CSE 250 Data Structures

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Lec 36: Memory Hierarchy

Announcements

- PA3 due last night (submissions close Tues @ 11:59PM)
- WA5 will be released today
- Course evaluation!
- This is the last week of recitations



LIES!

Lie #1: Accessing any element of an array of any length is O(1)

- This assumes the "RAM" model of computation
 - Simple, but not perfect
- Real-world hardware isn't this simple
 - Memory is hierarchical
 - Non-Uniform Memory Access (NUMA)

Lie #2: The constants don't matter...

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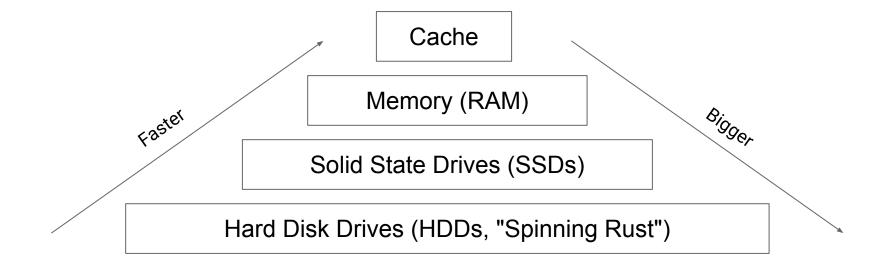
Memory Bounds (or Memory Complexity)

• The algorithm needs **O(...)** storage

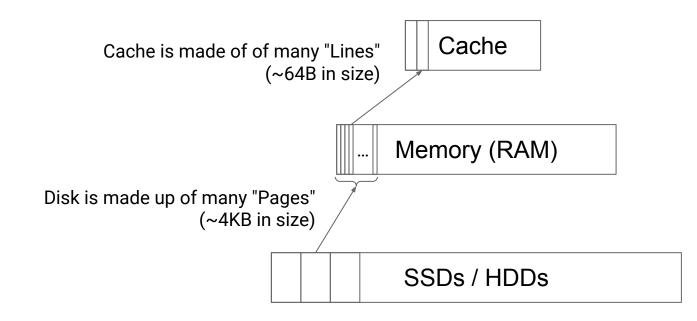
I/O Bounds (or I/O Complexity)

• The algorithm performs **O**(...) accesses to slower memory

The Memory Hierarchy (simplified)



The Memory Hierarchy (simplified)



In order to read an Array Entry:

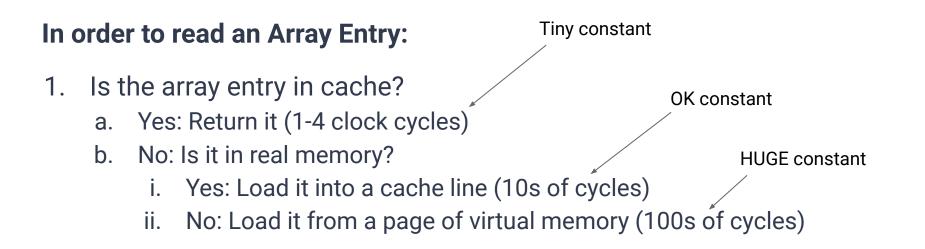
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 - i. Yes: Load it into a cache line (10s of cycles)
 - ii. No: Load it from a page of virtual memory (100s of cycles)



In practice, these constants do matter!

Ground Rules: Disk vs RAM

- 1. All data starts off in a file on disk
 - a. Need to load data into RAM before accessing it
 - b. Load data in 4KB pages
 - c. Amount of RAM is finite
- 2. Must describe 3 features of an algorithm
 - a. Number of instructions (runtime complexity)
 - b. Number of data loads (I/O complexity)
 - c. Number of pages of RAM required (memory complexity)

Note: Similar rules apply to any pair of levels in the hierarchy

Example:

2²⁰ records, 64 bytes each (8 byte key, 56 byte value)

64 MB of data total, 16,384 pages, 64 records per page

How many steps to binary search this data?

