Divide and Conquer

Divide-and-Conquer

- Divide-and-conquer.
 - Break up problem into several parts.
 - Solve each part recursively.
 - Combine solutions to sub-problems into overall solution.
- Most common usage.
 - Break up problem of size n into two equal parts of size $\frac{1}{2}$ n.
 - Solve two parts recursively.
 - Combine two solutions into overall solution in linear time.
- · Consequence.
- Brute force: n².
- Divide-and-conquer: n log n.
- Analysis.
- involves solving a recurrence relation that bounds the running time recursively in terms of the running time on smaller instances.

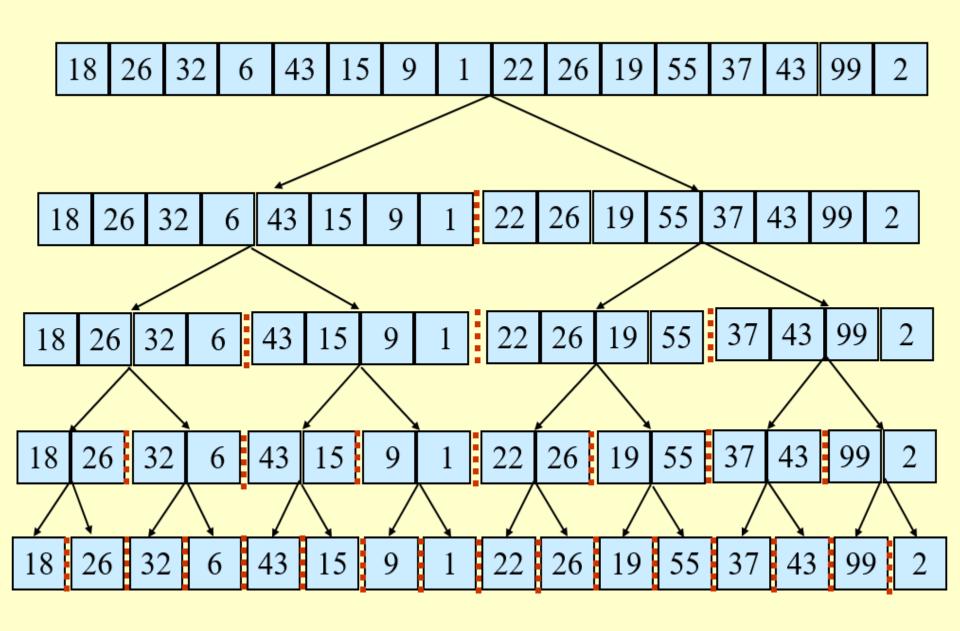
Merge Sort

Merge Sort

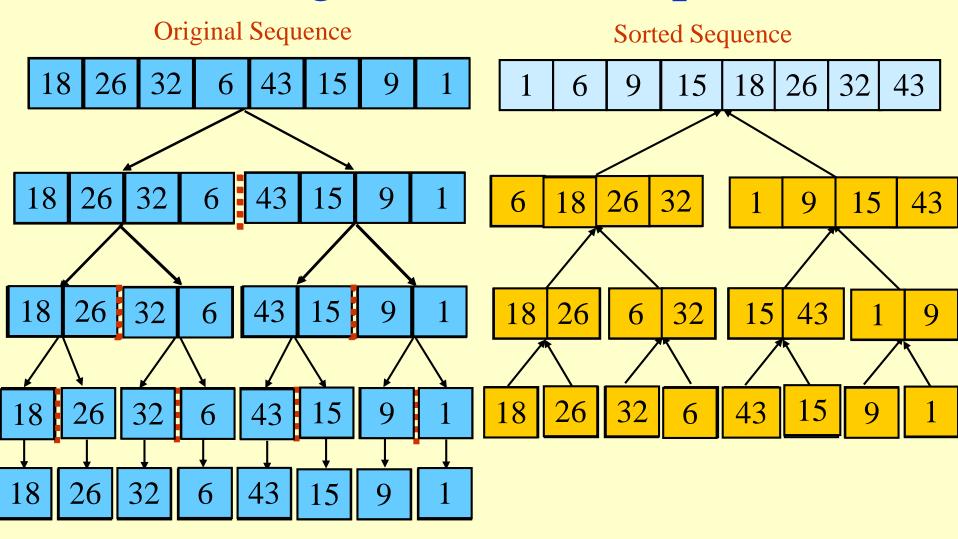
Sorting Problem: Sort a sequence of *n* elements into non-decreasing order.

