



Inverse problems and machine learning in medical physics

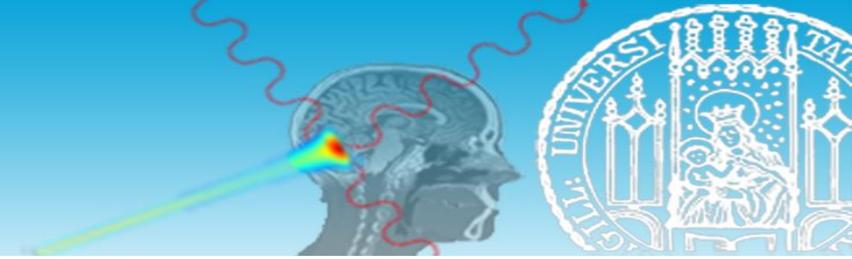
Introduction to the course

Dr. Chiara Gianoli

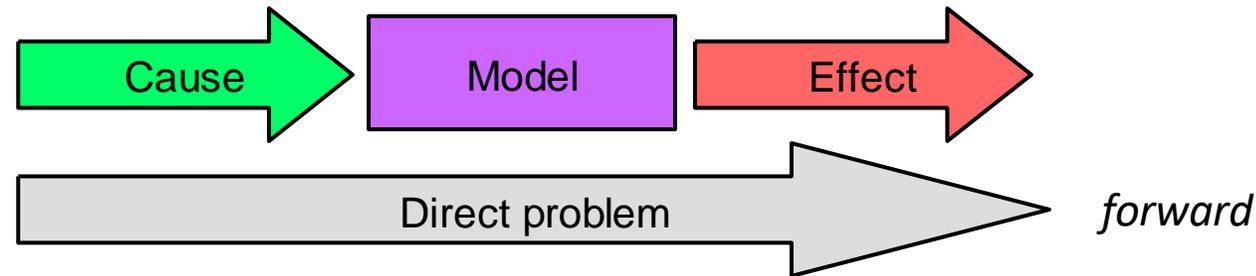
18/10/2022

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Direct and inverse problems in medical physics

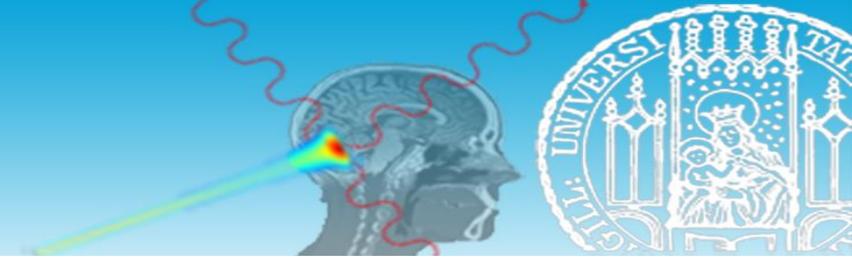


- Medical physics is a multidisciplinary field that includes computational science
- The computing power of machines is exploited to **understand** and **solve** complex problems of medical physics
- The direct problem is based on a mathematical model that links cause-effect of a certain phenomenon
 - The solution of the **direct problem** enables to explore the parameter space of the **consequences** (i.e., **understand the problem** by observing the consequences of the phenomenon)

