Lecture 11

Lecture 11: B+ Trees: An IO-Aware Index Structure

"If you don't find it in the index, look very carefully through the entire catalog"

- Sears, Roebuck and Co., Consumers Guide, 1897

Today's Lecture

- 1. Indexes: Motivations & Basics
- 2. B+ Trees

Lecture 11

1. Indexes: Motivations & Basics

What you will learn about in this section

- 1. Indexes: Motivation
- 2. Indexes: Basics
- 3. ACTIVITY: Creating indexes

Index Motivation

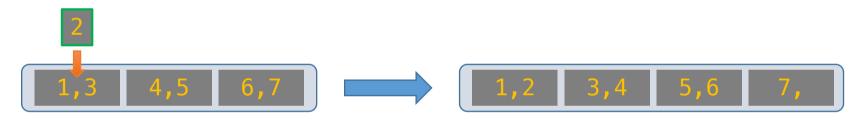
Person(<u>name</u>, age)

- Suppose we want to search for people of a specific age
- *First idea:* Sort the records by age... we know how to do this fast!
- How many IO operations to search over *N sorted* records?
 - Simple scan: O(N)
 - Binary search: O(log₂ N)

Could we get even cheaper search? E.g. go from $\log_2 N$ $\rightarrow \log_{200} N$?

Index Motivation

• What about if we want to **insert** a new person, but keep the list sorted?



- We would have to potentially shift *N* records, requiring up to ~ 2*N/P IO operations (where P = # of records per page)!
 - We could leave some "slack" in the pages...

Could we get faster insertions?

Index Motivation

- What about if we want to be able to search quickly along multiple attributes (e.g. not just age)?
 - We could keep multiple copies of the records, each sorted by one attribute set... this would take a lot of space

Can we get fast search over multiple attribute (sets) without taking too much space?

We'll create separate data structures called *indexes* to address all these points

Further Motivation for Indexes: NoSQL!

- NoSQL engines are (basically) *just indexes!*
 - A lot more is left to the user in NoSQL... one of the primary remaining functions of the DBMS is still to provide index over the data records, for the reasons we just saw!
 - Sometimes use B+ Trees (covered next), sometimes hash indexes (not covered here)

Indexes are critical across all DBMS types

Indexes: High-level

- An <u>index</u> on a file speeds up selections on the <u>search key</u> fields for the index.
 - Search key properties
 - Any subset of fields
 - is **not** the same as key of a relation
- Example:

Product(<u>name</u>, maker, price)

On which attributes would you build indexes?

More precisely

- An <u>index</u> is a data structure mapping <u>search keys</u> to <u>sets of rows in a</u> <u>database table</u>
 - Provides efficient lookup & retrieval by search key value- usually much faster than searching through all the rows of the database table
- An index can store the full rows it points to (*primary index*) or pointers to those rows (*secondary index*)
 - We'll mainly consider secondary indexes

Operations on an Index

- <u>Search</u>: Quickly find all records which meet some *condition on the search key attributes*
 - More sophisticated variants as well. Why?
- Insert / Remove entries
 - Bulk Load / Delete. Why?

Indexing is one the most important features provided by a database for performance

Conceptual Example

What if we want to return all books published after 1867? The above table might be very expensive to search over row-by-row...

Russian_Novels

BID	Title	Author	Published	Full_text
001	War and Peace	Tolstoy	1869	
002	Crime and Punishment	Dostoyevsky	1866	
003	Anna Karenina	Tolstoy	1877	

SELECT *
FROM Russian_Novels
WHERE Published > 1867

By_Yr_Index

Conceptual Example

Russian_Novels

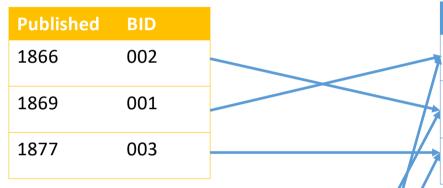
Published	BID
1866	002
1869	001
1877	003

Maintain an index for this, and search over that!

Why might just keeping the table sorted by year not be good enough?

