

Characteristics of Greedy

- It works by making the most optimal decision at any moment of time.
- The objective of this method is to ^{choose} ~~make~~ the optimal solⁿ for every sub-problem so that you can get optimal solⁿ for the overall problem.
- To solve a particular problem in an optimal way using greedy approach, there is a set or list of candidates C .
- Once a candidate is selected in the solⁿ, it is there forever, once a candidate is excluded from the solⁿ it is never considered.
- To construct the solⁿ in an optimal way, Greedy algorithm maintains two sets: One set contains candidates that have already been considered and chosen, while the other set contains candidates that have been considered but rejected.

Characteristics of Divide & Conquer

- The divide and conquer breaks a problem into sub-problems that are similar to original problem.
- The sub-problems are then solved and the answers are combined.
- Each sub-problem should be smaller than the original problem.
- Divide & Conquer has 3 parts: -
 - ① Divide the problem into number of sub-problems that are smaller instances of the same problem.

- ② Conquer the subproblem by solving them recursively
- ③ Combine the solutions to the subproblems into the solⁿ for the original problem.

Characteristics of Dynamic Programming

- ① It is used in optimization problems.
- ② Dynamic Programming solves problems by combining the solutions of subproblems.
- ③ Dynamic Programming solves each sub-problem just once and then saves its answer in a table, thereby avoiding the work of recomputing the answer every time.
- ④ Two main properties of Dynamic Programming are as follow:-

(a) Overlapping Sub-Problems

It is mainly used where the solⁿ of one sub-problem is needed repeatedly. The computed solⁿ are stored in a table so that these don't have to be re-computed. Hence, this technique is needed where over-lapping sub-problem exists.

(b) Optimal Sub-Structure

A given problem has optimal substructure property, if the optimal solⁿ of the given problem can be obtained using optimal solⁿ of its sub-problems.

1 Differences

Divide & Conquer

- ① It involves three steps:
Divide the problem, Conquer the sub problem and Combine the solⁿ
- ② They are recursive
- ③ It does more work on subproblems and hence has more time consumption
- ④ It is a top-down approach
- ⑤ In this subproblems are independent of each other
- ⑥ for e.g. Merge Sort & Binary Search etc.

Dynamic Programming

- ① It involves four steps:
Characterize the structure of optimal solⁿ
Recursively define the value of optimal solⁿ
Compute the value of optimal solⁿ and
Construct an optimal structure solⁿ from computed info.
- ② It is non Recursive
- ③ It solves subproblems only once and then stores it in table
- ④ It is a Bottom-up approach
- ⑤ In this sub-problems are interdependent
- ⑥ for e.g. Matrix Multiplication

Divide & Conquer.

- ① Optimises by breaking down a sub-problem into simpler versions of itself and using multi-threading & recursion to solve.
- ② Always finds the optimal soln but is slower than Greedy.
- ③ Requires some memory to remember recursive calls.

Greedy

- ① Optimises by making the best choice at the moment.
- ② Doesn't always find the optimal soln, but is very fast.
- ③ Requires almost no memory.

Greedy Method

- ① In greedy we make whatever choice seems best at that moment and then solve the sub-problems arising after the choice is made.
- ② In Greedy Method, sometimes there is no such guarantee of getting optimal solⁿ.
- ③ It is more efficient in terms of memory as it never look back or revise previous choices.
- ④ Greedy Methods are generally faster.
- ⑤ Less efficient than Dynamic Programming
- ⑥ For e.g. Fractional Knapsack

Dynamic Programming

- ① In Dynamic Programming, we choose at each step, but the choice may depend on the solⁿ to sub-problems.
- ② It is guaranteed that Dynamic Programming will generate an optimal solⁿ as it generally considers all possible cases and then chooses the best.
- ③ It requires dp table for memorization and it increases its memory complexity
- ④ Dynamic Programming is generally slower
- ⑤ More efficient than Greedy method
- ⑥ For e.g. 0/1 Knapsack Problem

Tower of Hanoi

Proof of No. of moves by mathematical induction.

Prove :- The min number of moves needed to move n disks from one pin in ToH problem is $2^n - 1$ for all natural no.

I $P(1)$ is true.

II Given: $P(k)$ is true.

Prove: $P(k+1)$ is true.

Now 1 disk.

$$2^1 - 1 = 2 - 1 \Rightarrow 1$$

Now min num of moves required to move k disks from one point to another in ToH problem is $2^k - 1$.

Assume we have $k+1$ disks.
we first move the top k disks
by $2^k - 1$ moves.

Now move the largest disk.
 \therefore 1 more move needed.

Again to move ~~$k+1$~~ k disks from second to third pin $2^k - 1$ moves needed.

∴ Total no. of moves

$$= 2^{k-1} - 1 + 1 + 2^k - 1$$

$$= 2^k + 2^k - 1 \Rightarrow 2(2^k) - 1$$

$$= 2^{k+1} - 1$$

∴ The min no. of move needed for $k+1$ disks is $2^{k+1} - 1$

∴ Min num of moves needed for n disks is $2^n - 1$

⇒ Time complexity $T(n) \propto$ No. of moves

For e.g. No. of moves needed for 3 disks will be

$$2^3 - 1 \Rightarrow 8 - 1 \Rightarrow 7$$

So, No. of moves needed for 4 disks will be

$$2^4 - 1 \Rightarrow 15$$

$$15 = 7 + 1 + 7$$

$$T(4) = T(3) + 1 + T(3)$$

$$T(4) = 2T(3) + 1$$

Now, Generalizing it ,

$$T(n) = 2T(n-1) + 1$$

Now by using substitution method.

$$T(n) = 2T(n-1) + 1 \quad \text{--- (1)}$$

$$= 2(2T(n-2) + 1) + 1$$

$$= 2^2(T(n-2)) + (1+2) \quad \text{--- (2)}$$

$$= 2^2(2T(n-3) + 1) + (1+2)$$

$$= 2^3(T(n-3)) + (1+2+4) \quad \text{--- (3)}$$

$$\Rightarrow 2^k T(n-k) + 2^k - 1$$

$$\Rightarrow 2^n T(0) + 2^n - 1$$

$$\Rightarrow 2^n + 2^n - 1$$

$$\boxed{\Rightarrow T(n) = O(2^n)}$$