

## Goals

- Practice Dynamic Programming Set-up
- Describe multi-exchange arguments

## Test Info

- All topics up to 3/5 except QuickSort
- All PSets up to 4
- Tuesday: review. Topics → Canvas Discussion until Sunday
- Wed @ noon → Frid @ 6 pm
- Reserve 3 hours
- No class 3/14
- 1 page, single sided handwritten notes

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Quiz on Canvas

No PSet 5

Scheduling Goal:

Minimize  $\sum_{j=1}^n w_j C_j \Leftarrow$  "Objective function"  
A

Greedy strategy: Order by largest  $w_i/t_i$

Q: If we get rid of assumption that all  $w_i/t_i$  are unique, what changes? Which change(s) make the proof fail?

Scheduling Goal:

$$\text{Minimize } \sum_{j=1}^n w_j C_j \quad \Leftarrow \text{"Objective function" } A$$

Greedy strategy: Order by largest  $w_i/t_i$

Q: If we get rid of assumption that all  $w_i/t_i$  are unique, why does our exchange argument fail?

A There might not be a unique solution

B) Objective function might not decrease from  $\sigma^*$  to  $\sigma^{*1}$

C) We can't create an ordering such that  $w_1/t_1 > w_2/t_2 > \dots > w_n/t_n$

All are true, but B ruins the proof.

Recall, we get a contradiction b/c  $\sigma^* = (\dots, k, j, \dots)$

$$\text{where } j < k \Rightarrow \frac{w_j}{t_j} > \frac{w_k}{t_k} \Rightarrow w_j t_k > w_k t_j$$

$$\Rightarrow A_{\sigma^*} - A_{\sigma^{*1}} = w_j t_k - w_k t_j > 0$$

Now, if  $w_i/t_i$  not unique:

$$j < k \Rightarrow \frac{w_j}{t_j} \geq \frac{w_k}{t_k} \Rightarrow w_j t_k \geq w_k t_j$$

$$\Rightarrow A_{\sigma^*} - A_{\sigma^{*'}} = w_j t_k - w_k t_j \geq 0$$

could be equal...  
not a contradiction

New Idea:

Still use **EXCHANGE** argument, but now need more exchanges.

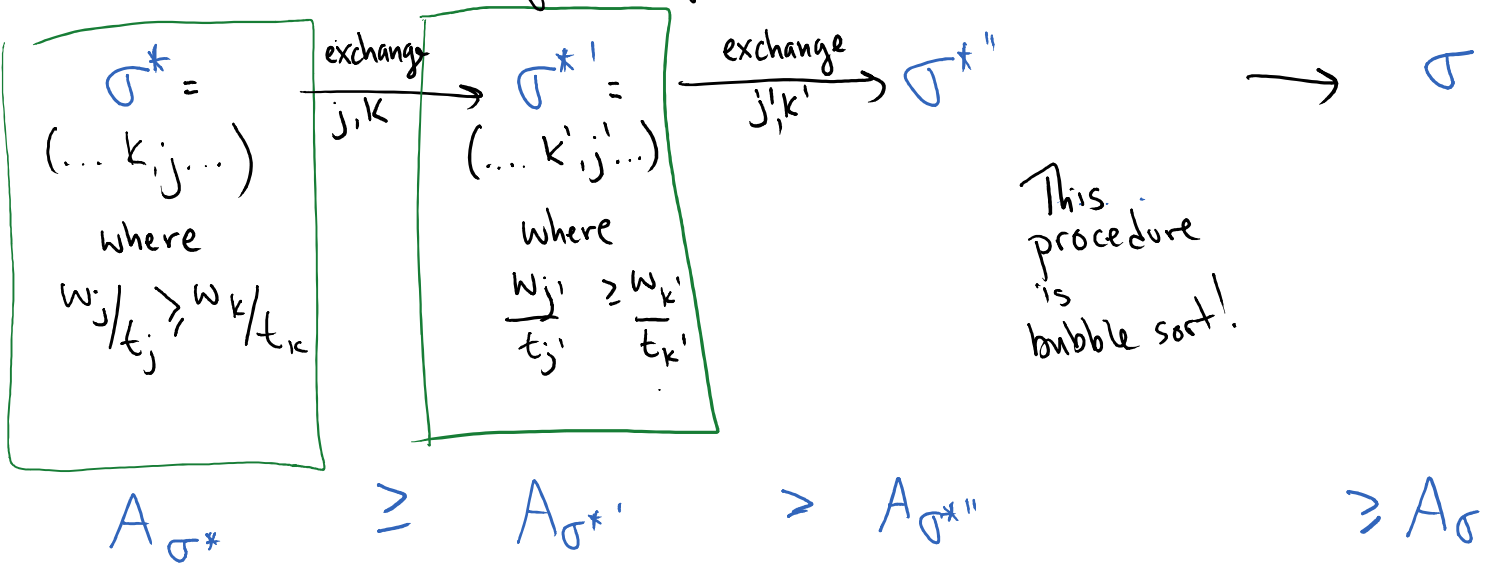
Choose some ordering s.t.

$$w_1/t_1 \geq w_2/t_2 \geq w_3/t_3 \dots \geq w_n/t_n$$

Let  $\sigma$  be strategy using this ordering

Let  $\sigma^*$  be any other strategy

We will show  $A_{\sigma^*} \geq A_{\sigma}$ , which means  $\sigma$  is optimal



This procedure is bubble sort!

