

# CSCI 136

## Data Structures & Advanced Programming

Lecture 32

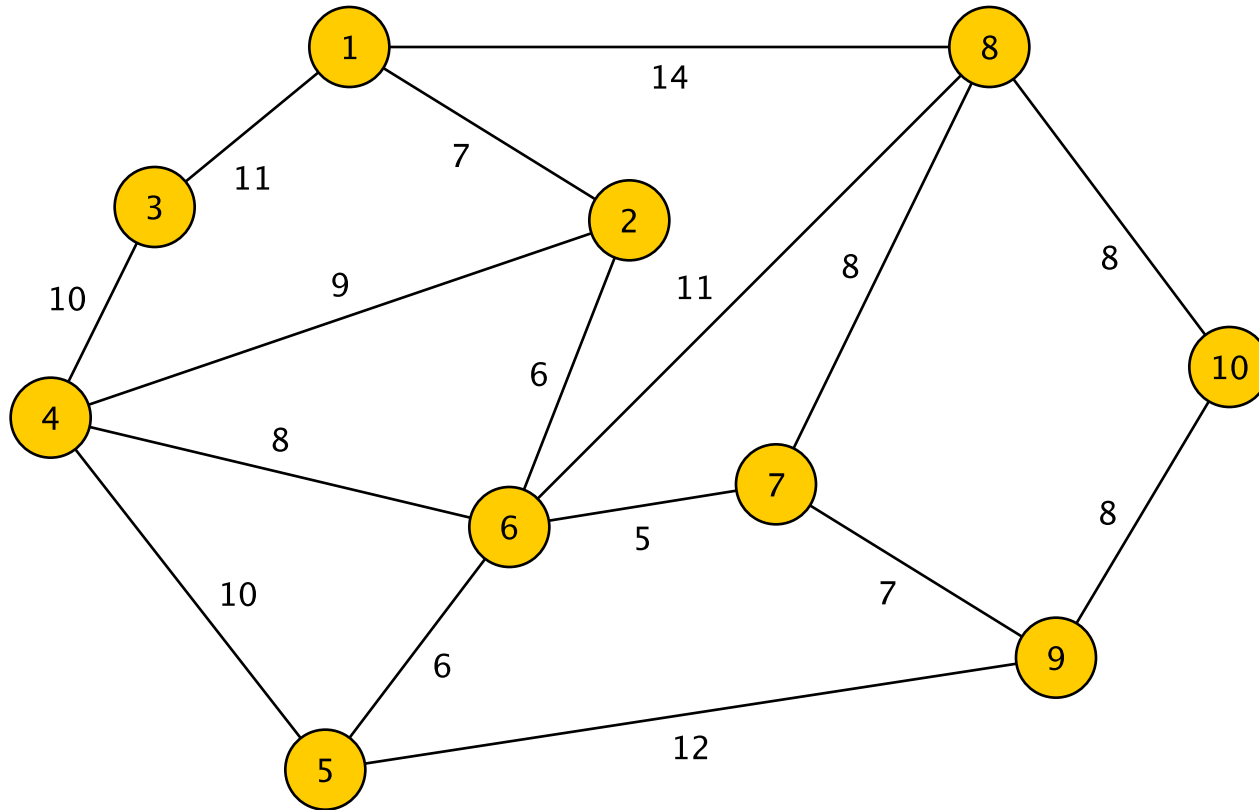
Fall 2018

Instructors: Bill and Sam

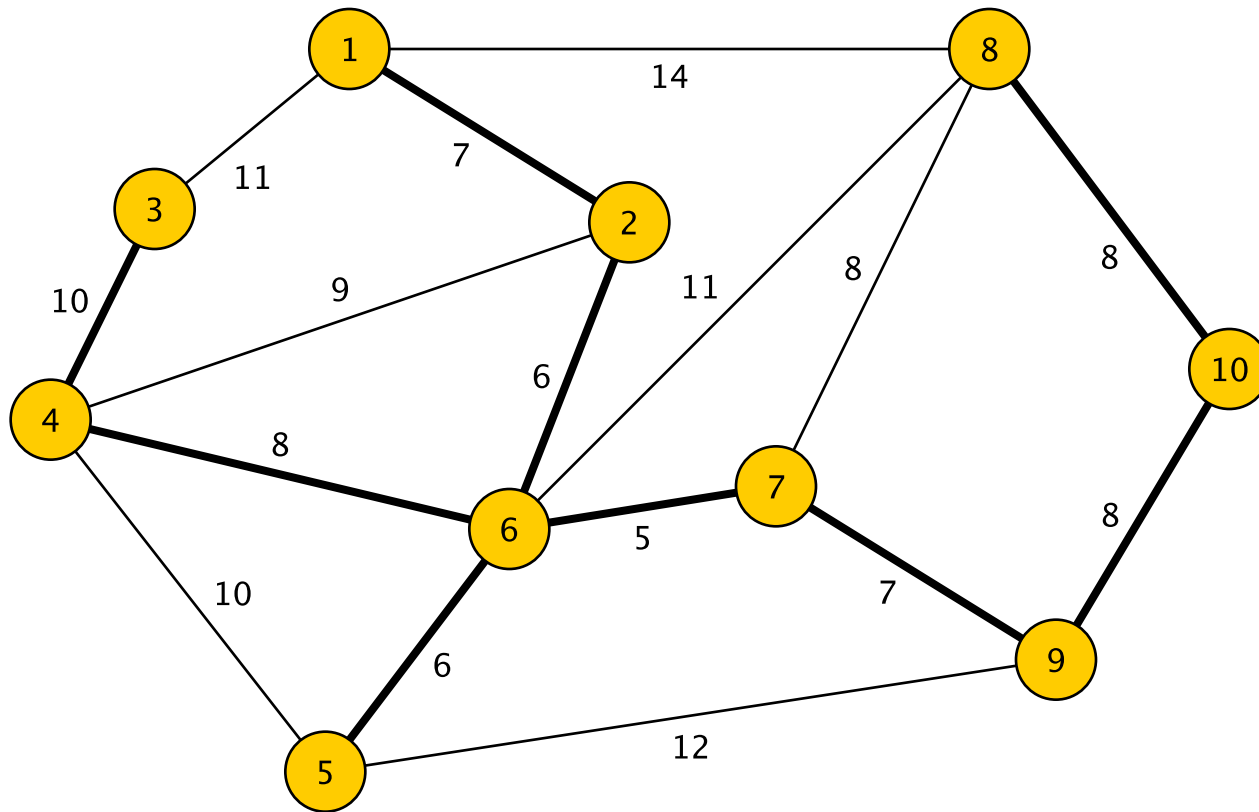
# Today's Outline

- More on Prim's Algorithm
- More Core Algorithms: Directed Graphs
  - Dijkstra's Algorithm

# Minimum-Cost Spanning Trees



# Minimum-Cost Spanning Trees



# Recall: An Amazing Fact

Thm: (Prim 1957) The greedy tree-growing algorithm always finds a minimum-cost spanning tree for any connected graph.

Contrast this with the greedy exam scheduling algorithm, which does *not* always find a minimum coloring

# Implementing Prim's Algorithm

- We'll “build” the MCST by marking its edges as “visited” in  $G$
- If a vertex is reached we mark it as visited
- How should we represent the list of possible new edges?
  - What operations are important to this list?
    - Add edges
    - Remove cheapest edge
- When we remove an edge from the list, check to ensure it has one end in each of  $V_1$  and  $V_2$

# ComparableEdge Class

- Values in a PriorityQueue need to implement Comparable
- We wrap edges of the PQ in a class called ComparableEdge
  - It requires the label used by graph edges to be of a Comparable type

# Prim : Space Complexity

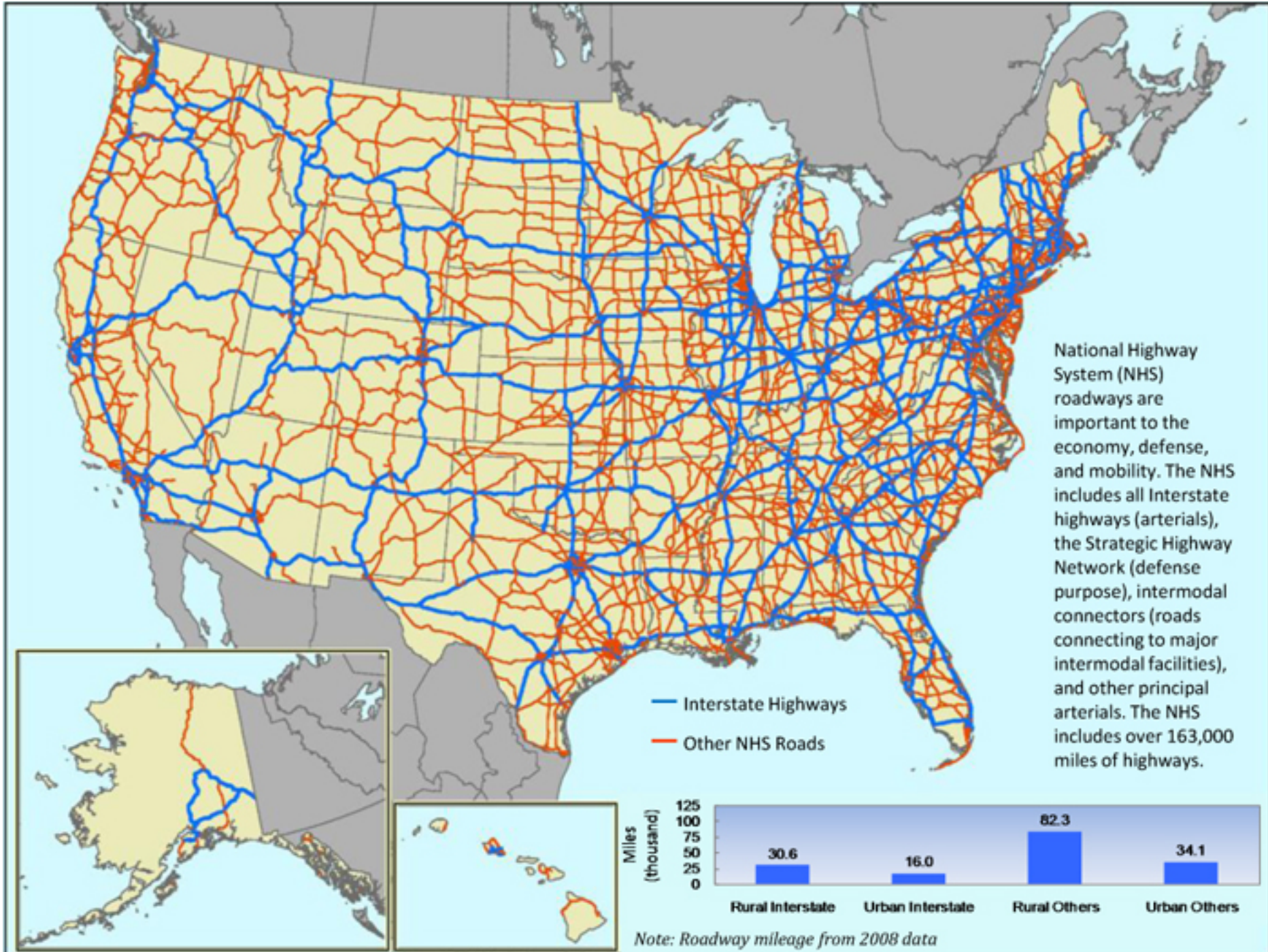
- Graph:  $O(|V| + |E|)$ 
  - Each vertex and edge uses a constant amount of space
- Priority Queue  $O(|E|)$ 
  - Each edge takes up constant amount of space
- Every other object (including the neighbor iterator) uses a constant amount of space
- Result:  $O(|V| + |E|)$ 
  - Optimal in Big-O sense!

# Prim : Time Complexity

Assume Map ops are  $O(1)$  time (not quite true!)

For each iteration of do ... while loop

- Add neighbors to queue:  $O(\log |E|)$  each
  - Iterator operations are  $O(1)$  [Why?]
  - Adding an edge to the queue is  $O(\log |E|)$
- Find next edge:  $O(\# \text{ edges checked} * \log |E|)$ 
  - Removing an edge from queue is  $O(\log |E|)$  time
  - All other operations are  $O(1)$  time
- $O(E \log E)$  time



# Single Source Shortest Paths

The Problem: Given a graph  $G$  and a starting vertex  $v$ , find, for *each* vertex  $u \neq v$  reachable from  $v$ , a shortest path from  $v$  to  $u$ .

