Course and Examination Fact Sheet: Spring Semester 2022

10,382: Econometrics of Big Data

ECTS credits: 4

Overview examination/s
(binding regulations see below)
Decentral - examination paper written at home (individual) (100%)
Examination time: term time

Attached courses
Timetable -- Language -- Lecturer
10,382,1.00 (GSERM) Econometrics of Big Data -- Englisch -- Spindler Martin, Hansen Christian Bailey

Course information

Course prerequisites
The course is a PhD level course. Basic knowledge of parametric statistical models and associated asymptotic theory is expected.

Learning objectives
Students will be introduced to several modern methods, largely coming from statistics and machine learning, which are useful for exploring high dimensional data and for building prediction models in high dimensional settings. Students will learn how to adapt high dimensional methods to the problem of doing valid inference about model parameters and illustrate applications of these proposals for doing inference about economically interesting parameters.

Course content
As in many other fields, economists are increasingly making use of high-dimensional models - models with many unknown parameters that need to be inferred from the data. Such models arise naturally in modern data sets that include rich information for each unit of observation (a type of "big data") and in nonparametric applications where researchers wish to learn, rather than impose, functional forms. High-dimensional models provide a vehicle for modeling and analyzing complex phenomena and for incorporating rich sources of confounding information into economic models.

Our goal in this course is two-fold. First, we wish to provide an overview and introduction to several modern methods, largely coming from statistics and machine learning, which are useful for exploring high-dimensional data and for building prediction models in high-dimensional settings. Second, we will present recent proposals that adapt high-dimensional methods to the problem of doing valid inference about model parameters and illustrate applications of these proposals for doing inference about economically interesting parameters.

Course structure and indications of the learning and teaching design
The course is set up as a weekly seminar in which we will discuss the topics indicated in the syllabus.

Lecture 1: Introduction to High-Dimensional Modeling
Lecture 2: Introduction to Distributed Computing for Very Large Data Sets
Lecture 3: Tree-based Methods
Lecture 4: An Overview of High-Dimensional Inference
Lecture 5: Penalized Estimation Methods
Lecture 6: Moderate p Asymptotics

Lecture 7: Examples

Lecture 8: Inference: Computation

Lecture 9: Introduction to Unsupervised Learning

Lecture 10: Very Large p Asymptotics

Course literature

Course notes and a list of readings provided at the beginning of the course.

Lecture 1: Introduction to High-Dimensional Modeling

- Hastie, T., R. Tibshirani, and J. Friedman (2009), The Elements of Statistical Learning: Data Mining, Inference, and Prediction, Springer. [Elements from Chapters 2, 5, 7, 8.7, 10]
- James, G., D. Witten, T. Hastie, and R. Tibshirani (2014), An Introduction to Statistical Learning with Applications in R, Springer. [Elements from Chapters 2, 3, 5, 7, 8.2]

Lecture 2: Introduction to Distributed Computing for Very Large Data Sets

- Hastie, T., R. Tibshirani, and J. Friedman (2009), The Elements of Statistical Learning: Data Mining, Inference, and Prediction, Springer. [Chapters 9, 10, 15, 16]
- James, G., D. Witten, T. Hastie, and R. Tibshirani (2014), An Introduction to Statistical Learning with Applications in R, Springer. [Chapter 8]

Lecture 3: Tree-based Methods

- Chernozhukov, V., C. Hansen, and M. Spindler (2015), "Valid Post-Selection and Post-Regularization Inference: An
Lecture 5: Penalized Estimation Methods

- Hastie, T., R. Tibshirani, and J. Friedman (2009), The Elements of Statistical Learning: Data Mining, Inference, and Prediction, Springer. [Chapters 3, 4, 5, 18]
- James, G., D. Witten, T. Hastie, and R. Tibshirani (2014), An Introduction to Statistical Learning with Applications in R, Springer. [Chapter 6]

Lecture 6: Moderate p Asymptotics

Lecture 7: Examples


Lecture 8: Inference: Computation

Lecture 9: Introduction to Unsupervised Learning

- Hastie, T., R. Tibshirani, and J. Friedman (2009), The Elements of Statistical Learning: Data Mining, Inference, and Prediction, Springer. [Chapter 14]
- James, G., D. Witten, T. Hastie, and R. Tibshirani (2014), An Introduction to Statistical Learning with Applications in R, Springer. [Chapter 10]

Lecture 10: Very Large p Asymptotics


**Additional course information**

**Only for PhD students of the University of St.Gallen**

PEF and P Econ students may register via regular bidding for the courses offered together by PEF and Global School in Empirical Research Methods (GSERM). Enrolment in a course is binding: students have to attend the course and take the exam. The credits will be shown on the scorecard.

