



# Algorithms Theory

## 13 – Bin Packing

P.D. Dr. Alexander Souza

# Bin packing

1. Problem definition and general observations
2. Approximation algorithms for the online bin packing problem
3. Approximation algorithms for the offline bin packing problem

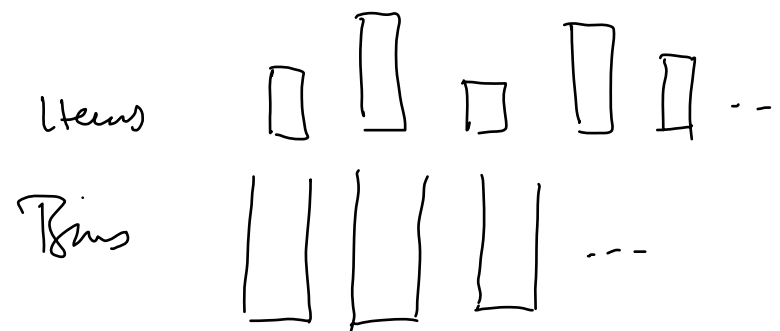
# Problem definition

## Given:

$n$  items with sizes

$s_1, \dots, s_n$

where  $0 < s_i \leq 1$  for  $1 \leq i \leq n$ .



## Goal:

Pack items into a minimum number of unit-capacity bins.

## Example:

7 items with sizes 0.2, 0.5, 0.4, 0.7, 0.1, 0.3, 0.8

# Problem definition

## Online bin packing:

Items arrive one by one. Each item must be assigned immediately to a bin, without knowledge of any future items. Reassignment is not allowed.

NEXT FIT  
FIRST FIT

## Offline bin packing:

All  $n$  items are known in advance, i.e. before they have to be packed.

FIRST FIT DECREASING

# Observations

- Bin packing is provably hard.  
(Offline bin packing is NP-hard.  
Decision problem is NP-complete.)
- There exists no online bin packing algorithm that always finds an optimal solution.

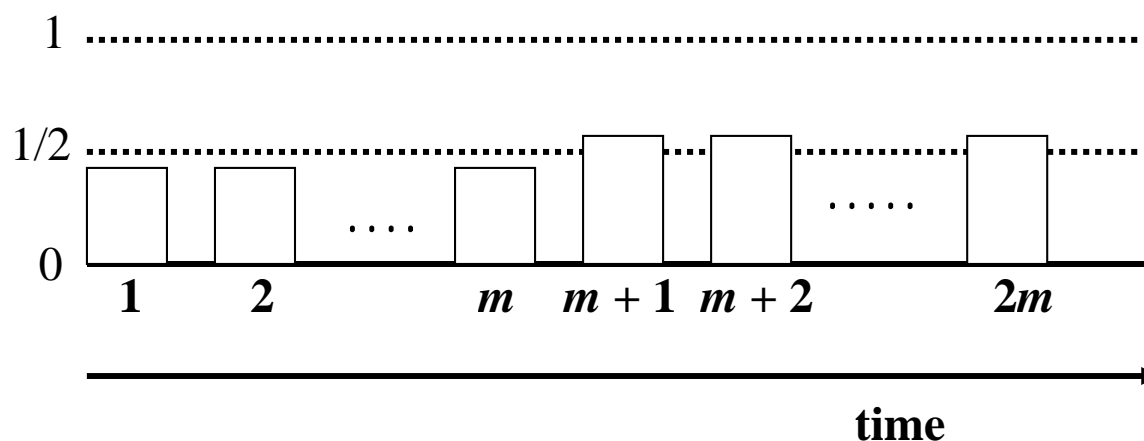
# Online bin packing

## Theorem 1:

There are inputs that force each online bin packing algorithm to use at least  $4/3 OPT$  bins where  $OPT$  is the minimum number of bins possible.

