

# Advanced Algorithms

## Lecture 8: Greedy algorithms

# Announcements

- HW 2 is out! Due next Friday — start early!
- HW 1 grades will be out by Monday  
[TA Vivek].

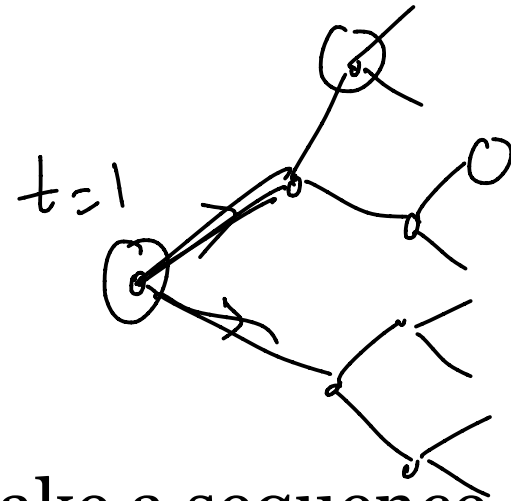
# Recap: dynamic programming

- Sequential decision making (eg. subset sum, paths in graphs, how much cake to eat, where-next in TSP tour, ...)  
↳ Set of vertices "remaining"
- Some resource "depleting" (sub-problem defined by "amount remaining")

- **Key:** past decisions lead to some "state"; we can then solve sub-problem starting at the state (ignoring past)

- One issue is memory.  
Time/memory trade-offs. (Traveling Salesman  $\rightarrow$  memory =  $O(n \cdot 2^n)$ .  
 $n \cdot 2^n$ )

# DP template

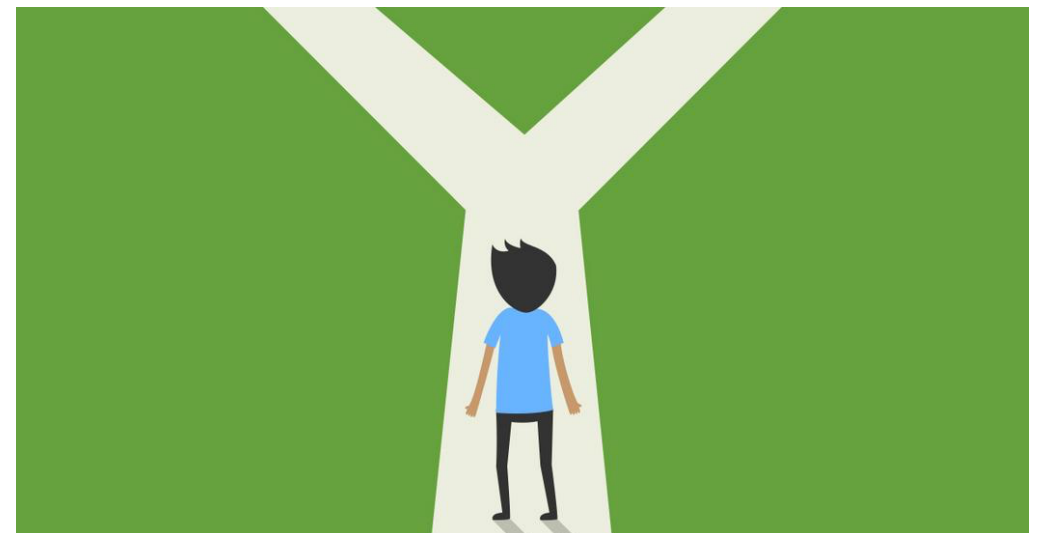


- We need to make a sequence of choices
- Try all choices at time 1. For each one,  
$$\text{cost} = \text{cost}(\text{choice 1}) + \text{cost}(\text{"remaining" problem})$$
- Then pick the best value of choice 1
- Key: figure out how to define/parametrize the *remaining problem*

