

# **Algorithms and Data Structures 2**

## **19 - Advanced design techniques**

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# Outline

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- **Recap on algorithm design techniques**
- **0-1 knapsack problem**
- **Dynamic programming**
- **Memoisation**
- **Fractional knapsack problem**
- **Greedy algorithms**

# Recap

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- We have already studied **three** algorithm design techniques

## 1. Incremental approach

- INSERTION-SORT, SELECTION-SORT, HEAPSORT

## 2. Divide and conquer

- MERGE-SORT, QUICKSORT

## 3. Randomisation

- Randomised QUICKSORT, universal hashing

# Knapsack problem

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- Given **weights** and **values** of **n** items, we need to put these items in a knapsack of capacity **W** to get the **maximum** total value in the knapsack
  - Classical problem in **combinatorial optimisation**
- There are **two** versions of this problem
  1. **0-1 knapsack problem**
    - Items are **indivisible**; you either take an item or not
  2. **Fractional knapsack problem**
    - Items are **divisible**: you can take any **fraction** of an item





# 0-1 knapsack problem

- Given two  $n$ -tuples of positive integers  $v[1,..,n]$  and  $w[1,..,n]$  and  $W > 0$  determine subset  $S \subseteq \{1,..,n\}$  that maximises subject to
  - $v_i$  is the value of  $i$ -th item
  - $w_i$  is the weight of  $i$ -th item

- Example**

Knapsack with  
capacity  $W = 16$



	Items	value	weight
1		5	12
2		3	5
3		4	3
4		2	1





# 0-1 knapsack problem (cont.)

- Given two  $n$ -tuples of positive integers  $v[1,..,n]$  and  $w[1,..,n]$  and  $W > 0$  determine subset  $S \subseteq \{1,..,n\}$  that maximises subject to
  - $v_i$  is the value of  $i$ -th item
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- Example**

Knapsack with  
capacity  $W = 16$



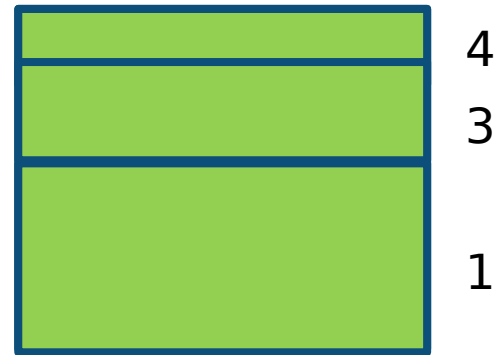
	Items	value	weight
1		5	12
2		3	5
3		4	3
4		2	1

# 0-1 knapsack problem (cont.)

- Given two  $n$ -tuples of positive integers  $v[1,..,n]$  and  $w[1,..,n]$  and  $W > 0$  determine subset  $S \subseteq \{1,..,n\}$  that maximises subject to
  - $v_i$  is the value of  $i$ -th item
  - $w_i$  is the weight of  $i$ -th item

- Example**

Knapsack with capacity  $W = 16$



Maximum value 11  
 $S = \{1,3,4\}$

Items                      value      weight

5                      12



3                      5

4                      3

2                      1

# 0-1 knapsack problem (cont.)

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- This is an **optimisation** problem
- Since there are **n** items, there are  **$2^n$**  possible combinations of items
- **Brute force algorithm**
  - Go through **all** combinations and find the one with maximum value and with total weight less or equal to  $W$
  - Running time is  **$O(2^n)$**
- **Can we do better?**