Conceptual Example

By_Yr_Index



By_Author_Title_Index

Author	Title	BID
Dostoyevsky	Crime and Punishment	002
Tolstoy	Anna Karenina	003
Tolstoy	War and Peace	001

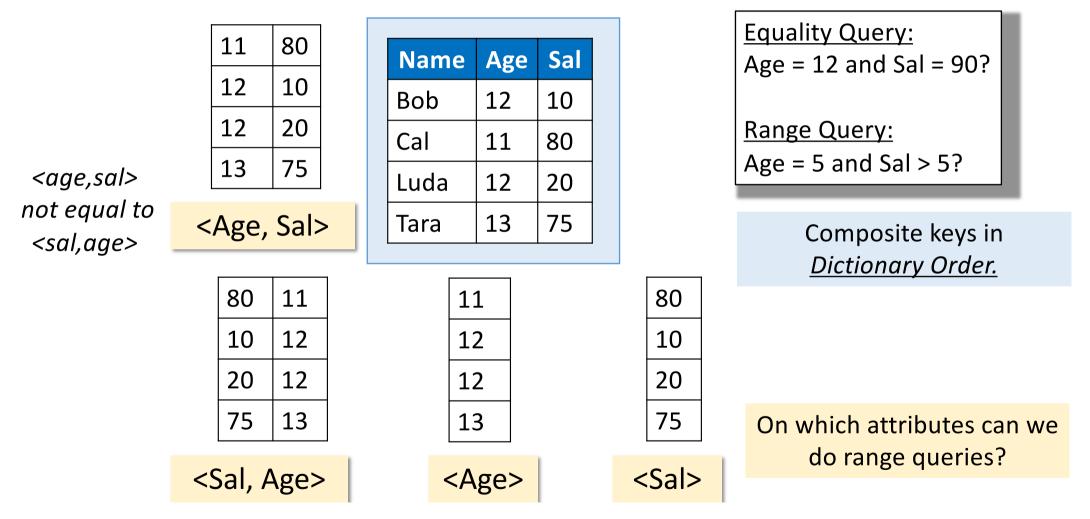
Russian_Novels

BID	Title	Author	Published	Full_text
001	War and Peace	Tolstoy	1869	
002	Crime and Punishment	Dostoyevsky	1866	
003	Anna Karenina	Tolstoy	1877	

Can have multiple indexes to support multiple search keys

Indexes shown here as tables, but in reality we will use more efficient data structures...

Composite Keys



Composite Keys

- Pro:
 - When they work they work well
 - We'll see a good case called "index-only" plans or **covering** indexes.
- Con:
 - Guesses? (time and space)

Covering Indexes

By_Yr_Index

Published	BID
1866	002
1869	001
1877	003

We say that an index is <u>covering</u> for a specific query if the index contains all the needed attributes*meaning the query can be answered using the index alone!*

The "needed" attributes are the union of those in the SELECT and WHERE clauses...

Example:

SELECT Published, BID
FROM Russian_Novels
WHERE Published > 1867

High-level Categories of Index Types

- B-Trees (covered next)
 - Very good for range queries, sorted data
 - Some old databases only implemented B-Trees
 - We will look at a variant called **B+ Trees**

The data structures we present here are "IO aware"

- Hash Tables (not covered)
 - There are variants of this basic structure to deal with IO
 - Called *linear* or *extendible hashing-* IO aware!

Real difference between structures: costs of ops determines which index you pick and why *Lecture 11 > Section 1 > ACTIVITY*

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Lecture 11 > Section 2

2. B+ Trees

What you will learn about in this section

- 1. B+ Trees: Basics
- 2. B+ Trees: Design & Cost
- 3. Clustered Indexes

B+ Trees

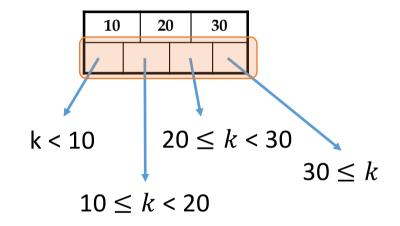
- Search trees
 - B does not mean binary!
- Idea in B Trees:
 - make 1 node = 1 physical page
 - Balanced, height adjusted tree (not the B either)
- Idea in B+ Trees:
 - Make leaves into a linked list (for range queries)

10	20)	30

Parameter *d* = the degree

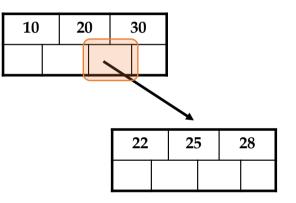
Each *non-leaf ("interior") node* has \geq d and \leq 2d *keys**