- *Divide*: Divide the *n*-element sequence to be sorted into two subsequences of *n*/2 elements each
- *Conquer:* Sort the two subsequences recursively using merge sort.
- *Combine*: Merge the two sorted subsequences to produce the sorted answer.

Merge Sort Example



Merge Sort – Example



Merge-Sort (A, p, r)

INPUT: a sequence of *n* numbers stored in array A **OUTPUT:** an ordered sequence of *n* numbers

```
MergeSort (A, p, r) // sort A[p..r] by divide & conquer1 if p < r2 then q \leftarrow \lfloor (p+r)/2 \rfloor3 MergeSort (A, p, q)4 MergeSort (A, q+1, r)5 Merge (A, p, q, r) // merges A[p..q] with A[q+1..r]
```

Initial Call: MergeSort(A, 1, n)

Procedure Merge

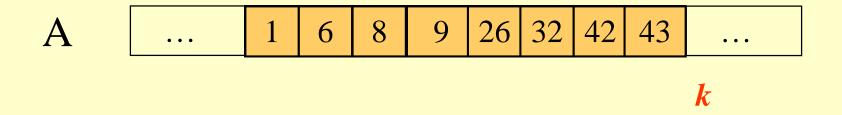
```
Merge(A, p, q, r)
1 n_1 \leftarrow q - p + 1
2 n_2 \leftarrow r - q
         for i \leftarrow 1 to n_1
             do L[i] \leftarrow A[p+i-1]
         for j \leftarrow 1 to n_2
             \operatorname{do} R[j] \leftarrow A[q+j]
      L[n_1+1] \leftarrow \infty
        R[n_2+1] \leftarrow \infty
         i \leftarrow 1
      j \leftarrow 1
10
         for k \leftarrow p to r
11
             do if L[i] \leq R[j] \leftarrow
12
13
                 then A[k] \leftarrow L[i]
14
                          i \leftarrow i + 1
                 else A[k] \leftarrow R[j]
15
                         j \leftarrow j + 1
16
```

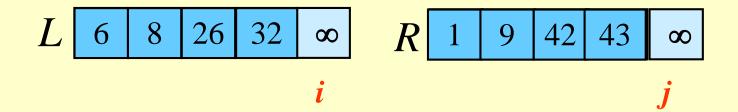
Input: Array containing sorted subarrays A[p..q] and A[q+1..r].

Output: Merged sorted subarray in A[p..r].

Sentinels, to avoid having to check if either subarray is fully copied at each step.

Merge – Example





Correctness of Merge

```
Merge(A, p, q, r)
1 n_1 \leftarrow q - p + 1
2 n_2 \leftarrow r - q
         for i \leftarrow 1 to n_1
             do L[i] \leftarrow A[p+i-1]
         for j \leftarrow 1 to n_2
             \operatorname{do} R[j] \leftarrow A[q+j]
       L[n_1+1] \leftarrow \infty
         R[n_2+1] \leftarrow \infty
         i \leftarrow 1
       j \leftarrow 1
10
         for k \leftarrow p to r
11
             do if L[i] \leq R[j]
12
                 then A[k] \leftarrow L[i]
13
14
                          i \leftarrow i + 1
                 else A[k] \leftarrow R[j]
15
                         j \leftarrow j + 1
16
```

Loop Invariant for the *for* **loop**

At the start of each iteration of the for loop:

Subarray A[p..k-1] contains the k-p smallest elements of L and R in sorted order. L[i] and R[j] are the smallest elements of L and R that have not been copied back into A.

Initialization:

Before the first iteration:

- •A[p..k-1] is empty.
- •i = j = 1.
- •L[1] and R[1] are the smallest elements of L and R not copied to A.

