

# Inverse problems and machine learning in medical physics

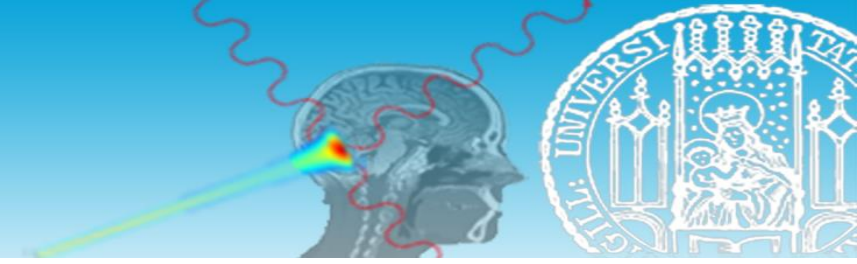
## Introduction to the course

Dr. Chiara Gianoli

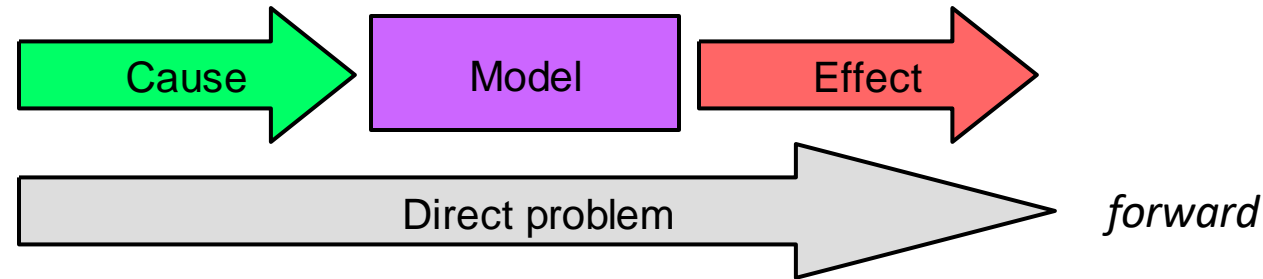
15/10/2024

[chiara.gianoli@physik.uni-muenchen.de](mailto:chiara.gianoli@physik.uni-muenchen.de)

# Direct and inverse problems in medical physics

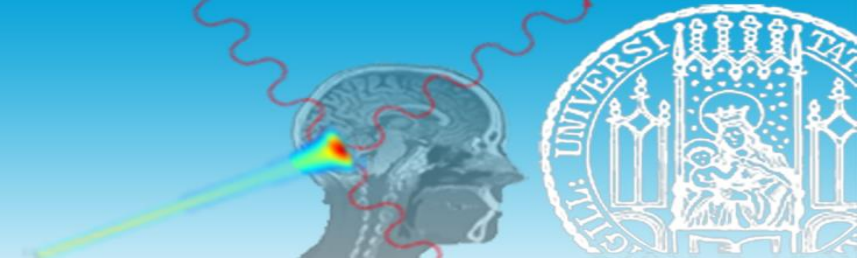


- Medical physics is a multidisciplinary field that includes computational science
- The computing power of machines is exploited to **understand** and **solve** complex problems of medical physics
- The direct problem is based on a mathematical model that links cause-effect of a certain phenomenon
  - 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)

