

# **RSE Curriculum**

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**1 WORK IN PROGRESS THIS IS NOT THE  
OFFICIAL STATEMENT OF THE  
COMMUNITY BUT THE CURRENT  
VERSION**

## 2 Why a RSE Curriculum?

The term Research Software Engineer, or RSE, emerged a little over 10 years ago as a way to represent individuals working in the research community but focusing on software development. The term has been widely adopted and there are a number of high-level definitions of what an RSE is. However, the roles of RSEs vary depending on the institutional context they work in. At one end of the spectrum, RSE roles may look similar to a traditional research role. At the other extreme, they resemble that of a software engineer in industry. Most RSE roles inhabit the space between these two extremes.

For the purpose of creating an RSE-Master Programm we identify the RSE as a person who creates or improves research software and/or the structures that the software interacts with in the computational environment of a research domain. In this spectrum we see skilled team member who may also choose to conduct own research as part of their role. But on the other end we also see paths for an RSE to specifically focus on a technical role as an alternative to a traditional research role because they enjoy and wish to focus on the development of research software.

For this task, to support research with/in the creation of digital tools, we structure this sample curriculum along three pillars (Goth et al. 2024):

- research skills: these are competencies that enable an RSE to effectively participate in the research domain.
- technical skills: these are competencies, that enable an RSE to create effective tools for research
- communication skills: these are skills that enable an RSE to effectively work and communicate with its peers and stakeholders across multiple domains.

### 3 Balancing Computer Science Fundamentals with Application Demands

Research Software Engineering is a fast-growing field and the curriculum should engender the development of both experts in RSE (Fachexperten), and multidisciplinary researchers that are capable of transferring high-level software-engineering concepts to their respective domains. This can be mapped to the typology of the German Computer Science Association (GI). (Gesellschaft für Informatik e.V. (GI) 2016) defines computer science programs with a three-fold typology:

- Type 1: Computer Science Programs: Computer Science is solely responsible.
- Type 2: Computer Science Programs with a Specific Application Area: Computer Science is responsible in coordination with the participating application discipline.
- Type 3: Interdisciplinary Programs with a Computer Science Component Equal in Weight to Other Participating Disciplines: Computer Science shares responsibility jointly with the participating disciplines.

Fully qualified computer science experts with a focus on research software fit type 1 of CS programs. Interdisciplinary researchers where the computer science background is rivalled by the domain expertise fit the type 3.

(At this point) the curriculum tries to support both aspects equally with two branches (or [tracks](#)) of the curriculum focussing on the different weights. However, both tracks share common modules and concepts such as the idea to teach RSE-specific Open Science Tooling, an advanced Software Engineering module with specific patterns and modelling techniques for RSE and dual-lab and dual-thesis ideas, where RSE-students interface with a domain or industry field to apply their cross-cutting research skills.

You can find the tracks as follows:

- Computer Science Generalist Track [here](#)
- Natural Science / Domain Science Track [here](#)

The RSE community has long used the three pillar structure for RSE specific skills. These are distributed in the modules as following:

### 3.1 Research skills

Implemented in the following components:

- [Domain Science Project](#)
- [Domain Science Wildcard Courses](#)
- [RSE Thesis](#)

### 3.2 Technical skills

Implemented in:

- [RSE Nuts and Bolts](#)
- [Data Science Foundations](#)
- [Programming Foundations](#)
- [Software Engineering Foundations](#)
- [RSE Software Engineering](#)
- [Scientific Computing Basics and High Performance Computing](#)

### 3.3 Communication skills

Implemented in:

- [RSE Management](#)
- [Domain Science Project](#) or [RSE Lab Project](#)
- [RSE Philosophy](#)
- [RSE Lecture Series](#)

## 4 Possible Job Roles for an RSE

### 4.1 Open Science RSE

Open science and FAIRness of data and software are increasingly important topics in research, as exemplified by the demand of an increasing amount of research funding agencies requiring openness. Hence, an Open Science RSE is required to have a deeper knowledge in **Research Culture (RC)** and how to distribute software publicly (**Software Reusability (SRU)**, **Software Publication (SP)**). Open Science RSEs can help researchers navigate the technical questions that come up when practising Open Science, such as:

- “How do I make my code presentable?”
- “How do I make my code citable?”
- “What do I need to do to make my software FAIR?”
- “How do I sustainably work with an (international) team on a large code base?”

Like the Data-focused RSE, they have a deep understanding of **Research Data Management (RDM)** topics.

### 4.2 Project/Community Manager RSEs

When research software projects become larger, they need someone who manages processes and people. In practice, this concerns change management for code and documentation, and community work to safeguard usability and adaptability, but also handling project governance and scalable decision-making processes. This gap can be filled by people who invest in the **Project Management (PM)**, **User Support (USERS)**, and **Team Management (TEAM)** skills.

Building a community around a research project is an important building block in building sustainable software (Segal 2009), so these RSEs play an important role, even if they do not necessarily touch much of the code themselves.

## 4.3 Teaching RSEs

RSEs interested in developing their **Teaching (TEACH)** skill can focus on teaching the next generation of researchers and/or RSEs and will play a vital role in improving the quality of research software. They need to have a good understanding of all RSE competencies relevant to their domain and additionally should have experience or training in the educational field.

## 4.4 User Interface/User Experience Designers for Research Software

Scientific software is a complex product that often needs to be refined in order to be usable even by other scientists. To facilitate this, there are people required that specialise in the **Documentation & Best Practices (DOCBB)** and probably the **Distribution (DIST)** competency with a focus on making end-user-facing software really reusable and hence FAIR. This task is supported by strong **Modelling (MOD)** skills to reason about the behaviour of potential users of the software.

# 5 Track 1: Computer Science Generalist Track

## 5.1 Suggested Study Plan

Type	Description	SWS	Sem 1	Sem 2	Sem 3	Sem 4
Lecture	RSE Lecture Series	2	3	0	0	0
Seminar	RSE Nuts and Bolts I	2	3	0	0	0
Lecture	RSE Management Lecture	2	2	0	0	0
Exercise	RSE Management Exercise	2	2	0	0	0
Lecture	Statistical Data Analysis	4	6	0	0	0
Exercise	Statistical Data Analysis Exercise	2	3	0	0	0
Lecture	Mathematical Foundations of Data Science	2	2	0	0	0
Exercise	Mathematical Foundations of Data Science Exercise	2	4	0	0	0
Seminar	RSE Philosophy	2	3	0	0	0
Lab	RSE Lab Project	8	0	12	0	0
Lecture	Scientific Computing Basics	2	0	3	0	0
Exercise	Scientific Computing Basics Exercise	2	0	3	0	0
Lecture	Distributed Systems	2	0	2	0	0
Exercise	Distributed Systems Exercise	2	0	4	0	0
Lecture	Design of Efficient Algorithms	2	0	2	0	0
Exercise	Design of Efficient Algorithms Exercise	2	0	4	0	0
Seminar	RSE Nuts and Bolts II	2	0	0	3	0
Lecture	Text2Data	2	0	0	4	0
Exercise	Text2Data Exercise	2	0	0	2	0
Lecture	High Performance Computing	2	0	0	3	0
Exercise	High Performance Computing Exercise	2	0	0	3	0
Lecture	Security, Information and Complexity	2	0	0	2	0



Type	Description	SWS	Sem 1	Sem 2	Sem 3	Sem 4
Exercise	Security, Information and Complexity Exercise	2	0	0	4	0
Seminar	Current topics in artificial intelligence	2	0	0	3	0
Seminar	RSE Software Engineering	2	0	0	3	0
Exercise	RSE Software Engineering Exercise	2	0	0	3	0
Thesis	RSE Master Thesis	0	0	0	0	30
	Total ECTS per semester		28	30	30	30

**Total Curriculum ECTS: 118**

## 5.2 Security, Information and Complexity

### 5.2.1 Introduction

This module deals with correctness, security and complexity of algorithms.

### 5.2.2 Contents

- Methods for secure and reliable transmission and processing of information, error-correcting coding methods
- Fundamentals of cryptographic systems, methods for information analysis, complexity aspects, applications
- Necessary foundations of mathematics and complexity theory are introduced alongside the topics

### 5.2.3 Learning Objectives

- Understand the mathematical foundations of secure and reliable information processing and their complexity-theoretical basis.
- Are familiar with the fundamentals of error-protected transmission and storage of data.
- Are capable of analysing the correctness, security, and complexity of methods

### 5.2.4 Examination methods

- Either a written exam (90 minutes).
- Or an oral examination (30 minutes).

**Lecture:** Security, Information and Complexity

**SWS:** 2 **ECTS:** 2

**Exercise:** Security, Information and Complexity Exercise

**SWS:** 2 **ECTS:** 4

### 5.2.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
sec_comp1	explain the necessity and methods of error-protected transmission and storage of data	Computer Science		describe and apply the taught methods to given examples	University of Potsdam	<a href="#">Link</a>
sec_comp2	analyse the correctness, security and complexity of algorithms	Computer Science		Submit a written analysis for a given algorithm	University of Potsdam	<a href="#">Link</a>

### 5.2.6 Sources & Implementations:

#### 5.2.6.1 Curricula

- [None](#)

#### 5.2.6.2 Courses

- [UP Sicherheit, Information und Komplexität](#)

### 5.2.6.3 Recommended Course Literature

- [TODO](#)

### 5.2.6.4 Programs

- [UP Computational Science Master](#)

## 5.3 Declarative Modelling

### 5.3.1 Introduction

This module deals with declarative problem-solving.

### 5.3.2 Contents

Declarative problem-solving methods use general problem-solving methods to automatically solve combinatorial (optimisation) problems. In contrast to traditional programming, no programmes are created to solve the problems; instead, the initial problems are simply modelled (formally). Today, general problem-solving systems are capable of solving problems involving several million variables. The resulting systems are now widely used in industry and the natural sciences. The course focuses on the application of declarative problem-solving methods (especially for solving scientific problems) in the context of modelling and implementation projects.

### 5.3.3 Learning Objectives

Students - are able to define and interpret the special features, limitations, terminology and doctrines in the field of declarative modelling. The students' knowledge and understanding forms the basis for the development and/or application of independent ideas in the field of declarative modelling in a research-oriented context. - have a broad, detailed and critical understanding of the latest knowledge in selected specialist areas in the field of declarative modelling - are able to apply their knowledge, understanding and problem-solving skills in new and unfamiliar situations that are related to the field of declarative modelling in a broader or multidisciplinary context.

### 5.3.4 Examination Methods

30 minute oral examination.

**Lecture:** Declarative Modelling

**SWS: 2 ECTS: 2**

**Lab:** Declarative Modelling Lab

**SWS: 2 ECTS: 4**

### 5.3.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
declarative mod 1	interpret the special features, limitations, terminology and doctrines in the field of declarative modelling	Computer Science		develop declarative models for new problems	University of Potsdam	<a href="#">Link</a>

### 5.3.6 Sources & Implementations:

#### 5.3.6.1 Curricula

- [None](#)

#### 5.3.6.2 Courses

- [UP Deklarative Modellierung](#)

#### 5.3.6.3 Recommended Course Literature

- [TODO](#)

#### 5.3.6.4 Programs

- [UP Computational Science Master](#)

## 5.4 RSE Computing

RSEs with expertise in HPC and other performance-critical computing domains specialize in optimizing code for efficient execution across various platforms, including clusters, cloud, edge, and embedded systems. They understand parallel programming models, hardware-specific optimizations, profiling tools, and platform constraints such as memory, energy, and latency. Their skills enable them to adapt software to diverse infrastructures, manage complex dependencies, and support researchers in accessing and using advanced computing resources effectively and sustainably.

### 5.4.1 Basic Scientific Computing

#### 5.4.1.1 Module Overview

This module provides an entry-level yet rigorous foundation in scientific computing for graduate students and researchers who need to **design, implement, and evaluate computational experiments**. Learners gain an awareness of the numerical underpinnings of modern simulation and data-driven research, with an emphasis on writing *reproducible, efficient, and trustworthy* code.

#### 5.4.1.2 Intended Learning Outcomes

By the end of the module participants will be able to

1. Benchmark small programs and interpret performance metrics in a research context.
2. Explain how approximation theory and floating-point arithmetic affect numerical accuracy and stability.
3. Identify when to use established simulation libraries (e.g. BLAS/LAPACK, PETSc, Trilinos) instead of custom code.
4. Write simple GPU kernels and describe the core principles of accelerator programming.
5. Submit and monitor batch & array jobs on a mid-size compute cluster.
6. Describe common HPC challenges—such as I/O bottlenecks, threading, and NUMA—and propose mitigation strategies.
7. Maintain research software through continuous benchmarking.

