

Lecture 8: Dynamic Programming Preliminaries

Study Chapter 6.1-6.3

Dynamic Programming



- *Dynamic Programming* is a technique for computing recurrence relations efficiently by storing partial or intermediate results
- Three keys to constructing a dynamic programming solution:
 1. Formulate the answer as a recurrence relation
 2. Consider all instances of the recurrence at each step
 3. Order evaluations so you will always have precomputed the needed partial results

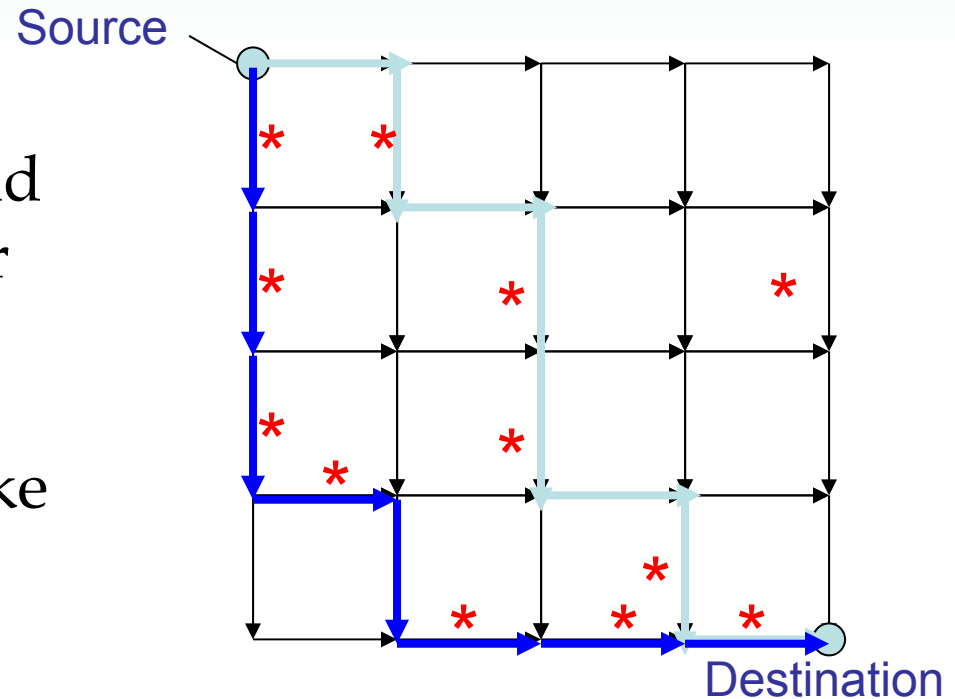


Manhattan Tourist Problem (MTP)



Imagine seeking a path from source to destination in a Manhattan-like city grid that maximizes the number of attractions (*) passed. With the following caveat—at every step you must make progress towards the goal.

We treat the city map as a graph, with “vertices” at each corner, and weighted edges along each block. The weights are the number of attractions along each block.



Manhattan Tourist Problem: Formulation



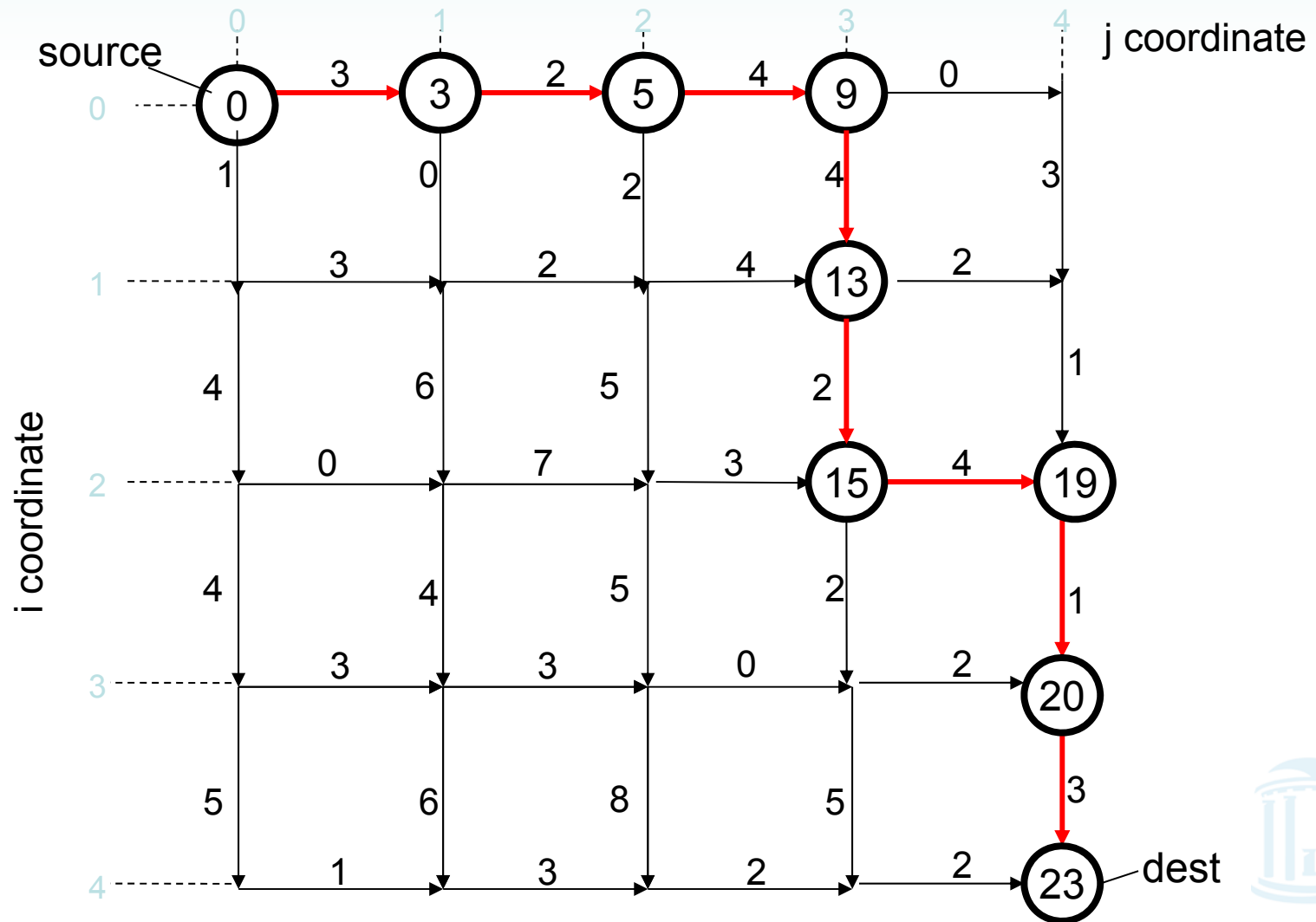
Goal: Find the maximum-weight path in a grid.

