

Greedy Algorithms

Introduction

- Greedy algorithms are a class of algorithms that make locally optimal choices at each step with the aim of finding a global optimum.
- These algorithms are simple, efficient, and often used for optimization problems where we seek the best solution from a set of possibilities.

Key Characteristics

- Greedy algorithms make decisions based on the current best option without considering the overall future consequences.
- They are greedy in the sense that they always choose the option that appears to be the best at the moment.
- Greedy algorithms do not always guarantee the globally optimal solution, but they often produce acceptable results quickly.

Basic Workflow

1. **Initialization:** Start with an empty solution or a solution with the initial state.
2. **Selection:** Choose the best available option based on a specific criteria.
3. **Feasibility Check:** Verify if the chosen option satisfies all constraints and requirements.
4. **Update:** Modify the solution and update the problem state.
5. **Termination:** Repeat steps 2 to 4 until a certain condition is met.

Kadane's Algorithm

- ▶ **Largest Sum Contiguous Subarray**
- ▶ Given an array `arr[]` of size **N**. The task is to find the sum of the contiguous subarray within a `arr[]` with the largest sum.

Pseudocode:

- ▶ *Initialize:*
 $\text{max_so_far} = \text{INT_MIN}$
 $\text{max_ending_here} = 0$
- ▶ *Loop for each element of the array*
- ▶ (a) $\text{max_ending_here} = \text{max_ending_here} + a[i]$
 (b) *if*($\text{max_so_far} < \text{max_ending_here}$)
 $\text{max_so_far} = \text{max_ending_here}$
 (c) *if*($\text{max_ending_here} < 0$)
 $\text{max_ending_here} = 0$
 return max_so_far

Code:

```
int maxSubArraySum(int a[], int size)
{
    int max_so_far = INT_MIN, max_ending_here = 0;

    for (int i = 0; i < size; i++) {
        max_ending_here = max_ending_here + a[i];
        if (max_so_far < max_ending_here)
            max_so_far = max_ending_here;

        if (max_ending_here < 0)
            max_ending_here = 0;
    }
    return max_so_far;
}
```

Jump Game

- ▶ You are given an integer array `nums`. You are initially positioned at the array's first index, and each element in the array represents your maximum jump length at that position.
- ▶ Return `true` if you can reach the last index, or `false` otherwise.

Solution

```
bool canJump(vector<int>& nums) {  
    int canJumpTill = 0;  
  
    for(int i=0; i<nums.size(); i++)  
    {  
        if(canJumpTill >= i)  
            canJumpTill = max(canJumpTill, i+nums[i]);  
        else  
            break;  
    }  
  
    return (canJumpTill >= nums.size()-1);  
}
```

Find Original Array From Doubled Array

- ▶ An integer array original is transformed into a doubled array changed by appending twice the value of every element in original, and then randomly shuffling the resulting array.
- ▶ Given an array changed, return original if changed is a doubled array. If changed is not a doubled array, return an empty array. The elements in original may be returned in any order.

Solution

```
multiset<int> mst;
for(auto e: a)
    mst.insert(e);
vector<int> ans;
while(mst.size())
{
    int x = *mst.begin();
    ans.push_back(x);
    mst.erase(mst.find(x));
    mst.erase(mst.find(2*x));
}
cout<<ans<<"\n";
```

