Course number G-LAS12 80010 LE10									
Course title (and course title in English)				Instructor's name, job title, and department of affiliation			Graduate School of Informatics Professor,KASHIMA HISASHI Graduate School of Informatics Senior Lecturer,TAKEUCHI KOH		
Group Interdisciplinary Graduate Courses Field(Classification) Statistics, Informatics and Data Science									
Language of instruction	English		Old group			Number of credits		2	
Number of weekly time blocks	1	Class style	ecture Face-to-fa	ace course)		Yea	r/semesters	2025 ·	First semester
Days and periods	Mon.1	Targ	<mark>et yea</mark> r G	raduate stude	nts	Elig	ible students	For sci	ence students
(Students of Graduate School of Informatics cannot take this course as liberal arts and general education course. Please register the course with your department.)									
[Overview and purpose of the course]									
primarily focus on supervised and unsupervised learning, with an emphasis on supervised learning. The course will cover essential theoretical concepts such as maximum likelihood estimation and Bayesian inference, as well as introduce the concept of Probably Approximately Correct (PAC) learning. Throughout the course, you will gain familiarity with various probabilistic models and predictive algorithms, including logistic regression, perceptrons, and neural networks. Additionally, we will touch upon advanced topics like semi-supervised learning, transfer learning, and sparse modeling, providing you with insights into the latest developments in the field of machine learning. In addition, opportunities for hands-on data analysis exercises will also be provided.									
[Course objectives] Understanding basic concepts, problems, and techniques of statistical learning and some of the recent tenior.									
[Course schedule and contents]]									
 Statistical Machine Learning Introduction to machine learning: historical perspective, basic concepts, and applications Regression and classification: linear regression, logistic regression, and neural networks. Inference framework and statistical learning theory: maximum likelihood estimation, regularization, Bayesian inference, Vapnik-Chervonenkis theory Model selection: performance measures, cross-validation, hyper-parameter selection 									
 Advanced to Semi-supervities Transfer lear Sparse mode Deep neural Graph learning 	opics ised learning ning ling networks ng						tinuo to Statistical		
Continue to Statistical Learning Theory(2)									

Statistical Learning Theory(2)

[Course requirements]

None

[Evaluation methods and policy]

Reports and/or final exam.

[Textbooks]

Instructed during class

[References, etc.]

(References, etc.)

Hastie, Friedman, Tibshirani [®] The Elements of Statistical Learning^a (Springer)

Shai Shalev-Shwartz and Shai Ben-David ^PUnderstanding Machine Learning: From Theory to Algorithms (Cambridge University Press)

[Study outside of class (preparation and review)]

Basic knowledge about probability and statistics

[Other information (office hours, etc.)]