### Packing Algorithms

An instance of a packing problem consists of:

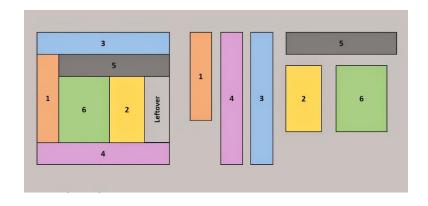
- 1. Items (associated with sizes, weights, profits).
- 2. Bins with limited capacity.
- 3. A set of constraints.

General Goal: Place items in the bin, items must not overlap with each other. Bins may not be filled beyond their capacity.

### Packing Problems

#### Variants:

- Multi-dimensional items.
- Limited number of bins pack as much as you can.
- Unlimited number of bins pack all items using minimal number of bins.
- Conflicting items cannot be placed together.
- · Cardinality constraints.
- · Cutting stock (minimizing wasted material)
- Class-constraints.
- · Online vs. offline
- · Variable bins.
- Many more...



### Popular Applications

- Physical items → boxes, tracks (shipping, delivery).
- Files → disks.
- Advertisements → commercial break/ magazines / web-pages.
- Orders → limited amount of material
- Jobs → processors.



### The Knapsack problem

- You are about to go to a camp.
- · There are many items you want to take.
- You have one knapsack. The total weight you can carry is at most W.
- Item i in your list has weight  $w_i$ , and value (benefit)  $b_i$ , that measures how much you really need it.
- You need to pack the knapsack in a way that maximizes the total value of the packed items.

### The Knapsack problem

Item#	Weight	Value		
1	1	8		
2	3	6		Max
3	5	5		weight
4	4	6		=8
			The state of the s	<u>-</u>

A possible packing: Items 2 and 3. Value: 11

An optimal packing: Items 1,2,4. Value: 20

The Knapsack problem is NP-hard.

## Greedy Algorithm for Knapsack

- 1. Consider the items in order of non-increasing  $b_i/w_i$  ratio  $b_1/w_1 \ge b_2/w_2 \ge ... \ge b_n/w_n$
- 2. Add items to the knapsack as long as there is space.

### Time Complexity:

O(n log n) (for sorting)
O(n) for packing loop.

 $\rightarrow$  O(n log n)

## Greedy Algorithm for Knapsack

Claim: The approximation ratio of Greedy is not bounded.

Proof: To get ratio c, consider the following instance:

There are two items: The knapsack has  $b_1=2$ ,  $w_1=1$  volume W=2c  $b_2=2c$ ,  $w_2=2c$ 

Greedy packs only the first item, value = 2. Optimal: Pack the second item, value=2c Ratio = c.

### Improved Algorithm for Knapsack

Take the maximum of Greedy and the most valuable item that fits by itself.

Theorem: The above algorithm is 2-approximation.

Proof: We assume w.l.o.g that no single item has weight more than W (these items can be removed in a preprocessing).

Sort the items such that  $b_1/w_1 \ge b_2/w_2 \ge ... \ge b_n/w_n$ .

Let B be the largest value of an item, and let G be the value computed by the greedy algorithm.

Let j be the first item that the greedy algorithm rejects.

## Improved Algorithm for Knapsack

$$ALG = max(B, G) \ge (B + G)/2$$

$$G = \sum_{i=1}^{j-1} b_i$$

(item j is the first to be rejected)

$$B \ge b_j$$

(B is the most profitable)

$$G+B \ge \sum_{i=1}^{J} b_i$$

opt 
$$\leftarrow \sum_{i=1}^{J} b_i$$

Because the first j items have the largest 'profit density'

$$\rightarrow$$
 ALG > opt/2

### Exact Solution - dynamic programming

Define a table M of size  $(n+1)\times(W+1)$ , where the (i, x) entry corresponds to the maximal profit that can be obtained from the first i items and a knapsack having capacity x.

The solution to the knapsack problem lies in M(n,W).

#### Base cases:

If i = 0, then there are no items to pack: M(0, x) = 0. If x < 0 then  $M(i, x) = -\infty$ .

### Exact Solution - dynamic programming

#### The DP recursion:

$$M(i, x) = \max \{ M(i-1, x), M(i-1, x-w_i) + b_i \}.$$

We take the maximum of two options:

- 1. M(i-1, x): not packing the i-th item.
- 2.  $M(i-1, x-w_i) + b_i$ : packing the i-th item.

