

# AMS 380 - Data Mining

Spring 2023; 2:00 PM - 3:20 PM; Tue & Thu

## Instructor Information

### Instructor

Hyunwook Koh, Ph.D.

### Email

[hyunwook.koh@stonybrook.edu](mailto:hyunwook.koh@stonybrook.edu)

### Office Location & Hours

B521; 3:00 PM - 4:30 PM; Mon & Wed  
(or by appointment)

## Course Information

### Course Description

Statistical learning refers to a set of tools for making sense of complex datasets. In recent years, we have seen a staggering increase in the scale and scope of data collection across virtually all areas of science and industry. As a result, statistical learning has become a critical toolkit for anyone who wishes to understand data - and as more and more of today's jobs involve data, this means that statistical learning is fast becoming a critical toolkit for everyone. This course provides a broad and less technical treatment of key topics in statistical learning (e.g., sparse methods for classification and regression, decision trees, boosting, support vector machines, deep learning). Each chapter includes an R lab. This course is appropriate for anyone who wishes to use contemporary tools for data analysis.

### Teaching Mode

All classes will be held in-person (B313).

### Course Materials

All the course materials (e.g., lecture notes, programming notes, assignments) can be found on **Brightspace**.

### Textbook

“An Introduction to Statistical Learning” by Gareth James, Daniela Witten, Trevor Hastie, Robert Tibshirani, 2<sup>ed</sup> edition, Springer, 2021 (Required) (freely available at <https://www.statlearning.com/>)

“The Elements of Statistical Learning” by Trevor Hastie, Robert Tibshirani, Jerome Friedman, 2<sup>ed</sup> edition, Springer, 2017 (Optional) (freely available at <https://hastie.su.domains/ElemStatLearn/>)

“Data Mining with R: Learning with Case Studies” by Luis Torgo, 2<sup>ed</sup> edition, Chapman & Hall, 2017 (Optional)

### Pre-requisite

AMS 311 (or instructor permission: contact Prof. Hyunwook Koh)

## **Academic Integrity**

Each student must pursue his or her academic goals honestly and be personally accountable for all submitted work. Representing another person's work as your own is always wrong. Faculty are required to report any suspected instances of academic dishonesty to the Academic Judiciary. For more comprehensive information on academic integrity, including categories of academic dishonesty, please refer to the academic judiciary website at <http://www.stonybrook.edu/uaa/academicjudiciary/>

## **School Policy on Attendance**

1. All students are required to attend every class.
2. If a student has over 20% unexcused absences, the student's final course grade will be an F.
3. Students should report the reason of absence to the professor in advance, or immediately after the absence.
4. When a student excuses his/her absence, the student must provide documentation of the reason for the absence to the professor.
5. The professor of the course reserves the right to excuse absences.
6. The professor may excuse the absence if the submitted documentation fulfills the following conditions: extreme emergencies, severe medical reasons with doctor's note, very important events.

## **Critical Incident Management**

Stony Brook University expects students to respect the rights, privileges, and property of other people. Faculty are required to report to the Office of Judicial Affairs any disruptive behavior that interrupts their ability to teach, compromises the safety of the learning environment, or inhibits students' ability to learn.

## **Course Evaluations**

Stony Brook University values student feedback in maintaining the high-quality education it provides and is committed to the course evaluation process, which includes a mid-semester assessment as well as an end-of-the-semester assessment, giving students a chance to provide information and feedback to an instructor which allows for development and improvement of courses. Please click the following link to access the course evaluation system: <http://stonybrook.campuslabs.com/courseeval/>

## **Teaching Assistant**

TBA

## **Grading**

Final grade =  $f(\text{Attendance [10\%]} + \text{Homework [10\%]} + \text{Midterm 1 [25\%]} + \text{Midterm 2 [25\%]} + \text{Final [30\%]})$ ; ABCDF grading; 3 credits

Total score	Final grade
94 - 100	A
90 - 93	A-
87 - 89	B+
84 - 86	B
81 - 83	B-
78 - 80	C+
75 - 77	C
72 - 74	C-
69 - 71	D+
66 - 68	D
61 - 65	D-
≤ 60	F

### Course Schedule (Tentative)

No.	Date	Topic	Homework
1	Feb 27	Introduction	TBA
2	Feb 29	Statistical Learning	TBA
3	Mar 5	Statistical Learning	TBA
4	Mar 7	Statistical Learning	TBA
5	Mar 12	Linear Regression	TBA
6	Mar 14	Linear Regression	TBA
7	Mar 19	Linear Regression	TBA
8	Mar 21	Classification	TBA
9	Mar 26	Classification	TBA
10	Mar 28	Classification	TBA
11	Apr 2	Resampling Methods	TBA
12	Apr 4	Midterm 1	
13	Apr 9	Resampling Methods	TBA
14	Apr 11	Linear Model Selection and Regularization	TBA
15	Apr 16	Linear Model Selection and Regularization	TBA
16	Apr 18	Linear Model Selection and Regularization	TBA
17	Apr 23	Moving Beyond Linearity	TBA

No.	Date	Topic	Homework
18	Apr 25	Moving Beyond Linearity	TBA
19	Apr 30	Moving Beyond Linearity	TBA
20	May 2	Tree-based Methods	TBA
21	May 7	Midterm 2	
22	May 9	Tree-based Methods	TBA
23	May 14	No Class (Correction Day)	
24	May 16	Tree-based Methods	TBA
25	May 21	Tree-based Methods	TBA
26	May 23	Deep Learning	TBA
27	May 28	Deep Learning	TBA
28	May 30	Deep Learning	TBA
29	Jun 4	Deep Learning	TBA
30	Jun 6	No Class (Memorial Day)	
31	TBA	Final	

### Exam Schedule (Tentative)

Date	Subject
Apr 4	Midterm 1
May 7	Midterm 2
TBA	Final