

# Sorting

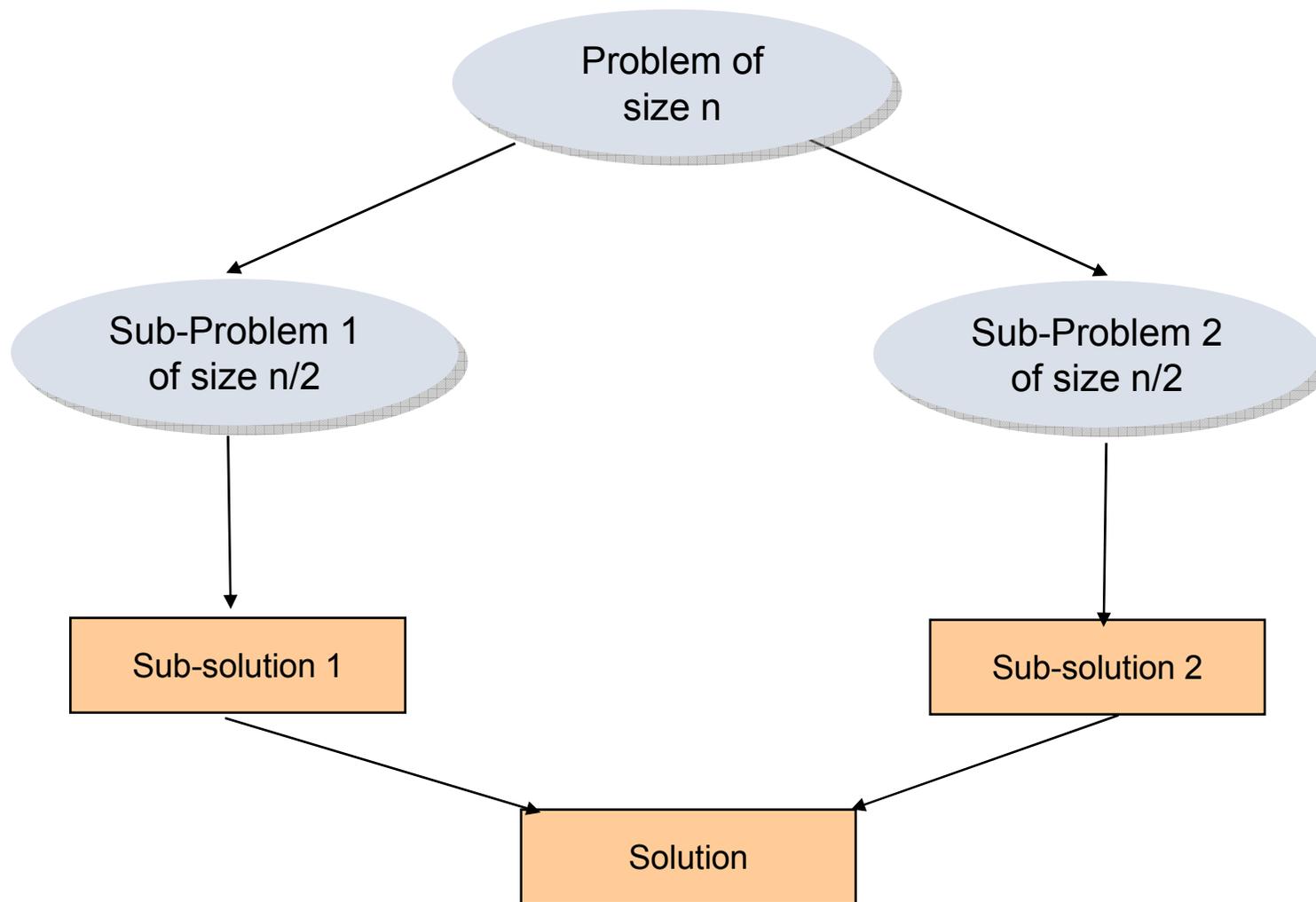
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# Overview

- » Divide and Conquer
- » Bubble Sort
- » Naïve Parallel Sort
- » Merging
- » Parallel Mergesort
- » Parallel Quicksort

# Divide and Conquer





# Parallel Divide and Conquer

- » Execute the sub tasks in parallel
- » Get sub solution to each sub task
- » Combine sub solutions into the solution
  - *The combination step sometime can be parallel, for example. Merging in Mergesort*



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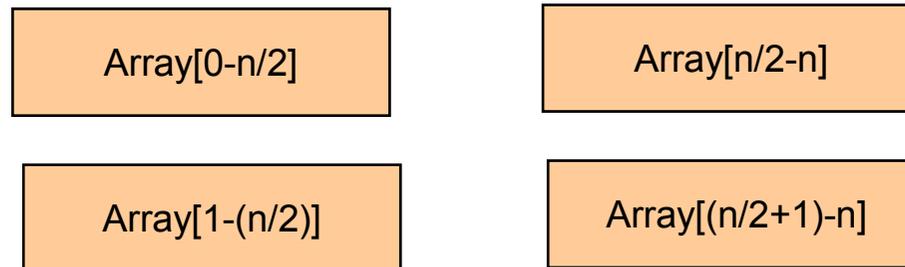


