

Reasoning and Decision Making under Uncertainty (5171040)

Module name english	Reasoning and Decision Making under Uncertainty					
Type of module	Pflichtmodul		Responsible for module		Prof. Dr. Frank Deinzer	
Lecturer	Prof. Dr. Frank Deinzer					
Language of instruction, L. of examination	Englisch		Semester		1	
SWS	4		Teaching and learning formats		Seminaristischer Unterricht	
ECTS-Credits	5		Type of examination		Portfolio	
Bonus benefits						
Workload	Workload (Total)	150	Attendance time	60	Self-Study time (incl. exam preparation)	90
Duration of module	1 Semester		Frequency		ME/OE	
Type of grading	Differenzierte Note		Verwendbarkeit		Artificial Intelligence	
Conditions for participation	none					
Recommended prerequisites						
Module's learning outcomes	<ul style="list-style-type: none"> - Students develop further knowledge and skills on the necessary mathematical foundations for understanding and developing algorithms for AI. - Students can apply the principles of Reinforcement Learning algorithms - Students can use the principles of modelling gents, environments and rewards. - Students understand the necessity of function approximations in learning. - Students understand the concepts of statistical sensor fusion - Students can realize sensor fusion applications - Students build on their acquired knowledge to master learning problems. 					
Module content	<p>The course is composed of 2 thematic blocks.</p> <p>Block A: Reinforcement Learning</p> <ol style="list-style-type: none"> 1. Basic Reinforcement Learning Concepts <ul style="list-style-type: none"> - Actions and States - Goals, Rewards, Returns and Episodes - Policies and Value Functions 2. Basic Reinforcement Learning Methods <ul style="list-style-type: none"> - Finite Markov Decision Processes - Dynamic Programming - Monte Carlo Methods 3. Advanced tabular learning Methods <ul style="list-style-type: none"> - Temporal-Difference Learning - Bootstrapping Methods 4. Learning in Continuous State and Action Spaces <ul style="list-style-type: none"> - On-Policy Approximation - Value-function Approximation - Off-Policy Approximation - Approximate Eligibility Traces 5. Value Function Approximation Case Studies <ul style="list-style-type: none"> - Computer Vision: Action planning - Mastering Games: Backgammon, Go 6. Applications and Exercises <p>Block B: Sensor Fusion</p> <ol style="list-style-type: none"> 1. Using Bayes for Sensor Data Fusion <ul style="list-style-type: none"> - Modeling and Estimation of Densities - Sensor Fusion over Time 2. Hidden Markov Models and Viterbi Algorithm 3. Recursive State Estimation <ul style="list-style-type: none"> - Gaussian Filters - Nonparametric Filters 4. Applications 					

Literature

1. Sutton, Barto. Reinforcement Learning - An Introduction. Bradford Books, 2018
2. Thorp. Beat the Dealer. Random House. 1966
3. Mitchell. Data Fusion: Concepts and Ideas. Springer. 2014
4. Thrun, Burgard, Fox: Probabilistic Robotics. MIT Press. 2005
5. Johnson, Freund, Miller. Miller & Freund's Probability and Statistics for Engineers. Pearson

Further specialized literature will be announced in the course.