

The divide-and-conquer design paradigm

- 1. Divide the problem (instance) into subproblems.
- 2. Conquer the subproblems by solving them recursively.
- 3. Combine subproblem solutions.



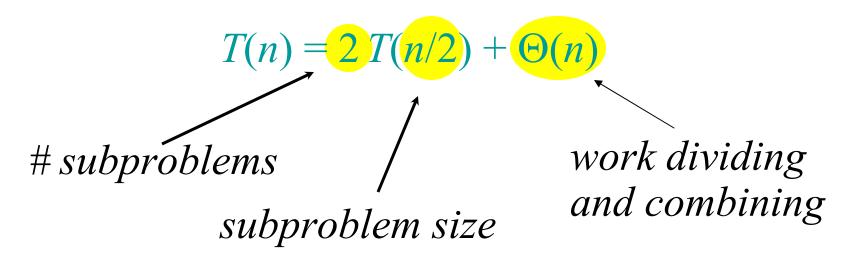
Merge sort

- 1. Divide: Trivial.
- 2. Conquer: Recursively sort 2 subarrays.
- 3. Combine: Linear-time merge.



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Master theorem (reprise)

$$T(n) = aT(n/b) + f(n)$$

CASE 1:
$$f(n) = O(n^{\log_b a} - \varepsilon)$$
, constant $\varepsilon > 0$
 $\Rightarrow T(n) = \Theta(n^{\log_b a})$.
CASE 2: $f(n) = \Theta(n^{\log_b a})$
 $\Rightarrow T(n) = \Theta(n^{\log_b a} \log_b a)$.

CASE 3: $f(n) = \Omega(n^{\log_b a + \varepsilon})$, constant $\varepsilon > 0$, and regularity condition

$$\Rightarrow T(n) = \Theta(f(n))$$
.

Merge sort: a = 2, $b = 2 \Rightarrow n^{\log_b a} = n^{\log_2 2} = n$ \Rightarrow CASE 2 $T(n) = \Theta(n \lg n)$.



Find an element in a sorted array:

- 1. Divide: Check middle element.
- 2. Conquer: Recursively search 1 subarray.
- 3. Combine: Trivial.



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Example: Find 9

3 5 7 8 9 12 15



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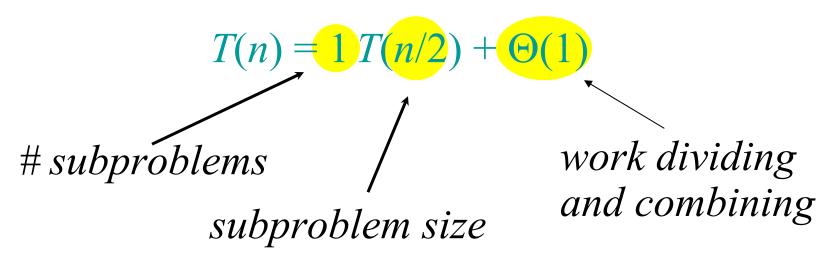
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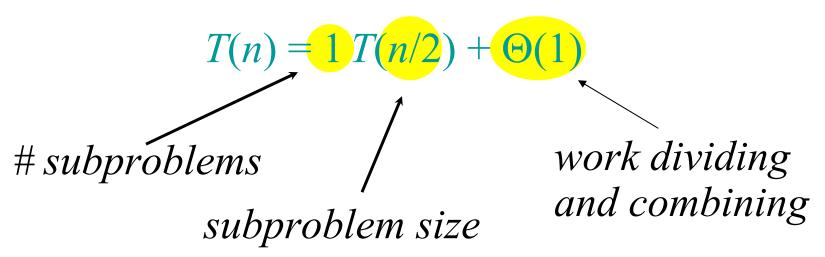


Recurrence for binary search





Recurrence for binary search



$$n^{\log_b a} = n^{\log_2 1} = n^0 = 1 \implies \text{CASE 2}(k = 0)$$

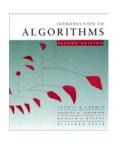
 $\Rightarrow T(n) = \Theta(\lg n)$.



Powering a number

Problem: Compute a^n , where $n \in \mathbb{N}$.

Naive algorithm: $\Theta(n)$.



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Divide-and-conquer algorithm:

$$a^{n} = \begin{cases} a^{n/2} \cdot a^{n/2} & \text{if } n \text{ is even;} \\ a^{(n-1)/2} \cdot a^{(n-1)/2} \cdot a & \text{if } n \text{ is odd.} \end{cases}$$



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$$T(n) = T(n/2) + \Theta(1) \implies T(n) = \Theta(\lg n)$$
.



Matrix multiplication

Input:
$$A = [a_{ij}], B = [b_{ij}].$$

Output: $C = [c_{ij}] = A \cdot B.$ $i, j = 1, 2, ..., n.$

$$\begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \cdot \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{nn} \end{bmatrix}$$

$$c_{ij} = \sum_{k=1}^{n} a_{ik} \cdot b_{kj}$$



Standard algorithm

for
$$i \leftarrow 1$$
 to n

do for $j \leftarrow 1$ to n

do $c_{ij} \leftarrow 0$

for $k \leftarrow 1$ to n

do $c_{ij} \leftarrow c_{ij} + a_{ik} \cdot b_{kj}$



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Running time = $\Theta(n^3)$



Divide-and-conquer algorithm

DEA:

 $n \times n$ matrix = 2×2 matrix of $(n/2) \times (n/2)$ submatrices:

$$\begin{bmatrix} r \mid s \\ -+- \\ t \mid u \end{bmatrix} = \begin{bmatrix} a \mid b \\ -+- \\ c \mid d \end{bmatrix} \cdot \begin{bmatrix} e \mid f \\ ---- \\ g \mid h \end{bmatrix}$$

$$C = A \cdot B$$

$$r = ae + bg$$

 $s = af + bh$
 $t = ce + dg$
 $u = cf + dh$
8 mults of $(n/2) \times (n/2)$ submatrices

