
Greedy Approach

Md. Tanvir Rahman

Lecturer, Dept. of ICT

Mawlana Bhashani Science and Technology University



Greedy Algorithm

- Greedy algorithms make the choice that looks best at the moment.
- This **locally optimal** choice may lead to a globally optimal solution (i.e. an optimal solution to the entire problem).

Greedy Algorithm

- A *greedy algorithm* always makes the choice that looks best at the moment
 - Some examples:
 - Playing cards
 - Invest on stocks
 - Choose a university
 - The hope: a locally optimal choice will lead to a globally optimal solution
 - For some problems, it works
- greedy algorithms tend to be easier to code

Greedy vs Dynamic Programming

Dynamic Programming	Greedy Approach
<ul style="list-style-type: none">• At each step, the choice is determined based on the solutions of sub problems	<ul style="list-style-type: none">• At each step, we quickly make a choice that currently looks best. A local greedy choice.
<ul style="list-style-type: none">• It is slower	<ul style="list-style-type: none">• It is faster
<ul style="list-style-type: none">• It is more complex	<ul style="list-style-type: none">• It is simple

Example: Making Change

- Instance: amount (in cents) to return to customer
- Problem: do this using fewest number of coins
- Example:
 - Assume that we have an unlimited number of coins of various denominations:
 - 1c (pennies), 5c (nickels), 10c (dimes), 25c (quarters), 1\$ (loonies)
 - Objective: Pay out a given sum \$5.64 with the smallest number of coins possible.

Example: Making Change

- E.g.:

$$\$5.64 = \$1 + \$1 + \$1 + \$1 + \$1 +$$

$$.25 + .25 + .10 +$$

$$.01 + .01 + .01 + .01$$

An Activity Selection Problem (Conference Scheduling Problem)

- **Input: A set of activities $S = \{a_1, \dots, a_n\}$**
- Each activity has start time and a finish time
 - $A_i = [s_i, f_i]$
- Two activities are compatible if and only if their interval does not overlap
- **Output: a maximum-size subset of mutually compatible activities**

The Activity Selection Problem

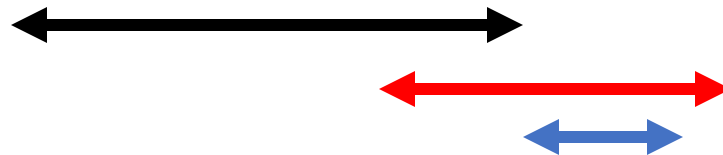
- Here are a set of start and finish times

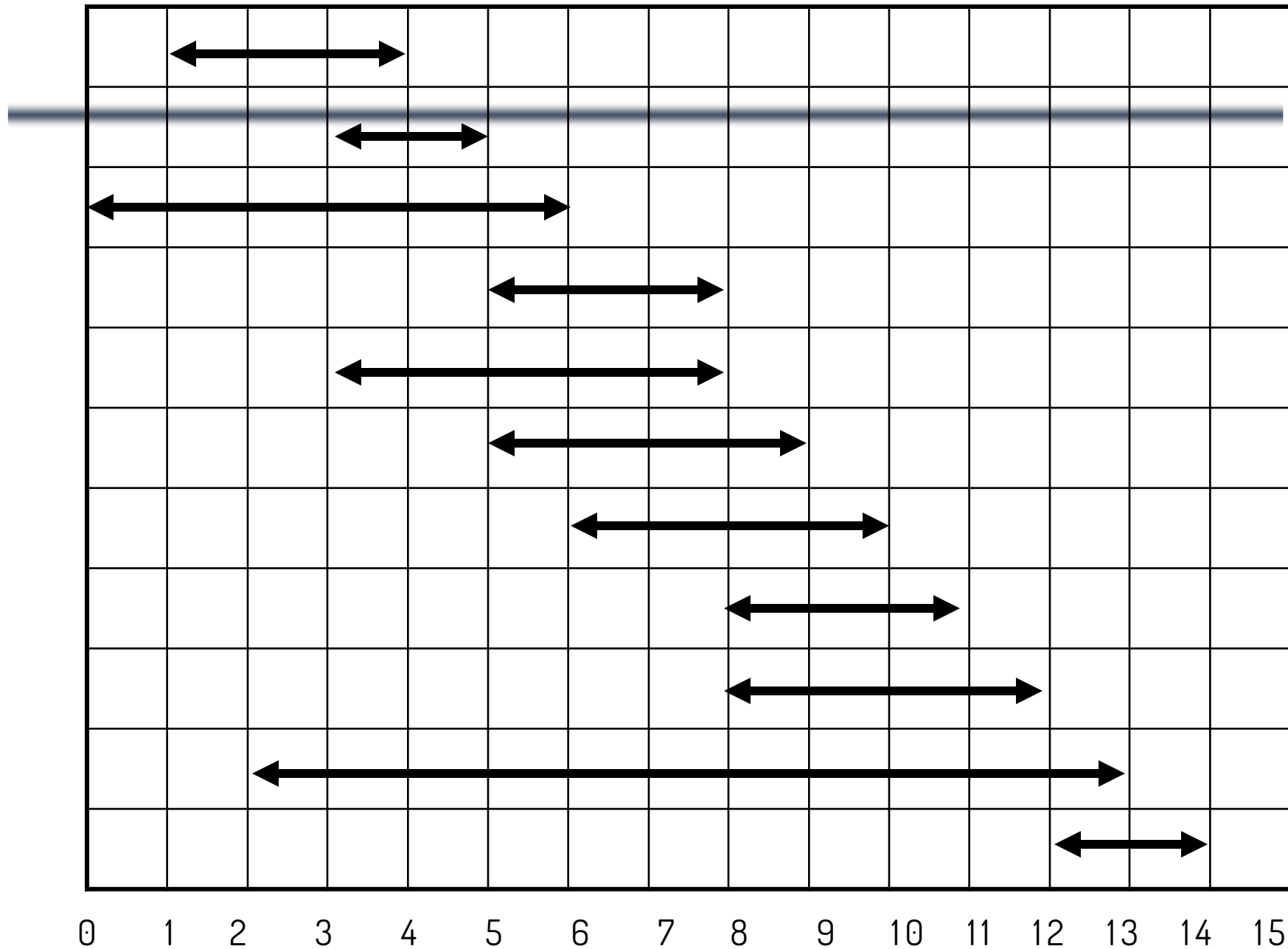
i	1	2	3	4	5	6	7	8	9	10	11
s_i	1	3	0	5	3	5	6	8	8	2	12
f_i	4	5	6	7	8	9	10	11	12	13	14

