# 6.006 Final Review Session II

Graph Algorithms

Dynamic Programming

## **Dynamic Programming**

Summary: \* DP ≈ "careful brute force" \*DP \approx guessing + recursion + memoization \*DP \approx dividing into reasonable # subproblems whose solutions relate — acyclicly— usually via guessing parts of solution \* time = # subproblems · time/subproblem treating recursive calls as 0(1) = # subproblems · # guess choices · time/guess - essentially an amortization
- count each subproblem only once;
after first time, costs O(1) via memoization \* DP often ~ shortest paths in some DAG

## **Dynamic Programming**

\* 5 easy steps to dynamic

(1) define subproblems
(2) guess (part of solution) programming: count # choices relate subprob. solutions compute time/subprob. recurse + memoize or build DP table bottom-up #subprobs.

- check subprobs. acyclic/topological order

5) solve original problem: = a subproblem
or by combining subprob. solutions (>) extra time)

## **Dynamic Programming**

#### Good practice problems:

Fall 2012 final: Problem 12 -- Optimizing Santa

Fall 2014 final: Problem 5 -- Hanging posters

> Fall 2014 final: Problem 6 -- Palindromes

> Spring 2014 final: Problem 7 -- 3D blocks

Spring 2014 final: Problem 10

Spring 2015 final: Problem 9-b

Link to 6.006 from previous semesters: http://courses.csail.mit.edu/6.006/

## Dynamic Programming (Fall 2014 final)

#### **Problem 6. Palindromes** [20 points]

Given a string  $x = x_1, ..., x_n$ , design an efficient algorithm to find the minimum number of characters that need to be inserted to make it a palindrome (recall that a palindrome is a string such as "racecar" the reads the same backwards). Analyze the running time of your algorithm and justify its correctness.

For example, when x = "ab3bd", we need to insert two characters (one "d" and one "a"), to get either the of the palindromes "dab3bad" or "adb3bda".

- Define subproblems. How many are there?
- 2. What are the possible guesses we have to consider?
- 3. Define recurrence.
- Compute time per subproblem.
- 5. Obtain running time.
- 6. Solve original problem

## Dynamic Programming (Spring 2014 final: Problem 7 -- 3D blocks)

**Problem 7.** [15 points] Suppose you are given a set of n rectangular three-dimensional blocks, where block  $B_i$  has length  $l_i$ , width  $w_i$ , and height  $h_i$ , all real numbers. You are supposed to determine the maximum height of a tower of blocks that is as tall as possible, using any subset of the blocks.

#### There are two constraints:

- 1. You are not allowed to rotate the blocks: the length always refers to the east-west direction, the width is always north-south, and the height is always up-down.
- 2. You are only allowed to stack block  $B_i$  on top of block  $B_j$  if  $l_i \leq l_j$  and  $w_i \leq w_j$ , that is, the two dimensions of the base of block  $B_i$  are no greater than those of block  $B_j$ .
- 1. Assume that I[i] <= I[j] when i < j (this assumption can be weakened if we sort all blocks according to I and breaking ties according to w)
- 2. Define subproblems. How many are there?
- 3. What are the possible guesses we have to consider?
- 4. Define recurrence.
- 5. Compute time per subproblem. Obtain running time.
- 6. Solve original problem

## All pairs shortest paths

```
Floyd-Warshall algorithm - O(|V|^3)
Subproblems: P(i,j,k):
```

Find shortest path from i to j using only nodes <=k as intermediate.

```
1 let dist be a |V| \times |V| array of minimum distances initialized to \infty (infinity)
2 for each vertex V
3   dist[v][v] \leftarrow 0
4 for each edge (u,v)
5   dist[u][v] \leftarrow w(u,v) // the weight of the edge (u,v)
6 for k from 1 to |V|
7   for i from 1 to |V|
8   for j from 1 to |V|
9         if dist[i][j] > \text{dist}[i][k] + \text{dist}[k][j]
10         dist[i][j] \leftarrow \text{dist}[i][k] + \text{dist}[k][j]
11   end if
```

