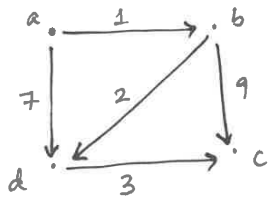


* Admin: See announcement about quizzes.

* Last time: Weighted adjacency matrices



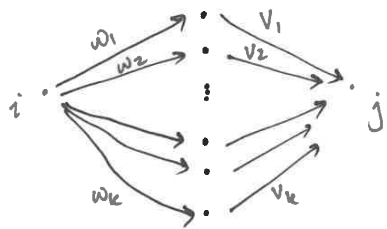
$$W = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & \infty & 7 \\ \infty & 0 & 9 & 2 \\ \infty & \infty & 0 & \infty \\ \infty & \infty & 3 & 0 \end{bmatrix} \end{matrix}$$

0 on diagonal } our conventions
 ∞ when no edge } for our application.

We want: least cost of travelling from i to j .

W tells us the least costs of travelling from i to j in 0 or 1 steps

→ From this, compute for ≤ 2 steps?



possible costs:

$$\begin{matrix} w_1 + v_1 \\ w_2 + v_2 \\ \vdots \\ w_k + v_k \end{matrix} \left. \vphantom{\begin{matrix} w_1 + v_1 \\ w_2 + v_2 \\ \vdots \\ w_k + v_k \end{matrix}} \right\} \text{depending on your route.}$$

$$\text{we want: } \min \{w_1 + v_1, w_2 + v_2, \dots, w_k + v_k\}$$

Note: Some "edges" can actually be weight=0 connections, which could include staying in the same spot, hence total length ≤ 2 .

We use the "min-plus" dot product.

addition + is replaced by min
 multiplication \times is replaced by +

as follows:

Let w and v be vectors. (same length)

$$w \odot v := \min \{w_1 + v_1, w_2 + v_2, \dots, w_k + v_k\}$$

denotes min-plus vector/matrix product.

* How to deal with " ∞ "?

Suppose c is a number (i.e. $c \neq \infty$)

$$\begin{matrix} c + \infty := \infty + c := \infty \\ \infty + \infty := \infty \end{matrix}$$

$$\begin{matrix} \min \{c, \infty\} := c \\ \min \{\infty, \infty\} := \infty \end{matrix}$$

REMEMBER:

" ∞ " is a placeholder symbol to denote a number bigger than anything else we encounter.

↳ * Matrix min-plus product

Let A & B be weighted adjacency matrices. Typically we'll take $A = B = W$ (our wtd adj matrix)

$$A \odot B = \begin{matrix} i \\ \text{---} \\ \text{---} \\ \text{---} \end{matrix} \odot \begin{matrix} j \\ \text{---} \\ \text{---} \\ \text{---} \end{matrix}$$

$$\begin{aligned} (i, j)^{\text{th}} \text{ entry} &= \text{min-plus dot product of } R_i \text{ \& } C_j \\ &= R_i \odot C_j \end{aligned}$$

So if $A=B=W$

$W \odot W = W^{\odot 2}$, and more generally, take

$$W^{\odot n} = n^{\text{th}} \text{ min-plus matrix power}$$

* Example

$$\begin{bmatrix} 0 & 1 & \infty & 7 \\ \infty & 0 & 9 & 2 \\ \infty & \infty & 0 & \infty \\ \infty & \infty & 3 & 0 \end{bmatrix} \odot \begin{bmatrix} 0 & 1 & \infty & 7 \\ \infty & 0 & 9 & 2 \\ \infty & \infty & 0 & \infty \\ \infty & \infty & 3 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 3 \\ & \end{bmatrix}$$

$\min \{0+0, 1+\infty, \infty+\infty, 7+\infty\}$
 $= \min \{0, \infty, \infty, \infty\}$
 $= 0$

$\min \{0+7, 1+2, \infty+\infty, 7+0\}$
 $= \min \{7, 3, \infty, 7\} = 3$

Q: (Aside): How to compute max costs?
→ See worksheet

* Consequence: The $(i,j)^{\text{th}}$ entry of $W \odot W$ is the min. cost of going from i to j in ≤ 2 steps

* Theorem: The $(i,j)^{\text{th}}$ entry of $W^{\odot n}$ is the min. cost of going from i to j in $\leq n$ steps.

③

* Remarks:

- For the purposes of computing min-costs, we will restrict ourselves to weights ≥ 0 .
- Usually we don't have self-loops of positive weight; even if we do, we'll take the cost of going from i to i as 0.

* Recall: If G has n vertices then the shortest path from i to j (if it exists) has length $\leq n$.

In fact if $i \neq j$, it has length $\leq (n-1)$.

Suppose $m \geq n$. Then:

$W^{\odot m}$ records min-costs of paths of length $\leq m$

* Since we have non-negative weights, shorter paths must have lesser (or equal) weights.

* Theorem: Let G have n vertices.

Suppose $m \geq n$. Then,

$$W^{\odot m} = W^{\odot (n-1)}$$

Explanation: If $i=j$ then the min cost from $i \rightarrow j$ is zero, already achieved in the length=0 path. If $i \neq j$, the min-cost from $i \rightarrow j$ is already achieved by a path of length $\leq n-1$. Beyond $(n-1)$, you don't add any new info.

④

* Partial order relations.

Def: Let S be a set; R a relation. on S .
We say that R is a total order if:

for any two $a, b \in S$: either
 $(a, b) \in R$ or
 $(b, a) \in R$ or ~~both~~ } and exactly one
 $a = b$ } of these holds.

R is transitive

Another way to define this:

R is a total order if

① R is reflexive

② R is anti-symmetric

③ R is transitive.

④ For each $a, b \in \mathbb{R}^S$, either $(a, b) \in R$ or $(b, a) \in R$.