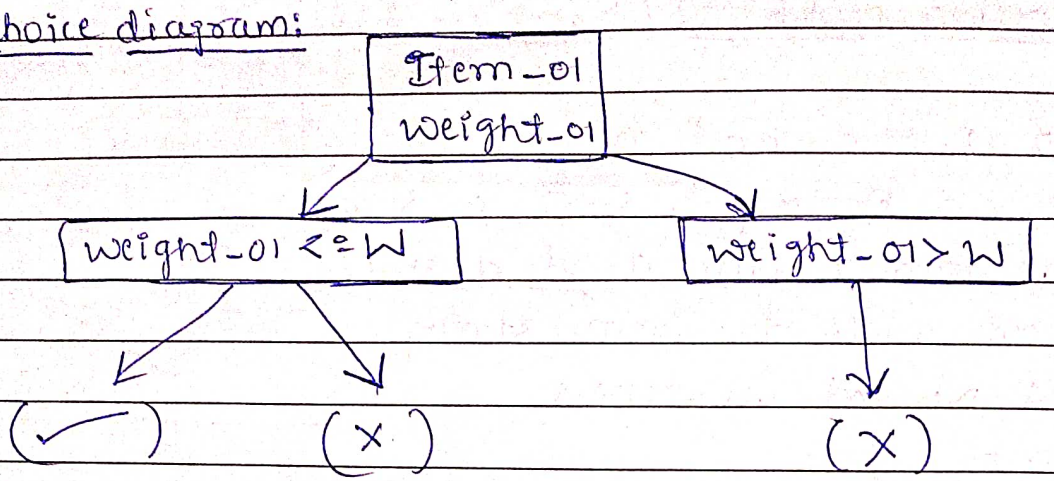
0-1 KnapSack : [Recursive function]

Given weights and value of n items, put these items in a knapsack of capacity W, to get the maximum total value in the knapsack.

Given : W : capacity of knapsack
n : Number of items
val[0.... n-1] : values associated with n items
weight[0.... n-1] : weight —"— —"—

* The Base condition : The Best way to choose a Base condition is to check the smallest valid input, In this case, we can have 2 options i.e, if n==0 which number of items is zero. then what we can include in knapsack, nothing we can, similarly, if the weight W==0 - the capacity of knapsack is only 0 then we cannot include or add even a single item, so inboth of these cases the maxProfit will be 0

⇒ Choice diagram:



In the above diagram, A item has a weight & value associated with it. For every item we have 2 choices,

① If the weight of the item <= than the weight of the knapsack. Again, here also we have 2 choices to include it in knapsack or not.

② If the weight of the item > than the weight of knapsack than obviously we cannot include / add it into the knapsack.

* Recursive Function:

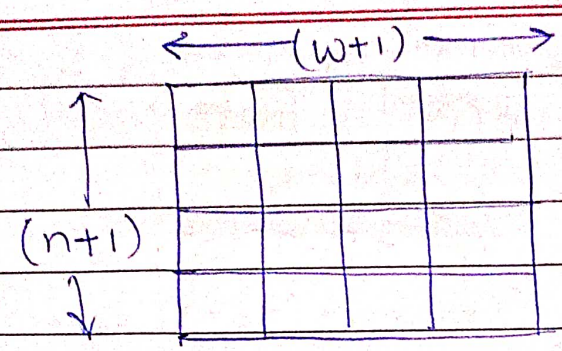
```
int knapSack(int W, int val[], int weight[], int n)
{
    if (n == 0 || W == 0)
        return 0;
    else if (weight[n-1] <= W)
        return max(val[n-1] + knapsack(W - weight[n-1],
                   val, weight, n-1),
                   knapSack(W, val, weight, n-1));
    else if (weight[n-1] > W)
        return knapsack(W, val, weight, n-1);
}
```

=> Time complexity : $O(2^n)$ As it checks for each subset.
=> Auxiliary space : O(1)

=> Memoization : int t[n][W];
memset(t, -1, sizeof(t));

```
int knapSack(int W, int val[], int weight[], int n)
{
  if(n==0 || W==0)
    return 0;
  if(t[n][W]!=-1)
    return t[n][W];
  else if(weight[n-1]<=W)
  {
    t[n][W] = max(val[n-1]+knapSack(W-weight[n-1],
              val, weight, n-1), knapSack(W, val,
              weight, n-1));
    return t[n][W];
  }
  else if(weight[n-1]>W)
  {
    t[n][W] = knapSack(W, val, weight, n-1);
    return t[n][W];
  }
}
```