# Divide and conquer algorithm

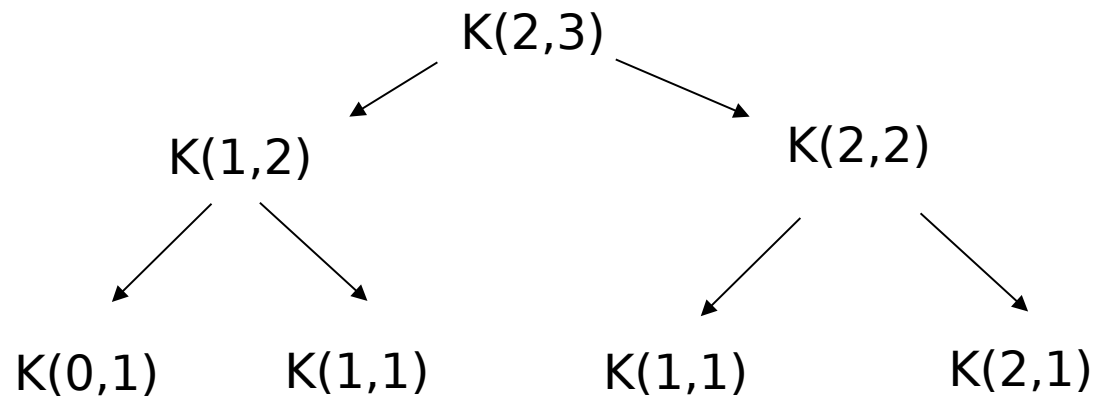
- Consider **all** subsets of items and calculate the total weight and value of each subset
    - Consider only subsets whose total weight is smaller than  **$W$**
  - From all such subsets, select the subset with **maximum** total value
    - For each item, select the **maximum** of two cases
1. An item is included in the optimal subset
    - Value of  **$n$ -th** item plus maximum value obtained by  **$n-1$**  items and  **$W$**  minus weight of  **$n$ -th** item
  2. An item is not included in the optimal set
    - Maximum value obtained by  **$n-1$**  items and  **$W$**  weight

# Pseudocode

```
KNAPSACK-REC( $W, w, v, n$ )  
  if  $n < 1$  or  $W = 0$   
    return 0  
  // weight of  $n$ -th item is more than  $W$   
  if  $w[n] > W$   
    return KNAPSACK-REC( $W, w, v, n-1$ )  
  else  
    //  $n$ -th item included  
     $a := v[n] + \text{KNAPSACK-REC}(W - w[n], w, v, n-1)$   
    //  $n$ -th item not included  
     $b := \text{KNAPSACK-REC}(W, w, v, n-1)$   
    return MAX( $a, b$ )
```

# Recursion tree

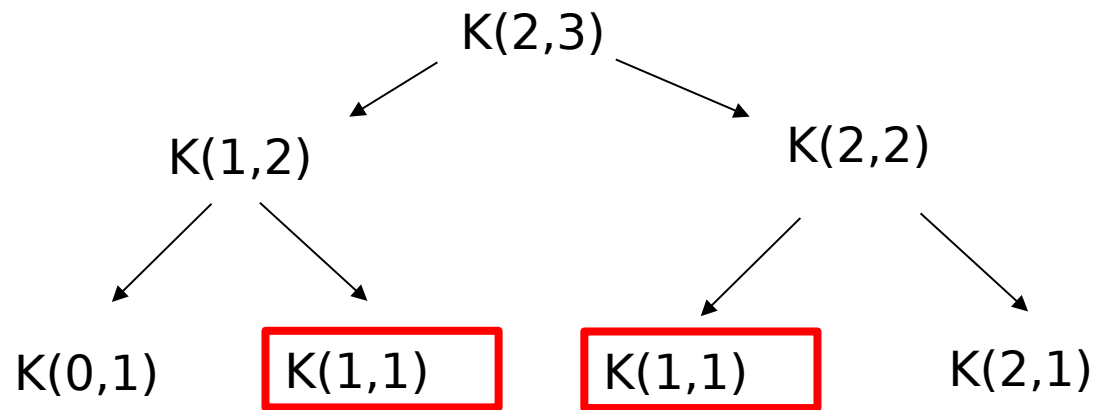
- $W = 2, n = 3, v = [5, 8, 7], w = [1, 1, 1]$
- $K(W, n) = \text{KNAPSACK-REC}(W, w, v, n)$



```
KNAPSACK-REC(W, w, v, n)
  if n < 1 or W = 0
    return 0
  // weight of n-th item is more than W
  if w[n] > W
    return KNAPSACK-REC(W, w, v, n-1)
  else
    // n-th item included
    a := v[n] + KNAPSACK-REC(W-w[n], w, v, n-1)
    // n-th item not included
    b := KNAPSACK-REC(W, w, v, n-1)
    return MAX(a, b)
```