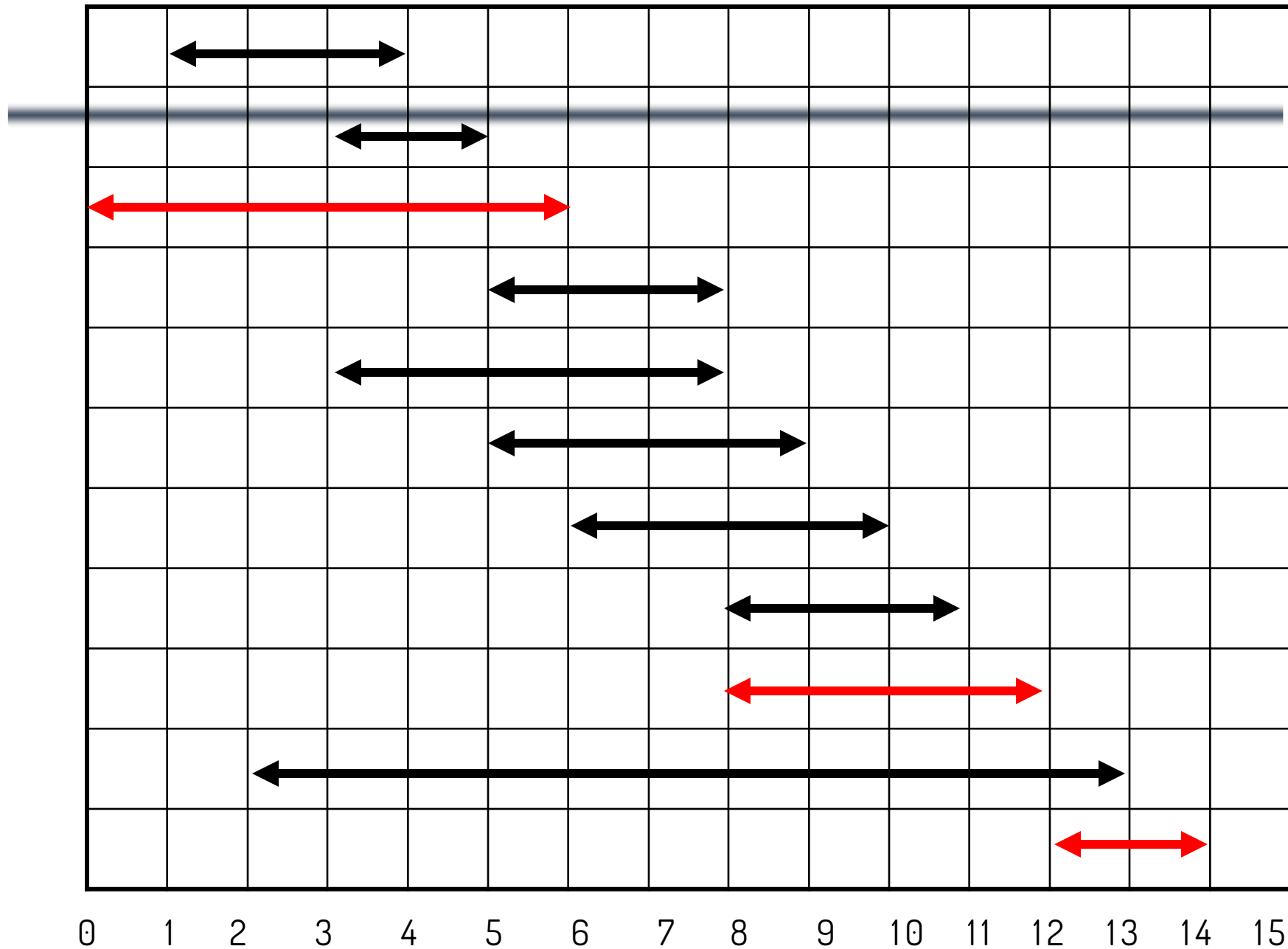
- What is the maximum number of activities that can be completed?
 - $\{a_3, a_7, a_{11}\}$ can be completed
 - $\{a_3, a_8, a_{11}\}$ can be completed
 - $\{a_3, a_9, a_{11}\}$ can be completed
 - But so can $\{a_1, a_4, a_8, a_{11}\}$ which is a larger set
 - But it is not unique, consider $\{a_2, a_4, a_9, a_{11}\}$

