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1 Question: Entropy of a Conditional Random Field

a)

Prove that the **expectation semiring** satisfies the semiring axioms:

Let $x_1, y_1, x_2, y_2, x_3, y_3 \in \mathbb{R}$ for this subsection.

Axiom 1

 $(\mathbb{R} \times \mathbb{R}, \oplus, \langle 0, 0 \rangle)$ is a commutative monoid with identity element $\langle 0, 0 \rangle$

Associativity and Commutativity of \oplus :

$$(\langle x_1, y_1 \rangle \oplus \langle x_2, y_2 \rangle) \oplus \langle x_3, y_3 \rangle \stackrel{!}{=} \langle x_1, y_1 \rangle \oplus (\langle x_3, y_3 \rangle \oplus \langle x_2, y_2 \rangle)$$

$$(\langle x_1, y_1 \rangle \oplus \langle x_2, y_2 \rangle) \oplus \langle x_3, y_3 \rangle = \langle x_1 + x_2, y_1 + y_2 \rangle \oplus \langle x_3, y_3 \rangle$$

$$= \langle (x_1 + x_2) + x_3, (y_1 + y_2) + y_3 \rangle$$

$$= \langle (x_1 + x_2) + x_3, (y_1 + y_2) + y_3 \rangle$$

$$= \langle x_1 + (x_2 + x_3), y_1 + (y_2 + y_3) \rangle$$

$$= \langle x_1, y_1 \rangle \oplus \langle x_2 + x_3, y_2 + y_3 \rangle$$

$$= \langle x_1, y_1 \rangle \oplus \langle x_3 + x_2, y_3 + y_2 \rangle$$

$$= \langle x_1, y_1 \rangle \oplus \langle x_3, y_3 \rangle \oplus \langle x_2, y_2 \rangle)$$

$$(def. \oplus)$$

$$(5)$$

$$= \langle x_1, y_1 \rangle \oplus \langle (x_3, y_3) \oplus \langle x_2, y_2 \rangle)$$

$$(def. \oplus)$$

$$(5)$$

 $\langle 0, 0 \rangle$ is the identity element:

$$\langle 0, 0 \rangle \oplus \langle x_1, y_1 \rangle \stackrel{!}{=} \langle x_1, y_1 \rangle$$

$$\langle 0, 0 \rangle \oplus \langle x_1, y_1 \rangle = \langle x_1 + 0, y_1 + 0 \rangle$$

$$= \langle x_1, y_1 \rangle$$

$$= \langle x_1 + 0, y_1 + 0 \rangle$$

$$= \langle x_1, y_1 \rangle \oplus \langle 0, 0 \rangle$$

$$(10)$$

$$(11)$$

$$= \langle x_1, y_1 \rangle \oplus \langle 0, 0 \rangle$$

$$(12)$$

Axiom 2

 $(\mathbb{R} \times \mathbb{R}, \otimes, \langle 1, 0 \rangle)$ is a monoid with identity element $\langle 1, 0 \rangle$

Associativity of \otimes :

$$(\langle x_1, y_1 \rangle \otimes \langle x_2, y_2 \rangle) \otimes \langle x_3, y_3 \rangle \stackrel{!}{=} \langle x_1, y_1 \rangle \otimes (\langle x_2, y_2 \rangle \otimes \langle x_3, y_3 \rangle)$$

$$(13)$$

$$(\langle x_1, y_1 \rangle \otimes \langle x_2, y_2 \rangle) \otimes \langle x_3, y_3 \rangle = \langle x_1 \cdot x_2, x_1 \cdot y_2 + y_1 \cdot x_2 \rangle \otimes \langle x_3, y_3 \rangle \quad (\text{def. } \otimes) \quad (14)$$

$$= \langle (x_1 \cdot x_2) \cdot x_3, (x_1 \cdot x_2) \cdot y_3 + (x_1 \cdot y_2 + y_1 \cdot x_2) \cdot x_3 \rangle$$
 (def. \otimes) (15)

