

## Week 7, Lecture 1

### Algorithm Analysis and Design

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### The Shortest Reliable Path Problem

Given a weighted graph  $G$ , a **src** node and a **dest** node, the shortest path from **src** to **dest** is to be found, that uses **at most  $k$  edges**.

Note: Dijkstra's Algorithm is not useful here as it doesn't keep track of the *number of edges traversed*.

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### The Dynamic Programming Solution

Since the problem can be broken down into several overlapping subproblems, we define:

Subproblem  $dist(v, i)$  (where  $i \leq k$ ) for each vertex  $v$ , to be the shortest length for a path from **src** to  $v$ .

Base cases:  $dist(v, 0) = \infty$ ,  $dist(src, 0) = 0$

The relation thus obtained is:

$$dist(v, i) = \min_{(u,v) \in E} \{dist(u, i-1) + l(u, v)\}$$

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### The Shortest Path All Vertex Pairs Problem

Given a weighted graph  $G$ , the shortest path between every pair of discrete vertices is to be found.

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### The Dynamic Programming Solution

#### Defining the Subproblems

Here, we observe that the **optimum substructure property** is followed.

Firstly, we try to find paths between node pairs that do not include *any* intermediate nodes whatsoever. Now, the shortest paths between all vertex pairs is

just the **edge weight**, if there exists an edge between a pair of vertices. Then, we *gradually* expand our scope and consider intermediate nodes too, and keep expanding this scope till all vertices of the graph are covered. Hence, we define the subproblem:  
 $dist(i, j, k)$  = the shortest path from node  $i$  to node  $j$  with the set of intermediate nodes being limited to  $\{1, \dots, k\}$   
**Base Cases** :  $dist(i, j, 0)$  length of edge  $ij$  if it exists, else  $\infty$ .

## Algorithm

Using a two dimensional memoization table, we obtain the following algorithm:

```
# let us conveniently name the memoization table 'dist'. 'Graph' is an
adjacency table (default value = infinity)
for i in range(0,V):
for j in range(0,V):
C[i][j] = Graph.edgeweight(i,j)

for k in range(0,V):
for i in range(0,V):
for j in range(0,V):
if(dist[i][j] > dist[i][k] + dist[k][j]):
dist[i][j] = dist[i][k] + dist[k][j]
return dist
```

The complexity of the above algorithm is naturally  $O(n^3)$  owing to the three nested loops.

## The Independent Sets In Trees Problem

Given a graph  $G = (V, E)$ , the largest subset of vertices  $S$  is to be found such that there exist no edges between them.

## The Dynamic Programming Solution

The layered structure of the tree itself provides the framework to define sub-problems, as the **optimum substructure property** is followed since for a tree to represent the optimal solution, *its subtrees need to represent their respective optimal sub-solution*.

Hence, we define the subproblem  $I(v)$  = largest independent set in the subtree with root  $v$ .

So, for a subtree with root  $v$ , the largest independent subset:

- Case 1: Includes  $v$   
Relation:  $I_1 = 1 + \sum \{\text{grandchildren } w \text{ of } v\} I(w)$
- Case 2: Does not include  $v$   
Relation:  $I_2 = \sum \{\text{children } w \text{ of } v\} I(w)$

and finally,  $I(v) = \max(I_1, I_2)$

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