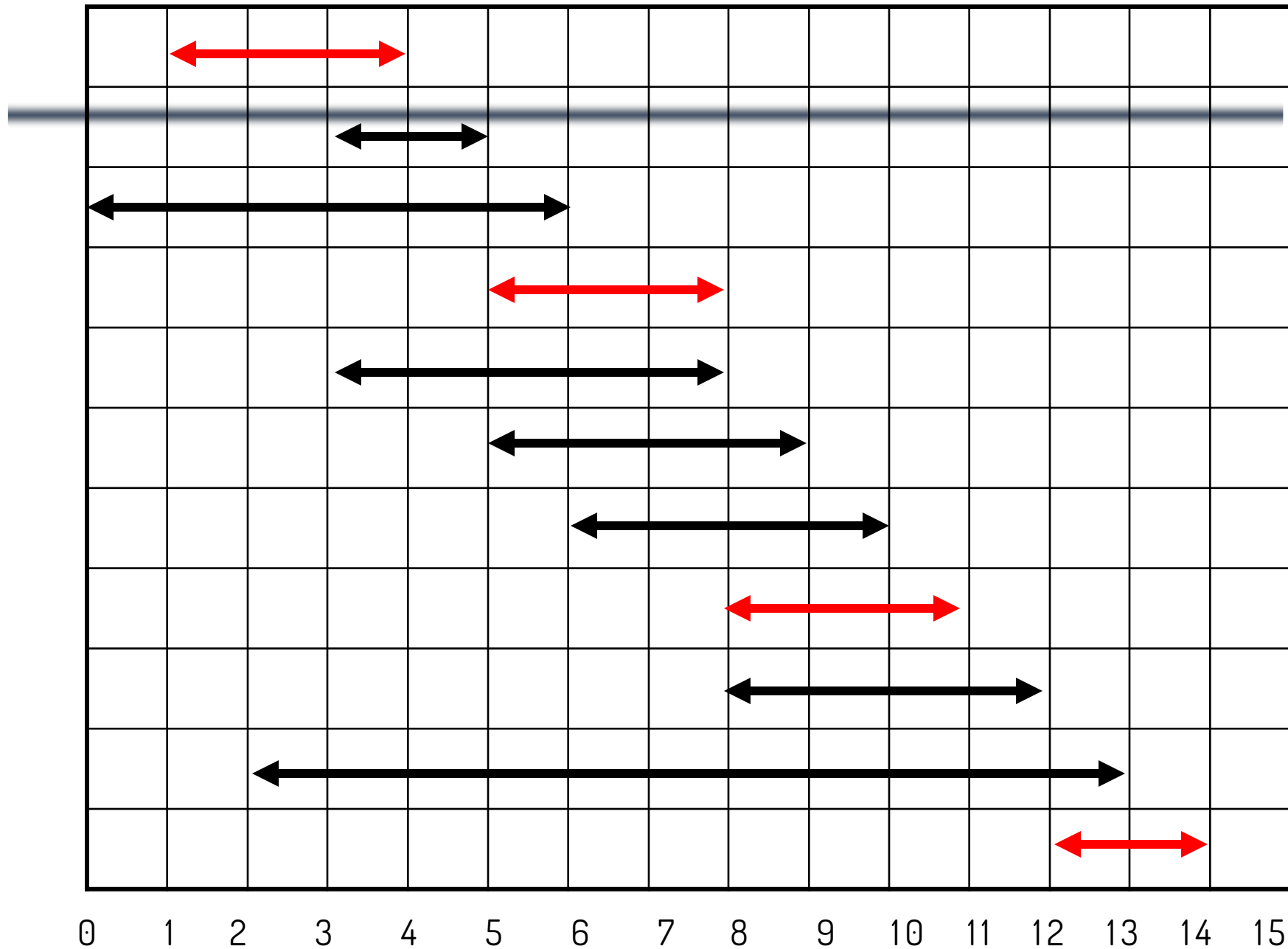
Interval Representation

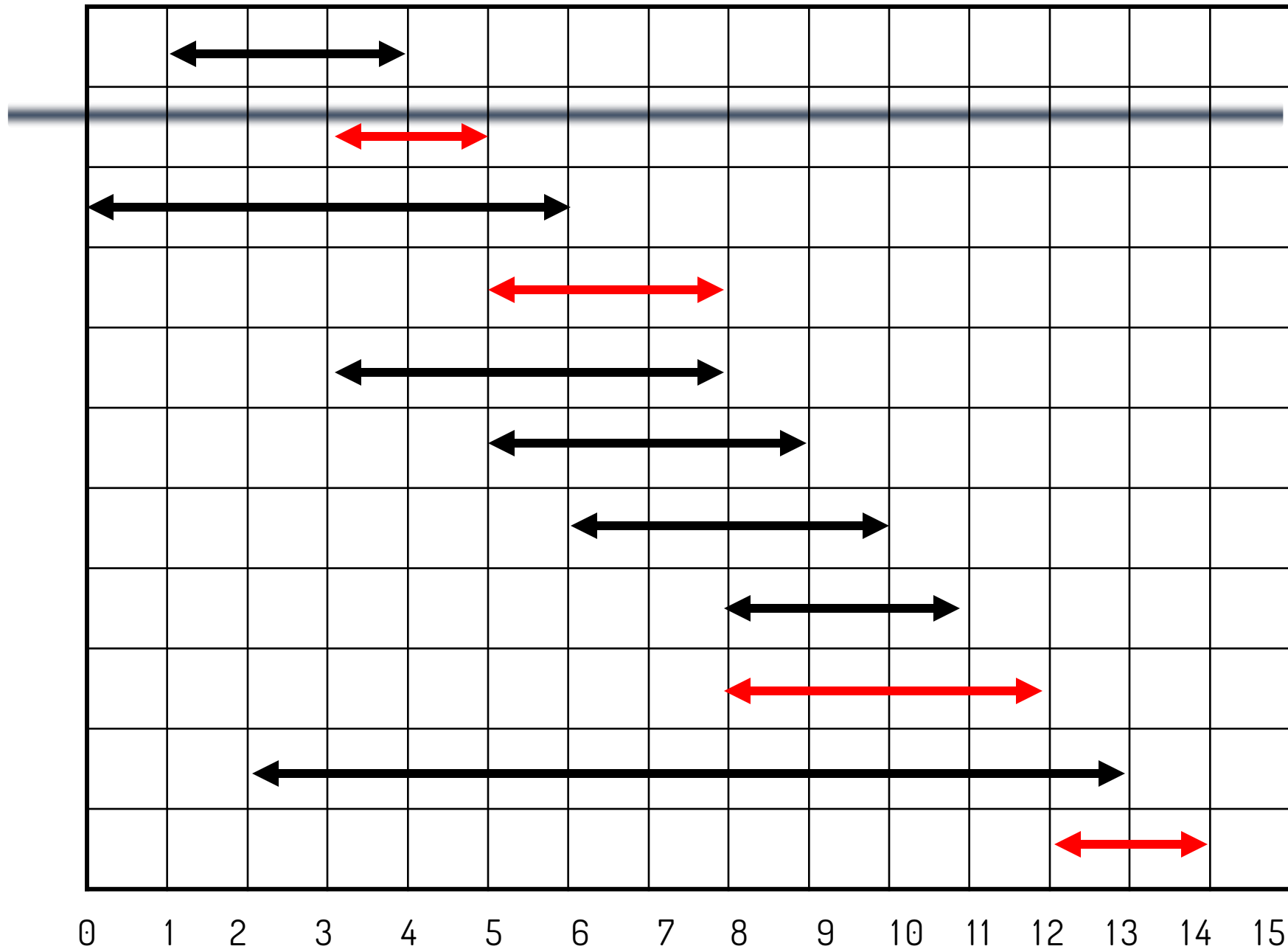
i	1	2	3	4	5	6	7	8	9	10	11
s_i	1	3	0	5	3	5	6	8	8	2	12
f_i	4	5	6	7	8	9	10	11	12	13	14