# Recursion tree

- $W = 2, n = 3, v = [5, 8, 7], w = [1, 1, 1]$
- $K(W, n) = \text{KNAPSACK-REC}(W, w, v, n)$



Subproblems are not independent

```
KNAPSACK-REC(W, w, v, n)
  if n < 1 or W = 0
    return 0
  // weight of n-th item is more than W
  if w[n] > W
    return KNAPSACK-REC(W, w, v, n-1)
  else
    // n-th item included
    a := v[n] + KNAPSACK-REC(W-w[n], w, v, n-1)
    // n-th item not included
    b := KNAPSACK-REC(W, w, v, n-1)
    return MAX(a, b)
```

# Overlapping subproblems

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- When subproblems **overlap**, then a divide and conquer algorithm repeatedly solves the common sub-subproblems
  - More work than necessary!
- Time complexity of this naive recursive solution is exponential ( **$O(2^n)$** )
- We will study how to speed this up later in this lecture

# Dynamic programming

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- **R. Bellman began the systematic study of dynamic programming in 1955**
  - The word “programming” refers to using a tabular solution method
- **Applies to optimisation problems in which we make a set of choices in order to arrive at an optimal solution**
- **Effective when a given subproblem may arise from more than one partial set of choices (overlapping subproblems)**
  - Store the solution to each subproblem in a table so they can be reused repeatedly later
  - We trade space for time

# Dynamic programming (cont.)

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- A dynamic programming (**DP**) algorithm consists of a sequence of **four** steps
  - 1. Structure**
    - Characterise the structure of an optimal solution
  - 2. Principle of optimality**
    - Recursively define the value of an optimal solution
  - 3. Bottom-up computation**
    - Compute the value of an optimal solution using a table
  - 4. Construction of an optimal solution**
    - Use computed information to construct an optimal solution

# Dynamic programming (cont.)

- A dynamic programming (**DP**) algorithm consists of a sequence of **four** steps

## 1. **Structure**

- Characterise the structure of an optimal solution

## 2. **Principle of optimality**

- Recursively define the value of an optimal solution

## 3. **Bottom-up computation**

- Compute the value of an optimal solution using a table

## 4. **Construction of an optimal solution**

- Use computed information to construct an optimal solution

Steps 1-3 form the basis of a dynamic programming solution to a problem

Step 4 can be omitted if only the value of an optimal solution is required



# DP algorithm for 0-1 knapsack

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## 1. Structure

- Construct a table  $V[0,..,n][0,..,W]$
- For  $1 \leq i \leq n$  and  $0 \leq j \leq W$ , entry  $V[i][j]$  stores the maximum total value of any subset of items  $\{1,..,i\}$  of combined weight at most  $j$
- Entry  $V[n][W]$  contains the maximum weight of the items that can fit into a knapsack of capacity  $W$

# DP algorithm for 0-1 knapsack

---

## 2. Principle of optimality

- Similar to recursive definition of the problem used in the divide and conquer algorithm
- Base:  $V[0][j] = 0$  with  $0 \leq j \leq W$  (set with no items)  
 $V[i][j] = -\infty$  with  $j < 0$  (illegal)
- Recursive step:  $V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$  where  $1 \leq i \leq n$  and  $0 \leq j \leq W$

# DP algorithm for 0-1 knapsack

---

## 3. Bottom-up computation

- Use **iteration** instead of recursion to compute table **V** row by row
- Use the characterisation defined in step 2

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7	weight
0									
1									
2									
3									
4									

items

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1								
2								
3								
4								

$i = 0$

Base:  $V[0][j] = 0$  for  $0 \leq j \leq 7$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2								
3								
4								

$i = 1$        $v[1] = 5$        $w[1] = 2$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$

$$\begin{aligned} V[1][2] &= \max(V[0][2], v[1] + V[0][2-w[1]]) \\ &= \max(0, 5 + V[0][2-2]) \\ &= \max(0, 5 + V[0][0]) \\ &= \max(0, 5 + 0) = 5 \end{aligned}$$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3								
4								

$$i = 2 \quad v[2] = 2 \quad w[2] = 3$$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$

$$\begin{aligned} V[2][5] &= \max(V[1][5], v[2] + V[1][5-w[2]]) \\ &= \max(5, 2 + V[1][5-3]) \\ &= \max(5, 2 + V[1][2]) \\ &= \max(5, 2 + 5) = 7 \end{aligned}$$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4								

