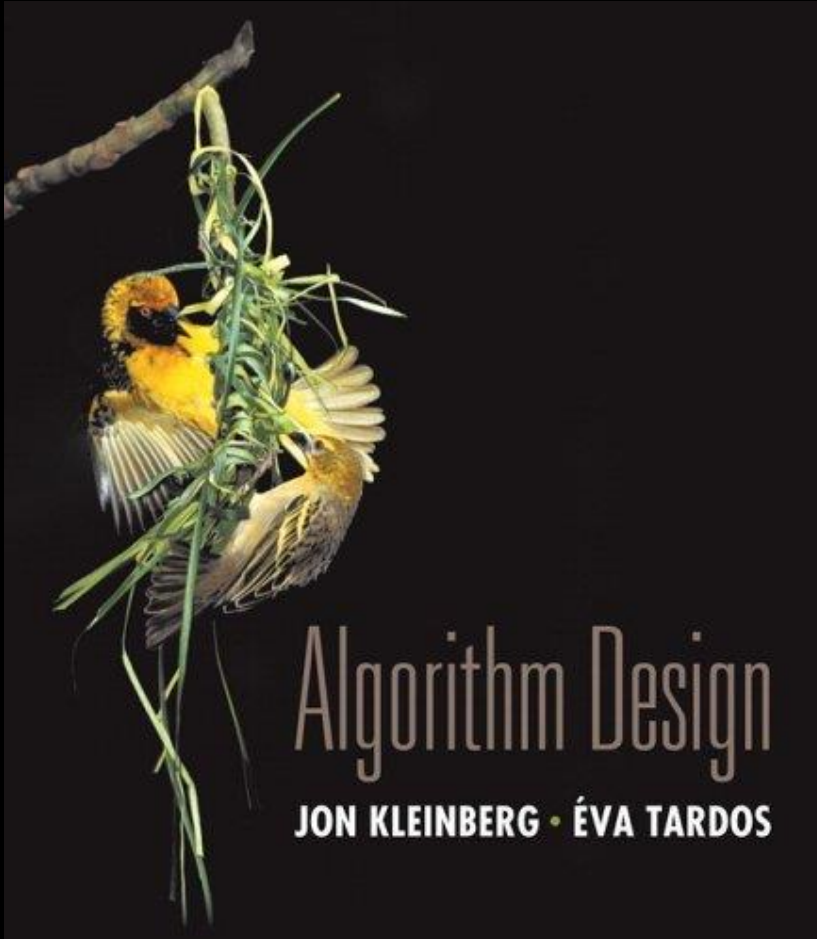


Chapter 6

Dynamic Programming



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6.1 Weighted Interval Scheduling

Weighted Interval Scheduling: DP 2 - Bottom-Up

Dynamic programming to fill up the memoization table M :

```
Input:  $n, s_1, \dots, s_n, f_1, \dots, f_n, v_1, \dots, v_n$ 
```

```
Sort jobs by finish times so that  $f_1 \leq f_2 \leq \dots \leq f_n$ .
```

```
Compute lastCompat(1), lastCompat(2), ..., lastCompat(n)
```

```
 $M = [] * n$ 
```

```
Iterative-Compute-Opt {
```

```
     $M[0] = 0$ 
```

```
    for  $j = 1$  to  $n$ 
```

```
        #Option of having item  $j$  in the optimal solution  
        of subproblem  $OPT(j)$ 
```

```
        If  $v_j + M[\text{lastCompat}(j)] > M[j-1]$ 
```

```
             $M[j] = v_j + M[\text{lastCompat}(j)]$ 
```

```
        #Option of not having item  $j$  in the optimal  
        solution of subproblem  $OPT(j)$ 
```

```
        Else
```

```
             $M[j] = M[j-1]$ 
```

```
}
```

Weighted Interval Scheduling: DP 2 - Bottom-Up

Obtaining the optimal **solution** for each subproblem, not just value:

- Array Sol stores optimal solution relative to $OPT(i)$ in $Sol(i)$

```
Iterative-Compute-Opt {
  M[0] = 0
  Sol[0] = {}
  for j = 1 to n
    #Option of having item j in the optimal solution
    #of subproblem  $OPT(j)$ 
    If  $v_j + M[\text{lastCompat}(j)] > M[j-1]$ 
      M[j] =  $v_j + M[\text{lastCompat}(j)]$ 
      Sol[j] = Sol[lastCompat(j)]  $\cup$  {j}
    #Option of not having item j in the optimal
    #solution of subproblem  $OPT(j)$ 
    Else
      M[j] = M[j-1]
      Sol[j] = Sol[j-1]

  Return Sol[n]
}
```

Weighted Interval Scheduling: DP 2 - Bottom-Up

Very important: When the code enters this "If" it does not mean that item j is in the optimal solution for the full problem...

