

Dynamic programming (part 1)

CS 146 - Spring 2017

Today

- Last time:
 - Bellman-For algorithm
 - Shortest paths with negative weights
- Floyd-Warshall algorithm
- main ideas behind dynamic programming

idea: relax ALL edges, $V - 1$ times

```
bellmanFord(G, s) {
```

```
    Map dist = new Map();
```

```
    for each vertex v of G
```

```
        dist.put(v, infinity)
```

```
    dist.put(s, 0)
```

```
    repeat  $V - 1$  times
```

```
        for each vertex v of G
```

```
            for neighbor w of v
```

```
                relax(v -> w)
```

```
    return dist;
```

```
}
```

initialize as before

one round of relaxation

time complexity:
 $O(VE)$

```

bellmanFord(G, s) {
    Map dist = new Map();
    for each vertex v of G
        dist.put(v, infinity)
    dist.put(s, 0)

```

Observation:
 if the shortest
 path tree has
 depth k ,
 there are no
 more changes
 to the distances
 after k rounds of
 relaxation

```

        repeat V - 1 times
            changed = false
            for each vertex v of G
                for neighbor w of v
                    if (dist(w) > dist(v) + w(v->w))
                        dist(w) = dist(v) + w(v->w)
                        changed = true
            if (!changed) break
    return dist;
}

```

Bellman-Ford with optimization

```

bellmanFord(G, s) {
    Map dist = new Map();
    for each vertex v of G
        dist.put(v, infinity)
    dist.put(s, 0)

    repeat V - 1 times
        for each vertex v of G
            for neighbor w of v
                relax(v -> w)

    for each vertex v of G
        for neighbor w of v
            if (dist(v) > dist(w) + w(v->w)
                error("negative cycle detected");

    return dist;
}

```

one more round of relaxation
to see if there are
any more changes

Bellman-Ford with cycle detection

In summary...

single source shortest-path problem

- 2 relaxation-based algorithms
 - Dijkstra's algorithm: $O(E \log V)$, edges non-neg
 - Bellman-Ford algorithm: $O(VE)$, neg edges OK

All-pairs shortest paths

- input: a weighted graph (potentially with negative weights)
- output: a dictionary or matrix of distances between every pair of vertices

give an algorithm for this problem and its runtime

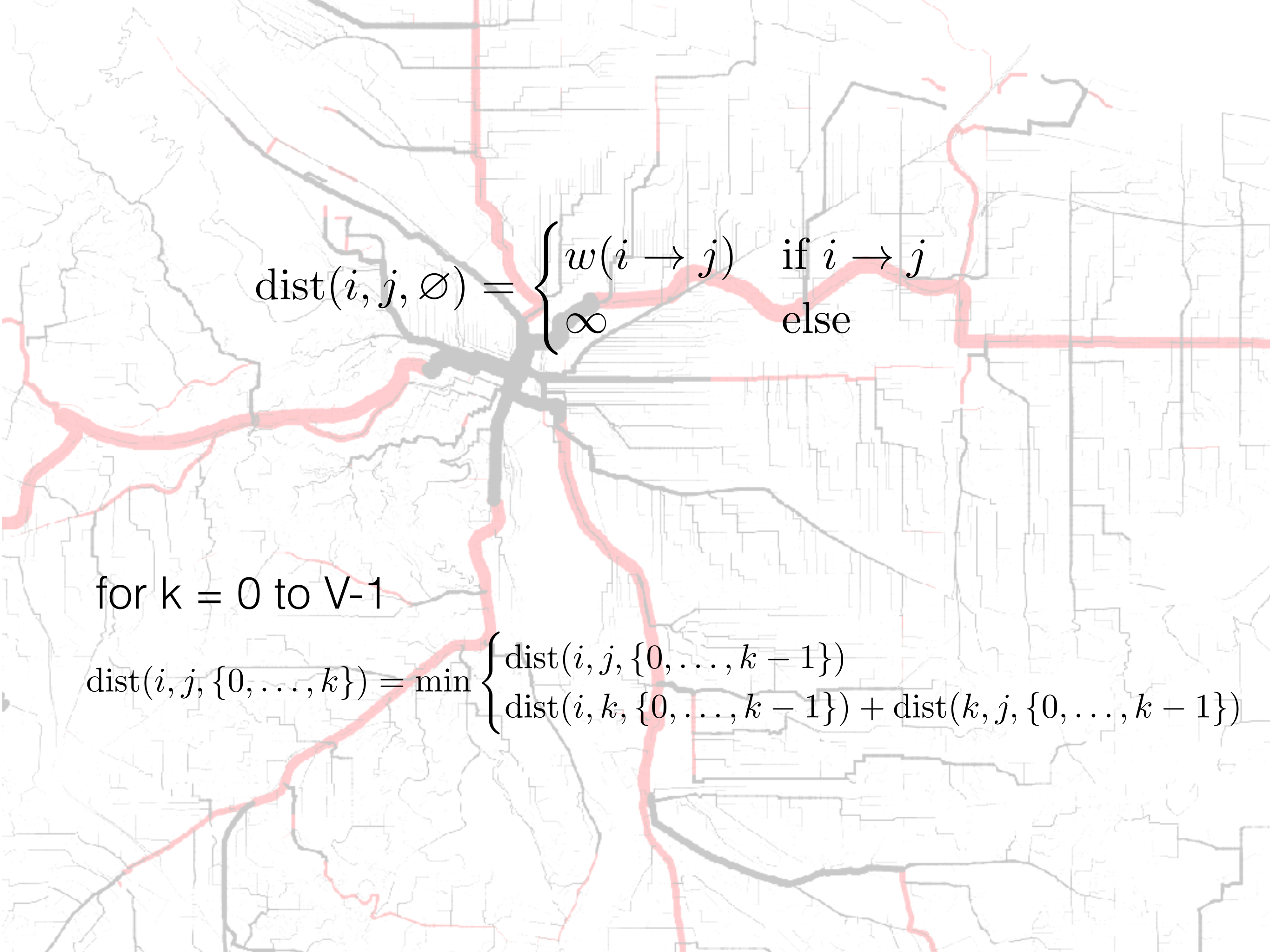
1st attempt: use Bellman-Ford multiple times

