

Dynamic Programming

Knapsack problem

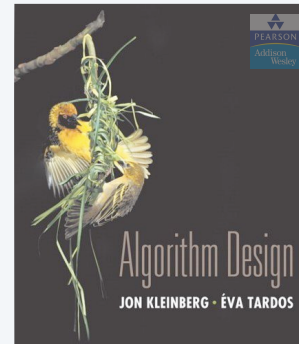
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Some slides created by or adapted from Dr. Kevin Wayne. For more information see

<http://www.cs.princeton.edu/~wayne/kleinberg-tardos>. Some code reused from [Python Algorithms](#) by Magnus Lie

Hetland.



SECTION 6.4

6. DYNAMIC PROGRAMMING I

- ▶ *weighted interval scheduling*
- ▶ *segmented least squares*
- ▶ *knapsack problem*
- ▶ *RNA secondary structure*

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Knapsack problem

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

Ex. $\{1, 2, 5\}$ has value 35.

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

Ex. $\{3, 5\}$ has value 46 (but exceeds weight limit).

i	v_i	w_i
1	1	1
2	6	2
3	18	5
4	22	6
5	28	7

knapsack instance
(weight limit $W = 11$)

Greedy by value. Repeatedly add item with maximum v_i .

Greedy by weight. Repeatedly add item with minimum w_i .

Greedy by ratio. Repeatedly add item with maximum ratio v_i / w_i .

Observation. None of greedy algorithms is optimal.

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Dynamic programming: false start

Def. $OPT(i) = \text{max profit subset of items } 1, \dots, i$.

Case 1. OPT does not select item i .

- OPT selects best of $\{1, 2, \dots, i-1\}$.

Case 2. OPT selects item i .

- Selecting item i does not immediately imply that we will have to reject other items.
- Without knowing what other items were selected before i , we don't even know if we have enough room for i .

optimal substructure property
(proof via exchange argument)

Conclusion. Need more subproblems!

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Dynamic programming: adding a new variable

Def. $OPT(i, w)$ = max profit subset of items $1, \dots, i$ with weight limit w .

Case 1. OPT does not select item i .

- OPT selects best of $\{1, 2, \dots, i-1\}$ using weight limit w .

Case 2. OPT selects item i .

- New weight limit = $w - w_i$.
- OPT selects best of $\{1, 2, \dots, i-1\}$ using this new weight limit.

optimal substructure property
(proof via exchange argument)

$$OPT(i, w) = \begin{cases} 0 & \text{if } i = 0 \\ OPT(i-1, w) & \text{if } w_i > w \\ \max\{OPT(i-1, w), v_i + OPT(i-1, w - w_i)\} & \text{otherwise} \end{cases}$$

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Knapsack problem: bottom-up

KNAPSACK($n, W, w_1, \dots, w_n, v_1, \dots, v_n$)

FOR $w = 0$ TO W

$M[0, w] \leftarrow 0$.

FOR $i = 1$ TO n

FOR $w = 1$ TO W

IF ($w_i > w$) $M[i, w] \leftarrow M[i-1, w]$.

ELSE $M[i, w] \leftarrow \max\{M[i-1, w], v_i + M[i-1, w - w_i]\}$.

RETURN $M[n, W]$.

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Knapsack problem: bottom-up demo

i	v_i	w_i
1	1	1
2	6	2
3	18	5
4	22	6
5	28	7

$$OPT(i, w) = \begin{cases} 0 & \text{if } i = 0 \\ OPT(i-1, w) & \text{if } w_i > w \\ \max\{OPT(i-1, w), v_i + OPT(i-1, w - w_i)\} & \text{otherwise} \end{cases}$$

		weight limit w											
		0	1	2	3	4	5	6	7	8	9	10	11
subset of items $1, \dots, i$	{ }	0	0	0	0	0	0	0	0	0	0	0	0
	{ 1 }	0	1	1	1	1	1	1	1	1	1	1	1
	{ 1, 2 }	0	1	6	7	7	7	7	7	7	7	7	7
	{ 1, 2, 3 }	0	1	6	7	7	18	19	24	25	25	25	25
	{ 1, 2, 3, 4 }	0	1	6	7	7	18	22	24	28	29	29	40
	{ 1, 2, 3, 4, 5 }	0	1	6	7	7	18	22	28	29	34	34	40

$OPT(i, w)$ = max profit subset of items $1, \dots, i$ with weight limit w .

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Knapsack problem: running time

Theorem. There exists an algorithm to solve the knapsack problem with n items and maximum weight W in $\Theta(nW)$ time and $\Theta(nW)$ space.

Pf.

weights are integers
between 1 and W

- Takes $O(1)$ time per table entry.
- There are $\Theta(nW)$ table entries.
- After computing optimal values, can trace back to find solution:
take item i in $OPT(i, w)$ iff $M[i, w] < M[i-1, w]$. ■

Remarks.

- Not polynomial in input size! ← "pseudo-polynomial"
- Decision version of knapsack problem is NP-COMplete. [CHAPTER 8]
- There exists a poly-time algorithm that produces a feasible solution that has value within 1% of optimum. [SECTION 11.8]

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