The insides of a DBMS



How is a SQL query executed?

Parse, Check, Rewrite, Optimize









Organize tables and records as groups of pages in a *logical file*

Manage the transfer of pages into RAM to provide the *illusion* of operating in memory

(De)allocate, read, write *pages* on one or more storage device(s)



Client-Server

Layered

Designed for disks

• A SQL client for applications and a server DBMS

- *Manages complexity*: each layer abstracts and hides complexity from the layer above so it can focus on one thing!
- Disks (HDDs) are slow and mechanical.
- No byte-level addressing or pointer-dereferencing. Disk access is *block-level* or page-level.
- An API to read page from disk to memory or write a page from memory to disk.

The DBMS Architecture

Disk Space Management

Disk Characteristics

Time to retrieve a page depends on location: **seek** (1-10ms) + *rotational delays* (10ms).

Disk *bandwidth* is high: I00MB/s < page/Ims.

Arrange pages of a files *sequentially*! (next block on same track, same cylinder, adjacent cylinder).



L1, L2 Cache 1 ns 2 ns 10 ns 100 ns Registers L3 Cache

Main Memory

Dubai



This campus

┮═┱──® This room



Your brain

1min

10min

1.5 hours

10⁶ ns (1ms)

Pluto

Disk

2 years

10⁹ ns (1s) Tape

Andromeda

HOW FAR AWAY

IS YOUR DATA?

2000 years Jim Gray's Analogy Read and write large chunks of sequential bytes

Sequential Access is fast

Amortize read and write costs

- Reads and updates *occur at the block level* a block has multiple records of many fields
- Typical Block/Page size (unit of transfer to and from storage) nowadays is 64-128 KB (postgres 8KB)

- "Next" disk block is fastest (no seek delays)
- *Spatial locality*: place things that are accessed together close to each other spatially

- Predict future access patterns & Prefetch blocks
- Cache popular blocks
- Buffer writes to sequential blocks

Disk & Storage Hierarchy Implications

Disk Space Manager

allocate/de-allocate a (sequence of) page(s)
read/write a page

API

A single physical file can be stripped with pages across multiple devices each managed by a file system. The Disk Space manager provides the illusion of single big physical file with a unified getPage(). IMPLEMENTAION OPTIONS

Deal directly with the storage device

- Better control over where you place blocks and when you retrieve them
- What happens if the disk changes?

Run over the OS-provided File System

- Easier to implement & portable
- Allocate single large "contiguous" file; assume sequential access is fast. Most FS optimize disk layout for sequential access

Database Files

How are the contents of a DB file physically represented?

A database *file* is a collection of *pages*, each containing a collection of *records*.

What are different file organizations?

Unordered, Sorted, Indexed, ...

How to choose the appropriate file organization?

Cost models and cost analysis

Main Questions



Record Layouts

Record Layout

A sequence of bytes that are interpreted by the DBMS into attribute types and values.

DESIGN GOALS

- I. Fast access to fields
- 2. Compact representation in both memory and disk
- 3. Handling both fixed and variable-length fields



Fixed Length Attributes

The *catalog* stores the schema; data type gives you length of each attribute.

The order of fields is often the same order of the table definition

- Arithmetic is very fast!
- Byte representation on disk and in memory are identical
- Compact

Option I: Use Delimiters



- Requires a full record scan to access each field
- What if the fields contain the delimiter?

Record Layout

Variable Length Attributes

Option 2: Use an array of field offsets



- Direct access to fields
- No need to escape delimiters



Record Layout

Fixed + Variable Length Attributes

- Direct access to fixed-length fields
- Compact representation, only pointers for variable-length fields
- No need to escape delimiters

... But what about nulls?

Option I: Null bit map



Record Layout

Handling Nulls

Option 2: Pointers can handle this naturally



Record Layout

Header Attribute Data

What goes in the header?

Header contains meta data:

- Does **not** contain schema
- Bit map for NULL values for fixed-length attributes
- Visibility information (revisit later in concurrency control)

What if we can't fit a record in a page?

