

*Note:* Your TA probably will not cover all the problems. This is totally fine, the discussion worksheets are deliberately made long so they can serve as a resource you can use to practice, reinforce, and build upon concepts discussed in lecture, readings, and the homework.

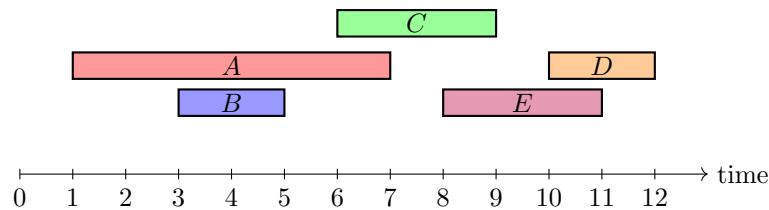
## Graphs and Greedy

### 1 Interval Scheduling

In lecture we saw the interval scheduling problem. For the review session, we will take a look at it again and focus on understanding the exchange argument.

You are given  $n$  intervals, where interval  $i$  has start time  $s_i$  and finish time  $f_i$ . Two intervals are *compatible* if they do not overlap. Your goal is to select the largest possible set of mutually compatible intervals.

**Example:** Consider the following 5 intervals:



The optimal solution is  $\{B, C, D\}$  with 3 intervals.

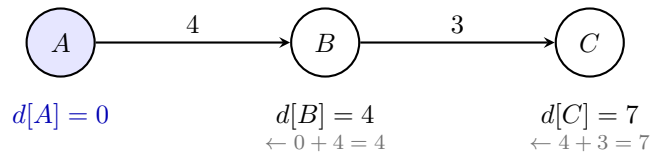
**Algorithm:** Sort intervals by finish time. Greedily select intervals: always pick the next interval that doesn't overlap with the last selected one.

Prove that this greedy algorithm is optimal using an exchange argument.

## 2 Quick Graph Algorithms Review

All four algorithms in the table below share a single core operation: **edge relaxation**. The key insight is that *global shortest paths are found by repeatedly applying a simple local update*.

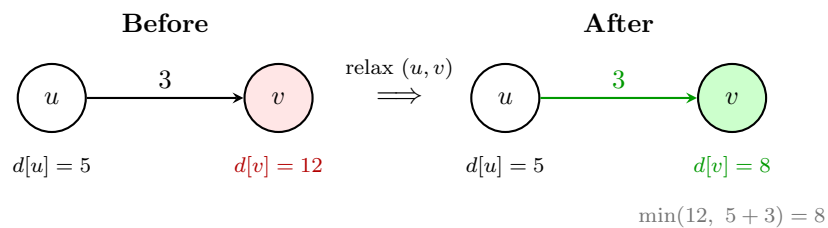
**Why “local”?** Consider the chain below. Node  $C$  cannot know its distance from  $A$  until  $B$  has figured out its *own* distance first. Once  $d[B]$  is settled,  $C$  simply asks: “What is  $B$ ’s distance, plus the edge weight to reach me from  $B$ ?”



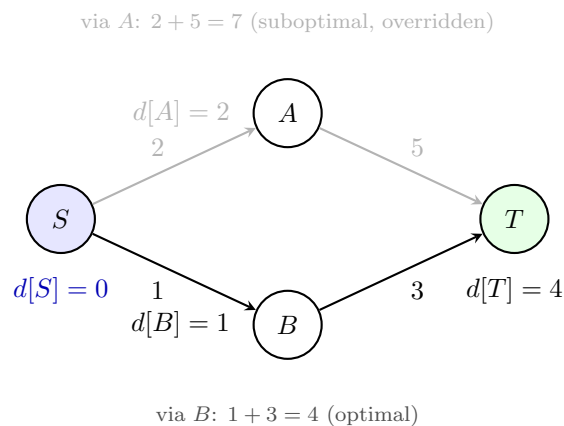
This operation — updating  $d[v]$  based on a neighbor  $u$  — is called **relaxing** edge  $(u, v)$ :

$$d[v] \leftarrow \min(d[v], d[u] + w(u, v))$$

**Relaxation in action.** A single relaxation step can improve a distance estimate:



**Multiple paths competing.** When two paths lead to the same node, relaxation automatically keeps only the shorter one:



The four algorithms below all apply this same relaxation rule — they differ only in *which edges they relax* and in *what order*.