Correctness of Merge

```
Merge(A, p, q, r)
1 n_1 \leftarrow q - p + 1
2 n_2 \leftarrow r - q
          for i \leftarrow 1 to n_1
              \operatorname{do} L[i] \leftarrow A[p+i-1]
         for j \leftarrow 1 to n_2
              \operatorname{do} R[j] \leftarrow A[q+j]
         L[n_1+1] \leftarrow \infty
         R[n_2+1] \leftarrow \infty
         i \leftarrow 1
        j \leftarrow 1
10
          for k \leftarrow p to r
11
              do if L[i] \leq R[j]
12
13
                  then A[k] \leftarrow L[i]
14
                           i \leftarrow i + 1
                  else A[k] \leftarrow R[j]
15
                           j \leftarrow j + 1
16
```

Maintenance:

Case 1: $L[i] \leq R[j]$

- •By LI, A contains p k smallest elements of L and R in sorted order.
- •By LI, L[i] and R[j] are the smallest elements of L and R not yet copied into A.
- •Line 13 results in A containing p k + 1 smallest elements (again in sorted order). Incrementing i and k reestablishes the LI for the next iteration.

Similarly for L[i] > R[j].

Termination:

- •On termination, k = r + 1.
- •By LI, A contains r p + 1 smallest elements of L and R in sorted order.
- •*L* and *R* together contain r p + 3 elements. All but the two sentinels have been copied back into *A*.

Analysis of Merge Sort

- Running time T(n) of Merge Sort:
- Divide: computing the middle takes $\Theta(1)$
- Conquer: solving 2 subproblems takes 2T(n/2)
- Combine: merging n elements takes $\Theta(n)$
- ◆ Total:

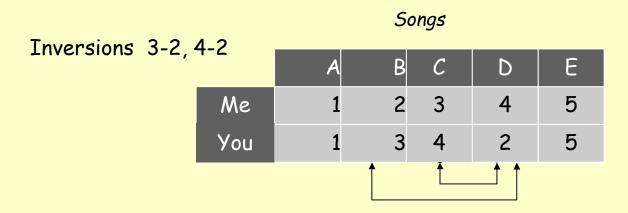
$$T(n) = \Theta(1)$$
 if $n = 1$
 $T(n) = 2T(n/2) + \Theta(n)$ if $n > 1$

```
\Rightarrow T(n) = \Theta(n \lg n)
```

Counting Inversions

Counting Inversions

- Music site tries to match your song preferences with others.
- You rank n songs.
- Music site consults database to find people with similar tastes.
- Similarity metric: number of inversions between two rankings.
- My rank: 1, 2, ..., n.
- Your rank: a₁, a₂, ..., a_n.
- Songs i and j inverted if i < j, but a_i > a_j.



Brute force: check all $\Theta(n^2)$ pairs i and j.

Applications

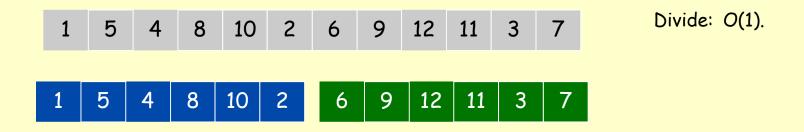
- Applications.
 - Voting theory.
 - . Collaborative filtering.
 - Measuring the "sortedness" of an array.
 - Sensitivity analysis of Google's ranking function.
 - Rank aggregation for meta-searching on the Web.
 - Nonparametric statistics (e.g., Kendall's Tau distance).

Divide-and-conquer.



Divide-and-conquer.

Divide: separate list into two pieces.



Divide-and-conquer.

- Divide: separate list into two pieces.
- Conquer: recursively count inversions in each half.



5-4, 5-2, 4-2, 8-2, 10-2

6-3, 9-3, 9-7, 12-3, 12-7, 12-11, 11-3, 11-7

Divide-and-conquer.

Divide: separate list into two pieces.

Conquer: recursively count inversions in each half.

Combine: count inversions where a_i and a_j are in different halves,

and return sum of three quantities.



Combine: ???

9 blue-green inversions 5-3, 4-3, 8-6, 8-3, 8-7, 10-6, 10-9, 10-3, 10-7

Total = 5 + 8 + 9 = 22.

