

HW 10 (due 05/31/2023 in class) solution

1. (30%) We can define Kullback Liebler divergence between two probability distribution $P(x)$ and $Q(x)$

$$D_{KL} = - \int P(x) \log \frac{P(x)}{Q(x)} dx$$

Find formula for the Kullback Liebler divergence between two gaussian distribution $P(x)$ and $Q(x)$.

$Q(x)$ is a Gaussian distribution zero mean and standard deviation 1 (for $\sigma=1$, and $\mu=0$) and $P(x)$ is a Gaussian distribution for mean μ and standard deviation σ

$$Q(x) = \frac{1}{\sqrt{2\pi}} \exp \left[-\frac{x^2}{2} \right]$$

$$P(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp \left[-\frac{(x-\mu)^2}{2\sigma^2} \right]$$

Answer:

$$D_{KL} = - \int \frac{1}{\sqrt{2\pi}\sigma} \exp \left[-\frac{(x-\mu)^2}{2\sigma^2} \right] \left(\log \left(\frac{1}{\sigma} \right) - \frac{(x-\mu)^2}{2\sigma^2} + \frac{x^2}{2} \right) dx$$

Using the formula

$$I1 = \int \frac{1}{\sqrt{2\pi}\sigma} \exp \left[-\frac{(x-\mu)^2}{2\sigma^2} \right] (x-\mu)^2 dx = \sigma^2$$

$$I2 = \int \frac{1}{\sqrt{2\pi}\sigma} \exp \left[-\frac{(x-\mu)^2}{2\sigma^2} \right] x dx = \mu$$

$$I1 \text{ and } I2 \rightarrow I3 = \int \frac{1}{\sqrt{2\pi}\sigma} \exp \left[-\frac{(x-\mu)^2}{2\sigma^2} \right] x^2 dx = \mu^2 + \sigma^2$$

After some manipulation, you will obtain

$$D_{KL} = \log(\sigma) + \frac{1}{2} - \frac{1}{2}(\sigma^2 + \mu^2)$$

This divergence is used in variational autoencoder.

Problem 2 composition of sigmoidal and cross entropy (20% each)

In the previous homework we define the cross entropy between two Bernouli distribution as

$$H(p, q) = p \log q + (1-p) \log(1-q)$$

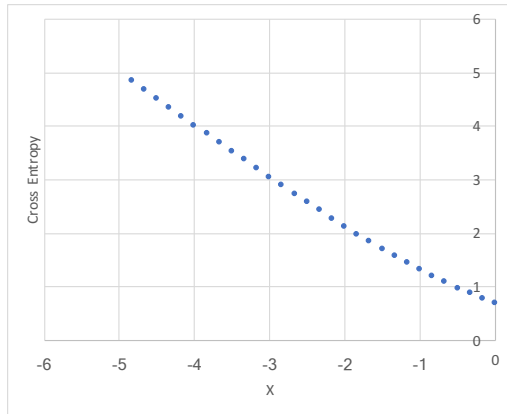
Now we would like you to plot the composition of such cross entropy and a sigmoidal function

$$\sigma(z) = \frac{1}{\exp(-z) + 1}$$

Define a new function $g(z) = H(p=1, \sigma(z))$ by setting $p=1$

(a) Plot $g(z)$ for $-5 \leq z \leq 0$ with at least 30 data point

(b) Find the approximation formula when $z \rightarrow -\infty$ for this function?



You can see that when $z \rightarrow -\infty$, the functional form is roughly linear

$$g = \log[\sigma(z)] = \log\left[\frac{1}{\exp(-z) + 1}\right] = \log\left[\frac{\exp(z)}{\exp(z) + 1}\right] \sim \log[\exp(z)] = z$$

You can verify this by plotting z against $g(z)$.

3. (10% each) (30%) Softmax calculation

In exploration and exploitation, it is possible to use the probability of taking action based on the following formula

$$p(a_i) = \exp(Q(a_i)/T) / \sum_i \exp\left(\frac{Q(a_i)}{T}\right)$$

The summation is sum over all possible actions, i.e, a_i

To gain some idea on this formula, we ask to compute the probability. The variable T is effective temperature. For simplicity, we consider the case of four possible actions a_1, a_2, a_3 , and a_4 . The corresponding Q values are given as

$$Q(a_1) = 10 \quad Q(a_2) = 11 \quad Q(a_3) = 12 \quad Q(a_4) = 13$$

For the probability distribution, $p(a_1), p(a_2), p(a_3)$, and $p(a_4)$ for various temperature

(a) low temperature, $T=1$

(b) intermediate temperature, $T=10$

(c) High temperature, $T= 100$

The probability is calculated and given below. I used excel to calculate

	T=1	T=10	T=100
pa1	0.0320586	0.21383822	0.24626259
pa2	0.08714432	0.23632778	0.24873757
pa3	0.23688282	0.26118259	0.25123743
pa4	0.64391426	0.28865141	0.25376241
sum	1	1	1

You can see that when $T=100$, $p(a1)=p(a2)=p(a3)=p(a4)$

(Note: As an example, you may think of these actions as moving up, down, left and right in 2D grid word. Here, Q differs roughly by $\sim 10\%$.)