

Dynamic Programming

Design and Analysis of Algorithms
Andrei Bulatov

Knapsack

The Knapsack Problem

Instance:

A set of n objects, each of which has a positive integer value v_i and a positive integer weight w_i . A weight limit W .

Objective:

Select objects so that their total weight does not exceed W , and they have maximal total value

Idea

A simple question: Should we include the last object into selection?

Let $\text{OPT}(n, W)$ denote the maximal value of a selection of objects out of $\{1, \dots, n\}$ such that the total weight of the selection doesn't exceed W

More general, $\text{OPT}(i, U)$ denote the maximal value of a selection of objects out of $\{1, \dots, i\}$ such that the total weight of the selection doesn't exceed U

Then

$$\text{OPT}(n, W) = \max\{ \text{OPT}(n - 1, W), \text{OPT}(n - 1, W - w_n) + v_n \}$$

Algorithm (First Try)

```
Knapsack(n,W)
set v1:=Knapsack(n-1,w)
set v2:=Knapsack(n-1,w- wn )
output max(v1,v2+vn)
```

Is it good enough?

Example

Let the values be 1,3,4,2, the weights 1,1,3,2, and $W = 5$

Recursion tree

Another Idea: Memoization

Let us store values $\text{OPT}(i,U)$ as we find them

We need to store (and compute) at most $n \times W$ numbers

We'll do it in a regular way:

Instead of recursion, we will compute those values starting from smaller ones, and fill up a table

Algorithm (Second Try)

```
Knapsack(n,W)
array M[0..n,0..W]
set M[0,w]:=0 for each w=0,1,...,W
for i=1 to n do
    for w=0 to W do
        set M[i,w]:= max{M[i-1,w],M[i-1,w-wi]+vi}
    endfor
endfor
```

Example

Example

Let the values be 1,3,4,2, the weights 1,1,3,2, and $W = 5$

$i \backslash w$	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	1	1	1	1	1
2	0	3	4	4	4	4
3	0	3	4	4	7	8
4	0	3	4	5	7	8

$$M[i,w] = \max\{ M[i-1, w], M[i-1, w - w_i] + v_i \}$$

Shortest Path

Suppose that every arc e of a digraph G has length
(or cost, or weight, or ...) $\text{len}(e)$

But now we allow negative lengths (weights)

Then we can naturally define the length of a directed path in G ,
and the distance between any two nodes

The s-t-Shortest Path Problem

Instance:

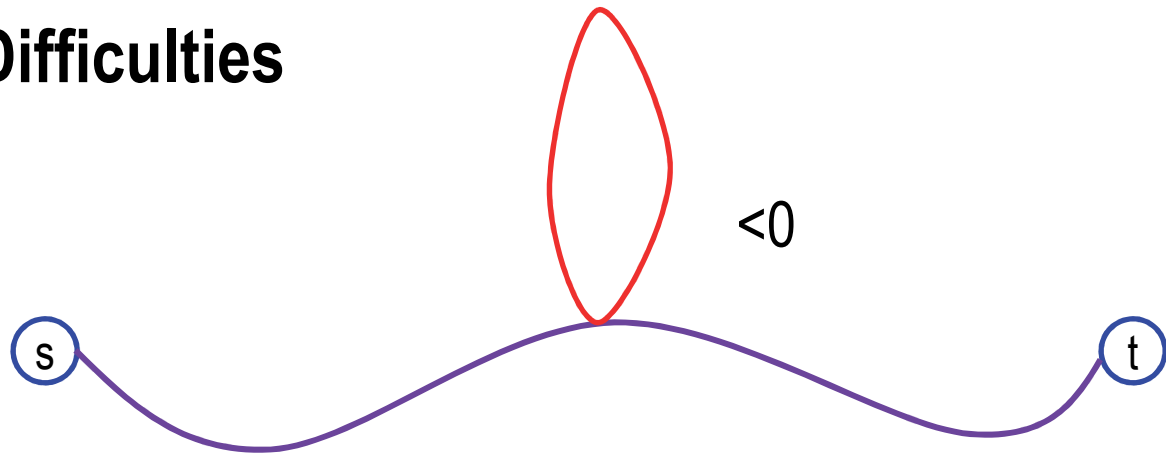
Digraph G with lengths of arcs, and nodes s, t

Objective:

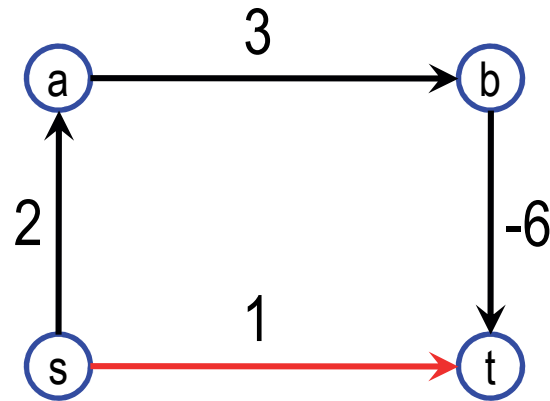
Find a shortest path between s and t

Shortest Path: Difficulties

Negative Cycles.



