

Eidgenössische Technische Hochschule Zürich Swiss Federal Institute of Technology Zurich

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Question 1:

- a) Prove that the expectation semiring satisfies the semiring axioms:
 - $(\mathbb{R} \times \mathbb{R}, \oplus, \mathbf{0})$ must be a commutative monoid with identity element $\mathbf{0}$:

$$(\langle x, y \rangle \oplus \langle x', y' \rangle) \oplus \langle x'', y'' \rangle = \langle x + x', y + y' \rangle \oplus \langle x'', y'' \rangle \tag{1}$$

$$= \langle x + x' + x'', y + y' + y'' \rangle \tag{2}$$

$$= \langle x, y \rangle \oplus \langle x' + x'', y' + y'' \rangle \tag{3}$$

$$= \langle x, y \rangle \oplus (\langle x', y' \rangle \oplus \langle x'', y'' \rangle) \tag{4}$$

$$\mathbf{0} + \langle x, y \rangle = \langle 0, 0 \rangle \oplus \langle x, y \rangle \tag{5}$$

$$= \langle 0 + x, 0 + y \rangle \tag{6}$$

$$=\langle x,y\rangle \tag{7}$$

$$= \langle x + 0, y + 0 \rangle \tag{8}$$

$$= \langle x, y \rangle + \mathbf{0} \tag{9}$$

$$\langle x, y \rangle + \langle x', y' \rangle = \langle x + x', y + y' \rangle \tag{10}$$

$$= \langle x' + x, y' + y \rangle \tag{11}$$

$$= \langle x', y' \rangle + \langle x, y \rangle \tag{12}$$

• $(\mathbb{R} \times \mathbb{R}, \otimes, \mathbf{1})$ must be a monoid with identity element 1:

$$(\langle x, y \rangle \otimes \langle x', y' \rangle) \otimes \langle x'', y'' \rangle = \langle x \cdot x', x \cdot y' + y \cdot x' \rangle \otimes \langle x'', y'' \rangle \tag{13}$$

$$= \langle x \cdot x' \cdot x'', x \cdot x' \cdot y'' + (x \cdot y' + y \cdot x') \cdot x'' \rangle \tag{14}$$

$$= \langle x \cdot x' \cdot x'', x \cdot x' \cdot y'' + x \cdot y' \cdot x'' + y \cdot x' \cdot x'' \rangle \quad (15)$$

$$= \langle x, y \rangle \otimes \langle x' \cdot x'', x' \cdot y'' + y' \cdot x'' \rangle \tag{16}$$

$$= \langle x, y \rangle \otimes (\langle x', y' \rangle \otimes \langle x'', y'' \rangle) \tag{17}$$

$$\mathbf{1} \otimes \langle x, y \rangle = \langle 1, 0 \rangle \otimes \langle x, y \rangle \tag{18}$$

$$= \langle 1 \cdot x, 1 \cdot y \rangle \tag{19}$$

$$=\langle x, y \rangle \tag{20}$$

$$= \langle x \cdot 1, y \cdot 1 \rangle \tag{21}$$

$$= \langle x, y \rangle \otimes \mathbf{1} \tag{22}$$

• Multiplication left and right distributes over addition:

 $\langle x, y \rangle \otimes (\langle x', y' \rangle \oplus \langle x'', y'' \rangle) = \langle x, y \rangle \otimes \langle x' + x'', y' + y'' \rangle$

$$= \langle x \cdot x' + x \cdot x'', x \cdot y' + x \cdot y'' + y \cdot x' + y \cdot x'' \rangle$$
(24)
$$= \langle x \cdot x', x \cdot y' + y \cdot x' \rangle \oplus \langle x \cdot x'', x \cdot y'' + y \cdot x'' \rangle$$
(25)
$$= (\langle x, y \rangle \otimes \langle x', y' \rangle) \oplus (\langle x, y \rangle \otimes \langle x'', y'' \rangle)$$
(26)
$$(\langle x, y \rangle \oplus \langle x', y' \rangle) \otimes \langle x'', y'' \rangle = \langle x + x', y + y' \rangle \otimes \langle x'', y'' \rangle$$
(27)
$$= \langle x \cdot x'' + x' \cdot x'', x \cdot y'' + x' \cdot y'' + y \cdot x'' + y' \cdot x'' \rangle$$
(28)
$$= \langle x \cdot x'', x \cdot y'' + y \cdot x'' \rangle \oplus \langle x' \cdot x'', x' \cdot y'' + y' \cdot x'' \rangle$$
(29)

