## 2309 CS462/SE462 Course Outline

Subject Code : CS462, SE462

Subject Title : Machine Learning

Course Type : Elective

Level : 3 Credits : 3

Teaching Activity : Lecture 30 hours

: Programming 15 hours

Prior Knowledge\* : CS110: COMPUTER PROGRAMMING

MATH101: CALCULUS I MATH102: CALCULUS II

MATH104: PROBABILITY AND STATISTICS

MATH100: LINEAR ALGEBRA

Class Schedule

 Class
 Week
 Time
 Classroom
 Date

 D1
 MON
 9:00-11:50
 C404
 2022/09/04-2022/12/17

Instructor : Zhao Qinglin Contact Number : (853)88972306

E-mail Address : qlzhao@must.edu.mo

Office : A212

Office Hour : Monday: 15:00-18:00

Tuesday: 15:00-18:00 Wednesday: 14:30-17:30 Friday: 9:00-10:00

### **COURSE DESCRIPTION**

Machine learning (ML) is one of the hottest fields of research today. It allows computational systems to adaptively improve their performance with experience accumulated from the data observed. It introduces the basics of learning theories, the design and analysis of learning algorithms, and some applications of machine learning.

### **TEXT BOOK**

Book title: Learning From Data

Author/Editor: Yaser S. Abu-Mostafa, Malik Magdon-Ismail, Hsuan-Tien Lin

Edition: 1

ISBN 10: 1600490069 ISBN 13: 9781600490064 Publisher: AMLBook

Date: 2012

## INTENDED LEARNING OUTCOMES

Upon the successful completion of this course students will be able to:

- 1. Understand when machines can learn.
- 2. Understand why machines can learn.
- 3. Understand how machines can learn.
- 4. Design simple algorithms that machines can learn better.
- 5. Design simple machine-learning algorithms by combining predictive features.
- 6. Design simple machine-learning algorithms by distilling hidden features.

# **Weekly Schedule**

Index	Торіс	Hours	Teaching Method
1	Basics of machine learning	3	Lecture
2	The learning problems	3	Lecture
3	Feasibility of learning	3	Lecture/lab
4	Theory of generalization	3	Lecture/lab
5	Linear models	3	Lecture/lab
6	Beyond basic linear models	3	Lecture/lab
7	Combatting overfitting-I	3	Lecture/lab
8	Combatting overfitting-II	3	Lecture/lab
9	Mid-exam	3	Lecture/lab
10	Putting it altogether: support vector machine	3	Lecture/lab
11	Soft-margin support vector machine	3	Lecture/lab
12	Bagging and boosting	3	Lecture/lab
13	Decision tree ensembles	3	Lecture/lab
14	Deep learning fundamentals-I	3	Lecture/lab
15	Deep learning fundamentals-II	3	Lecture/lab

### ASSESSMENT APPROACH

Assessment method	Weight(%)
1. Attendance and assignment	15%
2. Project	10%

3. Mid-term Exam	25%
4. Final Exam	50%
Total	100%

### **Guideline for Letter Grade:**

Marks	Grade	
93 - 100	A+	
88 - 92	A	
83 - 87	A-	
78 - 82	B+	
72 - 77	В	
68 - 71	B-	
63 - 67	C+	
59 - 62	С	
56 - 58	C-	
53 - 55	D+	
50 - 52	D	
Below 50	F	

#### Notes:

Students will be assessed on the basis of continuous assessment (i.e. coursework in the form of individual assignments and group work) and by an end of semester examination.

The coursework assessment items (e.g., assignment, project, and presentation etc.) evaluate students' ability to implement programming solution to practical problems using applying the data structures and algorithms.

Final examination will primarily evaluate students' overall understanding of main topics covered in the course (i.e., how to organize data and analyse algorithms when they develop programs).

### **ADDITIONAL READINGS**

Journals:

Conferences: NIPS, ICML, AAAI

Trade and other Publications:

### Website:

All source codes used in teaching and guidelines are available on moodle.