*except for root node, which can have between **1** and 2d keys

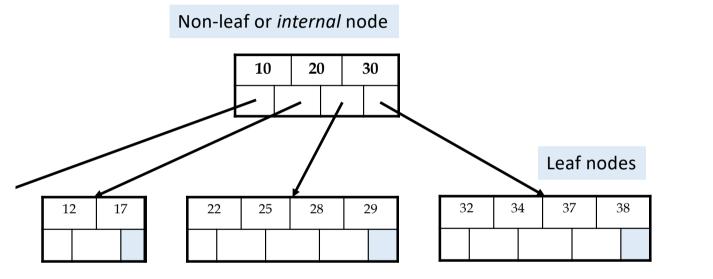


The *n* keys in a node define *n*+1 ranges

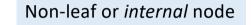
Non-leaf or *internal* node

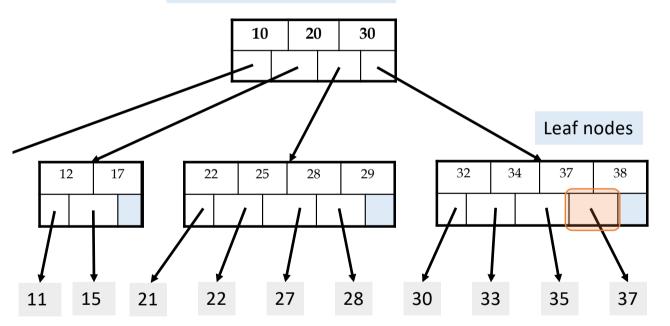


For each range, in a *non-leaf* node, there is a **pointer** to another node with keys in that range



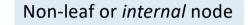
Leaf nodes also have between *d* and *2d* keys, and are different in that:

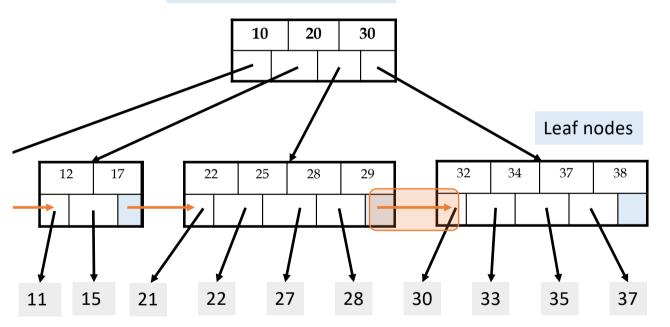




Leaf nodes also have between *d* and *2d* keys, and are different in that:

Their key slots contain pointers to data records

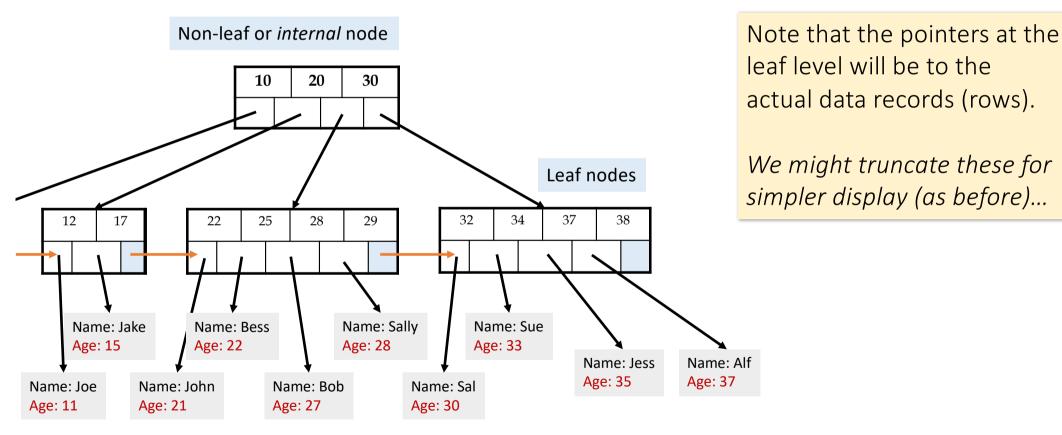




Leaf nodes also have between *d* and *2d* keys, and are different in that:

Their key slots contain pointers to data records

They contain a pointer to the next leaf node as well, *for faster sequential traversal*



