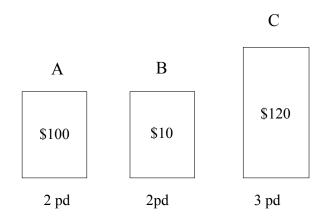
We already have seen an algorithmie technique in the Previous chapter names divide and conquer

In tems chapter, we introduce another such technique, colled Greedy algorithms

A greedy algorithm, mosty designed for an optimization problem, Starts with an empty solution and adds element to the subsolution based on a choice that Looks best at this moment Hoping that an optimal solution be achieved in this way. of the time, greedy solution to achieve optimality.

#### The Knapsack Problem...



Capacity of knapsack: K = 4

Fractional Knapsack Problem: Can take a fraction of an item.

0-1 Knapsack Problem:
Can only take or leave item. You can't take a fraction.

Geiven Plems with neight and xalve and a Knopsock with copuly k the objective to to put items inside the knopsock with moximum xolve.

#### Solution:

2 pd	2 pd C
\$100	\$80

#### Solution:

3 pd	
\$120	

#### The Fractional Knapsack Problem: Formal Definition

• Given K and a set of n items:

weight	$ w_1 $	W <sub>2</sub>	• • •	W <sub>n</sub>
value	$V_1$	<i>V</i> <sub>2</sub>	• • •	Vn

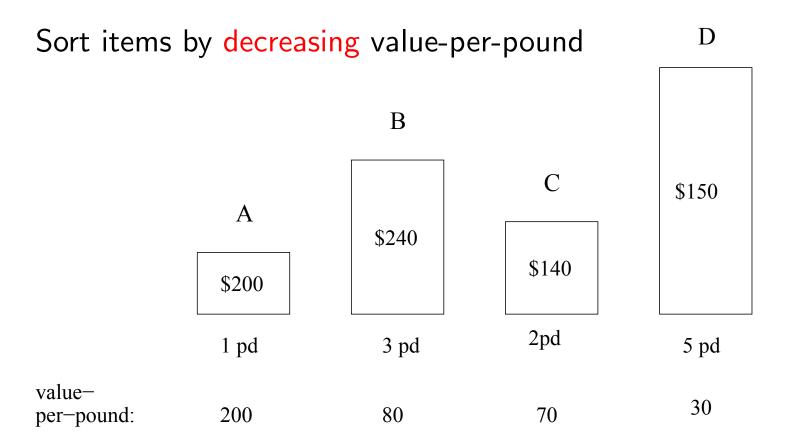
• Find:  $0 \le x_i \le 1$ ,  $i = 1, 2, \dots, n$  such that

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

and the following is maximized:

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

#### Greedy Solution for Fractional Knapsack



If knapsack holds K = 5 pd, solution is:

1	pd	Α
3	pd	В
1	pd	С

#### Greedy Solution for Fractional Knapsack

- Calculate the value-per-pound  $\rho_i = \frac{v_i}{w_i}$  for  $i = 1, 2, \ldots, n$ .
- Sort the items by decreasing  $\rho_i$ . Let the sorted item sequence be 1, 2, ..., i, ..., n, and the corresponding value-per-pound and weight be  $\rho_i$  and  $w_i$  respectively.
- Let k be the current weight limit (Initially, k = K). In each iteration, we choose item i from the head of the unselected list.
  - If  $k \ge w_i$ , set  $x_i = 1$  (we take item i), and reduce  $k = k w_i$ , then consider the next unselected item.
  - If  $k < w_i$ , set  $x_i = k/w_i$  ( we take a fraction  $k/w_i$  of item i), Then the algorithm terminates.

Running time:  $O(n \log n)$ .

#### Greedy Solution for Fractional Knapsack

- Observe that the algorithm may take a fraction of an item.
   This can only be the last selected item.
- We claim that the total value for this set of items is the optimal value.

#### Correctness

Given a set of n items  $\{1, 2, ..., n\}$ .

• Assume items sorted by per-pound values:  $\rho_1 \ge \rho_2 \ge ... \ge \rho_n$ .

Let the greedy solution be  $G = \langle x_1, x_2, ..., x_k \rangle$ 

•  $x_i$  indicates fraction of item i taken (all  $x_i = 1$ , except possibly for i = k).

