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Question 1: Exploring the Kleene Star

a)

$$\begin{aligned}
a^* &= \bigoplus_{n=0}^{\infty} a^n = a^0 \oplus \bigoplus_{n=1}^{\infty} a^n = 1 \oplus \bigoplus_{n=1}^{\infty} a^n = \\
&= 1 \oplus \bigoplus_{n=1}^{\infty} a \otimes a^{(n-1)} = 1 \oplus a \otimes \bigoplus_{n=1}^{\infty} a^{(n-1)} = \\
&= 1 \oplus a \otimes \bigoplus_{n=0}^{\infty} a^{(n)} = 1 \oplus a \otimes a^*
\end{aligned} \tag{1}$$

b)

$$\begin{aligned}
W_{\log} &= \langle \mathbb{R} \cup \{-\infty, \infty\}, \oplus_{\log}, +, -\infty, 0 \rangle \\
1 \oplus (a \otimes a^*) &= 1 \oplus (a + a^*) = \log(e^0 + e^{a+a^*}) \\
a^* &= \log(1 + e^{a+a^*}) \\
e^{a^*} &= 1 + e^{a+a^*} \\
e^{a^*} (1 - e^a) &= 1 \\
e^{a^*} &= \frac{1}{1 - e^a} \\
a^* &= \log\left(\frac{1}{1 - e^a}\right) \\
\text{Where } e^a &\leq 1
\end{aligned} \tag{2}$$

c)

$$\begin{aligned}
a^* &= 1 \oplus a \otimes a^* = 1 \oplus \langle xx^*, xy^* + yx^* \rangle \\
x^* &= xx^* + 1 \rightarrow x^* = \frac{1}{1-x} \\
y^* &= xy^* + yx^* \rightarrow y^* = \frac{y}{(1-x)^2} \\
a^* &= \left\langle \frac{1}{1-x}, \frac{y}{(1-x)^2} \right\rangle \tag{3}
\end{aligned}$$

d) 1. Proof that $\langle 2^{\Sigma^*}, \cup, \{\} \rangle$ is a commutative monoid.1.1) Distributivity of \oplus

$$\begin{aligned}
A \cup (B \cup C) &= A \cup \{x \mid x \in B \vee x \in C\} = \\
&\{x \mid x \in A \vee x \in B \vee x \in C\} = \\
&\{x \mid x \in A \vee x \in B\} \cup C = (A \cup B) \cup C
\end{aligned}$$

$$1.2) \mathbf{0} \oplus A = A \quad \{\} \cup A = \{x \mid x \in A\} = A$$

$$\begin{aligned}
1.3) \text{ Commutativity of } \oplus \quad A \cup B &= \{x \mid x \in A \vee x \in B\} = \\
&= \{x \mid x \in B \vee x \in A\} = B \cup A
\end{aligned}$$

2. Proof that $\langle 2^{\Sigma^*}, \otimes, \{\varepsilon\} \rangle$ is a monoid.

2.1) Distributivity of \otimes

$$\begin{aligned}
(A \otimes B) \otimes C &= \\
\{a \circ b \mid a \in A, b \in B\} \otimes C &= \\
\{a \cdot b \circ c \mid a \in A, b \in B, C \in C\} &= \\
= A \otimes \{b \circ c \mid b \in B, C \in C\} &= A \otimes (B \otimes C)
\end{aligned}$$

$$2.2) \mathbf{1} \otimes A = A$$

$$\{\varepsilon\} \otimes A = \{\varepsilon \circ a \mid a \in A\} = \{a \mid a \in A\} = A$$

3. Proof that \otimes distributes over \oplus

$$\begin{aligned}
A \otimes (B \oplus C) &= A \otimes \{x \mid x \in B \vee x \in C\} = \\
&= \{a \circ x \mid a \in A, x \in B \vee x \in C\} = \\
&= \{a \circ x \mid a \in A, x \in B\} \vee \{a \circ x \mid a \in A, x \in C\} \\
&= (A \otimes B) \oplus (A \otimes C)
\end{aligned}$$

4. Proof that $\mathbf{0}$ is the annihilator.

$$\{\} \otimes A = \{\}$$

A kleene star A^* for the language semiring is defined as:

$$A^* = \bigcup_{n=0}^{\infty} A^n = A^0 \cup A^1 \cup A^2 \dots \quad (4)$$

As we did in lecture 7, we can find A^n inductively. First:

$$\begin{aligned}
A^0 &= \mathbf{1} = \{\varepsilon\} \\
A^1 &= A \\
A^n &= \{s \circ a \mid s \in A^{n-1}, a \in A\}
\end{aligned} \quad (5)$$

Using what we proved in a):

$$A^* = \{\varepsilon\} \cup (A \otimes A^*) = (A \otimes A^*) \quad (6)$$

Question 2: Asterating the matrix

a) We have to proof that $\mathbf{1} \oplus a = \mathbf{1}$.