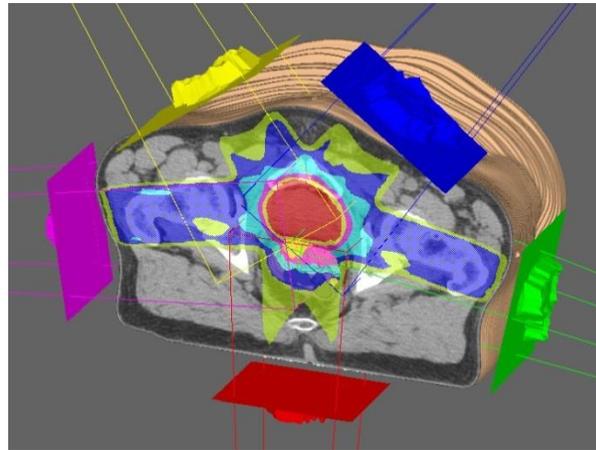
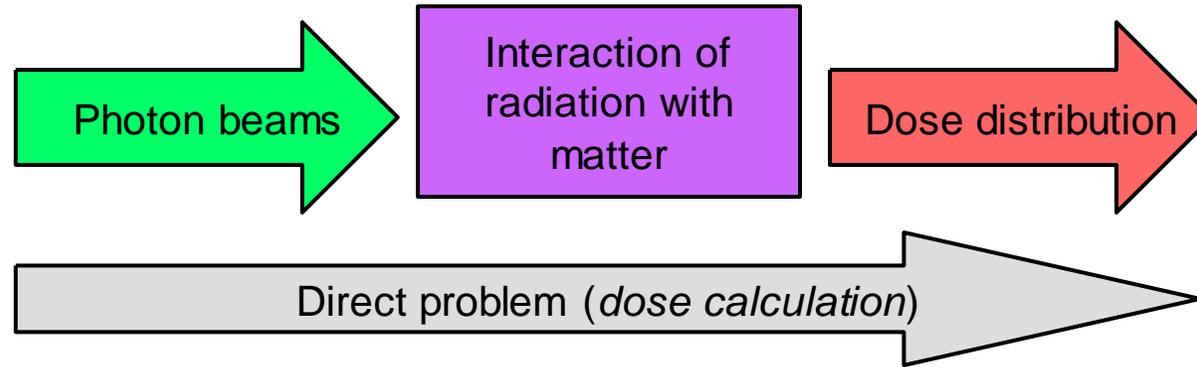


- The solution of the **inverse problem** requires the mathematical model of the direct problem that links cause-effect of a certain phenomenon to explore the parameter space of the **causes** (i.e., **solve the problem** by finding the causes of the phenomenon, when the causes are not directly observable)

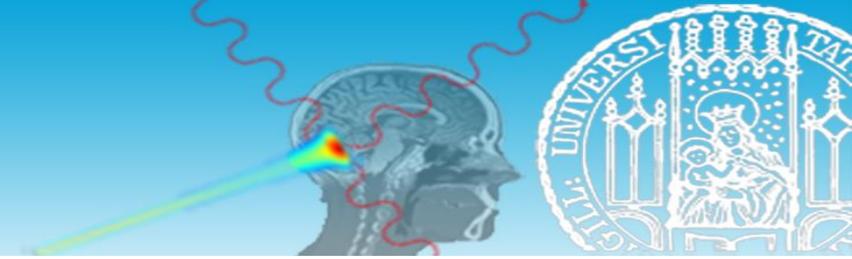
Direct and inverse problems in medical physics



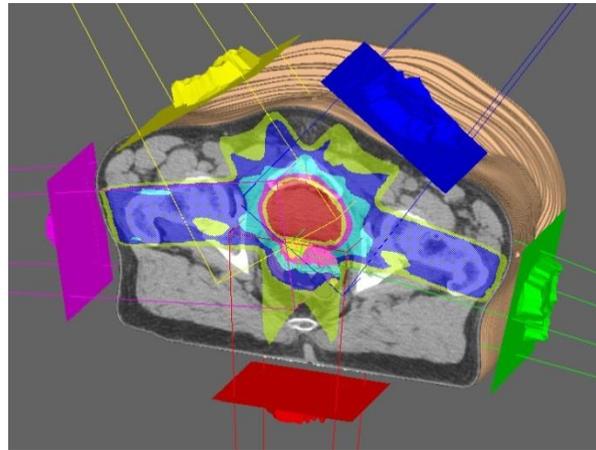
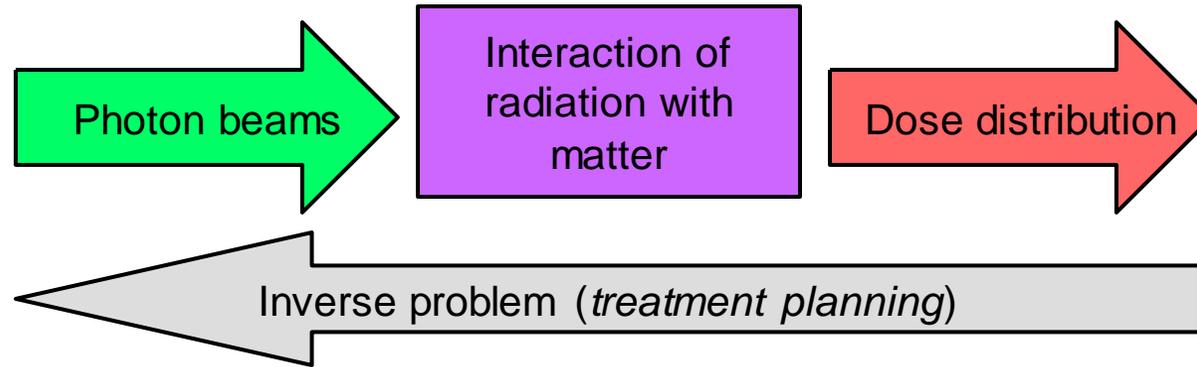
- Dose calculation is a **direct problem** in medical physics, typically based on analytical or Monte Carlo models



