

Complex-Network Modelling and Inference

Lecture 21: Path-problem algorithms

Matthew Roughan

`<matthew.roughan@adelaide.edu.au>`

https://roughan.info/notes/Network_Modelling/

School of Mathematical Sciences,
University of Adelaide

March 7, 2024

Bellman-Ford algorithm

- We want to solve

$$A^* = \left(A \otimes A^* \right) \oplus I$$

- One approach is successive iteration

$$A^{<k+1>} = \left(A \otimes A^{<k>} \right) \oplus I$$

Example: most reliable path

- Imagine links a subject to “problems”
- A message transits link e with IID probability $r_e \in [0, 1]$, which we call the link *reliability*
- The probability of successfully negotiating a path is

$$r_p = \prod_{e \in p} r_e$$

- So we want to solve

$$A_{ij}^* = \max_{p \in P_{ij}} \prod_{e \in p} r_e,$$

- The natural semiring to use is the Max-times or Viterbi Semiring

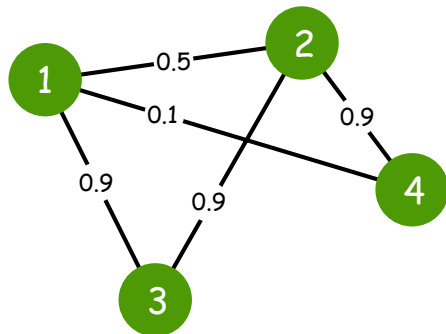
$$(\mathcal{S}, \oplus, \otimes, \bar{0}, \bar{1}) = ([0, 1], \max, \times, 0, 1)$$

Most reliable path example

$$(S, \oplus, \otimes, \bar{0}, \bar{1}) = ([0, 1], \max, \times, 0, 1)$$

$$A = \begin{pmatrix} 0.0 & 0.5 & 0.9 & 0.1 \\ 0.5 & 0.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 0.0 & 0.0 \\ 0.1 & 0.9 & 0.0 & 0.0 \end{pmatrix}$$

$$I = \begin{pmatrix} 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 1.0 \end{pmatrix}$$



Most reliable path example

We are calculating

$$A^{<1>} = \left(A \otimes A^{<0>} \right) \oplus I$$

Note $A^{<0>} = I$, so first calculate

$$\begin{aligned} A \otimes I &= \begin{pmatrix} 0.0 & 0.5 & 0.9 & 0.1 \\ 0.5 & 0.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 0.0 & 0.0 \\ 0.1 & 0.9 & 0.0 & 0.0 \end{pmatrix} \otimes \begin{pmatrix} 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 1.0 \end{pmatrix} \\ &= \begin{pmatrix} 0.0 & 0.5 & 0.9 & 0.1 \\ 0.5 & 0.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 0.0 & 0.0 \\ 0.1 & 0.9 & 0.0 & 0.0 \end{pmatrix} \\ &= A \end{aligned}$$

Most reliable path example

We are calculating

$$A^{<1>} = (A \otimes A^{<0>}) \oplus I$$

Now $A \otimes A^{<0>} = A$, so now

$$\begin{aligned} A \oplus I &= \begin{pmatrix} 0.0 & 0.5 & 0.9 & 0.1 \\ 0.5 & 0.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 0.0 & 0.0 \\ 0.1 & 0.9 & 0.0 & 0.0 \end{pmatrix} \oplus \begin{pmatrix} 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 1.0 \end{pmatrix} \\ &= \begin{pmatrix} 1.0 & 0.5 & 0.9 & 0.1 \\ 0.5 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.0 \\ 0.1 & 0.9 & 0.0 & 1.0 \end{pmatrix} \end{aligned}$$