# Bubble Sort

- » Bubble sort is a *very very* inefficient sorting algorithm. But for some reason, it shows up in almost all algorithm textbooks.
- » It swaps two adjacent items if the two items are not ordered until there are no items that can be swapped.
- » The complexity of bubble sort is  $\text{sqr}(n)$  as opposed to  $n \cdot \log(n)$  of most other well-known sorting algorithms such as mergesort or quicksort

# Parallel Bubble Sort

- » Divide the array into  $n$  sub-arrays so that each sub-array is assigned to a thread. Each thread executes a bubble sort
- » Re-divide the array into  $n$  sub-arrays by shifting one item away and assign them into the thread pool



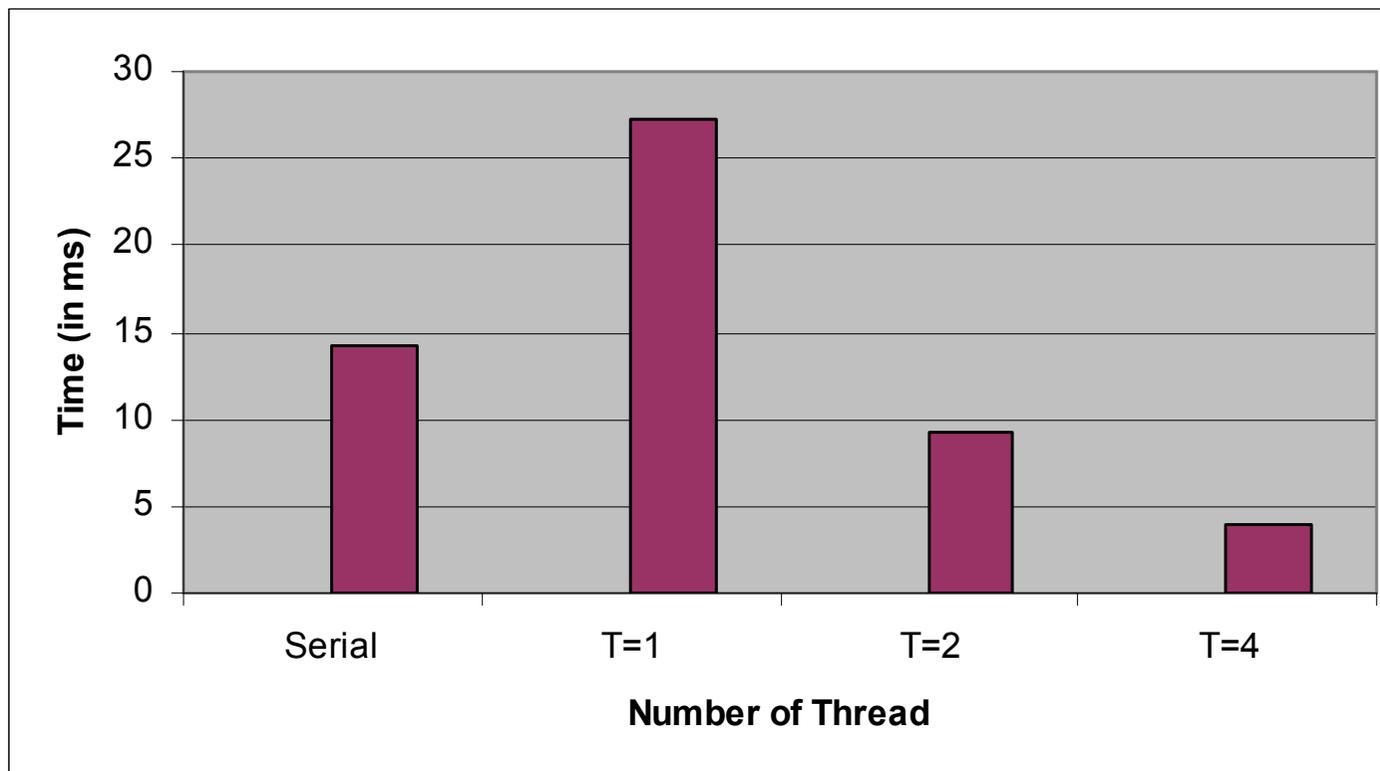
- » The algorithm stops when there is no swap going on, also called odd-even sort.

