# Flash Memory & Data Structures: What they don't want you to know

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### Flash 101

aka Flash memory is the future.

- 1. Advantages: Increased Throughput, Increased Robustness, Reduced Latency
- 2. Characteristics: Erase before write, Outof-place update, Weariness, Garbage collection
- 3. Used as: Raw flash and SSDs in Cache, primary tier
- 4. Problems: Slow random writes, Expensive

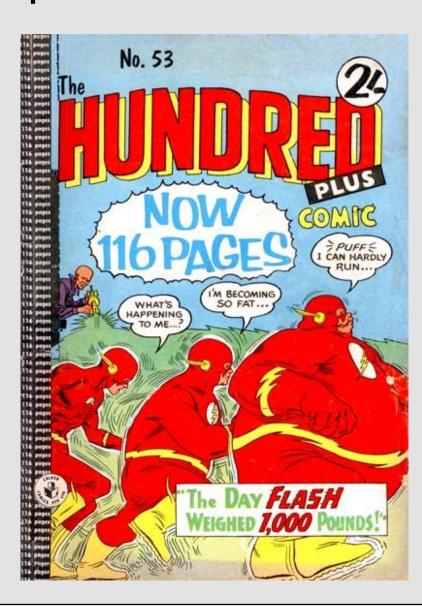


Figure 1. Flash getting fat [AusReprints.com]

# Indexing problem on flash memory

Standard B+-tree sucks: Writeamplification, faster usage of flash disk.

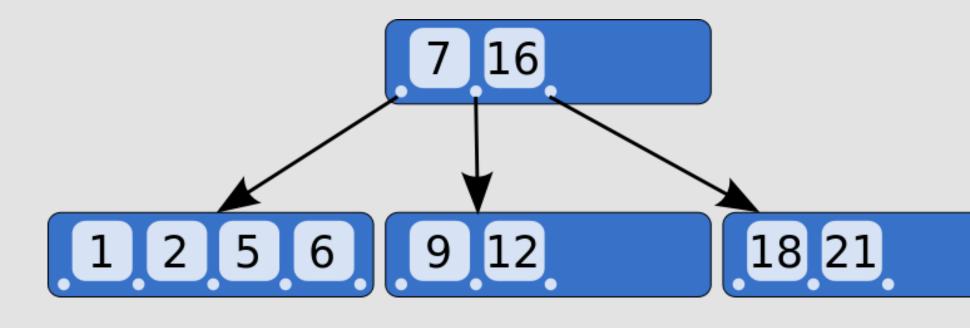


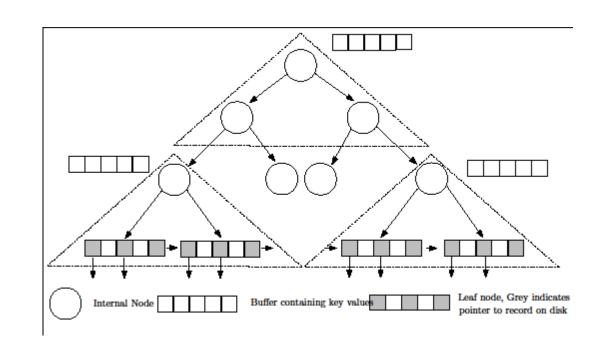
Figure 2. Sample B-tree [wikimedia.org]

How to construct and evaluate new data structures to choose the best?

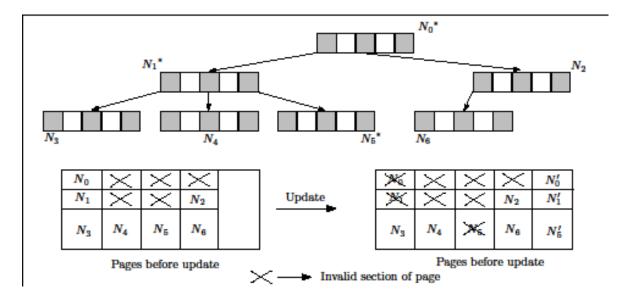
### **Cost analysis**

We theoretically evaluate the performance for insert, delete and modify operations.

Use general-cost flash memory model [2].



LA-tree [1]: Periodic buffering of I/O operations in the tree



μ-tree [4]: Variablesized nodes

FD-tree [5]: Converts random writes to

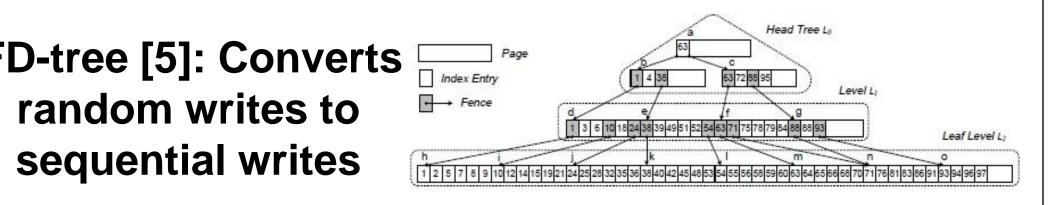


Figure 3. Proposed indexing structures

#### **Further trends:**

- . Block-aware layout
- 2. Self-tuning structure of the node
- 3. Intelligent FTL

# Comparison results

Table 1: Complexity of operations in various data structures in the general-cost model. (h = height of the tree)

| Data<br>Structure          | Search cost<br>(read I/Os)                | Update cost<br>(write I/Os)      | Reference |
|----------------------------|---|----------------------------------|-----------|
| B <sup>+</sup> -tree (FTL) | $O(log_{B_R} N)$                          | O(1)                             | [3]       |
| LA-tree                    | $O((1 + \frac{B_U}{k B_R}) \log_{B_R} N)$ | $O(\frac{\log_{BR}N}{k})$        | [1]       |
| FD-tree                    | $O(log_{B_R} N)$                          | $O(\frac{B_R}{B_W} log_{B_R} N)$ | [5]       |
| μ-tree                     | O(h)                                      | O(1)                             | [4]       |

N = Number of elements in tree  $B_R$  = Read block size  $B_W = Read block size$  $B_{II}$  = Size of buffer

#### Conclusions

- 1. FTL is useful but causes space and time overhead
- 2. Data structures like FD-tree can reduce write-amplification even without FTL, but are complex to implement
- 3. Simple workload-aware data structures (like FlashDB) can be very useful

#### Recommendations

- Need for developing more flash-aware algorithms
- Standardize performance claims over common workloads

# Literature cited

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