Dynamic Programming: Knapsack

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Data Structures and Algorithms
Algorithmic Toolbox

Outline

1. Problem Overview
2. Knapsack with Repetitions
3. Knapsack without Repetitions
4. Final Remarks

TV commercial placement

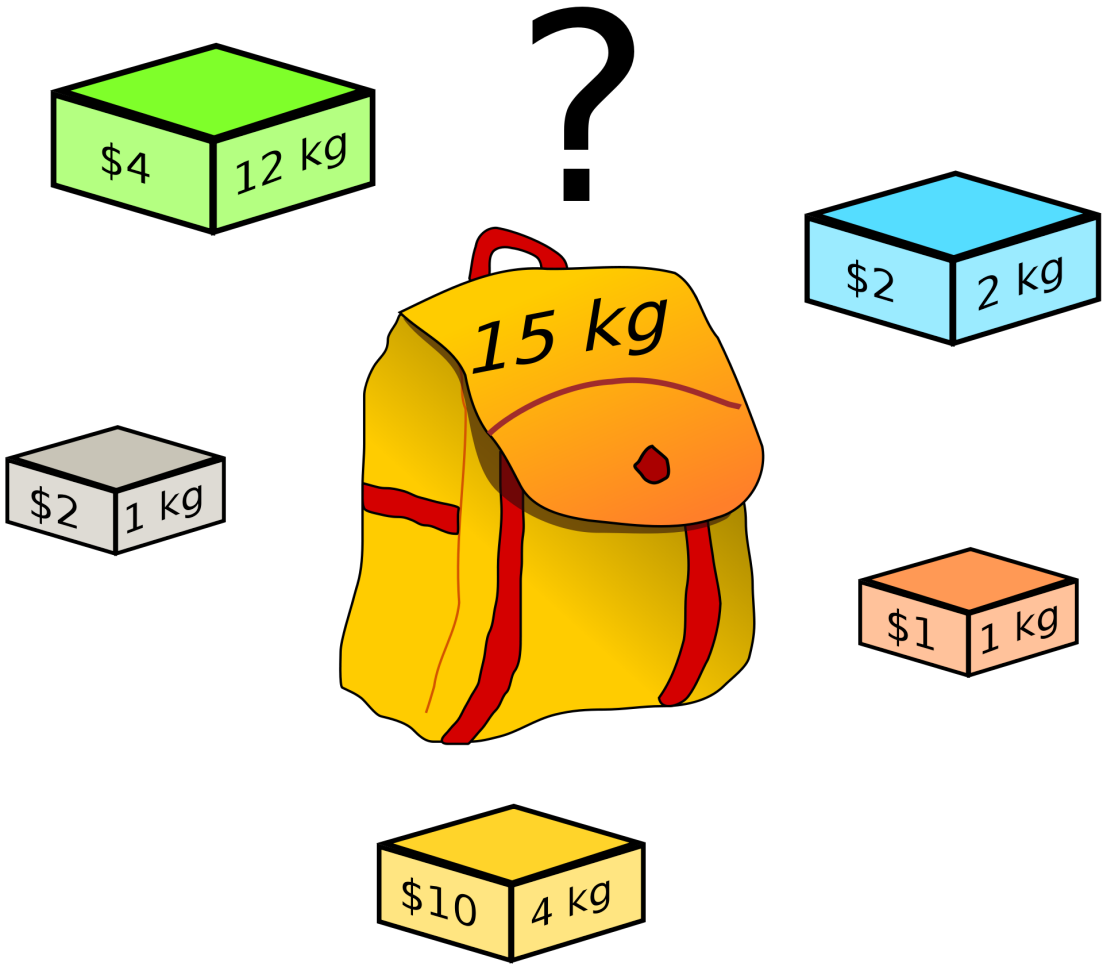
Select a set of TV commercials (each commercial has duration and cost) so that the total revenue is maximal while the total length does not exceed the length of the available time slot.

Optimizing data center performance

Purchase computers for a data center to achieve the maximal performance under limited budget.

Knapsack Problem

(knapsack is another word for backpack)



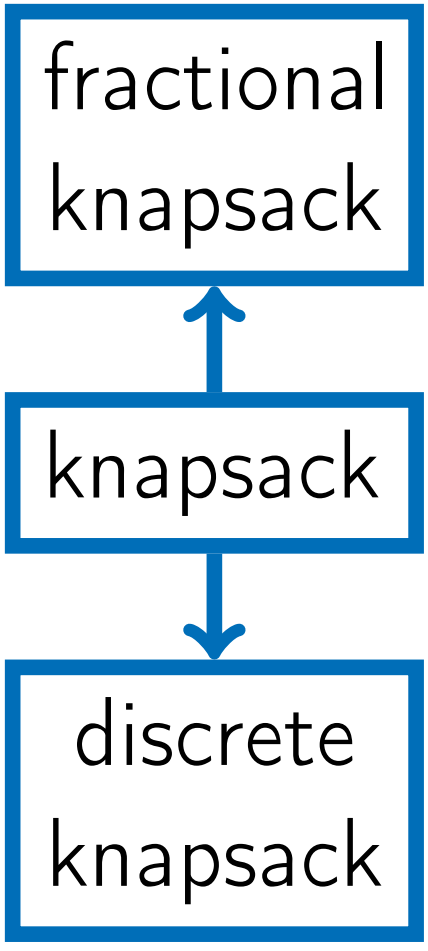
Goal

Maximize value ($) while limiting total weight (kg)