### 5.4.1.3 Syllabus (Indicative Content)

Week	Theme	Topics
1	<b>Benchmarking &amp; Profiling</b>	Timing strategies · micro vs. macro benchmarks · tooling overview
2	<b>Precision &amp; Approximation</b>	IEEE-754 recap · conditioning & stability · error propagation
3	<b>Scientific Libraries</b>	BLAS/LAPACK anatomy · hierarchical I/O libraries · overview of PETSc/Trilinos/Hypre
4	<b>GPU Primer</b>	Kernel model · memory hierarchy · CUDA/OpenCL/PyTorch lightning intro
5	<b>Working on a Cluster</b>	Slurm basics · job arrays · job dependencies · simple Bash launchers
6	<b>HPC Pitfalls</b>	I/O throughput · thread oversubscription · NUMA awareness
7	<b>Software Maintenance</b>	Regression + performance tests · continuous benchmarking pipelines

### 5.4.1.4 Teaching & Learning Methods

Short lectures (30%) are coupled with hands-on labs (70%). Students complete **weekly notebooks** and a **mini-project** that reproduces and optimises a published computational result.

### 5.4.1.5 Assessment

Component	Weight	Details
Continuous labs	40%	Weekly graded notebooks
Final mini-project	60%	Report, code, and benchmark suite

### 5.4.1.6 Prerequisites

- Basic programming in Python, C/C++, or Julia
- Undergraduate calculus & linear algebra

#### 5.4.1.7 Key Resources

ChatGPT fantasy

**Lecture:** Scientific Computing Basics

**SWS: 2 ECTS: 3**

**Exercise:** Scientific Computing Basics Exercise

**SWS: 2 ECTS: 3**

**Lecture:** High Performance Computing

**SWS: 2 ECTS: 3**

**Exercise:** High Performance Computing Exercise

**SWS: 2 ECTS: 3**

#### 5.4.2 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
comp_m1	Benchmark and profile computational code to evaluate performance and bottlenecks	Scientific Computing	rse_tooling_2	Submit benchmark reports comparing implementations and justifying trade-offs	RSE Curriculum Draft	<a href="#">Link</a>
comp_m2	Explain and apply principles of approximation theory and numerical precision in scientific computing	Scientific Computing		Answer conceptual questions and implement small examples highlighting precision trade-offs	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
comp_module_3	Explain floating-point arithmetic and its implications for scientific accuracy and performance	Scientific Computing	comp_module_2	Provide examples showing effects of precision loss and propose mitigations	RSE Curriculum Draft	<a href="#">Link</a>
comp_module_4	Describe common simulation libraries and numerical frameworks (e.g., BLAS, LAPACK, PETSc, Trilinos)	Scientific Computing		List relevant libraries for a task and justify choice or avoidance of custom implementations	RSE Curriculum Draft	<a href="#">Link</a>
comp_module_5	Compare interpreted and compiled languages in terms of performance and suitability for computing tasks	Scientific Computing		Write code samples in both types of language and explain their performance characteristics	RSE Curriculum Draft	<a href="#">Link</a>



ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
hpc_module_batch	Rule batch and array jobs on a cluster, including job dependencies	High-Performance Computing	rse_tooling_3	Submit job scripts using SLURM or similar systems demonstrating correct use of job arrays and dependencies	RSE Curriculum Draft	<a href="#">Link</a>
hpc_module_identify	Identify and manage common challenges in HPC systems (e.g., I/O bottlenecks, threading, NUMA memory)	High-Performance Computing	hpc_module_1	Provide performance logs and interpret bottlenecks in a real or simulated HPC task	RSE Curriculum Draft	<a href="#">Link</a>
hpc_module_shell	Use shell scripting (e.g., Bash) to automate HPC job submission	High-Performance Computing	rse_tooling_3	Submit scripts that automate the execution of HPC jobs and handle job logic	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
hpc_module_1	Understand and use the principles of accelerator programming (e.g., GPU kernels and frameworks)	High-Performance Computing		Submit a small CUDA or OpenCL program with documentation of the principles used	RSE Curriculum Draft	<a href="#">Link</a>
hpc_module_5	Maintain scientific computing software including use of continuous benchmarking	High-Performance Computing	comp_module_1	Provide benchmark and performance history for evolving versions of software	RSE Curriculum Draft	<a href="#">Link</a>

### 5.4.3 Sources & Implementations:

#### 5.4.3.1 Curricula

- [EUMaster4HPC](#)

#### 5.4.3.2 Courses

- [Viral Instructions Hardware](#)
- [HPC Computing](#)

#### 5.4.3.3 Recommended Course Literature

- [What every computer scientist should know about floating-point arithmetic](#)

#### **5.4.3.4 Programs**

- [HPC-carpentry](#)

## **5.5 Current Topics in Artificial Intelligence**

### **5.5.1 Introduction**

This module deals with recent research in artificial intelligence.

### **5.5.2 Contents**

The seminar addresses current research questions in the field of artificial intelligence.

### **5.5.3 Learning Objectives**

Students - are able to define and interpret the special features, limitations, terminology and doctrines in the field of artificial intelligence. The knowledge and understanding gained by students forms the basis for the development and/or application of independent ideas in the field of artificial intelligence in a research-oriented context - have a broad, detailed and critical understanding of the latest state of knowledge in selected specialist areas in the field of artificial intelligence - are able to apply their knowledge, understanding and problem-solving skills in new and unfamiliar situations that are related to the field of artificial intelligence in a broader or multidisciplinary context.

### **5.5.4 Examination Methods**

- 30 minute presentation with discussion

**Seminar:** Current topics in artificial intelligence

**SWS:** 2 **ECTS:** 3

### **5.5.5 Module Competences**

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
ai_1	describe and differentiate current research topics in AI	Computer Science		answer questions about those research topics	University of Potsdam	<a href="#">Link</a>

### 5.5.6 Sources & Implementations:

#### 5.5.6.1 Curricula

- [None](#)

#### 5.5.6.2 Courses

- [UP Aktuelle Themen der Künstlichen Intelligenz](#)

#### 5.5.6.3 Recommended Course Literature

- [TODO](#)

#### 5.5.6.4 Programs

- [UP Computational Science Master](#)

## 5.6 Formal Methods in Software Engineering

### 5.6.1 Introduction

This module covers advanced topics in Software Engineering with a focus on formal methods for specification, modeling, and verification.

### 5.6.2 Contents

The module includes the following topics:

- Software Quality Assurance: Formal methods for specifying and verifying system properties.
- Service Engineering: The role of formal methods in service-based architectures.
- System Design: Use of formal methods in system design, focusing on specification and verification.

### 5.6.3 Learning Objectives

- Understand and apply formal methods in system design and software engineering.
- Analyse theoretical and practical problems in modeling and implementation using formal methods.

### 5.6.4 Examination Methods

- Either a 90-minute written exam.
- Or a 20-30 minute oral examination.

**Lecture:** Formal Methods in Software Engineering Lecture covering formal methods in system design, software specification, and verification. **SWS: 2 ECTS: 2**

**Exercise:** Formal Methods Exercise Exercise for hands-on application of formal methods in system modeling and analysis. **SWS: 2 ECTS: 4**

### 5.6.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
formal_apply	Apply formal methods in system design and software engineering.	Computer Science	Basic knowledge of software engineering.	Submit application of formal methods for a given software system.	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
formal_Analysis_2	theoretical and practical problems in modelling and implementation using formal methods together with others.	Computer Science	Basic knowledge of formal methods.	Participate in self-regulated team exercises and presentations.	University of Potsdam	<a href="#">Link</a>

## 5.6.6 Sources & Implementations:

### 5.6.6.1 Curricula

- [None](#)

### 5.6.6.2 Courses

- [UP Formal Methods in Software Engineering](#)

### 5.6.6.3 Recommended Course Literature

- [None](#)

### 5.6.6.4 Programs

- [UP Computational Science Master](#)

## 5.7 Programming Languages and Compiler Techniques

### 5.7.1 Introduction

This module provides an in-depth understanding of programming languages and compiler technology. Topics covered include lexical analysis, syntax parsing, code generation, and optimisation techniques for compilers. The course includes both theoretical lectures and practical exercises, culminating in a project where students will build a simple compiler.

### 5.7.2 Contents

- Virtualization
- Programming Languages and Design
- Software System Security

### 5.7.3 Learning Objectives

- describe paradigms and tools for the specification, development, and quality assurance of modern software systems, as well as their application in different contexts
- use various approaches in programming languages and compiler technology

### 5.7.4 Examination Methods

- either a written exam (90-100 minutes)
- or an oral examination (20-30 minutes)
- or a project report (20-30 pages)

**Lecture:** Programming Languages and Compiler Technologies

**SWS:** 2 **ECTS:** 2

**Project:** Programming Languages and Compiler Technologies Project

**SWS:** 2 **ECTS:** 4

### 5.7.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
compiler_1	apply modelling techniques	Computer Science		develop solutions through technical methods	University of Potsdam	<a href="#">Link</a>
compiler_2	analyse problems in given software systems	Computer Science		discuss problems in a team	University of Potsdam	<a href="#">Link</a>

## 5.7.6 Sources & Implementations:

### 5.7.6.1 Curricula

- [None](#)

### 5.7.6.2 Courses

- [UP Programmiersprachen und Compilertechnologien](#)

### 5.7.6.3 Recommended Course Literature

- [TODO](#)

### 5.7.6.4 Programs

- [UP Computational Science Master](#)

## 5.8 RSE Nuts and Bolts

### 5.8.1 Introduction

This module, inspired by the [MIT Missing Semester](#), addresses the “nuts and bolts” often missing from traditional academic training in computing. It aims to provide students with practical skills and conceptual understanding for building robust, maintainable, and reproducible research software—key competencies in Research Software Engineering (RSE).



### 5.8.2 General Competencies

The module begins with general-purpose computing tools and techniques that are foundational for any research software engineer:

- Shell tools and scripting
- Command-line environments
- Editors and IDEs (e.g., Vim)
- Version control (Git)
- Data wrangling
- Debugging and profiling
- Metaprogramming
- Security and cryptography

### 5.8.3 RSE-Specific Topics

Building on these foundations, the module introduces RSE-specific concepts and good practices:

- **Version control and collaboration**
  - Git for code history, collaboration, and issue tracking
- **Virtualization concepts**
  - Containerization and environment management
- **The Data Life Cycle**
  - Managing research data and understanding data provenance
- **Good coding practices**
  - Reproducible and testable code
  - Meaningful documentation and error messages
  - Modular software design
  - Performance-conscious coding
  - Easily installable and distributable software
  - Coding standards, formatting, and linting
- **Software management planning**
  - Writing Data and Software Management Plans
  - Sustainable development and community involvement
- **Low-level programming**

- Introduction to a compiled language (e.g., C) to expose hardware-level concerns and efficient memory management
- **Long-term software maintenance**
  - Version tracking, bug management, and sustainability strategies
  - Building and maintaining research software communities

#### 5.8.4 Beyond the Basics

Finally, the module touches on practices that support the scholarly nature of research software:

- Software publication and citation (see SP in (Goth et al. 2024))
- Use of domain-specific repositories and registries (see DOMREP in (Goth et al. 2024))

By the end of this module, students will be well-equipped to design, develop, document, and maintain research software that meets high standards of quality, sustainability, and reproducibility.

The module is made up of two seminars that the students take at different stages in their master’s program: In the first seminar during their first semester, students mainly learn new concepts and get to know essential tools, whereas the second seminar in the third semester focuses on teaching others about research software and the development process of it (see TEACH in (Goth et al. 2024)).

**Seminar:** RSE Nuts and Bolts I This is an introductory class to essential techniques an RSE needs in everyday life. **SWS:** 2 **ECTS:** 3

**Seminar:** RSE Nuts and Bolts II This is an advanced class of RSE techniques that includes a teaching component as part of the preparation for working as an RSE in interdisciplinary teams. **SWS:** 2 **ECTS:** 3

#### 5.8.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tool1	Use literate programming tools (e.g., Quarto, Marimo, Pluto.jl, Jupyter) to combine code, results, and narrative	Research Software Engineering		Submit a literate notebook or document integrating code, visualizations, and explanatory text	Workshop Participants	<a href="#">Link</a>
rse_tool2	Use Python for visualization, scripting, templating, and integration tasks	Research Software Engineering		Submit a Python project demonstrating use of libraries for visualisation, web tasks, and templating	Workshop Participants	<a href="#">Link</a>
rse_tool3	Write and use Bash scripts for automation	Research Software Engineering		Submit shell scripts automating file manipulation or computational workflows	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling_4	Apply testing, debugging, and logging techniques to ensure software reliability	Research Software Engineering	rse_tooling_2	Submit logs, test cases, and debugging documentation for a non-trivial Python or Bash project	Workshop Participants	<a href="#">Link</a>
rse_tooling_5	Use workflow management tools (e.g., CWL, Nextflow) to design scalable, reproducible pipelines	Research Software Engineering	rse_tooling_3, rse_tooling_11	Submit a reproducible workflow including metadata and input/output definitions	Workshop Participants	<a href="#">Link</a>
rse_tooling_6	Estimate resource requirements for computational tasks using profiling and benchmarking	Research Software Engineering	rse_tooling_2, rse_tooling_5	Provide resource usage profiles and discuss optimization implications	Workshop Participants	<a href="#">Link</a>
rse_tooling_7	Use package managers and virtual environments (e.g., conda, nix) to manage software dependencies	Research Software Engineering		Submit environment definitions and reproducible setup instructions for a project	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling_8	Document and package software for usability and reusability, using generators and modular design	Research Software Engineering	rse_tooling_2	Submit user and developer documentation generated with Sphinx or similar, plus a reusable code module	Workshop Participants	<a href="#">Link</a>
rse_tooling_9	Communicate technical RSE topics effectively with non-technical audiences	Research Software Engineering		Prepare and deliver a presentation or write an article explaining RSE concepts to a general audience	Workshop Participants	<a href="#">Link</a>
rse_tooling_10	Apply authentication and authorization mechanisms (e.g., LDAP, ACLs, Active Directory)	Research Software Engineering		Configure and demonstrate access control for a multi-user service or application	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling_1	Make informed decisions about tooling and infrastructure (e.g., Jupyter vs scripts, local vs HPC/cloud)	Research Software Engineering	rse_tooling_1, rse_tooling_2, rse_tooling_3	Submit a comparative analysis justifying tooling and infrastructure choices for a research project	Workshop Participants	<a href="#">Link</a>
rse_tooling_2	Engage in practice collaborative development, including version control and code review	Research Software Engineering	rse_tooling_2	Submit a project with version history and documented code reviews	Workshop Participants	<a href="#">Link</a>
rse_tooling_12	Mentor others in research software engineering practices	Research Software Engineering	rse_tooling_12	Document a mentoring session, workshop, or support activity	Workshop Participants	<a href="#">Link</a>
rse_tooling_4	Deploy and maintain web servers for research applications	Research Software Engineering	rse_tooling_2	Deploy a working web application with setup and maintenance documentation	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling	Understand and manage file systems, including local and network-attached storage	Research Software Engineering		Document storage strategies and access mechanisms in a real-world setup	Workshop Participants	<a href="#">Link</a>

