

29 April 2024

$(1-\varepsilon)$  - Approximate Knapsack.

Plan

- \* Knapsack Problem
- \* Announcements
- \* Approximation Scheme
- \* Dynamic Program.

# Knapsack Problem

\* Given :

— list of  $n$  items.

↳ Each item  $i \in \{1, \dots, n\}$  has  
weight  $w_i$  & value  $v_i$

— weight bound  $W$

\* Find : Subset  $S \subseteq \{1, \dots, n\}$  maximizing

$$\sum_{i \in S} v_i$$

subject to

$$\sum_{i \in S} w_i \leq W.$$

5

4

7

3

4

10

2

6

11

15

3

4

$$W = 16$$

5

4

7

3

4

10

2

6

11

15

3

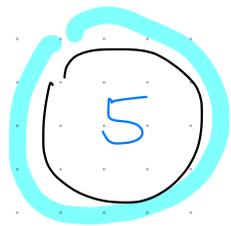
4

$$W = 16$$

$$S = \{5\}$$

$$V_s = 11$$

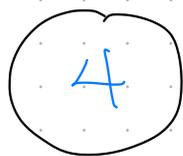
$$W_s = 15$$



4



3



10



6



15



4

$$W = 16$$

$$S = \{5\}$$

$$V_S = 11$$

$$W_S = 15$$

$$S = \{1, 2, 6\}$$

$$V_S = 15$$

$$W_S = 11$$

## Two facts about Knapsack

- (1) Knapsack is NP-Hard
- (2) Knapsack is solvable by dynamic programming  
in time  $O(n^2 v^*)$

where  $v^* = \max_{i \in \{1, \dots, n\}} v_i$

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Knapsack is "weakly" NP-Hard.

## Weak NP - Hardness

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on large value!

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### Idea

\* Reduce Knapsack on Large values to  
Knapsack on small values

\* Solution on "noisy instance" w/ small values  
approximates optimal solution on original instance

## Announcements

\* HW9 out now.

\* Fall 2024 TA Application due tonight!

\* Student Survey about CIS courses  
WIC/URMC

# DP Exact Algorithm.

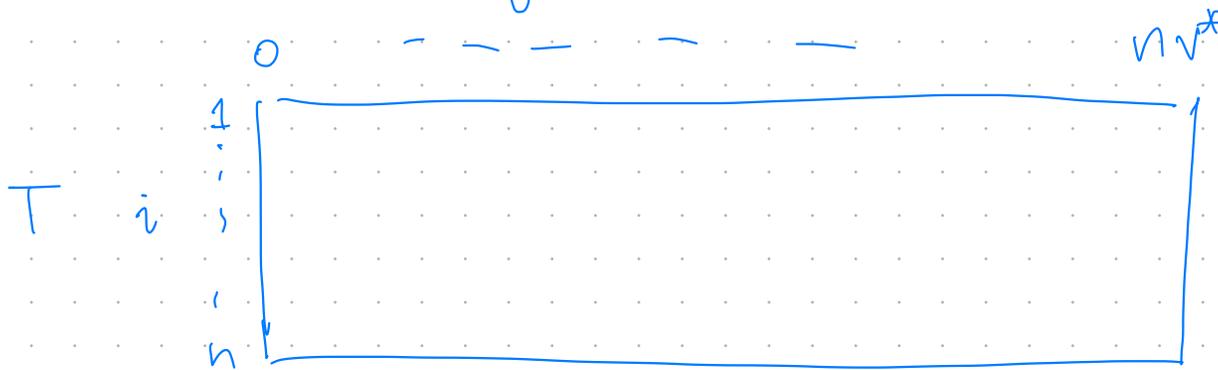
\* pseudo-polynomial time

Dynamic Program

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Dynamic Program



$$V^* = \max_{i \in \{1, \dots, n\}} V_i$$

(i.e.  $V_{\max}$ )

