

# A summary of courses at Uppsala University

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## 1 Courses taken

### 1.1 Undergraduate (kandidatprogram)

| Year | Period | Courses (credits, field, advanced)   |
|------|--------|--|
| 1    | 1      | Linear Algebra and Geometry I (5)<br>Introduction to Studies in Mathematics (5)<br>Single Variable Calculus M (5/10)<br>Honours Course in Mathematics (2.5/5)              |
|      | 2      | Scientific Computing I (5, Computer Science)<br>Algebra I (5)<br>Single Variable Calculus M (5/10)<br>Honours Course in Mathematics (2.5/5)                                |
|      | 3      | Linear Algebra II (5)<br>Computer Programming I (5, Computer Science)<br>Several Variable Calculus M (5/10)<br>Special Course in Mathematics II (2.5/5)                    |
|      | 4      | Logic and Proof Techniques I (5)<br>Algebra II (5)<br>Several Variable Calculus M (5/10)<br>Special Course in Mathematics II (2.5/5)<br>Affine and Projective Geometry (5) |
| 2    | 1      | Ordinary Differential Equations I (5)<br>Probability Theory I (5)<br>Computer Programming II (5, Computer Science)<br>Real Analysis (5/10)                                 |
|      | 2      | Fourier Analysis (5)<br>Inference Theory I (5)<br>Real Analysis (5/10)   |
|      | 3      | Linear Algebra III (5)<br>Stochastics (5)<br>Scientific Computing II (5, Computer Science)<br>Complex Analysis (5/10)  |

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|   | 4 | Basic Topology (5)<br>Regression Analysis (5)<br>Complex Analysis (5/10)   |
| 3 | 1 | Probability Theory II (5)<br>Functional Programming I (5, Computer Science, advanced)<br>Differential Geometry (5/10)<br>Multivariate Methods (5/10)                   |
|   | 2 | Inference Theory II (5)<br>Differential Geometry (5/10)<br>Multivariate Methods (5/10)   |
|   | 3 | Scientific Computing III (5, Computer Science, advanced)<br>Computer Intensive Statistics and Data Mining (5/10, advanced)<br>Degree Project C in Mathematics (7.5/15) |
|   | 4 | Computer Intensive Statistics and Data Mining (5/10, advanced)<br>Degree Project C in Mathematics (7.5/15)   |

## 1.2 Master's programme

All courses in the following table are on advanced level.

| Year | Period | Courses (credits, field)  |
|------|--------|---|
| 1    | 1      | Introduction to Data Science (5/10, Data Science)<br>Theoretical Statistics (5/10)<br>Integration Theory (5/10)   |
|      | 2      | Introduction to Data Science (5/10, Data Science)<br>Theoretical Statistics (5/10)<br>Integration Theory (5/10)<br>Generalised Linear Models (5)        |
|      | 3      | Statistical Machine Learning (5, Data Science)<br>Markov Processes (10)<br>Bayesian Statistics (5/10)   |
|      | 4      | High Performance and Parallel Computing (7.5, Computer Science)<br>Analysis of Time Series (10)<br>Bayesian Statistics (5/10)                           |
| 2    | 1      | Advanced Probabilistic Machine Learning (7.5, Data Science)<br>Accelerator-Based Programming (7.5, Computer Science)<br>Data Mining (7.5, Data Science) |
|      | 2      | Analysis of Categorical Data (5)<br>Database Design II (5, Computer Science)<br>Scientific Visualization (5, Computer Science)                          |
|      | 3      | Degree Project E in Mathematics (15/30)   |
|      | 4      | Degree Project E in Mathematics (15/30)   |

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## 2 Reflections

For a mathematics student who wants more computer science and software engineering, I would recommend the following courses (roughly in this order).

1. Computer Programming I and II for the basics of programming.
2. Scientific Computing I, II, and III for converting mathematical formulations into code.
3. High-Performance and Parallel Computing for understanding how computers works 'under the hood' and improving performance.
4. Functional Programming for a more mathematical approach to programming. Not as related to software engineering, but broadens the view on what programming can be.
5. Accelerator-Based Programming for GPU programming. Mostly of interest in high-performance computing, and to a lesser extent machine learning engineering.

In addition to these courses, a course focused on the practical side of software development would be very welcome. Such a course would introduce concepts and workflows such as containerization (Docker), packaging of software, version control (git/GitHub), basic terminal commands, etc. This course would preferably be given before High-Performance and Parallel Computing.

For a Master's programme, Computer Programming I and II, and Scientific Computing I and II, should be prerequisites.

The course High-Performance and Parallel Computing currently exists in two versions, a 7.5 and a 10 credit version. For the students of a Technical Mathematics programme, the 10 credit version should be strongly considered since the language in the course is C and the 10 credit course includes an introduction to C (which I did not know when applying).

I would also have liked to take the course Parallel and Distributed Programming, but unfortunately I was not able to fit it into my study schedule. This would be of interest to students highly interested in high-performance computing.

### 2.1 Note on study pace

Most of the time, I have had a higher than 100% study pace. The courses sum up to 355 credits, which makes the average study pace close to 120%. However, this is not evenly spread out, and the study pace in individual periods range from 100% to 150%.