- The Single Source Shortest Paths Problem
- Arises in many contexts, including network communications
- Uses edge weights (but we'll call them "lengths"): assume they are non-negative numbers
- Could be a directed or undirected graph

# Dijkstra's Algorithm

Dijkstra( $G, s$ ) //  $l(e)$  is the length of edge  $e$

let  $T \leftarrow (\{s\}, \emptyset)$  and PQ be an empty priority queue

for each neighbor  $v$  of  $s$ , add edge  $(s,v)$  to PQ with priority  $l(e)$

while  $T$  doesn't have all vertices of  $G$  and PQ is non-empty

repeat

$e \leftarrow \text{PQ.removeMin}()$  // skip edges with both ends in  $T$

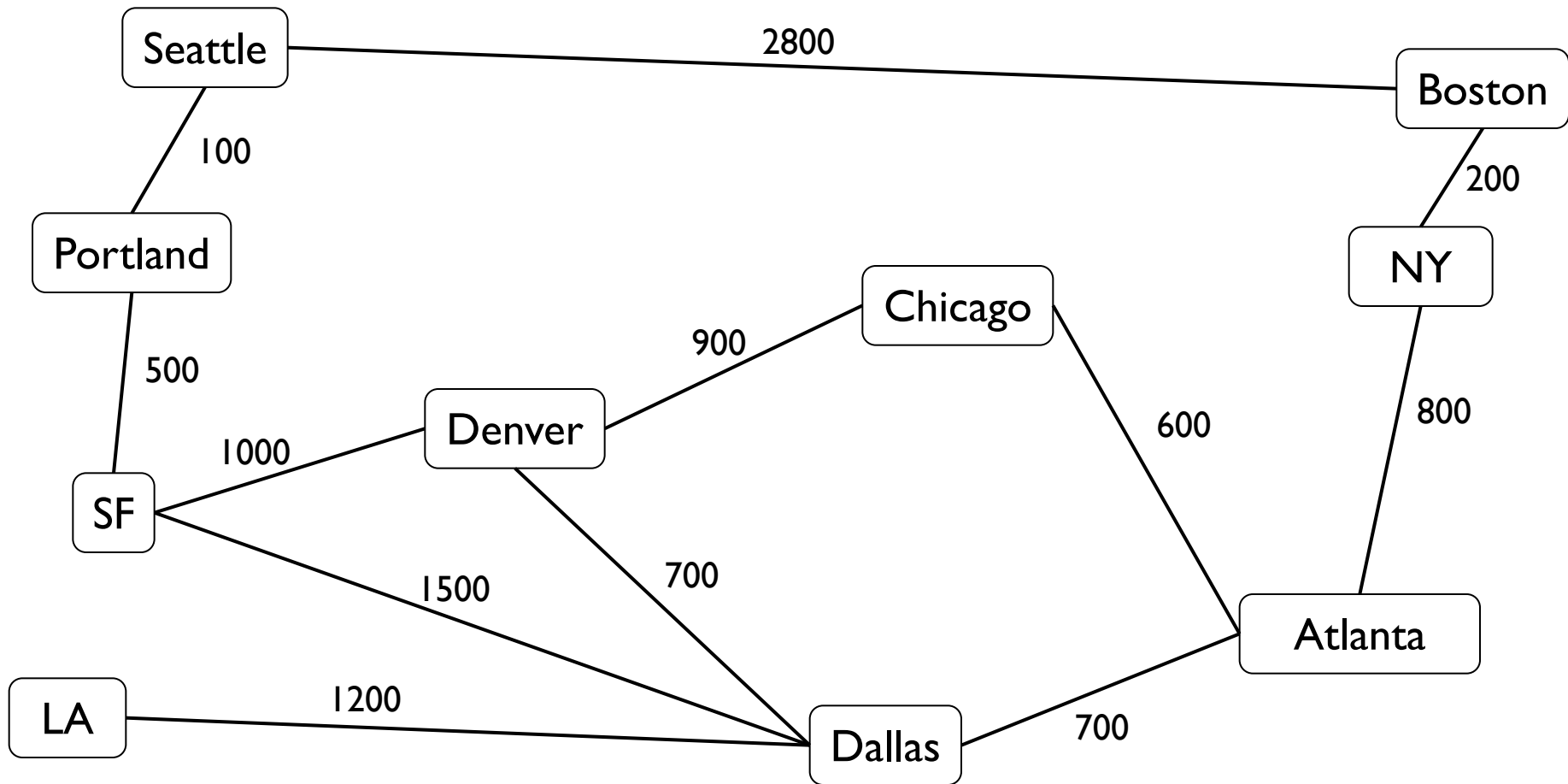
until PQ is empty or  $e=(u,v)$  for  $u \in T, v \notin T$