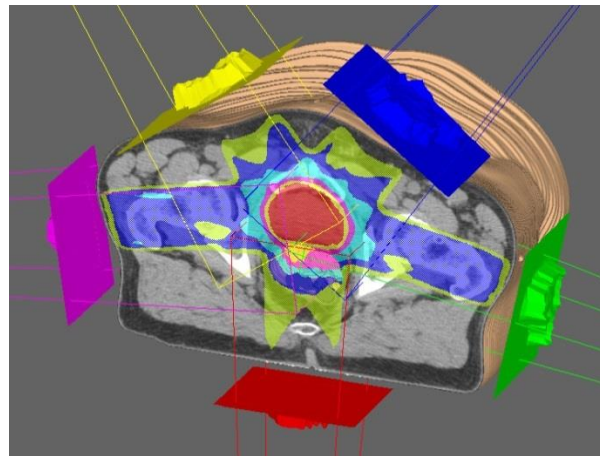
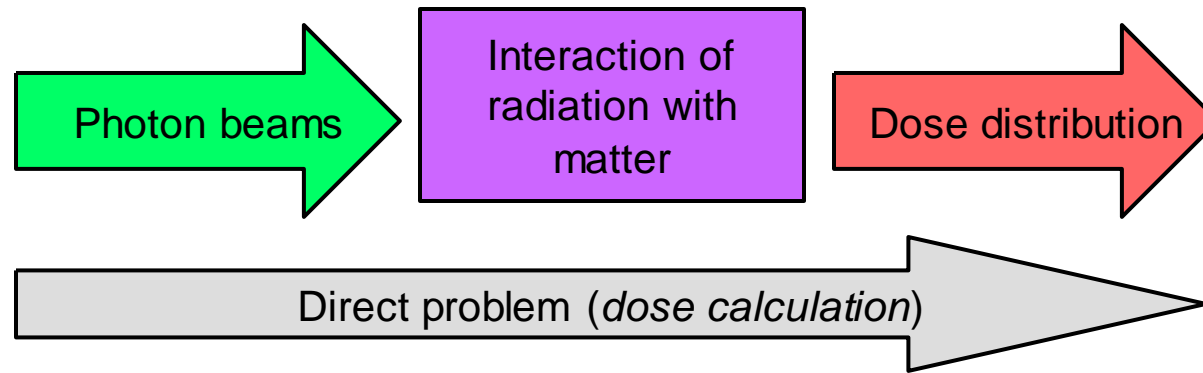


- The solution of the **inverse problem** requires the mathematical model of the direct problem that links cause-effect of a certain phenomenon to explore the parameter space of the **causes** (i.e., **solve the problem** by finding the causes of the phenomenon, when the causes are not directly observable)

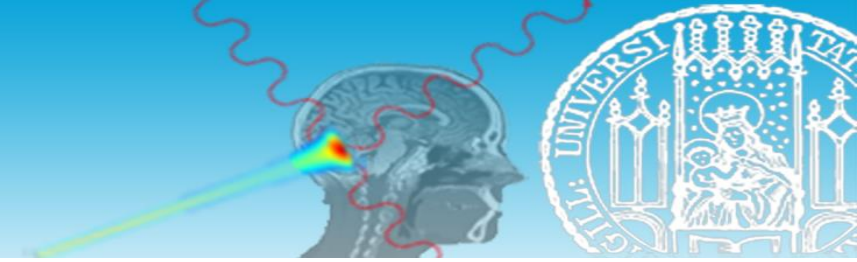
# Direct and inverse problems in medical physics



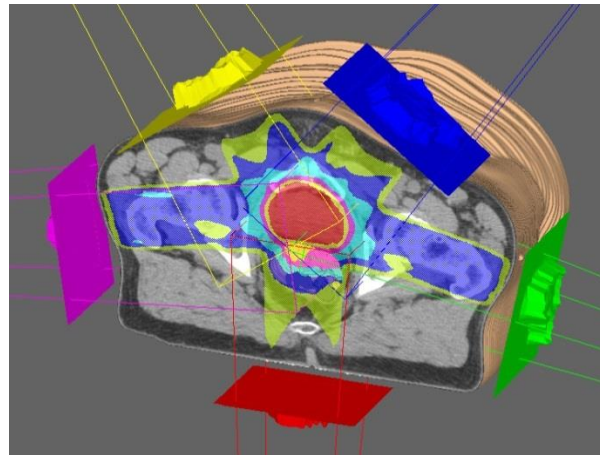
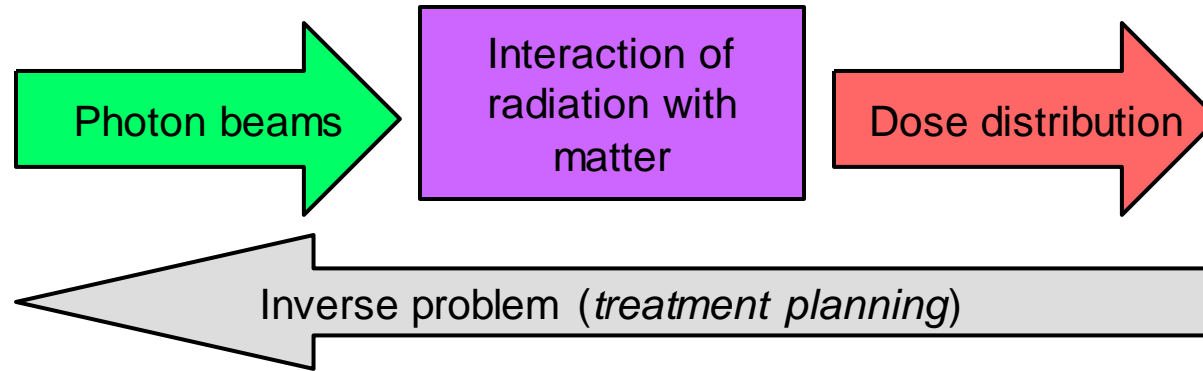
- Dose calculation is a **direct problem** in medical physics, typically based on analytical or Monte Carlo models



