Mathematics for Decision Making: An Introduction

Lecture 9

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Theorem (Efficiency of Ford's Algorithm)

If **c** is integer-valued and G = (V, A) has no negative-cost directed circuit, then Ford's Algorithm terminates after **at most** Cn^2 steps, where n = |V| and

 $C = m(\mathbf{c}) = 1 + 2||\mathbf{c}||_{\infty}$ with $||\mathbf{c}||_{\infty} = \max\{|c_{(a,b)}|: (a,b) \in A\}.$

 This is a typical statement of an efficiency result: We establish an upper bound for (some abstraction of) the running time, with a simple formula. Note we do not predict the precise running time; the algorithm could be much faster than that.

Proof.

- The first correction step, leading to a finite potential value y_v of a vertex, gives y_v that is at most $||\mathbf{c}||_{\infty}(n-1)$, because it is the cost of a simple directed path (with at most n-1 arcs!) from the root.
- In every later correction step, y_v gets reduced by at least 1, because **c** is integral
- In the end, the potential value y_v is the cost of a least-cost path; this may be negative, but we certainly have $y_v \ge -||\mathbf{c}||_{\infty}(n-1)$.
- Thus at most $1 + 2 \cdot ||\mathbf{c}||_{\infty} (n-1) \leq Cn$ steps per vertex
- Hence, at most *Cn*² correction steps in total.

Efficiency of Ford's Algorithm, II

- The theorem establishes that Ford's Algorithm is a **pseudo-polynomial algorithm**, i.e., its running time is bounded above by a polynomial expression in the "dimensions" (such as *n_k*) and the absolute values of its input data.
- Because the bound is monotonous in *C* and *n*, it is convenient to interpret this bound as an upper bound on the running time of the **worst case** that can happen among all problems (G, \mathbf{c}) with $|V| \le n$ and and $m(\mathbf{c}) \le C$.
- In a homework exercise, you show that there is a one-parameter family $\{(G_k = (V_k, A_k), \mathbf{c}^k) : k \in \mathbf{N}\}$ of networks with $n_k = |V_k| = 2k + 4$ vertices and $C_k = 2^k$, such that Ford's Algorithm (with a specific, clever, evil way of choosing **which** incorrect arc should be corrected) takes more than 2^k steps.
- This shows that the upper bound is "not too far off" from the worst case

Better efficiency classes:

- We are not happy with pseudo-polynomial algorithms, because for the same graph *G*, the running time might grow quickly if we just use "large numbers" (it might grow **exponentially in the number of digits** of the data $c_{(a,b)}$).
- Better are **polynomial algorithms**, where the running time is allowed to grow polynomially in the "dimensions" (such as n = |V| and m = |A|), but only **polynomially in the number of digits** of the data (such as $c_{a,b}$)
- Even better are strongly polynomial algorithms, where the worst-case running, time (#steps) is allowed to depend only on the dimensions, not on the data

Improving Ford's Algorithm: Ford–Bellman [1958]

- In a homework exercise, we saw that there are examples, in which a **particular** order of correcting arcs leads to very bad performance (many iterations).
- Let's try to find an order that is better.
- Let's rewrite the body of the while loop like this:
 - Choose an arc (v, w).

2 If (v, w) is incorrect, then correct it, updating predecessor information.

We call this **verifying** arc (v, w).

- We denote by S = ((v₁, w₁), (v₂, w₂),..., (v_k, w_k)) a sequence of arcs that we verify during Ford's Algorithm.
- Important observation:

Lemma

In Ford's Algorithm, after verifying the sequence S of arcs, for all directed paths P from r to v that are **embedded** in S, i.e.,

the arcs of P appear as a subsequence of S (i.e., in the right order, but not necessarily consecutively)

we have $y_v \leq c(P)$.

Improving Ford's Algorithm: Ford–Bellman [1958]

Proof.

Let $P = (v_0, a_0, v_1, a_1, v_2, ..., a_K, v_k)$ with $v_0 = r$ and $v_k = v$ be a directed path that is embedded in *S*.

• After verifying *a*₀ in some iteration *q*₀, we have

$$y_{\nu_1}^{(q_0)} \leq y_{\nu_0}^{(q_0-1)} + c_{\nu_0,\nu_1} = c_{\nu_0,\nu_1}.$$

• Then, after verifying a_1 in iteration $q_1 > q_0$, we have

$$y_{v_2}^{(q_1)} \le y_{v_1}^{(q_1-1)} + c_{v_1,v_2}$$
 (verification)
 $\le y_{v_1}^{(q_0)} + c_{v_1,v_2}$ (y_{v_1} possibly decreased between q_0 and $q_1 - 1$)
 $\le c_{v_0,v_1} + c_{v_1,v_2}$. (per above)

• and so on: induction yields $y_v \leq c(P)$.