# Example 1: Parallel Bubble Sort

```
» while (true) {
»     for (int i = 0; i < numOfThreads; i++) {
»         int endPos = (i+1)*len ;
»         if (endPos > MAX_LENGTH) endPos = MAX_LENGTH;
»         threadPool[i] = new BubbleSort(i*len +1, endPos);
»         threadPool[i].start();
»     }
»
»     for (int i = 0; i < numOfThreads; i++) { threadPool[i].join(); }
»
»     for (int i = 0; i < numOfThreads; i++) {
»         int endPos = (i+1)*len ;
»         if (endPos > MAX_LENGTH) endPos = MAX_LENGTH;
»         threadPool[i] = new BubbleSort(i*len, endPos);
»         threadPool[i].start();
»     }
»
»     for (int i = 0; i < numOfThreads; i++) {threadPool[i].join();}
»
»     boolean allDone = true;
»     for (int i = 0; i < numOfThreads; i++) {
»         if (threadPool[i].done == false) { allDone = false;break;}
»     }
»     if (allDone) break;
» }
```

# Parallel Bubble Sort Performance

- » Configuration: 4-core, 2.66Ghz, 4G RAM with Windows Vista 64-bit. The to be sorted array has 24K elements with “String” type, and each String is of length between 1-10.
- » The reason T=1 is worse than Serial is there are some “join” operations in T=1 in order to make the implementation universal.
- » If we just compare Serial/T=2/T=4, we got pretty reasonable speedup.





# Overview

- » Divide and Conquer
- » Bubble Sort
- » ***Naïve Parallel Sort***
- » Merging
- » Parallel Mergesort
- » Parallel Quicksort

# Take Advantage of Existing Sort Lib

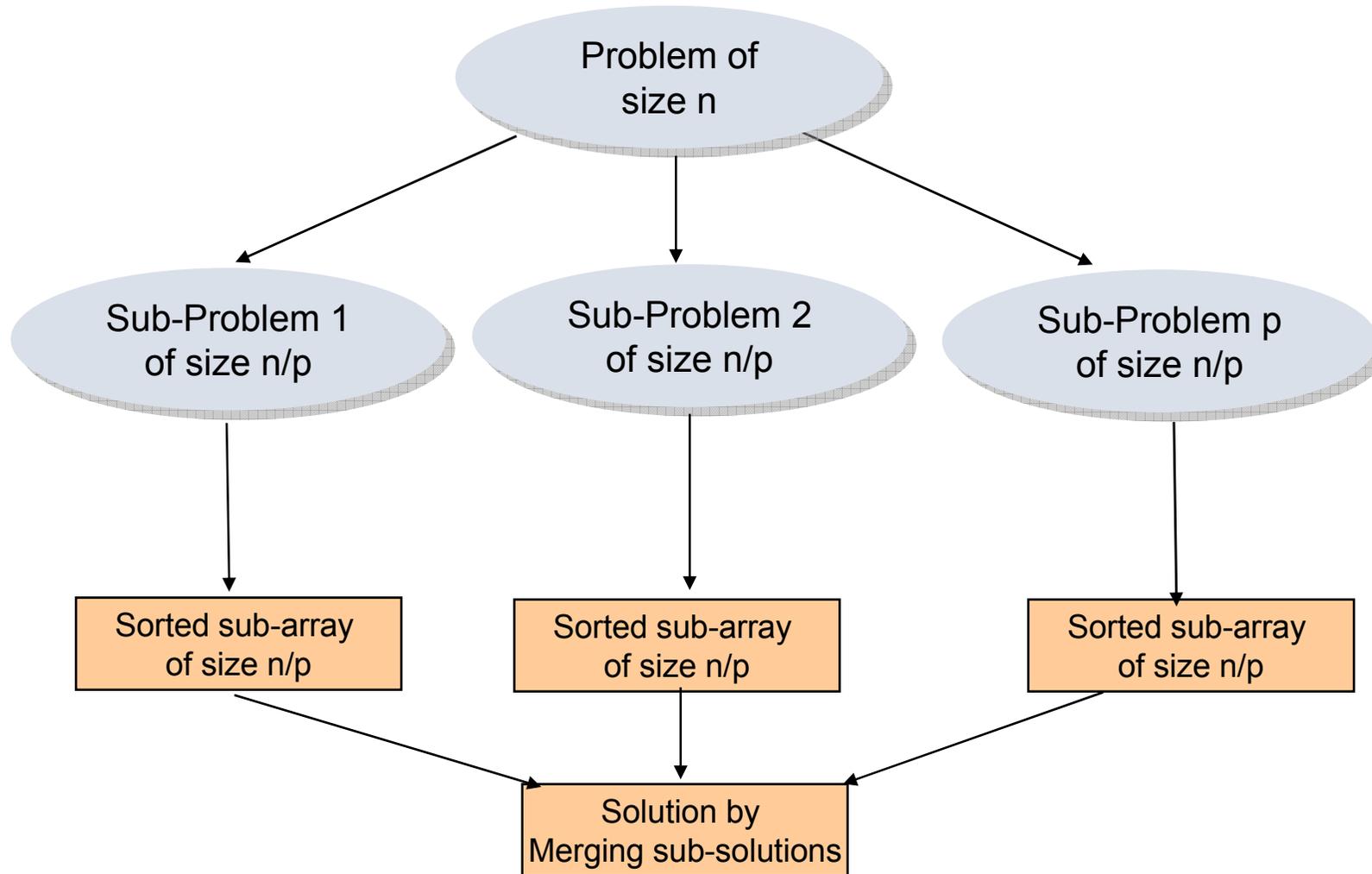
- » In `java.util.Arrays`, there are a series of “`sort()`” methods
  - *For arrays with primitive types, such as `int`, `long`, `float`, `double`, it performs “quicksort”*
    - *Quicksort has average performance of  $n \cdot \log(n)$ , worst case at  $n^2$ . Quicksort is a very interesting algorithm that its average performance is way better than the worst performance. Even today a lot of research has been conducted in different variations.*
  - *For other array types, it performs “mergesort”, an  $n \cdot \log(n)$  algorithm but sometimes not as fast as quicksort mainly because of its array copying operations*
- » We can divide the array into several sub-arrays and assign sub-arrays to several threads to perform “sort” in parallel among sub-arrays
- » After all sub-arrays are sorted, a “merging” operation is performed to merge all sorted sub-arrays