Problem Variations

knapsack

Problem Variations



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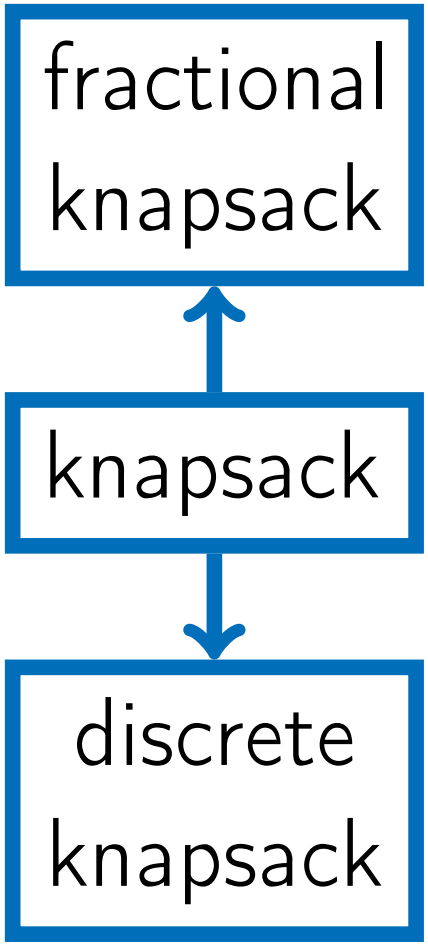
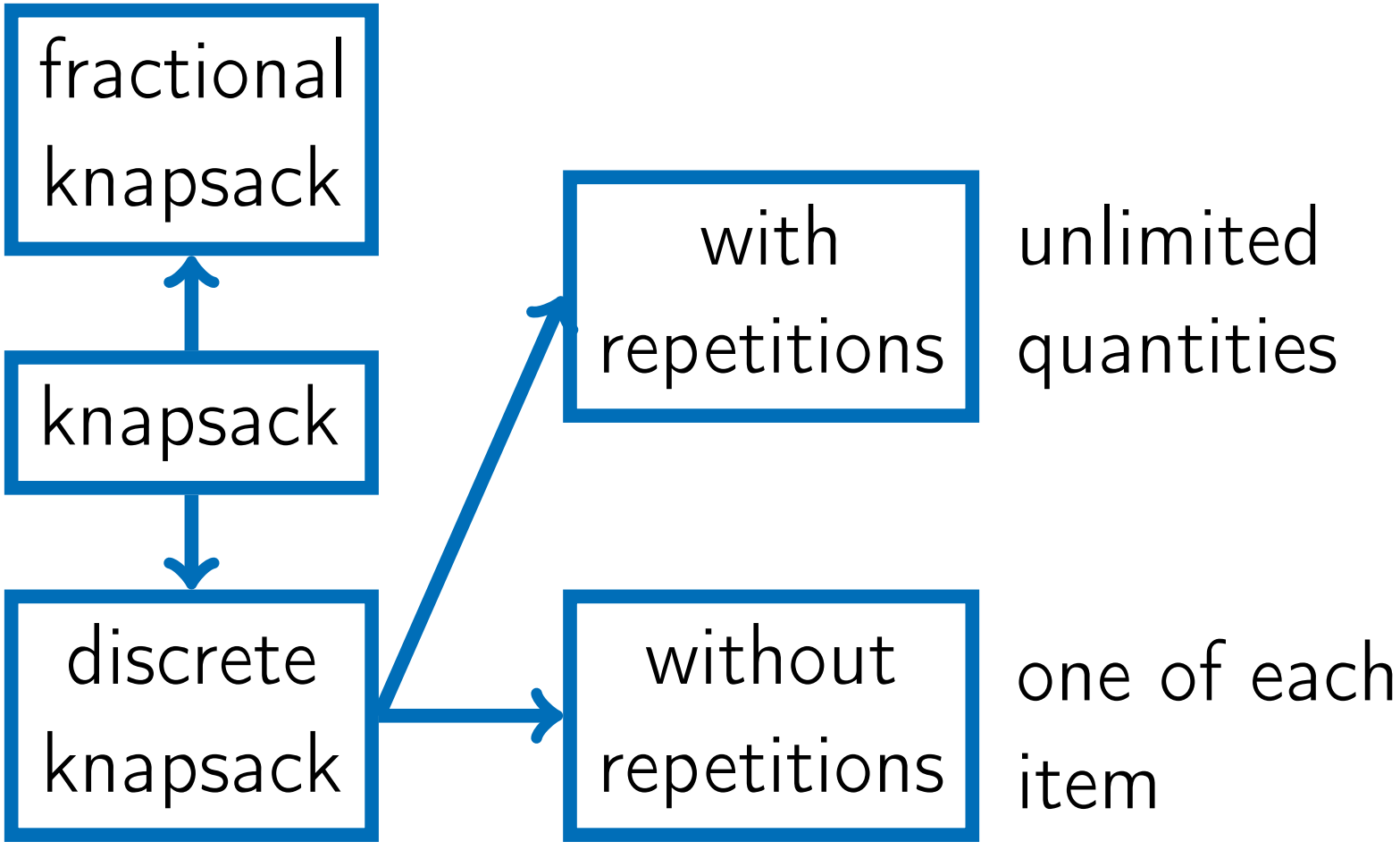


can take fractions of items

each item is either taken or not

Problem Variations



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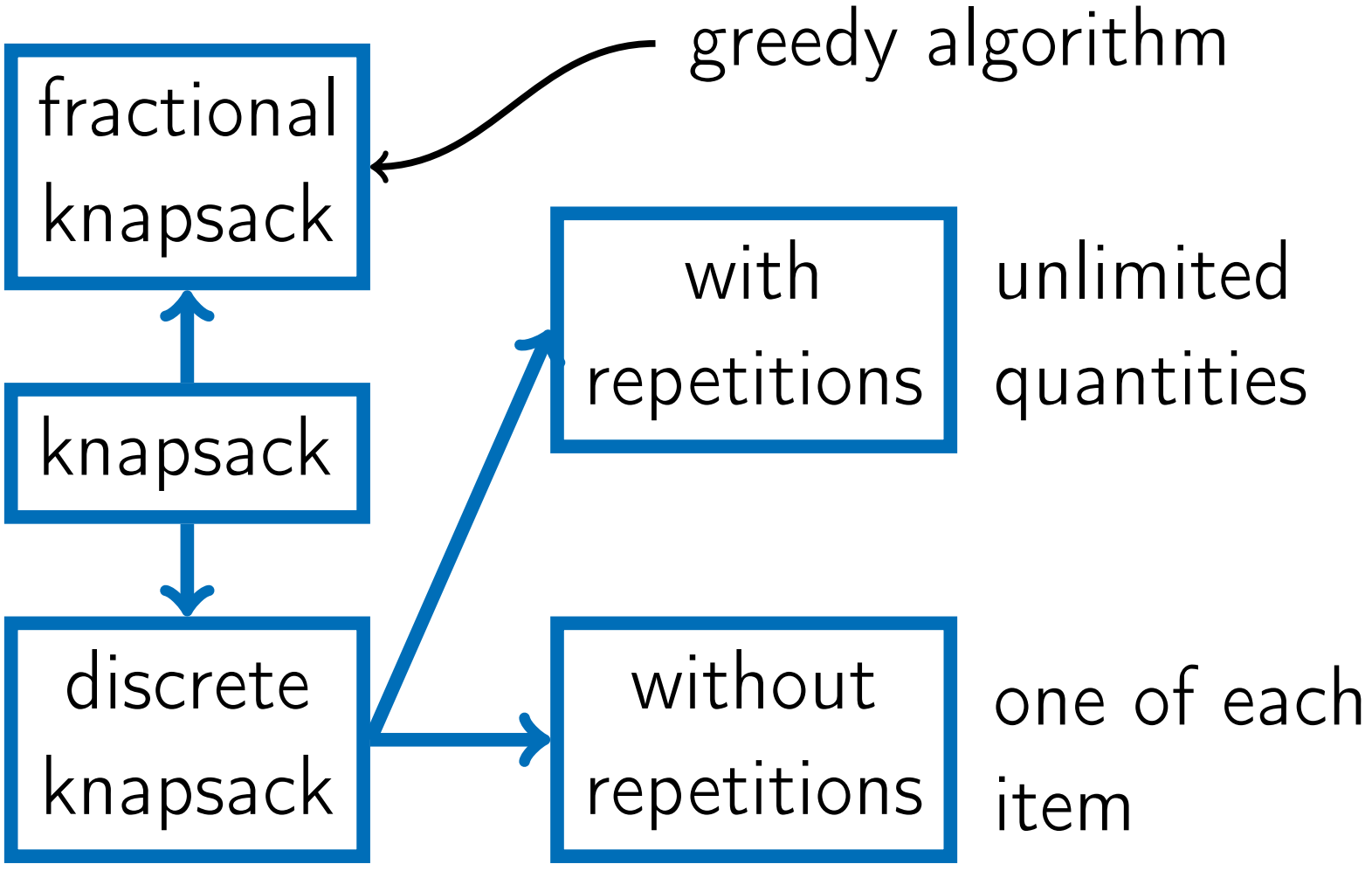