## Proof:

Assumption: online bin packing algorithm  $A$  always uses less than  $4/3 OPT$  bins



# Online bin packing

1st point of time:

$$OPT = m/2 \text{ and } \#bins(A) = b$$

$$\text{by assumption: } b < 4/3 \cdot m/2 = 2/3 m$$

Let  $b = b_1 + b_2$ , with

$$b_1 = \#bins \text{ containing one item}$$

$$b_2 = \#bins \text{ containing two items}$$

$$\text{There is: } b_1 + 2 b_2 = m, \text{ i.e. } b_1 = m - 2b_2$$

$$\text{Hence: } b = b_1 + b_2 = m - b_2 \quad (*)$$

# Online bin packing

2nd point of time:

$$OPT = m$$

$$\#bins(A) \geq b + m - b_1 = m + b_2$$

$$\text{Assumption: } m + b_2 \leq \#bins(A) < 4/3m$$

$$b_2 < m/3$$

$$\implies \text{ using (*): } b = m - b_2 > 2/3m$$



# Online bin packing

## Next Fit (NF), First Fit (FF), Best Fit (BF)

### Next Fit:

Assign an arriving item to the same bin as the preceding item. If it does not fit, open a new bin and place it there.

### Theorem 2:

(a) For all input sequences  $I$ :

$$NF(I) \leq 2 OPT(I).$$

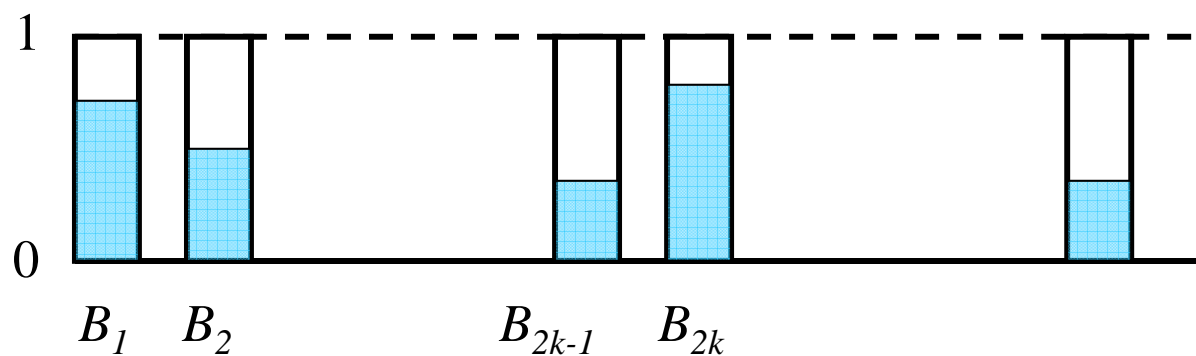
(b) There exist input sequences  $I$  such that:

$$NF(I) \geq 2 OPT(I) - 2.$$

# Next Fit

**Proof:** (a)

Consider two bins  $B_{2k-1}, B_{2k}$ ,  $2k \leq NF(I)$ .



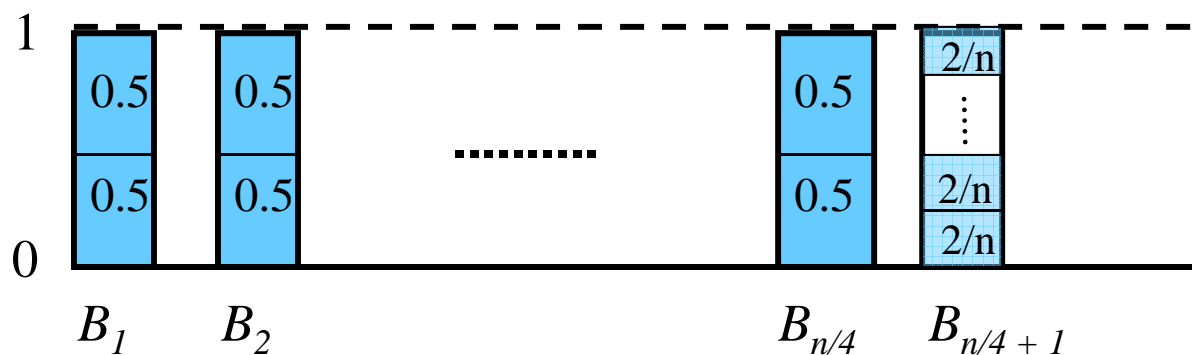
# Next Fit

**Proof:** (b)

Consider an input sequence  $I$  of length  $n$   
 ( $n \equiv 0 \pmod{4}$ ):

