

CSE 102 2-20-20

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Greedy Algorithms

Ex. Knapsack (continuous)

$$\text{maximize: } \sum_{i=1}^n x_i v_i$$

$$\text{subject to: } \sum_{i=1}^n x_i w_i \leq W \quad *$$

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

and $x = (x_1, x_2, \dots, x_n)$ satisfies

$$0 \leq x_i \leq 1.$$

Recall x is feasible iff
* is satisfied.

Greedy strategy:

make a locally optimal choice
then solve resulting sub-
problem.

must decide which object is
"best".

selection function: $f(i)$

at each step, maximize $f(i)$

Knapsack (v, w, W)

1.) $n \leftarrow \text{length}[v]$

2.) for $i \leftarrow 1$ to n

3.) $x[i] \leftarrow 0$

4.) $\text{weight} \leftarrow 0$

5.) while $\text{weight} < W$

* 6.) $i \leftarrow$ THE "BEST" REMAINING OBJECT

7.) if $\text{weight} + w[i] \leq W$

8.) $x[i] \leftarrow 1$

9.) $\text{weight} \leftarrow \text{weight} + w[i]$

10.) mark i as included.

11.) else

12.) $x[i] \leftarrow \frac{W - \text{weight}}{w[i]}$

13.) $\text{weight} \leftarrow W$

14.) return x

GREEDY
LOOP

THE "BEST" OBJECT IN (6) CAN BE INTERPRETED IN SEVERAL WAYS.

IN GENERAL WE DEFINE A SELECTION FUNCTION $f(i)$ WHICH ENCODES THE DESIRABILITY OF OBJECT i . LINE (6) THEN MAXIMIZES f OVER ALL REMAINING (UNMARKED) OBJECTS.

Possible selection func.

• $f(i) = v_i$

• $f(i) = \frac{1}{w_i}$

• $f(i) = \frac{v_i}{w_i}$

Ex. $n=5, W=10$

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

$f(i)$	x					Value
v_i	0	0	1	.5	1	14.6
$1/w_i$	1	1	1	1	0	15.6
v_i/w_i	1	1	1	0	.8	16.4

Theorem

if we maximize $\frac{v_i}{w_i}$ on line 6, then Knapsack returns an optimal solution.

Proof

Assume w.l.o.g that

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

Let $x = (x_1, x_2, \dots, x_n)$ be the solution returned by knapsack.

\exists all $x_i = 1$, then x is optimal. otherwise let j be the j^{th} index s.t. $x_j < 1$. so we have

$$x_i = 1 \quad (1 \leq i < j)$$

$$x_j < 1$$

$$x_i = 0 \quad (j < i \leq n)$$

and

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

□

let $V(x) = \sum_{i=1}^n x_i v_i$, the value of x .

let $y = (y_1, \dots, y_n)$ be any other feasible solution, and let

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

must show: $V(x) \geq V(y)$.

Since y is feasible

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

hence

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

Now

$$V(x) - V(y) = \sum_{i=1}^n x_i v_i - \sum_{i=1}^n y_i v_i$$

$$= \sum_{i=1}^n (x_i - y_i) v_i$$

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

observe

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

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

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

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

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

In all cases ($1 \leq i \leq n$):

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

Thus

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

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

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

$$\geq 0$$

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

since y was an arbitrary feasible solution, x is optimal. □

Routine:

- either sort on $\frac{v_i}{w_i}$ (cost $\Theta(n \log n)$) then run algorithm: $\Theta(n)$, so total cost $\Theta(n \log n)$
- or don't sort and search for "best" on each iteration:
cost = $\Theta(n \log n)$

what about Discrete Knapsack

Dyn. Prog had cost $\Theta(n \cdot (W+1))$

Exercise

modify g-greedy algorithm to break when can't steal all of ob_i . Same selection for

$$f(i) = \frac{v_i}{w_i} -$$

- find an Example in which g-greedy returns optimal.
- find an Example in which g-greedy is not optimal.

coin changing again

denominations $d = (d_1, \dots, d_n)$

Amount to be paid: N

Greedy strategy:

- from among all denom. whose addition would not cause sum to exceed N , choose largest
- stop when sum is N ,

we assume

$$1 = d_1 < d_2 < d_3 < \dots < d_n$$

Exercise (hard, but doable).

show that for

$$d = (1, 5, 10, 25, 100)$$

the g -greedy strategy yields
an optimal coin disbursement,
(for all N .)

Exercise (easy)

show if $d = (1, 10, 25, 100)$, g -greedy
strategy does not produce opt.
solution for some N .

Exercise (very hard)

characterize all $d = (d_1, \dots, d_n)$

s.t. g -greedy strategy yields

opt. soln for all $N \geq 0$.

what features tell us a
 g -greedy solution may work?

- optimal substructure.
- Greedy choice Property:
some g -greedy strategy works.

Minimum Weight Spanning Tree (MWST)

Given a weighted connected
Graph G find a MWST in
 G .