Conclusion:  $A_{\sigma^*} \geq A_{\sigma}$ , so greedy strategy is optimal!

Q What is the runtime of this greedy algorithm?

A)  $O(1)$     B)  $O(n)$     C)  $O(n \log n)$     D)  $O(n^2)$

Conclusion:  $A_{\sigma^*} \geq A_{\sigma}$ , so greedy strategy is optimal!

Q What is the runtime of this greedy algorithm?

- A)  $O(1)$     B)  $O(n)$     C)  $O(n \log n)$     D)  $O(n^2)$

↙ now some might have the same value.

$O(n \log n)$ . Need to sort  $n$  items, requires  $O(n \log n)$  time, bubble sort is only imagined to happen as part of proof. It doesn't actually occur.

To Create D.P. (dynamic programming) algorithm:

1. Think of form of optimal solution.

- WMIS on line:  $v_n \in S$  or  $n \notin S_n$

2. Optimal soln in terms of smaller soln? (For each case)



3. Create recurrence for objective function

$$A[k] = \max \left\{ \underset{(i)}{A[k-1]}, \underset{(ii)}{A[k-2] + w_k} \right\}$$

$\uparrow$   
 weight of  
 MWIS on  
 $G_k$

4. Store  $A$  in an array using for-loop

5. Work backwards through array to reconstruct optimal solution

# Dynamic Programming - More Practice

Knapsack Problem  $K(v_1, \dots, v_n, w_1, \dots, w_n, W)$

Input: •  $n$  items, each has

- value  $v_i$
- size  $w_i$

}  $\mathbb{Z}^+$

• Capacity  $W$

Output: A subset  $S \subseteq \{1, 2, \dots, n\}$  that maximizes  $\sum_{i \in S} v_i = V(S)$   
 and satisfies  $w(S) = \sum_{i \in S} w_i \leq W$

} constraints

} objective function

## Applications

- Cargo trucks

- Investments  $\left\{ \begin{array}{l} v_i = \text{expected yield \%} \\ w_i = \text{min investment amount} \\ W = \text{amount to invest} \end{array} \right.$

## Notation

- Let  $K_{i,R}$  be subproblem on 1st  $i$  items, with capacity  $R$ .
- **Solution**: satisfies constraints
- **Optimal solution**: satisfies constraints and maximizes value.



## 1. Form of optimal solution:

Let  $S$  be the optimal solution to  $K_{n,w}$

Two options:

- (i)  $n \notin S$
- (ii)  $n \in S$

## 2. Relate to optimal solution of subproblem

(i) If  $S$  is optimal solution to  $K_{n,w}$  and  $n \notin S$ , then  $\text{---}$  is optimal solution to  $K_{\text{---}, \text{---}}$

Pf: For contradiction assume  $\text{---}$  is not optimal solution to  $K_{\text{---}, \text{---}}$  . . . . .

(ii) If  $S$  is optimal solution to  $K_{n,w}$  and  $n \in S$  then  $\text{---}$  is optimal solution to  $K_{\text{---}, \text{---}}$ .

Pf: For contradiction assume  $\text{---}$  is not optimal solution to  $K_{\text{---}, \text{---}}$  . . . . .

(i) If  $S$  is the optimal soln to  $K_{n,w}$  and  $n \notin S$ , then  $S$  is optimal soln to  $K_{n-1,w}$

Pf: For contradiction, suppose  $S$  is not opt soln for  $K_{n-1,w}$ . Then there exists an optimal solution  $S'$  for  $K_{n-1,w}$  with  $V(S') > V(S)$ . But  $S'$  is also a solution to  $K_{n,w}$  with  $V(S') > V(S)$  and  $n \notin S'$ , so  $S$  is not the optimal solution, a contradiction

(ii) If  $S$  is opt. soln for  $K_{n,w}$  and  $n \in S$ , then  $S - \{n\}$  is opt soln for  $K_{n-1, w-w_n}$ .

Pf: Suppose for contradiction  $S - \{n\}$  is not optimal for  $K_{n-1, w-w_n}$ . Then there exists an optimal solution  $S'$  for  $K_{n-1, w-w_n}$  s.t.  $V(S') > V(S - \{n\})$ . But then  $S' + \{n\}$  is a soln to  $K_{n,w}$ , and  $V(S' + \{n\}) > V(S)$ , so  $S$  is not opt for  $K_{n,w}$ , a contradiction