

ESE 546, FALL 2023

HOMEWORK 5

KEQI WU [KEQIWU@SEAS.UPENN.EDU],
COLLABORATORS: NONE

Solution 1. Your solution goes here.

(a)

$$\begin{aligned} KL(q||p) &= \sum_{w \in W} q(w) \log \frac{q(w)}{p(w)} \\ &= \sum_{w \in W} q(w) \log \frac{q(w)}{\frac{e^{-\beta \Phi(w)}}{Z(\beta)}} \\ &= \sum_{w \in W} q(w) (\log q(w) - \log \frac{e^{-\beta \Phi(w)}}{Z(\beta)}) \\ &= \sum_{w \in W} q(w) [\log q(w) - (\log e^{-\beta \Phi(w)} - \log Z(\beta))] \\ &= \sum_{w \in W} q(w) [\log q(w) - (-\beta \Phi(w) - \log Z(\beta))] \\ &= \sum_{w \in W} q(w) [\log q(w) + \beta \Phi(w) + \log Z(\beta)] \\ &= \sum_{w \in W} q(w) \beta \Phi(w) + \sum_{w \in W} q(w) \log q(w) + \sum_{w \in W} q(w) \log Z(\beta) \\ &= \beta \sum_{w \in W} q(w) \Phi(w) + \sum_{w \in W} q(w) \log q(w) + \log Z(\beta) \sum_{w \in W} q(w) \\ &= \beta \sum_{w \in W} q(w) \Phi(w) + \sum_{w \in W} q(w) \log q(w) + \log Z(\beta) \end{aligned}$$

(b)

Since $\log Z(\beta) = \log \sum_{w \in W} \frac{e^{-\beta \Phi(w)}}{Z(\beta)}$, we have:

$$\begin{aligned}
\frac{d}{d\beta} \log Z(\beta) &= \frac{d}{d\beta} \log \sum_{w \in W} e^{-\beta \Phi(w)} \\
&= \frac{1}{\sum_{w \in W} e^{-\beta \Phi(w)}} \frac{d}{d\beta} \sum_{w \in W} e^{-\beta \Phi(w)} \\
&= \frac{1}{\sum_{w \in W} e^{-\beta \Phi(w)}} \sum_{w \in W} -\Phi(w) e^{-\beta \Phi(w)} \\
&= -\frac{1}{\sum_{w \in W} e^{-\beta \Phi(w)}} \sum_{w \in W} \Phi(w) e^{-\beta \Phi(w)} \\
&= -\sum_{w \in W} \frac{\Phi(w) e^{-\beta \Phi(w)}}{\sum_{w \in W} e^{-\beta \Phi(w)}} \\
&= -\sum_{w \in W} \Phi(w) \frac{e^{-\beta \Phi(w)}}{\sum_{w \in W} e^{-\beta \Phi(w)}} \\
&= -\sum_{w \in W} \Phi(w) p(w) \text{ as } p(w) = \frac{e^{-\beta \Phi(w)}}{Z} \\
&= -E_{w \sim p(w)}[\Phi(w)]
\end{aligned}$$

Therefore, $E_{w \sim p(w)}[\Phi(w)] = \frac{\partial \log Z(\beta)}{\partial \beta} \approx E_{w \sim q(w)}[\Phi(w)]$ ($p(w) \approx q(w)$) as required

(c)

$q^* = \min_{q \in Q_2} KL(q||p) = \sum_{w \in W} q(w) \log \frac{q(w)}{p(w)}$. $p(w) = 0$ implies $q(w) = 0$ as well, or $KL(q||p) = \infty$. Since $w_i \in \{0, 1\} \forall i = 1 \dots N$ and the probability of $w_i = 1$ is 0, $q(w = 1) = \prod_{i=1}^N q(w_i = 1) = 0$. Then, $q^*(w_i) = 1 - w_i$, and $q^*(w = 0) = \prod_{i=1}^N q(w_i = 0) = 1$, otherwise 0. Therefore, we have the distribution:

$$\begin{aligned}
 KL(q^*||p) &= \sum_{w \in W} q^*(w) \log \frac{q^*(w)}{p(w)} \\
 &= q^*(w = 0) \log \frac{q^*(w = 0)}{p(w = 0)} \\
 &= 1 \times \log \frac{1}{\frac{1}{2^N - 1}} \\
 &= \log(2^N - 1)
 \end{aligned}$$

$q^*(w)$ is not even close to $p(w)$. $q^*(x)$ is not 0 only when all $w_i = 0$ and $p(x) = \frac{1}{2^N - 1}$ when all w_i is not 0.