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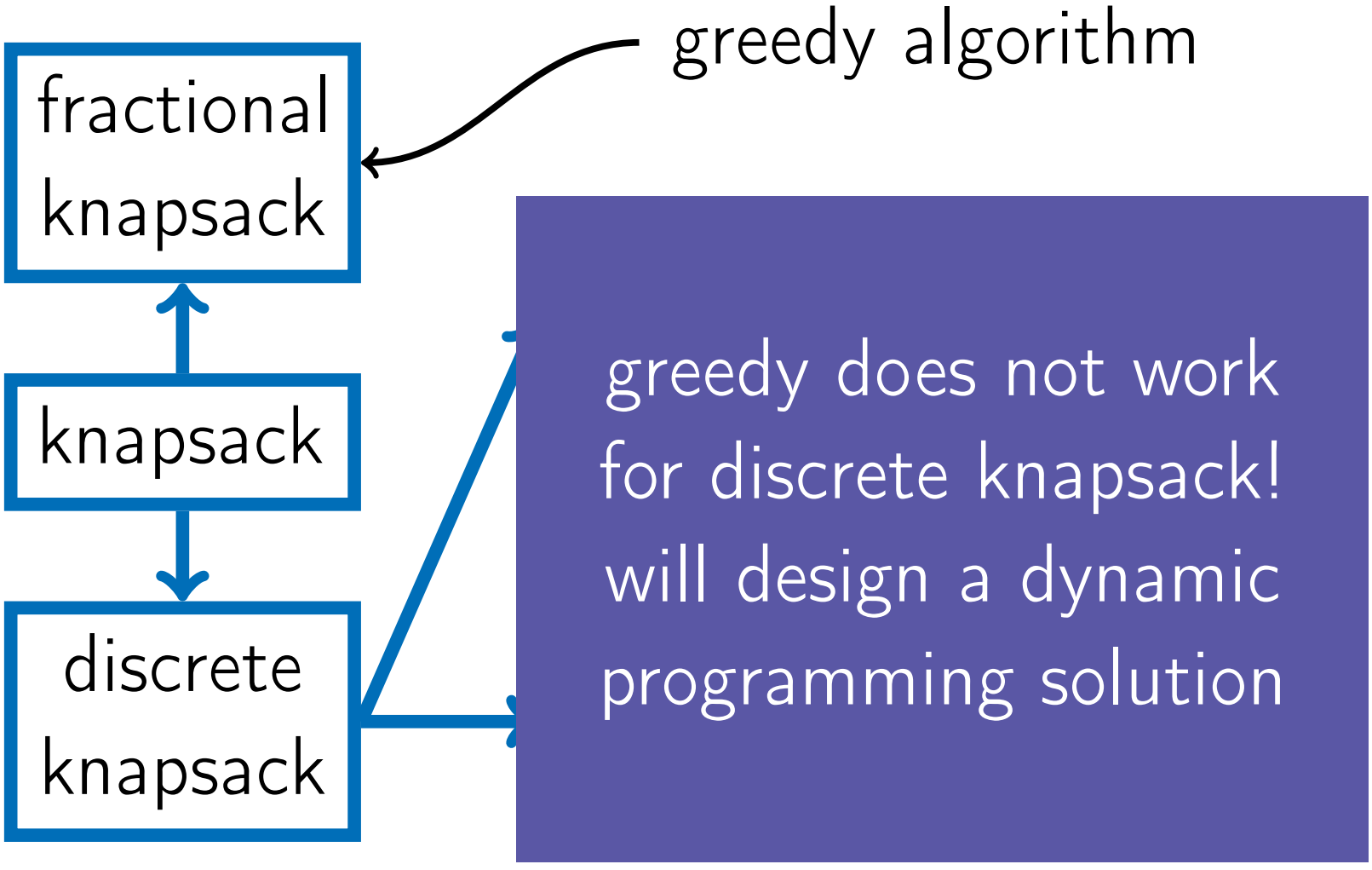
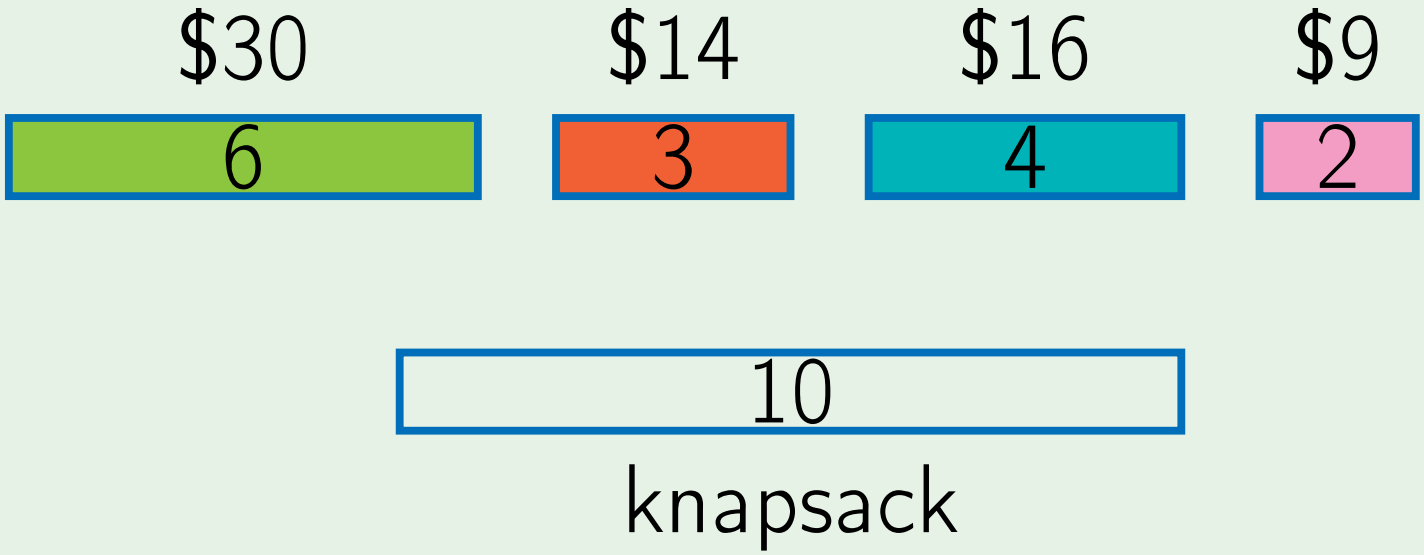
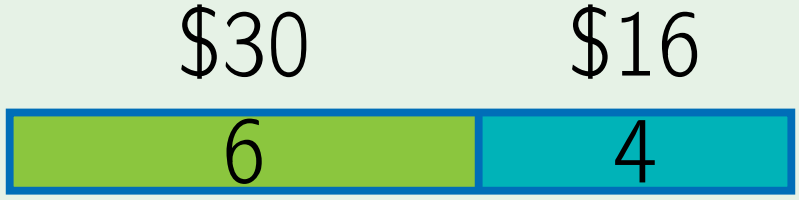


