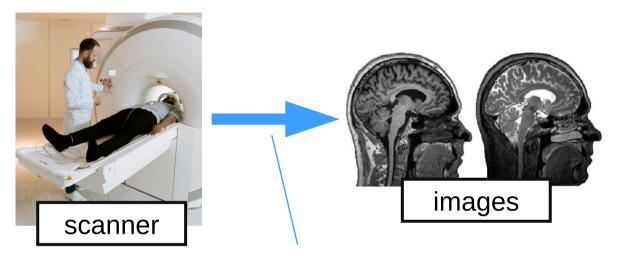
Medical Image Analysis NBE-E4010



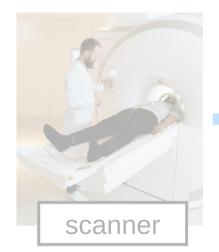
Koen Van Leemput Fall 2024

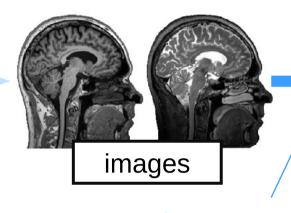
AI in medical imaging



- acquire images faster
- visualize more details

Al in medical imaging

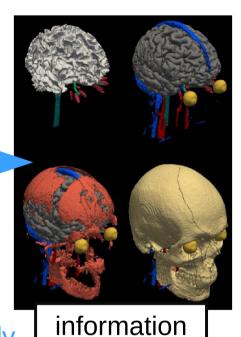




- expose the "unseeable"

- measure more consistently

- analyze images faster

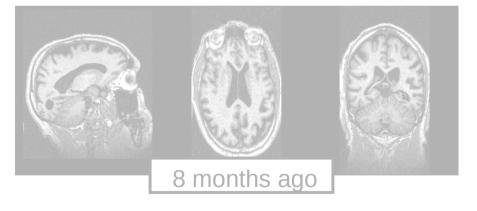


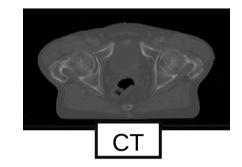
Exposing the "unseeable"

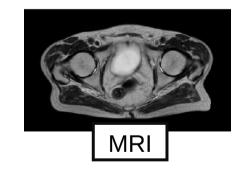


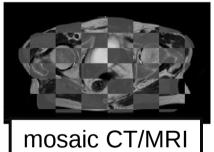
Exposing the "unseeable"





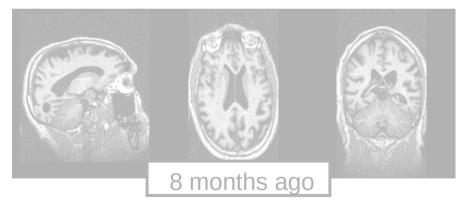


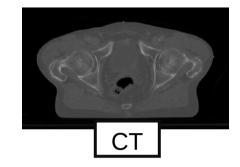


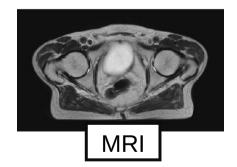


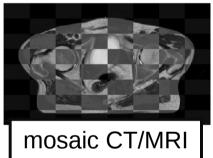
Exposing the "unseeable"