Direct and inverse problems in medical physics



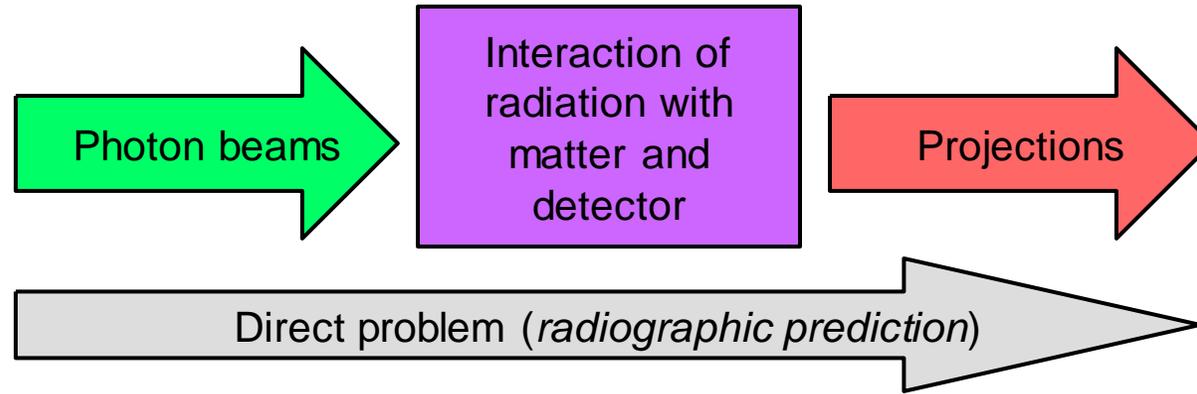
- Treatment planning is the **inverse problem** based on the analytical or Monte Carlo models of the direct problem (i.e., dose calculation)