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How many steps to binary search this data? log(2²⁰) = 20 steps

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Let's assume the target is at position 0

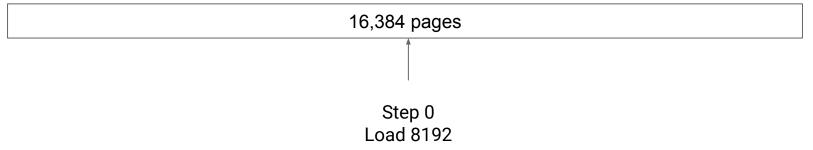
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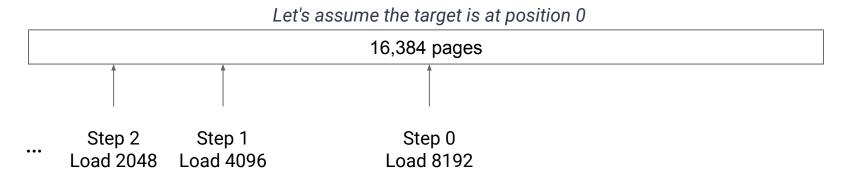
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Let's assume the target is at position 0 16,384 pages		
Step 1 Load 4096	Step 0 Load 8192	

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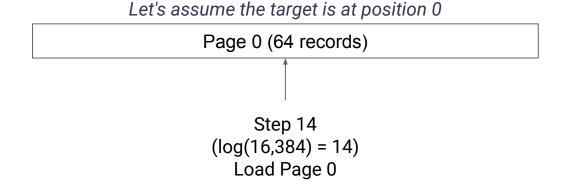
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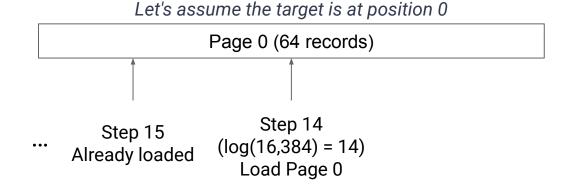
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What about I/O complexity?

Let's set up some variables:

- *n* total number of records
- **R** record size (in Bytes)
- **P** page size (in Bytes)
- **C** L**R/P** | records per page

Binary Search does log(*n*) steps broken into two stages:

Stage 1: Each request has to load a new page into memory

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Stage 1: Each request has to load a new page into memory

• log(n) - log(C) steps

Stage 2: The remaining requests all happen in the same page

• log(C) steps

Remember: Our page size is fixed...*C* records per page **Therefore:** The last **log(***C***)** binary search steps are all on the same page

Binary Search does O(log(n) - log(C)) loads from memory

Therefore: I/O complexity of Binary Search is log(n)

Binary Search Complexity:

- Runtime Complexity: **O(log(n))**
- Memory Complexity: **0(1)**
- I/O Complexity: **O(log(n))**

How can we improve on this?

Observations

Observation 1:

- Total size of records: 64MB = 2²⁰ x sizeof(key + data)
- Total size of keys only: 8MB = 2²⁰ x sizeof(key)

Observation 2:

- The first stage doesn't care what array index the record is at, just the page it is on
- Each page stores a contiguous range of keys...

Fence Pointers

Idea: Precompute the greatest key stored on each page

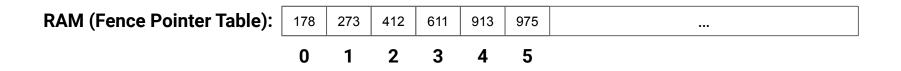
- *n* total records, *C* records per page, *n/C* keys required
- For our example, 2²⁰ records needs 2¹⁴ pages, therefore 2¹⁴ keys
 - 2²⁰ 64 byte records need 64MB memory
 - 2¹⁴ 8 byte keys only needs 512KB memory
- Call this a "Fence Pointer Table" and store it in memory

RAM: 2^{14} = 16,384 keys (Fence Pointer Table)

Disk: 16,384 pages (Actual Data)