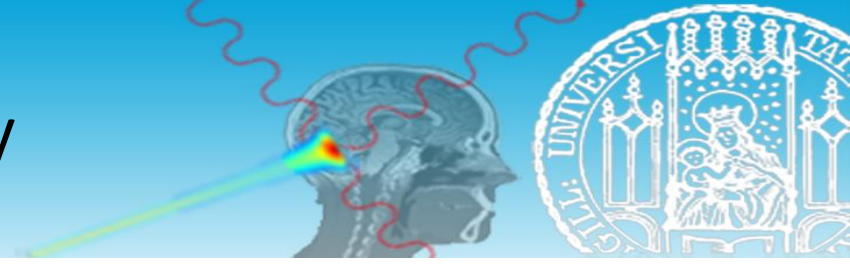
# Direct and inverse problems in medical physics



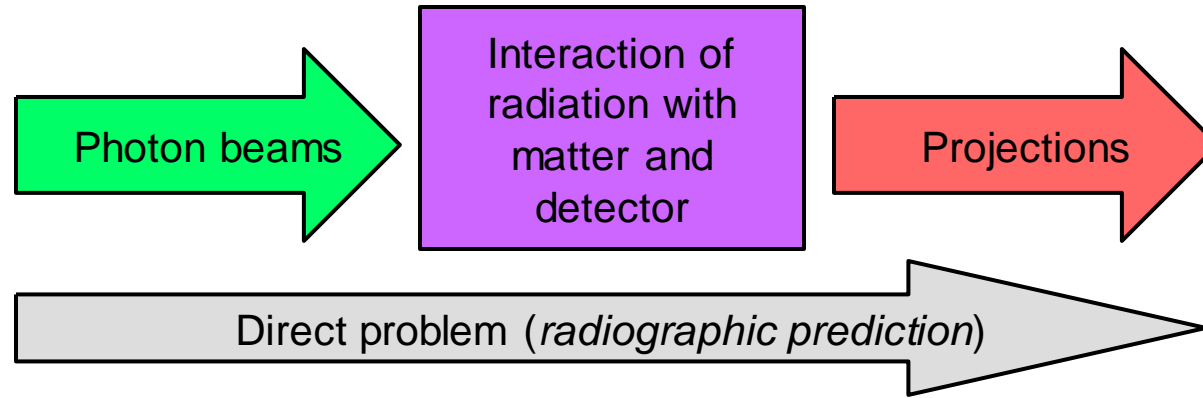
- Treatment planning is the **inverse problem** based on the analytical or Monte Carlo models of the direct problem (i.e., dose calculation)