8 mults of $(n/2)\times(n/2)$ submatrices

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Divide-and-conquer algorithm

IDEA:

 $n \times n$ matrix = 2×2 matrix of $(n/2) \times (n/2)$ submatrices:

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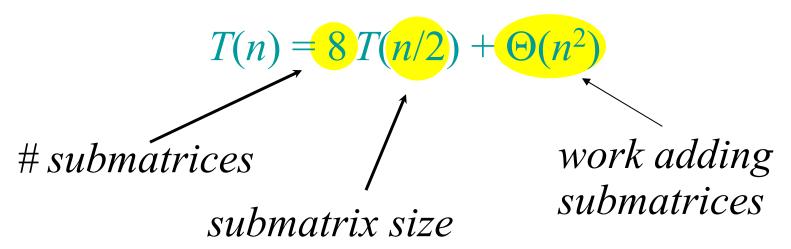
$$r = ae + bg$$

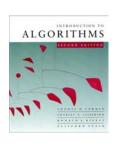
 $s = af + bh$
 $t = ce + dh$
 $u = cf + dg$
Solution $recursive$
8 mults of $(n/2) \times (n/2)$ submatrices
4 adds of $(n/2) \times (n/2)$ submatrices

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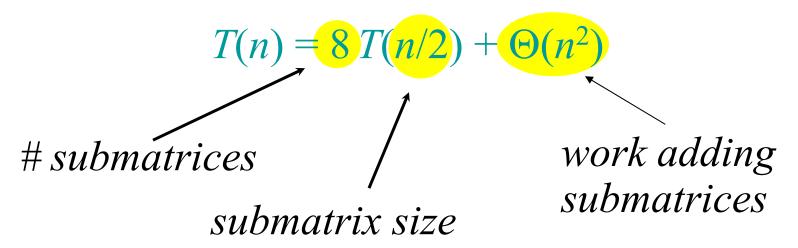


Analysis of D&C algorithm

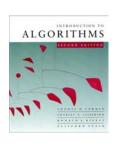




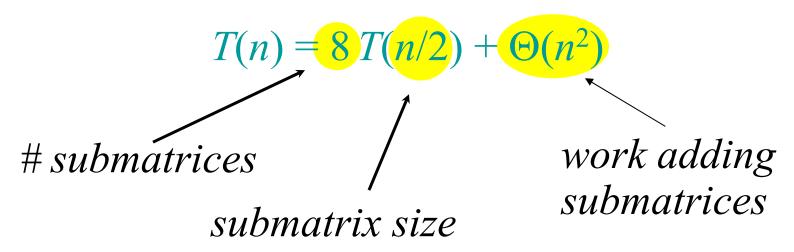
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Analysis of D&C algorithm



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No better than the ordinary algorithm.



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$$P_{1} = a \cdot (f - h)$$

 $P_{2} = (a + b) \cdot h$
 $P_{3} = (c + d) \cdot e$
 $P_{4} = d \cdot (g - e)$
 $P_{5} = (a + d) \cdot (e + h)$
 $P_{6} = (b - d) \cdot (g + h)$
 $P_{7} = (a - c) \cdot (e + f)$



• Multiply 2×2 matrices with only 7 recursive mults.

$$P_{1} = a \cdot (f - h)$$
 $r = P_{5} + P_{4} - P_{2} + P_{6}$
 $P_{2} = (a + b) \cdot h$ $s = P_{1} + P_{2}$
 $P_{3} = (c + d) \cdot e$ $t = P_{3} + P_{4}$
 $P_{4} = d \cdot (g - e)$ $u = P_{5} + P_{1} - P_{3} - P_{7}$
 $P_{5} = (a + d) \cdot (e + h)$
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$$r = P_5 + P_4 - P_2 + P_6$$

$$s = P_1 + P_2$$

$$t = P_3 + P_4$$

$$u = P_5 + P_1 - P_3 - P_7$$

7 mults, 18 adds/subs.
Note: No reliance on commutativity of mult!



• Multiply 2×2 matrices with only 7 recursive mults.

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$$r = P_{5} + P_{4} - P_{2} + P_{6}$$

$$= (a + d)(e + h)$$

$$+ d(g - e) - (a + b)h$$

$$+ (b - d)(g + h)$$

$$= ae + ah + de + dh$$

$$+ dg - de - ah - bh$$

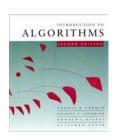
$$+ bg + bh - dg - dh$$

$$= ae + bg$$



Strassen's algorithm

- 1. Divide: Partition A and B into $(n/2)\times(n/2)$ submatrices. Form terms to be multiplied using + and -.
- 2. Conquer: Perform 7 multiplications of $(n/2)\times(n/2)$ submatrices recursively.
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$$T(n) = 7 T(n/2) + \Theta(n^2)$$

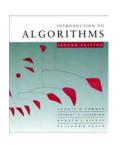


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The number 2.81 may not seem much smaller than 3, but because the difference is in the exponent, the impact on running time is significant. In fact, Strassen's algorithm beats the ordinary algorithm on today's machines for $n \ge 32$ or so.



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Best to date (of theoretical interest only): $\Theta(n^{2.376\cdots})$.



Conclusion

- Divide and conquer is just one of several powerful techniques for algorithm design.
- Divide-and-conquer algorithms can be analyzed using recurrences and the master method (so practice this math).
- The divide-and-conquer strategy often leads to efficient algorithms.