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# Dynamic Programming

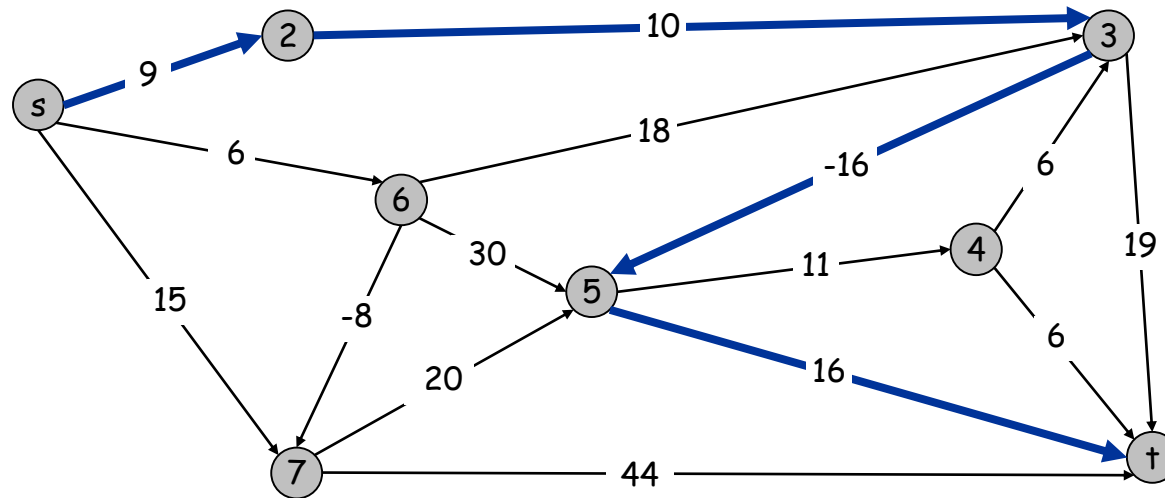
## Bellman-Ford Algorithm for Shortest Paths

# Shortest Paths

**Shortest path problem.** Given a directed graph  $G = (V, E)$ , with edge weights  $c_{vw}$ , find shortest path from node  $s$  to node  $t$ .

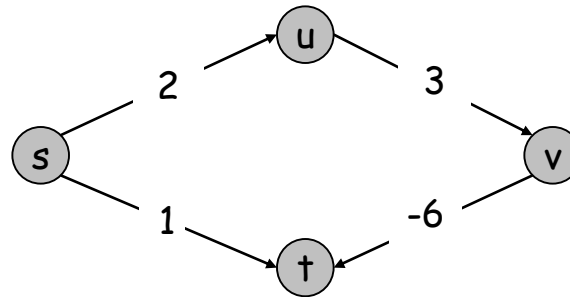
↖ allow negative weights

**Ex.** Nodes represent agents in a financial setting and  $c_{vw}$  is cost of transaction in which we buy from agent  $v$  and sell immediately to  $w$ .

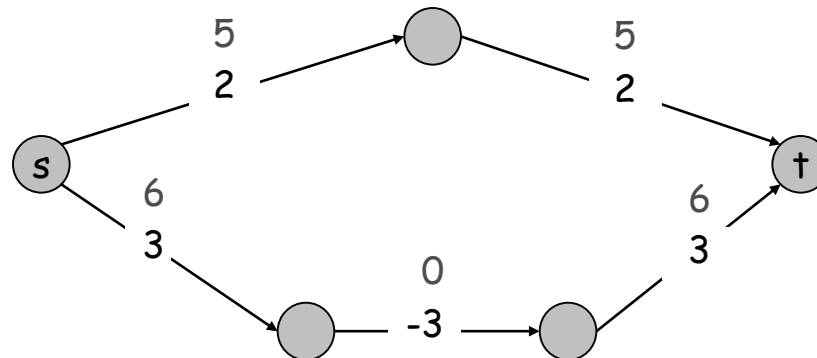


# Shortest Paths: Failed Attempts

Dijkstra. Can fail if negative edge costs.

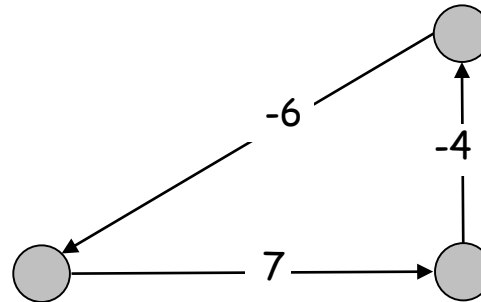


Re-weighting. Adding a constant to every edge weight can fail.

