DB 2

09 - Ordered Indexes (B+Trees)

Summer 2020

Torsten Grust Universität Tübingen, Germany Sequential scan (**Seq Scan**) and interpreted predicate evaluation go a long way. Large input tables call for significantly more **efficient support for value-based row access:**

```
SELECT i.b, i.c

FROM indexed AS i

WHERE i.a = 42 [i.c = 0.42] -- either filter on i.a or i.c
```

Assume column a is **primary key** in table indexed: expect query workload that frequently identifies rows via predicates a = k. **Indexes** can support such queries.



DBMS expects predicates a = k and creates an **index on column** a—a data structure associated with and maintained in addition to table indexed—to speed up evaluation:

CREATE INDEX indexed_a ON indexed USING btree (a);

- 2. When indexed is updated, indexed_a is maintained. ❖

¹ PostgreSQL chooses index name indexed_pkey but let's follow a <tαble>_<column> naming scheme here.



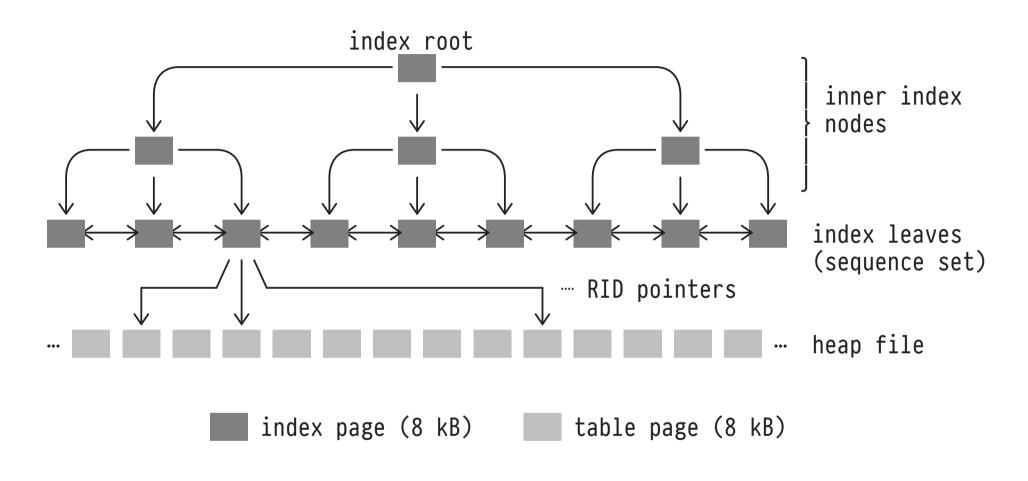
```
EXPLAIN VERBOSE
SELECT i.b, i.c
FROM indexed AS i -- 10° rows
WHERE i.a = 42; -- selection on key column a ⇒ ≤ 1 row will qualify

QUERY PLAN

Index Scan using indexed_a on indexed i (cost=0.42..8.44 rows=1 ...)
Output: b, c
Index Cond: (i.a = 42) --
```

- DBMS uses Index Scan (instead of Seq Scan), index scan will evaluate predicate i.a = k.
- System expects small result of a single row (rows=1),
 i.e., the predicate is assumed to be very selective.





Anatomy of a B+Tree

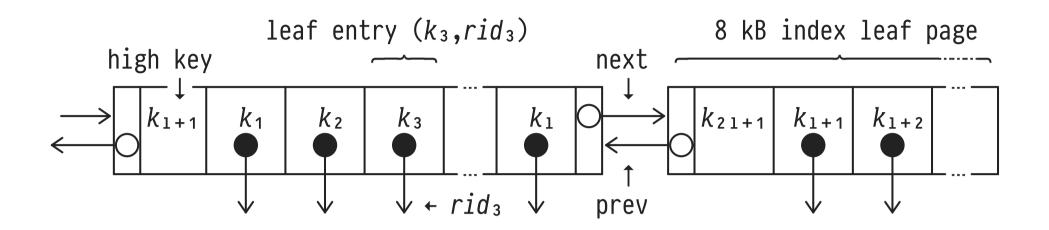


Notes on B+Tree anatomy:

- A B*Tree² index I on column T(a) is an ordered, n-ary (n
 > 2), balanced, block-oriented, dynamic search tree.
- Inner nodes and leaves are formed by 8 kB index pages.
- Each inner node holds n-1 values of column α (separators) that allow to navigate the search tree structure.
- Leaves form a bidirectional chain, the sequence set.
- Leaves use RIDs to point to rows in the heap file of table T: besides α column values, I holds no data of T.

² Invented by Bayer and McCreight (1969) at Boeing Labs. The "B" in "B+Tree" does *not* stand for Bayer, binary, balanced, block, or Boeing. (We tried to find out, but Rudolf Bayer wouldn't say.)





- Uses pointers prev/next to form the chained sequence set.
- Leaf entries are ordered by index keys k_i : $k_i \le k_{i+1}$.
- RID rid_i points to a row t of T with $t \cdot a = k_i$.
- The high key holds smallest key of next leaf (if any).

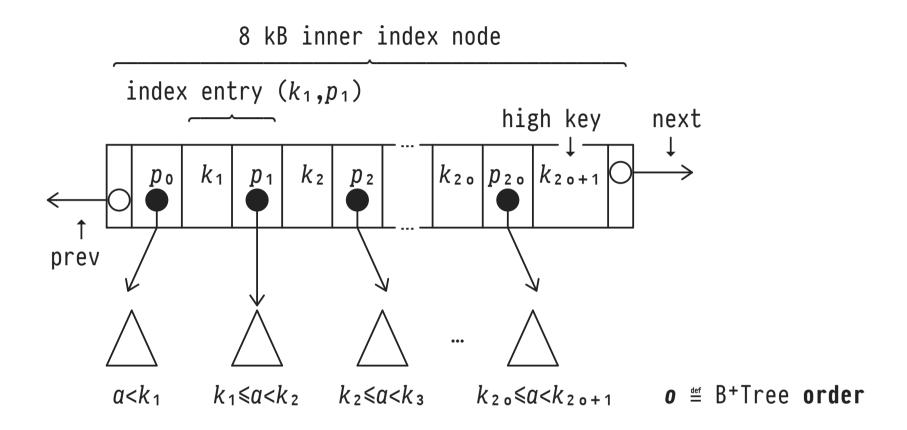


As described, a B+Tree is a **dense** index structure: every row t of T is represented by one leaf entry.

- The sequence set is ordered by keys $k_i \Rightarrow$ a binary search for a key $k = k_i$ may sound viable, **BUT** the search would
 - 1. need to inspect $\log_2(|T|)$ keys in the sequence set and access just as many pages \mathbb{Q} , and
 - 2. "jump around" the sequence set in an unpredictable fashion, thus leading to random I/O. \Box

B+Trees exploit the sequence set ordering and erect an n-ary search tree structure (n large!) atop the leaf entries.





