

# Greedy Algorithms - Part 1

## ▼ What are greedy algorithms?

- Algorithms where we go over each item  $X$  in some order and decide whether or not to include item  $X$  in the solution.
- We make decisions as we encounter each item, we're *greedy* about how we take/add stuff.

## ▼ What are some examples of greedy algorithms

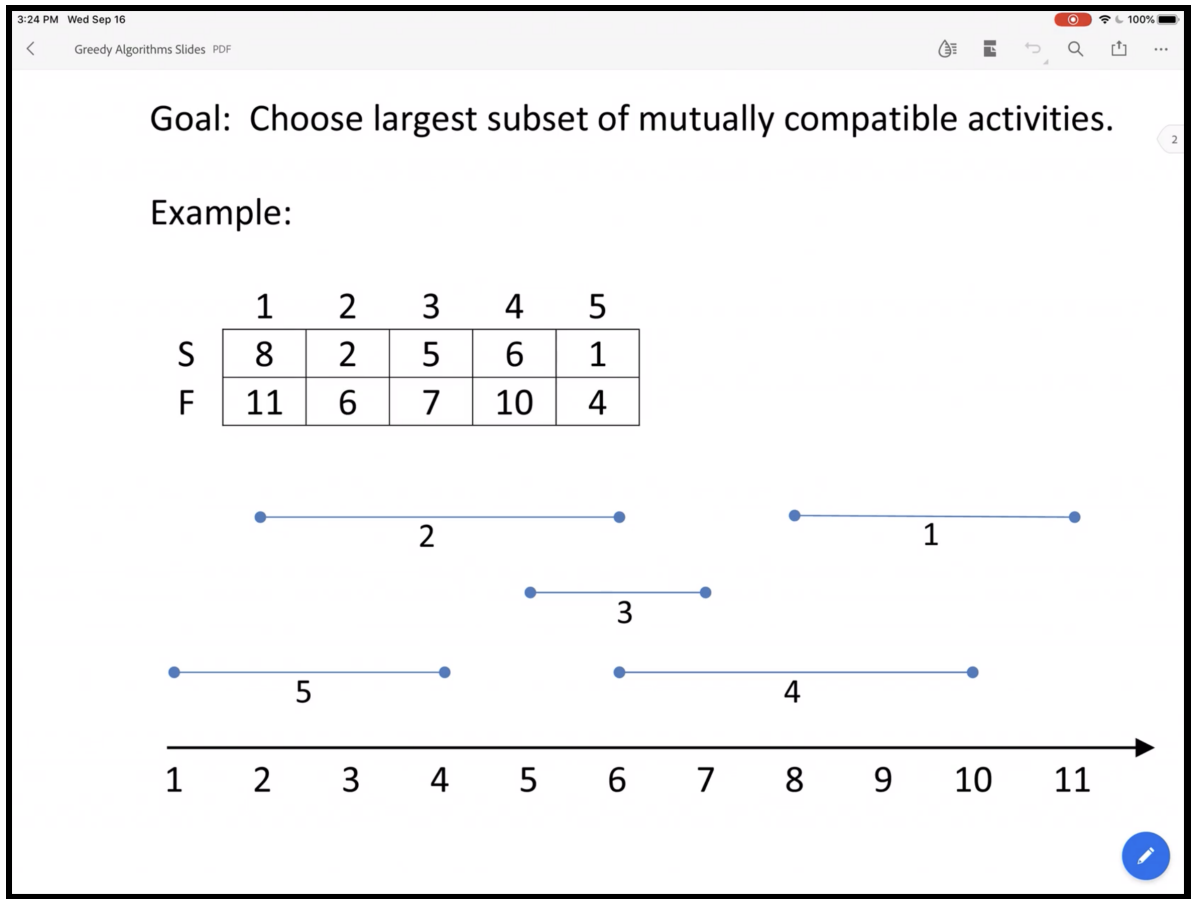
- Sorting: insertion sort, selection sort
- Minimum spanning tree: Kruskal, Prim
- Single-source shortest paths: Dijkstra's Algorithm

## ▼ Do greedy algorithms always work?

- **No.** There's a lot of 'gotchas' with greedy algorithms that might seem like they work but don't actually work.

