

Algorithms

Design by Induction

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Introduction

- It is not necessary to design the steps required to solve a problem from scratch
- It is sufficient to guarantee the following:
 - It is possible to solve one small instance or a few small instances of the problem (base case)
 - A solution to every problem/instance can be constructed from solutions to smaller problems/instances (inductive step)

Evaluating Polynomials

Problem

Given a sequence of real numbers $a_0, a_1, \dots, a_{n-1}, a_n$, and a real number x , compute the value of the polynomial

$$P_n(x) = a_0 + a_1x + \dots + a_{n-1}x^{n-1} + a_nx^n$$

Motivation: different approaches to the inductive step may result in algorithms of very different time complexities

Evaluating Polynomials (cont'd)

- Let $P_{n-1}(x) = a_0 + a_1x + \cdots + a_{n-1}x^{n-1}$
- Induction hypothesis (first attempt): We know how to evaluate a polynomial represented by the input $a_0, a_1, \cdots, a_{n-1}$, at the point x , i.e., we know how to compute $P_{n-1}(x)$
- $P_n(x) = P_{n-1}(x) + a_nx^n$
- Number of multiplications:
 - $n + (n - 1) + \cdots + 2 + 1 = n(n + 1) / 2$

$$P_n(x) = (((\cdots((a_0) + a_1x) + a_2x^2) + \cdots) + a_{n-1}x^{n-1}) + a_nx^n)$$

Evaluating Polynomials (cont'd)

- Let $P_{n-1}(x) = a_0 + a_1x + \cdots + a_{n-1}x^{n-1}$
- Induction hypothesis (second attempt): We know how to compute $P_{n-1}(x)$, and we know how to compute x^{n-1}
- $P_n(x) = P_{n-1}(x) + a_nx(x^{n-1})$
- Number of multiplications: $2n$

x^{n-2} is cached when computing this monomial



$$P_n(x) = (((\cdots((a_0 + a_1x) + a_2xx) + \cdots) + a_{n-1}xx^{n-2}) + a_nxx^{n-1})$$

Evaluating Polynomials (cont'd)

- Let $P'_{n-1}(x) = a_1 + a_2x + \cdots + a_{n-1}x^{n-2} + a_nx^{n-1}$
- Induction hypothesis (final attempt): We know how to evaluate a polynomial represented by the coefficients $a_1, a_2, \cdots, a_n$, at the point x , i.e., we know how to compute $P'_{n-1}(x)$
- $P_n(x) = a_0 + P'_{n-1}(x) \cdot x$

Evaluating Polynomials (cont'd)

- More generally
 - $P'_0(x) = a_n$
 - $P'_i(x) = a_{n-i} + P'_{i-1}(x) \cdot x$, for $1 \leq i \leq n$
- Number of multiplications: n

$$P_n(x) = a_0 + (a_1 + (a_2 + (\cdots + (a_{n-1} + (a_n)x)x)\cdots)x)x$$

Evaluating Polynomials (cont'd)

Algorithm Polynomial_Evaluation($\bar{a}$, x):

begin

$P := a_n;$

for $i := 1$ **to** n **do**

$P := x * P + a_{n-1};$

end

$(\bar{a} = a_0, a_1, \dots, a_n)$

This algorithm is known as ***Horner's rule***

Maximal Induced Subgraph

Problem

Given an undirected graph $G = (V, E)$ and an integer k , find an induced subgraph $H = (U, F)$ of G of maximum size such that all vertices of H have degree $\geq k$ (in H), or conclude that no such induced subgraph exists

Direct approach: remove vertices whose degree is $< k$

What would be the order of removals?

- all vertices of degree $< k$ first and then vertices whose degrees were reduced, or
- one vertex of degree $< k$ and then continue with affected vertices

Maximal Induced Subgraph (cont'd)

Let's think of proving a theorem that the algorithm exists

Induction hypothesis

We know how to find maximum induced subgraphs all of whose vertices have degrees $\geq k$, provided that the number of vertices is $< n$

Base case ($n = k + 1$)

Inductive step: $G = (V, E)$ is a graph with $n > k + 1$ vertices

- Case 1: all vertices have degrees $\geq k$
- Case 2: some vertex v has degree $< k$

Maximal Induced Subgraph (cont'd)

Algorithm Max_Ind_Subgraph(G, k):

begin

if number of nodes in $G \leq k$ **then**

 Max_Ind_Subgraph := 0

else if the degree of every vertex of $G \geq k$ **then**

 Max_Ind_Subgraph := G ;

else

 let v be a vertex of G with degree $< k$;

 Max_Ind_Subgraph := Max_Ind_Subgraph($G - v$, k);

end

Recursive version

Maximal Induced Subgraph (cont'd)

Algorithm Max_Ind_Subgraph(G, k):

begin

while the degree of some vertex v of $G < k$ **do**

$G := G - v;$

Max_Ind_Subgraph := G ;

end

Iterative version

One-to-One Mapping

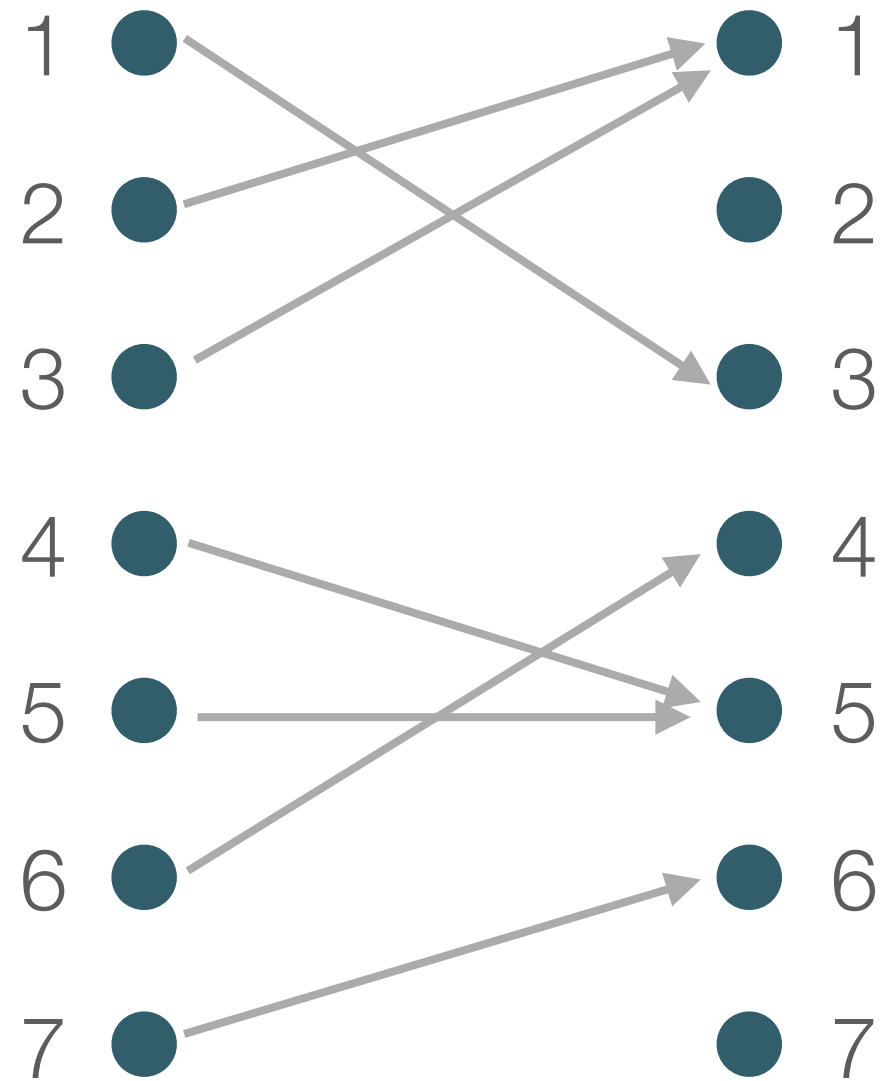
Problem

Given a finite set A and a mapping f from A to itself, find a subset $S \subseteq A$ with the maximum number of elements, such that

1. the function f maps every element of S to another element of S (i.e., f maps S into itself), and
2. no two elements of S are mapped to the same element (i.e., f is one-to-one with restricted to S)

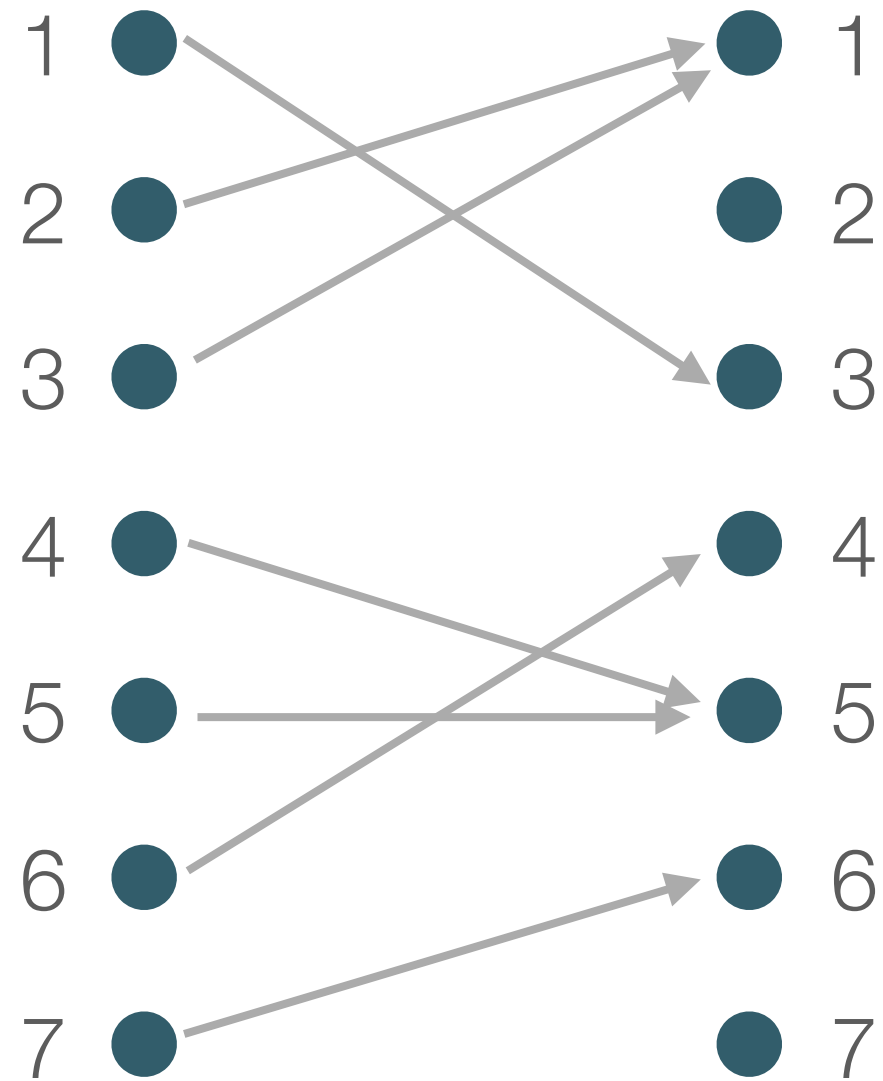
Direct approach: When i and j are both mapped to k , remove one of them. What would be the order of the removal. Is the result maximal?