Lecture 11 > Section 2 > B+ Tree basics

Some finer points of B+ Trees

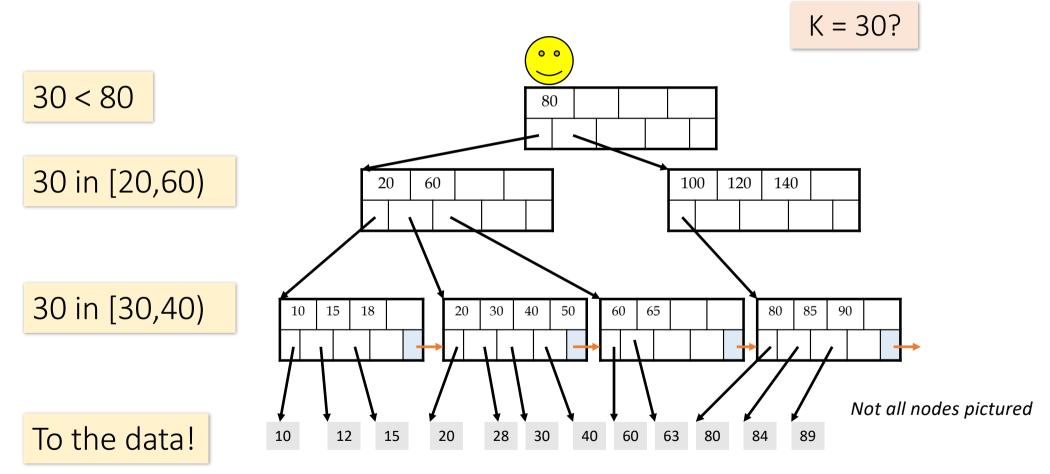
Searching a B+ Tree

- For exact key values:
 - Start at the root
 - Proceed down, to the leaf
- For range queries:
 - As above
 - Then sequential traversal

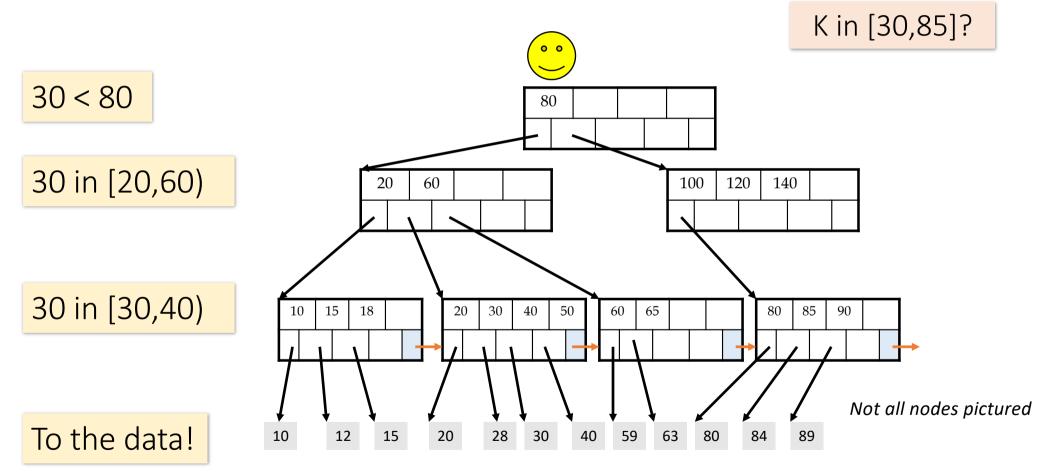
SELECT	name
FROM	people
WHERE	age = 25

SELECT	name
FROM	people
WHERE	20 <= age
AND	age <= 30

B+ Tree Exact Search Animation



B+ Tree Range Search Animation



B+ Tree Design

- How large is **d**?
- Example:
 - Key size = 4 bytes
 - Pointer size = 8 bytes
 - Block size = 4096 bytes

• We want each *node* to fit on a single *block/page*

• 2d x 4 + (2d+1) x 8 <= 4096 → d <= 170

NB: Oracle allows 64K = 2^16 byte blocks → d <= 2730

B+ Tree: High Fanout = Smaller & Lower IO

- As compared to e.g. binary search trees, B+ Trees have high fanout (between d+1 and 2d+1)
- This means that the depth of the tree is small → getting to any element requires very few IO operations!
 - Also can often store most or all of the B+ Tree in main memory!
- A TiB = 2⁴⁰ Bytes. What is the height of a B+ Tree (with fill-factor = 1) that indexes it (with 64K pages)?
 - $(2^{*}2730 + 1)^{h} = 2^{40} \rightarrow h = 4$

The <u>fanout</u> is defined as the number of pointers to child nodes coming out of a node

Note that fanout is dynamicwe'll often assume it's constant just to come up with approximate eqns!

The known universe contains ~10⁸⁰ particles... what is the height of a B+ Tree that indexes these?