## 5.8.6 Sources & Implementations:

### 5.8.6.1 Courses

- [MIT Missing Semester](#)
- [CodeRefinery](#)
- [INTERSECT Training Materials](#)
- [Digital Research Academy Materials \(Git, HPC, Reproducibility, Research Software\)](#)
- [Building Better Research Software \(SSI\)](#)
- [Docker for neuroscience \(jupyter book\)](#)

## 5.9 Design of Efficient Algorithms

### 5.9.1 Introduction

This module deals with the design of efficient algorithms.

### 5.9.2 Contents

Architecture and implementation of formal systems for developing efficient algorithms, correctness-preserving optimisation, programme verification and synthesis.

### 5.9.3 Learning Objectives

Students will be familiar with and understand basic techniques of formal programme development and its implementation.

### 5.9.4 Examination Methods

- Either a 90-minute written exam.
- Or a 30 minute oral examination.

**Lecture:** Design of Efficient Algorithms

**SWS:** 2 **ECTS:** 2

**Exercise:** Design of Efficient Algorithms Exercise

**SWS:** 2 **ECTS:** 4

### 5.9.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
eff_algorithms_1	discuss formal systems for the development of efficient algorithms and correctness-preserving optimisation	Computer Science		exemplary application of those formal methods	University of Potsdam	<a href="#">Link</a>

### 5.9.6 Sources & Implementations:

#### 5.9.6.1 Curricula

- [None](#)

#### 5.9.6.2 Courses

- [UP Entwurf effizienter Algorithmen](#)

#### 5.9.6.3 Recommended Course Literature

- [TODO](#)



#### 5.9.6.4 Programs

- [UP Computational Science Master](#)

## 5.10 RSE Philosophy

### 5.10.1 Introduction

The RSE master program is more than a computer science specialisation for researchers. People working as RSE are often involved in digitalization projects, institutional development or other non-technical tasks.

### 5.10.2 Contents

For a university level study program it is fitting that students learn an abstract high-level understanding of their field so that they can adapt technical models, communication frameworks and policy recommendations to the complex cases. For this they need a solid understanding in some of the more theoretical fields such as ...

- philosophy of science
- sociology of technology
- ethics and artificial intelligence
- human computer interaction
- digital humanities

### 5.10.3 General Competences

This module conveys competences in areas such as but not limited to ...

- conducting and leading research (NEW)
- understanding the research cycle (RC)
- interaction with users and stakeholders (USERS)

**Seminar:** RSE Philosophy This is an introductory class to ... **SWS:** 2 **ECTS:** 3

### 5.10.4 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_philosophy_1	Topology	Research Software Engineering		...		Link

### 5.10.5 Sources & Implementations:

#### 5.10.5.1 Courses

- **TODO**

## 5.11 Master's Thesis Module: Research Software Engineering Thesis

The master's thesis is the culminating component of the RSE programme. In this module, students apply the full spectrum of Research Software Engineering skills in a real-world research setting, demonstrating their ability to independently design, implement, and document a computational research contribution.

The thesis must address a research question in collaboration with a scientific or applied domain, but its core should include a substantial computational component. This may involve software development, data-intensive research, reproducibility infrastructure, or performance engineering — depending on the chosen topic and specialization.

Each thesis must be supervised jointly by:

- A domain expert (e.g., in physics, life sciences, or humanities)
- An RSE mentor (who ensures the quality and relevance of the computational contribution)

Students are expected to follow best practices in software engineering, version control, testing, and documentation. The final submission must include:

- A written thesis describing both the scientific and software contributions
- A structured, reproducible code repository
- A presentation and defense in a thesis colloquium

The colloquium serves as both a public communication exercise and a final evaluation, where students present their project and reflect on challenges and insights gained during the thesis.

**Thesis:** RSE Master Thesis

**SWS:** 0 **ECTS:** 30

## 5.12 RSE Management and Communication

### 5.12.1 Introduction

This module comprises the communication and management skills that are relevant for working in the interdisciplinary setting of RSE-professionals.

This includes but is not limited to:

- working in a team (see TEAM in (Goth et al. 2024))
- teaching RSE-basics (see TEACH in (Goth et al. 2024))
- project management (see PM in (Goth et al. 2024))
- interaction with users and other stakeholders (see USERS in (Goth et al. 2024))

### 5.12.2 Contents

- **research management**
  - research cycle
  - open science, FAIR, FAIR4RS
  - publication workflow: obstacles and embargoes
  - legal aspects of research data, e.g. GDPR
  - pseudonymization/anonymization methods for data privacy
  - public databases
- **quality control**
  - requirements engineering
  - trying goals with quality: test-driven development
  - behavior-driven development, Gherkin-Style acceptance testing
  - project folder organization
  - code review principles
- **communication and collaboration**
  - communication frameworks, e.g. AIDA, RACE, 7 C's
  - personality traits and their impact on cooperation
  - collaboration frameworks for remote work
  - realisation and benefits of pair programming and mob/ensemble programming
  - technical English writing skills: writing in issues and merge requests, code review...
  - conflict management, e.g. dealing with researchers that do not listen
  - how to address relevant stakeholders (i.e. users and SEs) with different background knowledge, experience and expectations
  - equity, diversity and inclusion principles

- **team management**

- challenges of transient teams (that only exist for 5-15 hours)
- effects of varying team sizes
- management depending on project size/type
- specialised roles in a software team
- intercultural and interdisciplinary differences
- team management methodologies
- importance of shared values in a RSE team
- dual goals: project vs. personal goal

- **time and project management**

- goal-setting
- project management methods, their strengths and weaknesses
- agile (not necessarily Scrum)
- Lean & Kanban (Small-Batch Philosophies)
- division of tasks into sub-tasks and task-dependencies
- iterative workflows
- continuous delivery
- communication with a manager/supervisor

**Lecture:** RSE Management Lecture This is an introductory lecture covering research, project and team management techniques an RSE needs in everyday life. **SWS: 2 ECTS: 2**

**Exercise:** RSE Management Exercise This is an exercise to apply and practice the taught methods with case-studys, role-plays etc. **SWS: 2 ECTS: 2**

### 5.12.3 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_13	Build and manage sustainable research software communities	Research Software Engineering, Community Engagement	rse_tooling_13	Document strategies used for user engagement, feedback, and community growth in a real project	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_9	Work in an agile software development process, including requirement gathering and iteration	Research Software Engineering		Submit a project that uses agile planning (e.g., user stories, sprints, stand-ups) and reflects on iteration outcomes	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_10	Define project scope, gather requirements, and manage stakeholder expectations	Research Software Engineering		Provide a requirements document and stakeholder communication log for a software project	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_11	Plan for software maintenance and long-term sustainability, including archiving strategies	Research Software Engineering	rse_practices_6	Submit a sustainability or exit plan describing how the software will be maintained or archived	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_maint_01	Explain particular implementation choices in a convincing manner	Research Software Engineering		Deliver a defense of chosen implementation decisions in a discussion with a domain expert who has limited technical knowledge (ideally oral examination or project presentation with potential ‘customer/user’ questions)	RSE Community	<a href="#">Link</a>
rse_maint_02	Exemplify and articulate shared team values and their impact on work	Research Software Engineering		Identify core team values and demonstrate how they influence key implementation decisions (e.g., design, communication, and collaboration)	RSE Community	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_management_03	Plan and manage projects using standard methods effectively and efficiently	Research Software Engineering		Develop a comprehensive project plan for a given project, including scope, milestones, risks, resources, and success criteria	RSE Community	<a href="#">Link</a>
rse_management_04	Discuss methods to set up a Diversity, Equity and Inclusion (DEI) framework in an RSE team	Research Software Engineering		Analyse and evaluate a DEI framework for a given project	RSE Community	<a href="#">Link</a>

## 5.12.4 Sources & Implementations:

### 5.12.4.1 Courses

- [RSE Leadership Course](#)

### 5.12.4.2 Recommended Course Literature

- [Remote Mob Programming](#)
- [Code with the Wisdom of the Crowd](#)
- [Collaboration Explained](#)
- [Team Topologies](#)

- [Technical Agile Coaching with the Samman method](#)
- [Lean Product and Process Development](#)
- [Extreme Programming Explained](#)

## 5.13 RSE Lab Module

*Applied Research Software Engineering in scientific institutions*

This lab module provides students with a hands-on opportunity to apply research software engineering principles to real-world scientific problems from the MINT disciplines (Mathematics, Informatics, Natural Sciences, and Technology). Students work on projects originating from active research contexts — such as simulations in physics, data analysis in chemistry, modeling in biology, or

**Lab:** RSE Lab Project

**SWS:** 8 **ECTS:** 12

### 5.13.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
project_1		Computer Science				Link

### 5.13.2 Sources & Implementations:

#### 5.13.2.1 Curricula

- [None](#)

#### 5.13.2.2 Courses

- [None](#)

#### 5.13.2.3 Recommended Course Literature

- [None](#)



#### 5.13.2.4 Programs

- [None](#)

**Lecture:** Mathematical Foundations of Data Science The module provides mathematical foundations in the field of Data Science. Topics include a selection from the areas of graph analysis, stochastic models, and signal analysis using wavelets. **SWS: 2 ECTS: 2**

**Exercise:** Mathematical Foundations of Data Science Exercise

**SWS: 2 ECTS: 4**

### 5.14 Statistical Data Analysis

This module focuses on the statistical study and quantitative analysis of the dependence between observed random variables (e.g., yield/production settings; lifespan/treatment type and injury type). Essential foundations for the statistical treatment of such relationships are provided by the linear regression model, which is studied in detail in the first part of the lecture. Within this framework, topics such as estimation, testing, and uncertainty quantification (analysis of variance) are addressed. In the second part, an introduction to advanced methods and approaches for examining relationships is offered, including nonlinear and nonparametric regression models. Additionally, questions of classification and dimensionality reduction are covered.

**Lecture:** Statistical Data Analysis

**SWS: 4 ECTS: 6**

### 5.15 Statistical Data Analysis

This module focuses on the statistical study and quantitative analysis of the dependence between observed random variables (e.g., yield/production settings; lifespan/treatment type and injury type). Essential foundations for the statistical treatment of such relationships are provided by the linear regression model, which is studied in detail in the first part of the lecture. Within this framework, topics such as estimation, testing, and uncertainty quantification (analysis of variance) are addressed. In the second part, an introduction to advanced methods and approaches for examining relationships is offered, including nonlinear and nonparametric regression models. Additionally, questions of classification and dimensionality reduction are covered.

