

Class 15

Data Systems on Modern Hardware: Multi-cores, Solid-State Drives, and Non-Volatile Memories

Dr. Subhadeep Sarkar

https://bu-disc.github.io/CS561/

Compute, Memory, and Storage Hierarchy

Traditional von-Neuman computer architecture

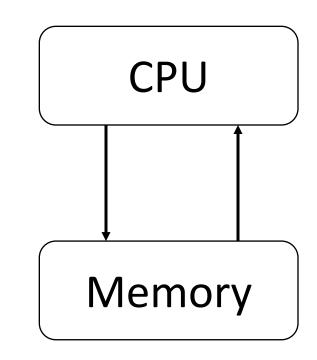
(i) assumes CPU is fast enough (for our applications)

(ii) assumes memory can keep-up with CPU and can hold all data

is this the case?



for (i): applications increasingly complex, higher CPU demand is the CPU going to be always fast enough?



Moore's law

Often expressed as: "X doubles every 18-24 months" where X is: "performance" CPU clock speed the number of transistors per chip

which one is it?



based on William Gropp's slides

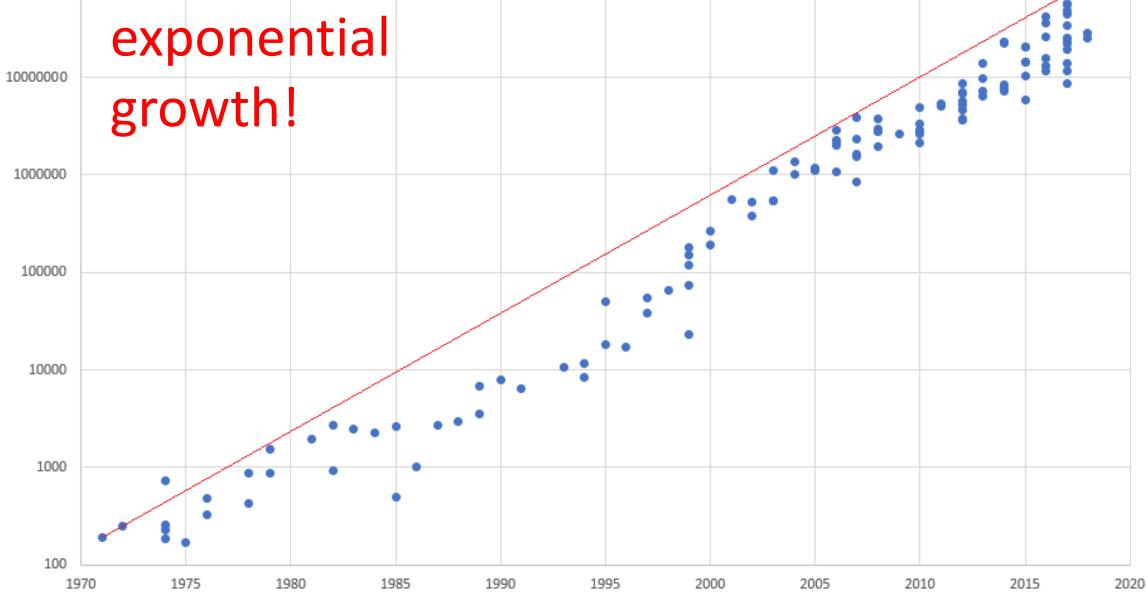
Moore's Law is Alive and Well! Transistors per Square Millimeter by Year

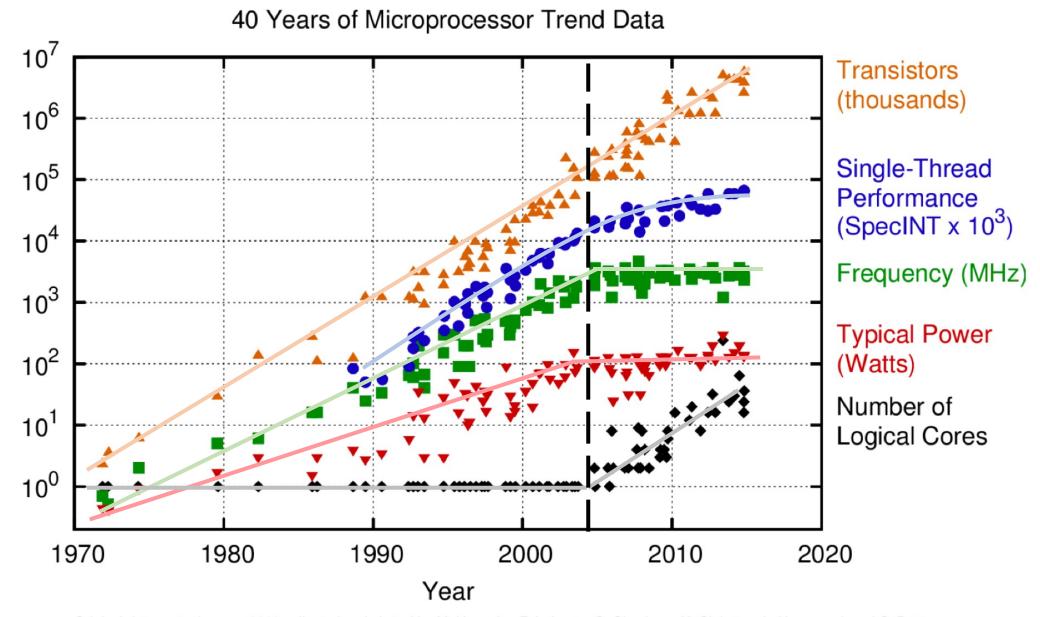
100000000

BC

UNI







Original data up to the year 2010 collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond, and C. Batten New plot and data collected for 2010-2015 by K. Rupp



Can (a single) CPU cope with increasing application complexity?

No, because CPUs (cores) are **not** getting faster!!!

.. but they are getting more and more (higher parallelism)

Research Challenges

how to handle them?

how to parallel program?



Compute, Memory, and Storage Hierarchy

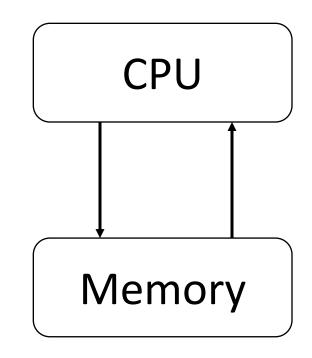
Traditional von-Neuman computer architecture

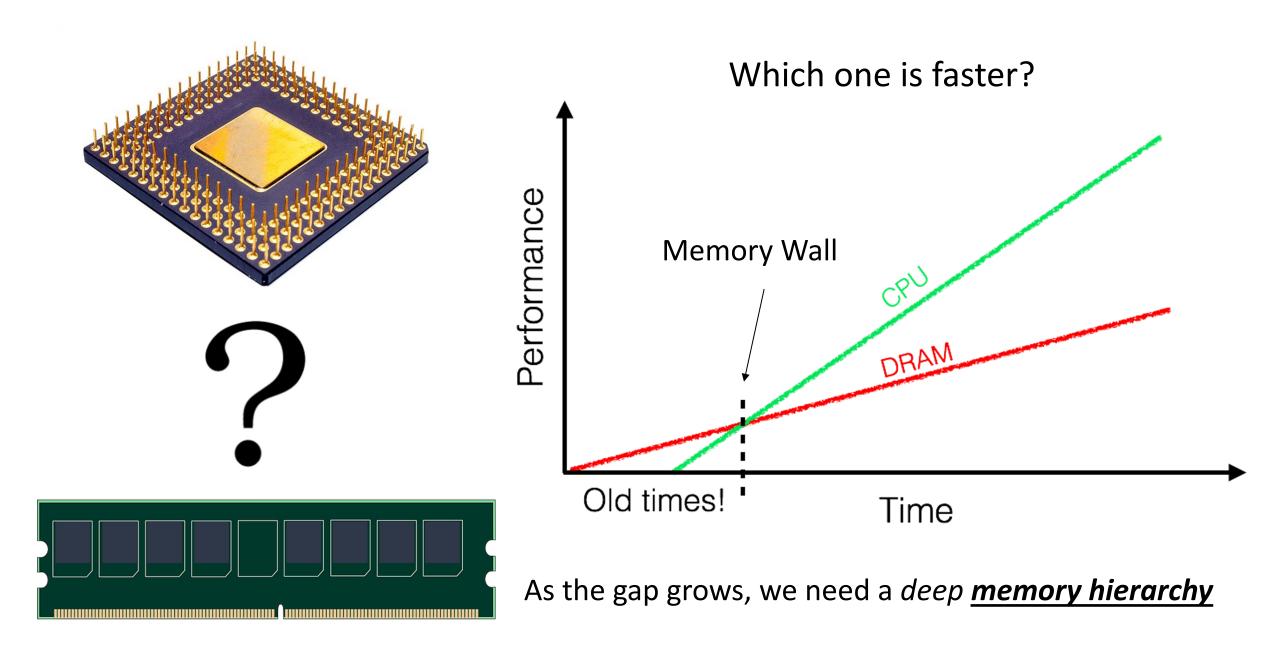
- (i) assumes CPU is fast enough (for our applications) *not always!*
- (ii) assumes memory can keep-up with CPU and can hold all data

is this the case?



for (ii): is memory faster than CPU (to deliver data in time)? does it have enough capacity?







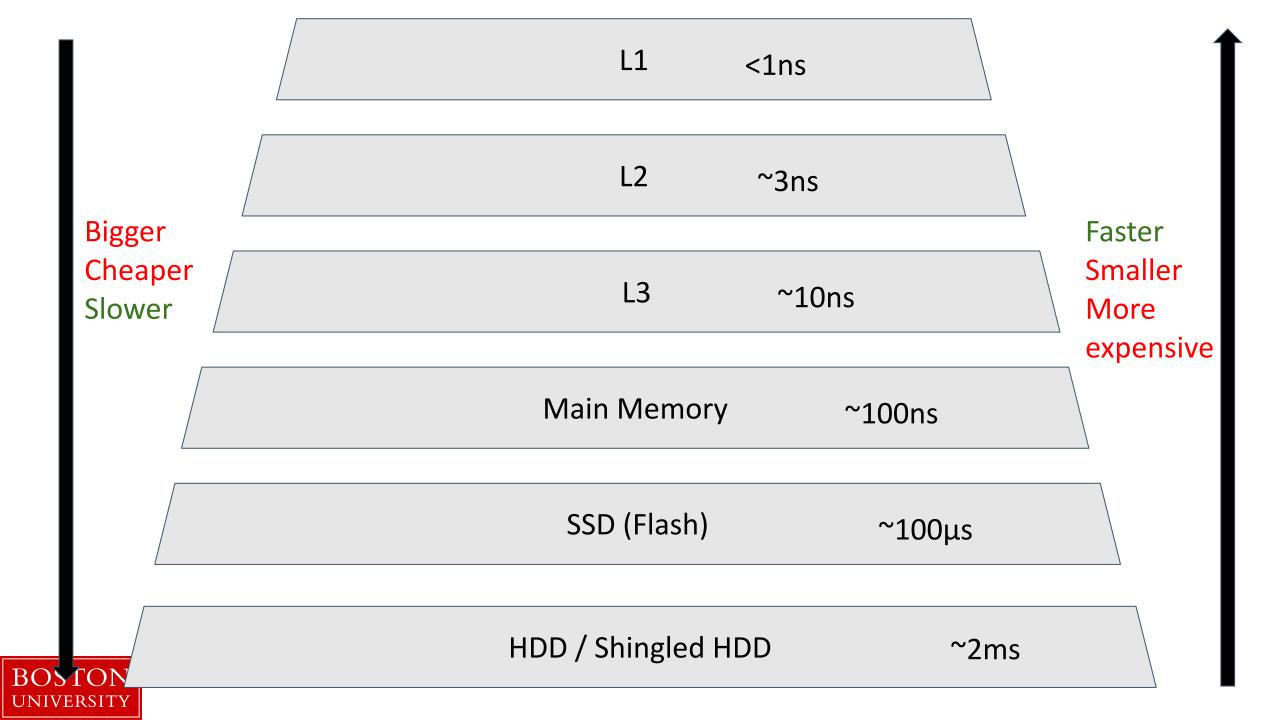
A single level of main memory is not enough