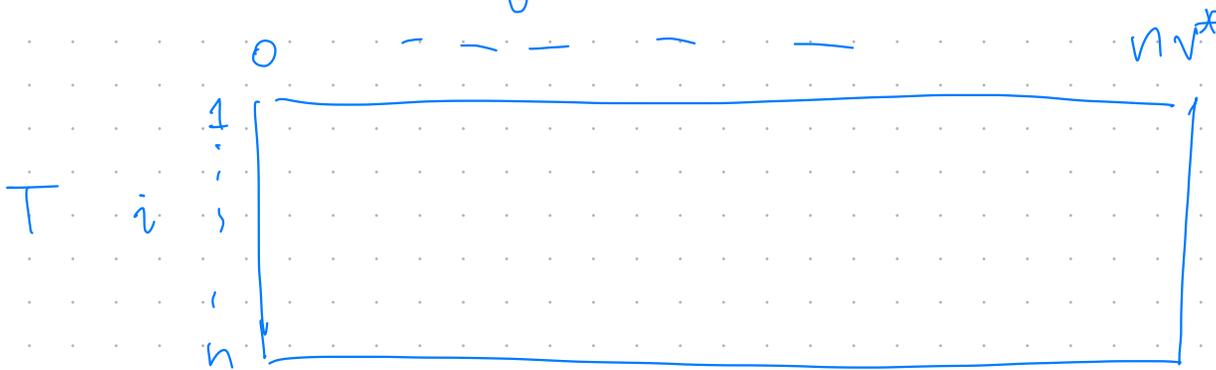
↳ n entries per value

$$\boxed{O(n^2 V^*) \text{ RT}}$$

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Dynamic Program



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\* Reduction, solve a "coarse" instance using exact DP

Reduce Knapsack to "Coarse" - Knapsack.

$$\max_{i \in \{1, \dots, n\}} \sum_{i \in S} v_i$$

s.t.

$$\sum_{i \in S} w_i \leq W$$



$$\max_{i \in \{1, \dots, n\}} \sum_{i \in S} \tilde{v}_i$$

s.t.

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Properties

$$* \quad \tilde{V} = \max_{i \in \{1, \dots, n\}} \tilde{v}_i \ll$$

$$V^* = \max_{i \in \{1, \dots, n\}} v_i$$

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## Properties

$$* \quad \tilde{V} = \max_{i \in \{1, \dots, n\}} \tilde{v}_i \ll V^*$$

$$V^* = \max_{i \in \{1, \dots, n\}} v_i$$

$$* \quad \exists b \in \mathbb{R}$$

s.t.  $\forall S \subseteq \{1, \dots, n\}$

$$b \cdot \sum_{i \in S} \tilde{v}_i \approx \sum_{i \in S} v_i$$

# Reduction

$$\tilde{v}_i = \left[ \frac{v_i}{b} \right] \quad \forall i \in \{1, \dots, n\}$$

## Reduction

$$\tilde{v}_i = \left\lfloor \frac{v_i}{b} \right\rfloor \quad \forall i \in \{1, \dots, n\}$$

## Approximation

$$v_i \stackrel{\textcircled{e}}{\leq} b \cdot \tilde{v}_i = b \cdot \left\lfloor \frac{v_i}{b} \right\rfloor \stackrel{\textcircled{u}}{\leq} v_i + b$$

## Reduction

Leave weights &  $W$   
the same.

$$\tilde{v}_i = \left\lceil \frac{v_i}{b} \right\rceil \quad \forall i \in \{1, \dots, n\}$$

## Approximation

$$v_i \leq b \cdot \tilde{v}_i = b \cdot \left\lceil \frac{v_i}{b} \right\rceil \leq v_i + b$$

Claim. Suppose  $S$  is maximum  $b$ -"coarse" solution, and  $S^*$  is maximum to original knapsack instance.

Then,

$$\sum_{i \in S} v_i \geq \sum_{j \in S^*} v_j - b n$$

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$$\sum_{i \in S} v_i \geq \sum_{j \in S^*} v_j - b/n$$

---

Pf.

①  $\sum_{i \in S^*} \tilde{v}_i \leq \sum_{i \in S} \tilde{v}_i$  // by maximum  $S$ .

