

COMP 7045

***Natural Language Processing & Large  
Language Model***

Instructors:

**Dr. Jing MA**

Teaching Assistant:

**Fu Rao**

# Outline

- General information
- Course contents
- Expected learning outcomes
- Assessment methods

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# Timetable

- Time of our classes
  - 13 Lectures:
    - Jan. 12 – Apr. 20
    - Mon 18:30 – 21:20
  - 2 Public Holidays
    - Feb. 16, Apr. 6
- Classroom of our classes
  - SCT502 (HSH Campus)
- Locations of our labs
  - RRS638, RRS735

2026	Sun	Mon	Tue	Wed	Thu	Fri	Sat	
January					1	2	3	
	4	5	6	7	8	9	10	
	11	12	13	14	15	16	17	
	18	19	20	21	22	23	24	
	25	26	27	28	29	30	31	
	1	The first day of January						
February	1	2	3	4	5	6	7	
	8	9	10	11	12	13	14	
	15	16	17	18	19	20	21	
	22	23	24	25	26	27	28	
	17-19 Lunar New Year holidays							
March	1	2	3	4	5	6	7	
	8	9	10	11	12	13	14	
	15	16	17	18	19	20	21	
	22	23	24	25	26	27	28	
	29	30	31					
April				1	2	3	4	
	5	6	7	8	9	10	11	
	12	13	14	15	16	17	18	
	19	20	21	22	23	24	25	
	26	27	28	29	30			
	3,4,7 Easter holidays							
	6 The day following Ching Ming Festival							
								9-10 Exchange Student Orientation
								10 First Semester Ends
								12 Second Semester Begins
								12 Second Semester Classes Begin
								12 First Day to Add/Drop Courses (Second Semester)
								14 Continuing Education School Board Meeting
								16-22 First Semester Make-up Examinations
								19 Transdisciplinary Undergraduate Programme Board Meeting
								24 Last Day to Add/Drop Courses (Second Semester)
								26 Senate Meeting
								4 Creative Arts School Board Meeting
								5 Arts and Social Sciences Faculty Board Meeting
								27 Communication School Board Meeting
								1-7 Christian Emphasis Week
								3 Business School Board Meeting
								3 Science Faculty Board Meeting
								4 Trimester III Tuition Payment Due Date
								9-21 Trimester II Examinations
								16 Senate Meeting
								23 Trimester III Classes Begin
								23 First Day to Add/Drop Courses (Trimester III)
								24 Council Meeting
								1 Creative Arts School Board Meeting
								1 Last Day to Add/Drop Courses (Trimester III)
								16 Arts and Social Sciences Faculty Board Meeting
								17 Chinese Medicine School Board Meeting
								22 Baccalaureate Service
								22 Continuing Education School Board Meeting
								24 Communication School Board Meeting
								25 Last Day of Classes (Second Semester)
								28 Business School Board Meeting

# Teaching Assistant

- **Mr. Fu Rao**
  - PhD student in Department of Computer Science
- **Course duty**
  - Assisting in marking course assessments
  - Project Support
  - Tutorial Support (Q&A)
- **Contact information**
  - Email: [csraofu@comp.hkbu.edu.hk](mailto:csraofu@comp.hkbu.edu.hk)
  - Office: DLB 625 6/F
  - Office Hour: (Fri) 2pm-4pm

# Reference Books

- **SPEECH and LANGUAGE PROCESSING**

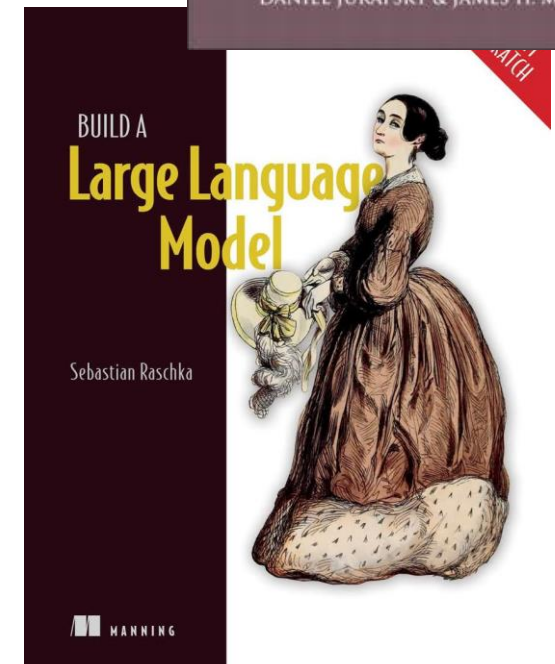
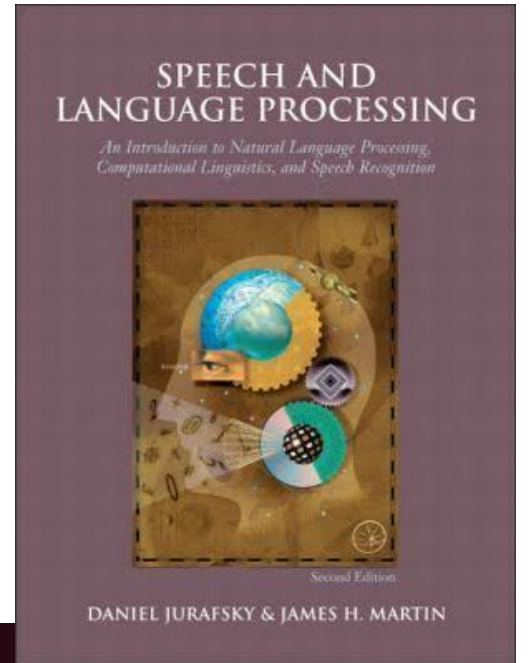
- [Daniel Jurafsky](#) and [James H. Martin](#), Speech and Language Processing: An Introduction to Natural Language Processing, Computational Linguistics and Speech Recognition with Language Models, 3rd edition. Prentice Hall. 2025.
- <https://web.stanford.edu/~jurafsky/slp3/>
- Raschka S. Build a Large Language Model (From Scratch). Simon and Schuster. 2024
- Christopher D. Manning and Hinrich Schütze. Foundations of Statistical Natural Language Processing. The MIT Press. 1999