This tells us the *most reliable path with 1 hop or less*

Most reliable path example

Second iteration

$$A^{<2>} = (A \otimes A^{<1>}) \oplus I$$

First calculate

$$\begin{aligned} A \otimes A^{<1>} &= \begin{pmatrix} 0.0 & 0.5 & 0.9 & 0.1 \\ 0.5 & 0.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 0.0 & 0.0 \\ 0.1 & 0.9 & 0.0 & 0.0 \end{pmatrix} \otimes \begin{pmatrix} 1.0 & 0.5 & 0.9 & 0.1 \\ 0.5 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.0 \\ 0.1 & 0.9 & 0.0 & 1.0 \end{pmatrix} \\ &= \begin{pmatrix} 0.81 & 0.81 & 0.9 & 0.45 \\ 0.81 & 0.81 & 0.9 & 0.9 \\ 0.9 & 0.9 & 0.81 & 0.81 \\ 0.45 & 0.9 & 0.81 & 0.81 \end{pmatrix} \end{aligned}$$

Most reliable path example

Second iteration

$$A^{<2>} = (A \otimes A^{<1>}) \oplus I$$

Now add the identity

$$\begin{aligned} (A \otimes A^{<1>}) \oplus I &= \begin{pmatrix} 0.81 & 0.81 & 0.9 & 0.45 \\ 0.81 & 0.81 & 0.9 & 0.9 \\ 0.9 & 0.9 & 0.81 & 0.81 \\ 0.45 & 0.9 & 0.81 & 0.81 \end{pmatrix} \oplus \begin{pmatrix} 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 1.0 \end{pmatrix} \\ &= \begin{pmatrix} 1.0 & 0.81 & 0.9 & 0.45 \\ 0.81 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.81 \\ 0.45 & 0.9 & 0.81 & 1.0 \end{pmatrix} \end{aligned}$$

This tells us the *most reliable path with 2 hops or less*

Notice that none of the reliabilities went down. They can't decrease, because we have more options when we allow longer paths. As values are monotonic and bounded, we know they must converge.

Most reliable path example

Third iteration

$$\begin{aligned} A^{<3>} &= (A \otimes A^{<2>}) \oplus I \\ &= \begin{pmatrix} 1.0 & 0.81 & 0.9 & 0.729 \\ 0.81 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.81 \\ 0.729 & 0.9 & 0.81 & 1.0 \end{pmatrix} \end{aligned}$$

This tells us the *most reliable path with 3 hops or less*

Most reliable path example

Forth iteration

$$\begin{aligned} A^{<4>} &= (A \otimes A^{<3>}) \oplus I \\ &= \begin{pmatrix} 1.0 & 0.81 & 0.9 & 0.729 \\ 0.81 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.81 \\ 0.729 & 0.9 & 0.81 & 1.0 \end{pmatrix} \end{aligned}$$

This tells us the *most reliable path with 4 hops or less*

Note that it is the same as the 3-hop version.

Most reliable path example

Fifth iteration

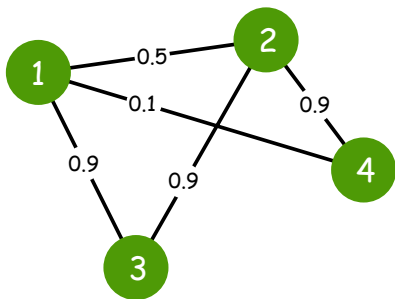
$$\begin{aligned} A^{<5>} &= (A \otimes A^{<4>}) \oplus I \\ &= \begin{pmatrix} 1.0 & 0.81 & 0.9 & 0.729 \\ 0.81 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.81 \\ 0.729 & 0.9 & 0.81 & 1.0 \end{pmatrix} \end{aligned}$$

This tells us the *most reliable path with 5 hops or less*

Note that it is the same as the 4-hop version.

