# HCMC UNIVERSITY OF TECHNOLOGY AND EDUCATION

Faculty of Electrical And Electronic Engineering

# **Department of Industrial Electronics**

## ELECTRONICS AND COMMUNICATION ENGINEERING TECHNOLOGY

**Level: Undergraduate** 

#### **SYLLABUS**

Course name: Machine Learning
 Course code: MALE321063

**3. Credits:** 3 (3/0/6)

Duration: 15 weeks (45h main course and 90h self-study)

#### 4. Instructors:

1- Nguyen Thanh Hai, PhD

2- Nguyen Manh Hung, PhD

3- Ngo Quoc Cuong, MEng

#### 5. Course conditions

Prerequisites: Programing Language

Corequisites: Statistic Theory

#### 6. Course description

This course provides students fundamental knowledge about pattern recognition and machine learning. This course introduces fundamental supervised and unsupervised learning algorithm as well as recommendation system.

#### 7. Course Goals

Goals	Goal description (This course provides students:)	ELOs
G1	Fundamental supervised and unsupervised learning algorithm	01 (H) 07 (M)
G2	An ability to identify, evaluate and analysis an regression/ classification system.	02 (M), 03 (M)
G3	An ability design an regression/ classification system.	10 (L) 11 (H)

<sup>\*</sup> Note: High: H; Medium: M; Low: L

#### 8. Course Learning Outcomes (CLOs)

CLOs		Description (After completing this course, students can have:)	Outcome
	G1.1	Have knowledge about linear regression models	1, 7
	G1.2	Have knowledge about non-linear regression models	1, 7
G1	G1.3	Have knowledge about clustering problem	1, 7
	G1.4	Have knowledge about dimensional reduction	1, 7
	G1.5	Have knowledge about recommendation systems	1, 7

G2.1 Have ability to identify an image based recognition		2, 3	
G2	G2.2	Have ability to evaluate and analysis an natural language based recognition	2, 3
	G3.1	Have ability to design an regression or classification system	10, 11
G3	G3.2	Have ability to validate and verify and an regression or classification sytems	10, 11

## 9. Study materials

#### - Textbooks:

[1] Christopher M. Bishop, *Pattern Reconition and Machine Learning*, 2nd ed Springer, 2007.

#### - References:

- [2] Duda, Richard, Peter Hart, and David Stork. *Pattern Classification*. 2nd ed. New York, NY: Wiley-Interscience, 2000.
- [3] Hastie, T., R. Tibshirani, and J. H. Friedman. *The Elements of Statistical Learning: Data Mining, Inference and Prediction*. New York, NY: Springer, 2001.

#### 10. Sudent Assessments

- Grading points: 10
- Planning for students assessment is followed:

Type	Contents	Linetime	Assessment techniques	CLOs	Rates (%)
	Midterms				50
Exam01	Short exercise	Begin/ end of each week	Short question in class	G1.1- G1.5	10
Exam02	Coding exercise	week 11	Individual paper assessment in class	G2.1- G2.2,	20
Exam03	Project report	week 15	Individual paper assessment in class	G3.1, G3.3,	20
	Final exam				50
Final Exam	Final report	week 16	Project report	G1.1- G1.5 G2.1- G2.2	

#### 11. Course details:

Week	Contents	CLOs
1	Chapter 1: < Machine Learning Introduction> (3/0/6)	

	A/ Contents and teaching methods: (3)	
	Contents:	
	1.1 Machine learning introduction	
	1.2 Unsupervised learning	G1.1
		0111
	1.3 Supervised learning	
	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/Self-study contents: (6) + Interpretation methods	G1.1
	+ Inductive methods	G1.1
	Chapter 2: < LINEAR REGRESSION > (6/0/12)	
	A/ Contents and teaching methods: (3)	
	Contents:	
	2.1 Linear regression	
	2.2 Object function	G1.1
	2.3 Optimization	
2	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/ Self-study contents: (6)	
	+ Object function derivative	G1.1
	+ Exercises	
	Chapter 2: < LINEAR REGRESSION (cont.)> (6/0/12)	
	A/ Contents and teaching methods:(3)	
	Contents:	
	2.4 Multivariable problems	
	2.5 Multivariable features	G1.1
2	2.6 Optimal multivariable function	
3	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/ Self- study contents: (6)	
	+ Standard equation	G1.1
	+ Optimal by standard equation	
	Chapter 3: < CLASSIFICATION> (3/0/6)	
	A/ Contents and teaching methods: (3)	
4	Contents:	G1.1
	3.1 Classification	
	3.2 Object function for classified problems	
	3.3 Optimal objet function for classified problems	