One-to-One Mapping (cont'd)



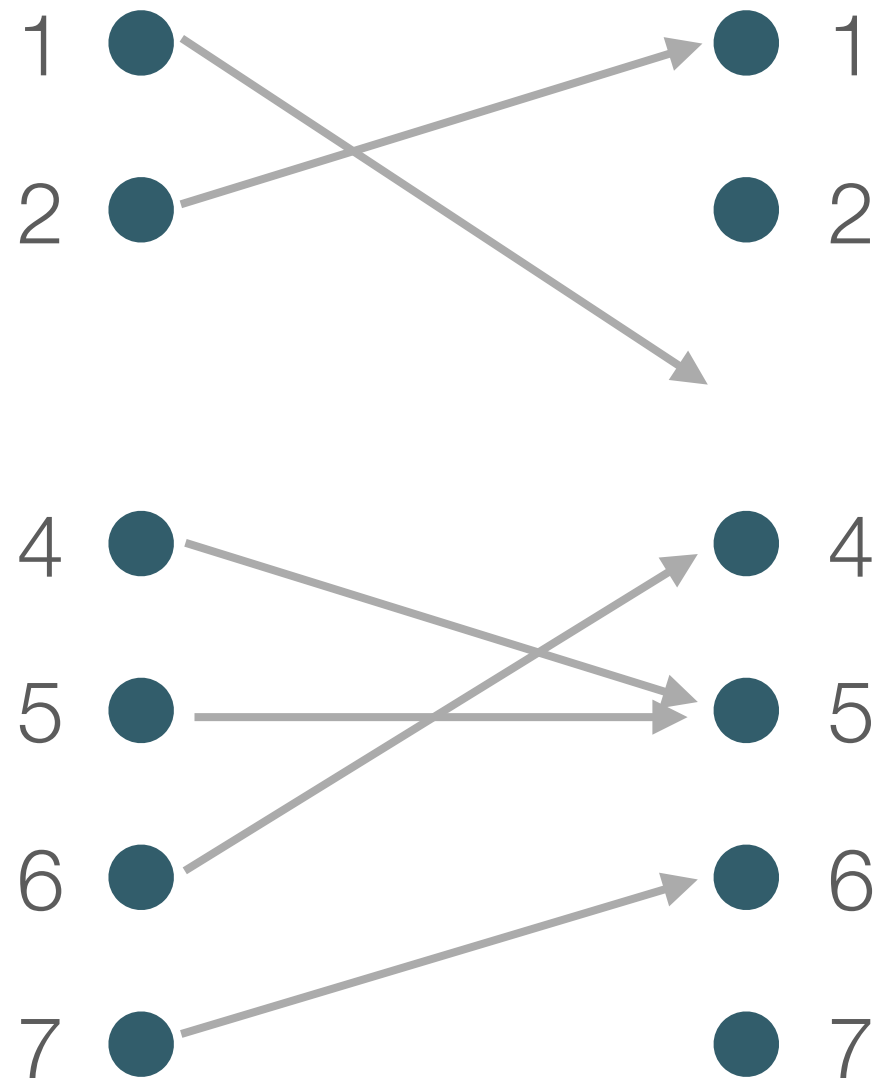
What would be the order of the removal?
Is the result maximal?

One-to-One Mapping (cont'd)



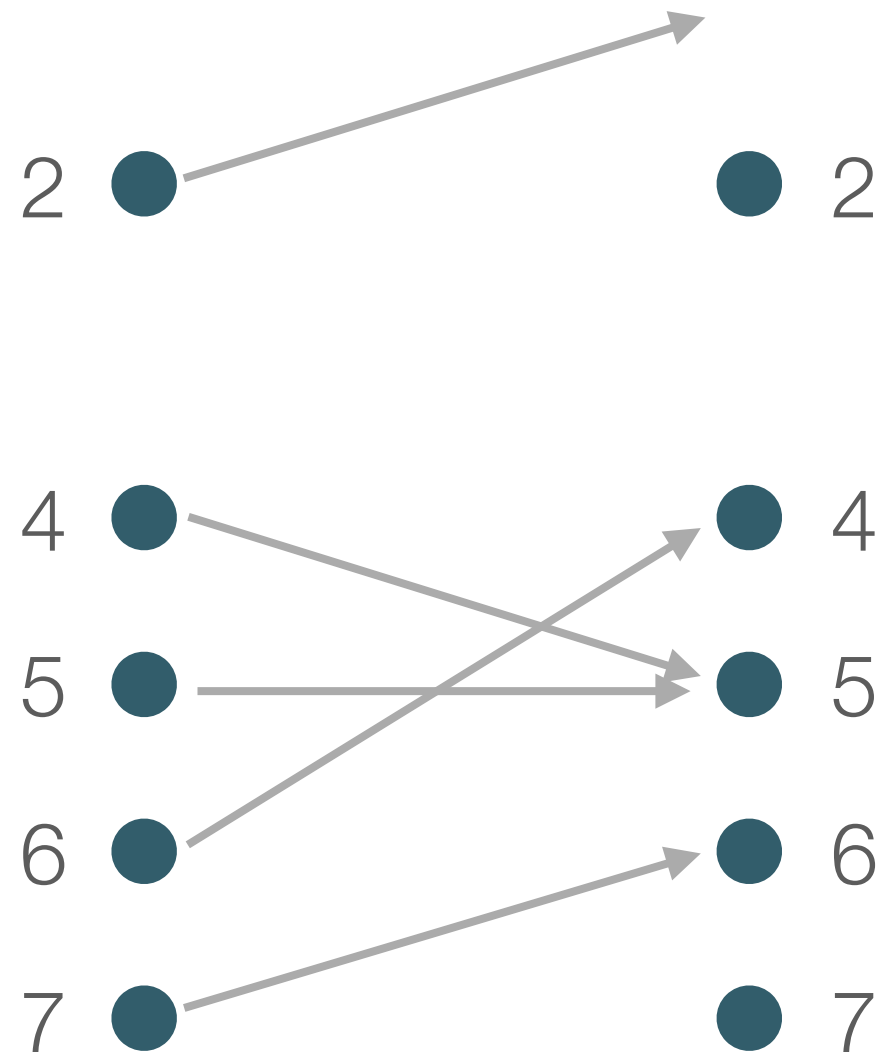
Case 1: remove 3 first

One-to-One Mapping (cont'd)



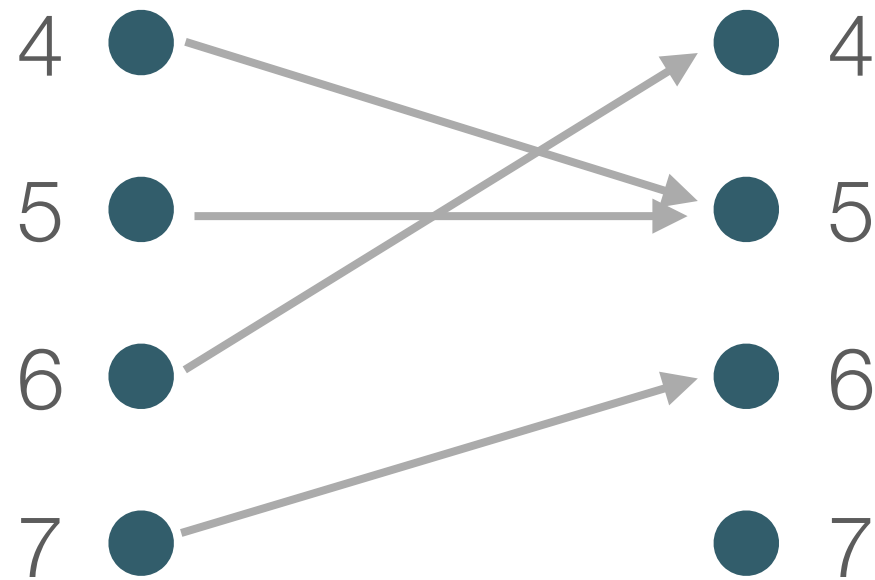
Case 1: remove 3 first

One-to-One Mapping (cont'd)



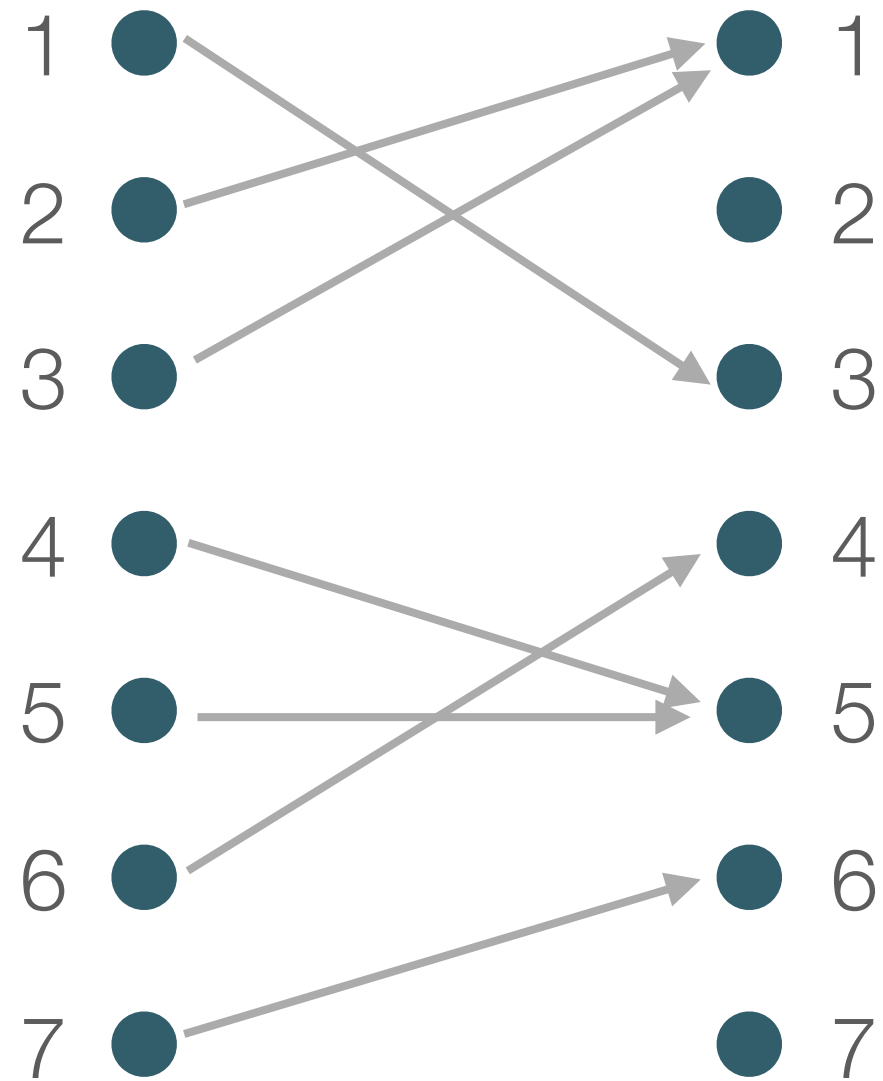
Case 1: remove 3 first

One-to-One Mapping (cont'd)