Example



Example

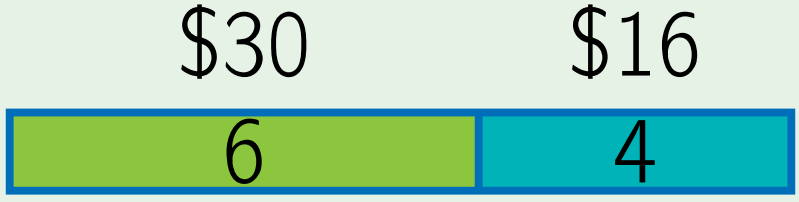
w/o repeats



total: $46

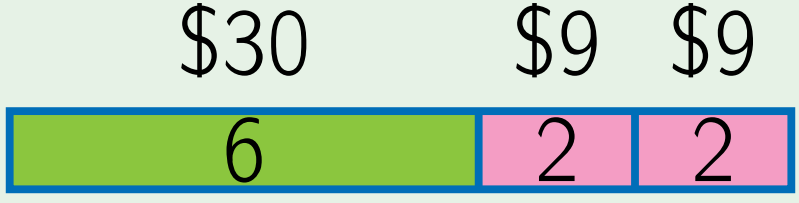
Example

w/o repeats



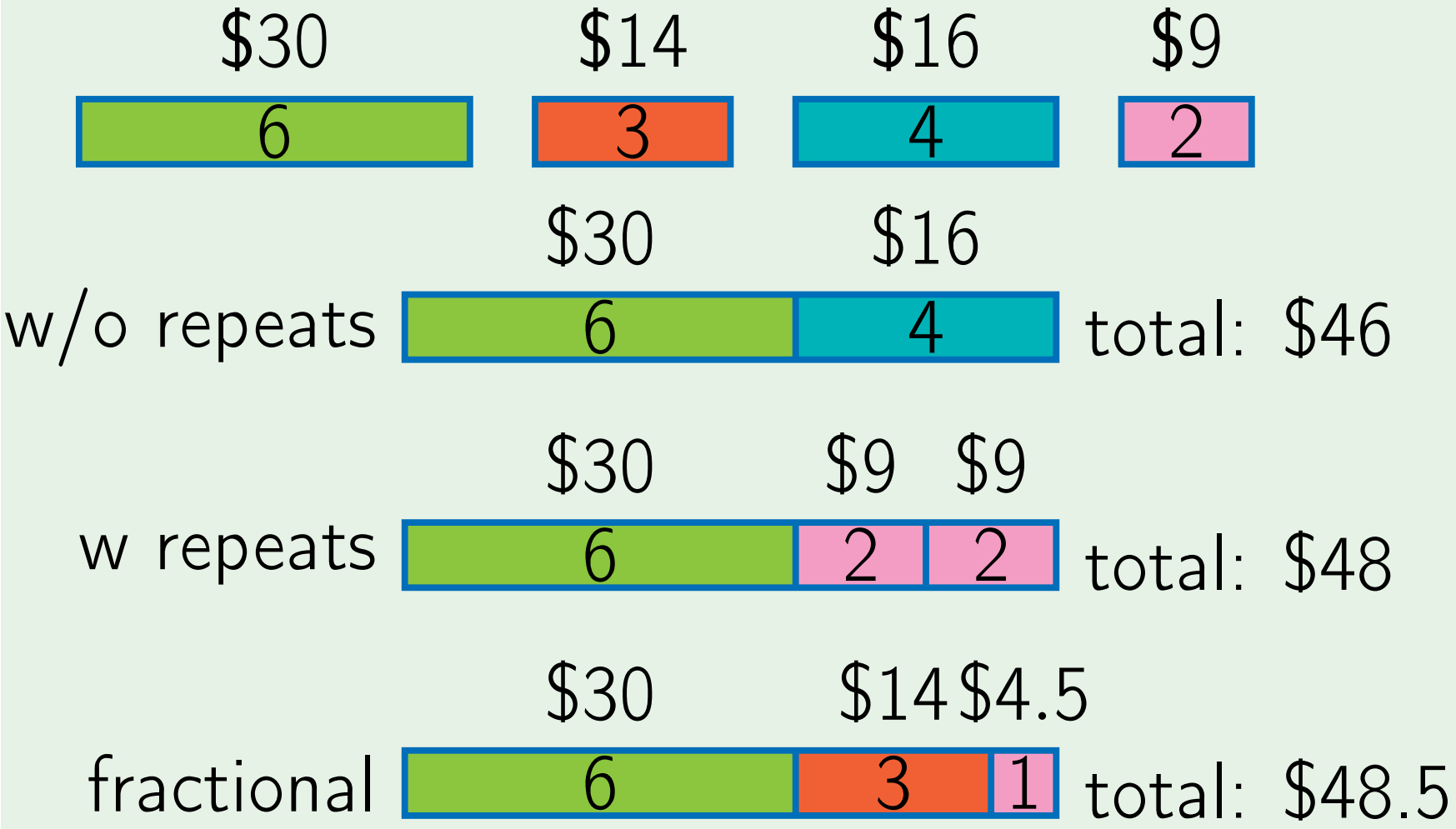
total: $46

w repeats



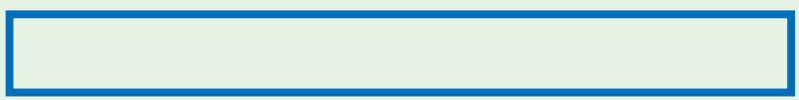
total: $48

Example


$30
$14
$16
$9
6
3
4
2
$30
$16
w/o repeats
6
4
total: $46
$30
$9
$9
w repeats
6
2
2
total: $48
$30
$14
$4.5
fractional
6
3
1
total: $48.5

Why does greedy fail for the discrete knapsack?

Example



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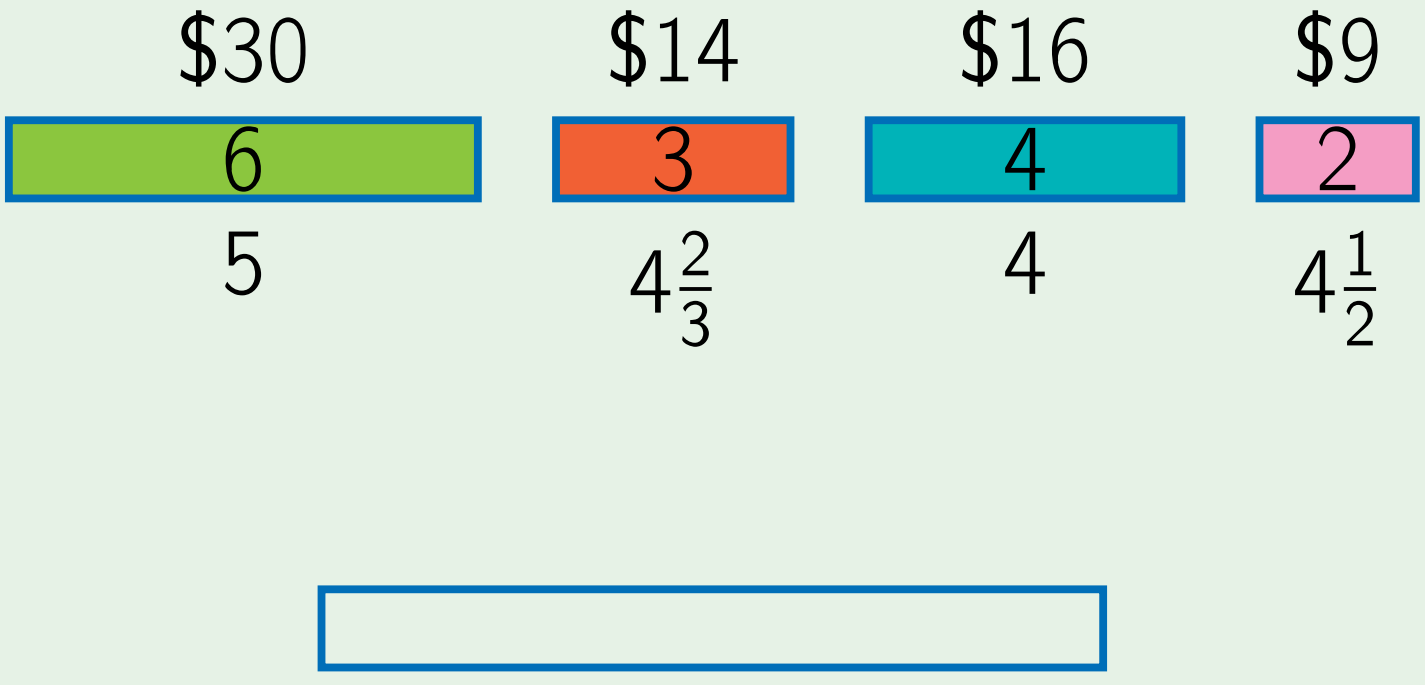

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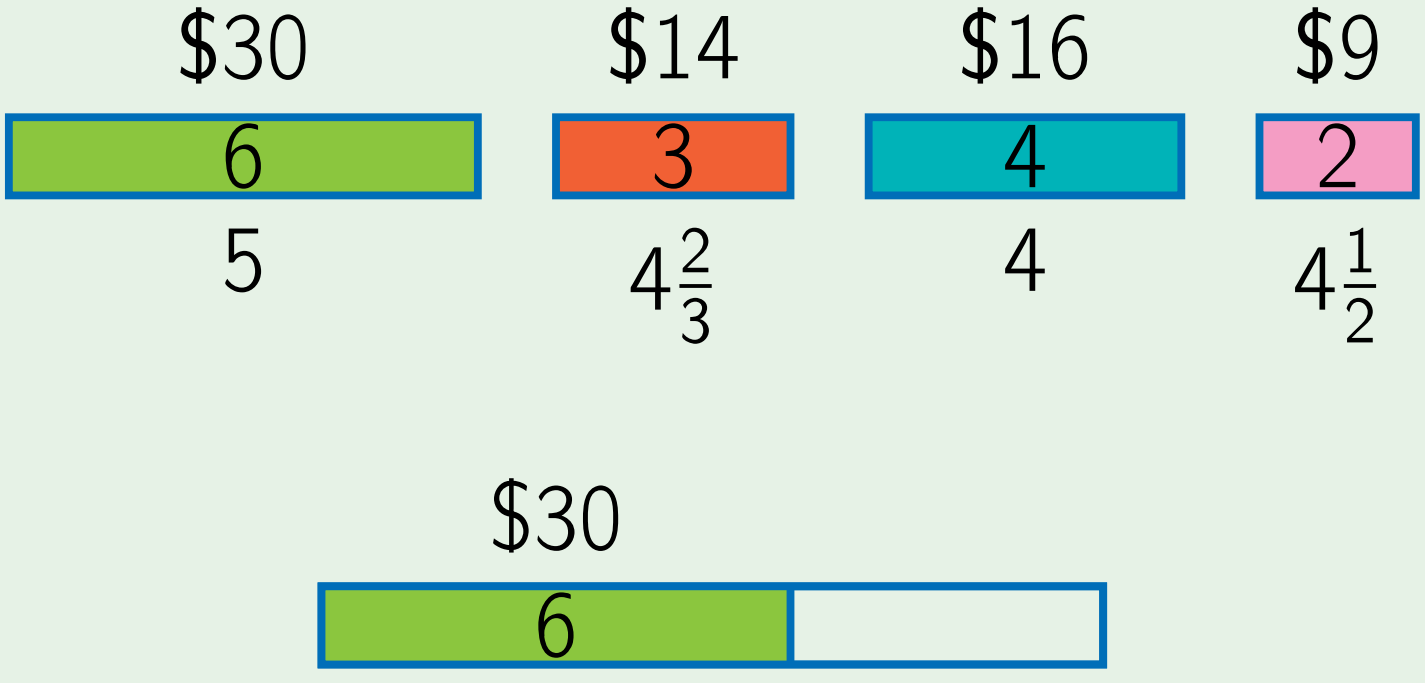