All other PhD students should register for the courses offered by Global School in Empirical Research Methods (GSERM), both via bidding and via GSERM for:

- courses for the curriculum and optional courses with an examination. These will be listed on the scorecard under optional work (only possible if all required elective courses have already been completed).

Please register only via GSERM for:

- optional courses without an examination and
- optional courses if not all required elective courses have been completed (not shown on the scorecard).

In the case of the President’s Board having to implement new directives due to the SARS-CoV-2 pandemic in SpS2022, the course information listed above will be changed as follows:

- This course may take place in a digital version via Zoom.

There will be no change in the examination.
Examination information

Examination sub part/s

1. Examination sub part (1/1)

Examination time and form
Decentral - examination paper written at home (individual) (100%)
Examination time: term time

Remark
take-home final exam

Examination-aid rule
Term papers

Written work must be written without outside help according to the known citation standards, and a declaration of authorship must be attached, which is available as a template on the StudentWeb.

Documentation (quotations, bibliography, etc.) must be carried out universally and consistently according to the requirements of the chosen/specified citation standard such as e.g. APA or MLA.

The legal standard is recommended for legal work (cf. by way of example: FORSTMOser, P., OGOREK R., SCHINDLER B., Juristisches Arbeiten: Eine Anleitung für Studierende (the latest edition in each case), or according to the recommendations of the Law School).

The reference sources of information (paraphrases, quotations, etc.) that has been taken over literally or in the sense of the original text must be integrated into the text in accordance with the requirements of the citation standard used. Informative and bibliographical notes must be included as footnotes (recommendations and standards e.g. in METZGER, C., Lern- und Arbeitsstrategien (latest edition)).

For all written work at the University of St.Gallen, the indication of page numbers is mandatory, regardless of the standard chosen. Where page numbers are missing in sources, the precise designation must be made differently: chapter or section title, section number, article, etc.

Supplementary aids
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Examination languages
Question language: English
Answer language: English

Examination content

The examination will cover the content of the lectures, in particular:

- High-Dimensional Modeling
- Distributed Computing for Very Large Data Sets
- Tree-based Methods
- High-Dimensional Inference
- Penalized Estimation Methods
- Moderate p Asymptotics
- Inference Computation
- Unsupervised Learning
- Very Large p Asymptotics
- and selected examples discussed in the seminar

Examination relevant literature
The relevant literature is listed in the syllabus and will be discussed in class.

Please note

Please note that only this fact sheet and the examination schedule published at the time of bidding are binding and takes precedence over other information, such as information on StudyNet (Canvas), on lecturers’ websites and information in lectures etc.

Any references and links to third-party content within the fact sheet are only of a supplementary, informative nature and lie outside the area of responsibility of the University of St.Gallen.

Documents and materials are only relevant for central examinations if they are available by the end of the lecture period (CW21) at the latest. In the case of centrally organised mid-term examinations, the documents and materials up to CW 12 are relevant for testing.

Binding nature of the fact sheets:

- Course information as well as examination date (organised centrally/decentrally) and form of examination: from bidding start in CW 04 (Thursday, 27 January 2022);
- Examination information (regulations on aids, examination contents, examination literature) for decentralised examinations: in CW 12 (Monday, 21 March 2022);
- Examination information (regulations on aids, examination contents, examination literature) for centrally organised mid-term examinations: in CW 12 (Monday, 21 March 2022);
- Examination information (regulations on aids, examination contents, examination literature) for centrally organised examinations: two weeks before the end of the registration period in CW 15 (Monday, 11 April 2022).