

Comprehensive **Roadmap of Machine Learning for Computer Vision**, organized by **Learning Priority** and **Necessary Prerequisites**.

This guide moves from fundamentals to advanced, so it can build layer by layer.







## **Complete Study Path for Computer Vision**

This roadmap is structured into progressive phases from foundation to frontier.

### **Phase 1: Prerequisites & Core Knowledge (Foundational Concepts)**

These are must-know concepts before diving into computer vision:

**Goal:** Build mathematical and programming fluency.

-  **Mathematics:**
  - ✓  **Linear Algebra & Calculus:** Vectors, matrices, derivatives
  - ✓  **Probability & Statistics:** Distributions, Bayes' theorem, evaluation metrics
-  **Core ML Principles:** Supervised vs. unsupervised learning, overfitting, bias-variance tradeoff, loss functions
-  **Programming:** Python mastery (NumPy, Matplotlib, Pandas)
-  **Evaluation Metrics:** Accuracy, precision, recall, F1-score





**Priority Level:**  Essential

**Prerequisite for:** All models and architectures

### **Phase 2: ML Model Fundamentals**

Before working with images, understand basic model types:

**Goal:** Being familiar with Machine Learning.

-  **Regression and Classification Models:** Logistic regression, decision trees, k-NN
-  **Gradient-Based Optimization:** Backpropagation, loss functions
-  **Feature Engineering:** Scaling, normalization, data splitting
-  **Evaluation:** Accuracy, precision, recall, F1-score, confusion matrix







**Priority Level:**  High

**Prerequisite for:** Neural networks, CNNs

## **Phase 3: Classical Computer Vision**

Before using ML and DL, Interacting with the image.

**Goal:** Understand rule-based image processing and feature detection.

-  **Introduction** to Computer Vision and image types (RGB, grayscale, etc.)
-  **Image manipulation** with OpenCV (resize, crop, filters)
-  **Image Filters:** Edge detection (Sobel, Canny), blurring
-  **Feature Extraction:** SIFT, SURF, ORB
-  **Image Geometry:** Homographies, Camera Calibration
-  **Libraries:** OpenCV, PIL




**Priority Level:**  Essential

**Prerequisite for:** Neural networks, CNNs

## **Phase 4: Deep Learning Foundations**

Start exploring computer vision with these essentials:

**Goal:** Learn to train neural networks from scratch.

-  **Neural Networks:** Feedforward networks, activation functions, backpropagation, training
-  **Convolutional Neural Networks (CNNs):** Convolutions, Filters, pooling, feature maps (activation maps)
-  **Frameworks:** PyTorch or TensorFlow for building models

**Projects:**

- Build a handwritten digit classifier (MNIST)
- Create a custom image classifier









**Priority Level:**  Critical

**Prerequisite for:** Image classification, object detection, segmentation

## **Phase 5: Core CV (Computer Vision) Tasks**

Now ready to tackle real-world visual challenges:

**Goal:** Master practical computer vision applications.

Task	Goal - Implement	Example Models/Techniques
 Image Classification	Label whole image (TorchVision or Keras)	CNNs, ResNet, EfficientNet, ViT
 Object Detection	Detect and localize objects in image (Detectron2)	YOLO, SSD, Faster R-CNN
 Semantic Segmentation	Pixel-level class labeling (MMDetection)	U-Net, DeepLab
 Instance Segmentation	Segment individual objects	Mask R-CNN
 Pose Estimation	OpenPose, HRNet (MediaPipe)	MediaPipe
 Face Recognition	FaceNet, ArcFace (InsightFace)	InsightFace
 Image Captioning	Generate descriptive text	CNN+RNN, Vision Transformers
 Image Generation	Synthesize new images	GANs, Diffusion Models

**Projects:**

- Detect objects in street scenes
- Segment medical images





**Priority Level:**  Advanced

**Prerequisite for:** Application-specific CV systems

## Phase 6: Advanced & Generative Models & Architectures

Once fluent with vision tasks, expand into cutting-edge systems:

**Goal:** Explore emerging architectures and creative AI.

-  **Vision Transformers (ViT, Swin Transformer):** Attention-based models for image data
-  **Multimodal Learning:** Combine vision with language (e.g., CLIP, BLIP, Flamingo)
-  **Generative Models:** GANs, VAEs, Diffusion-based Synthesis Models
-  **Self-Supervised Learning:** Learn representations without labels

**Projects:**

- Train StyleGAN to generate portraits
- Build an image captioning system with BLIP







**Priority Level:**  Expert

**Prerequisite for:** Research, AI applications, multimodal systems

## **Phase 7: Applications Deployment & Optimization**

Finally, bring it all into practice:

**Goal:** Productionize and refine vision systems.

-  **Model Evaluation & Tuning:** Cross-Validation, Hyperparameter Search
-  **Model Acceleration:** Quantization, Pruning
-  **Edge Deployment:** Using Models on Mobile or Embedded Devices (ONNX, TensorRT,
-  **Explainability & Ethics:** Bias detection, model transparency (Grad-CAM, SHAP)
-  Hyperparameter Tuning: Optuna, Ray Tune
-  **Toolkits:** OpenCV, Detectron2, MMDetection, HuggingFace

**Projects:**

- Deploy models to mobile
- Create interpretable vision systems

**Priority Level:**  Practical

**Prerequisite for:** Production-level ML systems