Input: A weighted grid \mathbf{G} with two distinct vertices, one labeled “*source*” and the other labeled “*destination*”

Output: The best path (= greatest total weight) in \mathbf{G} from “*source*” to “*destination*”



MTP as a Dynamic Program



MTP Strategy



- Instead of solving the Manhattan Tourist problem directly, (i.e. the path from $(0,0)$ to (n,m)) we will solve a more general problem: find the best path from $(0,0)$ to any arbitrary vertex (i,j) .
- If the best path from $(0,0)$ to (n,m) passes through some vertex (i,j) , then the path from $(0,0)$ to (i,j) must be the best. Otherwise, you could increase your path weight by changing it.



MTP: Simple Recursive Program



What's wrong with this approach?

$MT(n,m)$

if $n = 0$ and $m = 0$

return 0

if $n = 0$

return $MT(0,m-1) + \text{weight of edge from } (0,m-1) \text{ to } (0,m)$

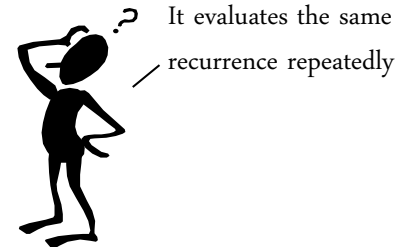
if $m = 0$

return $MT(n-1, 0) + \text{weight of edge from } (n-1,0) \text{ to } (n,0)$

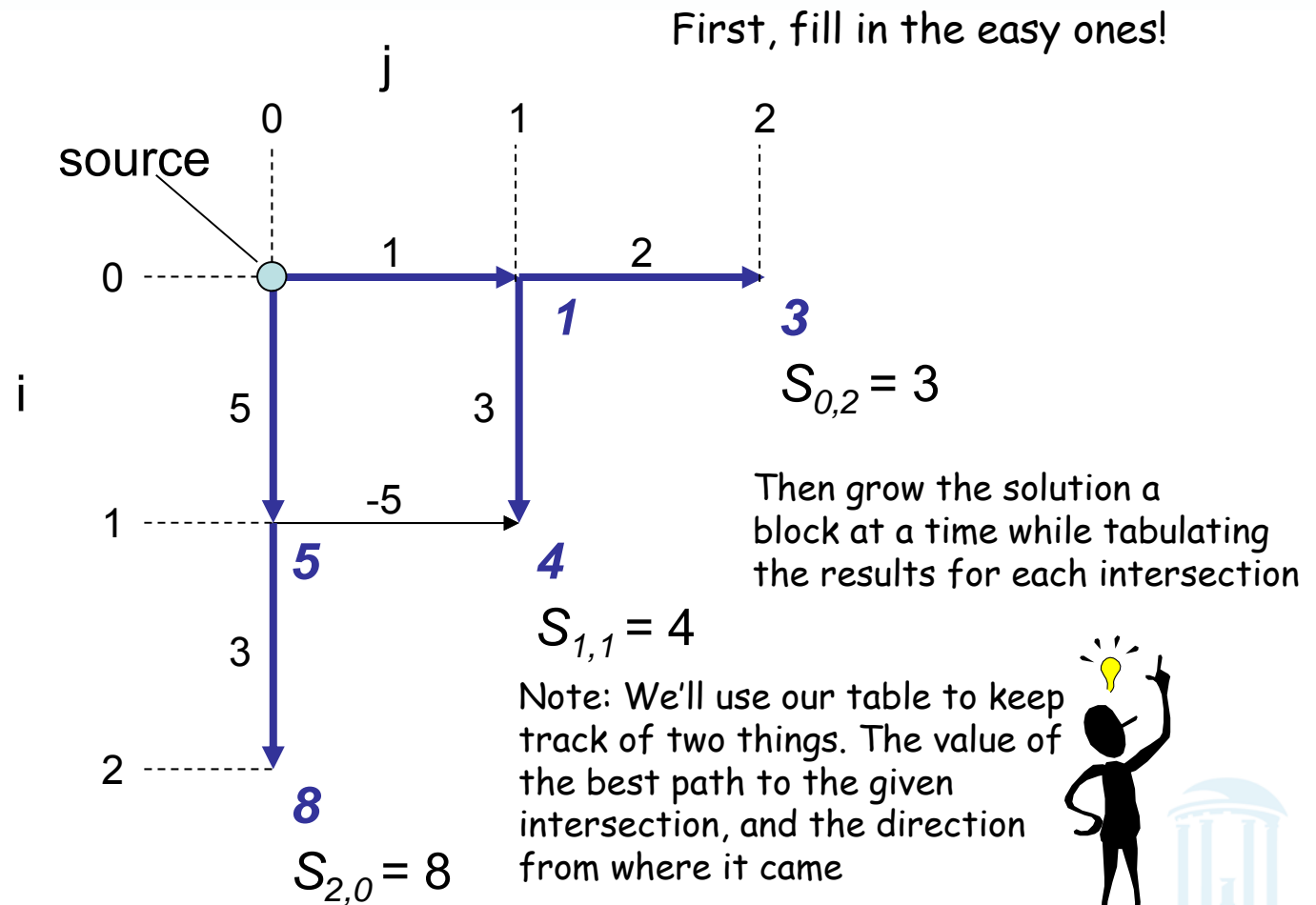
$x \leftarrow MT(n-1,m) + \text{weight of edge from } (n-1, m) \text{ to } (n,m)$

$y \leftarrow MT(n,m-1) + \text{weight of edge from } (n, m-1) \text{ to } (n,m)$