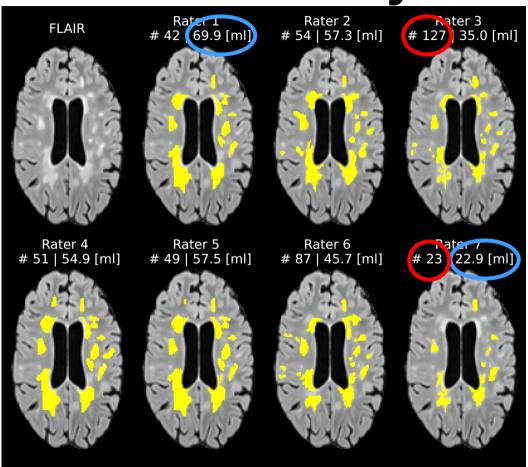




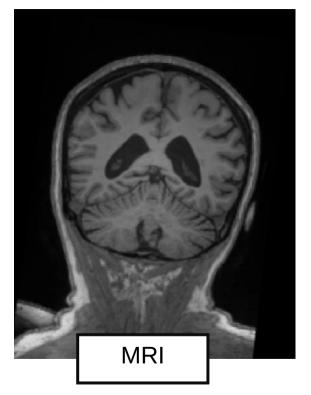
Measuring more consistently

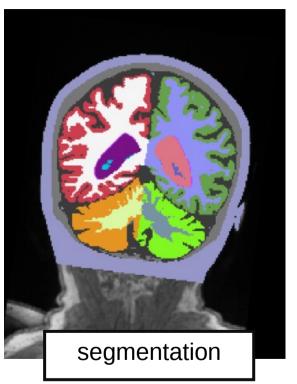
Quantifying lesions in multiple sclerosis (MS):

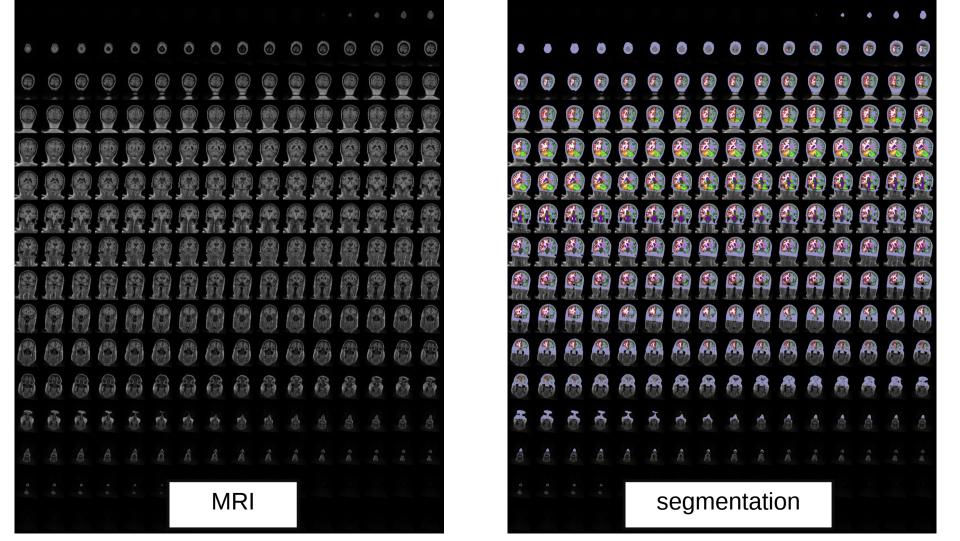
- number (#)
- volume (ml)



Analyzing images faster



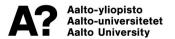




Want to know more?



Artificial Intelligence in medical imaging: From research to clinical practice - Koen Van Leemput



https://youtu.be/iiw4j-Frljo

Learning objectives

After this course you should be able to:

- ✓ Implement image smoothing and interpolation techniques
- ✓ Use spatial coordinate systems in medical images
- ✔ Perform landmark-based and intensity-based image registration
- ✓ Select the most appropriate similarity measure for specific image registration problems
- ✓ Implement rigid, affine and nonlinear spatial transformation models
- ✓ Solve segmentation problems using generative models
- Perform image segmentation using discriminative methods (neural nets)
- Weigh the advantages and limitations of generative vs. discriminative techniques in medical image analysis

A "do" course



Who are we?

- ✓ Koen Van Leemput, teacher (Professor, Neuroscience and Biomedical Engineering)
- ✓ Ida Granö, TA (Doctoral Researcher, Neuroscience and Biomedical Engineering)
- ✓ Matias Vuokko, TA (Doctoral Researcher, Neuroscience and Biomedical Engineering)
- Amir Hassankhani, TA (Doctoral Researcher, Neuroscience and Biomedical Engineering)











Who are you?

- ✓ Life Science Technologies: 21
- Exchange studies: 10
- Electrical Engineering: 6
- Aalto Doctoral Programme in Science: 4
- Science and Technology: 2
- Chemical Engineering: 1
- Engineering: 1
- Computer, Communication and Information Sciences: 1
- Chemical Technology: 1
- Non-degree Studies (Arts and Culture): 1

Physical attendance:

- not "required" in lectures or exercise sessions
- but but...



