

Weakly Approximating Knapsack in Subquadratic Time

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Knapsack

- n items with weights $\{w_i\}_i$ and profits $\{p_i\}_i$
- a knapsack with capacity t
- maximize total profit subject to the capacity constraint

$$\max \left\{ \sum_{i=1}^n p_i x_i : \sum_{i=1}^n w_i x_i \leq t, x_i \in \{0, 1\} \right\}$$

PTAS

- Asks for a subset S of items with

$$p(S) \geq OPT/(1 + \varepsilon)$$

$$w(S) \leq t$$

- Can be done in $\tilde{O}(n + (\frac{1}{\varepsilon})^2)$ time
[Chen, Lian, Mao & Zhang '24][Mao '24]

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[Chen, Lian, Mao & Zhang '24][Mao '24]
- No $O((n + \frac{1}{\varepsilon})^{2-\delta})$ -time algorithm for any constant $\delta > 0$,
under the (min, +)-convolution conjecture
[Künnemann, Paturi & Schneider '17]
[Cygan, Mucha, Węgrzycki & Włodarczyk '19]

Resource Augmentation

- Asks for a subset S of items with

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Weak Approximation

- Asks for a subset S of items with

$$p(S) \geq OPT/(1 + \varepsilon)$$

$$w(S) \leq (1 + \varepsilon)t$$

- Can it be done in $\tilde{O}(n + (\frac{1}{\varepsilon})^{2-\delta})$ time for some constant $\delta > 0$?

Related Problems

- Subset Sum:
Standard approx: no $O((n + \frac{1}{\varepsilon})^{2-\delta})$ -time algorithm
Weak approx: solved in $\tilde{O}(n + 1/\varepsilon)$ time

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Weak approx: ?

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Weak approx: ? **solved in $\tilde{O}(n + (\frac{1}{\varepsilon})^{7/4})$ time**

Value Function

- Let I be a set of items.
- Define $f_I : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ as follows.

$$f_I(x) = \max \left\{ \sum_{i \in I} p_i z_i : \sum_{i \in I} w_i z_i \leq x, z_i \in \{0, 1\} \right\}$$

I

w	2	3
p	3	4



0	1	2	3	4	5
0	0	3	4	4	7

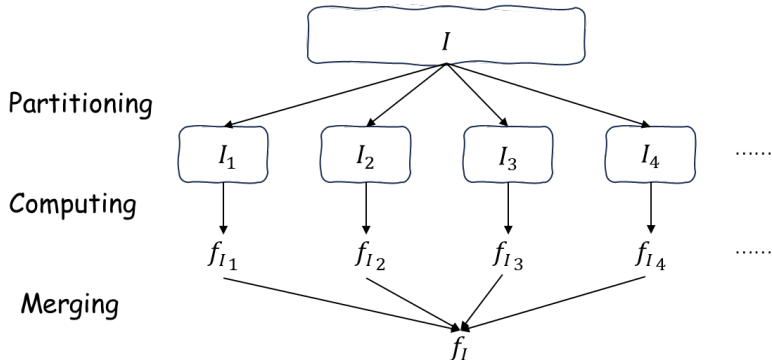
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- Weak approx Knapsack \Rightarrow Compute f' weak approx f_I :

Framework for Value Function



Max-Plus Convolution

- Given f_{I_1} and f_{I_2}

$$f_{I_1 \cup I_2}(x) = \max \{ f_{I_1}(x_1) + f_{I_2}(x_2) : x_1 + x_2 = x \}$$

- $f_{I_1 \cup I_2} = f_{I_1} \oplus f_{I_2} : (\max, +)\text{-convolution}$

 f_I

	7	
	?	

 f_{I_1}

0	1	2	3	4	5

 f_{I_2}

0	1	2	3	4	5

$$? = \max \{ \text{red square} + \text{red square}, \text{blue square} + \text{blue square}, \dots \}$$