Advantages

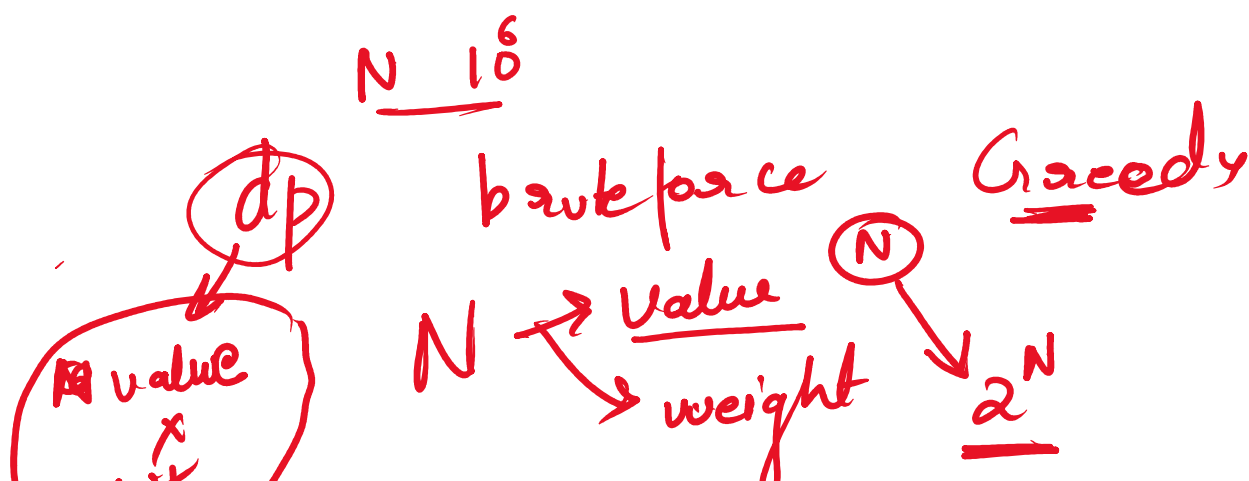
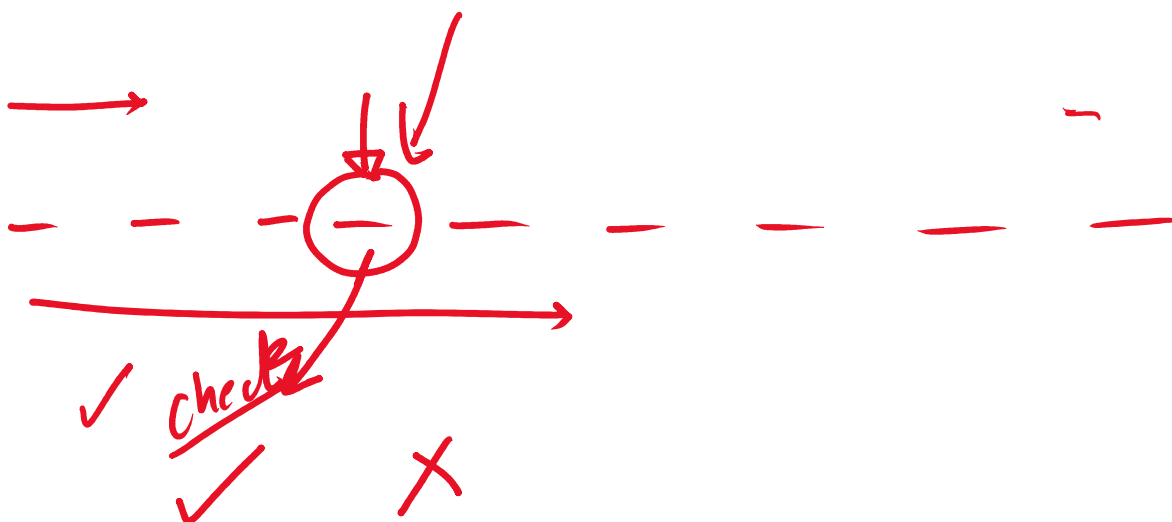
- Greedy algorithms are relatively easy to implement and efficient in terms of runtime complexity.
- They work well for problems where making the locally optimal choice at each step leads to a globally optimal solution.
- They can provide quick solutions for large-scale problems when other methods might be computationally expensive.

Limitations

- Greedy algorithms might not always produce the best solution, as they lack global awareness.
- There's a risk of getting stuck in a suboptimal solution, especially if the locally optimal choices accumulate.
- Careful analysis is required to ensure that greedy algorithms indeed lead to optimal or acceptable solutions.



