



<b>Title:</b>	Robotics & Autonomous Systems <b>APPROVED</b>
<b>Long Title:</b>	Robotics & Autonomous Systems
<b>Module Code:</b>	COMP9069
<b>Duration:</b>	1 Semester
<b>Credits:</b>	5
<b>NFQ Level:</b>	Expert
<b>Field of Study:</b>	Computer Science
<b>Valid From:</b>	Semester 1 - 2019/20 ( September 2019 )
<b>Module Delivered in</b>	<a href="#">2 programme(s)</a>
<b>Module Coordinator:</b>	Sean McSweeney
<b>Module Author:</b>	Sean McSweeney
<b>Module Description:</b>	Robotics and autonomous systems has the potential to transform many industries such as manufacturing, construction and logistics. Traditional automated system design requires highly controlled more-or-less stationary environments for correct operation, such systems have a limited number of applications. The integration of machine learning into robotic systems allows robots to overcome this constraint and thus operate in unconstrained environments. Recent development in robotics middle-ware that facilitate the application of machine learning approaches has allowed the development of robots that can modify behaviour with changing environmental conditions, continuously improve operation and adapt to system failures. This module will focus on utilizing contemporary robotics middle-ware and the application of machine learning to both articulated systems (e.g. robotic arms) and autonomous systems (e.g. quad-copters and rovers).
<b>Learning Outcomes</b>	
<i>On successful completion of this module the learner will be able to:</i>	
LO1	Develop and simulate models for articulated and autonomous robotic systems.
LO2	Evaluate the applicability of machine learning in robotics.
LO3	Adapt machine learning algorithms to robotic motion control and autonomous applications.
LO4	Appraise the application of deep learning to robotic systems.
<b>Pre-requisite learning</b>	
<b>Module Recommendations</b>	
<i>This is prior learning (or a practical skill) that is strongly recommended before enrolment in this module. You may enrol in this module if you have not acquired the recommended learning but you will have considerable difficulty in passing (i.e. achieving the learning outcomes of) the module. While the prior learning is expressed as named CIT module(s) it also allows for learning (in another module or modules) which is equivalent to the learning specified in the named module(s).</i>	
<b>Incompatible Modules</b>	
<i>These are modules which have learning outcomes that are too similar to the learning outcomes of this module. You may not earn additional credit for the same learning and therefore you may not enrol in this module if you have successfully completed any modules in the incompatible list.</i>	
No incompatible modules listed	
<b>Co-requisite Modules</b>	
No Co-requisite modules listed	
<b>Requirements</b>	
<i>This is prior learning (or a practical skill) that is mandatory before enrolment in this module is allowed. You may not enrol on this module if you have not acquired the learning specified in this section.</i>	
No requirements listed	



**Module Content & Assessment**

**Indicative Content**

**Modelling and Simulating Robots and Autonomous Systems**

Spatial descriptions and transformations, forward kinematics, inverse kinematics, jacobian matrices, modelling non-rigid robots, autonomous system kinematics. Uncertainty in robotic models. Simulation and programming tools and environments such as V-REP, ROS, Gazebo.

**Reinforcement Learning**

Elements of RL, Finite Markov Decision Processes, Policies and Value Functions, Partially Observable MDPs, Inverse Reinforcement Learning, Bellman Equations, Optimal Value Functions, Model Based vs Model Free Algorithms, Dynamic Programming, Monte Carlo Methods, Temporal-Difference Prediction and Q Learning.

**Reinforcement Learning in Robotic Systems**

Searching for parametric motor primitives, adapting parametric motor primitives to changing conditions, control prioritisation for motor primitives. Autonomous systems map building, localisation, path planning, obstacle avoidance and navigation in dynamic environments.

**Deep Reinforcement Learning in Robotics**

Radial Basis Function Artificial Neural Networks, Policy Gradient, TD Lambda, and Deep Q-Learning applications in robotic systems. Usage of OpenAI Gym, Tensorflow.

Assessment Breakdown	%
Course Work	100.00%

Course Work				
Assessment Type	Assessment Description	Outcome addressed	% of total	Assessment Date
Project	Project developing a simulation model of an articulated or autonomous robotic system and evaluating of the fidelity of the model developed.	1,2	40.0	Week 7
Project	Project applying machine learning to a robotic or autonomous system, iterating and evaluating the methodology applied to environmental and system changes.	3,4	60.0	Sem End

No End of Module Formal Examination

**Reassessment Requirement**

**Coursework Only**

*This module is reassessed solely on the basis of re-submitted coursework. There is no repeat written examination.*