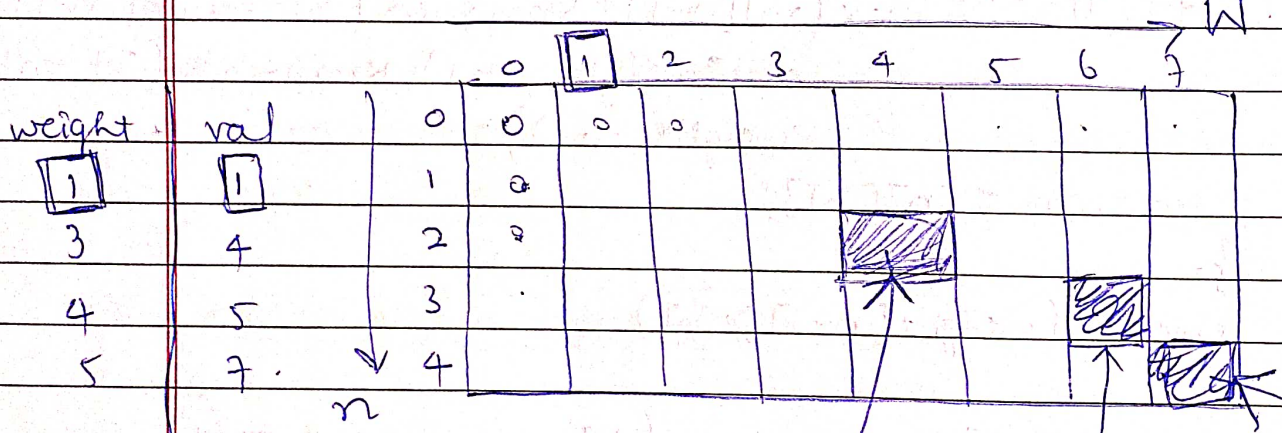
* Time complexity of above memoized function is : O(N*M)
* Space complexity of above function is : O(N*M)



⇒ Top Down DP: Here, in Top down DP, the Base condition of Recursion becomes the initialization.

ex: val[] = [1, 4, 5, 7] W = 7
weight[] = [1, 3, 4, 5] n = 4

So, we have to create a 2D matrix of size dp[n+1][W+1].



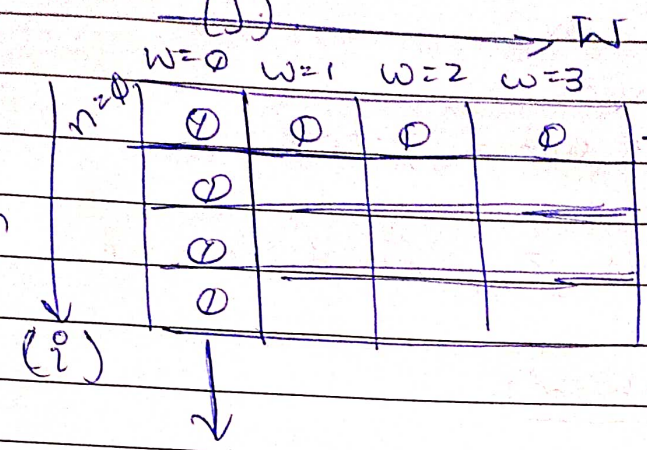
This value is a solution of subproblem i.e., when Knapsack capacity is 4.

W = 4
val[] = [1, 4] and
weight[] = [1, 3]

W = 6
val[] = [1, 4, 5]
weight[] = [1, 3, 4]

W = 7
val[] = [1, 4, 5, 7]
weight[] = [1, 3, 4, 5]

The Base condition of Recursion becomes the initialization of dp[][], i.e, when n==0 or W==0. Initialize the dp[0][0] matrix with 0.



