

# **Large Language Models**

### **Audience course Large Language Models**

The course Large Language Models is intended for software engineers, data scientists, and technical professionals who want to work with large language models (LLMs).

#### **Prerequisites Large Language Models Course**

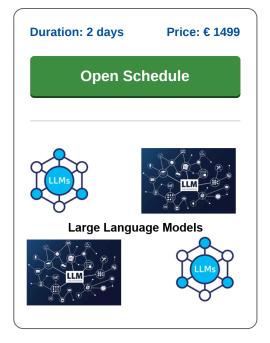
To participate in the course, a basic understanding of Python and machine learning is required. Familiarity with neural networks or natural language processing is useful.

#### **Realization training Large Language Models**

The course is led by an experienced trainer and includes a mix of theory and handson exercises. Demonstrations and case studies involving LLMs are used to illustrate key concepts.

#### **Large Language Models Certificate**

After successfully completing the course, attendants receive a certificate of participation in the course Large Language Models.



# **Content Large Language Models**

The course Large Language Models from SpiralTrain provides a comprehensive understanding of large language models (LLMs), from foundational architectures like transformers to advanced topics like fine-tuning, safety, deployment, and real-world applications. Participants will explore open models, tools, and cutting-edge use cases across domains.

#### Intro to LLMs

This module introduces LLMs and their evolution, from GPT to BERT and T5. It explains transformers, attention, and tokenization. Participants explore training objectives, scaling laws, and key differences between pretraining and fine-tuning. Open source vs proprietary models are also compared.

#### **Model Architectures**

Participants learn LLM types: decoders, encoder-decoders, and key models like GPT, LLaMA, and PaLM. It covers training pipelines, optimizers, and precision formats. Tools like Hugging Face and Deepspeed are introduced, plus tuning techniques like LoRA and incontext learning.

## Training LLMs

This module focuses on preparing and fine-tuning data for LLMs. It includes tokenizer setup, adapters, SFT, and avoiding overfitting. Participants learn metrics for evaluation, model alignment, and benchmarking. Hugging Face tools and best practices are emphasized.

## **LLM Deployment**

Participants learn how to serve and optimize LLMs for production. Topics include quantization, distillation, and cloud deployment (AWS, Azure, GCP). Also covered are LangChain integration, embeddings, caching, and reducing inference costs with scalable strategies.

#### **Safety and Bias**

Focus is on LLM safety: identifying and reducing bias, model auditing, and prompt attacks. Topics include explainability, red teaming, and moderation. Legal and privacy concerns are discussed, with strategies for responsible LLM deployment.

#### **LLM Use Cases**

The course ends with real-world LLM applications in code, legal, health, and education. Use cases include RAG systems, agents, and plugins. Participants explore enterprise integration, model evaluation, and research trends shaping the future of LLMs.

Locations



# **Modules Large Language Models**

Module 1: Intro to LLMs	Module 2: Model Architectures	Module 3: Training LLMs
What are LLMs?	Decoder vs Encoder-Decoder Models	Dataset Creation and Curation
Transformer architecture	GPT, LLaMA, T5, and PaLM	Tokenizer Customization
Training Objectives (causal, masked)	Training Pipeline Overview	Data Preprocessing
Evolution of LLMs (GPT, BERT, T5)	Optimizers (Adam, Adafactor)	Fine-Tuning with Hugging Face
Open Source vs Proprietary LLMs	Precision (FP32, FP16, quantization)	SFT (Supervised Fine-Tuning)
Tokenization and Vocabulary	Transformers (HF), Megatron, Deepspeed	Adapters and LoRA
Attention Mechanism	Parameter vs Instruction Suning	Evaluation Metrics
Model Scaling Laws	LoRA and QLoRA	Avoiding Overfitting
Transfer Learning	In-context Learning	Model Alignment
Pretraining vs Fine-Tuning	Reinforcement Learning with HF	Model Evaluation and Benchmarking
Module 4: LLM Deployment	Module 5: Safety and Bias	Module 6: LLM Use Cases
Inference Optimization	Understanding Model Biases	Coding Assistants
	Understanding Model Biases Mitigation Strategies	Coding Assistants Al for Legal and Finance
Model Distillation		
Model Distillation Quantization Techniques	Mitigation Strategies	Al for Legal and Finance
Inference Optimization Model Distillation Quantization Techniques Hosting on AWS, GCP, Azure Using Model Gateways	Mitigation Strategies Model Auditing	Al for Legal and Finance Education and Learning
Model Distillation Quantization Techniques Hosting on AWS, GCP, Azure	Mitigation Strategies Model Auditing Adversarial Prompts	Al for Legal and Finance Education and Learning Health Care and Biotech
Model Distillation Quantization Techniques Hosting on AWS, GCP, Azure Using Model Gateways LangChain and Semantic Search	Mitigation Strategies Model Auditing Adversarial Prompts User Privacy	Al for Legal and Finance Education and Learning Health Care and Biotech Chatbots and Agents
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