Consider any optimal solution  $O = \langle y_1, y_2, ..., y_n \rangle$ 

- $y_i$  indicates fraction of item i taken in O (for all i,  $0 \le y_i \le 1$ ).
- Knapsack must be full in both G and O:  $\sum_{i=1}^{n} x_i w_i = \sum_{i=1}^{n} y_i w_i = K.$

Consider the first item i where the two selections differ.

• By definition, solution G takes a greater amount of item i than solution O (because the greedy solution always takes as much as it can). Let  $x = x_i - y_i$ .

#### Correctness...

Consider the following new solution O' constructed from O:

- For j < i, keep  $y'_j = y_j$ .
- Set  $y_i' = x_i$ .
- In O, remove items of total weight  $xw_i$  from items i+1 to n, resetting the  $y'_i$  appropriately.

This is always doable because  $\sum_{j=i}^{n} x_j = \sum_{j=i}^{n} y_j$ 

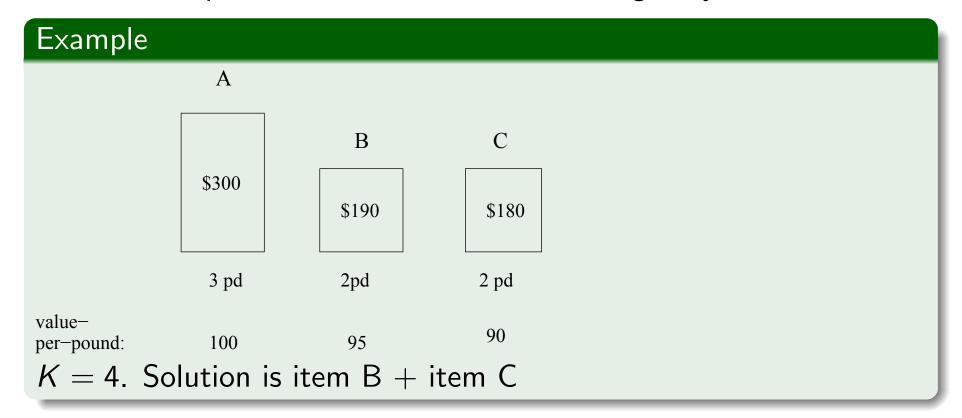
- The total value of solution O' is greater than or equal to the total value of solution O (why?)
- Since O is largest possible solution and value of O' cannot be smaller than that of O, O and O' must be equal.
- Thus solution O' is also optimal.

By repeating this process, we will eventually convert O into G, without changing the total value of the selection.

Therefore *G* is also optimal!

#### Greedy solution for 0-1 Knapsack Problem?

The 0-1 Knapsack Problem does not have a greedy solution!



#### Question

Suppose we tried to prove the greedy algorithm for 0-1 knapsack problem **does** construct an optimal solution. If we follow exactly the same argument as in the fractional knapsack problem where does the proof fail?

# **Greedy Scheduling**

https://cs.pomona.edu/classes/cs140/

# Scheduling (ignoring concurrency)

You have a shared resource
For example, a processor
You have many jobs that need to use the resource

#### Each job j has:

- A <u>Priority</u> P<sub>i</sub> that stands for the job's importance
- A <u>Duration</u> D<sub>i</sub> that stands for the length of time to run the job

In what sequence should we complete the jobs?

# Scheduling (ignoring concurrency)

#### In what sequence should we complete the jobs?

- What is our criteria? What do we want to optimize?
- Let's start by looking at job j's completion time C<sub>i</sub>
- Given three jobs:  $D_1 = 1$ ,  $D_2 = 2$ ,  $D_3 = 3$
- What is the completion time for each if they are scheduled in order?



What is the completion time of Job 5?