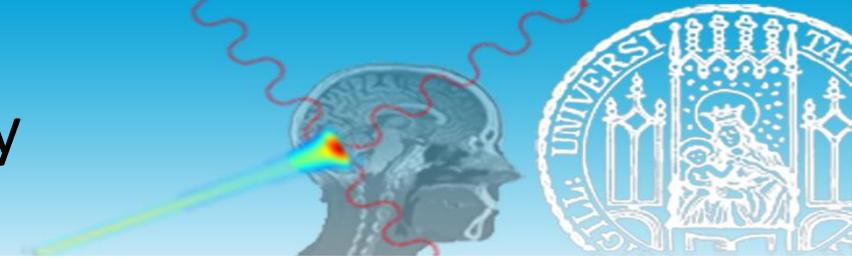
Examples in radiation oncology



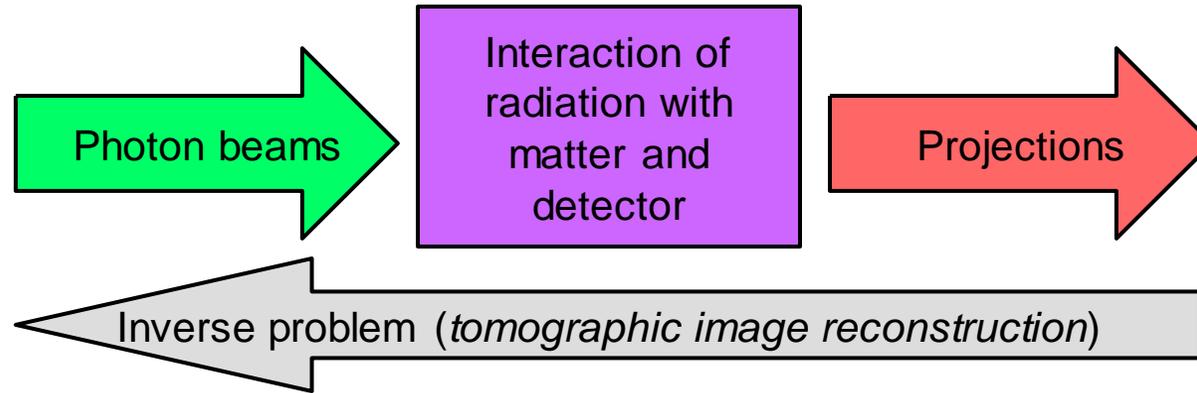
- Radiography prediction or calculation is a **direct problem** in medical physics, typically based on algebraic models



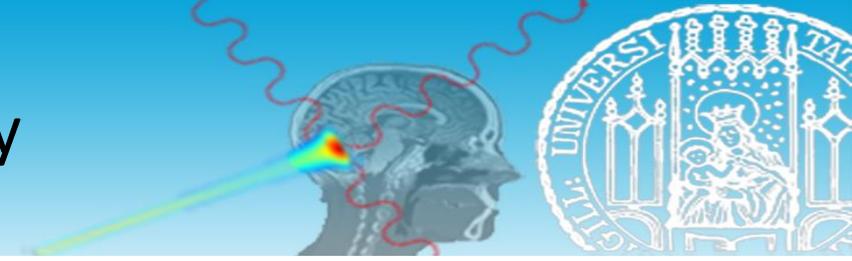
Examples in radiation oncology



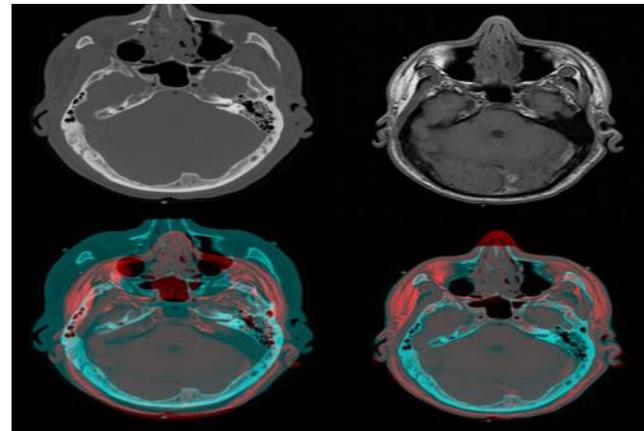
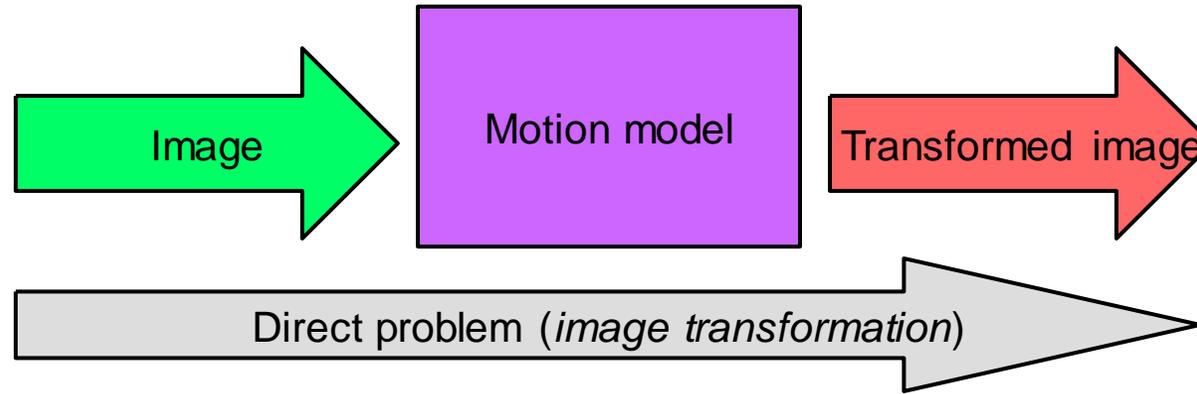
- Tomographic image reconstruction is the **inverse problem** based on the algebraic models of the direct problem (i.e., radiography prediction)