Max-Plus Convolution

- $f_{I_1 \cup I_2} = f_{I_1} \oplus f_{I_2}$: (max, +)-convolution
- *General Case*: can be computed in $O(n^2)$ time
(min, +)-convolution conjecture: no $O(n^{2-\delta})$ -time algorithm

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[Chi, Duan, Xie & Zhang '22]

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- *Bounded Monotone*: can be computed in $\tilde{O}(n^{3/2})$ time
[Chi, Duan, Xie & Zhang '22]
 - Given monotone functions f_1 and f_2 , a weak approximation of $f_1 \oplus f_2$ can be computed in $\tilde{O}((\frac{1}{\varepsilon})^{3/2})$ time
 - We can partition items into several groups

The Reduced Problem

- For some $\alpha \geq 1$,
 - $w_i \in [1, 2]$
 - $p_i \in [1, 2]$
 - $t = \Theta(\frac{1}{\alpha\epsilon})$
 - absolute err = $O(\frac{1}{\alpha})$
- When $\alpha \geq 1/\epsilon^{2/3}$, the solution size is small and it can be tackled using color coding
- The difficult case is when $1 \leq \alpha \leq 1/\epsilon^{2/3}$
- Compute the value function of this set

Value Function with Bounded Efficiency

- item efficiency = $\frac{p_i}{w_i}$
- If item efficiencies are same,
weak approx f_I = weak approx sumset sum \rightarrow in $\tilde{O}(\frac{1}{\epsilon})$ time

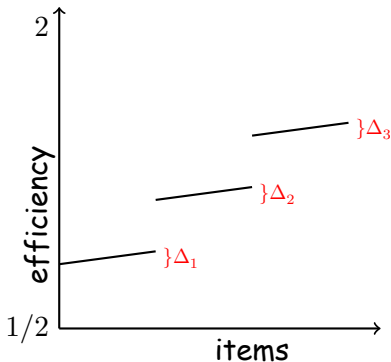
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 f_I can be approximated in $\tilde{O}((\frac{1}{\epsilon})^2 \Delta)$ time.

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- If item efficiencies are in $[\rho, \rho + \Delta]$, using 2D-FFT,
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 - Item efficiencies are in $[1/2, 2] \Rightarrow$ still quadratic time

Items with Similar Efficiencies

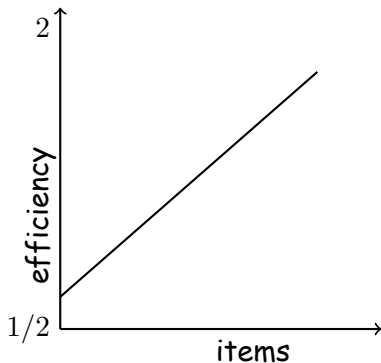


- partition into groups
- compute their value function in time

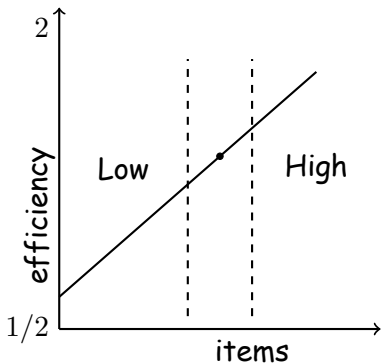
$$\tilde{O}\left(\left(\frac{1}{\varepsilon}\right)^2 \sum \Delta_i\right)$$

Items with Different Efficiencies

- use a proximity bound

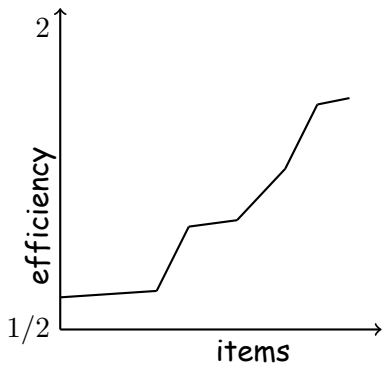


Items with Different Efficiencies

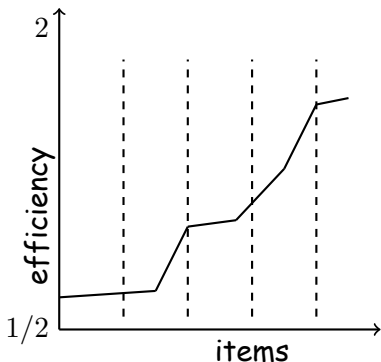


- use a proximity bound
- partition into three parts
- only a few items from Low part will contribute; and only a few items from High part will not contribute
- allow a larger approx factor