- for each vertex v
 - run Bellman-Ford with v as the source
 - record distances from v to every other vertex
- time: $O(V^2 E)$
- space: $O(V^2)$

seems like a lot of work being redone, can we do better?

2nd attempt: recursion + memoization

- record every shortest path between each pair as we go along number the vertices $0, 1, 2, 3, \dots V - 1$
- subproblem: **$\text{dist}(i, j, S)$ = distance from i to j using only vertices in set S to connect them**
- base case?
- problem in terms of subproblem?



The background is a map of a city street network. A thick grey path highlights a specific route through the center of the map. A red path highlights another route, which is more extensive and covers a larger area of the map. The red path starts from the left, goes through the center, and then branches out to the right and top right. The grey path is a more direct route through the center.

$$\text{dist}(i, j, \emptyset) = \begin{cases} w(i \rightarrow j) & \text{if } i \rightarrow j \\ \infty & \text{else} \end{cases}$$

for $k = 0$ to $V-1$

$$\text{dist}(i, j, \{0, \dots, k\}) = \min \begin{cases} \text{dist}(i, j, \{0, \dots, k-1\}) \\ \text{dist}(i, k, \{0, \dots, k-1\}) + \text{dist}(k, j, \{0, \dots, k-1\}) \end{cases}$$

Recall longest increasing subsequence

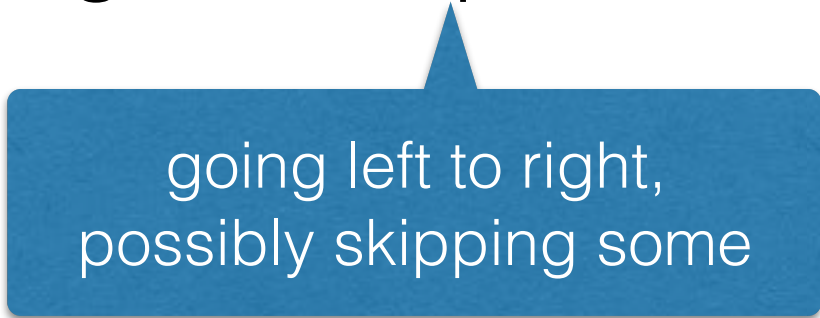
Given a sequence of numbers

for example: 2, 4, 3, 5, 1, 7, 6, 9, 8

What is the length of the longest increasing subsequence?