(d)

$p(x) = (1 - \prod_i^N w_i) \frac{1}{2^N - 1}$, then:

$$\begin{aligned}
 KL(q||p) &= \sum_{w \in W} q(w) \log \frac{q(w)}{p(w)} \\
 &= \sum_w (1 - w_1) \prod_{i=2}^N \log \frac{(1 - w_i) \prod_{i=2}^N q_i(w_i)}{(1 - \prod_{i=1}^N w_i) \frac{1}{2^N - 1}} \\
 &= \sum_{w_2..w_N} \prod_{i=2}^N q_i(w_i) \log \prod_{i=2}^N q_i(w_i) \frac{1}{2^N - 1} \\
 &= \sum_{w_2..w_N} \prod_{i=2}^N q_i(w_i) (\sum_{j=2}^N \log q_j(w_j)) + \log(2^N - 1)
 \end{aligned}$$

The first term is the negative. the uniform distribution minimizes the KL divergence, where $q^*(w_i) = \frac{1}{2}$, then $q^*(w) = (1 - w_1) \prod_{i=2}^N q^*(w_i) = \frac{1-w_1}{2^{N-1}}$. Therefore:

$$\begin{aligned}
 KL(q^*||p) &= - \sum_{w_2..w_N} \log(2^{N-1}) \frac{1}{2^N - 1} + \log(2^N - 1) \\
 &= -\log(2^{N-1}) + \log(2^N - 1) \\
 &= \log \frac{2^N - 1}{2^{N-1}}
 \end{aligned}$$

Now $q^*(w)$ looks similar to $p(w)$

Solution 2. Your solution goes here.

(i)

check code

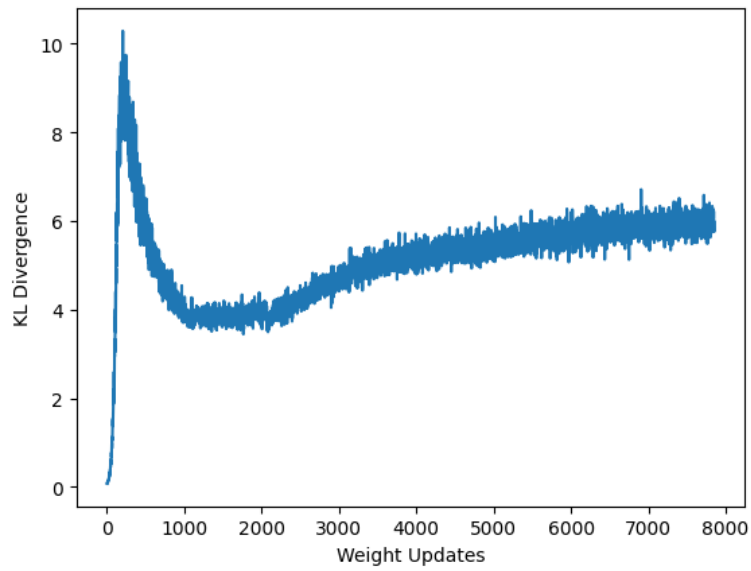
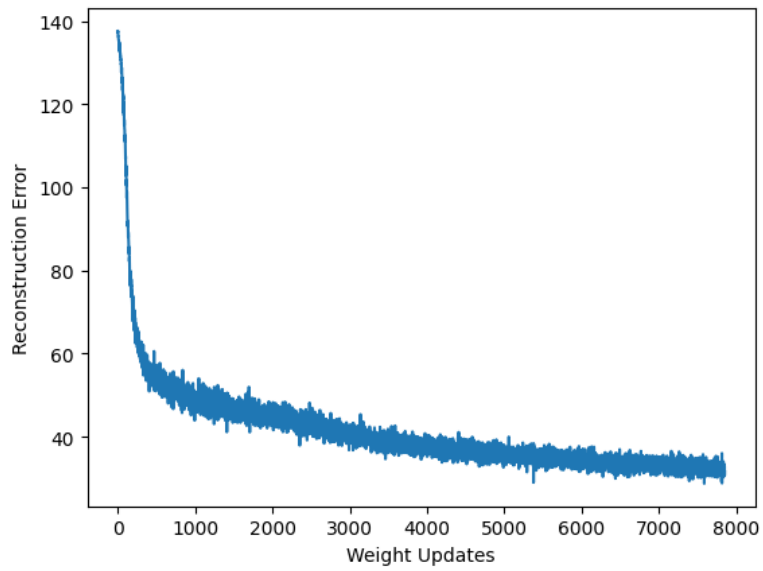
(ii)