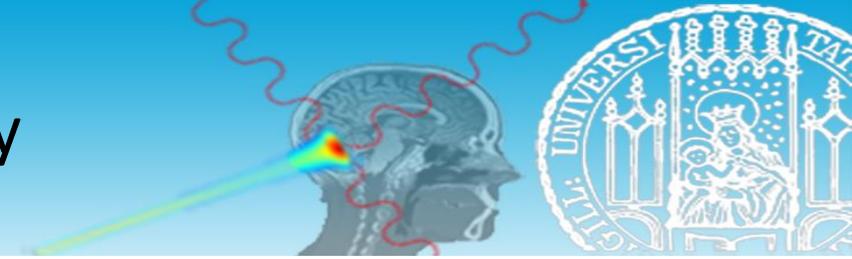
Examples in radiation oncology



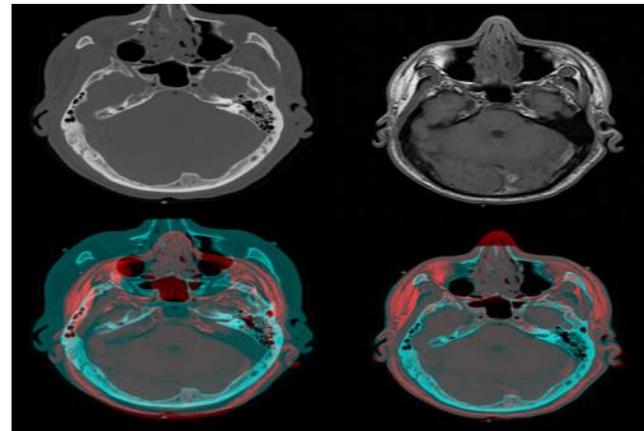
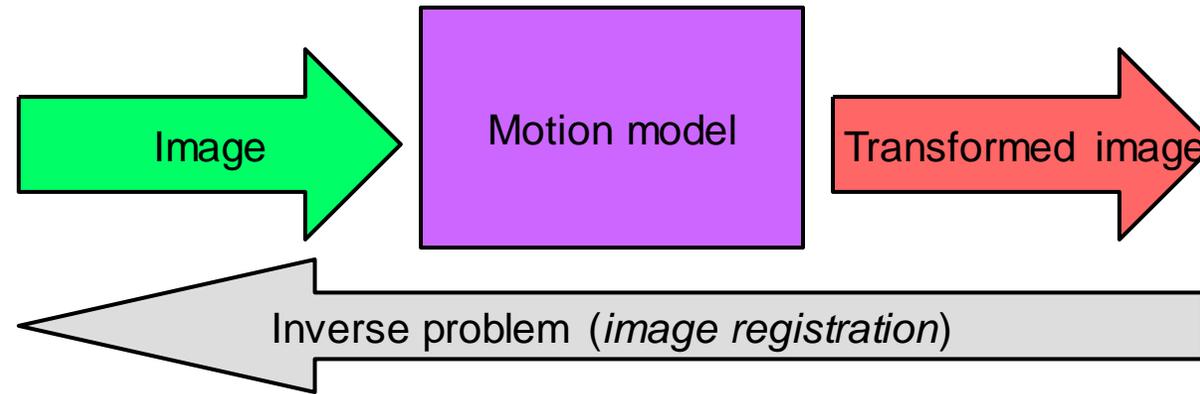
- Image transformation or deformation is a **direct problem** in medical physics, typically based on motion models