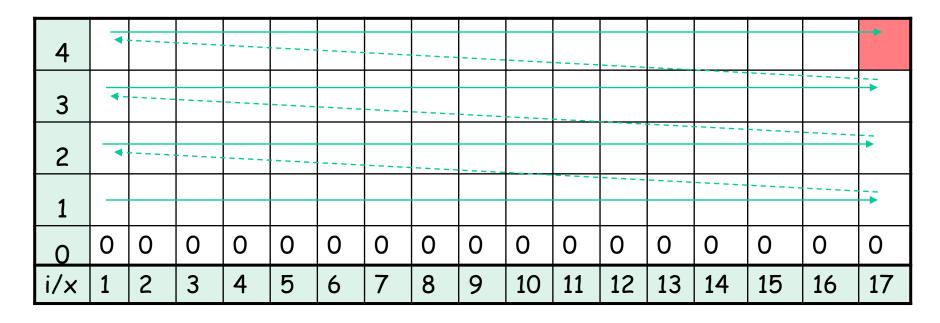
n						* -
•••						
i					<b>→</b> ↑	
i-1						
0	0	0	0	0	 0	 0
	1	2	3	4	 ×	 W

The solution lies here

## Knapsack DP Example

Assume n=4 and W=17.

Weights: {2, 4, 7, 10} Values: {3, 7, 9, 16}



In class exercise - complete the table

### Knapsack DP Example

Assume n=4 and W=17.

Weights: {2, 4, 7, 10} Values: {3, 7, 9, 16}

4	0	3	3	7	7	10	10	10	12	16	16	19	19	23	23	26	26
3	0	3	3	7	7	10	10	10	12	12	12	12	19	19	19	19	19
2	0	3	3	7	7	10	10	10	10	10	10	10	10	10	10	10	10
1	0	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
i/x	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17

$$M(3,9) = max\{M(2,9), M(2,2)+9\} = 3+9=12$$
  
 $M(4,12) = max\{M(3,12), M(3,2)+16\}=3+16=19.$ 

### The Bin Packing Problem

- Input: Items of sizes 0 < s<sub>i</sub> < 1
- Output: A feasible packing in bins of size 1
- Goal: minimize number of bins used.

### Example:

Input:

0.45 0.3 0.2 0.45

0.25 0.3 0.2 0.7

A packing in 3 bins:

0.7 0.25

0.45 0.45

0.3 0.3 0.2 0.2

### Next-fit Algorithm:

- 1. Open an active bin.
- 2. For all i=1,2,...,n:
  - If possible, place a<sub>i</sub> in the current active bin;
  - Otherwise, open a new active bin and place a; in it.

```
Example: The input: {0.3, 0.9, 0.2}. Next-fit packing (three bins): (0.3), (0.9), (0.2).
```

Theorem: Next-fit is 2-approximation to BP Proof: An optimal algorithm must use at least  $\Sigma_i a_i$  bins (why?).

Analysis of Next Fit (cont'): Assume that Next Fit uses h bins. The sum of items sizes in two consecutive bins is greater than 1 (otherwise, we can put them together).

Case 1: h is even: 
$$c(B_1) + c(B_2) > 1$$
  $c(B_1) + c(B_2) > 1$   $c(B_3) + c(B_4) > 1$   $c(B_3) + c(B_4) > 1$   $c(B_{h-1}) + c(B_h) > 1$   $c(B_{h-2}) + c(B_{h-1}) > 1$   $c(B_{h-2}) + c(B_h) > 1$   $c(B_h) > 1$ 

In both cases, we can obtain  $h \leq \lceil 2\Sigma_i a_i \rceil \leq 2opt$ 

Remark: it can be shown that  $h \le 2opt-1$ 

- Is the analysis tight? consider an instance with 4n items {1/2, 1/2n, 1/2, 1/2n, ...}.
- Next-fit will put any two consecutive items in a bin.
- Total number of bin used: 2n.
- An optimal packing in n+1 bins: n bins, each with 1/2+1/2, one bin for the tiny items.
- The ratio:  $2n/(n+1) \rightarrow 2$  as n grows.

First fit algorithm: place the next item in the first open bin that can accommodate it. Open a new bin only if no open bin has enough room.

Theorem:  $h_{ff} \le 1.7$ opt +2 (proof not here)

First fit Decreasing: sort the items from largest to smallest. Run FF according to the resulting order.

Theorem:  $h_{ffd} \le 1.222$ opt + 3 (proof not here)

■ No additive-error approximation is known for bin packing. That is, the best known is  $(1+\delta)$ opt.

# Unit Fractions Bin Packing

- A Unit Fraction: A fraction of the form 1/i for an integer i.
- Input: integers  $w_1, w_2, ..., w_n$ .
- Goal: Bin packing of the unit fractions  $\{1/w_1, 1/w_2, ..., 1/w_n\}$ .
- Let  $H(W) = \left[\sum_{i \in W} \frac{1}{w_i}\right]$ . Clearly,  $OPT(W) \ge H(W)$ .
- We will see: An algorithm that uses at most H(W)+1 bins (additive error of one for any input).