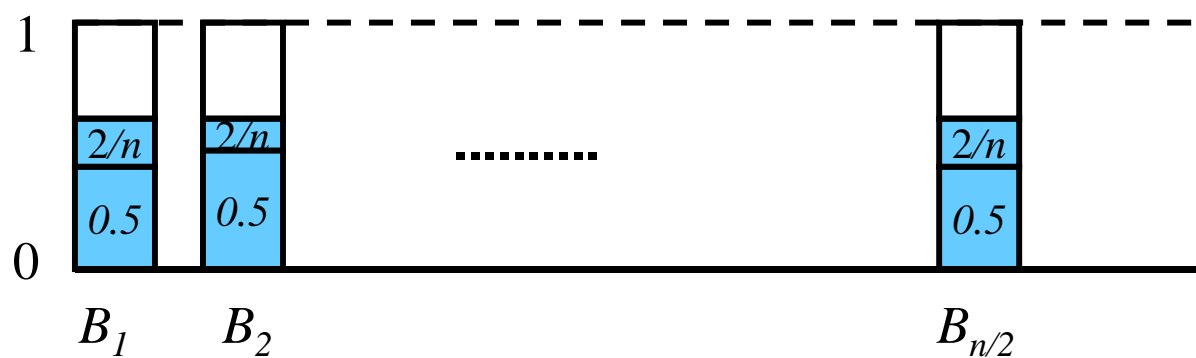
$0.5, 2/n, 0.5, 2/n, 0.5, \dots, 0.5, 2/n$

Optimal packing:



# Next Fit

Next Fit yields:



$$NF(I) =$$

$$OPT(I) =$$

# First Fit

## First Fit:

Assign an arriving item to the first bin (i.e. that was opened earliest) in which it fits. If there is no such bin, open a new one and place it there.

## Observation:

At each point in time there is at most one bin that is less than half full.

$$\rightarrow FF(I) \leq 2OPT(I)$$

# First Fit

## Theorem 3:

(a) For all input sequences  $I$ :

$$FF(I) \leq \lceil \frac{17}{10} OPT(I) \rceil$$

*involved*

(b) There exist input sequences  $I$  such that:

$$FF(I) \geq \frac{17}{10} (OPT(I) - 1)$$

(b') There exist input sequences  $I$  such that:

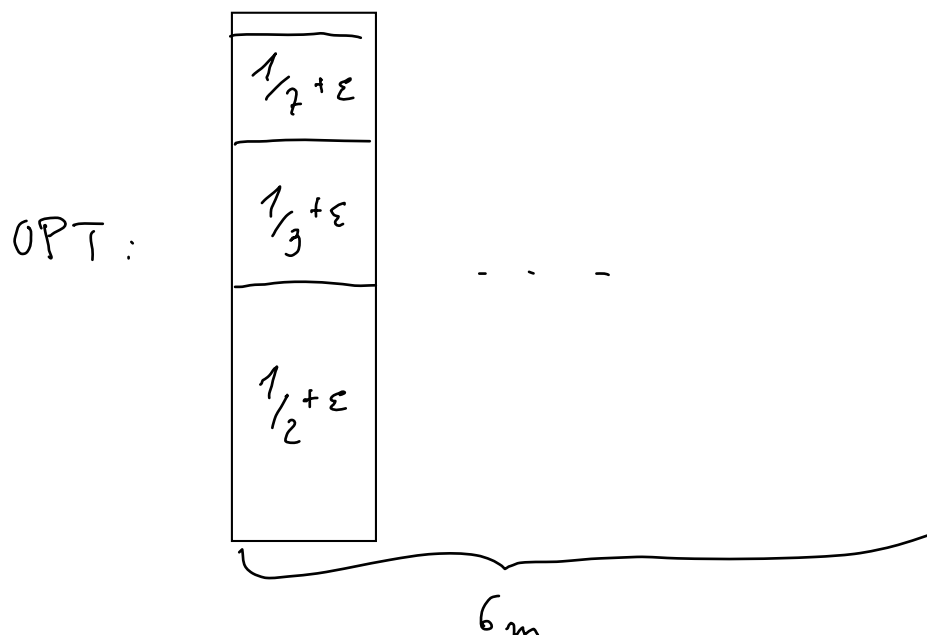
$$FF(I) = \frac{10}{6} OPT(I)$$

# First Fit

Proof (b`): Input sequence of length  $3 \cdot 6m$  :