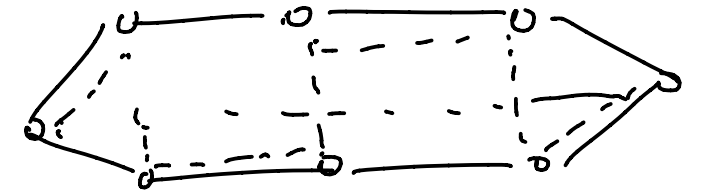
# Greedy algorithms

# Greedy paradigm

- Need to make a sequence of decisions
- “Myopic” choice — make irrevocable decision based on current state
- For each choice, associate *value* (only fn of present), make choice that has best value



# Most “natural” algorithms



- E.g., traveling salesman problem (travel to closest unvisited node)
- Coin change: you are given coins of denominations 1c, 5c, 10c, 20c, 25c, 50c. Make change for say 75c using the fewest # of coins

40¢ → greedy uses 3 coins  
↳ opt is to do 20+20¢.

- More complex problems... chess?

**Moral:** greedy algorithms “typically” aren’t optimal, but give useful insights...

# Scheduling jobs

**Problem:** suppose we have  $n$  jobs, with processing times  $p_1, p_2, \dots, p_n$ .  
Find the best order of scheduling them so as to minimize the sum of  
“completion times”

$$\underline{p_1, p_2, p_3.}$$

Completion times:

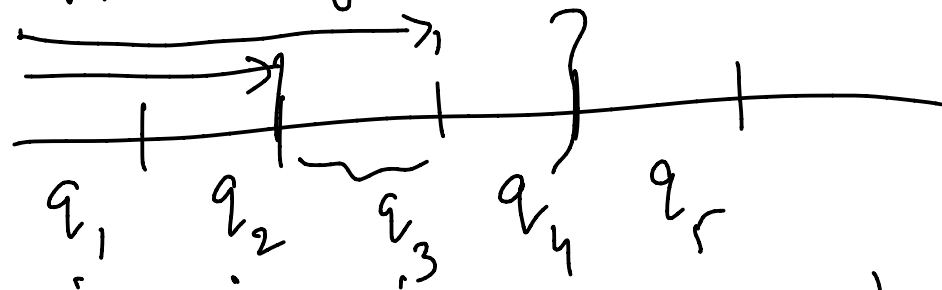
$$p_1 ; p_1 + p_2 ; p_1 + p_2 + p_3 .$$



# Key: correctness

## Proof 1: closed form

Suppose jobs are done in some order:  $(\sigma_1, \sigma_2, \dots, \sigma_n)$  permutation of  $(1, 2, \dots, n)$



$$p_{\sigma_i} = q_i$$

$$q_1$$

$$q_1 + q_2$$

$$\vdots$$

$$q_1 + q_2 + \dots + q_n$$

$$\frac{q_1 + q_2 + \dots + q_n}{n} \rightarrow S$$

Claim: To minimize  $S$ , we should

$$\text{Set } q_1 \leq q_2 \leq \dots \leq q_n.$$

$\therefore$  we must process jobs in increasing order of  $p_i$ .

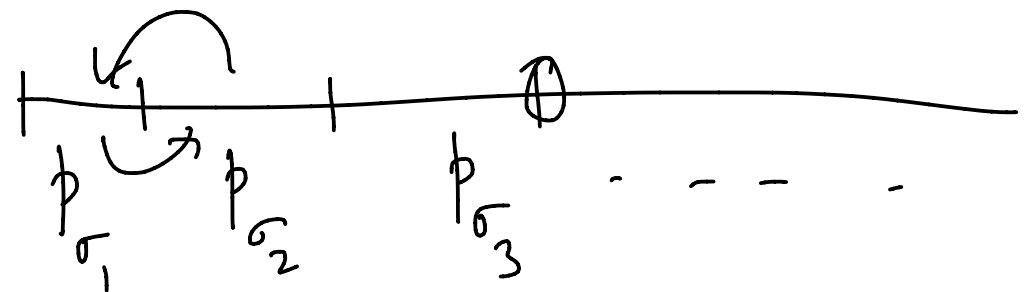
# Key: correctness

**Proof 2: swapping — structure of opt solution**

$p_1, p_2, \dots, p_n$ .

