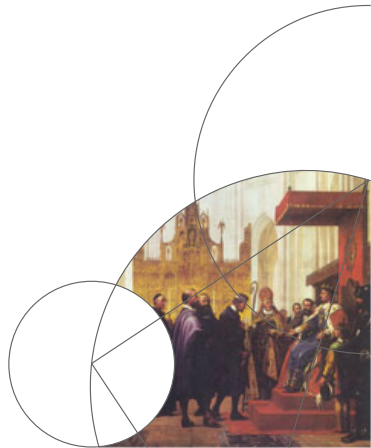




Introduction to course formats and requirements

Elements of machine learning

Jens Petersen



About us

Lecturers

- Jens Petersen, Assistant Professor, IMAGE Section
- Marleen de Bruijne, Professor, Machine Learning Section
- Silas Ørting, Postdoc, IMAGE Section



Teaching Assistants

- Nanna Munk Berg
- Casper Lisager Frandsen
- Shakir Yousefi



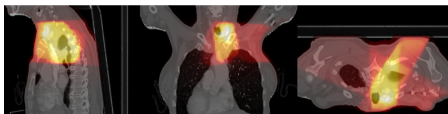
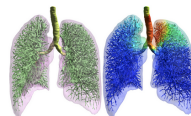
About me

Assistant Professor, IMAGE section at DIKU and Department of Oncology at Rigshospitalet

I also teach the Applied Programming course.

Research

- Medical image analysis
 - Segmentation
- (Interactive) Machine learning
- Applications
 - Radiation therapy of cancer
 - Chest imaging



About you

Please fill out the questionnaire at
<https://absalon.ku.dk/courses/56253/quizzes>



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Course is tailored to the Computer Science, data science
specialization...

We will not teach Python unless you feel there is a need for it...



Some advice from us to you

- Follow the course website on Absalon regularly
- Start in ample time with the assignments! (you will need not only time to think but also time to compute)



Format

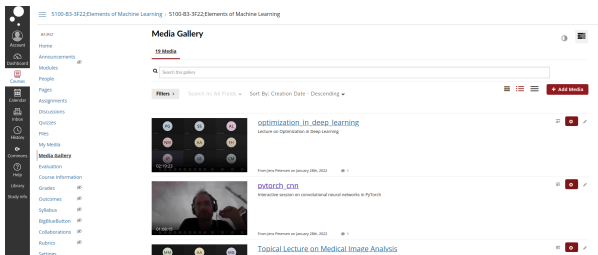
Classic format of lectures and exercise classes

- Lectures Mondays and Fridays (10-12)
- Exercise classes Tuesdays (13-17)



Lectures

- We prefer interaction, so please raise your hand and comment or ask questions.
- We do not expect recording lectures will be possible, but last years lectures can be found in the Media Gallery.
- We may decide to experiment with flipped classroom in some weeks.



Exercise classes and online help

Absalon

- Use the Absalon course room discussion board throughout the weeks.
- We check and answer questions posted here outside exercise hours.

We are considering an online option in addition to the physical exercise class, likely using Discord

- Online meeting point during and outside exercise hours.
- TA will host meetings you can join during exercise hours to discuss answers to specific questions



Assignments

- We will have 4 assignments (with handins approximately biweekly)
- Every student must submit an individual report
 - Group discussions are allowed
- Final grade is based on an average of the points achieved on each assignment
 - Preliminary points will be given by TAs and final will be given by teachers at the end of the course.
 - So in principle you do not have to hand in every assignment (but it is a good idea to do so!)
- For the re-exam the grades of the handed in assignments count along with the result of the oral exam. You may resubmit assignments prior to the re-exam.



Assignments - what is expected?

We do not want to read your code!

- Please describe what you have implemented in words, illustrations and/or possibly pseudo-code to such a degree that we could potentially implement it ourself.
- We do, however, want your code attached to the handin to allow us to settle plagiarism issues.

Describe your results!

- Do not just display bare graphs, numbers, or equations!
 - Explain what to read from the graph, how you arrived at the solution, and components of equations.



Please provide feedback

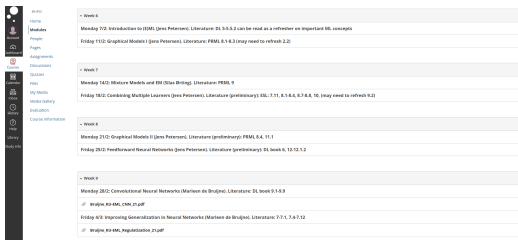
Third time the course is held

- Let us know early if something works/does not work.
- We will gather feedback both from TAs and you as the course runs and in the final weeks.



Schedule and content

An updated schedule and a list of planned content can be found on the course Absalon page.



The screenshot shows the Absalon course page for 'Introduction to Deep Learning'. The left sidebar contains navigation links: Home, Modules, Progress, Assignments, Discussions, Quizzes, Files, My Media, Media Gallery, Evaluation, and Course Information. The main content area displays the course schedule for Week 1, Week 2, Week 3, and Week 4. Each week section lists the day, time, and topic, along with the literature used.