- Now let us design a sequence *S* of arcs such that **every possible minimum-cost path** is embedded in *S*
- Minimum-cost paths are simple directed paths, so they contain at most n-1 arcs (where n = |V|)
- Simple construction: Let S_i be any ordering of the arcs A. Then the sequence

$$S = (S_1, \ldots, S_{n-1})$$

has the desired property. We say that we make n-1 passes through the graph.

• We call this refined algorithm the Ford-Bellman algorithm.

Ford–Bellman algorithm

Input: A digraph G = (V, A) with arc costs, starting node *r* **Output:** If *G* has a negative cycle, output "negative cycle!"; otherwise output a predecessor vector **p**, which encodes minimum-cost paths from *r* to all other nodes.

- Initialize y and p
- Set i := 0
- Solution While i < n and **y** is not a feasible potential:
 - Set *i* := *i* + 1
 - For $(v, w) \in A$ (in arbitrary order):

If (v, w) is incorrect, then correct it, updating predecessor information.

If i = n, return "negative cycle!"; otherwise, return **p**.

Improving Ford's Algorithm: Ford–Bellman [1958]

Theorem (Correctness and efficiency of Ford–Bellman)

The Ford–Bellman algorithm is correct. It terminates after at most $m \cdot n$ arc verifications.

Proof.

Correctness follows from the above lemma:

- If there is no negative cycle, after the arc-verification sequence S, for every minimum-cost path P_v we have y_v ≤ c(P_v) because P_v is embedded in S. Thus the *while* loop terminates with i < n.
- If there is a negative cycle, we know there does not exist a feasible potential, so the *while loop* terminates because of *i* = *n*.

The bound on the number of verifications is obvious.

- Thus it is a strongly combinatorial algorithm.
- In the general case, no algorithm with a better running time bound is known.

The case of topologically sortable graphs

- Suppose that we can order the vertices of the directed graph *G* = (*V*,*A*) "from left to right", so that all arcs go from left to right.
- In other words, suppose there is an ordering v_1, \ldots, v_n of V such that for any arc $(v_i, v_j) \in A$ we have i < j.
- Such an ordering is called a topological sort of G.

Observation:

- All directed paths in *G* are embedded in the arc-correction sequence
 S = (*L*₁,..., *L_n*) where *L_i* is an arbitrary ordering of the arcs leaving vertex *v_i*
- Therefore, Ford's Algorithm has the correct answer after running this arc-correction sequence *S*.

Where do topologically sortable graphs come from?

- In some applications, the graphs have a natural topological sort because the vertices are **layered**, for instance by "time", and there are only arcs that go from "now" to "later".
- This is related to the idea of **dynamic programming** (with respect to time or other "increasing" parameters)
- Which (other) directed graphs have a topological sort? Complete answer on the next slide.

Characterization of Topological Sortability

Lemma (Topological Sortability Lemma)

A directed graph has a topological sort if and only if it is acyclic (has no directed circuit)

Proof of the Topological Sortability Lemma.

- If there is a topological sort v_1, \ldots, v_n , there clearly is no directed circuit.
- For the converse, we first show that there is a suitable choice for v₁, i.e., a vertex with no predecessor, i.e., no incoming arc.
 - Suppose, to the contrary, that every vertex has a predecessor.
 - Let $w_1 \in V$ be arbitrary; pick a predecessor of w_1 and call it w_2 .
 - Pick a predecessor of w₂ and call it w₃.
 - This produces an infinite sequence $w_1, w_2, \dots \in V$.
 - However, *V* is finite, so there is some i < j with $w_i = w_j$.
 - Thus there is a directed circuit $(w_j, w_{j-1}, \dots, w_{i+1}, w_i)$ in *G*, a contradiction.
- Continue inductively on a graph G₁ where we have removed v₁ (and the arcs originating from v₁).

This proof suggests an efficient algorithm that constructs a topological sort or detects a directed circuit (homework).

Bellman's Algorithm for the Acyclic Case

Bellman's Algorithm ("Dynamic Programming")

Input: A digraph G = (V, A) with arc costs, starting node *r* **Output:** If *G* has a cycle, output "cycle!"; otherwise output a predecessor vector **p**, which encodes minimum-cost paths from *r* to all other nodes.

- Sind a topological sort v_1, \ldots, v_n of *G*; if there is none, return "cycle!".
- Initialize y and p

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    For i = 1 to n:
    Scan vertex v<sub>i</sub>, i.e., do for all arcs (v<sub>i</sub>, w) ∈ A:
    If (v<sub>i</sub>, w) is incorrect, then correct it, updating predecessor information.
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Return p.