strictly

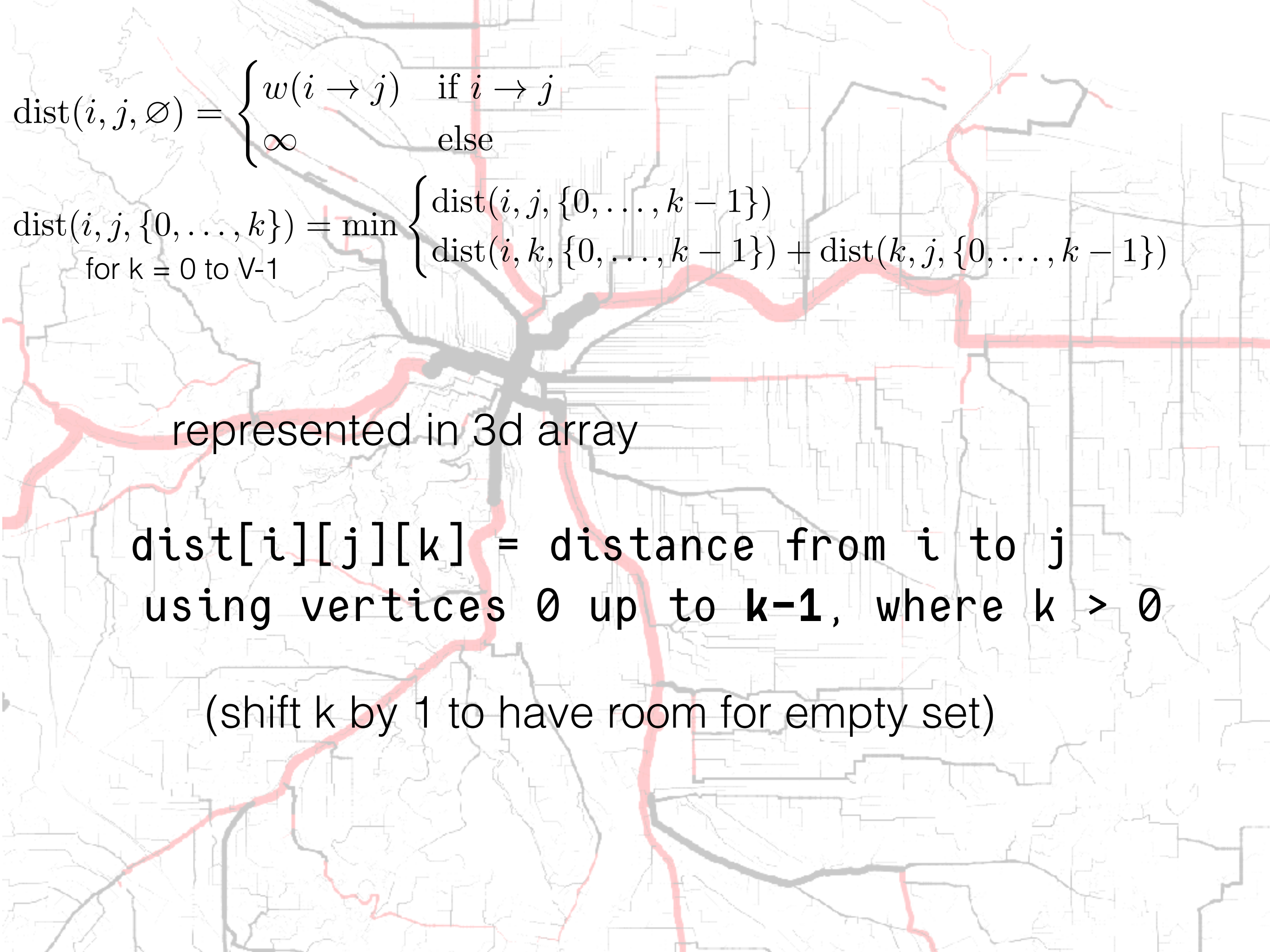


going left to right,
possibly skipping some

Making recursion more time efficient

(by using more space efficient)

	memoization (top-down)	dynamic programming (bottom up)
store subproblems in ...	dictionary	array (2D if subpb has 2 params)
subproblem input	dictionary key	array index
subproblem output	dictionary value	entry at given index
code structure	recursive	for-loops


$$\text{dist}(i, j, \emptyset) = \begin{cases} w(i \rightarrow j) & \text{if } i \rightarrow j \\ \infty & \text{else} \end{cases}$$

$$\text{dist}(i, j, \{0, \dots, k\}) = \min_{\text{for } k = 0 \text{ to } V-1} \begin{cases} \text{dist}(i, j, \{0, \dots, k-1\}) \\ \text{dist}(i, k, \{0, \dots, k-1\}) + \text{dist}(k, j, \{0, \dots, k-1\}) \end{cases}$$

represented in 3d array

$\text{dist}[i][j][k]$ = distance from i to j
using vertices 0 up to $k-1$, where $k > 0$