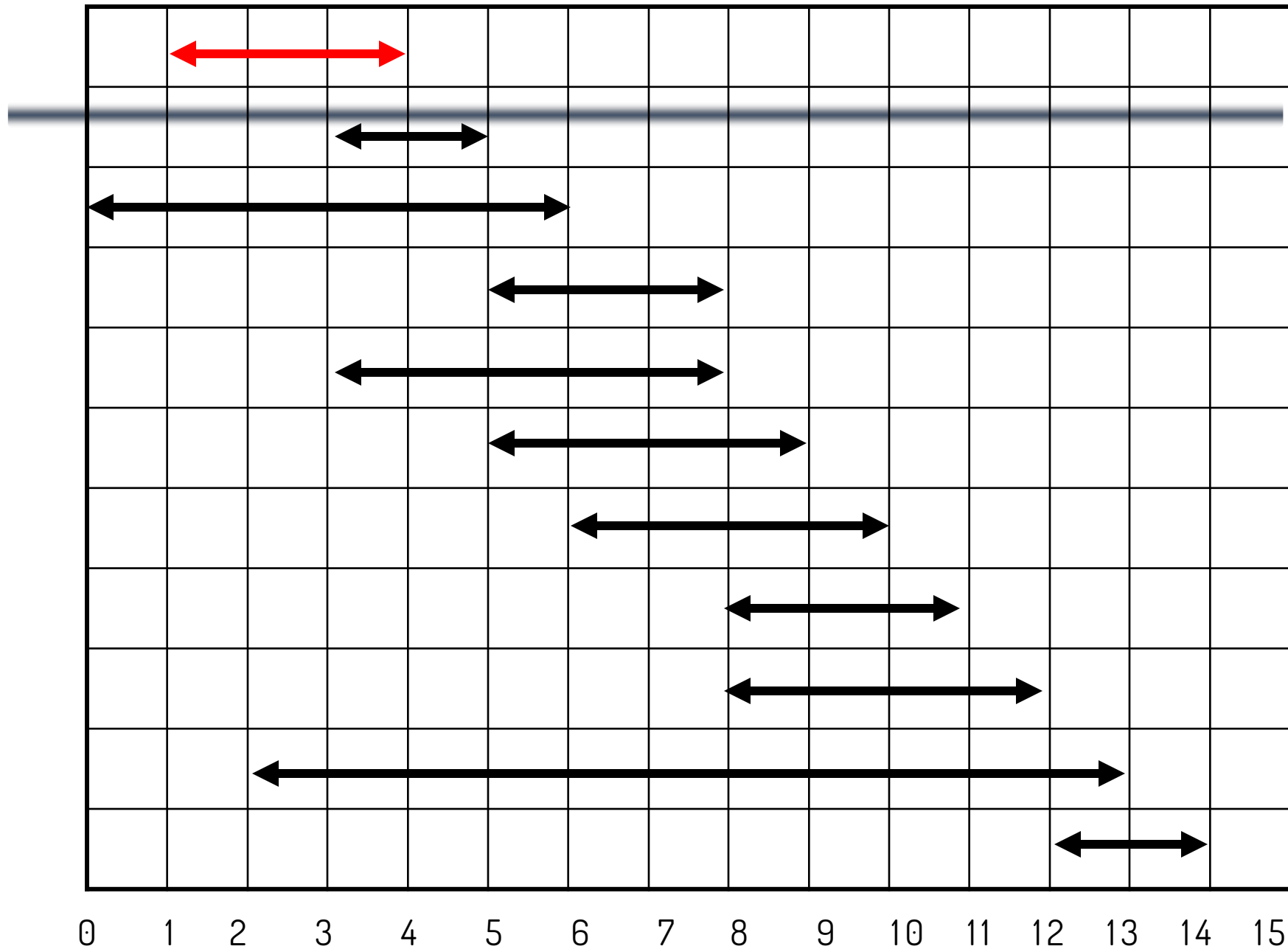


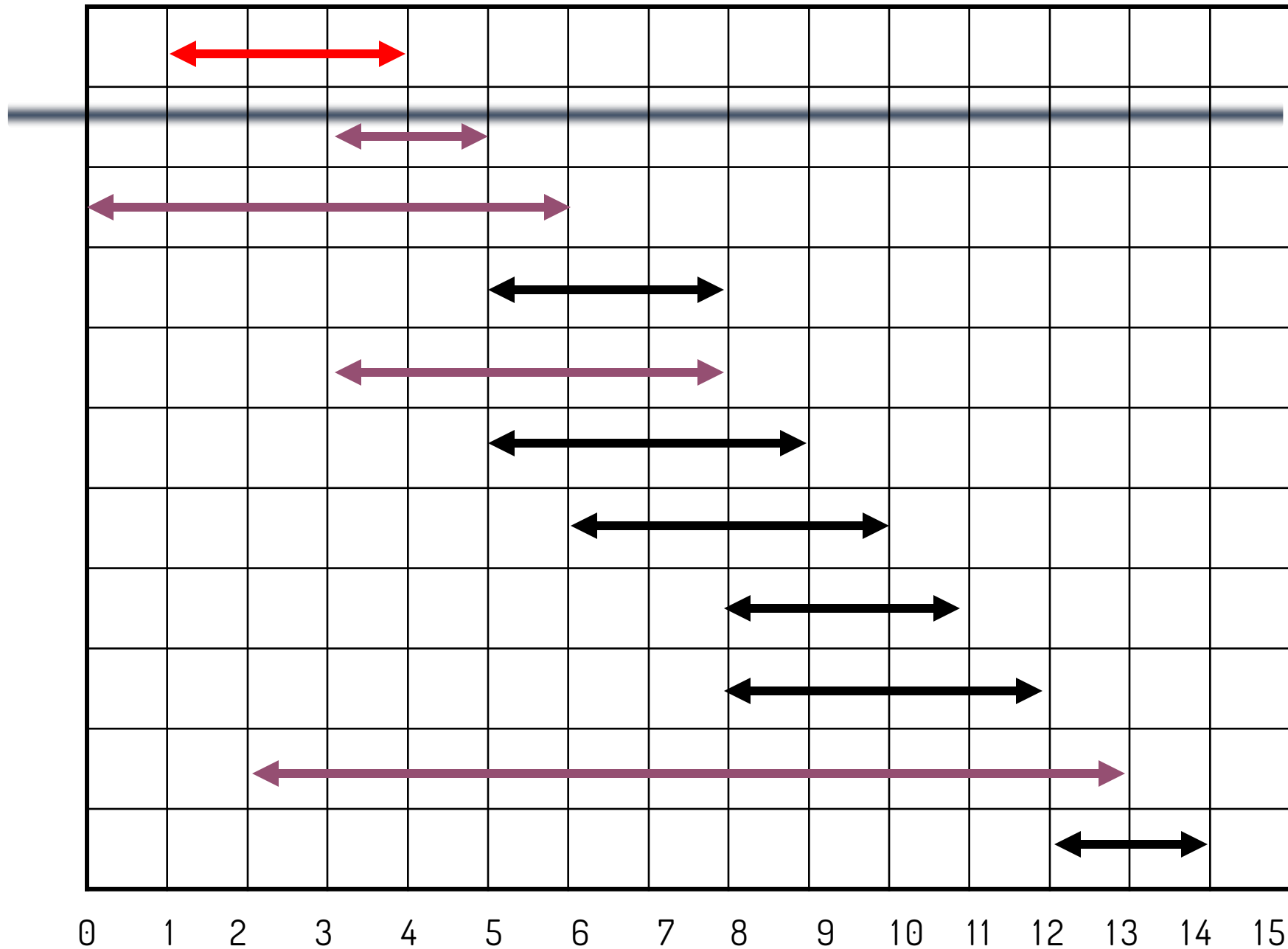


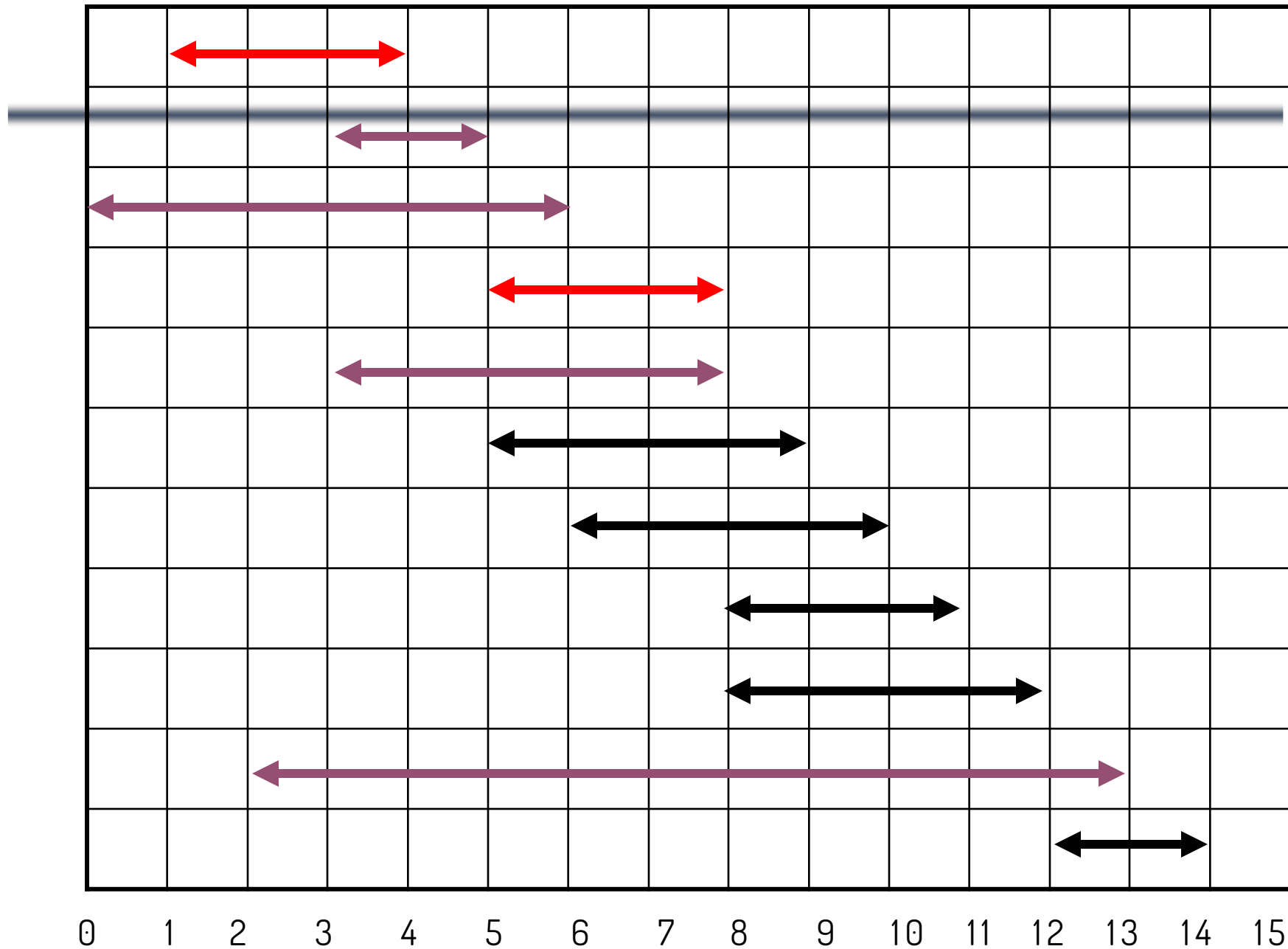


