

Section 3.1: The Knapsack Problem

Knapsack

Input:

Knapsack with a capacity $B \in \mathbb{Z}^+$

Items $I = \{1, 2, \dots, n\}$

Item i has size $s_i \in \mathbb{Z}^+$ and value $v_i \in \mathbb{Z}^+$

Objective:

Find a set of items with total size $\leq B$
and largest possible total value

Greedy alg.

Consider items in order of decreasing s/v ratio

Does not have any constant approximation factor:

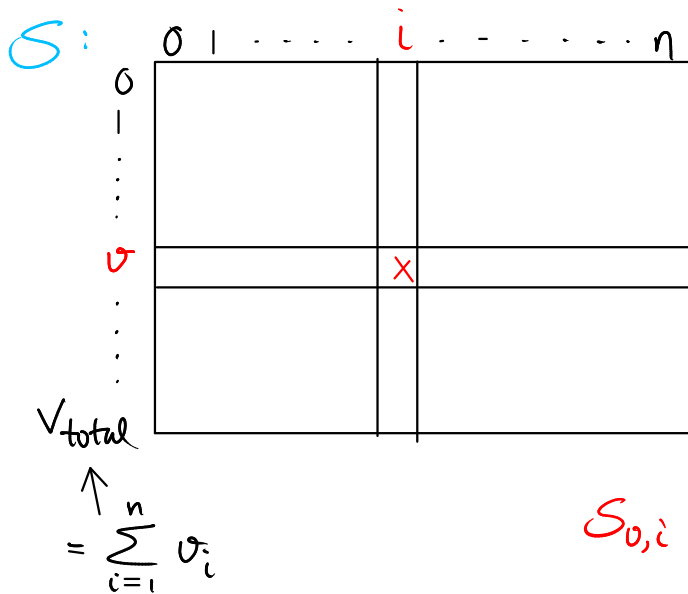
Ex:

i	1	2
v_i	1	$B-1$
s_i	1	B

$$\frac{v_1}{s_1} = 1 > \frac{v_2}{s_2} = 1 - \frac{1}{B}$$

$$\Rightarrow \text{Greedy} = 1 = \frac{1}{B-1} \cdot \text{OPT}$$

Dynamic prg alg:



$S_{v,i}$: smallest possible total size of a subset of $\{1, \dots, i\}$ with total value v , i.e.,

$$S_{v,i} = \min_{I \subseteq \{1, \dots, i\}} \left\{ \sum_{j \in I} s_j \mid \sum_{j \in I} v_j = v \right\}$$

Ex:

$B=5$

i	1	2	3
v_i	3	1	2
s_i	2	3	2

Optimal total value \rightarrow

$i \backslash v_i$	0	1	2	3
0	0	0	0	0
1	∞	∞	3	3
2	∞	∞	∞	2
3	∞	2	2	2
4	∞	∞	5	5
5	∞	∞	∞	4
6	∞	∞	∞	7

$4 \leq B$
 $7 > B$

How to fill the table

$S:$

	0	$i-1$	i	n
0	0
∞				
$\sigma - v_i$			X	
\vdots				
σ			X	X
\vdots				
V_{total}				
∞				

$V_{total} = \sum_{i=1}^n v_i$

$S_{\sigma,i}$: smallest possible total size of a subset of $\{1, \dots, i\}$ with total value σ , i.e.,

$$S_{\sigma,i} = \min_{I \subseteq \{1, \dots, i\}} \left\{ \sum_{j \in I} s_j \mid \sum_{j \in I} v_j = \sigma \right\}$$

If $i=0$ and $1 \leq \sigma \leq V_{total}$
 $S_{\sigma,i} = \infty$

Otherwise,

$$S_{\sigma,i} = \begin{cases} S_{\sigma,i-1}, & \text{if } 0 \leq \sigma < v_i \\ \min \left\{ \underbrace{S_{\sigma,i-1}}_{\text{best solution without item } i}, \underbrace{S_{\sigma-v_i,i-1} + s_i}_{\text{best solution with item } i} \right\}, & \text{if } \sigma \geq v_i \end{cases}$$