- The **separator** keys k_i are ordered: $k_i \leq k_{i+1}$.
- Page pointers p_j point to index (leaf or inner) nodes.



- Space in inner nodes is used economically: in a B+Tree of order o, any inner node—but the root node—is guaranteed to hold between o and 2 × o (≝ fan-out F) index entries.
- Given predicate $t \cdot a = k$, perform binary search inside node to find B+Tree subtree with $k_i \leq k < k_{i+1}$.
- B+Tree is **balanced**: subtrees \triangle are of identical height.
- Path length s from B+Tree root to leaf node predictable:

$$|T| \times 1/F \times \cdots \times 1/F = 1 \Leftrightarrow s = \log_F(|T|)$$
s times



A B+Tree is *the* index structure to support the evaluation of these kinds of conditions:

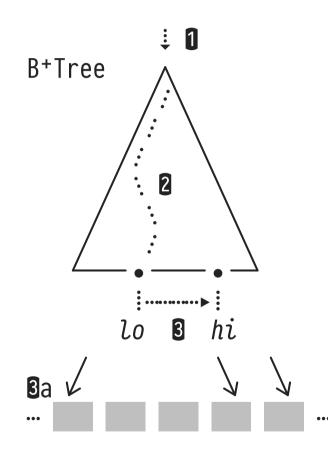
- 1. Range predicates: $lo \leq a \leq hi$
- 2. Half-open ranges: $lo \leq a$ or $a \leq hi$
- 3. Equality predicates: $\alpha = lo$
- An Index Scan on index I for column T(a) is parameterized by such a condition (PostgreSQL EXPLAIN: Index Cond).
- Index Scan uses *lo* to navigate the search tree structure and locate the start of relevant sequence set section.

³ Half-open ranges are special range predicates where $hi = \infty$ ($lo = \infty$). Equality predicates are special range predicates where lo = hi.



An index scan accesses the B+Tree index and the heap file:

heap file



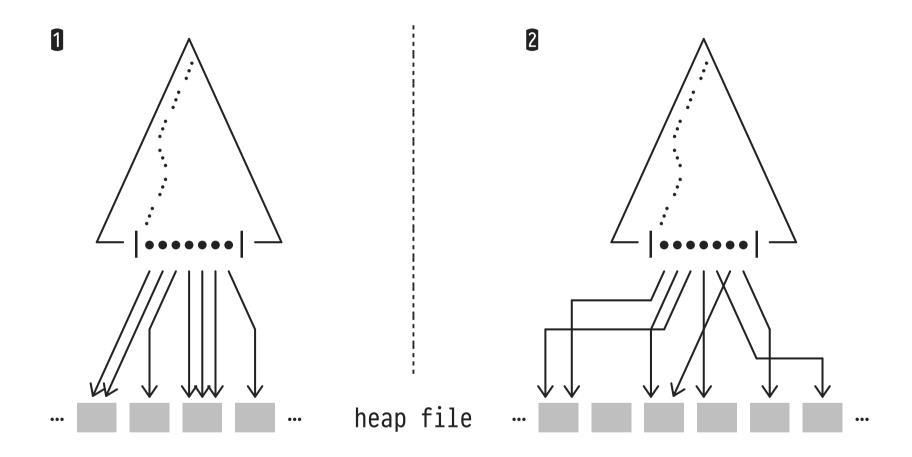
- 1 Enter at B+Tree root page
- 2 Use key lo to navigate the inner nodes (search tree) until we reach the leaf level
- 8 Scan leaf entries in the sequence set section $lo \le a \le hi$, extract RIDs
 - **B**a For each RID, access heap file for table *T* and return matching row



Phase 2 runs a vanilla traversal of a 2×o-way search tree:

```
Search(lo):
                                         returns entry point
  return TreeSearch(lo, root);
                                         for scan of
                                          sequence set
TreeSearch(lo, node):
  if (node is a leaf)
   return node;
  switch lo
     case lo < k_1
      return TreeSearch(lo, p_0);
                                         use binary search
     case k_i \leq lo < k_{i+1}
                                          to implement
      return TreeSearch(lo, p<sub>i</sub>);
                                         subtree choice
     case k_{20} \leq lo
       return TreeSearch(lo, p_{2o});
```





- 1 Order of leaf entry keys $k_i \equiv \text{row order in heap file.}$
- ② Order of k_i in sequence set and row order do not match.



Index I for column T(a) is **clustered** if the order of leaf entries coincides with T's row order (i.e., both I's sequence set and T's heap file are ordered by a):

Given entries $\langle k_i, p_i \rangle$ and $\langle k_j, p_j \rangle$, $k_i \leq k_j \Rightarrow p_i \leq p_j$.

- An Index Scan over a clustered index
 - 1. collects matching rows from adjacent heap file pages $(\Rightarrow \text{ sequential I/O } \bigcirc)$,
 - 2. will find many matching rows on each accessed heap file page (\Rightarrow less page I/O \circlearrowleft).

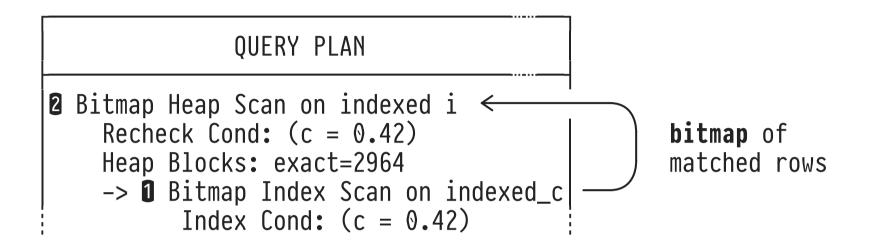


Sad fact: only *one*—among the many possible—indexes for a table may be clustered. Most indexes are non-clustered.

- An Index Scan over a non-clustered index
 - 1. will find matching rows potentially scattered across all heap file pages (\Rightarrow random I/O \heartsuit),
 - 2. will find few matching rows on each accessed heap file page and may access the same page more than once (\Rightarrow) as many page I/Os as matching rows \mathbb{Q}).

PostgreSQL addresses this challenge through RID sorting, implemented via Bitmap Index Scan & Bitmap Heap Scan.