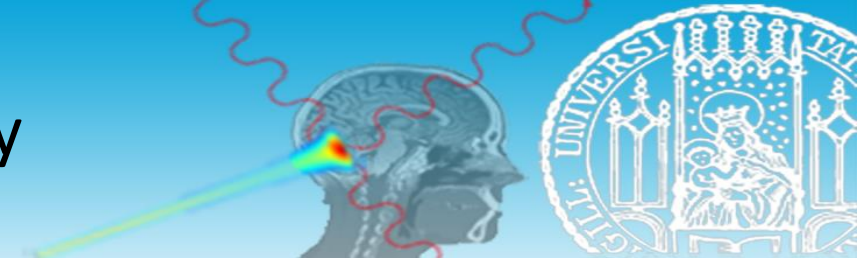
# Examples in radiation oncology



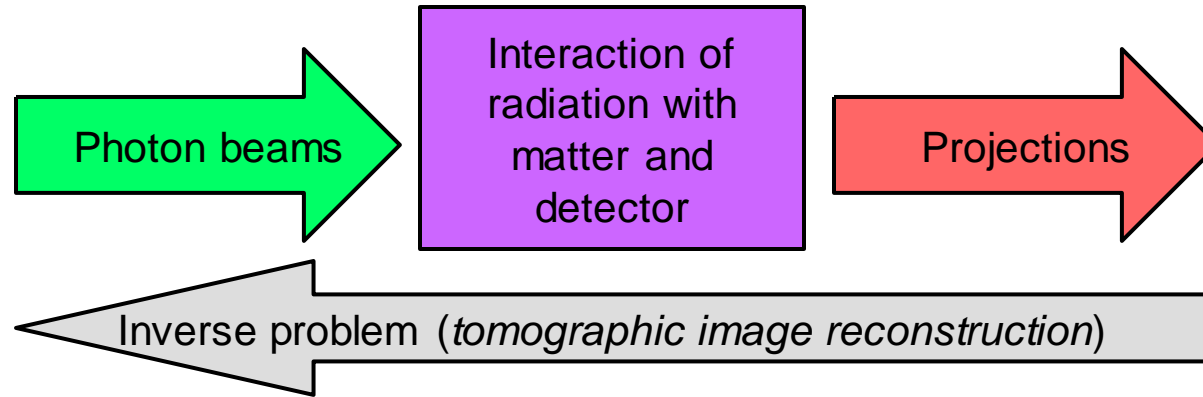
- Radiography prediction or calculation is a **direct problem** in medical physics, typically based on algebraic models



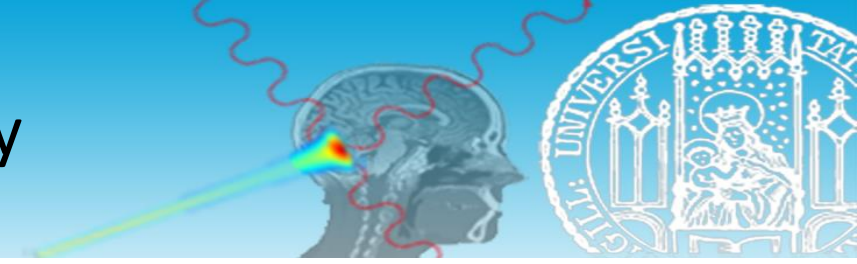
# Examples in radiation oncology



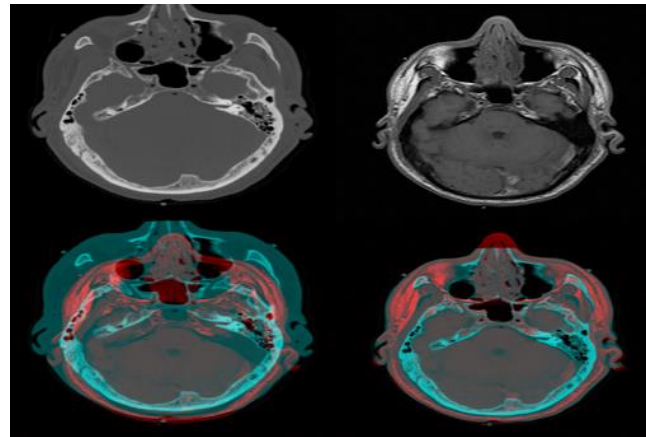
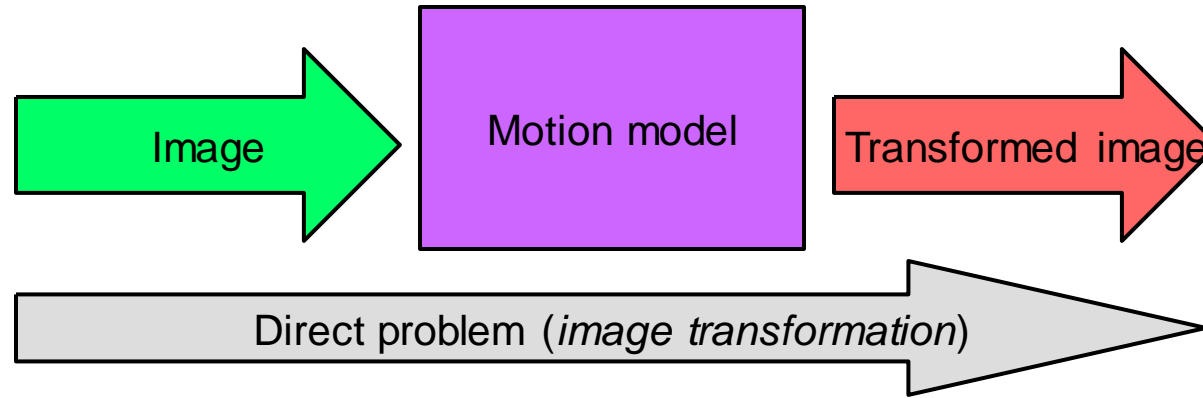
- Tomographic image reconstruction is the **inverse problem** based on the algebraic models of the direct problem (i.e., radiography prediction)