Fence Pointer Example

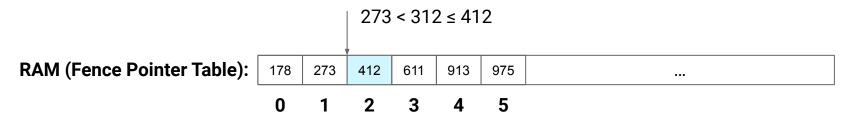
Binary Search for 321





Fence Pointer Example

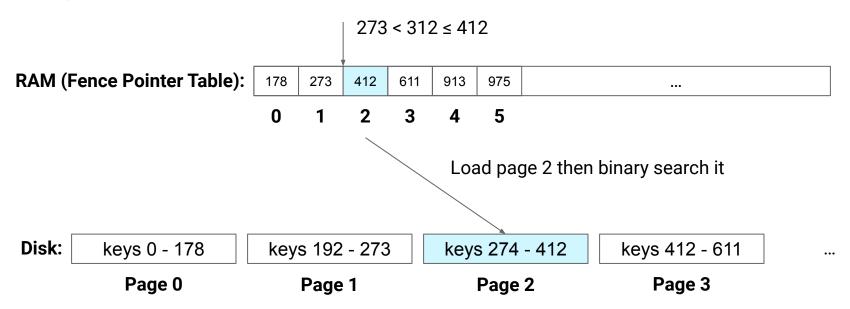
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Fence Pointer Example

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Binary Search with Fence Pointers

Step 1: Binary search the fence pointer table

- O(log(n) log(C)) steps
- All in memory, 0 disk reads

Step 2: Load page

• 1 step, 1 disk read

Step 3: Binary search within page

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Runtime: O(log(n))

I/O: O(1)

Memory?

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- O(log(C)) steps
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Runtime: O(log(n))

I/O: O(1)

Memory: O(n)

We need the entire fence pointer table in memory at all times :(

Records per page, C, is a constant, size of the fence pointer table is n / C

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I/O Com

 O(n) is not ideal... and what if the fence pointer table
 Load
 gets too big for memory?

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Idea: What if we binary search the fence pointers on disk?

With our current example:

- We can store 512 8B keys per 4KB page (2⁹ keys per page)
- 2^{20} records / 64 records per page = 2^{14} pages of records
- 2¹⁴ fence pointer keys = 2⁵ pages
- Binary search of the pointer key pages will require **log(2⁵) = 5 loads**

In general: log(n) - log(C) - log(keys/page)

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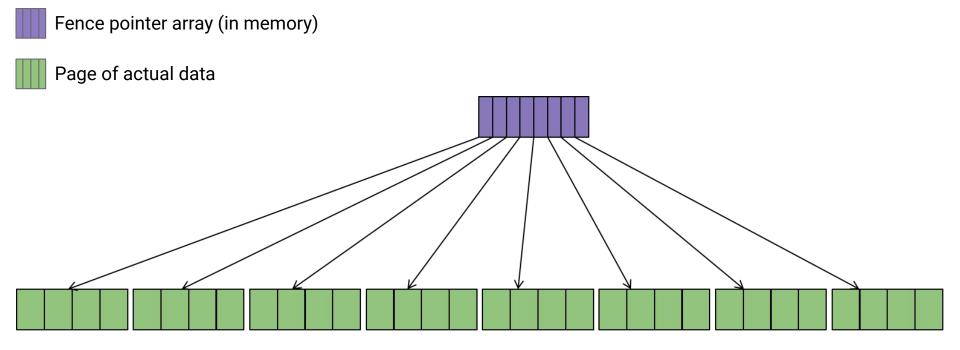
IO Complexity: $\log(n) - \log(C_{data}) - \log(C_{kev}) = O(\log(n))$

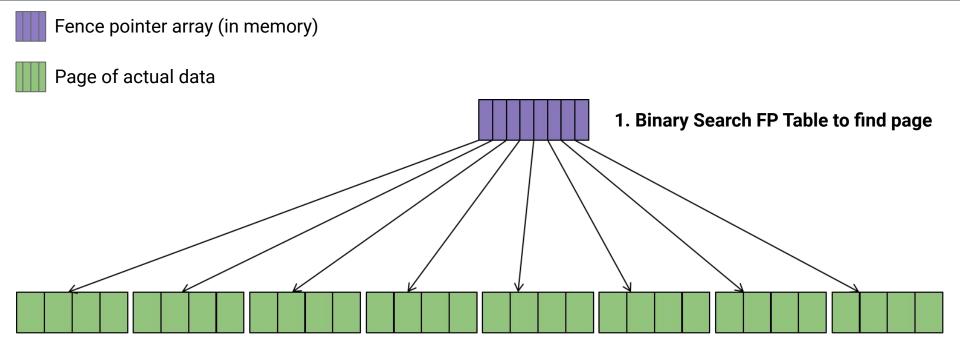
- C_{data} = records per page (ie: 64)
 C_{key} = keys per page (ie: 512)