$i = 3$        $v[3] = 1$        $w[3] = 4$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$



# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4						

$$i = 4 \quad v[4] = 4 \quad w[4] = 1$$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$

$$\begin{aligned} V[4][1] &= \max(V[3][1], v[4] + V[3][1-w[4]]) \\ &= \max(0, 4 + V[3][1-1]) \\ &= \max(0, 4 + V[3][0]) \\ &= \max(0, 4 + 0) = 4 \end{aligned}$$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5					

$$i = 4 \quad v[4] = 4 \quad w[4] = 1$$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$

$$\begin{aligned} V[4][2] &= \max(V[3][2], v[4] + V[3][2-w[4]]) \\ &= \max(5, 4 + V[3][2-1]) \\ &= \max(5, 4 + V[3][1]) \\ &= \max(5, 4 + 0) = 5 \end{aligned}$$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9				

$$i = 4 \quad v[4] = 4 \quad w[4] = 1$$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$

$$\begin{aligned} V[4][3] &= \max(V[3][3], v[4] + V[3][3-w[4]]) \\ &= \max(5, 4 + V[3][3-1]) \\ &= \max(5, 4 + V[3][2]) \\ &= \max(5, 4 + 5) = 9 \end{aligned}$$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	

$$i = 4 \quad v[4] = 4 \quad w[4] = 1$$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$

$$\begin{aligned} V[4][6] &= \max(V[3][6], v[4] + V[3][6-w[4]]) \\ &= \max(7, 4 + V[3][6-1]) \\ &= \max(7, 4 + V[3][5]) \\ &= \max(7, 4 + 7) = 11 \end{aligned}$$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

$$i = 4 \quad v[4] = 4 \quad w[4] = 1$$

Recursive step:

$$V[i][j] = \max(V[i-1][j], v[i] + V[i-1][j-w[i]])$$

$$\begin{aligned} V[4][7] &= \max(V[3][7], v[4] + V[3][7-w[4]]) \\ &= \max(7, 4 + V[3][7-1]) \\ &= \max(7, 4 + V[3][6]) \\ &= \max(7, 4 + 7) = 11 \end{aligned}$$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  $W = 7$ 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

A knapsack with capacity  $W = 7$  and four items as in  $v$  and  $w$  can contain a combination of items with total value at most  $V[4][7] = 11$

Which items are in the optimal solution?

# KNAPSACK algorithm

- DP algorithm for 0-1 knapsack
- Running time is  $O(nW)$

```
KNAPSACK(W, w, v, n)
  for j = 0 to W
    V[0][j] := 0
  for i = 1 to n
    for j = 0 to W
      if w[i] ≤ j
        V[i][j] := MAX(V[i-1][j], v[i] + V[i-1][j-w[i]])
      else
        V[i][j] := V[i-1][j]
  return V[n][W]
```

# DP algorithm for 0-1Knapsack

## 4. Construction of an optimal solution

- Use table  $V$  to construct an optimal solution

```
i := n
k := W
while i > 0 and k > 0
    if V[i][k] != V[i-1][k]
        mark i as in the knapsack
        i := i - 1
        k := k - w[i]
    else
        i := i - 1
```



# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

```
i := n
k := W
while i > 0 and k > 0
  if V[i][k] != V[i-1][k]
    mark i as in the knapsack
    i := i - 1
    k := k - w[i]
  else
    i := i - 1
```

$V[4][7] \neq V[3][7]$

Mark 4

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

```
i := n
k := W
while i > 0 and k > 0
    if V[i][k] != V[i-1][k]
        mark i as in the knapsack
        i := i - 1
        k := k - w[i]
    else
        i := i - 1
```

Repeat the loop

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

```
i := n
k := W
while i > 0 and k > 0
  if V[i][k] != V[i-1][k]
    mark i as in the knapsack
    i := i - 1
    k := k - w[i]
  else
    i := i - 1
```

$V[3][6] = V[2][6]$

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

```
i := n
k := W
while i > 0 and k > 0
  if V[i][k] != V[i-1][k]
    mark i as in the knapsack
    i := i - 1
    k := k - w[i]
  else
    i := i - 1
```