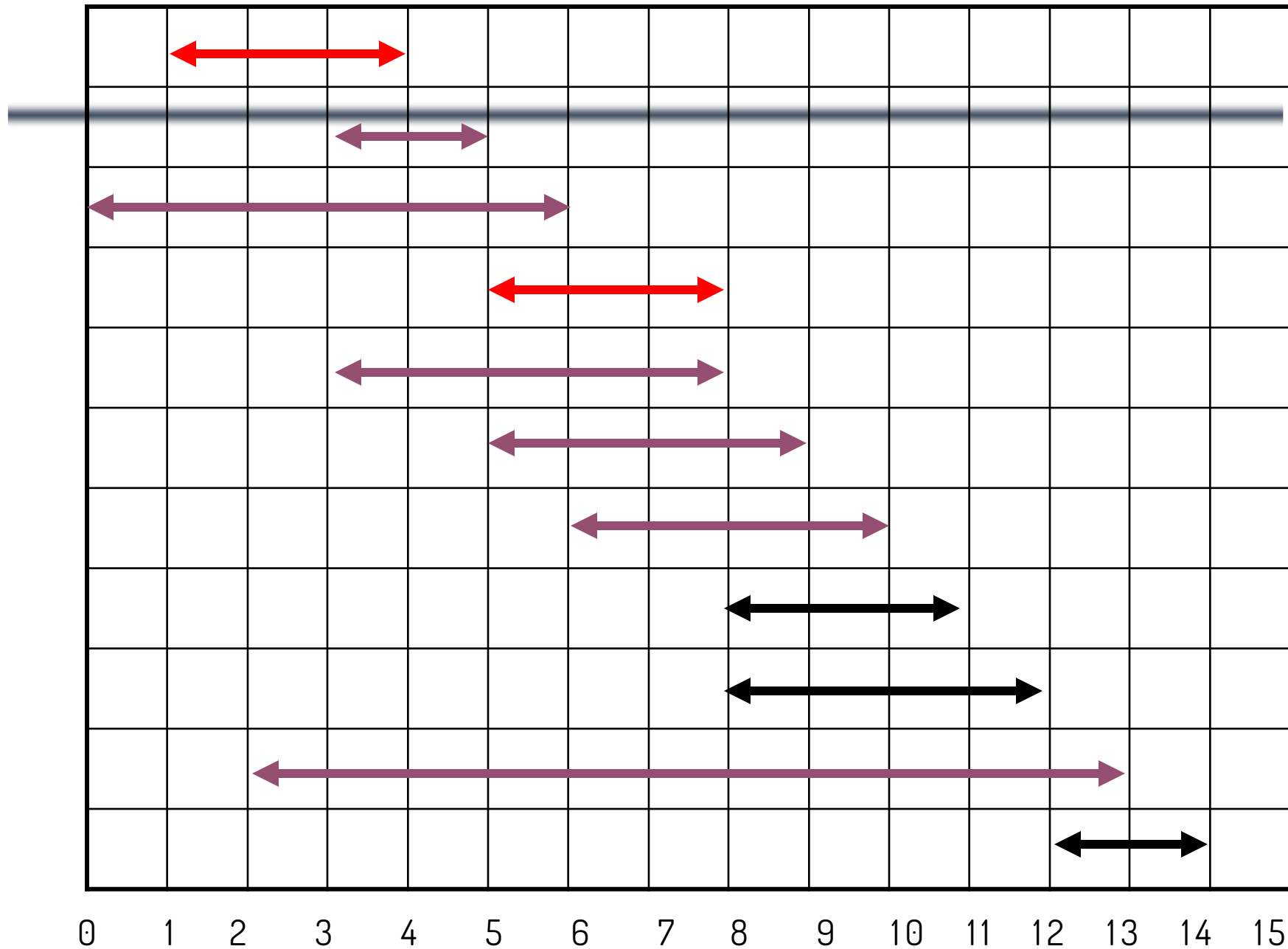
Early Finish Greedy

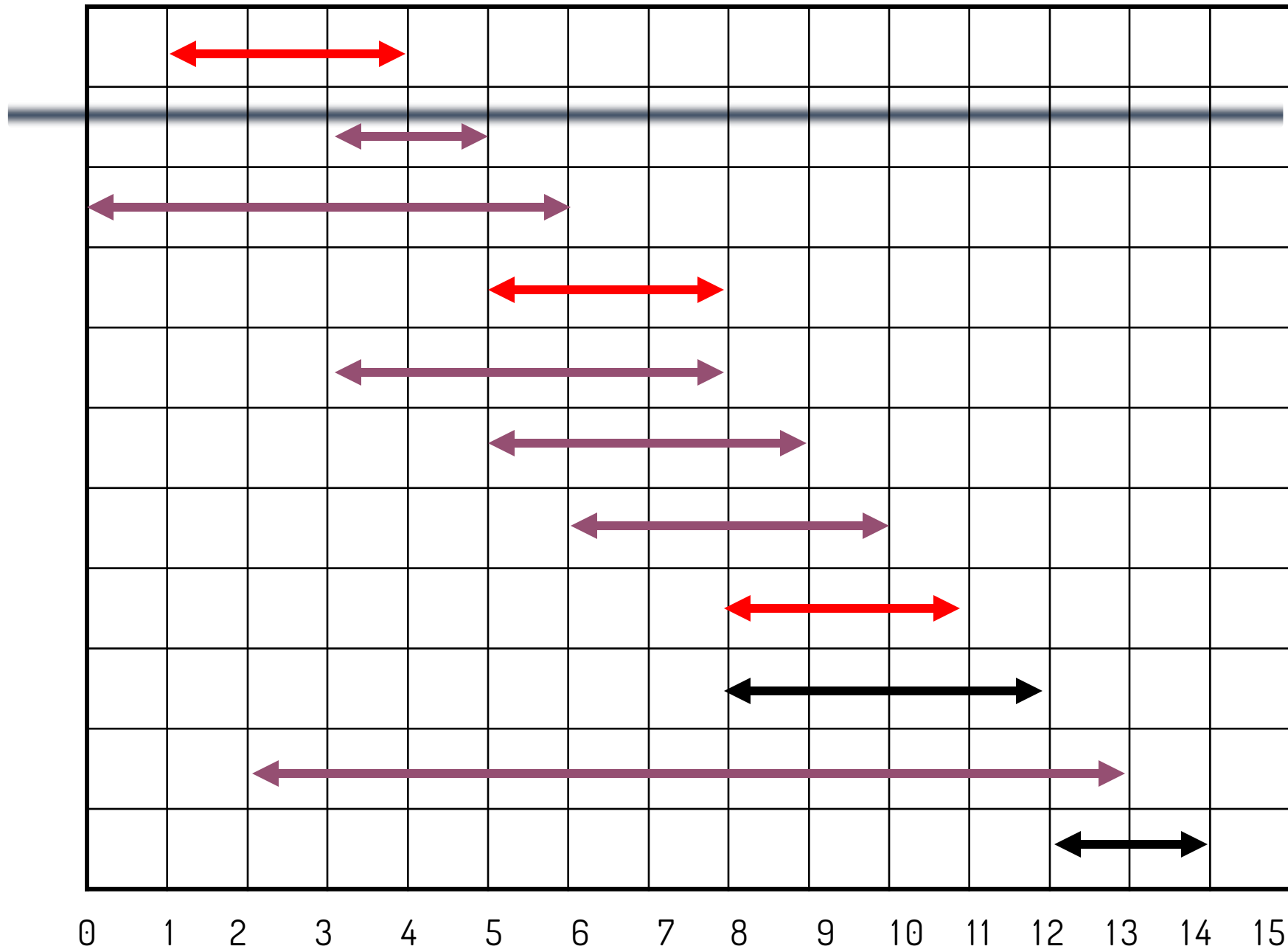
- Select the activity with the earliest finish
- Eliminate the activities that could not be scheduled
- Repeat!