Teaching form

Lectures:

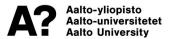
- ✓ Mondays 12.15-14.00
- Lecture recordings will be made available
- ✓ Two guest lectures: Aino Nieminen (Disior) and Eero Salli (HUS Radiology)

Exercises:

- ✓ Thursdays 12.15-14.00
- ✓ Python + Jupyter notebooks
- Group-work (max. 3 students per group)
- 2 weeks to submit a group report (6 reports in total)
- Peer grading using FeedbackFruits

Student presentations:

- Students present their fellow students' exercise reports
- ✓ Instead of the lectures on 21 Oct and 25 Nov



Teaching form

Lectures:

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Course grading:

- peer grading quality (10%)
- teachers' final score on reports (80%)
- student presentation (10%)

Student presentations:

- Students present their fellow students' exercise reports
- ✓ Instead of the lectures on 21 Oct and 25 Nov



Logistics

https://leempko.github.io/mia/

- Course material (book, slides)
- Schedule

https://mycourses.aalto.fi/

- Discussion fora



Week

2

3

Date

Mon 2

Mon 9

Thu 12

Mon 16

Thu 19

Sep

Sep

Sep

Thu 26

Mon 30

Sep

Thu 3

Mon 7

Mon 21

Oct

Oct

Oct

Sep

Sep

Sep

Sep

Activity

Lecture

Lecture

Exercise

Lecture

Exercise

Exercise

Lecture

Exercise

Lecture

Lecture

Location

F239a

F239a

F239a

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- Mon 23
- Lecture
- F239a

Nonlinear registration

Landmark-based registration

registration

- section 2.4 in the book, excluding Gauss-Newton optimization (html pdf) Intensity-based registration - slides: html pdf Mutual Information-based registration

Topic

Introduction to the course

Smoothing and interpolation

Image smoothing and interpolation

Coordinate systems, linear spatial

transformations, landmark-based

- submission deadline: Wed 9 Oct: at 23:59 - sections 2.2.2 and 2.4 in the book. especially Gauss-Newton optimization

- Announcements



- F239a
 - - Mutual Information-based registration (cont.)

Model-based segmentation I

Student presentations of the first three

- slides: html pdf - submission deadline: Wed 9 Oct: at 23:59

- slides: pdf

23:59

23:59

(html pdf)

- slides: html pdf

- slides: html pdf

- slides: html pdf

- jupyter notebook example - chapter 1 in the book (html pdf)

- submission deadline: Wed 25 Sep at

- sections 2.1-2.3 in the book (html pdf)

- submission deadline: Wed 2 Oct at

- sections 3.1-3.3 in the book (html pdf) - probability refresher; html pdf

Exercises (input data, report submissions, peer grading)

- Thu 24
- Y202a Exercise
- exercises
 - Nonlinear registration
- 23:59
 - submission deadline: Wed 6 Nov at

- Oct Oct
- Lecture
- F239a
- Model-based segmentation II
- - sections 3.4-3.5 in the book (html pdf) - slides: html pdf

Python and Jupyter Notebooks

- Python: https://lectures.scientific-python.org/
- Jupyiter Notebooks: https://www.dataquest.io/blog/jupyter-notebook-tutorial/



Image Smoothing and Interpolation

Linear regression

-0.75

Let $\mathbf{x}=(x_1,\dots,x_D)^T$ denote the spatial position in a D-dimensional space. In medical imaging, D is typically 2 or 3. Given N measurements $\{r_n\}_{n=1}^N$ at locations $\{\mathbf{x}_n\}_{n=1}^N$, a frequent task is to predict the value i at a new location \mathbf{x} . A simple model, known as lemph[linear regression], uses the function value $\mathbf{y}(\mathbf{x}_i, \mathbf{x}_i) = u_0 + u_1, \mathbf{x}_i + \dots + u_D \mathbf{x}_i + \dots + u_D \mathbf{x}_i$

as its prediction, where w_0, \dots, w_D are tunable weights that need to be estimatated from the available measurements. A more general form uses nonlinear functions of the input locations instead:

$$y(\mathbf{x}; \mathbf{w}) = w_0 + \sum_{m=1}^{M-1} w_m \phi_m(\mathbf{x}).$$

which greatly increases the flexibility of the model. Here the functions $\phi_m(\mathbf{x})$ are known as \emph{basis functions}, and it is often convenient to define an additional ``dummy'' basis function $\phi_n(\mathbf{x}) = 1$, so that the model can be written as

$$y(\mathbf{x}; \mathbf{w}) = \sum_{m=0}^{M-1} w_m \phi_m(\mathbf{x}),$$

where $\mathbf{w} = (w_0, \dots, w_{M-1})^T$ are M tunable parameters.