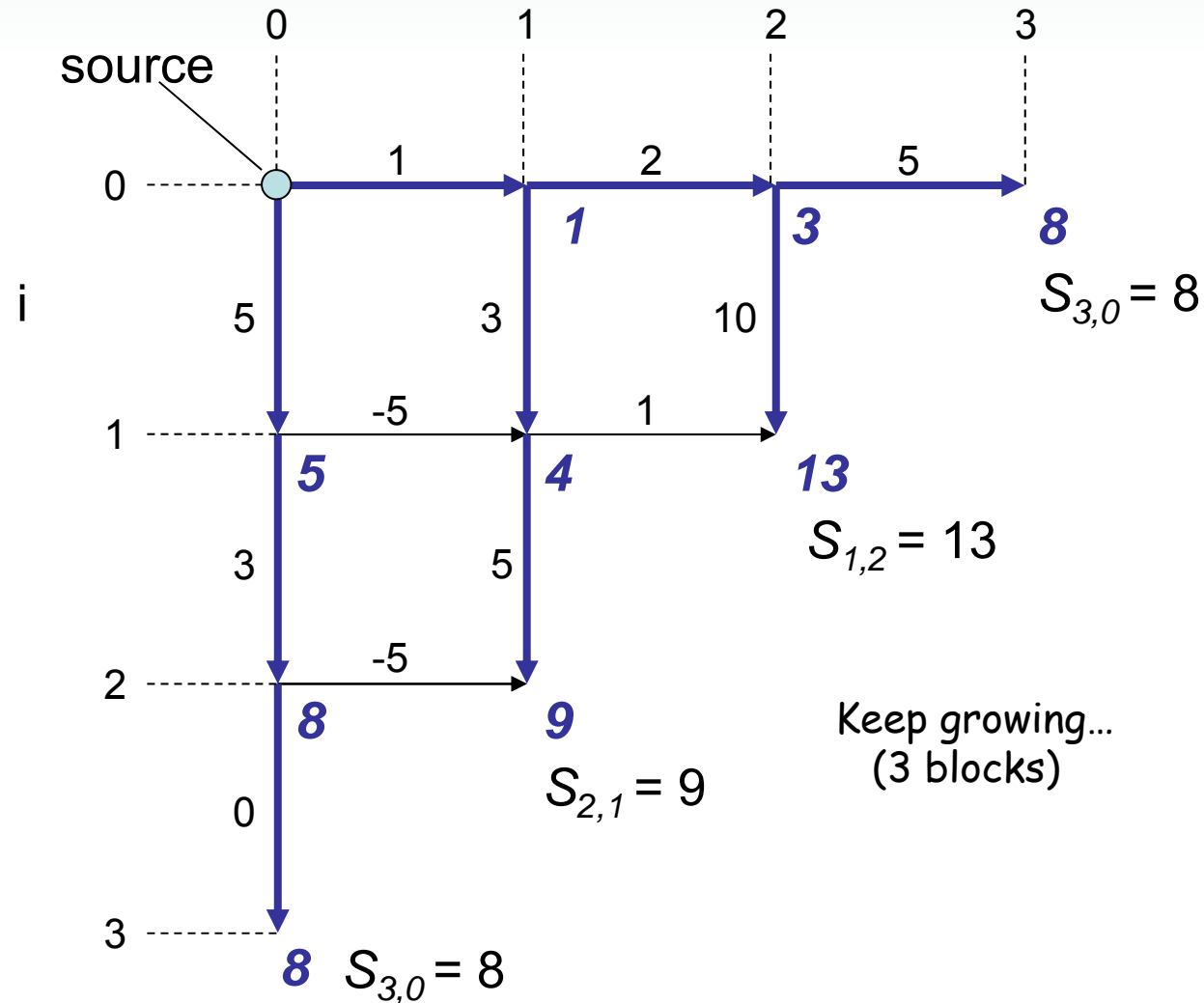
return $\max(x,y)$