# Any-fit Decreasing for UFBP

- 1. Sort the items such that  $1/w_1 \ge 1/w_2 \ge \cdots \ge 1/w_n$
- 2. Pack the items in this order, each item is placed in any open bin that can accommodate it, or in a new bin, if none exists.

Theorem: The number of bins used is at most

$$1 + \left\lceil \sum_{i} \frac{1}{w_{i}} \right\rceil \leq 1 + OPT$$

Proof idea: After packing all the items of size at least 1/k:

- (i) There are at most k-1 non-full bins, and
- (ii) Each of the non-full bins is at least 1-1/k full.

Details: In Class

# Any-fit Decreasing for UFBP

Remark: The analysis is tight (the alg. is not optimal)

Example: 
$$\frac{1}{2}$$
,  $\frac{1}{3}$ ,  $\frac{1}{4}$ ,  $\frac{1}{4}$ ,  $\frac{1}{4}$  - in decreasing order.

- Will be packed in three bins:

$$\left\{\frac{1}{2},\frac{1}{3}\right\} \quad \left\{\frac{1}{3},\frac{1}{4},\frac{1}{4}\right\} \quad \left\{\frac{1}{4}\right\}$$

- Can be packed in two bins:

$$\left\{\frac{1}{2}, \frac{1}{4}, \frac{1}{4}\right\} \quad \left\{\frac{1}{3}, \frac{1}{3}, \frac{1}{4}\right\}$$

### Online Bin Packing

```
The input: A sequence of items (numbers), a_1, a_2, ..., a_n, such that for all i, 0 < a_i < 1
```

The goal: 'pack' the items in bins of size 1. Use as few bins as possible.

Example: The input: 1/2, 1/3, 2/5, 1/6, 1/5, 2/5.

Optimal packing in two bins:

(1/2, 1/3, 1/6), (2/5, 2/5, 1/5).

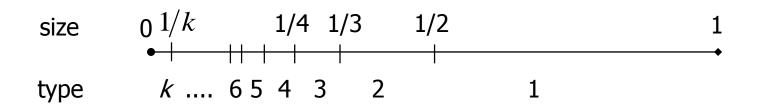
Legal packing in three bins:

(1/2, 1/3), (2/5, 1/6, 1/5), (2/5)

Online BP:  $a_i$  must be packed before we know  $a_{i+1},...,a_n$ 

## The HARMONIC-k Algorithm

Classify items into k intervals according to size (1/2,1] one item per bin (1/3,1/2] two items per bin ... (1/k,1/(k-1)] k-1 items per bin (0,1/k] use NF



### The HARMONIC Algorithm

- Each bin contains items from only one class: i items of type i per bin
- Items of last type are packed using NEXT FIT: use one bin until next item does not fit, then start a new bin
- Keeps k-1 bins open

- Let X be the number of bins for (1/2,1]
  - Those bins are full by more than 1/2
- Let Y be the number of bins for (1/3,1/2)
  - Those bins are full by more than 2/3
- Let T be the number of bins for (0,1/3]
  - Those bins are full by more than 2/3

Let W be the total size of all items Then W>X/2+2Y/3+2T/3

#### Other bounds:

- $OPT \ge X$  (items larger than 1/2)
- OPT  $\geq$  (X+2Y)/2 (items larger than 1/3)

• 
$$H3 \le X+Y+T(+2) \le (3(W+X/6))/2(+2)$$
  
  $\le 3W/2+X/4(+2) \le 1.75OPT(+2)$ 

Asymptotically, this is neglected.

- Let X be the number of bins for (1/2,1]
  - Those bins are full by more than 1/2
- Let Y be the number of bins for (1/3,1/2)
  - Those bins are full by more than 2/3
- Let Z be the number of bins for (1/4, 1/3)
  - Those bins are full by more than 3/4
- Let T be the number of bins for (0,1/4)
  - Those bins are full by more than 3/4
- Let W be the total size of all items
   Then W>X/2+2Y/3+3Z/4+3T/4

#### Other bounds:

- $OPT \ge X$  (items larger than 1/2)
- OPT  $\geq$  (X+2Y)/2 (items larger than 1/3)

```
    H4 ≤ X+Y+Z+T (+3) ≤ (4(W+X/4+Y/12))/3(+3)
    ≤ 4·W/3+X/3+Y/9 (+3) =
    =4·W/3+(x/18+Y/9)+5·X/18 (+3)
    ≤ 31·OPT /18 (+3) ≈ 1.7222·OPT (+3)
```

- Theorem: For any k, Harmonic-k is at most 1.691 competitive.
- Proof: C. C. Lee and D. T. Lee. A simple online bin-packing algorithm. Journal of the ACM 32 (3) July 1985. (beyond our scope. Available in the course web-page).