We need a *memory hierarchy*



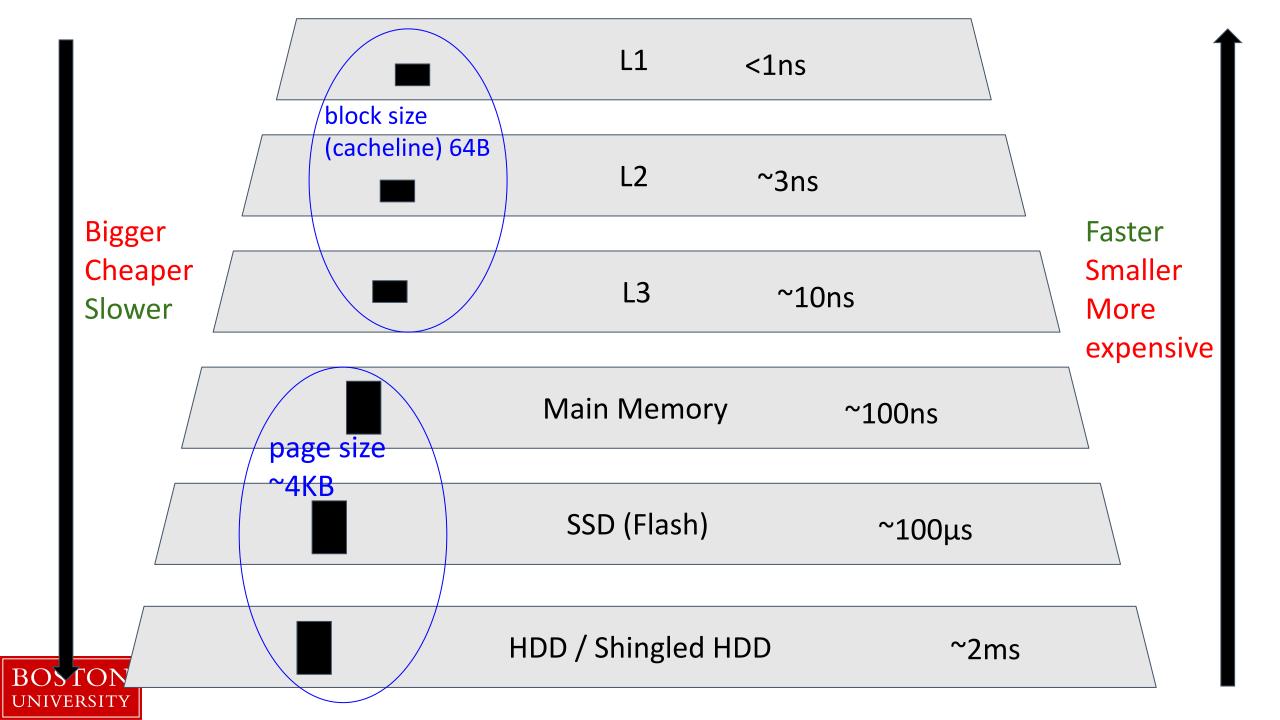
What is the memory hierarchy ?