# Naïve Parallel Sorting

- » We can divide the array into several sub-arrays and assign different sub-arrays to different threads to perform serial “sort” (in existing library) in parallel among sub-arrays
- » After all sub-arrays are sorted, a “merging” operation is performed to merge all sorted sub-arrays

# Naïve Parallel Sorting

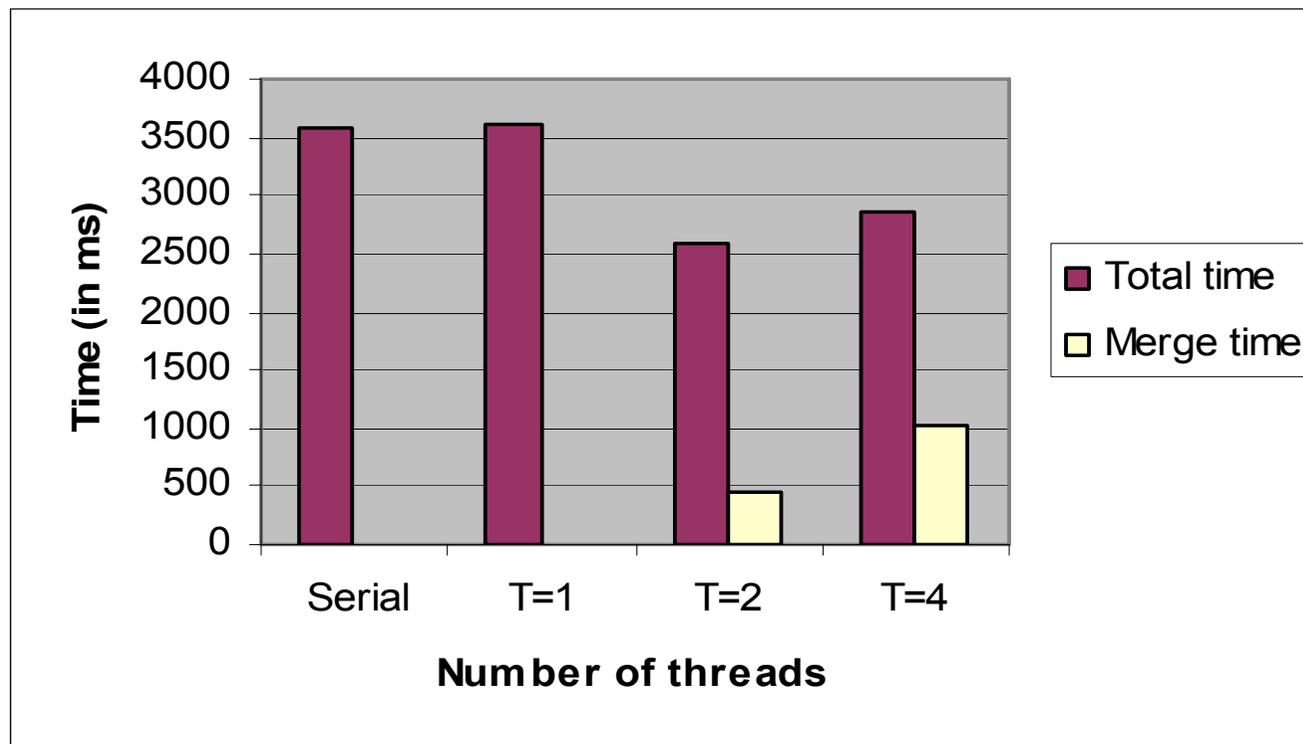


## Example 2: Naïve Parallel Sort

```
» for (int i = 0; i < numOfThreads; i++) {
»     int endPos = (i+1)*len;
»     if (endPos > MAX_LENGTH) endPos = MAX_LENGTH;
»     threadPool[i] = new MergeSort1(i*len, endPos); // java.util.Arrays.sort()
»     threadPool[i].start();
» }
»
» for (int i = 0; i < numOfThreads; i++) {threadPool[i].join();}
» for (int i = 0; i < (numOfThreads-1); i++) {
»     String[] tgt = merge(stringArray, 0, (i+1)*len, stringArray, (i+1)*len, (i+2)*len);
»     System.arraycopy(tgt, 0, stringArray, 0, tgt.length);
» }
» }
```