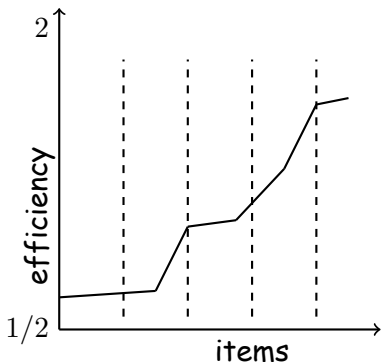
General Case



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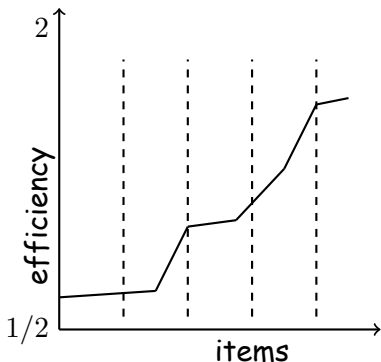


General Case



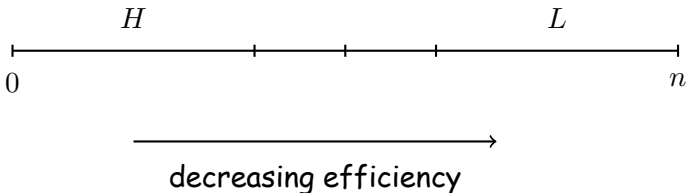
- partition into groups of fix size τ
- $\Delta \leq \frac{1}{\tau}$ (Good):
Compute directly
- $\Delta \geq \frac{1}{\tau}$ (Bad):
Use a larger factor

General Case



- partition into groups of fix size τ
- $\Delta \leq \frac{1}{\tau}$ (Good):
Compute directly
- $\Delta \geq \frac{1}{\tau}$ (Bad):
Use a larger factor
- In $\tilde{O}(n + (\frac{1}{\varepsilon})^{11/6})$ time

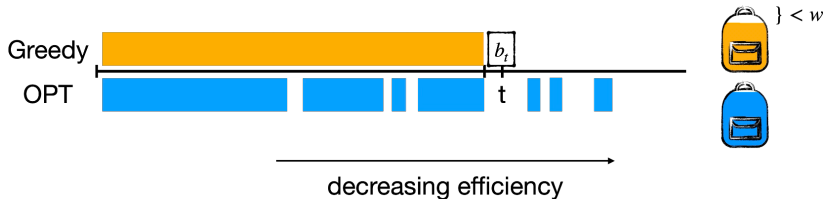
Proximity Bound



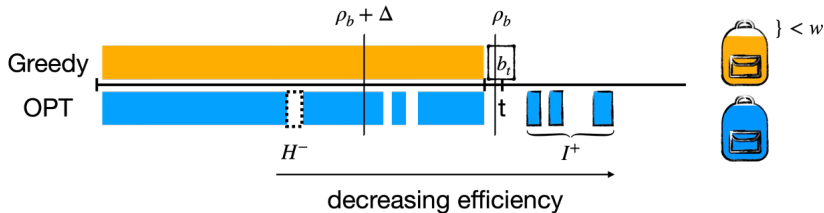
- Most items in H will be used
- Most items in L will not be used

Efficiency-Based Proximity

- Greedy Solution: add items until could not add
- Breaking item: the first item that the greedy solution does not use

