# City University of Hong Kong Course Syllabus

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

| Part I Course Overv                         | riew                           |
|---|--------------------------------|
| 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: (Course Code and Title)      | Nil                            |
| Precursors: (Course Code and Title)         | Nil                            |
| Equivalent Courses: (Course Code and Title) | Nil                            |
| Exclusive Courses: (Course Code and Title)  | 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 and results of the probability theory and statistical inference for machine learning, classical and cutting-edge methods and theories for regression and classification, and theoretical foundation of kernel methods.

### 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   |          | ery-en   |           |
|-----|--|-------------|----------|----------|-----------|
|     |  | (if         |          | ılum re  |           |
|     |  | applicable) | learnir  | ig outco | omes      |
|     |  |             | (please  | e tick   | where     |
|     |  |             | approp   | riate)   |           |
|     |  |             | A1       | A2       | <i>A3</i> |
| 1.  | Familiarize the key concepts of statistical inference theory | 20%         | <b>✓</b> |          |           |
|     | for machine learning   |             |          |          |           |
| 2.  | Familiarize fundamental statistical regression and           | 20%         | ✓        |          |           |
|     | classification models  |             |          |          |           |
| 3.  | Able to formulate kernel machines and regularization         | 20%         | ✓        | ✓        |           |
|     | forms  |             |          |          |           |
| 4.  | Comprehend key concepts and results in learning theory       | 20%         | ✓        | ✓        |           |
| 5.  | Able to apply taught machine learning methods to conduct     | 20%         |          | <b>✓</b> | <b>√</b>  |
|     | real life data analytics problem via software                |             |          |          |           |
|     |  | 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)

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

| TLA                | Brief Description  |          | O No.    | Hours/week |          |          |                 |
|--------------------|--|----------|----------|------------|----------|----------|-----------------|
|                    |  | 1        | 2        | 3          | 4        | 5        | (if applicable) |
| Lecture            | Introduce fundamental theories and principles in statistical learning to students  | <b>√</b> | <b>√</b> | <b>√</b>   | <b>√</b> |          | 26 hours/sem    |
| Laboratory<br>work | Assist students to develop the data analytics skill base and use learned knowledge to address real problems through lab activities |          | <b>√</b> | <b>√</b>   | ✓        | <b>✓</b> | 13 hours/sem    |

Lectures cover not only the narrowly focused techniques in engineering economy but also the wider issues of the environment that affect engineering economic decision making. Students are expected to participate in class discussion when needed.

## 4. Assessment Tasks/Activities (ATs)

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

| Assessment Tasks/Activities      | nent Tasks/Activities CILO No. |          |          | Weighting | Remarks |        |  |  |
|----------------------------------|--------------------------------|----------|----------|-----------|---------|--------|--|--|
|                                  | 1                              | 2        | 3        | 4         | 5       |        |  |  |
| Continuous Assessment: <u>50</u> | Continuous Assessment: 50 %    |          |          |           |         |        |  |  |
| Group Project                    |                                | ✓        | ✓        |           | ✓       | 0-30%  |  |  |
|                                  |                                |          |          |           |         |        |  |  |
|                                  |                                |          |          |           |         |        |  |  |
|                                  |                                |          |          |           |         |        |  |  |
| Individual Coursework            | ✓                              | <b>✓</b> | <b>✓</b> | <b>✓</b>  |         | 20-30% |  |  |
|                                  |                                |          |          |           |         |        |  |  |
|                                  |                                |          |          |           |         |        |  |  |
| Examination: 50 % (durat         | ion:                           | 2 hou    | rs , i   | f appl    | licable | e)     |  |  |
| Examination                      | ✓                              | <b>√</b> | <b>√</b> | <b>√</b>  | ✓       | 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 in Semester A 2022/23 and thereafter

| Assessment Task          | Criterion | Excellent (A+, A, A-) | Good<br>(B+, B) | Marginal (B-, C+, C) | Failure<br>(F)    |
|--------------------------|-----------|-----------------------|-----------------|----------------------|-------------------|
| 1. Group Project         | 0-30%     | High                  | Significant     | Basic                | Not even reaching |
|                          |           |                       |                 |                      | marginal levels   |
| 2. Individual Coursework | 20-30%    | High                  | Significant     | Basic                | Not even reaching |
|                          |           |                       |                 |                      | marginal levels   |
| 3. Examination           | 50%       | High                  | Significant     | 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. Group Project         | 0-30%     | High        | Significant | Moderate    | Basic    | Not even reaching |
|                          |           |             |             |             |          | marginal levels   |
| 2. Individual Coursework | 20-30%    | High        | Significant | Moderate    | Basic    | Not even reaching |
|                          |           |             |             |             |          | marginal levels   |
| 3. Examination           | 50%       | High        | Significant | Moderate    | 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 Estimators; Bayes Estimator; EM Algorithm; Bootstrap; Asymptotic Theory
- Regression: Linear and Nonlinear Regression; Kernel Smoothing; Local Polynomial; Cubic Splines;
   Regression Splines; Asymptotic Inference
- Classification: Misclassification Error; Discriminant Analysis; Logistic Regression; CART; Bagging;
   Random Forest; Estimation Consistency; Asymptotic Normality
- Kernel Machines: Bayes Rule; Fisher Consistency; Loss Functions; Empirical Risk Minimization; Regularization Form; SVM; Reproducing Kernel Hilbert Space; Function Approximation

### 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. Burger                        |
|----|--|
| 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.)

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