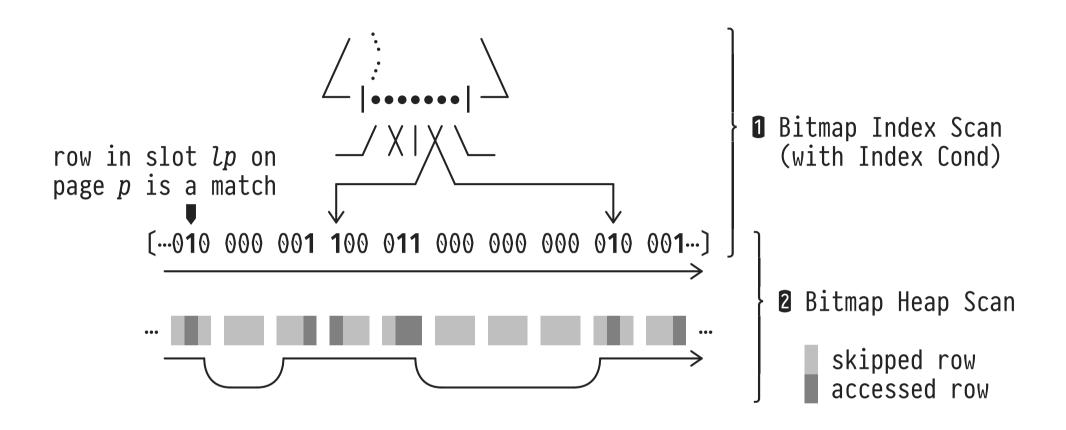




- **1 Bitmap Index Scan:** perform Index Scan and create **bitmap** that encodes *heap file locations* of rows matching the Index Cond. Do *not* access rows in heap file yet.
- Bitmap Heap Scan: scan heap file once, only access those rows (pages) that have been marked 1 in the bitmap.

Bitmap Index Scan & Bitmap Heap Scan: Row-Level Bitmap

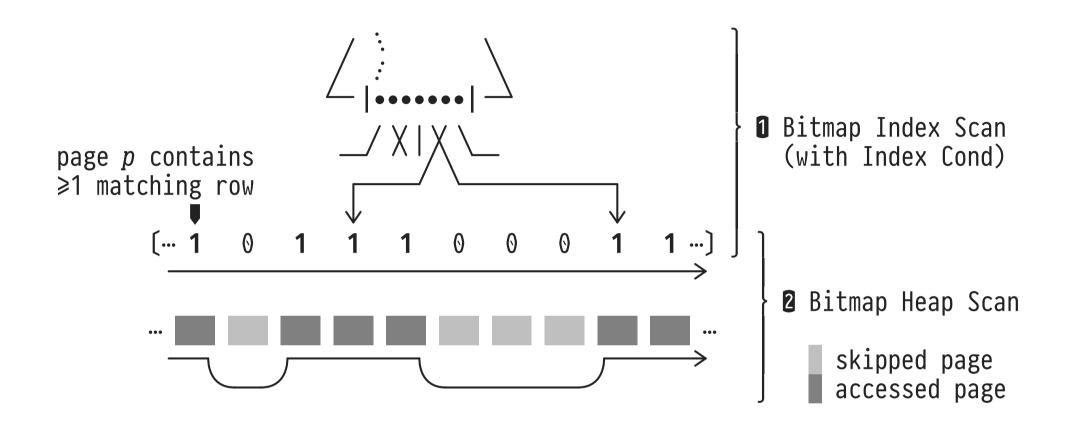




Bitmap Heap Scan performs one sequential scan (with skips) of the heap file, regardless of RID order in sequence set.

Bitmap Index Scan & Bitmap Heap Scan: Page-Level Bitmap





Working memory tight \Rightarrow build page-level bitmap. \triangle In \bigcirc , need to recheck condition for all rows on accessed pages.



If the workload depends on top performance of particular predicates supported by non-clustered index I, we may

physically reorder the rows of underlying table's T**heap file** to coincide with the key order in I's sequence set (i.e., *I* will become a *clustered* index⁴):

```
CLUSTER [VERBOSE] \langle T \rangle USING \langle I \rangle;
CLUSTER <T>; -- re-cluster once T's rows get out of order
```

 \bullet **!** Subsequent updates on T can destroy the perfect clustering. (May need to re-cluster T in intervals.)

⁴ At a price, of course: formerly *clustered* indexes on *T* will turn into *non-clustered* indexes.



B+Trees...

- 1. economically utilize space in inner/leaf nodes (minimum node occupancy 50%, typical fill factor 67%),
- 2. are balanced trees and thus require a predictable number of page I/Os to traverse from root to sequence set enables query optimizer to forecast B+Tree access cost.

DBMSs maintain properties 1. and 2. when rows are inserted into/deleted from an B+Tree-indexed table.⁵

 $^{^{5}}$ Some real B+Tree implementations of row deletion deviate from the textbook to keep things simpler.

B⁺Tree Insertion for New Entry $\langle k, rid \rangle$ (Sketch)



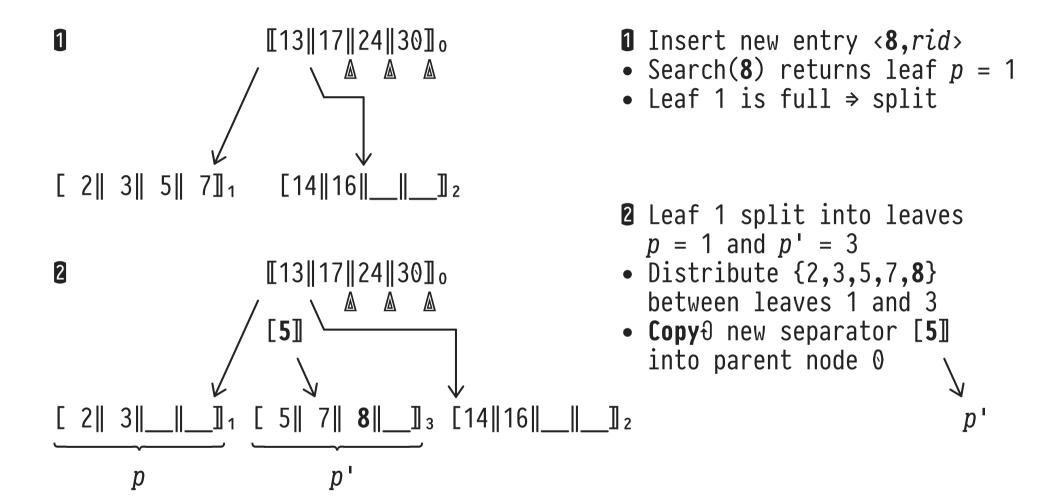
- 1. Use Search(k) to **find leaf page** p which should hold the entry for k.
- 2. If p has **enough space** to hold new entry (i.e., at most $2 \times o 1$ entries in p), **simply insert** $\langle k, rid \rangle$ into p.
- 3. Otherwise, node p must be **split** into p and p' and a new **separator** has to be inserted \heartsuit into the parent of p.

