

MEMORY HIERARCHY (PART III) CACHE CONSCIOUS DATA-PROCESSING

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OVERVIEW

| ☐ In previous lectures, we show the basic concepts and analysis of memory hierarchy. |
|---|
| In this lecture, we show some design principles to make use of memory hierarchy for processing big data. We will discuss following two cases. □ Array access patterns □ Spatial locality designs □ Temporal locality designs |
| ☐ Big data sorting |

MOTIVATING QUESTION

Suppose we have a 10000x10000 matrix of integers, which is stored in a 2-dimentional array A[10000][10000]. We want to set all of its value to 1. Cache size = 5000 integers. How would you write your code?

```
Solution X
for (int i=0;i<10000;i++)
for(int j=0;j<10000;j++)
A[i][j]=1;
```

Solution Y for (int j=0;j<10000;j++) for(int i=0;i<10000;i++) A[i][j]=1;



Which one is better, X or Y?

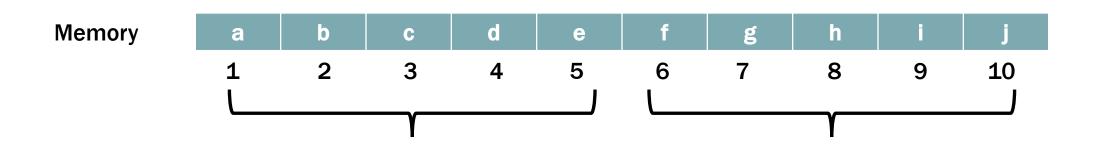
ACCESSING AN ARRAY

- We may access an array with different access patterns
 - □ Access pattern means the position sequence of accessing an array.
 - ☐ Different access patterns may impact the utility of cache

We have a 10-integer array stored in main memory, cache size = 5 integers, transfer size (cache line)= 5 integers

Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10





Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

Cache a b c d e

Cache Miss: 1
Cache Hit: 0

Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

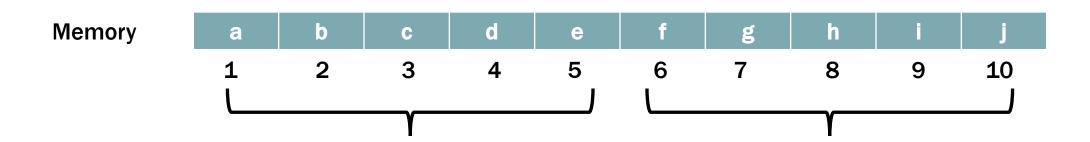
Cache a b c d e

Cache Miss: 1
Cache Hit: 1

Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

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Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

