

# STA623 - Bayesian Data Analysis - Assignment 1

22 - 26 September 2025

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## Assignment

Please email your typed or scanned solutions before 23:59 on Monday 24 November 2025 to BOTH [mhenrion@mlw.mw](mailto:mhenrion@mlw.mw) and [biostat-unima@unima.ac.mw](mailto:biostat-unima@unima.ac.mw).

Please include **STA623 - Assignment 1** in the subject line. Please include your code, model output and graphs. Please comment any submitted code.

## Notation

Please try to use the following notation where possible.

- $X, Y, Z$  - random variables
- $x, y, z$  - measured / observed values
- $\bar{X}, \bar{Y}, \bar{Z}$  - sample mean estimators for  $X, Y, Z$
- $\bar{x}, \bar{y}, \bar{z}$  - sample mean estimates of  $X, Y, Z$
- $\hat{T}, \hat{t}$  - given a statistic  $T$ , estimator and estimate of  $T$
- $P(A)$  - probability of an event  $A$  occurring
- $f_X(\cdot), f_Y(\cdot), f_Z(\cdot)$  - probability mass / density functions of  $X, Y, Z$
- $p(\cdot)$  - used as a shorthand notation for pmfs / pdfs if the use of this is unambiguous
- $X \sim F$  -  $X$  distributed according to distribution function  $F$
- $E[X], E[Y], E[Z], E[T]$  - the expectation of  $X, Y, Z, T$  respectively

Table 1: Please use the random seed associated with your name / ID. Solutions using other data than those generated using your seed will not be accepted.

Student	ID	Seed
Eric Mangani	MSC/BIO/STAT/08/23	1899
Satiel Ngwira	MSC/BIO/STAT/17/23	1845
Ausbin Kutumani	MSC/BIO/STAT/J/01/25	1845
Chikondi Moyo	MSC/BIO/STAT/J/03/25	1608
Kenneth Kachiphaphi	MSC/BIO/STAT/J/04/25	1316
Steven Kaunda	MSC/BIO/STAT/J/06/25	1408
Felix Msamira	MSC/BIO/STAT/J/07/24	1005
Eliams Moyo	MSC/BIO/STAT/S/02/24	2616
Loveness Soko	MSC/BIO/STAT/S/04/24	2587
Filudi Nakutuwa	MSC/BIO/STAT/S/07/24	2472
Ephat Chitsulo	MSC/BIO/STAT/S/08/24	2100
Alex Kachitsa	MSC/BIO/STAT/S/09/24	1970
Steven Chiyembe	MSC/BIO/STAT/S/10/2024	2387
Charity Hamuza	MSC/BIO/STAT/S/12/24	2268
Cassim Nanyumba	MSC/BIO/STAT/S/13/24	1935
Hastings Malunga	MSC/BIO/STAT/S/14/24	1296
Osward Kaposi	MSC/BIO/STAT/S/15/24	1472
Seti Evance	MSC/BIO/STAT/S/16/24	1344
Edward Kamphongwe	MSC/BIO/STAT/S/17/24	2184
Chikondi Banda	MSC/BIO/STAT/S/19/24	2688
Steven Nanga	MSC/BIO/STAT/S/23/24	1920
Chikumbutso Banda	NA	1560

## Exercise

For the exercise below, you will need to specify a seed value. You will be given individual seed numbers according to the table on the previous page. **You have to use your own individual seed value** – your data (and hence your results) will be unique to you and different from those of your colleagues.

Assume you observe some data  $y_1, \dots, y_n$  for the waiting times (in hours) from arrival to be seen by a doctor at a large hospital's A&E department.

1. Why could an  $\text{Exponential}(\lambda)$  sampling model be a reasonable assumption? [5 marks]

For the rest of this exercise, assume that the data are exponentially distributed:

$$Y_1, \dots, Y_n \sim \text{Exp}(\lambda)$$

2. Run the code below to generate the `dat` data frame. In the first line, you have to specify a random seed. You are each given a different seed value (meaning no two of you have the same dataset). **Be sure to change the first line to include your individual seed value!** Print out the number of data observations in your dataset, the average lambda value used for your dataset, the average waiting time  $\bar{y}$  (as per the `wait` column in the `dat` data frame) and the number of male patients (as per variable `sex`). [5 marks]

```
set.seed(0000) # REPLACE 0000 with your individual seed value!
# Solutions using the seed value 0000 will not be accepted.

n<-rpois(n=1,lambda=100)
es<-rnorm(1,mean=1.2,sd=0.075)

sex<-sample(x=c("Male","Female"),size=n,prob=c(0.5,0.5),replace=TRUE)
lambda<-rgamma(n=n,shape=10,rate=ifelse(sex=="Male",1.5,1.5*es))

dat<-data.frame(
  sex=sex,
  wait=rexp(n=n,rate=lambda)
)
```

3. Write computer code (and submit a print-out of this code with your assignment) that fits the model resulting from a  $\Gamma(a, b)$  prior and an  $\text{Exp}(\lambda)$  sampling model to the data `dat`. You can choose your own values `a, b` for the prior. Make sure the model estimates WAIC while sampling. [20 marks]

4. Do some diagnostic checks on the results: show the trace plot for  $\lambda|y_1, \dots, y_n$  and plot an estimate of the posterior based on the MCMC results. Compute the Gelman-Rubin potential scale reduction factor. Do you see evidence for non-convergence? [20 marks]
5. Interpret your results:
  - What is the posterior mean of  $\lambda|y_1, \dots, y_n$ ? [5 marks]
  - What is the posterior median of  $\lambda|y_1, \dots, y_n$ ? [5 marks]
  - Compute a 95% Bayesian confidence interval for your posterior estimate of  $\lambda|y_1, \dots, y_n$ . [15 marks]
  - How does your prior compare to your posterior? [5 marks]
  - Do your computational results agree with the theoretical posterior distribution from question 1 above? [5 marks]
6. Now assume that the rate parameter  $\lambda$  depends on the patient's sex:

$$\log(\lambda) = \beta_0 + \beta_1 \cdot x_{female}$$

where  $x_{female} = 1$  if the patient is female, 0 otherwise. Assume a  $\mathcal{N}(0.5, 5)$  prior for  $\beta_0$  and a  $\mathcal{N}(0, 5)$  prior for  $\beta_1$ . Write code that fits the resulting model using MCMC and compare the posterior distributions of  $\lambda_{male} = \exp(\beta_0)$  and  $\lambda_{female} = \exp(\beta_0 + \beta_1)$ . Compare the WAIC from this model to the WAIC from the model you fitted in question 3. Which model do you recommend? [15 marks]