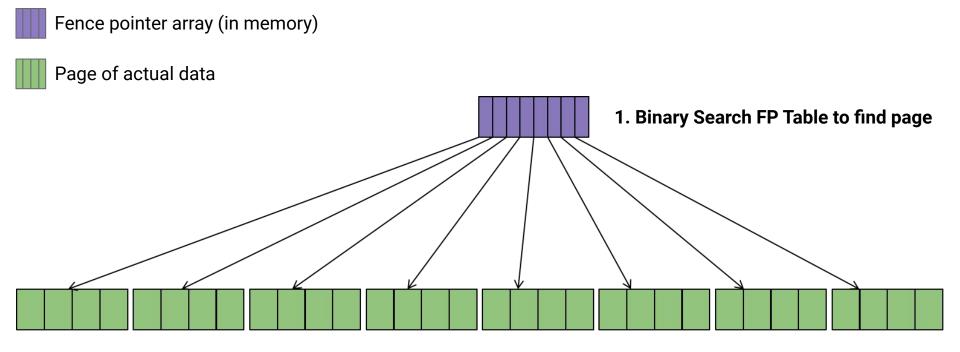
Can we improve our search of the on-disk Fence Pointer Table...?

Idea: A fence pointer table for our fence pointer table!

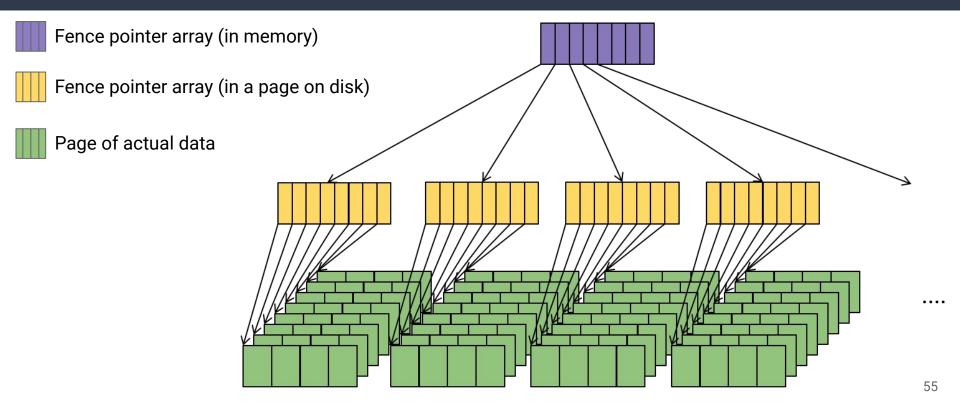
(and if that fence pointer table is too big...a fence pointer table for that table...and so on and so on and so on...until we have one that fits in memory)