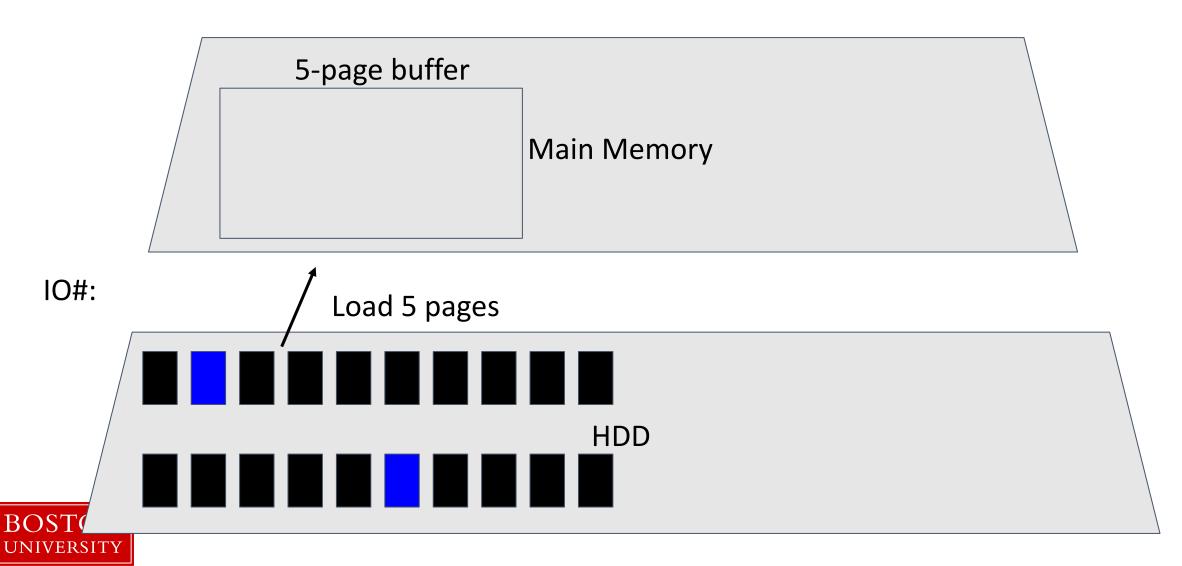


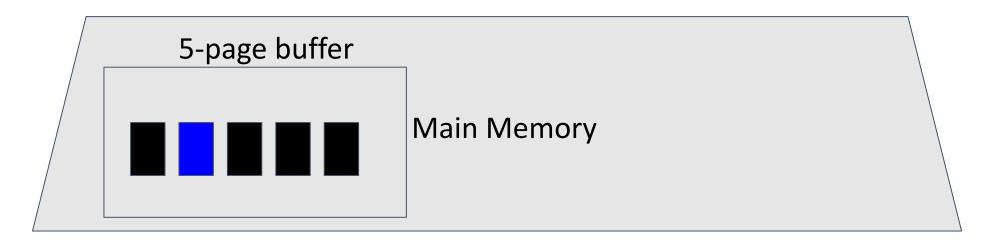


Access Granularity

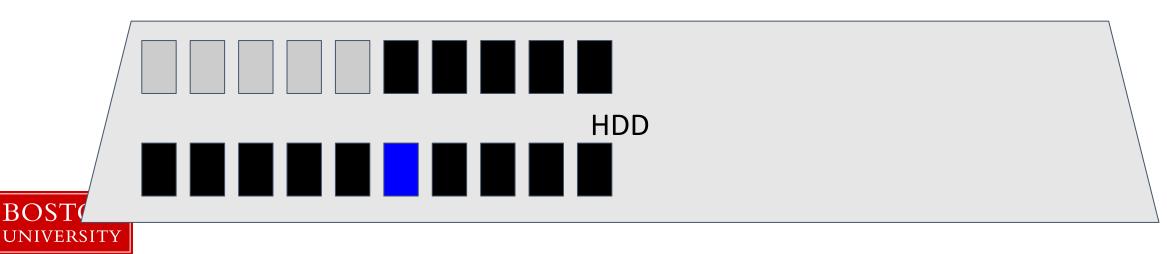


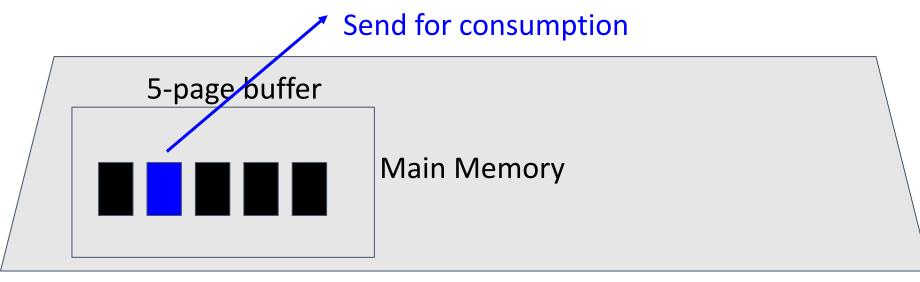




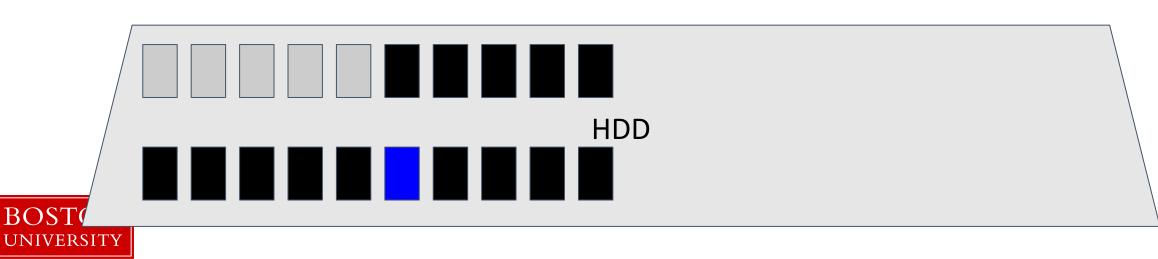


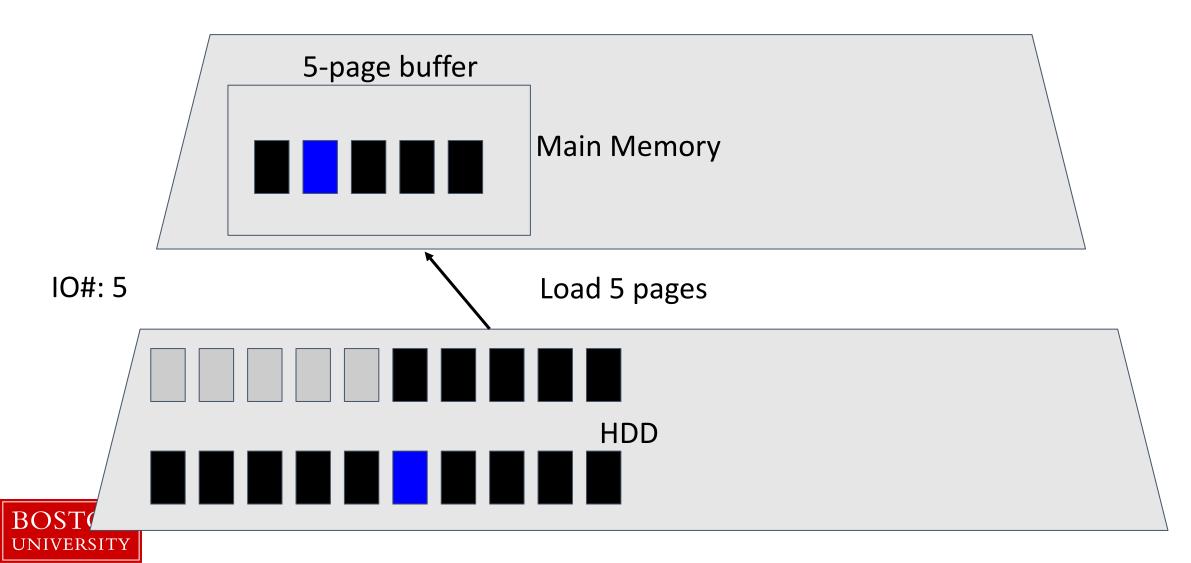
IO#: 5

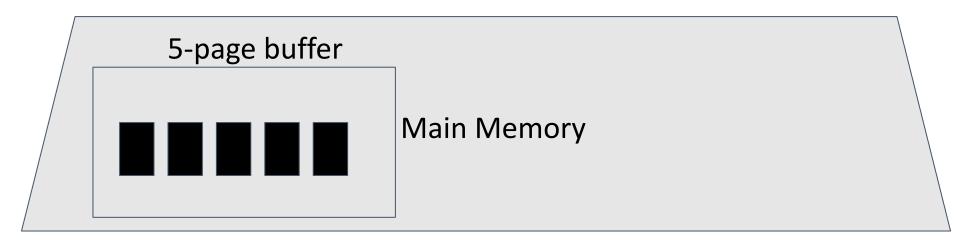


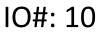


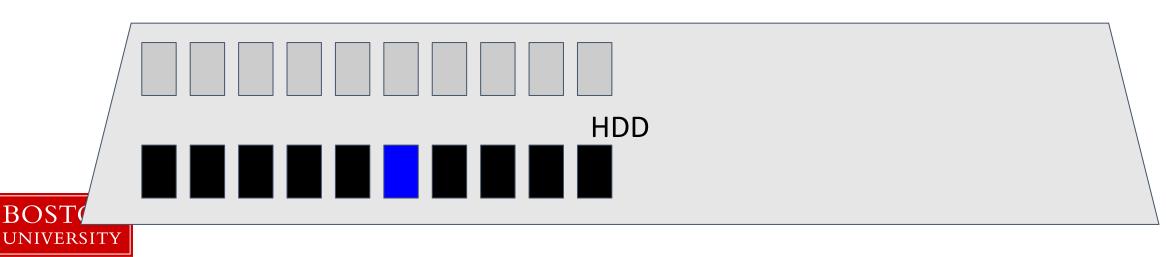


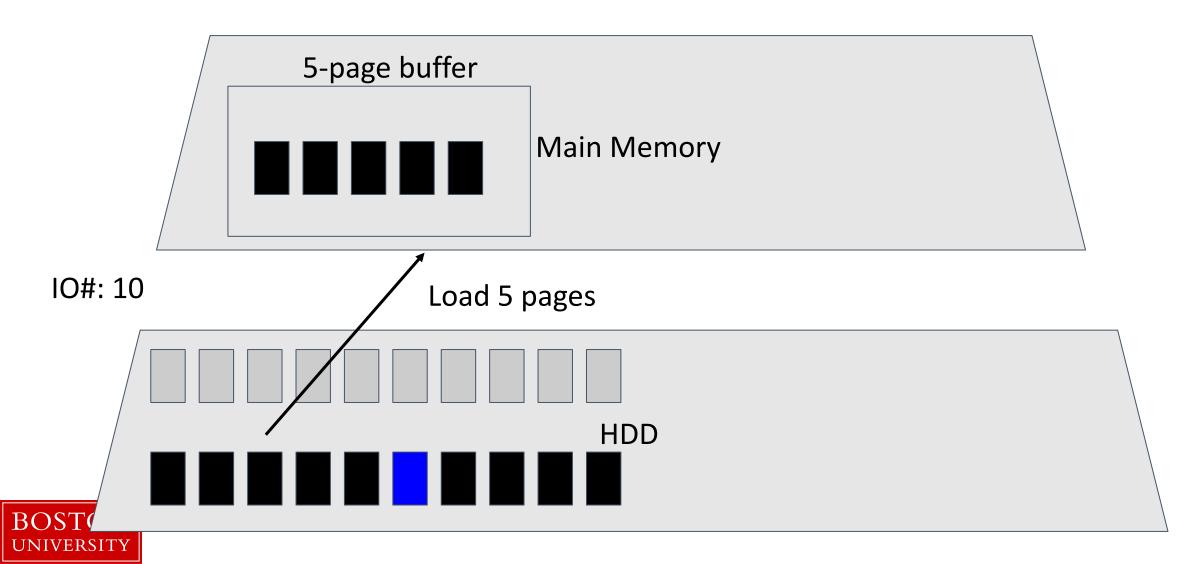


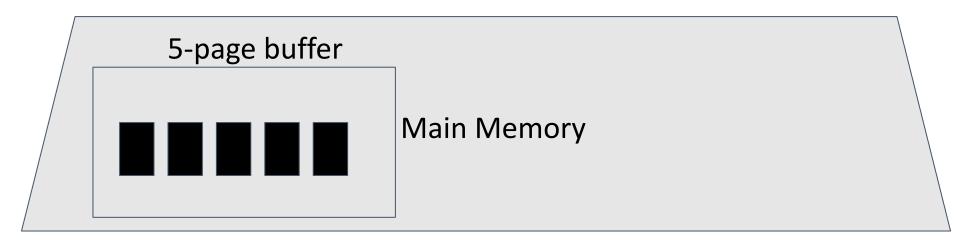