**Exercise:** Statistical Data Analysis Exercise

**SWS: 2 ECTS: 3**

## 5.16 Text to Data

### 5.16.1 Contents

- Introduction to textual data and Natural Language Processing (NLP): key concepts and use cases
- Text acquisition and preprocessing: tokenisation, stemming/lemmatisation, stopwords, named entity recognition, text cleaning
- Text representations: bag-of-words, TF-IDF, word embeddings (Word2Vec, GloVe), contextualised embeddings (BERT, RoBERTa, GPT variants)
- Text mining and information extraction: key terms, topics, sentiment analysis, relation extraction
- Data-to-insight pipelines: from raw text to structured data (CSV/SQL), feature engineering for text data
- Modelling and evaluation: classification, regression, sequence models, transformers; evaluation metrics (accuracy, F1, MCC, AUC)
- Prototyping and reproducibility: experiment tracking, versioning, reproducible pipelines (Docker/Kubernetes, MLflow, DVC)
- Risk management and ethics: bias, fairness, data protection (GDPR), transparency, interpretability (LIME, SHAP)
- Deployment and practical applications: API-based models, batch vs real-time processing, monitoring
- Project work: from a research question to data collection, modelling, evaluation and documentation

### 5.16.2 Learning outcomes

- Understand how text data is collected, preprocessed, and transformed into usable formats
- Be able to select and justify different text representations
- Apply NLP methods to extract relevant information from text
- Develop, train and evaluate text-based models with attention to reproducibility and ethics
- Assess model limitations, interpretability, and risk of errors
- Build a reproducible, scalable text-data pipeline
- Communicate results clearly in reports, presentations, and prototype designs

**Exercise:** Text2Data Exercise

**SWS:** 2 **ECTS:** 2

## 5.17 Text to Data

### 5.17.1 Contents

- Introduction to textual data and Natural Language Processing (NLP): key concepts and use cases
- Text acquisition and preprocessing: tokenisation, stemming/lemmatisation, stopwords, named entity recognition, text cleaning
- Text representations: bag-of-words, TF-IDF, word embeddings (Word2Vec, GloVe), contextualised embeddings (BERT, RoBERTa, GPT variants)
- Text mining and information extraction: key terms, topics, sentiment analysis, relation extraction
- Data-to-insight pipelines: from raw text to structured data (CSV/SQL), feature engineering for text data
- Modelling and evaluation: classification, regression, sequence models, transformers; evaluation metrics (accuracy, F1, MCC, AUC)
- Prototyping and reproducibility: experiment tracking, versioning, reproducible pipelines (Docker/Kubernetes, MLflow, DVC)
- Risk management and ethics: bias, fairness, data protection (GDPR), transparency, interpretability (LIME, SHAP)
- Deployment and practical applications: API-based models, batch vs real-time processing, monitoring
- Project work: from a research question to data collection, modelling, evaluation and documentation

### 5.17.2 Learning outcomes

- Understand how text data is collected, preprocessed, and transformed into usable formats
- Be able to select and justify different text representations
- Apply NLP methods to extract relevant information from text
- Develop, train and evaluate text-based models with attention to reproducibility and ethics
- Assess model limitations, interpretability, and risk of errors
- Build a reproducible, scalable text-data pipeline
- Communicate results clearly in reports, presentations, and prototype designs

**Lecture:** Text2Data

**SWS:** 2 **ECTS:** 4

### 5.17.3 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_datascience_1	Assess comprehensive, detailed, and specialized knowledge of selected fundamentals in the field of Data Science	Data Science		Demonstrate knowledge through theoretical exams and practical assignments	University of Potsdam	<a href="#">Link</a>
gen_datascience_2	Demonstrate an in-depth understanding of selected Data Science methods	Data Science	gen_datascience_1	Apply Data Science methods in practical projects and case studies	University of Potsdam	<a href="#">Link</a>
gen_datascience_3	Analyze and assimilate data and inference problems, develop and implement solutions, and assess solution quality	Data Science	gen_datascience_2	Solve complex inference problems and present implemented solutions with evaluation	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_datascience_2	Develop new ideas and methods, weigh alternatives under incomplete information, and evaluate them considering different evaluation criteria	Data Science	gen_datascience_1	Present projects showcasing creative problem-solving and alternative evaluations under uncertainty	University of Potsdam	<a href="#">Link</a>
gen_statistics_1	Review comprehensive, detailed, and specialized understanding of the linear regression model based on the latest research	Data Science, Statistics		Apply linear regression models to practical problems and interpret results	University of Potsdam	<a href="#">Link</a>
gen_statistics_2	Understand fundamental concepts and methods of nonparametric statistics	Data Science, Statistics	gen_statistics_1	Solve problems involving nonparametric methods and explain applied techniques	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_statistics_3	Solve complex statistical data analysis problems, evaluate alternative modeling approaches according to various criteria, and use statistical software packages for analysis	Data Science, Statistics	gen_statistics_2	Develop solutions for complex data problems using appropriate statistical methods and software	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_statistics_4	Demonstrate academic competences including self-organization, planning skills (identifying work steps), scientific thinking and working techniques (developing solutions for complex questions), discussion of methods, verification of hypotheses, application of mathematical and statistical methods, and use of software packages	Data Science, Statistics	gen_statistics_2	Document project workflows demonstrating planning, analysis, evaluation, and use of statistical software tools	University of Potsdam	<a href="#">Link</a>

#### 5.17.4 Sources & Implementations:

##### 5.17.4.1 Curricula

- [Empfehlungen Masterstudiengänge Data Science](#)

##### 5.17.4.2 Courses

- [Statistical Data Analysis \(MATVMD837\)](#)
- [Mathematical Foundations of Data Science \(MAT-DSAM8B\)](#)

- [Programmieren für Data Scientists Python](#)

#### 5.17.4.3 Programs

- [UP Data Science](#)

## 5.18 RSE Lecture Series

### 5.18.1 Introduction

This lecture series covers current RSE research and provides room for guest lectures and varying focus topics.

### 5.18.2 Contents

vary between semesters and weeks, but are all related to RSE

### 5.18.3 Learning Objectives

- students get a broad image of RSE in the wild

### 5.18.4 Examination Methods

none?

**Lecture:** RSE Lecture Series

**SWS:** 2 **ECTS:** 3

### 5.18.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_0	describe RSE activities	Computer Science		name x different fields and analyse their differences		Link



## 5.18.6 Sources & Implementations:

### 5.18.6.1 Curricula

- [None](#)

### 5.18.6.2 Courses

- [None](#)

### 5.18.6.3 Recommended Course Literature

- [None](#)

### 5.18.6.4 Programs

- [None](#)

## 5.19 RSE-Software Engineering

This module extends the Classical Software Engineering Module with research specific learnings. This includes but is not limited to

- software re-use (see SRU in (Goth et al. 2024))
- creating documented code building blocks (see DOCBB in (Goth et al. 2024))
- building distributable software (see DIST in (Goth et al. 2024))
- research specific programming languages
- research specific code requirements (scalability, functional programming, ...)
- Adapting the software life cycle to research (see SWLC in (Goth et al. 2024))
- Software behaviour awareness and analysis (see MOD in (Goth et al. 2024))
- Research specific Engineering Patterns

Also, the seminar provides room for reflection and discussions of SE lab experiences.

**Seminar:** RSE Software Engineering

**SWS:** 2 **ECTS:** 3

**Exercise:** RSE Software Engineering Exercise

**SWS:** 2 **ECTS:** 3

### 5.19.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_1	Apply good coding practices including formatting, linting, and modular design	Research Software Engineering	rse_tooling_2	Submit a code project demonstrating modularity, consistent formatting, and use of linters	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_2	Write code and documentation that supports reproducibility in research	Research Software Engineering	rse_tooling_1, rse_tooling_4	Submit a project with data, software, and instructions allowing full reproduction of results	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_3	Organise files and name code artifacts using clear, consistent conventions	Research Software Engineering		Submit a software repository with a structured layout and consistent naming scheme	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_4	Version control code and collaborate using platforms like GitHub or GitLab	Research Software Engineering	rse_tooling_12	Participate in a collaborative coding project using Git-based workflows and merge requests	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_5	Write effective documentation and user-facing error messages	Research Software Engineering	rse_tooling_8	Provide documentation and example error handling demonstrating clarity and user support	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_12	Write performant code suitable for use in compute-intensive contexts	Research Software Engineering	rse_tooling_2	Submit benchmark results comparing an optimized version of code with a naive implementation	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practices_6	Publish code and software in trusted repositories and package managers	Research Software Engineering	rse_practices_6	Publish software to a repository (e.g., GitHub, PyPI, CRAN) and register it with a long-term archive (e.g., Zenodo)	RSE Curriculum Draft	<a href="#">Link</a>
rse_practices_6	Apply licensing and publishing strategies to make software reusable and citable	Research Software Engineering		Submit a software project with an appropriate open license and published DOI (e.g., via Zenodo)	RSE Curriculum Draft	<a href="#">Link</a>
rse_practices_7	Apply principles of Open Source and FAIR (Findable, Accessible, Interoperable, Reusable) software	Research Software Engineering	rse_practices_6	Review or create a software project and evaluate its compliance with FAIR/Open Source principles	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice4	Submit a data-driven project showing clear data organisation, metadata, and reproducibility within a software project in accordance with best practices	Research Software Engineering, Data Management			RSE Curriculum Draft	<a href="#">Link</a>

## 5.19.2 Sources & Implementations:

### 5.19.2.1 Courses

- [TODO](#)

## 5.20 Classical Software Engineering

*To summarise the vast range of the skills a software engineer is typically equipped with, we refer to the Guide to the Software Engineering Body of Knowledge (Bourque, Fairley, and IEEE Computer Society 2014). Because research software engineering is an interface discipline, RSEs are often stronger in topics more commonly encountered in research software contexts (e.g., mathematical and engineering foundations) than in other areas (e.g., software engineering economics). However, they bring a solid level of competence in all software engineering topics. Therefore, RSEs can set and analyse software requirements in the context of open-ended, question-driven research. They can design software so that it can sustainably grow, often in an environment of rapid turnover of contributors. They are competent in implementing solutions themselves in a wide range of technologies fit for different scientific applications. They can formulate and implement various types of tests, they can independently maintain software and automate operations of the integration and release process. They can provide working, scalable, and future-proof solutions in a professional context and with common project and software management techniques, adapted to the needs of the research environment. Finally, as people who have often gained significant research experience in a particular discipline, they combine the necessary foundations from their domain with software engineering skills to develop complex software.*(Goth et al. 2024)

This module tries to lay the foundations for the advanced RSE software engineering training.

- Bourque, Pierre, Richard E. Fairley, and IEEE Computer Society. 2014. *Guide to the Software Engineering Body of Knowledge (SWEBOK(R)): Version 3.0*. 3rd ed. Washington, DC, USA: IEEE Computer Society Press.
- Gesellschaft für Informatik e.V. (GI). 2016. “Empfehlungen Für Bachelor- Und Masterprogramme Im Studienfach Informatik an Hochschulen.” GI-Empfehlungen.
- Goth, F, R Alves, M Braun, LJ Castro, G Chourdakis, S Christ, J Cohen, et al. 2024. “Foundational Competencies and Responsibilities of a Research Software Engineer [Version 1; Peer Review: Awaiting Peer Review].” *F1000Research* 13 (1429). <https://doi.org/10.12688/f1000research.157778.1>.
- Segal, Judith. 2009. “Some Challenges Facing Software Engineers Developing Software for Scientists.” In *Proceedings of the 2009 ICSE Workshop on Software Engineering for Computational Science and Engineering*. IEEE. <https://doi.org/10.1109/secse.2009.5069156>.

## 5.21 Software Engineering I

Basic concepts of software engineering, software and product life cycle, process models for the design of large software systems, semantic aspects of domain description, hierarchy, parallelism, real-time and embedded systems as fundamental paradigms, organizational principles of complex software systems, design by contract, patterns in modeling and design methods of quality assurance, evolution and re-engineering, selected languages and tools for process- and object-oriented modeling, methods and languages for object-oriented design, architectures and architectural patterns of software systems, architecture of enterprise applications, design and implementation models in the object-oriented paradigm, e.g., Java 2 SE, design patterns, software testing methods.

**Lecture:** Software Engineering I

**SWS:** 2 **ECTS:** 4

**Exercise:** Software Engineering I Exercise

**SWS:** 2 **ECTS:** 2

## 5.22 Software Engineering 2

The module covers a selection of advanced topics in the field of software engineering, such as software quality assurance, service engineering, virtualization, programming languages and design, and formal methods in system design.

