

• a more developed solution now

?

in 2 next Thursday. See Dieter email for logistics.

day: Advice for structuring solutions.

Greedy: ~~2 paradigms: 2 Greedy strategies by Exchange arguments~~

- Describe your algorithm at a high-level (ignoring implementation details) [1-3 sentences]
- State a claim that implies your algorithm is correct
 - ↳ Typically a statement that, for every other solution, greedy does at least as well as it.
- * Prove the claim
- Give a refined implementation including any implementation details (if needed)
- Argue running time. [1-2 sentences]

GSA proofs

- Clearly state what is to be proven by induction.
(Usually your claim should look like "at every step, greedy stays ahead ..."
~~induction~~ & induction is on ~~#~~ steps.)
- Give an inductive proof (Base case, inductive step.)

↳ The hard part is finding the right way that greedy "stays ahead" & formulating it precisely.

Exchange Arguments

- Locate an exchange in the other solution that makes it closer to greedy.
- Prove that exchange makes the other solution no worse.

Greedy with GSA example: HW5 #4 (b).

The problem: Input: n intervals $(s_i, f_i] \in \mathbb{R}$
 Output: A set S of points s.t. $\forall i (s_i, f_i] \cap S \neq \emptyset$
 and $|S|$ is as small as possible.

Algo: Find ~~largest~~ earliest j s.t. f_j is smallest. Add f_j to S .
 Discard intervals ~~which~~ that intersect with S & repeat until no more intervals.

Correctness: Let g_1, g_2, \dots, g_k be the points in greedy solution
 Let t_1, t_2, \dots, t_l be the points in an arbitrary solution S .

~~Observe~~ Claim For all $i = 1, 2, \dots, n$, $g_i \geq t_i$, where
 $g_i = +\infty$ if $i > k$ and $t_i = +\infty$ if $i > l$.

Implies my algo correct since ~~if~~ ~~if~~ $k > l$, then we would have
 g_k is finite ~~and~~ $t_k = +\infty$, but claim says $g_k \geq t_k$.

Proof of claim: By induction on i .

Base case: $i=1$. Let f_j be earliest end of an interval. We have $g_1 = f_j$.
 Some point in S is in $(s_j, f_j]$.
 In particular, $t_1 \leq f_j = g_1$.

Induct. step: We know $g_{i-1} \geq t_{i-1}$, and we can assume wlog both are finite.
 (if either is infinite, g_{i-1} is infinite, & so g_i is infinite & $g_i \geq t_i$ follows.)
 Likewise assume g_i is finite.

Let $(s_j, f_j]$ s.t. $f_j = g_i$.
 By construction, $s_j > g_{i-1} \geq t_{i-1} \geq t_{i-2} \geq \dots \geq t_1$.
 So there is another point in S that is in $(s_j, f_j]$.
 This implies t_i is finite and $t_i \leq f_j = g_i$. □

Implementation: Sort intervals ~~to~~ to have increasing order of f_j .
 Let $e = -\infty$, $S = \emptyset$.
 For each interval $(s_j, f_j]$:
 if $e < s_j$:
 $S \leftarrow S \cup \{f_j\}$
 $e \leftarrow f_j$
 Return S

Running Time: Sort in $O(n \log n)$ time, then $O(n)$ work
 \Rightarrow overall $O(n \log n)$.

Greedy with Exchange example: HW6 #1.

Problem: Input: n jobs taking time $t_1 \dots t_n$ with weights $w_1 \dots w_n$

Output: An order of the jobs so that

$$\sum_{i=1}^n C_i w_i$$

is minimized, where C_i is total time of jobs completed before (including) the i -th.

Algo: Sort jobs in decreasing order of $\frac{w_i}{t_i}$.

Correctness: Claim This order minimizes $\sum_{i=1}^n C_i w_i$.

Proof: Exchange argument. Fix an arbitrary schedule.

Suppose jobs l and r are adjacent in that schedule, (l before r)
and out of order with our solution. $\left(\frac{w_l}{t_l} \leq \frac{w_r}{t_r}\right)$

Consider swapping their order.

Let C_i be completion times pre-swap, C_i' post-swap.

We have $C_i = C_i'$ for all $i \neq l, r$.

$$\text{So } \sum_i C_i w_i - \sum_i C_i' w_i = \cancel{C_l - C_l'} (C_l - C_l') w_l + (C_r - C_r') w_r.$$

$C_l - C_l'$ is $-t_r$. $C_r - C_r'$ is t_l .

$$\text{So } (\cancel{C_l - C_l'}) w_l + (C_r - C_r') w_r = -t_r w_l + t_l w_r.$$

Since $\frac{w_l}{t_l} \leq \frac{w_r}{t_r}$, this is positive.

i.e. $\sum_i C_i w_i \geq \sum_i C_i' w_i$. So the new order is no worse.

Since any solution with no adjacent out-of-order pairs is the greedy solution, this proves the greedy solution is optimal. \square

Running Time: Sorting takes $O(n \log n)$ time.

Network Flow: (reductions to problems solved using network flow)

- State the problem you reduce to. (Max flow, min cut, proj. selection, etc.) [Est. 1 sentence]
- * Describe the reduction. Given input to original problem, describe how to build input to problem you reduce to. Be declarative (the utxs are —, the edges are —, ...)
- ↳ Be careful with pictures. Don't rely on them exclusively. [Est. 3-6 sentences]
- Say how to solve original problem given a solution of problem reduced-to. [Est. 1-2 sentences]
- Prove correctness. Relate solutions ^{of orig.} ~~to~~ ~~the~~ problem to solutions of reduced-to problem [Est. 4-6 sentences]
 - May need to make "wlog" assumptions, such as integrality of a maximum flow.
- Running Time Analysis:
 - Compute size of output instance in terms of size of input instance. [Est. 1-2 sentences]
 - ⚡ Plug into running time of reduced-to problem's algorithm.

Network Flow reduction example • HW 8 #3

Problem: Input: n components to be bought from Alpha or Omega.

Alpha cost: $\alpha_1, \dots, \alpha_n$

Omega cost: $\omega_1, \dots, \omega_n$

Incompat cost: $c(i, j)$ if i from Alpha, j from Omega

Output: where to buy each item to minimize total cost.

Reduce to min cut.

Make flow network with vertices s, t, v_1, \dots, v_n ;

edges $s \rightarrow v_i$ for each $i = 1 \dots n$ with cap ω_i

$v_i \rightarrow t$ " with cap α_i

$v_i \rightarrow v_j$ for each i, j with cap $c(i, j)$.

Find min-cap. ~~st~~ st-cut: (S, T) .

For each i , if v_i in S , buy item i from Alpha
if v_i in T , buy item i from Omega

Correctness: Given a way to buy items from Alpha & Omega (sets A, Ω that partition $[n]$),
make a cut $S = \{s\} \cup A$, $T = \{t\} \cup \Omega$.

Its capacity is $\sum_{i \in A} \alpha_i + \sum_{i \in \Omega} \omega_i + \sum_{i \in A, j \in \Omega} c(i, j) = \text{cost of the buying strategy.}$

~~Each cut likewise produces a buying strategy whose cost is the capacity of the cut.~~

Each cut likewise produces a buying strategy whose cost is the capacity of the cut.
 \Rightarrow Finding a min cut gives a buying strategy with minimum cost. \square

Running Time: $O(n)$ nodes, $O(n^2)$ edges.

Using min-cut algo from class, running time is $O(n) \cdot O(n^2) = O(n^3)$.