Cache a b c d e

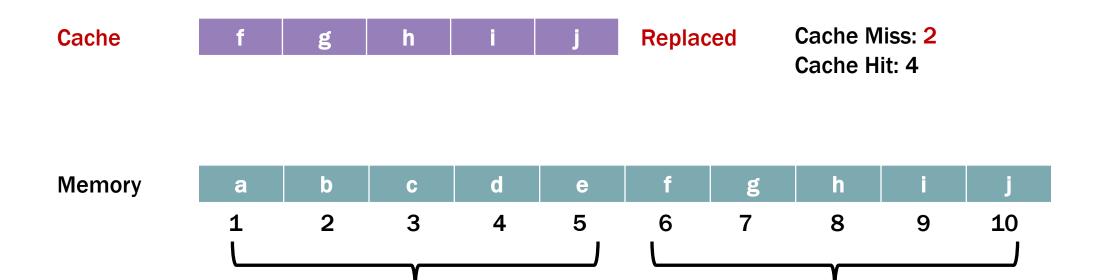
Cache Miss: 1
Cache Hit: 3

Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

Cache a b c d e

Cache Miss: 1
Cache Hit: 4

Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10



Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

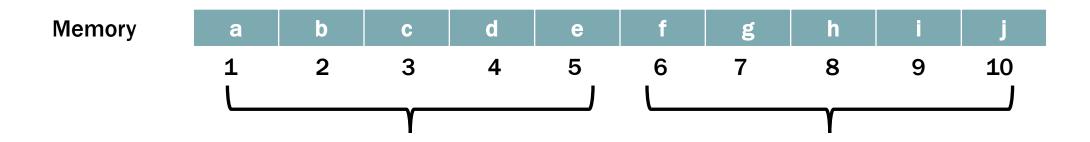
Cache f g h i j

Cache Miss: 2
Cache Hit: 5

Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

Cache f g h i j

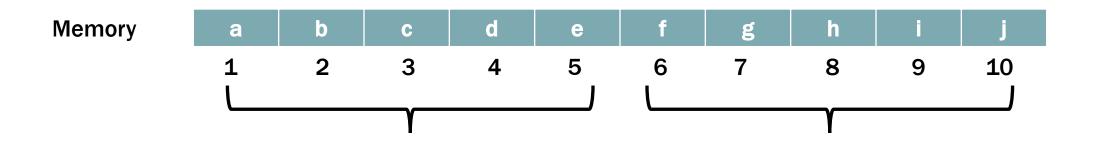
Cache Miss: 2



Access pattern: 1, 2, 3, 4, 5, 6, 7, 8, 9, 10

Cache f g h i j

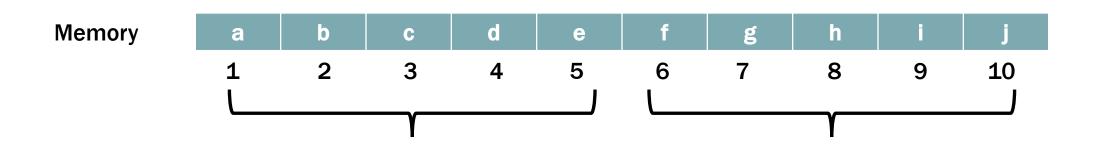
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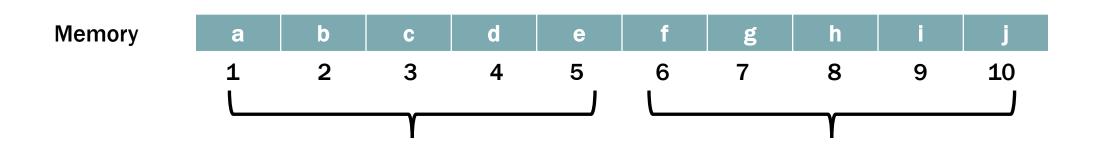
Cache f g h i j

Cache Miss: 2



Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10



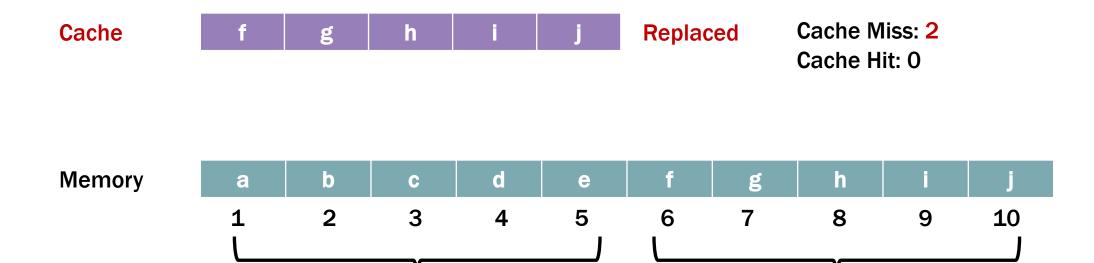


Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

Cache a b c d e

Cache Miss: 1
Cache Hit: 0

Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

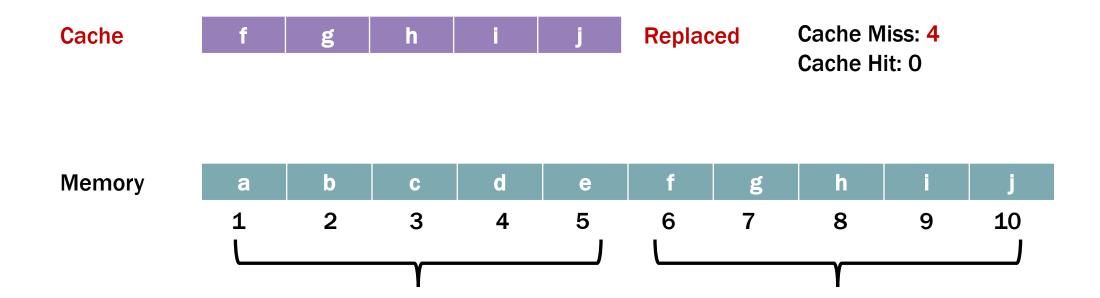


Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

Cache

a b c d e Replaced Cache Miss: 3
Cache Hit: 0

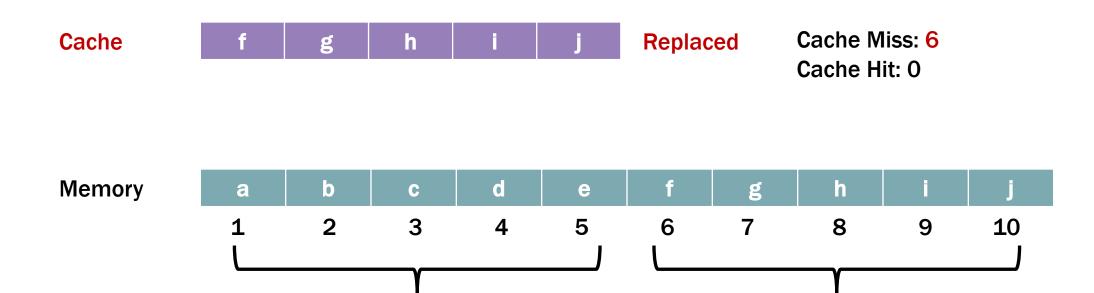
Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10



Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

Cache a b c d e Replaced Cache Miss: 5
Cache Hit: 0

Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

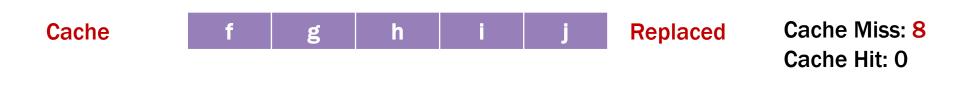


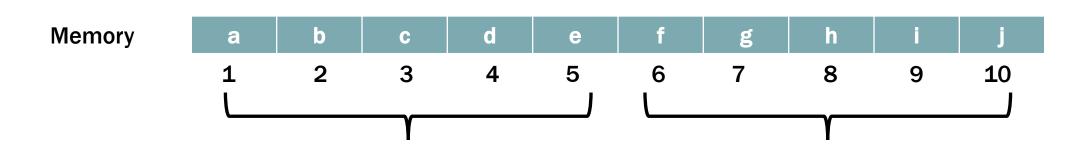
Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

Cache

a b c d e Replaced Cache Miss: 7
Cache Hit: 0

Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

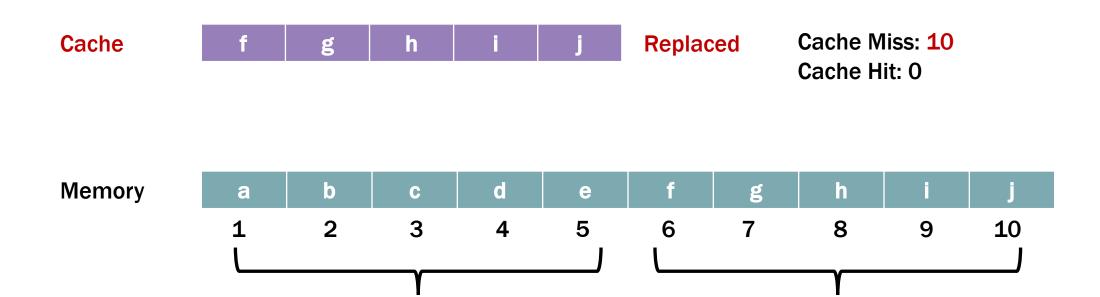




Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

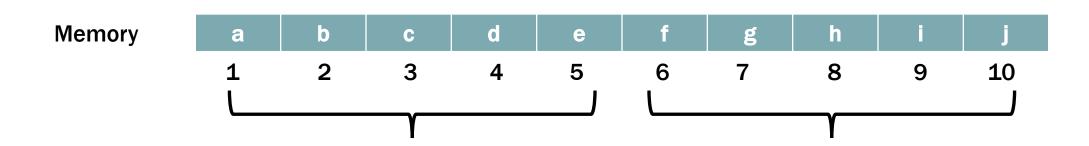
Cache a b c d e Replaced Cache Miss: 9
Cache Hit: 0

Access pattern: 1, 6, 2, 7, 3, 8, 4, 9, 5, 10

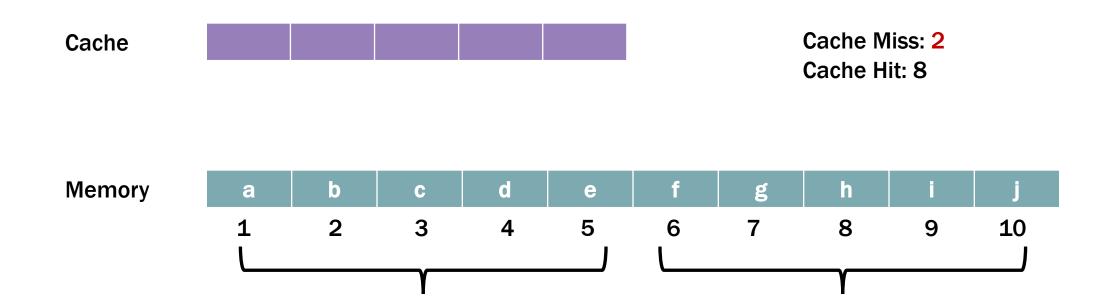


Access pattern: 1, 6, 8, 7, 7, 8, 9, 9, 6, 10





Access pattern: 1, 6, 8, 7, 7, 8, 9, 9, 6, 10



SUMMARY

- Access pattern with spatial locality (pattern 1) or temporal locality (pattern 3) is cache friendly.
- ☐ The adversarial access (pattern 2) may cause the worst case read performance

■ When data is big, the impact is very significant

Whenever possible, we store the data in a way that the access pattern has certain locality.