This is still a label-correcting algorithm; but it's a one-pass algorithm.

Theorem (Correctness and Efficiency of Bellman's Algorithm)

Bellman's algorithm is correct. It terminates after m = |A| arc verification steps.

Another important special case is to allow directed cycles, but to require that all arc costs are **nonnegative**.

- Again, we use an arc-correction sequence that corresponds to the idea of scanning the vertices in some ordering v₁, v₂,..., v_n (i.e., first verifying all arcs leaving v₁, then all arcs leaving v₂, etc.)
- This time, however, we do not determine this ordering a priori
- Rather, when v₁, v₂,..., v_i have been determined and scanned, we choose v_{i+1} as an unscanned vertex v with minimum potential value y_v (at that time).

The resulting algorithm is called Dijkstra's Algorithm.

Dijkstra's Algorithm

Input: A digraph G = (V, A) with nonnegative arc costs, starting node *r* **Output:** A predecessor vector **p**, encoding minimum-cost paths from *r* to all nodes.

- Initialize y, p.
- Set S := V.

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    While S ≠ Ø:

Choose v ∈ S with y<sub>v</sub> minimum.

Set S := S \ {v}.

Scan vertex v, i.e., do for all arcs (v, w) ∈ A:

If (v, w) is incorrect, then correct it, updating predecessor information.
```

Dijkstra's Algorithm: Correctness

- We use the notation v_1, v_2, \ldots, v_n for the ordering of the nodes
- We denote by $\mathbf{y}^{(i)}$ the value of \mathbf{y} at the point when v_i is chosen to be scanned.

Lemma (Monotonicity of potentials of scanned nodes)

For all i < k we have $y_{v_i}^{(i)} \leq y_{v_k}^{(k)}$.

Proof.

- Suppose the contrary, i.e., there exist i < k with $y_{v_i}^{(i)} > y_{v_k}^{(k)}$.
- Fix such a *i* and choose *k* minimal with this property, i.e., *v_k* is the earliest-chosen vertex after *v_i* that, at the time of its scanning, had a smaller potential than the vertex *v_i* at the time of its scanning.
- But by the minimal choice in the algorithm, we have $y_{v_i}^{(i)} \le y_{v_k}^{(i)}$.
- So y_{v_k} must have been lowered while scanning some vertex v_j with i < j < k.
- This arc correction made $y_{v_k}^{(k)} = y_{v_k}^{(j+1)} = y_{v_j}^{(j)} + c_{v_j, v_k}$.
- Because $c_{v_j,v_k} \ge 0$, we have $y_{v_j}^{(j)} \le y_{v_k}^{(k)} < y_{v_i}^{(i)}$.
- This is a contradiction to the definition of k.

Theorem

Dijkstra's Algorithm is correct.

Proof.

We prove that, after all vertices have been scanned, we have a feasible potential y^{n+1} :

- Suppose not, i.e., for some $(v_i, v_k) \in A$, we have $y_{v_i}^{(n+1)} + c_{v_i, v_k} < y_{v_k}^{(n+1)}$.
- But directly after scanning vertex v_i , we certainly did have $y_{v_i}^{(i+1)} + c_{v_i,v_k} \ge y_{v_k}^{(i+1)}$.
- Since we never increase the potentials, y_{v_i} must have been lowered afterwards! Say, it was lowered the last time when scanning vertex v_j (with i < j).
- Thus $y_{v_i}^{(i+1)} > y_{v_i}^{(n+1)} = y_{v_i}^{(j+1)} = y_{v_j}^{(j)} + c_{v_j,v_i} \ge y_{v_j}^{(j)}$
- On the other hand, by the Lemma, because v_j was scanned after v_i , we have $y_{v_j}^{(j)} \ge y_{v_i}^{(i)}$, a contradiction $(y_{v_i}^{(i+1)} > y_{v_i}^{(i)})$.

Theorem (Efficiency of Dijkstra's Algorithm)

Dijkstra's Algorithm terminates after m = |A| arc verification steps.

- Let's try out Dijkstra's Algorithm in practice; we expect that the running time essentially only depends, linearly, on the number of arcs.
- We try on examples with the same number of arcs, but different numbers of vertices.
- Result: There is a great dependence on the number of vertices, and we are not happy with the running time for large, sparse graphs (many vertices, few arcs)
- Where is the running time spent? Our coarse abstraction of running time (number of arc verification steps) does not give the answer.
- To find this out in the practical program, it is strongly recommended to find this out by measuring time, rather than thinking or guessing.
- Every modern, reasonable programming system has a facility for measuring how much running time is spent in parts of the program; this is called a (time) profiler.
- In the case of C, the GCC toolchain (compiler/linker option -pg) and the gprof tool provide a (sampling) time profiler.