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Fall'19 CSCE 629

Analysis of Algorithms

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Lecture 13

- Bellman-Ford algorithm:
Shortest path with negative weight

Credit: based on slides by A.Smith and K.Wayne

Logistics

■ Write up your algorithm

- Start with “**Idea**”: short description of the key to your algorithm. [Your peer should be able to take it and come up with an algorithm with *a bit*, if any, thought]
- Pseudocode: write in a nice format as you’d do in a real program, even if you are using plain-English descriptions [Your peer should be able to implement your algorithm **without** thinking other than implementation details]
- Comments in your pseudocode are helpful

■ When studying

- Identify what you **don’t** understand
- Test yourself via (a) rederiving algorithms, proofs from class and (b) exercises in book.
- Read alternative explanations (e.g. KT,E,DPV)

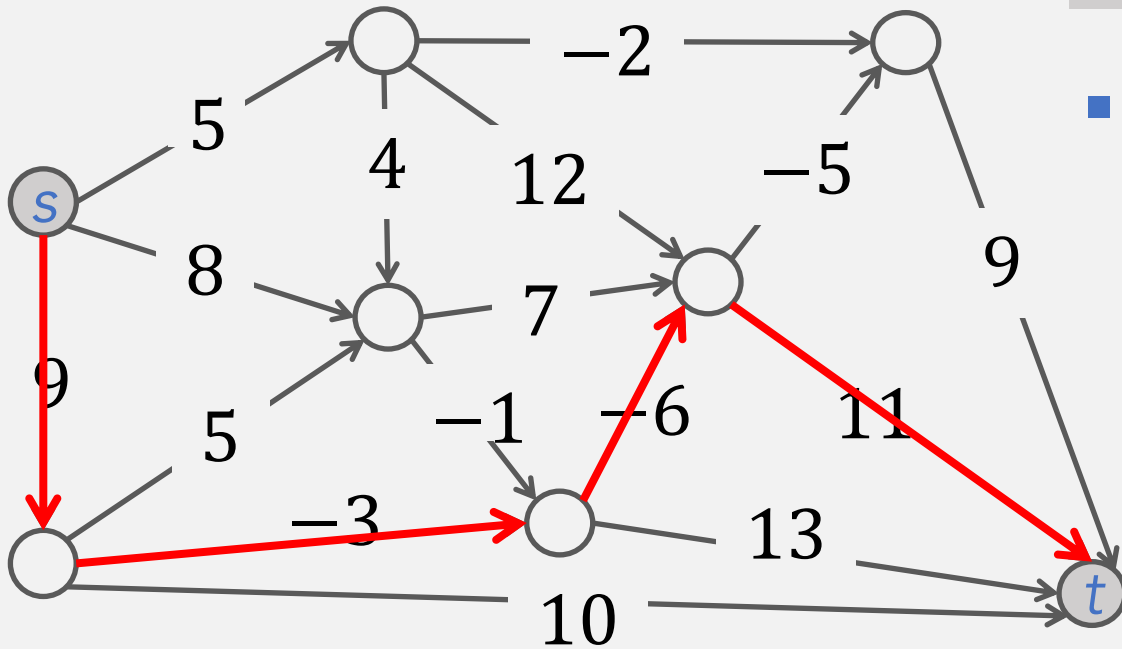
Recall: shortest path problem

- **Input.** graph G , node s and t
- **Output.** $dist(s, t)$

- Every edge has a **length** l_e (can be negative)
- Length of a path $l(P) = \sum_{e \in P} l_e$
- **Distance** $dist(u, v) = \min_{P: u \rightsquigarrow v} l(P)$