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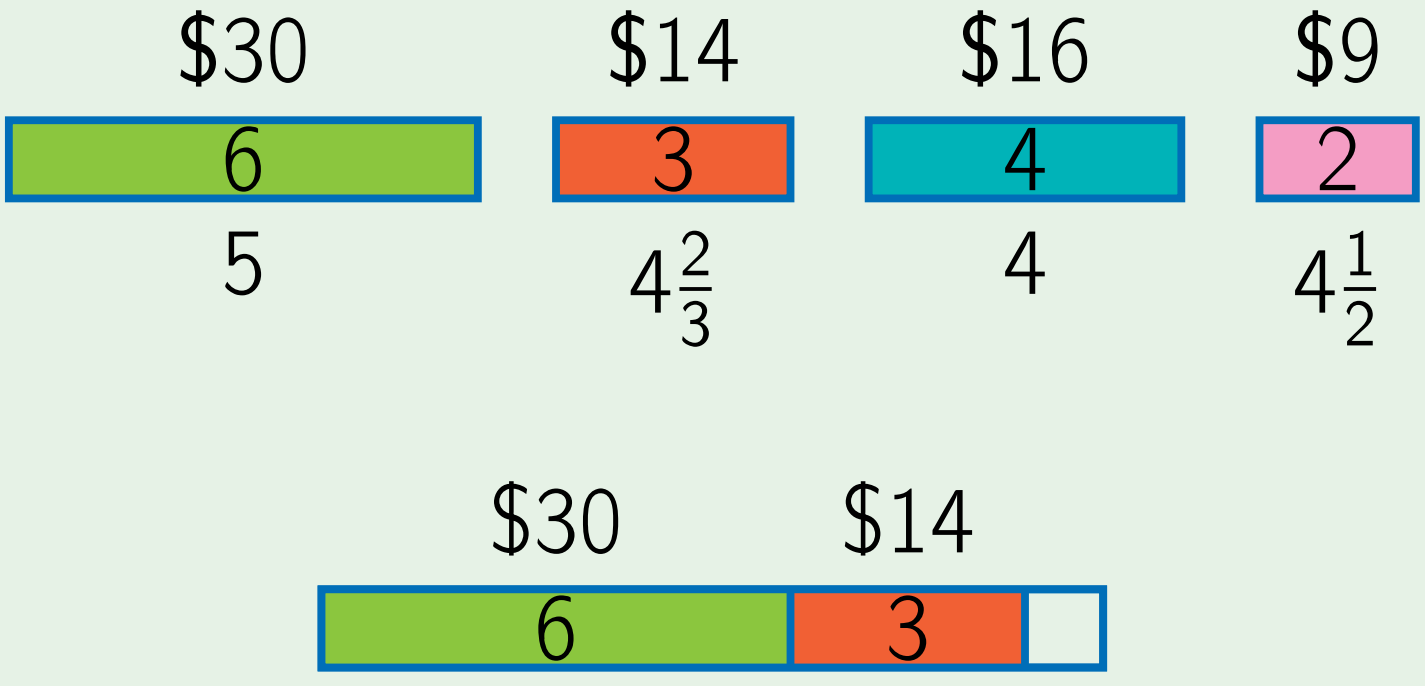

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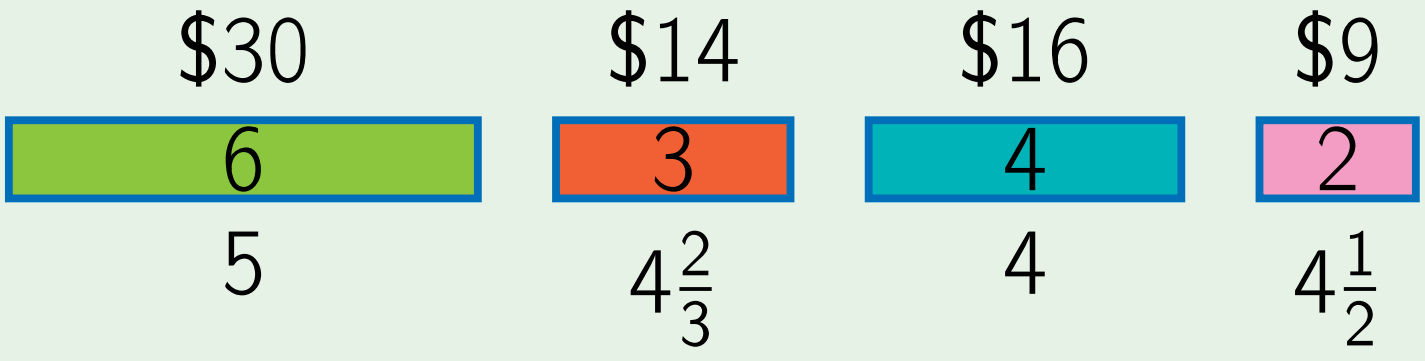

taking an element of maximum value per unit of weight is not safe!

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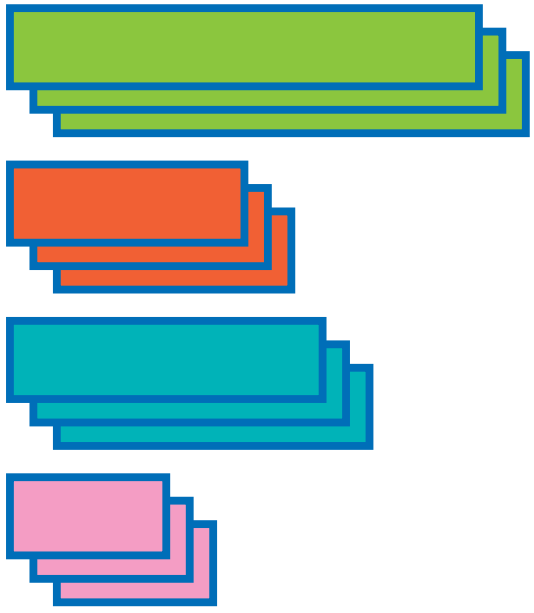
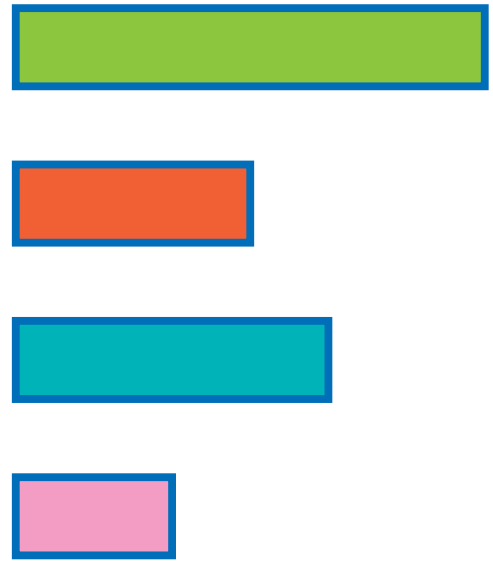
With repetitions:

unlimited quantities



Without repetitions:

one of each item



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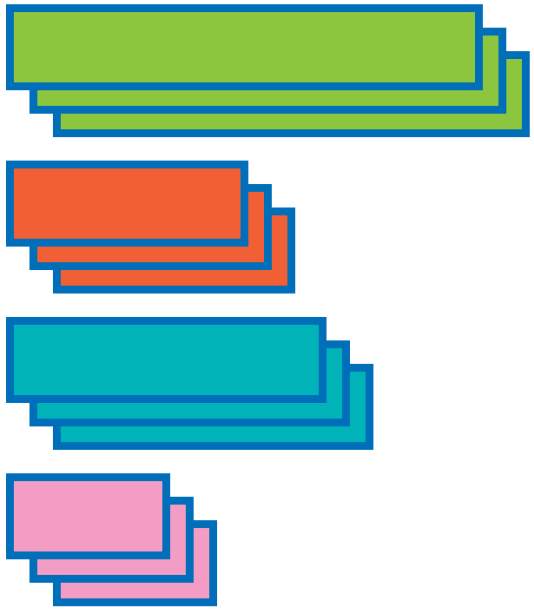

Without repetitions:

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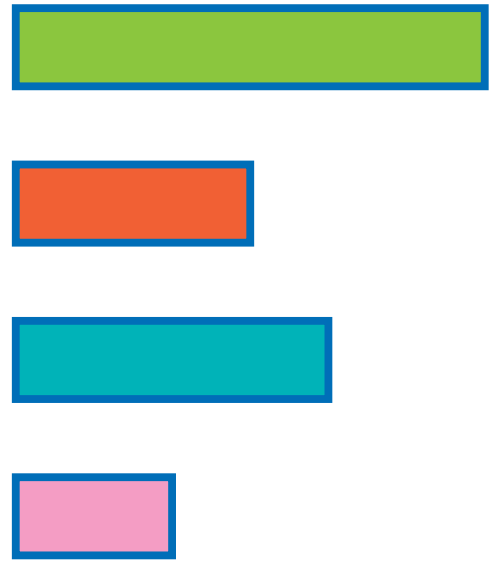

Knapsack with repetitions problem

Input: Weights $w_1, \ldots, w_n$ and values $v_1, \ldots, v_n$ of n items; total weight W (v_i's, w_i's, and W are non-negative integers).

Output: The maximum value of items whose weight does not exceed W. Each item can be used any number of times.

Subproblems

- Consider an optimal solution and an item in it:

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- Consider an optimal solution and an item in it:

- If we take this item out then we get an optimal solution for a knapsack of total weight $W - w_i$.

Subproblems

Let $value(w)$ be the maximum value of knapsack of weight w.

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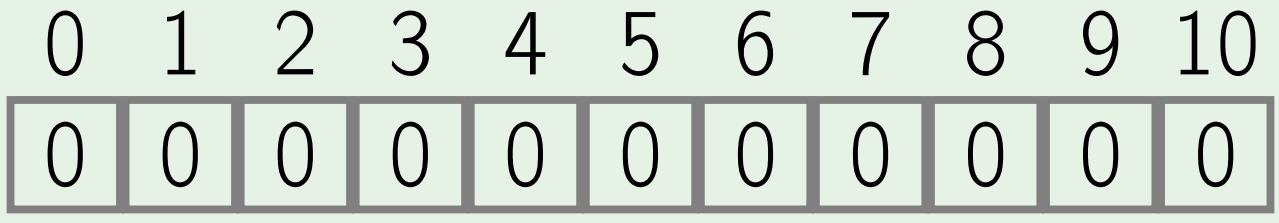
$$value(w) = \max_{i\,:\,w_i \le w} \{value(w - w_i) + v_i\}$$

Knapsack(W)

```
value(0) ← 0
for w from 1 to W:
   value(w) ← 0
   for i from 1 to n:
      if w_i ≤ w:
         val ← value(w − w_i) + v_i
         if val > value(w):
            value(w) ← val
return value(W)
```

Example: $W = 10$



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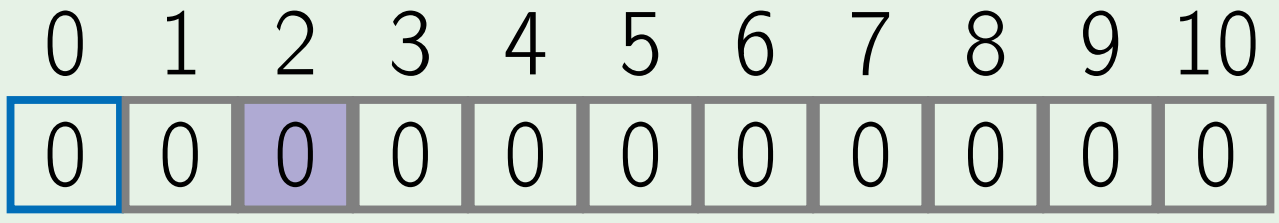

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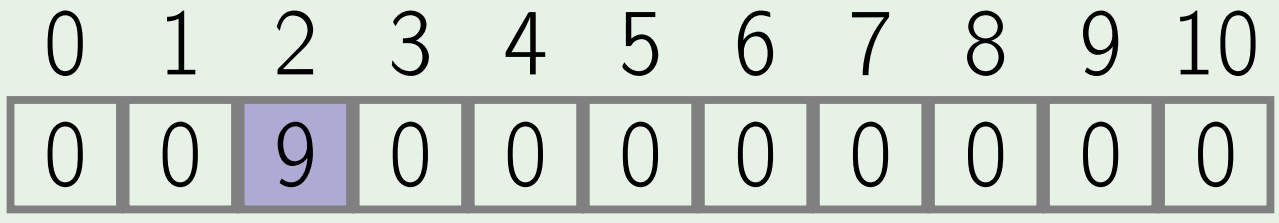

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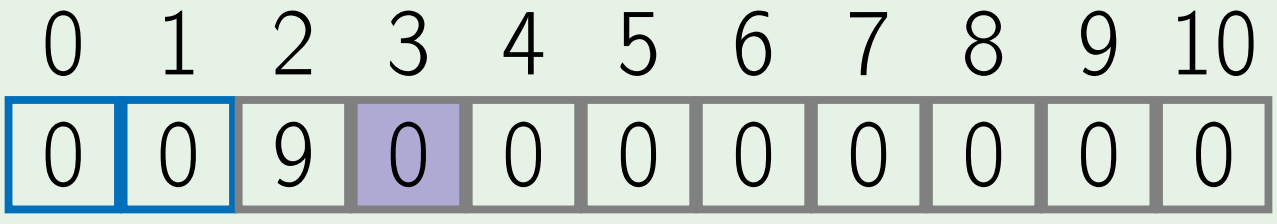

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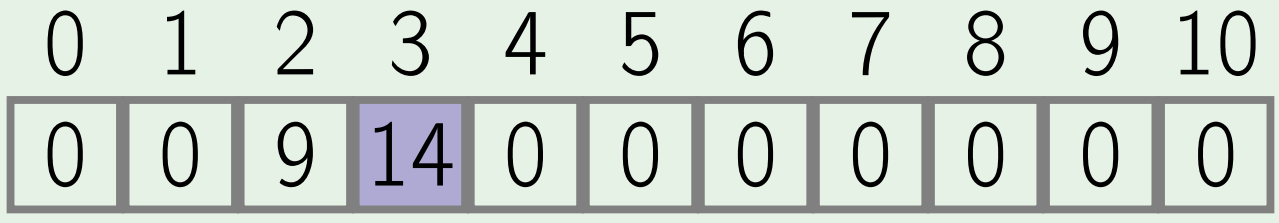