Thank You



→ weight α

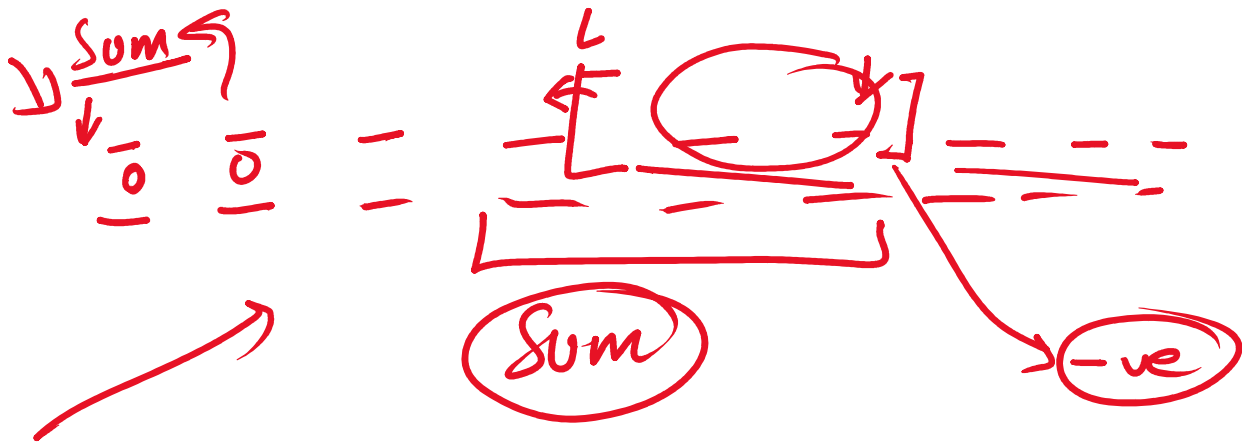
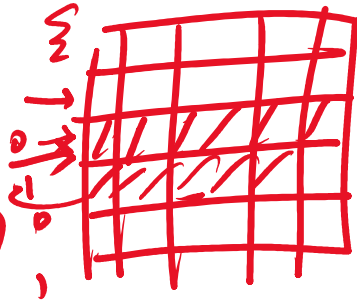
10⁶

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dp[n][m] is 0

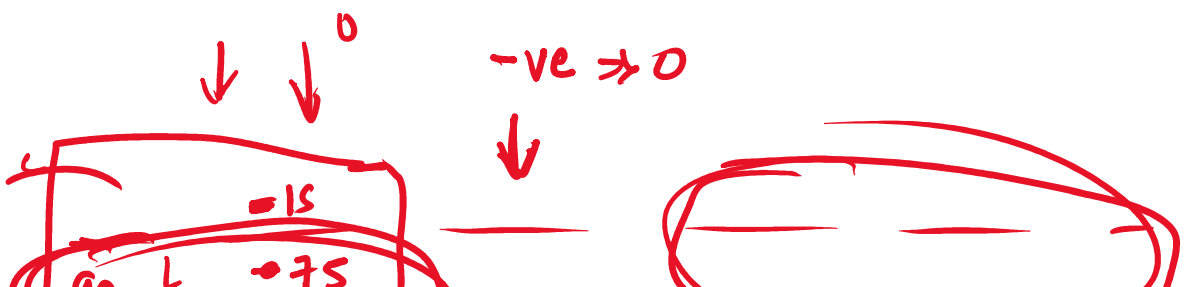
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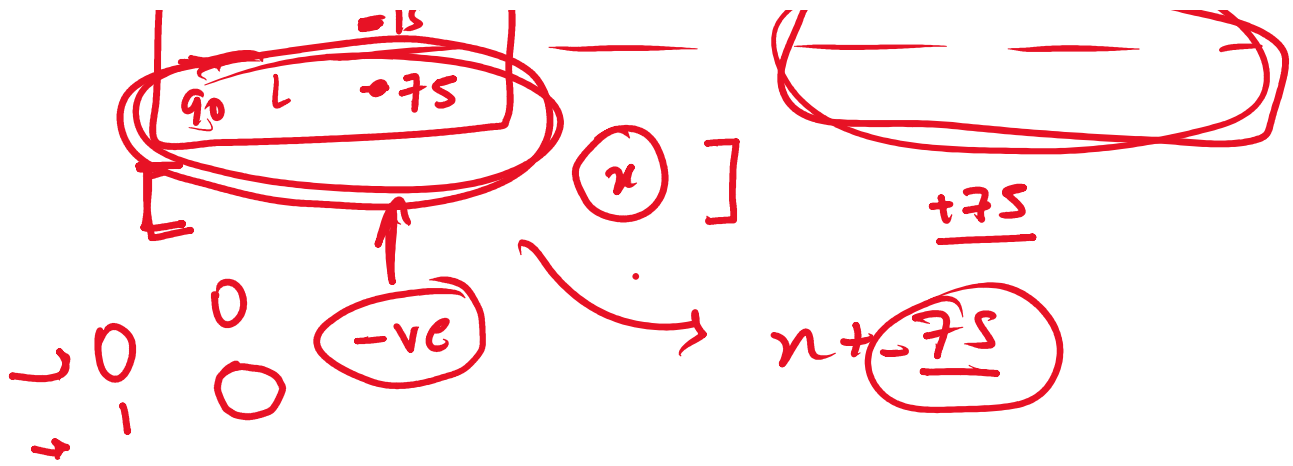
dp[2][m] 10⁸