Counting Inversions: Combine

Combine: count blue-green inversions

- Assume each half is sorted.
- ${}_{\circ}$ Count inversions where a_i and a_j are in different halves.
- Merge two sorted halves into sorted whole.

to maintain sorted invariant

2	11	16	17	23	25
6	3	2	2	0	0

13 blue-green inversions: 6 + 3 + 2 + 2 + 0 + 0

Count: O(n)

Merge: O(n)

$$T(n) \leq T\Big(\left\lfloor n/2\right\rfloor\Big) + T\Big(\left\lceil n/2\right\rceil\Big) + O(n) \implies \mathrm{T}(n) = O(n\log n)$$

Counting Inversions: Implementation

Pre-condition. [Merge-and-Count] A and B are sorted. Post-condition. [Sort-and-Count] L is sorted.

```
Sort-and-Count(L) {
if list L has one element return 0 and the
list L

Divide the list into two halves A and B
    (r<sub>A</sub>, A) ← Sort-and-Count(A)
    (r<sub>B</sub>, B) ← Sort-and-Count(B)
    (r , L) ← Merge-and-Count(A, B)

return r = r<sub>A</sub> + r<sub>B</sub> + r and the sorted list L
}
```

Quick Sort

Quick Sort

```
Quicksort(A, p, r)

1 if p < r

2 then q \leftarrow Partition(A, p, r)

3 Quicksort(A, p, q - 1)

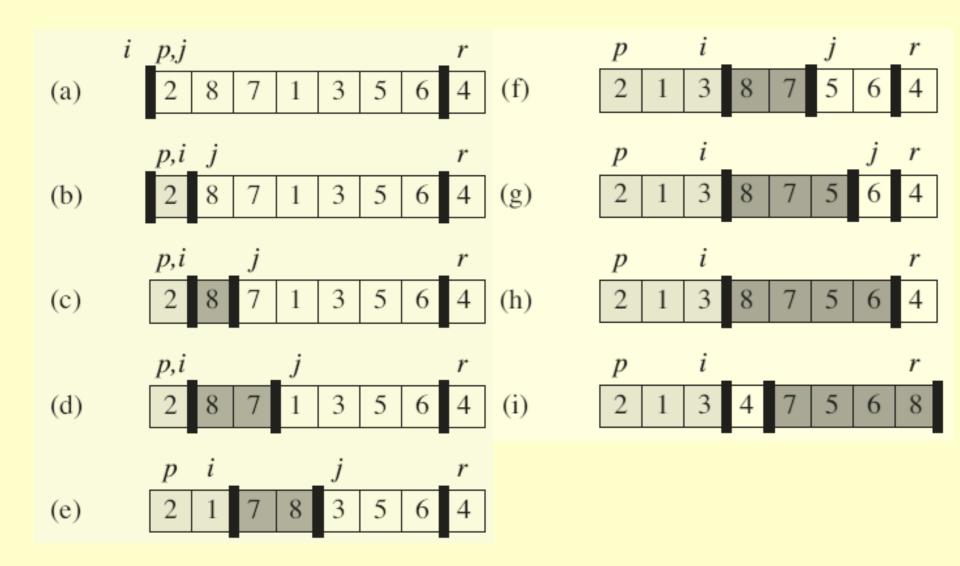
4 Quicksort(A, q + 1, r)
```

Initial call is Quicksort(A, 1, n)

Quick Sort

```
Partition(A, p, r)
1 \times A[r]
2 i \leftarrow p - 1
3 for j \leftarrow p to r - 1
         do if A[j] \leq x
                 then i \leftarrow i + 1
                        exchange A[i] \leftrightarrow A[j]
7 exchange A[i + 1] \leftrightarrow A[r]
8 \text{ return } i + 1
```

Quick Sort: Example



Partitioning

- Select the last element A[r] in the subarray A[p..r] as the pivot the element around which to partition.
- As the procedure executes, the array is partitioned into four (possibly empty) regions.
 - 1. A[p..i] All entries in this region are $\leq pivot$.
 - 2. A[i+1..j-1] All entries in this region are > pivot.
 - 3. A[r] = pivot.
 - 4. A[j..r-1] Not known how they compare to *pivot*.
- The above hold before each iteration of the *for* loop, and constitute a *loop invariant*. (4 is not part of the LI.)

- Use loop invariant.
- Initialization:
 - » Before first iteration
 - A[p..i] and A[i+1..j-1] are empty Conds. 1 and 2 are satisfied (trivially).
 - r is the index of the pivot Cond. 3 is satisfied.