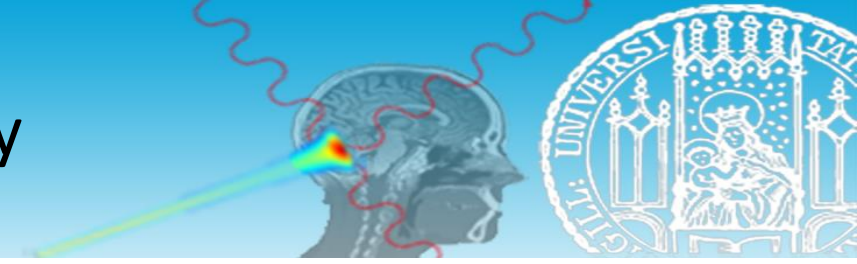
# Examples in radiation oncology



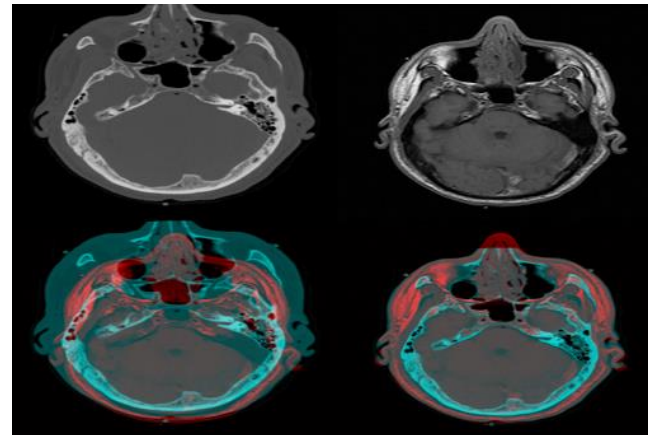
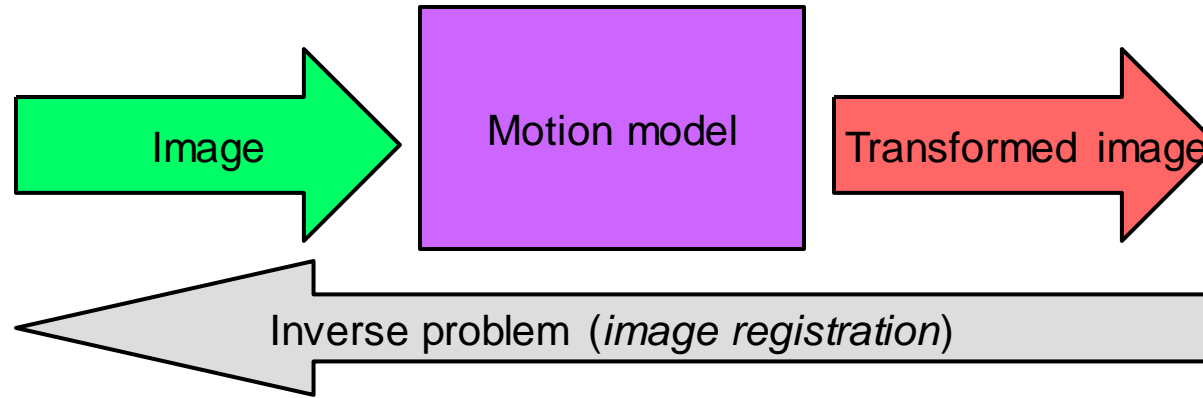
- Image transformation or deformation is a **direct problem** in medical physics, typically based on motion models