• Multiplication by **0** annihilates $\mathbb{R} \times \mathbb{R}$:

$$\mathbf{0} \otimes \langle x, y \rangle = \langle 0, 0 \rangle \otimes \langle x, y \rangle \tag{31}$$

$$= \langle 0 \cdot x, 0 \cdot y \rangle \tag{32}$$

$$= \langle 0, 0 \rangle \tag{33}$$

$$= \mathbf{0} \tag{34}$$

$$= \langle 0, 0 \rangle \tag{35}$$

$$= \langle x \cdot 0, y \cdot 0 \rangle \tag{36}$$

$$= \langle x, y \rangle \otimes \langle 0, 0 \rangle \tag{37}$$

$$= \langle x, y \rangle \otimes \mathbf{0} \tag{38}$$

 $= (\langle x, y \rangle \otimes \langle x'', y'' \rangle) \oplus (\langle x', y' \rangle \otimes \langle x'', y'' \rangle)$

(23)

(30)

b) Our initial graph looks like Fig. 1.

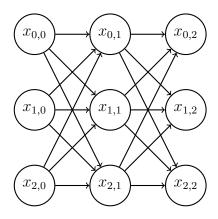


Figure 1: The initial graph

Where the columns represent the words in \mathbf{w} and the rows represent different tags. We use the algorithm from the script:

Algorithm 1: Forward pass

- 1 $\beta(\mathbf{w}, t_0) = 1$ 2 for $n = 1 \rightarrow N$ do
- 3 | $\beta(\mathbf{w}, t_n) = \sum_{t_{n-1} \in \mathcal{T}} \exp(\operatorname{score}_{\theta}(\langle \langle t_{n-1}, t_n \rangle \rangle, \boldsymbol{w})) \otimes \beta(\mathbf{w}, t_{n-1})$
- 4 end

When we now lift the CRF into the expectation semiring, the forward propagation algorithm changes to:

Algorithm 2: Forward pass

$$\beta(\mathbf{w}, t_0) = \langle 1, 0 \rangle$$

2 for
$$n=1 \rightarrow N$$
 do

3
$$\beta(\mathbf{w}, t_n) = \bigoplus_{t_{n-1} \in \mathcal{T}} \langle w, -w \log w \rangle \otimes \beta(\mathbf{w}, t_{n-1})$$

4 end

Where $w = \exp(\operatorname{score}_{\theta}(\langle\langle t_n, t_{n+1}\rangle\rangle, \boldsymbol{w})).$

The output of the forward algorithm lifted into the semiring will yield:

$$\bigoplus_{t_{1:N} \in T^n} \bigotimes_{n=1}^N \langle w, -w \log w \rangle \tag{39}$$

We want to show that the result of the forward propagation lifted in the semiring is the same as the unnormalized Entropy:

$$H_u(T_w) = -\sum_{\mathbf{t} \in \mathcal{T}^N} \exp(score_{\boldsymbol{\theta}}(\mathbf{t}, \boldsymbol{w})) score_{\boldsymbol{\theta}}(\mathbf{t}, \boldsymbol{w})$$
(40)

We show this by induction. Starting with the base case where N=1:

$$\bigoplus_{t_1 \in T^1} \bigotimes_{n=1}^{1} \langle \boldsymbol{w}, -\boldsymbol{w} \log \boldsymbol{w} \rangle = \bigoplus_{t_1 \in T^1} \langle \exp(\operatorname{score}_{\theta}(\langle t_0, t_1 \rangle, \boldsymbol{w})), \qquad (42)$$

$$- \exp(\operatorname{score}_{\theta}(\langle t_0, t_1 \rangle, \boldsymbol{w})) \log(\exp(\operatorname{score}_{\theta}(\langle t_0, t_1 \rangle, \boldsymbol{w}))) \rangle \qquad (43)$$

$$= \bigoplus_{t \in T} \left\langle \exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})), \qquad (44)$$

$$- \exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})) \log(\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))) \right\rangle \qquad (44)$$

$$= \left\langle \sum_{t \in T} \exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})), \qquad (45)$$

Our induction hypothesis is the following:

$$\bigoplus_{t_{1:i} \in T^{i}} \bigotimes_{n=1}^{i} \langle w, -w \log w \rangle = \left\langle \sum_{t \in T^{i}} \exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})), -\sum_{t \in T^{i}} \exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})) \log(\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))) \right\rangle$$
(46)

Meaning we assume that $\beta(\mathbf{w}, t_i)$ corresponds to the unnormalized entropy of all sequences of length i.