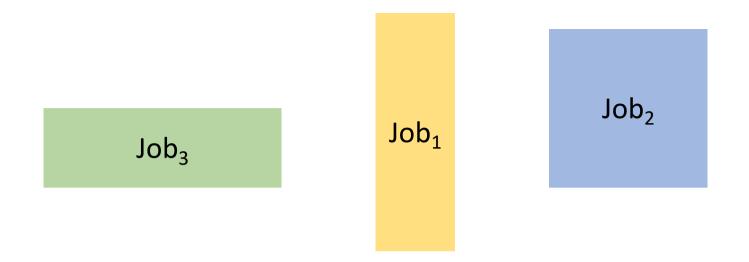
### Scheduling

Optimization objective: minimize the weighted sum of completion times

$$S_{cost} = \min[\sum_{j=1}^{N} P_j C_j]$$

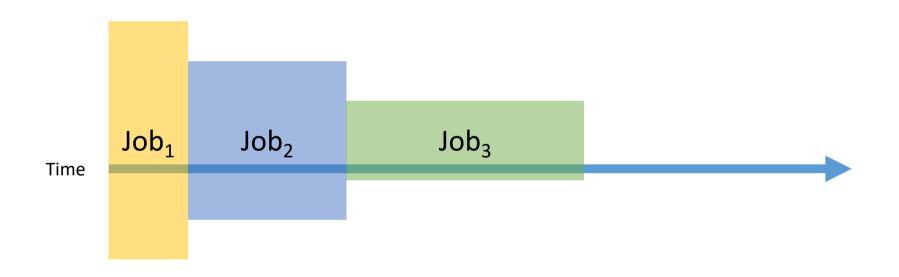
What is the weighted sum of completion times if we schedule the following jobs in order?

Job	$J_1$	J <sub>2</sub>	J <sub>3</sub>
Duration	D <sub>1</sub> = 1	$D_2 = 2$	$D_3 = 3$
Priority	$P_1 = 3$	$P_2 = 2$	$P_3 = 1$



Time

#### Exercise Question 1, 2, and 3



### Scheduling

Calculate the weighted sum of completion times for the following jobs if they are scheduled in the order: 1, 2, 3.

Job	J <sub>1</sub>	J <sub>2</sub>	J <sub>3</sub>
Duration	$D_1 = 1$	$D_2 = 2$	$D_3 = 3$
Priority	$P_1 = 3$	$P_2 = 2$	$P_3 = 1$
Completion			
Weight			

Weighted sum of completion times: ?

#### Greedy Scheduling

Our process for creating a greedy scheduling algorithm

- 1. Look at some special cases for our problem
- 2. Describe some possible greedy criteria
- 3. Compare our greedy criteria
- 4. Select the "best" one
- 5. Prove correctness if possible

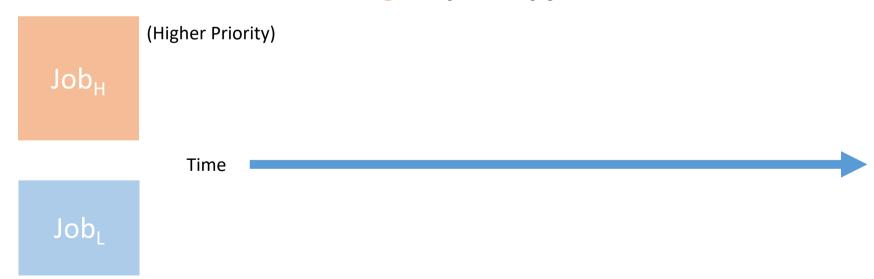
### Greedy Scheduling

Goal: devise a greedy algorithm to minimize the weighted sum of completion times

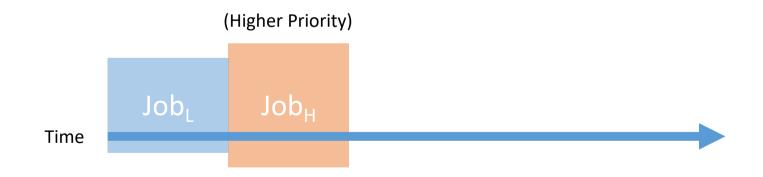
Why are we approaching this problem with a greedy algorithm?

- It's a pretty easy way to start.
- Compare the approach we go through in these slides with a Divide and Conquer approach

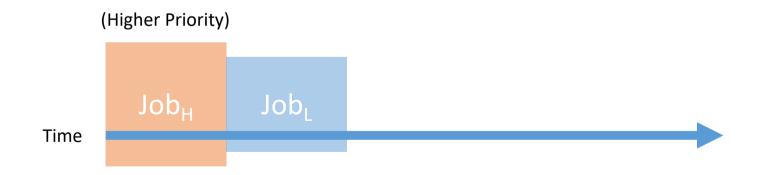
- These jobs have different priorities (P<sub>H</sub> and P<sub>L</sub>)
- Do we schedule the lower or higher priority job first?