Maintenance:

- » Case 1: A[j] > x
 - Increment *j* only.
 - LI is maintained.

```
Partition(A, p, r)

x, i := A[r], p-1;

for j := p to r - 1 do

if A[j] \le x then

i := i + 1;

A[i] \leftrightarrow A[j]

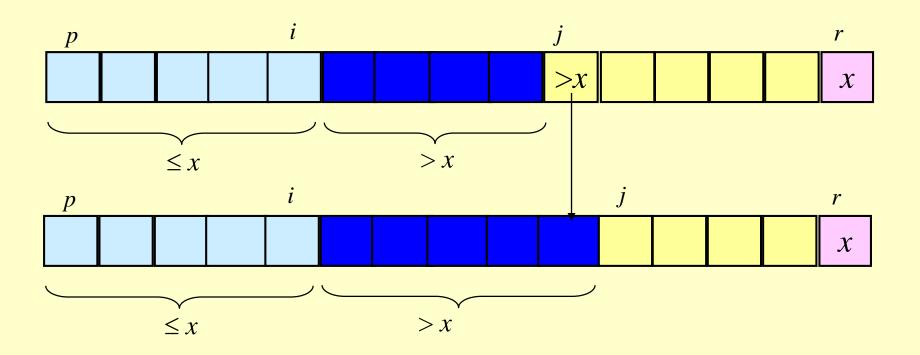
fi

od;

A[i + 1] \leftrightarrow A[r];

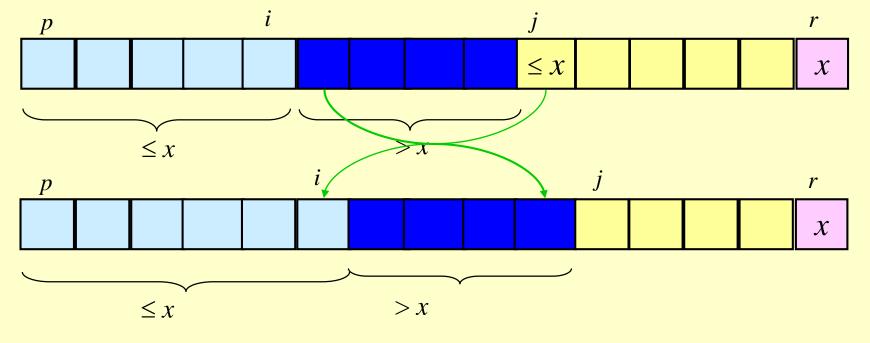
return i + 1
```

Case 1:



- Case 2: $A[j] \le x$
 - » Increment i
 - \gg Swap A[i] and A[j]
 - Condition 1 is maintained.
 - » Increment *j*
 - Condition 2 is maintained.

- \rightarrow A[r] is unaltered.
 - Condition 3 is maintained.



Termination:

- » When the loop terminates, j = r, so all elements in A are partitioned into one of the three cases:
 - $A[p..i] \leq pivot$
 - A[i+1..j-1] > pivot
 - A[r] = pivot
- The last two lines swap A[i+1] and A[r].
 - » *Pivot* moves from the end of the array to between the two subarrays.
 - » Thus, procedure *partition* correctly performs the divide step.

Time Complexity of Partition

- ◆ PartitionTime(*n*) is given by the number of iterations in the *for* loop.
- Time complexity of Partition algorithm is $\Theta(n)$.
- $\Theta(n)$: n = r p + 1.

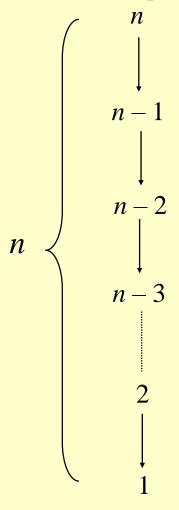
Algorithm Performance

Running time of quicksort depends on whether the partitioning is balanced or not.

- Worst-Case Partitioning (Unbalanced Partitions):
 - » Occurs when every call to partition results in the most unbalanced partition.
 - » Partition is most unbalanced when
 - Subproblem 1 is of size n-1, and subproblem 2 is of size 0 or vice versa.
 - $pivot \ge$ every element in A[p..r-1] or pivot < every element in A[p..r-1].
 - » Every call to partition is most unbalanced when
 - Array *A*[1..*n*] is sorted or reverse sorted!