ans = max(ans, sum)

$\left. \begin{array}{l} \text{sum} < 0 \\ \text{sum} = 0 \end{array} \right\}$





dp[2][n]

Init:

④ sel \rightarrow $sum += a[i]$
 feas \rightarrow $ans = \max(ans, sum)$
 ④ Updat \rightarrow $sum = 0$ ($sum < 0$)
 Termin

Ans

min

min



DP

n^2

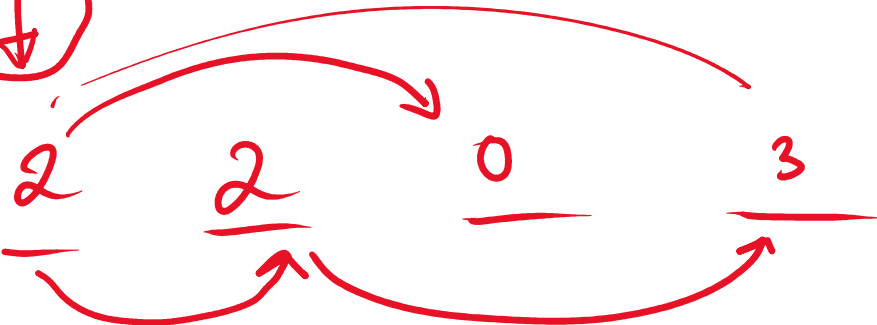
vis



~~0~~

max ind

↓



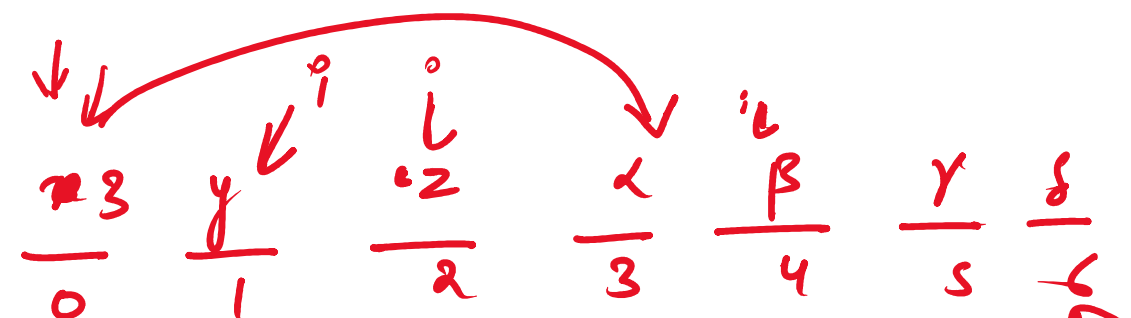
n^2

0	1	2	3
2	2	0	3
↑	<u>↑</u>		↑
	<u>↑</u>		

↑ Iteration ↑
 └──────────┘
max = 2

2 2
 S 10 0 0 0 0 0 0 1
 t1 t1 ~~0~~ -1

S 0 0 0 1 0 0 0 0 10 ✓
 t1 → t1

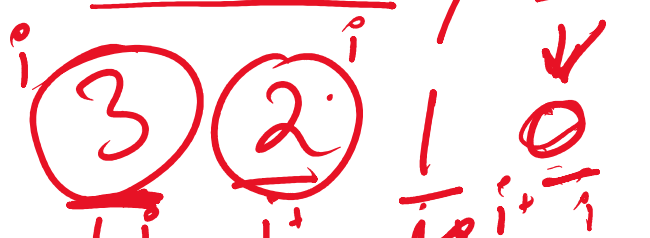


max index Reached $\geq i$

2 char false

max index = ϕ 3

(0 + 3) \rightarrow $i + a[i]$



$0 + 3 \geq 3$

$\max \geq i$



$\max \geq i$

$1 + 2$



Ini \rightarrow max index $\Rightarrow 0$

