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- Practicalities
- Examination
- Course improvements

Bayesian Statistics and Data Analysis Course information

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Course information

The aims of this course are that, after this course you should:

- 1. have knowledge in basic concepts, philosophy, and perspectives in Bayesian Statistics,
- 2. derive posterior distributions in simple situations,
- 3. derive and use predictive distributions,
- 4. identify and formulate Bayesian probabilistic models for analysis and predictions,
- 5. estimate models using contemporary computer-based methods for posterior approximations,
- understand and use basic principles for decisions under uncertainty.
- have knowledge about and be able to use Bayesian methods for model comparisons,
- 8. be able to critically evaluate Bayesian methods,
- 9. report, orally and in writing, a Bayesian statistical analysis



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Pre-requisites

- Basic probability theory
 - probability, probability density, distribution
 - sum, product rule, and Bayes' rule
 - expectation, mean, variance, median
- Basic linear algebra and calculus
- Basic visualisation techniques (R or Python)
 - histogram, density plot, scatter plot

First assignment is a recap.



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Course Outline

Two main parts:

- Core Content (9 lecture blocks)
- Assignments (8 individual assignments)
- Mini-project: do your own Bayesian data analysis (2-3 students)

Exact dates and details; see the course page.



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Core Content

- Every week: lectures/and assignments (approx. 2-4h)
 - Online video material and reading assignments (approx. 2-4h, 50-90 pages a week)
 - Lecture(s): present overall theory and content (overview)
 - Assignment(s): Computational and theoretical individual work. Start monday morning every week!
- An individual assignment (approx. 12-16h). Deadline Sundays 23.59.
- Recommended workflow for each week
 - Do the reading assignments
 - Watch the videos (although, optional)
 - Do self-study exercises
 - Start with the assignment
 - Attend lecture (bring questions!)
 - Attend datalabs (bring questions!)
 - Submit the assignment



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Assignments



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Assignments

- Core components and concepts and state-of-the-art methods
- 2. Warning! There might be bugs in the assignments!
- All labs can be turned in a three times (but use the first).See Studium for details.
- 4. We will mark and return each assignment within 10 working days.
- 5. Important! Do not write your name anywhere
- 6. Important! Do the assignment evaluation!
- 7. Three one hour zoom seminars per week with individual help on assignments.
- 8. We supply grading information for all assignments. Although, they may change!



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R vs Python and Colab

- We strongly recommend using R in the course as there are more packages for Stan and statistical analysis in general in R
- If you are already fluent in Python, but not in R, then using Python may be easier, but it can still be more useful to learn also R
- We supply a Google Colab template with everything pre-installed.
- We also supply a knitR LATEX template suitable for Overleaf.



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Stan

- Stan is a probabilistic programming framework (PPF) and ecosystem
- 40+ developers, 100+ contributors, 100K+ users
- R, Python, Julia, Scala, Stata, Matlab, command line interfaces
- More than 120 R packages using Stan
- Many packages to support diagnostics and workflow
- Can be used for frequentist inference as well
- Alternative PPF exists, Turing (Julia), Pyro (PyTorch), etc.





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Mini-project



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Mini-project

- See project instructions on webpage for details.
- Data analysis of choice on real data.
- 2-3 students.
- Supply a half-page project proposal of data and problem in the middle of the course.
- Ideally, use a model not presented in this course.
- Project will last two weeks (half time) but start earlier.
- The project should use Stan.
- Approximate 40 hours of work per student.
- The project should result in a 4 page report (PDF) using the ICML LaTeX template (see course page).
- Project oral presentation (10-15 minutes)



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Practicalities



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Practicalities

- Course page/content: Github please do a PR if something is wrong!
- Communication: Studium
- Schedule: Time Edit/Studium
- Assignments submissions: Studium
- Acknowledgements: Aki Vehtari
- Teaching assistant: Väinö Yrjänäinen



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Literature

 Book: Gelman, Carlin, Stern, Dunson, Vehtari & Rubin: Bayesian Data Analysis, Third Edition. (online pdf available)



• Additional articles and blog posts



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Examination



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Examination

- 1. To pass (G): All labs, mini-project, and project review need to be passed
- 2. To pass with distinction (VG): 7/10 VG points
- 3. If everything is correct in an assignment (¿90%), 1 VG point is awarded *on the first submission deadline*.
- 4. The mini-project is worth 2 VG-points (if it is passed with distinction).
- 5. Ph.D. students: I suggest you get VG to pass the course. Make the project a potential paper.



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Course improvements since last time

• Multiple short TA sessions over zoom