

Algorithms and Data Structures

Laboratory work #2

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Classes plan

- 1. Previous homework problem #1005 "Stone Pile"
- 2. Problem #1401 "Gamers"
- 3. Task for homework
- 4. Explanation of test 1



- Link to the problem's description
 https://acm.timus.ru/problem.aspx?space=1&num=1005&localege=en
- You have a number of stones with known weights w_1 , ..., w_n . Write a program that will rearrange the stones into two piles such that weight difference between the piles is minimal.
- Input contains the number of stones n (1 ≤ n ≤ 20) and weights of the stones w_1 , ..., w_n (integers, 1 ≤ w_i ≤ 100000) delimited by white spaces.
- Your program should output a number representing the minimal possible weight difference between stone piles.



0-1 Knapsack problem

- number x_i of copies of each kind of item restricted to zero or one
- set of n items numbered from 1 up to n, each with a weight w_i and a value v_i
- maximum weight capacity W

Goal:

- maximize $\sum_{i=1}^{n} v_i x_i$
- subject to $\sum_{i=1}^{n} w_i x_i \leq W$
- where $x_i \in \{0, 1\}$



- Interpretation of problem as 0-1 Knapsack problem
 - $v_i = w_i$
 - maximize $\sum_{i=1}^{n} w_i x_i$
 - $\sum_{i=1}^{n} w_i x_i \leq \frac{\sum_{i=1}^{n} w_i}{2}$ smaller pile
- Result: $\sum_{i=1}^{n} w_i 2 \sum_{i=1}^{n} w_i x_i$

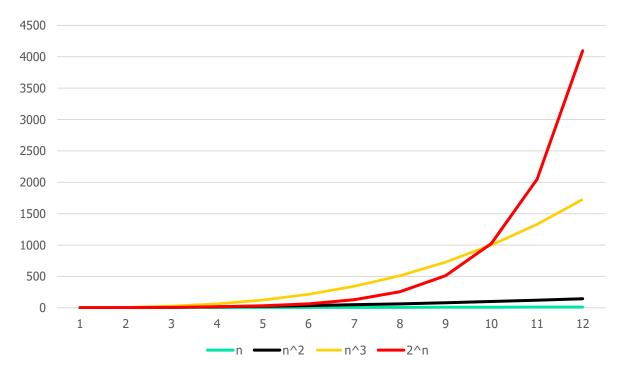


NP-complete problem!

- Solutions of 0-1 Knapsack problem
 - Brute-force search
 - "Greedy" approximation algorithm
 - Dynamic programming



- Brute-force algorithm
 - Let's check all possible combinations of stones in piles!
 - Total 2^{n-1} different combinations!





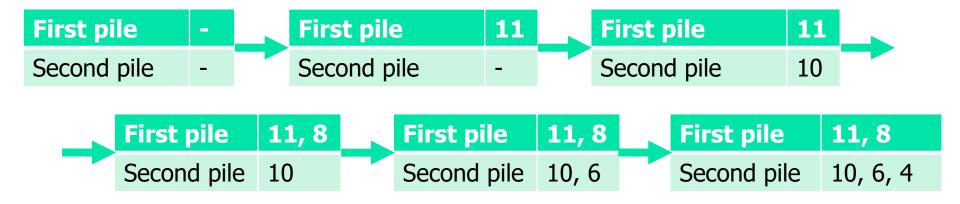
- Brute-force algorithm
 - Complexity is $O(2^n)$
 - With a sufficiently large number of objects, the problem becomes unsolvable by this method in an acceptable time
 - Good news: we have only 20 stones and 2¹⁹ combinations



- "Greedy" approximation algorithm
 - Sort all elements by value
 - Take elements with maximum values
 - In our case take stones of maximum weights to the first pile until adding of any new stone will exceed half the sum of all weights
 - Idea from knapsack greedy algorithm



- Example 1
 - Let stones have weights {8, 11, 10, 4, 6}
 - Sorted weights {11, 10, 8, 6, 4}
 - Sum of weights is 39, half sum is 19