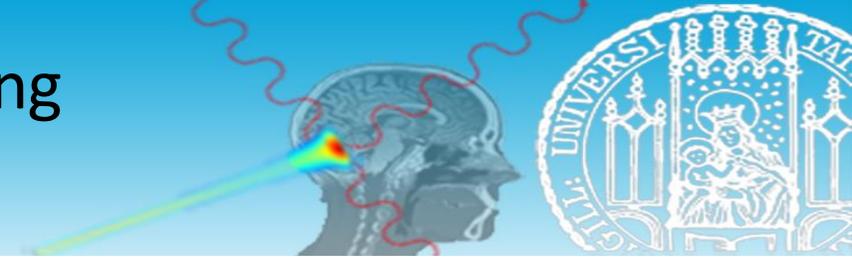
Examples in radiation oncology



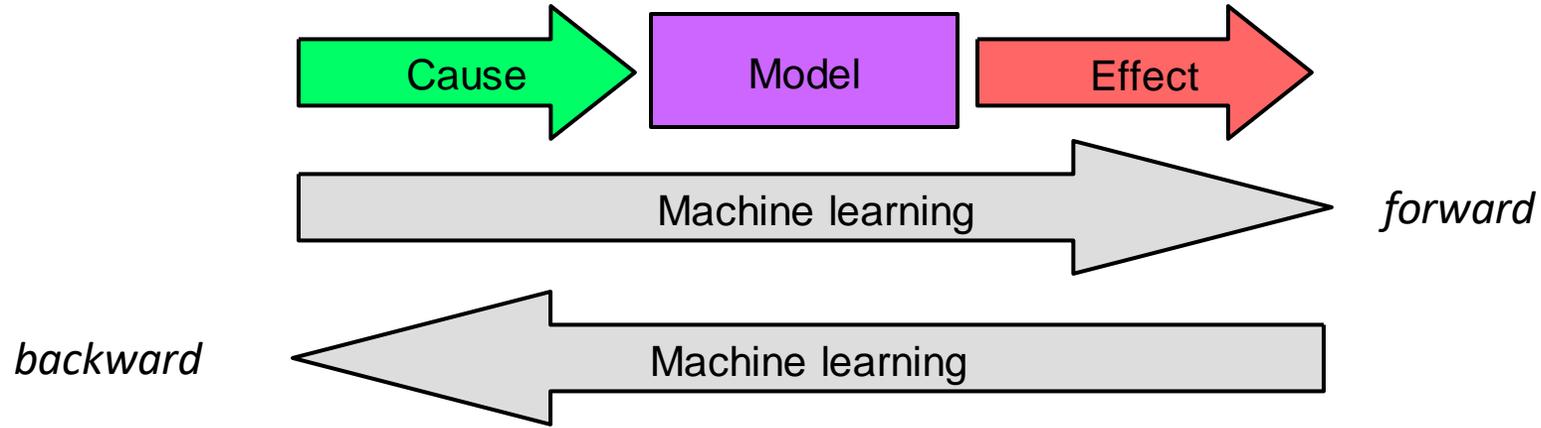
- Image registration is the **inverse problem** based on the motion models of the direct problem (i.e., image transformation)



