

# Bayesian Statistics and Probabilistic Programming – Spring 2022

## Course project guidelines

### Goals, extension, format

In addition to the merely academic purpose of grading, the course project should be an opportunity of practicing and orienting several skills to the common aim of gathering and sorting information on a specific subject and, then, composing a short, but self-contained and informative report on it.

An orientative figure for the expected length could be around fifteen pages, **plus code**. All submitted documents should be packed together as a .zip file.

**(1) A written report** in pdf format is required, including an abstract, a context introduction where the project scope is delimited and its goal stated, then a well-structured text body, plus full references, for articles or books, following the usual conventions in scholarly documents, for online documents specify an accessible, unbroken link. Difficult-to-find documents should be included in the submitted

The length figure is just a *roughly orientative* estimate, meaning that a three page report, say, will certainly be deemed as much too short, whereas a forty five pages report would be an undue anticipation of your Master thesis.

**(2) Code** (a jupyter or Rmd notebook) is also required and should be fully operational, including data or pointer to an online source thereof. Any needed auxiliary material should be properly mentioned and described: I will run your submitted code expecting no errors or suspicious warnings appear. Try to be exhaustive, up to the point of fastidiousness in documenting code. It will be judged (and graded) better the more detailed, illustrative and clarifying you manage to be in this respect.

Both **individual or group projects** are possible. In collective projects each author is responsible of and should be able to answer questions about the whole project. Meaning there should be no allocation of different parts to

different team members.

Send email with your intended project topic (and, if applicable, the list of group members) so I can give you the go ahead.

## Possible subjects

In principle, any Data Analysis topic is possible, provided it is connected with, or seen from a Bayesian and/or Probabilistic Programming perspective. Thus if you know of, or have seen some possible subject, you can propose it by sending me an email. As possible suggestions of lines along which you can search for a theme, I propose the following classification:

### A) Methodology-oriented projects:

1. Sampling schemes/approximation methods. Variants and extensions of sampling or approximation methods seen -or to be seen this month- in class. Some of these: Integrated Nested Laplace Approximation (INLA), Variational Inference, Langevin Markov Chain Monte Carlo, Adaptive MCMC, Reversible Jump MCMC.
2. Other models: Bayesian versions of CART, or clustering, or neural networks, or other aspects of inference: Bayesian model selection (in particular predictor selection, e.g. *spike-slab* method). Model averaging.
3. Centered on a probabilistic programming language, e.g., Edward(2), Pyro, Tensorflow-probability, Turing.

### B) Data-oriented projects:

Projects centered on analyzing a given dataset or collection of datasets. This variety of project should include treatment applying several methods, with a detailed comparison and discussion of their respective strengths and weaknesses. Code should be thoroughly documented.

If in your job or research work you are using Bayesian Statistical methods, a possible course project is an informative description of some relevant problem and its treatment.

## Caution

Beware of too ambitious projects which, as stated above, should better go to your Master thesis.