# Naïve Parallel Sort Performance

- » Configuration: 4-core, 2.66Ghz, 4G RAM with Windows Vista 64-bit. The to be sorted array has 2.4M elements with “String” type, and each String is of length between 1-10.
- » When number of threads increases to 4, no performance gain. Merging takes substantial amount of time



# Advantage of Naïve Parallel Sort

- » It's simple, no need to understand the internals of serial sorting algorithms
- » It can work with any existing serial sorting algorithms
- » Reasonable performance when number of processors is small
  - $O(n \cdot \log(n/p)/p + n) = O(n \cdot \log(n)/p - n \log(p)/p + n) = O(n \cdot \log(n)/p)$



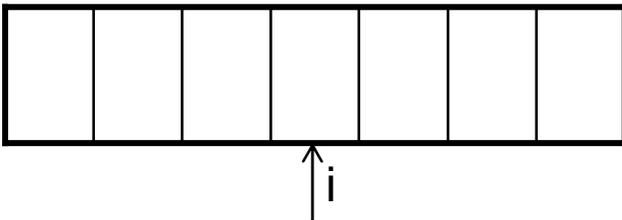
# Overview

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- » Bubble Sort
- » Naïve Parallel Sort
- » ***Merging***
- » Parallel Mergesort
- » Parallel Quicksort

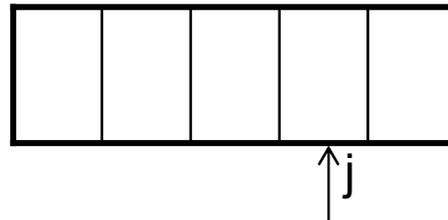
# Merging: The Serial Algorithm

- » How do we do serial merging
- » Two sorted arrays, merge them into a bigger sorted array
  - $O(m+n)$

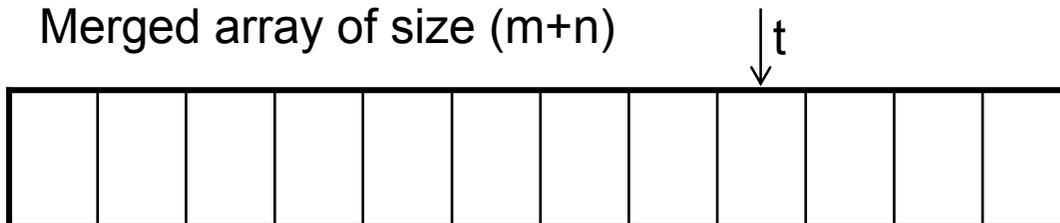
Sorted array 1 of size m



Sorted array 2 of size n



Merged array of size (m+n)



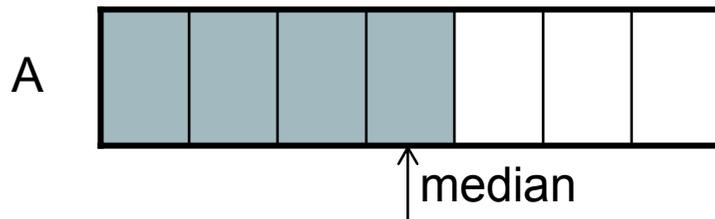
# Merging: Serial Algorithm

```
static String[] merge(String[] src1, int firstStart, int firstEnd, String[] src2, int secondStart, int
    secondEnd) {
    »     String[] tgt = new String[secondEnd-secondStart+firstEnd-firstStart];
    »     int fi = firstStart;
    »     int si = secondStart;
    »     for (int i = 0; i < tgt.length; i++) {
    »         if (fi >= firstEnd) {
    »             System.arraycopy(src2, si, tgt, i, (secondEnd-si));
    »             break;
    »         }
    »         if (si >= secondEnd) {
    »             System.arraycopy(src1, fi, tgt, i, (firstEnd-fi));
    »             break;
    »         }
    »         if (src1[fi].compareTo(src2[si]) <= 0) {
    »             tgt[i] = src1[fi]; fi++;
    »         } else {
    »             tgt[i] = src2[si]; si++;
    »         }
    »     }
    »     return tgt;
    » }
```

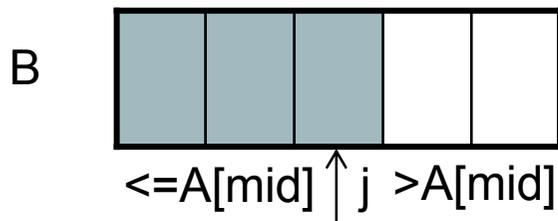
# Parallel Merging: The Parallel Algorithm

- » How do we do parallel merging? Divide and Conquer
  - Divide the longer array  $A$  into two halves using the middle item
  - Do a binary search on  $B$  using the media of  $A$  and then divide the shorter array  $B$  using the middle item in the longer array
  - Merge the 1<sup>st</sup> half of  $A$  with the first part of  $B$ ; and second half of  $B$  with second part of  $A$  in parallel