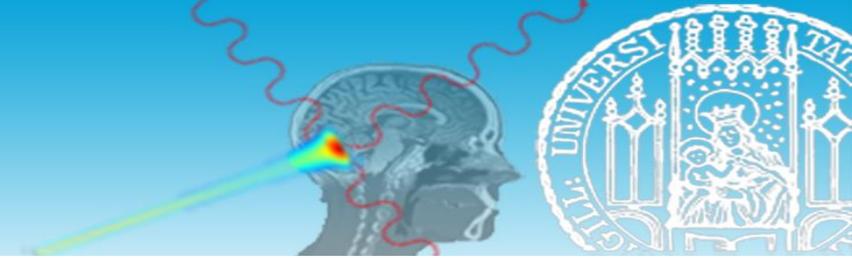
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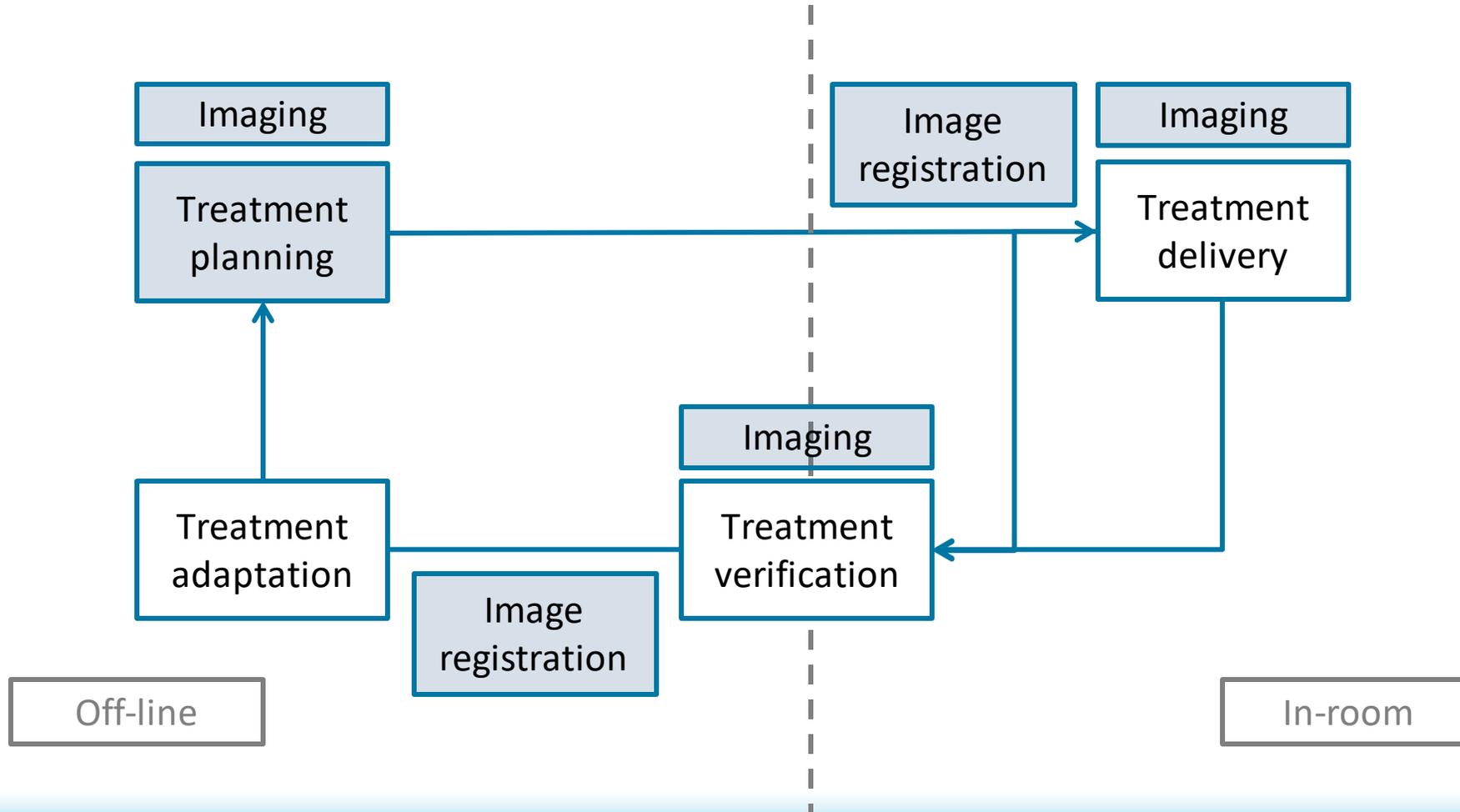
- Machine learning is based on numerical optimization
- The numerical optimization aims at finding the model that links cause-effect or effect-cause
 - to solve the direct problem when the causes are the inputs and the consequences are the targets/outputs
 - to solve the inverse problem when the consequences are the inputs and the causes are the targets/outputs with no explicit knowledge of the model of the direct problem



Topics of the lectures

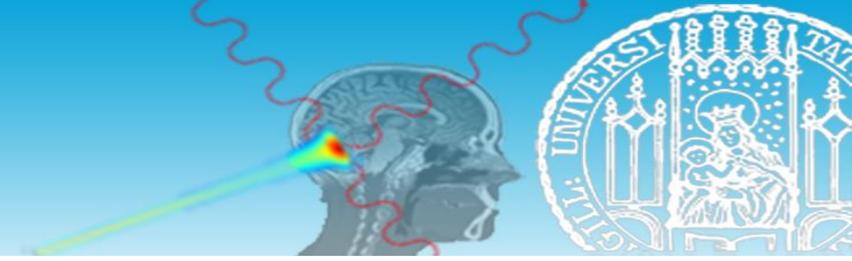


- Lectures cover topics in **radiation oncology** where inverse problems play fundamental role





