



Lecture 1: Introduction to Computer Vision & Image Formation

CIS 6217 – Computer Vision for Data Representation
College of Computer Science, King Khalid University

● Learning Outcomes

- Define computer vision and explain its importance in modern applications.
- Understand the historical context and evolution of computer vision.
- Describe the basics of image formation: geometry, photometry, and representation.
- Recognize the relationship between pixels, color spaces, and digital image structure.

● What is Computer Vision?

- A field of AI that enables computers to interpret and process visual information from the world.
- Mimics human visual perception using algorithms and models.
- Bridges image/video data and decision-making processes.

● Applications of Computer Vision

- Self-driving cars and autonomous navigation.
- Medical imaging and diagnostics.
- Facial recognition and biometrics.
- Augmented and Virtual Reality.
- Robotics, drones, and industrial automation.

● Historical Development of CV

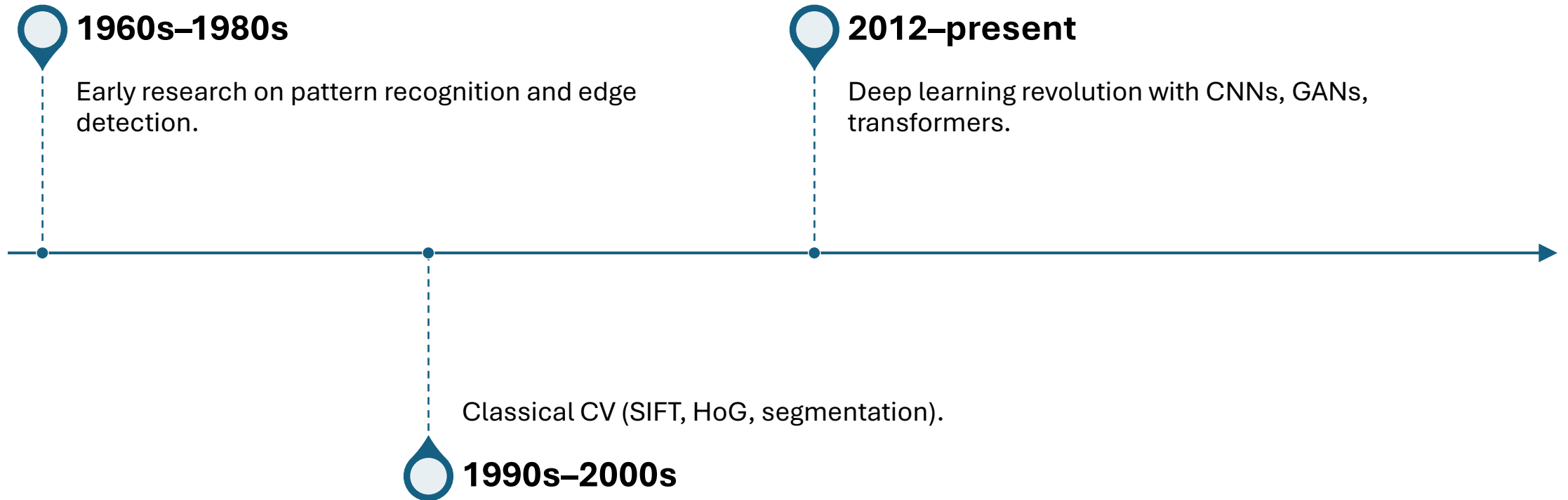


Image Formation

Forsyth & Ponce defines image formation as combined process “*Geometry tells us where the light goes; radiometry tells us how much arrives.*”

● Image Formation: Geometry

- Geometry explains how 3D scenes are mapped onto a 2D image plane.



Pinhole camera model: light rays projected onto a 2D plane.



Perspective projection and vanishing points.



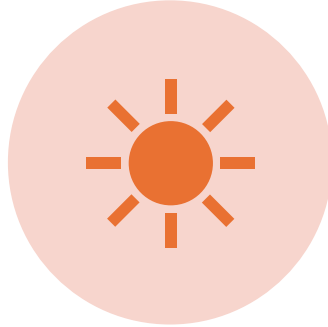
Parameters: focal length, field of view, aperture size.

● Image Formation: Photometry

- Radiometry provides the physical quantities (energy per area, per solid angle) that explain brightness and contrast in images.



INTERACTION OF LIGHT WITH
OBJECTS: REFLECTION,
ABSORPTION, TRANSMISSION.



BRIGHTNESS, CONTRAST, AND
INTENSITY LEVELS.



ILLUMINATION CONDITIONS
IMPACT CAPTURED IMAGE
QUALITY.

Digital Image

A **digital image** is a **2D array of intensity values** (pixels).

● Digital Image Representation

- An image is stored as a 2D array (matrix) of pixels.
- Each pixel represents intensity (grayscale) or RGB values (color).
- Resolution: number of pixels along width and height.
- Aspect ratio: proportional relationship between dimensions.

● Intensity & Color Representation

- Grayscale images: each pixel has a single intensity value (brightness).
- Color images: typically represented using channels (RGB = Red, Green, Blue).
- Each channel usually stored with 8 bits (values 0–255).
- Thus, a color pixel = [R, G, B].

● Color Models

- RGB: Red, Green, Blue (additive model).
- HSV: Hue, Saturation, Value (better for color-based segmentation).
- YCbCr: Luminance + chrominance components (used in video compression).

● Digital Image as a Matrix

- In memory, an image is stored as a matrix (array).
 - Grayscale: $H \times W$ matrix.
 - Color (RGB): $H \times W \times 3$ array.
- Example: A 256×256 RGB image = $256 \times 256 \times 3 = \sim 200\text{K}$ values.

● Image Data in CV

- Stored as multidimensional arrays (height × width × channels).
- `Image.shape = (H, W, 3)` for RGB images.
- Libraries: OpenCV, Pillow, PyTorch, TensorFlow.

● Python Example: Image Handling

- Load and display an image with OpenCV.
- Convert image to grayscale.
- Explore pixel values and channels.

● References

- Computer Vision: A Modern Approach – Forsyth & Ponce (2010)