if  $e=(u,v)$  for  $u \in T, v \notin T$

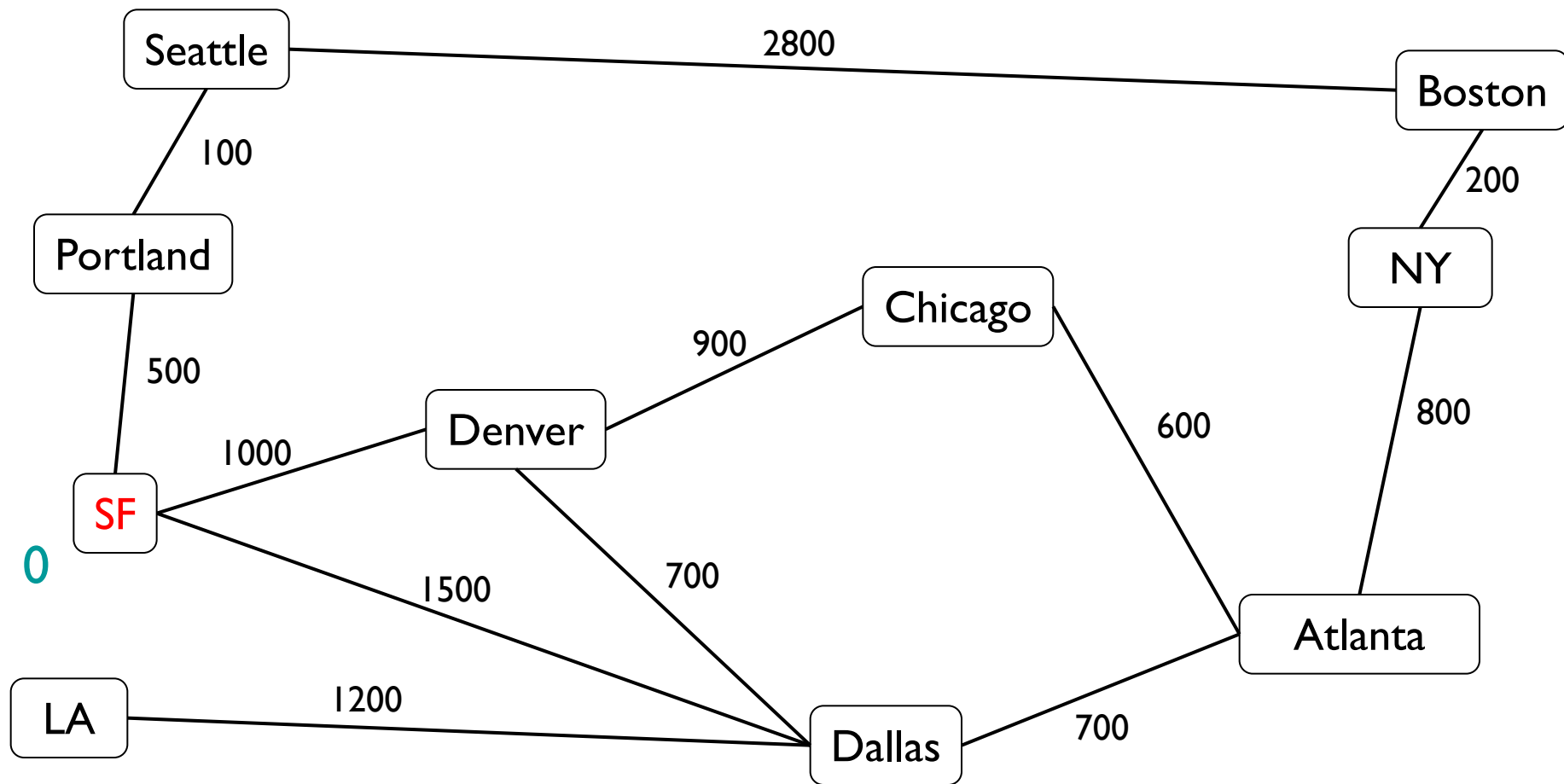
    add  $e$  (and  $v$ ) to  $T$

    for each neighbor  $w$  of  $v$

        add edge  $(v,w)$  to PQ with weight/key  $d(s,v) + l(v,w)$

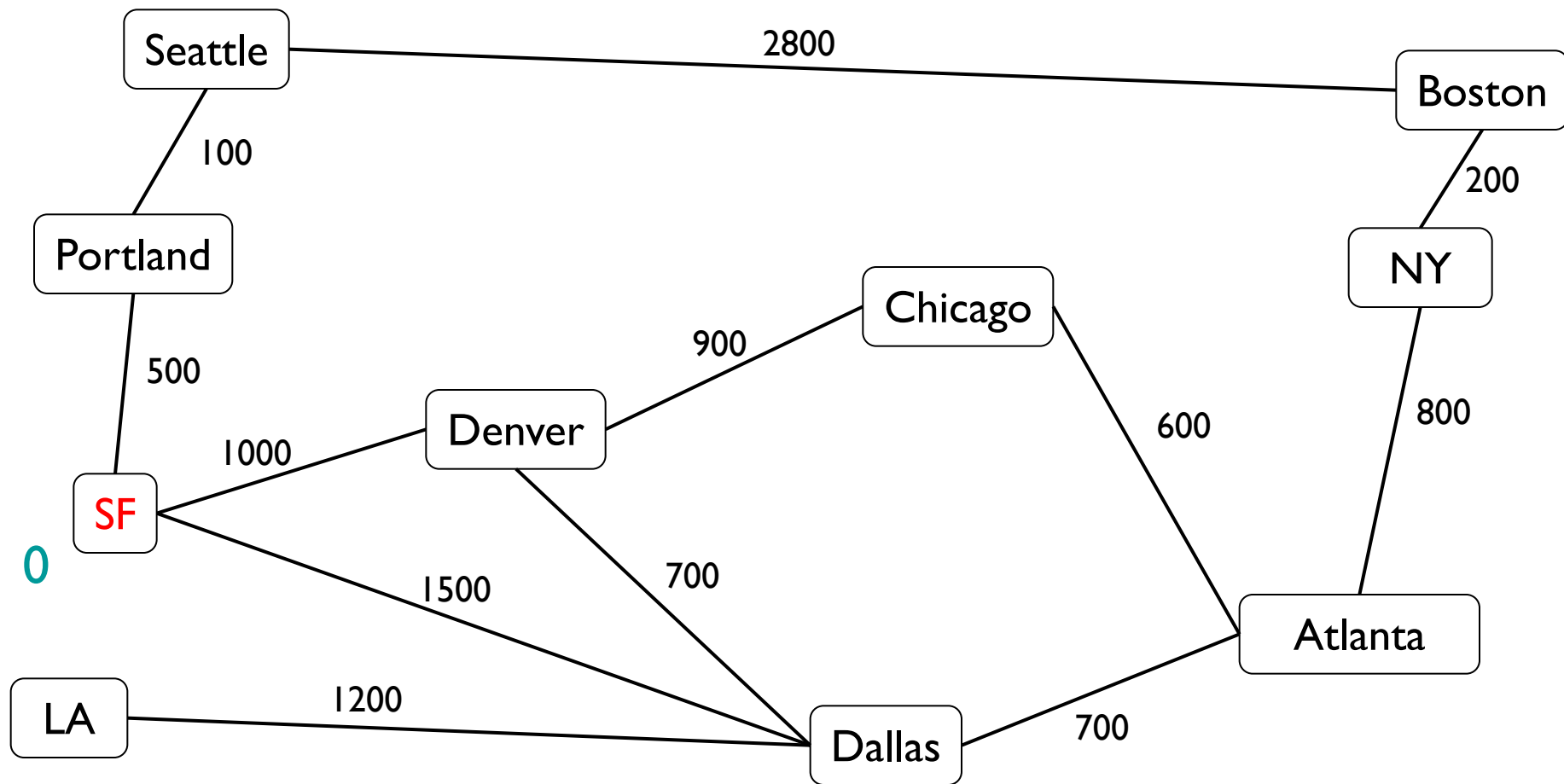


# Dijkstra's Algorithm



Priority Queue





0

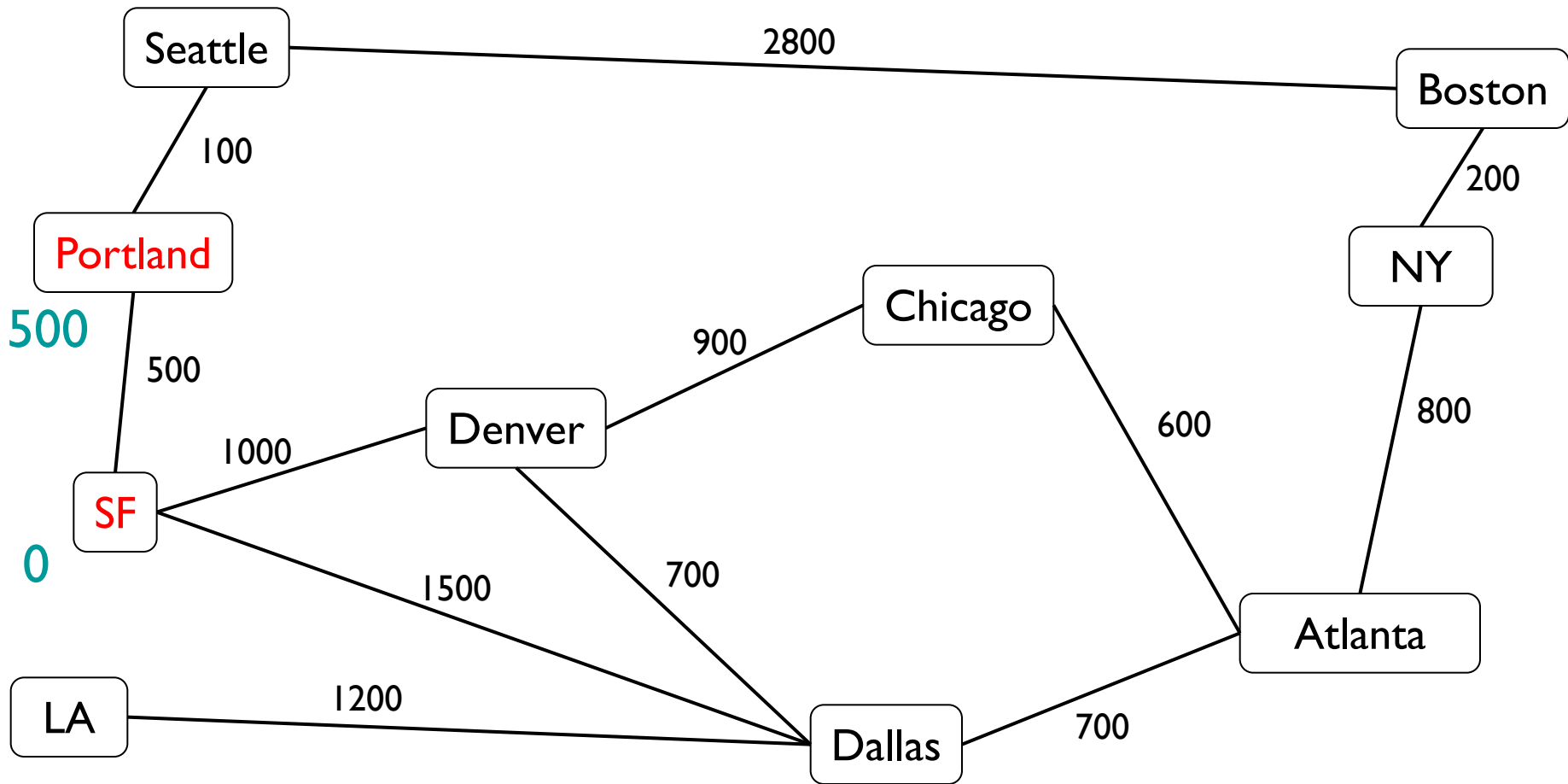
## Priority Queue



SF->Port;  
500

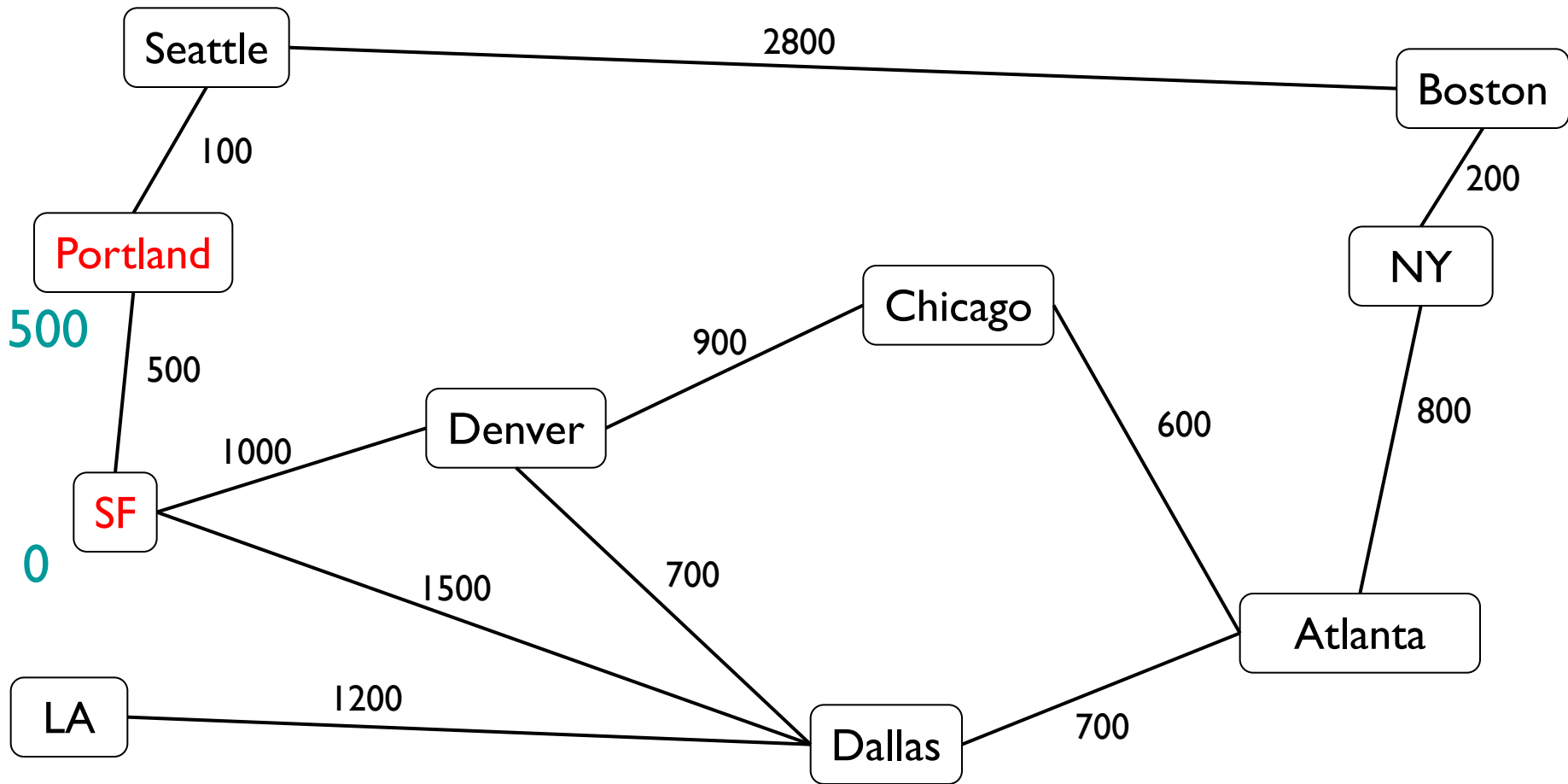
SF->Den;  
1000

SF->Dal  
1500



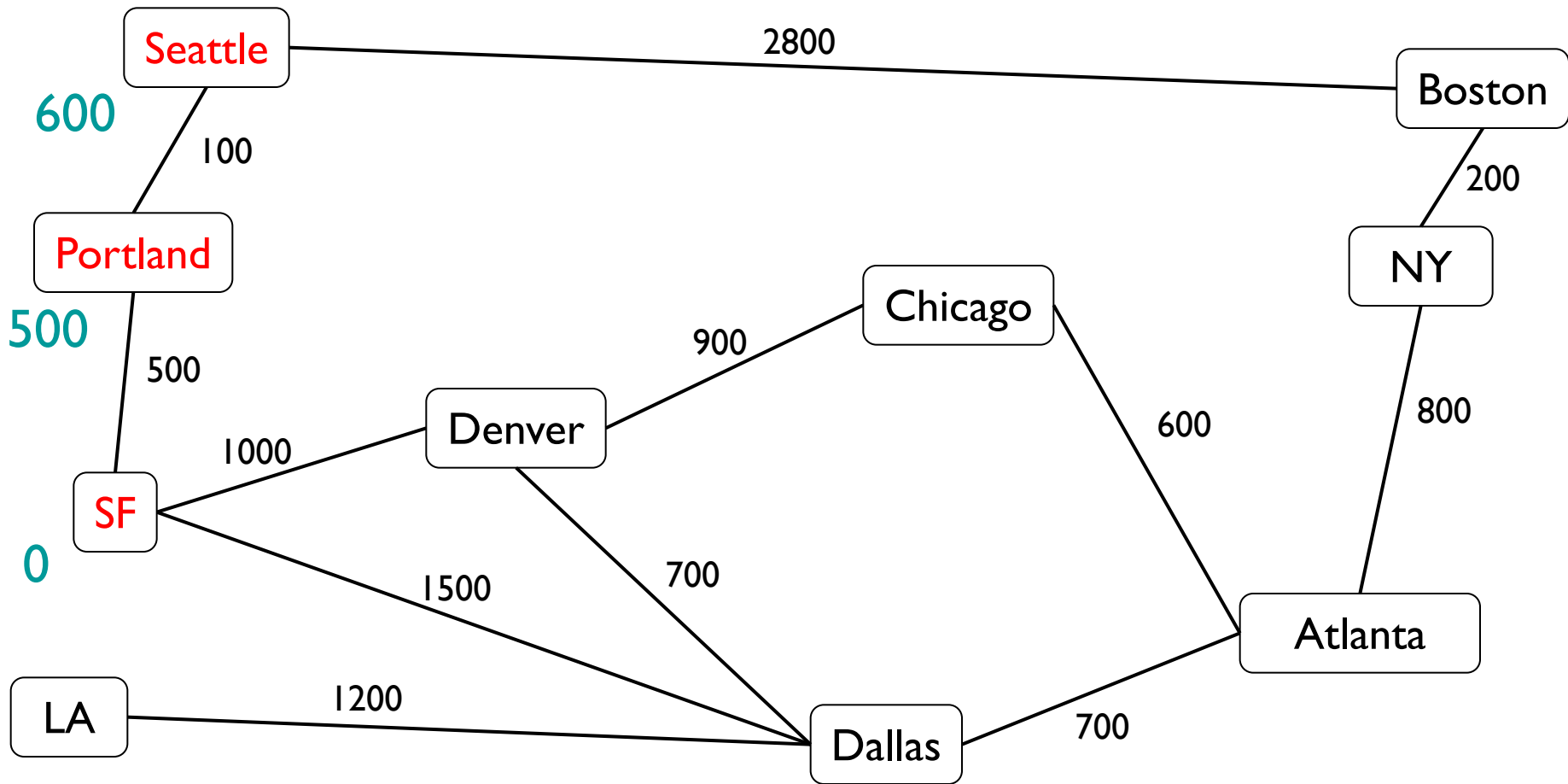
Current: 500 SF->Port (need to add Port's neighbors to PQ)