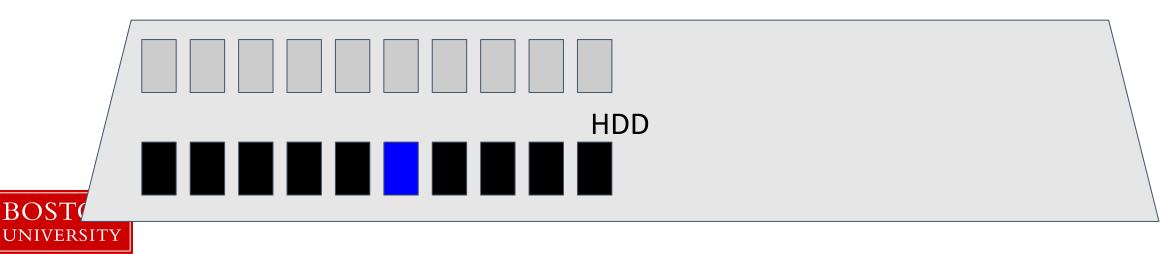


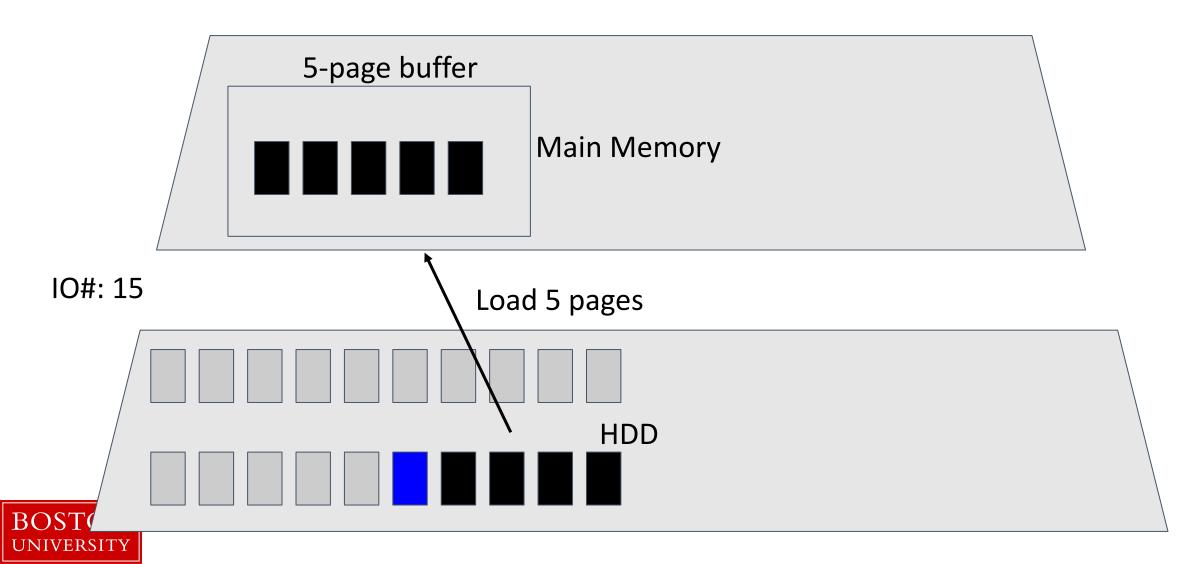


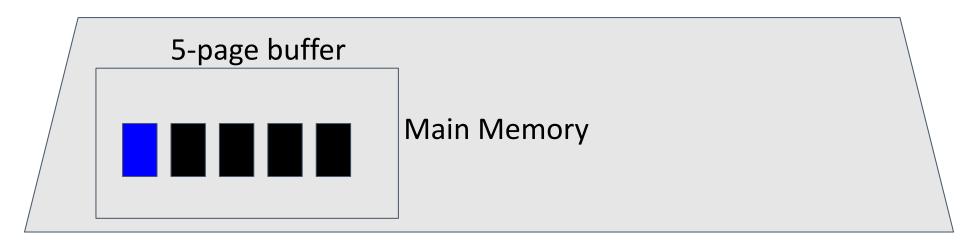


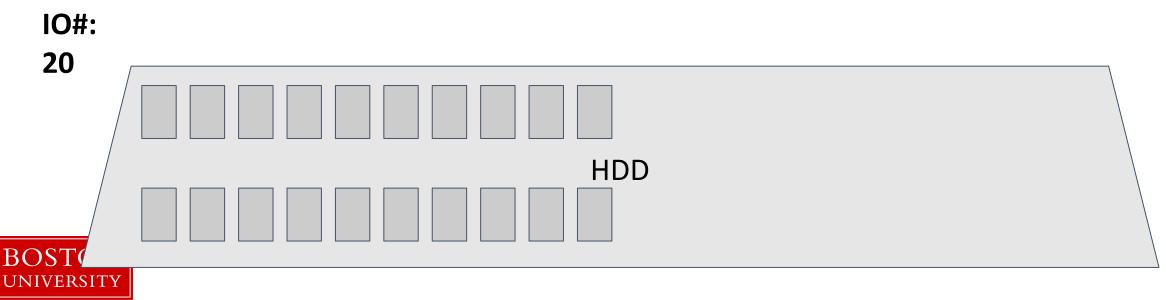


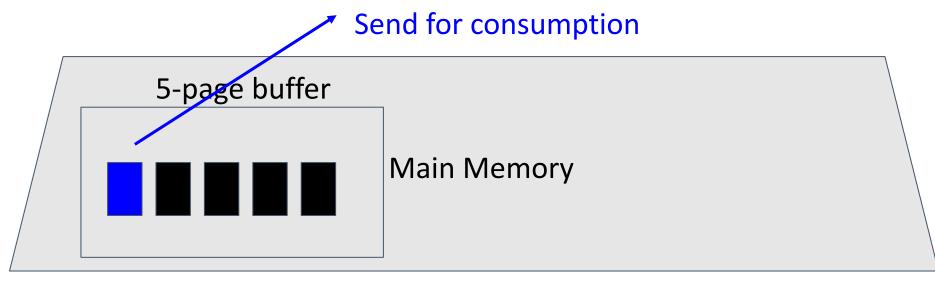
IO#: 15

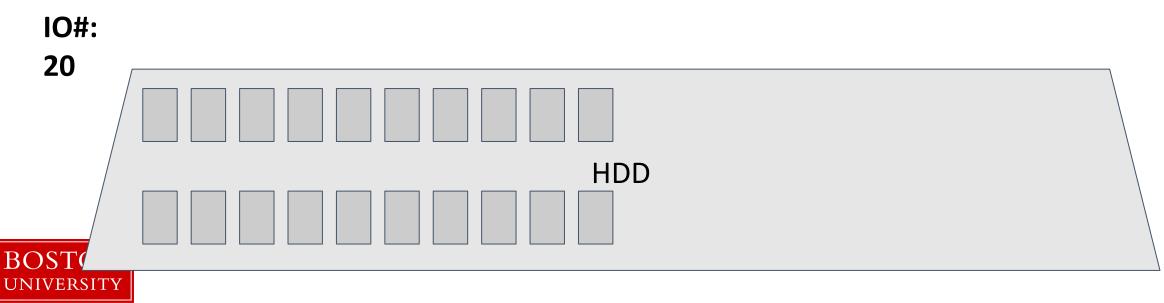






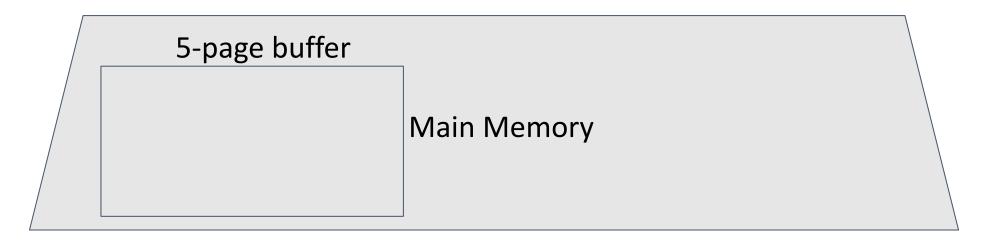




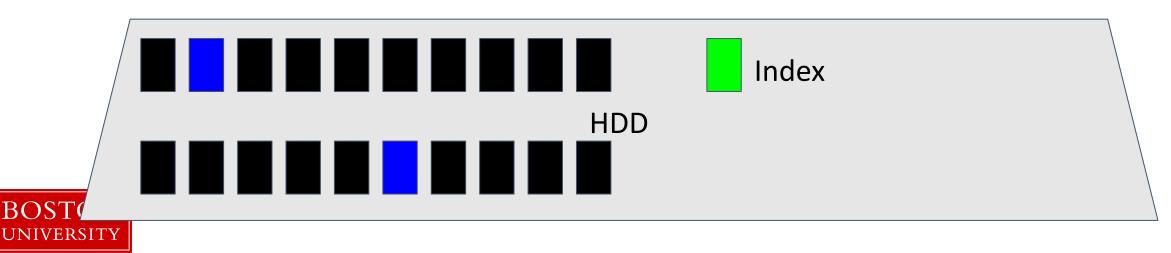


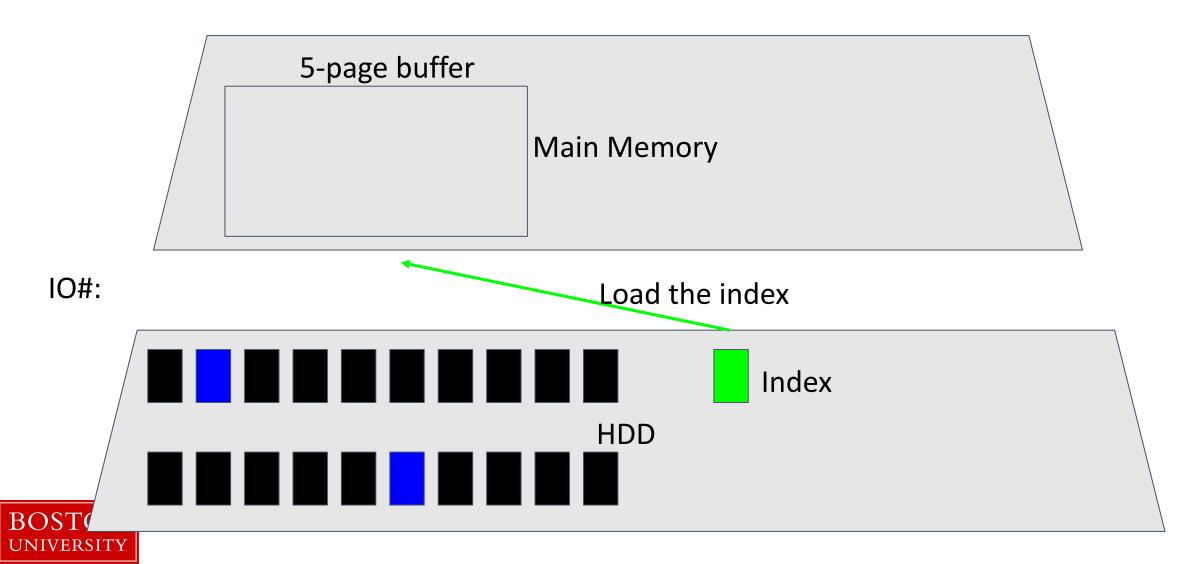
What if we had an oracle (index)?

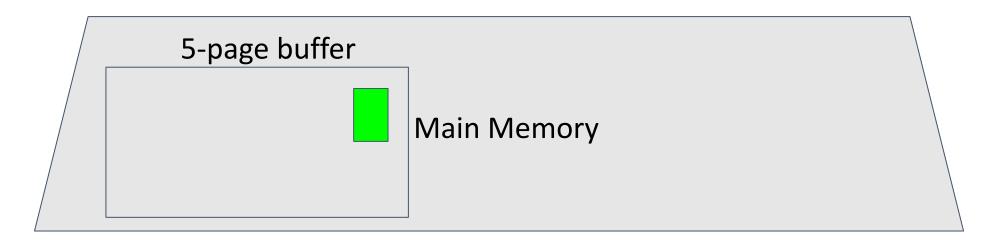




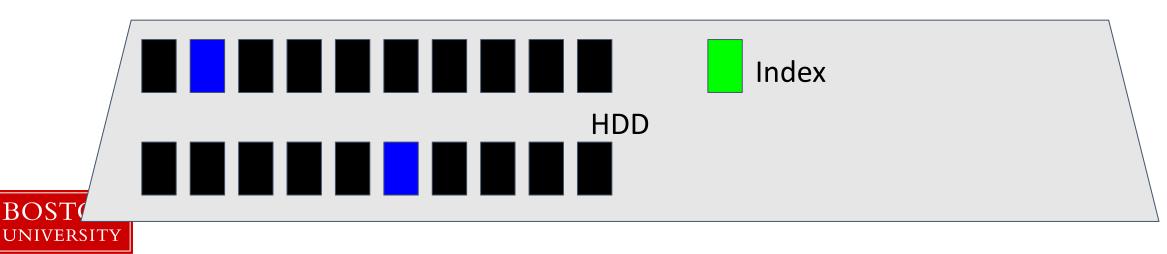
IO#:

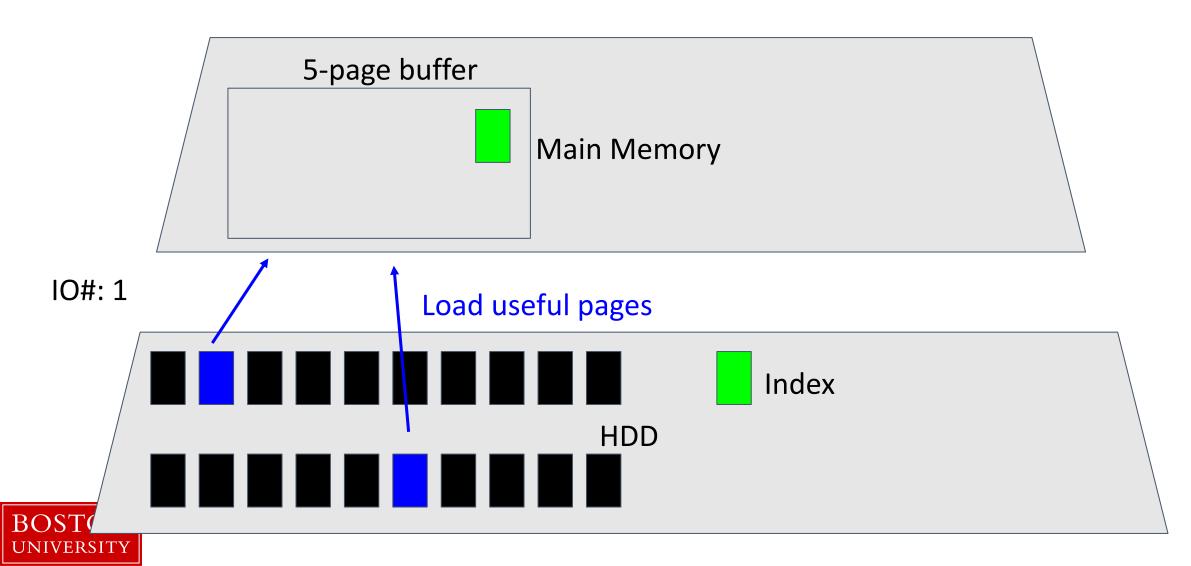


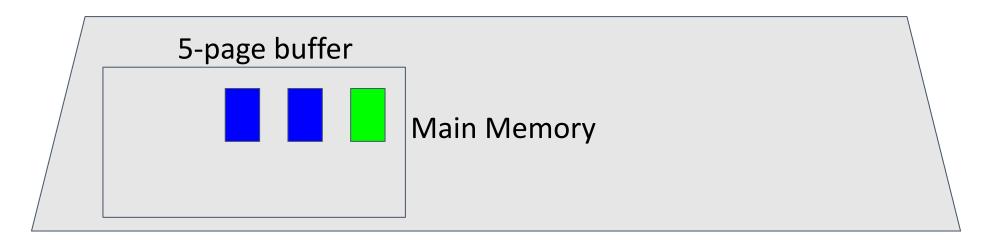




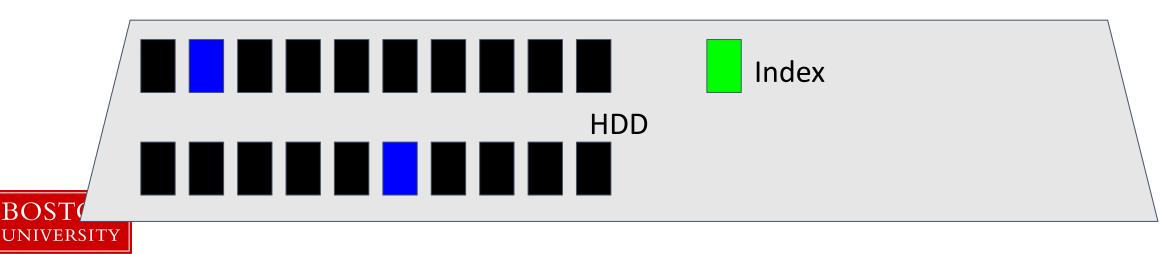








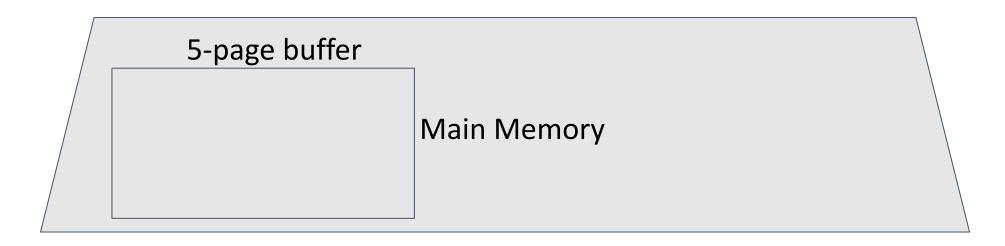




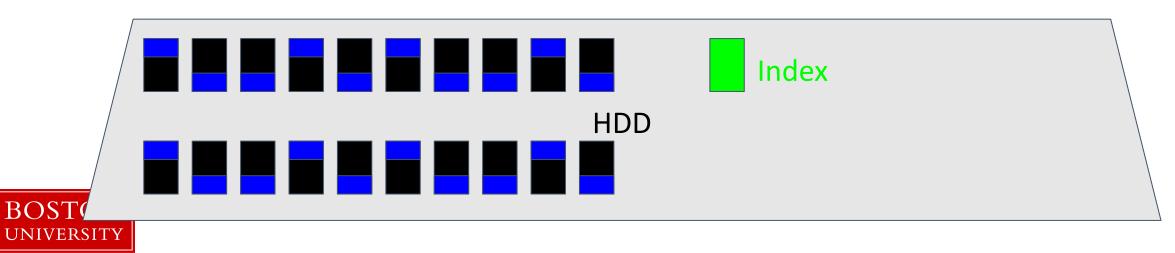
What if useful data is in all pages?



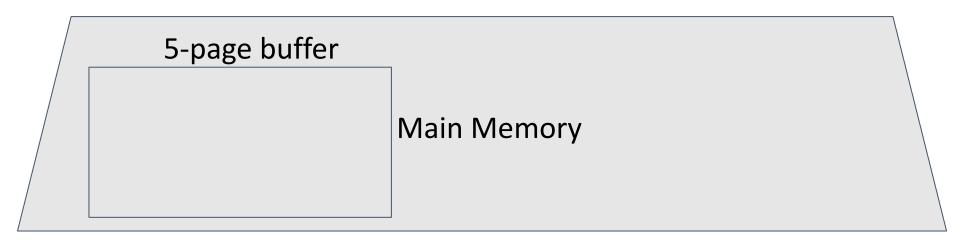
Scan or Index ?



IO#:

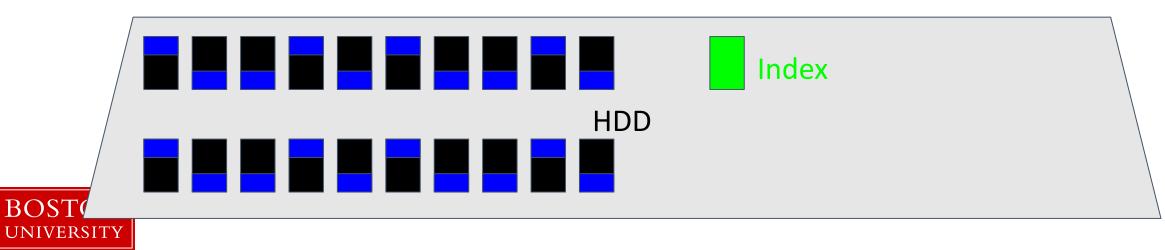


Scan or Index ?

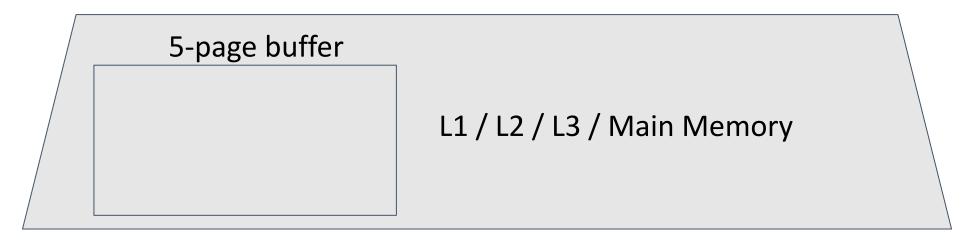


IO#: 20 with scan

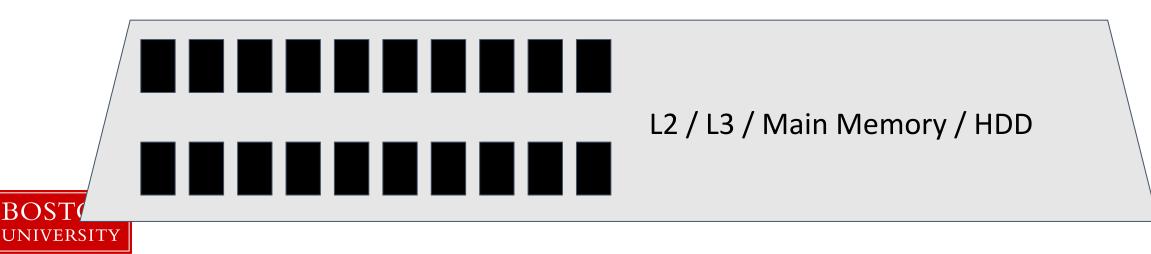
IO#: 21 with index



Same analysis for any two memory levels!!

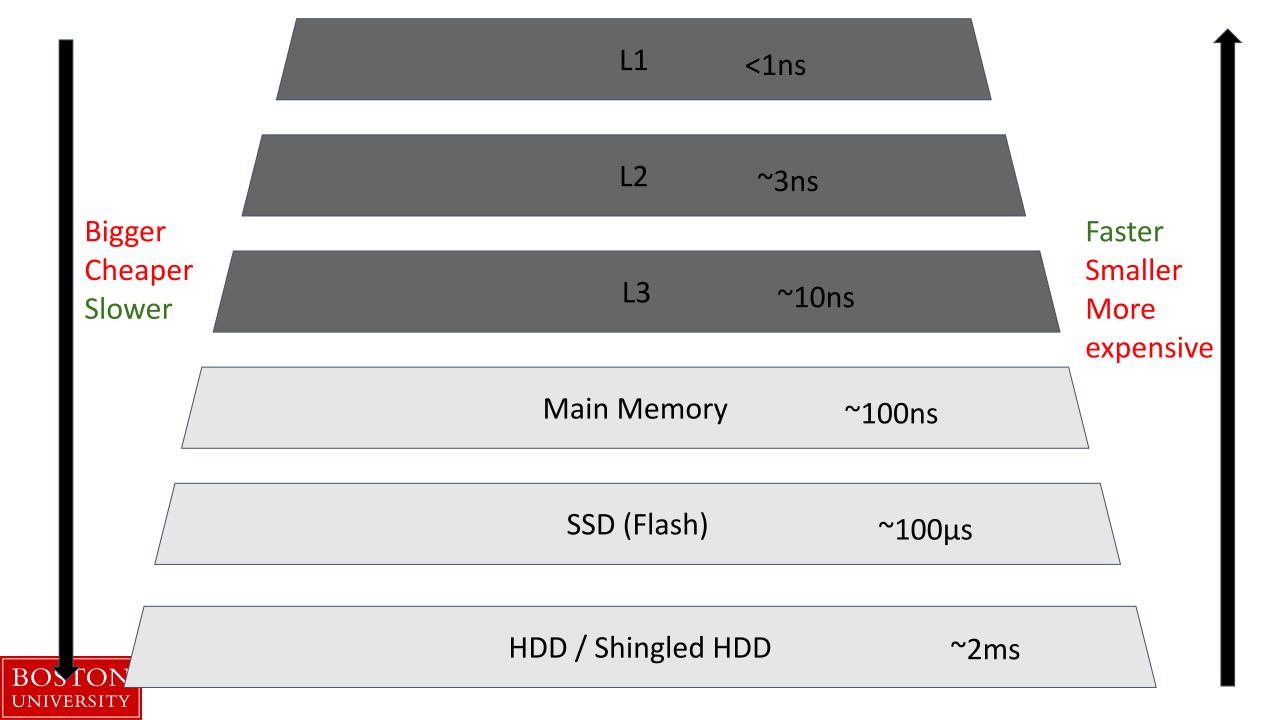


IO#:



Cache Hierarchy



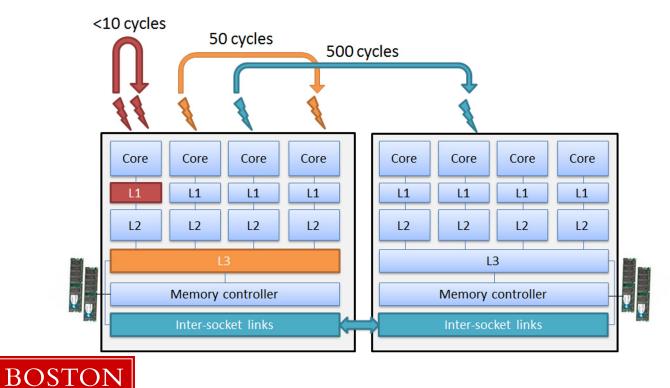


Cache Hierarchy

What is a core?

What is a socket?

UNIVERSITY



L1



L3

Cache Hierarchy

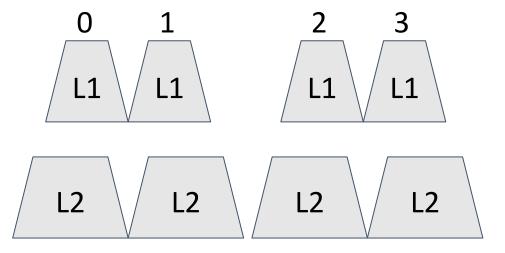
Shared Cache: L3 (or LLC: Last Level Cache)

L3 is physically distributed in multiple sockets

L2 is physically distributed in every core of every socket

Each *core* has its own **private** L1 & L2 cache All levels need to be *coherent**

what does coherent mean?





A core reads faster when data are in its L1

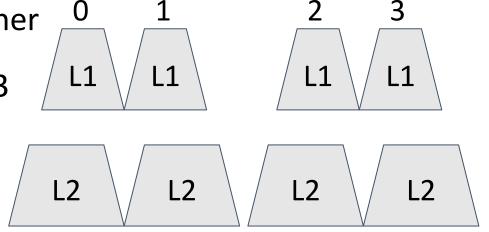
If it does not fit, it will go to L2, and then in L3

Can we control where data is placed?