$$= \langle (x_1 \cdot x_2) \cdot x_3, x_1 \cdot x_2 \cdot y_3 + x_1 \cdot y_2 \cdot x_3 + y_1 \cdot x_2 \cdot x_3 \rangle$$
 (diss. ·) (16)

$$= \langle (x_1 \cdot x_2) \cdot x_3, x_1 \cdot (x_2 \cdot y_3 + y_2 \cdot x_3) + y_1 \cdot x_2 \cdot x_3 \rangle$$
 (diss. ·) (17)

$$= \langle x_1 \cdot (x_2 \cdot x_3), x_1 \cdot (x_2 \cdot y_3 + y_2 \cdot x_3) + y_1 \cdot (x_2 \cdot x_3) \rangle$$
 (ass. ·)

$$= \langle x_1, y_1 \rangle \otimes \langle x_2 \cdot x_3, x_2 \cdot y_3 + y_2 \cdot x_3 \rangle \tag{19}$$

$$= \langle x_1, y_1 \rangle \otimes (\langle x_2, y_2 \rangle \otimes \langle x_3, y_3 \rangle) \tag{def. } \otimes) \tag{20}$$

$\langle 1, 0 \rangle$ is the identity element:

$$\langle x_1, y_1 \rangle \otimes \langle 1, 0 \rangle \stackrel{!}{=} \langle x_1, y_1 \rangle \tag{21}$$

$$\langle x_1, y_1 \rangle \otimes \langle 1, 0 \rangle = \langle x_1 \cdot 1, x_1 \cdot 0 + y_1 \cdot 1 \rangle \tag{def. } \otimes) \tag{22}$$

$$=\langle x_1, y_1 \rangle \tag{23}$$

$$= \langle 1 \cdot x_1, 1 \cdot y_1 + 0 \cdot x_1 \rangle \tag{24}$$

$$= \langle 1, 0 \rangle \otimes \langle x_1, y_1 \rangle \tag{def. } \otimes) \tag{25}$$

Axiom 3

 \otimes distributes left and right over \oplus :

$$\langle x_1, y_1 \rangle \otimes (\langle x_2, y_2 \rangle \oplus \langle x_3, y_3 \rangle) \stackrel{!}{=} (\langle x_1, y_1 \rangle \otimes \langle x_2, y_2 \rangle) \oplus (\langle x_1, y_1 \rangle \otimes \langle x_3, y_3 \rangle) \tag{26}$$

$$\langle x_1, y_1 \rangle \otimes (\langle x_2, y_2 \rangle \oplus \langle x_3, y_3 \rangle) = \langle x_1, y_1 \rangle \otimes \langle x_2 + x_3, y_2 + y_3 \rangle \tag{def. } \oplus) \tag{27}$$

$$= \langle x_1 \cdot (x_2 + x_3), x_1 \cdot (x_2 + x_3) + y_1 \cdot (y_2 + y_3) \rangle$$
 (def. \otimes) (28)

$$= \langle (x_1 \cdot x_2) + (x_1 \cdot x_3), (x_1 \cdot y_2) + (y_1 \cdot x_2) + (x_1 \cdot y_3) + (y_1 \cdot x_3) \rangle$$
 (diss. ·) (29)

$$= \langle x_1 \cdot x_2, (x_1 \cdot y_2) + (y_1 \cdot x_2) \rangle \oplus \langle x_1 \cdot x_3, (x_1 \cdot y_3) + (y_1 \cdot x_3) \rangle$$
 (def. \oplus) (30)

$$= (\langle x_1, y_1 \rangle \otimes \langle x_2, y_2 \rangle) \oplus (\langle x_1, y_1 \rangle \otimes \langle x_3, y_3 \rangle)$$
 (def. \otimes) (31)

$$(\langle x_2, y_2 \rangle \oplus \langle x_3, y_3 \rangle) \otimes \langle x_1, y_1 \rangle \stackrel{!}{=} (\langle x_2, y_2 \rangle \otimes \langle x_1, y_1 \rangle) \oplus (\langle x_3, y_3 \rangle \otimes \langle x_1, y_1 \rangle)$$
(32)