➡ SF->Den; 1000      SF->Dal 1500



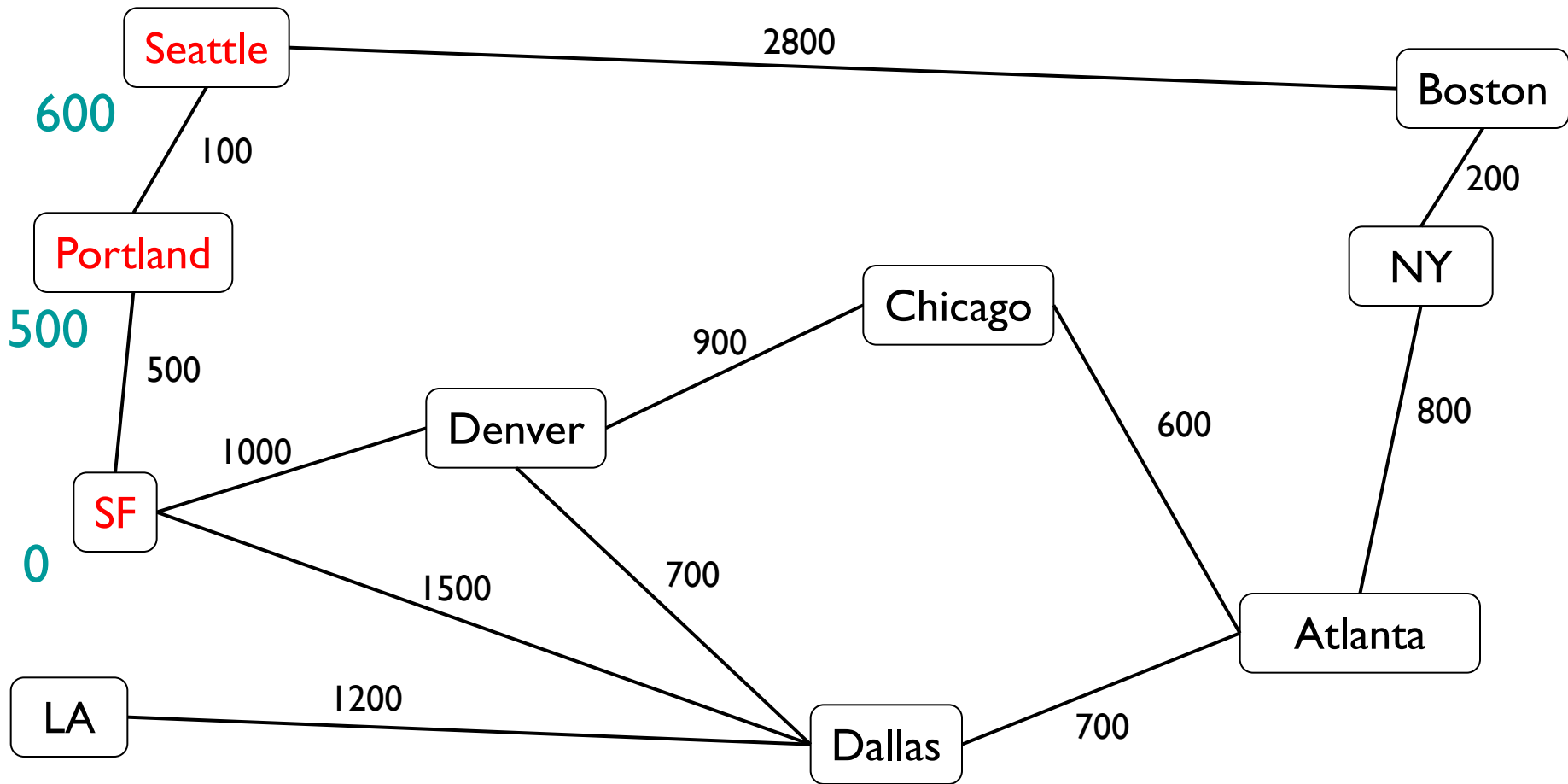
Current: 500 SF->Port

➡ SF->Port->Sea; 600      SF->Den; 1000      SF->Dal; 1500

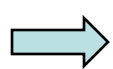


Current: 600 SF->Port->Sea

➡ SF->Den;      SF->Dal  
1000              1500



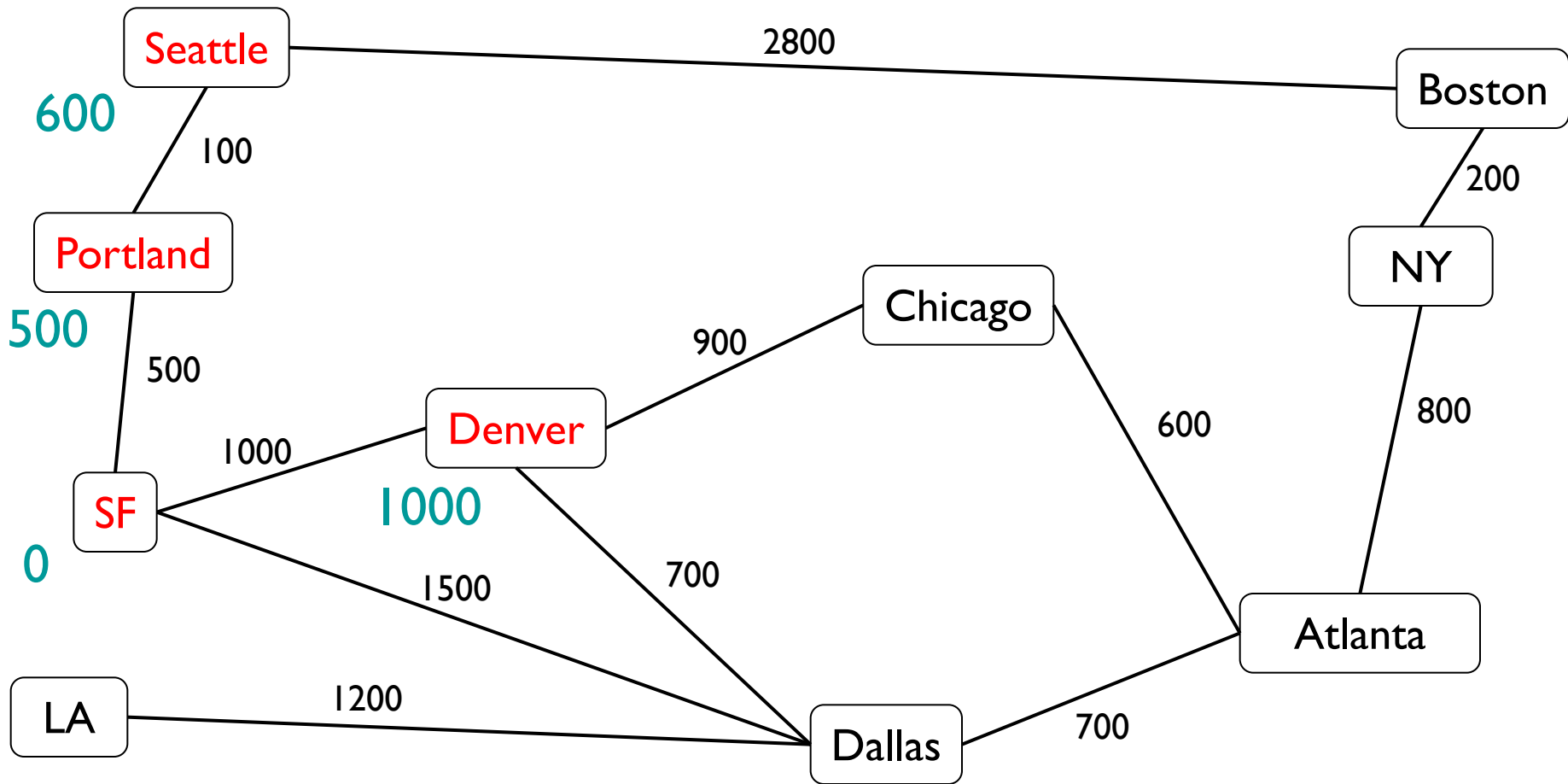
Current: 600 SF->Port->Sea



SF->Den;  
1000

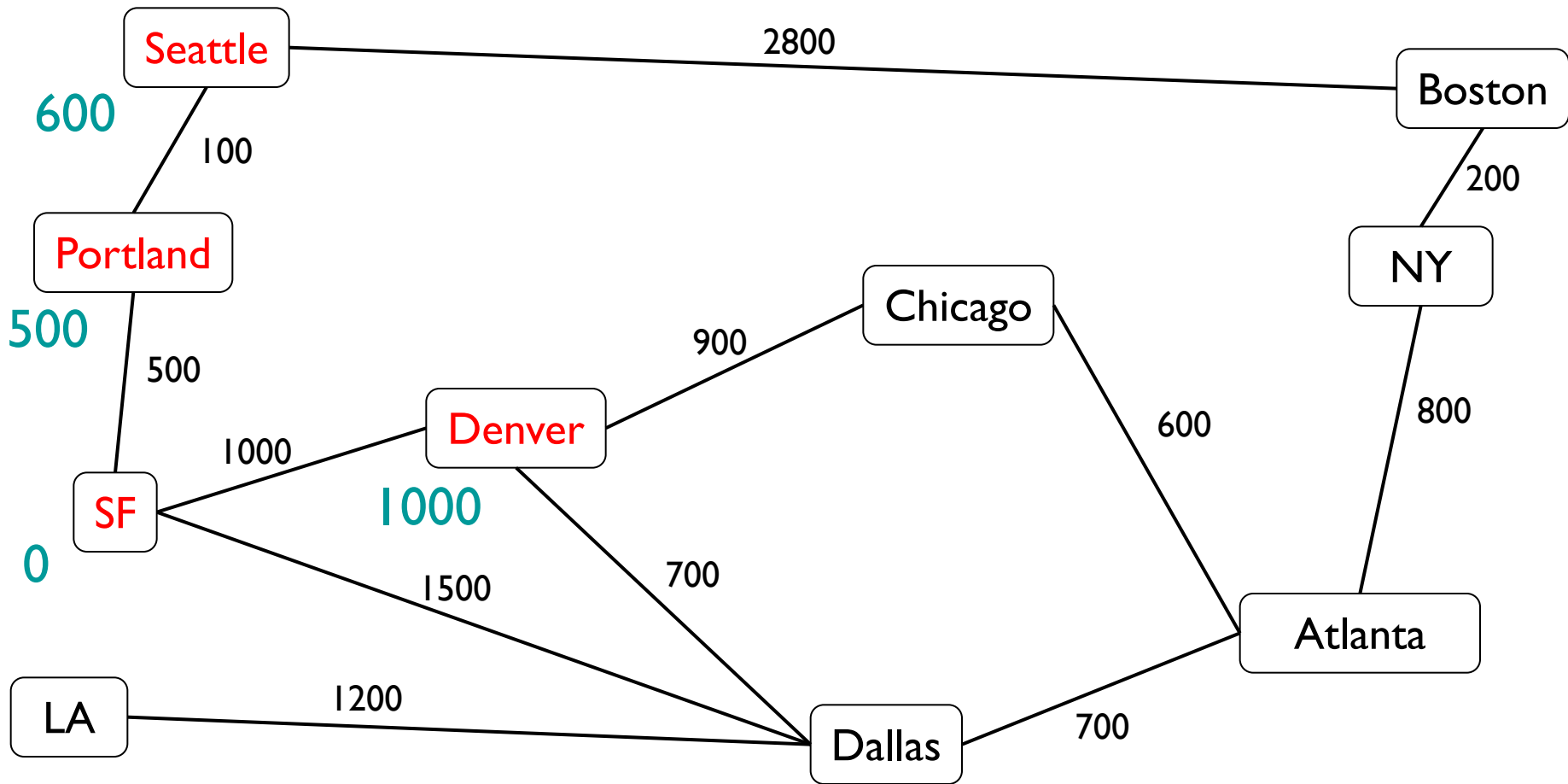
SF->Dal;  
1500

SF->Port->Sea->Bos  
3400



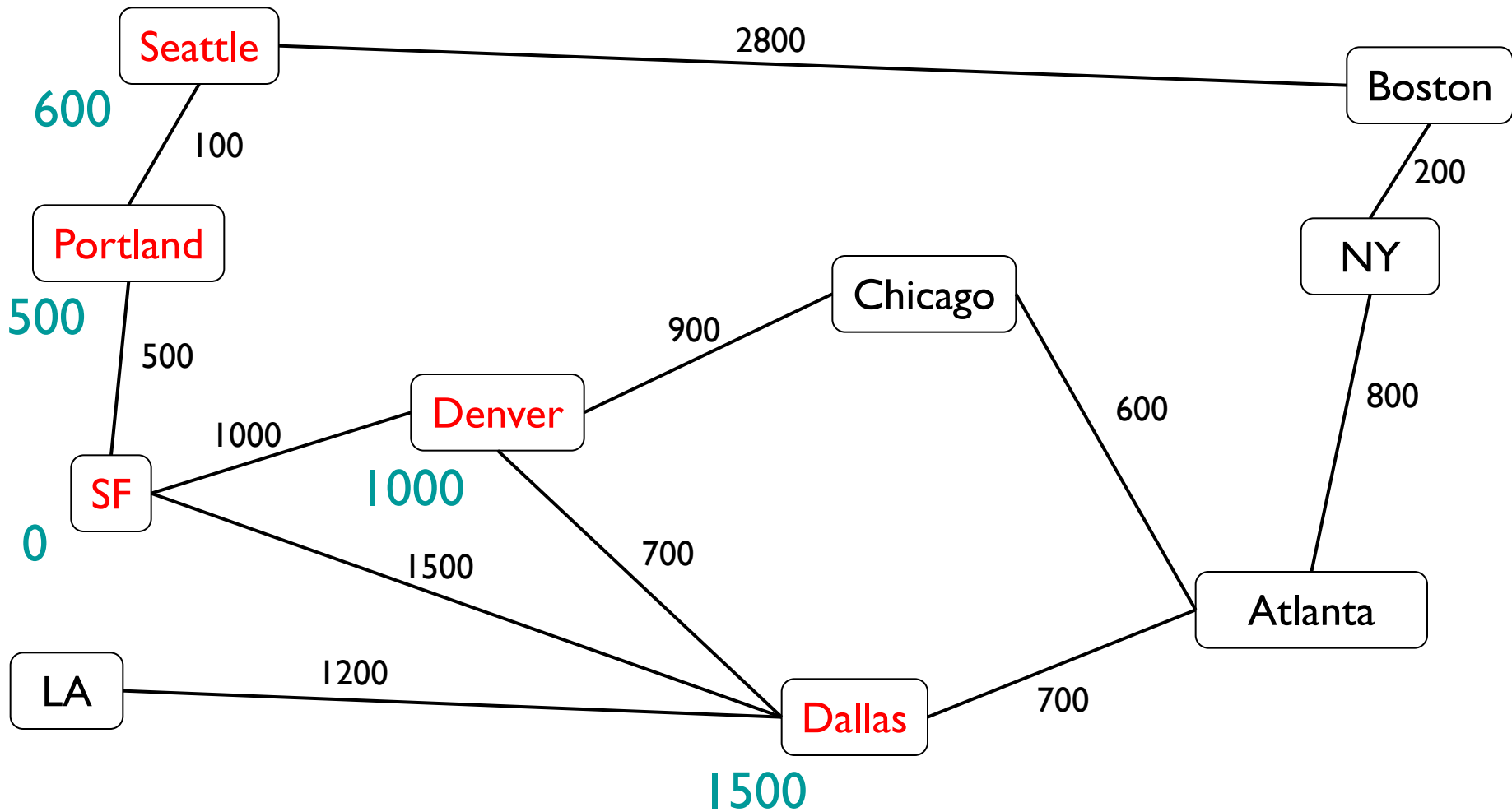
Current: 1000 SF->Den