- **Recommended Reading**

- Steven Bird, Ewan Klein, and Edward Loper. Natural Language Processing with Python. 1st edition, O'Reilly Media; 2009. (<http://www.nltk.org/book/>)
- Lane H and Dyshel M. Natural Language Processing in Action. Simon and Schuster. 2025.
- Jason Brownlee. Deep Learning for Natural Language Processing. Machine Learning Mastery, 2018.

- **We will release the lecture notes, tutorial materials (every week)**

- Slides cannot replace the textbook.
- They are at most a guideline.



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# Course Contents

- This course aims at teaching students the state-of-the-art NLP data analytics, including **techniques, software, applications**, and **perspectives** with textual data.
- The class will cover, but not be limited to, the following *topics*:
  - Text preprocessing such as segmentation, stemming, etc;
  - Statistical Language Models;
  - Syntactic Analysis such as Tagging and Parsing;
  - Word Embedding such as word2vec, GloVe, ELMo, etc;
  - Deep learning model to solve some NLP tasks;
  - Attention and Transformer;
  - Neural Language Models
  - Large Language Models
  - NLP tasks and applications;
- The assignments and project may involve various application domains



# Tutorial Subjects

Grasp basic knowledge about Python and Pytorch/Tensorflow

We will use Google Colab to train & test some models

Some local platforms such as Anaconda

The content cover the following topics:

- **Textural processing**: regular expression, text segmentation, tagging, stemming, etc
- Create a statistical language model, generate text.
- Use a space vector model to represent query and document, information retrieval system.
- Word embedding and visualization tools
- Train toy deep learning models: RNN and CNN
- Train a transformer-based model for classification task
- LLM fine-tuning, a toy LLM for text summarization

# Course Expectations

- Understand what is Natural Language Processing (NLP) and what are the main characteristics of NLP.
- Get familiar with the infrastructure and techniques of NLP, which are associated with real-world data and applications;
- Introduce some the current key issues and the evolution of NLP, and apply software tools for NLP analytics technology
- Get to know the research challenges in the area of NLP and machine learning.
- Gain insights into Large Language Models, train and use LLMs to solve practical problems

# Class Arrangements (may be adjusted as needed)

Week	Date	Course Topic
1	Jan 12	Course Information & Introduction to NLP
2	Jan 19	Text preprocessing
3	Jan 26	Statistical Language Models
4	Feb 2	Syntactic Analysis
5	Feb 9	Lab1 : Hugging Face & Colab, Text Preprocessing, Train a Language Model for Text Generation
6	Feb 23	Quiz/In-class Exercise, Word Embedding
7	Mar 2	Deep Neural Networks, Attention and Transformer
8	Mar 9	Neural Language Models
9	Mar 16	Lab2: Train a Text Classification Model, Word Embedding and Visualization, Bert Model and Train your Neural Network
10	Mar 23	Large Language Model
11	Mar 30	Quiz, Lab3: Prompt Engineering, Train A toy LLM
12	Apr 13	NLP Applications, course review
13	Apr 20	Mini-Project Presentation

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# Pre-Requisites

- Excitement about language!
- Willingness to learn

Mandatory	Desirable
<ul style="list-style-type: none"><li>• Basics of probability calculus</li><li>• English grammar</li><li>• Data Structures Algorithms</li><li>• Python programming</li></ul>	<ul style="list-style-type: none"><li>• NLP</li><li>• Machine Learning</li><li>• Deep learning</li></ul>

## This course will NOT cover:

- How to do basic Coding
- How to understand the mathematical mechanisms of NLP models, machine learning and deep learning.
- Generative models for modalities other than text

# Learning Outcomes

## ■ **Knowledge**

- Describe the fundamental concepts and methodologies of natural language processing and large Language Models
- Explain the advantages and limitations of methods developed for different scenarios
- Identify relevant language processing techniques to meet real-world needs

## ■ **Professional Skills**

- Apply specific methods and techniques in a number of natural language applications
- Evaluate the solutions designed to technical problems

## ■ **Attitude**

- Engage in life-long independent learning for problem-solving and advance state-of-the art solutions

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# Assessment Methods

- **Evaluation Objective**
  - Spread evaluation over the whole course, not just one exam or one report
- **Continuous Assessment (50%)**
  - **(Take-home / In-class) Assignment(s) + Quizz(es) + Course mini-project**
  - To assess students' mastery of the concepts, techniques and their applications
- **Examination (50%)**
  - Final examination
  - To measure the extent to which the students have reached all of the learning outcomes
- **Important Notices:**
  - **Plagiarism:** Students who plagiarized and who were plagiarized will be given **zero mark**.
  - **Final Exam:** In order to pass this course, students should attain **at least 30% of the final examination mark**.



# More about Projects and Assignments

- **Individual/Group project.**

- You need to do an oral presentation and submit a report.
- Choose a topic, literature review, implementation, result analysis
- Can use any programming languages & third-party libraries
- Detailed schedule will be announced later
  - Report submission (Tentative deadline: End of week-13)
  - Report presentation ~10 minutes
  - will be graded by the instructor, TA , and other students.

- **Assignments & quiz**

- release before recess week