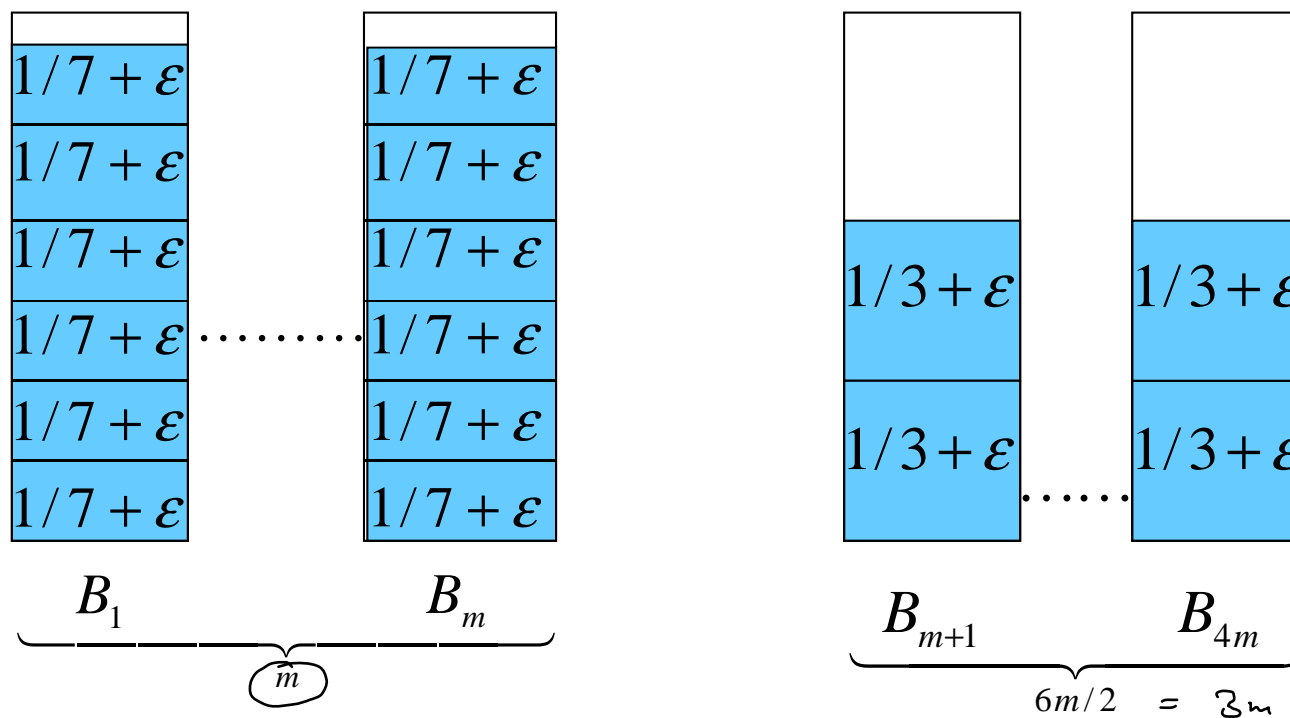
$$\underbrace{1/7 + \varepsilon, \dots, 1/7 + \varepsilon}_{6m}, \underbrace{1/3 + \varepsilon, \dots, 1/3 + \varepsilon}_{6m},$$