Splitting happens recursively ○ and may eventually lead to a split of the root node (increasing B+Tree height).

 \circ **Distribute** the entries of p and new entry $\langle k, rid \rangle$ onto pages p and p'.

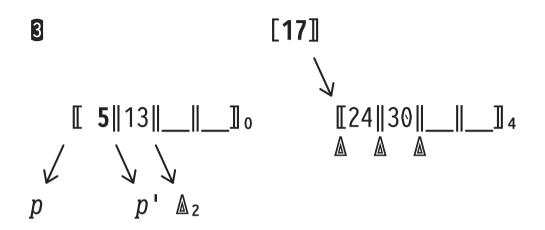
B*Tree Insertion and Leaf Node Split

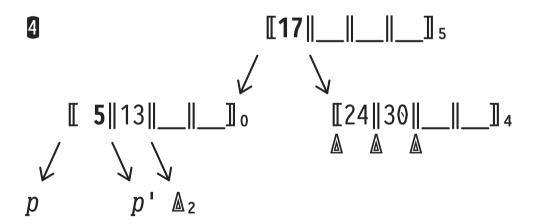




B*Tree Insertion and Inner Node Split







- Inner node 0 (here: root)
 is full ⇒ split
- Inner node 0 splits into old node 0 and new p'' = 4
- Distribute {5,13,24,30} A between nodes 0 and 4
- Move 1 new separator [17]
 into parent of
 node 0
- 4 Split node 0 has been the old root
- Create new root node 5, has [17] as only entry
- B+Tree height has increased

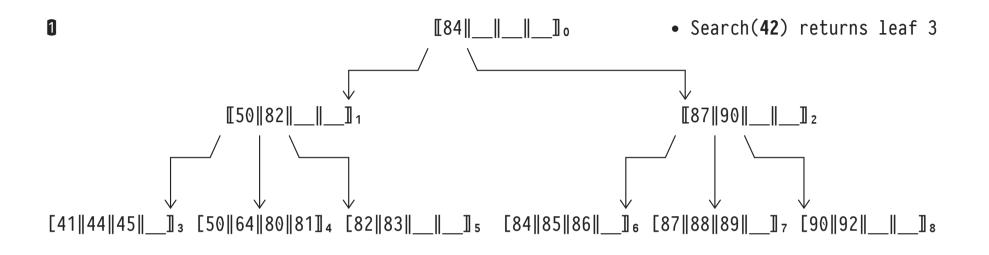


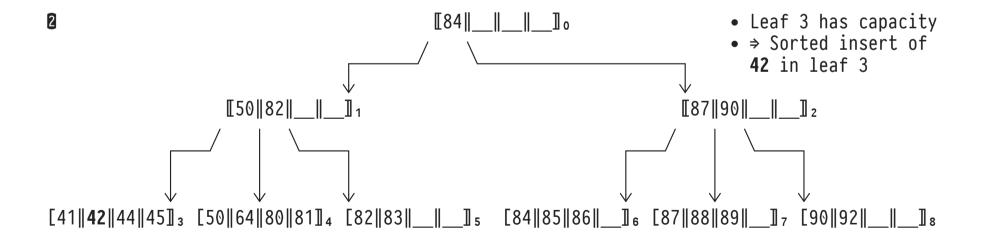
- Splitting starts at the leaf level and continues upward as long as inner index nodes are fully occupied (holding 2×0 entries).
- In Unlike during a *leaf* split, an *inner* node split **moves** the new separator [**sep**] discriminating between p and p' upwards and recursively inserts it into the parent. **Q:** Why?
- Q: How often do you expect a root node split to happen?

⁶ A leaf node split **copies** the new separator upwards, i.e., the entry [sep] also remains at the leaf level.

