

SOLUTIONS FOR EXERCISES 2

Solution 2.1. Sampling from truncated normal:

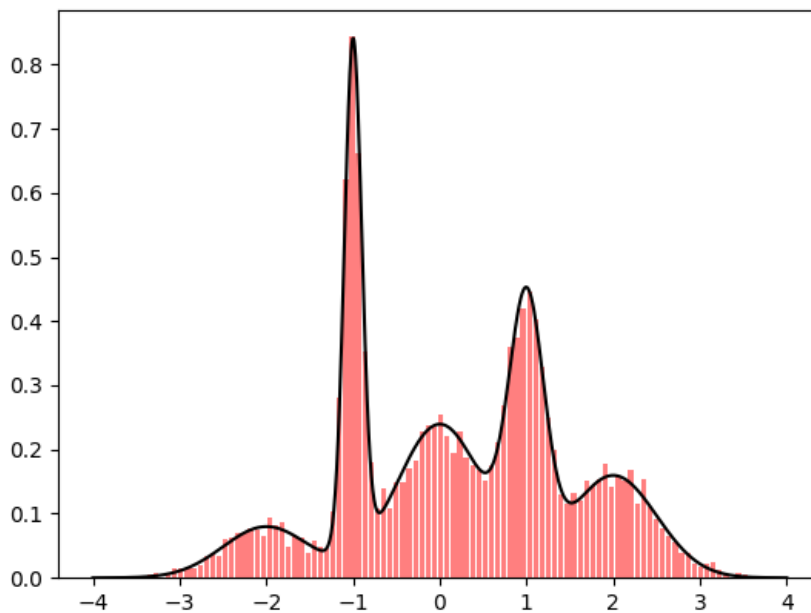
```
1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 n = 100000
5
6 a = 2.5
7 x_accepted = np.array([])
8
9 while len(x_accepted) < n:
10
11     x = np.random.normal(0, 1)
12     if -a <= x <= a:
13         x_accepted = np.append(x_accepted, x)
14
15 plt.hist(x_accepted, bins=100, density=True)
16 plt.show()
```

Solution 2.2. The following code will generate data from the model

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 n = 10000
5
6 x = np.random.uniform(-10, 10, n)
7
8 a = 0.5
9 b = 0.1
10 sigma_0 = 0.15 # this is the standard deviation, not the variance!
11
12 y = a * np.cos(x) + b + sigma_0 * np.random.normal(0, 1, n)
13
14 plt.scatter(x, y, color='k', alpha=1, s=0.05)
15 plt.show()
```

Solution 2.3. The following code will solve the exercise:

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3
4
5 def discrete(s, w): # draws a single sample from a discrete
                     # distribution defined on s with
                     # probabilities w
6
7     cw = np.cumsum(w)
8     sample = []
9
10    u = np.random.uniform(0, 1)
11
12    for k in range(len(cw)):
13        if cw[k] > u:
14            sample = s[k]
15            break
```



```

16     return sample
17
18
19 s = np.array([0, 1, 2, 3, 4]) # support of the discrete distribution (
                                indices)
20 w = np.array([0.1, 0.2, 0.3, 0.2, 0.2]) # weights of the discrete
                                           distribution (probabilities)
21 mu = np.array([-2, -1, 0, 1, 2]) # mean of the Gaussian components
22 sigma = np.array([0.5, 0.1, 0.5, 0.2, 0.5]) # standard deviation of
                                                the Gaussian components
23
24 N = 10000 # number of samples to draw
25 x = np.zeros(N) # initialize the array to store the samples
26
27 for i in range(N):
28     samp = discrete(s, w) # sample from the discrete distribution
29     x[i] = np.random.normal(mu[samp], sigma[samp], 1) # sample from
                                                         the Gaussian with the sampled
                                                         index
30
31 plt.hist(x, bins=100, density=True, rwidth=0.8, color='r', alpha=0.5)
32 plt.show()

```

If you want to visualise the mixture density with the samples, remove the last plotting lines (last two lines above) but add the following:

```

1 def mixture_of_gauss(xx, mu, sigma, w):
2     yy = np.zeros(len(xx))
3
4     for i in range(len(mu)):
5         yy += w[i] * np.exp(-0.5 * (xx - mu[i])**2 / sigma[i]**2) / np
                                     .sqrt(2 * np.pi * sigma[i]
                                     **2)
6

```

```
7     return yy
8
9 xx = np.linspace(-4, 4, 1000)
10 yy = mixture_of_gauss(xx, mu, sigma, w)
11
12 plt.hist(x, bins=100, density=True, rwidth=0.8, color='r', alpha=0.5)
13 plt.plot(xx, yy, 'k-')
14 plt.show()
```

which will give you the plot shown above.