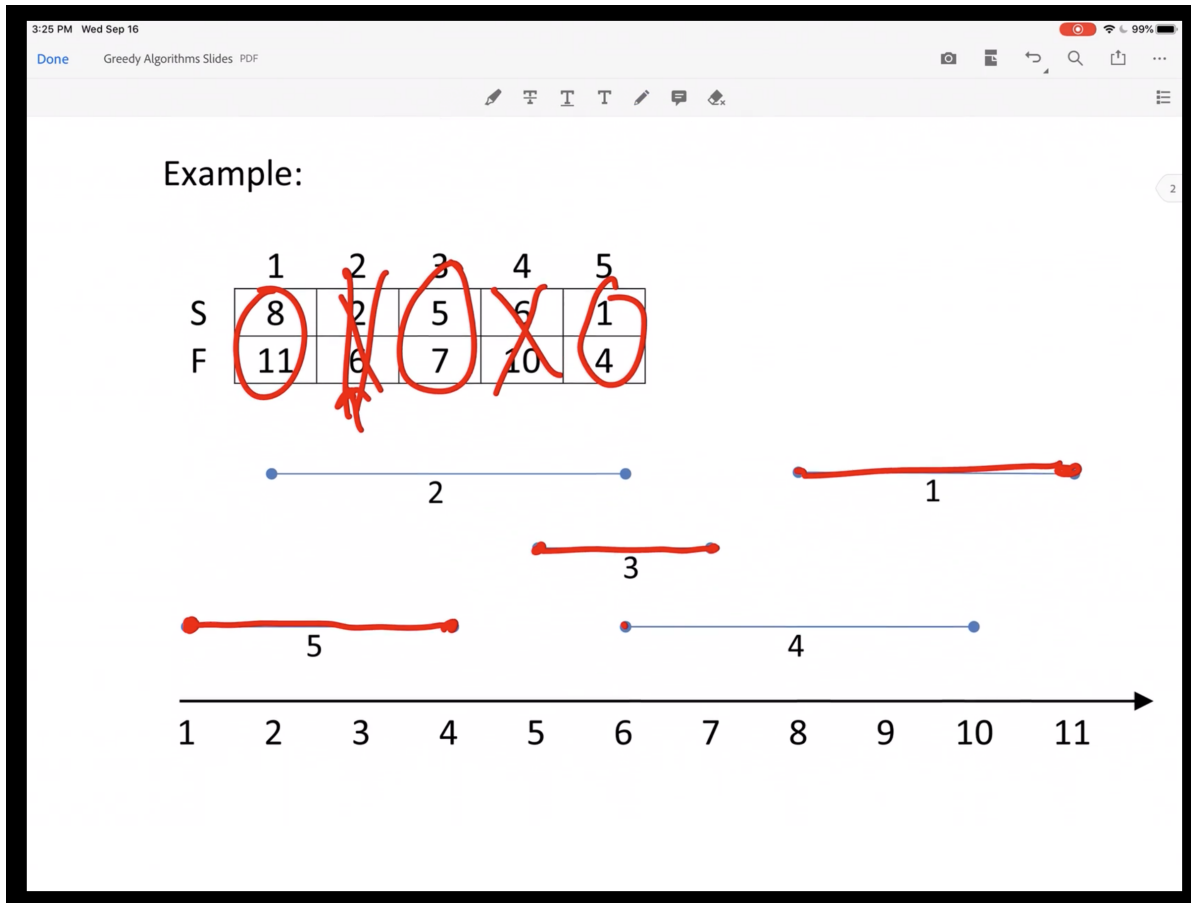
## ▼ What's the definition of the activity selection problem?

- You have a list of Activities  $\{1 \dots n\}$ , their Start times  $S[1 \dots n]$ , and the finish times  $F[1 \dots n]$
- We'll say that activities  $j$  and  $k$  are compatible if either  $F[j] \leq S[k]$  or  $F[k] \leq S[j]$ .
- Compatible jobs fit with out overlap, conflicting activities don't.
- Goal: Choose the largest subset of mutually compatible activities.