Not necessary to fill in the ∞ -entries
 ($A[i]$ corresponds to column i):

Alg 3.1:

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A[1] ← { (0,0), (s1, v1) }
for i ← 2 to n
    A[i] ← A[i-1]
    for each (S,V) ∈ Aprev
        if S + si ≤ B
            A[i] ← A[i] ∪ { (S+si, V+vi) }
    Remove dominated pairs from A[i]
Return max(S,V) ∈ A[n] { V }

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B=5

i	1	2	3
v _i	3	1	2
s _i	2	3	2

$$A[1] = \{ (0,0), (3,2) \}$$

$$\begin{aligned}
 A[2] &= A[1] \cup \{ (1,3), (4,5) \} \\
 &= \{ (0,0), (1,3), (3,2), (4,5) \}
 \end{aligned}$$

$$\begin{aligned}
 A[3] &= A[2] \cup \{ (2,2), (3,5), (5,4) \} \\
 &= \{ (0,0), (1,3), (2,2), (3,2), (3,5), (4,5), (5,4) \}
 \end{aligned}$$

dominated by



Analysis

Running time: $O(n \cdot V_{\text{total}})$

Input size: $O(\log B + n(\log M + \log S))$, where

$$M = \max_{1 \leq i \leq n} \{v_i\} \quad \text{and} \quad S = \max_{1 \leq i \leq n} \{s_i\}.$$

Poly. time?

Ex: Consider a family of instances where

$$V_{\text{total}} = 2^n \quad \text{and}$$

$$B, S \leq 2^n$$

Then

Running time $T(n) \in \Omega(n \cdot 2^n)$ and

Input size $S(n) \in O(n^2)$

$$\Rightarrow T(n) \in \Omega((S(n))^{c\sqrt{n}})$$

No

But if the **numeric part** of the input (i.e., B, v_i, s_i) were **written in unary**, the input size would be $\Theta(B + V_{\text{total}} + S_{\text{total}})$, and the running time would be poly. in the input size.

Hence, the running time is **pseudopolynomial**.

Note: if V_{total} is poly. in n for all possible input instances, the dyn. prg. alg. is poly.

Leading to the following idea...

Idea for approximation algorithm:

Round values st. there are only a poly. number of (equidistant) values:

- Choose a value μ
- Round down each item value to the nearest multiple of μ
- Do dyn. prg. on the rounded values

How to choose μ ?

- Approximation:

When rounding, each item loses a value of less than μ . Hence, the value of any solution is changed by less than $n\mu$.

Thus, if we want a precision of ϵ ,

$$\mu = \frac{\epsilon M}{n}$$

will do, since then $n\mu = \epsilon M \leq \epsilon \cdot \text{OPT}$.

(We will add more detail to this argument in the proof of Thm 3.5.)

- Running time:

$$n \cdot \frac{V_{\text{total}}}{\mu} \leq n \cdot \frac{nM}{\mu} = n \cdot nM \cdot \frac{n}{\epsilon M} = \frac{1}{\epsilon} \cdot n^3$$

Since each rounded value is a multiple of μ , we might as well scale by a factor of $\frac{1}{\mu}$ s.t. the possible values will be $1, 2, \dots, \lfloor \frac{V_{total}}{\mu} \rfloor$ instead of $\mu, 2\mu, \dots, \lfloor \frac{V_{total}}{\mu} \rfloor \mu$:

Alg 3.2

$$M \leftarrow \max_{1 \leq i \leq n} v_i$$

$$\mu = \frac{\epsilon M}{n}$$

for $i \leftarrow 1$ to n

$$v_i' \leftarrow \lfloor \frac{v_i}{\mu} \rfloor$$

Do dyn. prg. with values v_i' (and sizes s_i)