Greediness fails



Adding constant weight to all arcs fails

Shortest Path: Observations

Assumption

There are no negative cycles

Lemma

If graph G has no negative cycles, then there is a shortest path from s to t that is simple (i.e. does not repeat nodes), and hence has at most $n - 1$ arcs

Proof

If a shortest path P from s to t repeats a node v , then it also include a cycle C starting and ending at v .

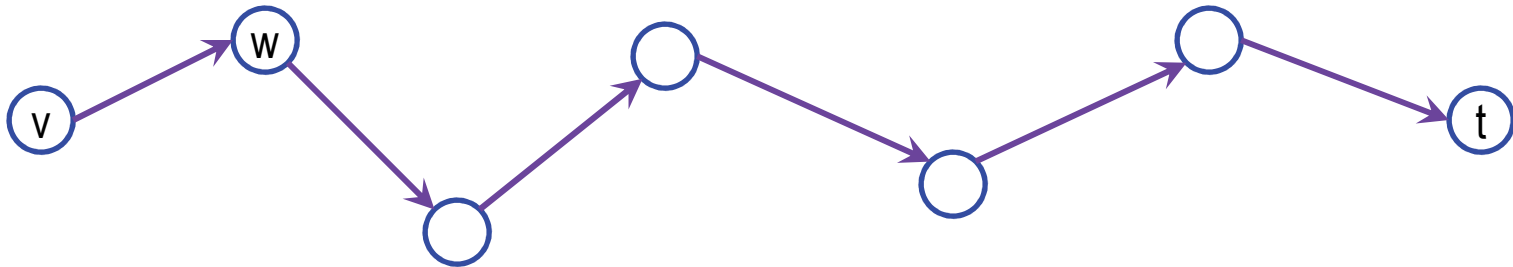
The weight of the cycle is non-negative, therefore removing the cycle makes the path shorter (no longer).

QED

Shortest Path: Dynamic Programming

We will be looking for a shortest path with increasing number of arcs

Let $OPT(i,v)$ denote the minimum weight of a path from v to t using at most i arcs



Shortest $v - t$ path can use $i - 1$ arcs. Then $OPT(i,v) = OPT(i - 1,v)$
 Or it can use i arcs and the first arc is vw . Then

$$OPT(i,v) = \text{len}(vw) + OPT(i - 1,w)$$

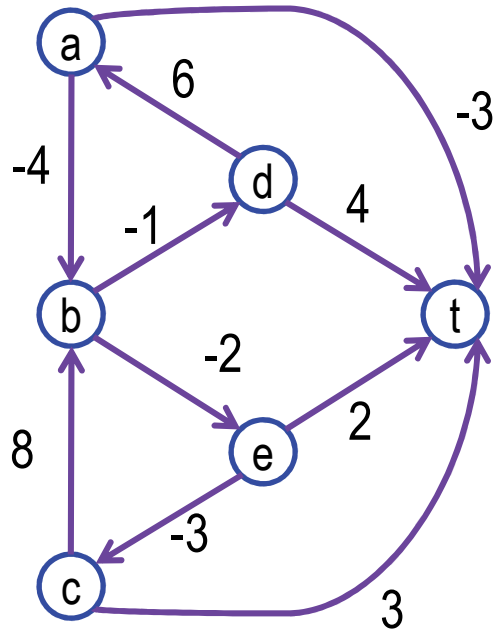
$$OPT(i,v) = \min \{ OPT(i - 1,v), \min_{w \in V} \{ OPT(i - 1,w) + \text{len}(vw) \} \}$$

Shortest Path: Algorithm

Shortest-Path(G, s, t)

```
set  $n := |V|$  /*number of nodes in G
array  $M[0..n-1, V]$ 
set  $M[0, t] := 0$  and  $M[0, v] := \infty$  for each  $v \in V - \{t\}$ 
for  $i = 1$  to  $n - 1$  do
    for  $v \in V$  do
        set  $M[i, v] := \min\{M[i-1, v], \min_{w \in V} \{M[i-1, w] + \text{len}(vw)\}\}$ 
    endfor
endfor
return  $M[n-1, s]$ 
```

Example



	t	a	b	c	d	e
0	0	∞	∞	∞	∞	∞
1	0	-3	∞	3	4	2
2	0	-3	0	3	3	0
3	0	-4	-2	3	3	0
4	0	-6	-2	3	2	0
5	0	-6	-2	3	0	0

$$M[i,v] = \min\{ M[i-1, v], \min_{w \in V} \{ M[i-1, w] + \text{len}(vw) \} \}$$

Shortest Path: Soundness and Running Time

Theorem

The ShortestPath algorithm correctly computes the minimum cost of an s-t path in any graph that has no negative cycles, and runs in $O(n^3)$ time

Proof.