Sorted array 1 of size  $m$



Sorted array 2 of size  $n$



Merged array of size  $(m+n)$



# Example 3: Implementation of Parallel Merging

```
» static void pMerge(String[] tgt, int tStart, String[] src1, int start1, int end1, String[] src2, int start2, int end2)
» {
»     if ((end1-start1) < (end2-start2)) {
»         pMerge(tgt, tStart, src2, start2, end2, src1, start1, end1);
»         return;
»     }
»     if ((end1-start1) < (SERIAL_THRESHOLD+1)) {
»         merge(tgt, tStart, src1, start1, end1, src2, start2, end2);
»         return;
»     }
»     int halfLen1 = (end1-start1)>>>1;
»     int med = start1 + halfLen1;
»     int par2 = find(src1[med], src2, start2, end2);    // binary search
»     PMerge pm1 = new PMerge(tgt, tStart, src1, start1, med, src2, start2, par2);
»     PMerge pm2 = new PMerge(tgt, tStart+ halfLen1+(par2-start2), src1, med, end1, src2, par2,end2);
»     try {
»         pm1.join();
»         pm2.join();
»     } catch (InterruptedException ie) {
»         System.out.println(ie);
»     }
» }
```

# Improvement of Naïve Parallel Sort

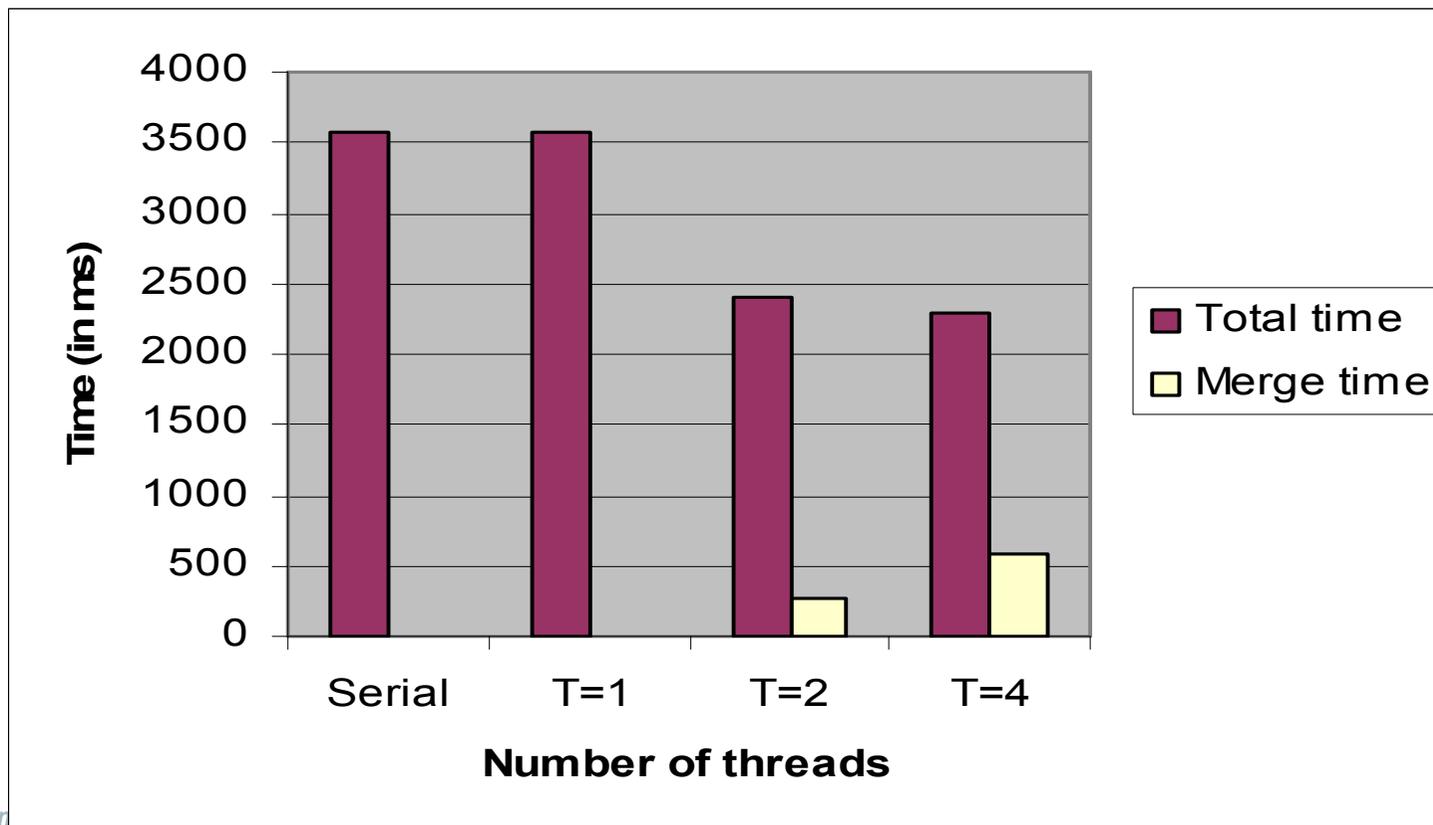
- » Replacement of serial merging with parallel merging in the naïve algorithm
- » It can work with any sorting algorithms
- » Reasonably good performance without much work
  - $O(n \cdot \log(n/p)/p + n/p) = O(n \cdot \log(n)/p - n \log(p)/p + n/p) = O(n \cdot \log(n)/p)$