Worst-case Partition Analysis

Recursion tree for worst-case partition



Running time for worst-case partitions at each recursive level:

$$T(n) = T(n-1) + T(0) + PartitionTime(n)$$

$$= T(n-1) + \Theta(n)$$

$$= \sum_{k=1 \text{ to } n} \Theta(k)$$

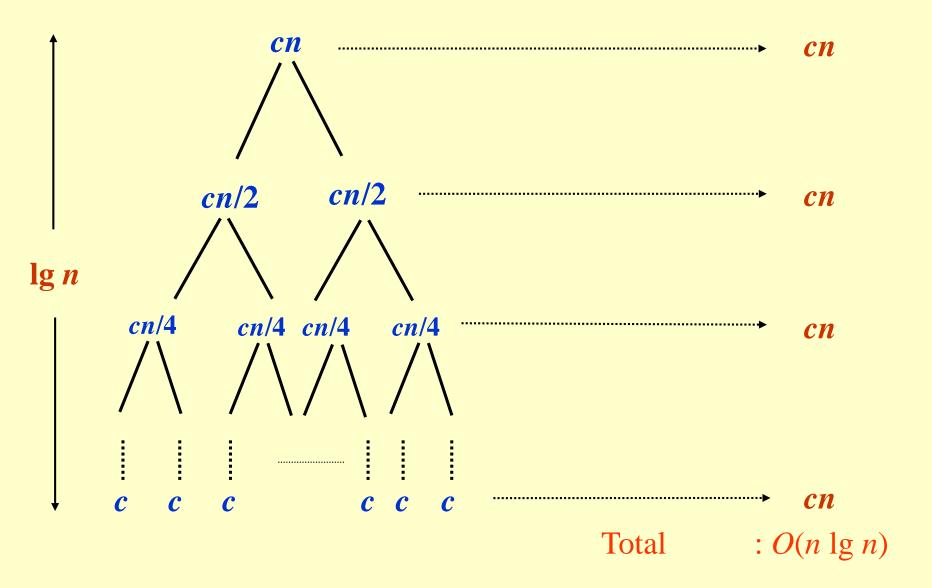
$$= \Theta(\sum_{k=1 \text{ to } n} k)$$

$$= \Theta(n^2)$$

Best-case Partitioning

- Size of each subproblem $\leq n/2$.
 - » One of the subproblems is of size $\lfloor n/2 \rfloor$
 - » The other is of size $\lceil n/2 \rceil 1$.
- Recurrence for running time
 - » T(n) ≤ 2T(n/2) + PartitionTime(n)= 2T(n/2) + Θ(n)
- By using the case 2 of Master's method, the solution to the above recurrence is $T(n) = \Theta(n \lg n)$

Recursion Tree for Best-case Partition



Fast Integer Multiplication

Integer Addition

Addition. Given two *n*-bit integers a and b, compute a+b. Grade-school. $\Theta(n)$ bit operations.

1	1	1	1	1	1	0	1	
	1	1	0	1	0	1	0	1
+	0	1	1	1	1	1	0	1

Remark. Traditional addition algorithm is optimal.

Integer Multiplication

Multiplication. Given two *n*-bit integers a and b, compute $a \times b$. Grade-school. $\Theta(n^2)$ bit operations.

```
1 0 1 0 1 0 1
              \times 0 1 1 1 1 1 0 1
                1 1 0 1 0 1 0 1
              0 0 0 0 0 0 0 0
            1 1 0 1 0 1 0 1
         1 1 0 1 0 1 0 1
       1 1 0 1 0 1 0 1
     1 1 0 1 0 1 0 1
   1 1 0 1 0 1 0 1
 0 0 0 0 0 0 0
0 1 1 0 1 0 0 0 0 0 0 0 0 0 1
```

Q. Is traditional multiplication algorithm optimal?

Divide-and-Conquer Multiplication: Warmup

To multiply two n-bit integers a and b:

- Multiply four $\frac{1}{2}n$ -bit integers, recursively.
- Add and shift to obtain result.

$$a = 2^{n/2} \cdot a_1 + a_0$$

$$b = 2^{n/2} \cdot b_1 + b_0$$

$$ab = \left(2^{n/2} \cdot a_1 + a_0\right) \left(2^{n/2} \cdot b_1 + b_0\right) = 2^n \cdot a_1 b_1 + 2^{n/2} \cdot \left(a_1 b_0 + a_0 b_1\right) + a_0 b_0$$