B+ Trees in Practice

- Typical order: d=100. Typical fill-factor: 67%.
 - average fanout = 133
- Typical capacities:
 - Height 4: 133⁴ = 312,900,700 records
 - Height 3: 133³ = 2,352,637 records
- Top levels of tree sit *in the buffer pool*:
 - Level 1 = 1 page = 8 Kbytes
 - Level 2 = 133 pages = 1 Mbyte
 - Level 3 = 17,689 pages = 133 MBytes

<u>Fill-factor</u> is the percent of available slots in the B+ Tree that are filled; is usually < 1 to leave slack for (quicker) insertions

Typically, only pay for one IO!

Simple Cost Model for Search

- Let:
 - *f* = fanout, which is in [d+1, 2d+1] (we'll assume it's constant for our cost model...)
 - **N** = the total number of *pages* we need to index
 - **F** = fill-factor (usually ~= 2/3)
- Our B+ Tree needs to have room to index **N / F** pages!
 - We have the fill factor in order to leave some open slots for faster insertions
- What height (*h*) does our B+ Tree need to be?
 - h=1 \rightarrow Just the root node- room to index f pages
 - h=2 \rightarrow f leaf nodes- room to index f² pages
 - h=3 \rightarrow f² leaf nodes- room to index f³ pages
 - ...
 - $h \rightarrow f^{h-1}$ leaf nodes- room to index f^h pages!

→ We need a B+ Tree of height h = $\left[\log_f \frac{N}{F}\right]!$

Simple Cost Model for Search

- Note that if we have **B** available buffer pages, by the same logic:
 - We can store L_B levels of the B+ Tree in memory
 - where L_B is the number of levels such that the sum of all the levels' nodes fit in the buffer:
 - $B \ge 1 + f + \dots + f^{L_B 1} = \sum_{l=0}^{L_B 1} f^l$
- In summary: to do exact search:
 - We read in one page per level of the tree
 - However, levels that we can fit in buffer are free!
 - Finally we read in the actual record

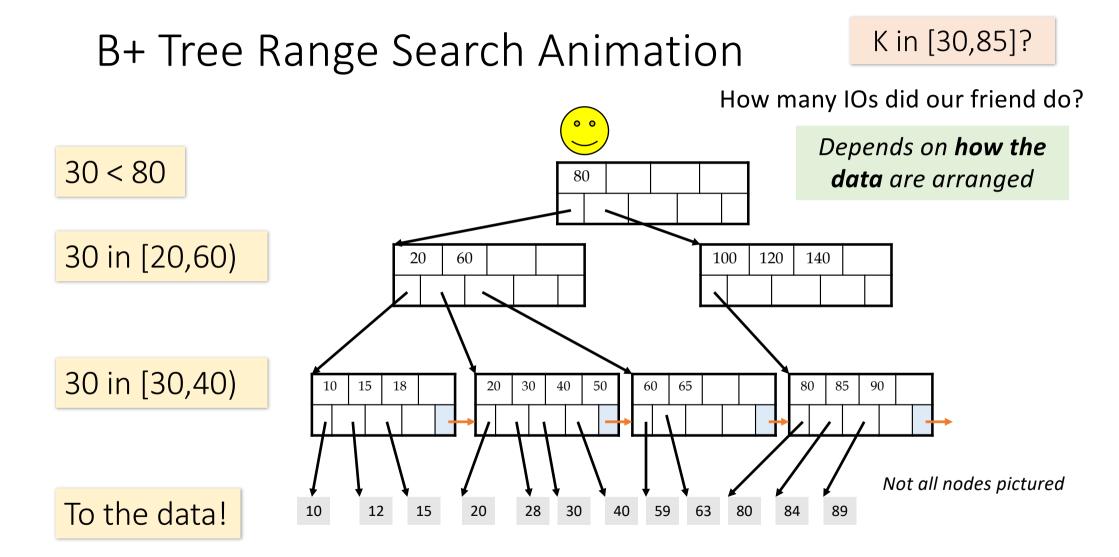
IO Cost: $\left[\log_{f} \frac{N}{F}\right] - L_{B} + 1$ where $B \ge \sum_{l=0}^{L_{B}-1} f^{l}$

Simple Cost Model for Search

- To do range search, we just follow the horizontal pointers
- The IO cost is that of loading additional leaf nodes we need to access + the IO cost of loading each *page* of the results- we phrase this as "Cost(OUT)"