**The institute reserves the right to alter the nature and timings of assessment**

**Module Workload**

<b>Workload: Full Time</b>				
<i>Workload Type</i>	<i>Workload Description</i>	<i>Hours</i>	<i>Frequency</i>	<i>Average Weekly Learner Workload</i>
Lecture	Lecture delivering theory underpinning learning outcomes.	2.0	Every Week	2.00
Lab	Practical computer-based lab supporting learning outcomes.	2.0	Every Week	2.00
Independent & Directed Learning (Non-contact)	Independent & directed learning	3.0	Every Week	3.00
Total Hours				7.00
Total Weekly Learner Workload				7.00
Total Weekly Contact Hours				4.00

<b>Workload: Part Time</b>				
<i>Workload Type</i>	<i>Workload Description</i>	<i>Hours</i>	<i>Frequency</i>	<i>Average Weekly Learner Workload</i>
Lecture	Lecture delivering theory underpinning learning outcomes.	2.0	Every Week	2.00
Lab	Practical computer-based lab supporting learning outcomes.	2.0	Every Week	2.00
Independent & Directed Learning (Non-contact)	Independent & directed learning	3.0	Every Week	3.00
Total Hours				7.00
Total Weekly Learner Workload				7.00
Total Weekly Contact Hours				4.00

Module Resources
<i>Recommended Book Resources</i>
<ul style="list-style-type: none"> <li>• Sutton, Richard S and Barto, Andrew G 1998, <i>Reinforcement learning: An introduction</i>, MIT press Cambridge [ISBN: 9780262193986]</li> </ul>
<i>Supplementary Book Resources</i>
<ul style="list-style-type: none"> <li>• Jens Kober and Jan Peters 2014, <i>Learning Motor Skills From Algorithms to Robot Experiments</i>, Springer International Publishing [ISBN: 9783319031941]</li> <li>• Todd Hester 2013, <i>TEXPLORE: Temporal Difference Reinforcement Learning for Robots and Time-Constrained Domains</i>, Springer International Publishing [ISBN: 9783319011677]</li> </ul>
<i>Recommended Article/Paper Resources</i>
<ul style="list-style-type: none"> <li>• Kober, Jens and Bagnell, J Andrew and Peters, Jan 2013, <i>Reinforcement learning in robotics: A survey</i>, The International Journal of Robotics Research, 32, no 11, pp 1238-1274</li> <li>• Cully, Antoine and Clune, Jeff and Tarapore, Danesh and Mouret, Jean-Baptiste 2015, <i>Robots that can adapt like animals</i>, Nature Research, 521, pp 503-507</li> <li>• Ijspeert, Auke Jan 2008, <i>Central pattern generators for locomotion control in animals and robots: a review</i>, Elsevier Journal on Neural networks, Vol 21, No 4, pp 642-653</li> </ul>
<i>Supplementary Article/Paper Resources</i>
<ul style="list-style-type: none"> <li>• Chatzilygeroudis, Konstantinos and Rama, Roberto and Kaushik, Rituraj and Goepf, Dorian and Vassiliades, Vassilis and Mouret, Jean-Baptiste 2017, <i>Black-Box Data-efficient Policy Search for Robotics</i>, IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)</li> <li>• Cutler, Mark and How, Jonathan P 2015, <i>Efficient reinforcement learning for robots using informative simulated priors</i>, IEEE International Conference on Robotics and Automation (ICRA), pp 2605-2612</li> <li>• Abbeel, Pieter and Coates, Adam and Quigley, Morgan and Ng, Andrew Y 2007, <i>An application of reinforcement learning to aerobatic helicopter flight</i>, Advances in neural information processing systems, pp 1-8</li> </ul>
<i>Other Resources</i>
<ul style="list-style-type: none"> <li>• Website: 2017 Curated List of Reinforcement Learning Applications in Robotics <a href="https://github.com/Phylliade/awesome-mac-hine-learning-robotics">https://github.com/Phylliade/awesome-mac-hine-learning-robotics</a></li> <li>• Website: 2017 Curated List of Reinforcement Learning Resources <a href="https://github.com/aikorea/awesome-rl#bo oks">https://github.com/aikorea/awesome-rl#bo oks</a></li> <li>• Website: 2017 Curated list of open source robotics simulators and libraries. <a href="https://github.com/jslee02/awesome-robot-ics-libraries">https://github.com/jslee02/awesome-robot-ics-libraries</a></li> </ul>

**Module Delivered in**

<b>Programme Code</b>	<b>Programme</b>	<b>Semester</b>	<b>Delivery</b>
CR_EINMS_9	<a href="#"><u>Certificate in Intelligent Manufacturing Systems</u></a>	2	Mandatory
CR_KARIN_9	<a href="#"><u>Master of Science in Artificial Intelligence</u></a>	2	Elective