2D ARRAYS

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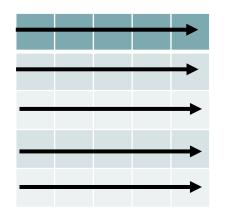
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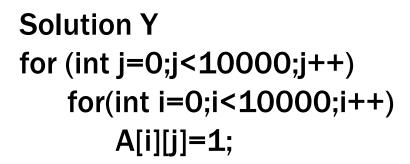
2D ARRAYS

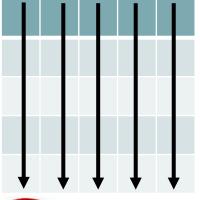
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Solution X
for (int i=0;i<10000;i++)
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Cache friendly

Which one is better, X or Y?





Always evict cache



In the big data era, it happens very often that the data volume is too big to hold in main memory.

Suppose we have 1 TB of integers stored in disk and we want to sort the array. My computer only has 16GB main memory, what should I do?



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Suppose we have 1 TB of integers stored in disk and we want to sort the array. My computer only has 16GB main memory, what should I do?

We have learnt a lot of sorting algorithms, can we simply use them?



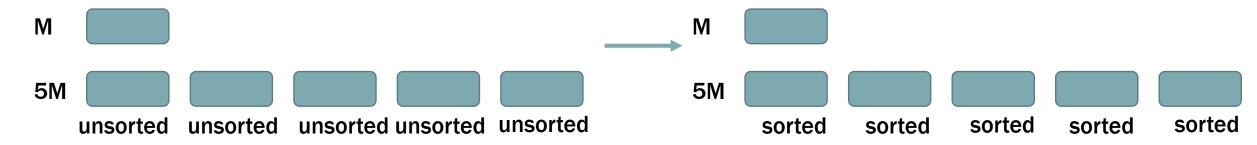
- We can't allocate enough large array in memory when coding.

The problem of sorting a big array larger than the main memory is called external memory sorting.

Suppose the size of main memory is M, the array size is 5M.



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Step 1:

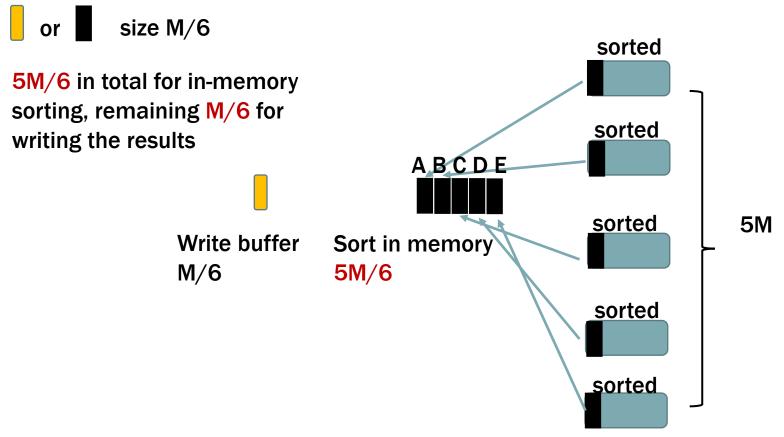
Cut 5M arrays into 5 parts. Each part is put in main memory to sort (using any sorting algorithm we learnt, e.g., quick sort)

