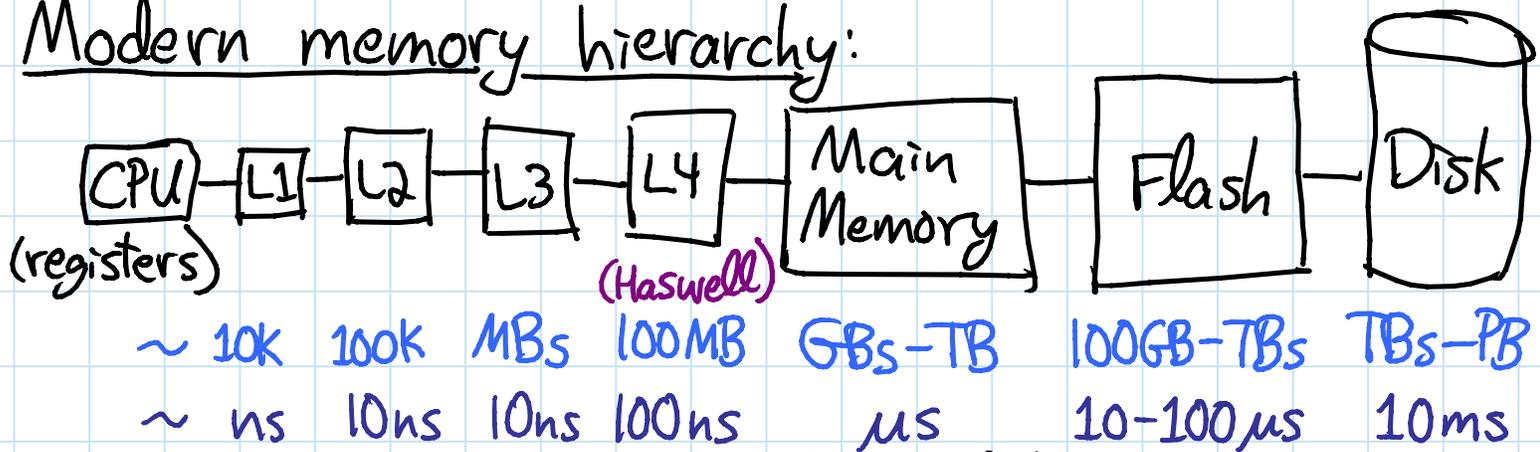


TODAY: Cache-oblivious algorithms I (of 2)

- memory hierarchy
- external memory vs. cache oblivious models
- scanning
- divide & conquer
  - median finding
  - matrix multiplication
- LRU block replacement

So far we've viewed all word operations & all memory accesses as equal cost...

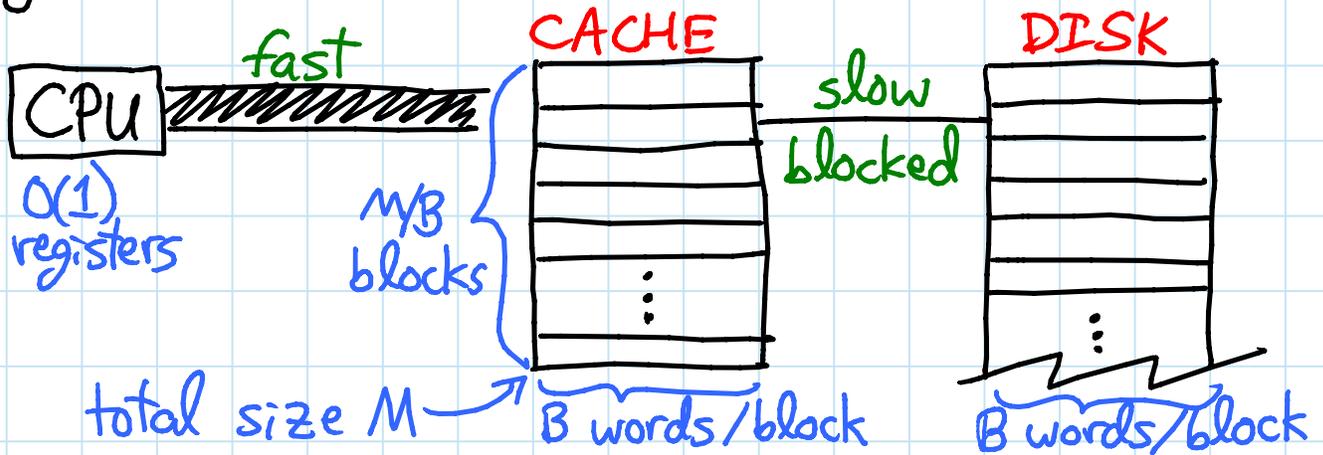
## Modern memory hierarchy:



- bigger but slower latency:  
distance travel & physical seek on disk
- bandwidth usually matched (RAID etc.)
  - blocking to mitigate latency:
    - when fetching a word of data, get entire block containing it
    - idea: amortize latency over whole block
    - ⇒ amortized cost per word
    - =  $\frac{\text{latency}}{\text{block size}} + \frac{1}{\text{bandwidth}}$
    - set roughly equal via block size
  - to work, we need algorithms to use all elements in a block (spatial locality) & re-use blocks in cache (temporal locality)

# External-memory model: [Aggarwal & Vitter 1988]

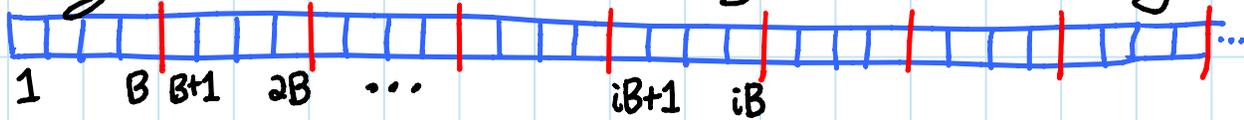
- just 2 levels:



- cache accesses free (just count computation)
- $\Rightarrow$  count memory transfers between cache  $\leftrightarrow$  disk  
= # blocks read from/written to disk
- algorithm explicitly reads & writes blocks

Cache-oblivious model: <sup>→ FFTW → L in CURS</sup> [Frigo, Leiserson, Prokop, <sup>→ M.Eng.</sup> Ramachandran 1999]

- algorithm doesn't know  $B$  or  $M$  (!)
- accessing a word in memory (blocked array:



automatically fetches entire block containing it  
& evicts (writes) least recently used (LRU)  
block from cache if full

(more like real caches)

⇒ every algorithm is a cache-oblivious algorithm

- new measurement & objective:

minimize # memory transfers

Why?

- cooler
- often possible
- "cleaner" algorithms, & implementations
- automatic "tuning"
- optimize all levels of memory hierarchy (each with their own  $B$  &  $M$ )

## Scanning:

Single scan: e.g. for  $i$  in range( $N$ ):

Sum +=  $A[i]$

- assume array  $A$  stored contiguously in memory
- external memory: align  $A$  with block start<sup>0</sup>  
⇒  $\lceil N/B \rceil$  memory transfers



- cache oblivious: can't control alignment
  - still  $\leq \lceil N/B \rceil + 1 = N/B + O(1)$

$O(1)$  parallel scans: (assuming  $N/B = \Omega(1)$ )

- e.g. reversing  $A[0:n]$ :

[Bentley]

for  $i$  in range( $\lfloor N/2 \rfloor$ ):

swap  $A[i] \leftrightarrow A[N-i-1]$



- keep one block  $\ni A[i]$  & one  $\ni A[N-i-1]$   
⇒  $O(N/B + 1)$  memory transfers (assuming  $N/B \geq 2$ )

## Divide & conquer approach: → cache oblivious

- algorithm divides problem down to  $O(1)$  size
- analysis considers recursion at which
  - problem fits in cache i.e.  $\leq M$
  - problem fits in  $O(1)$  blocks i.e.  $O(B)$
- TODAY: one example of each

## Median finding / order statistics:

- recall  $O(N)$ -time deterministic algorithm: [L2]

① view array as partitioned into columns of 5 like blocks, but  $O(1)$  size ←

② sort each column → median

③ recursively find median of column medians

④ partition array by  $x$  ( $\leq x, > x$ )

⑤ recurse on one side

- memory transfer analysis:  $MT(N)$

① free

② scan  $\Rightarrow O(N/B + 1)$

③  $MT(N/5) \sim$  if we coalesce  $N/5$  medians into a consecutive array (via 2 parallel scans)

④ 3 parallel scans  $\Rightarrow O(N/B + 1)$

⑤  $MT(\frac{7}{10}N)$

$$\Rightarrow MT(N) = MT(N/5) + MT(\frac{7}{10}N) + O(N/B + 1)$$

- usual base case:  $MT(O(1)) = O(1)$ 
  - $\Rightarrow MT(N) \geq \# \text{ leaves } L(N) \text{ in recursion}$
  - $L(N) = L(N/5) + L(\frac{7}{10}N)$ 

$$N^\alpha = (N/5)^\alpha + (\frac{7}{10}N)^\alpha$$

$$1 = (1/5)^\alpha + (7/10)^\alpha$$
  - $\Rightarrow \alpha \approx 0.83978$
  - $\Rightarrow MT(N) \geq N^{0.8} = \omega(N/B)$  if  $B = \omega(B^{0.2})$