... it only mean that item j belongs to the optimal solution for subproblem $OPT(j)$

```

#Option of having item j in the optimal solution
of subproblem  $OPT(j)$ 
If  $v_j + M[\text{lastCompat}(j)] > M[j-1]$ 
     $M[j] = v_j + M[\text{lastCompat}(j)]$ 
     $Sol[j] = Sol[\text{lastCompat}(j)] \cup \{j\}$ 
#Option of not having item j in the optimal
solution of subproblem  $OPT(j)$ 
Else
     $M[j] = M[j-1]$ 
     $Sol[j] = Sol[j-1]$ 

Return  $Sol[n]$ 
}
```

Weighted Interval Scheduling: DP 2 - Bottom-Up

Very important 2: Never use Sol in the recurrence relation, 99% chance of getting it wrong

First write recurrence relation/algo just to fill up memoization table M, then add Sol

```
of subproblem OPT(j)
If  $v_j + M[\text{lastCompat}(j)] > M[j-1]$ 
     $M[j] = v_j + M[\text{lastCompat}(j)]$ 
     $\text{Sol}[j] = \text{Sol}[\text{lastCompat}(j)] \cup \{j\}$ 
#Option of not having item j in the optimal
solution of subproblem OPT(j)
Else
     $M[j] = M[j-1]$ 
     $\text{Sol}[j] = \text{Sol}[j-1]$ 

Return Sol[n]
}
```

Maior subsequência crescente

Maior subsequência crescente: encontrando o tamanho

Input: $n, a_1, \dots, a_n,$

$M(1) \leftarrow 1$

For $j=2$ to n

$M(j) \leftarrow 1$

For $i=1$ to $j-1$ // faz $OPT(i) \leftarrow \max_i \{ 1+OPT(i) \mid i < j \text{ e } a(j) > a(i) \}$

If $A(i) < A(j)$ and $1+M(i) > M(j)$ then

$M(j) \leftarrow 1+M(i)$

End If

End For

End For

$\max \leftarrow 0, \text{maxi} \leftarrow 0$

For $i=1$ to n // encontra indice do maior $OPT(i)$

If $M(j) > \max$, set $\max \leftarrow M(j)$ and $\text{maxi} \leftarrow i$

End For

Return $M(\text{maxi})$

Maior subsequência crescente: encontrando o tamanho

Input: $n, a_1, \dots, a_n,$

$M(1) \leftarrow 1$

$Sol(1) = \{a_1\}$

For $j=2$ to n

$M(j) \leftarrow 1$

$Sol(j) \leftarrow a_j$

For $i=1$ to $j-1$ // faz $OPT(i) \leftarrow \max_i \{ 1+OPT(i) \mid i < j \text{ e } a(j) > a(i) \}$

If $A(i) < A(j)$ and $1+M(i) > M(j)$ then

$M(j) \leftarrow 1+M(i)$

$Sol(j) \leftarrow$ concatenation of $Sol(i)$ plus a_j

End If

End For

End For

$max \leftarrow 0, \text{maxi} \leftarrow 0$

For $i=1$ to n // encontra índice do maior $OPT(i)$

If $M(j) > max$, set $max \leftarrow M(j)$ and $\text{maxi} \leftarrow i$

End For

Return $Sol(\text{maxi})$

6.4 Knapsack Problem

Knapsack Problem: Bottom-Up

```
Input:  $n, w_1, \dots, w_N, v_1, \dots, v_N$ 

for  $w = 0$  to  $W$ 
     $M[0, w] = 0$ 

for  $i = 1$  to  $n$ 
    for  $w = 1$  to  $W$ 
        if  $(w_i > w)$ 
             $M[i, w] = M[i-1, w]$ 
        else
            if  $v_i + M[i-1, w-w_i] > M[i-1, w]$ 
                 $M[i, w] = v_i + M[i-1, w-w_i]$ 
            else
                 $M[i, w] = M[i-1, w]$ 

return  $M[n, W]$ 
```

Knapsack Problem: Bottom-Up

```
Input:  $n, w_1, \dots, w_N, v_1, \dots, v_N$ 

for  $w = 0$  to  $W$ 
     $M[0, w] = 0$ 

for  $i = 1$  to  $n$ 
    for  $w = 1$  to  $W$ 
        if ( $w_i > w$ )
             $M[i, w] = M[i-1, w]$ 
             $Sol[i, w] = Sol[i-1, w]$ 
        else
            if  $v_i + M[i-1, w-w_i] > M[i-1, w]$ 
                 $M[i, w] = v_i + M[i-1, w-w_i]$ 
                 $Sol[i, w] = Sol[i-1, w-w_i] \cup \{i\}$ 
            else
                 $M[i, w] = M[i-1, w]$ 
                 $Sol[i, w] = Sol[i-1, w]$ 

return  $Sol[n, W]$ 
```