➡ SF->Dal; 1500      SF->Port->Sea->Bos 3400



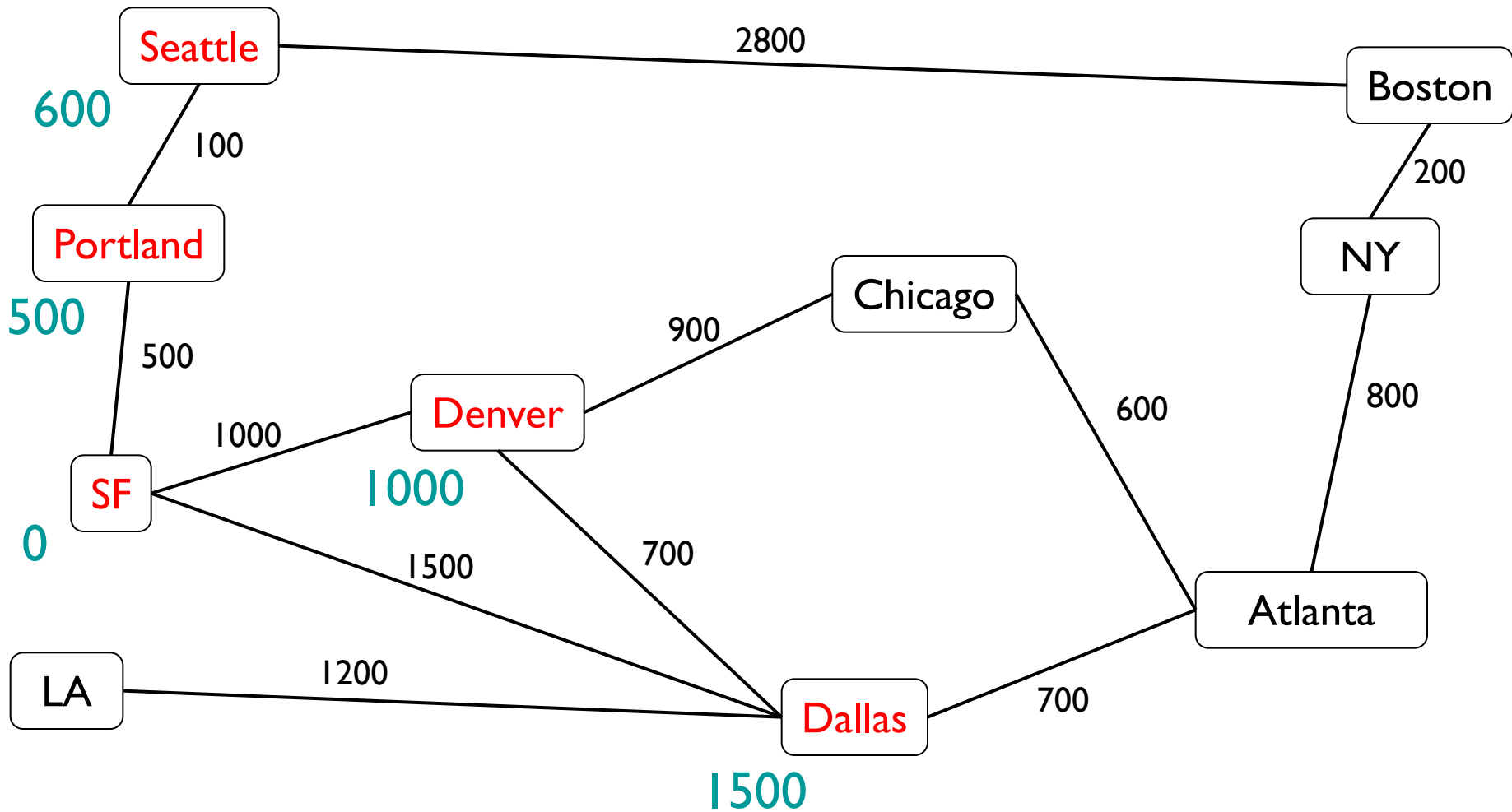
Current: 1000 SF->Den

➡ SF->Dal; 1500     
 SF->Den->Dal; 1700     
 SF->Den->Chi; 1900     
 SF->Port->Sea->Bos 3400



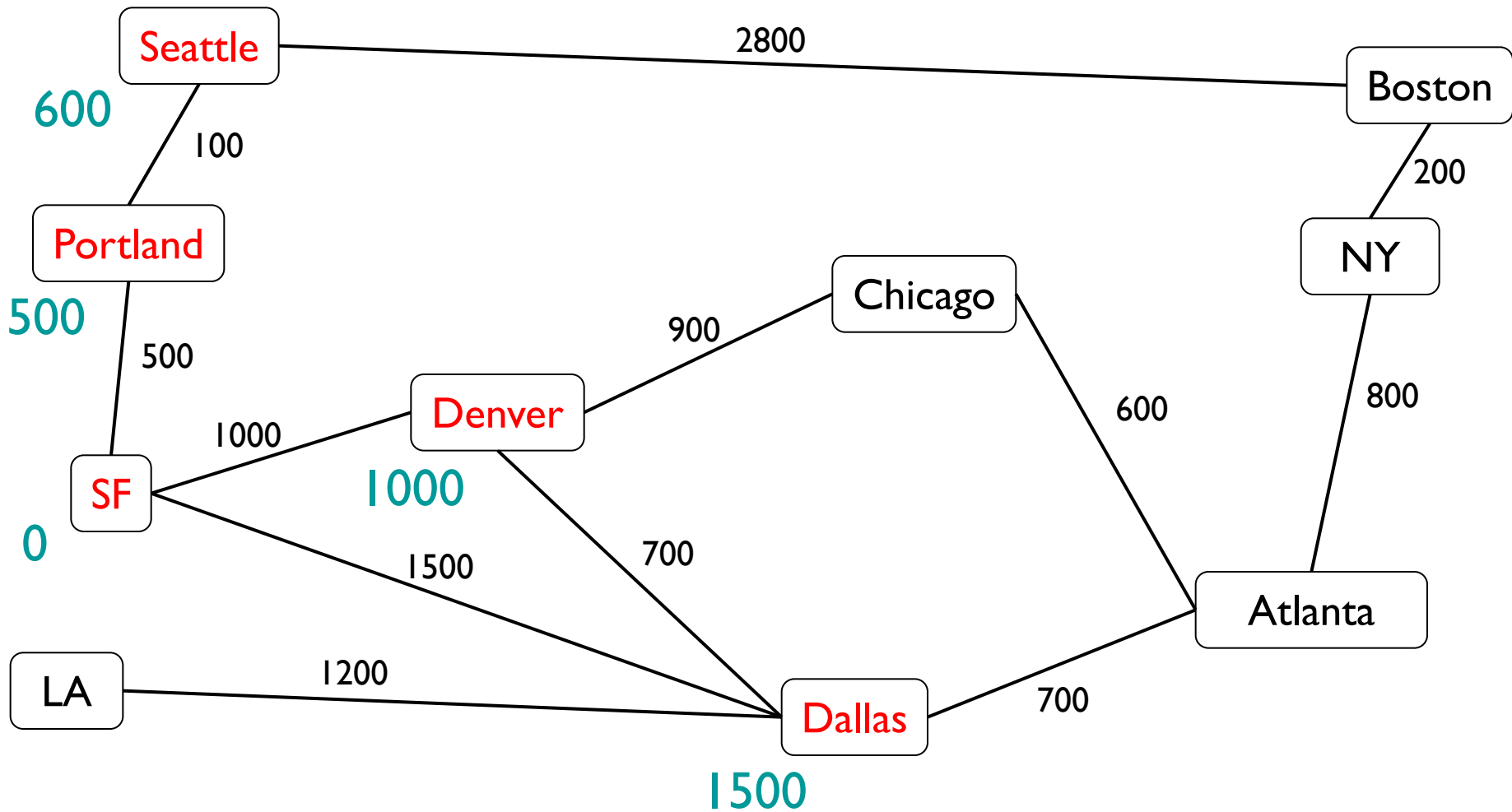
Current: 1500 SF->Dal

➡ SF->Den->Dal; 1700      SF->Den->Chi; 1900      SF->Port->Sea->Bos 3400



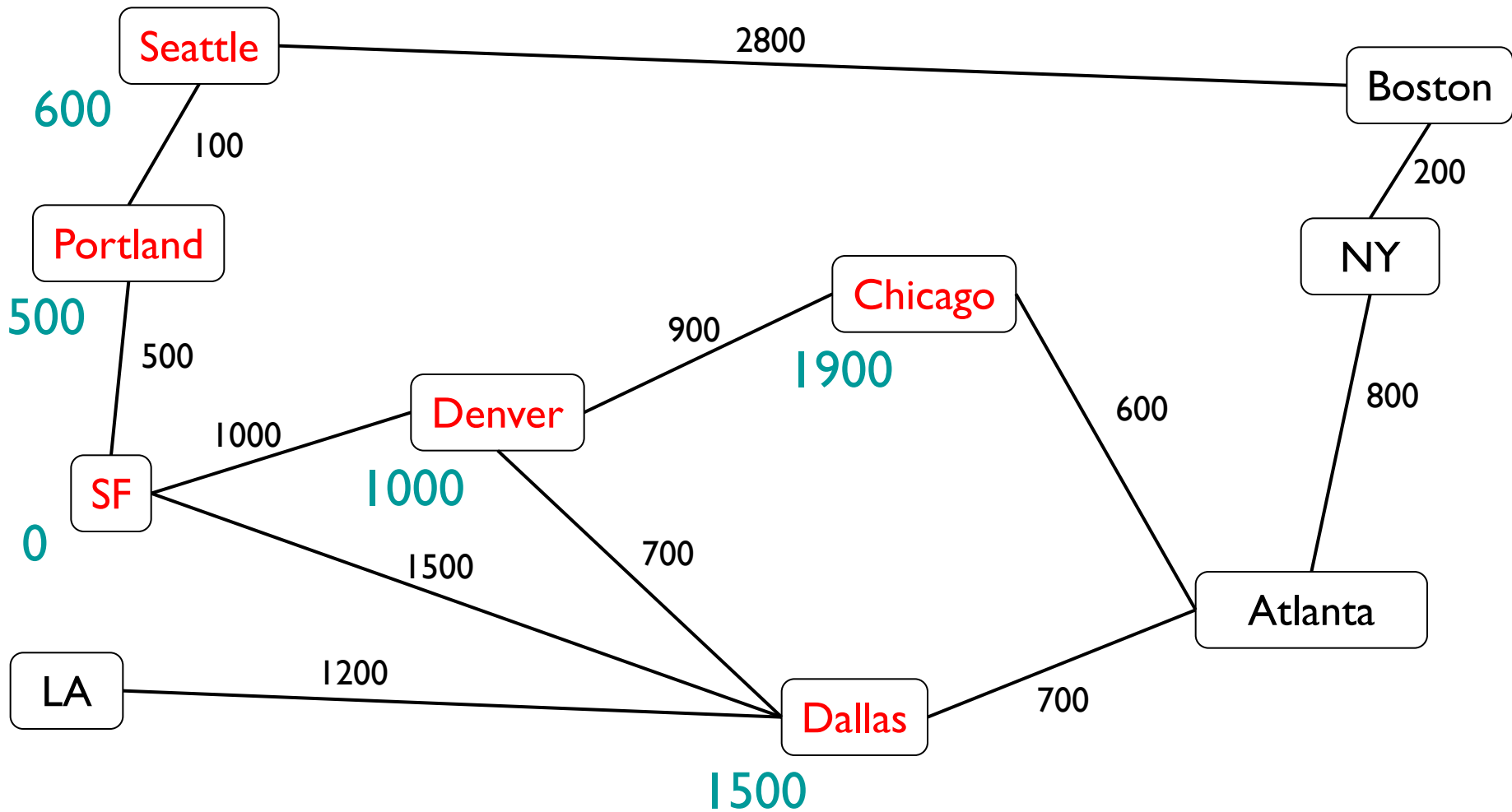
Current: 1500 SF→Dal

→ SF→Den→Dal; 1700      SF→Den→Chi; 1900      SF→Dal→Atl; 2200      SF→Dal→LA; 2700      SF→Port→Sea→Bos 3400