- stronger base case:  $MT(O(B)) = O(1)$ 
  - $\Rightarrow \# \text{ leaves } L(N) = (N/B)^\alpha = o(N/B)$
  - cost at each level of recursion tree decreases geometrically down  
(a little tricky to prove — better to use substitution method like L2)
  - $\Rightarrow$  dominated by root cost  $O(N/B + 1)$
  - $\Rightarrow MT(N) = O(N/B + 1)$

Matrix multiplication:

$$N \left\{ \begin{array}{|c|} \hline Z \\ \hline \end{array} \right\} = N \left\{ \begin{array}{|c|} \hline X \\ \hline \end{array} \right\} \cdot N \left\{ \begin{array}{|c|} \hline Y \\ \hline \end{array} \right\}$$

Standard algorithm:

- ideal memory layout:

- X stored in row-major order

- Y stored in column-major order

- Z stored in either, say row-major

- each  $z_{ij}$  costs  $\Theta(N/B + 1)$

- upper bound: 2 parallel scans

- X row  $i$  gets re-used in all  $z_{i*}$

(assuming  $N/B \geq 3$ )

- but Y column  $j$  gets read for every  $z_{ij}$

(assuming  $M < N^2 = \text{size}(Y)$ )

-  $MT(N) = \Theta(N^3/B + N^2)$

- NOT OPTIMAL

Block algorithm:

$$\begin{array}{|c|c|} \hline z_{11} & z_{12} \\ \hline z_{21} & z_{22} \\ \hline \end{array} = \begin{array}{|c|c|} \hline x_{11} & x_{12} \\ \hline x_{21} & x_{22} \\ \hline \end{array} \cdot \begin{array}{|c|c|} \hline y_{11} & y_{12} \\ \hline y_{21} & y_{22} \\ \hline \end{array} = \begin{array}{|c|c|} \hline x_{11}y_{11} + x_{12}y_{21} & x_{11}y_{12} + x_{12}y_{22} \\ \hline x_{21}y_{11} + x_{22}y_{21} & x_{21}y_{12} + x_{22}y_{22} \\ \hline \end{array}$$

- store matrices in recursive block layout:

$$\boxed{X} = \boxed{x_{11}} \boxed{x_{12}} \boxed{x_{21}} \boxed{x_{22}}$$

↑ recursive layouts

- order of blocks doesn't matter

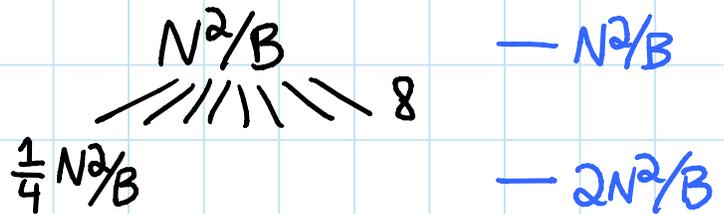
- key: each block is stored consecutively

$$\Rightarrow MT(N) = \underbrace{8 \cdot MT(N/2)}_{\text{recursion}} + \underbrace{O(N^2/B + 1)}_{\text{addition is 3 parallel scans}}$$

- base cases:  $MT(O(1)) = O(1)$   
 $MT(O(B)) = O(1)$   
 $MT(\sqrt{M/3}) = O(M/B)$

$\Rightarrow 3 \sqrt{M/3} \times \sqrt{M/3}$  fit in cache

- recursion tree:



$$\underbrace{O(M/B) \quad O(M/B) \quad \dots}_{\# \text{leaves} = 8 \lg(N/\sqrt{M}) = O((N/\sqrt{M})^3)} = O\left(\frac{N^3}{B\sqrt{M}}\right)$$

- geometrically increasing cost down tree  
 (like Master Theorem)

$\Rightarrow$  dominated by leaf level

$\Rightarrow MT(N) = O\left(\frac{N^3}{B\sqrt{M}}\right) \leftarrow$  ASYMPTOTICALLY OPTIMAL

- generalizes to non-powers of 2 & non-square matrices
- similar algorithms & analyses for
  - Strassen's algorithm
  - FFT

## Why LRU block replacement strategy?

$$LRU_M \leq 2 \cdot OPT_{M/2}$$

[Sleator & Tarjan 1985]

RESOURCE AUGMENTATION  
(changing  $M$ )

Proof:

- partition block access sequence into maximal phases of  $M/B$  distinct blocks
- LRU spends  $\leq M/B$  memory transfers/phase
- OPT must spend  $\geq \frac{M}{2}/B$  memory transfers per phase: at best, starts phase with entire  $M/2$  cache with needed items, but there are  $M/B$  blocks during phase, so  $\leq$  half free

ONLINE ALGORITHMS - comparing regular "online" algorithm (can't see the future) against offline/prescient optimal algorithm

- changing  $M$  by factor of 2 doesn't affect bounds like  $O\left(\frac{N^2}{B\sqrt{M}}\right)$

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