sel $\rightarrow i + a[i]$

feas check $\rightarrow i < \max \text{ index}$

update $\rightarrow \max \text{ index} = \max(\max, \text{sel})$

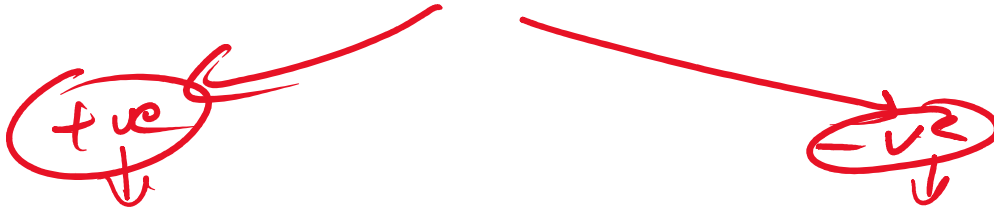
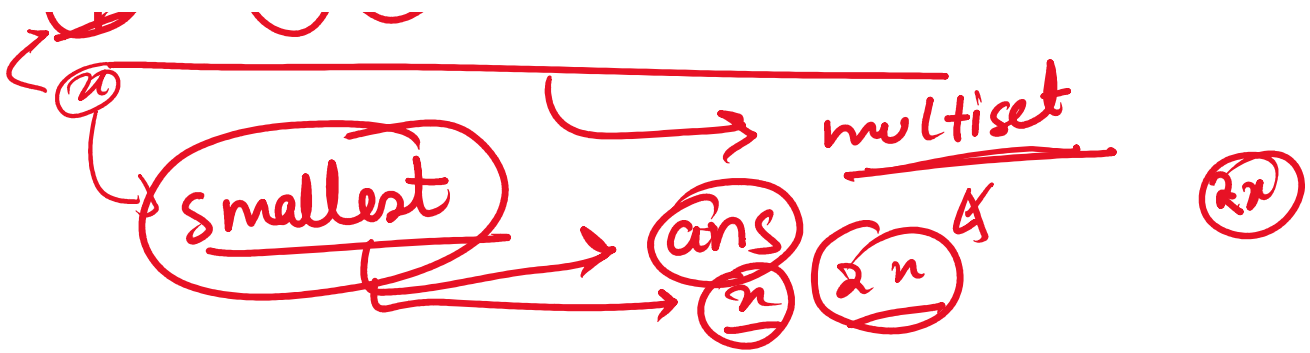
feasible update Term \rightarrow max index = $\max(\max, i + a[i])$
 $\leq n$

1 22 3 4 6 8 10 20

Original

$\{d \quad \beta \quad \gamma \quad \delta\}$
 $\left[\begin{array}{ccccccccc} d & \beta & \gamma & \delta & 2d & 2\beta & 2\gamma & 2\delta \\ 2 & 10 & 1 & 3 & 4 & 20 & 2 & 6 \end{array} \right]$
 $2 \quad 10 \quad 1 \quad 3$
 d

~~2~~ 10 1 3
 2 10 1 3 4 20 26
~~1~~ 2 (2) (3) 4 6 (10) 20
 (20) 1 11 12 13 14 15 16 17 18 19 21 22 23 24 25 27 28 29 30



m |

multiset.erase(2n)

multiset.erase(multiset.
find(2n))