- These jobs have different priorities (P<sub>H</sub> and P<sub>L</sub>)
- Do we schedule the lower or higher priority job first?



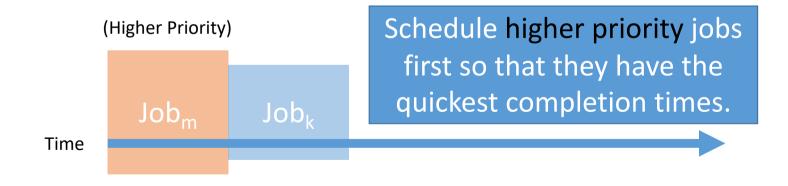
- These jobs have different priorities (P<sub>H</sub> and P<sub>L</sub>)
- Do we schedule the lower or higher priority job first?



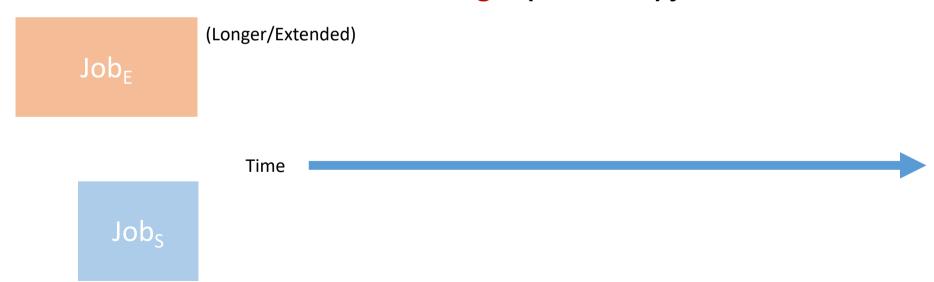
Schedule with Lower Priority First

Schedule with Higher Priority First

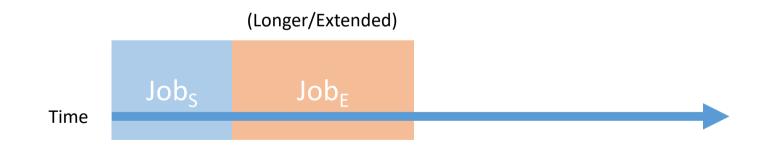
- These jobs have different priorities (P<sub>H</sub> and P<sub>L</sub>)
- Do we schedule the lower or higher priority job first?



- These jobs have different durations (D<sub>E</sub> and D<sub>S</sub>)
- Do we schedule the shorter or longer (Extended) job first?



- These jobs have different durations (D<sub>E</sub> and D<sub>S</sub>)
- Do we schedule the shorter or longer (Extended) job first?



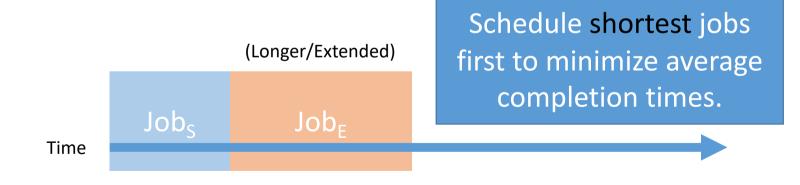
- These jobs have different durations (D<sub>E</sub> and D<sub>S</sub>)
- Do we schedule the shorter or longer (Extended) job first?



Schedule with Shorter Job First

Schedule with Longer Job First

- These jobs have different durations (D<sub>E</sub> and D<sub>S</sub>)
- Do we schedule the shorter or longer (Extended) job first?



#### 2. Describe some possible greedy criteria

What do we do when in the more general case:

- 1. Schedule highest priority first
- 2. Schedule shortest duration first

$$P_i > P_j \ and \ D_i > D_j$$
 (job i has higher priority and longer duration)

What are some simple scoring functions that aggregate our criteria?