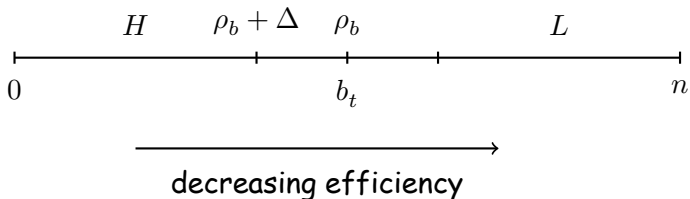


Efficiency-Based Proximity



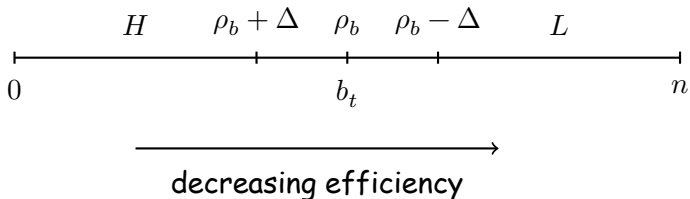
- Weight: $I^+ - H^- \leq w_b$
- Efficiency: $H^- - I^+ \geq \Delta$
- $w(H^-) \leq \frac{p_b}{\Delta}$

Efficiency-Based Proximity



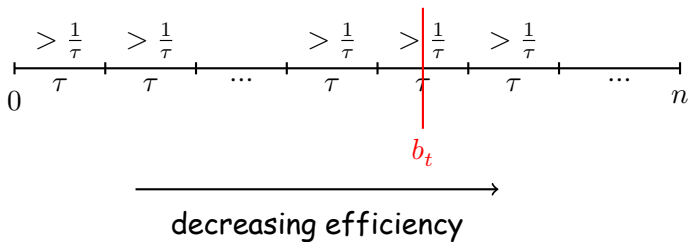
- $w(H^-) \leq \frac{p_{\max}}{\Delta} \leq \frac{2}{\Delta}$

Efficiency-Based Proximity

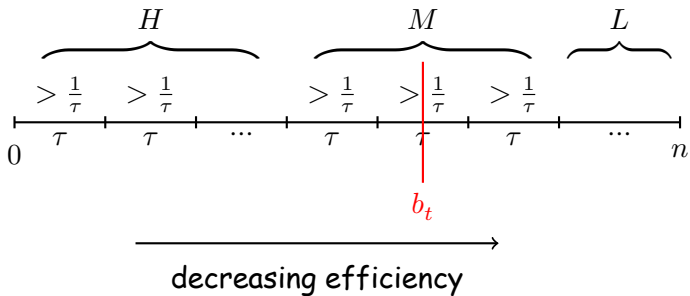


- $w(H^-) \leq \frac{p_{\max}}{\Delta} \leq \frac{2}{\Delta}$
- $w(L^+) \leq \frac{p_{\max}}{\Delta} \leq \frac{2}{\Delta}$

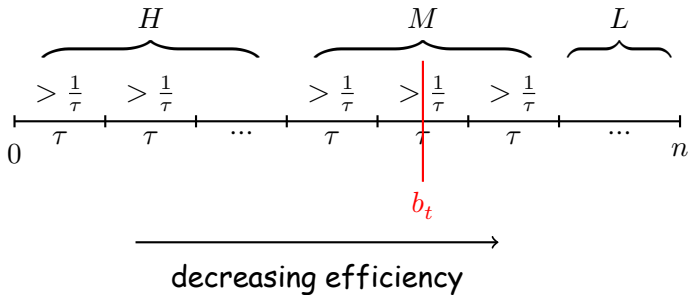
Dealing with Bad Groups



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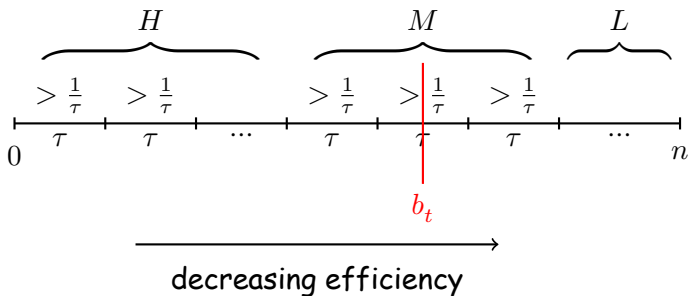