(shift k by 1 to have room for empty set)

Floyd-Warshall algorithm for all-pairs shortest path algorithm

```
int[][][] dist = new int[V][V][V+1]

for (int i = 0; i < V; i++)
    for (int j = 0; j < V; j++)
        dist[i][j][0] = i -> j ? w(i -> j) : (i == j ? 0 : inf)

for (int k = 1; k < V+1; k++)
    for (int i = 0; i < V; i++)
        for (int j = 0; j < V; j++)
            dist[i][j][k] = min(dist[i][j][k-1],
                                dist[i][k][k-1] + dist[k][j][k-1])

return dist;
```

time and
space
complexity?

How to reconstruct the paths?

- record choices made at every step
- start at the beginning or the end (whichever is appropriate) and follow choices to reconstruct path
- same approach as with Dijkstra's algorithm, making change problem, recursive problems etc.

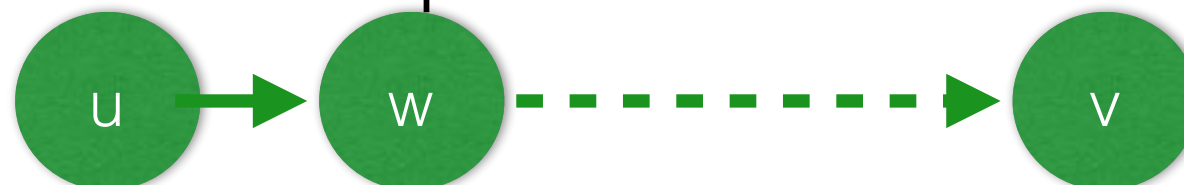
record choices made at every step:
next[][s][V] stores shortest path tree **to** s
(s is the sink or destination, rather than source)

in other words...

$$\text{next}[u][v][V] = w$$

means

w comes right after u on
shortest path from u to v



```
int[][][] dist = new int[V][V][V+1]; int[][][] next = new int[V][V][V+1]
for (int i = 0; i < V; i++)
    for (int j = 0; j < V; j++)
        dist[i][j][0] = i -> j ? w(i -> j) : (i == j ? 0 : INF);
        next[i][j][0] = i -> j ? j : null;
for (int k = 1; k < V+1; k++)
    for (int i = 0; i < V; i++)
        for (int j = 0; j < V; j++)
            if (dist[i][j][k-1] < dist[i][k][k-1] + dist[k][j][k-1]) {
                dist[i][j][k] = dist[i][j][k-1];
                next[i][j][k] = next[i][j][k-1];
            } else {
                dist[i][j][k] = dist[i][k][k-1] + dist[k][j][k-1];
                next[i][j][k] = next[i][k][k-1];
            }
return dist;
```

record choices made at every step:
next[][s][V] stores shortest path tree **to** s
(s is the sink or destination, rather than source)

Floyd-Warshall algo with path reconstruction (1/2)



`next[u][v][v] = w`

```
reconstruct-path(int u, int v, int[][] next) {  
    List<Integer> path = new ArrayList<>();  
    path.add(u);  
    for (int w = u; next[w][v] != null; w = next[w][v])  
        path.add(next[w][v]);  
    return path;  
}
```

```
shortest-path-with-Floyd-Warshall(G, u, v) {  
    int[][][] next = Floyd-Warshall(G);  
    return reconstruct-path(u, v, next[][][v]);  
}
```

sample usage
of reconstruct-path

(not efficient: if only
need one path, use
Dijkstra)

Floyd-Warshall algo with path reconstruction (2/2)

Making recursion more time efficient

(by using more space efficient)

	memoization (top-down)	dynamic programming (bottom up)
store subproblems in ...	dictionary	array (2D if subpb has 2 params)
subproblem input	dictionary key	array index
subproblem output	dictionary value	entry at given index
code structure	recursive	for-loops

(see reading)

can be
automated

?