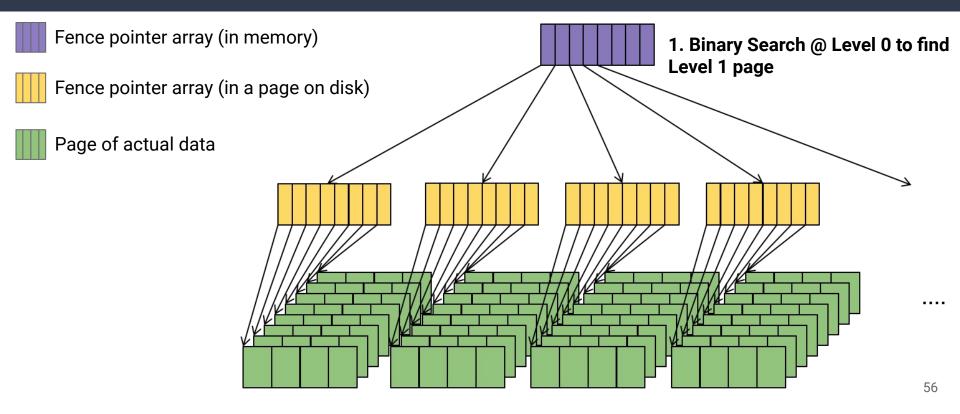


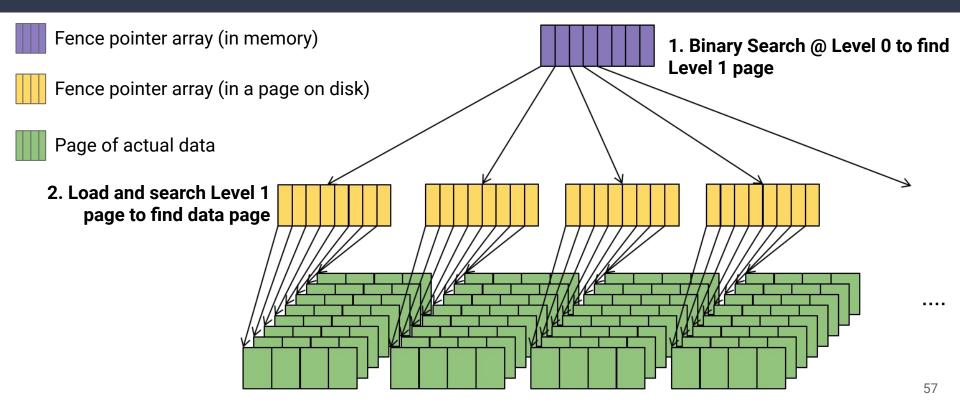


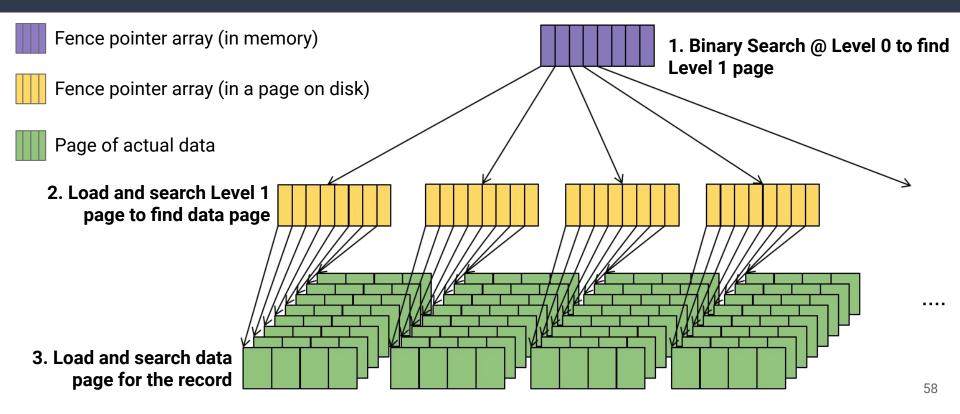


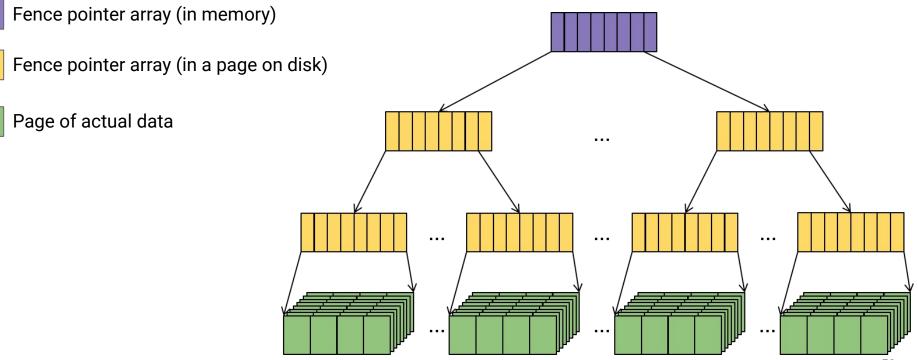
2. Load page and binary search for record