$$\underbrace{1/2 + \varepsilon, \dots, 1/2 + \varepsilon}_{6m}$$



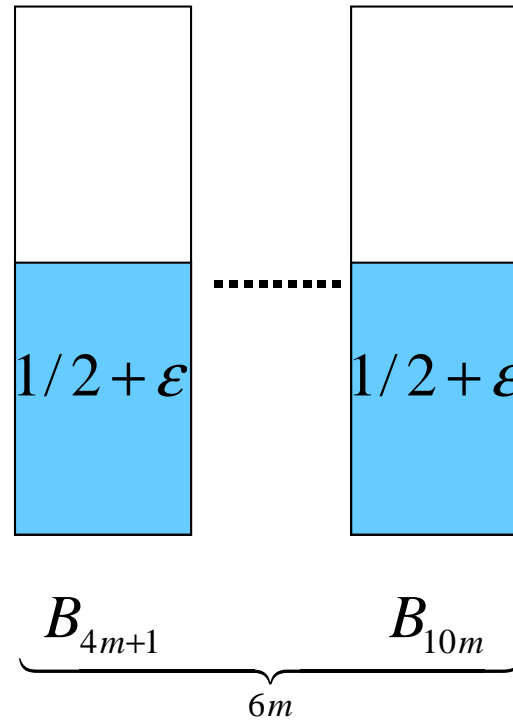
# First Fit

First Fit yields:





# First Fit



$$OPT = 6m$$

$$FF = m + 3m + 6m = 10m$$



# Best Fit

## Best Fit:

Assign an arriving item to the bin in which it fits best (i.e. where it leaves the smallest empty space).

Performance of BF and FF is similar.

Running times on input sequences of length  $n$  :

NF	$O(n)$		
FF	$O(n^2)$	→	$O(n \log n)$
BF	$O(n^2)$	→	$O(n \log n)$
	↑		↑
	naive		data structures

# Offline bin packing

Prior to the packing,  $n$  and  $s_1, \dots, s_n$  are known in advance.

An optimal packing can be found by exhaustive search. *exponential time*

## **Approach to an offline approximation algorithm:**

Initially sort the items in decreasing order of size and assign the larger items first!

**First Fit Decreasing (FFD)** resp. **FFNI**

**Best Fit Decreasing (BFD)**

# First Fit Decreasing

Theorem  $FFD(I) \leq \frac{4}{3} \cdot OPT(I) + \frac{1}{3}$

## Lemma 1:

Let  $I$  be an input sequence of  $n$  objects with sizes

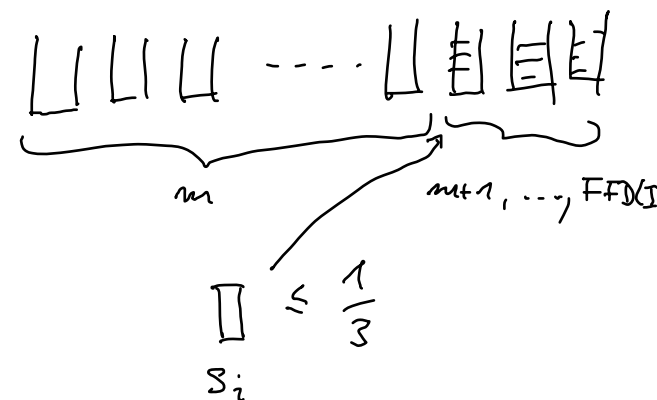
$$s_1 \geq s_2 \geq \dots \geq s_n$$

and let  $m = OPT(I)$ .

