INFDEV036A - Algorithms Lesson Unit 6

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Today

- ► Why is my code slow?
 - **▶** Empirical and complexity analysis
- ► How do I order my data?
 - **▶** Sorting algorithms
- ► How do I structure my data?
 - ► Linear, tabular, recursive data structures
- ► How do I represent relationship networks?
 - **▶** Graphs

More detailed agenda

- What is dynamic programming?
 - ► Fibonacci example; memoization and bottom-up approaches
- ▶ How can we find the shortest paths between **all pairs** of nodes in a graph?
 - ► Floyd Warshall algorithm

Dynamic Programming

Fibonacci sequence, Memoization, Bottom up, General idea

Definition

- ► Sequence of integer numbers, built according to the following rules
 - ▶ Elements 0 and 1 are 1
 - ► Any other element is the result of the sum of the two preceding elements in the sequence
- Recursive formulation

$$F(0) = 1$$

$$F(1) = 1$$

$$F(n) = F(n-1) + F(n-2)$$

Recursive function

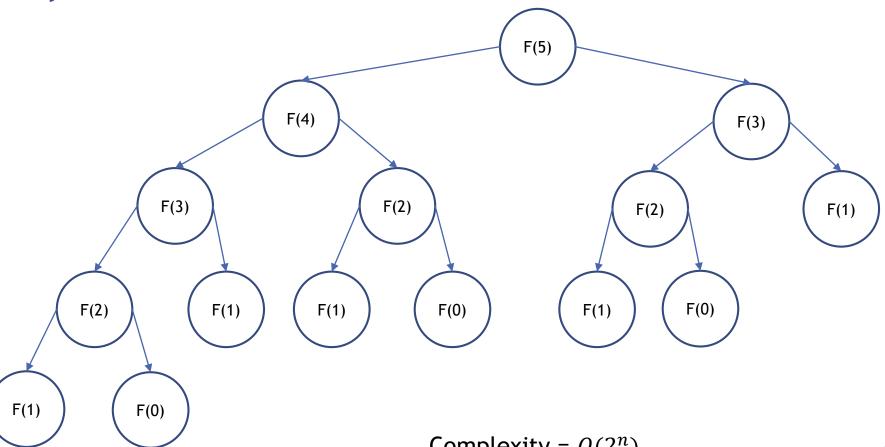
```
function fib(n)

if n <= 1 return 1

return fib(n - 1) + fib(n - 2)
```

n	Time (ms)
30	4
35	50
40	551
45	3792
>50	Happy watching!

Why so slow?



Complexity = $O(2^n)$

Why so slow?

"Did I ever tell you what the definition of insanity is? Insanity is doing the exact... same *** thing... over and over again expecting... things to change..."

- ▶ We are doing the same thing over and over again
 - \triangleright F(3) computed twice, F(2) computed three times...
- Ideas on how to speed up the process?
 - ▶ We could save the result of sub-problems (= memoization)!

Memoization

- Save the result of sub-problems into a data structure (called lookup table)
- Before making a recursive call, we check the lookup table...
 - ► Sub-problem never solved yet? → Make the recursive call
 - ► Sub-problem already solved? → Read the result stored in the table
- ▶ Add the result of the current recursive call to the lookup table

Memoized algorithm (recursive)

► Suppose m is a map object which maps each value of fib that has already been calculated to its result

```
var m := map(0 \rightarrow 1, 1 \rightarrow 1)
function fib(n)
if key n is not in map m
m[n] := fib(n - 1) + fib(n - 2)
return m[n]
```

Memoization

```
var m := map(0 → 0, 1 → 1)
function fib(n)
  if key n is not in map m
    m[n] := fib(n - 1) + fib(n - 2)
  return m[n]
```

- Time complexity?
 - \blacktriangleright Accessing the lookup table requires O(1)
 - ightharpoonup Sub-problems are n
 - ▶ Time complexity then is O(n): we compute n times sub-problems that require O(1) time
- Space complexity?
 - ightharpoonup O(n) memory space to save the sub-problems results

Bottom up algorithm (iterative)

- Other possible approach
 - ▶ Build the result of the computation starting from the base case of the recursion
 - ▶ At each iteration save the intermediate results to use at the next step
- ▶ Same time complexity O(n) as the recursive version, but only O(1) of space complexity

```
function fib(n)
  if n = 0
    return 1
  else
    var previousFib := 1, currentFib := 1
    repeat n - 1 times // loop is skipped if n = 1
      var newFib := previousFib + currentFib
      previousFib := currentFib
      currentFib := newFib
    return currentFib
```

All pairs shortest paths problem

Shortest path problem

- Remember Dijkstra?
 - ► Shortest paths starting from ONE node to ALL other nodes
- What if we need shortest paths from multiple source nodes?
 - ► Floyd-Warshall: shortest paths starting from ALL nodes to ALL nodes
 - ► Example of dynamic programming