Repeat the loop

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

```
i := n
k := W
while i > 0 and k > 0
    if V[i][k] != V[i-1][k]
        mark i as in the knapsack
        i := i - 1
        k := k - w[i]
    else
        i := i - 1
```

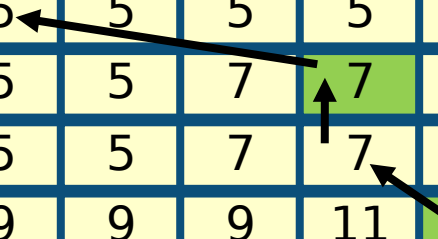
$V[2][6] \neq V[1][6]$

Mark 2

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11



```
i := n
k := W
while i > 0 and k > 0
    if V[i][k] != V[i-1][k]
        mark i as in the knapsack
        i := i - 1
        k := k - w[i]
    else
        i := i - 1
```

Repeat the loop

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

```
i := n
k := W
while i > 0 and k > 0
    if V[i][k] != V[i-1][k]
        mark i as in the knapsack
        i := i - 1
        k := k - w[i]
    else
        i := i - 1
```

$V[1][3] \neq V[0][3]$

Mark 1

# Example

- Find a combination of items that maximise the value of a knapsack with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$

	0	1	2	3	4	5	6	7
0	0	0	0	0	0	0	0	0
1	0	0	5	5	5	5	5	5
2	0	0	5	5	5	7	7	7
3	0	0	5	5	5	7	7	7
4	0	4	5	9	9	9	11	11

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    mark i as in the knapsack
    i := i - 1
    k := k - w[i]
  else
    i := i - 1
```

Termination

Optimal combination is  $\{1, 2, 4\}$



# Top-down with memoisation

---

- **Alternative way to implement a dynamic programming approach**
- **Recursive algorithms with overlapping subproblems are made more efficient by saving the result of each subproblem (usually in an array or hash table)**
  - The algorithm first checks whether it has previously solved this subproblem
  - If so, it returns the saved value, saving further computation at this level
  - If not, the procedure computes the value in the usual manner
- **We say that the recursive procedure has been memoised**
  - From “memo”
  - The algorithm remembers what results it has already computed

# Example

- **$T[0, \dots, W][0, \dots, n]$**  is initialised to **-1**

