

# CSSS/STAT 564: Assignment 2

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## 0.1 Instructions

1. Fork this repository to your account
2. Edit the file `solutions.Rmd` with your solutions to the problems.
3. Submit a pull request to have it graded. Include either or both a HTML and PDF file.

For updates and questions follow the Slack channel: `#assignment2`.

This assignment will require the following R packages:

```
library("rstan")
```

## 0.2 How much does the prior influence the posterior?

In this problem, we aim for a greater understanding of the influence of a normal prior on inference about the mean of a normal distribution.

Suppose you observe  $n$  single data points which you model as coming from a normal distribution. Consider the case in which the mean of the distribution is  $\mu$  (unknown), but the variance,  $\sigma^2$ , is known:

$$x|\mu \sim N(\mu, \sigma^2)$$

You provide a prior distribution for  $\mu$ ,

$$\mu \sim N(\mu_0, \sigma_0^2)$$

A normal prior for  $\mu$  is conjugate to a normal likelihood, meaning that the posterior distribution is also normal:

$$\mu|x \sim N\left(\frac{\frac{1}{\sigma_0^2}\mu_0 + \frac{n}{\sigma^2}\bar{x}}{\frac{1}{\sigma_0^2} + \frac{n}{\sigma^2}}, \frac{1}{\frac{1}{\sigma_0^2} + \frac{n}{\sigma^2}}\right)$$

**Q1** In your own words, what is a conjugate prior? What are the advantages and disadvantages of such a prior? Suppose that instead of a normal prior for  $\mu$ , a non-conjugate distribution such as the Laplace distribution were used. Write down the posterior density function in this case. Why might a normal prior be preferred?

**Q2** Describe the effects of the following on inference about  $\mu$ :

- Lots of data vs. not much data
- Strong prior information vs. weak prior information
- Lots vs. little noise in the data
- Data that is incompatible with the prior vs. data that is compatible with the prior

A normal prior with finite variance necessarily contributes information to the model; the question is how much. We now investigate a way of understanding the information in a normal prior in terms of implicit additional observed data.

**Q3** The data in `data/pr2data.csv` were generated from a normal distribution with a certain mean and variance. First, fit a linear model with just an intercept and report the estimated mean and variance, as well as a centered 97% confidence interval for  $\mu$ . Interpret the confidence interval appropriately.

**Q4** Now you're told that the variance of the normal distribution that generated these data is 100 (this should be no surprise if you've done the previous question). Fit a Bayesian model for the mean assuming a normal likelihood with standard deviation 10, and with prior  $N(10, 0.1^2)$ . (In the usual notation for the normal model, the second parameter is the variance; but in *Stan*, the second parameter is the standard deviation, because that is unfortunately the convention in *R*. So here, the standard deviation is 0.1 and the variance is 0.01.) Would you characterize this prior as "highly informative," "modestly informative," or "uninformative," and why? Provide the same output (estimate, variance/standard deviation of estimate, confidence interval) you did in **Q3**, and interpret the 97% confidence interval appropriately (hint: use the *print* function for Stan models).

**Q5** Now, generate  $10^4$  independent observations from a  $N(10, 10^2)$  distribution, center them to have a mean of 10 (this is cheating a bit, we fully admit), concatenate them to the data from `data/pr2data.csv`, and fit both a standard linear model with just a mean intercept and a Bayesian model with normal likelihood and an improper flat prior. Provide the same outputs as before (no need to interpret the confidence intervals again, though).

**Q6** What do you think is going on here? How can we interpret the prior in terms of a number of additional data points drawn from a distribution? Relate this distribution back to the formula for the posterior distribution  $\mu|x$ , given before **Q1**.

**Q7** What simplifying assumptions were made in this problem for illustrative purposes, and what do you think might happen to inference about  $\mu$  (the estimation of the mean and the standard error of that mean) if those assumptions were relaxed?