- Compares all possible paths through the graph between each pair of vertices
- Every combination of edges is tested
- It works by incrementally improving an estimate on the shortest path between two vertices, until the estimate is optimal
- Based on a recursive idea

Consider

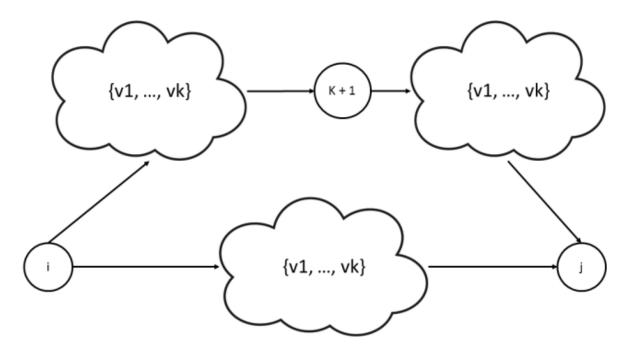
- ▶ A graph *G* with vertices *V* numbered from 1 to *N*
- A function shortestPath(i, j, k) which returns the shortest possible path from i to j using vertices only from the set $\{1, 2, ..., k\}$ as intermediate points along the way

▶ Goal

- Find the shortest path from each i to each j using only vertices from 1 to k + 1
- ► What could this shortest path be?

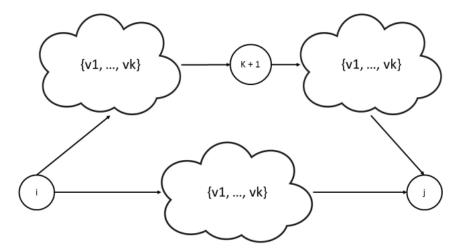
- ▶ Goal
 - Find the shortest path from each i to each j using only vertices from 1 to k + 1
 - ▶ What could this shortest path be?
 - 1. a path that only uses vertices in the set $\{1, ..., k\}$
 - 2. a path that goes from i to k + 1 and then from k + 1 to j
 - In other words... can we improve the shortest path between i and j if we pass through the vertex k+1?

▶ Can we improve the shortest path between i and j if we pass through the vertex k+1?



 \blacktriangleright Can we improve the shortest path between i and j if we pass through the vertex

k + 1?



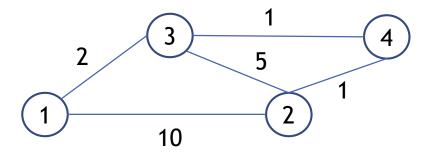
shortestPath(i, j, k + 1)= min(shortestPath(i, j, k), shortestPath(i, k + 1, k) + shortestPath(k + 1, j, k))

- ▶ What is the base (easiest) case of shortestPath(i, j, k)?
 - \triangleright $k = 0 \rightarrow$ the path between i and j does not involve any other vertex
 - ▶ Length of the path = weight of the edge w(i,j)
 - ▶ Remember that if two vertices are not directly connected, then $w(i,j) = \infty$

- \blacktriangleright What is the base (easiest) case of shortestPath(i, j, k)?
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 - ▶ Length of the path = weight of the edge w(i,j)
 - ▶ Remember that if two vertices are not directly connected, then $w(i,j) = \infty$
- ► Combination of the two formulas → core of the Floyd Warshall algorithm

```
\begin{split} & shortestPath(i,j,0) = w(i,j) \\ & shortestPath(i,j,k+1) \\ &= min(shortestPath(i,j,k), shortestPath(i,k+1,k) + shortestPath(k+1,j,k)) \end{split}
```

Floyd-Warshall and dynamic programming



We are recomputing the same paths multiple times, for example:

- ightharpoonup sp(1,2,4) = min(sp(1,2,3),sp(1,4,3) + sp(4,2,3))
- ightharpoonup sp(1,2,3) = min(sp(1,2,2), sp(1,3,2) + sp(3,2,2))
- Arr sp(1,4,3) = min(sp(1,4,2),sp(1,3,2) + sp(3,4,2))

Floyd-Warshall memoization

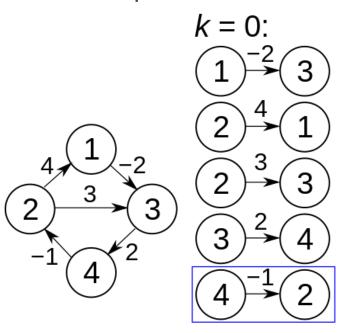
- ► How do we avoid to compute the same thing multiple times?
 - Saving the intermediate results into a matrix!
 - This matrix contains the length of the path from vertex i to vertex j
- The algorithm uses an iterative bottom up approach
 - ▶ Start by: filling the matrix with the base case solutions
 - ▶ 0 is the distance between a vertex and itself
 - w(i,j) is the distance between adjacent vertices
 - ▶ ∞ is the distance between non-adjacent vertices
 - ▶ Iterative part: filling in the matrix by iteratively expanding the set of intermediate vertices to use in the path