IO Cost:
$$\left[\log_{f} \frac{N}{F}\right] - L_{B} + Cost(OUT)$$

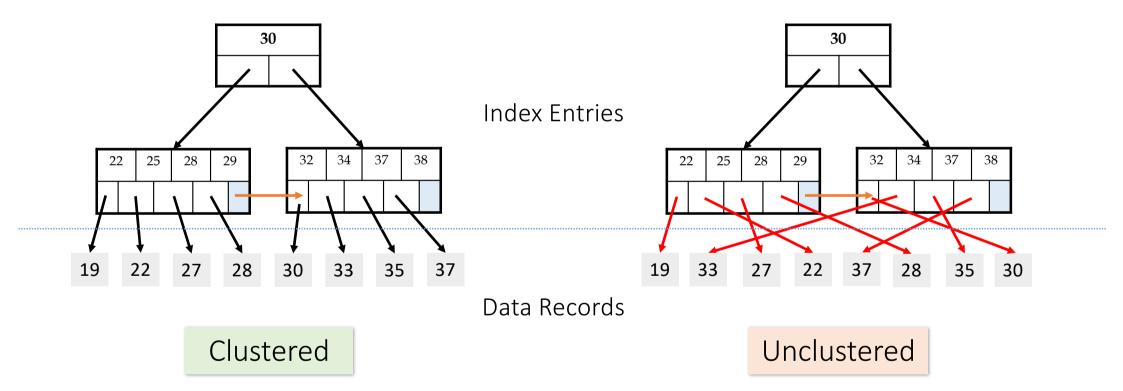
where $B \ge \sum_{l=0}^{L_{B}} f^{l}$



Clustered Indexes

An index is <u>clustered</u> if the underlying data is ordered in the same way as the index's data entries.

Clustered vs. Unclustered Index



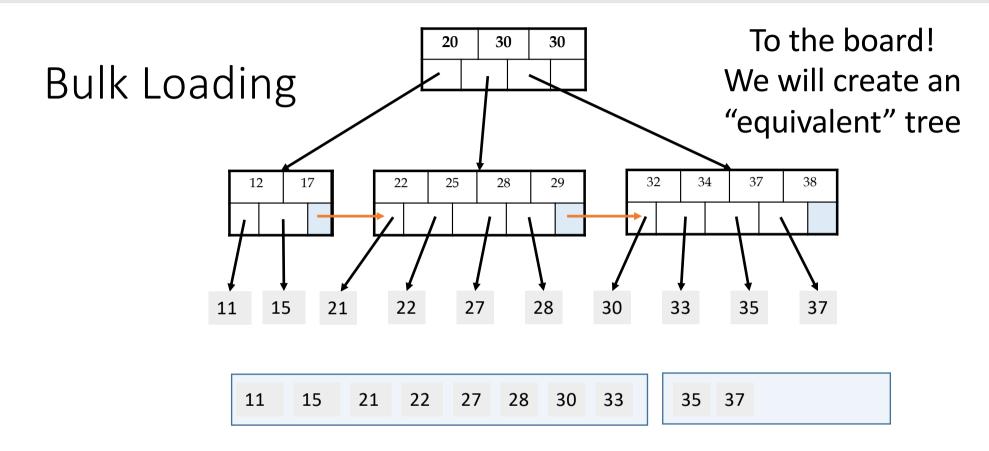
Clustered vs. Unclustered Index

- Recall that for a disk with block access, sequential IO is much faster than random IO
- For exact search, no difference between clustered / unclustered
- For range search over R values: difference between 1 random IO + R sequential IO, and R random IO:
 - A random IO costs ~ 10ms (sequential much much faster)
 - For R = 100,000 records- difference between ~10ms and ~17min!

Fast Insertions & Self-Balancing

- We won't go into specifics of B+ Tree insertion algorithm, but has several attractive qualities:
 - ~ Same cost as exact search
 - **Self-balancing:** B+ Tree remains **balanced** (with respect to height) even after insert

B+ Trees also (relatively) fast for single insertions! However, can become bottleneck if many insertions (if fill-factor slack is used up...)



Input: Sorted File Output: B+ Tree

Message: Bulk Loading is faster!

Summary

- We covered an algorithm + some optimizations for sorting largerthan-memory files efficiently
 - An *IO aware* algorithm!
- We create indexes over tables in order to support fast (exact and range) search and insertion over multiple search keys
- **B+ Trees** are one index data structure which support very fast exact and range search & insertion via *high fanout*
 - *Clustered vs. unclustered* makes a big difference for range queries too