Faculty: Information Technology	
Department: Data Science and Artificial Intelligence	Program: Bachelor
Academic year:	Semester:



Course Plan

First: Course Information

Course No.:	Course Title: Advan	iced	Credit Hou	rs: 3	Theoretical:3	Practical:0
1505320 Machine Learning						
Prerequisite No. and Title: 1505311						
Machine Learning	/0300103	Section	<i>No.: 3</i>	Lectur	re Time:	
Statistics and Proba	ıbilities					
Level in JNQF	7					
	sity Requ	irement	☐ Elective University Requirement			
Type Of Course:	□ Obligatory Facult	ement	□ Elective Faculty Requirement			
25pc of commen	■ Obligatory Specialization Requirement			\Box <i>Elec</i>	tive Specialization	Requirement
	☐ Ancillary course					
Type of Learning:	□ Face-to-Face Le ■ Blended Learning □ Online Learning	ng (2 Fac			•	

Second: Instructor's Information

Course Coordinat	or:				
Name:		Academic Rank:			
Office Number:		Extension Number:	Email:		
Course Instructor					
Name:		Academic Rank:			
Office Number:		Extension Number:	Email:		
Office Hours:	Sunday Mondo	ay Tuesday Wednesday	y Thursday		



Third: Course Description

This course intends to extend the basic knowledge and skills that have been introduced in the introduction to machine learning course. The main objective of this course is to provide students with extensive knowledge of unsupervised learning, semi-supervised learning, and reinforcement learning. Moreover, this course introduces three significant data analysis tools which are they: KEEL, WEKA, and Orange.

Fourth: Course Objectives

- 1. Introducing the student to the fundamental concepts of unsupervised learning, semi-supervised learning, reinforcement learning.
- 2. Developing the student's ability to analyze unlabeled datasets.
- 3. Expanding the student's skills of analyzing unlabeled datasets and perform several main tasks such as association, clustering, discretization, and feature selection.
- 4. Providing the student with the skills of utilizing several data analysis tools such as KEEL, Orange, and WEKA.

Fifth: Learning Outcomes

Level descriptor according to (JNQF)	CILOs Code	CILOs If any CLO will not be assessed in the course, mark NA.	Associated PILOs Code Choose one PILO for each CILO*	Assessment method Choose at least two methods
	K1	Outline the main types of learning in machine learning.	PK1	Mid-term ExamFinal Exam
Knowledge	K2	Identify the main machine learning unsupervised tasks such as association analysis and clustering.	PK2	Mid-term ExamFinal Exam
	К3	Describe the main data analysis tools used in the domain of machine learning.	PK4	Mid-term ExamFinal ExamQuizzes
	S1	Improve The ability to analyze unsupervised datasets.	PS1	Mid-term ExamFinal Exam
	S2	Choose the best tool to handle a given problem.	PS2	Mid-term ExamFinal Exam
Skills	S3	Improve the quality of the data by performing the required preprocessing steps.	PS2	Mid-term ExamFinal Exam
	S4	Solving real-world problem through the main steps of the scientific research.	PS3	 Quizzes Mid-term Exam Final Exam
Competencies	C1	Develop effective communication skills needed for group collaboration.	PC1	Course project and participation

^{*}CILOs: Course Intended Learning Outcomes; PILOs: Program Intended Learning Outcomes; For each CILO, the PILO could be the same or different.



P10F004 scs issue:03 Issue Date: 20/10/2023

Sixth: Learning Resources

Main Reference:	Understanding Machine Learning: From Theory to Algorithms							
Author: Shai Shalev- Shai Ben-David	Shwartz and	Issue No.:2 nd	Print: 2014	Publication Year: 2014				
Additional Sources and Websites:								
Teaching Type:	Classroom	☐ Laboratory	□ Workshop	MS Teams Moodle				

Seventh: Course Structure

Lecture Date	Course Intended Teaching Outcomes (CILOs)	Topics	Teaching Procedures	Teaching Methods**	References***	
		Course Syllabus Discussion	Face-to-Face	Lecture, In-class Questions	Moodle	
Week 1	K1	Revision	Face-to-Face	Lecture, In-class Questions	Lecture notes	
	K2	Revision	asynchronous	Video	YouTube	
	K3, S4 K3, S4	Supervised learning special topic: Multi Label Ranking	Face-to-Face	Lecture, In-class Questions	Lecture notes	
Week 2	K3, 54	Label Ranking	Face-to-Face	Lecture, In-class Questions	Lecture notes	
	K3, S2, S4	Multi Label Classification	asynchronous	Video	YouTube	
	K1, K2, S1, S3, S4 K1, K3, S1, S3, S4	Problem Transformation Methods	Face-to-Face	Lecture, In-class Questions	Lecture notes	
Week 3	K1, K3, S1, S3, S4	Algorithm Adaptation Methods	Face-to-Face	Lecture, In-class Questions	Lecture notes	
	K1, K3, S1, S3, S4	Evaluation metrics for Multi Label Classification	asynchronous	Quiz	Moodle	
K3, S4		Introduction to Association Analysis	Face-to-Face	Lecture, In-class Questions	Lecture notes	
Week 4	K3, S4	Apriori Algorithm	Face-to-Face	Lecture, In-class Questions	Lecture notes	
	K3, S4	Predictive Apriori Algorithm	asynchronous	Video	YouTube	
Week 5	K3, S1, S4	FP-growth Algorithm	Face-to-Face	Lecture, In-class Questions	Lecture notes	
WICEN 3	K3, S1, S4	Applications of Association analysis	asynchronous	Self-reading	Internet	
	K3, S1, S4	WEKA tutorial	Face-to-Face	Lecture, In-class Questions	Lecture notes	
Week 6	K3, S1, S4	WEKA tutorial	Face-to-Face	Lecture, In-class Questions	Lecture notes	
	K3, S1, S4	WEKA tutorial	asynchronous	Self-reading	Internet	