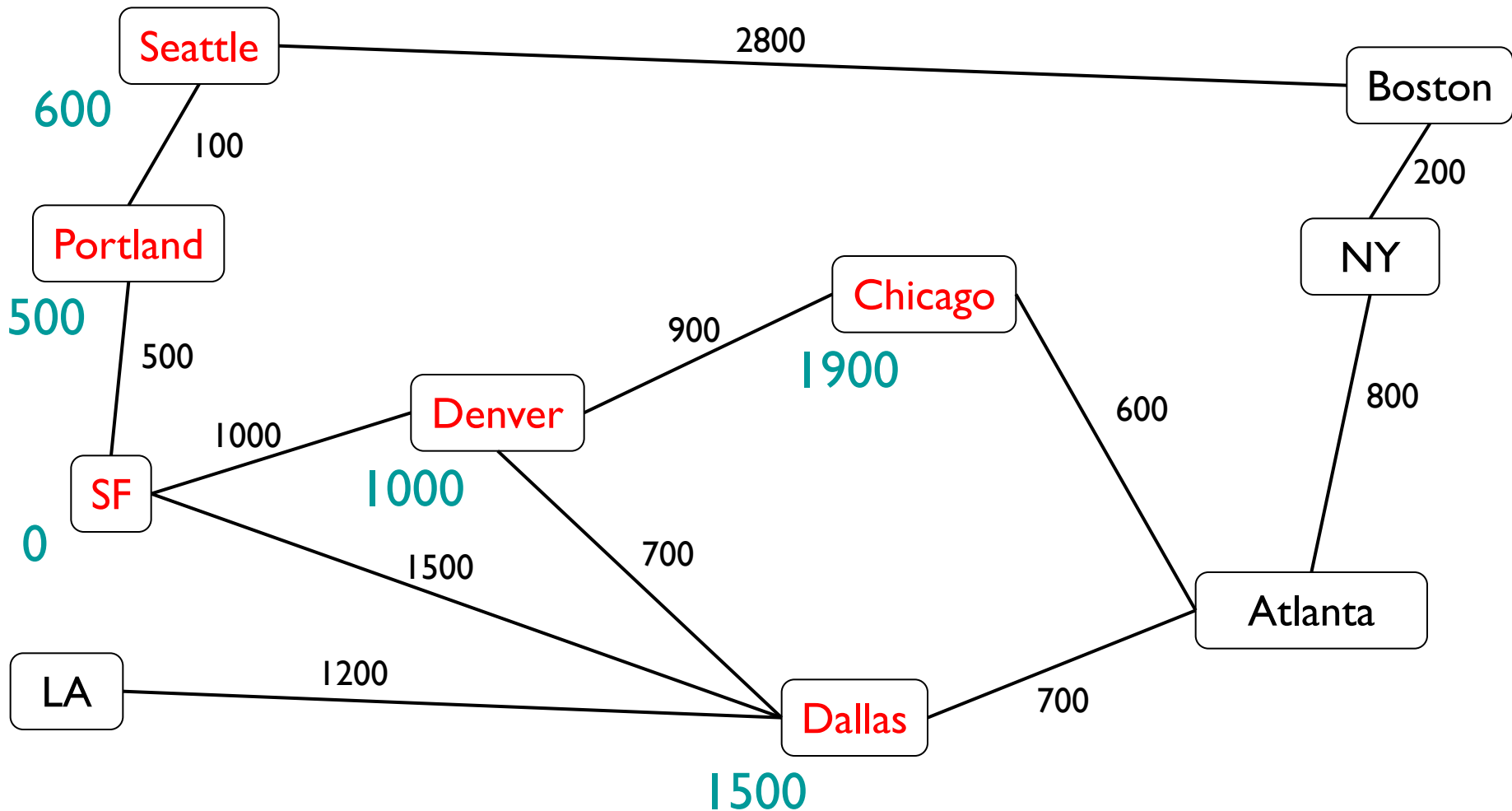
Current: 1700 SF->Den->Dal (we already have Dallas!)

➡ SF->Den->Chi; 1900      SF->Dal->Atl; 2200      SF->Dal->LA; 2700      SF->Port->Sea->Bos 3400

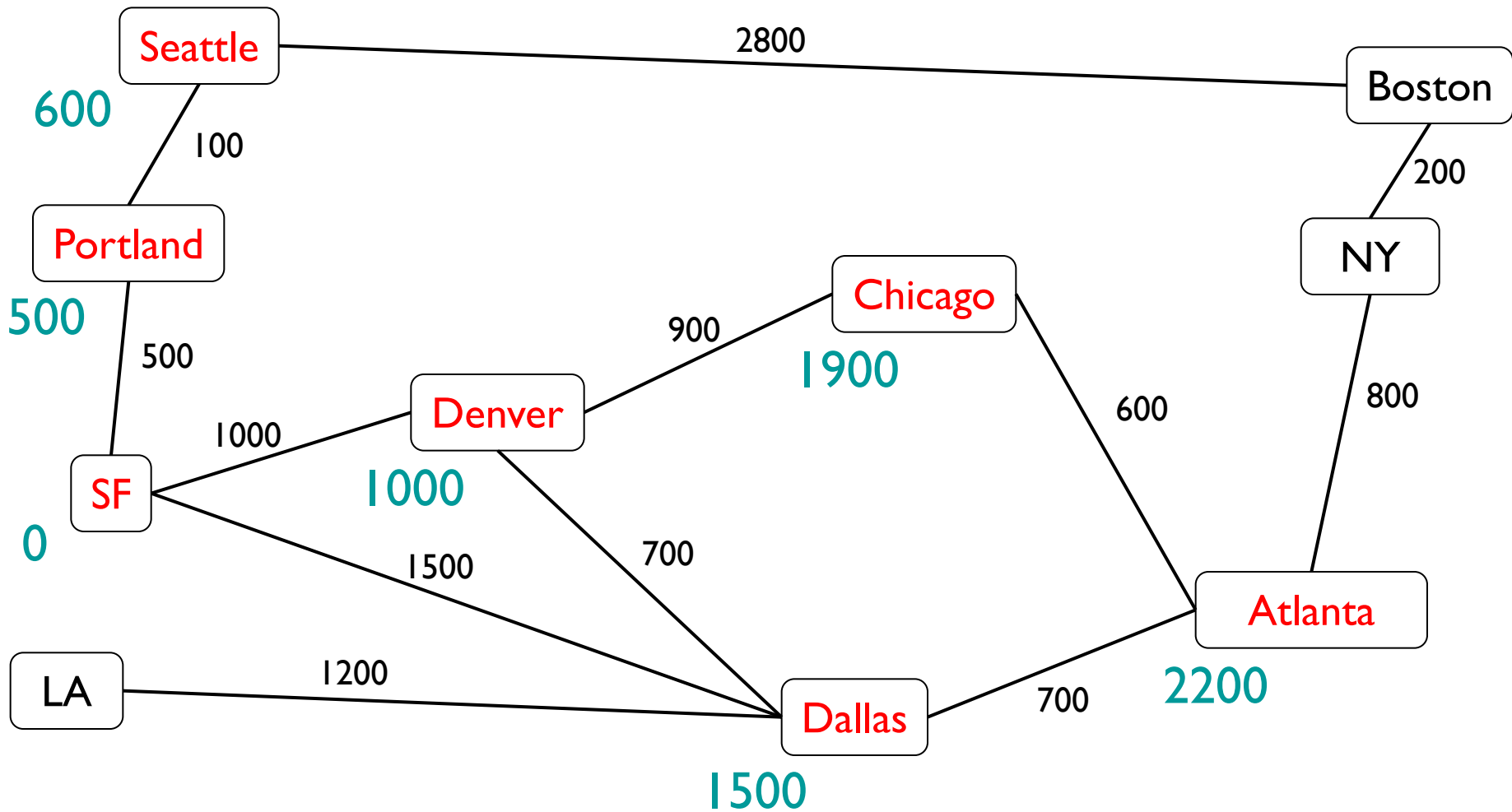


Current: 1900 SF->Den->Chi

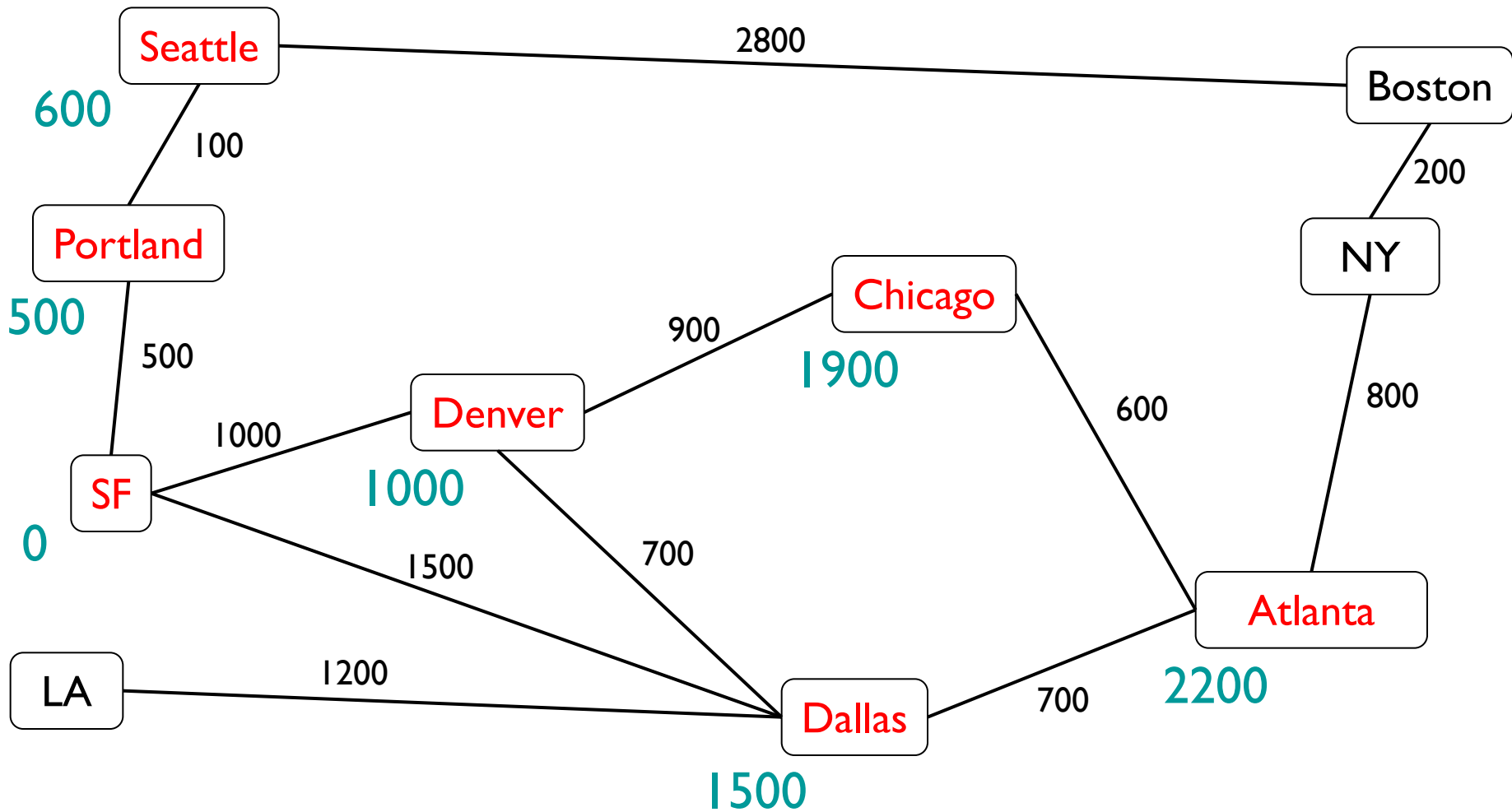
➡ SF->Dal->Atl; 2200      SF->Dal->LA; 2700      SF->Port->Sea->Bos 3400



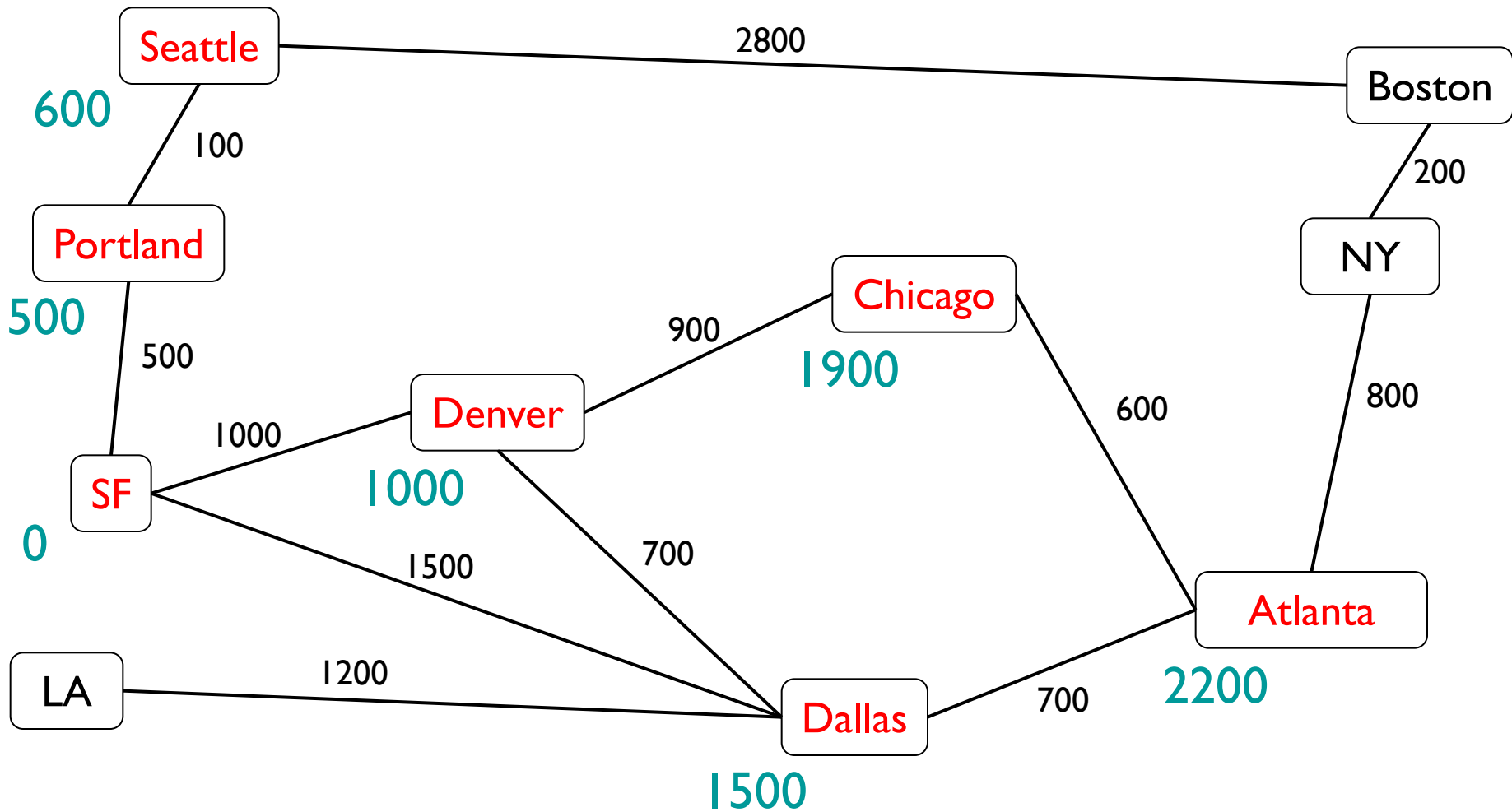
➡ SF->Dal->Atl; 2200      SF->Den->Chi->Atl; 2500      SF->Dal->LA; 2700      SF->Port->Sea->Bos 3400



