Algorithms for Programming Contests - Week 08

Prof. Dr. Javier Esparza
Pranav Ashok, A. R. Balasubramanian,
Tobias Meggendorfer, Philipp Meyer,
Mikhail Raskin,
conpra@in.tum.de

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Fibonacci Numbers

fib(0) = 1

Definition (Fibonacci Numbers)

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fib(1) = 1

fib(n) = fib(n-1) + fib(n-2)

procedure FIB(n)
```

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if n \le 2 then return 1

else

return FIB(n-1) + FIB(n-2)
```

Fibonacci Numbers

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fib(0) = 1

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```
procedure \operatorname{FiB}(n)

if n \leq 2 then return 1

if DP[n] \neq 0 then

DP[n] \leftarrow \operatorname{FiB}(n-1) + \operatorname{FiB}(n-2)

return DP[n]
```

What is the minimum number of coins to make 40 cents?



What are the subproblems?

What is the minimum number of coins to make 40 cents?



What are the subproblems? Add 1 coin to the solutions for 40 - 25¢, 40 - 20¢, 40 - 1¢

What is the minimum number of coins to make 40 cents?



What are the subproblems? Add 1 coin to the solutions for 40 - 25¢, 40 - 20¢, 40 - 1¢

Let dp[i] = "What is the least amount of coins I need to make $i \notin$?"

What is the minimum number of coins to make 40 cents?



What are the subproblems? Add 1 coin to the solutions for 40 - 25¢, 40 - 20¢, 40 - 1¢

Let dp[i] = "What is the least amount of coins I need to make $i \Leftrightarrow$?" $dp[i] = \min(dp[i-25], dp[i-20], dp[i-1]) + 1$

General Approach

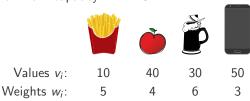
- Find recursive subproblems (smaller numbers, fewer nodes, ...)
- 2 Solve subproblem, cache solution
- 3 Assemble bigger solution

Maximum capacity W = 10



Values v_i : 10 40 30 50 Weights w_i : 5 4 6 3





Possible approach: Build 2-dimensional table dp[n][W] where dp[i][w] considers a backpack of size w < W and items $1, \ldots, i$ only.

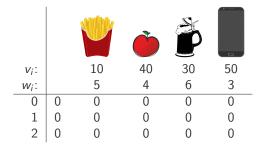
Maximum capacity W = 10



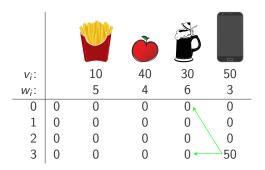
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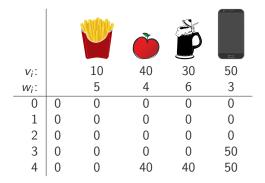
$$dp[0, w] \leftarrow 0$$
 for all $w \leq W$
 $dp[i, w] \leftarrow \max(dp[i-1, w], dp[i-1, w-w_i] + v_i)$



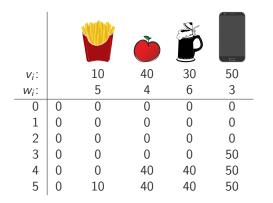
$$\max \left(\begin{array}{l} dp[i-1,w], \\ dp[i-1,w-w_i] + v_i \end{array} \right)$$



$$\max\left(\begin{array}{c}dp[i-1,w],\\dp[i-1,w-w_i]+v_i\end{array}\right)$$



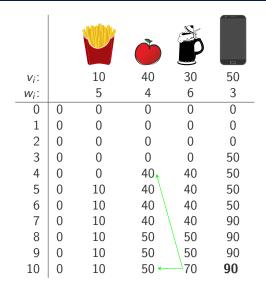
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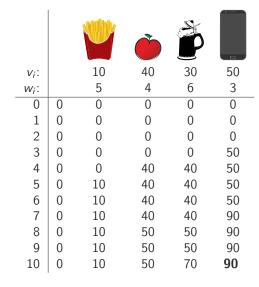
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V_i :		10	40	30	50
W_i :		5	4	6	3
0	0	0	0	0	0
1	0	0	0	0	0
2	0	0	0	0	0
3	0	0	0	0	50
4	0	0	40	40	50
5	0	10	40	40	50
6	0	10	40	40	50
7	0	10	40	40	90
8	0	10	50	50	90
9	0	10	50	50	90
10	0	10	50	70	90

$$\max\left(\begin{array}{c}dp[i-1,w],\\dp[i-1,w-w_i]+v_i\end{array}\right)$$

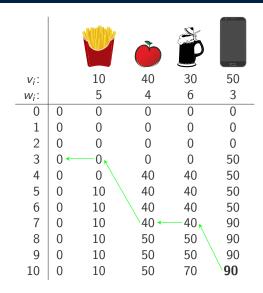


$$\max \left(\begin{array}{c} dp[i-1,w], \\ dp[i-1,w-w_i] + v_i \end{array} \right)$$



$$\max \left(\begin{array}{l} dp[i-1,w], \\ dp[i-1,w-w_i] + v_i \end{array} \right)$$

How to compute the solution?



$$\max \left(\begin{array}{l} dp[i-1,w], \\ dp[i-1,w-w_i] + v_i \end{array}
ight)$$

How to compute the solution? predecessor array storing incoming edge