```
KNAPSACK-MEMO(W, w, v, n, T)
  if n < 1 or W = 0
    return 0
  if T[W][n] != -1
    return T[W][n]
  if w[n] > W
    T[W][n] := KNAPSACK-MEMO(W, w, v, n-1)
  else
    a := v[n] + KNAPSACK-MEMO(W-w[n], w, v, n-1)
    b := KNAPSACK-MEMO(W, w, v, n-1)
    T[W][n] := MAX(a, b)
  return T[W][n]
```

# Fractional knapsack

---

- Same as 0-1 knapsack but **fractional** amounts of each object can be included in the knapsack
- Formally, maximise  $\sum v_i f_i$  subject to  $\sum w_i f_i \leq W$  where  $0 \leq f_i \leq 1$
- **How to find a solution**
  - Sort items in non decreasing order of their value/weight **ratio**
  - When an object is considered choose a **maximal** amount such that the knapsack capacity is not violated
- **Greedy approach**

# Greedy algorithms

---

- **Applies to **optimisation** problems in which we make a set of choices in order to arrive at an optimal solution**
  - Each choice is made in a **locally optimal** manner
- **A greedy approach provides an optimal solution for many problems much more quickly than a dynamic programming approach**
- **We cannot always easily tell whether a greedy approach will be effective**
  - Only when the **greedy choice property** holds: an optimal solution can be obtained by making the greedy choice at each step

# Example

---

- Find a combination of items that maximise the value of a **fractional knapsack** with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$
  - Compute the value/weight ratios  $r = [5/2, 2/3, 1/4, 4/1] = [2.5, 0.67, 0.25, 4]$
  - Sort  $v$  and  $w$  according to  $r$ :  $v = [4, 5, 2, 1]$  and  $w = [1, 2, 3, 4]$

# Example

- Find a combination of items that maximise the value of a **fractional knapsack** with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$
  - Compute the value/weight ratios  $r = [5/2, 2/3, 1/4, 4/1] = [2.5, 0.67, 0.25, 4]$