# Examples in radiation oncology

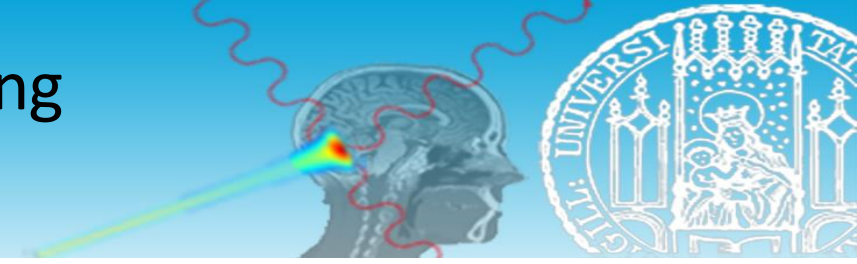


- Image registration is the **inverse problem** based on the motion models of the direct problem (i.e., image transformation)

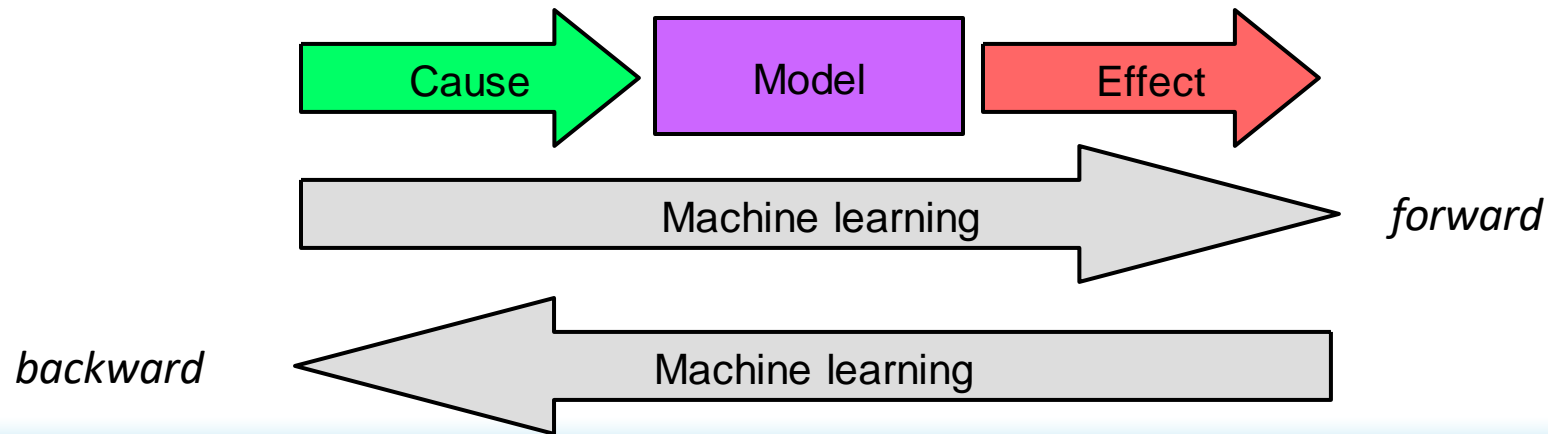




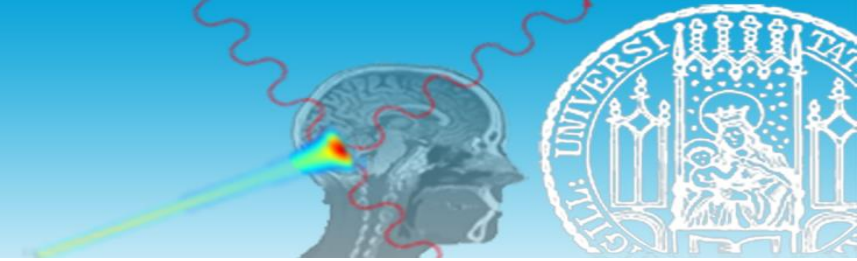
# Inverse problems and machine learning in medical physics