**Lecture:** Software Engineering II

**SWS:** 2 **ECTS:** 2

**Exercise:** Software Engineering II Exercise

**SWS:** 2 **ECTS:** 4

### 5.22.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_programming_1	Understanding the fundamental concepts of software engineering	Computer Science		Demonstrate understanding through theoretical assessments and practical examples	University of Potsdam	<a href="#">Link</a>
gen_programming_2	Applying various approaches of software engineering	Computer Science	gen_programming_1	Complete assignments or projects using different software engineering methods	University of Potsdam	<a href="#">Link</a>
gen_programming_3	Identifying and utilize essential technologies and tools for specification, component-based development, and quality assurance of modern software systems	Computer Science	gen_programming_1	Work with selected tools and technologies in practical exercises and case studies	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_programming_4	Demonstrate an in-depth understanding and ability to apply various approaches of software engineering	Computer Science	gen_programming_1 gen_programming_2	Successfully complete advanced projects employing different software engineering methods	University of Potsdam	<a href="#">Link</a>
gen_programming_5	Understand the characteristics of a wide range of technologies and tools for specification, component-based development, and quality assurance of modern software systems, and apply them in various contexts	Computer Science	gen_programming_3	Apply appropriate technologies and tools in complex case studies and demonstrate their use in different application scenarios	University of Potsdam	<a href="#">Link</a>

## 5.22.2 Sources & Implementations:

### 5.22.2.1 Curricula

- [Computing Curricula 2020](#)

### 5.22.2.2 Courses

- [Software Engineering I](#)



### 5.22.2.3 Programs

- [UP Computational Science Master](#)

## 5.23 Example Module: Fundamentals of Computer Science

This is an example module to showcase the integration pipeline

### 5.23.1 Basics of Computer Science

#### 5.23.1.1 Basic Concepts

- Introduction to computer science, basic concepts of operating systems using UNIX/Linux as an example
- From problem to algorithm: concept of an algorithm, design of algorithms, pseudocode, refinement, brute-force algorithms, models and modeling, graphs and their representation, simple algorithms on graphs, analysis of algorithms (correctness, termination, runtime)
- Implementation of algorithms (e.g., using Python)
- Programming paradigms: procedural, object-oriented, and functional programming; recursion versus iteration
- From program to process: assembly languages, assembler, compiler, interpreter, syntax and semantics of programming languages
- Limits of algorithms: computability, decidability, undecidability

**Lecture:** Basic Programming

**SWS:** 2 **ECTS:** 2

**Exercise:** Basic Programming Exercise

**SWS:** 2 **ECTS:** 4

### 5.23.2 Applied Programming

#### 5.23.2.1 Procedural Programming Concepts

*Programming with an imperative-procedural language (such as C):*

- Data types, type casting, control structures, functions and procedures, parameter passing paradigms, call stack
- Pointers, arrays, strings, structured types

- Errors and their handling
- Dynamic memory management
- Program libraries

### 5.23.2.2 Programming in an Object-Oriented Language (e.g., Java)

- Classes, objects, constructors
- Inheritance, polymorphism, abstract classes/interfaces
- Exceptions and exception handling
- Namespaces (packages)
- Generic classes and types
- Program libraries

**Lecture:** Applied Programming

**SWS: 2 ECTS: 2**

**Exercise:** Applied Programming Exercise

**SWS: 2 ECTS: 4**

### 5.23.3 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
ex_programming_mod1	Using modular programming in imperative-procedural programming language (e.g., C) and an object-oriented language (e.g., Java) with confidence	Computer Science		Submit working programs in both languages demonstrating syntax and language-specific features	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
ex_programming_mod1C3	Implement basic data structures and algorithms	Computer Science	ex_programming_mod1C2	Submit a project with implemented algorithms and data structures (e.g., lists, trees, sorting)	University of Potsdam	<a href="#">Link</a>
ex_programming_mod1C3	Distinguish between error types and handle them appropriately in code	Computer Science	ex_programming_mod1C2	Demonstrate error handling techniques in submitted code (e.g., input validation, error codes, exceptions)	University of Potsdam	<a href="#">Link</a>
ex_programming_mod1C4	Identify and use appropriate library functions in programming tasks	Computer Science	ex_programming_mod1C3	Integrate external libraries in coding tasks and document their usage	University of Potsdam	<a href="#">Link</a>
ex_programming_mod1C5	Use basic functions and mechanisms of operating systems using UNIX/Linux as an example	Computer Science	ex_programming_mod1C4	Demonstrate file handling, permissions, and process control using UNIX/Linux commands	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
ex_prog_createing_mod1C6	Creating and refine simple algorithms using semi-formal notation	Computer Science		Submit pseudocode or flowcharts for given algorithmic problems	University of Potsdam	<a href="#">Link</a>
ex_prog_evaluating_and_mod1C7	Evaluating and compare algorithms using runtime analysis	Computer Science	ex_programming_mod1C6	Provide time complexity comparisons for multiple algorithmic solutions	University of Potsdam	<a href="#">Link</a>
ex_prog_implement_mod1C8	Implement simple algorithms using imperative and functional programming styles (e.g., in Python)	Computer Science	ex_programming_mod1C6	Submit code demonstrating both imperative and functional styles for the same problem	University of Potsdam	<a href="#">Link</a>
ex_prog_distinguishing_mod1C9	Distinguishing between programming paradigms and identify their characteristics	Computer Science	ex_programming_mod1C6	Classify given code snippets by paradigm and justify the classification	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
C10	Express simple programs in an assembly language	Computer Science		Translate simple high-level logic into assembler code	University of Potsdam	<a href="#">Link</a>
C11	Discuss the limits of algorithms, including computability and decidability	Computer Science		Write a short essay or present on concepts such as the Halting Problem or undecidability	University of Potsdam	<a href="#">Link</a>

#### 5.23.4 Sources & Implementations:

##### 5.23.4.1 Curricula

- [Computing Curricula 2020](#)

##### 5.23.4.2 Courses

- [UP Grundlagen der Programmierung](#)
- [UP Praxis der Programmierung](#)
- [Python for Psychologists](#)
- [Grundlagen der Informatik](#)

##### 5.23.4.3 Programs

- [UP Computational Science Master](#)

## 5.24 Distributed Systems

### 5.24.1 Introduction

This module deals with distributed IT systems.

### 5.24.2 Contents

The module covers a selection of the following topics:

- Reliability of distributed systems: Concepts of distributed file systems, synchronization techniques for reliable distributed applications, concepts of load balancing in high-availability clusters,
- Example: Sensor networks: Routing in sensor networks, operating systems for sensor networks, security in sensor networks,
- Secure internet protocols (IP security (IPsec), Pretty Good Privacy (PGP), Secure Socket Layer (SSL), Transport Layer Security (TLS), Secure Shell (SSH), DNS security (DNSsec)), secure IPv6 networks.

### 5.24.3 Learning Objectives

- can evaluate existing distributed systems in terms of reliability and security and identify vulnerabilities.
- can correctly identify reliability and security requirements when designing new distributed systems and consider them early in the development process.

### 5.24.4 Examination Methods

- either 120 min written exam
- or 20-30 min oral examination

**Lecture:** Distributed Systems

**SWS:** 2 **ECTS:** 2

**Exercise:** Distributed Systems Exercise

**SWS:** 2 **ECTS:** 4

### 5.24.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
dist_systems1	analyze existing distributed systems in terms of reliability and security and identify vulnerabilities	Computer Science		Submit written analysis of existing distributed systems	University of Potsdam	<a href="#">Link</a>
dist_systems2	identify reliability and security requirements when designing new distributed system and consider them early in the development process	Computer Science		discuss necessary requirements for a design of a distributed system	University of Potsdam	<a href="#">Link</a>

## 5.24.6 Sources & Implementations:

### 5.24.6.1 Curricula

- [None](#)

### 5.24.6.2 Courses

- [UP Verteilte Systeme](#)

### 5.24.6.3 Recommended Course Literature

- [TODO](#)

#### 5.24.6.4 Programs

- [UP Computational Science Master](#)



## 6 Track 2: Domain Science / Natural Science Track

### 6.1 Suggested Study Plan

Type	Description	SWS	Sem 1	Sem 2	Sem 3	Sem 4
Lecture	RSE Lecture Series	2	3	0	0	0
Seminar	RSE Nuts and Bolts I	2	3	0	0	0
Lecture	RSE Management Lecture	2	2	0	0	0
Exercise	RSE Management Exercise	2	2	0	0	0
Lecture	Statistical Data Analysis	4	6	0	0	0
Exercise	Statistical Data Analysis Exercise	2	3	0	0	0
Lecture	Basic Programming	2	2	0	0	0
Exercise	Basic Programming Exercise	2	4	0	0	0
Lecture	Wildcard Science I	2	2	0	0	0
Seminar	RSE Philosophy	2	3	0	0	0
Lab	Wildcard Science Lab I	2	0	4	0	0
Lecture	Wildcard Science II	2	0	2	0	0
Lecture	Computational Wildcard Science	2	0	3	0	0
Lab	Computational Wildcard Science Lab	2	0	3	0	0
Lab	Science Lab Project	4	0	6	0	0
Lecture	Applied Programming	2	0	2	0	0
Exercise	Applied Programming Exercise	2	0	4	0	0
Lecture	Scientific Computing Basics	2	0	3	0	0
Exercise	Scientific Computing Basics Exercise	2	0	3	0	0
Lab	Wildcard Science Lab II	2	0	0	4	0
Seminar	RSE Nuts and Bolts II	2	0	0	3	0
Lecture	Text2Data	2	0	0	4	0
Exercise	Text2Data Exercise	2	0	0	2	0
Lecture	Software Engineering I	2	0	0	4	0
Exercise	Software Engineering I Exercise	2	0	0	2	0
Lecture	High Performance Computing	2	0	0	3	0
Exercise	High Performance Computing Exercise	2	0	0	3	0
Seminar	RSE Software Engineering	2	0	0	3	0

Type	Description	SWS	Sem 1	Sem 2	Sem 3	Sem 4
Exercise	RSE Software Engineering Exercise	2	0	0	3	0
Thesis	RSE Master Thesis	0	0	0	0	30
	Total ECTS per semester		30	30	31	30

**Total Curriculum ECTS: 121**

## 6.2 Wildcard Science Module

This module offers RSE students the opportunity to deepen their understanding of a scientific discipline outside of their home domain. Students choose a science module — such as physics, chemistry, biology, or earth sciences — and engage with its research practices, core questions, and data/software challenges.

The goal is to help students become better collaborators by gaining first-hand exposure to the terminology, logic, and needs of another scientific domain. This broadens the student's ability to apply RSE skills in interdisciplinary teams and unfamiliar environments.

The module may consist of lectures, lab sessions, and domain-specific mini-projects. RSEs are encouraged to reflect on how software engineering, data handling, reproducibility, and tooling intersect with the chosen discipline.

This module is deliberately flexible to accommodate institutional offerings and student interests as well as providing the option to stay attached to the identity of the chosen discipline.

**Lecture:** Wildcard Science I

**SWS: 2 ECTS: 2**

**Lecture:** Wildcard Science II

**SWS: 2 ECTS: 2**

**Lab:** Wildcard Science Lab I

**SWS: 2 ECTS: 4**

**Lab:** Wildcard Science Lab II

**SWS: 2 ECTS: 4**

### 6.2.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
None		Wildcard Domain Science				<a href="#">Link</a>

## 6.2.2 Sources & Implementations:

### 6.2.2.1 Programs

- [Uni Würzburg Master Physik](#)
- [Uni Würzburg Master Chemie](#)

## 6.3 RSE Computing

RSEs with expertise in HPC and other performance-critical computing domains specialize in optimizing code for efficient execution across various platforms, including clusters, cloud, edge, and embedded systems. They understand parallel programming models, hardware-specific optimizations, profiling tools, and platform constraints such as memory, energy, and latency. Their skills enable them to adapt software to diverse infrastructures, manage complex dependencies, and support researchers in accessing and using advanced computing resources effectively and sustainably.

### 6.3.1 Basic Scientific Computing

#### 6.3.1.1 Module Overview

This module provides an entry-level yet rigorous foundation in scientific computing for graduate students and researchers who need to **design, implement, and evaluate computational experiments**. Learners gain an awareness of the numerical underpinnings of modern simulation and data-driven research, with an emphasis on writing *reproducible, efficient, and trustworthy* code.

### 6.3.1.2 Intended Learning Outcomes

By the end of the module participants will be able to

1. Benchmark small programs and interpret performance metrics in a research context.
2. Explain how approximation theory and floating-point arithmetic affect numerical accuracy and stability.
3. Identify when to use established simulation libraries (e.g. BLAS/LAPACK, PETSc, Trilinos) instead of custom code.
4. Write simple GPU kernels and describe the core principles of accelerator programming.
5. Submit and monitor batch & array jobs on a mid-size compute cluster.
6. Describe common HPC challenges—such as I/O bottlenecks, threading, and NUMA—and propose mitigation strategies.
7. Maintain research software through continuous benchmarking.

### 6.3.1.3 Syllabus (Indicative Content)

Week	Theme	Topics
1	<b>Benchmarking &amp; Profiling</b>	Timing strategies · micro vs. macro benchmarks · tooling overview
2	<b>Precision &amp; Approximation</b>	IEEE-754 recap · conditioning & stability · error propagation
3	<b>Scientific Libraries</b>	BLAS/LAPACK anatomy · hierarchical I/O libraries · overview of PETSc/Trilinos/Hypre
4	<b>GPU Primer</b>	Kernel model · memory hierarchy · CUDA/OpenCL/PyTorch lightning intro
5	<b>Working on a Cluster</b>	Slurm basics · job arrays · job dependencies · simple Bash launchers
6	<b>HPC Pitfalls</b>	I/O throughput · thread oversubscription · NUMA awareness
7	<b>Software Maintenance</b>	Regression + performance tests · continuous benchmarking pipelines

### 6.3.1.4 Teaching & Learning Methods

Short lectures (30%) are coupled with hands-on labs (70%). Students complete **weekly notebooks** and a **mini-project** that reproduces and optimises a published computational result.