A postgres page ~8KB

- Generally, a tuple cannot exceed the size of a single page.
- Use separate overflow storage page and store pointer to overflow page for field value
- Store pointers to external files (no durability or transaction guarantees)

Page Layouts

Page Layout

A page has records of the same relation. Each record is uniquely identified by a *rid* DESIGN GOALS

- I. Fast access to records
 - By **rid** (page id, location in page)
- 2. Compact representation in both memory and disk
 - 3. Efficient handling of record deletes/inserts
 - 4. Handles fragmentation



Page Layout What goes in the header?

Header contains meta data:

- Number of records
- Free space
- Maybe a next/last pointer
- Bitmaps, Slot Table
- Page size
- Checksum
- DBMS version
- Visibility information
- Compression details



Option 1: Packed Fixed-Length Format

Pack records densely

```
rid = (page id, 'location in page'') =
(page id, slot #)
```

Insertions

• Just append

Deletions

- Re-arrange & update rids
- What about other files (e.g. indexes) that may reference rids?



Option 2: Unpacked Fixed-Length Format

Bitmap denotes "slots" with records

rid = (page id, slot #)

Insertions

• Find first empty slot

Deletions

• Clear bit



Each entry in the slot directory is a pointer to a record's beginning and its length.

Record id = page id, slot location in directory

Option 3: Slotted Pages



Set slot directory entry to null Doesn't impact other records

Delete a record



Place record in free space Reuse available free slots Update the free space pointer

Insert a record



Place record in free space Create new slot at the end & update slot count Update the free space pointer

Insert a record – grow slots



Pays off to allow some degree of fragmentation

Compact records on the page but you don't need to change a record's slot #!

Reorganize page

File Organizations

A Database (Logical) File

A collection of pages, each containing a collection of records. A file can be a table or an index.

Could span multiple physical OS files or even devices.

Supports an API for

- Insert, delete or modify record
- Fetch by record id (page #, slot#)
- Scan all records (with filter condition)

A Database File

Collection of records in *no* specific order

Can grow or shrink by allocating or deallocating page.

Supports record level operations by keeping track of:

- the pages in a file
- the free space on pages
- the records on a page

The *catalog* keeps track of a heap's header page ID and heap file name (i.e., the mapping from a table to the heap file)





A Heap File

To find a specific page, you must sequentially scan the lists.

Each page keeps track of the number of free slots in itself.



Option 1: Linked Listed Heap File

How do I find page 5? How do I find enough space for a record of length *x* bytes?



One more level of indirection: Use the page-directory to find the location of page p.

Each directory entry also includes additional information like # of free bytes/slots in a page.

Cache page directory!

Option 2: Heap File Page Directory



All together now

Unordered Heap Files Records placed arbitrarily across pages

Clustered Heap Files Records and pages are grouped

Sorted Files Pages and records are in sorted order

Index Files B+ Trees, Linear Hashing, ... May contain records or *point to records in other files*

Other Organizations

How to choose the best file organization?

Identify your access patterns/workloads

Create a model to quantify your trade-offs

Is it a read-mostly workload, or write-heavy? Are there many equality-searches? range-searches? full scans?

- Estimate in a principled way the costs.
 Crude & insightful (not complex & perfect)
- Identify assumptions upfront

How to choose the best file organization?

Identify your access		Heap File	Sorted
patterns/workloads	Scan		
	Equality-search		
	Range-search		
Create a model to quantify your trade-offs	Single Record Insert		
	Single Record Delete		

Heap File vs. Sorted

Identify your access patterns/workloads

Create a model to quantify your trade-offs

Heap File vs. Sorted

Estimate in a principled way the costs. *Crude & insightful* (not complex & perfect)

- Time to read or write a block/page (T)
- Number of blocks/pages (B)
- Number of records per page (R)
- Conduct *average-case* analysis

Identify assumptions upfront

- Heap files append inserts
- Sorted files are packed (always compact after deletion)
- Sorted files are sorted by search key
- Ignore sequential vs. random IO, inmemory costs, ...

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search		
Range-search		
Single Record Insert		
Single Record Delete		

Cost of a scan

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search	Is it $B \times T$ also?	Does sorting help?
Range-search		
Single Record Insert		
Single Record Delete		

Cost of an equality-search

What is the probability P(i) that I find the record with the search key on page i?

How many pages will I visit on average to hit the search key?

Assume uniformly random keys
$$P(i) = \frac{1}{B}$$

If you start from beginning to end scanning all pages, at each page there is I/B chance that it has the search key, and you stop the scan.

