## **EXERCISES 2**

Exercise 2.1. Sample from a truncated Normal with

$$\bar{p}(x) = \mathcal{N}(x; 0, 1) \mathbf{1}_{x \in [-0.8, 0.8]}(x)$$

using  $q(x) = \mathcal{N}(x; 0, 1)$ . Plot the histogram and the unnormalised density. Can you plot the pdf? You can use np.random.normal for sampling from q(x). Try to choose M and the rejection procedure clearly!

Exercise 2.2. Simulate data from the nonlinear model

$$p(x) = \text{Unif}(x; -10, 10)$$
  
$$p(y|x) = \mathcal{N}(y; a\cos(x) + b, \sigma^2),$$

with a=0.5, b=0.5, and  $\sigma=0.15$ . You can use np.random.normal for this. Scatter plot (x,y) samples and discuss the behaviour. Plot p(y) using samples/histogram, do you think p(y) is computable any other way?

**Exercise 2.3.** Sample the following mixture of Gaussians using only uniforms:

$$p(x) = \sum_{k=1}^{5} w_k q_k(x),$$

where

$$q_k(x) = \mathcal{N}(x; \mu_k, \sigma_k^2).$$

Use the Box-Müller transform and a transformation to sample Gaussians with a certain mean and covariance – and inversion to sample from the discrete distribution. The weights are defined as

$$w_1 = 0.1, w_2 = 0.2, w_3 = 0.3, w_4 = 0.2, w_5 = 0.2,$$

and

$$\mu_1 = -2, \ \mu_2 = -1, \ \mu_3 = 0, \ \mu_4 = 1, \ \mu_5 = 2$$

and

$$\sigma_1 = 0.5, \ \sigma_2 = 0.1, \ \sigma_3 = 0.5, \ \sigma_4 = 0.2, \ \sigma_5 = 0.5.$$

Note that these are standard deviations, not variances. What would you do if each Gaussian was truncated around their mean  $[\mu_k - 0.1, \mu_k + 0.1]$ ? Think about combining your code with a rejection sampler, if mixture components themselves are hard to sample from.