Then, all items placed by FFD into bins

$$B_{m+1}, B_{m+2}, \dots, B_{FFD(I)}$$

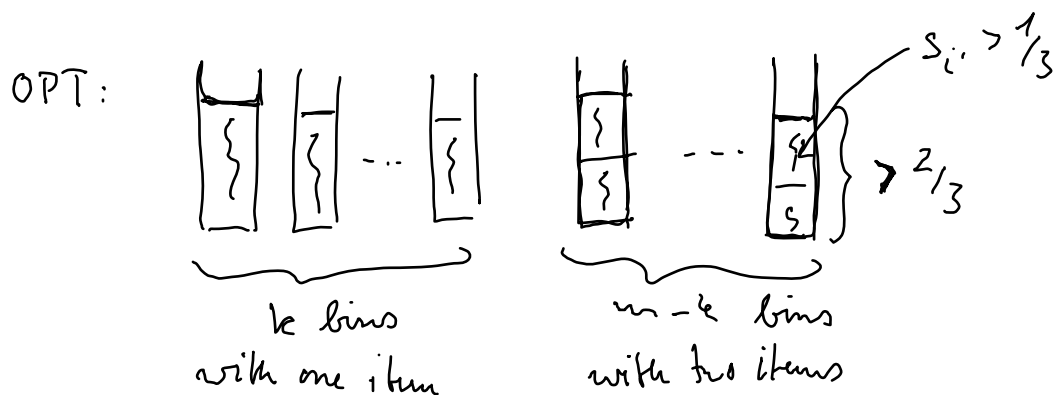
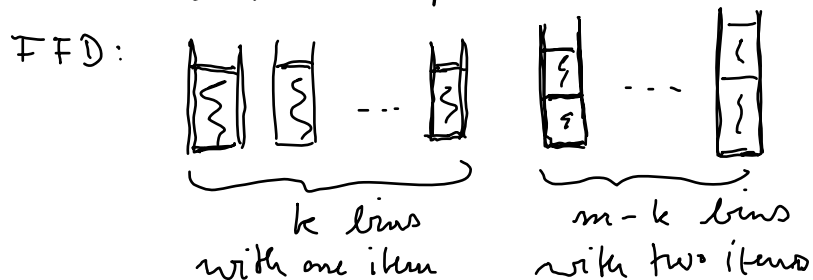
are of size at most  $\frac{1}{3}$ .



# First Fit Decreasing

**Proof:** Let  $s_i$  be the first item which is placed into bin  $m+1$ .  
 We will show  $s_i \leq 1/3$  implying  $s_{i+1}, \dots, s_m \leq 1/3$ .

Assume:  $s_i > 1/3$   
 when FFD places  $i$  into bin  $m+1$  the first  $m$  bins have the following configuration



Item  $i$  can not be packed into any of the first  $k$  bins (FFD has proved that this is not possible).  
 And item  $i$  can not be packed into any of the last  $m-k$  bins because these bins are packed to an extent of  $> 2/3$  and the size  $s_i > 1/3$ .

$$\text{OPT}(I) > m \lfloor \frac{1}{3} \rfloor$$

# First Fit Decreasing

## Lemma 2:

Let  $I$  be an input sequence of  $n$  objects with sizes

$$s_1 \geq s_2 \geq \dots \geq s_n$$

and let  $m = OPT(I)$ .

Then the number of items placed by FFD into bins

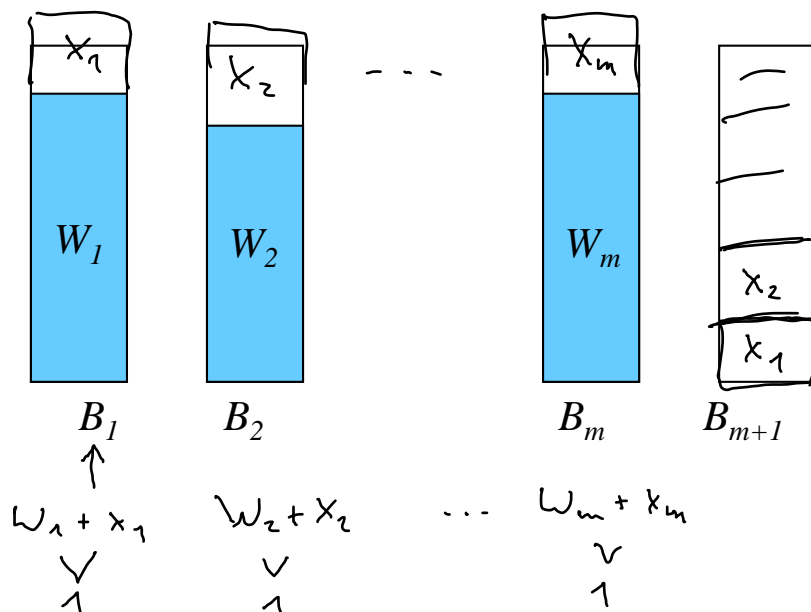
$$\underline{B_{m+1}}, \underline{B_{m+2}}, \dots, \underline{B_{FFD(I)}}$$

is at most  $m - 1$ .