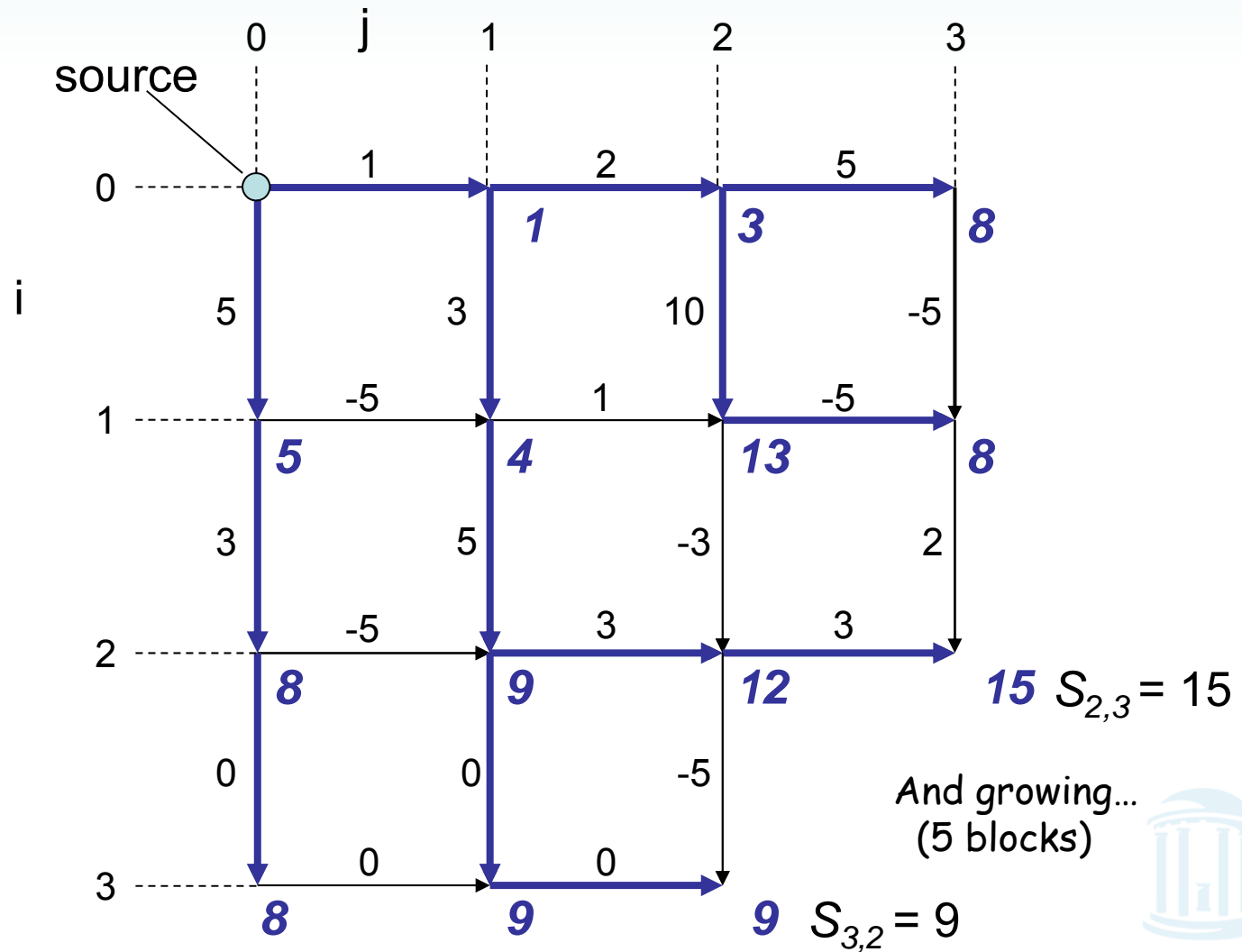
MTP: Dynamic Programming (cont'd)