  - Sort  $v$  and  $w$  according to  $r$ :  $v = [4, 5, 2, 1]$  and  $w = [1, 2, 3, 4]$
  - Add maximal amount of **4** to knapsack (available space  $7 - 1 = 6$ )

# Example

- Find a combination of items that maximise the value of a **fractional knapsack** with maximum capacity  **$W = 7$**

- $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$
- Compute the value/weight ratios  $r = [5/2, 2/3, 1/4, 4/1] = [2.5, 0.67, 0.25, 4]$
- Sort  $v$  and  $w$  according to  $r$ :  $v = [4, 5, 2, 1]$  and  $w = [1, 2, 3, 4]$
- Add maximal amount of **4** to knapsack (available space  $7 - 1 = 6$ )
- Add maximal amount of **5** to knapsack (available space  $6 - 2 = 4$ )

– Solution =  $4 * 1 + 5 * 1$

# Example

- Find a combination of items that maximise the value of a **fractional knapsack** with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$
  - Compute the value/weight ratios  $r = [5/2, 2/3, 1/4, 4/1] = [2.5, 0.67, 0.25, 4]$
  - Sort  $v$  and  $w$  according to  $r$ :  $v = [4, 5, 2, 1]$  and  $w = [1, 2, 3, 4]$
  - Add maximal amount of **4** to knapsack (available space  $7 - 1 = 6$ )
  - Add maximal amount of **5** to knapsack (available space  $6 - 2 = 4$ )
  - Add maximal amount of **2** to knapsack (available space  $4 - 3 = 1$ )
  - Solution =  $4 * 1 + 5 * 1 + 2 * 1$



# Example

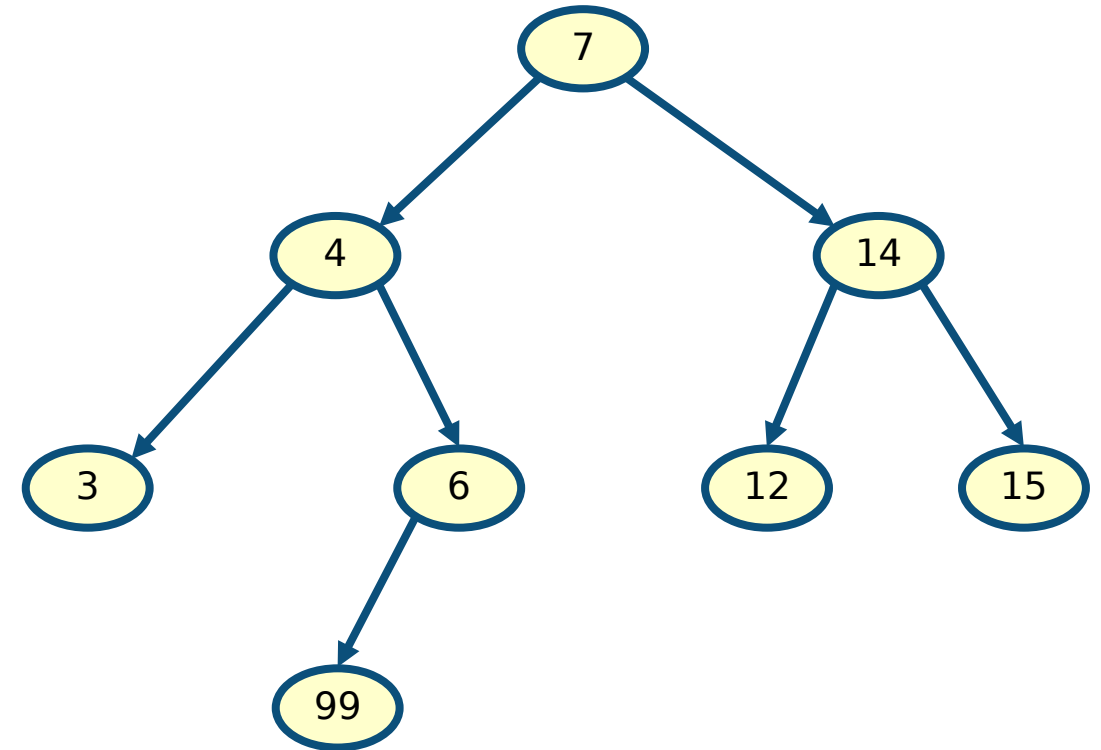
- Find a combination of items that maximise the value of a **fractional knapsack** with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$
  - Compute the value/weight ratios  $r = [5/2, 2/3, 1/4, 4/1] = [2.5, 0.67, 0.25, 4]$
  - Sort  $v$  and  $w$  according to  $r$ :  $v = [4, 5, 2, 1]$  and  $w = [1, 2, 3, 4]$
  - Add maximal amount of **4** to knapsack (available space  $7 - 1 = 6$ )
  - Add maximal amount of **5** to knapsack (available space  $6 - 2 = 4$ )