For the tropical semiring:

$$\min(0, a) = 0, \forall a \in R_{\geq 0}.$$

For the arctic semiring:

$$\max(0, a) = 0, \forall a \in R_{\leq 0}.$$

b) The adjacency matrix M of graph G contains the weight of the paths of length 1 (weight of each transition between nodes). To start our proof by induction, the base case is $M^1 = M$. We assume that M^{n-1} encodes the sum of all paths of length exactly $n-1$. Because $M^n = M^{n-1} \otimes M$ and by the definition of matrix multiplication, we have that:

$$(M^n)_{i,k} = \bigoplus_{j=0}^{N-1} (M^{n-1})_{i,j} \otimes M_{j,k} \quad (7)$$

This is the semiring sum of all the paths between i and k of length $n-1$ and length 1, i.e, it is the semiring-sum of all paths from the node i to node k in graph G of length exactly n .

c)

d) If the transition matrix is over a 0-closed semiring, the shortest path weight depends only on the paths of at most $N-1$.

The semiring-sum of paths between node i and node k of length at most $N-1$ is:

$$\bigoplus_{j=0}^{N-1} (M^n)_{i,k}.$$

Combining these two results, we derive that:

$$M^* = \bigoplus_{n=0}^{\infty} (M^n) = \bigoplus_{n=0}^{N-1} (M^n).$$

e) For the algorithm we will use two equations:

$$M^* = \bigoplus_{n=0}^{N-1} M^n \text{ and } M^n = M^{n-1} \otimes M$$

We initialize matrix M^n and its kleene star M^* as identity matrices.

Then we run a loop N times such that at each iteration:

- Overwrite M^n with $M^n \otimes M$.
- Overwrite M^* with $M^* \oplus M^n$.

The resulting M^* will be our kleene star.

The runtime of the matrix multiplication is cubic with N . Because the algorithm computes matrix multiplication N times, the runtime of the algorithm will be $O(N^4)$.

f)

$$a \oplus a = (\mathbf{1} \otimes a) \oplus (\mathbf{1} \otimes a) = (\mathbf{1} \oplus \mathbf{1}) \otimes a = \mathbf{1} \otimes a = a \quad (8)$$

We use first the identity property and second the distributivity property. Third, because its a 0-closed semiring and $\mathbf{1}$ is an element of the semiring, $\mathbf{1} \oplus \mathbf{1} = \mathbf{1}$. Last, we use the identity property again.

g)

h)

i) First, we see that for any x : $\left\| \frac{x}{\|x\|_2} \right\|_2 = 1$. Second, if x is a constant, then $\|Ax\| = \|A\|\|x\|$. For this, when we calculate the supremum of $\frac{\|Ax\|_2}{\|x\|_2}$, we can rewrite it as $\sup_{x \neq 0} \left\| A \frac{x}{\|x\|_2} \right\|_2 = \sup_{\|x\|_2=1} \|Ax\|_2$.

Next, we will use the Single Value Decomposition of the matrix $A = U\Sigma V^T$ where U and V are orthogonal and Σ is the diagonal matrix with $\sigma_1, \dots, \sigma_n$ values. Then:

$$\sup_{\|x\|_2=1} \|Ax\|_2 = \sup_{\|x\|_2=1} \|U\Sigma V^T x\|_2 = \sup_{\|a\|_2=1} \|\Sigma a\|_2 = \sup \sqrt{\sum_{i=1}^n \sigma_i^2 a_i^2} = \sigma_{\max}. \quad (9)$$

j)

$$\begin{aligned} \left\| A^* - \sum_{n=0}^K A^n \right\|_2 &= \left\| \sum_{n=0}^{\infty} A^n - \sum_{n=0}^K A^n \right\|_2 = \left\| \sum_{n=K+1}^{\infty} A^n \right\|_2 \\ &\leq \sum_{n=K+1}^{\infty} \|A^n\|_2 \leq \sum_{n=K+1}^{\infty} \|A\|_2^n = \sum_{n=K+1}^{\infty} \sigma_{\max}(A)^n. \end{aligned}$$

We can rewrite the previous result in terms of geometric series:

$$\sum_{n=K+1}^{\infty} \sigma_{\max}(A)^n = \sum_{n=0}^{\infty} \sigma_{\max}(A)^n - \sum_{n=0}^K \sigma_{\max}(A)^n$$

This subtraction will go to 0 when K goes to ∞ if $|\sigma_{\max}| < 1$ because then we can calculate the limit as:

$$\lim_{K \rightarrow \infty} \sum_{n=0}^{\infty} \sigma_{\max}(A)^n - \sum_{n=0}^K \sigma_{\max}(A)^n = \lim_{K \rightarrow \infty} \frac{1}{1 - \sigma_{\max}(A)} - \frac{1 - \sigma_{\max}(A)^{K+1}}{1 - \sigma_{\max}(A)} = \lim_{K \rightarrow \infty} \frac{\sigma_{\max}(A)^{K+1}}{1 - \sigma_{\max}(A)} = 0$$

k) The big-O bound of the truncation error is $O(\sigma_{\max}(A)^K)$.

Question 3: Implementation of a Neural Transducer

Link to colab notebook:

https://colab.research.google.com/drive/1ymghMoIvWvVLeUs63ruBn1PTGI_Tl6E9?usp=sharing