Most reliable path example

$$A^{<1>} = A$$
$$A^{<2>} = \begin{pmatrix} 1.0 & 0.81 & 0.9 & 0.45 \\ 0.81 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.81 \\ 0.45 & 0.9 & 0.81 & 1.0 \end{pmatrix}$$
$$A^{<3>} = \begin{pmatrix} 1.0 & 0.81 & 0.9 & 0.729 \\ 0.81 & 1.0 & 0.9 & 0.9 \\ 0.9 & 0.9 & 1.0 & 0.81 \\ 0.729 & 0.9 & 0.81 & 1.0 \end{pmatrix}$$



- $A^* = A^{<3>}$
- In a real algorithm we also need to keep track of the predecessor nodes
- In a real problem we hope that it converges before we reach $A^{<n-1>}$

Section 1

Path-problem algorithm properties

Fixed-point iteration

- We want to solve

$$A^* = \left(A \otimes A^* \right) \oplus I$$

- One approach is successive iteration

$$\begin{aligned} A^{<0>} &= I \\ A^{<k+1>} &= \left(A \otimes A^{<k>} \right) \oplus I \end{aligned}$$

- When does
 - ▶ the iteration converge to a *fixed-point*?
 - ▶ the equation have a unique result?
 - ▶ if the equation has more than one result, then which one would the iteration find?

Fixed-point iteration

Lemma

If we take

$$\begin{aligned}A^{<0>} &= I \\A^{<k+1>} &= \left(A \otimes A^{<k>} \right) \oplus I\end{aligned}$$

then $A^{<k+1>} = A^{(k+1)} = I \oplus A \oplus \dots \oplus A^k$.

Proof.

By induction. The $k = 0$ case is true by definition. Assume it is true for k , then

$$\begin{aligned}A^{<k+1>} &= \left(A \otimes A^{<k>} \right) \oplus I \\&= \left(A \otimes (I \oplus A \oplus \dots \oplus A^k) \right) \oplus I \\&= I \oplus A \oplus \dots \oplus A^{k+1}\end{aligned}$$

Note that distributivity and commutativity of \oplus is required. □

q -Stability

Take an arbitrary semiring $(S, \oplus, \otimes, \bar{0}, \bar{1})$ where we define powers using these operators, *i.e.*,

$$a^0 = \bar{1}, \quad \text{and} \quad a^k = a \otimes a^{k-1}$$

and we define

$$\begin{aligned} a^{(q)} &= a^0 \oplus a^1 \oplus a^2 \oplus \dots \oplus a^q \\ a^* &= a^0 \oplus a^1 \oplus a^2 \oplus \dots \end{aligned}$$

Definition (q -stability)

If there exists a q such that $a^{(q)} = a^{(q+1)}$ then we say a is q -stable. We say the semiring S is q -stable if every $a \in S$ is q -stable.

q -Stability

Lemma

If a is q -stable, then $a^* = a^{(q)}$.

Proof.

If a is q -stable, then $a^{(q)} = a^{(q+1)}$, and by induction we get $a^{(q)} = a^{(q+t)}$ for all $t \geq 0$. Take $t \rightarrow \infty$ and we get the result. \square

q -Stability

Lemma

If $\bar{1}$ is an annihilator for \oplus ,

$$a \oplus \bar{1} = \bar{1} \oplus a = \bar{1}, \quad \forall a \in S,$$

then S is 0-stable.

Examples:

- Boolean: $a \text{ OR } T = T \text{ OR } a = T$
- Min-plus: $\min(a, 0) = \min(0, a) = 0$
- Viterbi: $\max(a, 1) = \max(1, a) = 1$
- lots of others ...

q -Stability

Lemma

If $\bar{1}$ is an annihilator for \oplus , i.e.,

$$a \oplus \bar{1} = \bar{1} \oplus a = \bar{1}, \quad \forall a \in S,$$

then S is 0-stable.

Proof.

$$\begin{aligned} a^{(1)} &= \bar{1} \oplus a \\ &= \bar{1} \\ &= a^{(0)} \end{aligned}$$



q -Stability of matrix semirings

Lemma

If S is 0-stable, then $M_n(S)$ is $(n - 1)$ -stable, that is

$$A^* = A^{(n-1)} = I \oplus A \oplus \dots \oplus A^{n-1}$$

- Intuition: annihilation means we can ignore paths with loops, and so the longest possible path has $n - 1$ hops, so we don't need any higher powers.
- Or, extend the idea that “shortest paths are built from shortest paths”

$$\begin{aligned}(a \otimes x \otimes b) \oplus (a \otimes b) &= a \otimes (\bar{1} \oplus x) \otimes b, \text{ by distributivity} \\ &= a \otimes \bar{1} \otimes b, \text{ by annihilation} \\ &= a \otimes b, \text{ as } \bar{1} \text{ is multiplicative identity}\end{aligned}$$

- Primary consequence is that the fixed-point iteration will always converge in at most $n - 1$ steps.