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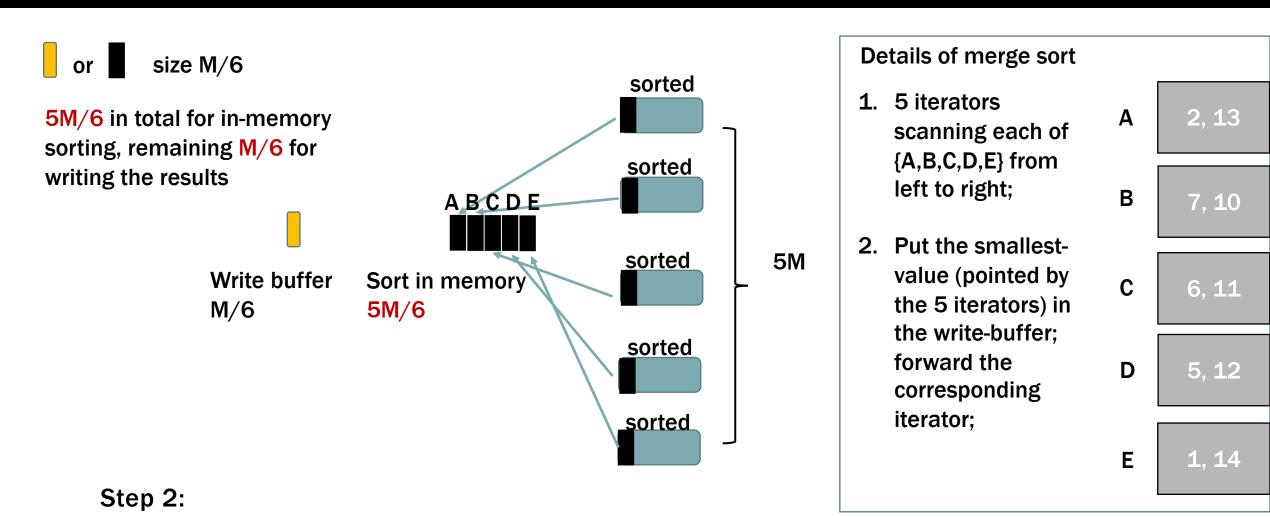
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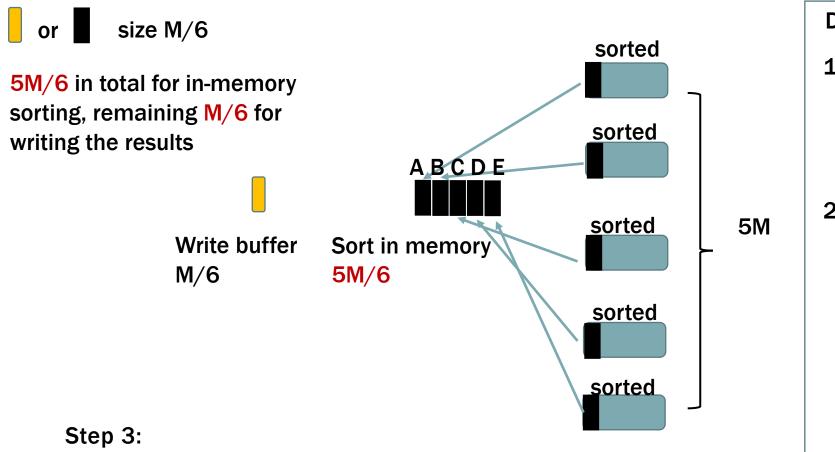


Step 2:

Apply 5-way merge sort to the first M/6 of each sorted part.

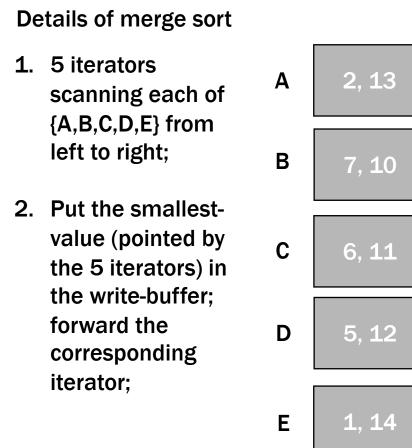


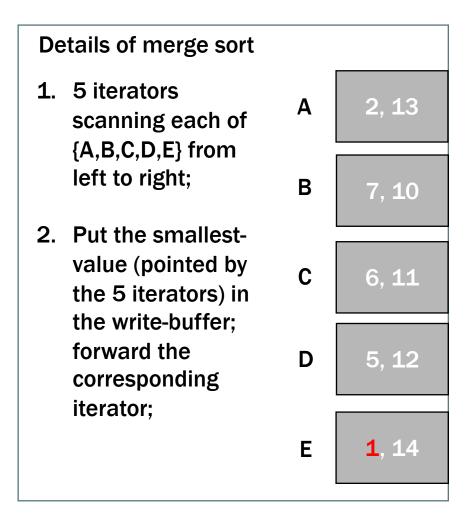
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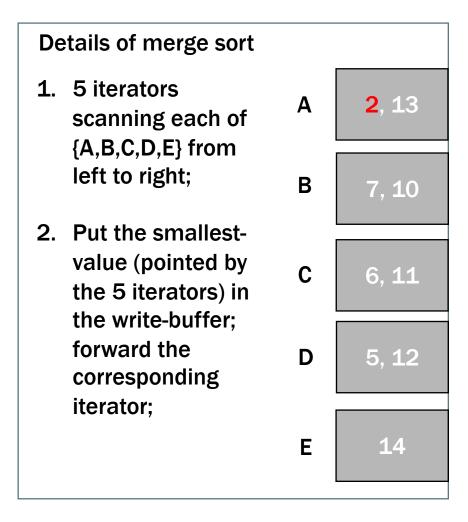
Whenever one of {A,B,C,D,E} is empty, refill it from the source part.

Whenever the write-buffer is full, write it to the disk.