➡ SF->Den->Chi->Atl; 2500      SF->Dal->LA; 2700      SF->Port->Sea->Bos 3400

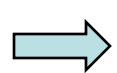
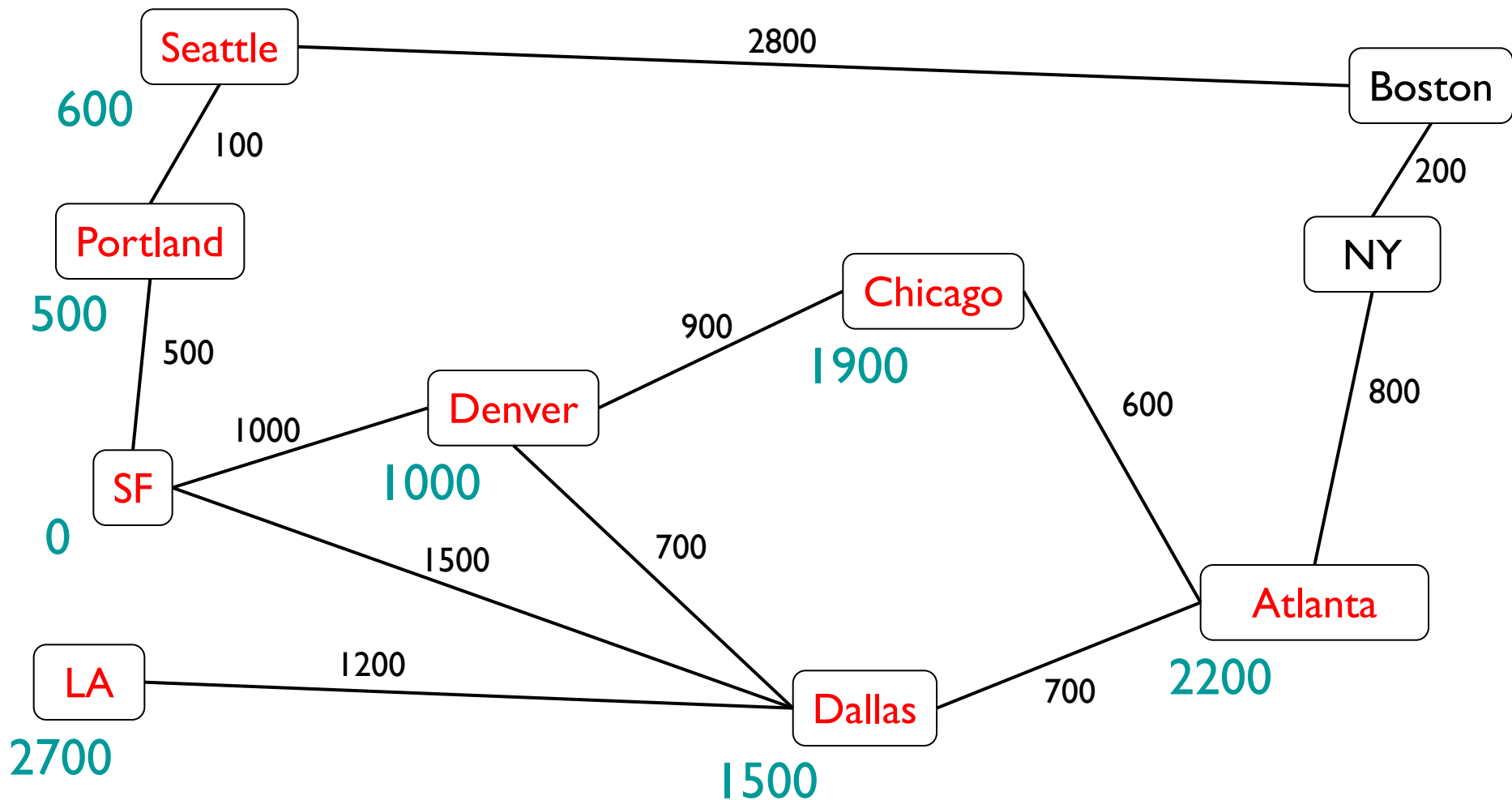


➡ SF->Den->Chi->Atl; 2500      SF->Dal->LA; 2700      SF->Dal->Atl->NY; 3000      SF->Port->Sea->Bos 3400



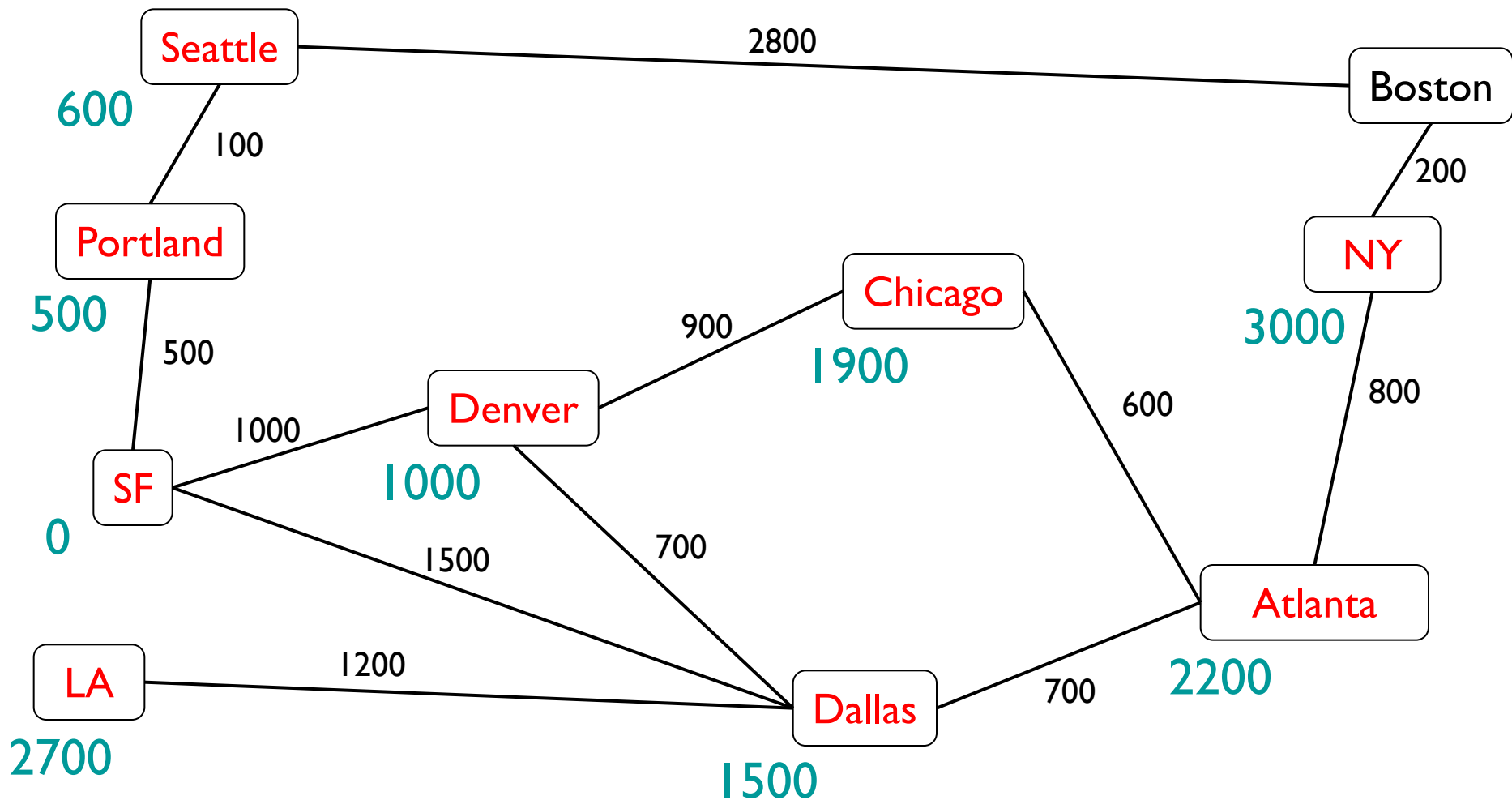
Current: 2500 SF->Den->Chi->Atl

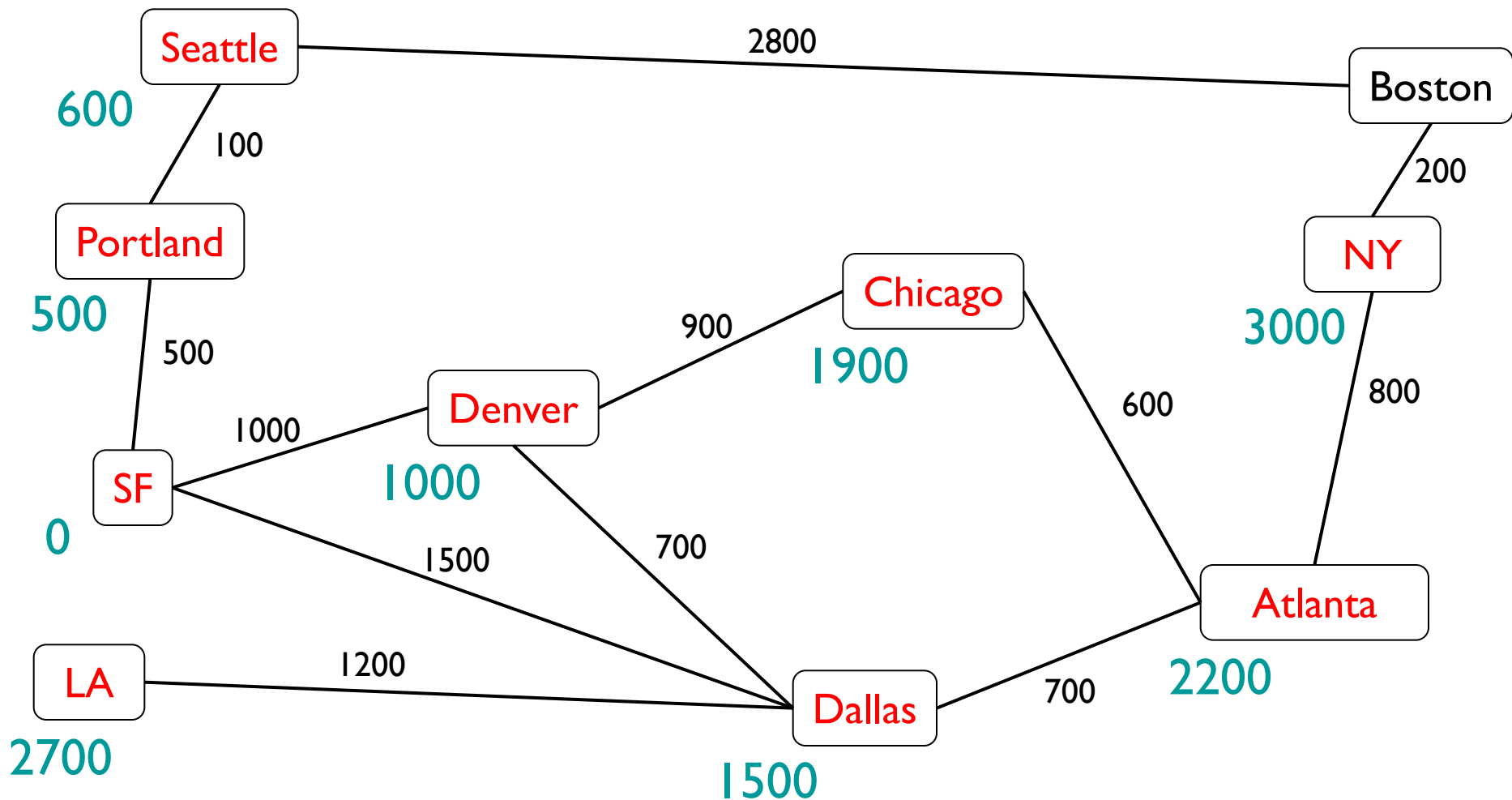
➡ SF->Dal->LA; 2700      SF->Dal->Atl->NY; 3000      SF->Port->Sea->Bos 3400