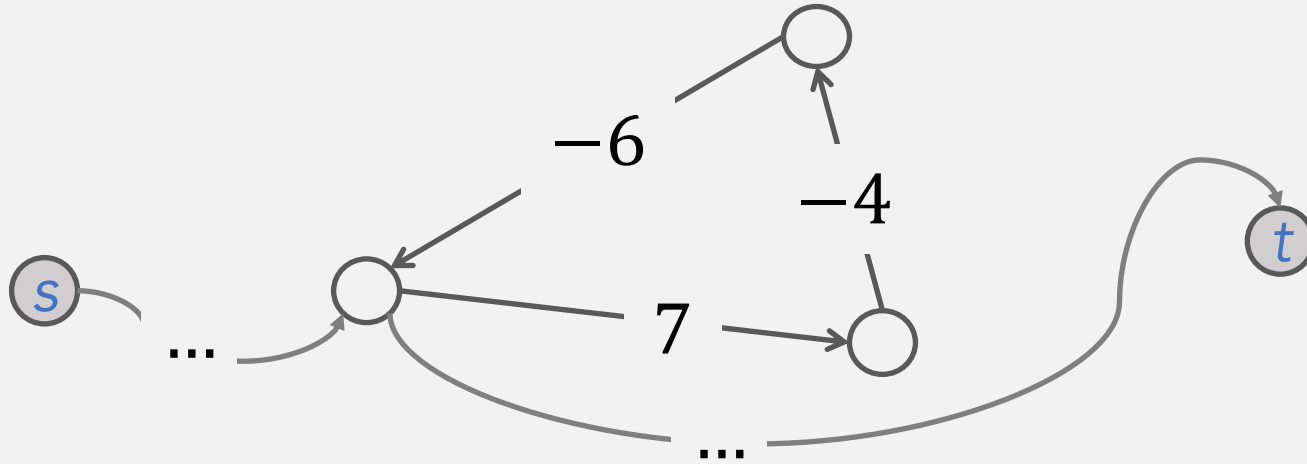
- **Special cases**

- All edge of equal length: **BFS** $O(m + n)$
- DAG: **DP** in topological order $O(m + n)$



Length of shortest path: $dist(s, t) = 9 - 3 - 6 + 11 = 11$

A technical issue: negative length cycles



■ Observation.

- If some $s \rightsquigarrow t$ path contains a **negative length cycle**, there **does not** exist a shortest $s \rightsquigarrow t$ path;
- Otherwise there exists a **simple** (i.e., no repetition node) path $\leq n - 1$ edges

■ For simplicity: assuming G has no NegativeLengthCycle

- can be detected with little overhead

DP1: develop a recursion

Def. for all $i = 0, \dots, n - 1, v \in V$

$\text{OPT}(i, v) :=$ length of shortest $v \rightsquigarrow t$ path P using $\leq i$ edges

- **Case 1.** P uses at most $i - 1$ edges
 - $\text{OPT}(i, v) := \text{OPT}(i - 1, v)$
 - **Case 2.** P uses **exactly** i edges
 - If (v, w) the first edge, then OPT uses (v, w) and then selects best $w \rightsquigarrow t$ path using $\leq i - 1$ edges
- Goal. Find $\text{OPT}(n - 1, s)$
 - Basis: $c(i, v) = 0$ or ∞ if $i = 0$
 - Recursion: how to define $\text{OPT}(i, v)$ recursively?

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

DP2: build up solutions

	V	t	s		v	n		
i	0	0	∞	∞	∞	∞	∞	∞
1	0							
	0							
	0							
	0							
i	0							
	0							
	0							
$n - 1$	0							

- Subproblems: $O(n^2)$
- Memoization data structure
 - 2-D array $M[0, \dots, n - 1, v_1, \dots, v_n]$
- Dependencies
 - Each $OPT(i, v)$ depends on subproblems on the row above
- Evaluation order
 - Row by row, each row arbitrary order