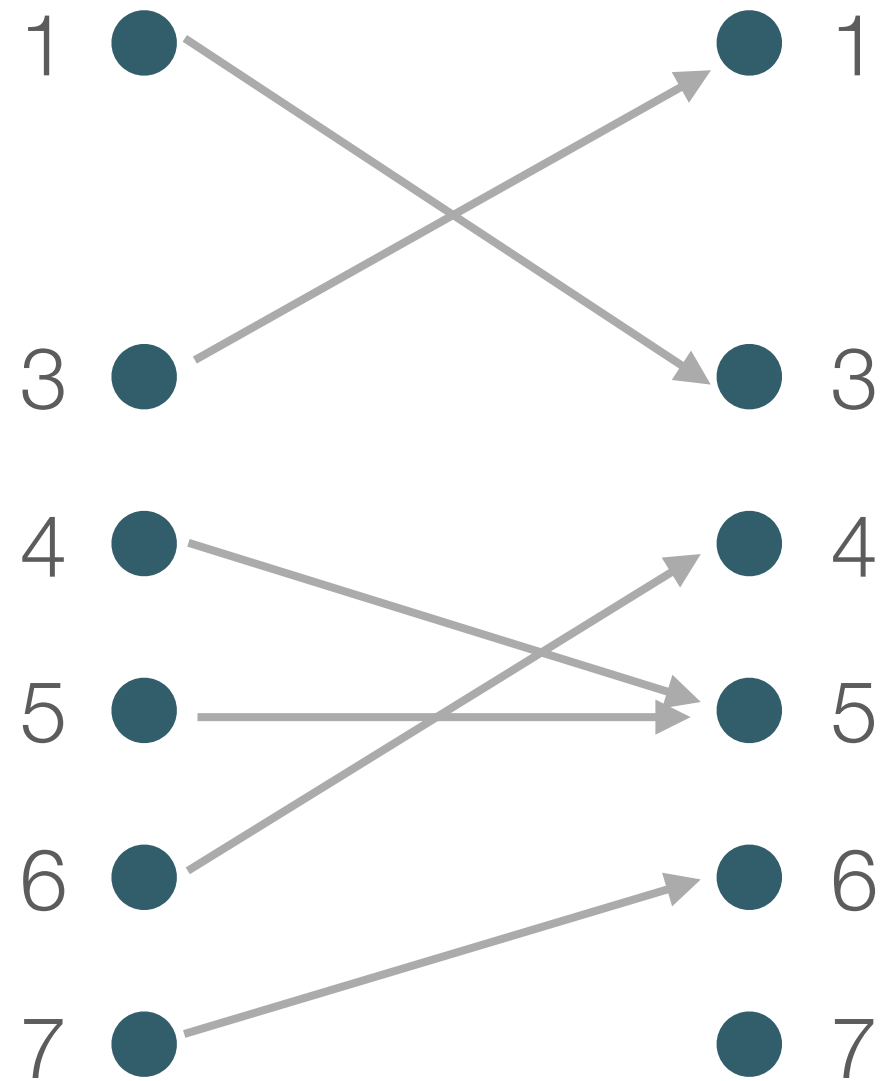
Case 1: remove 3 first

One-to-One Mapping (cont'd)



Case 2: remove 2 first

One-to-One Mapping (cont'd)



Case 2: remove 2 first

One-to-One Mapping (cont'd)

We can reduce the size of the problem by finding an element that does not belong to S

Inductive hypothesis

We know how to solve the problem for sets of $n - 1$ elements

Base case (one element)

Inductive step (n elements): any element i that has no other element mapped to it cannot belong to S

Case 1: there is such element

Case 2: there is no such element

One-to-One Mapping (cont'd)

Algorithm Mapping(f, n):

begin

$S := A;$

for $j := 1$ **to** n **do** $c[j] := 0;$

for $j := 1$ **to** n **do** increment $c[f[j]];$

for $j := 1$ **to** n **do**

if $c[j] = 0$ **then** put j in Queue;

while Queue is not empty **do**

 remove i from the top of Queue;

$S := S - \{ i \};$

 decrement $c[f[i]];$

if $c[f[i]] = 0$ **then** put $f[i]$ in Queue;

end

Celebrity

Problem

Given an $n \times n$ adjacency matrix, determine whether there exists an i (the “celebrity”) such that all the entries in the i -th column (except for the ii -th entry) are 1, and all the entries in the i -th row (except for the ii -th entry) are 0

- Celebrity: someone who is known by everyone but does not know anyone
- ij -th entry is one: i knows j , i.e., j is known by i
- A celebrity corresponds to a sink (with in-degree $n - 1$ and out-degree 0) of the directed graph
- Every directed graph has at most one sink

Celebrity (cont'd)

- Assume that we can find the celebrity among the $n - 1$ persons
- There are three possibilities
 - The celebrity is among the first $n - 1$ (ask two more questions)
 - The celebrity is the n -th person (ask $2(n - 1)$ questions in the n -th step and $O(n^2)$ in total)
 - There is no celebrity

Celebrity (cont'd)

- Instead of identifying a celebrity, it is easier to identify someone as a noncelebrity
- Ask if i knows j :
 - Yes: i is noncelebrity
 - No: j is noncelebrity
- Always remove one person with one question
- Check whether the last one is a celebrity or not

Celebrity (cont'd)

Algorithm Celebrity(Know):

begin

$i := 1;$

$j := 2;$

$\text{next} := 3;$

while $\text{next} \leq n + 1$ **do**

if $\text{Know}[i, j]$ **then** $i := \text{next}$

else $j := \text{next};$

$\text{next} := \text{next} + 1;$

if $i = n + 1$ **then** $\text{candidate} := j$

else $\text{candidate} := i;$

...

Celebrity (cont'd)

Algorithm Celebrity(Know):

...

wrong := false;

k := 1;

Know[candidate, candidate] := false;

while not wrong and $k \leq n$ **do**

if Know[candidate, k] **then** wrong := true;

if not Know[k, candidate] **then**

if candidate $\neq$ k **then** wrong := true;

 k := k + 1

if not wrong **then** celebrity := candidate

else celebrity := 0

end

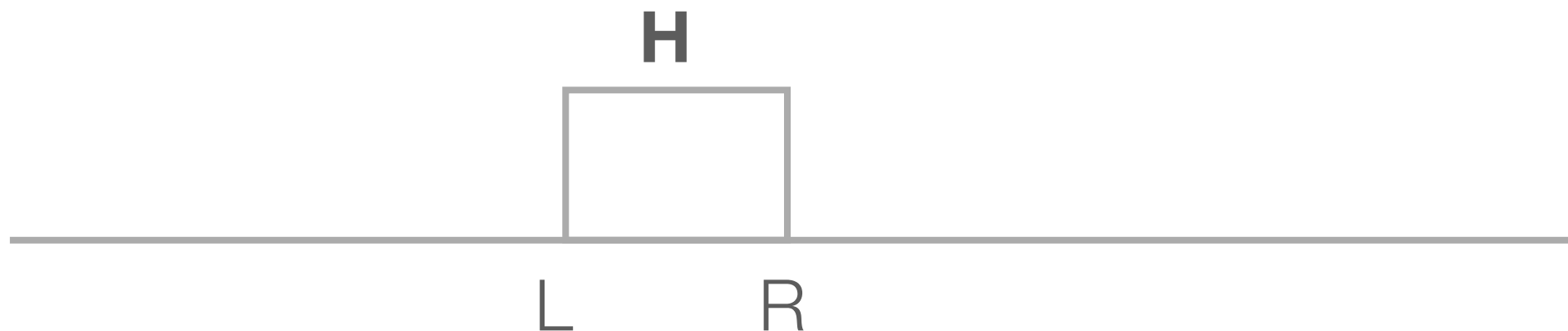
The Skyline Problem

Problem

Given the exact locations and shapes of several rectangular buildings in a city, draw the skyline (in two dimensions) of these buildings, eliminating hidden lines

