City University of Hong Kong Course Syllabus

offered by School of Data Science with effect from Semester A 2022/23

Part I Course Overview

Course Title:	Advanced Statistics
Course Code:	SDSC8004
Course Duration:	One semester
Credit Units:	3
Level:	R8
Medium of Instruction:	English
Medium of Assessment:	English
Prerequisites:	Nil
Precursors:	Nil
Equivalent Courses:	Nil
Exclusive Courses:	Nil

1. Abstract

This course aims to provide students with a solid foundation of statistical concepts, theory, and methods including probability theory, statistical estimation and inference methods, and multivariate statistics. It also aims to provide students with a rigorous introduction to the theory and implementation of statistical models such as linear models, generalized linear models, and nonparametric models. Emphasis will be placed on rigorous mathematical derivations, understanding of the fundamentals of statistics but implementation of the statistical methods via computer programming in R or Matlab will be an important part of the course as well.

2. Course Intended Learning Outcomes (CILOs)

No.	CILOs	Weighting*		very-eni	
		(if	curricu	ılum rel	lated
		applicable)		ig outco	
				e tick	where
			approp	riate)	-
			Al	A2	A3
1.	Prove and apply various fundamental results in probability	20%	\checkmark	\checkmark	
	theory.				
2	Understand and implement the techniques of parametric	250/			
2.	inference such as maximum likelihood estimation and Bayesian inference.	25%	v	v	
3.	Understand and derive key results in the theory of linear models and linear model selection.	25%	\checkmark	\checkmark	
	Understand and derive key results in the theory of				
4.	nonparametric statistical models and methods such as the	20%	\checkmark	\checkmark	
	bootstrap method and Gaussian process regression.				
5.	Implement statistical inference methods and modelling	10%	\checkmark	~	
5.	methodologies with computer codes.	1070		•	
		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 self-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. Teaching and Learning Activities (TLAs)

TLA	Brief Description	CIL	CILO No.		Hours/week (if applicable)		
		1	2	3	4	5	
Large Class Activities	Lectures, software training, and in-class exercises	\checkmark	\checkmark	~	~	~	26 hours/semester
Assignments	Exercises provide students with the opportunities to familiarize and apply the statistical tools learnt during the lectures	~	~	\checkmark	\checkmark	\checkmark	12 hours/semester

4. Assessment Tasks/Activities (ATs)

Assessment Tasks/Activities	CILO No.				-	Weighting*	Remarks
	1	2	3	4	5		
Continuous Assessment: <u>75</u> %							
Midterm						25%	
Students will be assessed via the midterm							
their understanding of concepts, theory,	\checkmark	\checkmark	\checkmark	\checkmark			
and methods learned in class, textbooks,	•	v	•	•			
reading materials and their ability to apply							
subject-related knowledge.							
Two assignments						50%	
Students will work individually to derive							
or prove results in probability and	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark		
statistical theory, and apply statistical	·	·	·		·		
methods to analyse data with the help of							
software.							
Examination: 25 % (duration: 3 Hours							
Examination						25%	
Students will be assessed via the							
examination their understanding of							
concepts, theory, and methods learned in	\checkmark	\checkmark	\checkmark	\checkmark			
class, textbooks, reading materials and							
their ability to apply subject-related							
knowledge.							
						1000/	

100%

5. Assessment Rubrics

Applicable to students admitted in Semester A 2022/23 and thereafter

Assessment Task	Criterion	Excellent (A+, A, A-)	Good (B+, B)	Marginal (B-, C+, C)	Failure (F)
1. Coursework	Midterm and assignments	High	Moderate	Basic	Not even reaching marginal levels
2. Examination	Examination questions are designed to assess student's level of achievement of the intended learning outcomes. Students will need to demonstrate understanding of various elements of statistical theory and methods taught in the course through precise mathematical exposition.	High	Moderate	Basic	Not even reaching marginal levels

Applicable to students admitted before Semester A 2022/23

Assessment Task	Criterion	Excellent	Good	Fair	Marginal	Failure
		(A+, A, A-)	(B+, B, B-)	(C+, C, C-)	(D)	(F)
1. Coursework	Midterm and assignments	High	Significant	Moderate	Basic	Not even reaching marginal levels
2. Examination	Examination questions are designed to assess student's level of achievement of the intended learning outcomes. Students will need to demonstrate understanding of various elements of statistical theory and methods taught in the course through precise mathematical exposition.	High	Significant	Moderate	Basic	Not even reaching marginal levels

Part III Other Information

1. Keyword Syllabus

• Probability theory and distributions (probability space, random variables, expectation, inequalities, and convergence of random variables)

• Parametric statistical inference theory and methods (maximum likelihood estimation, Fisher's scoring, Fisher information, consistency and limiting distribution of maximum likelihood estimators, statistical decision theory, Rao-Blackwell theorem, minimum variance unbiased estimation, Bayesian inference)

• Multivariate statistics (covariance matrix estimation, James-Stein estimator, principle components analysis), linear model theory (least squares, Gauss Markov theorem, ridge regression, leave-one-out cross validation, optimal design of experiments), variable selection methods (Bayesian information criterion, LASSO, LARS).

• Nonparametric statistical models and methods (bootstrap, Gaussian process models, local polynomial regression, kernel methods)

2. Reading List

2.1 Compulsory Readings

1.	Wasserman, L. (2013). All of statistics: a concise course in statistical inference. Springer
	Science & Business Media.
2.	Keener, R. W. (2011). Theoretical statistics: Topics for a core course. Springer.
3.	Resnick, S. I. (2013). A probability path. Springer Science & Business Media.
4.	Casella, G., & Berger, R. L. (2002). Statistical inference (Vol. 2). Pacific Grove, CA:
	Duxbury.
5.	Rasmussen, C. E., & Williams, C. K. (2006). Gaussian Process Regression for Machine
	Learning. The MIT Press

2.2 Additional Readings

NIL