Soundness follows by induction from the recurrent relation for the optimal value.

DIY.

Running time:

We fill up a table with n^2 entries. Each of them requires $O(n)$ time

Shortest Path: Soundness and Running Time

Theorem

The ShortestPath algorithm can be implemented in $O(mn)$ time

A big improvement for sparse graphs

Proof.

Consider the computation of the array entry $M[i,v]$:

$$M[i,v] = \min\{ M[i-1, v], \min_{w \in V} \{ M[i-1, w] + \text{len}(vw) \} \}$$

We need only compute the minimum over all nodes w for which v has an edge to w

Let n_v denote the number of such edges

Shortest Path: Running Time Improvements

It takes $O(n_v)$ to compute the array entry $M[i,v]$.

It needs to be computed for every node v and for each i , $1 \leq i \leq n$.

Thus the bound for running time is

$$O\left(n \sum_{v \in V} n_v\right) = O(nm)$$

Indeed, n_v is the outdegree of v , and we have the result by the Handshaking Lemma.

QED

Shortest Path: Space Improvements

The straightforward implementation requires storing a table with n^2 entries

It can be reduced to $O(n)$

Instead of recording $M[i,v]$ for each i , we use and update a single value $M[v]$ for each node v , the length of the shortest path from v to t found so far

Thus we use the following recurrent relation:

$$M[v] = \min\{ M[v], \min_{w \in V} \{ M[w] + \text{len}(vw) \} \}$$

Shortest Path: Space Improvements (cntd)

Lemma

Throughout the algorithm $M[v]$ is the length of some path from v to t , and after i rounds of updates the value $M[v]$ is no larger than the length of the shortest from v to t using at most i edges

Shortest Path: Finding Shortest Path

In the standard version we only need to keep record on how the optimum is achieved

Consider the space saving version.

For each node v store the first node on its path to the destination t

Denote it by $\text{first}(v)$

Update it every time $M[v]$ is updated

Let P be the **pointer graph** $P = (V, \{(v, \text{first}(v)) : v \in V\})$

Shortest Path: Finding Shortest Path

Lemma

If the pointer graph P contains a cycle C , then this cycle must have negative cost.

Proof

If $w = \text{first}(v)$ at any time, then $M[v] \geq M[w] + \text{len}(vw)$

Let v_1, v_2, \dots, v_k be the nodes along the cycle C , and (v_k, v_1) the last arc to be added

Consider the values right before this arc is added

We have $M[v_i] \geq M[v_{i+1}] + \text{len}(v_i v_{i+1})$ for $i = 1, \dots, k-1$ and

$$M[v_k] > M[v_1] + \text{len}(v_k v_1)$$

Adding up all the inequalities we get

$$0 > \sum_{i=1}^{k-1} \text{len}(v_i v_{i+1}) + \text{len}(v_k v_1)$$

Shortest Path: Finding Shortest Path (cntd)

Lemma

Suppose G has no negative cycles, and let P be the pointer graph after termination of the algorithm. For each node v , the path in P from v to t is a shortest v - t path in G .

Proof

Observe that P is a tree.

Since the algorithm terminates we have $M[v] = M[w] + \text{len}(vw)$, where $w = \text{first}(v)$.

As $M[t] = 0$, the length of the path traced out by the pointer graph is exactly $M[v]$, which is the shortest path distance.

QED

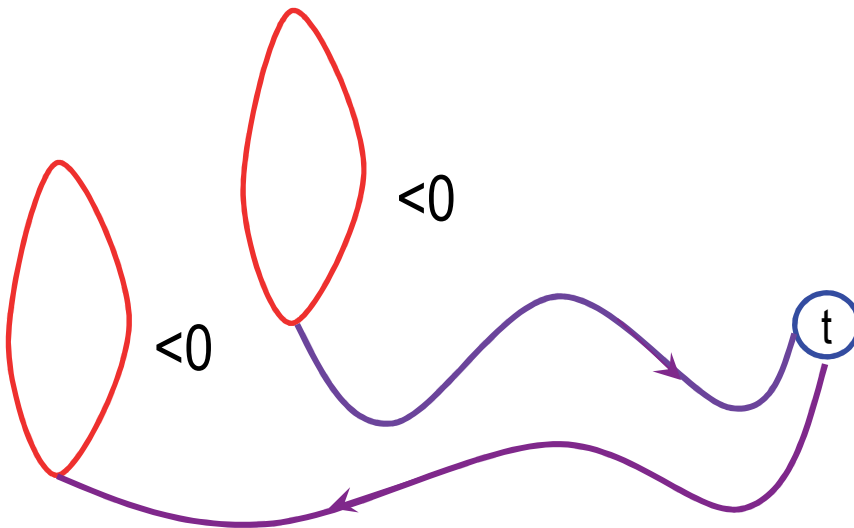
Shortest Path: Finding Negative Cycles

Two questions:

- how to decide if there is a negative cycle?
- how to find one?