We would like to avoid going to L2 and L3 altogether

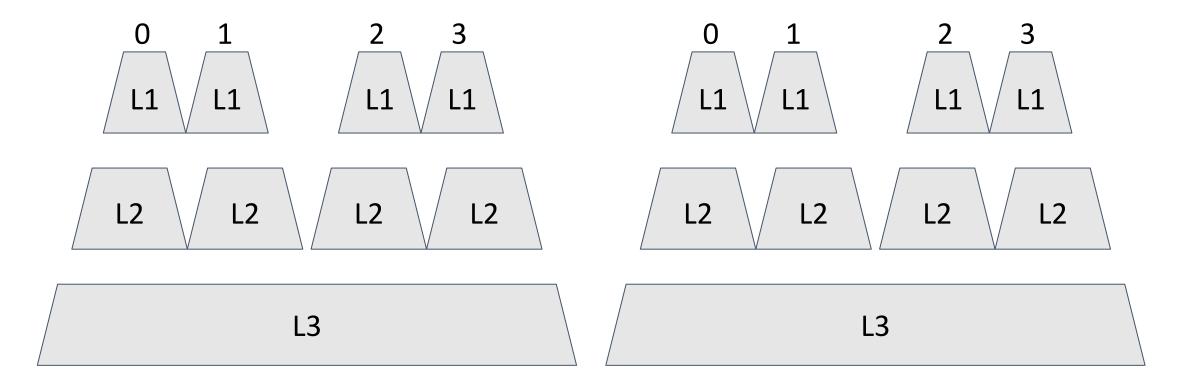
But, at least we want to avoid to remote L2 and L3

And remember: this is only one socket. We have **multiple** sockets!

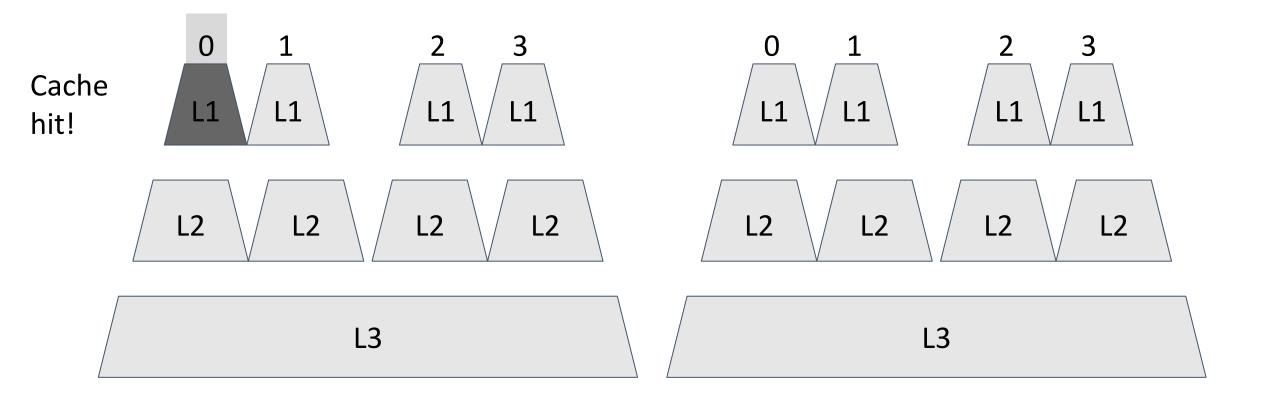




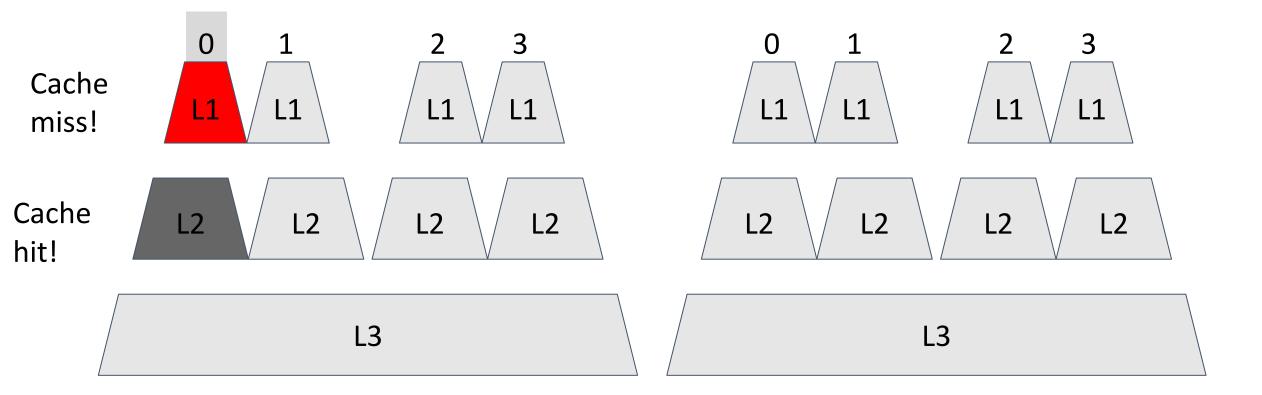




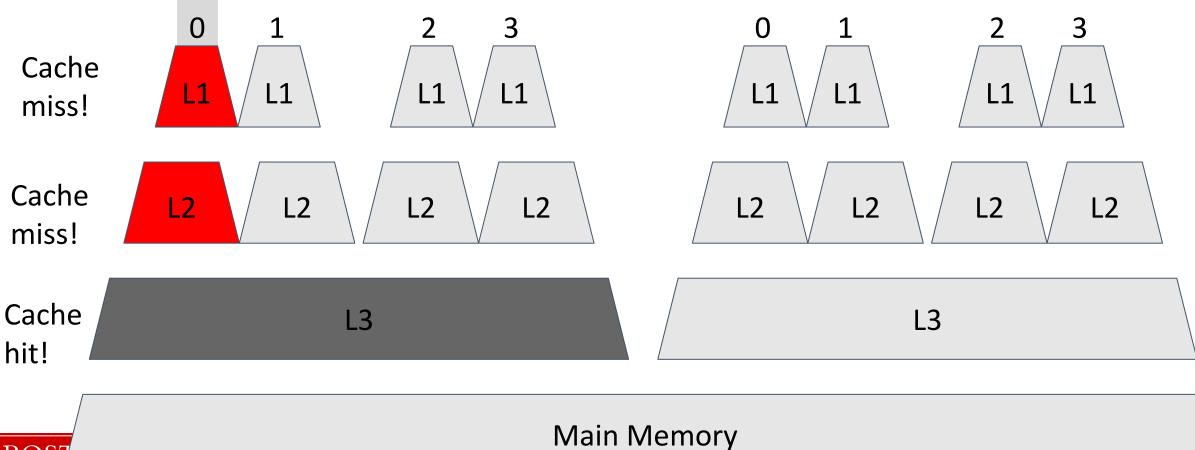




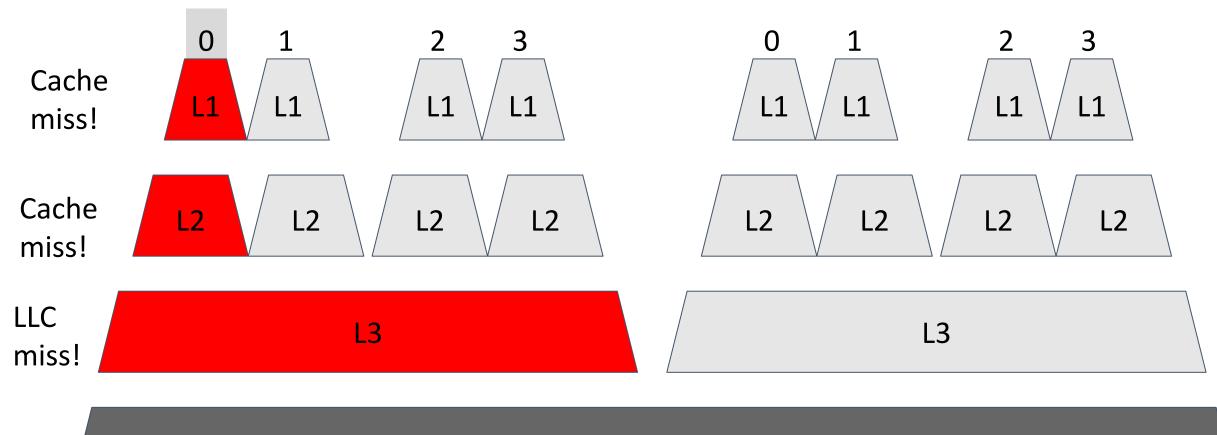




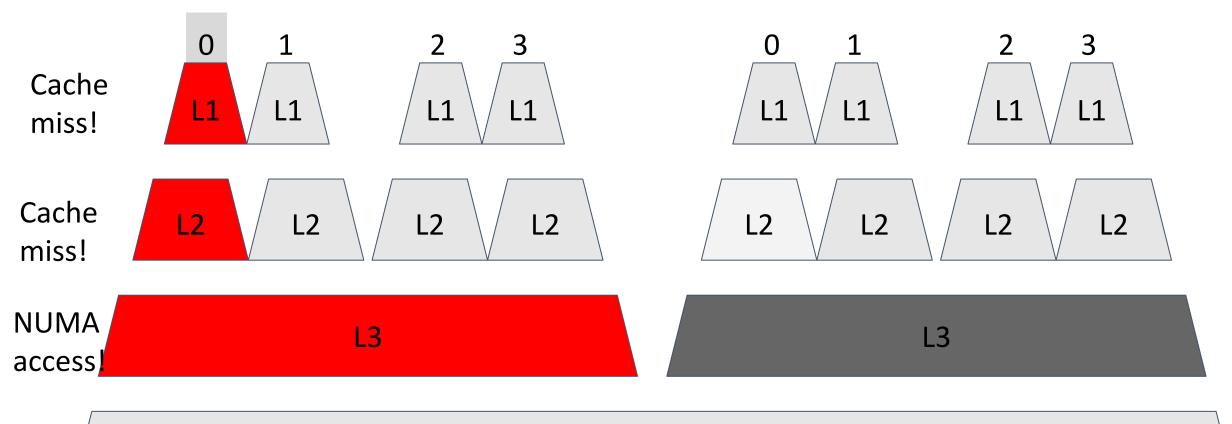










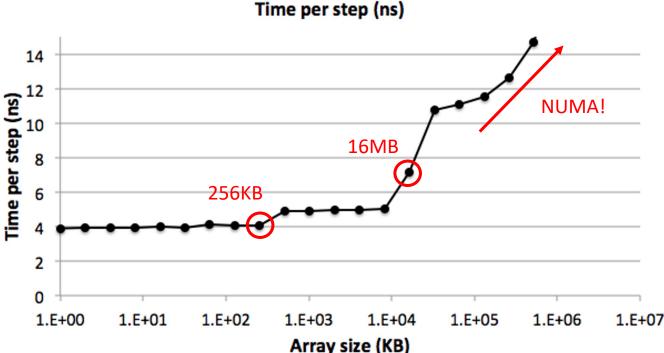




Why knowing the cache hierarchy matters

```
int arraySize;
for (arraySize = 1024/sizeof(int) ; arraySize <= 2*1024*1024*1024/sizeof(int) ; arraySize*=2)
// Create an array of size 1KB to 4GB and run a large arbitrary number of operations
{
    int steps = 64 * 1024 * 1024; // Arbitrary number of steps
    int* array = (int*) malloc(sizeof(int)*arraySize); // Allocate the array
    int lengthMod = arraySize - 1;
    // Time this loop for every arraySize
    int i;
    for (i = 0; i < steps; i++)
        {
            array[(i * 16) & lengthMod]++;
            // (x & lengthMod) is equal to (x % arraySize)
```