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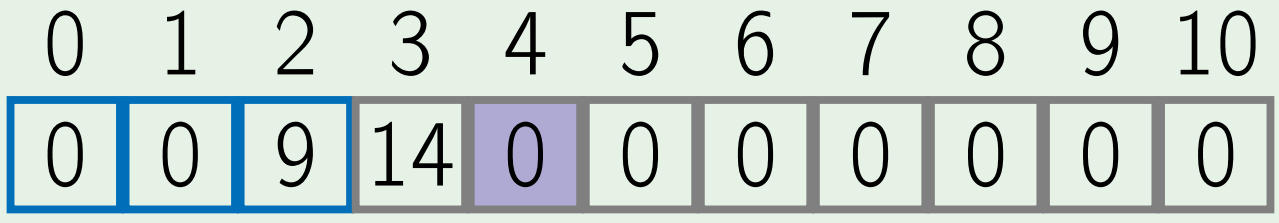

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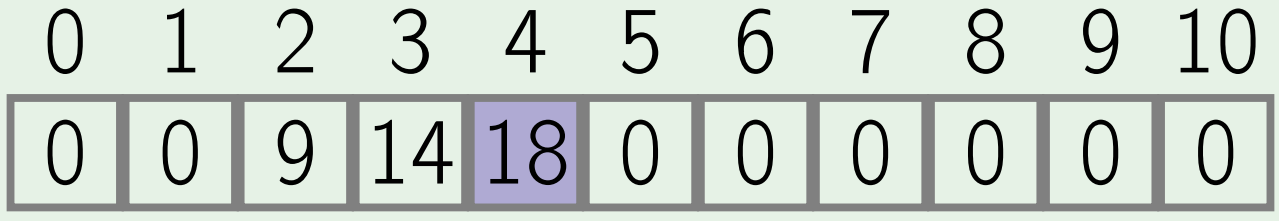

Example: $W = 10$

0	1	2	3	4	5	6	7	8	9	10
0	0	9	14	18	23	30	32	39	44	48

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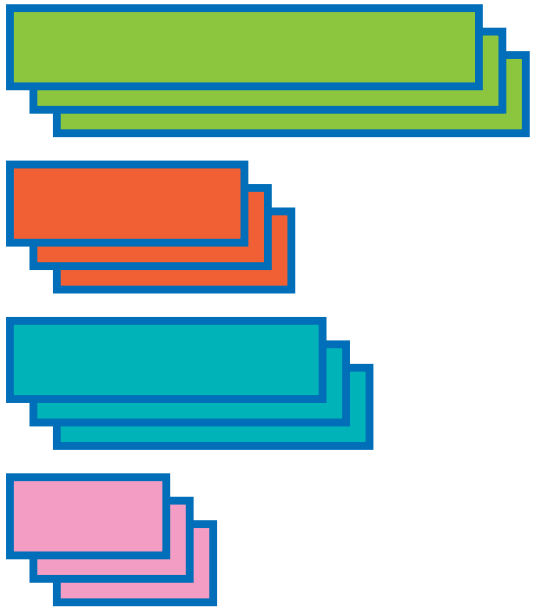
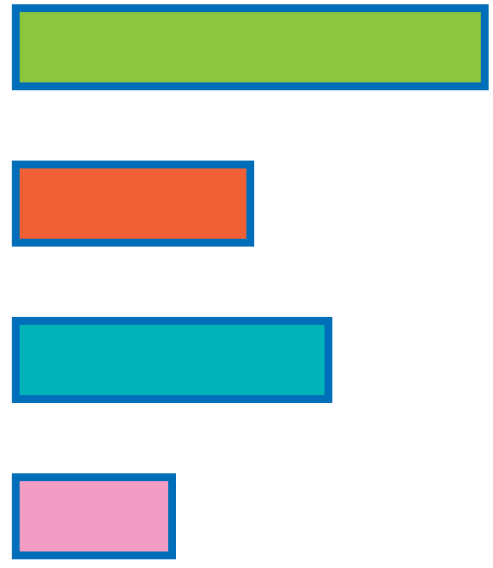


Without repetitions:

one of each item



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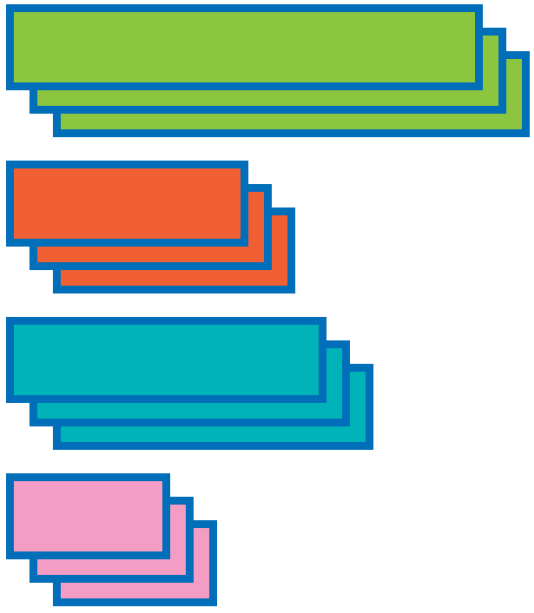

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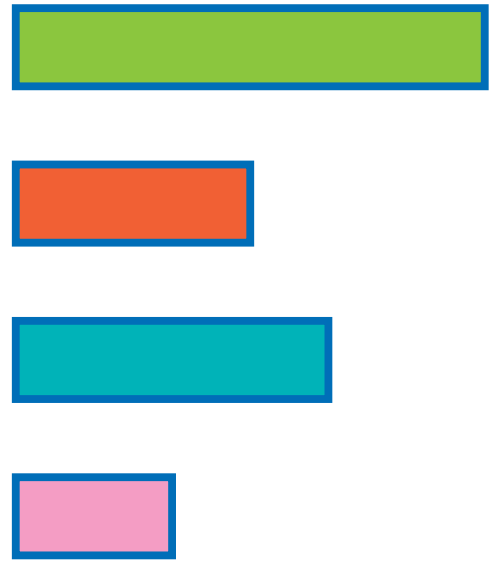

Knapsack without repetitions problem

Input: Weights $w_1, \ldots, w_n$ and values $v_1, \ldots, v_n$ of n items; total weight W (v_i's, w_i's, and W are non-negative integers).

Output: The maximum value of items whose weight does not exceed W. Each item can be used at most once.

Same Subproblems?

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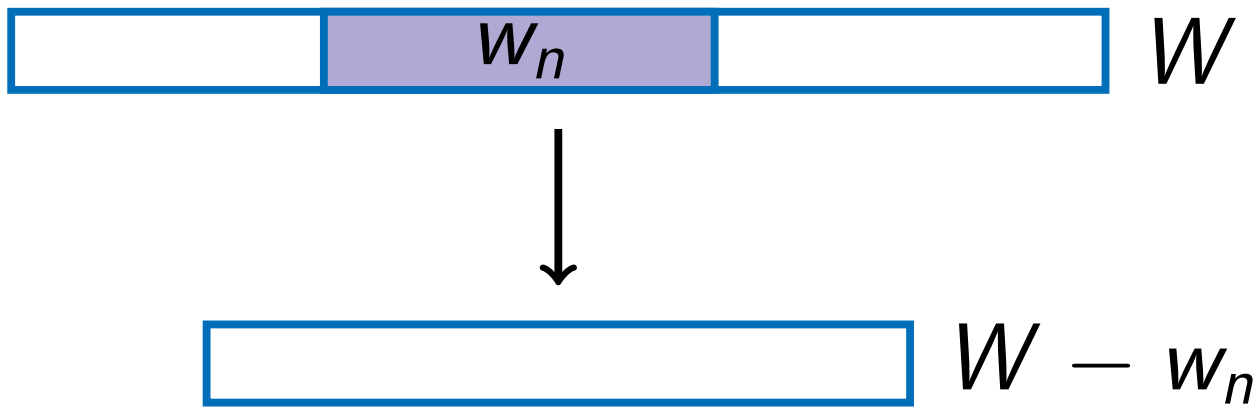