Uniqueness

Theorem

If A is q -stable, then A^ exists (and hence solves the equations)*

$$X = AX \oplus I$$

and A^ is the “least” solution.*

Proof.

Existence is shown by stability (above), and that it is a solution to the equations in the last lecture.

The remaining issue is the “least.” To answer this, we need to define an ordering. □

Binary Relations

Definition

A *relation* from A to B is a binary operator $x\mathcal{R}y$ for $x \in A$ and $y \in B$ that either “holds”, or does not hold. We say it is a relation on A if $A = B$.

Examples

- “is the mother of”
- “is a friend of”
- “is a multiple of”
- =
- \leq
- \subset
- ...

Orderings

An ordering is a special type of relation on A (that we'll denote \preceq) with the properties

- *reflexivity*: $x \preceq x$ for all $x \in A$
- *anti-symmetry*: if $x \preceq y$ and $y \preceq x$ then $x = y$
- *transitivity*: if $x \preceq y$ and $y \preceq z$ then $x \preceq z$

Further, two elements of A are said to be *comparable* if either $x \preceq y$ or $y \preceq x$, and *incomparable* otherwise. If all elements of A are comparable then \preceq defines a *total order*, otherwise it is a *partial order*.

Examples

- \leq defines total order on \mathbb{R}
- \subset defines a partial order for the set of subsets of A

Orderings

We will write ordering relation here as \preceq or \succeq , where we interpret

$$a \preceq b$$

as meaning b is “at least as good” as a , or the preferred path to a in path problem.

- Note that this can be far from the standard meaning
 - ▶ e.g., in the shortest-path problem, better means shorter so

$$2 \succeq 3$$

- ▶ e.g., in path properties algebra, where we deal with subsets, it becomes the subset relation

Orderings

An *idempotent*, *commutative*, and *associative* operator \oplus defines ordering as follows:

$$x \preceq y, \quad \text{if and only if } x \oplus y = y$$

Generally, in *path algebras* like this, we want \oplus to be selective, and hence, idempotent, and hence the \oplus operator defines an ordering, with $\bar{0}$ as the least element, *i.e.*,

$$\bar{0} \leq x, \quad \text{for all } x \in S$$

One implication is that

$$y \preceq x \oplus y$$

$$x \preceq x \oplus y$$

i.e., if we can choose between two paths, the result should be at least as good as either of the choices.

Isotonic operators

Definition (Isotonic)

We say that an operator $a \bullet b$ is *isotonic* if

$$b \preceq c \Rightarrow a \bullet b \preceq a \bullet c$$

for all $a \in S$.

- Intuition: think of \otimes as extending a path, then this says for an isotonic \otimes if we extend two alternate paths with the same link, the preferred ordering of the extended paths doesn't change.
- When \oplus is idempotent (and commutative and associative), then it and \otimes are isotonic.

Isotonic operators

Lemma

For semiring $(S, \oplus, \otimes, \bar{0}, \bar{1})$, if \oplus is also idempotent it defines an ordering \preceq , i.e.,

$$x \preceq y, \quad \text{if and only if } x \oplus y = y$$

and both \oplus and \otimes are isotone with respect to this ordering.

Proof.

The relation \preceq is an ordering because it is

- reflexive: $x \oplus x = x$ through idempotence
- anti-symmetric: $x \oplus y = y \oplus x$ through commutativity
- transitive: by associativity, i.e., , if $x \oplus y = y$ and $y \oplus z = z$ then

$$x \oplus z = x \oplus (y \oplus z) = (x \oplus y) \oplus z = y \oplus z = z$$



Isotonic operators

Proof.