• Result is
$$|(11 + 8) - (10 + 6 + 4)| = 1$$



- Example 2
 - Let stones have weights {8, 13, 10, 6, 4, 10}
 - Sorted weights {13, 10, 10, 8, 6, 4}
 - Sum of weights is 51, half sum is 25

First	pile	-		First pile	13		First pile	13, 10	
Seco	nd pile	-		Second pile	-		Second pile	-	
First pile		ile	13, 10 10				irst pile	13, 10	
-	Second pile					9	Second pile	10, 8, 6, 4	

• Result is
$$|(13 + 10) - (10 + 8 + 6 + 4)| = 5$$



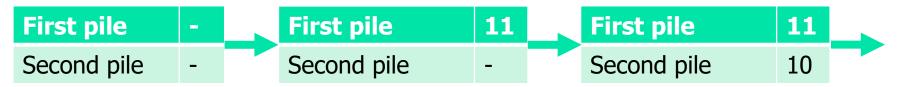
- Example 2
 - Result = 5
 - Result is not optimal!

First pile	13, 8, 4		
Second pile	10, 10, 6		

• Result is |(13 + 8 + 4) - (10 + 10 + 6)| = 1



- Other "greedy" approach
 - Sort all elements by weight
 - Add next element to pile with less total weight
 - In example 1 {11, 10, 8, 6, 4} result is 3





Result is not optimal!



- "Greedy" approximation algorithm
 - Algorithm calculates "approximate" result
 - Not applicable if accurate result is required
 - Very fast
 - Sorting: $O(n \log n)$
 - Iterate over sorted weights: O(n)
 - Total: $O(n \log n)$



Dynamic programming

$$W = \frac{\sum_{i=1}^{n} w_i}{2}$$

- Table t with size (N + 1) × (W + 1)
 - Starting from 0
- Numerate all stones from 1 till N
- t[i][j] maximum possible sum of weights for stones with numbers { 1 .. i }, restricted by j
- The result is t[N][W]
- How to fill the table?



Remember example 1

num	1	2	3	4	5
stone weight	8	11	10	4	6

- Table will have size 6 × 20
- t[4][5] = 4; t[3][15] = 11; t[4][16] = 11 + 4 = 15
- Let's iterate over table rows by i = { 0 .. N } and columns by w = { 0 .. W }
- Fill next element
 - t[0][w] = 0
 - t[i][w] = ?



- Case 1. w_i > w
 - No sense to use this stone
 - t[i][w] = t[i 1][w]
- Case 2. w_i <= w</p>
 - What means to add stone?
 - Check maximum sum for all previous stones, if restrictions was (w – w_i) – so we have enough place to add current stone
 - $t[i][w] = t[i-1][w-w_i] + w_i$



- Case 2. w_i <= w (continue)</p>
 - $t[i][w] = t[i 1][w w_i] + w_i$
 - It is possible, that to add current stone we remove stone with greater weight
 - Need compare this value with maximum without current stone
 - $t[i][w] = max(t[i-1][w-w_i] + w_i, t[i-1][w])$
- By this 3 rules we can fill the table!



- Dynamic programming
 - Time complexity is O(n * W)
 - It is called pseudo-polynomial time
 - Memory complexity is O(n * W)
 - Is O(n * W) always better than $O(2^n)$?
 - In worst case for our problem:
 - Brute-force: $2^{19} \approx 500\ 000$
 - Dynamic: $20 * \frac{100000}{2} * 20 \approx 2000000$



- Link to the problem's description <u>https://acm.timus.ru/problem.aspx?space=1&num=1401&local</u> e=en
- Mr. Chichikov argued with some blunderers that he would be able to prove that it is impossible to pave the 512 × 512 square checker-board with the figures:

 Once one of those blunderers happened to be not so silly and he claimed that he was able to pave the 512 × 512 square checker-board without the upper right cell with those figures. Chichikov blurted out that he could pave any 2ⁿ × 2ⁿ square checker-board without one arbitrary cell with those figures.



