# **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.

#### **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**

- 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

#### **Module 7: Image Embeddings**

- 19. Image Embeddings
  - Slides
  - Video
- 20. Vision Transformers
  - Slides
  - Video

- 21. CLIP (Contrastive Language-Image Pretraining)
  - Slides
  - Video

### **Module 8: Object Detection and Image Segmentation**

- 22. Approaches to Object Detection
  - Slides
  - Video
- 23. Approaches to Image Segmentation
  - Slides
  - Video

# **Module 9: Image Generation**

- 24. Image Generation with Diffusion Models
  - Slides
  - Video

# **Module 10: Interpretability**

- 25. Interpretability with Class Activation Mapping
  - Slides
  - Video

# **GitHub Repository**

Notebooks will be extended and published on Kaggle before May 2025 <a href="https://github.com/andandandand/practical-computer-vision">https://github.com/andandandandand/practical-computer-vision</a>