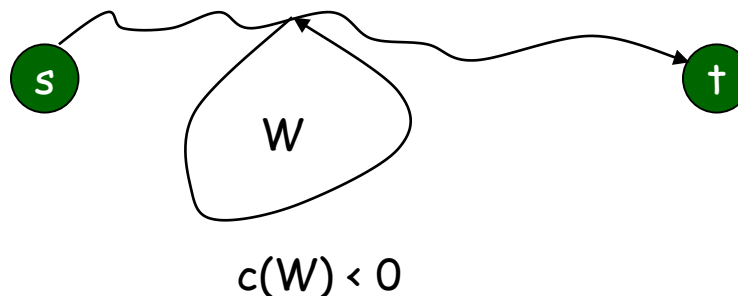


# Shortest Paths: Negative Cost Cycles

Negative cost cycle.



**Observation.** If some path from  $s$  to  $t$  contains a negative cost cycle, there does not exist a shortest  $s$ - $t$  path; otherwise, there exists one that is simple.



# Shortest Paths: Dynamic Programming

**Def.**  $OPT(i, v)$  = length of shortest  $v$ - $t$  path  $P$  using at most  $i$  edges.

- Case 1:  $P$  uses at most  $i-1$  edges.
  - $OPT(i, v) = OPT(i-1, v)$
- Case 2:  $P$  uses exactly  $i$  edges.
  - if  $(v, w)$  is first edge, then  $OPT$  uses  $(v, w)$ , and then selects best  $w$ - $t$  path using at most  $i-1$  edges

$$OPT(i, v) = \begin{cases} \infty & \text{if } i = 0 \\ \min \left\{ OPT(i-1, v), \min_{(v, w) \in E} \{ OPT(i-1, w) + c_{vw} \} \right\} & \text{otherwise} \end{cases}$$

- $OPT(0, t) = 0$

**Remark.** By previous observation, if no negative cycles, then  $OPT(n-1, v)$  = length of shortest  $v$ - $t$  path.

# Shortest Paths: Implementation

```
Shortest-Path( $G, t$ ) {  
    foreach node  $v \in V$   
         $M[0, v] \leftarrow \infty$   
     $M[0, t] \leftarrow 0$   
  
    for  $i = 1$  to  $n-1$   
        foreach node  $v \in V$   
             $M[i, v] \leftarrow M[i-1, v]$   
            foreach edge  $(v, w) \in E$   
                 $M[i, v] \leftarrow \min \{ M[i, v], M[i-1, w] + c_{vw} \}$   
}
```

**Analysis.**  $\Theta(mn \log(n))$  time,  $\Theta(n^2 \log(n))$  space.

**Finding the shortest paths.** Maintain a "successor" for each table entry. We will see it shortly.

# Shortest Paths: Practical Improvements

## Practical improvements.

- Maintain only one array  $M[v]$  = shortest  $v$ - $t$  path that we have found so far.
- No need to check edges of the form  $(v, w)$  unless  $M[w]$  changed in previous iteration.

**Theorem.** Throughout the algorithm,  $M[v]$  is length of some  $v$ - $t$  path, and after  $i$  rounds of updates, the value  $M[v]$  is no larger than the length of shortest  $v$ - $t$  path using  $\leq i$  edges.

## Overall impact.

- Memory:  $O((m + n)\log n)$ .
- Running time:  $O(mn \log n)$  worst case, but substantially faster in practice.

# Bellman-Ford: Efficient Implementation

```
Push-Based-Shortest-Path( $G, s, t$ ) {  
  foreach node  $v \in V$  {  
     $M[v] \leftarrow \infty$   
    successor[ $v$ ]  $\leftarrow \phi$   
  }  
  
   $M[t] = 0$   
  for  $i = 1$  to  $n-1$  {  
    foreach node  $w \in V$  {  
      if ( $M[w]$  has been updated in previous iteration) {  
        foreach node  $v$  such that  $(v, w) \in E$  {  
          if ( $M[v] > M[w] + c_{vw}$ ) {  
             $M[v] \leftarrow M[w] + c_{vw}$   
            successor[ $v$ ]  $\leftarrow w$   
          }  
        }  
      }  
    }  
    If no  $M[w]$  value changed in iteration  $i$ , stop.  
  }  
}
```



# Summary

## Greedy Algorithms:

- Interval Scheduling (a.k.a. Activity-Selection) [CLRS, Ch16.1]
- Shortest Paths in a Graph, Dijkstra's Algorithm [CLRS, Ch24.3]
- Minimum Spanning Tree [CLRS, Ch 23.1-2]
- Huffman code [CLRS, Ch16.3]

## Dynamic Programming:

- Shortest path graphs, Bellman-Ford algorithm [CLRS, Ch24.1]

## Next Lecture. Dynamic Programming:

- Weighted interval scheduling
- Knapsack
- Longest common subsequence [CLRS] Ch15.4
- RNA secondary structure