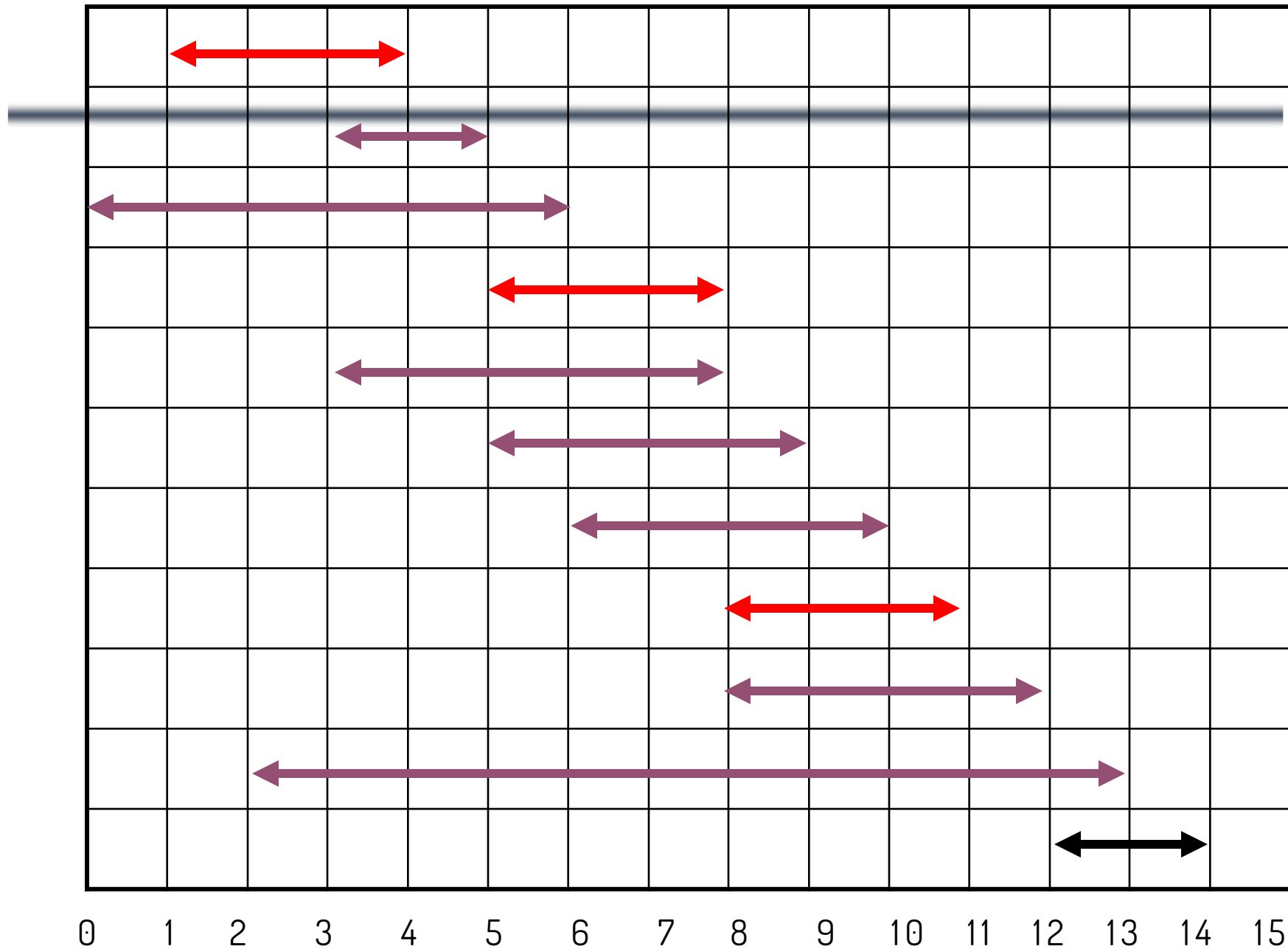


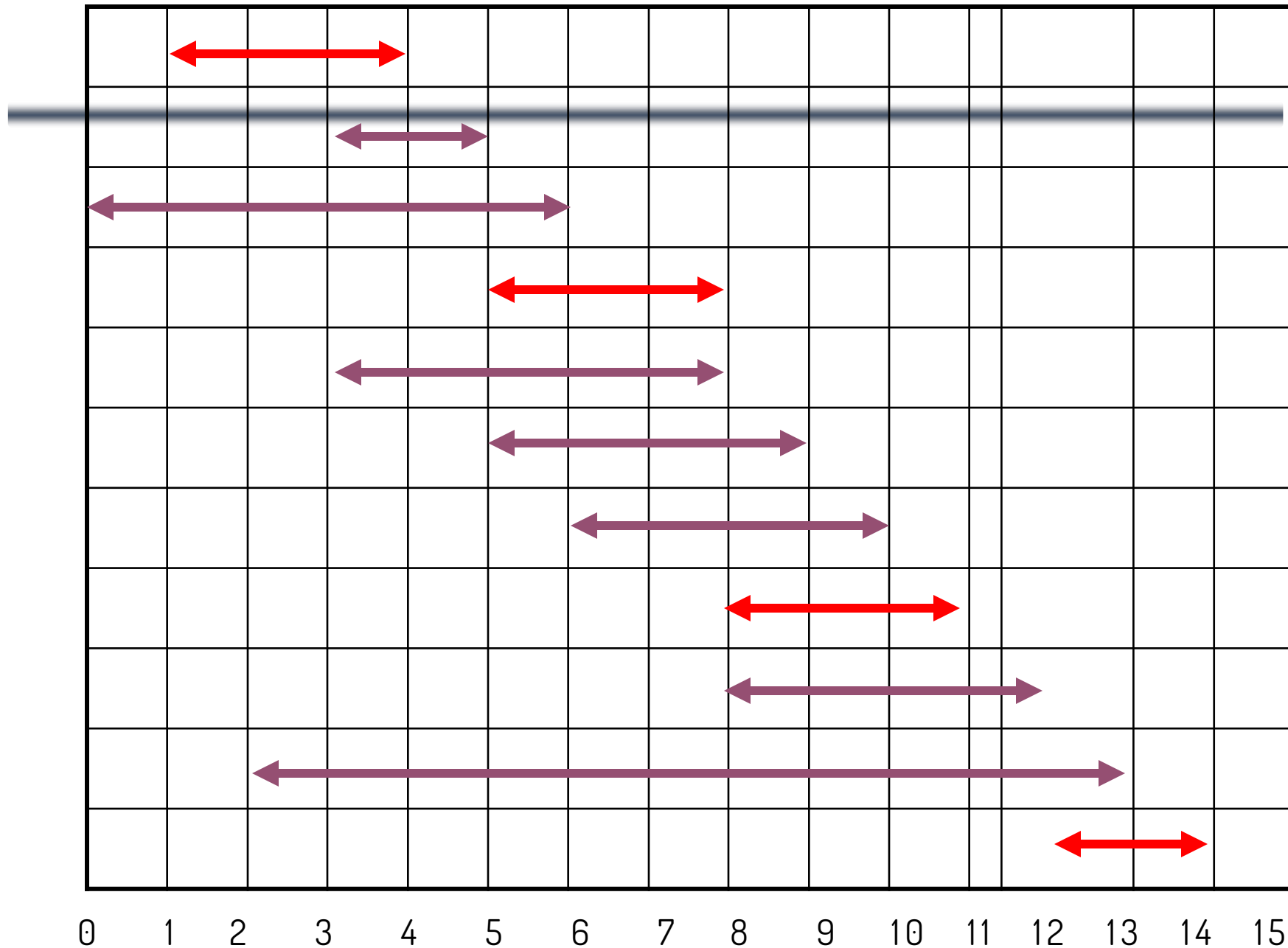












Assuming activities are sorted by finish time

GREEDY-ACTIVITY-SELECTOR(s, f)

```
1   $n \leftarrow \text{length}[s]$ 
2   $A \leftarrow \{a_1\}$ 
3   $i \leftarrow 1$ 
4  for  $m \leftarrow 2$  to  $n$ 
5      do if  $s_m \geq f_i$ 
6          then  $A \leftarrow A \cup \{a_m\}$ 
7               $i \leftarrow m$ 
8  return  $A$ 
```

Why it is Greedy?

