CH08-320201 Algorithms and Data Structures

Lecture 5/6 — 20 Feb 2018

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This & that

- Medical excuse policy refined:
 - http://vis.jacobs-university.de/teaching/ads/18s/
- Anybody signed up anew?
 - email me to add you to Moodle
- Auditing
- Midterm
 - March 20, 9:45am (no lecture on that day then)

Today

- Apply recurrences to Divide & Conquer
 - Power of a number
 - Fibonacci numbers
 - Maximum-subarray problem
 - Matrix multiplication

1.6 Apply recurrences to Divide & Conquer

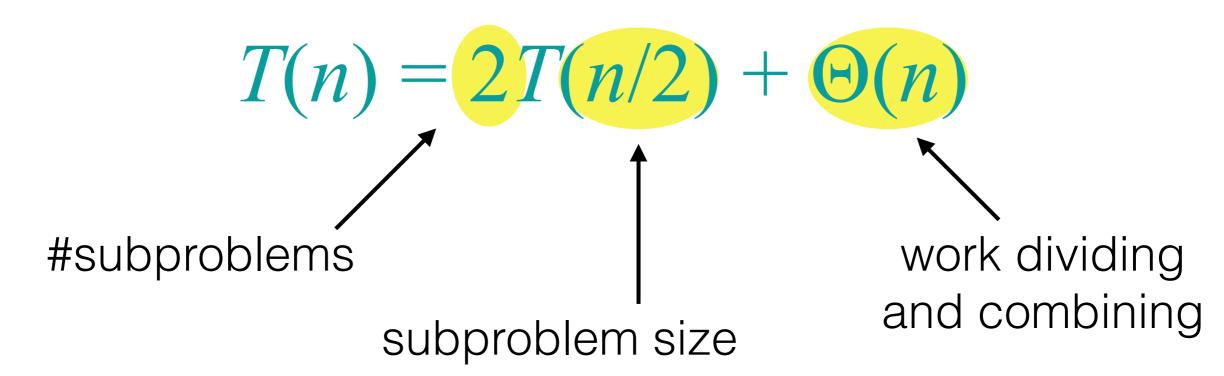
Recall: Divide & Conquer

Design paradigm:

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

Recall: Merge Sort

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



Master method on Merge Sort

$$T(n) = 2T(n/2) + n$$

$$a = 2, b = 2$$

$$n^{\log_b a} = n$$

$$f(n) = n$$

Case 2:
$$f(n) = \Theta(n)$$
,
Thus, $T(n) = \Theta(n \lg n)$.

Power of a number

- Problem:
 - Input: numbers $a \in \mathbf{R}$ and $n \in \mathbf{N}$.
 - Output: aⁿ
- Naive algorithm:
 - $T(n) = \Theta(n)$.
- Divide & Conquer:

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

Recurrence:

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

Solution:

-
$$a = 1$$
, $b = 2$, $n^{\log_b a} = 1$, $f(n) = \Theta(1)$ —> **Case 2**.

- Thus, $T(n) = \Theta(\lg n)$

Recursive definition:

$$F_{n} = \begin{cases} 0 & \text{if } n = 0; \\ 1 & \text{if } n = 1; \\ F_{n-1} + F_{n-2} & \text{if } n \ge 2. \end{cases}$$

Output: return the *n*-th Fibonacci number

Naive algorithm:

Implement the recursion as in the definition.

FIBONACCI

- 1 **if** (n < 2)
- 2 return n;
- 3 else return Fibonacci (n 1) + Fibonacci (n 2)

$$T(n) = T(n-1) + T(n-2) + \Theta(1).$$

$$T(n) = T(n-1) + T(n-2) + \Theta(1).$$

Lower bound:

$$T(n-1) \approx T(n-2)$$

 $T(n) = 2T(n-2) + \Theta(1)$
 $T(n) = 4T(n-4) + \Theta(1)$
 $T(n) = 8T(n-6) + \Theta(1)$

. . .

$$T(n) = 2^k T(n - 2k) + \Theta(1)$$
 —> $n - 2k = 0 \Rightarrow k = n/2$
 $T(n) = 2^{n/2} T(0) + \Theta(1)$ —> $T(n) = \Omega(2^{n/2})$,
i.e., exponential time

Side note: A tighter lower bound

$$\mathsf{T}(\mathsf{n}) = \Omega(\mathcal{\Phi}^{\mathsf{n}}),$$

since #leaves in rec. tree = FIBONACCI (n) x $\Theta(1)$

Φ is the golden ratio

$$\phi = (1 + \sqrt{5})/2$$

Bottom up approach:

Avoid recursion, i.e.,

compute F_0 , F_1 , F_2 , ..., F_n in the given order instead, forming each number by summing the two previous.

