

# Exercise Sheet 6

Machine Learning 2, SS16

June 2, 2016

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## Exercise 1

(a)

Define

$$y := x^T W - b^T$$

$$\begin{aligned} p_\theta(x) &= \sum_{h \in \{-1,0,1\}^N} p(x, h) \\ &= \sum_{h \in \{-1,0,1\}^N} \frac{1}{Z} \exp(yh + x^T a) \\ &= \frac{1}{Z} \exp(x^T a) \sum_{h \in \{-1,0,1\}^N} \exp(yh) \\ &= \frac{1}{Z} \exp(x^T a) \sum_{h \in \{-1,0,1\}^N} \exp\left(\sum_{i=1}^N y_i h_i\right) \\ &= \frac{1}{Z} \exp(x^T a) \sum_{h \in \{-1,0,1\}^N} \prod_{i=1}^N \exp(y_i h_i) \end{aligned}$$

Because the expression  $\exp(y_i h_i)$  only depends on the  $i$ 'th component of  $h$ , we can rewrite the sum and product to get:

$$\begin{aligned} p_\theta(x) &= \frac{1}{Z} \exp(x^T a) \prod_{i=1}^N \sum_{h \in \{-1,0,1\}} \exp(y_i h_i) \\ &= \frac{1}{Z} \exp(x^T a) \exp\left(\log\left(\prod_{i=1}^N \sum_{h \in \{-1,0,1\}} \exp(y_i h_i)\right)\right) \\ &= \frac{1}{Z} \exp(x^T a) \exp\left(\sum_{i=1}^N \log\left(\sum_{h \in \{-1,0,1\}} \exp(y_i h_i)\right)\right) \\ &= \frac{1}{Z} \exp(x^T a) \exp\left(\sum_{i=1}^N \log(1 + e^{y_i} + e^{-y_i})\right) \\ &= \frac{1}{Z} \exp(x^T a) \exp\left(\sum_{i=1}^N \log(1 + 2\cosh(y_i))\right) \\ &= \frac{1}{Z} \exp(x^T a + \sum_{i=1}^N \log(1 + 2\cosh(w_i x - b_i))) \end{aligned}$$

$$\begin{aligned}
&= \frac{1}{Z} \exp(x^T a + \sum_{i=1}^N \log(2(\frac{1}{2} + \cosh(w_i x - b_i)))) \\
&= \frac{1}{Z} \exp(x^T a + \sum_{i=1}^N \log(2) + (\frac{1}{2} + \cosh(w_i x - b_i))) \\
&= \frac{1}{Z} \exp(N \log(2) + x^T a + \sum_{i=1}^N (\frac{1}{2} + \cosh(w_i x - b_i))) \\
&= \frac{1}{Z} 2^N \exp(x^T a + \sum_{i=1}^N (\frac{1}{2} + \cosh(w_i x - b_i)))
\end{aligned}$$

With

$$Z' := \frac{2^N}{Z}$$

the desired result follows.

**(b)**

First compute gradients of  $F$ :

$$\nabla_a F$$