



RESEARCH  
DATA UNIT



Scientific AI  
Hamprecht Lab, IWR,  
Heidelberg University



SCIENTIFIC  
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UNIVERSITÄT  
HEIDELBERG  
ZUKUNFT  
SEIT 1386

# AI in research software: Best practices

*Research Data Unit:*

*Scientific AI group:*

*Scientific Software Center:*

Dr. Georg Schwesinger and Dr. Sebastian Zangerle

Peter Lippmann

Dr. Inga Ulusoy

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# 1. Requirements of “ML-based science”

## *What this course is not*

- An introduction to data science
- An introduction to machine learning
- A course about different ML algorithms
- A course about different ML training approaches and libraries
- ...

# *What this course is*

- A best practices guide to creating machine learning based research software (MLBRS)
- A recommendation on how to manage and prepare your data
- A recommendation on how to train your models
- An introduction to software engineering best practices for MLBRS
- A guideline on how to generate independently reproducible scientific results using data-based approaches
- A guideline on how to publish your data and your models



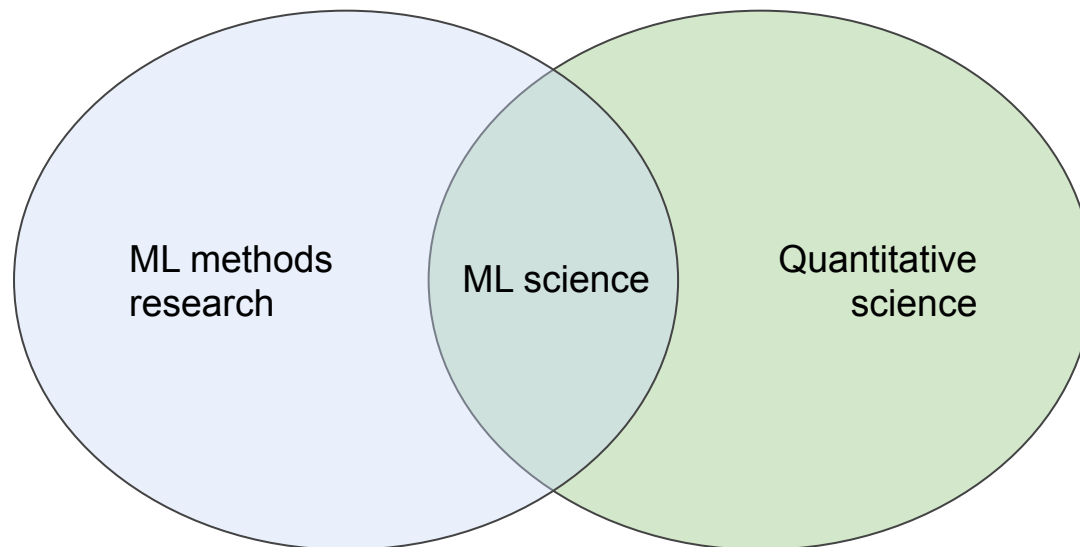
# What is special about research software based on data? (“ML-based science”)

# ML science

- Scientific research that uses machine learning models to extend scientific knowledge
- Answers a scientific question by using ML
- No restriction on algorithm, method, library, domain

*Contrary to:*

- ML methods research: Research on ML methods and algorithms with the goal to improve the field of ML



# Research software

“... software that is developed and used in the context of research...”

## **Shifting requirements**

*A scientific question is answered using computation/simulation, but the way the problem is solved changes as part of the research process.*

## **Passed along researchers**

*Initially developed for one purpose but then often organically extended depending on the researcher's needs.*