 $\sum_{i=1}^{B} i \times P(i) = \sum_{i=1}^{B} i \frac{1}{B} \approx \frac{B}{2}$

Cost of an equality-search - Heap File





Cost of an equality-search - Sorted File

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search	$B/_2 \times T$	$\log_2 B \times T$
Range-search		
Single Record Insert		
Single Record Delete		

Cost of an equality-search

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search	$B/2 \times T$	$\log_2 B \times T$
Range-search	Do we find every value in the range?	Find smallest value & scan right
Single Record Insert		
Single Record Delete		

Cost of a range-search - key in [x, y]

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search	$B/_2 \times T$	$\log_2 B \times T$
Range-search	$B \times T$	$(\log_2 B + \# pages) \times T$
Single Record Insert		
Single Record Delete		

Cost of a range-search

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search	$B/_2 \times T$	$\log_2 B \times T$
Range-search	$B \times T$	$(\log_2 B + \# pages) \times T$
Single Record Insert	Append at the end	Find, insert, shift
Single Record Delete		

Cost of an insert

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search	$B/_2 \times T$	$\log_2 B \times T$
Range-search	$B \times T$	$(\log_2 B + \# pages) \times T$
Single Record Insert	2 <i>T</i>	$(\log_2 B + 2\frac{B}{2}) \times T$
Single Record Delete		

Cost of an insert

	Heap File	Sorted
 Scan	$B \times T$	$B \times T$
Equality-search	$B/_2 \times T$	$\log_2 B \times T$
Range-search	$B \times T$	$(\log_2 B + \# pages) \times T$
Single Record Insert	2 <i>T</i>	$(\log_2 B + B) \times T$
Single Record Delete	Find, delete, write	Find, delete, shift

Cost of a delete

	Heap File	Sorted
Scan	$B \times T$	$B \times T$
Equality-search	$B/_2 \times T$	$\log_2 B \times T$
Range-search	$B \times T$	$(\log_2 B + \# pages) \times T$
Single Record Insert	2 <i>T</i>	$(\log_2 B + B) \times T$
Single Record Delete	$(B/2 + 1) \times T$	$(\log_2 B + B) \times T$

Cost of a delete

	Heap File	Sorted	
Scan	$B \times T$	$B \times T$	
Equality-search	$B/_2 \times T$	$\log_2 B \times T$	
Range-search	$B \times T$	$(\log_2 B + \# pages) \times T$	
Single Record Insert	2 <i>T</i>	$(\log_2 B + B) \times T$	
Single Record Delete	$(B/2 + 1) \times T$	$(\log_2 B + B) \times T$	

A workload may have a different distribution of each of these tasks! What happens when we change some of our assumptions? We can do better with indexes

Heap File vs. Sorted

Row Stores vs. Column Stores

Row Stores



Header	employeeID	name	department	salary
Header	employeeID	name	department	salary
Header	employeeID	name	department	salary
Header	employeeID	name	department	salary

Column Stores

Header						Header				
name	name	name	name	name	employeeID	employeeID	employeeID	employeeID	employeeID	
name	name	name	name	name	employeeID	employeeID	employeeID	employeeID	employeeID	
name	name	name	name	name	employeeID	employeeID	employeeID	employeeID	employeeID	
name	name	name	name	name	employeeID	employeeID	employeeID	employeeID	employeeID	
Header						Header				
department	department	department	department	department	salary	salary	salary	salary	salary	
department	department	department	department	department	salary	salary	salary	salary	salary	
department	department	department	department	department	salary	salary	salary	salary	salary	
department	department	department	department	department	salary	salary	salary	salary	salary	

Header									
salary	salary	salary	salary	salary					
salary	salary	salary	salary	salary					
salary	salary	salary	salary	salary					
salary	salary	salary	salary	salary					

- Good performance for writes (inserts, updates, deletes) transactional workloads
- Good for queries that need the entire tuple.
- Not ideal for scanning large portions of only subsets of a table's attributes.

When to use Row-Stores?

- Ideal for analytical workloads where read-only queries perform large scans over a subset of the table's attributes.
- Better data compression
- Easy schema expansion: add a new column doesn't entail rewrite
- Same column can be replicated with different orders

• Slow for inserts, updates, deletes --- tuple stitching or joining overhead.

When to use Column-Stores?