$$(\langle x_2, y_2 \rangle \oplus \langle x_3, y_3 \rangle) \otimes \langle x_1, y_1 \rangle = (\langle x_2 + x_3, y_2 + y_3 \rangle) \otimes \langle x_1, y_1 \rangle \qquad (\text{def. } \oplus) \qquad (33)$$

$$= \langle (x_2 + x_3) \cdot x_1, (x_2 + x_3) \cdot y_1 + (y_2 + y_3) \cdot x_1 \rangle$$
 (def. \otimes) (34)

$$= \langle (x_2 \cdot x_1) \cdot (x_3 \cdot x_1), (x_2 \cdot y_1) + (x_3 \cdot y_1) + (y_2 \cdot x_1) + (y_3 \cdot x_1) \rangle$$
 (diss. ·) (35)

$$= \langle x_2 \cdot x_1, x_2 \cdot y_1 + y_2 \cdot x_1 \rangle \oplus \langle x_3 \cdot x_1, x_3 \cdot y_1 + y_3 \cdot x_1 \rangle \tag{def. } \oplus) \tag{36}$$

$$=(\langle x_2, y_2 \rangle \otimes \langle x_1, y_1 \rangle) \oplus (\langle x_3, y_3 \rangle \otimes \langle x_1, y_1 \rangle) \tag{def. } \otimes) \tag{37}$$

Axiom 4

$$\langle 0, 0 \rangle \otimes \langle x_1, y_1 \rangle \stackrel{!}{=} \langle 0, 0 \rangle \stackrel{!}{=} \langle x_1, y_1 \rangle \otimes \langle 0, 0 \rangle \tag{38}$$

$$\langle 0, 0 \rangle \otimes \langle x_1, y_1 \rangle = \langle 0 \cdot x_1, 0 \cdot y_1 + 0 \cdot x_1 \rangle \tag{def. } \otimes) \tag{39}$$

$$= \langle 0, 0 \rangle \tag{40}$$

$$= \langle x_1 \cdot 0, x_1 \cdot 0 + y_1 \cdot 0 \rangle \tag{41}$$

$$= \langle x_1, y_1 \rangle \otimes \langle 0, 0 \rangle \tag{42}$$

With this, we have proven that the **expectation semiring** is a semiring.

b)

The definition of the forward algorithm is as follows: Where $\forall t_i \in T$ is implicit. We raise this

Algorithm 1: Forward algorithm

```
\overline{\beta(\mathbf{w}, t_0)} = \mathbb{1}

for i = 1 to N do

| \beta(\mathbf{w}, t_i) = \bigoplus_{t_{i-1} \in T} exp(score(t_{i-1}, t_i, \mathbf{w})) \otimes \beta(\mathbf{w}, t_{i-1})

end
```

into the expectation semiring by replacing the \oplus and \otimes with the corresponding operations in the expectation semiring. We also replace the initialization of $\beta(w, t_0)$ with the neutral element of multiplication in the expectation semiring. Which yields the following algorithm:

Algorithm 2: Forward algorithm in the expectation semiring

We want to prove that the result of the forward algorithm lifted in the expectation semiring computes the unnormalized Entropy defined as:

$$H_u(T_{\mathbf{w}}) = -\sum_{t \in T^N} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w})$$
(43)

(11)

The lifted forward algorithm is then equal to:

$$\bigoplus_{t_{1:N} \in T^N} \bigotimes_{n=1}^N \langle w, -w \cdot \log(w) \rangle \tag{45}$$

We prove the statement by induction on the length of the sequence N. The base case is trivial, since the forward algorithm only computes the score of the first token:

$$\bigoplus_{t_{1:1} \in T^1} \bigotimes_{n=1}^1 \langle w, -w \cdot \log(w) \rangle \tag{46}$$

$$= \bigoplus_{t_1 \in T} \langle w, -w \cdot \log(w) \rangle \tag{47}$$

$$(\text{def. } \mathbf{w}) = \bigoplus_{t_1 \in T} \langle \exp(score(t_0, t_1, \mathbf{w})), -\exp(score(t_0, t_1, \mathbf{w})) \cdot score(t_0, t_1, \mathbf{w}) \rangle$$
(48)

$$= \bigoplus_{t \in T} \langle \exp(score_{\theta}(t, \mathbf{w})), -exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \rangle$$
 (49)

$$(\text{def.} \oplus) = \left\langle \sum_{t \in T^1} \exp(score_{\theta}(t, \mathbf{w})), -\sum_{t \in T^1} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right\rangle$$
(50)

