We decided this derivation is too much work for a small part of an assignment. We provide the original question and the solution in this document.

This is a very good exercise to check your understanding of integration and Analysis, so we encourage you to try it yourself.

**Todo:** Compute the Entropy of the normal distribution.

$$\mathcal{N}(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp(-\frac{(x-\mu)^2}{2\sigma^2}).$$

## Solution:

The entropy of a Gaussian can be computed without performing any integrals directly:

$$\begin{split} H\left[\mathcal{N}\right] &= -\int\limits_{-\infty}^{\infty} \log\left(N(x;\mu,\sigma^2)\right) \cdot N(x;\mu,\sigma^2) \,\mathrm{d}x \\ &= -\int\limits_{-\infty}^{\infty} \left[ -\frac{(x-\mu)^2}{2\sigma^2} - \frac{1}{2}\log\left(2\pi\sigma^2\right) \right] \cdot N(x;\mu,\sigma^2) \,\mathrm{d}x \\ &= \frac{1}{2\sigma^2} \int\limits_{-\infty}^{\infty} (x-\mu)^2 \cdot N(x;\mu,\sigma^2) \,\mathrm{d}x + \frac{1}{2}\log\left(2\pi\sigma^2\right) \int\limits_{-\infty}^{\infty} N(x;\mu,\sigma^2) \,\mathrm{d}x \\ &= \frac{1}{2} + \frac{1}{2}\log\left(2\pi\sigma^2\right) = \frac{1}{2}\log\left(2\pi\sigma^2\right) \end{split}$$

Todo: Find the Kullback-Leibler divergence between two Gaussian distributions.

$$D_{KL}\left(\mathcal{N}(\mu_1, \sigma_1^2)||\mathcal{N}(\mu_2, \sigma_2^2)\right)$$
.

## Solution:

Let's first simplify the expression for the Kullback-Leibler divergence, by introducing the notation:

$$\mathcal{N}_1(x) = \mathcal{N}(x; \mu_1, \sigma_1^2) \text{ and } \mathcal{N}_2(x) = \mathcal{N}(x; \mu_2, \sigma_2^2)$$
:

$$D_{KL}(\mathcal{N}_1||\mathcal{N}_2) = \int_{-\infty}^{\infty} \mathcal{N}_1(x) \cdot \log\left(\frac{\mathcal{N}_1(x)}{\mathcal{N}_2(x)}\right) dx$$

$$= -H[\mathcal{N}_1] - \int_{-\infty}^{\infty} \log\left(\mathcal{N}_2(x)\right) \mathcal{N}_1(x) dx$$

$$= -\frac{\log\left(2\pi e \sigma_1^2\right)}{2} - \int_{-\infty}^{\infty} \left(-\frac{\log(2\pi \sigma_2^2)}{2} - \frac{(x - \mu_2)^2}{2\sigma_2^2}\right) \cdot \mathcal{N}_1(x) dx$$

$$= -\frac{\log\left(2\pi e \sigma_1^2\right)}{2} + \frac{\log(2\pi \sigma_2^2)}{2} + \int_{-\infty}^{\infty} \frac{(x - \mu_2)^2}{2\sigma_2^2} \cdot \mathcal{N}_1(x) dx$$

$$= \log\left(\frac{\sigma_2}{\sigma_1}\right) - \frac{1}{2} + \int_{-\infty}^{\infty} \frac{(x - \mu_2)^2}{2\sigma_2^2} \cdot \mathcal{N}_1(x) dx$$

$$(1)$$

For space reasons, we continue to compute only the last integral

$$\int_{-\infty}^{\infty} (x - \mu_2)^2 \cdot \mathcal{N}(x; \mu_1, \sigma_1^2) \, \mathrm{d}x = \int_{-\infty}^{\infty} \underbrace{\frac{(x - \mu_1 + \mu_1 - \mu_2)^2}{(x - \mu_1)^2 + (\mu_1 - \mu_2)^2}}_{= (x - \mu_1)^2 + (\mu_1 - \mu_2)^2} \cdot \mathcal{N}(x; \mu_1, \sigma_1^2) \, \mathrm{d}x$$

$$= \int_{-\infty}^{\infty} (x - \mu_1)^2 \mathcal{N}(x; \mu_1, \sigma_1^2) \, \mathrm{d}x$$

$$= \sigma_1^2$$

$$+ (\mu_1 - \mu_2)^2 \int_{-\infty}^{\infty} \mathcal{N}(x; \mu_1, \sigma_1^2) \, \mathrm{d}x$$

$$= 1$$

$$+ 2(\mu_1 - \mu_2) \int_{-\infty}^{\infty} (x - \mu_1) \mathcal{N}(x; \mu_1, \sigma_1^2) \, \mathrm{d}x$$

$$= \sigma_1^2 + (\mu_1 - \mu_2)^2$$

Plugging this back into the previous equation yields

$$D_{KL}\left(\mathcal{N}(\mu_1, \sigma_1^2) || \mathcal{N}(\mu_2, \sigma_2^2)\right) = \log\left(\frac{\sigma_2}{\sigma_1}\right) - \frac{1}{2} + \frac{\sigma_1^2 + (\mu_1 - \mu_2)^2}{2\sigma_2^2} \,.$$

Note that the divergence is 0, if  $\mu_1 = \mu_2$  and  $\sigma_1 = \sigma_2$ .