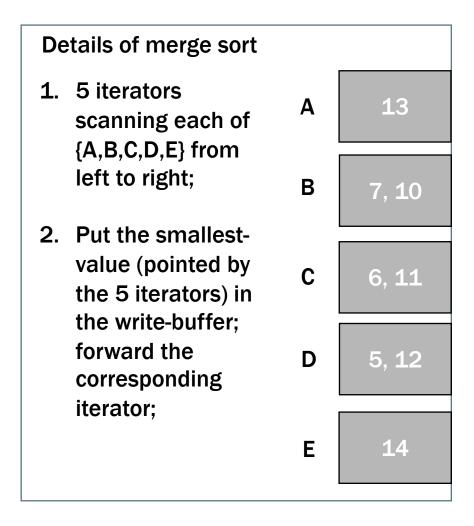


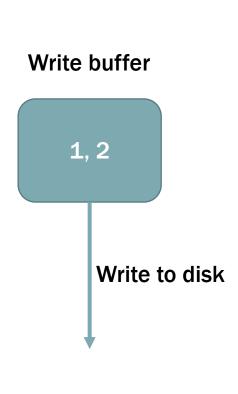
Write buffer (size 2)

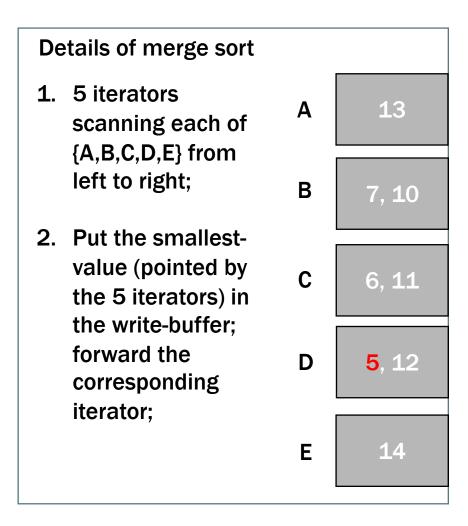


Write buffer

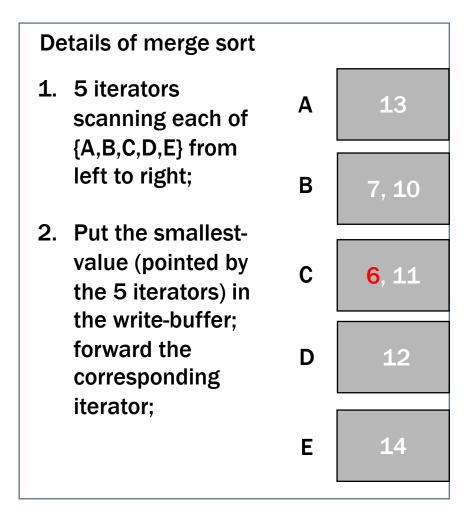
1





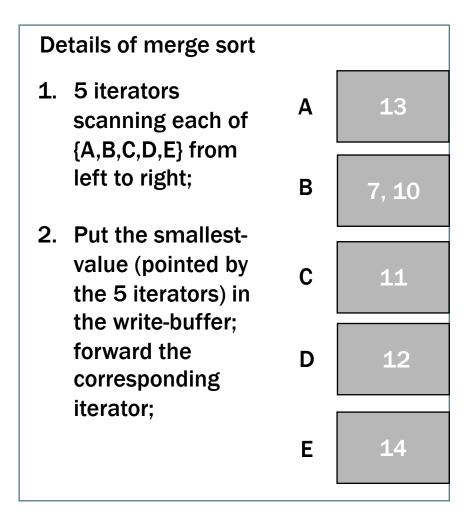


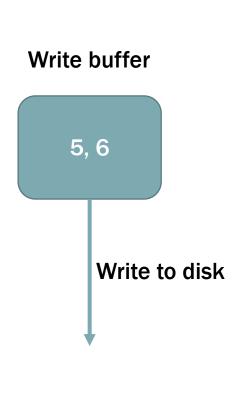


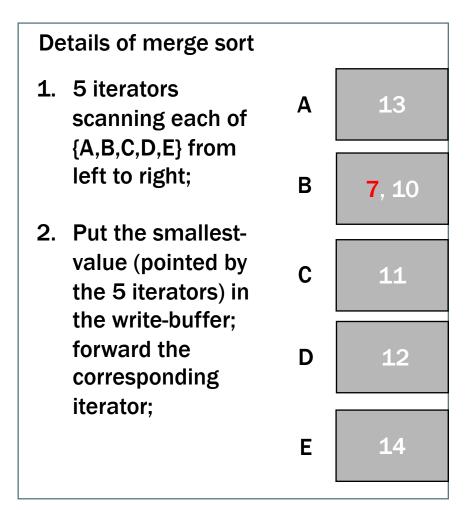


Write buffer

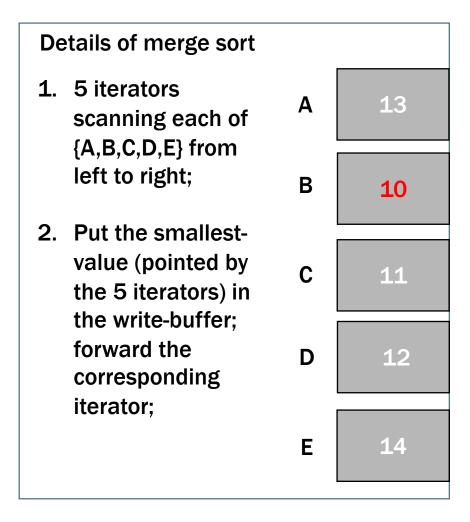
5





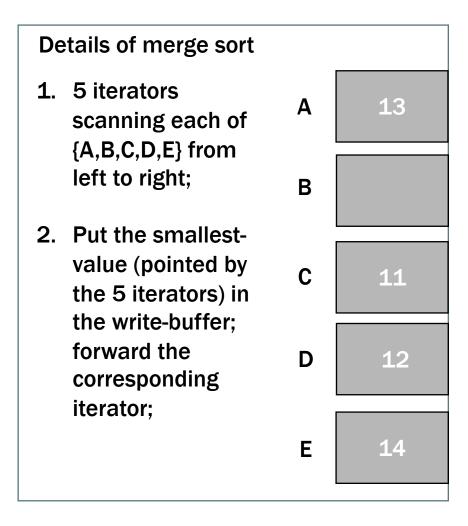






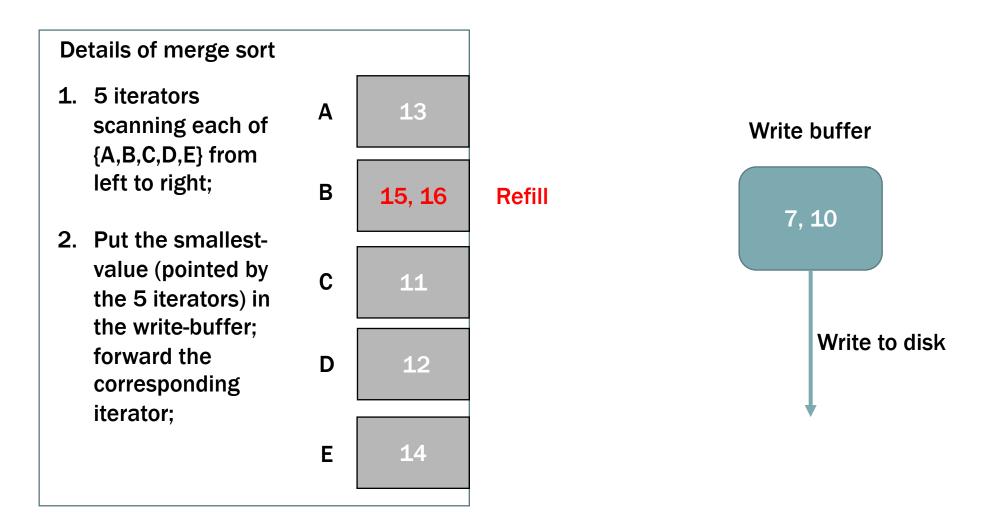
Write buffer

7



Write buffer

7, 10



What's the cost?



Let N be the data size, M be the memory size, and the B be the page (transfer unit) size.

□ Step 1: Read all the N items once and write back to the disk once, incurring O(N/B) I/Os.