Top-Down vs Bottom-Up

Top-Down - Memoization

- Recursive computation
- Save results as they appear (HashMap)

Straight-forward to implement
Only computing relevant subproblems
Good for sparse statespace

Bottom-Up

- Fill table for all smaller subproblems first
- Save results in array

Better cache locality Good for dense statespace Top-Down vs. Bottom-Up

More Examples

- Floyd-Warshall: All Pairs Shortest Paths
- Dijkstra: Single Source Shortest Path
- Longest Common Subsequence of two strings
- Edit Distance of two strings

☐ Top-Down vs. Bottom-Up

Longest Increasing Subsequence

Problem

Given a sequence of numbers

0, 8, 4, 12, 2, 10, 6, 14, 1, 9, 5, 13, 3, 11, 7, 15

What is the length of the longest increasing subsequence?

Problem

Given a sequence of numbers

What is the length of the longest increasing subsequence?

Answer

6. But how? What are the subproblems?

Problem

Given a sequence of numbers

What is the length of the longest increasing subsequence?

Answer

6. Compute the solution for shorter sequences and build up.

s[i] := "What is the length of the longest increasing subsequence with length s[i] that ends with the value v[i]."

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$$v[i] \mid 0, 8, 4, 12, 2, 10, 6, 14, 1, 9 \\ s[i] \mid 1 2$$

Observation

Dynamic Programming
 ☐ Top-Down vs. Bottom-Up

Longest Increasing Subsequence

s[i] := "What is the length of the longest increasing subsequence with length s[i] that ends with the value v[i]."

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Top-Down vs. Bottom-Up

Longest Increasing Subsequence

s[i] := "What is the length of the longest increasing subsequence with length s[i] that ends with the value v[i]."

Observation

If there is a longest increasing subsequence at a smaller index i and v[i] < v[j], then there is a sequence of length s[i] + 1 at index j. $\Rightarrow s[j]$ is one longer than the maximum sequence ending at values smaller than v[j].

Easy algorithm

Compute s[i] by looking at all smaller s[i].

Top-Down vs. Bottom-Up

Longest Increasing Subsequence

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Easy algorithm

Compute s[i] by looking at all smaller s[i]. $\mathcal{O}(n^2)$

```
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Dynamic Programming
Top-Down vs. Bottom-Up
```

s[i] := "What is the length of the longest increasing subsequence with length s[i] that ends with the value v[i]."

Keep track of representative sequences by ascending length. m[i] := "What is the index j with a minimum value v[j], s.t. there is a LIS of length i ending at v[i]."

```
1 2 3 4
8 4 12 2
                      10
                          6 14
m[i]
```

```
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Dynamic Programming

Top-Down vs. Bottom-Up
```

```
0 1 2 3 4 5 6 7 8 9
0 8 4 12 2 10 6 14 1 9
m[i]
   parent[0] = -1
   maxlength = 0
   for i in [0..n-1]:
     # Binary Search for largest j, $s.t.
     # v[m[j]] < v[i] and j < i
     j = search()
     parent[i] = m[j]
     m[j + 1] = i
     if j + 1 > maxlength:
       maxlength = j + 1
   return maxlength
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m[i] := "What is the index j with a minimum value v[j], s.t. there is a LIS of length i ending at v[i]."
```

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