## EECS 336: Introduction to Algorithms <br> Lecture 5 <br> Dynamic Programming (cont) <br> Bellman-Ford

Reading: 6.4, 6.8
"guide to dynamic programming" (Canvas)

## Last time:

- Dynamic Programming (a framework)
- Integer Knapsack

Today:

- Shortest Paths.


## Suggested Approach

I. identify subproblem in english
$\mathrm{OPT}(i)=$ "optimal schedule of $\{i, \ldots, n\}$ (sorted by start time)"
II. specify subproblem recurrence
$\mathrm{OPT}(i)=\max \left(\mathrm{OPT}(i+1), v_{i}+\right.$ OPT(next(i)))
III. solve original problem (from subproblems)

Optimal Interval Schedule $=$ OPT(1)
IV. identify base case
$\operatorname{OPT}(n+1)=0$
V. write iterative DP.
(see last thurs)
VI. analyze runtime.
$O(n \log n)$
VII. (for homework) implement iterative DP.
(any language most students can read. e.g., Python)

## Shortest Paths with Negative Weights

"e.g., currency exchange: nodes are currencies, path weights are exchange rates, minimize product of path weights."

Note: to minimize product of path weights, can minimize sum of logs of path weights.

Example: $\quad r_{1} r_{2}=2^{\log _{2} r_{1} 2^{\log _{2} r_{2}}=}$ $2^{\log _{2} r_{1}+\log _{2} r_{2}}$.

Note: if $r \leq 1$ then $\log r$ is negative.
Example:


## Try Dynamic programming

$\mathrm{OPT}(v)$
$=$ shortest path from $v$ to $t$.

$$
=\min _{u \in N(v)}[\underbrace{c(v, u)}_{\text {weight }}+\operatorname{OPT}(u)] .
$$

Example:


Subproblems have cyclic dependencies!

## Imposing measure of progress

"parameterize subproblems to keep track of progress"

Lemma: if $G$ has no negative cycles, then minimum cost path is simple (i.e., does not repeat nodes); therefore, it has at most $n-1$ edges.

Proof: (contradiction)

- let $P$ be the min-length path with fewest number of edges.
- suppose (for contradiction) that $P$ is not simple.

$$
\Rightarrow P \text { repeats a vertex } v \text {. }
$$

- no negative cycle $\Rightarrow$ path from $v$ to $v$ nonnegative.
$\Rightarrow$ can "splice out" cycle and not increase length.
$\Rightarrow$ new path has fewer edges than $p$.

Idea: if simple path goes $s \leadsto v \rightarrow u \leadsto t$ then $u-t$ path has one fewer edge than $v-t$ path.

## Part I: identify subproblem in english

$\operatorname{OPT}(v, k)$
$=$ "length of shortest path from $v$ to $t$ with
at most $k$ edges."

## Part II: write recurrence

$\operatorname{OPT}(v, k)$

$$
=\min _{u \in N(v)}[c(v, u)+\operatorname{OPT}(u, k-1)]
$$

Correctness: lemma + induction.

## Part III: solve original problem Part VI: Runtime

- minimum cost path $=\operatorname{OPT}(s, n-1)$.

$$
\begin{aligned}
T(n, m) & =\overbrace{\text { "size of table" }}^{n^{2}} \times \overbrace{\text { "cost per entry" }}^{n} \\
& =O\left(n^{3}\right)
\end{aligned}
$$

## Part IV: base case

(better accounting: $T(n, m)=O\left(n^{2}+n m\right)=$ $O(n m))$

- for all $k: \operatorname{OPT}(t, k)=0$.
- for all $v \neq t: \operatorname{OPT}(v, 0)=\infty$.


## Part V: iterative DP

Algorithm: Bellman-Ford

1. initialize
for all $k: \mathrm{OPT}[t, k]=0$.
for all $v \neq t: \operatorname{OPT}[v, 0]=\infty$.
2. for $k=1$ up to $n-1$,
for all $v$
$\operatorname{OPT}[v, k]=\min _{u \in N(v)} \operatorname{OPT}(u, k-$ 1).
3. return $\operatorname{OPT}[s, n-1]$.

Example:


|  | 0 | 1 | 2 | 3 |
| :---: | :---: | :---: | :---: | :---: |
| $s$ | $\infty$ | $\infty$ | 3 | 2 |
| $a$ | $\infty$ | 2 | 1 | 1 |
| $b$ | $\infty$ | -2 | -2 | -2 |
| $t$ | 0 | 0 | 0 | 0 |