B+Tree Insertion Example: Insert <42, rid>

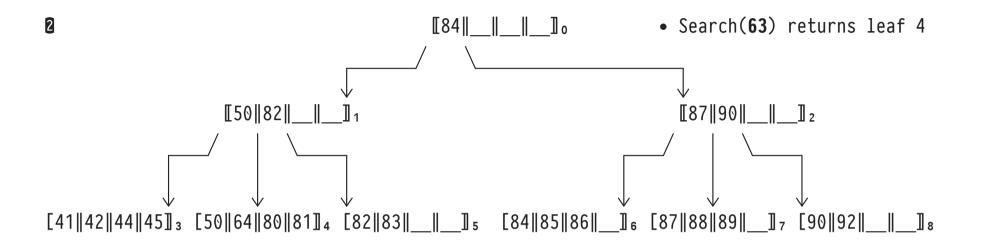


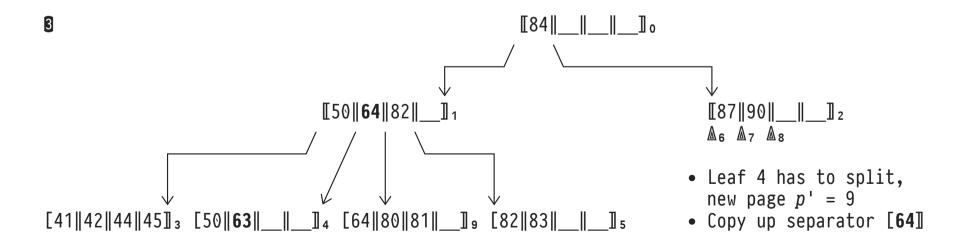




B*Tree Insertion Example: Insert <63, rid>

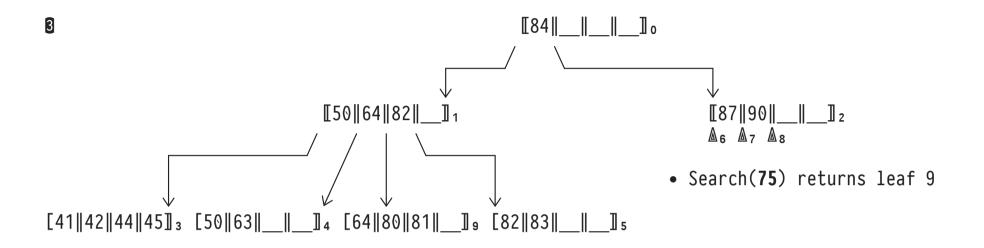


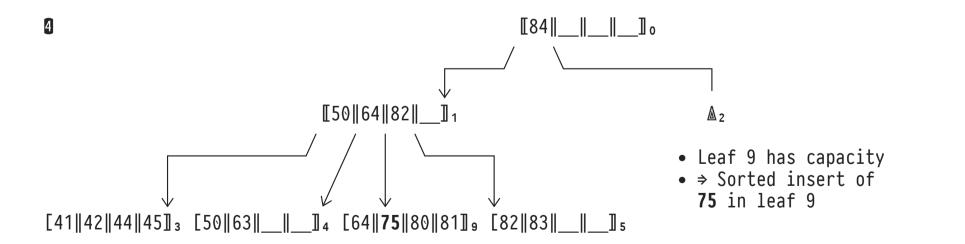




B*Tree Insertion Example: Insert <75, rid>

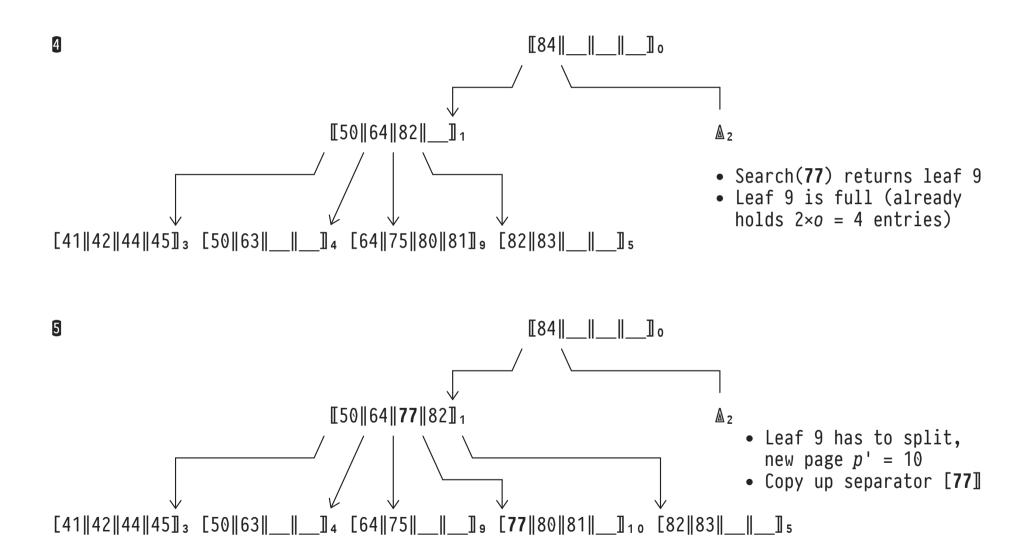






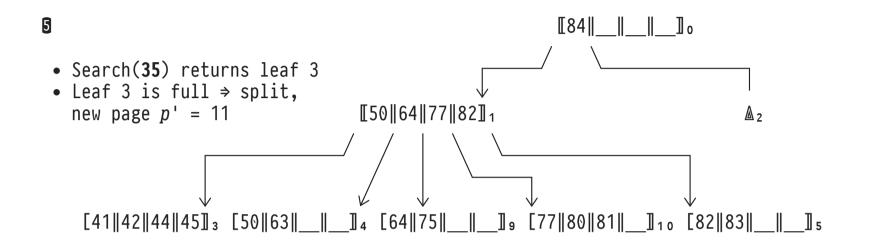
B*Tree Insertion Example: Insert <77, rid>

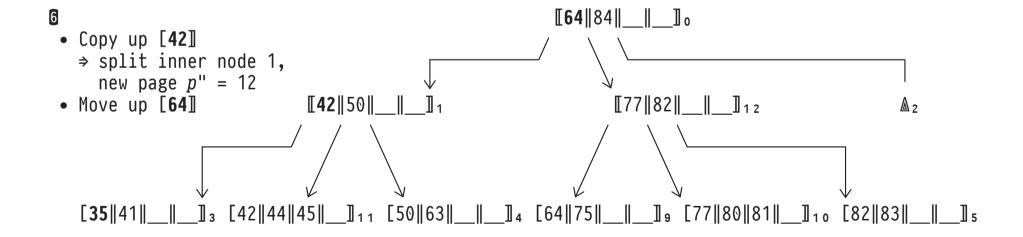




B*Tree Insertion Example: Insert <35, rid>









```
TreeInsert(<k,rid>,node):
  if (node is a leaf)
      return LeafInsert(<k,rid>,node);
  else
      switch k
          case k < k_1
           | \langle sep, ptr \rangle \leftarrow TreeInsert(\langle k, rid \rangle, p_0);
          case k_i \leq k < k_{i+1}
                                                                 see Search()
           [ <sep,ptr> + TreeInsert(<k,rid>,pi);
          case k_{20} \leq k
           \langle sep,ptr \rangle \leftarrow TreeInsert(\langle k,rid \rangle,p_{2o});
      if (sep = 1)
                                                                 upwards split?
          return <1,1>;
          return InnerInsert(<sep,ptr>,node);
                                                                 ⇒ ves
```

⁷ Note: $\langle sep,ptr \rangle \equiv [sep]$ in our discussion above.

B⁺Tree Insertion Algorithm (2)



```
LeafInsert(\langle k, rid \rangle, node):

if (node has < 2×o entries)

insert \langle k, rid \rangle into node;

return \langle 1, 1 \rangle; \} \langle 1, _{-} \rangle = no upwards split required

else

[\langle k_1, rid_1 \rangle, ..., \langle k_{2 \circ + 1}, rid_{2 \circ + 1} \rangle] \leftarrow entries of node \cup \langle k, rid \rangle;

node \leftarrow [\langle k_1, rid_1 \rangle] ... \langle k_2, rid_2 \rangle] \rangle

p' \leftarrow [\langle k_0, rid_0 \rangle] ... \rangle | \langle k_0, rid_0 \rangle];

return \langle k_0, rid_0, rid_
```

• Copy upwards: entry $\langle k_{o+1}, rid_{o+1} \rangle$ remains in leaf p'.

B⁺Tree Insertion Algorithm (3)

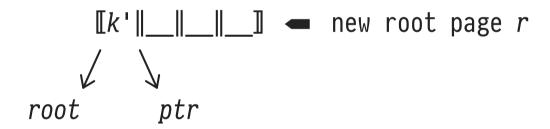


• **Move** upwards: new entry $\langle k_{o+1}, p' \rangle$ returned for insertion at parent. No entry $\langle k_{o+1}, _ \rangle$ remains at level of node/p'.