Pave 2 × 2 square



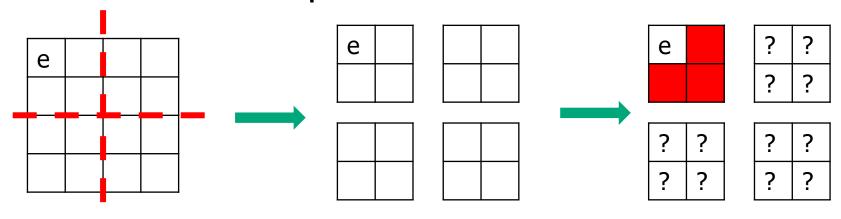






- e empty cell
- Let's name the function to fill 2 × 2 square pave2On2()
- Arguments coordinates of top left cell and empty cell

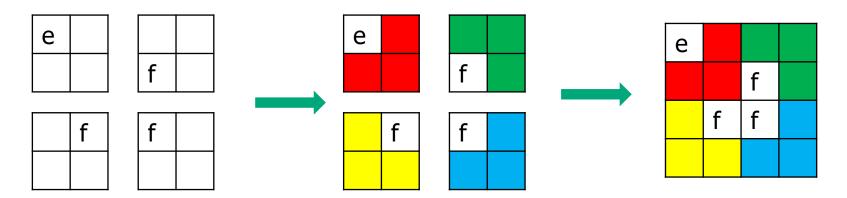
Pave 4 ×4 square

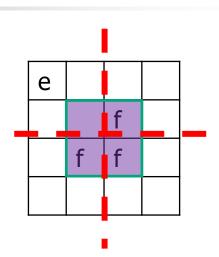






- "Middle" square
- Fake empty cells
- Provide fake empty cells to pave2On2()





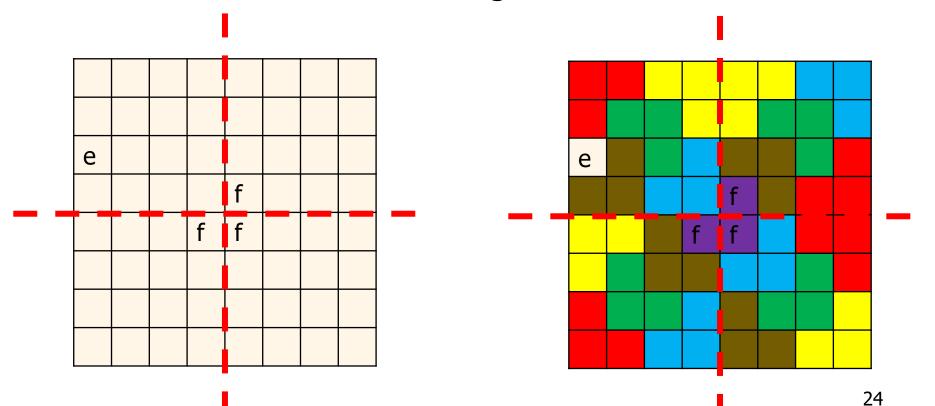


- Pave 4 × 4 square
 - Pave "middle" square
 - pave4On4()
 - Arguments are the same as for pave2On2()





- Pave 8 × 8 square
 - Let's use the same algorithm!





- Pave 2ⁿ × 2ⁿ square
 - Introduce paveSquare() recursive function
 - Let current size of square is n × n
 - If n = 2 easy to pave it based on empty cell
 - Otherwise split square on 4 parts, each has size $\frac{n}{2} \times \frac{n}{2}$
 - If this part have empty cell call paveSquare() for this part with real empty cell
 - Otherwise choose fake empty cell in middle square and use to call paveSquare() for this part
 - Fill middle square



- Pave 2ⁿ × 2ⁿ square
 - Complexity?
 - There are other ways to pave 2ⁿ × 2ⁿ square this algorithm is only example and prove that it is always possible
 - Feel free to use this algorithm of paving or choose another one