Week 1
Monday 7:00: Introduction to DNNs (Jens Petersen). Literature: DL 5.5.5.2 can be read as a refresher on important DL concepts
Friday 11:00: Graphical Models I (Jens Petersen). Literature: PRML 6.3-6.3.3 (may need to refresh 2.2)

Week 2
Monday 14:00: Mixture Models and EM (Sara Bråting). Literature: PRML 9
Friday 18:00: Combining Multiple Learners (Jens Petersen). Literature (preliminary): EBC 7.7.1, 8.5.8.6, 8.7.8.8, 10. (may need to refresh 9.2)

Week 3
Monday 21:00: Graphical Models II (Jens Petersen). Literature (preliminary): PRML 8.4, 11.1
Friday 20:00: Feedforward Neural Networks (Jens Petersen). Literature (preliminary): DL book 6, 12.12.1.2

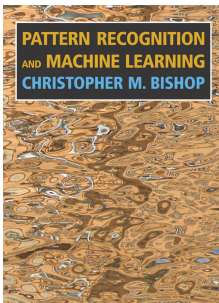
Week 4
Monday 20:00: Convolutional Neural Networks (Marlene de Bruijn). Literature: DL book 9.1-9.9
dl-bruijn_K9-EM_CNN_21.pdf
Friday 6:00: Improving Generalization in Neural Networks (Marlene de Bruijn). Literature: 7.7.1, 7.9.7.12
dl-bruijn_K9-EM_Regularization_21.pdf

Link: <https://absalon.ku.dk/courses/56253/modules>



Choice of textbooks (and why?)

No single textbook covers all of our chosen topics



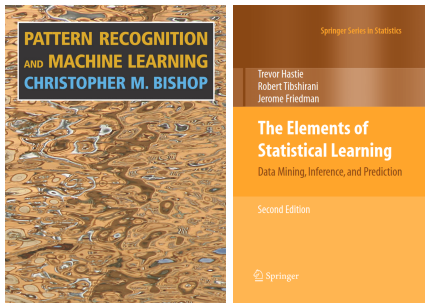
Available freely at:

- <https://www.microsoft.com/en-us/research/uploads/prod/2006/01/Bishop-Pattern-Recognition-and-Machine-Learning-2006.pdf>



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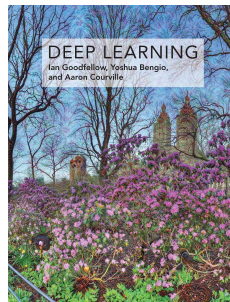
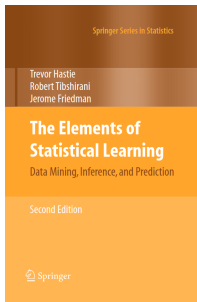
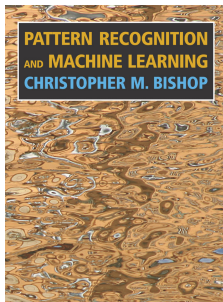
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- <http://www.deeplearningbook.org/>



Why Python?

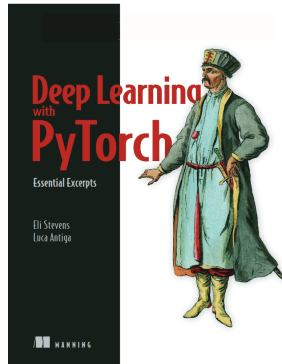
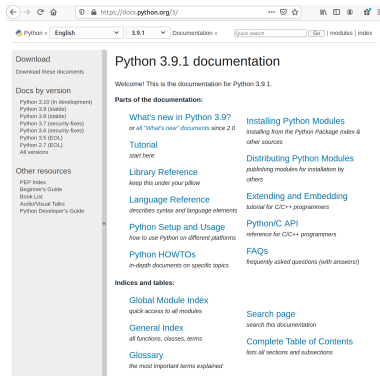
Has become the goto language for doing ML

- Keras, TensorFlow/PyTorch, Scikit-learn
- Scipy
- Pandas
- Numpy
- Matplotlib

Platform independent, big and growing community.



Python resources



Available freely at:

<https://docs.python.org/>
<https://pytorch.org/assets/deep-learning/Deep-Learning-with-PyTorch.pdf>



Scientific programming

Software and libraries you may need

- PyTorch, Scikit-learn - deep learning and machine learning
- Scipy - software for mathematics, science, and engineering
- Pandas - data manipulation, analysis and cleaning toolkit
- Numpy - high performance tools for handling multidimensional arrays
- Matplotlib - plotting library
- Jupyter - Jupyter Notebooks will be used in teaching to demonstrate python code

Consider downloading a distribution of data science libraries like <https://www.anaconda.com/distribution/> to minimize library conflicts.



Some notes on the first assignment

Now available at

<https://absalon.ku.dk/courses/56253/assignments>

- The rest of today we will be used for introducing basic ML concepts
- We will not have time to cover all of the content relevant for the first exercise before tomorrows exercise class (8/2)
 - The material for Friday (11/2) and Monday (14/2) is the most relevant for exercise 1. So consider reading this to prepare for tomorrow.
- We are planning the remaining exercises based on content that will have been covered by the time they are given.



Questions?

