## 4.13

## Intro

In this exercise, we will determine the minimum amount of samples we must drawn from a particular distribution to have a certain posterior credible interval. This situation is quite common in practice. For example, when we take a number of measures in a lab experience we need to report it as a single number plus a degree of uncertainty. One common approach is to report the sample average together with the sample standard deviation. The posterior credible interval is a more Bayesian approach to this situation.

## Solution

a)

The r.v. is Gaussian and has the following distribution:  $X \sim \mathcal{N}(\mu, \sigma^2 = 4)$ . Since  $\mu$  is unknown, we treat it as a r.v. as well. Its prior is  $\mu \sim \mathcal{N}(\mu_0, \sigma_0^2 = 9)$ . The posterior is  $\mu \sim \mathcal{N}(\mu_n, \sigma_n^2)$ , so we know we are dealing with Bayesian inference of Gaussian parameters. This is important because from section 4.6.1 we have an expression that relates the three standard deviations:

$$V_N^{-1} = V_0^{-1} + N\Sigma^{-1}$$

$$\frac{1}{\sigma_n^2} = \frac{1}{\sigma_0^2} + \frac{N}{\sigma^2}$$

$$N = \left[\sigma^2 \left(\frac{1}{\sigma_n^2} - \frac{1}{\sigma_0^2}\right)\right]$$
(1)

Now, we need to find the posterior variance  $\sigma_n^2$  to determine the number of samples. As you may already guess, the posterior variance is deeply related to the credible interval. From the problem statement we know that |u-l|=1. Using the hint of the author about Gaussians we conclude that

$$|u - l| = \mu_n + 1.96\sigma_n - (\mu_n - 1.96\sigma_n) = 3.92\sigma_n$$

$$1 = 3.92\sigma_n$$

$$\sigma_n = \frac{1}{3.92}$$
(2)

With the three variances ( $\sigma_0^2=9,\ \sigma_n^2=\frac{1}{3.92^2},\ \sigma^2=4$ ) at hand, we can calculate N.

$$N = \left\lceil \sigma^2 \left( \frac{1}{\sigma_n^2} - \frac{1}{\sigma_0^2} \right) \right\rceil = \lceil 61.02 \rceil = 62 \tag{3}$$

## Conclusion

In this exercise we determined the number of samples necessary to have a posterior credible interval of width equal to 1. In order to do that we performed Bayesian inference in the mean parameter of the original Gaussian distribution  $X \sim \mathcal{N}(\mu, \sigma^2 = 4)$ . We discovered that the minimum number of samples for  $p(l \leq \mu \leq u) \geq 0.95$  is N = 62.

It is worth noting that in (1) the number of necessary samples increases if the original standard deviation  $(\sigma)$  increases. The number of samples also increases if we want to get a posterior with better precision  $(\frac{1}{\sigma_x^2})$ .