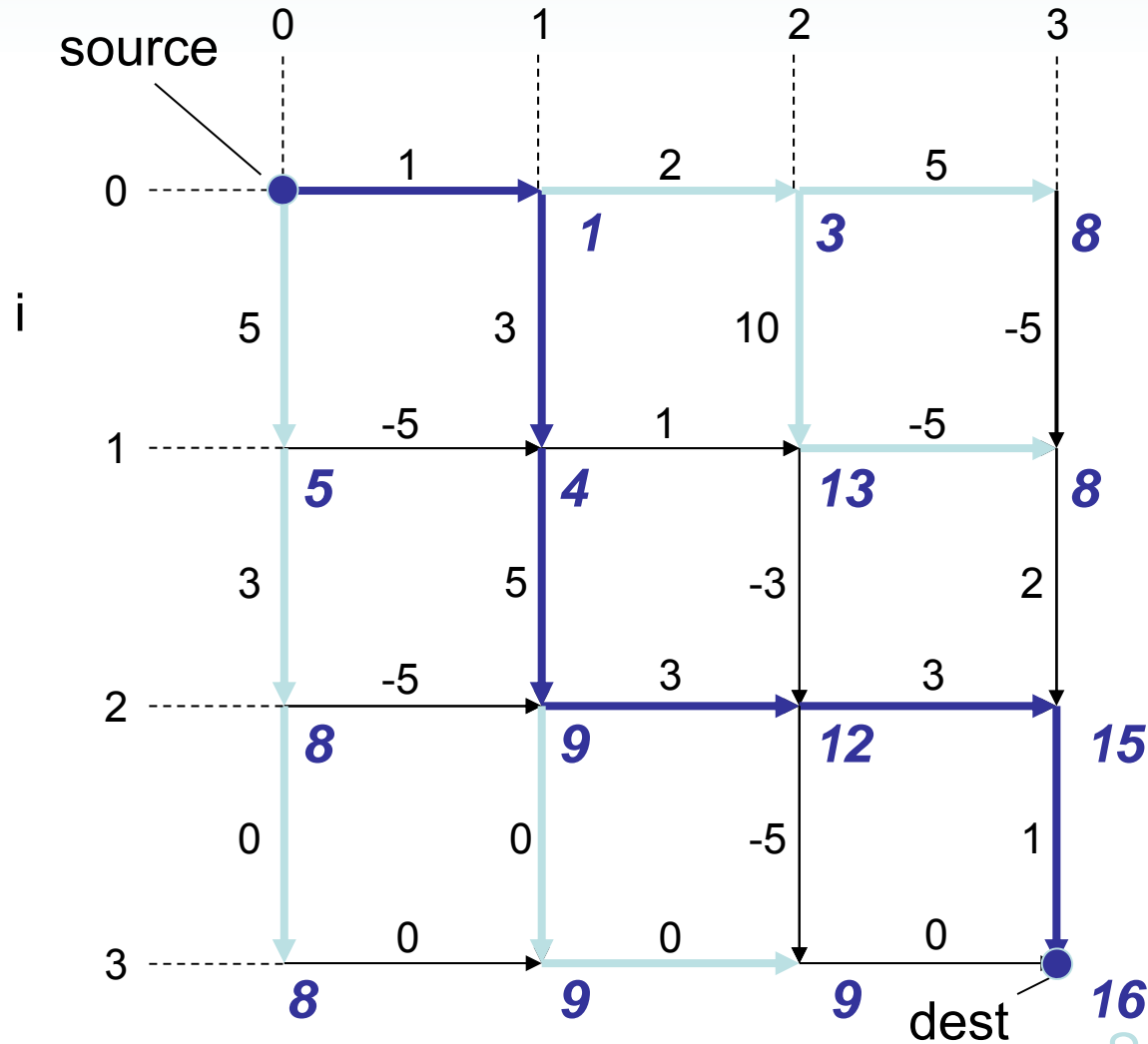
MTP: Dynamic Programming (cont'd)



MTP: Dynamic Programming (cont'd)



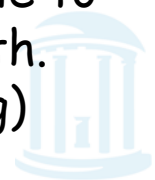
MTP: Dynamic Programming (cont'd)



Once the "destination" node (intersection) is reached, we're done.

Our table will have the answer of the maximum number of attractions stored in the entry associated with the destination.