- Greedy in the sense that it leaves as much opportunity as possible for the remaining activities to be scheduled
- The greedy choice is the one that maximizes the amount of unscheduled time remaining

Huffman Codes

- Widely used technique for data compression
- Assume the data to be a sequence of characters
- Looking for an effective way of storing the data
- ***Binary character code***
 - Uniquely represents a character by a binary string
- Lossless Compression Technique.
- https://www.youtube.com/watch?v=co4_ahEDCho

Fixed-Length Codes

E.g.: *Lets say*, Data file containing 100,000 characters

	a	b	c	d	e	f
Frequency (thousands)	45	13	12	16	9	5

- 3 bits needed
- $a = 000$, $b = 001$, $c = 010$, $d = 011$, $e = 100$, $f = 101$
- Requires: $100,000 \cdot 3 = 300,000$ bits

Huffman Codes

- Idea:
 - Use the frequencies of occurrence of characters to build a optimal way of representing each character

	a	b	c	d	e	f
Frequency (thousands)	45	13	12	16	9	5

Variable-Length Codes

E.g.: Data file containing 100,000 characters

	a	b	c	d	e	f
Frequency (thousands)	45	13	12	16	9	5

- Assign short codewords to frequent characters and long codewords to infrequent characters
- $a = 0, b = 101, c = 100, d = 111, e = 1101, f = 1100$
- $(45 \cdot 1 + 13 \cdot 3 + 12 \cdot 3 + 16 \cdot 3 + 9 \cdot 4 + 5 \cdot 4) \cdot 1,000$
 $= 224,000 \text{ bits}$

Encoding with Binary Character Codes

- Encoding
 - Concatenate the codewords representing each character in the file
- *E.g.:*
 - $a = 0, b = 101, c = 100, d = 111, e = 1101, f = 1100$
 - $abc = 0 \cdot 101 \cdot 100 = 0101100$

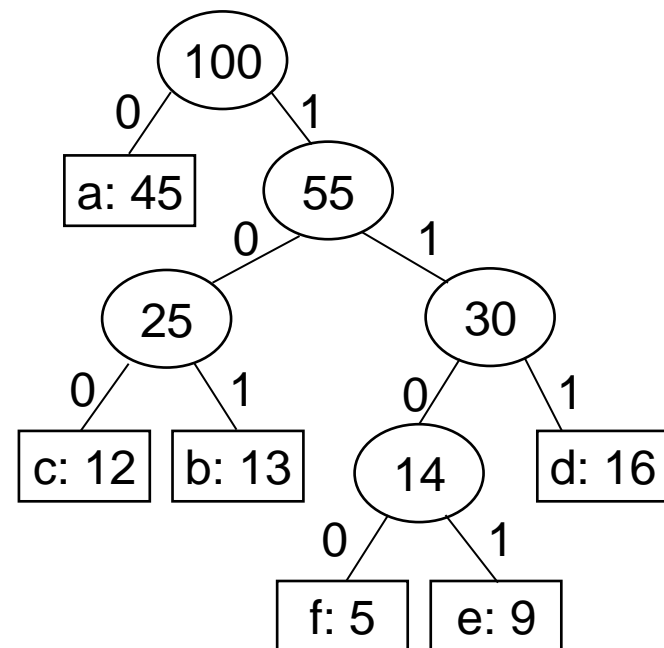
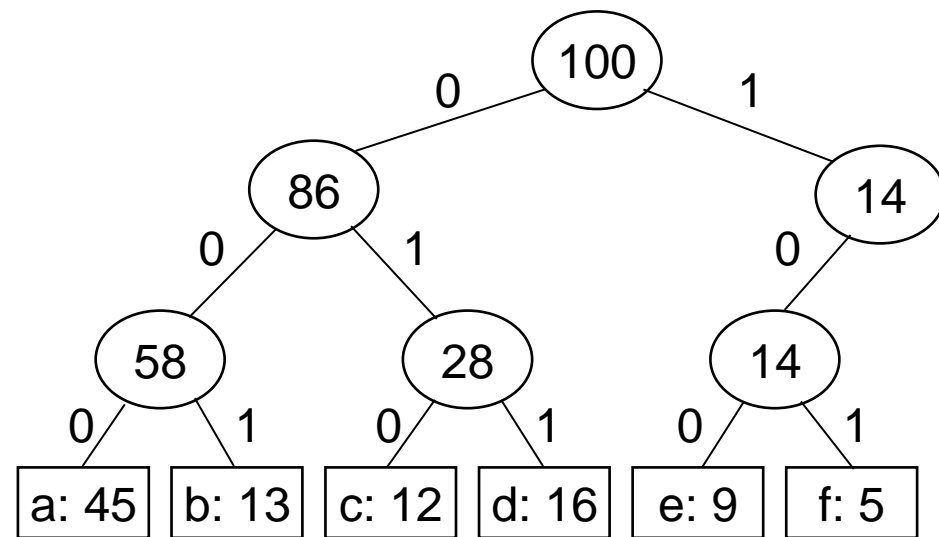
Decoding with Binary Character Codes

- Prefix codes simplify decoding
 - No codeword is a prefix of another \Rightarrow the codeword that begins an encoded file is unambiguous
- Approach
 - Identify the initial codeword
 - Translate it back to the original character
 - Repeat the process on the remainder of the file
- *E.g.:*
 - $a = 0, b = 101, c = 100, d = 111, e = 1101, f = 1100$
 - $001011101 =$

$0 \cdot 0 \cdot 101 \cdot 1101 = \text{aabe}$

Prefix Code Representation

- Binary tree whose leaves are the given characters
- Binary codeword
 - the path from the root to the character, where 0 means “go to the left child” and 1 means “go to the right child”
- Length of the codeword
 - Length of the path from root to the character leaf (depth of node)



Example

f: 5 e: 9 c: 12 b: 13 d: 16 a: 45