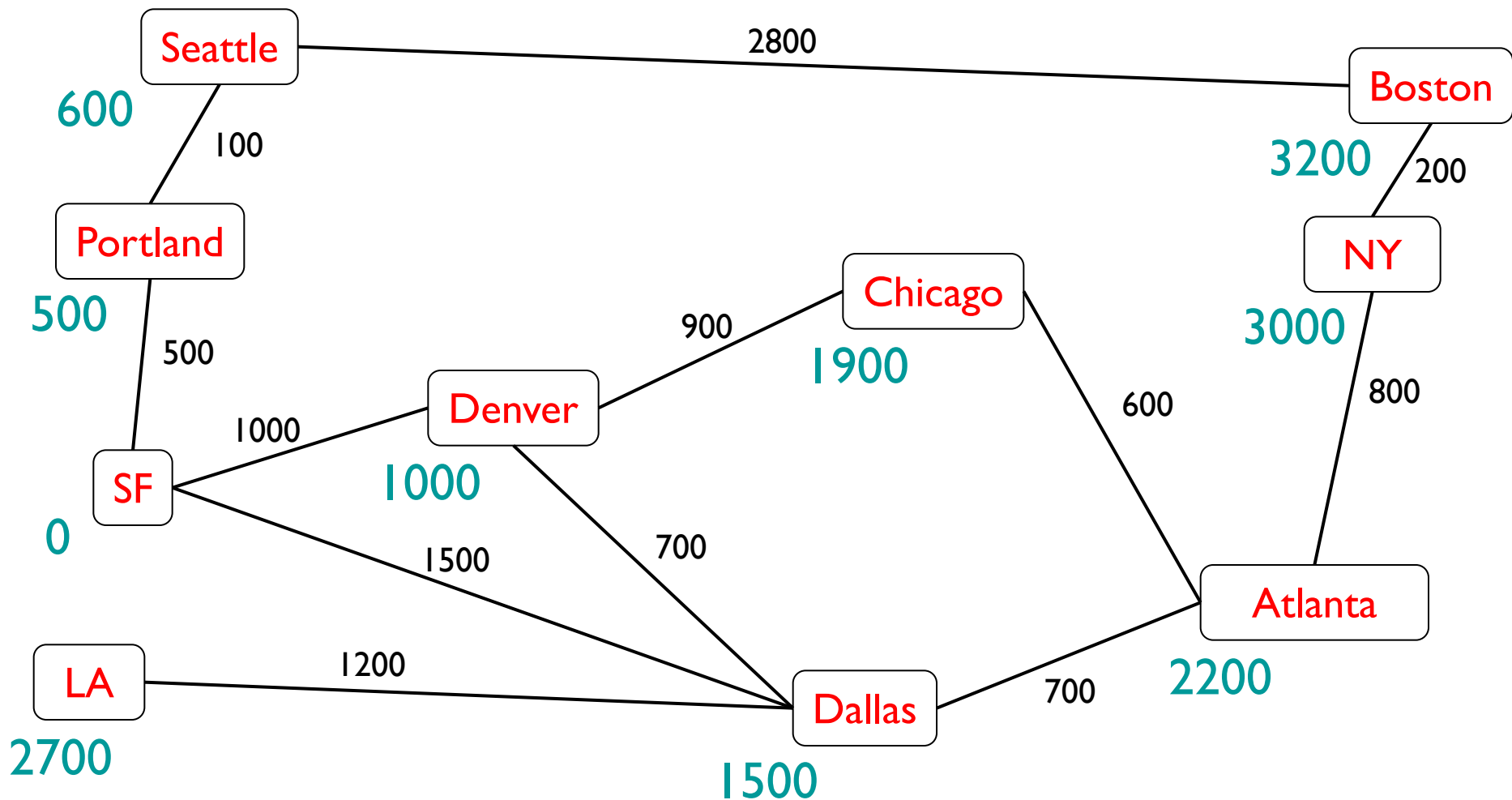
SF->Dal->Atl->NY;  
3000

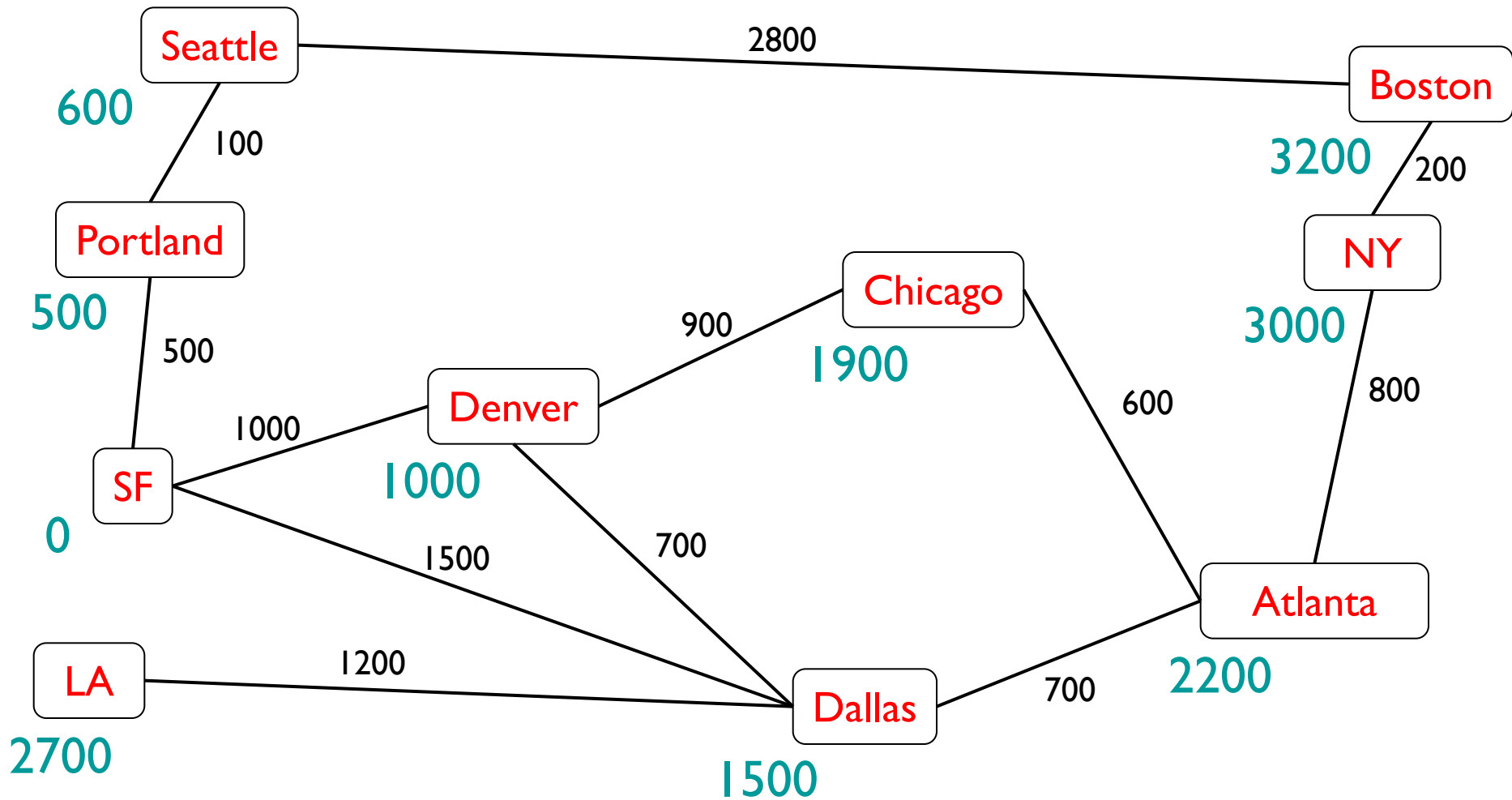
SF->Port->Sea->Bos  
3400



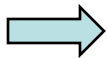


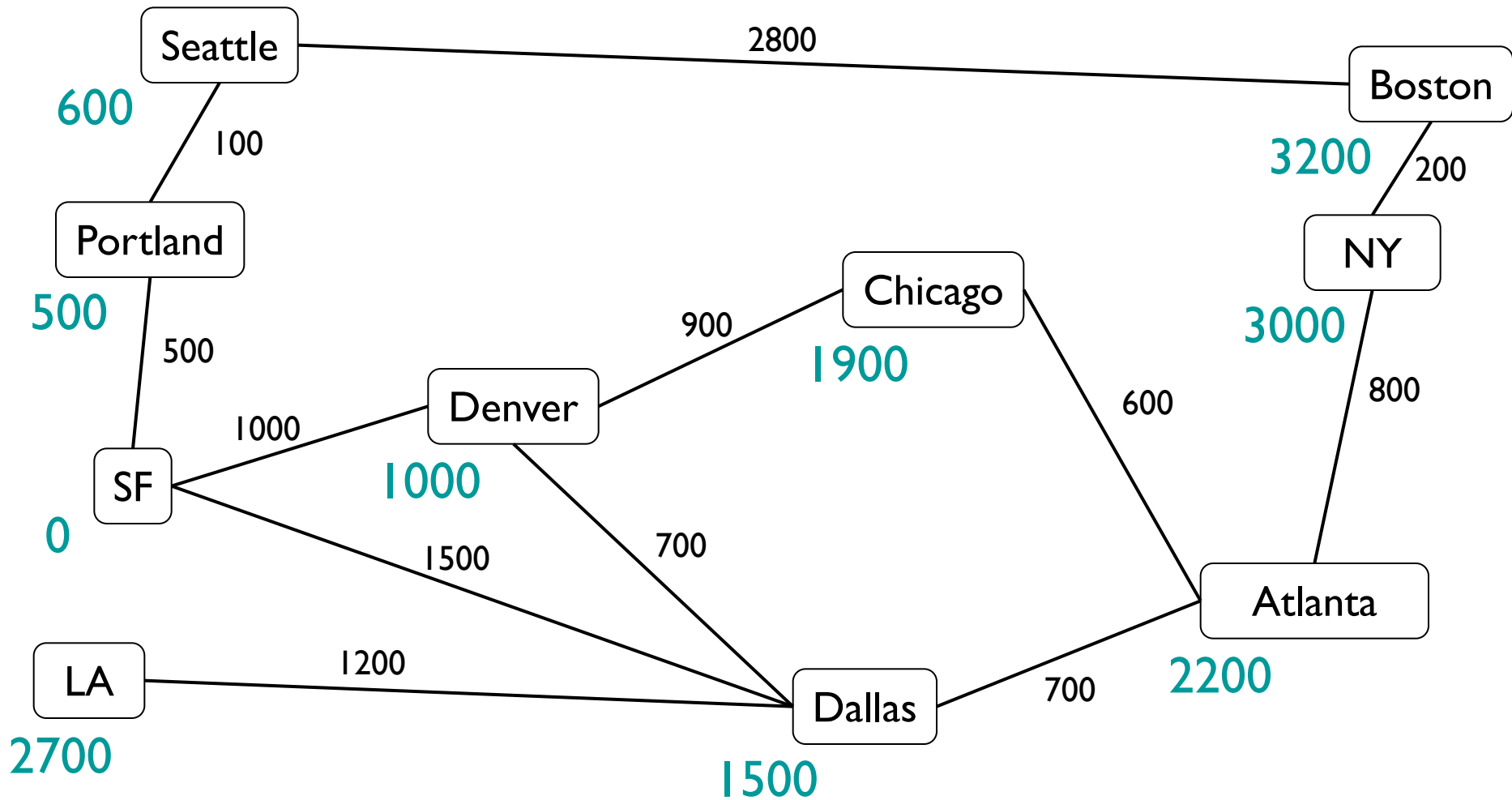
➡ SF->Dal->Atl->NY->Bos; 3200  
 SF->Port->Sea->Bos 3400



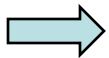


Current: 3400 SF->Port->Sea->Bos





Current:



# Dijkstra: What Do We Return?

- As we find a new vertex  $v$  to add to the tree of shortest paths, add edges  $e=(v,w)$  to a map.
- Precisely:
  - Use the PQ association( $X,Y$ ) edgeInfo where
    - $X$  is  $d(s,v) + l(v,w)$
    - $Y$  is the edge  $e=(v,w)$
  - Add the key/value pair  $(w, \text{edgeInfo})$  to the map
- So the map entry with key  $w$  tells us the edge the best path used to get from the tree to  $w$

# Dijkstra: Space Complexity

- Graph:  $O(|V| + |E|)$ 
  - Each vertex and edge uses a constant amount of space
- Priority Queue  $O(|E|)$ 
  - Each edge takes up constant amount of space
- Result:  $O(|V| + |E|)$ 
  - Optimal in Big-O sense!

# Dijkstra : Time Complexity

Assume Map ops are  $O(1)$  time

Across *all* iterations of outer while loop

- Edges are added to and removed from the priority queue
  - But any edge is added (and removed) at most once!
  - Total PQ operation cost is  $O(|E| \log |E|)$  time
  - All other operations take constant time
- Thus time complexity is  $O(|E| \log |E|)$

# Improvements?

- Dijkstra's can be improved to  $O(|E| + |V| \log |V|)$  using a *Fibonacci heap*
  - Can “decrease key” in  $O(1)$
- Prim's can be improved to  $O(|E| \alpha(|E|))$  time
  - Where  $\alpha$  is the incredibly-slowly-growing “inverse Ackermann” function
- It is not known if either of these can be improved further

# Let's look at the code

# After Thanksgiving

- Maps!!!
- Start reviewing for the final