We use the "links" back in the table to recover the path. (Backtracking)



MTP: Recurrence



Computing the score for a point (i,j) by the recurrence relation:

$$s_{i,j} = \max \left\{ \begin{array}{l} s_{i-1,j} + \text{weight of the edge between } (i-1, j) \text{ and } (i, j) \\ s_{i,j-1} + \text{weight of the edge between } (i, j-1) \text{ and } (i, j) \end{array} \right.$$

Path to the intersection from the left

Path to the intersection from above

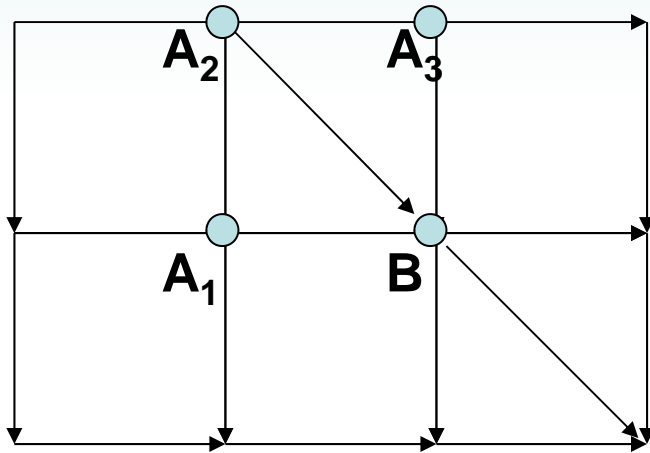
The running time is $O(nm)$ for a n by m grid

(You visit all intersections once, and perform 2 tests)

(n = # of rows, m = # of columns)



Manhattan Is Not A Perfect Grid



What about diagonals?

Broadway, Greenwich, etc.

- Easy to fix. Just adds more recursion cases.
- The score at point B is given by:

$$s_B = \max \left\{ \begin{array}{l} s_{A_1} + \text{weight of the edge } (A_1, B) \\ s_{A_2} + \text{weight of the edge } (A_2, B) \\ s_{A_3} + \text{weight of the edge } (A_3, B) \end{array} \right.$$



Generalizing Manhattan to a Directed Graph



Computing the score for point x is given by the recurrence relation:

$$s_x = \max_{\text{of}} \left\{ s_y + \text{weight of vertex } (y, x) \text{ where } y \in \text{Predecessors}(x) \right.$$

- Predecessors (x) – set of vertices having edges leading to x
- In a graph $G(\mathbf{V}, \mathbf{E})$
(\mathbf{V} is the set of all vertices and \mathbf{E} is the set of all edges)
each edge and each vertex is considered once

Traveling in the Grid



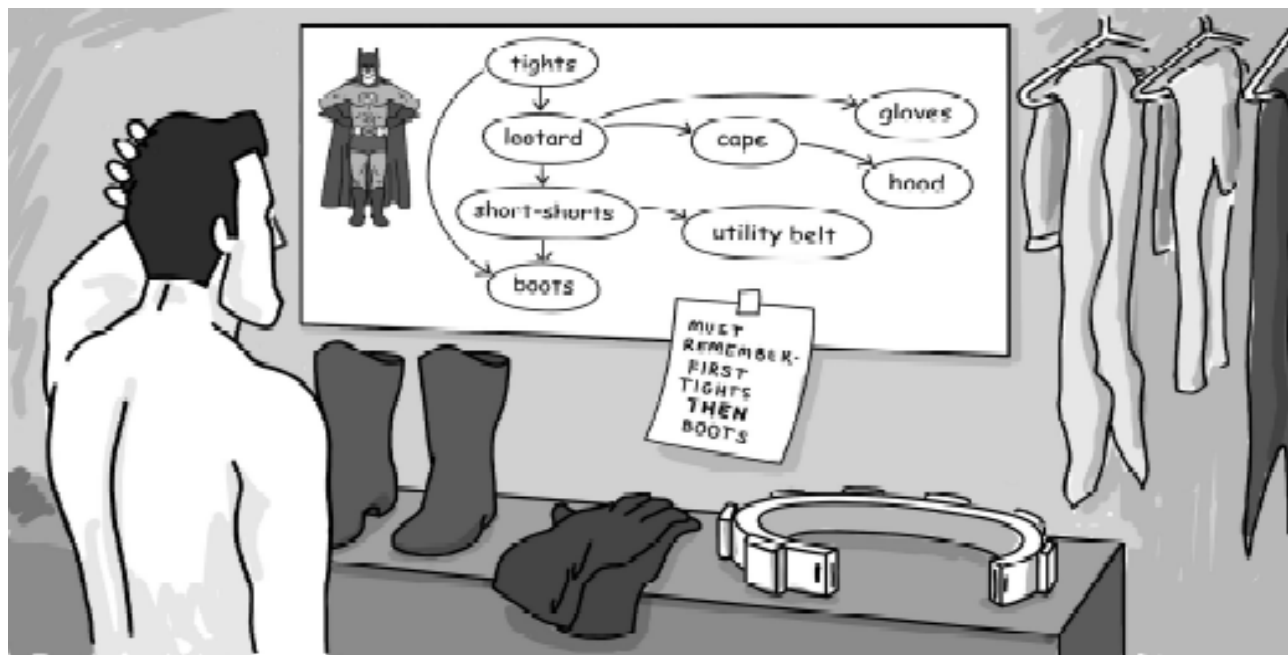
- The only hitch is that one must decide on an *order* to visit the vertices
- We must assure that by the time the vertex x is analyzed, the values, s_y , for all its predecessors, y , should be computed – otherwise we are in trouble.
- We need to traverse the vertices in some order
- How to find such order for any directed graph?