Insert(<k,rid>) is the top-level B+Tree insertion routine:

```
Insert(\langle k, rid \rangle):
\langle k', ptr \rangle \leftarrow \text{TreeInsert}(\langle k, rid \rangle, root); \} root \equiv \text{old root page}
if (k' \neq \bot)
r \leftarrow [root|k'|ptr|\_||\_||\_||; r \equiv \text{new root page}
root \leftarrow r
```

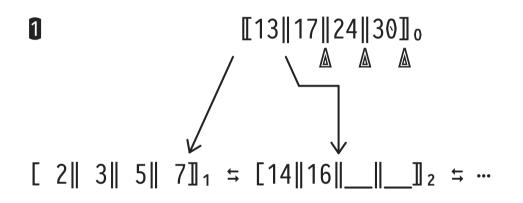


• Note: Insert() may leave us with a new root node that violates the minimum occupancy rule. 「\(ツ)/¬

B+Tree Insertion: Redistribution (1)



Can improve average occupancy and delay height increase on B+Tree insertion through **redistribution**:

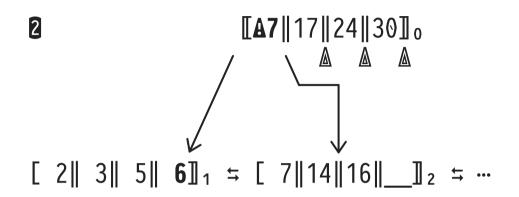


- 1 Insert new entry <6,rid>
- Search(6) returns leaf 1
- Leaf 1 is full, but its right **sibling** 2 has capacity

- Push entry from overflowing node to sibling and ! update separator in parent node to reflect this redistribution.

B⁺Tree Insertion: Redistribution (2)





- 2 Push entry <7, rid'> to leaf 2
- Place <**6**, rid> in leaf 1
- Update separator (13 → **7**) in parent node 0
- B+Tree remains at height 1

- ullet Inspecting node sibling involves additional page I/0. laphi
- Actual implementations use redistribution on the index leaf level only (if at all).

7 B+Tree Deletion of Entry With Key k (Sketch)



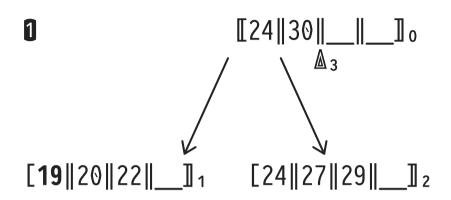
- 1. Use Search(k) to **find the leaf** p holding entry $\langle k, rid \rangle$.
- 2. Simply delete $\langle k, rid \rangle$ from $p.^8$
- 3. If *p* now holds < *o* entries, leaf *p* **underflows**. Any sibling of *p* with spare entries?
 - \circ Yes, use **redistribution** to move an entry into p.
 - \circ No, merge p and a sibling leaf p' of o entries. Delete \circ the now obsolete separator of p and p' in their parent node.

Deletion propagates upwards and may eventually leave the root node empty (decreases B+Tree height).

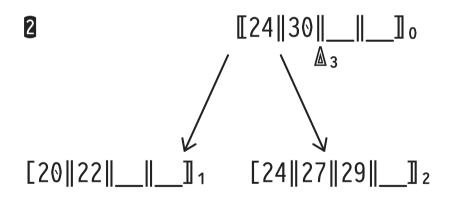
⁸ Q: If $\langle k, rid \rangle$ is the leftmost entry in p, do we need to update the associated separator entry in p's parent node? Why not?

B*Tree Deletion (No Underflow)





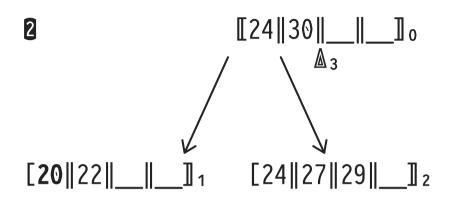
- 1 Delete entry with key k = 19
- Search(19) returns leaf 1
- Leaf 1 has > o entries, node will not underflow



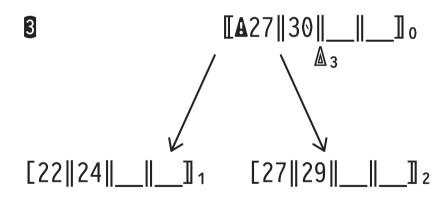
2 Simply delete entry <19,rid> from leaf 1

B+Tree Deletion and Redistribution





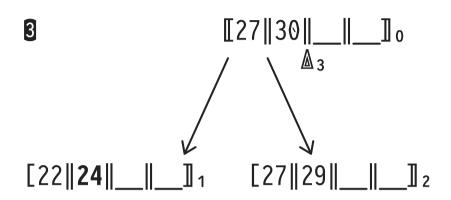
- **2** Delete entry with key k = 20
- Search(20) returns leaf 1
- Leaf 1 has minimum occupancy of o entries ⇒ will underflow



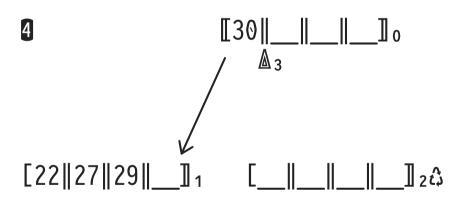
- Sibling p' = 2 has one entry to spare \Rightarrow redistribution
- Move entry <24,rid'> from leaf 2 to leaf 1
- Update separator (24 → 27)
 in parent node 0

B*Tree Deletion and Leaf Node Merging





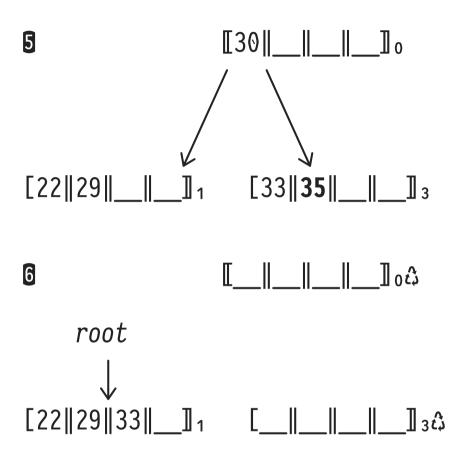
- **3** Delete entry with key k = 24
- Search(24) returns leaf 1
- Leaf 1 has minimum occupancy, no sibling with spare entries



- Merge leaf nodes 1 and 2,
 mark empty page 2 as garbage
- In parent 0, delete0 obsolete separator [27]

B*Tree Deletion and Leaf Node Merging (Empty Root)



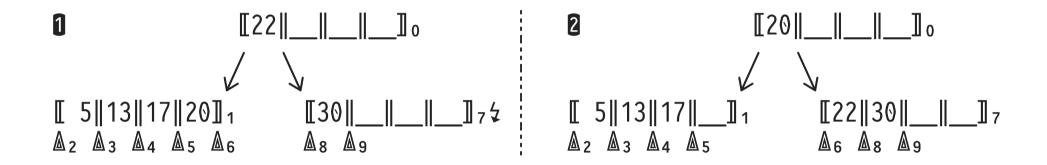


- **5** Delete entry with key k = 35
- Search(35) returns leaf 3
- Leaf 3 has minimum occupancy, no sibling with spare entries

- 6 Merge leaf nodes 1 and 3, mark empty page 3 as garbage
- In parent 0, delete0 obsolete separator [30]
- Old root empty (⇒ garbage), mark page 1 as the new root
- B+Tree height decreases