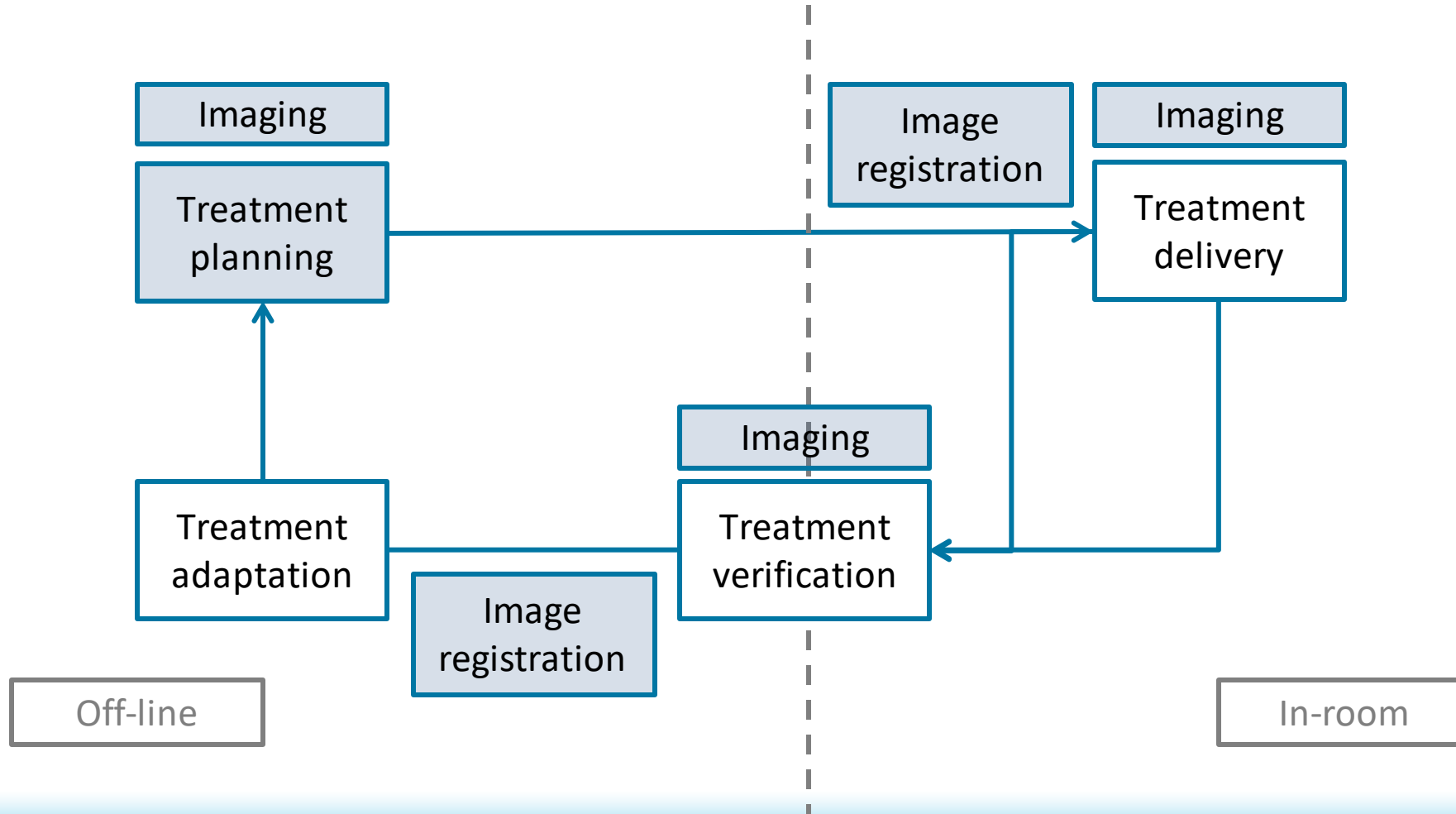
- Machine learning is based on **numerical optimization** (of the parameters of the functions that describe the neural network...)
- The numerical optimization aims at **finding the model** that links cause-effect or effect-cause
  - to solve the direct problem when the **causes** are the **inputs** 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

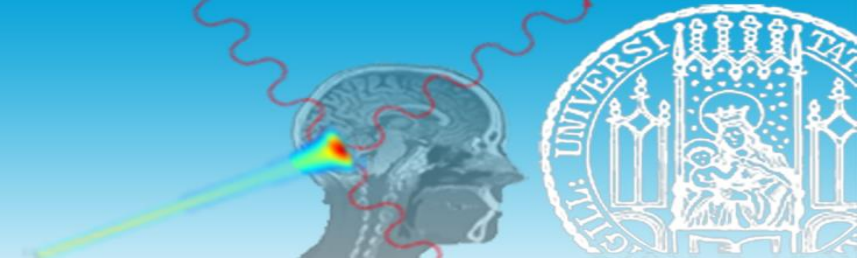


- Lectures cover topics in **radiation oncology** where inverse problems play fundamental role