	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/ Self- study contents: (6)	G1.1
	+ Calculating derivation for target function	G1.1
	+ Exercises	
	Chapter 4: < OVERFITING > (3/0/6)	
	A/ Contents and teaching methods: (3)	
	Contents:	
	4.1 Overfitting problem	
	4.2 Identification overfitting problem.	G3.1
_	4.3 Handle overfitting problem	
5	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/ Self- study contents: (6)	
	+ Calculating gradient in a regilization	G3.1 - G3.2
	+ Exercises	33.1 33.2
	Chapter 5: < NON-LINEAR SYSTEMS > (9/0/18)	
	A/ Contents and teaching methods: (3)	
	Contents:	
	5.1 Neuron network	
	5.2 Object function	G1.2
6	Teaching methods:	
	+ Questions and answers	
	+ Guide	
	B/ Self- study contents: (6)	G1.2
	+ Reinforce the knowledge learned	G1.2
	Chapter 5: < NON-LINEAR SYSTEMS (cont.) > (9/0/18)	
	A/ Contents and teaching methods: (3)	
	Contents:	
	5.3 Forward propagation	
	5.4 Backward propagation	G1.2
7	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/ Self- study contents: (6)	
	+ Reinforce the knowledge learned	G1.2
	+ Exercises	
8	Chapter 5: < NON-LINEAR SYSTEMS (cont.) > (9/0/18)	
	(2/0/10)	

	A/ Contents and teaching methods: (3)	
	Contents:	
	5.5 Suport Vector Machine	
	5.6 Using support vector machine  Teaching methods:  + Theoretical lectures  + Questions	G1.2
	B/ Self- study contents: (6)	G1.2
	+ Exercises	G2.1
	Chapter 6: < MODEL SELECTION> (3/0/6)	
	A/ Contents and teaching methods: (3)	
	Contents:	
	6.1 Select model size	
	6.2 Select model parameters	
9	6.3 Dataset Evaluation	G3.2
9	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/ Self- study contents: (6)	
	+ Reinforce the knowledge learned	
	Chapter 7: < CLUSTERING > (3/0/6)	
	A/ Contents and teaching methods: (3)	
	Contents:	
	7.1 Unsupervised learning	
	7.2 Clustering problems	
	7.3 K-means algorithm	G1.3
10	7.4 Graph method	01.5
	Teaching methods:	
	+ Theoretical lectures	
	+ Questions	
	B/ Self- study contents: (6)	
	+ Spectral clustering method	G2.1
	PROJECT REPORT	
	A/ Contents and teaching methods: (3)	
	Contents:	
11	1 Project report	G2.1
	Teaching methods:	G2.2 G3.1
	+ Presentation	U3.1
	+ Evaluation	
·		1

	B/ Self- study contents: (6) + Reinforce the knowledge learned	G2.1 G2.2
	> (6/0/12)	G3.2
	A/ Contents and teaching methods: (3) Contents: 8.1 Dimensonal reduction	
12	8.2 Princible Component Analysis (PCA)  Teaching methods:  + Theoretical lectures  + Questions	G1.4
	B/Self- study contents: (6) + LDA method + Exercises	G1.4
	Chapter 8: < DIMENSIONAL REDUCTION OF DATA (cont.) > (6/0/12)	
13	A/ Contents and teaching methods: (3)  Contents:  8.3 LLE method  8.4 Iso-Map method  Teaching methods:  + Theoretical lectures  + Questions	G1.4
	B/ Self- study contents: (6) + Progam the LLE- Iso Map methods + Exercises	G1.4
	Chapter 9: < RECOMMENDATION SYSTEMS > (3/0/6)	
14	A/Contents and teaching methods: (3) Contents:  9.1 Introduced about recommendation systems 9.2 Hidden models 9.3 Program recommendation systems Teaching methods: + Theoretical lectures + Questions	G1.4
	B/Self- study contents: (6) + Reinforce the knowledge learned + Exercises	G1.4
	PROGRAMMING EXERCISES REPORT	
15	A/ Contents and teaching methods: (3) Contents:	G2.1 G2.2 G3.1

1 Recommendation systems	G3.2
2 Hidden models	
3 Recommendation systems programing	
Teaching methods:	
Report – Assess	
B/Self-study contents: (6)	
+ Reinforce the knowledge learned	G3.1
+ Group Discussion	G3.2

# 12. Learning ethics:

- Home assignments and projects must be done by the students themselves. Plagiarism found in the assessments will get zero point

# 13. First approved date:

# 14. Approval level:

<b>Dean Department</b>	Instructor
------------------------	------------

## 15. Syllabus updated process

200 Synusus apautou process			
1 <sup>st</sup> time: Updated content dated	Instructors		
2 <sup>st</sup> time: Updated content dated	Head of department		