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

Use closed form (rounded to next integer):

$$F_n = \phi^n / \sqrt{5}$$

$$\phi = (1 + \sqrt{5})/2$$

(proof by induction).

Compute by "Power of a number" recursion.

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

But: numerical problems may occur (floating-point arithmetic).

Use matrix representation:

$$\begin{bmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n$$

(proof by induction).

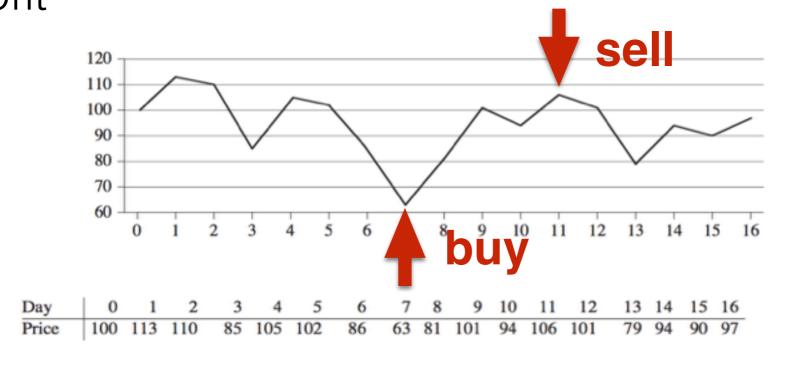
Compute by "Power of a number" recursion (using a generalization to 2x2 matrices).

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

And: uses integers only (no floating-point errors).

- Motivation
 Scenario buy & sell stock
 - Input: a sequence of numbers

Output: subsequence that results in the highest profit



Brute-Force algorithm

Try every pair of days.

Number of pairs:

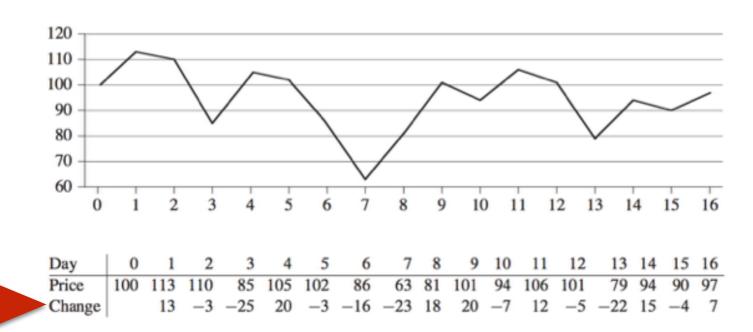
$$\binom{n}{k} = \frac{n!}{2!(n-2)!} = \frac{n(n-1)}{2} = \Theta(n^2)$$

Transformation:

- consider daily change in price instead

—> Maximum-subarray problem

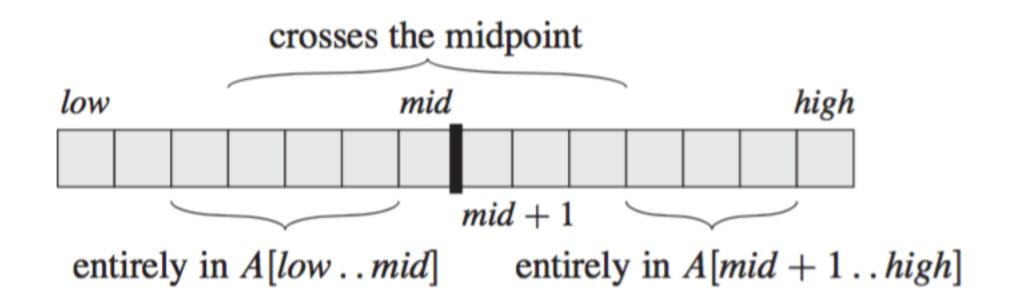
- Input: a sequence of numbers <a1, a2,, an> of positive and negative numbers (otherwise it does not make sense)
- Output: contiguous, non-empty subsequence $< a_i, a_{i+1}, \ldots, a_j > 0$ with $i \ge 1$, $j \le n$, so that $\sum_{k=i}^j a_k$ is maximized