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□ Step 2&3: During merge and writeback, each item is read from and written to the disk exactly once, incurring O(N/B) I/Os.

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Does this analysis apply to all situations?



To make the analysis hold, we should have

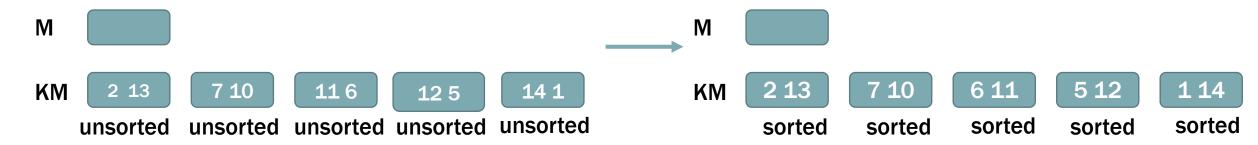
☐ The size of write buffer is at least one page size (size B)

☐ The process of in-memory merge sort should be fit in main memory.

Does this analysis apply to all situations?

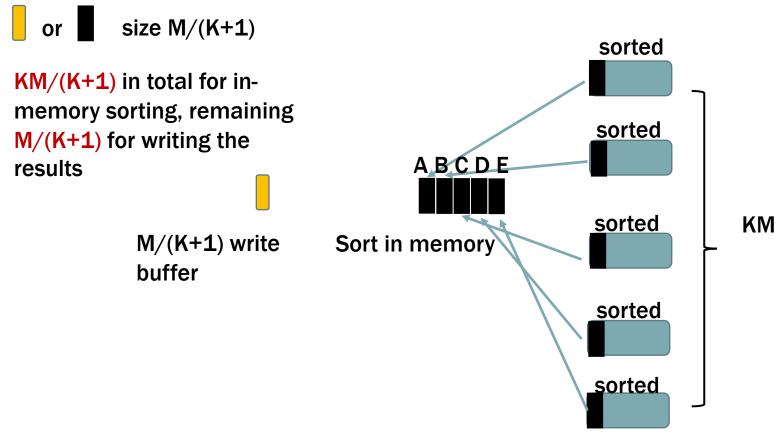


Suppose the size of main memory is M, the array size is KM(=N).



