

Modules - Practical Computer Vision with PyTorch

This is a practical online course that will be offered through OpenHPI in May 2025. Learners will gain hands-on experience with modern deep learning techniques for solving computer vision problems, including image classification, object detection, and image generation.

Trailer Video

An overview of the course's content

[Trailer video](#)

Prerequisites

A basic knowledge of programming and high-school level math is sufficient to follow the course. The following resources serve as helpful references:

Programming Prerequisites

- Python Fundamentals
 - [Basic knowledge of Python](#)
 - [Writing functions and classes](#)

Mathematics Prerequisites

- Statistics and Probability
 - [Mean, median, and mode](#)
 - [Probability](#)
 - [Variance and standard deviation](#)
- Linear Algebra
 - [Dot product](#)
 - [Matrix multiplication](#)
- Calculus
 - [Derivatives](#)

Course Modules

Each module is paired with at least one practical programming notebook. Notebooks are currently in development and you can find them [here](#).

Module 1: Foundations

1. Intro to Common Computer Vision Tasks
 - [Slides](#)
 - [Video](#)
2. Digital Image Representation in PIL and NumPy
 - [Slides](#)
 - [Video](#)
3. Image Tensors in PyTorch
 - [Slides](#)
 - [Video](#)

Module 2: Neural Networks Fundamentals

4. Introduction to Neural Networks
 - [Slides](#)
 - [Video](#)
5. Building a Multilayer Perceptron for Regression in PyTorch
 - [Slides](#)
 - [Video](#)
6. Matrix Multiplication and Network Shape
 - [Slides](#)
 - [Video](#)

Module 3: Training and Evaluation

7. Training a Feedforward Network for Classification in PyTorch
 - [Slides](#)
 - [Video](#)
8. Performance Metrics for Classification and Experiment tracking
 - [Slides](#)
 - [Video](#)
9. PyTorch Datasets and DataLoaders
 - [Slides](#)
 - [Video](#)

Module 4: Convolutional Neural Networks

10. Fundamentals of Convolutions

- [Slides](#)
- [Video](#)

11. Pooling in Neural Networks

- [Slides](#)
- [Video](#)

12. Upsampling and Channel Mixing with Convolutions

- [Slides](#)
- [Video](#)

Module 5: Training Techniques for Image Classification

13. Normalizing Input Values, and Inference with Pretrained Models

- [Slides](#)
- [Video](#)

14. Usage of Binary Cross Entropy Loss

- [Slides](#)
- [Video](#)

15. Skip Connections

- [Slides](#)
- [Video](#)

Module 6: Model Optimization and Interpretability

16. Data Augmentation

- [Slides](#)
- [Video](#)

17. Regularization with Dropout and Batch Normalization

- [Slides](#)
- [Video](#)

18. Transfer Learning and Fine Tuning Pre-trained Models

- [Slides](#)
- [Video](#)

19. Class Activation Mapping

- [Slides](#)
- [Video](#)

Module 7: Image Embeddings

20. Image Embeddings

- [Slides](#)
- [Video](#)

21. Vision Transformers

- [Slides](#)
- [Video](#)

22. CLIP (Contrastive Language-Image Pretraining)

- [Slides](#)
- [Video](#)

Module 8: Object Detection and Image Segmentation

23. Approaches to Object Detection

- [Slides](#)
- [Video](#)

24. Approaches to Image Segmentation

- [Slides](#)
- [Video](#)

Module 9: Image Generation

25. Image Generation with Diffusion Models

- [Slides](#)
- [Video](#)

GitHub Repository

Notebooks will be extended and published on Kaggle before May 2025

<https://github.com/andandandand/practical-computer-vision>