Theorem 3.5

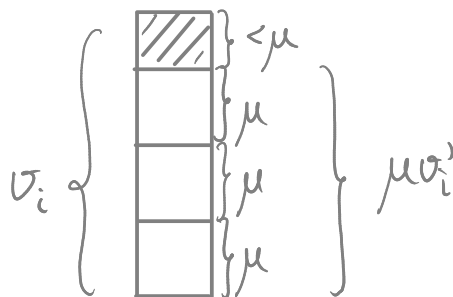
Alg. 3.2 is a $(1-\epsilon)$ -approx. alg. with a running time poly. in both input size and $\frac{1}{\epsilon}$

Proof:

Approximation ratio:

For each item i , $\mu v_i'$ equals v_i rounded down to the nearest multiple of μ . (*)

Thus, $v_i - \mu v_i' < \mu$ (each item "loses" less than μ in the rounding.) (**)



Let A be the set of items selected by Alg. 3.2

This is an optimal solution to the instance with values v_i' , and hence, to the instance with values $\mu v_i'$. (***)

Let O be the set of items in an optimal solution to the original instance with values v_i .

The total value produced by Alg. 3.2 is

$$\begin{aligned}\sum_{i \in A} v_i &\geq \sum_{i \in A} \mu v_i', \quad \text{by } (*) \\ &\geq \sum_{i \in O} \mu v_i', \quad \text{by } (***) \\ &> \sum_{i \in O} (v_i - \mu), \quad \text{by } (**) \\ &\geq \left(\sum_{i \in O} v_i \right) - n\mu, \quad \text{since } |O| \leq n \\ &= \text{OPT} - n \cdot \frac{\epsilon M}{n} \\ &= \text{OPT} - \epsilon M \\ &\geq (1 - \epsilon) \text{OPT}, \quad \text{since } \text{OPT} \geq M\end{aligned}$$

Running time:

$O(\frac{1}{\epsilon} \cdot n^3)$ as proven above.

□

According to Thm 3.5, Alg. 3.2 is a
fully polynomial time approximation scheme (FPTAS)
also poly. in $1/\epsilon$ poly. in input size Family $\{A_\epsilon\}$ of alg., where A_ϵ has precision ϵ .
($(1-\epsilon)$ -approx. alg for max. problems,
($(1+\epsilon)$ -approx. alg for min. problems)

Def. 3.4

Def. 3.3

Thus, Thm 3.5 could also be stated like this:

Theorem 3.5: Alg 3.2 is a FPTAS

Multiple Knapsack problem: Fixed #knapsacks

Bin Packing can be seen as a dual version of Multiple Knapsack.

Bin Packing

Input: n items with sizes between 0 and 1.

Objective: Pack items in bins of size 1,
using as few bins as possible.

Simple approximation algorithms:

Alg.	Running time	Asymp. approx. factor
Next-Fit	$O(n)$	2
First-Fit	$O(n \log n)$	1.7
Best-Fit	— " —	— " —
Next-Fit - Decreasing	— " —	≈ 1.69
First-Fit - Decreasing	— " —	$1.\bar{2}$
Best-Fit - Decreasing	— " —	— " —

Approximation scheme?

Can we do the same kind of rounding for Bin Packing as we did for Knapsack?

No:

Assume that each item size is rounded up to the nearest multiple of μ , for some $0 < \mu < 1$.
(We need to round up to ensure the packing will be valid.)

Ex:

Let $a = \max\{k\mu \mid k \in \mathbb{Z} \wedge k\mu \leq \frac{1}{2}\}$.

Then, $\frac{1}{2} - \mu < a \leq \frac{1}{2}$.

Assume $\mu < \frac{1}{6}$. (An approx. scheme should work for any $\mu > 0$.)

Then, $\frac{1}{3} < a \leq \frac{1}{2}$.

Consider an input consisting of

- m items of size $a - \mu/2 \rightarrow$ rounded up to $a > \frac{1}{3}$
- m items of size $1 - a + \mu/2$

For this instance, the items fit pairwise in m bins, but for the rounded instance, $m + \frac{m}{2}$ bins are needed.

This would yield an approx. factor of at least $\frac{3}{2}$.