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

DP2: build up solutions

	V	t		s		v		n
i	0	0	∞	∞	∞	∞	∞	∞
1								
i								
$n - 1$								

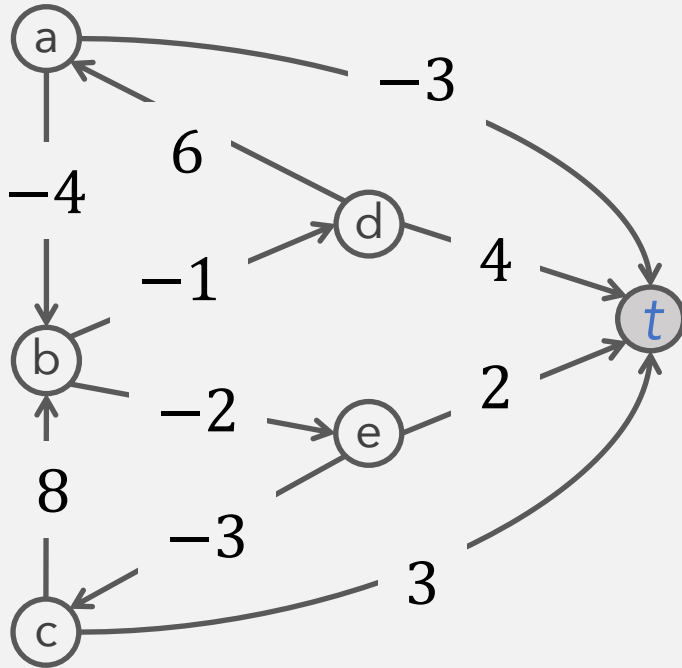
$SPLen(G, t)$

```
// M[i, v] memoize subproblem values
// M[0, t] = 0 and M[0, v] = ∞ otherwise
For i = 1, ..., n - 1 // row by row
  For v ∈ V // any order
    M[i, v] ← M[i - 1, v] // case 1
  For edge (v → w) ∈ E // case 2
    M[i, v]
    ← min{M[i, v], M[i - 1, w] + lvw}
```

This actually gives us $dist(v, t)$ for all $v \in V$

- **Analysis:** $O(n^2)$ space; $O(mn)$ time [visit all edges for each i]
- **Finding a shortest path:** maintain a "successor" for each entry

Example



	<i>v</i>	<i>t</i>	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>
<i>i</i> 0	0	∞	∞	∞	∞	∞	∞
1	0	-3	∞	3			
2	0						
3	0						
4	0						
5	0						

0	∞	∞	∞	∞	∞
0	-3	∞	3	4	2
0	-3	0	3	3	0
0	-4	-2	3	3	0
0	-6	-2	3	2	0
0	-6	-2	3	0	0

For $i = 1, \dots, n - 1$ // row by row

For $v \in V$ // any order

$M[i, v] \leftarrow M[i - 1, v]$ // case 1

For edge $(v \rightarrow w) \in E$ // case 2

$M[i, v] \leftarrow \min\{M[i, v], M[i - 1, w] + l_{vw}\}$

A simple but impactful improvement

Maintain only one array $M[v]$ = length of shortest $v \rightsquigarrow t$ path found so far.
No need to check edge (v, w) unless $M[w]$ changed in previous iteration.

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

- **Memory:** $O(m + n)$.
- **Running time:** $O(mn)$ worst case, but faster in practice.
- ➔ **Bellman-Ford algorithm:** efficient implementation
 - Application: (distance-vector) routing protocol on Internet [more to come]

Single-source shortest paths with negative weights

Year	Worst case	Discovered by
1955	$O(n^4)$	Shimbel
1956	$O(mn^2W)$	Ford
1958	$O(mn)$	Bellman, Moore
1983	$O(n^{3/4}m\log W)$	Gabow
1989	$O(mn^{1/2}\log(nW))$	Gabow-Tarjan
1993	$O(mn^{1/2}\log W)$	Goldberg
2005	$O(n^{2.38}W)$	Sankowski, Yuster-Zwisch
2016	$O(n^{10/7}\log W)$	Cohen-Madry-Sankowski-Vladu
20XX	???	You???

weights between $[-W, W]$