## Example 4: Improvement of Naïve Parallel Sort

```
» for (int i = 0; i < numOfThreads; i++) {
»     int endPos = (i+1)*len;
»     if (endPos > MAX_LENGTH) endPos = MAX_LENGTH;
»     threadPool[i] = new MergeSort1(i*len, endPos); // java.util.Arrays.sort()
»     threadPool[i].start();
» }
»
» for (int i = 0; i < numOfThreads; i++) {threadPool[i].join();}
» for (int i = 0; i < (numOfThreads-1); i++) {
»     String[] tgt = PMerge.pmerge(stringArray, 0, (i+1)*len, stringArray, (i+1)*len,
» (i+2)*len);
»     System.arraycopy(tgt, 0, stringArray, 0, tgt.length);
» }
» }
```

# Improvement of Naïve Parallel Sort: Performance

- » Configuration: 4-core, 2.66Ghz, 4G RAM with Windows Vista 64-bit. The to be sorted array has 2.4M elements with “String” type, and each String is of length between 1-10.
- » Merge time has been substantially reduced. When number of threads increases to 4, some marginal performance gain.



# Naïve Algorithm Not Good Enough

- » Still a lot of room to improve
  - *Need to work the parallel algorithm inside out*
- » The good old Divide and Conquer technique
- » Recursive and Parallel
  - *Study the classical recursive algorithm and make it parallel*
- » Both merge-sort and quick-sort can be made into parallel quite easily
  - *Simply change the divide-and-conquer recursiveness into divide-and-conquer parallelism*



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- » ***Parallel Mergesort***



# MergeSort: Classical and Parallel Algorithms

## » Divide-and-Conquer

```
1. Mergesort(A) {
2.     Divide a given array A into two parts B and C
3.     Mergesort(B);
4.     Mergesort(C);
5.     Merge B and C into a new array D;
6.     Copy D to A
7. }
```

## » Divide-and-Conquer in Parallel

```
1. Mergesort(A) {
2.     Divide a given array A into two parts B and C
3.     Spawn Mergesort(B);
4.     Spawn Mergesort(C);
5.     join(); // waiting spawned threads to finish
6.     Parallel Merge B and C into a new array D
7.     Copy D to A
8. }
```

# Example 5: Parallel Mergesort

```
» private final void pSort() {
»     if (endPos <= startPos)
»         return;
»     if ((endPos - startPos) < (SERIAL_THRESHOLD+1)) {
»         Arrays.sort(stringArray, startPos, endPos);
»         return;
»     }
»     int halfLen = (endPos - startPos) >>>1;
»     int medPos = startPos + halfLen;
»     MergeSort3 ms1 = new MergeSort3(stringArray, startPos, medPos);
»     MergeSort3 ms2 = new MergeSort3(stringArray, medPos, endPos);
»     try {
»         ms1.join();
»         ms2.join();
»     } catch (InterruptedException ie) {
»     }
»     String[] tgt = new String[endPos-startPos];
»     PMerge.MAX_LENGTH = MAX_LENGTH;
»     PMerge pm = new PMerge(tgt, 0, stringArray, startPos, medPos, stringArray, medPos, endPos);
»     try {
»         pm.join();
»     } catch (InterruptedException ie) {
»     }
»     System.arraycopy(tgt, 0, stringArray, startPos, tgt.length);
» }
```