Our induction hypothesis is that the forward algorithm computes the unnormalized entropy for sequences of length i:

$$\bigoplus_{t_{1:i} \in T^i} \bigotimes_{n=1}^i \langle w, -w \cdot \log(w) \rangle = \left\langle \sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})), -\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right\rangle$$

Induction step $(i \rightarrow i + 1)$:

$$\bigoplus_{t_{1\cdot i+1} \in T^{i+1}} \bigotimes_{n=1}^{i+1} \langle w, -w \cdot \log(w) \rangle \tag{51}$$

$$= \bigoplus_{t_{i+1} \in T} \left(\bigoplus_{t_{1:i} \in T^i} \bigotimes_{n=1}^i \langle w, -w \cdot \log(w) \rangle \right) \otimes \langle w, -w \log(w) \rangle$$
 (52)

$$(\text{def. IH}) = \bigoplus_{t_{i+1} \in T} \left\langle \sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})), -\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right\rangle$$
(53)

$$\otimes \langle w, -w \log(w) \rangle \tag{54}$$

$$(\text{def. w}) = \bigoplus_{t_{i+1} \in T} \left\langle \sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})), -\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right\rangle$$
(55)

$$\otimes \langle \exp(score(t_i, t_{i+1}, \mathbf{w})), -\exp(score(t_i, t_{i+1}, \mathbf{w})) \cdot score(t_i, t_{i+1}, \mathbf{w}) \rangle$$
 (56)

$$\left(\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w}))\right) \cdot \exp(score(t_i, t_{i+1}, \mathbf{w}))$$
(57)

$$= \sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w}) + score(t_i, t_{i+1}, \mathbf{w}))$$
(58)

Further we also have:

$$\left(\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w}))\right) \cdot \left(-\exp(score(t_i, t_{i+1}, \mathbf{w})) \cdot score(t_i, t_{i+1}, \mathbf{w})\right)$$
(59)

$$+ \left(-\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right) \cdot \exp(score(t_i, t_{i+1}, \mathbf{w}))$$
 (60)

$$= -\sum_{t \in T^{i}} exp(score_{\theta}(t, \mathbf{w}) + score(t_{i}, t_{i+1}, \mathbf{w})) \cdot score(t_{i}, t_{i+1}, \mathbf{w})$$
(61)

$$-\sum_{t \in T_i} exp(score_{\theta}(t, \mathbf{w}) + score(t_i, t_{i+1}, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w})$$
(62)

$$= -\sum_{t \in T^{i}} exp(score_{\theta}(t, \mathbf{w}) + score(t_{i}, t_{i+1}, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \cdot score(t_{i}, t_{i+1}, \mathbf{w})$$
(63)

Using the equalities from equation 57 to equation 57 we can rewrite the induction step as:

$$56 \Rightarrow \bigoplus_{t_{i+1} \in T} \left\langle \left(\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})) \right) \cdot \exp(score(t_i, t_{i+1}, \mathbf{w})), \right.$$
(64)

$$\left(\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w}))\right) \cdot \left(-\exp(score(t_i, t_{i+1}, \mathbf{w})) \cdot score(t_i, t_{i+1}, \mathbf{w})\right)$$
(65)

$$+ \left(-\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right) \cdot \exp(score(t_i, t_{i+1}, \mathbf{w})) \right)$$
 (66)

(with 63) =
$$\bigoplus_{t_{i+1} \in T} \left\langle \left(\sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w})) \right) \cdot \exp(score(t_i, t_{i+1}, \mathbf{w})), \right.$$
(67)

$$-\sum_{t \in T^{i}} exp(score_{\theta}(t, \mathbf{w}) + score(t_{i}, t_{i+1}, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \cdot score(t_{i}, t_{i+1}, \mathbf{w})$$
(68)