In order to find suitable values of the parameters of the model, the following energy can be minimzed with respect to w:

$$E(\mathbf{w}) = \sum_{n=1}^{N} \left(t_n - \sum_{n=1}^{M-1} w_m \phi_m(\mathbf{x}_n) \right)^2,$$

which simply sums of the squared distances between the measurements t_n and the model's predictions $y(\mathbf{x}_n; \mathbf{w})$.

```
In [4]: #
        import numpy as np
        from matplotlib import pyplot as plt
        plt.ion()
        ns = np.arange( N ).reshape( -1, 1 )
        A = np.cos(np.pi * (ns + 0.5) * np.arange(3) / N
        A[:, 0] *= 1/np.sqrt(2) # DC component is scaled differently
        plt.figure()
        plt.plot( ns, A )
Out[4]: [<matplotlib.lines.Line2D at 0x7fdf2fc61a60>,
         <matplotlib.lines.Line2D at 0x7fdf2fc6la90>
         <matplotlib.lines.Line2D at 0x7fdf2fc6fc70>]
          0.75
          0.50
         0.25
          0.00
         -0.25
         -0.50
```



Jupyter notebooks at Aalto

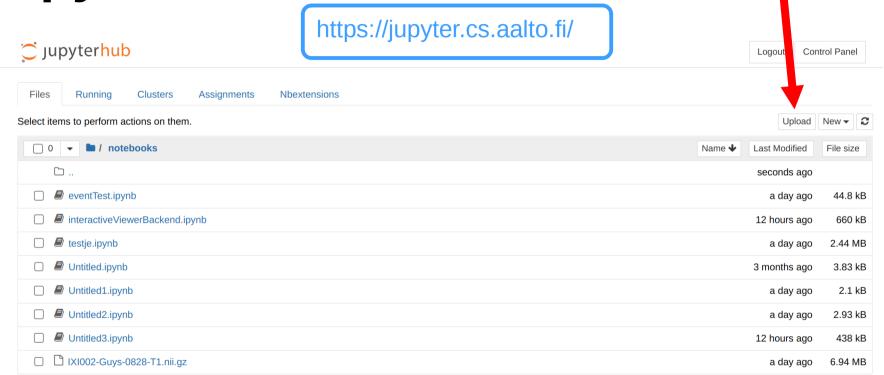
https://jupyter.cs.aalto.fi/

Server Options

- Python: General use (JupyterLab) v6.1.4
- Python: General use (classic notebook) v6.1.4
- R: General use (JupyterLab) v5.0.25-jh401
- O Julia: General use (JupyterLab) v5.0.16-jh401
- (testing) Python: General use (JupyterLab) v6.0.0
- Old version (Junyterlah) vs 0.26

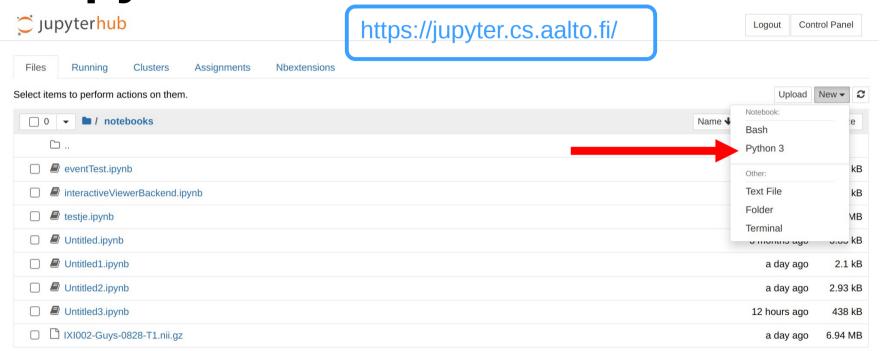


Jupyter notebooks at Aalto





Jupyter notebooks at Aalto





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