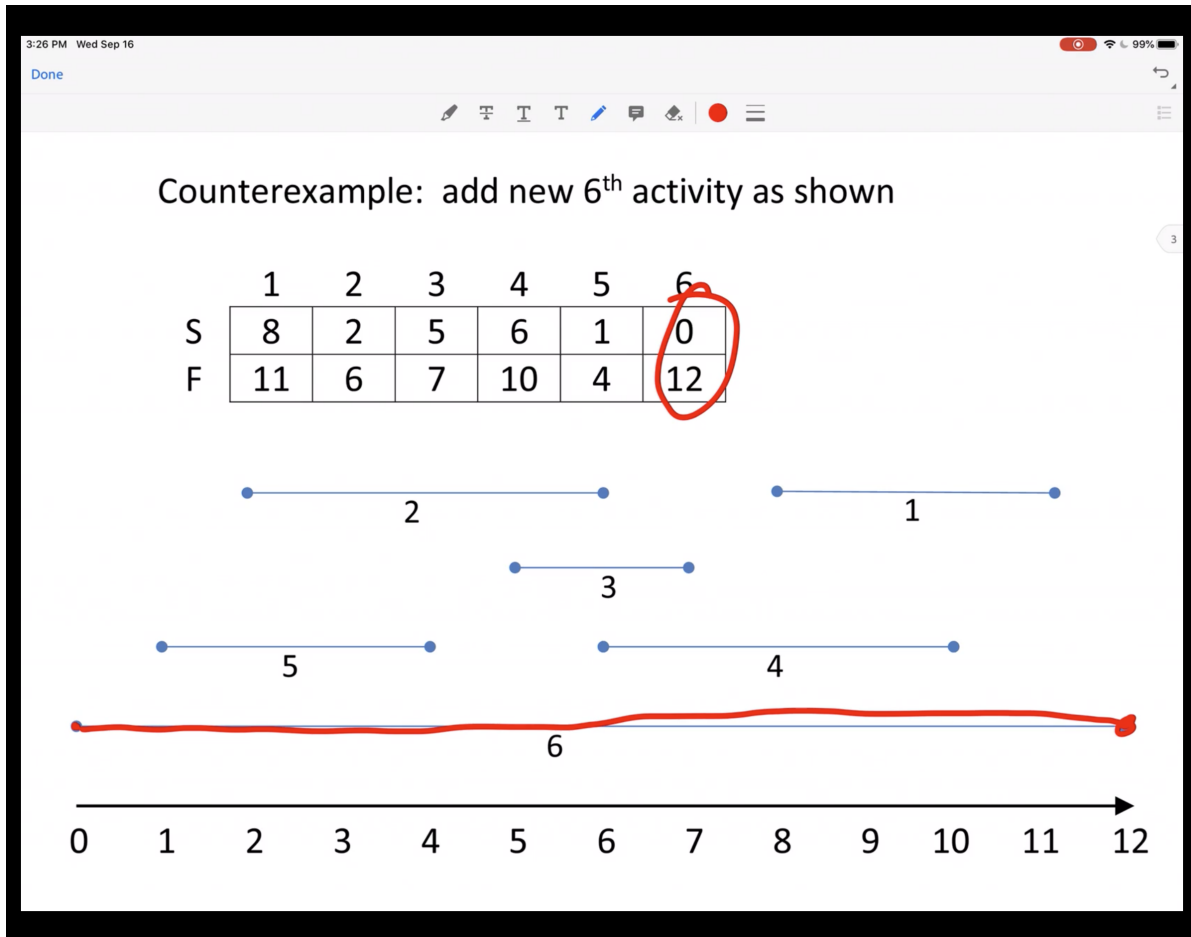


▼ First greedy approach to activity selection

- Let's go over each item in order of earliest start time first. If it fits with the rest of the items, we add it. If it doesn't, then it gets left out.
- This actually works on our first example! This algorithm would select jobs 5, 3, and 1, which is the greatest set of compatible activities.