Dealing with Bad Groups



- $w(H^-) \leq 2\tau$
- $w(M^+) \leq 3 \times 2\tau = 6\tau$
- $w(L^+) \leq 2\tau$

Dealing with Bad Groups



$$f_I(t) = f_H(t_H) + f_M(t_M) + f_L(t_L)$$

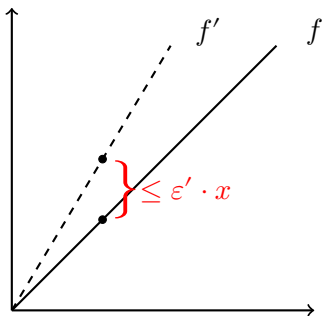
- $t_H \geq w(H) - 2\tau$
- $t_M \leq 6\tau$
- $t_L \leq 2\tau$

Dealing with Groups

$$f_I(t) = f_H(t_H) + f_M(t_M) + f_L(t_L)$$

$$\text{err} = O\left(\frac{1}{\alpha}\right) \Rightarrow \varepsilon' = \frac{1}{\alpha\tau}$$

- $f_M, f_L: t_M \leq 6\tau, t_L \leq 2\tau$



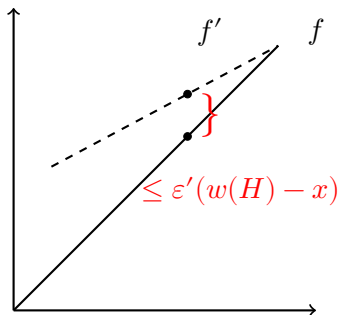
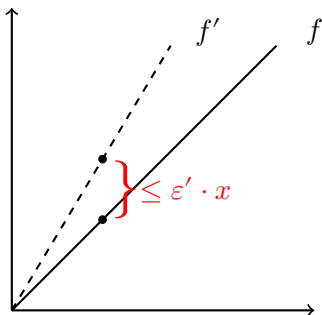
Dealing with Groups

$$f_I(t) = f_H(t_H) + f_M(t_M) + f_L(t_L)$$

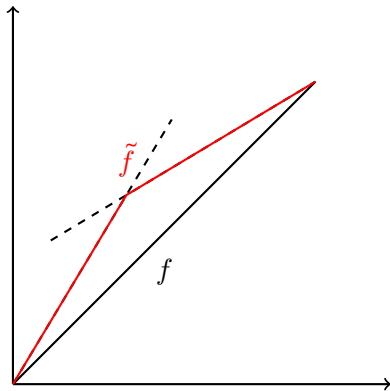
$$\text{err} = O\left(\frac{1}{\alpha}\right) \Rightarrow \varepsilon' = \frac{1}{\alpha\tau}$$

- $f_M, f_L: t_M \leq 6\tau, t_L \leq 2\tau$

- $f_H: t_H \geq w(H) - 2\tau$



A Good Approximation for All Capacities



Theorem

There is a $\tilde{O}(n + (\frac{1}{\varepsilon})^{11/6})$ -time weak approximation scheme for Knapsack.

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Can be improved to $\tilde{O}(n + (\frac{1}{\varepsilon})^{7/4})$ by grouping the items in a more careful way.

Future Work

- $\tilde{O}(n + (\frac{1}{\varepsilon})^{3/2})$ time?
 - same as the Bounded Monotone (max,+) Convolution.
- $\tilde{O}(n + (w + p)^{2-\delta})$ time for exact algorithms?
 - under the (min, +)-convolution conjecture, there is no $O(n + w^{2-\delta})$ (or $O(n + p^{2-\delta})$) -time algorithm

Thank you !