Introduction to AI, LLMs and NLP

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This two-week course offers an accessible introduction to Artificial Intelligence, Natural Language Processing (NLP), and Large Language Models (LLMs) for bachelor's and master's students from any discipline. Starting from the foundations of human language (how sounds, words, and grammar create meaning), the course gradually moves toward computational methods that support real-world products such as chatbots, translators, and voice assistants. Through

interactive labs and a team mini-project, participants gain practical experience with tools such as spaCy and Hugging Face Transformers. By the end, students will understand both the theoretical and industry-relevant aspects of modern NLP and will be able to design and present a small working prototype of their own.

Duration: 2 weeks (10 days, 5 hours 30 mins per day)
Audience: Bachelor's and Master's students, mixed backgrounds. Structure:
Lecture/discussion (60 min); Break (30 min) Hands-on/lab (90 min); Wrap-up and discussion (30 min)

1 Day 1 - Human Language and Al

- Lecture: What Is Language?
 Exploring Linguistic Levels such as Phonetics and Phonology, Morphology, Syntax, Semantics, and Pragmatics with relevant examples from morphology-rich and analytic languages.

 Relevance with NLP and Human to Machine connection from early linguistic approaches (Chomsky, rule-based) to statistical to deep learning. Use of NLP within industry (chatbots, speech assistants, translation).
- Lab: Short multi-language annotation (POS, morphology, syntactic chunks).
- Outcomes: Recognise complexity of human language; map linguistic layers to modern NLP tasks.

2 Day 2 - Words, Meaning and Data

- Lecture: Tokens, types, lemmas. Frequency, Zipf's law. Corpora: corpus design and sampling bias. Vector space model.
- Lab: explore word frequencies and collocations in a news vs. social media corpora (Python/Colab), discuss domain differences.
- Outcomes: Link language data to statistical patterns.

3 Day 3 - From Rules to Statistics

- Lecture: Show evolution from rule-based NLP to statistical methods.
 Regular expressions, simple grammars. N-grams, language modelling, prob ability basics.
- Lab: Build a bigram language model on a small corpus, generate "silly sentences". Regex patterns to find dates/emails/hashtags.
- Outcomes: Understand the basics of NLP methods.

4 Day 4 - Modern NLP

- Lecture: Preprocessing: tokenization, stopwords, stemming vs. lemma tization. POS tagging and named entity recognition (NER). Introduction to spaCy.
- Lab: groups manually label entities in sample tweets, then compare with spaCy output. Explore spaCy pipeline components and visualize dependency trees.
- Outcomes: Understand typical NLP workflow. Understand standard preprocessing and tagging tasks. Get hands-on with spaCy.

5 Day 5 - Word Meaning and Embeddings

- Lecture: Distributional hypothesis, vector semantics, Word2Vec/GloVe intuition, cosine similarity.
- Lab: Explore pre-trained embeddings (gensim or spaCy vectors): nearest neighbours, analogy tasks. Make word clusters.
- Outcomes: Understand distributional semantics and vector space mod els. Experiment with pre-trained embeddings.

6 Day 6 - Deep Learning for NLP

- Lecture: Fundamentals of feed-forward nets, RNNs, and attention. Sequence-to-sequence tasks (translation, summarisation).
- Lab: run a pre-built sentiment classifier notebook; tweak hyperparameters or input size to observe changes. Compare RNN vs. bag-of-words baseline.
- Outcomes: Grasp fundamentals of neural nets, sequence modelling and attention.

7 Day 7 - Transformers and LLMs

- Lecture: Transformer architecture. Pre-training vs. fine-tuning. Prompt Engineering methods.
- Lab: Compare outputs of a small open model on HuggingFace with a large one. Compare outputs from the same prompt.
- Outcomes: Understand the Transformer architecture. Practice basic prompting and model comparison.

8 Day 8 - Mini-Project Build Day (pt. 1)

- Lecture: Project design and dataset selection.
 Teams define a business-relevant NLP/LLM application and outline a processing pipeline.
- Lab: Begin implementation. Example options: Chatbot for a local business, Sentiment dashboard, Domain-specific NER, Simple text summarizer.
- Outcomes: Apply the week's knowledge to a practical, industry-relevant task. Develop coding skills with natural language coding.

9 Day 9 - Mini-Project (pt. 2) and Presentation · Lecture:

Q&A and final mentoring.

 Lab: Continue projects. Team presentations (10 min each + 5 min Q&A); peer feedback. Outcomes: Communicate technical work and relate skills to real-world career paths.

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10 Day 10 - Ethics, Bias and Responsible Al

- Lecture: Al safety, bias, fairness, privacy, environmental impact, explainability and limitations.
- Lab: Inspect bias in a sentiment model (e.g., gendered names) and propose mitigation steps.
- Outcomes: Critically evaluate social and ethical implications of NLP and LLMs.