

CS207 Design and Analysis of Algorithms

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Dynamic Programming

Dynamic Programming

- ▶ Particularly useful for optimization problems. Solution space. Constraints. Many feasible solutions, each of which satisfies the constraints, and has a value. We wish to find one feasible solution that minimises (maximizes) value.
- ▶ Steps:
 - ▶ Formulate the structure of an optimal solution
 - ▶ Recursively define the value of an optimal solution
 - ▶ Compute the value of an optimal solution, typically in a bottom-up fashion
 - ▶ Construct an optimal solution from computed information
- ▶ Works particularly when the following properties hold:
 - ▶ Optimal Substructure: an optimal solution to the problem contains within it optimal solutions to subproblems
 - ▶ Overlapping subproblems: recursive algorithm for the problem solves the same subproblems repeatedly

The 0-1 Knapsack Problem

- ▶ Ali Baba's brother Kassim goes into the cave of the 40 thieves.
- ▶ His donkey can carry a weight of at most W .
- ▶ He finds n metal ingots arranged in order.
- ▶ Ingot i is labelled by its (integer) weight w_i and its (integer) value v_i .
- ▶ Kassim's avarice knows no bounds.
- ▶ He will carry home the largest possible value.
- ▶ We need an algorithm for him to run.
- ▶ A trivial algorithm looks at each of the 2^n possible subsets: check if the subset has a total weight of at most W ; declare it the “best-so-far” if the sum of its members' values exceeds those of the subsets seen earlier; the “best-so-far” in the end is the best. This takes $O(n \cdot 2^n)$ time, because a subset and its complement together cover all n items.

0-1 Knapsack Solution: Recursive formulation

- ▶ Look at item n .
- ▶ Two possibilities: take it, or leave it
- ▶ If we take it, our residual capacity decreases to $W - w_n$.
- ▶ If we leave it, our residual capacity remains W .
- ▶ Suppose we have solved
 1. the subproblem on $n - 1$ items with maximum weight $W - w_n$, and know $A[n - 1, W - w_n]$, the maximum value that can be carried off in that case, and
 2. the subproblem on $n - 1$ items with maximum weight W , and know $A[n - 1, W]$, the maximum value that can be carried off in that case.

0-1 Knapsack Solution: Recursive formulation

- ▶ If $A[n - 1, W - w_n] + v_n \geq A[n - 1, W]$, then the solution of (1) plus item n is our solution; set
$$A[n, W] = A[n - 1, W - w_n] + v_n$$
- ▶ Otherwise, the solution of (2) is our solution; set
$$A[n, W] = A[n - 1, W] + v_n$$

Recursive Binary Knapsack

w , v and B are global arrays; B is initialized to all 0

Algorithm 1 RBK(n , W)

```
1: if ( $n = 0$  or  $W = 0$ ) return 0;  
2:  $a = \text{RBK}(n - 1, W - w_n) + v_n$ ;  
3:  $b = \text{RBK}(n - 1, W)$ ;  
4: if  $a \geq b$  then  
5:    $B[n, W] = 1$ ; return  $a$ ;  
6: else  
7:   return  $b$ ;  
8: end if
```

Binary Knapsack Print Solution

Algorithm 2 BKPS(n, W)

```
1: if ( $n = 0$  or  $W = 0$ ) return;  
2: if  $B[n, W]$  then  
3:   BKPS( $n - 1, W - w_n$ ); Print  $n$ ;  
4: else  
5:   BKPS( $n - 1, W$ );  
6: end if  
7: return
```

- ▶ BKPS runs in $O(n)$ time; each call spawns only one recursive call, and that too with a decremented n
- ▶ RBK runs in $\Omega(nW)$ time; usually much more, because there are overlapping subproblems that are repeatedly solved
- ▶ Not good!

Iterative Binary Knapsack

Algorithm 3 IBK(W, w, v, n)

```
1: Initialize 2-D array  $A[0 \dots n, 0 \dots W]$  to all zero
2: for  $i = 1$  to  $n$  do
3:   for  $w = 1$  to  $W$  do
4:      $A[i, w] = A[i - 1, w]; B[i, w] = 0;$ 
5:     if  $w \geq w_i$  and  $A[i - 1, w] < A[i - 1, w - w_i] + v_i$  then
6:        $A[i, w] = A[i - 1, w - w_i] + v_i; B[i, w] = 1;$ 
7:     end if
8:   end for
9: end for
```

- ▶ IBK runs in $O(nW)$ time
- ▶ The total length of the input is
 $\log W + \sum_{i=1}^n w_i + \sum_{i=1}^n v_i + \log n = \Omega(n + \log W)$
- ▶ IBK can be an exponential time algorithm in the worst case
- ▶ Not for nothing was Kassim still undecided when the thieves returned!
- ▶ 0-1 Knapsack is an NP-complete problem; a polynomial time algorithm is not known

Memoized Binary Knapsack

A and B are global $(n + 1) \times (W + 1)$ arrays; B is initialized to all 0; A to all -1

Algorithm 4 MBK(n, W)

```
1: if ( $A[n, W] = -1$ ) then
2:   if ( $n = 0$  or  $W = 0$ ) then
3:      $A[n, W] = 0$ 
4:   else
5:      $a = \text{MBK}(n - 1, W - w_n) + v_n$ ;  $b = \text{MBK}(n - 1, W)$ ;
6:     if  $a \geq b$  then
7:        $A[n, W] = a$ ;  $B[n, W] = 1$ ;
8:     else
9:        $A[n, W] = b$ ;
10:    end if
11:  end if
12: end if
13: return  $A[n, W]$ 
```

Analysis

- ▶ MBK runs in $O(nW)$ time
- ▶ Asymptotically as fast as IBK

The most precious metal may not feature in the optimum solution

- ▶ Suppose $W = 50$, $w_1 = 10$, $w_2 = 20$, $w_3 = 30$, $v_1 = 60$, $v_2 = 100$, $v_3 = 120$
- ▶ Items 1, 2 and 3 cost 6, 5, 4 respectively per unit weight
- ▶ Item 1 is the most precious metal
- ▶ But $\{2, 3\}$ is the optimal solution