We want a function for which jobs with a bigger score are scheduled first:

- Score increases for higher priorities
- Score increases for shorter times
- 1. Greedy Criterion 1:  $P_i D_i$  (take the difference)
- 2. Greedy Criterion 2:  $P_i/D_i$  (take the ratio)

Jobs will be ordered from biggest to smallest value

Job with same duration	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
Job 1: P=2, D=1		
Job 2: P=5, D=1		

• Jobs will be ordered from biggest to smallest value

	Job with same duration	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
	Job 1: P=2, D=1	1	2
Highest priority	Job 2: P=5, D=1	4	5
	Total weighted sum		

Jobs will be ordered from biggest to smallest value

	Job with same duration	Difference Metric $(P_i - D_i)$	Ratio Metric ( $P_i/D_i$ )	
	Job 1: P=2, D=1	1	2	
Highest priority	Job 2: P=5, D=1	4	5	
	Total weighted sum	5*1 + 2*2 = 9	5*1 + 2*2 = 9	Same Result

Job with same priority	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
Job 1: P=1, D=3		
Job 2: P=1, D=4		
Total weighted sum		

Jobs will be ordered from biggest to smallest value

	Job with same duration	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )	
	Job 1: P=2, D=1	1	2	
Highest priority	Job 2: P=5, D=1	4	5	
	Total weighted sum	5*1 + 2*2 = 9	5*1 + 2*2 = 9	Same Result

	Job with same priority	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
Shortest time	Job 1: P=1, D=3	-2	1/3
	Job 2: P=1, D=4	-3	1/4
	Total weighted sum		

Jobs will be ordered from biggest to smallest value

	Job with same duration	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )	
	Job 1: P=2, D=1	1	2	
Highest priority	Job 2: P=5, D=1	4	5	
	Total weighted sum	5*1 + 2*2 = 9	5*1 + 2*2 = 9	Same Result

	Job with same priority	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )	
Shortest time	Job 1: P=1, D=3	-2	1/3	
	Job 2: P=1, D=4	-3	1/4	
	Total weighted sum	1*3 + 1*7 = 10	1*3 + 1*7 = 10	Same Result

- Let's try to get them to disagree.
- Why does it matter if they don't produce the same result?
- One scoring metric must be better than the other

 Apply the two greedy algorithms and calculate their weighted sum of completion times

## 3. Compare our greedy criteria

Jobs will be ordered from biggest to smallest metric value

Job	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
Job 1: P=3, D=5		
Job 2: P=1, D=2		
Total weighted sum		

# 3. Compare our greedy criteria

Jobs will be ordered from biggest to smallest metric value

Job	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
Job 1: P=3, D=5	-2	3/5
Job 2: P=1, D=2	-1	1/2
Total weighted sum		

Which job goes first?

# 3. Compare our greedy criteria

Jobs will be ordered from biggest to smallest metric value

Job	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
Job 1: P=3, D=5	-2	3/5
Job 2: P=1, D=2	-1	1/2
Total weighted sum		

Which job goes first?

What is the priority sum?

#### 4. Select the "best" one

Jobs will be ordered from biggest to smallest metric value

Job	Difference Metric ( $P_i - D_i$ )	Ratio Metric ( $P_i/D_i$ )
Job 1: P=3, D=5	-2	3/5
Job 2: P=1, D=2	-1	1/2
Total weighted sum	1*2 + 3*7 = 23	3*5 + 1*7 = 22

Which job goes first?

What is the priority sum?

Which criteria is better?

## 5. Prove correctness if possible

Is criteria 2 optimal?

We don't know yet.

Claim: Criteria 2 is optimal for minimizing the weighted sum of completion times.

We're going to prove this using an exchange argument!

## Exchange Arguments

Consider your greedy solution, G

- Consider an alternative solution, A
  - A can be anything that is not G
  - Create A by changing G in some way

- Compare these solutions
  - Show that turning A into G makes A get better

#### Proof

- Assume that we have no ties (all  $P_i/D_i$  are distinct numbers)
- Fix an arbitrary input with n jobs
- Let's perform a proof using an exchange argument contradiction

Let G = the greedy schedule and A = the (alternative) optimal schedule

- Let's assume that A must be better than G (assume greedy is not optimal)
- To perform the contradiction, we must show that G is better than A, thus contradicting the purported optimality of A

