### 4/1/4. Course Specification:

# **COURSE SPECIFICATIONS**Form

Course Title: Reinforcement Learning

Course Code: 14016474-3

Date: 2018 -10 - 21.	<b>Institution</b> : Umm Al-Qura University
<b>College</b> : College of Computer and Information	Systems <b>Department</b> : Department of Computer Science

## A. Course Identification and General Information

1. Course title and code: Reinforcement Learning 14016474-3			
2. Credit hours: <u>3</u>			
3. Program(s) in which the course is offered	d. Master of Computer Science (Artif	<u>icial Intelligence)</u>	
(If general elective available in many progra	ams indicate this rather than list pro	grams)	
4. Name of faculty member responsible for	r the course <u>Dr. Mohsin Bilal</u>		
5. Level/year at which this course is offered	d: <u>2</u>		
6. Pre-requisites for this course (if any):			
7. Co-requisites for this course (if any):			
8. Location if not on main campus:			
9. Mode of Instruction (mark all that apply)	١٠		
a. Traditional classroom	percentage?	100	
a. Traditional Classiconii	percentage:	100	
b. Blended (traditional and online)	percentage?		
c. E-learning	percentage?		
d. Correspondence	percentage?		
f. Other	percentage?		
Comments:			

#### **B** Objectives

#### 1. The main objective of this course

This course provides knowledge of Reinforcement Learnings and their application in real world applications.

2. Describe briefly any plans for developing and improving the course that are being implemented. (e.g. increased use of the IT or online reference material, changes in content as a result of new research in the field)

The contents will be prepared from globally recognized text books, web-based reference materials and latest research in the field. Practical home works and a term project related to latest tools and techniques will also be designed. At the end of the course, a seminar day can be announced in which students can present their course projects and literature review.

# **C. Course Description** (Note: General description in the form used in the program's bulletin or handbook)

#### **Course Description:**

This course covers a range of adaptive learning systems, reinforcement learning and unsupervised methods, particularly as used in RL systems. By the end of the module the student should have a grasp of modern learning techniques and the issues involved in dealing with real-world data. The main techniques covered in the course are basic reinforcement learning, dynamic programming, Monte Carlo methods, Q-learning, function approximation, unsupervised and constructive methods, radial basis and other local functions, classifier systems as compared to RL systems.

1. Topics to be Covered			
List of Topics	No. of Weeks	Contact hours	
Reinforcement learning framework	1	3	
Bandit problems and action selection	1	3	
Dynamic programming methods	1	3	
Monte-Carlo methods	2	6	
Temporal difference methods	1	3	
Q-learning and eligibility traces	1	3	
Environment modeling	2	6	
Actor-critic applications	1	3	
Planning in the RL context	1	3	
Constructive methods - nets that grow	2	6	
Evaluating performance of algorithms	1	3	

#### 2. Course components (total contact and credit hours per semester):

		Lecture	Tutorial	Laboratory/ Studio	Practical	Other	Total
Contact	Planned	42					42
Hours	Actual	42					42
Cuadit	Planned	3					3
Credit	Actual	3					3

3. Individual study/learning hours expected for students per week.

9-12

4. Course Learning Outcomes in NQF Domains of Learning and Alignment with Assessment Methods and Teaching Strategies

On the table below are the five NQF Learning Domains, numbered in the left column.

<u>First</u>, insert the suitable and measurable course learning outcomes required in the appropriate learning domains (see suggestions below the table). <u>Second</u>, insert supporting teaching strategies that fit and align with the assessment methods and targeted learning outcomes. <u>Third</u>, insert appropriate assessment methods that accurately measure and evaluate the learning outcome. Each course learning outcomes, assessment method, and teaching strategy should fit in together with the rest to form an integrated learning and teaching process. (Courses are not required to include learning outcomes from each domain.)

**Curriculum Map** 

Code	NQF Learning Domains	Course Teaching	Course Assessment	
#	And Course Learning Outcomes	Strategies	Methods	
1.0	Knowledge			
1.1	Understand reinforcement learning framework	Lectures	Exam, Quizzes	
1.2	Understand basic and advanced reinforcement learning techniques	Lectures	Exams, Homework, Quizzes	
1.3	Understand self-growing net algorithms	Lectures	Exam, Homework, Quizzes	
1.4	Identify current tools for reinforcement learning in real world applications	Lectures, Project	Exam, Homework, Project report	
2.0	Cognitive Skills			
2.1	Ability to evaluate how effective a particular RL algorithm is	Lecture, Case studies	Exams, Quizzes, Homework	
2.2	Describe multiple criteria for analyzing RL algorithms and evaluate algorithms on these metrics	Lecture, Case studies,	Exams, Quizzes, Homework	
	Evaluate a real-world problem to be formulated as RL problem	Group discussion, Project	Project Report, Project presentation	
3.0	Interpersonal Skills & Responsibility	<u> </u>	·	

