

# Math 182 Lecture 13

## Chapter 7 Greedy Algorithms

### §7.1 An activity selection problem

Greedy: making locally optimal choice

Standard example: cashier making change.

$S = \{a_1, \dots, a_n\}$  proposed activities

$a_i$ : start time  $s_i$ , finish time  $f_i$ :

$$0 \leq s_i < f_i < \infty$$

$a_i$  takes place during  $[s_i, f_i]$ )

We say  $a_i, a_j$  compatible if

$$[s_i, f_i] \cap [s_j, f_j] = \emptyset.$$

The **activity-selection problem**, we want to find a maximum-size subset of  $S = \{a_1, \dots, a_n\}$  of mutually-compatible activities. Note that we want to maximize the *number* of activities scheduled, not necessarily the total *duration* of the activities which are scheduled.

By convention we'll assume  
 $f_1 \leq f_2 \leq \dots \leq f_{n-1} \leq f_n$ .

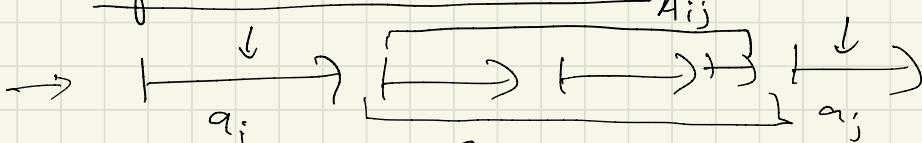
Example:

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	8	8	10	11	12	14	16

$\{a_3, a_9, a_{11}\}$  mult. comp. size 3

$\{a_1, a_4, a_8, a_{11}\}$  mult. comp. size 4  
or  $\{a_2, a_7, a_9, a_{11}\}$  optimal sols.

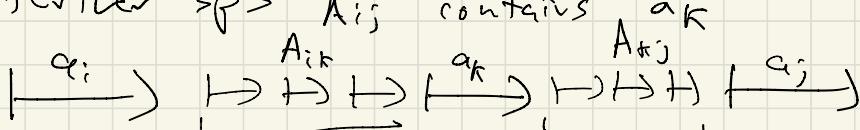
Optimal substructure  $S = \{a\text{|| activities}\}$



Let  $S_{i:j} = \{a\text{|| activities starting after } a_i \text{ and ending before } a_j\}$

want activities between  $a_i$  and  $a_j$  to be optimal sol +  $S_{i:j}$ .

further  $S_{i:j}$   $A_{i:j}$  contains  $a_k$



define  $A_{ik} := A_{i:j} \cap S_{ik}$

$A_{kj} := A_{i:j} \cap S_{kj}$

Then  $A_{ik}/A_{kj}$  optimal sol to  $S_{ik}/S_{kj}$ .

$$A_{ij} = A_{ik} \cup \{a_k\} \cup A_{kj}$$

$$\Rightarrow |A_{ij}| = |A_{ik}| + |A_{kj}| + 1.$$

Let  $c[i,j] :=$  size of optimal sol of  $S_{ij}$

$$c[i,j] = \begin{cases} 0 & \text{if } S_{ij} = \emptyset \\ \max_{a_k \in S_{ij}} c[i,k] + c[k,j] + 1 & \text{if } S_{ij} \neq \emptyset \end{cases}$$

Making greedy choice

(This would work)  
 although we can do better bcs we can say which  $k$  gives us a max

Sps we wanted to commit to one activity which is guaranteed to be part of some optimal sol.

Idea: commit to activity w/ earliest finish time.

Need to show activity w/ earliest stopping time always part of sue optimal solution.

Define:  $S_K := \{a_i \in S : s_i \geq f_k\}$   
 $= \{\text{all activities which start after } a_k \text{ finishes}\}.$

Theorem Consider any nonempty subproblem  $S_K$ , and let  $a_m$  be an activity in  $S_K$  w/ earliest finish time.

Then  $a_m$  is included in some maximize-size subset of mutually compatible activities of  $S_k$ .

Proof: "Exchange argument"

(Idea: the greedy choice is just as good as any other choice which is part of optimal sol.)

Let  $A_k \subseteq S_k$  be maximum size compatible subset. Let  $a_j \in A_k$  be activity w/ first start time.

Case 1:  $a_j = a_m$ . Done ✓.