This machine has: 256KB L2 per core 16MB L3 per socket





}



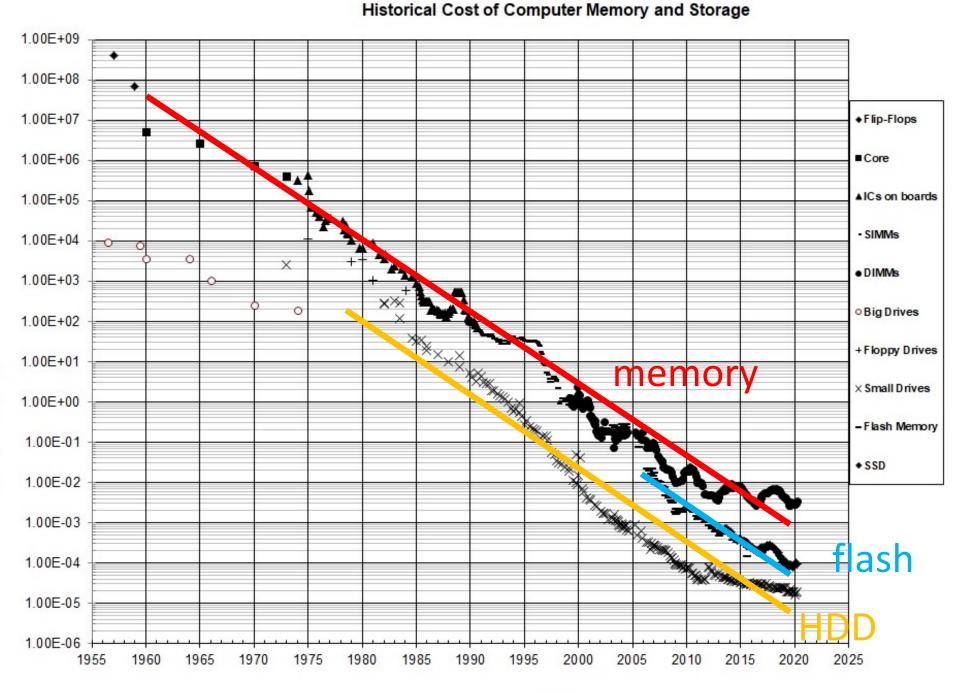
Why not just stay in memory?



Cost!

what else?

Memory Price (\$/MB)





Why not stay in memory?

Rephrase: what is missing from memory hierarchy?

Durability (data survives between restarts)

Capacity (enough capacity for data-intensive applications)



Main Memory

SSD (Flash)

HDD

Shingled Disks



Таре

Main Memory

SSD (Flash)

HDD

Shingled Disks



Таре

Hard Disk Drives

Secondary durable storage that support both *random* and *sequential* access

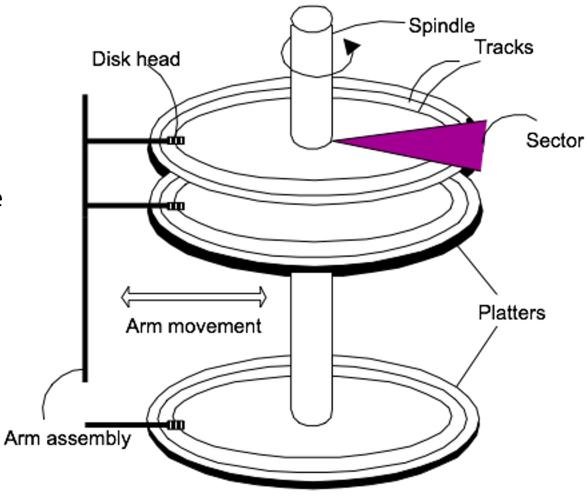
Data organized on pages/blocks (across tracks)

Multiple *tracks* create an (imaginary) *cylinder*

Disk access time: seek latency + rotational delay + transfer time (0.5-2ms) + (0.5-3ms) + <0.1ms/4KB

Sequential >> random access (~10x)

Goal: avoid random access





Seek time + Rotational delay + Transfer time

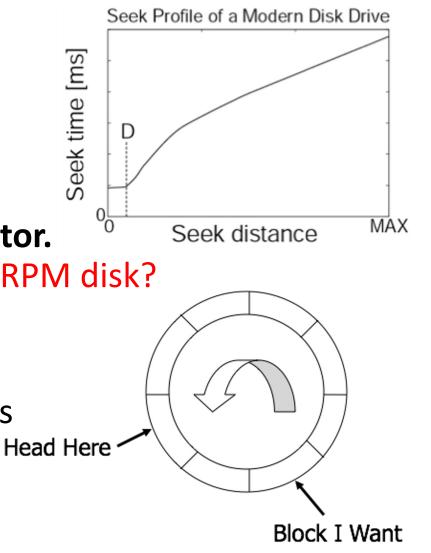
<u>Seek time</u>: the **head** goes to the right **track**

Short seeks are dominated by "settle" time (D is on the order of hundreds or more)

<u>Rotational delay</u>: The **platter** rotates to the right **sector**. ¹⁰ What is the min/max/avg rotational delay for 10000RPM disk?

min: 0, max: 60s/10000=6ms, avg: 3ms

<u>Transfer time</u>: <0.1ms / page \rightarrow more than 100MB/s





Sequential vs. Random Access

Bandwidth for Sequential Access (assuming 0.1ms/4KB):

0.04ms for 4KB \rightarrow **100MB/s**

Bandwidth for Random Access (4KB):

0.5ms (seek time) + 3ms (rotational delay) + 0.04ms = 3.54ms

4KB/3.54ms → **1.16MB/s**



Flash

Secondary durable storage that support both *random* and *sequential* access

Data organized on pages (similar to disks) which are further grouped to erase blocks

Main advantage over disks: random read is now much more efficient

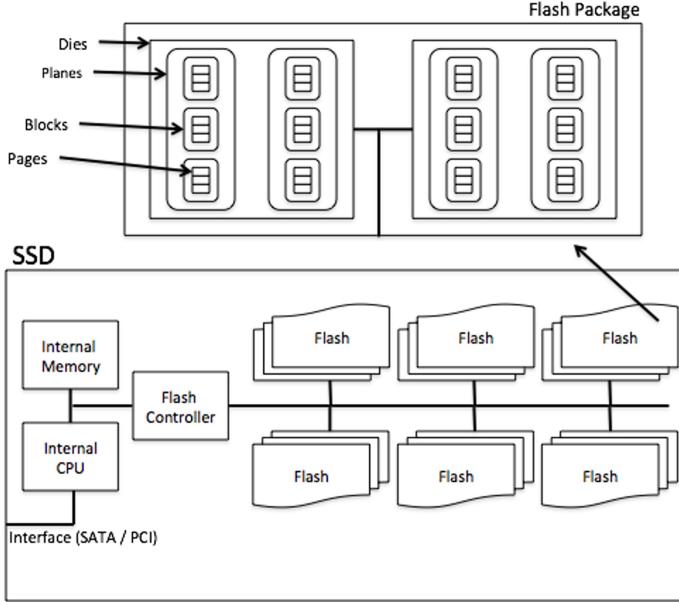
BUT: Not as fast random writes!

Goal: avoid random writes





The internals of flash



BOST

UNIVERSIT

interconnected flash chips

no mechanical limitations

maintain the block API compatible with disks layout

internal parallelism for both read/write

complex software driver

Flash access time

... depends on:

device organization (internal parallelism)

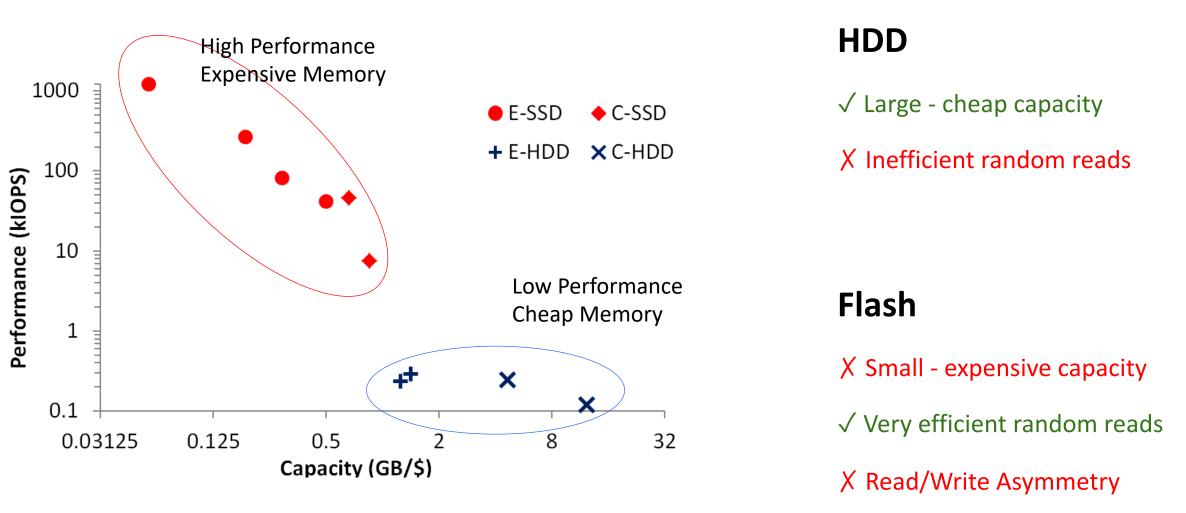
software efficiency (*driver*)

bandwidth of flash packages

the Flash Translation Layer (FTL), a complex device driver (firmware) which tunes performance and device lifetime



Flash vs HDD





Main Memory

Flash

HDD

Shingled Disks



Таре

Tapes

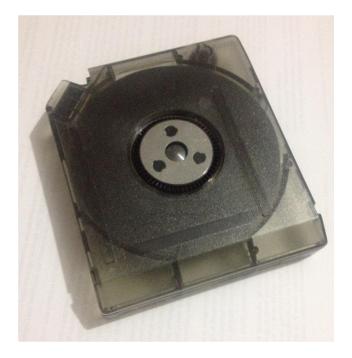
Data size grows exponentially!

Cheaper capacity:

Increase density (bits/in²) Simpler devices

Tapes:

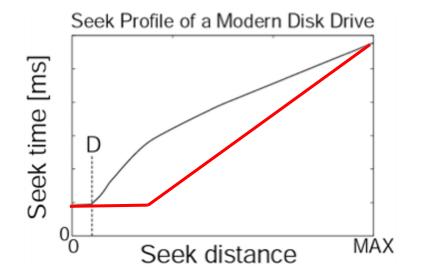
Magnetic medium that allows only **sequential access** (yes like an old-school tape)!