???



DAG: Directed Acyclic Graph

- Since most cities are not perfect regular grids, we represent paths in them as a DAGs
- DAG for *Dressing in the morning* problem



Topological Ordering



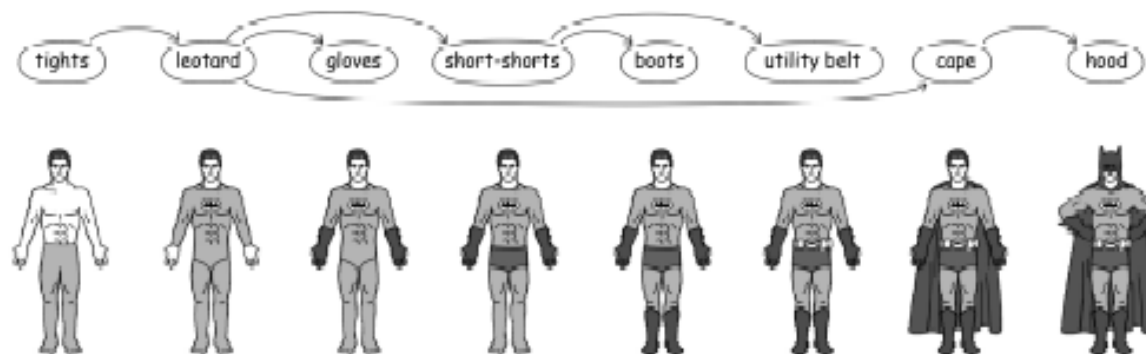
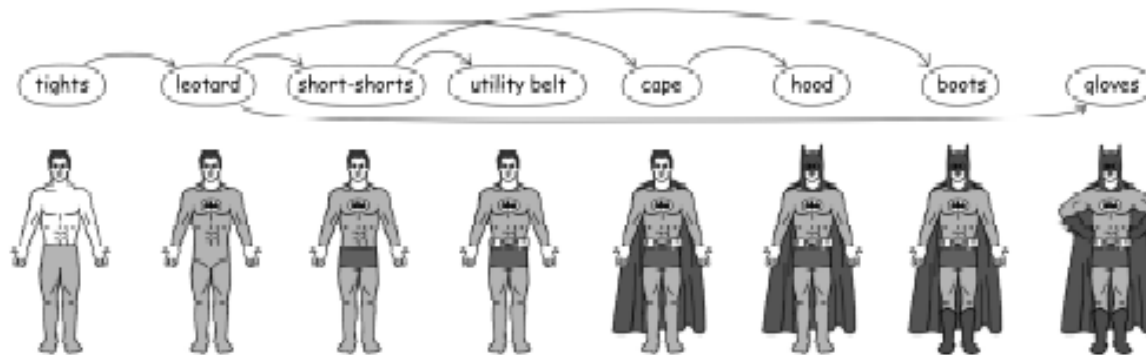
- A numbering of vertices of the graph is called *topological ordering* of the DAG if every edge of the DAG connects a vertex with a smaller label to a vertex with a larger label
- In other words, if vertices are positioned on a line in an increasing order of labels then all edges go from left to right.



Topological Ordering



- 2 different topological orderings of the DAG



Best Path in DAG Problem



- Goal: Find highest weight path between two vertices in a weighted DAG
- Input: A weighted DAG G with source and destination vertices
- Output: A highest weight path in G from source to destination



Longest Path in DAG: Dynamic Programming



- Suppose vertex v has indegree 3 and predecessors $\{u_1, u_2, u_3\}$
- Longest path to v from source is:

$$s_v = \max \text{ of } \left\{ \begin{array}{l} s_{u_1} + \text{weight of edge from } u_1 \text{ to } v \\ s_{u_2} + \text{weight of edge from } u_2 \text{ to } v \\ s_{u_3} + \text{weight of edge from } u_3 \text{ to } v \end{array} \right.$$

In General:

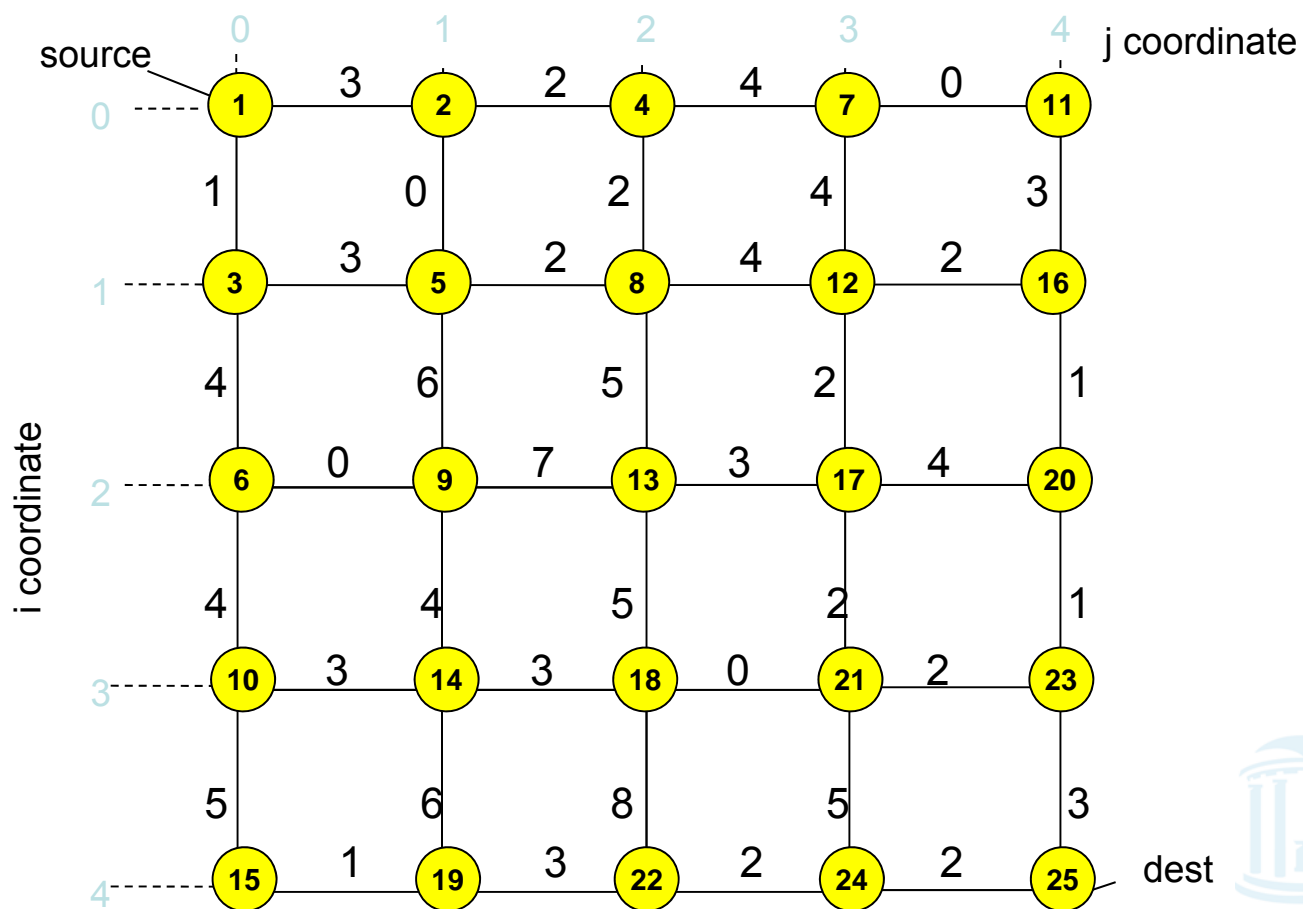
$$s_v = \max_u (s_u + \text{weight of edge from } u \text{ to } v)$$



Evaluation order



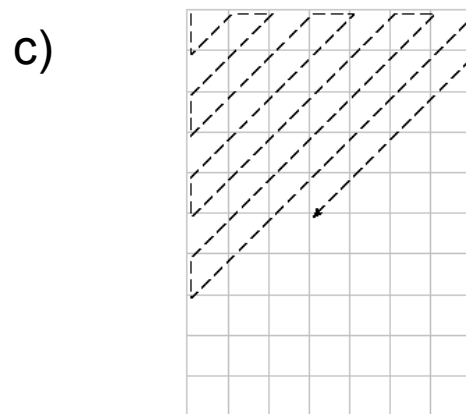
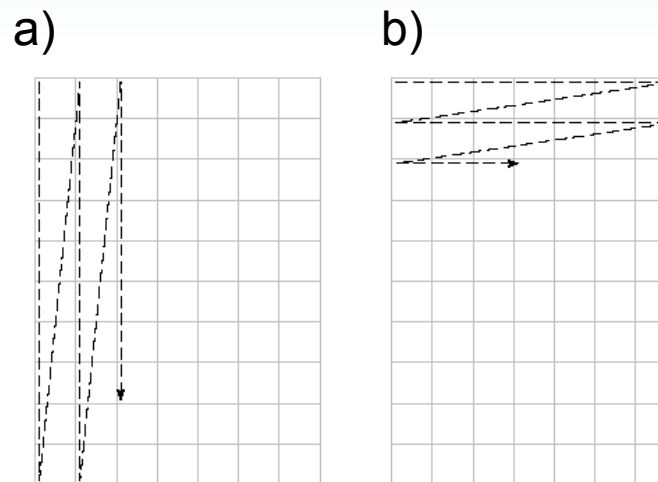
- *Any* topological ordering of vertices will work!



Traversing the Manhattan Grid



- We chose to evaluate our table in a particular order. Uniform distances from the source (all points one block away, then 2 blocks away, etc.)
- Other strategies:
 - a) Column by column
 - b) Row by row
 - c) Along diagonals
- This choice can have performance implications



Next Time



- Return to biology
- Solving sequence alignments using
Dynamic Programming

