

Foundations of Machine Learning

Target Audience: Students and Professionals Preparing for ML

Why This Course?

Machine Learning requires more than just coding skills—it demands a strong foundation in **mathematics**, algorithms, and programming. Many learners struggle with ML not because the concepts are too advanced, but because they lack the **mathematical** intuition or coding fluency needed to implement and understand the models.

This course bridges that gap. By combining **Python programming** with **essential mathematics for ML**, you will build the skills necessary to transition confidently into advanced machine learning.

By the end of this course, you will:

- Strengthen your Python skills for mathematical and algorithmic problem-solving.
- Gain mastery of the core math topics essential for ML.
- Implement mathematical concepts in Python through coding exercises and projects.
- Use Git and GitHub to manage and showcase your work.
- Be fully prepared to take on advanced ML algorithms and techniques.

What Will You Learn?

1. Python Foundations for Math and ML

- Python programming for scientific computing.
- Explore essential libraries: NumPy, Pandas, and matplotlib.
- Work with Jupyter notebooks for experimenting with math and algorithms.

2. Mathematical Foundations

- Linear Algebra: Vectors, matrices, operations, eigenvalues, and eigenvectors.
- Probability & Statistics: Random variables, distributions, expectation, variance, hypothesis testing.
- Calculus & Optimization: Derivatives, gradients, and the foundations of optimization methods.

3. Algorithms for ML Preparation

- Implementation of classical algorithms in Python.
- Time and space complexity (Big-O analysis).
- Numerical methods relevant for ML.

4. Optimization Methods

- Gradient descent: intuition, mathematics, and implementation.
- Variants of gradient descent (stochastic, mini-batch).
- Convergence, pitfalls, and practical considerations.

5. Programming Practice & Projects

Extensive hands-on coding throughout the course.

- Mini-projects to reinforce each major concept.
- Application-focused exercises linking math to ML.

6. Git & Version Control

- Introduction to Git and GitHub.
- Tracking changes, branching, and collaborative coding.
- Building a GitHub portfolio of your projects.

7. Additional Concepts as Needed

- Entropy and Information Gain (for decision trees).
- Distance and Similarity Metrics (for kNN and clustering).
- Log-Likelihood and Expectation (for probabilistic models).
- Matrix Factorization and Eigen Decomposition (for dimensionality reduction).

Course Differentiators

- 24-week intensive program combining depth and hands-on practice.
- Integrated approach combining Python coding with mathematical rigor.
- Heavy emphasis on programming, projects, and practical problem-solving.
- Git and GitHub usage throughout to instill industry-ready practices.
- Designed as a bridge course to prepare you for ML.

Course Benefits

- Build the core math intuition required for Al and ML.
- Develop problem-solving skills that connect theory to practice.
- Gain confidence by completing mini-projects and maintaining a GitHub portfolio.
- Be ML-ready by the end of the course, with a solid foundation to tackle advanced topics.

Who Should Enroll?

- Students aiming to pursue ML or Al.
- Professionals looking to strengthen their math and coding foundations.
- Learners who want to confidently bridge the gap between Python programming and ML.

Course Duration: 24 weeks /48 weeks

Mode: Online

Instructor Support: Live sessions and Q&A opportunities

Get ML-Ready with Python and Math

This course gives you the tools to **connect coding, math, and algorithms**, setting you up for success in advanced machine learning.

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Master AI & Machine Learning From Fundamentals to Advanced Practice

A Comprehensive 24-Week Deep Dive for Software Professionals

Go beyond the basics and become a proficient AI & Machine Learning

practitioner. This extended, in-depth course is meticulously designed for developers, testers, team leads, and software professionals aiming for a robust skill set and a confident transition into specialized AI/ML roles.

Course Principles & Objectives

- Mastery Through Depth & Breadth: Build a comprehensive understanding, covering not just core concepts but also advanced algorithms, specialized domains (NLP, CV), practical MLOps, and ethical considerations.
- **Rigorous Conceptual Foundations:** Gain deep intuition behind the math and algorithms, enabling you to innovate, troubleshoot complex problems, and critically evaluate different approaches.
- **Dual Framework Proficiency:** Achieve practical mastery in **both** major Deep Learning frameworks: **PyTorch** *and* **TensorFlow/Keras**.
- **Production-Ready Skills:** Focus on the entire ML lifecycle, including robust evaluation, experiment tracking, versioning, deployment strategies, and monitoring concepts (MLOps).
- Strategic Career Advancement: Equip yourself with the advanced knowledge and extensive portfolio needed to target specialized AI/ML engineer, data scientist, or research-oriented roles.