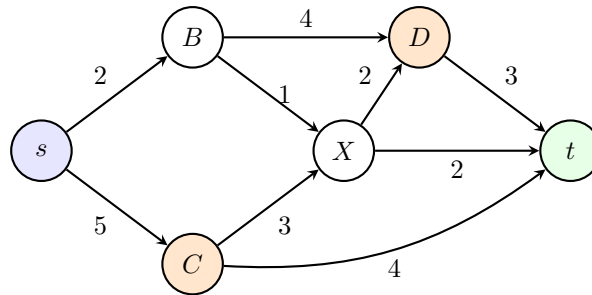
Algorithm	Key Features	Runtime
DFS	Cycle detection, topological sort, SCCs	$O( V  +  E )$
BFS	Layer-by-layer (“onion”); shortest paths in <i>unweighted</i> graphs	$O( V  +  E )$
Dijkstra	Like BFS but with non-uniform layer sizes; fails with negative edge weights	$O(( V  +  E ) \log  V )$
Bellman-Ford	Handles negative edges; detects negative weight cycles	$O( V  \cdot  E )$

### 3 Eat Before Class

You are given a map of the Berkeley campus represented as a directed, weighted graph  $G = (V, E)$  with positive edge weights. You start at your home node  $s \in V$ , and you need to get to your class at node  $t \in V$ . However, you are hungry and want to grab food on the way. You are given a set of nodes  $F \subseteq V$  that represent your favorite food locations.

Your task is to design an efficient algorithm to find the length of the shortest path from  $s$  to  $t$  that visits at least one food location  $f \in F$ .

**Example:** Consider the following graph where  $F = \{C, D\}$ .



In this example, the shortest path from  $s$  to  $t$  without any restrictions is  $s \rightarrow B \rightarrow X \rightarrow t$  with length  $2 + 1 + 2 = 5$ . However, this path does not visit any node in  $F$ . The shortest path that visits a food location is  $s \rightarrow B \rightarrow X \rightarrow D \rightarrow t$  with length  $2 + 1 + 2 + 3 = 8$ , which visits  $D \in F$ . Another valid path is  $s \rightarrow C \rightarrow t$  with length  $5 + 4 = 9$ , which visits  $C \in F$ . The shortest valid path is  $s \rightarrow B \rightarrow X \rightarrow D \rightarrow t$  with length 8.

Describe your algorithm, prove its correctness, and analyze its runtime.

# Divide & Conquer and Dynamic Programming

## 1 Maximum Subarray Sum

In homework we saw this problem. For review, we will try to focus on learning how recursion works.

Given an array  $A$  of  $n$  integers, the *maximum subarray sum* is the largest sum of any contiguous subarray of  $A$  (including the empty subarray). In other words, the maximum subarray sum is:

$$\max_{i \leq j} \sum_{k=i}^j A[k]$$

For example, the maximum subarray sum of  $[-1, 3, 4, -2, 7, 8, -3, -1]$  is 20, the sum of the contiguous subarray  $[3, 4, -2, 7, 8]$ .

-1	3	4	-2	7	8	-3	-1
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sum = 3 + 4 + (-2) + 7 + 8 = 20

Design an  $O(n \log n)$ -time divide-and-conquer algorithm for this problem.

## 2 Knapsack with Repetition

You have a knapsack with capacity  $W$  and  $n$  types of items. Item  $i$  has weight  $w_i$  and value  $v_i$ . You may use each item type as many times as you like (unlimited supply). Find the maximum total value of items that fit in the knapsack.

**Example:** Suppose  $W = 10$  and we have:

Item	Weight $w_i$	Value $v_i$
1	6	30
2	3	14
3	4	16