## **Development Practices**

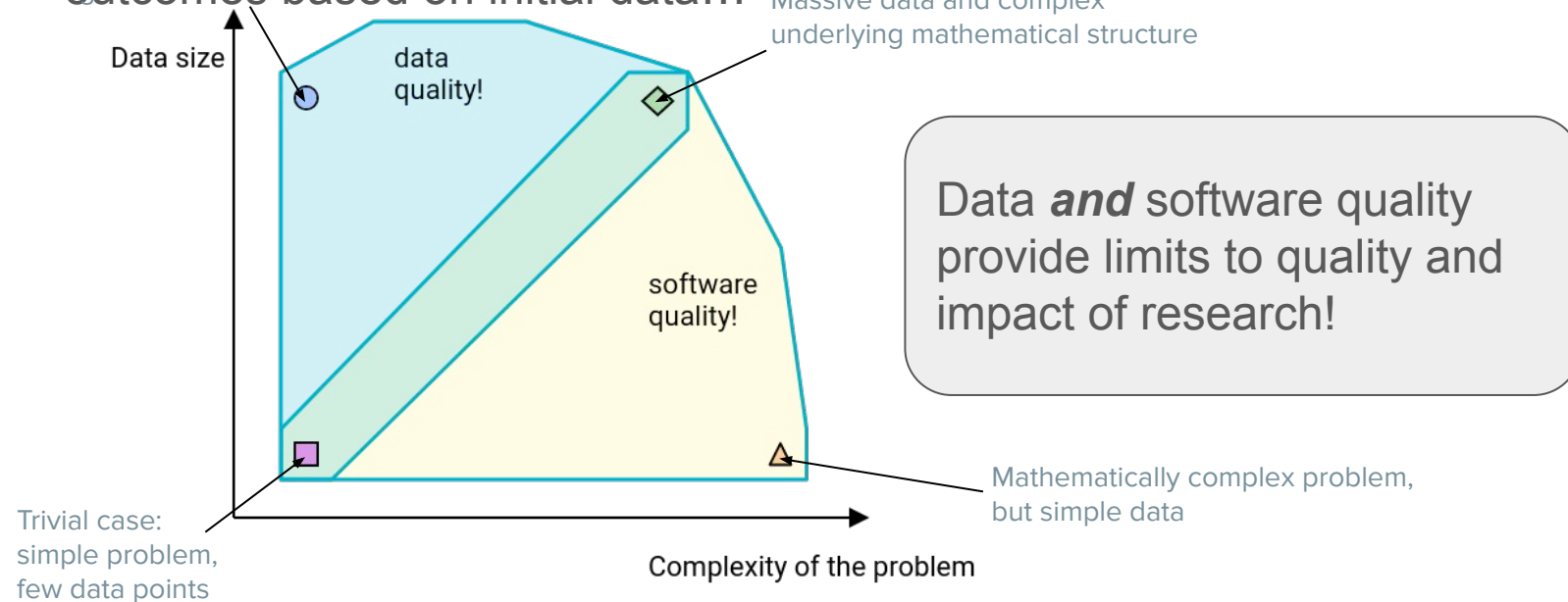
*Often created by researchers that have no fundamental training in software engineering and inherit practices from those around them.*

# ML-based research software

“... software that is developed and used in the context of research and predicts outcomes based on initial data...”

Mathematically simpler problem,  
but large amount of data

Massive data and complex  
underlying mathematical structure





# MLBRS: Data

## Data is foundation for..

*...model training, decision making and/or predictions.*

## Different kinds of data

*For example, numerical data, textual data, images, audio, video.*

## Metadata

*What is relevant metadata and should be included on the data card?*

## Availability and licensing

*Will the data be publicly available to the community? What license does/will the dataset have?*

## Legal considerations

*Where does the data come from? Is it licensed? Is it public or private data? In what form is the data stored and processed?*

## Ethical considerations

*Does the data exploit work of others? Does it break some sort of confidentiality? Will it impact in a possible harmful way or can it be misconstrued to do harm?*

## Bias

*Is there an inherent bias in the data itself, due to the data collection approach, or other reasons?*

# MLBRS: Software

## Purpose

*Will the software be more widely used, be an in-house code, or one-person software?*

## Software engineering best practices

*Does the software follow software engineering best practices (version control, testing, documentation, ...)?*

## Usability and reproducibility

*Does the software include documentation on how models can be trained, and keeps track of training parameters? Does the software help to generate model cards and provide models in transferable format?*

## Accuracy and reliability

*Does the software create robust and consistent results, even though it is based on a non-deterministic process?*

## Legal considerations

*Does the software incorporate third-party models and/or code?*

## Legal considerations

*What license is the software published under?  
What license are models published under?*

## Security

*Is the software secure against data injection?*

# Reproducibility

- Provide data to enable others to reproduce findings
- Provide code to enable others to reproduce findings

→ ***Computational reproducibility (i)***

- Make sure your findings are true findings, and do not arise from problems with your data/code

→ ***Independent reproducibility (ii)***

Research software engineering generally targets (i), but with MLBRS we target (ii)

## ***Why should you care?***

Your research integrity, scientific best conduct (malpractice), can have long-lasting detrimental effect on science (impact on others and the field), affects society!

# Key aspects

Legal aspects

Software  
quality

Reproducibility  
of the model's  
predictions

Reproducibility  
of the model  
training

Documentati  
on on data  
collection,  
data  
cleaning,  
feature  
selection

Documentation on  
model training,  
hyperparameter  
tuning, model  
testing

Robustness of  
the model(s)

Ethical  
aspects

Model bias

Software  
security

Data bias

Data leakage