  - Add maximal amount of **2** to knapsack (available space  $4 - 3 = 1$ )
  - Add maximal amount of **1** to knapsack (available space  $1 - (4 * \frac{1}{4}) = 0$ )
  - Solution =  $4 * 1 + 5 * 1 + 2 * 1 + 1 * \frac{1}{4}$

# Example

- Find a combination of items that maximise the value of a **fractional knapsack** with maximum capacity  **$W = 7$** 
  - $v = [5, 2, 1, 4]$  and  $w = [2, 3, 4, 1]$
  - Compute the value/weight ratios  $r = [5/2, 2/3, 1/4, 4/1] = [2.5, 0.67, 0.25, 4]$
  - Sort  $v$  and  $w$  according to  $r$ :  $v = [4, 5, 2, 1]$  and  $w = [1, 2, 3, 4]$
  - Add maximal amount of **4** to knapsack (available space  $7 - 1 = 6$ )
  - Add maximal amount of **5** to knapsack (available space  $6 - 2 = 4$ )
  - Add maximal amount of **2** to knapsack (available space  $4 - 3 = 1$ )
  - Add maximal amount of **1** to knapsack (available space  $1 - (4 * \frac{1}{4}) = 0$ )
  - Solution =  $4 * 1 + 5 * 1 + 2 * 1 + 1 * \frac{1}{4} = 11.25$

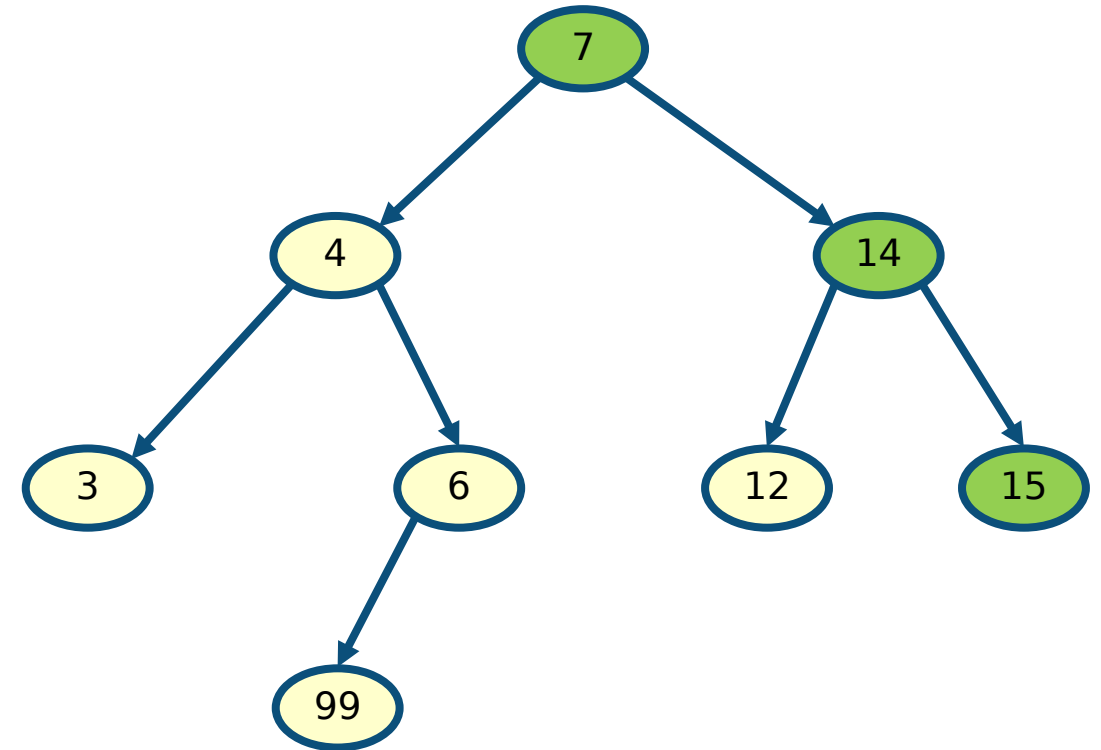
# Another example

- Find the path from the root to a leaf with the **maximum** sum in the tree on the right
- Pick the maximum at each step



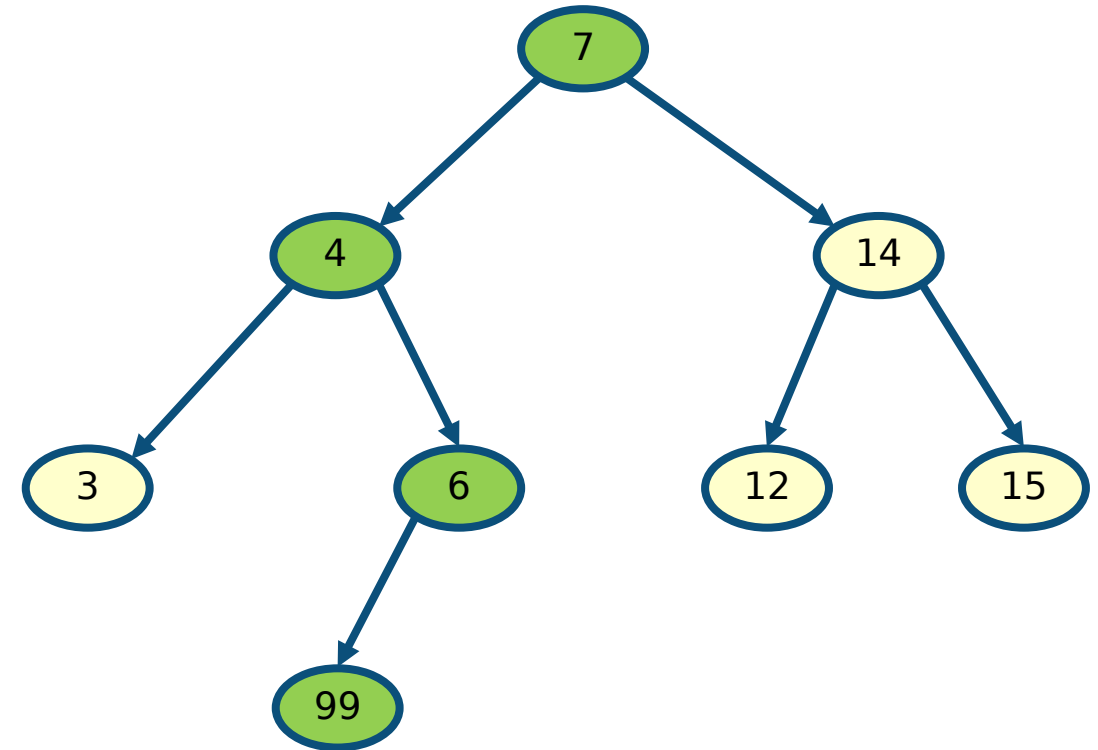
# Another example

- Find the path from the root to a leaf with the **maximum** sum in the tree on the right
- Pick the maximum at each step
  - $7 + 14 + 15 = 36$



# Another example

- Find the path from the root to a leaf with the **maximum** sum in the tree on the right
- Pick the maximum at each step
  - $7 + 14 + 15 = 36$
- The correct solution is
  - $7 + 4 + 6 + 99 = 116$
- The **greedy choice property** does not hold in this case



# Summary

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- **Recap on algorithm design techniques**
- **0-1 knapsack problem**
- **Dynamic programming**
- **Memoisation**
- **Fractional knapsack problem**
- **Greedy algorithms**