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- Assignments
- Mini-project
- Practicalities
- Examination
- Course improvements

Bayesian Statistics and Data Analysis Course information

Måns Magnusson Department of Statistics, Uppsala University



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Who am I?

- 1. Associate professor in Statistics
- 2. Teaching: Machine Learning and Bayesian Statistics
- 3. Research: Machine Learning and Bayesian Statistics
 - 3.1 Bayesian Statistics and computational statistics
 - 3.2 Data-centric machine learning
 - 3.3 Statistical inference from textual data
 - 3.4 Applications: Law, sociology, political science, and medicine



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Section 1

Course information



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Course information

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



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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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- Basic probability theory
 - probability, probability density, distribution
 - sum, product rule, and Bayes' rule
 - expectation, mean, variance, median



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- Basic probability theory
 - probability, probability density, distribution
 - sum, product rule, and Bayes' rule
 - expectation, mean, variance, median
- Basic linear algebra and calculus



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- 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
- Note! This is a masters course in Statistics.



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First assignment is a recap.



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Who are you? (very brief)

- 1. Name
- 2. What is your expectation on this course? Why did you choose it?



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Course Outline

Two main parts:

• Core Content (9 lecture blocks)



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Course Outline

Two main parts:

- Core Content (9 lecture blocks)
- Assignments (8 individual assignments)



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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 (1-3 students)



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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 (1-3 students)

Exact dates and details; see the course page.



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Core Content

- Every week: lectures (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. (approx. 12-16h). Deadline Sundays 23.59.
 Start monday morning every week!



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 Start monday morning every week!
- 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 Zoom datalabs (bring questions! Helps with debugging Stan code)
 - Submit the assignment



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Section 2



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 Core components and concepts and state-of-the-art methods



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- Core components and concepts and state-of-the-art methods
- 2. Warning! There might be bugs in the assignments!



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- 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 (See Studium for details), but only first time can give VG:
 - 3.1 The week of the assignment
 - 3.2 The last day of the course
 - 3.3 2-4 weeks after the course



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- 5. Important! Do not write your name anywhere



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- 6. Important! Do the assignment evaluation!



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- 7. Three one hour zoom seminars per week with individual help on assignments. *Note!* Don't send us e-mail, instead use the Zoom seminars.



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- 7. Three one hour zoom seminars per week with individual help on assignments. *Note!* Don't send us e-mail, instead use the Zoom seminars.
- 8. We supply grading information for all assignments. Although, they may change!
- 9. We have quite strict formatting guidelines. Read carefully!
- Language models (eg ChatGPT) is not allowed. I will include "white herrings". Only what you can read in the assignment is included.



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 We strongly recommend using R in the course as there are more packages for Stan and statistical analysis in general in R



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- All data for the assignments are included in the R package. You can access the data data or data-raw in the R package folder in the repo if you need to get the data without using the bsda R package.



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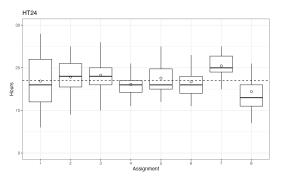
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- We also supply a knitR LATEX template suitable for Overleaf.



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Course Workload

- Conclusions from 2024
 - Slightly too large workload in assignment 7



Student course workload



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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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Section 3

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.



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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.



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- 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.



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- 2-3 students.
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- 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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- 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)
- The first author is corresponding author



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- 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)
- The first author is corresponding author
- This year.
 - Test to implement complex models in Stan as project reach out to me.
 - Mini-project discussions: after lectures and at Thursday Zoom sessions (first from next week.



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Master theses using Stan

- New thing this year: Joint supervision and work with researchers and students (PhD and masters)
- For students that want to work closely to research(ers)
- Headed by me and Sara Hamis at IT
- First step: a joint hackathon in mid november (morning + lunch).



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- Six project (three from stats and three from IT) on Stan and Bayes:



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Master theses using Stan

- New thing this year: Joint supervision and work with researchers and students (PhD and masters)
- For students that want to work closely to research(ers)
- Headed by me and Sara Hamis at IT
- First step: a joint hackathon in mid november (morning + lunch).
- Six project (three from stats and three from IT) on Stan and Bayes:
- At Statistics:
 - Modelling of pollen data in Ancient greece (with Anton Bonnier)
 - Using the Gumbel Softmax to do missing data imputation with discrete parameters (with Jakob Torgander)
 - Modelling proteins in pharmacometrics



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Section 4

Practicalities



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 Course page/content: Github – please do a PR or submit an issue if something is wrong!

https://github.com/MansMeg/BSDA/issues



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Communication: Studium



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• Communication: Studium

Schedule: Time Edit/Studium

• Assignments submissions: Studium



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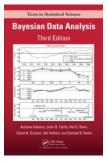
• Teaching assistant: Väinö Yrjänäinen



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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 (see reading list per week)



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1. To pass (G): All labs, mini-project, and project review need to be passed



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- 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



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- 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*.



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- 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.
- 6. Reassesment of grades (supply form to course admin)
- 7. Failing the course: You will need to redo all assignments and mini-project next year.
- 8. Large language models, e.g. chatGPT, are not allowed. "White herrings" are included!



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Section 6

Course improvements



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Course improvements since last time

- Better balance between Assignment 7 and 8.
- White herrings and Large Language Model use.
- New idea (test) of mini project.



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Section 6

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• Put in the work, it is worth it.



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- Put in the work, it is worth it.
- There is alot to do but it will be worth it if you really put in work.



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- Put in the work, it is worth it.
- There is alot to do but it will be worth it if you really put in work.
- Be prepared that the tempo and workload is very high.



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- Put in the work, it is worth it.
- There is alot to do but it will be worth it if you really put in work.
- Be prepared that the tempo and workload is very high.
- Attend the lectures and start with the assignments early.
 Because they are really fun if you get passed the time pressure aspect



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- Start the assignments as soon as possible. It is perfectly reasonable to be done with the assignments before the weekend, so aim for that.



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- Be prepared that the tempo and workload is very high.
- Attend the lectures and start with the assignments early.
 Because they are really fun if you get passed the time pressure aspect
- Start the assignments as soon as possible. It is perfectly reasonable to be done with the assignments before the weekend, so aim for that.
- Do the course properly, it is worth the effort



Questions?

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Questions?