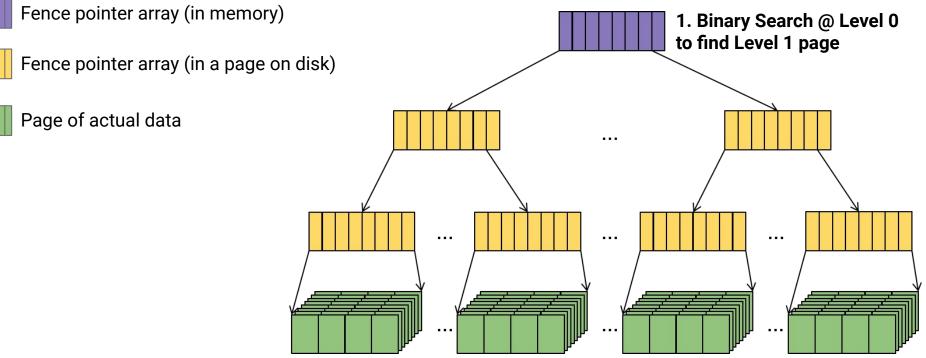


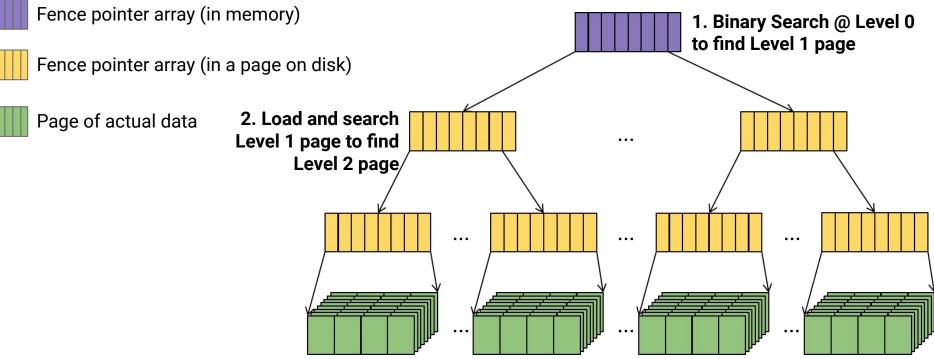


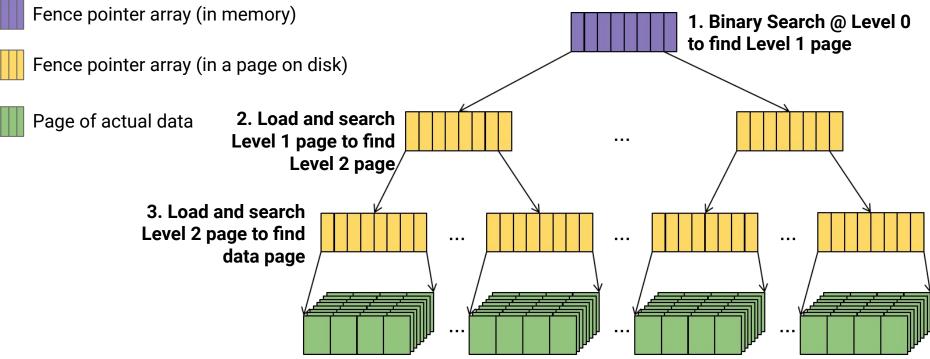


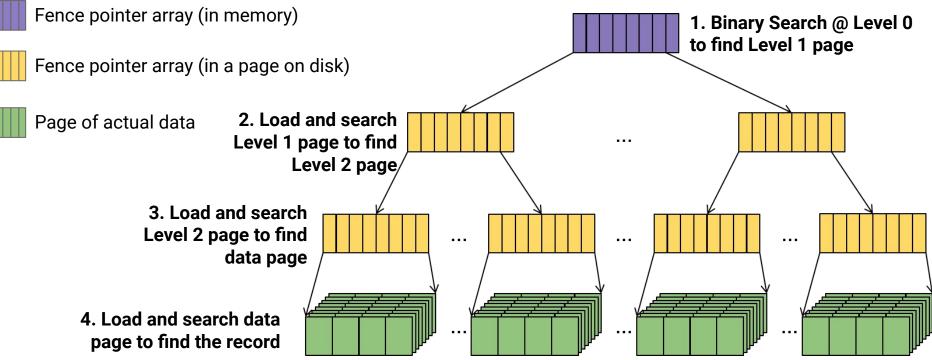




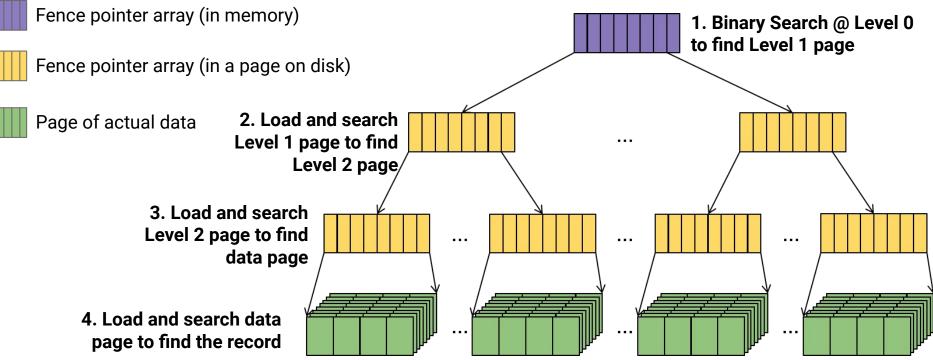








Improving on Fence Pointers ISAM Index



IO Complexity:

- 1 read at L0 (or assume already in memory)
- 1 read at L1
- 1 read at L2
- ...
- 1 read at L_{max}
- 1 read at data level

How many levels will there be (this isn't a binary tree...)

• Level 0: 1 page w/C_{key} keys

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- Level 1: Up to C_{key} pages w/ C_{key}^{2} keys

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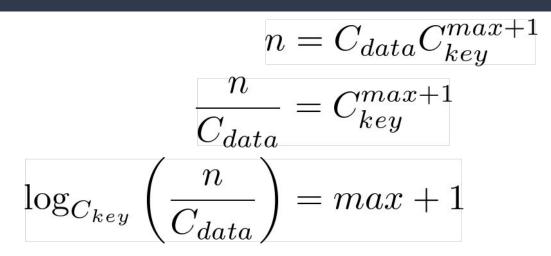
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- Level max: Up to C_{key}^{max} pages w/ C_{key}^{max+1} keys

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- Level max: Up to C_{kev}^{max} pages w/ C_{kev}^{max+1} keys
- Data Level: Up to C_{key}^{max+1} pages w/ $C_{data}^{max+1}C_{key}^{max+1}$ records

 $n = C_{data} C_{key}^{max+1}$