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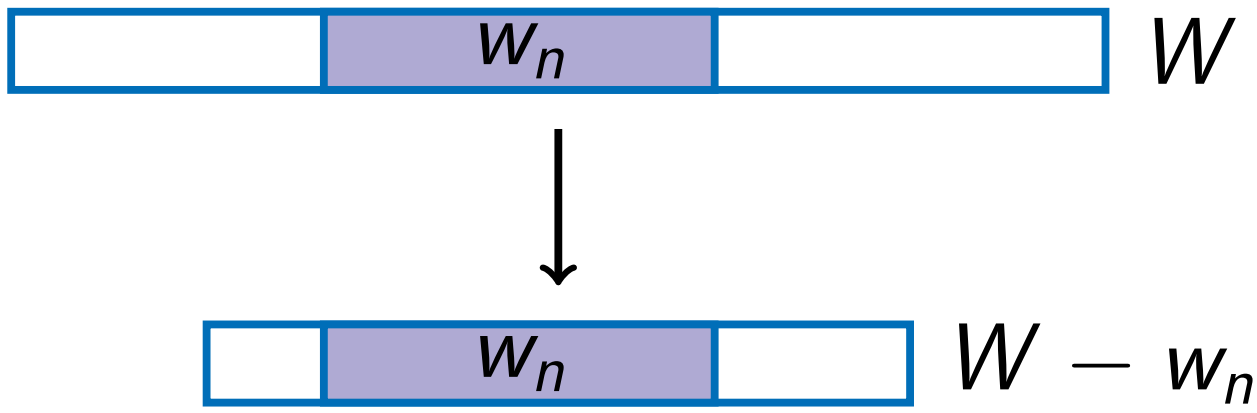

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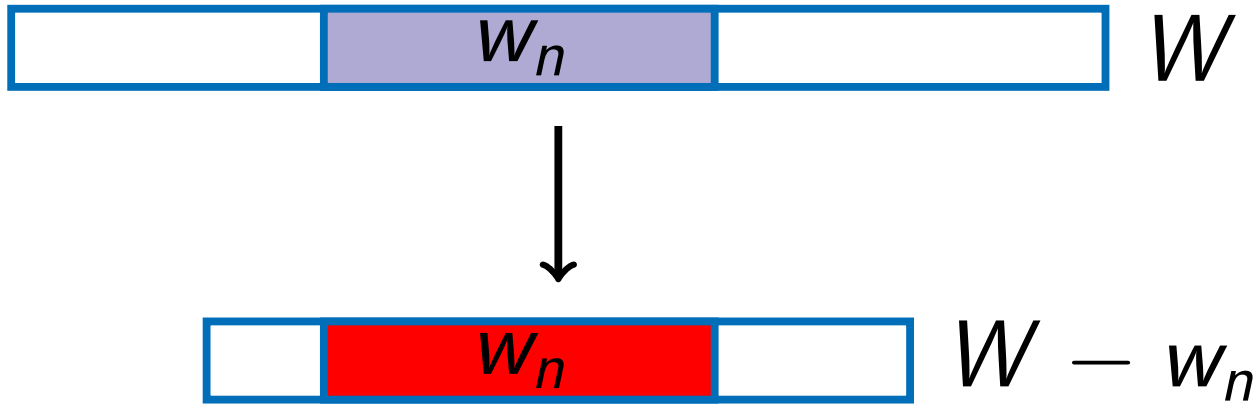

Subproblems

- If the n-th item is taken into an optimal solution:

  then what is left is an optimal solution for a knapsack of total weight $W - w_n$ using items $1, 2, \ldots, n-1$.

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- If the n-th item is taken into an optimal solution:

  then what is left is an optimal solution for a knapsack of total weight $W - w_n$ using items $1, 2, \ldots, n-1$.
- If the n-th item is not used, then the whole knapsack must be filled in optimally with items $1, 2, \ldots, n-1$.

Subproblems

For $0 \leq w \leq W$ and $0 \leq i \leq n$, $value(w, i)$ is the maximum value achievable using a knapsack of weight w and items $1, \ldots, i$.

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For $0 \leq w \leq W$ and $0 \leq i \leq n$, $value(w, i)$ is the maximum value achievable using a knapsack of weight w and items $1, \ldots, i$.

The i-th item is either used or not: $value(w, i)$ is equal to

$$\max\{value(w - w_i, i-1) + v_i, value(w, i-1)\}$$

Knapsack(W)

```
initialize all value(0, j) ← 0
initialize all value(w, 0) ← 0
for i from 1 to n:
  for w from 1 to W:
    value(w, i) ← value(w, i − 1)
    if w_i ≤ w:
      val ← value(w − w_i, i − 1) + v_i
      if value(w, i) < val
        value(w, i) ← val
return value(W, n)
```

Example: reconstructing a solution

$30	$14	$16	$9
6	3	4	2

	0	1	2	3	4	5	6	7	8	9	10
0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	30	30	30	30	30
2	0	0	0	14	14	14	30	30	30	44	44
3	0	0	0	14	16	16	30	30	30	44	46
4	0	0	9	14	16	23	30	30	39	44	46

Optimal solution:

1	2	3	4

Example: reconstructing a solution

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6	3	4	2

	0	1	2	3	4	5	6	7	8	9	10
0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	30	30	30	30	30
2	0	0	0	14	14	14	30	30	30	44	44
3	0	0	0	14	16	16	30	30	30	44	46
4	0	0	9	14	16	23	30	30	39	44	46

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1	0	0	0	0	0	0	30	30	30	30	30
2	0	0	0	14	14	14	30	30	30	44	44
3	0	0	0	14	16	16	30	30	30	44	46
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1	0	0	0	0	0	0	30	30	30	30	30
2	0	0	0	14	14	14	30	30	30	44	44
3	0	0	0	14	16	16	30	30	30	44	46
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2	0	0	0	14	14	14	30	30	30	44	44
3	0	0	0	14	16	16	30	30	30	44	46
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0	0	0	0	0	0	0	0	0	0	0	0
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2	0	0	0	14	14	14	30	30	30	44	44
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	0	1	2	3	4	5	6	7	8	9	10
0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	30	30	30	30	30
2	0	0	0	14	14	14	30	30	30	44	44
3	0	0	0	14	16	16	30	30	30	44	46
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6	3	4	2

	0	1	2	3	4	5	6	7	8	9	10
0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	30	30	30	30	30
2	0	0	0	14	14	14	30	30	30	44	44
3	0	0	0	14	16	16	30	30	30	44	46
4	0	0	9	14	16	23	30	30	39	44	46

Optimal solution:

1	2	3	4
1	0	1	0

Outline

1. Problem Overview
2. Knapsack with Repetitions
3. Knapsack without Repetitions
4. Final Remarks

Memoization

Knapsack(w)

```
if w is in hash table:
  return value(w)
value(w) ← 0
for i from 1 to n:
  if w_i ≤ w:
    val ← Knapsack(w − w_i) + v_i
    if val > value(w):
      value(w) ← val
insert value(w) into hash table with key w
return value(w)
```

What Is Faster?

- If all subproblems must be solved then an iterative algorithm is usually faster since it has no recursion overhead.

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- If all subproblems must be solved then an iterative algorithm is usually faster since it has no recursion overhead.
- There are cases however when one does not need to solve all subproblems: assume that W and all w_i's are multiples of 100; then $value(w)$ is not needed if w is not divisible by 100.

Running Time

- The running time $O(nW)$ is not polynomial since the input size is proportional to $\log W$, but not W.

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- In other words, the running time is $O(n2^{\log W})$.
- E.g., for

$$W = 71\,345\,970\,345\,617\,824\,751$$

(twenty digits only!) the algorithm needs roughly 10^{20} basic operations.

Running Time

- The running time $O(nW)$ is not p
- I
- E

later, we'll learn why
solving this problem
in polynomial time costs \$1M!

(twenty digits only!) the algorithm needs roughly 10^{20} basic operations.