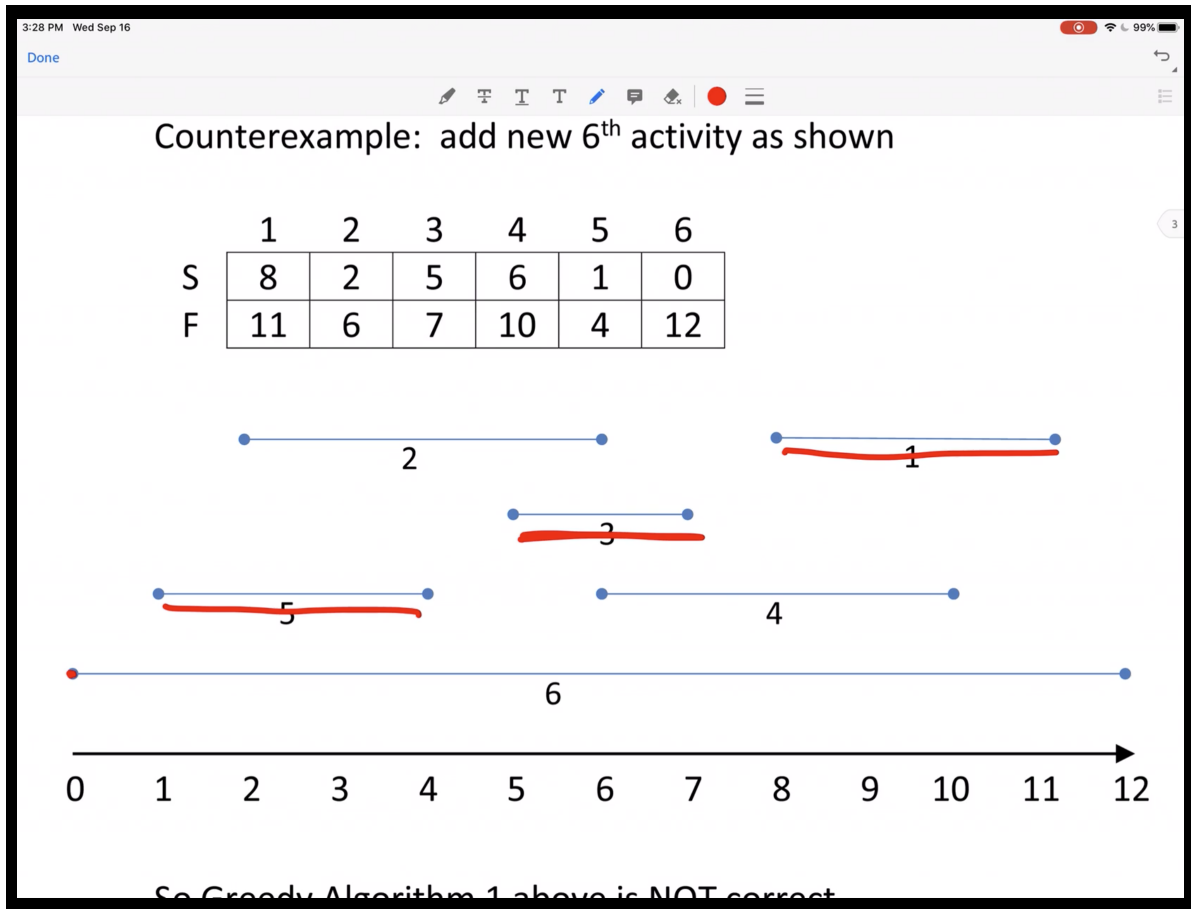


- However, there's a good counterexample to this problem. This wouldn't work right here (remember that the length of the job doesn't matter)