Suppose  $\sigma_1, \sigma_2, \dots, \sigma_n$  is opt ordering.

Claim:  $p_{\sigma_1} \leq p_{\sigma_2}$



Proof: Suppose if possible that  $p_{\sigma_1} > p_{\sigma_2}$ . Now swap  $\sigma_1$  &  $\sigma_2$ .

→ completion time of  $\sigma_3, \sigma_4, \dots$  will remain the same.

→ Sum of completion times of  $\sigma_1$  &  $\sigma_2$  changes from  $p_{\sigma_1} + (p_{\sigma_1} + p_{\sigma_2})$  to  $p_{\sigma_2} + (p_{\sigma_2} + p_{\sigma_1})$

If  $p_{\sigma_1} > p_{\sigma_2}$ , then swapping gives a strictly better solution than optimum soln!  
↓  
contradiction.

Next claim:  $p_{\sigma_2} \leq p_{\sigma_3}$

Exactly the same proof: swapping  $\sigma_2$  &  $\sigma_3$  does not change completion times of other jobs..

Can keep doing this  $\rightarrow p_{\sigma_1} \leq p_{\sigma_2} \leq \dots$

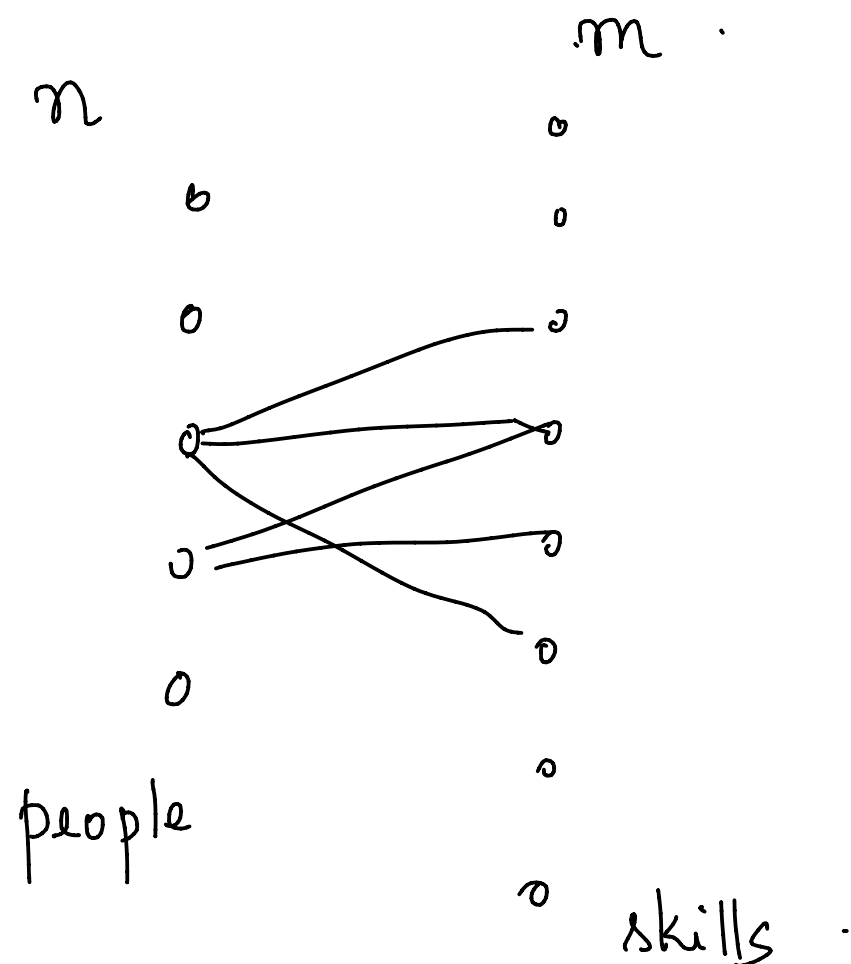
# Key: correctness

**Proof 3:** induction — most common

**Idea:** prove by induction that first  $k$  choices are “correct”

# Set cover

**Problem:** suppose we have  $n$  people, and  $m$  “desired skills”; each person has a subset of the skills. Pick the smallest subset of people such that every skill is covered