### Proof

Let G = the greedy schedule and A = the optimal schedule

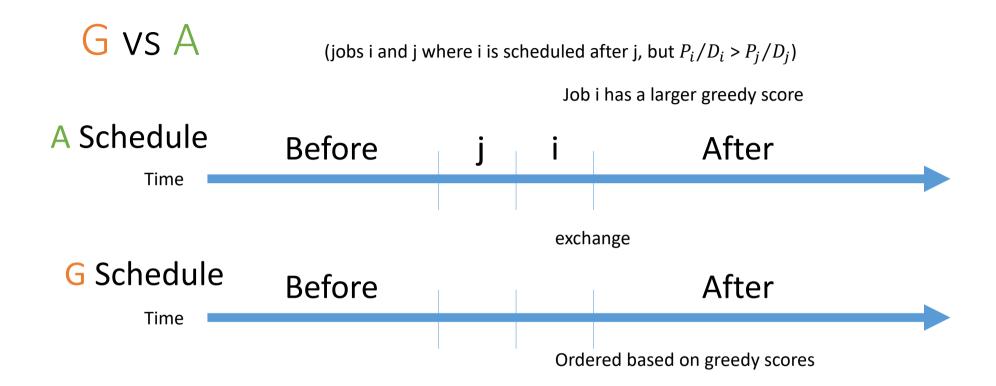
- Assume that:  $P_1/D_1 > P_2/D_2 > ... > P_n/D_n$
- We can just rename all jobs after we calculate their scores...
- Thus, G is just job 1 followed by job 2 etc. (1, 2, ..., n)

Reorde	Ratio	Length	Weight	Job ID
4	0.3	4	1	1
3	1.3	6	8	2
1	6.0	1	6	3
5	0.2	5	1	4
6	0.1	9	1	5
2	2.3	3	7	6

#### Proof

Let G = the greedy schedule and A = the optimal schedule

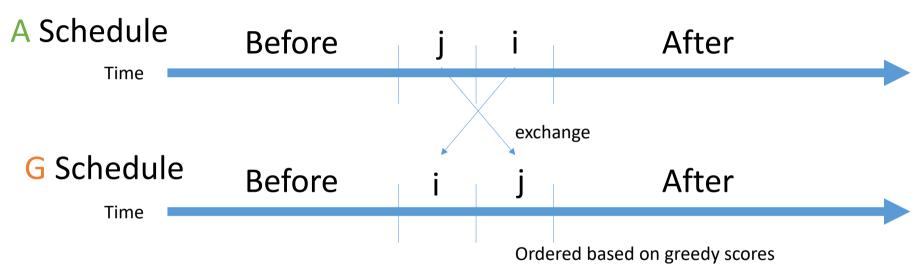
- Assume that:  $P_1/D_1 > P_2/D_2 > ... > P_n/D_n$
- We can just rename all jobs after we calculate their scores...
- Thus, G is just job 1 followed by job 2 etc. (1, 2, ..., n)
- For A there must be at least two jobs that are "out of order"
  - Specifically, jobs i and j where i is scheduled after j, but S<sub>i</sub> > S<sub>j</sub> (for example, Job<sub>5</sub> after Job<sub>6</sub>)
- The greedy schedule is the only schedule where the jobs are in order





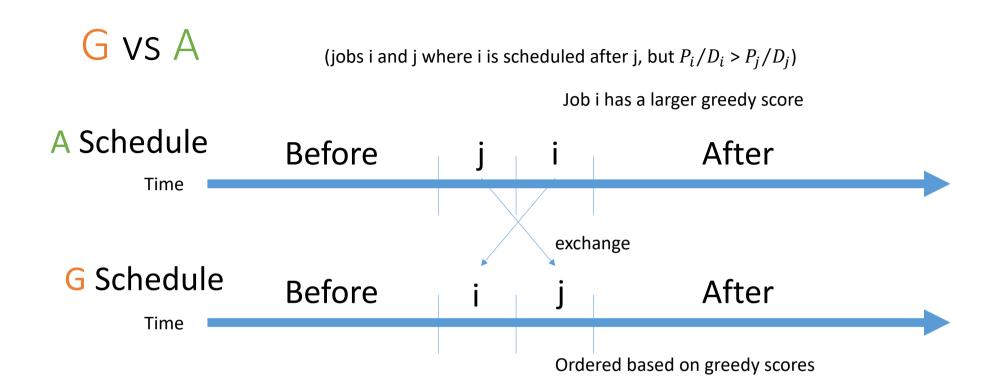
(jobs i and j where i is scheduled after j, but  $P_i/D_i > P_j/D_j$ )