#### 6.3.1.5 Assessment

Component	Weight	Details
Continuous labs	40%	Weekly graded notebooks
Final mini-project	60%	Report, code, and benchmark suite

#### 6.3.1.6 Prerequisites

- Basic programming in Python, C/C++, or Julia
- Undergraduate calculus & linear algebra

#### 6.3.1.7 Key Resources

ChatGPT fantasy

**Lecture:** Scientific Computing Basics

**SWS:** 2 **ECTS:** 3

**Exercise:** Scientific Computing Basics Exercise

**SWS:** 2 **ECTS:** 3

**Lecture:** High Performance Computing

**SWS:** 2 **ECTS:** 3

**Exercise:** High Performance Computing Exercise

**SWS:** 2 **ECTS:** 3

#### 6.3.2 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
comp_module_1	Explain benchmark and profile computational code to evaluate performance and bottlenecks	Scientific Computing	rse_tooling_2	Submit benchmark reports comparing implementations and justifying trade-offs	RSE Curriculum Draft	<a href="#">Link</a>
comp_module_2	Explain and apply principles of approximation theory and numerical precision in scientific computing	Scientific Computing		Answer conceptual questions and implement small examples highlighting precision trade-offs	RSE Curriculum Draft	<a href="#">Link</a>
comp_module_3	Explain floating-point arithmetic and its implications for scientific accuracy and performance	Scientific Computing	comp_module_2	Provide examples showing effects of precision loss and propose mitigations	RSE Curriculum Draft	<a href="#">Link</a>
comp_module_4	Describe common simulation libraries and numerical frameworks (e.g., BLAS, LAPACK, PETSc, Trilinos)	Scientific Computing		List relevant libraries for a task and justify choice or avoidance of custom implementations	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
comp_module_5	Compare interpreted and compiled languages in terms of performance and suitability for computing tasks	Scientific Computing		Write code samples in both types of language and explain their performance characteristics	RSE Curriculum Draft	<a href="#">Link</a>
hpc_module_3	Run batch and array jobs on a cluster, including job dependencies	High-Performance Computing	rse_tooling_3	Submit job scripts using SLURM or similar systems demonstrating correct use of job arrays and dependencies	RSE Curriculum Draft	<a href="#">Link</a>
hpc_module_2	Identify and manage common challenges in HPC systems (e.g., I/O bottlenecks, threading, NUMA memory)	High-Performance Computing	hpc_module_1	Provide performance logs and interpret bottlenecks in a real or simulated HPC task	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
hpc_module_3	Understand shell scripting (e.g., Bash) to automate HPC job submission	High-Performance Computing	rse_tooling_3	Submit scripts that automate the execution of HPC jobs and handle job logic	RSE Curriculum Draft	<a href="#">Link</a>
hpc_module_4	Understand and use the principles of accelerator programming (e.g., GPU kernels and frameworks)	High-Performance Computing		Submit a small CUDA or OpenCL program with documentation of the principles used	RSE Curriculum Draft	<a href="#">Link</a>
hpc_module_5	Maintain scientific computing software including use of continuous benchmarking	High-Performance Computing	comp_module_1	Provide benchmark and performance history for evolving versions of software	RSE Curriculum Draft	<a href="#">Link</a>

### 6.3.3 Sources & Implementations:

#### 6.3.3.1 Curricula

- [EUMaster4HPC](#)

#### 6.3.3.2 Courses

- [Viral Instructions Hardware](#)



- [HPC Computing](#)

#### 6.3.3.3 Recommended Course Literature

- [What every computer scientist should know about floating-point arithmetic](#)

#### 6.3.3.4 Programs

- [HPC-carpentry](#)

## 6.4 Science Lab Module

*Applied Research Software Engineering in STEM sciences*

This lab module provides students with a hands-on opportunity to apply research software engineering principles to real-world scientific problems from the MINT disciplines (Mathematics, Informatics, Natural Sciences, and Technology). Students work on projects originating from active research contexts — such as simulations in physics, data analysis in chemistry, modeling in biology, or

**Lab:** Science Lab Project

**SWS:** 4 **ECTS:** 6

### 6.4.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
None		Wildcard Domain Science				<a href="#">Link</a>

### 6.4.2 Sources & Implementations:

#### 6.4.2.1 Programs

- [Uni Potsdam Master Bioinformatik](#)

## 6.5 RSE Nuts and Bolts

### 6.5.1 Introduction

This module, inspired by the [MIT Missing Semester](#), addresses the “nuts and bolts” often missing from traditional academic training in computing. It aims to provide students with practical skills and conceptual understanding for building robust, maintainable, and reproducible research software—key competencies in Research Software Engineering (RSE).

### 6.5.2 General Competencies

The module begins with general-purpose computing tools and techniques that are foundational for any research software engineer:

- Shell tools and scripting
- Command-line environments
- Editors and IDEs (e.g., Vim)
- Version control (Git)
- Data wrangling
- Debugging and profiling
- Metaprogramming
- Security and cryptography

### 6.5.3 RSE-Specific Topics

Building on these foundations, the module introduces RSE-specific concepts and good practices:

- **Version control and collaboration**
  - Git for code history, collaboration, and issue tracking
- **Virtualization concepts**
  - Containerization and environment management
- **The Data Life Cycle**
  - Managing research data and understanding data provenance
- **Good coding practices**
  - Reproducible and testable code
  - Meaningful documentation and error messages
  - Modular software design

- Performance-conscious coding
- Easily installable and distributable software
- Coding standards, formatting, and linting
- **Software management planning**
  - Writing Data and Software Management Plans
  - Sustainable development and community involvement
- **Low-level programming**
  - Introduction to a compiled language (e.g., C) to expose hardware-level concerns and efficient memory management
- **Long-term software maintenance**
  - Version tracking, bug management, and sustainability strategies
  - Building and maintaining research software communities

#### 6.5.4 Beyond the Basics

Finally, the module touches on practices that support the scholarly nature of research software:

- Software publication and citation (see SP in (Goth et al. 2024))
- Use of domain-specific repositories and registries (see DOMREP in (Goth et al. 2024))

By the end of this module, students will be well-equipped to design, develop, document, and maintain research software that meets high standards of quality, sustainability, and reproducibility.

The module is made up of two seminars that the students take at different stages in their master’s program: In the first seminar during their first semester, students mainly learn new concepts and get to know essential tools, whereas the second seminar in the third semester focuses on teaching others about research software and the development process of it (see TEACH in (Goth et al. 2024)).

**Seminar:** RSE Nuts and Bolts I This is an introductory class to essential techniques an RSE needs in everyday life. **SWS:** 2 **ECTS:** 3

**Seminar:** RSE Nuts and Bolts II This is an advanced class of RSE techniques that includes a teaching component as part of the preparation for working as an RSE in interdisciplinary teams. **SWS:** 2 **ECTS:** 3

### 6.5.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tool1	Use literate programming tools (e.g., Quarto, Marimo, Pluto.jl, Jupyter) to combine code, results, and narrative	Research Software Engineering		Submit a literate notebook or document integrating code, visualizations, and explanatory text	Workshop Participants	<a href="#">Link</a>
rse_tool2	Use Python for visualization, scripting, templating, and integration tasks	Research Software Engineering		Submit a Python project demonstrating use of libraries for visualization, web tasks, and templating	Workshop Participants	<a href="#">Link</a>
rse_tool3	Write and use Bash scripts for automation	Research Software Engineering		Submit shell scripts automating file manipulation or computational workflows	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling_4	Apply testing, debugging, and logging techniques to ensure software reliability	Research Software Engineering	rse_tooling_2	Submit logs, test cases, and debugging documentation for a non-trivial Python or Bash project	Workshop Participants	<a href="#">Link</a>
rse_tooling_5	Use workflow management tools (e.g., CWL, Nextflow) to design scalable, reproducible pipelines	Research Software Engineering	rse_tooling_3, rse_tooling_11	Submit a reproducible workflow including metadata and input/output definitions	Workshop Participants	<a href="#">Link</a>
rse_tooling_6	Estimate resource requirements for computational tasks using profiling and benchmarking	Research Software Engineering	rse_tooling_2, rse_tooling_5	Provide resource usage profiles and discuss optimization implications	Workshop Participants	<a href="#">Link</a>
rse_tooling_7	Use package managers and virtual environments (e.g., conda, nix) to manage software dependencies	Research Software Engineering		Submit environment definitions and reproducible setup instructions for a project	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling_8	Document and package software for usability and reusability, using generators and modular design	Research Software Engineering	rse_tooling_2	Submit user and developer documentation generated with Sphinx or similar, plus a reusable code module	Workshop Participants	<a href="#">Link</a>
rse_tooling_9	Communicate technical RSE topics effectively with non-technical audiences	Research Software Engineering		Prepare and deliver a presentation or write an article explaining RSE concepts to a general audience	Workshop Participants	<a href="#">Link</a>
rse_tooling_10	Apply authentication and authorization mechanisms (e.g., LDAP, ACLs, Active Directory)	Research Software Engineering		Configure and demonstrate access control for a multi-user service or application	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling_1	Make informed decisions about tooling and infrastructure (e.g., Jupyter vs scripts, local vs HPC/cloud)	Research Software Engineering	rse_tooling_1, rse_tooling_2, rse_tooling_3	Submit a comparative analysis justifying tooling and infrastructure choices for a research project	Workshop Participants	<a href="#">Link</a>
rse_tooling_2	Engage in practice collaborative development, including version control and code review	Research Software Engineering	rse_tooling_2	Submit a project with version history and documented code reviews	Workshop Participants	<a href="#">Link</a>
rse_tooling_12	Mentor others in research software engineering practices	Research Software Engineering	rse_tooling_12	Document a mentoring session, workshop, or support activity	Workshop Participants	<a href="#">Link</a>
rse_tooling_4	Deploy and maintain web servers for research applications	Research Software Engineering	rse_tooling_2	Deploy a working web application with setup and maintenance documentation	Workshop Participants	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_tooling	Understand and manage file systems, including local and network-attached storage	Research Software Engineering		Document storage strategies and access mechanisms in a real-world setup	Workshop Participants	<a href="#">Link</a>

## 6.5.6 Sources & Implementations:

### 6.5.6.1 Courses

- [MIT Missing Semester](#)
- [CodeRefinery](#)
- [INTERSECT Training Materials](#)
- [Digital Research Academy Materials \(Git, HPC, Reproducibility, Research Software\)](#)
- [Building Better Research Software \(SSI\)](#)
- [Docker for neuroscience \(jupyter book\)](#)

## 6.6 RSE Philosophy

### 6.6.1 Introduction

The RSE master program is more than a computer science specialisation for researchers. People working as RSE are often involved in digitalization projects, institutional development or other non-technical tasks.

### 6.6.2 Contents

For a university level study program it is fitting that students learn an abstract high-level understanding of their field so that they can adapt technical models, communication frameworks and policy recommendations to the complex cases. For this they need a solid understanding in some of the more theoretical fields such as ...



- philosophy of science
- sociology of technology
- ethics and artificial intelligence
- human computer interaction
- digital humanities

### 6.6.3 General Competences

This module conveys competences in areas such as but not limited to ...

- conducting and leading research (NEW)
- understanding the research cycle (RC)
- interaction with users and stakeholders (USERS)

**Seminar:** RSE Philosophy This is an introductory class to ... **SWS:** 2 **ECTS:** 3

### 6.6.4 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_philosophy_1	TODO	Research Software Engineering		...		Link

### 6.6.5 Sources & Implementations:

#### 6.6.5.1 Courses

- TODO

## 6.7 Master's Thesis Module: Research Software Engineering Thesis

The master's thesis is the culminating component of the RSE programme. In this module, students apply the full spectrum of Research Software Engineering skills in a real-world research setting, demonstrating their ability to independently design, implement, and document a computational research contribution.

The thesis must address a research question in collaboration with a scientific or applied domain, but its core should include a substantial computational component. This may involve

software development, data-intensive research, reproducibility infrastructure, or performance engineering — depending on the chosen topic and specialization.

Each thesis must be supervised jointly by:

- A domain expert (e.g., in physics, life sciences, or humanities)
- An RSE mentor (who ensures the quality and relevance of the computational contribution)

Students are expected to follow best practices in software engineering, version control, testing, and documentation. The final submission must include:

- A written thesis describing both the scientific and software contributions
- A structured, reproducible code repository
- A presentation and defense in a thesis colloquium

The colloquium serves as both a public communication exercise and a final evaluation, where students present their project and reflect on challenges and insights gained during the thesis.

**Thesis:** RSE Master Thesis

**SWS:** 0 **ECTS:** 30

## 6.8 RSE Management and Communication

### 6.8.1 Introduction

This module comprises the communication and management skills that are relevant for working in the interdisciplinary setting of RSE-professionals.