Another view: we are given

$$S_1, S_2, \dots, S_n \subseteq \{1, 2, \dots, m\}$$

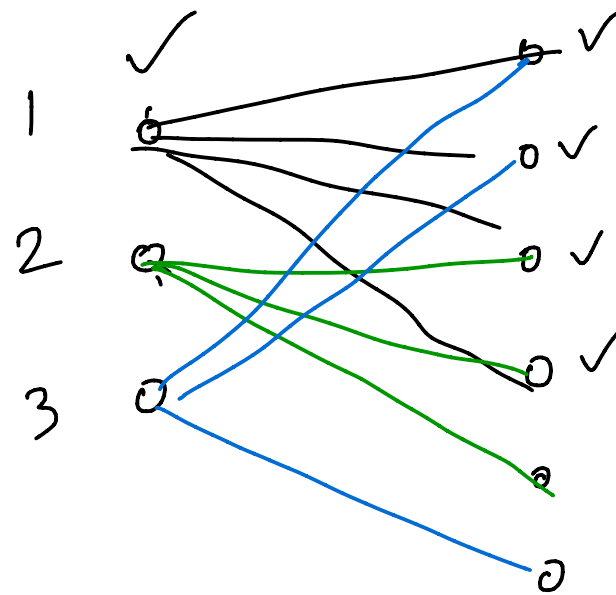
↓  
[m].

find some  $i_1, i_2, \dots, i_k$  s.t.

$$S_{i_1} \cup S_{i_2} \cup \dots \cup S_{i_k} = [m].$$

# Greedy algorithm

- At each time, choose the person with the largest number of uncovered skills (breaking ties arbitrarily)
- Is this optimal?



greedy soln picks all- 3 people.

opt solution only has  $\{2, 3\}$ .

# Example

→ Greedy is not always optimal.

# How bad can greedy be?

[Approx. algorithm].

**Surprising theorem!** suppose there is an optimum solution that uses  $k$  people. Then the greedy algorithm does not use more than  $k \log n$ .