Increasing disk density

Very difficult to differentiate between tracks "settle" time becomes Writing a track affects neighboring tracks Create different readers/writers Interleave writes tracks

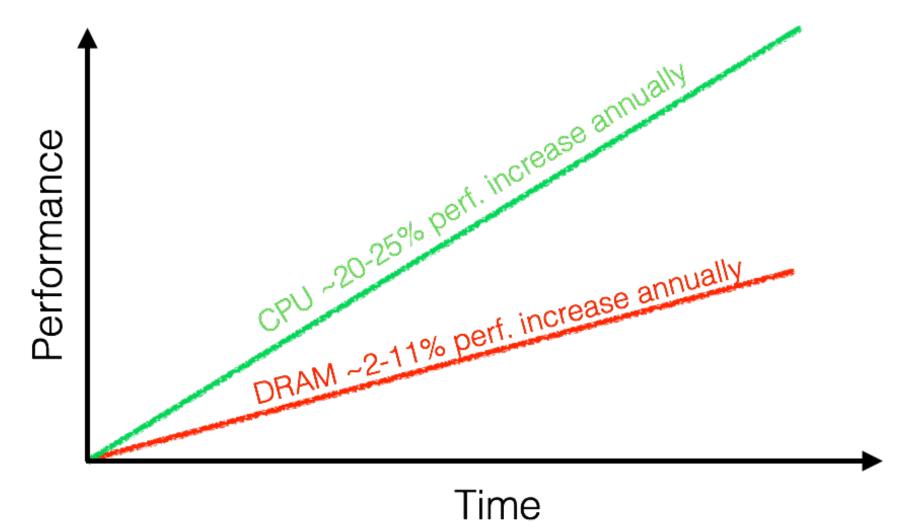


Conventional Writes



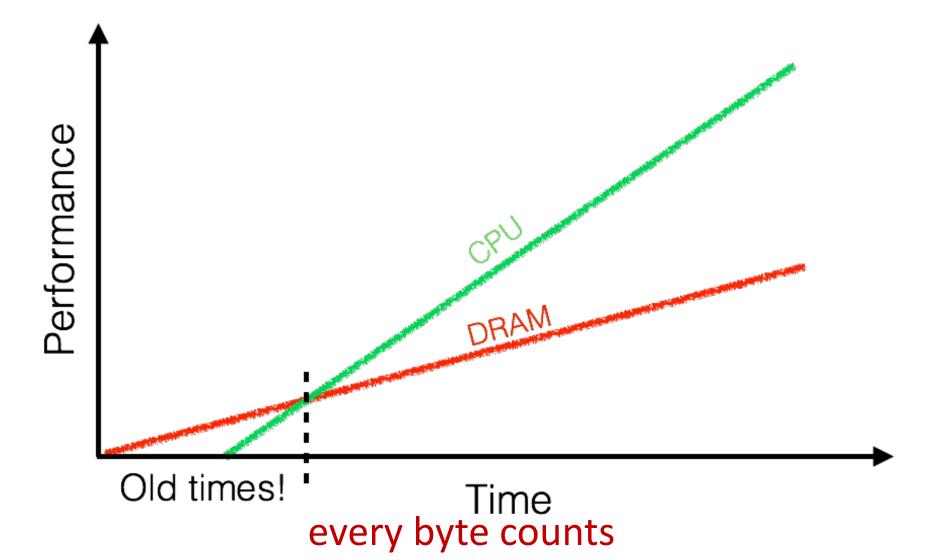
Memory & Storage Walls

Memory Wall





Memory Wall





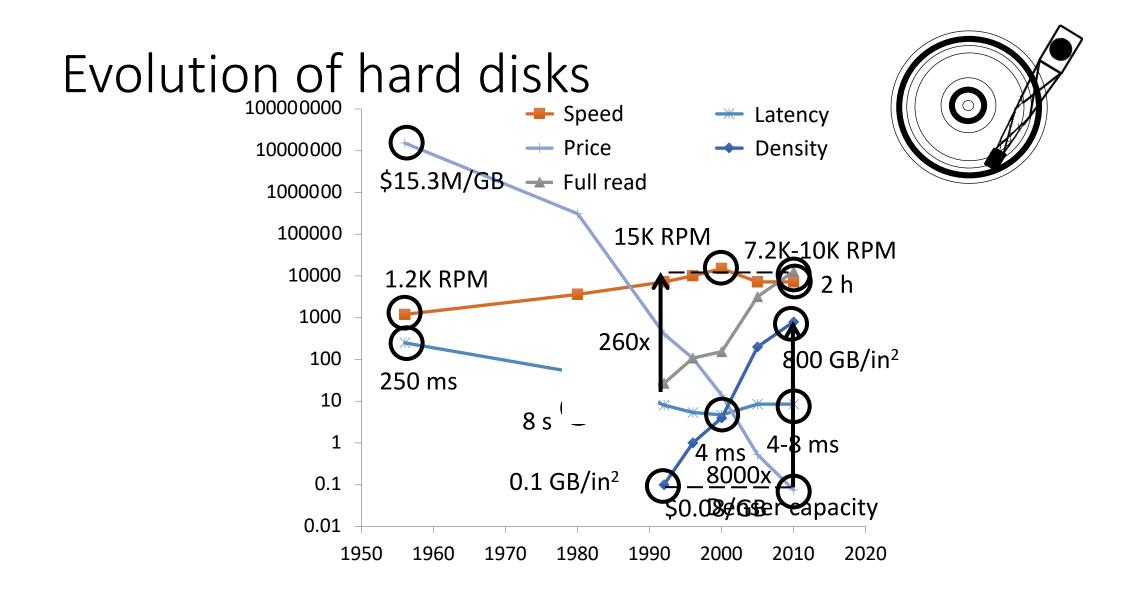
Storage Wall



HDD

- ✓ capacity
- ✓ sequential access
- × random access
- × latency plateaus





disks become larger but – relatively – slower

Storage Wall

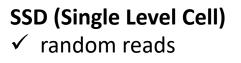


HDD

BOSTON

UNIVERSITY

- ✓ capacity
- \checkmark sequential access
- × random access
- × latency plateaus



✓ low latency × capacity

- × endurance
- × read/write asymmetry

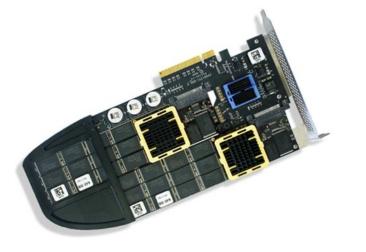
SSD (Multi Level Cell)

- ✓ capacity
- × endurance (worse)

"Tape is Dead, Disk is Tape, Flash is Disk"

[Jim Gray 2007]

 Storage without mechanical limitations

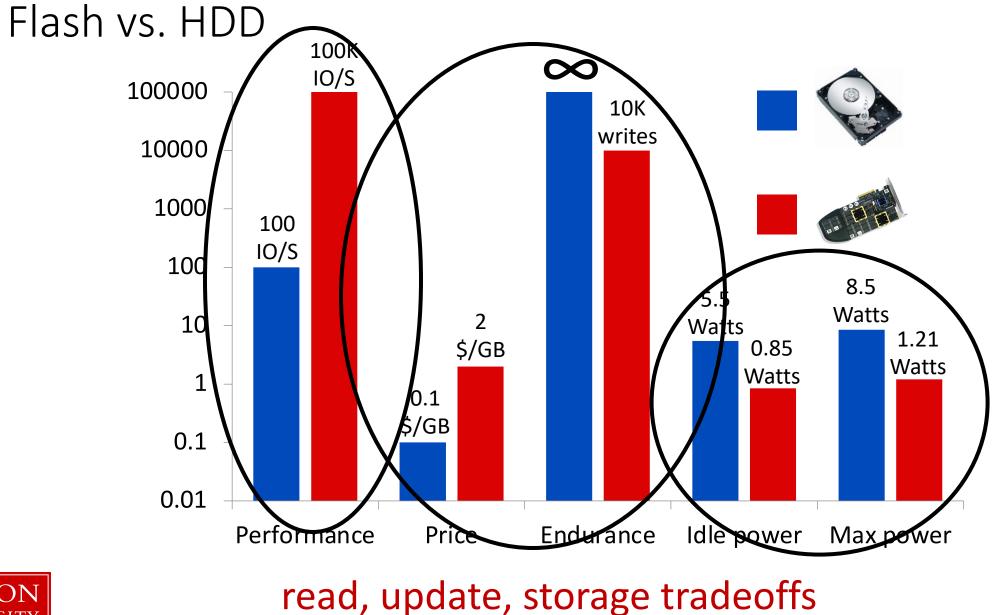


Several technologies

- Flash
- Phase Change Memory (IBM)
- Memristor (HP)



flash vs. hard disks?





Storage Wall



HDD

- ✓ capacity
- ✓ sequential access
- × random access
- × latency plateaus

SSD (Single Level Cell)

- ✓ random reads
- ✓ low latency
- × capacity
- × endurance
- × read/write asymmetry

SSD (Multi Level Cell)

- ✓ capacity
- × endurance (worse)

HDD (Shingled Magnetic Rec.)

- ✓ capacity
- × read/write asymmetry



every byte counts

Technology Trends & Research Challenges

(1) From fast single cores to increased parallelism

(2) From slow storage to efficient random reads

(3) From **infinite** endurance to **limited endurance**

(4) From symmetric to asymmetric read/write performance



Technology Trends & Research Challenges

How to exploit increasing parallelism (in compute and storage)?

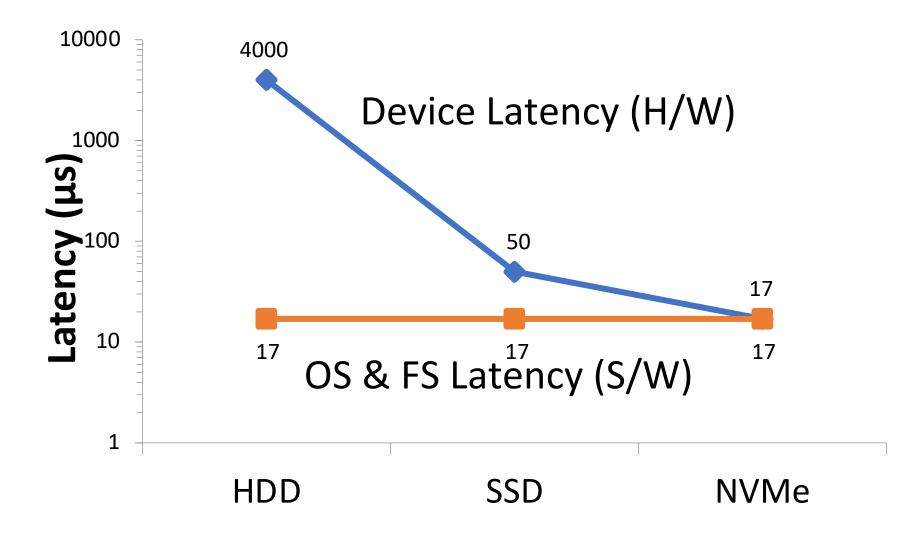
How to redesign systems for efficient random reads? e.g., no need to aggressively minimize index height!

How to reduce *write amplification* (physical writes per logical write)?

How to write algorithms for asymmetric storage?



Even faster devices are available (NVMe) How to use them when the software stack is too slow?







Class 15

Data Systems on Modern Hardware: Multi-cores, Solid-State Drives, and Non-Volatile Memories

Dr. Subhadeep Sarkar

https://bu-disc.github.io/CS561/