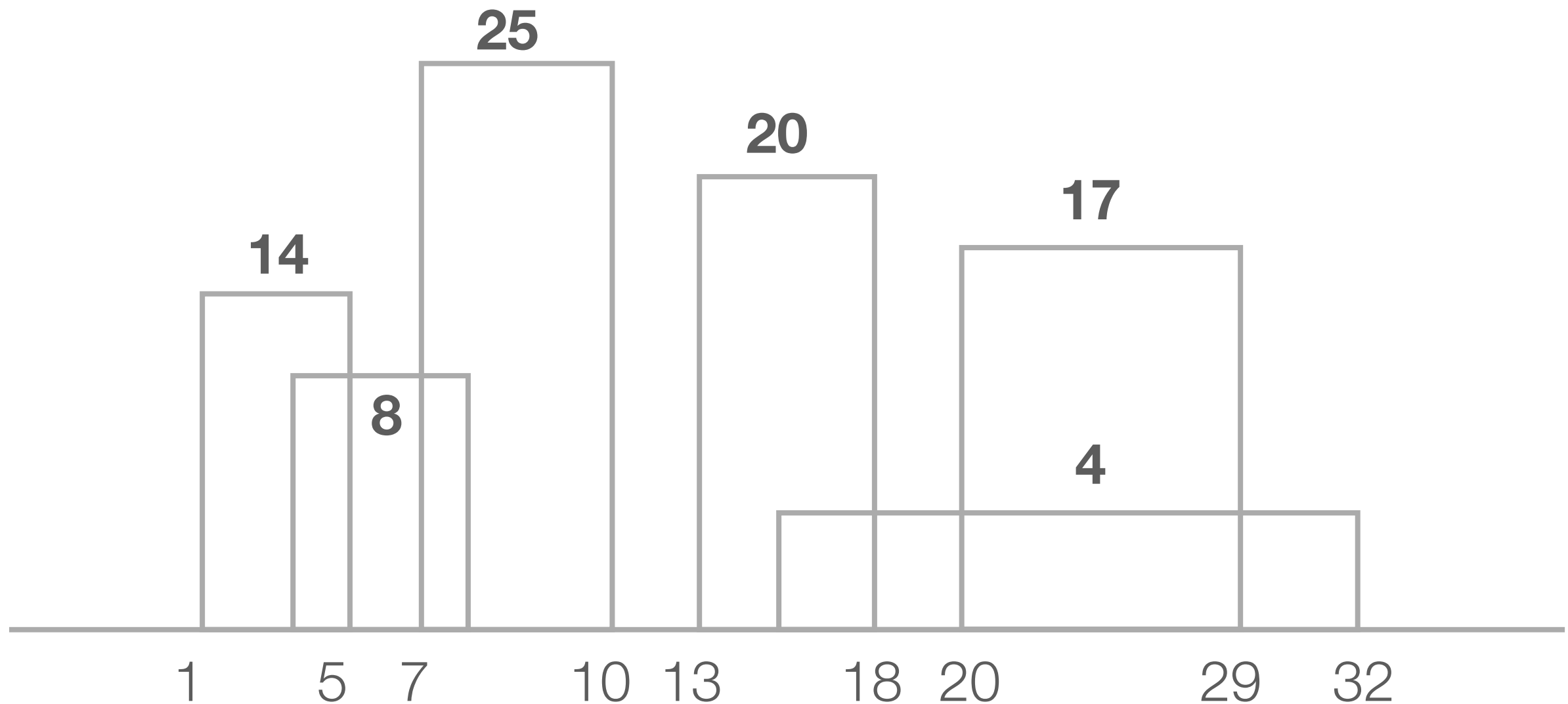
Motivation: different approaches to the inductive step may result in algorithms of very different time complexities

A building is represented by a triple $(L, \mathbf{H}, R)$



The Skyline Problem (cont'd)

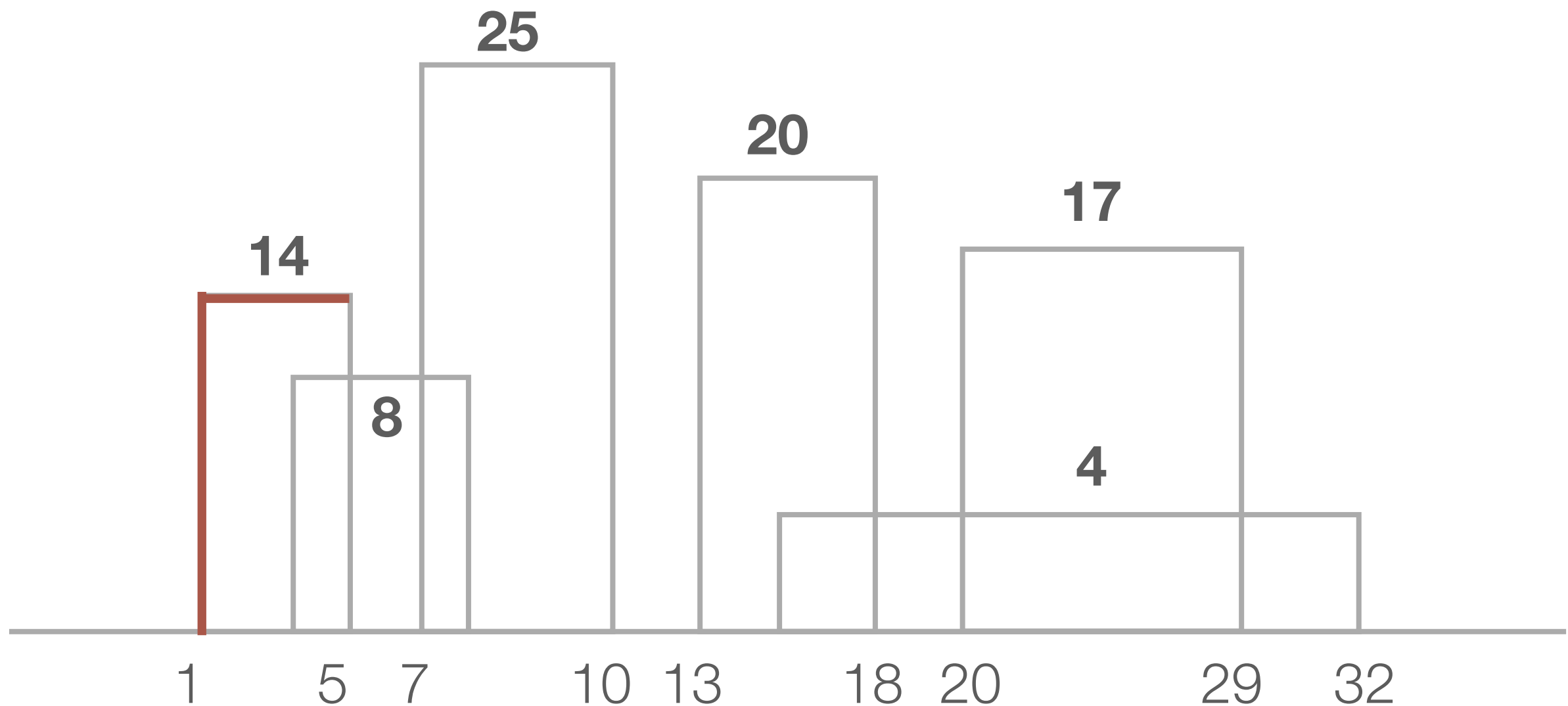
Example:



The Skyline Problem (cont'd)

Example:

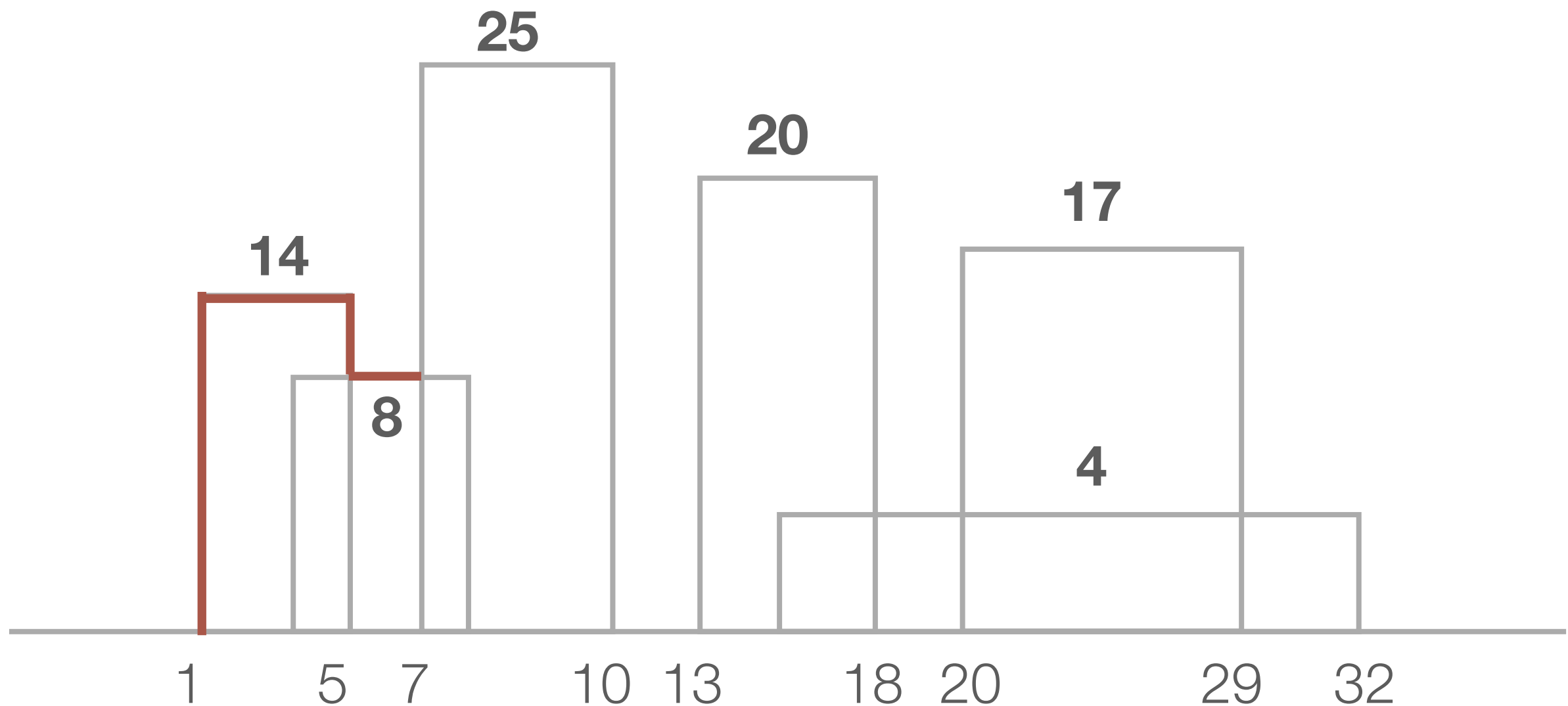
(1, **14**, 5



The Skyline Problem (cont'd)

Example:

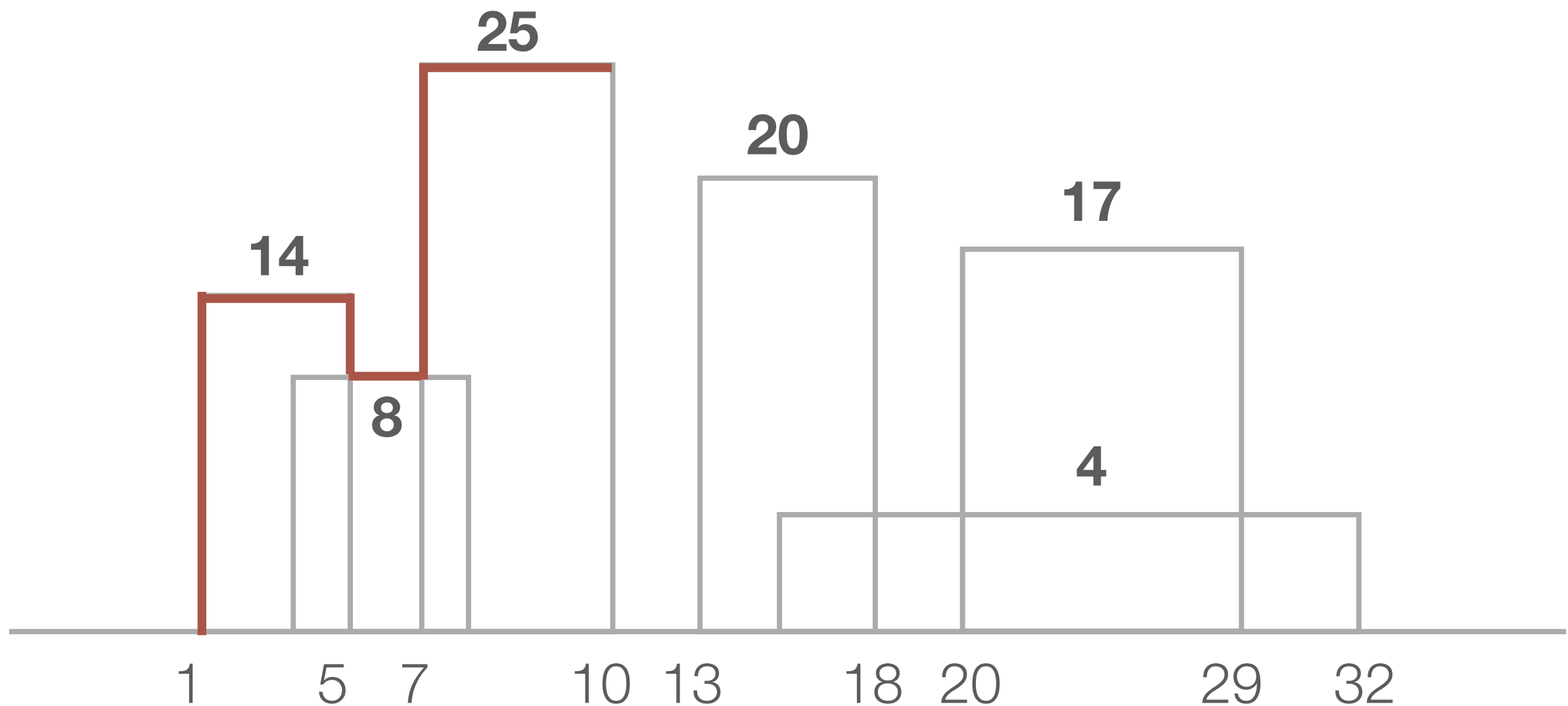
(1, **14**, 5, **8**, 7



The Skyline Problem (cont'd)

Example:

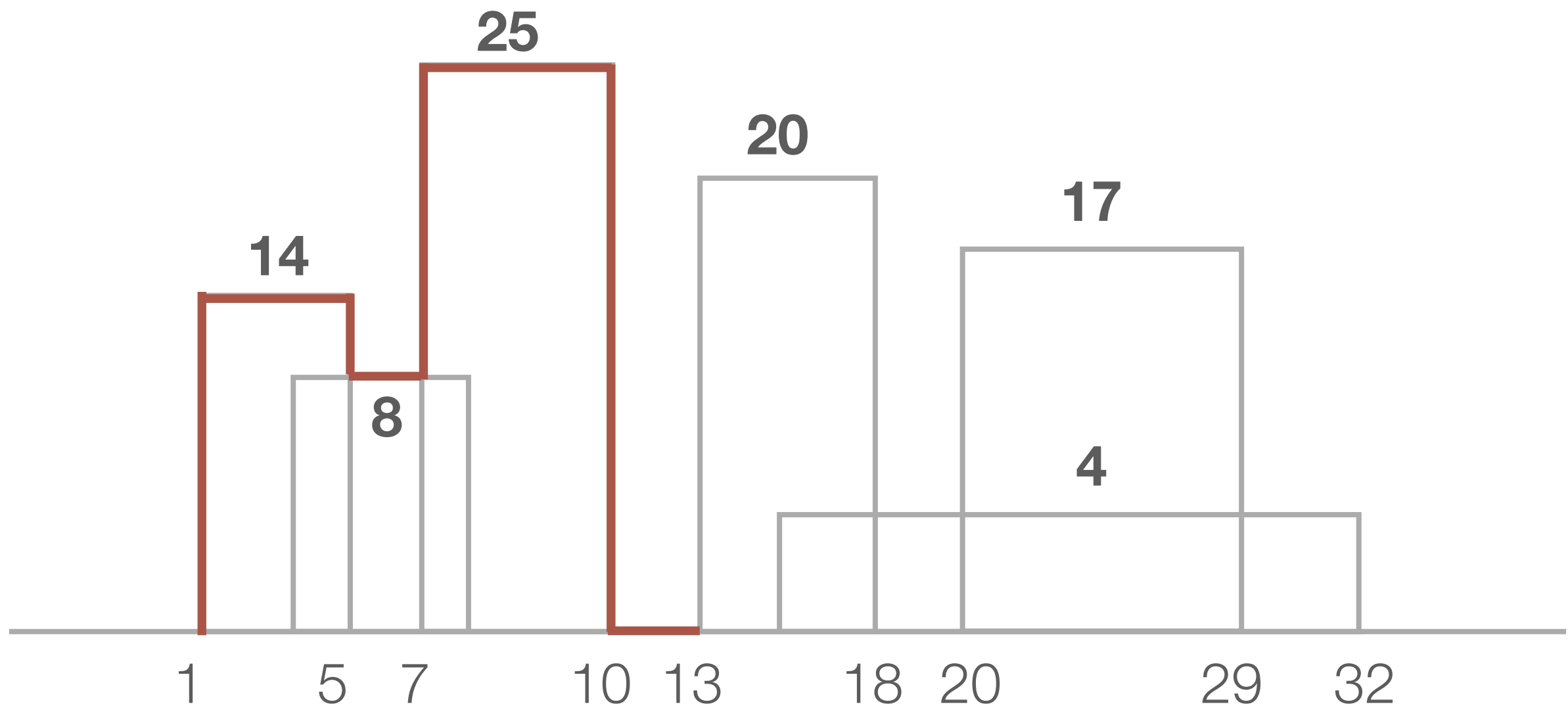
(1, **14**, 5, **8**, 7, **25**, 10



The Skyline Problem (cont'd)

Example:

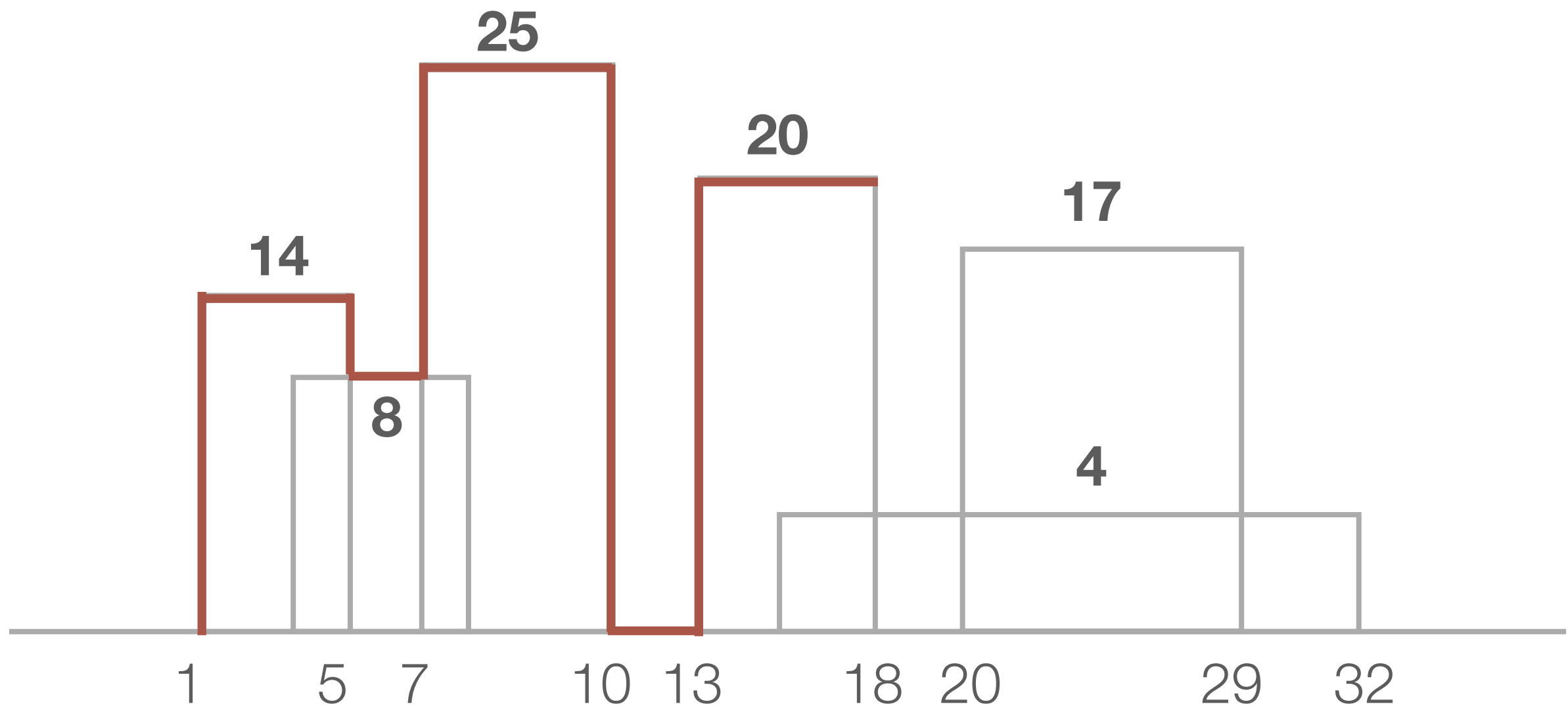
(1, **14**, 5, **8**, 7, **25**, 10, **0**, 13



The Skyline Problem (cont'd)

Example:

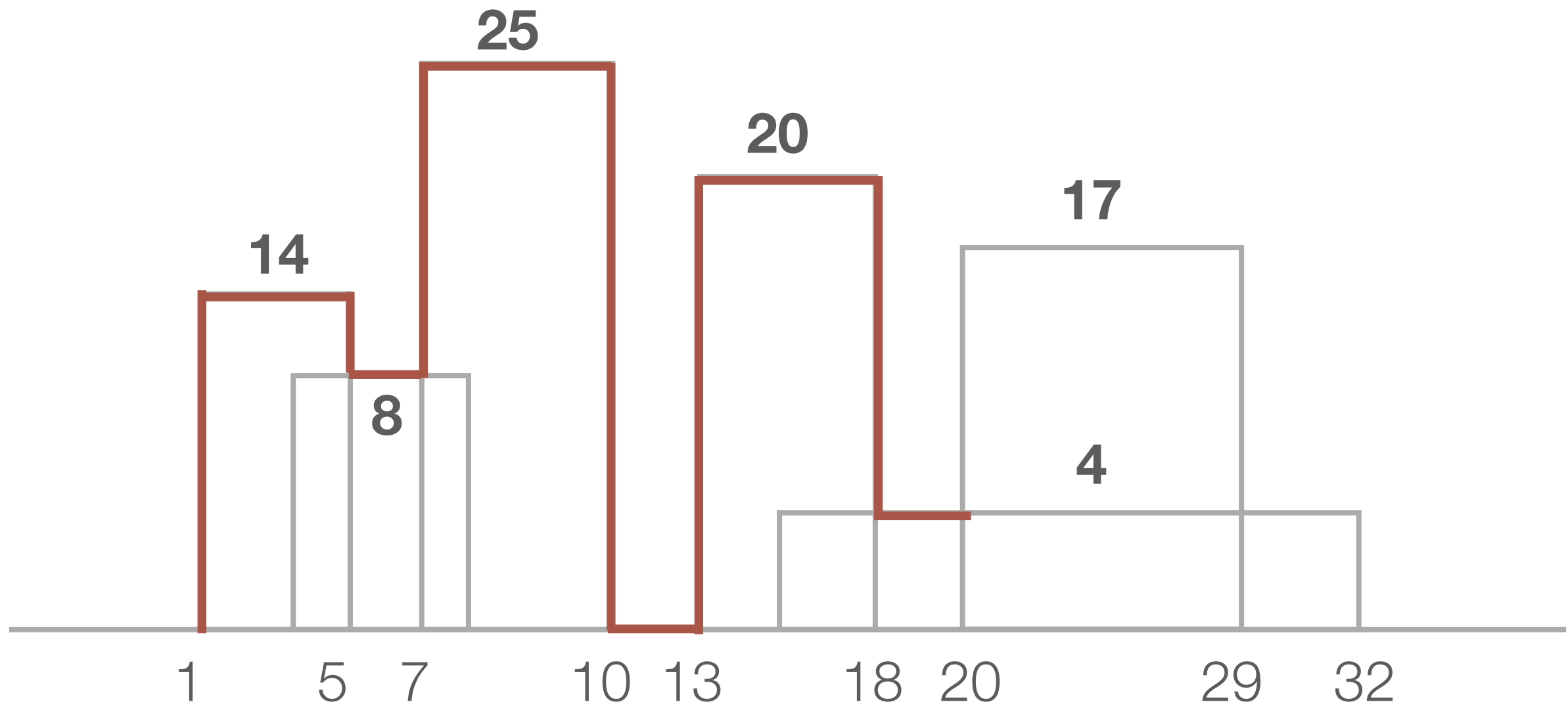
(1, **14**, 5, **8**, 7, **25**, 10, **0**, 13, **20**, 18



The Skyline Problem (cont'd)

Example:

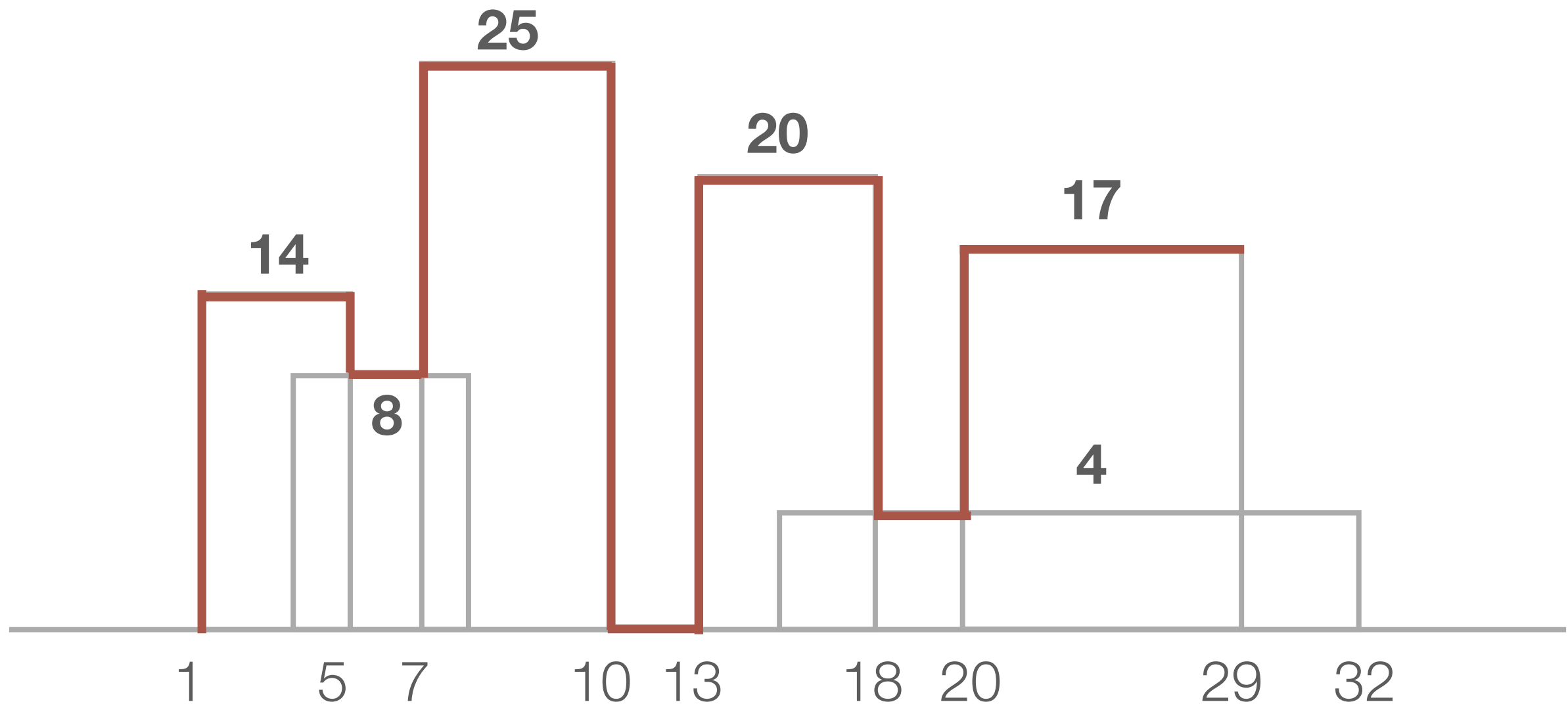
(1, **14**, 5, **8**, 7, **25**, 10, **0**, 13, **20**, 18, **4**, 20)



The Skyline Problem (cont'd)

Example:

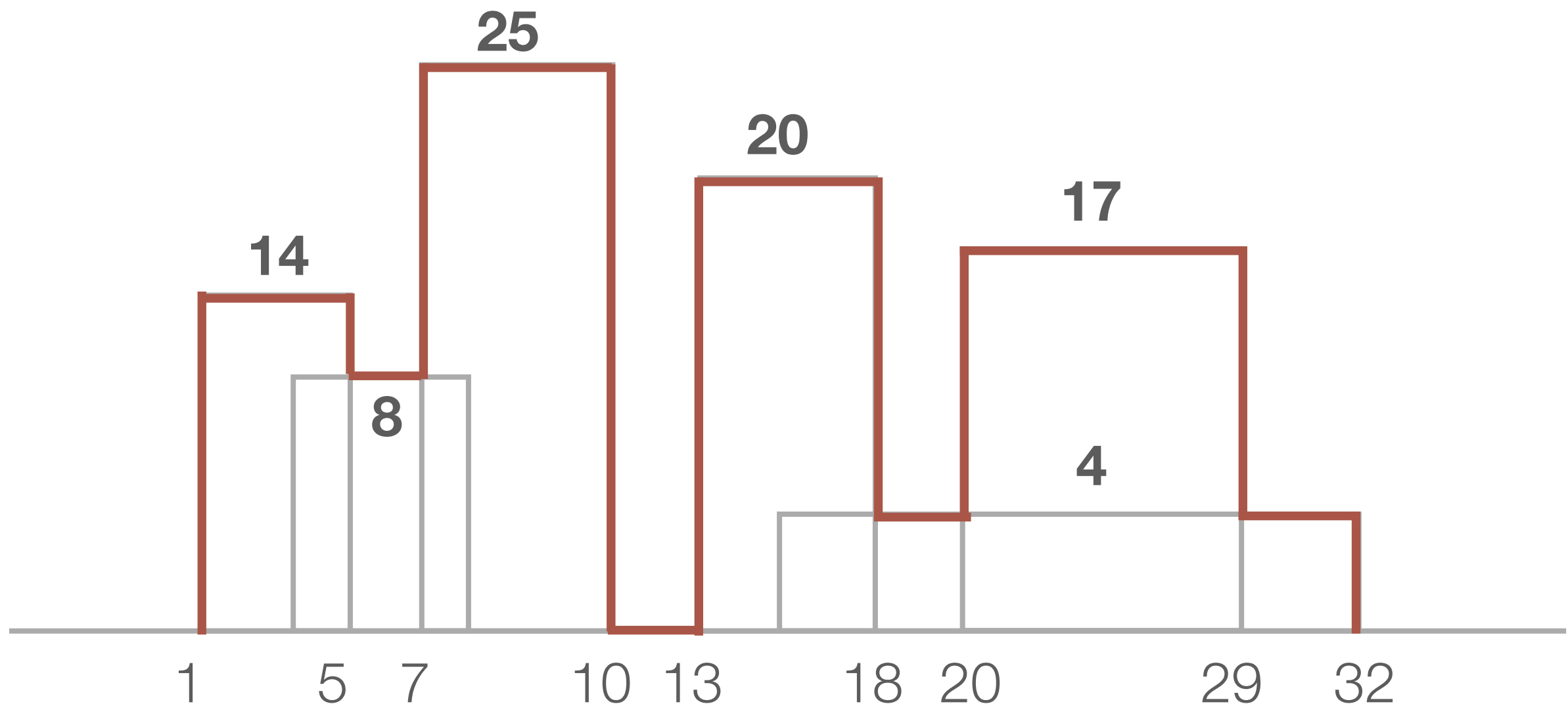
(1, **14**, 5, **8**, 7, **25**, 10, **0**, 13, **20**, 18, **4**, 20, **17**, 29)



The Skyline Problem (cont'd)

Example:

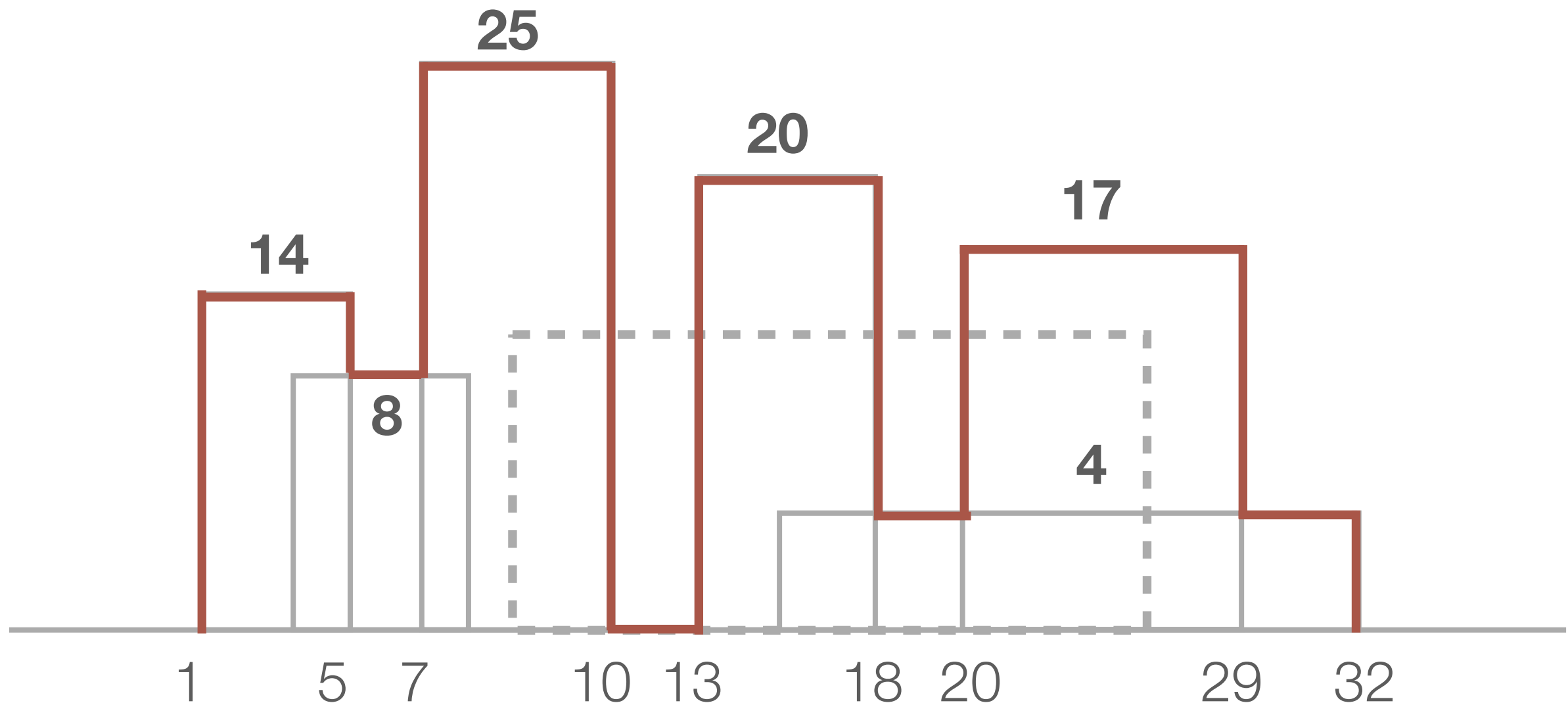
(1, **14**, 5, **8**, 7, **25**, 10, **0**, 13, **20**, 18, **4**, 20, **17**, 29, **4**, 32)



The Skyline Problem (cont'd)

Induction hypothesis: We know how to solve the problem for $n - 1$ buildings

(1, **14**, 5, **8**, 7, **25**, 10, **0**, 13, **20**, 18, **4**, 20, **17**, 29, **4**, 32)
(9, **10**, 27)



The Skyline Problem (cont'd)

- Adding one building at a time:
 - $T(1) = O(1)$
 - $T(n) = T(n - 1) + O(n), n \geq 2$
- Time complexity: $O(n^2)$

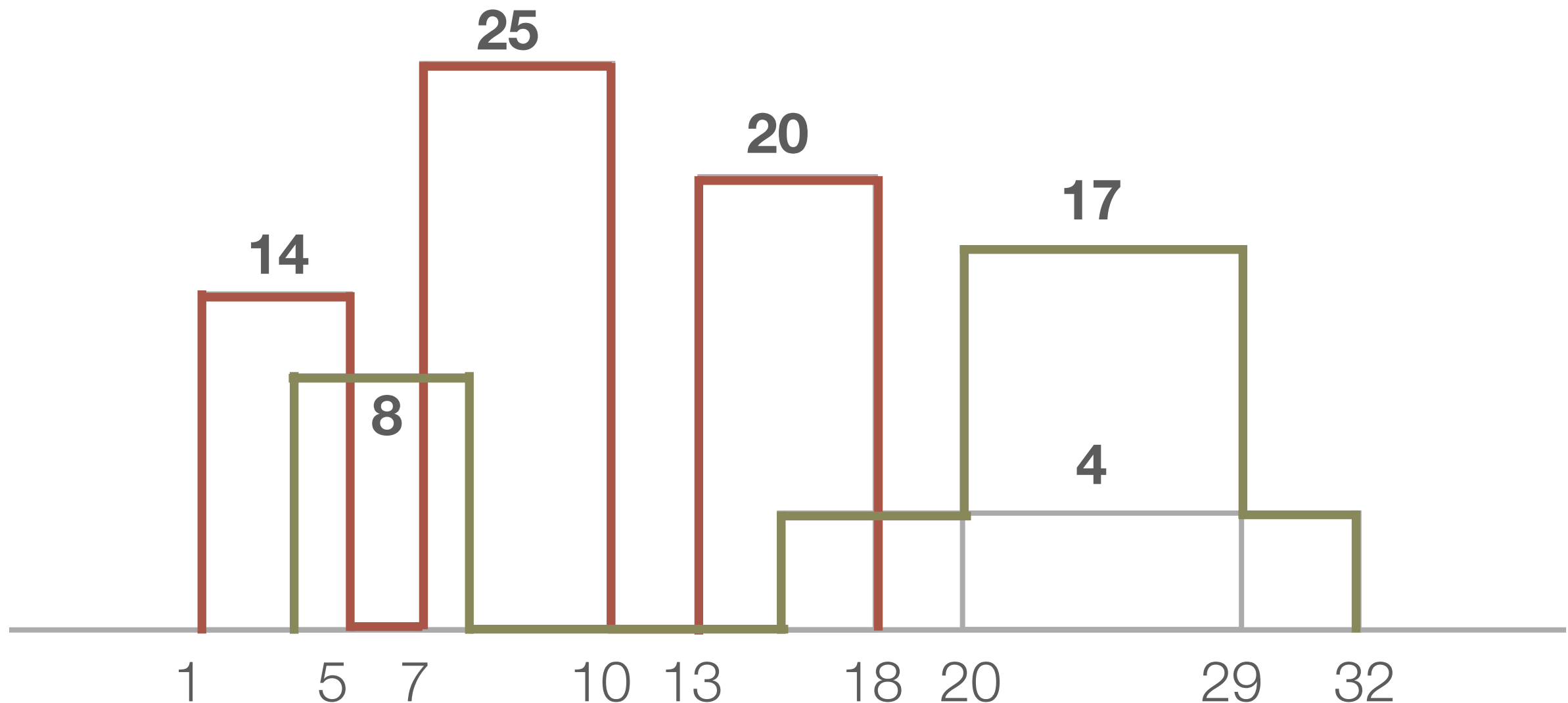
The Skyline Problem (cont'd)

Induction hypothesis: We know how to solve the problem for $n / 2$ buildings

(1, **14**, 5, **0**, 7, **25**, 10, **0**, 13, **20**, 18)

(3, **8**, 8, **0**, 15, **4**, 20, **17**, 29, **4**, 32)

divide-and-conquer



The Skyline Problem (cont'd)

- Merge two skylines every round:
 - $T(1) = O(1)$
 - $T(n) = 2T(n / 2) + O(n), n \geq 2$
- Time complexity: $O(n \log n)$

Balance Factors in Binary Trees

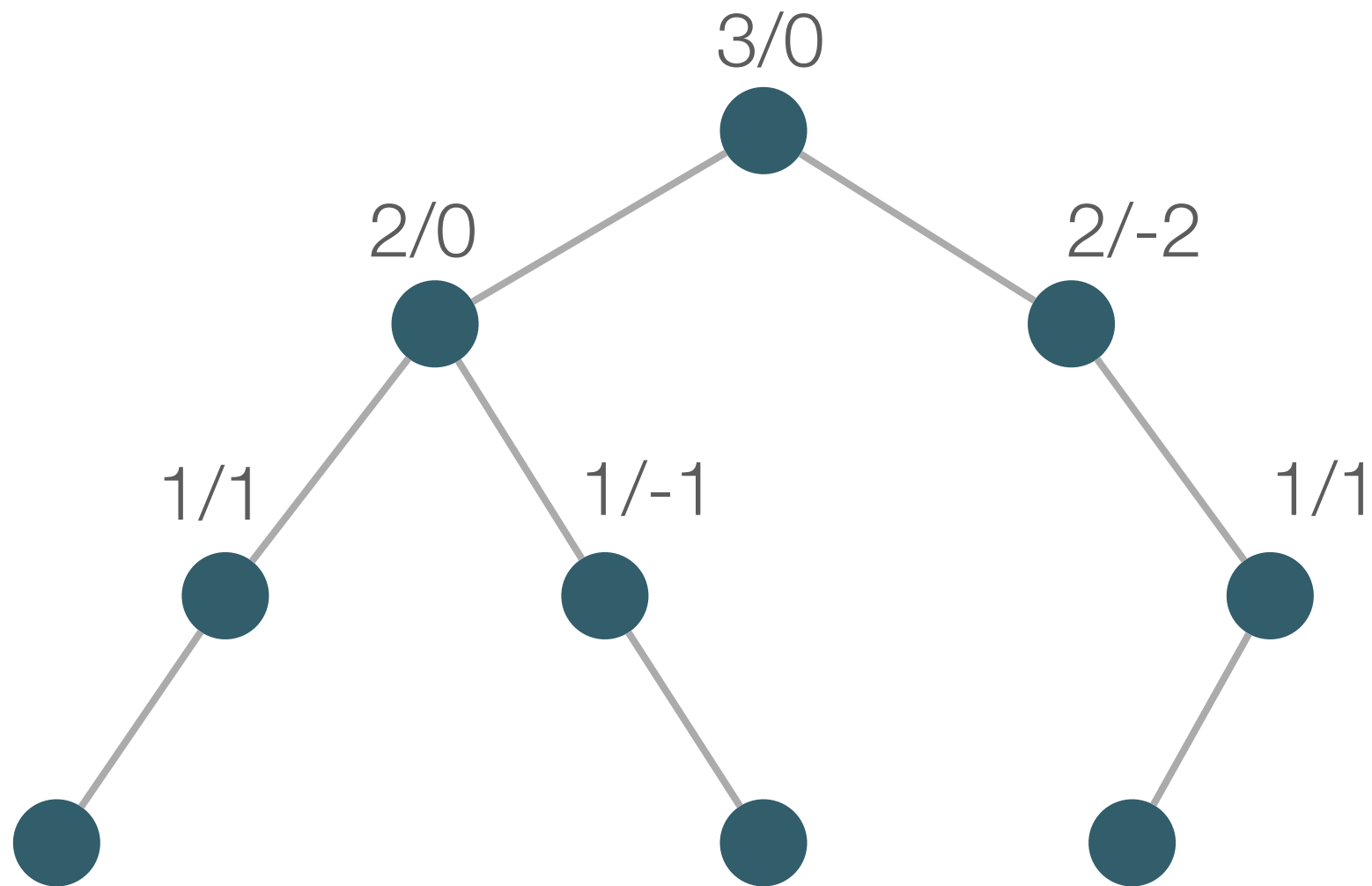
Problem

Given a binary tree T with n nodes, compute the balance factors of all nodes

The ***balance factor*** of a node is defined as the difference between the height of the node's left subtree and the height of the node's right subtree