$$k = 5 \rightsquigarrow \leq 60$$

$$n = 10000$$



$$\begin{array}{c} \ln n \\ \parallel \\ 12 \end{array}$$

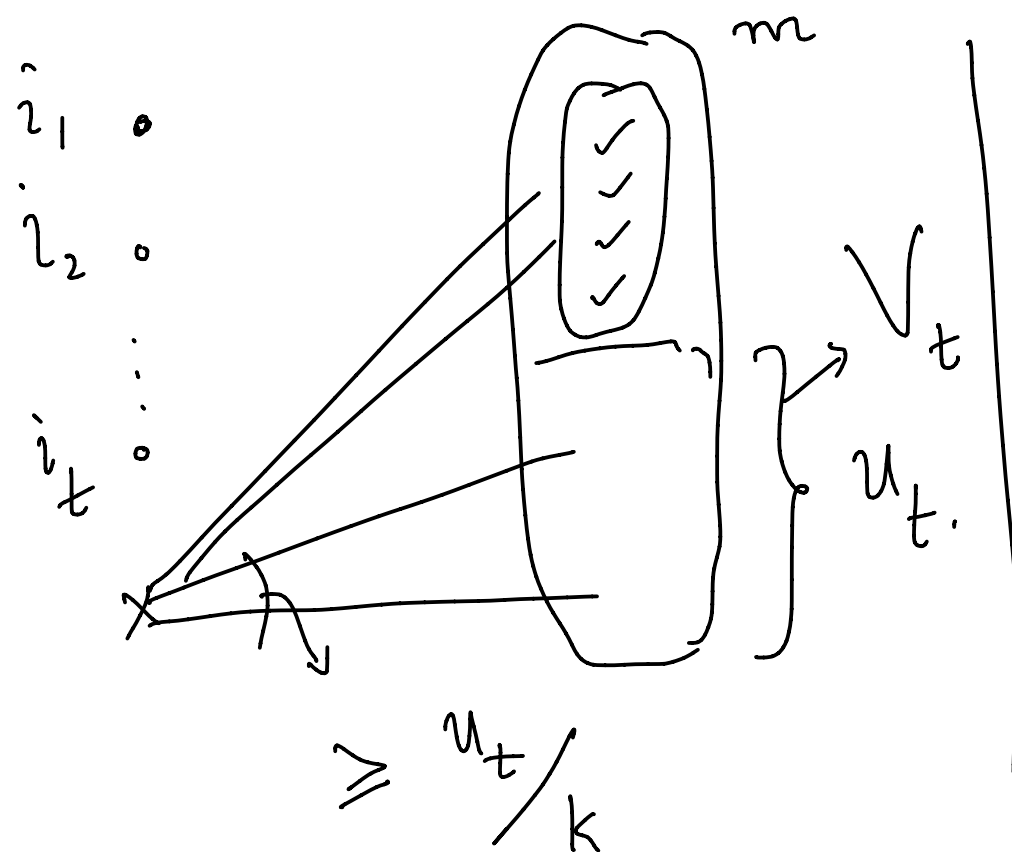
Challenge in the proof: showing that greedy alg. works well just by knowing the existence of a good soln.



# Proof

**Key idea:** many skills are covered at each step!

Formally: Suppose we have  $u_t$  uncovered skills at iteration  $t$ . Then we claim that  $u_{t+1} \leq u_t - \frac{u_t}{k}$ .



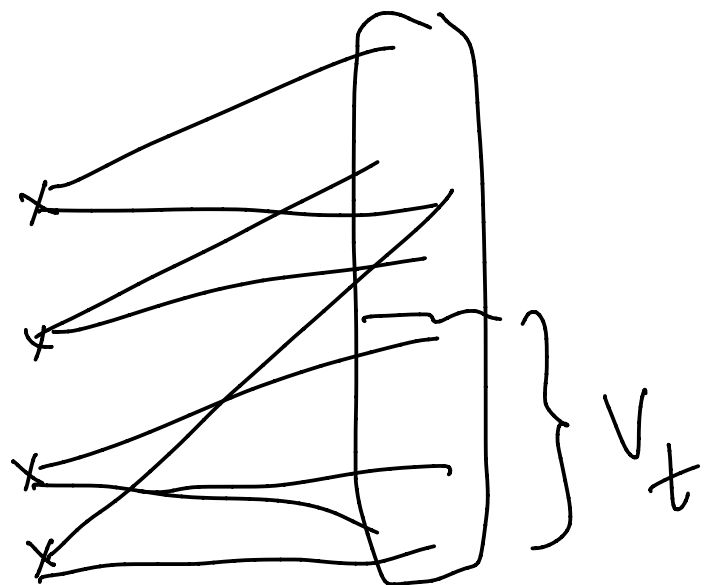
Suffices to show that there exists a person with at least  $\frac{u_t}{k}$  uncovered skills.

# Proof

Let  $S_1, S_2, \dots, S_k$  be the skill-sets of the optimal soln.  $(j_1, j_2, \dots, j_k)$

$$S_1 \cup S_2 \cup \dots \cup S_k = [m] \supseteq V_t.$$

$$(S_1 \cap V_t) \cup (S_2 \cap V_t) \cup \dots \cup (S_k \cap V_t) = V_t.$$



$$\Downarrow$$

one of  $|S_i \cap V_t| \geq \frac{|V_t|}{k}.$

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