

Winter Term 21/22

# Adversarial Self-Supervised Learning with Digital Twins

## Org & Introduction

Prof. Dr. Holger Giese ([holger.giese@hpi.uni-potsdam.de](mailto:holger.giese@hpi.uni-potsdam.de))

Christian Medeiros Adriano ([christian.adriano@hpi.de](mailto:christian.adriano@hpi.de)) - **“Chris”**

He Xu ([he.xu@hpi.de](mailto:he.xu@hpi.de))

- Weekly Hours: **4**
- Credit Points: **6**
- Teaching Form: **Project Seminar**
- Enrolment Type: **Compulsory Elective Module** (“Wahlpflichtmodul”)
- Course Language: **English**
- Study Programs and Modules:
  - **IT-Systems Engineering MA**
    - Mandatory module : „*IT-Systems Engineering Analysis*”
    - Mandatory module: „*IT-Systems Engineering Design*”
    - Specialization module(s): „*Software Architecture & Modeling Technology*”
  - **Data Engineering MA**
  - **Digital Health MA**

- Enrollment deadline: **22.10.2021**
  - Cancellation deadline for enrollment: **30.01.2022**
  
- Introductory meeting: **27.10.2021 [NOW]**
  
- Meetings:
  - *Lectures - scheduled*
  - *Update meetings – on demand, usually weekly*
  
- Final Presentations at end of the semester: **To be decided**
  - *We will be present at the lecture room, but we will also be joining via Zoom.*

# Communicantion Plan

Motive	Content	Medium
<b>Artifacts</b>	Source code, Data Documentation, Wiki	Github - <a href="https://github.com/orgs/hpi-sam/">https://github.com/orgs/hpi-sam/</a>
<b>Papers</b>	Copyrighted material	Bib-Admin
<b>Messaging ad hoc</b>	Questions, Suggestions, Sharing	Our Slack group: <a href="https://adversarialre-o743758.slack.com">adversarialre-o743758.slack.com</a>
<b>Official communications</b>	Schedule, Orientations, Administrative issues	Email <a href="mailto:christian.adriano@hpi.de">christian.adriano@hpi.de</a>
<b>Meetings</b>	Lectures, Status, Work meetings	Zoom, Skype
<b>Emergency</b>	Call, SMS, messaging	Chris mobile number (check Chris' Slack profile)

- Work **alone or in groups** on **one selected topic/project**.
- Each team has on-demand update meetings.

## **Project Execution: [60% of final grade]**

- Weekly update meeting
- Intermediary Presentations

## **Written deliverables: [30% of final grade]**

- Final report on findings
  - Length: approx. 10 pages ACM Format per team participants
  - Some parts must be attributable to each individual author

## **Final Presentations: [10% of final grade]**

- Presentation on findings
- Questions and feedback for other students' presentation

# Road Map (1/2)

1. Intro and Course Organization
2. Self-Adaptive Systems (mRubis)

**Week-1  
Organization**

## Objectives

Setup the environment  
Form groups

3. Digital Twins
4. Model-Free Reinforcement Learning
5. Model-Based Reinforcement Learning

**Week-2  
Foundations**

Acquire conceptual knowledge  
General architecture of projects

6. Underspecification & Generalization
7. Simulation and Sim2Real
8. Robust RL
9. Safe RL

**Week-3  
Challenges**

Understand a phenomenon  
Select and describe the  
problems

10. Adversarial Training
11. Continual & Curriculum Learning
12. Transfer & Meta-Learning RL
13. Representation Learning & Causal RL

**Week-4  
Solutions**

Understand solution space  
(degrees of freedom)  
Select solutions  
Choose evaluation methods  
Plan experiments

## ■ **Project Phase 1: Learn fundamentals - Lectures**

- Goal: learn fundamentals
- Deadline: Mid-End of December

## ■ **Project Phase 2: Present Proposal - Reading and Writing**

- Goal: learn about the state of art of one application area

## ■ **Project Phase 3: Apply a method - Coding and Evaluation**

- Goal: learn to apply and evaluate a method
- Present update in weekly meetings

## ■ **Final Presentations** in one session in **February 2022**

## ■ **Submission of final report** one week after the presentation

**Team size:** up to four people.

## **Project proposal in two stages:**

### **1- State-of-art** (one page, double column) – in 6 weeks (First week of December)

- Each person covers at least five well-selected papers (group covers at least 20 papers)

### **2- Plan** - first draft in 8 weeks (before New Years Break)

- Detail the problem (what is it? why should I care?, why is it challenging?)
- Describe the dataset (source, size, main features, cite any papers that used it)
- Determine the metrics and algorithms to be used (preliminary insights, it might change)
- Discuss how you will evaluate your results (benchmarks and baselines)



END