Job i has a larger greedy score



How does the exchange affect the completion time for:

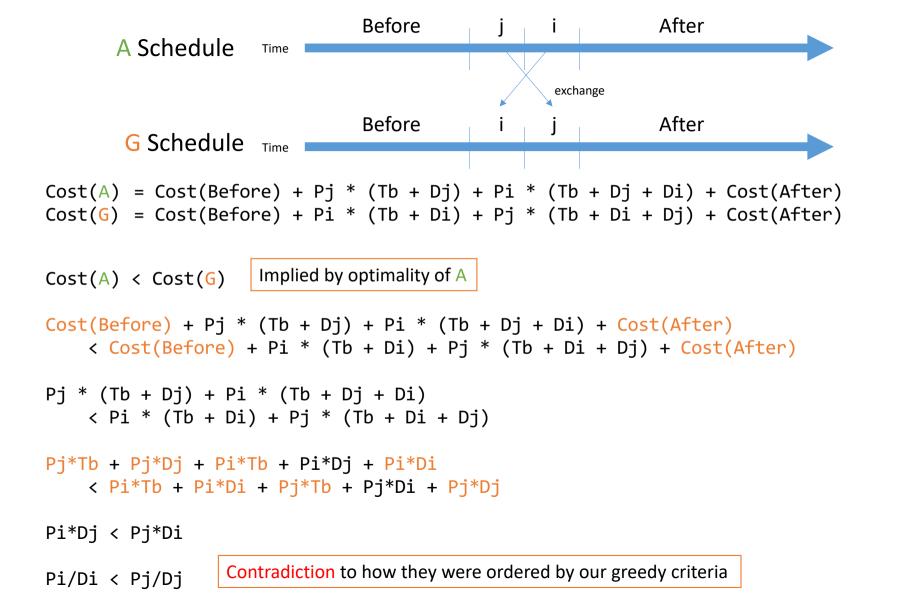
- 1. Jobs other than i and j?
- 2. Job i
- 3. Job j



What is the weighted sum of completion times for each schedule?

```
Before
                                                             After
      A Schedule
                     Time
                                                  exchange
                               Before
                                                             After
       G Schedule
Cost(A) = Cost(Before) + Pj * (Tb + Dj) + Pi * (Tb + Dj + Di) + Cost(After)
Cost(G) = Cost(Before) + Pi * (Tb + Di) + Pj * (Tb + Di + Dj) + Cost(After)
                     Implied by optimality of A
Cost(A) < Cost(G)
Cost(Before) + Pj * (Tb + Dj) + Pi * (Tb + Dj + Di) + Cost(After)
    < Cost(Before) + Pi * (Tb + Di) + Pj * (Tb + Di + Dj) + Cost(After)</pre>
```

```
After
                               Before
       A Schedule
                     Time
                                                  exchange
                                                            After
                               Before
       G Schedule
Cost(A) = Cost(Before) + Pj * (Tb + Dj) + Pi * (Tb + Dj + Di) + Cost(After)
Cost(G) = Cost(Before) + Pi * (Tb + Di) + Pj * (Tb + Di + Dj) + Cost(After)
                    Implied by optimality of A
Cost(A) < Cost(G)
Cost(Before) + Pj * (Tb + Dj) + Pi * (Tb + Dj + Di) + Cost(After)
    < Cost(Before) + Pi * (Tb + Di) + Pj * (Tb + Di + Dj) + Cost(After)
Pj * (Tb + Dj) + Pi * (Tb + Dj + Di)
    < Pi * (Tb + Di) + Pi * (Tb + Di + Di)
Pj*Tb + Pj*Dj + Pi*Tb + Pi*Dj + Pi*Di
    < Pi*Tb + Pi*Di + Pj*Tb + Pj*Di + Pj*Dj</pre>
```



# Summary of Greedy Scheduling

- Given n jobs, each with a priority and a duration
- Give each job a score based on their ratio of priority to duration
- Schedule jobs in <u>decreasing</u> order of their <u>score</u>
- This gives us an optimal schedule

- What do we do if we're given more jobs while these are running?
- Any issues with this scheme?
  - Some jobs might always be postponed.