This includes but is not limited to:

- working in a team (see TEAM in (Goth et al. 2024))
- teaching RSE-basics (see TEACH in (Goth et al. 2024))
- project management (see PM in (Goth et al. 2024))
- interaction with users and other stakeholders (see USERS in (Goth et al. 2024))

## 6.8.2 Contents

- **research management**
  - research cycle
  - open science, FAIR, FAIR4RS
  - publication workflow: obstacles and embargoes
  - legal aspects of research data, e.g. GDPR
  - pseudonymization/anonymization methods for data privacy
  - public databases
- **quality control**
  - requirements engineering
  - trying goals with quality: test-driven development
  - behavior-driven development, Gherkin-Style acceptance testing
  - project folder organization
  - code review principles
- **communication and collaboration**
  - communication frameworks, e.g. AIDA, RACE, 7 C's
  - personality traits and their impact on cooperation
  - collaboration frameworks for remote work
  - realisation and benefits of pair programming and mob/ensemble programming
  - technical English writing skills: writing in issues and merge requests, code review...
  - conflict management, e.g. dealing with researchers that do not listen
  - how to address relevant stakeholders (i.e. users and SEs) with different background knowledge, experience and expectations
  - equity, diversity and inclusion principles
- **team management**
  - challenges of transient teams (that only exist for 5-15 hours)
  - effects of varying team sizes
  - management depending on project size/type
  - specialised roles in a software team
  - intercultural and interdisciplinary differences
  - team management methodologies
  - importance of shared values in a RSE team
  - dual goals: project vs. personal goal
- **time and project management**
  - goal-setting
  - project management methods, their strengths and weaknesses
  - agile (not necessarily Scrum)

- Lean & Kanban (Small-Batch Philosophies)
- division of tasks into sub-tasks and task-dependencies
- iterative workflows
- continuous delivery
- communication with a manager/supervisor

**Lecture:** RSE Management Lecture This is an introductory lecture covering research, project and team management techniques an RSE needs in everyday life. **SWS: 2 ECTS: 2**

**Exercise:** RSE Management Exercise This is an exercise to apply and practice the taught methods with case-studys, role-plays etc. **SWS: 2 ECTS: 2**

### 6.8.3 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_8	Build and manage sustainable research software communities	Research Software Engineering, Community Engagement	rse_tooling_13	Document strategies used for user engagement, feedback, and community growth in a real project	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_9	Work in an agile software development process, including requirement gathering and iteration	Research Software Engineering		Submit a project that uses agile planning (e.g., user stories, sprints, stand-ups) and reflects on iteration outcomes	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practices_10	Define project scope, gather requirements, and manage stakeholder expectations	Research Software Engineering		Provide a requirements document and stakeholder communication log for a software project	RSE Curriculum Draft	<a href="#">Link</a>
rse_practices_11	Plan for software maintenance and long-term sustainability, including archiving strategies	Research Software Engineering	rse_practices_6	Submit a sustainability or exit plan describing how the software will be maintained or archived	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_maint_01	Explain particular implementation choices in a convincing manner	Research Software Engineering		Deliver a defense of chosen implementation decisions in a discussion with a domain expert who has limited technical knowledge (ideally oral examination or project presentation with potential ‘customer/user’ questions)	RSE Community	<a href="#">Link</a>
rse_maint_02	Exemplify and articulate shared team values and their impact on work	Research Software Engineering		Identify core team values and demonstrate how they influence key implementation decisions (e.g., design, communication, and collaboration)	RSE Community	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_management_03	Plan and manage projects using standard methods effectively and efficiently	Research Software Engineering		Develop a comprehensive project plan for a given project, including scope, milestones, risks, resources, and success criteria	RSE Community	<a href="#">Link</a>
rse_management_04	Discuss methods to set up a Diversity, Equity and Inclusion (DEI) framework in an RSE team	Research Software Engineering		Analyse and evaluate a DEI framework for a given project	RSE Community	<a href="#">Link</a>

## 6.8.4 Sources & Implementations:

### 6.8.4.1 Courses

- [RSE Leadership Course](#)

### 6.8.4.2 Recommended Course Literature

- [Remote Mob Programming](#)
- [Code with the Wisdom of the Crowd](#)
- [Collaboration Explained](#)
- [Team Topologies](#)

- [Technical Agile Coaching with the Samman method](#)
- [Lean Product and Process Development](#)
- [Extreme Programming Explained](#)

**Lecture:** Mathematical Foundations of Data Science The module provides mathematical foundations in the field of Data Science. Topics include a selection from the areas of graph analysis, stochastic models, and signal analysis using wavelets. **SWS: 2 ECTS: 2**

**Exercise:** Mathematical Foundations of Data Science Exercise

**SWS: 2 ECTS: 4**

## 6.9 Statistical Data Analysis

This module focuses on the statistical study and quantitative analysis of the dependence between observed random variables (e.g., yield/production settings; lifespan/treatment type and injury type). Essential foundations for the statistical treatment of such relationships are provided by the linear regression model, which is studied in detail in the first part of the lecture. Within this framework, topics such as estimation, testing, and uncertainty quantification (analysis of variance) are addressed. In the second part, an introduction to advanced methods and approaches for examining relationships is offered, including nonlinear and nonparametric regression models. Additionally, questions of classification and dimensionality reduction are covered.

**Lecture:** Statistical Data Analysis

**SWS: 4 ECTS: 6**

## 6.10 Statistical Data Analysis

This module focuses on the statistical study and quantitative analysis of the dependence between observed random variables (e.g., yield/production settings; lifespan/treatment type and injury type). Essential foundations for the statistical treatment of such relationships are provided by the linear regression model, which is studied in detail in the first part of the lecture. Within this framework, topics such as estimation, testing, and uncertainty quantification (analysis of variance) are addressed. In the second part, an introduction to advanced methods and approaches for examining relationships is offered, including nonlinear and nonparametric regression models. Additionally, questions of classification and dimensionality reduction are covered.

**Exercise:** Statistical Data Analysis Exercise



**SWS: 2 ECTS: 3**

## **6.11 Text to Data**

### **6.11.1 Contents**

- Introduction to textual data and Natural Language Processing (NLP): key concepts and use cases
- Text acquisition and preprocessing: tokenisation, stemming/lemmatisation, stopwords, named entity recognition, text cleaning
- Text representations: bag-of-words, TF-IDF, word embeddings (Word2Vec, GloVe), contextualised embeddings (BERT, RoBERTa, GPT variants)
- Text mining and information extraction: key terms, topics, sentiment analysis, relation extraction
- Data-to-insight pipelines: from raw text to structured data (CSV/SQL), feature engineering for text data
- Modelling and evaluation: classification, regression, sequence models, transformers; evaluation metrics (accuracy, F1, MCC, AUC)
- Prototyping and reproducibility: experiment tracking, versioning, reproducible pipelines (Docker/Kubernetes, MLflow, DVC)
- Risk management and ethics: bias, fairness, data protection (GDPR), transparency, interpretability (LIME, SHAP)
- Deployment and practical applications: API-based models, batch vs real-time processing, monitoring
- Project work: from a research question to data collection, modelling, evaluation and documentation

### **6.11.2 Learning outcomes**

- Understand how text data is collected, preprocessed, and transformed into usable formats
- Be able to select and justify different text representations
- Apply NLP methods to extract relevant information from text
- Develop, train and evaluate text-based models with attention to reproducibility and ethics
- Assess model limitations, interpretability, and risk of errors
- Build a reproducible, scalable text-data pipeline
- Communicate results clearly in reports, presentations, and prototype designs

**Exercise:** Text2Data Exercise

**SWS: 2 ECTS: 2**

## 6.12 Text to Data

### 6.12.1 Contents

- Introduction to textual data and Natural Language Processing (NLP): key concepts and use cases
- Text acquisition and preprocessing: tokenisation, stemming/lemmatisation, stopwords, named entity recognition, text cleaning
- Text representations: bag-of-words, TF-IDF, word embeddings (Word2Vec, GloVe), contextualised embeddings (BERT, RoBERTa, GPT variants)
- Text mining and information extraction: key terms, topics, sentiment analysis, relation extraction
- Data-to-insight pipelines: from raw text to structured data (CSV/SQL), feature engineering for text data
- Modelling and evaluation: classification, regression, sequence models, transformers; evaluation metrics (accuracy, F1, MCC, AUC)
- Prototyping and reproducibility: experiment tracking, versioning, reproducible pipelines (Docker/Kubernetes, MLflow, DVC)
- Risk management and ethics: bias, fairness, data protection (GDPR), transparency, interpretability (LIME, SHAP)
- Deployment and practical applications: API-based models, batch vs real-time processing, monitoring
- Project work: from a research question to data collection, modelling, evaluation and documentation

### 6.12.2 Learning outcomes

- Understand how text data is collected, preprocessed, and transformed into usable formats
- Be able to select and justify different text representations
- Apply NLP methods to extract relevant information from text
- Develop, train and evaluate text-based models with attention to reproducibility and ethics
- Assess model limitations, interpretability, and risk of errors
- Build a reproducible, scalable text-data pipeline
- Communicate results clearly in reports, presentations, and prototype designs

**Lecture:** Text2Data

**SWS:** 2 **ECTS:** 4

### 6.12.3 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_datascience_1	Assess comprehensive, detailed, and specialized knowledge of selected fundamentals in the field of Data Science	Data Science		Demonstrate knowledge through theoretical exams and practical assignments	University of Potsdam	<a href="#">Link</a>
gen_datascience_2	Demonstrate an in-depth understanding of selected Data Science methods	Data Science	gen_datascience_1	Apply Data Science methods in practical projects and case studies	University of Potsdam	<a href="#">Link</a>
gen_datascience_3	Analyze novel data assimilation and inference problems, develop and implement solutions, and assess solution quality	Data Science	gen_datascience_2	Solve complex inference problems and present implemented solutions with evaluation	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_datascience_2	Develop new ideas and methods, weigh alternatives under incomplete information, and evaluate them considering different evaluation criteria	Data Science	gen_datascience_1	Present projects showcasing creative problem-solving and alternative evaluations under uncertainty	University of Potsdam	<a href="#">Link</a>
gen_statistics_1	Review comprehensive, detailed, and specialized understanding of the linear regression model based on the latest research	Data Science, Statistics		Apply linear regression models to practical problems and interpret results	University of Potsdam	<a href="#">Link</a>
gen_statistics_2	Understand fundamental concepts and methods of nonparametric statistics	Data Science, Statistics	gen_statistics_1	Solve problems involving nonparametric methods and explain applied techniques	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_statistics_3	Solve complex statistical data analysis problems, evaluate alternative modeling approaches according to various criteria, and use statistical software packages for analysis	Data Science, Statistics	gen_statistics_2	Develop solutions for complex data problems using appropriate statistical methods and software	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_statistics_4	Demonstrate academic competences including self-organization, planning skills (identifying work steps), scientific thinking and working techniques (developing solutions for complex questions), discussion of methods, verification of hypotheses, application of mathematical and statistical methods, and use of software packages	Data Science, Statistics	gen_statistics_2	Document project workflows demonstrating planning, analysis, evaluation, and use of statistical software tools	University of Potsdam	<a href="#">Link</a>

#### 6.12.4 Sources & Implementations:

##### 6.12.4.1 Curricula

- [Empfehlungen Masterstudiengänge Data Science](#)

##### 6.12.4.2 Courses

- [Statistical Data Analysis \(MATVMD837\)](#)
- [Mathematical Foundations of Data Science \(MAT-DSAM8B\)](#)

- [Programmieren für Data Scientists Python](#)

#### 6.12.4.3 Programs

- [UP Data Science](#)

### 6.13 RSE Lecture Series

#### 6.13.1 Introduction

This lecture series covers current RSE research and provides room for guest lectures and varying focus topics.

#### 6.13.2 Contents

vary between semesters and weeks, but are all related to RSE

#### 6.13.3 Learning Objectives

- students get a broad image of RSE in the wild

#### 6.13.4 Examination Methods

none?

**Lecture:** RSE Lecture Series

**SWS:** 2 **ECTS:** 3

#### 6.13.5 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_0	describe RSE activities	Computer Science		name x different fields and analyse their differences		Link

### 6.13.6 Sources & Implementations:

#### 6.13.6.1 Curricula

- [None](#)

#### 6.13.6.2 Courses

- [None](#)

#### 6.13.6.3 Recommended Course Literature

- [None](#)

#### 6.13.6.4 Programs

- [None](#)

## 6.14 RSE-Software Engineering

This module extends the Classical Software Engineering Module with research specific learnings. This includes but is not limited to

- software re-use (see SRU in (Goth et al. 2024))
- creating documented code building blocks (see DOCBB in (Goth et al. 2024))
- building distributable software (see DIST in (Goth et al. 2024))
- research specific programming languages
- research specific code requirements (scalability, functional programming, ...)
- Adapting the software life cycle to research (see SWLC in (Goth et al. 2024))
- Software behaviour awareness and analysis (see MOD in (Goth et al. 2024))
- Research specific Engineering Patterns

Also, the seminar provides room for reflection and discussions of SE lab experiences.