Case 2:  $a_j \neq a_m$ . Consider "exchanged" set  $A'_k := A_k \setminus \{a_j\} \cup \{a_m\}$ .

$A'_k$  activities are disjoint bcs activities in  $A_k \setminus \{a_j\}$  are disjoint. Also since  $f_m \leq f_j \leq S_i \quad \forall i \in A_k \setminus \{a_j\}$  and  $i \Rightarrow$  disjoint w/  $A_k \setminus \{a_j\}$ .

Also  $|A_k| = |A'_k|$  thus  $A'_k$  also optimal sol. which contains  $a_m$ . 

Remark: Since we know a choice guaranteed to work, better to implement as top-down instead of bottom-up.

## Recursive greedy algorithm

$$S = \langle s_0, s_1, s_2, \dots, s_n \rangle$$

$$f = \langle f_0, f_1, f_2, \dots, f_n \rangle$$

$n = \# \text{ of original activities}$

$k = \text{index for subproblem } S_k \text{ to solve.}$

RECURSIVE-ACTIVITY-SELECTOR( $s, f, k, n$ )

- 1  $m = k + 1$
- 2 **while**  $m \leq n$  and  $s[m] < f[k]$  // find the first activity in  $S_k$  to finish
- 3      $m = m + 1$
- 4 **if**  $m \leq n$
- 5         **return**  $\{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)$
- 6 **else return**  $\emptyset$

(call Rec-Act-Set ( $s, f, 0, n$ ) will  
find optimal sol for original problem  $S$ )

Line 1 we first consider activity  $a_{k+1}$ .

Lines 2-3 iterate through  $a_{k+1}, \dots, a_m$   
until find comp. activity w/ earliest  
start time

If this happens, make this greedy  
choice in line 5.

O/w if no activities left, return  $\emptyset$ .

Running time:  $\Theta(n)$  bcs m

takes values  $1, \dots, m$  exactly once  
across all recursive calls.

# Iterative greedy algorithm

$$S = \langle s_1, \dots, s_n \rangle$$
$$f = \langle f_1, \dots, f_n \rangle$$

GREEDY-ACTIVITY-SELECTOR( $s, f$ )

```
1   $n = s.length$ 
2   $A = \{a_1\}$ 
3   $k = 1$ 
4  for  $m = 2$  to  $n$ 
5      if  $s[m] \geq f[k]$ 
6           $A = A \cup \{a_m\}$ 
7           $k = m$ 
8  return  $A$ 
```

$\Theta(n)$

Line 2 make first greedy choice  
Line 3 says first problem to  
consider is  $s_1$ .

Lines 4-7 iterates through activities

$k$  keeps track of  $s_k$

if  $a_m$  compatible w/  $a_k$   
choose (greedily)  $a_m$  in line 6.

# Chapter 8 Elementary graph algorithms

## §8.1 Representations of graphs

Given a graph  $G = (V, E)$

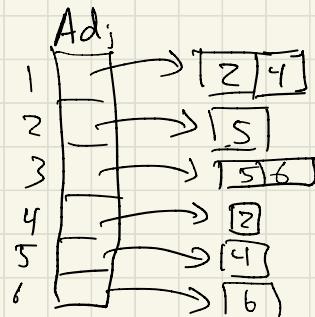
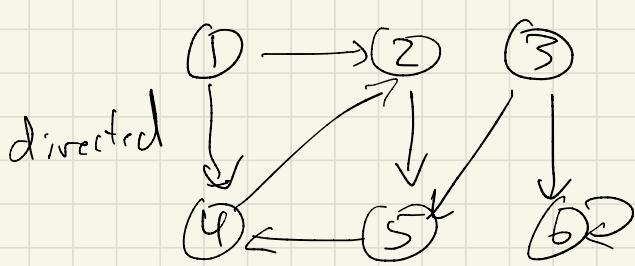
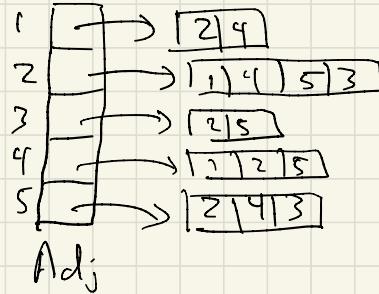
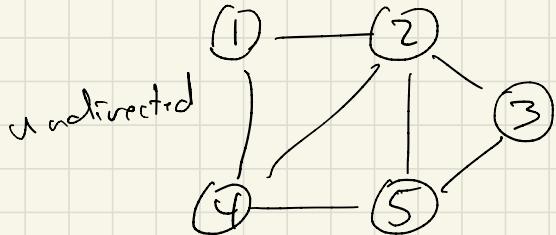
$$V = \{\text{vertices}\}$$

$$E = \{\text{edges}\} \subseteq V \times V$$

can represent  $G$  in two ways:

### 1: Adjacency-list representation

Graph has an array  $\text{Adj}$  of length  $|V|$  s.t. for every  $v \in V$   $\text{Adj}[v]$  is a list of all vertices  $v$  is adjacent to.



In pseudocode will write  
 G. Adj  
 or G. Adj[u]

Remarks • In general we'll use adjacency-lists.

- They use  $\Theta(V+E)$  memory.
- Good way of representing "sparse" graphs w/  $|E| \ll |V|^2$
- Disadvantage: asking " $(u,v) \in E?$ " can't be done in constant time.

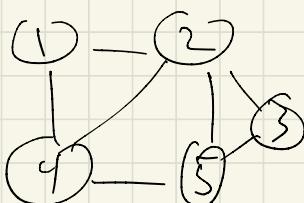
2: Adjacency-matrix representation.

Given  $G = (V, E)$  define  
 the adjacency matrix  $A = (a_{ij})$   
 of size  $|V| \times |V|$  by

$$a_{ij} := \begin{cases} 0 & \text{if } (i,j) \notin E \\ 1 & \text{if } (i,j) \in E \end{cases}$$

where  $V = \{1, \dots, |V|\}$ .

Ex:



1	0	1	0	1	0
2	1	0	1	0	1
3	0	1	0	0	1
4	1	1	0	0	1
5	0	1	1	1	0

1 2 3 4 5

Remarks •  $\Theta(|V|^2)$  memory

- Question " $(i,j) \in E?$ " can be answered in  $\Theta(1)$  time.

## §8.2 Breadth-first search (BFS)

Goal: Given  $G = (V, E)$  and distinguished  $s \in V$  called the source.  
Want to visit all nodes which are "reachable" from  $s$ .

BFS does the following:

- (1) Compute distance to  $s$  of all vertices reachable from  $s$ .
- (2) It produces a "breadth-first tree" of all vertices reachable from  $s$  w/ root  $s$ . (i.e., will produce spanning tree of connected component containing  $s$ )
- (3) For any  $v$  reachable from  $s$ , the unique simple path from  $s$  to  $v$  in tree is a shortest path from  $s$  to  $v$  in  $G$ .

"Breadth-first" refers to fact that algorithm will visit all vertices distance 1 away, then distance 2 away, then distance 3, etc.

BFS tags vertices with three attributes:

color: vertex white if undiscovered  
 v. color gray or black if discovered already  
 grey vertices are on the frontier  
 might have white neighbors.  
 black vertices have gray or black ~~vert~~  
 neighbors.  
distance best known distance to s.  
 v.d initialized to  $\infty$ .

predecessor the neighbor which first discovered  
 v.  
 v. $\pi$  will be initialized to NIL.

BFS( $G, s$ )

```

1  for each vertex  $u \in V, V \setminus \{s\}$ 
2     $u.\text{color} = \text{WHITE}$ 
3     $u.d = \infty$ 
4     $u.\pi = \text{NIL}$ 
5   $s.\text{color} = \text{GRAY}$ 
6   $s.d = 0$ 
7   $s.\pi = \text{NIL}$ 
8   $Q = \emptyset$ 
9  ENQUEUE( $Q, s$ )
10 while  $Q \neq \emptyset$ 
11    $u = \text{DEQUEUE}(Q)$ 
12   for each  $v \in G.\text{Adj}[u]$ 
13     if  $v.\text{color} == \text{WHITE}$ 
14        $v.\text{color} = \text{GRAY}$ 
15        $v.d = u.d + 1$ 
16        $v.\pi = u$ 
17       ENQUEUE( $Q, v$ )
18    $u.\text{color} = \text{BLACK}$ 

```