3.1	Work effectively in groups to accomplish a common goal and show leadership qualities	Small group discussion, research activities, Projects	Project Report, Group presentation
3.2	Act ethically and responsibly with high moral standards	Research activities, Project	Anti-plagiarism Presentation
4.0	Communication, Information Technology, Numerical		
4.1	Ability to communicate clearly in oral and written form with range of audiences	Project research activities, Project	Project Report, Group presentation
4.2	Demonstrate the ability to apply recent tools in RL application	Project	Project Report, Project presentation
4.3	Demonstrate the ability to use mathematical and statistical techniques in the design and analysis of RL algorithms	Lecture, Case studies	Exams, homework
5.0	Psychomotor (if any)		
5.1			

5. Assessment Task Schedule for Students During the Semester				
	Assessment task (i.e., essay, test, quizzes, group project, examination, speech, oral presentation, etc.)	Week Due	Proportion of Total Assessment	
1	Quiz 1	2	3.3%	
2	Homework 1	3	3.3%	
3	Quiz 2	4	3.3%	
4	Homework 2	5	3.3%	
5	Midterm Exam	7	20%	
6	Quiz 3	8	3.3%	
7	Homework 3	9	3.3%	
8	Project	11	30%	
9	Final Exam	14	30%	

#### D. Student Academic Counseling and Support

- 1. Arrangements for availability of faculty and teaching staff for individual student consultations and academic counseling. (include the time teaching staff are expected to be available per week)
- iii. Office Hours for student counseling and support Three hours/week
- iv. Availability of teaching Staff on e-learning resources like uqu20/Piazza

#### **E Learning Resources**

- 1. List Required Textbooks
- iii. Csaba Szepesvari and Ronald Brachman, Algorithms for Reinforcement Learning, Morgan and Claypool Publishers, latest edition.
- iv. M.Wiering and M.V.Otterlo, Reinforcement Learning: State-of-the-art, Springer Science & Business Media, latest edition.
- 2. List Essential References Materials (Journals, Reports, etc.)
  - iv. Recent Papers in Reinforcement Learnings related journals
- 3. List Electronic Materials, Web Sites, Facebook, Twitter, etc.

ii.

- 4. Other learning material such as computer-based programs/CD, professional standards or regulations and software.
  - ii. MATLAB, Python, or similar software

#### F. Facilities Required

Indicate requirements for the course including size of classrooms and laboratories (i.e. number of seats in classrooms and laboratories, extent of computer access, etc.)

- 1. Accommodation (Classrooms, laboratories, demonstration rooms/labs, etc.)
  - ii. One classroom (25 seats)
- iii. One lab (25 PCs)
- 2. Technology resources (AV, data show, Smart Board, software, etc.)
- iii. Whiteboard
- iv. Internet connection
- v. Anti-plagiarism software
- 3. Other resources (specify, e.g. if specific laboratory equipment is required, list requirements or attach list)

#### **G** Course Evaluation and Improvement Procedures

1. Strategies for Obtaining Student's Feedback on Effectiveness of Teaching

- v. At the end of semester, course evaluation forms will be filled by the students electronically or on paper. The evaluation forms will be anonymous.
- 2. Other Strategies for Evaluation of Teaching by the Instructor or the Department
  - iii. Course file of the course will be maintained and evaluated by some senior faculty member.
  - iv. Instructor evaluation is performed for every semester
- 3. Procedures for Teaching Development
  - iii. Constant reading of new books and research papers, attending related conferences and workshops, participation in the research groups and blogs etc.
- 4. Procedures for Verifying Standards of Student's Achievement (e.g. check marking by an independent member teaching staff of a sample of student's work, periodic exchange and remarking of tests or a sample of assignments with staff members at another institution)
  - iii. A random sample from the marked papers may be evaluated by an independent senior faculty member.
- iv. Departmental quality assurance committee can review the students grades and course files to make sure that high standard of teaching is maintained.
- 5. Describe the planning arrangements for periodically reviewing course effectiveness and planning for developing it.
  - iii. Department has curriculum committee that periodically review courses.
  - iv. Faculty council review offer program as per need.

Name of Course Instructor: <u>Dr Muhammad Arif</u>

Signature: <u>Dr Muhammad Arif</u>	Date Completed: Oct. 22, 2018	
Program Coordinator:		
Signature:	Date Received:	