## All pairs shortest paths

Johnson's algorithm -  $O(|V||E|+|V|^{2*}log|V|)$ 

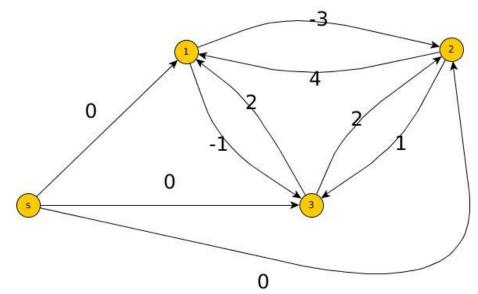
#### Idea:

- Reduce to a graph that has non-negative weights: w<sub>h</sub>(u,v)= w(u,v)+h(u)-h(v)
   (Find appropriate function h using Bellman-Ford) O(|V||E|)
- Run Dijkstra's algorithm |V| times to compute δ<sub>h</sub>(u,v) O(|V||E|+|V|<sup>2\*</sup>log|V|)
- Recover the original shortest path lengths  $\delta(u,v)=\delta_h(u,v)+h(v)-h(u)-O(|V|^2)$

## Example of solving difference constraints

Adapted from Problem 9 in Spring 14 Final: Solve the following difference constraints:

$$x_1 \le x_2 + 4$$
 $x_2 \le x_1 - 3$ 
 $x_1 \le x_3 + 2$ 
 $x_3 \le x_1 - 1$ 
 $x_2 \le x_3 + 2$ 
 $x_3 \le x_2 + 1$ 

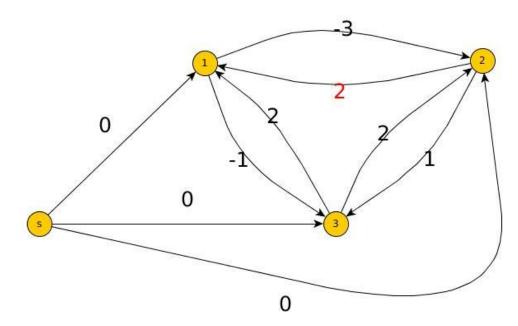


## Example of solving difference constraints

Adapted from **Problem 9 in Spring 14 Final**: Solve the following difference

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 $x_3 \le x_1 - 1$ 
 $x_2 \le x_3 + 2$ 
 $x_3 \le x_2 + 1$ 



## Shortest paths problems

Adapted from **Spring 11 final Problem 5**: Compute all pairs shortest paths in a graph G=(V,E,w) where the weight of paths that have **at least 11 edges** is **doubled**.

- Create a graph G'=(V',E',w') that has 12 copies of the graph G
- For every edge {u,v} in E, add {u<sub>i</sub>,v<sub>i+1</sub>} (0<= i <=10). Also add {u<sub>11</sub>,v<sub>11</sub>}
- Run Floyd-Warshall on G'
- For each  $\{u,v\}$  in E pick the smallest among  $\{\delta(u_0,v_i)|0 \le i \le 10\}U\{2*\delta(u_0,v_1)\}$

**Alternatively:** Modify Floyd-Warshall to keep track of shortest paths between i and j, having exactly k edges (k<=10) or >10 edges.

## **Graph Representations**

#### Adjacency Matrices:

Adjacency matrix A is a 2D matrix,  $[1, \ldots, n, 1, \ldots, n]$ .

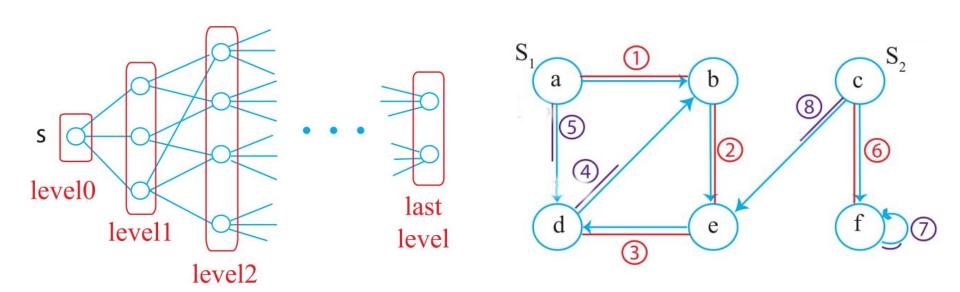
$$A[i,j] = \begin{cases} 1 & \text{if } (i,j) \in E \\ 0 & \text{if } (i,j) \notin E \end{cases}$$