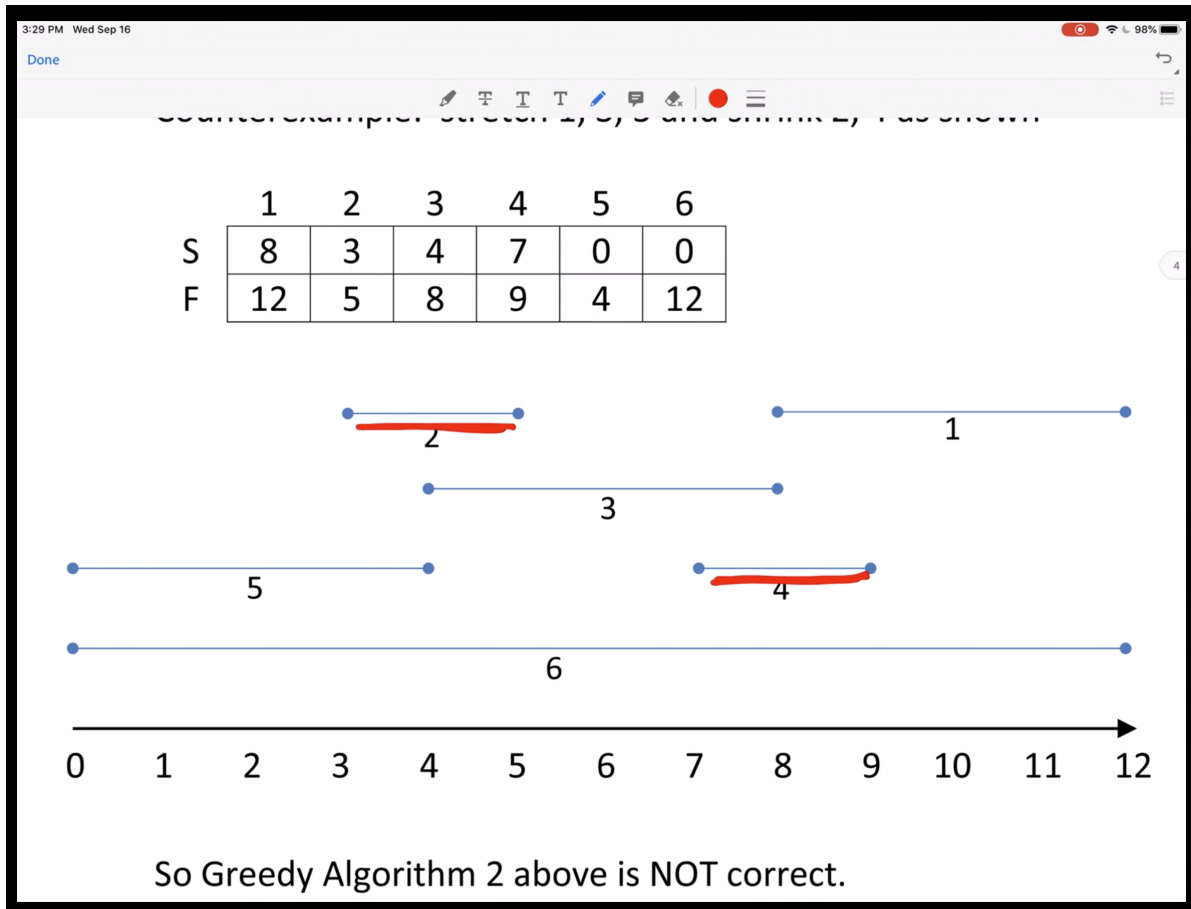


▼ Second greedy algorithm

- Let's go over each item in order of *shortest job first* ( $F[k] - S[k]$ ).
- This would work for our little counterexample actually! It would select items, 3, 5, and 1.



- However, like the other approaches, there's a counterexample:



- So this greedy algorithm is not correct.
- ▼ Third greedy algorithm
- The third idea is to select items in order of their finish time. We want to choose the job with the earliest finish time first.
  - This is the greedy strategy that works!
  - We can't just say that 'there's no counter example lol'
- ▼ Contradiction argument proof
- Suppose this algorithm isn't correct
  - Consider its first incorrect choice of some activity  $k$ .
  - So there must be a better next choice, say some activity  $k'$  which leads to a larger solution subset  $A'$ .

- By compatibility, every future  $k''$  in  $A'$  has  $F[k'] \leq S[k'']$ . Also  $F[k] \leq S[k'']$ .
- Therefore solution  $A' - \{k'\} \cup \{k\}$  is an equally good solution, thus contradicting that  $K'$  is a better choice than  $k$ .
- Therefore, this algorithm is indeed correct.

#### ▼ What's the fractional knapsack problem?

- It's the 0-1 knapsack problem but without the 'we have to take the whole item' constraint.

#### ▼ Algorithm one

- For each item in order of descending  $P[k]$ , let's add each item as a whole and then add the last item as a fraction of the weight.

Counterexample:

for each object  $k$  in order of **descending  $P[k]$**  {  
 $F[k] = \min (M/W[k], 1)$ ;  
 $M = M - F[k]*W[k]$ ;  
 $Total = Total + F[k]*P[k]$ ;  
}

	1	2	3	4	5	6	7	8	9
P	14	30	36	9	33	40	12	42	35
W	4	5	8	2	6	10	3	12	7
F						2/5		1	

<u>M</u>	<u>Total</u>	<u>k</u>	<u>F[k]</u>
16	0	8	1
4	42	6	2/5
0	58		

All other  $F[k] = 0$

- So this algorithm is not optimal. We put some of the heaviest items in first, which weren't the most profitable.

▼ Algorithm two

- Let's add items to the knapsack in order of increasing weight.
- This is not optimal either.

▼ Algorithm three

- Let's add things in order of descending  $P[k] / W[k]$  (highest profit to weight ratio).
- We're adding items in order of the most profit per unit of weight.
- This is actually the solution! We compute the value  $R$  (profit over weight value) and then just sort by that.