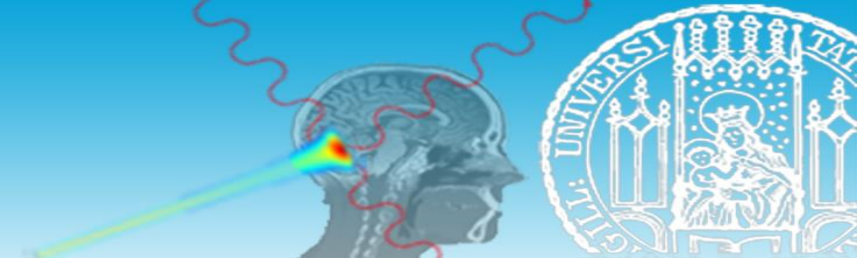
# Lectures and tutorials



- Both lectures and tutorials (by Ines Butz) are given in presence (*Kleiner Physiksaal*)
  - During tutorials, step by step and “real time” coding in [Python](#) is displayed (participation with personal laptop is suggested)
- Slides of the lectures and supporting material of the tutorials are uploaded on the course website (generally one day in advance)
- A remote access to the lectures and tutorials is available upon request ([Chiara.Gianoli@physik.uni-muenchen.de](mailto:Chiara.Gianoli@physik.uni-muenchen.de))
- No recording of the lectures and tutorials



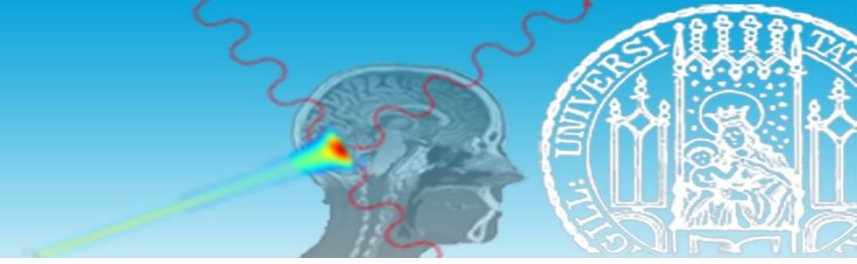
# Exercises



- The exercises are not mandatory
- The exercise are 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 exercises
- The exercises consist in the implementation of a **numerical optimization algorithm** and a **neural network 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)



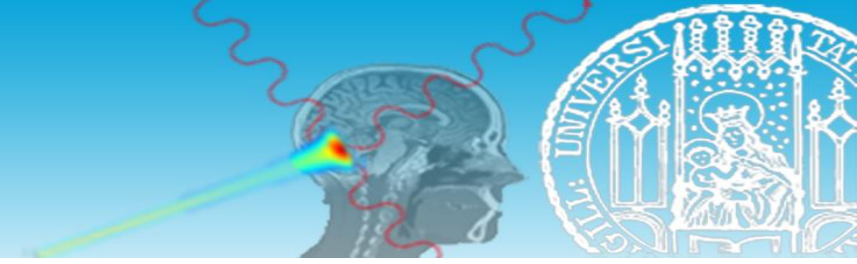
# Exam



- 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



15.10.2024	Introduction to the course	Lecture
22.10.2024	Fundamentals of tomographic imaging - Analytical image reconstruction	Lecture
29.10.2024	Numerical image reconstruction - Transmission and emission imaging in radiation oncology	Lecture
5.11.2024	<i>Introduction and imaging fundamentals in Python</i>	<i>Tutorial</i>
12.11.2024	Imaging in ion beam therapy – Ion imaging	Lecture
19.11.2024	Tomographic image reconstruction for ion imaging	Lecture
26.11.2024	<i>Tomographic image reconstruction algorithm in Python</i>	<i>Tutorial</i>
3.12.2024	Introduction to machine learning	Lecture
10.12.2024	<i>Introduction and machine learning fundamentals in Python</i>	<i>Tutorial</i>
17.12.2024	<i>Machine learning applications in Python</i>	<i>Tutorial</i>
7.1.2025	Machine learning for tomographic image reconstruction or “deep reconstruction”	Lecture
14.1.2025	Treatment planning - Machine learning in treatment planning	Lecture
21.1.2025	Image registration - Machine learning for image registration	Lecture
28.1.2025	Artificial intelligence in adaptive radiation therapy	Lecture
4.2.2025	Robotics in radiation therapy	Lecture
to be decided	Exam and handing in of the exercise	