

• Exam 2 on Thursday

continue -- Matrix Chain mult. ...

Greedy Algorithms

Ex. Continuous Knapsack

Goal :

(1) maximize: $\sum_{i=1}^n x_i v_i$ ↙ objective function

(2) subject to: $\sum_{i=1}^n x_i w_i \leq W$

where $w_i > 0, v_i > 0, w_i > 0$ ($1 \leq i \leq n$).

Also $0 \leq x_i \leq 1$ ($1 \leq i \leq n$). (instead of $x_i \in \{0, 1\}$ for discrete version.)

A vector $x = (x_1, x_2, \dots, x_n)$ is called a feasible solution if (2) is satisfied without regard to (1)

Defn

A g-greedy strategy consists of making some (locally) optimal choice, then solving the sub-instance resulting from that choice.

for cont. Knapsack: g-ab the "best" object first.

what does "best" mean?

see Java code for greedy Knapsack.

In general, we must define a selection function $f(i)$ that ranks objects $i = 1, \dots, n$ by their desirability.

Possible selection fns for C.K.P.

- $f(i) = v_i$: sort objects by decreasing values.
- $f(i) = \frac{1}{w_i}$: sort objects by increasing weight.
- $f(i) = \frac{v_i}{w_i}$: sort objects by decreasing value density.

Ex. $n=5$, $v=(2, 3, 6.6, 4, 6)$

$w=(1, 2, 3, 4, 5)$, $W=10$

i	1	2	3	4	5
v_i	2	3	6.6	4	6
w_i	1	2	3	4	5
$\frac{v_i}{w_i}$	2	1.5	2.2	1	1.2

$f(i)$	x					Value
v_i	0	0	1	.5	1	14.6
$\frac{1}{w_i}$	1	1	1	1	0	15.6
$\frac{v_i}{w_i}$	1	1	1	0	.8	16.4

Theorem

If we maximize $\frac{v_i}{w_i}$ in our greedy algorithm (i.e. sort by decreasing value densities.), then the greedy algorithm returns an optimal solution.

Proof.

Assume objects are already sorted by desirability:

$$\frac{v_1}{w_1} \geq \frac{v_2}{w_2} \geq \frac{v_3}{w_3} \geq \dots \geq \frac{v_n}{w_n}$$

Let $x = (x_1, x_2, x_3, \dots, x_n)$ be the soln. returned by the greedy algorithm.

If all $x_i = 1$, then obviously x is optimal. otherwise, let j be first index for which $x_j < 1$.

Thus we have (from the algorithm)

$$x_i = 1 \quad \text{for } 1 \leq i < j$$

$$x_j < 1$$

$$x_i = 0 \quad \text{for } j < i \leq n$$

Also

$$\sum_{i=1}^n x_i w_i = W.$$

Let $y = (y_1, y_2, y_3, \dots, y_n)$ be any feasible solution, i.e.

$$\sum_{i=1}^n y_i w_i \leq W,$$

notation:

$$V(y) = \sum_{i=1}^n y_i v_i \quad (\text{value of } y)$$

$$V(x) = \sum_{i=1}^n x_i v_i \quad (\text{value of } x).$$

□

we must show that

$$V(x) \geq V(y)$$

we have

$$\begin{aligned} \sum_{i=1}^n (x_i - y_i) w_i &= \sum_{i=1}^n x_i w_i - \sum_{i=1}^n y_i w_i \\ &= V - \sum_{i=1}^n y_i w_i \geq 0 \end{aligned}$$

Also

$$\begin{aligned} V(x) - V(y) &= \sum_{i=1}^n (x_i - y_i) v_i \\ &= \sum_{i=1}^n (x_i - y_i) \cdot w_i \cdot \left(\frac{v_i}{w_i} \right) \end{aligned}$$

observe

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$$\bullet \quad i < j \Rightarrow x_i = 1 \Rightarrow (x_i - y_i) \geq 0 \quad \& \quad \frac{v_i}{w_i} \geq \frac{v_j}{w_j}$$

$$\therefore (x_i - y_i) \left(\frac{v_i}{w_i} \right) \geq (x_i - y_i) \left(\frac{v_j}{w_j} \right)$$

$$\bullet \quad i > j \Rightarrow x_i = 0 \Rightarrow (x_i - y_i) \leq 0 \quad \& \quad \frac{v_i}{w_i} \leq \frac{v_j}{w_j}$$

$$\therefore (x_i - y_i) \left(\frac{v_i}{w_i} \right) \geq (x_i - y_i) \left(\frac{v_j}{w_j} \right)$$

$$\bullet \quad i = j \Rightarrow (x_i - y_i) \left(\frac{v_i}{w_i} \right) = (x_i - y_i) \left(\frac{v_j}{w_j} \right)$$

In all cases :

$$(x_i - y_i) \left(\frac{v_i}{w_i} \right) \geq (x_i - y_i) \left(\frac{v_j}{w_j} \right) \quad (1 \leq i \leq n)$$

$$\therefore V(x) - V(y) = \sum_{i=1}^n (x_i - y_i) \cdot \left(\frac{v_i}{w_i} \right) \cdot w_i$$

$$\geq \sum_{i=1}^n (x_i - y_i) \left(\frac{v_j}{w_j} \right) w_i$$

$$= \left(\frac{v_j}{w_j} \right) \cdot \sum_{i=1}^n (x_i - y_i) w_i$$

$$\geq 0$$

∴ $V(x) \geq V(y)$

∴ x is optimal !!!