Step 1:

Cut KM arrays into K parts. Each part is put in main memory to sort (using any sorting algorithm we learnt, e.g., quick sort)



Step 2:

Apply K-way merge sort to the first M/(K+1) of each sorted part.

THE RANGE OF K

☐ Condition 1:

A write buffer has at least one page size (denoted by B).

Hence, we have M/(K+1)>=B, giving us $K\leq=M/B-1$

☐ Condition 2:

The process of K-way merge-sort should be fit in remaining memory size (excluding write buffer).

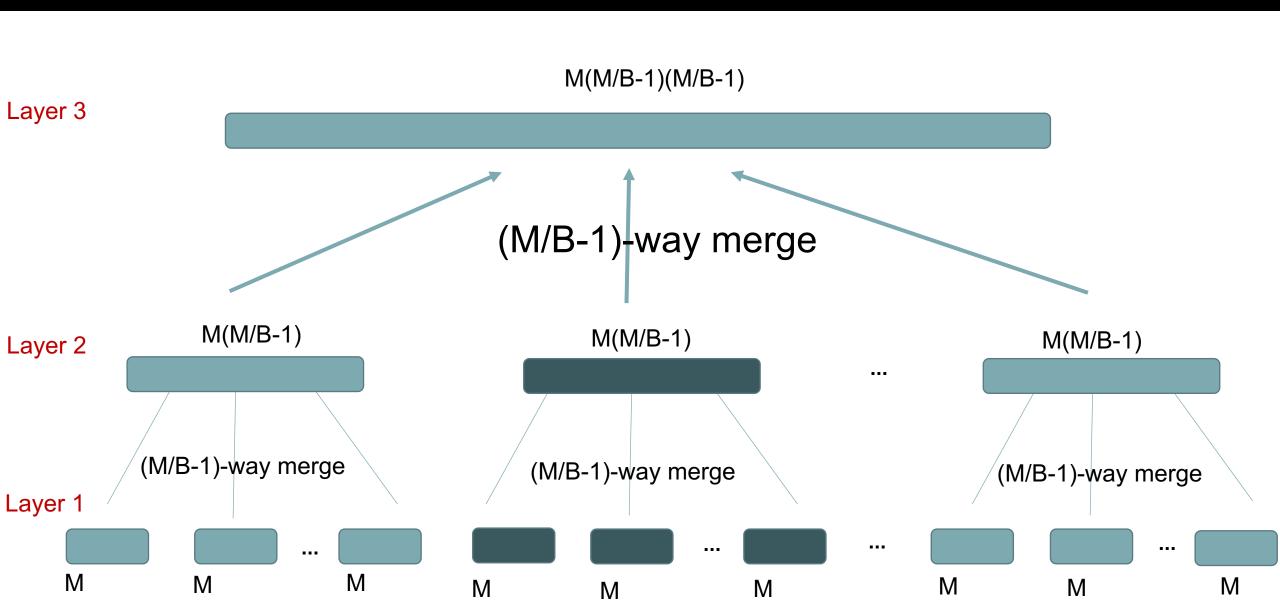
Hence, $K \times B \le KM/(K+1)$, also giving us $K \le M/B-1$

EXTEND TO GENERAL CASE

☐ The cost analysis holds when N is at most (M/B-1)M.

□ We can extend the method to general N by recursively applying the (M/B-1)-way merge.

RECURSIVELY APPLY (M/B-1)-WAY MERGE



GENERAL ANALYSIS

 \Box If there are L layers, then M(M/B-1)^{L-1}= N because N items should be sorted, giving

$$L=O(\log \frac{N/M}{M/B})$$

$$M(M/B-1)^{L-1} = N$$

$$\Leftrightarrow (M/B-1)^{L-1} = N/M$$

$$\Leftrightarrow (M/B-1) = N/M$$

$$\Leftrightarrow \log_{M/B-1}(M/B-1)^{L-1} = \log_{M/B-1} N/M$$

$$\Leftrightarrow L - 1 = \log_{M/B - 1} N/M$$

$$\Leftrightarrow L = 1 + \log_{M/B-1} N/M$$

$$\Leftrightarrow L = O(\log_{M/B} N/M)$$

- ☐ The merges in each layer costs O(N/B) I/Os.
- \square Hence, the total cost is O((N/B) $\log_{M/B}^{N/M}$) I/Os.

We finish Memory Hierarchy! (



Next lecture:

Column Store