Lectures



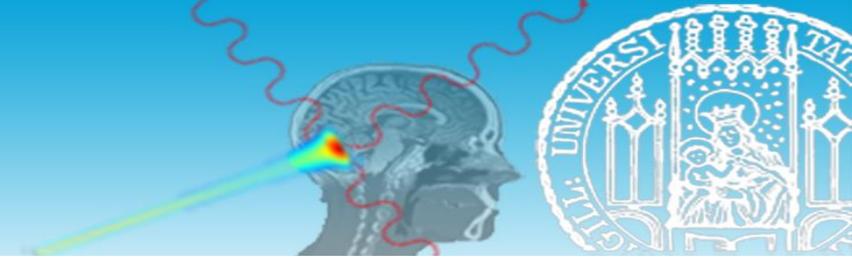
- Lectures are given in presence (*Kleiner Physiksaal*)
- Slides are uploaded on the course website (generally one day in advance)
- The lecture can be followed on line at this ZOOM link:

<https://lmu-munich.zoom.us/meeting/register/tJwsce2pqz8tEtLbIcBO5x5nK9IYFNE4pci0>

- A ZOOM registration of the lectures is available upon request (Chiara.Gianoli@physik.uni-muenchen.de)



Tutorials



- Tutorials are given in presence (*Kleiner Physiksaal*) by [Ines Butz](#) and uploaded on the course website
 - The first tutorial presents fundamentals of [Python](#) with particular focus on [linear algebra](#)
 - The second tutorial presents a guided implementation of a [numerical optimization algorithm](#) for tomographic image reconstruction
 - Step by step and “real time” coding in [Python](#) is displayed (participation with personal laptop is suggested)



- The exercise is not mandatory
- The exercise is assigned to bridge theory and practice of the course (and to get a bonus in the exam...)
- The aim of the tutorials is to support the implementation of the exercise
- The exercise consists in the implementation of a **numerical optimization algorithm** for **tomographic image reconstruction**
 - The “preparatory” algorithms can be similar to those presented in the tutorials
 - The “core” algorithm is not the one presented in the tutorials
- The code can be implemented in **Python**, **Octave (Matlab)** or **c++** (with libraries, and previous experience)



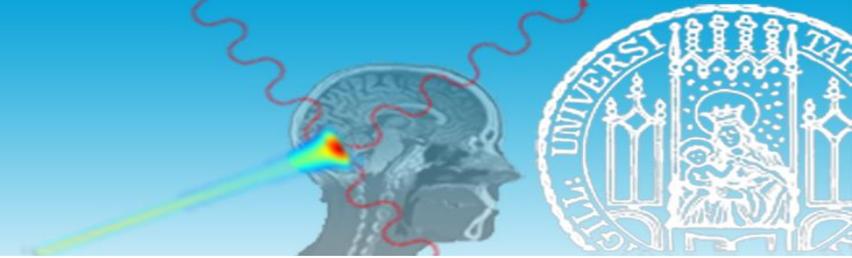
Exam



- The written exam is made by open questions (1 or 2) and multiple choice questions (9 or 10) with different weights
 - Open questions can consider pseudo-codes but no coding
 - Multiple choice question can consider calculations (a calculator is allowed but not necessarily needed...)
- The date of the exam is democratically decided via on line poll



Outline of the lectures



18.10.2022	Introduction to the course	Lecture
25.10.2022	Analytical image reconstruction	Lecture
8.11.2022	<i>Introduction to Python</i>	<i>Tutorial</i>
15.11.2022	Numerical image reconstruction	Lecture
22.11.2022	<i>Tomographic image reconstruction algorithm in Python</i>	<i>Tutorial</i>
29.11.2022	Transmission and emission imaging in radiation oncology	Lecture
6.12.2022	<i>Tomographic image reconstruction algorithm in Python – Exercise assignment</i>	<i>Tutorial</i>
13.12.2022	Ion imaging in ion beam therapy	Lecture
20.12.2022	Treatment planning and adaptive radiation therapy	Lecture
10.1.2023	Introduction to machine learning - Machine learning in tomographic image reconstruction	Lecture
17.1.2023	Machine learning in adaptive radiation therapy	Lecture
24.1.2023	Deformable image registration	Lecture
31.1.2023	Image registration in radiation oncology	Lecture
7.2.2023	Robotics in radiation oncology	Lecture
to be decided	Exam and handing in of the exercise	