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n γ_{key}^{max+1} C_{data}



$$\begin{split} n &= C_{data} C_{key}^{max+1} \\ \frac{n}{C_{data}} &= C_{key}^{max+1} \\ \log_{C_{key}} \left(\frac{n}{C_{data}}\right) &= max+1 \\ \log_{C_{key}} (n) - \log_{C_{key}} (C_{data}) &= max+1 \end{split}$$

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Number of Levels: $O\left(\log_{C_{key}} (n) \right)$

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$$\begin{split} n &= C_{data}C_{key}^{max+1} \\ \hline n \\ \hline C_{data} &= C_{key}^{max+1} \\ \hline \log_{C_{key}}\left(\frac{n}{C_{data}}\right) &= max+1 \\ \hline \log_{C_{key}}(n) - \log_{C_{key}}(C_{data}) &= max+1 \\ \hline \text{Note this isn't base 2!} \\ \text{Number of Levels:} \quad O\left(\boxed{\log_{C_{key}}(n)}\right) \end{split}$$

Like Binary Search, but "Cache-Friendly"

- Still takes **O(log(n))** steps
- Still requires **O(1)** memory (1 page at a time)
- Now requires $\log_{Ckey}(n)$ loads from disk $(\log_{Ckey}(n) \ll \log_2(n))$

What if the data changes?