Course Title:

Machine Learning and Lab

2022-1st Semester

[Sylabus]

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	Category	Major selection (major selection)		Department or Division	Department of Statistics	
	Number(section)	47771(01)		Name		
	Title	Machine Learning and Lab		Phone		
	Title	IMachine Learning and Lab	Instructor	E-mail		
Course	Credit(Hours)	3 Credit(4 Hours)				
	Туре	Lecture+Experiment/Practice		Homepage		
	Time(Room)	Tue 06,07,08/33-710, Thu 01/33-710		Office Hours		
	school year	3/4 year	Assistant	name & phone		
Evaluatio		ethod	absolute evaluation			
	☐ Attendance (1	0%) 🗆 Portfolio (0%)	☐ Participation (10%)			
Grading	□ Assignment (2	0%) □ Quiz (0%)	☐ Midterm Report (0%) ☐ Midterm Exam (30%)			
	☐ Final Report ((0%) ☐ Final Exam (30%) ☐ Other (0%)				
Тур	e Lecture a	and Practice , PBL , Foreign Langu	age			
Teaching	Method Lecture	, Practice				
	It is consid	dered plagiarism to draw any idea or	any language	from someone el	se wihout adequately crediting that	
Plagiarism	Plagiarism Policy source in your work. It doesn't matter whether the source is a published author, another student, a Web site without clear authorship, a Web site that sells academic papers, or any other person: Taking credit for antone else's work					
Anv student		, and it is unacceptable in all acaden elements are to contact the instructor to get		-		
		s by calling 02-6490-6273 to discuss t				
		Course C)bjectives			
	s is a foreign langua e will be presented	age class conducted in English. in English)				
1.		of various machine learning method ns of various machine learning metho				
o Understa	nding the evaluatior	n of machine learning methods the performances of machine learning				
o Applying	appropriate machin	e learning methods to real data	ig memous)			
(Building sc	(Building sounds skills in machine learning)					
Common Constitution						
Course Description Textbooks and Reference Materials In this course, based on R or Python, there are many learning						
problems The methodology used is introduced here. The main topics are			An Introduction to Statistical Learning: with applications in R			
decision trees and nerves. Various supervised learning techniques and evaluation methods			G. James, D. Witten, T. Hastie, and R. Tibshirani,			
There are self-learning techniques such as the dimension			Springer.			
reduction method.						
Specialty competency						
	Specialt	y competency		Represe <u>nt</u>	ative competency	
		y competency cal Modeling		Represent	ative competency Primary	

Programming

Specialty competency	Representative competency
Problem Solving	Secondary
Collaboration	
Global Competence	
Ethics in Statistical Practice and Communication	

Course Title: Machine Learning and Practice

2022year 1st Semester

[Weekly Lesson Plan]

Week	Contents	Teaching Method	Teaching Materials	Requirements, Assignments, etc.
One	1. Overview of Machine Learning: 1-1. What is machine learning, 1-2. What is model evaluation is it 1. Introduction to Statistical Learning 1-1. What is Statistical Learning, 1-2. Assessing Model Accuracy		Introduction to Statistical LearningCh. 1 & 2.	
2	2. Linear Model I 2-1. Simple star regression model, 2-2. Multiple Linear Regression Model I 2. Linear Regression Part I 2-1. Simple Linear regression, 2-2. Multiple Linear Regression		Introduction to Statistical LearningCh. 3	
3	3. Linear Model II 3-1. Multiple Linear Regression Model II, 3-2. Extension of the linear model 3. Linear Regression II 3-1. Multiple Linear Regression Part 2, 3-2. Other Considerations in the Regression Models		Introduction to Statistical LearningCh. 3 continued	
4	4. Categorization I 4-1. Introduction to Categorization, 4-2. logistic regression model 4. Classification 4-1. An Overview of Classification, 4-2. Logistic Regression		Introduction to Statistical LearningCh. 4	
5	5. Categorization II 5-1. Generative Model, 5-2. of various categorization methods. compare 5. Classification II 5-1. Generative Models 5-2. A Comparison of Classification Method		Introduction to Statistical LearningCh. 4 continued	
6	6. Resampling 6-1. Cross-Validation, 6-2. bootstrap 6. Resampling Methods 6-1. Cross-Validation, 6-2. Bootstrap		Introduction to Statistical LearningCh. 5	
7	7. Linear Model Selection and Normalization I 7-1. Subset Selection Method, 7-2. shrinkage model 7. Linear Model Selection and Regularization 7-1. Subset Selection, 7-2. Shrinkage Methods		Introduction to Statistical LearningCh. 6	
8	Midterm exam Midterm			
9	9. Linear Model Selection and Normalization II 9-1. Dimensional Reduction, 9-2. High dimensionality and multicollinearity 9. Linear Model Selection and Regularization 9-1. Dimension Reduction Methods, 9-2. Considerations in High Dimensions 10. Tree Model I		Introduction to Statistical LearningCh. 6 Continued	
	10-1. Introduction to Nonlinear Methods, 10-2.		Introduction to	

	Introduction to the tree method		
10	10. Tree Methods I 10-1. An Overview of Non-linear Methods, 10-2. An Overview of Tree Methods	Statistical LearningCh. 7 & 8	

11	11. Tree II 11-1. Bagging method, 11-2. Random Forest, 11-3. booth chat 11. Tree Methods II 11-1. Bagging, 11-2. Random Forests, 11-3. Boosting	Introduction to Statistical LearningCh. 8 continued
12	Supplementary Week	
13	13. Support Vector Machine 13-1. Margin and support vectors, 13-2. support vector machine 13. Support Vector Machine 13-1. Margins, Classifiers, Support vectors, and Support Vector Classifiers, 13-2. Support Vector Machines	Introduction to Statistical LearningCh. 9
14	14. Unsupervised Learning 14-1. Principal Component Analysis, 14-2. Clustering Analysis 14. Unsupervised Learning 14-1. Principal Component Analysis, 14-2. Clustering Methods	Introduction to Statistical LearningCh. 10.
15	15. Neural Networks 15-1. Introduction of Neural Networks, 15-2. of neural networks learning 15. Neural Networks 15-1. An Overview of Neural Networks, 15-2. Neural Network Learning	Lecture Slides
16	Finals Final Exam	