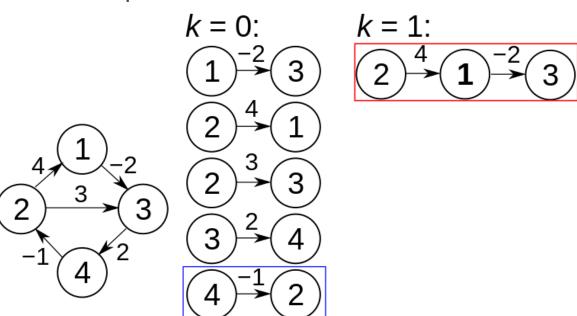
- lnitialization (k = 0)
 - \blacktriangleright shortestPath(i, j, 0) = w(i, j)
- Loop
 - ightharpoonup compute shortestPath(i,j,k) for all (i,j) pairs when k=1
 - ▶ See if passing for vertex 1 improves some shortest paths
 - ightharpoonup compute shortestPath(i,j,k) for all (i,j) pairs when k=2
 - ▶ See if passing for vertex 2 improves some shortest paths
 - ...
 - ▶ compute shortestPath(i,j,k) for all (i,j) pairs when k = N
 - ▶ See if passing for vertex N improves some shortest paths
 - After this, we have found the shortest path for all (i, j) pairs using **any** intermediate vertices

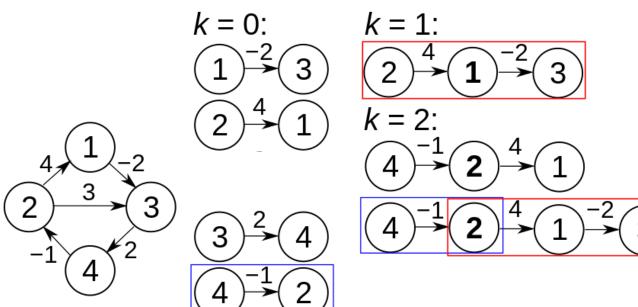
```
dist \leftarrow |V| \times |V| matrix of minimum distances initialized to \infty
for each vertex v
  dist[v][v] \leftarrow 0
for each edge (u,v)
  dist[u][v] \leftarrow w(u,v)
for k from 1 to |V|
  for i from 1 to |V|
    for j from 1 to |V|
       if dist[i][j] > dist[i][k] + dist[k][j]
         dist[i][j] \leftarrow dist[i][k] + dist[k][j]
       end if
```

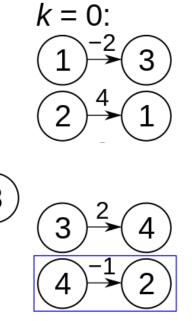
- Previous code only computed the <u>minimum distance</u> between pairs of vertices, but not the <u>actual minimum path</u>
- Small changes to the algorithm make it possible to save also the information on the actual path

- Complexity?
 - ► Three nested loops
 - ► Each loop performs |V| iterations
 - ▶ $|V| \times |V| \times |V|$ iterations in total $\rightarrow o(|V|^3)$
 - ▶ Or, if we call n the number of nodes (|V|), $O(n^3)$









$$k = 1:$$

$$2 \xrightarrow{4} 1 \xrightarrow{-2} 3$$

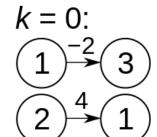
$$k = 2:$$

$$4 \xrightarrow{-1} 2 \xrightarrow{4} 1$$

$$4 \xrightarrow{-1} 2 \xrightarrow{4} 1$$

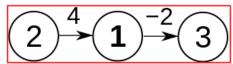
$$k = 3:$$

$$1 \xrightarrow{-2} 3 \xrightarrow{2} 4$$



$$\frac{2}{3}$$

$$k = 1$$
:



$$4 \xrightarrow{-1} 2 \xrightarrow{4} 1$$

$$4 \xrightarrow{-1} 2 \xrightarrow{4} 1 \xrightarrow{-2} 3$$

$$k = 3$$
:

$$1)^{-2} 3 \xrightarrow{2} 4$$

$$2 \xrightarrow{4} 1 \xrightarrow{-2} 3 \xrightarrow{2} 4$$

$$k = 4$$
:

$$\begin{array}{c} 3 \\ \hline \end{array} \begin{array}{c} 2 \\ \hline \end{array} \begin{array}{c} -1 \\ \hline \end{array} \begin{array}{c} 2 \\ \end{array} \begin{array}{c} \end{array}$$

$$3 \xrightarrow{2} 4 \xrightarrow{-1} 2 \xrightarrow{4} 1$$

$$1 \xrightarrow{-2} 3 \xrightarrow{2} 4 \xrightarrow{-1} 2$$

Homework

- Study the slides
- Answer the MC questions on GrandeOmega
- Implement Floyd-Warshall algorithm
- Do the sample practical assessment
 - most likely will be published this Friday on N@tschool
- ► [optional] Complete third exercise of practical assignment (about graphs)
- Next week (week 8)
 - Available in our offices for doubts both on MC questions and implementations (sample exams/assessments)