The operator \oplus can be seen to be isotonic from the definition of $a \oplus b \preceq a \oplus c$, i.e.,

$$\begin{aligned}(a \oplus b) \oplus (a \oplus c) &= (a \oplus a) \oplus (b \oplus c), && \text{associativity and commutativity} \\ &= a \oplus (b \oplus c), && \text{idempotence} \\ &= a \oplus c, && \text{because } b \preceq c\end{aligned}$$

Hence $(a \oplus b) \preceq (a \oplus c)$ follows from $b \preceq c$.

A similar result follows for \otimes using distributivity. □

Least solutions

Lemma

For semiring $(S, \oplus, \otimes, \bar{0}, \bar{1})$, if \oplus and \otimes are isotone and x and y are stable, then

$$x \preceq y \Rightarrow x^* \preceq y^*.$$

Proof.

The proof follows by noting that if x and y are stable they can be expanded as a finite sequence of \oplus and \otimes operations with themselves, each of which preserves the original order. □

Hence, if \oplus is idempotent, everything else works!

Solutions

Lemma

For semiring $(S, \oplus, \otimes, \bar{0}, \bar{1})$, if \oplus is idempotent, then the equation

$$y = (a \otimes y) \oplus \bar{1}$$

has a solution $y = a^*$ and this is the *least* possible solution.

Proof.

We have already shown that idempotence implies stability, and hence that $y = a^*$ is a solution to the above equation. Now presume that y_0 is any solution to the equation, we can repeatedly substitute it into the equation itself to get

$$y_0 = (a \otimes [(a \otimes y_0) \oplus \bar{1}]) \oplus \bar{1} = a^2 \otimes y_0 \oplus (a \oplus \bar{1})$$



Solutions

Proof.

Repeating results in

$$y_0 = a^k y_0 \oplus (\bar{1} \oplus a \oplus a^2 \oplus \dots \oplus a^{k-1})$$

If $k > q$ we have $a^{(q)} = a^*$ and hence

$$y_0 = a^k y_0 \oplus a^*$$

and hence (because $x \oplus y \succeq y$)

$$y_0 \succeq a^*$$

So a^* is the least solution. □

This is a special case of the more general equation $y = (a \otimes y) \oplus b$, which has solution $y = a^* b$.

Transitive Closure

A^* is a special case of a *transitive closure*

- Take some *relation* between members of the set
 - ▶ as expressed by links in the graph
- Extend the relation to a consistent relation over all pairs
 - ▶ as expressed by paths between pairs

Section 2

Issues

So we are finished?

- We have a VERY general approach
 - ▶ define a semiring with a idempotent \oplus
 - ▶ extend it to adjacency matrices
 - ▶ and we know we can solve path problems
- We can even solve a particular column of A^* at a time by solving

$$y = Ay \oplus e_k$$

- ▶ so we can start to apply techniques for doing fast linear algebra
 - ▶ e.g., Gauss-Jordan, and better techniques
 - ▶ algorithms, e.g., Dijkstra, can be seen as solution techniques from linear algebra
- Does it cover everything?

Is everything a semiring

- Many simple, obvious binary operators are not semigroup operators (and hence can't be used in semiring).
 - ▶ e.g., the average $a \bullet b = (a + b)/2$ is not associative

$$(1 \bullet 2) \bullet 3 = \frac{9}{4} \neq \frac{7}{4} = 1 \bullet (2 \bullet 3)$$

- Some viable operations don't distribute over others
 - ▶ e.g., multi-objective optimisations are often performed by

$$\min \text{objective}_1 \text{ subject to } \text{objective}_2 < C$$

can't translate this into a semiring (as far as I know) except for special cases

- But, there are a vast set of possibilities, especially when we start to compose semirings, e.g., take lexicographic products ...

Further reading I



B. Carré, *Graphs and networks*, vol. 135, 1979.