check code

```
Epoch [1/50], Loss: 116.9643
Epoch [2/50], Loss: 75.2627
Epoch [3/50], Loss: 63.2424
Epoch [4/50], Loss: 58.6469
Epoch [5/50], Loss: 56.1024
Epoch [6/50], Loss: 54.4040
Epoch [7/50], Loss: 53.0902
Epoch [8/50], Loss: 51.9848
Epoch [9/50], Loss: 51.2476
Epoch [10/50], Loss: 50.4601
Epoch [11/50], Loss: 49.9479
Epoch [12/50], Loss: 49.5056
Epoch [13/50], Loss: 49.1121
Epoch [14/50], Loss: 48.4276
Epoch [15/50], Loss: 47.8799
Epoch [16/50], Loss: 47.3366
Epoch [17/50], Loss: 46.7608
Epoch [18/50], Loss: 46.1849
Epoch [19/50], Loss: 45.4961
Epoch [20/50], Loss: 45.0622
Epoch [21/50], Loss: 44.5507
Epoch [22/50], Loss: 44.0811
Epoch [23/50], Loss: 43.7420
Epoch [24/50], Loss: 43.4028
Epoch [25/50], Loss: 43.0608
Epoch [26/50], Loss: 42.8091
Epoch [27/50], Loss: 42.4771
Epoch [28/50], Loss: 42.2195
Epoch [29/50], Loss: 42.0273
Epoch [30/50], Loss: 41.7019
Epoch [31/50], Loss: 41.4625
Epoch [32/50], Loss: 41.3099
Epoch [33/50], Loss: 41.0869
Epoch [34/50], Loss: 40.8619
Epoch [35/50], Loss: 40.6930
Epoch [36/50], Loss: 40.4849
Epoch [37/50], Loss: 40.2223
Epoch [38/50], Loss: 40.1150
Epoch [39/50], Loss: 39.9316
Epoch [40/50], Loss: 39.8060
Epoch [41/50], Loss: 39.5413
Epoch [42/50], Loss: 39.4373
Epoch [43/50], Loss: 39.2686
Epoch [44/50], Loss: 39.2336
Epoch [45/50], Loss: 38.9682
Epoch [46/50], Loss: 38.8976
Epoch [47/50], Loss: 38.7460
Epoch [48/50], Loss: 38.6323
Epoch [49/50], Loss: 38.4808
Epoch [50/50], Loss: 38.3998
```

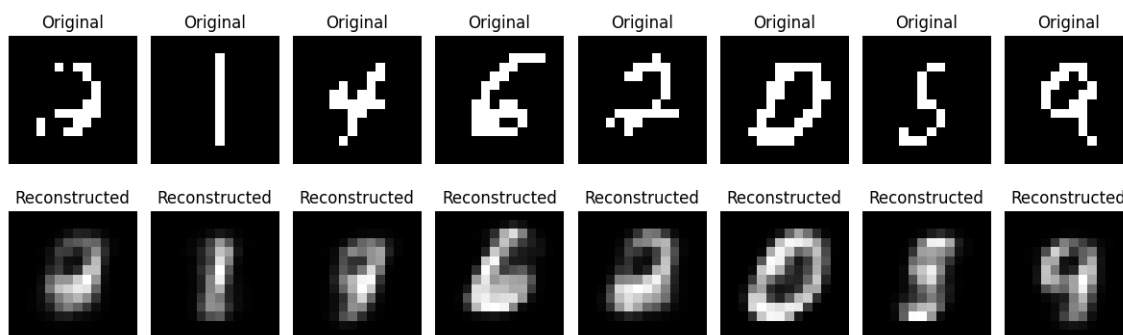
(iii)

check code



(iv)

check code



(v)

check code

