

**City University of Hong Kong
Course Syllabus**

**offered by School of Data Science
with effect from Semester A 2024/25**

Part I Course Overview

Course Title:	Statistical Machine Learning I
Course Code:	SDSC5001
Course Duration:	One Semester
Credit Units:	3
Level:	P5
Medium of Instruction:	English
Medium of Assessment:	English
Prerequisites: <i>(Course Code and Title)</i>	Nil
Precursors: <i>(Course Code and Title)</i>	Nil
Equivalent Courses: <i>(Course Code and Title)</i>	Nil
Exclusive Courses: <i>(Course Code and Title)</i>	Nil

Part II Course Details

1. Abstract

This course focuses on the theoretical foundation and fundamental methods in statistical machine learning, covering the key concepts of the probability theory and statistical inference for machine learning, classical and cutting-edge methods and theories for regression and classification, and popular methods for unsupervised learning.

2. Course Intended Learning Outcomes (CILOs)

(CILOs state what the student is expected to be able to do at the end of the course according to a given standard of performance.)

No.	CILOs	Weighting (if applicable)	Discovery-enriched curriculum related learning outcomes (please tick where appropriate)		
			A1	A2	A3
1.	Summarize the key concepts of statistical inference theory for machine learning	20%	✓		
2.	Build fundamental statistical regression and classification models	20%	✓		
3.	Formulate kernel machines and regularization forms	20%	✓	✓	
4.	Describe the underlying bias-variance tradeoff in each learning method	20%	✓	✓	
5.	Apply taught machine learning methods to solve real life data analytics problem via software	20%		✓	✓
		100%			

A1: Attitude

Develop an attitude of discovery/innovation/creativity, as demonstrated by students possessing a strong sense of curiosity, asking questions actively, challenging assumptions or engaging in inquiry together with teachers.

A2: Ability

Develop the ability/skill needed to discover/innovate/create, as demonstrated by students possessing critical thinking skills to assess ideas, acquiring research skills, synthesizing knowledge across disciplines or applying academic knowledge to real-life problems.

A3: Accomplishments

Demonstrate accomplishment of discovery/innovation/creativity through producing /constructing creative works/new artefacts, effective solutions to real-life problems or new processes.

3. Learning and Teaching Activities (LTAs)

(LTAs designed to facilitate students' achievement of the CILOs.)

LTA	Brief Description	CILO No.					Hours/week (if applicable)
		1	2	3	4	5	
Lecture	In lectures, students will learn fundamental theories and principles in statistical machine learning.	✓	✓	✓	✓		26 hours/semester
Laboratory	In labs, students will learn software of statistical machine learning and apply taught methods to address real-world problems.		✓	✓	✓	✓	13 hours/semester

4. Assessment Tasks/Activities (ATs)

(ATs are designed to assess how well the students achieve the CILOs.)

Assessment Tasks/Activities	CILO No.					Weighting	Remarks
	1	2	3	4	5		
Continuous Assessment: <u>50</u> %							
<u>Group Project</u>		✓	✓		✓	20%	
<u>Homework assignments</u>	✓	✓	✓	✓		30%	
Examination: <u>50</u> % (duration: <u>2 hours</u> , if applicable)							
<u>Examination</u>	✓	✓	✓	✓		50%	
						100%	

5. Assessment Rubrics

(Grading of student achievements is based on student performance in assessment tasks/activities with the following rubrics.)

Applicable to students admitted before Semester A 2022/23 and in Semester A 2024/25 & thereafter

Assessment Task	Criterion	Excellent (A+, A, A-)	Good (B+, B, B-)	Fair (C+, C, C-)	Marginal (D)	Failure (F)
1. Group Project	20%	High	Significant	Moderate	Basic	Not even reaching marginal levels
2. Homework assignments	30%	High	Significant	Moderate	Basic	Not even reaching marginal levels
3. Examination	50%	High	Significant	Moderate	Basic	Not even reaching marginal levels

Applicable to students admitted from Semester A 2022/23 to Summer Term 2024

Assessment Task	Criterion	Excellent (A+, A, A-)	Good (B+, B)	Marginal (B-, C+, C)	Failure (F)
1. Group Project	20%	High	Significant	Basic	Not even reaching marginal levels
2. Homework assignments	30%	High	Significant	Basic	Not even reaching marginal levels
3. Examination	50%	High	Significant	Basic	Not even reaching marginal levels

Part III Other Information (more details can be provided separately in the teaching plan)

1. Keyword Syllabus

(An indication of the key topics of the course.)

- Review of Probability Theory
- Statistical Inference: Method of Moments, Maximum Likelihood Estimation, Bootstrap
- Regression: Linear and Nonlinear Regression, Kernel Smoothing, Local Polynomial, Cubic Splines, Regression Splines, Gaussian Process Regression
- Classification: Misclassification Error, Discriminant Analysis, Logistic Regression, CART, Bagging, Random Forest, Boosting
- Kernel Machines: Loss Functions, Regularization Form, SVM
- Unsupervised Learning: Principle Component Analysis, Clustering

2. Reading List

2.1 Compulsory Readings

(Compulsory readings can include books, book chapters, or journal/magazine articles. There are also collections of e-books, e-journals available from the CityU Library.)

1.	Statistical Inference by George Casella and Roger L. Berger
2.	The Elements of Statistical Learning by Hastie, Tibshirani, and Friedman, Springer
3.	Lecture Notes

2.2 Additional Readings

(Additional references for students to learn to expand their knowledge about the subject.)

NIL