Lemma

It suffices to find negative cycles C such that t can be reached from C

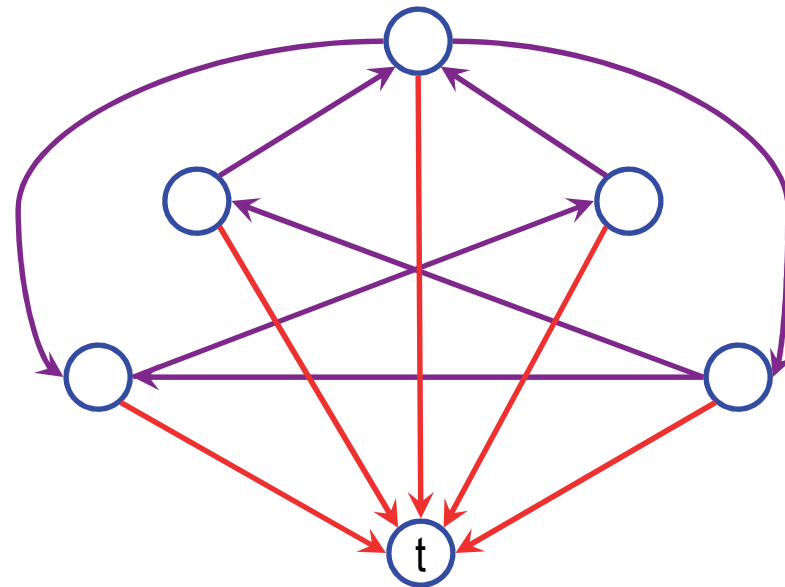


Shortest Path: Finding Negative Cycles

Proof

Let G be a graph

The augmented graph, $A(G)$, is obtained by adding a new node and connecting every node in G with the new node



As is easily seen, G contains

a negative cycle if and only if $A(G)$ contains a negative cycle C such that t is reachable from C

QED

Shortest Path: Finding Negative Cycles (cntd)

Extend $\text{OPT}(i,v)$ to $i \geq n$

If the graph G does not contain negative cycles then

$$\text{OPT}(i,v) = \text{OPT}(n-1,v) \text{ for all nodes } v \text{ and all } i \geq n$$

Indeed, it follows from the observation that every shortest path contains at most $n-1$ arcs.

Lemma

There is no negative cycle with a path to t if and only if

$$\text{OPT}(n,v) = \text{OPT}(n-1,v)$$

Proof

If there is no negative cycle, then $\text{OPT}(n,v) = \text{OPT}(n-1,v)$ for all nodes v by the observation above

Shortest Path: Finding Negative Cycles (cntd)

Proof (cntd)

Suppose $\text{OPT}(n,v) = \text{OPT}(n-1,v)$ for all nodes v .

Therefore

$$\begin{aligned}\text{OPT}(n,v) &= \min\{ \text{OPT}(n-1,v), \min_{w \in V} \{ \text{OPT}(n-1,w) + \text{len}(vw) \} \} \\ &= \min\{ \text{OPT}(n,v), \min_{w \in V} \{ \text{OPT}(n,w) + \text{len}(vw) \} \} \\ &= \text{OPT}(n+1,v) \\ &= \dots\end{aligned}$$

However, if a negative cycle from which t is reachable exists, then

$$\lim_{i \rightarrow \infty} \text{OPT}(i,v) = -\infty$$

QED

Shortest Path: Finding Negative Cycles (cntd)

Let v be a node such that $\text{OPT}(n,v) \neq \text{OPT}(n-1,v)$.

A path P from v to t of weight $\text{OPT}(n,v)$ must use exactly n arcs

Any simple path can have at most $n-1$ arcs, therefore P contains a cycle C

Lemma

If G has n nodes and $\text{OPT}(n,v) \neq \text{OPT}(n-1,v)$, then a path P of weight $\text{OPT}(n,v)$ contains a cycle C , and C is negative.

Proof

Every path from v to t using less than n arcs has greater weight.

Let w be a node that occurs in P more than once.

Let C be the cycle between the two occurrences of w

Deleting C we get a shorter path of greater weight, thus C is negative