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Pf.

$$\textcircled{1} \quad \sum_{i \in S^*} \tilde{v}_i \leq \sum_{i \in S} \tilde{v}_i \quad // \text{ by maximum } S$$

By the inequalities established:

$$\sum_{i \in S^*} v_i \leq b \cdot \sum_{i \in S^*} \tilde{v}_i \leq b \cdot \sum_{i \in S} \tilde{v}_i \leq \sum_{i \in S} v_i + b |S|$$

by  $\textcircled{2}$ 
by  $\textcircled{1}$ 
by  $\textcircled{2}$

So. if we can solve the  $b$ -coarse Knapsack problem, we can obtain approximation.

$$V_{\text{approx}} = \sum_{i \in S} v_i$$

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Approximation Ratio

$$\frac{V_{\text{approx}}}{V_{\text{exact}}} \geq \frac{V_{\text{exact}} - b_n}{V_{\text{exact}}} = 1 - \frac{b_n}{V_{\text{exact}}}$$

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$$r \geq 1 - \epsilon \quad \text{for all} \quad \epsilon \geq \frac{b_n}{\sum_{j \in S^*} v_j}$$

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↳ An  $\epsilon$ -approximation algorithm for any  $\epsilon < 1$ .

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Given any  $\epsilon > 0$ , algorithm produces a  
 $(1-\epsilon)$ -approximate solution to Knapsack

in time  $O(\text{poly}(n) \cdot f(1/\epsilon))$  for some  $f(\cdot)$ .

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Fully PTAS

$O(\text{poly}(n, 1/\epsilon))$

Given  $\varepsilon$ ,

$$b \leq \frac{\varepsilon \cdot V_{\text{exact}}}{n}$$

choose

$$b \leq \frac{\varepsilon \cdot V^*}{n}$$

$V_{\text{exact}}$  at least the max value  
(pre-process to remove  
infeasible values)

Given  $\epsilon$ ,  
choose

$$b \leq \frac{\epsilon \cdot v^*}{n}$$

$$\Rightarrow \tilde{v}_i \leq \frac{v_i^*}{\frac{\epsilon \cdot v^*}{n}} = \boxed{\frac{n}{\epsilon}}$$

Max coarse  
value

$(1-\epsilon)$ -Approximate Knapsack  $(W, w_1, \dots, w_n, v_1, \dots, v_n)$

$$b = \frac{\epsilon \cdot \max_{i \in \{1, \dots, n\}} v_i^*}{n}$$

$$\tilde{v}_i \leftarrow \left\lceil \frac{v_i}{b} \right\rceil \text{ for all } i \in \{1, \dots, n\}$$

Return  $S \leftarrow$  Knapsack  $(W, w_1, \dots, w_n, \tilde{v}_1, \dots, \tilde{v}_n)$

# Dynamic Program for Knapsack

\* 2D table  $T[i, u]$

minimum weight

$$\sum_{j \in S_{i,u}} w_j$$

where

$$S_{i,u} \subseteq \{1, \dots, i\}$$

and

$$\sum_{j \in S_{i,u}} v_j \geq u$$

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and

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- $i$  tracks subset of elements considered.
- $u$  tracks value

Eventually return max value  $u$   
that doesn't exceed weight  $W$

# Dynamic Program for Knapsack

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and

$$\sum_{j \in S_{i,u}} v_j \geq u$$

Cases.

$$\textcircled{1} \quad i \notin S_{i,u} \quad \Rightarrow \quad T[i, u] = T[i-1, u]$$

$$\textcircled{2} \quad i \in S_{i,u} \quad \Rightarrow \quad T[i, u] = w_i + T[i-1, u - v_i]$$

Need to handle  
case where  $< 0$ .

## Knapsack DP Values

Initialize  $T[i, u] = +\infty$  for  $i \in \{1, \dots, n\}$   
 $u \in \{0, \dots, n v^*\}$

For  $i = 1, \dots, n$   $T[i, u_0] = 0$  for  $u_0 \leq 0$ .

For  $i = 1, \dots, n$ .

For  $u = 1, \dots, \sum_{j=1}^i v_j$

$$T[i, u] = \min \begin{cases} T[i-1, u], \\ w_i + T[i-1, u - v_n] \end{cases}$$

$+\infty$  for infeasible  $u$ .

$0$  for  $u_0 \leq 0$ .

Return

$$\max u \quad \text{s.t.} \quad T[n, u] \leq W.$$

Claim. DP returns the correct value.

Claim. DP runs in  $O(n^2 \checkmark)$  time.

$\Rightarrow$  Theorem. Knapsack has an FPTAS.

For any  $\epsilon > 0$ .

$(1-\epsilon)$ -approximate algorithm

running in  $O(n^3/\epsilon)$  time.