#### **Adjacency Lists**

Array Adj of |V| linked lists

• for each vertex  $u \in V$ , Adj[u] stores u's neighbors, i.e.,  $\{v \in V \mid (u,v) \in E\}$ . (u,v) are just outgoing edges if directed.

## BFS / DFS



## BFS / DFS

```
DFS (V, Adj)
BFS (V,Adj,s):
                                                    parent = \{\}
    level = \{ s: 0 \}
    parent = \{s : None \}
                                                    for s in V:
    i = 1
                                                         if s not in parent:
    frontier = [s]
                                                               parent [s] = None
    while frontier:
         next = []
                                                               DFS-visit (Adj, s)
         for u in frontier:
                                            DFS-visit (Adj, s):
            for v in Adj [u]:
               if v not in level:
                                                  for v in Adj [s]:
                   level[v] = i
                                                        if v not in parent:
                   parent[v] = u
                   next.append(v)
                                                               parent [v] = s
         frontier = next
                                                               DFS-visit (Adj, v)
         i + =1
```

## BFS / DFS Applications

BFS DFS

• Shortest paths in unweighted graphs

- Edge Classification
- Topological Sort
- Cycle Detection

Runtimes: O(V+E) [using adjacency lists]

## Single Source Shortest Paths

**Path:** A sequence of vertices  $(v_0, v_1, \dots, v_k)$ , such that,

for all  $0 \le i \le k$ ,  $(v_i, v_{i+1})$  is in E.

**Simple Path:** A sequence of vertices  $(v_0, v_1, ..., v_k)$ , such that,

For all  $0 \le i \le k$ ,  $(v_i, v_{i+1})$  is in E and

for all  $0 \le i$ ,  $j \le k$ , if i = -j then  $v_i = -j$ .

## Single Source Shortest Paths

**Problem:** Given a graph G and a vertex s, find the shortest paths from s to each other vertex in G.

#### First Try

Brute Force Algorithm: Try every possible path. Running time =  $2^{\Omega(n)}$ 

## Relaxation

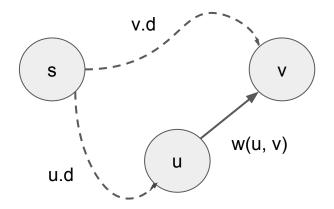
Would it be shorter if I use (u, v)?

```
relax(u, v):

if u.d + w(u, v) < v.d

v.d = u.d + w(u, v)

v.parent = u
```



## Shortest Paths in DAGs

```
SP_DAG(G):
                                                  Running Time: O(V+E)
    initialization(V)
    topo_sort(G)
    for each u in V in topological order:
         for each v in Adj[u]:
             relax(u, v)
```

## Bellman - Ford Algorithm

Fix an ordering of edges  $\langle e_1, e_2, \dots, e_{|E|} \rangle$ .

#### **Bellman-Ford(G):**

initialization(V)

for i = 1 to |V| - 1:

for j = 1 to |E|:

relax(e<sub>i</sub>)

#### **Running Time:** O(VE)

Can handle negative weights.

Can find negative weight cycles.

## Dijkstra's Algorithm

- Works out one node at a time
- Maintains a priority queue of best path weights to nodes
- Runtime is O(V\*(EXTRACT-MIN)+E\*(DECREASE+KEY))
- With a Fibonacci heap, this is O(V log V + E)

# Dijkstra's: example problem

A large metropolitan area's police departments would like to compute the shortest path from any police station to any intersection in the area, where each portion of a street between intersections has a specified time-to-travel. How can one find these shortest paths to each intersection in O(S+N log N), for S street segments and N intersections?