國立台灣科技大學 114學年 第2學期　課程大綱

Spring 2026 NTUST Course Outline

授課教師：陳郁堂

Instructor:Chen, Yie-Tarng

課程名稱：大型語言模型與應用

Course Title：Large Language Models and Applications

2026/11/3

課程代號：ET5347701 Course Code 學分數：3 Credits	必選修：選修/半學年 Required/Electve:Elective/Half Yr. 先修課程： Prerequisites
節次教室：F2(IB-408) F3(IB-408) R4(IB-408) Time/Location	
專業核心能力： Core Professional Competencies	1. 運用數學、科學及工程知識的能力。 2. 設計與執行實驗，以及分析與解釋數據的能力。 5. 培養學生具備執行跨領域專案整合、領導及溝通之能力。 6. 發掘、分析、應用研究成果及因應複雜且整合性工程問題的能力。
課程網址： Course Website	
課程宗旨： Course Objectives	This course provides an in-depth introduction to Large Language Models (LLMs) and generative models, covering their theoretical foundations, technical architectures, model fine-tuning, and practical development with real-world applications. The curriculum explores the internal mechanisms of LLMs in detail, including language pretraining, transfer learning, and other core techniques, and examines how fine-tuning methods can be applied to adapt models to specific tasks. In addition to language models, the course also introduces generative models—particularly text-to-image generation—and discusses their training methodologies, enabling students to understand state-of-the-art technologies while developing hands-on practical skills.
課程大綱： Outline of Lectures	

1. LLM Fundamentals (optional)
• Pytorch: Essential programming skills
• Neural Networks: Introduction to concepts
2. The LLM Architecture
• Transformer Overview: Encoder-decoder vs. decoder-only models
• Tokenization: Converting text to tokens
• Attention Mechanisms: Self-attention and scaled dot-product attention
• Text Generation Techniques: Greedy decoding, beam search, top-k sampling, nucleus sampling
3. Building an Instruction Dataset
• Synthetic Data Generation
• Dataset Improvement Techniques: Evol-Instruct, regex filtering
• Prompt Templates
4. Pre-training Models
• Data Pipeline and vocabulary setup
• Causal vs. Masked Language Modeling
5. Supervised Fine-Tuning
• Full fine-tuning, LoRA, QLoRA
• Axolotl and DeepSpeed
6. Preference Alignment
• RLHF and preference datasets
• PPO-based optimization
• GRPO optimization
7. LLM Inference and Serving
• Latency Decomposition: Prefill vs Decode
• KV Cache Management and Memory Optimization
• Continuous batching and scheduling
• Long-context serving and million-token context
• vLLM and SGLang architectures
• GPU memory hierarchy and KV offloading
• Quantization Methods: GPTQ, AWQ, GGUF
• Speculative Decoding for fast inference
8. Evaluation and Benchmarking
• LLM-as-a-Judge evaluation frameworks
• Benchmark datasets: MMLU, HumanEval
• RAG evaluation techniques: RAGAS
• Automated evaluation pipelines and metrics
9. Running LLMs
• API usage vs local execution
• Prompt Engineering
10. Retr

授課方式： Method of Instruction	講授 Lecture：100% 分組討論 Group discussion：0% 案例研討 Case study：0% 操做練習 Practical exercises：0% 講授 Lecture：%
教科書： Textbooks	Class notes
參考書目： References	N/A
修課須知： Notice	
評量方式： Grading	• Term Exam: 45% • Final Project 25% • Paper Presentation: 10% • Homework and Programs: 20%
備註說明： Notes	Students should be familiar with Python and PyTorch and have basic deep learning courses