Divide & Conquer:

Idea: Divide A into 2 pieces —> maximum subarray is either

- 1. entirely in the subarray A[low ... mid],
- 2. entirely in the subarray A[mid+1 ... high], or
- 3. crossing the midpoint



FIND-MAX-CROSSING-SUBARRAY (A, low, mid, high)

```
left-sum = -\infty
   sum = 0
   for i = mid downto low
        sum = sum + A[i]
        if sum > left-sum
                                                                   A[mid+1..j]
            left-sum = sum
                                                                                        high
                                           low
                                                               mid
            max-left = i
    right-sum = -\infty
    sum = 0
                                                                    mid + 1
                                                                               j
    for j = mid + 1 to high
10
                                                         A[i ..mid]
        sum = sum + A[j]
11
        if sum > right-sum
12
13
            right-sum = sum
            max-right = j
14
    return (max-left, max-right, left-sum + right-sum)
```

Runtime: $\Theta(n)$.

FIND-MAXIMUM-SUBARRAY (A, low, high)

```
\Theta(1) \begin{bmatrix} 1 & \text{if } high == low \\ 2 & \text{return } (low, high, A[low]) & \text{//} base case: only one element} \\ 2 & \text{else } mid = \lfloor (low + high)/2 \rfloor \\ 4 & (left-low, left-high, left-sum) = \\ & FIND-MAXIMUM-SUBARRAY(A, low, mid) \\ 5 & (right-low, right-high, right-sum) = \\ & FIND-MAXIMUM-SUBARRAY(A, mid + 1, high) \\ 6 & (cross-low, cross-high, cross-sum) = \\ & FIND-MAX-CROSSING-SUBARRAY(A, low, mid, high) \\ 7 & \text{if } left-sum \geq right-sum \text{ and } left-sum \geq cross-sum \\ 8 & \text{return } (left-low, left-high, left-sum) \\ 9 & \text{elseif } right-sum \geq left-sum \text{ and } right-sum \geq cross-sum \\ 10 & \text{return } (right-low, right-high, right-sum) \\ 11 & \text{else return } (cross-low, cross-high, cross-sum) \end{bmatrix}
```

• Recurrence:

$$- T(n) = 2T(n/2) + \Theta(n)$$

• Solution:

-
$$a = 2$$
, $b = 2$, $n^{\log_b a} = n$, $f(n) = \Theta(n)$ —> **Case 2**.

