62000
Mozart	Music	40000
Einstein	Physics	95000
Gold	Physics	87000
	Srinivasan Katz Brandt Kim Wu Singh El Said Califieri Mozart Einstein	Srinivasan Comp. Sci. Katz Comp. Sci. Brandt Comp. Sci. Kim Elec. Eng. Wu Finance Singh Finance El Said History Califieri History Mozart Music Einstein Physics

 Hash structure after insertion of "Gold" and "El Said" records



Insert Katz record

 dept_name
 h(dept_name)

 Biology
 0010 1101 1111 1011 0010 1100 0011 0000

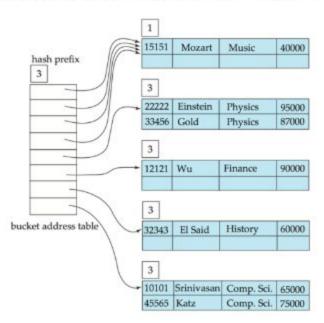
 Comp. Sci.
 1111 0001 0010 0100 1001 0011 0110 1101

 Elec. Eng.
 0100 0011 1010 1100 1100 0110 1101 1111

 Finance
 1010 0011 1010 0000 1100 0110 1001 1111

76766	Crick	Biology	72000
10101	Srinivasan	Comp. Sci.	65000
45565	Katz	Comp. Sci.	75000
83821	Brandt	Comp. Sci.	92000
98345	Kim	Elec. Eng.	80000
12121	Wu	Finance	90000
76543	Singh	Finance	80000
32343	El Said	History	60000
58583	Califieri	History	62000
15151	Mozart	Music	40000
22222	Einstein	Physics	95000
33465	Gold	Physics	87000

Hash structure after insertion of "Katz" record

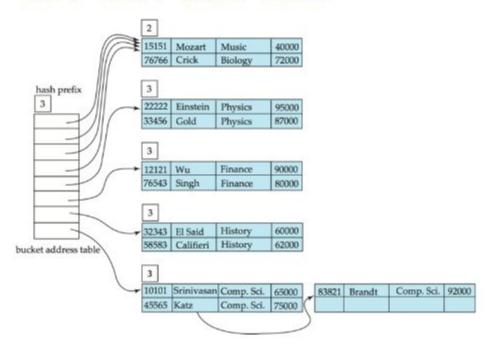


 Insert "Singh", "Califieri", "Crick", "Brandt" records dept_name h(dept_name)

Biology Comp. Sci. Elec. Eng. Finance History Music Physics 0010 1101 1111 1011 0010 1100 0011 0000 1111 0001 0010 0100 1001 0011 0110 1101 0100 0100 0011 1101 1101 0100 0011 1101 1101 1101 0011 1010 1101 0011 1101 1111 1100 0111 1110 1101 1101 1111 1111 0011 1111 1100 0111 1110 1101 1100 1100 1100 1111 1100 1100 1100 0000 0001

1	76766	Crick	Biology	72000
1	10101	Srinivasan	Comp. Sci.	65000
	45565	Katz	Comp. Sci.	75000
ł	83821	Brandt	Comp. Sci.	92000
1	98345	Kim	Elec. Eng.	80000
1	12121	Wu	Finance	90000
1	76543	Singh	Finance	80000
1	32343	El Said	History	60000
1	58583	Califieri	History	62000
1	15151	Mozart	Music	40000
ł	22222	Einstein	Physics	95000
1	33465	Gold	Physics	87000

Hash structure after insertion of "Singh", "Califieri", "Crick", "Brandt" records

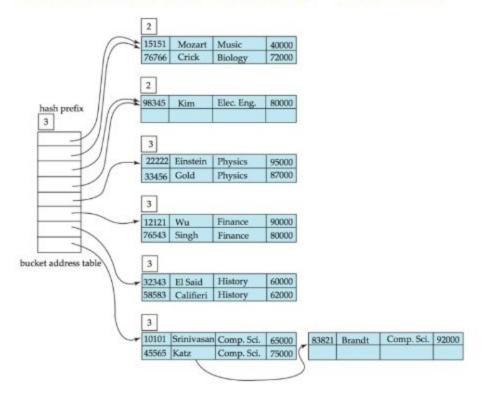


h(dept_name) dept name Biology 0010 1101 1111 1011 0010 1100 0011 0000 Comp. Sci. 1111 0001 0010 0100 1001 0011 0110 1101 Elec. Eng. 0100 0011 1010 1100 1100 0110 1101 1111 Finance 1010 0011 1010 0000 1100 0110 1001 1111 History 1100 0111 1110 1101 1011 1111 0011 1010 Music 0011 0101 1010 0110 1100 1001 1110 1011 Physics 1001 1000 0011 1111 1001 1100 0000 0001

76766	Crick	Biology	72000
10101	Srinivasan	Comp. Sci.	65000
45565	Katz	Comp. Sci.	75000
83821	Brandt	Comp. Sci.	92000
98345	Kim	Elec. Eng.	80000
12121	Wu	Finance	90000
76543	Singh	Finance	80000
32343	El Said	History	60000
58583	Califieri	History	62000
15151	Mozart	Music	40000
22222	Einstein	Physics	95000
33465	Gold	Physics	87000

Insert Kim record

Hash structure after insertion of "Kim" record



 dept_name
 h(dept_name)

 Biology
 0010 1101 1111 1011 0010 1100 0011 0000

 Comp. Sci.
 1111 0001 0010 0100 1001 0011 0110 1101

 Elec. Eng.
 0100 0011 1010 1100 1100 0110 1011 1111

 Finance
 1010 0011 1010 0000 1100 0110 1001 1111

 History
 1100 0111 1110 1101 1011 1111 0011

 Music
 0011 0101 1010 0110 1100 1001 1110 1001

1001 1000 0011 1111 1001 1100 0000 0001

Physics

76766	Crick	Biology	72000
10101	Srinivasan	Comp. Sci.	65000
45565	Katz	Comp. Sci.	75000
83821	Brandt	Comp. Sci.	92000
98345	Kim	Elec. Eng.	80000
12121	Wu	Finance	90000
76543	Singh	Finance	80000
32343	El Said	History	60000
58583	Califieri	History	62000
15151	Mozart	Music	40000
22222	Einstein	Physics	95000
33465	Gold	Physics	87000

Bitmap Indices:

- Bitmap indices are a special type of index designed for efficient querying on multiple keys
- Records in a relation are assumed to be numbered sequentially from, say, 0
- \circ Given a number n it must be easy to retrieve record n

Particularly easy if records are of fixed size

- Applicable on attributes that take on a relatively small number of distinct values
 - Applicable on attributes that take on a relatively small number of distinct value
 - For example: gender, country, state, . . .
 For example: income-level (income broken up into a small number of levels such as 0-9999, 10000-19999, 20000-50000, 50000- infinity)
- A bitmap is simply an array of bits

- In its simplest form a bitmap index on an attribute has a bitmap for each value of the attribute
 - o 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

				Bitmaps for gender			Bitmaps for
record number	ID	gender	income_level	m	10010		income_level
0	76766	m	L1	f	01101	L1	10100
1	22222	f	L2			L2	01000
2	12121	f	L1			L3	00001
3	15151	m	L4			L4	00010
4	58583	f	L3			L5	00000
						-	00000

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