Now we proceed with the induction step, where $i \to i + 1$:

$$\bigoplus_{t_{1:N-1} \in T^N} \bigotimes_{n=1}^{N} \langle w, -w \log w \rangle \tag{47}$$

$$= \bigoplus_{t_{1:N-1} \in T^{N-1}} \bigoplus_{t_n \in T} \bigotimes_{n=1}^{N} \langle w, -w \log w \rangle \tag{48}$$

$$= \bigoplus_{t_{1:N-1} \in T^{N-1}} \bigotimes_{n=1}^{N-1} \exp(\operatorname{score}_{\theta}(\langle t_{n-1}, t_n \rangle, \boldsymbol{w})), -\exp(\operatorname{score}_{\theta}(\langle t_{n-1}, t_n \rangle, \boldsymbol{w})) \log(\exp(\operatorname{score}_{\theta}(\langle t_{n-1}, t_n \rangle, \boldsymbol{w}))) \log(\exp(\operatorname{score}_{\theta}(\langle t_{n-1}, t_n \rangle, \boldsymbol{w})))) \log(\exp(\operatorname{score}_{\theta}(\langle t_n, t_n \rangle, \boldsymbol{w}))) \log(\exp(\operatorname{score}_{\theta}(\langle t_n, t_n \rangle, \boldsymbol{w})))) \log(\operatorname{score}_{\theta}(\langle t_n, t_n \rangle, \boldsymbol{w}))) \log(\exp(\operatorname{score}_{\theta}(\langle t_n, t_n \rangle, \boldsymbol{w})))) \log(\operatorname{score}_{\theta}(\langle t_n, t_n \rangle, \boldsymbol{w}))) \log(\operatorname{sco$$

c) We want to prove:

$$H(T_w) = Z(\boldsymbol{w})^{-1} H_U(T) + \log(Z(\boldsymbol{w}))$$
(51)

$$H(T_w) = -\sum_{t \in TN} p(t \mid w) \cdot \log(p(t \mid w))$$
 (def. H)

$$= -\sum_{t \in T^{N}} \frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))}{Z(\boldsymbol{w})} \log \left(\frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))}{\sum_{t' \in T^{N}} \exp(\operatorname{score}_{\theta}(t', \boldsymbol{w}))} \right)$$
(def. p)

(53)

$$= -\sum_{t \in T^{N}} \frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))}{Z(\boldsymbol{w})} \log \left(\frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))}{Z(\boldsymbol{w})} \right)$$
(54)

$$= -\sum_{t \in TN} \frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))}{Z(\boldsymbol{w})} (\operatorname{score}_{\theta}(t, \boldsymbol{w}) - \log Z(\boldsymbol{w}))$$
(55)

$$= -\sum_{t \in T^{N}} \frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})) \operatorname{score}_{\theta}(t, \boldsymbol{w}) - \exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})) \log Z(\boldsymbol{w}))}{Z(\boldsymbol{w})}$$
(56)

$$= -\sum_{t \in T^N} \frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})) \operatorname{score}_{\theta}(t, \boldsymbol{w})}{Z(\boldsymbol{w})} + \sum_{t \in T^N} \frac{\exp(\operatorname{score}_{\theta}(t, \boldsymbol{w})) \log Z(\boldsymbol{w}))}{Z(\boldsymbol{w})}$$

$$= H_U(T_{\boldsymbol{w}})Z(\boldsymbol{w})^{-1} + \frac{\log(Z(\boldsymbol{w}))}{Z(\boldsymbol{w})} \sum_{t \in T^N} \exp(\operatorname{score}_{\theta}(t, \boldsymbol{w}))$$
(58)

$$= H_U(T_{\boldsymbol{w}})Z(\boldsymbol{w})^{-1} + \log(Z(\boldsymbol{w}))$$
(59)

(60)

(57)