Mandatory task

- Prepare source code to solve problem #1401 "Gamers"
 https://acm.timus.ru/problem.aspx?space=1&num=1401&loc ale=en
- Pass tests on Timus system for this problem https://acm.timus.ru/submit.aspx?space=1&num=1401
- 3. Prepare a report with algorithm complexity and explanation Use template.docx to prepare report and send it to hduitmo.ads@yandex.ru with correct subject



Task for homework

You can solve following problem to get extra 2 points:

1. Problem #2025 "Line Fighting"

https://acm.timus.ru/problem.aspx?space=1&num=2025&locale=en

N.B. Report for this problem should contain explanation, why your approach to split fighters on teams is optimal



Task 1

- Complexity of polynomic function
- Simpler understanding is that for

$$p(n) = a_k n^k + a_{k-1} n^{k-1} + \dots + a_0$$
, if $a_k > 0$:

- $p(n) = O(n^m)$, for $m \ge k$
- $p(n) = \Omega(n^m)$, for $m \le k$
- $p(n) = \Theta(n^m)$, for m = k

Examples:

- $f(N) = 15N^4 + 16N \Rightarrow f(N) = O(N^5)$
- $f(N) = 75N + 8 \Rightarrow f(N) = \Theta(N)$
- $f(N) = 12N^3 + 6N^2 + 3N + 1 \Rightarrow f(N) = \Omega(N)$



Task 2

Complexity of function in source code

```
for (int i = 0; i < n; i++) {
    for (int j = 0; j < 1; j++) {
        // Complexity of function() is Θ(1)
        function(n, 1);
    }
}</pre>
```

Correct answer:

- function() will be called I times for each start of internal loop
- Internal loop will be started n times
- Total n*m times
- Complexity of function() is constant
- $\Theta(n*l)$



Task 2

Complexity of function in source code

```
for (int i = 0; i < n / 2 + 2; i++) {
    // Complexity of function() is ⊕(log n)
    function();
}</pre>
```

Correct answer:

- function() will be called (n / 2 + 2) times for each start of internal loop
- Dependence is linear
- Complexity of function() is Θ(log n)
- $\Theta(n \log n)$



Task 2

Complexity of function in source code

```
for (int i = 0; i < n; i++) {
    i *= m;
}</pre>
```

Correct answer:

- How many iterations?
- i will change in following way $0 \Rightarrow 1 \Rightarrow m \Rightarrow m^2 \Rightarrow ... \Rightarrow m^k \Rightarrow n$
- k iterations, where k is solution of $m^k >= n => k \ge \log_m n$
- $\Theta(\log_m n)$

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Explanation of test 1

Task 3

- Time of recursion T(N) = 4T(N/2) + 1
- Depth of recursion tree is log N
- On each level of recursion tree: $1 \Rightarrow 4 \Rightarrow 6 \Rightarrow 4^x$ elements
- On the last level of tree:

$$x = \log N \Rightarrow 4^x = 4^{\log N} = 2^{\log N^2} = N^2$$
 elements

• Geometric series from 1 to N^2 with step equal to 4, its sum

$$\frac{1*(4^{\log N+1}-1)}{4-1} = \frac{4N^2-1}{3} = \Theta(N^2)$$

Examples:

- $T(N) = 4T\left(\frac{N}{2}\right) + 1 \Rightarrow T(N) = \Theta(N^2)$
- $T(N) = 2T\left(\frac{N}{2}\right) + O(N) \Rightarrow T(N) = \Theta(N \log N)$
- $T(N) = 4T\left(\frac{N}{4}\right) + 1 \Rightarrow T(N) = \Theta(N)$
 - Note, depth of recursion tree in this case is log₄ N



- Task 4
- Stable sort
 - The order of equivalent elements is guaranteed to be preserved
 - Stability can be important for some cases, when equal elements are distinguishable
- William's vs Floyd's methods for building heap
 - Consecutive inserts vs sift down the roots of each subtree
 - Complexity O (N logN) vs O(N)
- Bottom-up approach for merge sort
 - Bottom-up: sort little groups in the whole array, then move to the larger groups
 - Doesn't required recursion

Thank you!