→ It means that we may have the capacity of Bag but no items present as n=0

It means it does matter how much elements / items we can include when we have bag capacity as 0. We cannot include a single it. so, it is 0.

```
int knapSack(int W, int val[], int weight[], int n)
{  int dp[n+1][W+1];
   for(int i=0; i<n+1; i++)
      for(int j=0; j<W+1; j++)
         if(i==0 || j==0)
            dp[i][j] = 0;

   for(int i=1; i<n+1; i++)
      for(int j=1; j<W+1; j++)
      {
         if(weight[i-1] <= j)
            t[i][j] = max(val[i-1] + t[i-1][j-weight
                          [i-1], t[i-1][j]);
         else
            t[i][j] = t[i-1][j];
      }  return dp[n][W];
}
```

0/1 Knapsack : Bottom Up [Tabulation]

```
int knapsack(int W, int wt[], int val[],
             int n)
{
    int DP[n+1][W+1];
    for(int i=0; i<=n; i++)
      for(int j=0; j<=W; j++)
        if(i==0 || j==0)
          DP[i][j] = 0;

    for(int i=1; i<=n; i++)
      for(int j=1; j<=W; j++)
      {
        if(wt[i-1] <= j)
          DP[i][j] = max(val[i-1] +
                         DP[i-1][j-wt[i-1]],
                         DP[i-1][j]);
        else
          DP[i][j] = DP[i-1][j];
      }

    return DP[n][W];
}
```

Initialization

⇒ Tracing of the Above code with the following input data :

W = 4, wt = | 4 | 5 | 1 |
n = 3, val = | 1 | 2 | 3 |

max profit: 3

DP:

n ↓ / W → j	0	1	2	3	4
i 0	0	0	0	0	0
1	0	1	1	1	1
2	0	1	1	1	1
3	0	3	3	3	3

i = 1
j = x, 2, 3, 4, 5

wt[i-1] ⇒ wt[0] = 4 <= j ⇒ 4 <= 1

↓

false, so,

DP[i][j] = 0

Again, 4 <= 2, false DP[1][2] = 0
4 <= 3, false DP[1][3] = 0
4 <= 4, true max(1, 0) = 1

val[i-1] ⇒ val[1-1] ⇒ val[0] = (1 + 0, 0)

DP[i-1][j-wt[i-1]] ⇒ DP[0][0] = 0

↓ ↓

[0][4-4] = [0][0] DP[0][4] = 1

wt[i-1] ⇒ wt[0] = 4

i = 2
j = X, 1, 2, 3, 4, 5.

wt[i-1] => wt[1] = 5 <= 1 false, DP[2][1] = 0
5 <= 2 " DP[2][2] = 0
5 <= 3 " DP[2][3] = 0
5 <= 4 " DP[2][4] = 0

i = 3
j = X, 1, 2, 3, 4, 5

① wt[i-1] => wt[3-1] => wt[2] = 1

1 <= 1, true,

max(val[i-1] + DP[i-1][j - wt[i-1]], DP[i-1][j]);

val[i-1] => val[3-1] => val[2] = 3
DP[i-1][j-wt[i-1]]
↓ ↓ 3-1 = 2 => wt[2] = 1
[2] [1-1] => DP[2][0] = 0
DP[2][1] = 0.

max(0+3, 0) = 3

(ii) $K=2$, true $\rightarrow$ $wt[i-1] \Rightarrow 1 <= 3$

$val[i-1] = 3$.
$DP[i-1][j-wt[i-1]]$
$[2][2-1] \Rightarrow DP[2][1] = 0$

$DP[i-1][j] \Rightarrow DP[2][2]$
0.

$\max(3+0, 0) = 3$.

(iii) $K=3$, true. $wt[i-1] = 1 <= 3$.

$val[i-1] = 3$
$DP[i-1][j-wt[i-1]]$
$[2][3-1] \Rightarrow DP[2][2] = 0$

$DP[i-1][j] = [2][3] = 0$

$\max(3+0, 0) = 3$, $i = 3$

(iv) $K=4$, $wt[i-1] \Rightarrow 1 <= 3$ true.

$val[i-1] = 3$
$DP[i-1][j-wt[i-1]]$
$[2][4-wt[3-1]] = DP[2][4-1]$
$DP[2][3] \Rightarrow 0$

$DP[i-1][j] = DP[2][4]$
$\downarrow$
0.

$\max(3+0, 0) = 3$

return $DP[n][W] = 3$