# First Fit Decreasing

## Proof:

Assumption: FFD places more than  $m - 1$  items, say  $x_1, \dots, x_m$ , into extra bins.



$$OPT \geq \sum_{i=1}^n s_i \geq \sum_{j=1}^m (W_j + x_j) > m \quad \text{⚡}$$

# First Fit Decreasing

## Theorem:

For all input sequences  $I$ :

$$FFD(I) \leq (4 \text{OPT}(I) + 1) / 3. = \frac{4}{3} \cdot \text{OPT}(I) + \frac{1}{3}$$

Proof  $FFD(I) \leq m + \lceil \frac{m-1}{3} \rceil \leq m + \frac{m-1}{3} + \frac{2}{3} = \frac{4}{3} \cdot m + \frac{1}{3} = \frac{4}{3} \cdot \text{OPT}(I) + \frac{1}{3}$   $\square$

## Theorem:

1. For all input sequences  $I$ :

$$FFD(I) \leq \textcircled{11/9} \text{OPT}(I) + 4. \quad \text{difficult}$$

2. There exist input sequences  $I$  such that:

$$FFD(I) = 11/9 \text{OPT}(I).$$

$OPT = m$

Lemma 1+2:

At most  $m-1$  items each having size at most  $1/3$  are packed into extra bins by FFD.