	K3, S1, S4	Introduction to Clustering	Face-to-Face	Lecture, In-class Questions	Lecture notes
Week 7	K3, S4	K-means Algorithm	Face-to-Face	Lecture, In-class Questions	Lecture notes
	K2, S1, S2, S4	K-means Algorithm	asynchronous	Self-reading	Internet
K2, S1, S2 Week 8		Hierarchal Clustering Algorithms	Face-to-Face	Lecture, In-class Questions	Lecture notes
WEEK 0	K2, S1, S2, S4	Hierarchal Clustering Algorithms	Face-to-Face	Lecture, In-class Questions	Lecture notes
		Midterm	Exam		
	K2, S1, S2, S4	KEEL tutorial	Face-to-Face	Lecture, In-class Questions	Lecture notes
Week 9	K2, S1, S2, S4	KEEL tutorial	Face-to-Face	Lecture, In-class Questions	Lecture notes
	K2, S1, S2, S4	KEEL tutorial	asynchronous	Video	YouTube
	K2, S1, S2, S4,	Semi-supervised learning	Face-to-Face	Lecture, In-class Questions	Lecture notes
Week 10	K2, S5	Semi-supervised learning	Face-to-Face	Lecture, In-class Questions	Lecture notes
	K2, S4	Semi-supervised learning	asynchronous	Quiz	Moodle
	K2, S1, S2, S4	Introduction to Reinforcement learning	Face-to-Face	Lecture, In-class Questions	Lecture notes
Week 11	K2, S1, S2, S4	Application of Reinforcement learning	Face-to-Face	Lecture, In-class Questions	Lecture notes
K2, S1, S2, S4		Application of Reinforcement learning	asynchronous	Video	YouTube
	K2, S1, S2, S4		Face-to-Face	Lecture, In-class Questions	Lecture notes
Week 12	K2, S1, S2, S4	Policy-based reinforcement learning.	Face-to-Face	Lecture, In-class Questions	Lecture notes
	K2, S1, S2, S4	Model-based reinforcement learning.	asynchronous	Video	YouTube
	K2, S1, S2, S4	Orange tutorial	Face-to-Face	Lecture, In-class Questions	Lecture notes
Week 13	K2, S1, S2, S4	Orange tutorial	Face-to-Face	Lecture, In-class Questions	Lecture notes
	K2, S3	Orange tutorial	asynchronous	Self-reading	Internet
		Revision	Face-to-Face	Lecture, In-class Questions	Lecture notes
Week 14	K1, K2, K3, S1, S2, S3, S4	Revision	Face-to-Face	Lecture, In-class Questions	Lecture notes
		Revision	asynchronous	Self-reading	Internet
		Final E	xam		

 $[\]hbox{*Teaching procedures: (Face-to-Face, synchronous, and asynchronous).}$



^{***} Reference: (Pages of the book, recorded lecture, video....)

^{**} Teaching methods: (Lecture, video....).

Eighth: Assessment Methods

Methods	Online Blended Learning Learning	Face-To- Face	Specific Course Output to be assessed **If any CILO will not be assessed in the course, mark NA.								
			Learning	K1	К2	К3	S1	S2	S3	S4	C1
First Exam											
Second Exam											
Mid-term Exam		30		✓	√	√	✓	✓	✓	√	
Participation											
Asynchronous Activities		5		✓	✓	√	√	√	√	√	
Quizzes		5				✓	✓	✓	✓	✓	
Assignments		5				✓	✓	✓	✓	✓	
Group presentation		5				√	√	✓	✓	✓	✓
Final Exam		50		√	✓	✓	✓	✓	√	✓	
Total out of 100		100									



Ninth: Course Policies

- All course policies are applied to all teaching patterns (online, blended, and face-to-face Learning) as follows:
 - a. Punctuality.
 - b. Participation and interaction.
 - c. Attendance and exams.
- Academic integrity: (cheating and plagiarism are prohibited).