Motivation: an example of why we must ***strengthen the hypothesis*** (and hence the problem to be solved)

Balance Factors in Binary Trees (cont'd)



h/b: h = height, b = balance factor

Balance Factors in Binary Trees (cont'd)

- **Induction hypothesis:** We know how to compute balance factors of all nodes in trees that have $< n$ nodes
 - We still need to compute the heights of the children of a node when computing the balance factor of the node
- **Induction hypothesis:** We know how to compute balance factors ***and heights*** of all nodes in trees that have $< n$ nodes

Note: in the inductive step for a tree of n nodes, both the balance factor and the height of the root have to be computed

Maximum Consecutive Subsequence

Problem

Given a sequence $x_1, x_2, \dots, x_n$ of real numbers (not necessarily positive) find a subsequence $x_i, x_{i+1}, \dots, x_j$ (of consecutive elements) such that the sum of the numbers in it is maximum over all subsequences of consecutive elements

Example:

Sequence: (2, -3, 1.5, -1, 3, -2, -3, 3)

Maximum subsequence: (1.5, -1, 3)

Maximum Consecutive Subsequence (cont'd)

Induction hypothesis

We know how to find the maximum subsequence in sequences of size $< n$

Consider a sequence $S = (x_1, x_2, \dots, x_n)$ of size $n > 1$

The maximum subsequence of $S' = (x_1, x_2, \dots, x_{n-1})$ is $S_M' = (x_i, x_{i+1}, \dots, x_j)$ for certain i and j such that $1 \leq i \leq j \leq n - 1$ (by induction)

Case 1: $j = n - 1$



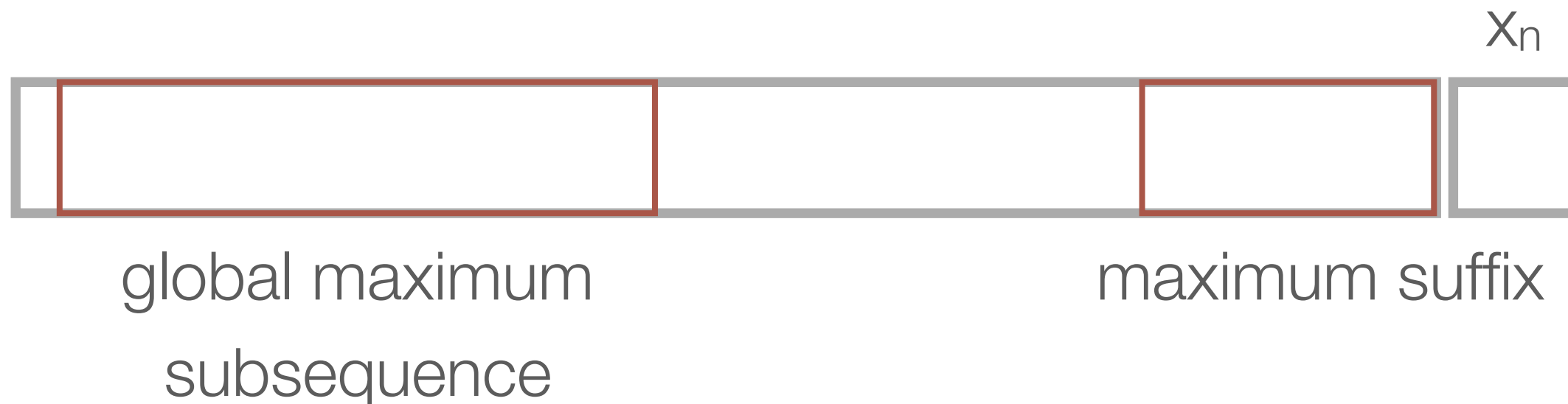
Case 2: $j < n - 1$



Maximum Consecutive Subsequence (cont'd)

Stronger induction hypothesis

We know how to find, in sequences of size $< n$, the maximum subsequence overall ***and the maximum subsequence that is a suffix***



Maximum Consecutive Subsequence (cont'd)

Algorithm Max_Consec_Subseq(X, n):

begin

Global_Max := 0;

Suffix_Max := 0;

for i := 1 **to** n **do**

if $X[i] + \text{Suffix_Max} > \text{Global_Max}$ **then**

 Suffix_Max := Suffix_Max + X[i];

 Global_Max := Suffix_Max

else if $X[i] + \text{Suffix_Max} > 0$ **then**

 Suffix_Max := Suffix_Max + X[i]

else

 Suffix_Max := 0

end

Strengthening the Induction Hypothesis

- Denote a theorem by P
- The induction hypothesis can be denoted by $P(< n)$
- The proof must conclude that $P(< n) \Rightarrow P(n)$
- We can add another assumption Q , under which the proof becomes easier
 - $[P \text{ and } Q](< n) \Rightarrow P(n)$ v.s. $P(< n) \Rightarrow P(n)$
- We need to prove $[P \text{ and } Q](< n) \Rightarrow [P \text{ and } Q](n)$

The Knapsack Problem

Problem

Given an integer K and n items of different sizes such that the i -th item has an integer size k_i , find a subset of the items whose sizes sum to exactly K , or determine that no such subset exists

Denote the problem by $P(n, K)$ where n is the number of items and K is the size of the knapsack

$P(i, k)$ denotes the problem with the first i items and a knapsack of size k

The Knapsack Problem (cont'd)

Induction hypothesis

We know how to solve $P(n - 1, K)$

Base case ($P(1, K)$): trivial

Inductive step ($P(n, K)$):

- Case 1: There is a solution for $P(n - 1, K)$
- Case 2: There is no solution for $P(n - 1, K)$
 - n -th item must be included
 - Need to solve $P(n - 1, K - k_n)$

The Knapsack Problem (cont'd)

Induction hypothesis

We know how to solve $P(n - 1, K)$ for all $0 \leq k \leq K$

K in this induction hypothesis can be any number

Base case ($P(1, k)$): trivial

Inductive step: $P(n, k)$ has a solution if one of the following subproblem has a solution

- $P(n - 1, k)$
- $P(n - 1, k - k_n)$

The time complexity is exponential!

The Knapsack Problem (cont'd)

$P(n, k)$ has a solution if one of the following subproblem has a solution

- $P(n - 1, k)$
- $P(n - 1, k - k_n)$

The time complexity is exponential.

There are only nK different possible problems.

Why the complexity is exponential?

How to reduce the time complexity?

The Knapsack Problem (cont'd)

- We can reduce the time complexity by solving all the nK problems with a table construction
- The table is constructed iteratively
- Each entry is computed from a combination of other entries above it or to the left of it in the table
- This technique is called ***dynamic programming***

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$																	
$k_2=3$																	
$k_3=5$																	
$k_4=6$																	

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$																	
$k_3=5$																	
$k_4=6$																	

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	?											
$k_3=5$																	
$k_4=6$																	

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3=5$	O	-	O	O	-	?											
$k_4=6$																	

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I	-	I	-	-	-	-	-	-
$k_4=6$	O	-	O	O	-	O	I	O	O	I	O	I	-	I	I	-	I

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I	-	I	-	-	-	-	-	-
$k_4=6$	O	-	O	O	-	O	I	O	O	I	O	I	-	I	I	-	I

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I	-	I	-	-	-	-	-	-
$k_4=6$	O	-	O	O	-	O	I	O	O	I	O	I	-	I	I	-	I

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I	-	I	-	-	-	-	-	-
$k_4=6$	O	-	O	O	-	O	I	O	O	I	O	I	-	I	I	-	I

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I	-	I	-	-	-	-	-	-
$k_4=6$	O	-	O	O	-	O	I	O	O	I	O	I	-	I	I	-	I

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$k_1=2$	O	-	I	-	-	-	-	-	-	-	-	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-	-	-	-	-	-	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I	-	I	-	-	-	-	-	-
$k_4=6$	O	-	O	O	-	O	I	O	O	I	O	I	-	I	I	-	I

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8
$k_1=2$	O	-	I	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I
$k_4=6$	O	-	O	O	-	O	I	O	O

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8
$k_1=2$	O	-	I	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I
$k_4=6$	O	-	O	O	-	O	I	O	O

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8
$k_1=2$	O	-	I	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I
$k_4=6$	O	-	O	O	-	O	I	O	O

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8
$k_1=2$	O	-	I	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I
$k_4=6$	O	-	O	O	-	O	I	O	O

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

	0	1	2	3	4	5	6	7	8
$k_1=2$	O	-	I	-	-	-	-	-	-
$k_2=3$	O	-	O	I	-	I	-	-	-
$k_3=5$	O	-	O	O	-	O	-	I	I
$k_4=6$	O	-	O	O	-	O	I	O	O

I: a solution containing this item has been found

O: a solution without this item has been found

-: no solution has not yet been found

Trace back to see what items are included

The Knapsack Problem (cont'd)

Algorithm Knapsack(S, K):

begin

$P[0, 0].\text{exist} := \text{true};$

for $k := 1$ **to** K **do**

$P[0, k].\text{exist} := \text{false};$

for $i := 1$ **to** n **do**

for $k := 0$ **to** K **do**

$P[i, k].\text{exist} := \text{false};$

if $P[i - 1, k].\text{exist}$ **then**

$P[i, k].\text{exist} := \text{true}; P[i, k].\text{belong} := \text{false}$

else if $k - S[i] \geq 0$ **then**

if $P[i - 1, k - S[i]].\text{exist}$ **then**

$P[i, k].\text{exist} := \text{true}; P[i, k].\text{belong} := \text{true}$

end