• **Redistribution** is also defined for **inner nodes**. Suppose we encounter underflow **①** during ♡ deletion propagation:

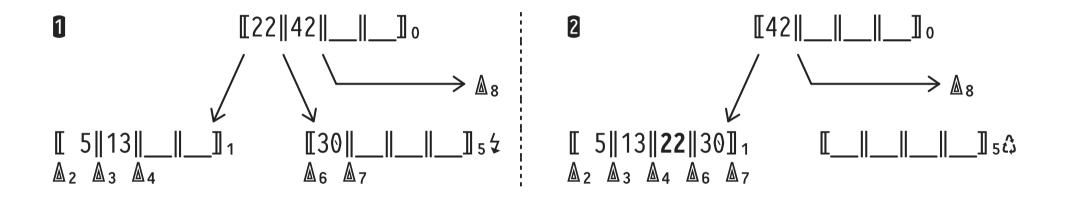


• Inner node 1 has two spare entries. "Rotate entry [20] through parent" to underflowed inner node 7.

NB: Semantics of subtree \triangle_6 (holds index entries with $k \ge 20 \land k < 22$) are preserved.



• Likewise, inner nodes may also be merged. The underflow in ① cannot be handled by redistribution:



• Note how the separator 22 has been **pulled down** \odot from the parent to discriminate between subtrees \triangle_4 and \triangle_6 :

 $\circ \Delta_4$: $k \ge 13 \land k < 22$

 \circ \triangle ₆: k ≥ 22 ∧ k < 30



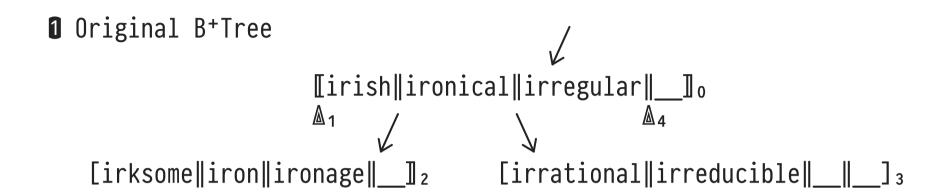
The higher the fan-out F, the more index entries fit in a B+Tree of fixed height. How to maximize F?

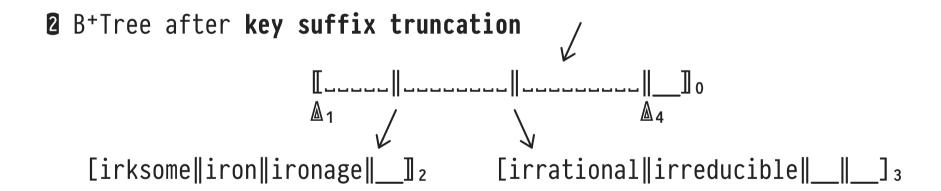
- For entries $\langle k,p \rangle$ in indexes over text/char columns, we may have $|k| \gg |p|$. Can we reduce the size of k?
- Search() and TreeInsert() do *not* inspect the actual key values but only use </≤ to direct tree traversals.
 - $\circ \Rightarrow$ May shorten (truncate) string keys as long as the ordering relation is preserved.
 - This applies to index entries in inner nodes only. Leaf level keys remain as is.

 $^{^9}$ The implementation (thus size) of page pointers p is prescribed by the DBMS. Nothing to win here.

B⁺Trees: Key Suffix Truncation







While truncating, preserve the separator semantics.

B⁺Trees: Key Prefix Compression



Observation: string keys within a B+Tree inner node often share a common prefix.

- $\$ Store common prefix only once (e.g., as " k_0 ").
- Violating the 50% occupancy rule can help compression.



Grab a hot cup of **P** and start a war on Stack Overflow: 10

Q: Which order of operations is better?

```
1 CREATE TABLE T (...);
2 INSERT INTO T VALUES (<5 \times 10^6 \text{ rows}>);
feats CREATE INDEX I ON T USING btree (...);
```

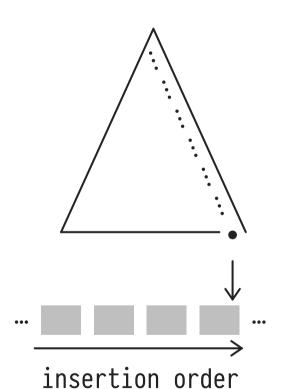
or

```
1 CREATE TABLE T (...);
3 CREATE INDEX I ON T USING btree (...);
2 INSERT INTO T VALUES (<5 \times 10^6 \ rows>);
```

¹⁰ See, for example, https://stackoverflow.com/questions/5910486/indexes-on-a-table-database



If insertions happen in index key order (i.e., ascending values of k), we observe a particular B+Tree access pattern:



- TreeInsert() will always traverse path :, will always hit the righmost leaf.
- ⇒ Fix rightmost leaf in buffer, insert next entry right there (no traversal from root). Node splits only occur along path :.
- We effectively create a clustered index.

heap file (sorted on keys k)





```
SELECT i.b, i.c

FROM indexed AS i

WHERE i.a = 42 [i.c = 0.42] -- either filter on i.a or i.c
```

Indexes in MonetDB play a secondary role and are *not* organized in tree shapes.

MMDBMSs try to exploit that data resides in directly-adressable memory and primarily aim to avoid access to separate index data structures (to avoid pointer chasing and potential cache misses).

Using EXPLAIN on Q_8 : Filter on Column a



 MonetDB uses algebra.thetaselect(..., 42:int, "==") to implement the predicate filter.

Using EXPLAIN on Q_8 : Filter on Column c^{11}



- Plan is nearly identical (modulo access to the a BAT).
- MonetDB appears to use the same algebra.thetaselect(..., 42:sht, "==") MAL operation.

Note how MonetDB maps the domain of type numeric(3,2) of column c, i.e., the set $N_{3,2} = \{-9.99, ..., 9.99\}$ with $|N_{3,2}| = 1999$, to a 16-bit value of type :sht. Nifty.



When MonetDB constructs a BAT t, a family of tail column **properties** prop(t) is derived/maintained: 12

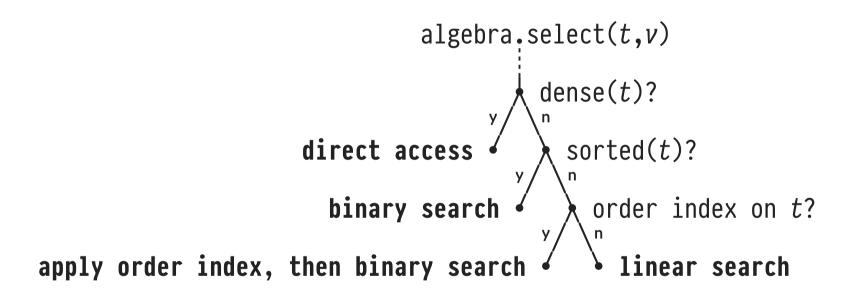
BAT Property prop(t)	Description
dense (tails of type :oid only)	ascending values, no gaps
key	unique values
sorted	ascending values
revsorted	descending values
nil/nonil	at least one/no nil value

- Use bat.info(t) to inspect current properties of t.
- Incomplete: t's tail may be sorted although sorted(t) = false (\Rightarrow but not \Leftrightarrow).