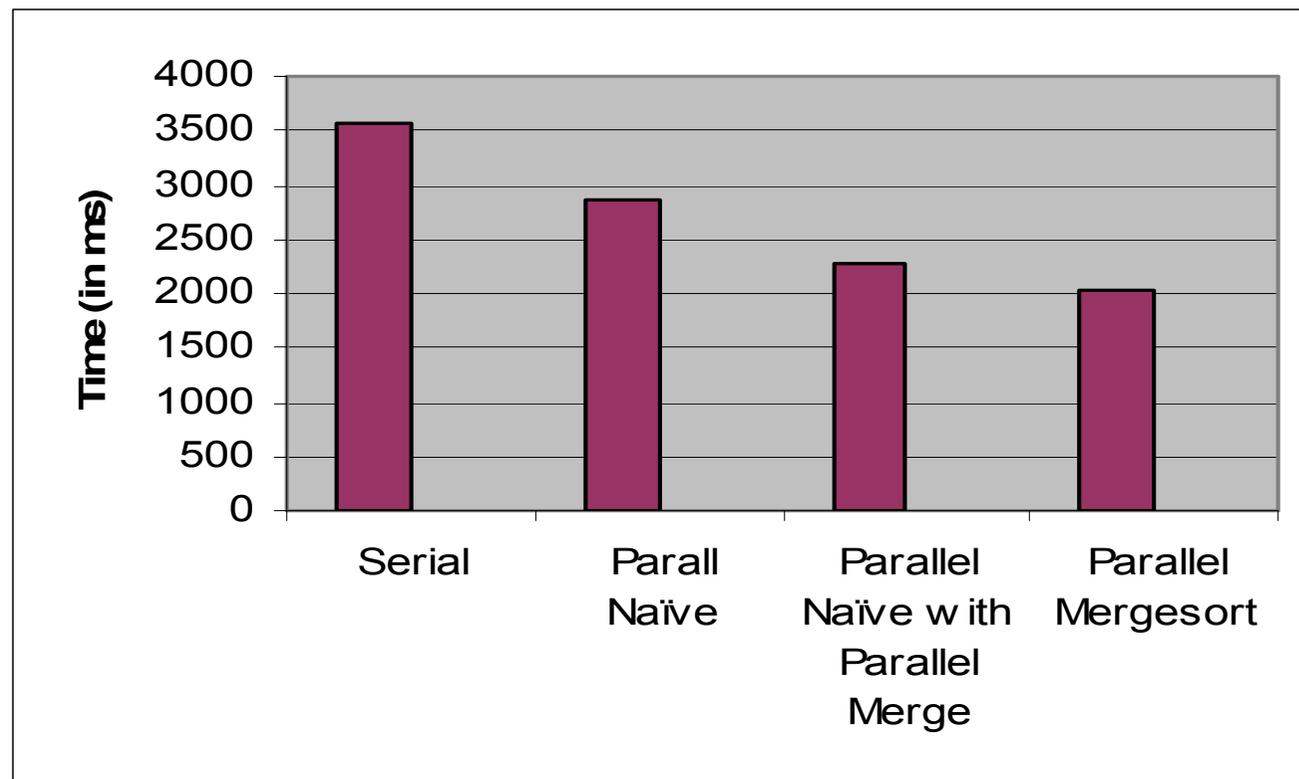


## Example 5: Parallel MergeSort

- » A parallel implementation of divide-and-conquer
  - *The recursion is implemented in parallel*
- » When the parallelism is reached to its threshold, i.e., the number of threads is larger than the number of physical processors, the sub-array sort is done using Java's `Arrays.sort()`

# Performance of Parallel MergeSort

- » With same configuration machine, number of cores is 4, performance comparison among different algorithms.
- » Sorting in general is a memory intensive, when more CPUs are added, memory bandwidth becomes bottleneck





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# Quicksort: Classical and Parallel Algorithms

## » Divide-and-Conquer

```
1. Quicksort(A, l, r) {  
2.     s = Partition(A, l, r); s is the split point  
3.     Quicksort(A, l, s-1);  
4.     Quicksort(A, s+1, r);  
5. }
```

## » Divide-and-Conquer in Parallel

```
1. Quicksort(A, l, r) {  
2.     s = ParallelPartition(A, l, r);  
3.     Spawn Quicksort(A, l, s-1);  
4.     Spawn Quicksort(A, s+1, r);  
5.     join(); // waiting spawned threads to finish  
6. }
```

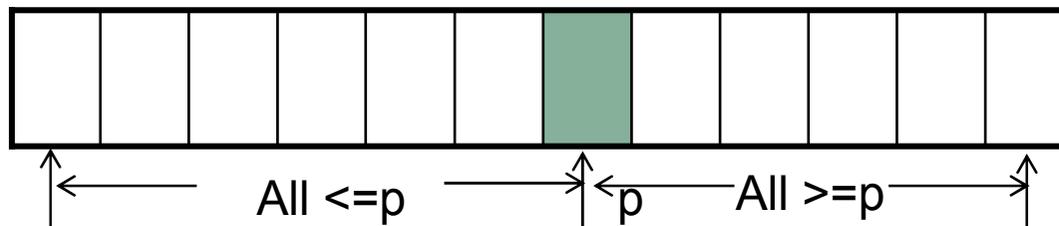
# Quicksort

- » The key to the quicksort algorithm is the partition.
- » Find a pivot point  $p$  such that left are all  $\leq$  and right are all  $\geq$

Array of size  $(m+n)$



Partitioned sub arrays of size  $(m+n)$





# Parallel Quicksort

- » Again, we look at the classical Quicksort algorithm, try to add parallelism into it.
- » It's the familiar divide-and-conquer idea used in MergeSort: the recursive divide-and-conquer part is, again, done in parallel.
- » Similar to MergeSort where key is to parallelize merging, the key here is to parallelize the partition

# Parallel Partition

- » Suppose we have  $p$  processors and we divide the entire array into  $p$  segments and each processor is assigned a segment.
- » We pick a pivot point across all segments. Each processor  $i$  try to partition its own segment into two parts,  $L_i$  and  $U_i$ .
- » All  $L_i$ s are merged and  $U_i$ s are merged.



Split point



# Parallel Partition

- » The complexity of serial partition is linear. The complexity of serial partition is  $n/p$  on average if the segments are evenly distributed.
- » But we lose some of desired property of quicksort, such as in-place sort. In the second phase of parallel partition, we have to do merging, which essentially is an array-copy operation.