(with 57) =
$$\bigoplus_{t_{i+1} \in T} \left\langle \sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w}) + score(t_i, t_{i+1}, \mathbf{w})), \right\rangle$$
(69)

$$-\sum_{t \in T^{i}} exp(score_{\theta}(t, \mathbf{w}) + score(t_{i}, t_{i+1}, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \cdot score(t_{i}, t_{i+1}, \mathbf{w}) \rangle$$
(70)

$$(\text{def.} \oplus) = \left\langle \sum_{t_{i+1} \in T} \sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w}) + score(t_i, t_{i+1}, \mathbf{w})), \right.$$
(71)

$$-\sum_{t_{i+1} \in T} \sum_{t \in T^i} exp(score_{\theta}(t, \mathbf{w}) + score(t_i, t_{i+1}, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \cdot score(t_i, t_{i+1}, \mathbf{w}) \rangle$$
(72)

$$= \left\langle \sum_{t \in T^{i+1}} exp(score_{\theta}(t, \mathbf{w})), -\sum_{t \in T^{i+1}} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right\rangle$$
(73)

Concluding the induction.

From this, we can follow that:

$$\bigoplus_{t_{1:N} \in T^N} \bigotimes_{n=1}^N \left\langle w, -w \cdot \log(w) \right\rangle = \left\langle \sum_{t \in T^N} exp(score_{\theta}(t, \mathbf{w})), -\sum_{t \in T^N} exp(score_{\theta}(t, \mathbf{w})) \cdot score_{\theta}(t, \mathbf{w}) \right\rangle$$

This shows that the second part of the pair is equal to the unnormalized entropy.

c)

To prove:

$$H(T_w) \stackrel{!}{=} Z(w)^{-1} H_U(T_w) + \log Z(w)$$
 (74)

$$H(T_w) = -\sum_{t \in T^N} p(t|w) \cdot \log p(t|w)$$
 (def. H) (75)

$$= -\sum_{t \in T^N} \frac{exp(score_{\theta}(t, w))}{Z(w)} \cdot \log(\frac{exp(score_{\theta}(t, w))}{Z(w)})$$
 (def. p) (76)

$$= -\sum_{t \in TN} \frac{exp(score_{\theta}(t, w))}{Z(w)} \cdot (score_{\theta}(t, w) - \log(Z(w)))$$
 (77)

$$= -\sum_{t \in T^N} \frac{exp(score_{\theta}(t, w)) \cdot (score_{\theta}(t, w) - \log(Z(w)))}{Z(w)}$$
(78)

$$= \sum_{t \in T^N} \frac{exp(score_{\theta}(t, w)) \cdot (-score_{\theta}(t, w) + \log(Z(w)))}{Z(w)}$$
(79)

$$= \sum_{t \in T^N} \frac{exp(score_{\theta}(t, w)) \cdot \log(Z(w))}{Z(w)}$$
(80)

$$-\sum_{t \in T^N} \frac{exp(score_{\theta}(t, w)) \cdot score_{\theta}(t, w)}{Z(w)}$$
(81)

$$= \sum_{t \in T^N} \frac{exp(score_{\theta}(t, w)) \cdot \log(Z(w))}{Z(w)}$$
(82)

$$-\sum_{t \in T^N} (exp(score_{\theta}(t, w)) \cdot score_{\theta}(t, w)) \cdot Z(w)^{-1}$$
(83)

$$= H_U(T_w) \cdot Z(w)^{-1} + \sum_{t \in T^N} \frac{exp(score_{\theta}(t, w)) \cdot \log(Z(w))}{Z(w)}$$
 (def. H_U) (84)

$$= H_U(T_w) \cdot Z(w)^{-1} + \frac{\log(Z(w))}{Z(w)} \cdot \sum_{t \in T^N} exp(score_{\theta}(t, w))$$
(85)

$$= H_U(T_w) \cdot Z(w)^{-1} + \frac{\log(Z(w))}{Z(w)} \cdot Z(w)$$
 (def. $Z(w)$) (86)

$$= Z(w)^{-1}H_U(T_w) + \log Z(w)$$
(87)