

# Dynamic Programming.

## 0/1 Knapsack, Recursive

How to identify whether it is DP problem or not

- choice
- optimal  $\begin{cases} \text{Max} \\ \text{Min} \\ \text{Larger} \\ \text{Smaller} \end{cases}$

Recursive  $\Rightarrow$  Memoize  $\Rightarrow$  ~~DP Down~~ Bottom up

(DP)

Recursive Code

$\downarrow$   
make choice diagram

After writing choice diagram Code (writing will be easy)

- Dynamic programming is used to solve a wide variety of discrete optimization problems such as scheduling, string-editing, packaging and inventory management
- Break problems into sub-problems and combine their solutions to larger problem.
- In contrast to divide and conquer there may be relationship across subproblems.

## 0/1 Knapsack problem.

- We are given a knapsack of capacity  $C$  and a set of  $n$  objects numbered  $1, 2, \dots, n$ . Each object  $i$  has  $w_i$  and profit  $P_i$ .

## Steps to Design a Dynamic programming Algorithm

1. Characterize optimal Structure
2. Recursively define the value of an optimal solution
3. Compute the value bottom up.
4. If needed Construct an optimal solution.

### Knapsack

- Given  $n$  objects and a "Knapsack"
- Item  $i$  weighs  $w_i > 0$  kilograms and has value  $v_i$
- Knapsack has Capacity of  $W$  kilograms.
- Goal: fill knapsack so as to maximize total value.

Item	Value	Weight
1	1	1
2	6	2
3	18	5
4	22	6
5	28	7

Knapsack weight  
 $W = 11$

Ex:  $\{3, 4\}$  has value 40

Greedy :- repeatedly add item with maximum  $(v_i/w_i)$

Ex  $\{5, 2, 1\}$  achieves only value of 35  $\Rightarrow$

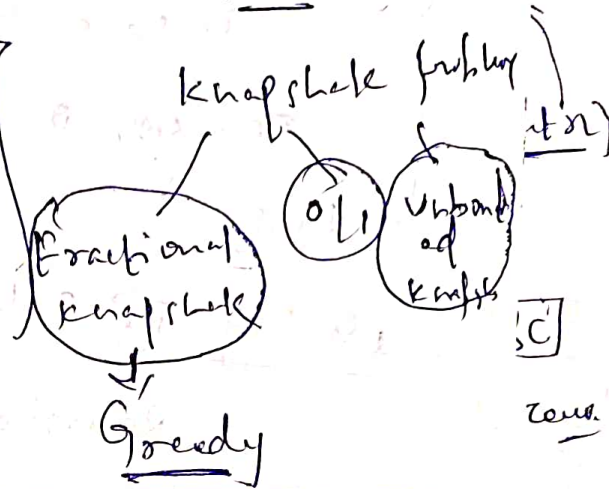
Greedy not optimal.



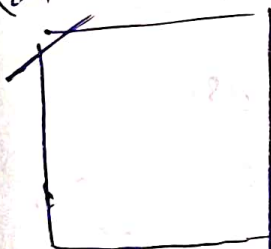
# 0-1 Knapsack problems.

- 1) Subset Sum
- 2) Equal Sum partition
- 3) Count of subset sum
- 4) minimum subset sub Diff
- 5) Target Sum
- 6) # of subset  $\leq$  gives d/f.

• all problems are inter connected each other



Max profit



Bag (Knapsack)

• keeps item such that profit should be max

Problem

	$I_1$	$I_2$	$I_3$	$I_4$
wt	1	3	4	5
val	1	4	5	7
val				

Constraint  $\Rightarrow$  Bag fix  
fixed weight

- $\bigcirc$   $P_1$   $W_1$
- $\bigcirc$   $P_2$   $W_2$
- $\bigcirc$   $P_3$   $W_3$
- $\bigcirc$   $P_4$   $W_4$



knapsack

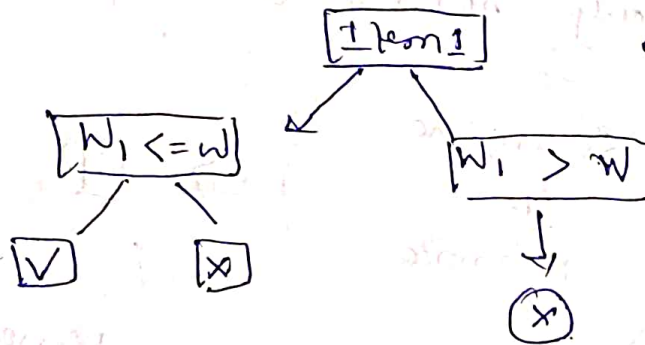
• Unbounded

item  $\rightarrow$  unlimited supply



0.5, 0.6  
Fractional (half/divide)  
 $\rightarrow$  Greedy  
0/1  $\rightarrow$  keep/don't keep

## Choice Diagram



- Base Cond is imp in recursive solutions
  - Choice diagrams
- code

## Algorithm

→ return max profit

int knapsack ( int wt[], int val[], int w, int n )

wt Array

val Array

Array size

// think of Base Condition.

// Think of the smallest Valid Input for getting BC

// array size may be zero, weight may be zero.

Base Cond [ if ( n == 0 || w == 0 ) // smallest Valid output - return 0; ]

if ( wt[n-1] <= w ) // choice diagram.

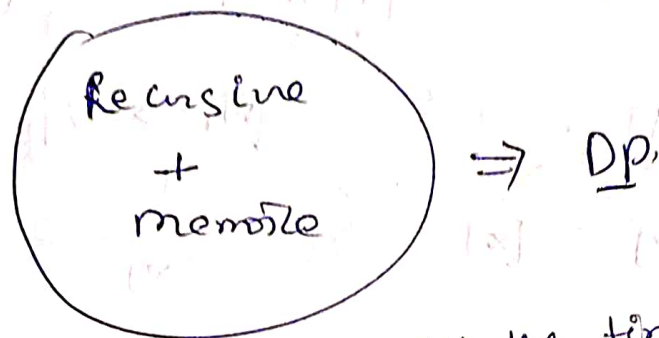
return max( val[n-1] + knapsack( wt, val, w - wt[n-1], n-1 ), knapsack( wt, val, w, n-1 ) )

else if ( wt[n-1] > w )

return knapsack( wt, val, w, n-1 );

Choice Diagram

- To memoize after identifying recursive code it needs only 210th most line of code.



- W and n changes all the time, so we need to apply matrix (Bottom up to those (W and n))

knapsack ( wt[], val[], (W), (n) )  
 $\searrow$   $\searrow$   
 wt wt[n-1]  $\searrow$  n-1

int t[n+1][W+1]  
 $\searrow$  matrix initialize with -1.  
 $\searrow$  memset ( t, -1, size of t )

-1	-1	-1	-1
-1	-1	-1	-1
-1	-1	-1	-1

- DP with Bottom up and memoize will have same time complexity

- Apply memoization only for the parameter which changes.