Thus at most  $\lceil \frac{m-1}{3} \rceil$  extra bins





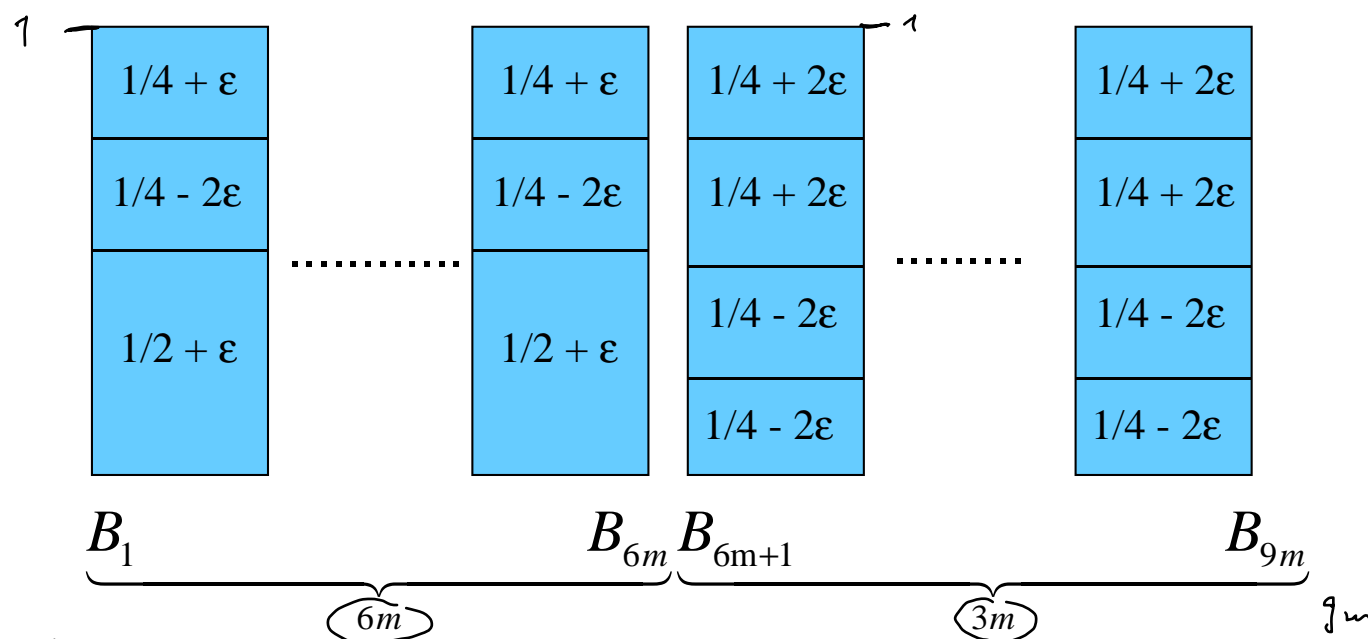
# First Fit Decreasing

**Proof (b):** Input sequence of length  $3 \cdot 6m + 12m$ :

4 types

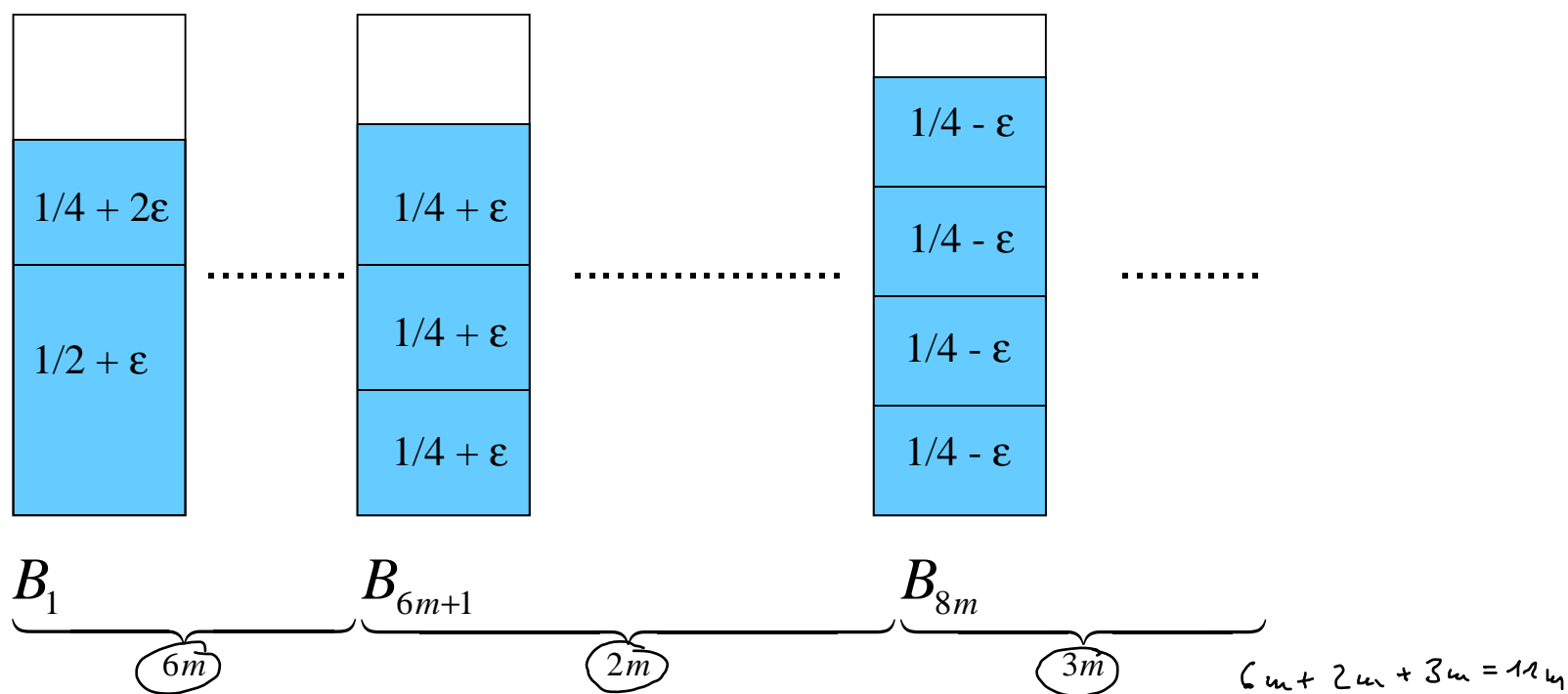
$$\left. \begin{array}{l} \underbrace{1/2 + \varepsilon, \dots, 1/2 + \varepsilon}_{6m}, \underbrace{1/4 + 2\varepsilon, \dots, 1/4 + 2\varepsilon}_{6m} \\ \underbrace{1/4 + \varepsilon, \dots, 1/4 + \varepsilon}_{6m}, \underbrace{1/4 - 2\varepsilon, \dots, 1/4 - 2\varepsilon}_{12m} \end{array} \right\} \text{in decreasing order of size}$$

Optimal packing:



# First Fit Decreasing

*First Fit Decreasing* yields:



$$OPT(I) = 9m$$

$$FFD(I) = 11m$$