**Seminar:** RSE Software Engineering

**SWS:** 2 **ECTS:** 3

**Exercise:** RSE Software Engineering Exercise

**SWS:** 2 **ECTS:** 3



### 6.14.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_1	Apply good coding practices including formatting, linting, and modular design	Research Software Engineering	rse_tooling_2	Submit a code project demonstrating modularity, consistent formatting, and use of linters	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_2	Write code and documentation that supports reproducibility in research	Research Software Engineering	rse_tooling_1, rse_tooling_4	Submit a project with data, software, and instructions allowing full reproduction of results	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_3	Organise files and name code artifacts using clear, consistent conventions	Research Software Engineering		Submit a software repository with a structured layout and consistent naming scheme	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_4	Version control code and collaborate using platforms like GitHub or GitLab	Research Software Engineering	rse_tooling_12	Participate in a collaborative coding project using Git-based workflows and merge requests	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_5	Write effective documentation and user-facing error messages	Research Software Engineering	rse_tooling_8	Provide documentation and example error handling demonstrating clarity and user support	RSE Curriculum Draft	<a href="#">Link</a>
rse_practice_12	Write performant code suitable for use in compute-intensive contexts	Research Software Engineering	rse_tooling_2	Submit benchmark results comparing an optimized version of code with a naive implementation	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practices_6	Publish code and software in trusted repositories and package managers	Research Software Engineering	rse_practices_6	Publish software to a repository (e.g., GitHub, PyPI, CRAN) and register it with a long-term archive (e.g., Zenodo)	RSE Curriculum Draft	<a href="#">Link</a>
rse_practices_6	Apply licensing and publishing strategies to make software reusable and citable	Research Software Engineering		Submit a software project with an appropriate open license and published DOI (e.g., via Zenodo)	RSE Curriculum Draft	<a href="#">Link</a>
rse_practices_6	Apply principles of Open Source and FAIR (Findable, Accessible, Interoperable, Reusable) software	Research Software Engineering	rse_practices_6	Review or create a software project and evaluate its compliance with FAIR/Open Source principles	RSE Curriculum Draft	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
rse_practice_1	Submit a data-driven project showing clear data organisation, metadata, and reproducibility within a software project in accordance with best practices	Research Software Engineering, Data Management		Submit a data-driven project showing clear data organisation, metadata, and reproducibility	RSE Curriculum Draft	<a href="#">Link</a>

## 6.14.2 Sources & Implementations:

### 6.14.2.1 Courses

- [TODO](#)

## 6.15 Classical Software Engineering

*To summarise the vast range of the skills a software engineer is typically equipped with, we refer to the Guide to the Software Engineering Body of Knowledge (Bourque, Fairley, and IEEE Computer Society 2014). Because research software engineering is an interface discipline, RSEs are often stronger in topics more commonly encountered in research software contexts (e.g., mathematical and engineering foundations) than in other areas (e.g., software engineering economics). However, they bring a solid level of competence in all software engineering topics. Therefore, RSEs can set and analyse software requirements in the context of open-ended, question-driven research. They can design software so that it can sustainably grow, often in an environment of rapid turnover of contributors. They are competent in implementing solutions themselves in a wide range of technologies fit for different scientific applications. They can formulate and implement various types of tests, they can independently maintain software and automate operations of the integration and release process. They can provide working, scalable, and future-proof solutions in a professional context and with common project and software management techniques, adapted to the needs of the research environment. Finally, as people who have often gained significant research experience in a particular discipline, they combine the necessary foundations from their domain with software engineering skills to develop complex software.*(Goth et al. 2024)

This module tries to lay the foundations for the advanced RSE software engineering training.

- Bourque, Pierre, Richard E. Fairley, and IEEE Computer Society. 2014. *Guide to the Software Engineering Body of Knowledge (SWEBOK(R)): Version 3.0*. 3rd ed. Washington, DC, USA: IEEE Computer Society Press.
- Gesellschaft für Informatik e.V. (GI). 2016. “Empfehlungen Für Bachelor- Und Masterprogramme Im Studienfach Informatik an Hochschulen.” GI-Empfehlungen.
- Goth, F, R Alves, M Braun, LJ Castro, G Chourdakis, S Christ, J Cohen, et al. 2024. “Foundational Competencies and Responsibilities of a Research Software Engineer [Version 1; Peer Review: Awaiting Peer Review].” *F1000Research* 13 (1429). <https://doi.org/10.12688/f1000research.157778.1>.
- Segal, Judith. 2009. “Some Challenges Facing Software Engineers Developing Software for Scientists.” In *Proceedings of the 2009 ICSE Workshop on Software Engineering for Computational Science and Engineering*. IEEE. <https://doi.org/10.1109/secse.2009.5069156>.

## 6.16 Software Engineering I

Basic concepts of software engineering, software and product life cycle, process models for the design of large software systems, semantic aspects of domain description, hierarchy, parallelism, real-time and embedded systems as fundamental paradigms, organizational principles of complex software systems, design by contract, patterns in modeling and design methods of quality assurance, evolution and re-engineering, selected languages and tools for process- and object-oriented modeling, methods and languages for object-oriented design, architectures and architectural patterns of software systems, architecture of enterprise applications, design and implementation models in the object-oriented paradigm, e.g., Java 2 SE, design patterns, software testing methods.

**Lecture:** Software Engineering I

**SWS:** 2 **ECTS:** 4

**Exercise:** Software Engineering I Exercise

**SWS:** 2 **ECTS:** 2

## 6.17 Software Engineering 2

The module covers a selection of advanced topics in the field of software engineering, such as software quality assurance, service engineering, virtualization, programming languages and design, and formal methods in system design.

**Lecture:** Software Engineering II

**SWS:** 2 **ECTS:** 2

## Exercise: Software Engineering II Exercise

SWS: 2 ECTS: 4

### 6.17.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_programming_1	Understanding the fundamental concepts of software engineering	Computer Science		Demonstrate understanding through theoretical assessments and practical examples	University of Potsdam	<a href="#">Link</a>
gen_programming_2	Applying various approaches of software engineering	Computer Science	gen_programming_1	Complete assignments or projects using different software engineering methods	University of Potsdam	<a href="#">Link</a>
gen_programming_3	Identifying and utilize essential technologies and tools for specification, component-based development, and quality assurance of modern software systems	Computer Science	gen_programming_1	Work with selected tools and technologies in practical exercises and case studies	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
gen_programming_4	<b>Demonstrating</b> an in-depth understanding and ability to apply various approaches of software engineering	Computer Science	gen_programming_1 gen_programming_2	Successfully complete advanced projects employing different software engineering methods	University of Potsdam	<a href="#">Link</a>
gen_programming_5	<b>Understanding</b> the characteristics of a wide range of technologies and tools for specification, component-based development, and quality assurance of modern software systems, and apply them in various contexts	Computer Science	gen_programming_3	Apply appropriate technologies and tools in complex case studies and demonstrate their use in different application scenarios	University of Potsdam	<a href="#">Link</a>

## 6.17.2 Sources & Implementations:

### 6.17.2.1 Curricula

- [Computing Curricula 2020](#)

### 6.17.2.2 Courses

- [Software Engineering I](#)

### 6.17.2.3 Programs

- [UP Computational Science Master](#)

## 6.18 Example Module: Fundamentals of Computer Science

This is an example module to showcase the integration pipeline

### 6.18.1 Basics of Computer Science

#### 6.18.1.1 Basic Concepts

- Introduction to computer science, basic concepts of operating systems using UNIX/Linux as an example
- From problem to algorithm: concept of an algorithm, design of algorithms, pseudocode, refinement, brute-force algorithms, models and modeling, graphs and their representation, simple algorithms on graphs, analysis of algorithms (correctness, termination, runtime)
- Implementation of algorithms (e.g., using Python)
- Programming paradigms: procedural, object-oriented, and functional programming; recursion versus iteration
- From program to process: assembly languages, assembler, compiler, interpreter, syntax and semantics of programming languages
- Limits of algorithms: computability, decidability, undecidability

**Lecture:** Basic Programming

**SWS:** 2 **ECTS:** 2

**Exercise:** Basic Programming Exercise

**SWS:** 2 **ECTS:** 4

### 6.18.2 Applied Programming

#### 6.18.2.1 Procedural Programming Concepts

*Programming with an imperative-procedural language (such as C):*

- Data types, type casting, control structures, functions and procedures, parameter passing paradigms, call stack
- Pointers, arrays, strings, structured types



- Errors and their handling
- Dynamic memory management
- Program libraries

#### 6.18.2.2 Programming in an Object-Oriented Language (e.g., Java)

- Classes, objects, constructors
- Inheritance, polymorphism, abstract classes/interfaces
- Exceptions and exception handling
- Namespaces (packages)
- Generic classes and types
- Program libraries

**Lecture:** Applied Programming

**SWS:** 2 **ECTS:** 2

**Exercise:** Applied Programming Exercise

**SWS:** 2 **ECTS:** 4

#### 6.18.3 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
ex_programming_mod1	Using modular programming in imperative-procedural programming language (e.g., C) and an object-oriented language (e.g., Java) with confidence	Computer Science		Submit working programs in both languages demonstrating syntax and language-specific features	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
ex_programming_mod1C3	Implement basic data structures and algorithms	Computer Science	ex_programming_mod1C2	Submit a project with implemented algorithms and data structures (e.g., lists, trees, sorting)	University of Potsdam	<a href="#">Link</a>
ex_programming_mod1C3	Distinguish between error types and handle them appropriately in code	Computer Science	ex_programming_mod1C2	Demonstrate error handling techniques in submitted code (e.g., input validation, error codes, exceptions)	University of Potsdam	<a href="#">Link</a>
ex_programming_mod1C4	Identify and use appropriate library functions in programming tasks	Computer Science	ex_programming_mod1C3	Integrate external libraries in coding tasks and document their usage	University of Potsdam	<a href="#">Link</a>
ex_programming_mod1C5	Use basic functions and mechanisms of operating systems using UNIX/Linux as an example	Computer Science	ex_programming_mod1C4	Demonstrate file handling, permissions, and process control using UNIX/Linux commands	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
ex_prog_createing_mod1C6	Creating and refine simple algorithms using semi-formal notation	Computer Science		Submit pseudocode or flowcharts for given algorithmic problems	University of Potsdam	<a href="#">Link</a>
ex_prog_evaluating_and_mod1C7	Evaluating and compare algorithms using runtime analysis	Computer Science	ex_programming_mod1C6	Provide time complexity comparisons for multiple algorithmic solutions	University of Potsdam	<a href="#">Link</a>
ex_prog_implement_mod1C8	Implement simple algorithms using imperative and functional programming styles (e.g., in Python)	Computer Science	ex_programming_mod1C6	Submit code demonstrating both imperative and functional styles for the same problem	University of Potsdam	<a href="#">Link</a>
ex_prog_distinguish_mod1C9	Distinguishing between programming paradigms and identify their characteristics	Computer Science	ex_programming_mod1C6	Classify given code snippets by paradigm and justify the classification	University of Potsdam	<a href="#">Link</a>

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
C10	Express simple programs in an assembly language	Computer Science		Translate simple high-level logic into assembler code	University of Potsdam	<a href="#">Link</a>
C11	Discuss the limits of algorithms, including computability and decidability	Computer Science		Write a short essay or present on concepts such as the Halting Problem or undecidability	University of Potsdam	<a href="#">Link</a>

#### 6.18.4 Sources & Implementations:

##### 6.18.4.1 Curricula

- [Computing Curricula 2020](#)

##### 6.18.4.2 Courses

- [UP Grundlagen der Programmierung](#)
- [UP Praxis der Programmierung](#)
- [Python for Psychologists](#)
- [Grundlagen der Informatik](#)

##### 6.18.4.3 Programs

- [UP Computational Science Master](#)

## 6.19 Wildcard Computational Science

This module offers RSE students the opportunity to deepen their understanding of computational methods specific to a science discipline. Students choose a science module — such as physics, chemistry, biology, or earth sciences — and engage with its computational practices, core questions, and data/software challenges.

The goal is to apply the general competences acquired in the general programming and software engineering courses to the practices and special needs of the chosen discipline. Computational Physics might face different algorithmic or conceptual challenges than computational chemistry. This module is intended for the case that the institution offers such a specialized computational course.

**Lecture:** Computational Wildcard Science

**SWS:** 2 **ECTS:** 3

**Lab:** Computational Wildcard Science Lab

**SWS:** 2 **ECTS:** 3

### 6.19.1 Module Competences

ID	Description	Disciplines	Prerequisites	Evidence	Author	Source
None		Wildcard Domain Science				<a href="#">Link</a>

### 6.19.2 Sources & Implementations:

#### 6.19.2.1 Programs

- [Uni Potsdam Master Bioinformatik](#)

## 7 Glossary

**C** A general-purpose programming language often used for system-level development.

**Cpp** C++ — an extension of C that supports object-oriented programming.

**DIST** Software distribution — the practice of packaging and delivering software and its dependencies.

**DOCBB** Documentation and best practices — ensuring code is understandable and maintainable.

**DOMREP** Domain repositories — platforms that store and share domain-specific research data.

**HPC** High-Performance Computing — using supercomputers and parallel processing for complex tasks.

**MOD** Modularity — the design principle of separating software into interchangeable, functional components.

**NEW** Novel research — work that contributes original insights to a scientific domain.

**PM** Project Management — planning, executing, and overseeing projects effectively.

**Python** A high-level programming language widely used in data science and scripting.

**R** A programming language and environment for statistical computing and graphics.

**RSE** Research Software Engineer — someone who applies software engineering skills to scientific research.

**SP** Software publication — the process of preparing and disseminating software artifacts.

**SRU** Software reuse — the practice of using existing software components in new projects.

**STEM** Science, Technology, Engineering, and Mathematics.

**SWREPOS** Software repositories — systems for storing and managing software code and versions.

**TEAM** Teamwork — the ability to collaborate effectively in a group setting.

**TEACH** Teaching — the skill of communicating knowledge and helping others learn.

**USERS** End users — the scientists or researchers who rely on software tools.