- Thus, $T(n) = \Theta(n \lg n)$

Problem

```
- Input: A = [a_{ij}], B = [b_{ij}]. 
- Output: C = [c_{ij}] = A \cdot B. i, j = 1, 2, \dots, 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:

$$\begin{aligned} & \textbf{for } i = 1 \textbf{ to } n \\ & do \textbf{ for } j = 1 \textbf{ to } n \\ & do \textbf{ } c_{ij} = 0 \\ & \textbf{ for } k = 1 \textbf{ to } n \\ & \textbf{ do } c_{ij} = c_{ij} + a_{ik} \times b_{kj} \end{aligned}$$

$$T(n) = \Theta(n^3)$$

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

Divide & Conquer:

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

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

$$C = A \cdot B$$

Combining subproblem solutions:

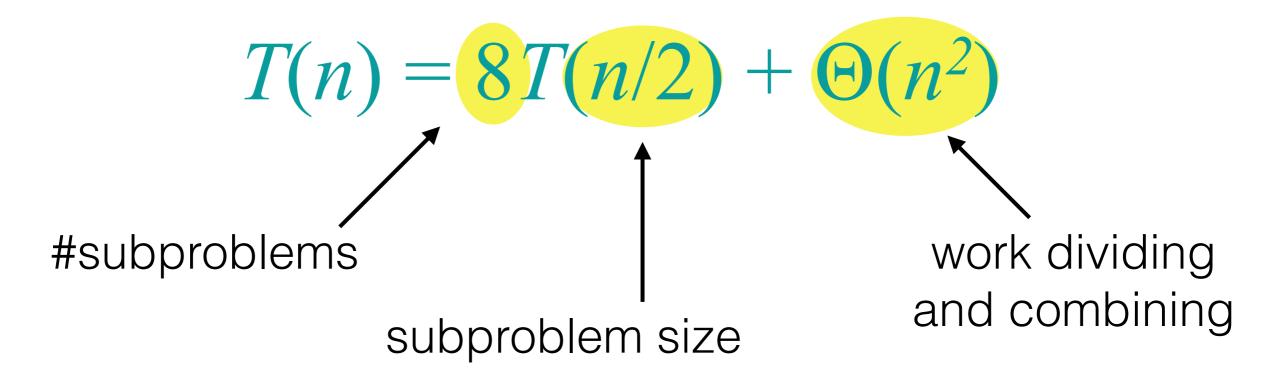
$$r = ae + bg$$

 $s = af + bh$
 $t = ce + dg$
 $u = cf + dh$

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

* with index calculations, otherwise $\Theta(n^2)$ for copying entries

Recurrence:



$$n^{\log_b a} = n^{\log_2 8} = n^3$$

Case 1:
$$T(n) = \Theta(n^3)$$
.

No better than the ordinary algorithm.

- Lessons Learned:
 - + additions goes away (constant factor)
 - +multiplications not —> recursive case (they make the tree "bushy")
- What to do?
 - Try to reduce #multiplications
 - OK to have more additions

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

$$C = A \cdot B$$

Strassen's idea:

Multiply matrices with 7 multiplications and 18 additions.

$$P_1 = a \cdot (f - h)$$
 $r = P_5 + P_4 - P_2 + P_6$
 $P_2 = (a + b) \cdot h$ $s = P_1 + P_2$
 $P_3 = (c + d) \cdot e$ $t = P_3 + P_4$
 $P_4 = d \cdot (g - e)$ $u = P_5 + P_1 - P_3 - P_7$
 $P_5 = (a + d) \cdot (e + h)$
 $P_6 = (b - d) \cdot (g + h)$
 $P_7 = (a - c) \cdot (e + f)$

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

Strassen's idea:

Multiply matrices with 7 multiplications and 18 additions.

$$P_{1} = a \cdot (f - h) \qquad r = P_{5} + P_{4} - P_{2} + P_{6}$$

$$P_{2} = (a + b) \cdot h \qquad = (a + d)(e + h)$$

$$P_{3} = (c + d) \cdot e \qquad + d(g - e) - (a + b)h$$

$$P_{4} = d \cdot (g - e) \qquad + (b - d)(g + h)$$

$$P_{5} = (a + d) \cdot (e + h) \qquad = ae + ah + de + dh$$

$$P_{6} = (b - d) \cdot (g + h) \qquad + dg - de - ah - bh$$

$$P_{7} = (a - c) \cdot (e + f) \qquad + bg + bh - dg - dh$$

$$= ae + bg$$

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.

3. Combine:

Form C using + and – on $(n/2)\times(n/2)$ submatrices.

Complexity of Strassen's algorithm:

Recurrence:

$$T(n) = 7 T(n/2) + \Theta(n^2)$$

$$n^{\log_b a} = n^{\log_2 7} = n^{2.81}$$

Case 1:
$$T(n) = \Theta(n^{\lg 7})$$
.

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

Best known algorithm:

Latest improvement in 2014 [1]:

 $O(n^{2.3728639})$

Only of theoretical interest.

Most approaches that are faster than Strassen are not used in practice.

They are only faster for very large n.

One cannot get better than O(n²), cf. **Case 3**.

[1] Francois LeGall, Powers of Tensors and Fast Matrix Multiplication, 30 Jan 2014.

1.7 Conclusion

Conclusion

- Definitions
- First example of an algorithm (InsertionSort)
- Asymptotic analysis
- First powerful concept (Divide&Conquer)
- Solve recurrences for analysis