Additional properties *nokey*, *nosorted*, *norevsorted* give "proofs" (tail positions) why property does not hold. Example: nosorted = 3 = tail value for row 300 < tail value for row 200.



MAL operations inspect BAT properties at *query runtime*, select one of several efficient implementations:



• This is coined **tactical optimization** (as opposed to strategical query optimization at *query compile time*).

The Tactics of algebra.select: dense(t)



If input BAT t is **dense**, use **positional access** and **slicing** to evaluate equality and range selections:



head tail

000 3900
100 4000
200 4100
300 4200 --- offset 3 = 4200-3900
400 4300 --- hseqbase(t)

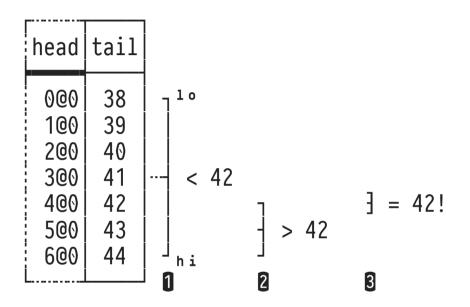
algebra.select(t,40@0,42@0,t,t,f)

head	tail] 			
100 200 300 400		‡ 	≡	algebi	a.slice(t,1,3)

The Tactics of algebra.select: sorted(t)



algebra.select(t,42)



Binary Search:

- Test middle value (pivot) between limits lo and hi
- Recurse into upper or lower partition based on test
- Finishes in $\log_2(|t|)$ steps

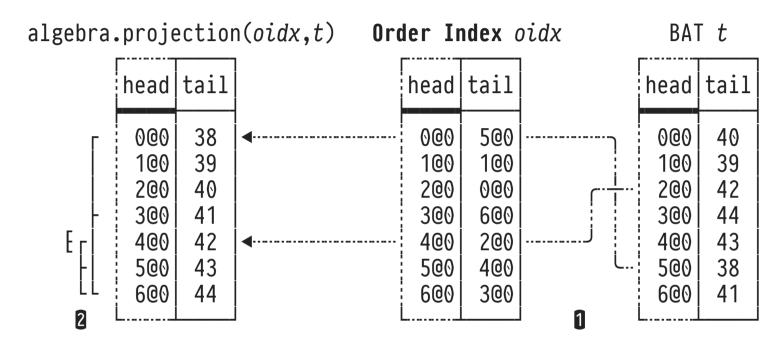
NB: Unpredictable branches (

≤ 42?) and jumps of pivot position less than ideal for CPU.

The Tactics of algebra.select: Order Indexes







• Row $[i@0,j@0] \in oidx$: value at offset j is ith largest in tail. Tactic: ① Apply oidx, ② then use binary search.

Creating Order Indexes (On the Fly)



MonetDB may automatically create a temporary order index to support predicates $lo \le a \le hi$ or other order-sensitive queries (e.g., ORDER BY, GROUP BY).

 Check current properties of column BATs and presence of indexes in MonetDB system table sys.storage:

<pre>sql> SELECT column, sorted, revsorted, "unique", orderidx FROM sys.storage('sys', 'indexed');</pre>										
column	•	revsorted	•	•						
a b c	true null false	1	true null null	0 0 0						

Creating Order Indexes (Manually)



If this seems beneficial for the query workload, clients may manually create an order index.

• ♪ Order indexes are **static** (i.e., not maintained under updates—costly) ⇒ underlying table must be *read-only*:

```
<create and populate table T>
sql> ALTER TABLE T SET READ ONLY;
sql> CREATE ORDERED INDEX I ON T(\alpha);
```

 \circ Order index I is made persistent (in a *.torderidx disk file) and will be used by future algebra.select()s on column α .



With **column cracking**, ¹³ MonetDB introduced a **self-organizing** (partially) ordered index structure.

- A cracker index for column α is created/updated as a byproduct of processing range predicates $lo \leq \alpha \leq hi$.
 - \circ In the cracker index, the α values \in [lo,hi] are stored physically contiguous (yet unordered, in general).
- If the query workload focuses only on a subset of column α , that part is indexed with fine granularity (while the other parts remain largely non-indexed).

¹³ "Database Cracking", S. Idreos, M. Kersten, S. Manegold. Proc. CIDR, Asilomar (CA, USA), 2007.

Column Cracking As a By-Product of Query Processing



1 BAT α		Γα	2 Cracker BAT (Ir			index)		3 Cr	acker	BAT	ex)				
	head	tail		head	tail						head	tail				
	0@0 1@0	17 3		0@0 1@0	4 3		≤ 5		S ₁		0@0 1@0	2		\leq	3	S 4
	2@0 3@0	8		2@0 3@0	2 6	=	> 5		S ₂		2@0 3@0	4 6		> >	_	S ₅
	4@0	2	\longrightarrow	4@0	8				3 2	\longrightarrow	400	8			J	36
	500	15	$Q_{\mathtt{i}}$	500	15			_		$Q_{\mathtt{j}}$	500	12		≥	10	S 7
	600	13 4		600 700	13 17		≥ 1	9	S ₃		600	13 17		≥	14	S o

•
$$Q_i$$
: ... WHERE $a > 5$ AND $a < 10$ Result: slice s_2

800

•
$$Q_i$$
: ... WHERE $a > 5$ AND $a < 10$ Result: slice s_2 Result: slices $s_5 + s_6 + s_7$

Column Cracking Notes



- $\forall x \in s_i$, $y \in s_{i+1}$: x < y: a fully cracked column ($\forall_i | s_i | = 1$) is completely ordered. This is uncommon (workload skew).
- First cracking step (①→②) copies source BAT. All further steps physically reorganize the cracker BAT.
- MonetDB implements slicing in terms of *views* ¹⁴ of the cracker BAT, no data copying involved. Cost free.
- Physical cracker index reorganization ("tail shuffling") can be efficiently performed *in-situ*.

¹⁴ A possible BAT view: (source BAT, first row, last row).



Reorganize column vector a[] between row offsets start and end, relocate its elements in-situ:

• \star Either a[start] \geq hi \wedge a[end] < hi or start = end.