Benefits & Your Capabilities

- **Build an Extensive, High-Impact Portfolio:** Complete numerous sophisticated projects covering diverse ML/DL tasks, including NLP, Computer Vision, and potentially Generative Models, showcasing advanced capabilities.
- Command Advanced ML & DL Techniques: Confidently implement, evaluate, and tune a wide array of algorithms: Linear/Logistic Models, SVM, Naive Bayes, Tree Ensembles (Random Forests, XGBoost, LightGBM), Advanced Unsupervised Learning, Anomaly Detection, and complex Deep Learning architectures (CNNs, RNNs/LSTMs, Transformers).
- Become Framework Fluent: Seamlessly work with both PyTorch and TensorFlow/Keras, understanding their strengths and choosing the right tool for the job.
- Implement End-to-End MLOps Practices: Apply practical skills in experiment tracking (MLflow/W&B), data/model versioning (DVC concept), advanced evaluation, monitoring concepts, and diverse deployment strategies (APIs, Docker, Cloud Platforms Intro).
- Specialize in High-Demand Areas: Gain significant practical experience in Natural Language Processing (NLP) using Transformers and the Hugging Face ecosystem, and Advanced Computer Vision (CV) including Object Detection/Segmentation concepts.
- Champion Responsible AI: Understand and apply techniques for fairness assessment, model explainability (SHAP/LIME), bias mitigation, and ethical AI development.

• Accelerate Your AI/ML Career: Possess the deep knowledge and practical skills to stand out in the job market, targeting a wider range of AI/ML roles and contributing at a higher level.

Course Content Highlights

Modules 1-2: In-Depth Foundations & Advanced Data Techniques:

- Principles: ML Lifecycle Mastery, Advanced EDA, Robust Data Cleaning/Preprocessing, Feature Engineering & Selection Strategies, Advanced Math Intuition (Linear Algebra, Calculus, Probability/Stats).
- Tools: Python (NumPy, Pandas), Visualization Libraries, Scikit-learn Advanced Features.

Modules 3-5: Comprehensive Supervised & Unsupervised Learning:

- Principles: Deep Dive into Linear Models, SVM Margins & Kernels, Probabilistic Classification (Naive Bayes), Instance-Based Learning (KNN), Advanced Tree Ensembles & Boosting Algorithms (GBM, XGBoost), Advanced Clustering (DBSCAN), Anomaly Detection, Dimensionality Reduction (PCA, t-SNE/UMAP).
- *Tools:* Scikit-learn, XGBoost, LightGBM.

Module 6: Deep Learning Mastery (Dual Framework):

- Principles: Advanced NN Training (Optimization, Backpropagation nuances, Regularization deep dive), Architectural Patterns.
- Tools: PyTorch AND TensorFlow/Keras (Extensive practical coverage in both).

Module 7: Advanced Computer Vision (CV):

- Principles: Advanced CNN Architectures (ResNet concept), Transfer Learning (Feature Extraction vs. Fine-Tuning), Object Detection & Image Segmentation Fundamentals.
- *Tools:* PyTorch/Keras, torchvision, tf.keras.applications.

Module 8: Natural Language Processing (NLP) with Transformers:

- Principles: Word Embeddings (Word2Vec/GloVe), Sequence Modeling (LSTMs),
 Attention Mechanisms, Transformer Architecture, Pre-trained Models
 (BERT/GPT concepts), Fine-tuning Transformers for downstream tasks.
- *Tools:* PyTorch/Keras, **Hugging Face transformers**, Tokenizers.

Module 9: Introduction to Generative Models:

• *Principles:* Autoencoders, Variational Autoencoders (VAEs), Generative Adversarial Networks (GANs) concepts.

Module 10: Practical MLOps:

- Principles: Experiment Tracking, Code/Data/Model Versioning, ML Testing, Monitoring for Drift, Deployment Strategies (APIs, Containers, Cloud), CI/CD concepts for ML.
- *Tools:* MLflow/W&B, DVC concept, Docker concept, Cloud Platform Intro (AWS/GCP/Azure).

Module 11: Responsible AI in Practice:

- Principles: Fairness Metrics, Explainability Techniques (SHAP/LIME), Bias Mitigation Strategies, Privacy Concepts.
- Tools: SHAP/LIME libraries.

Learning Approach

- · Immersive live online sessions (2 x 2 hours per week for deeper engagement).
- Extensive hands-on coding demonstrations in **both PyTorch and TensorFlow/Keras**.
- · Challenging programming assignments and a substantial, multi-faceted portfolio.
- Emphasis on deep conceptual understanding, mathematical intuition, and practical tradeoffs.
- · Comprehensive capstone project allowing for specialization (e.g., NLP, CV, MLOps focus).

Prerequisites: Solid programming experience (any language - Python covered). Familiarity with basic software development concepts. No prior ML experience required, but a strong desire to learn is essential.

Invest in Your Future. Become an AI/ML Expert.

Course Duration: 24 weeks . Total course duration - 48 weeks

Mode: Online

Instructor Support: Live sessions, one-on-one feedback, and project reviews