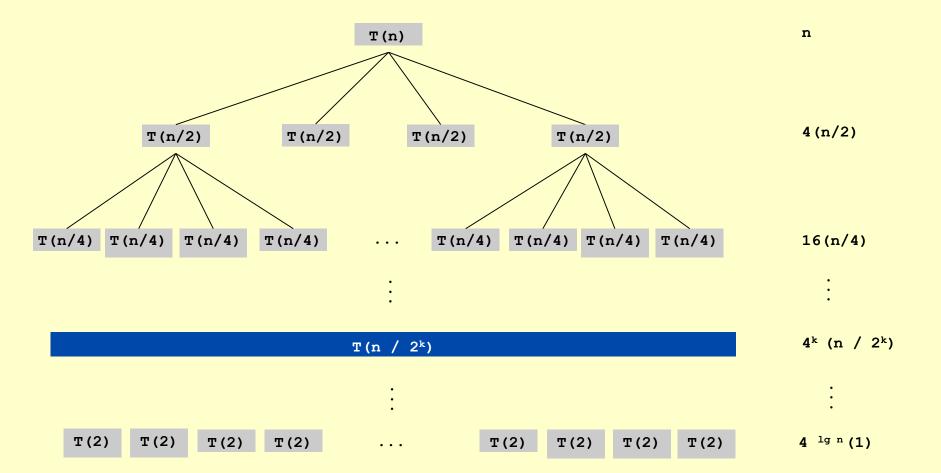
Ex.
$$a = 10001101$$
 $b = 11100001$

$$T(n) = \underbrace{4T(n/2)}_{\text{recursive calls}} + \underbrace{\Theta(n)}_{\text{add, shift}} \Rightarrow T(n) = \Theta(n^2)$$

Recursion Tree

$$T(n) = \begin{cases} 0 & \text{if } n = 0\\ 4T(n/2) + n & \text{otherwise} \end{cases}$$

$$T(n) = \sum_{k=0}^{\lg n} n \, 2^k = n \left(\frac{2^{1+\lg n} - 1}{2 - 1} \right) = 2n^2 - n$$



Karatsuba Multiplication

To multiply two n-bit integers a and b:

- Add two $\frac{1}{2}n$ bit integers.
- Multiply three $\frac{1}{2}n$ -bit integers, recursively.
- Add, subtract, and shift to obtain result.

$$a = 2^{n/2} \cdot a_1 + a_0$$

$$b = 2^{n/2} \cdot b_1 + b_0$$

$$ab = 2^n \cdot a_1 b_1 + 2^{n/2} \cdot (a_1 b_0 + a_0 b_1) + a_0 b_0$$

$$= 2^n \cdot a_1 b_1 + 2^{n/2} \cdot ((a_1 + a_0)(b_1 + b_0) - a_1 b_1 - a_0 b_0) + a_0 b_0$$
1
2
1
3
3

$$T(n) \leq \underbrace{T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + T(1 + \lceil n/2 \rceil)}_{\text{recursive calls}} + \underbrace{O(n)}_{\text{add, subtract, shift}} \Rightarrow T(n) = O(n^{\lg 3}) = O(n^{1.585})$$

Karatsuba Multiplication

```
Recursive-Multiply(x,y):

Write x = x_1 \cdot 2^{n/2} + x_0
y = y_1 \cdot 2^{n/2} + y_0
Compute x_1 + x_0 and y_1 + y_0
p = \text{Recursive-Multiply}(x_1 + x_0, y_1 + y_0)
x_1y_1 = \text{Recursive-Multiply}(x_1, y_1)
x_0y_0 = \text{Recursive-Multiply}(x_0, y_0)
Return x_1y_1 \cdot 2^n + (p - x_1y_1 - x_0y_0) \cdot 2^{n/2} + x_0y_0
```

$$T(n) \leq \underbrace{T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + T(1 + \lceil n/2 \rceil)}_{\text{recursive calls}} + \underbrace{O(n)}_{\text{add, subtract, shift}} \Rightarrow T(n) = O(n^{\lg 3}) = O(n^{1.585})$$

Karatsuba: Recursion Tree

$$T(n) = \begin{cases} 0 & \text{if } n = 0\\ 3T(n/2) + n & \text{otherwise} \end{cases}$$

$$T(n) = \sum_{k=0}^{\lg n} n \left(\frac{3}{2}\right)^k = n \left(\frac{\left(\frac{3}{2}\right)^{1+\lg n} - 1}{\frac{3}{2} - 1}\right) = 3n^{\lg 3} - 2n$$

