

Lab 11

Optical Flow

Learning Objectives:

- *By the end of this lab, students will be able to:*
- *Understand the concept of optical flow and its role in motion analysis.*
- *Apply dense optical flow techniques using the Farneback method.*
- *Track sparse feature points between frames using the Lucas-Kanade method.*
- *Visualize and interpret motion information in videos using OpenCV.*

Theoretical Background:

What is Optical Flow?

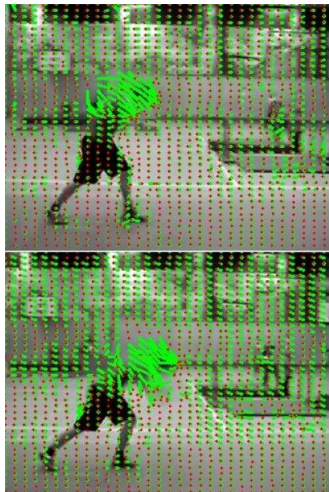
Optical flow is the pattern of **apparent motion** of objects, surfaces, and edges in a visual scene caused by the **relative motion** between an observer and the scene.

It provides information about:

- The **direction** in which pixels are moving.
- The **speed** of pixel movement.

Optical flow assumes (constraint):

- Brightness constancy: pixel intensity remains the same between consecutive frames.
- Small motion between frames.
- Smooth motion in neighboring pixels.



Optical flow vectors



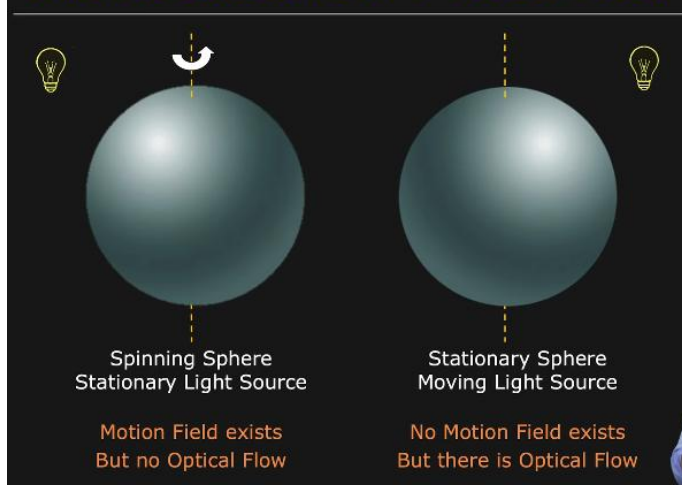
What is the difference between (Optical flow and Motion field) ?

Motion Field: The true physical motion of objects (or points) in the 3D world projected onto the 2D image plane.

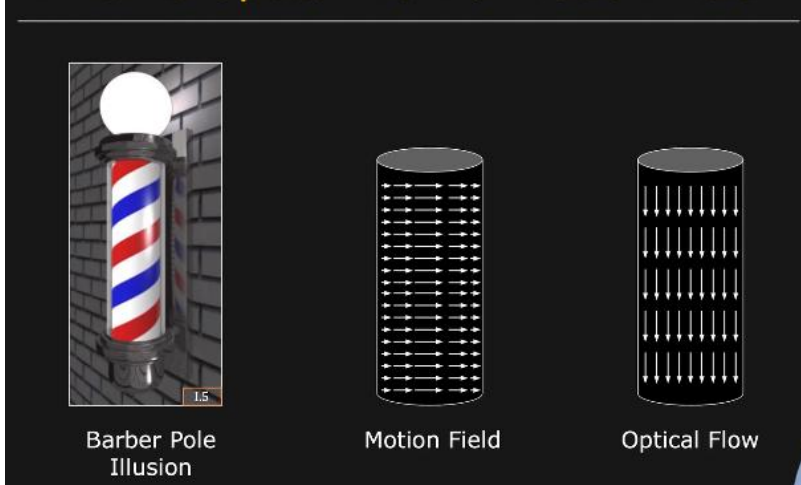
Optical Flow: The apparent motion computed from image intensity changes between two frames.

Ideally, Optical Flow = Motion Field, However Sometimes its not

When is Optical Flow \neq Motion Field?



When is Optical Flow \neq Motion Field?



Types of Optical Flow Estimation:

- Dense Optical Flow:**
 - Calculates motion vectors for **every pixel** in the frame.
 - Example method: **Farneback's method**.
 - Useful for capturing global motion fields.
- Sparse Optical Flow:**
 - Calculates motion vectors for **selected key points** (e.g., corners).
 - Example method: **Lucas-Kanade method**.
 - More efficient for tracking prominent features in videos.



(a) Sparse Optical Flow – Lukas Kanade



(b) Dense Optical Flow - Gunnar Farneback

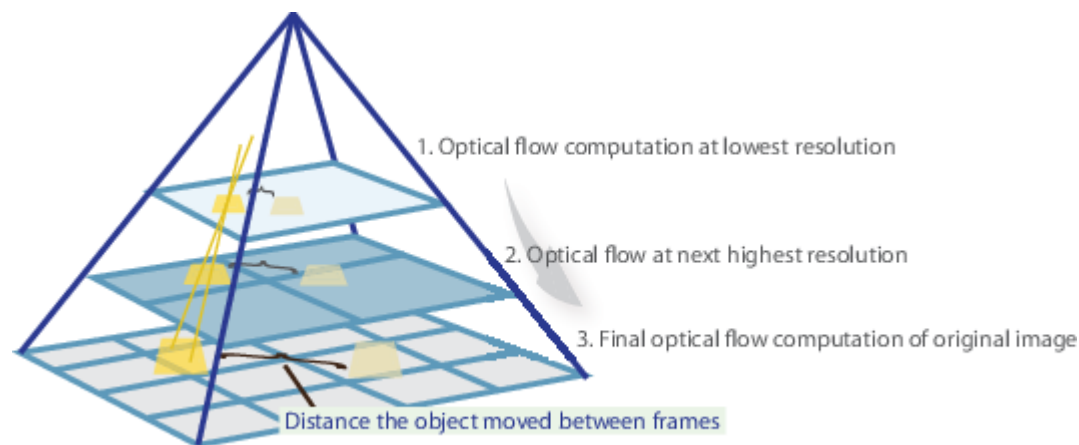
Exercise 1: Dense Optical Flow with Farneback Method

Objective:

Compute dense optical flow for all pixels between two consecutive frames.

Method Overview:

The Farneback algorithm approximates pixel neighborhoods with polynomials to estimate motion.



Steps:

1. Read two consecutive frames from a video.

2. Convert both frames to grayscale:

```
prev_gray = cv.cvtColor(prev_frame, cv.COLOR_BGR2GRAY)
next_gray = cv.cvtColor(next_frame, cv.COLOR_BGR2GRAY)
```

3. Compute the dense optical flow:

```
flow = cv.calcOpticalFlowFarneback(prev_gray, next_gray, None,
                                    pyr_scale=0.5, levels=3, winsize=15,
                                    iterations=3, poly_n=5, poly_sigma=1.2,
                                    flags=0)
```

4. Visualize the:

- Use color coding (hue for direction, brightness for magnitude).

Farneback Function Parameters Overview:

- `pyr_scale`: Image scale (<1) for pyramid building.
- `levels`: Number of pyramid layers.
- `winsize`: Window size for averaging.
- `iterations`: Iterations at each pyramid level.
- `poly_n, poly_sigma`: Parameters for polynomial expansion smoothing.
- `flags`: Optional flags like `OPTFLOW_USE_INITIAL_FLOW` or `OPTFLOW_FARNEBACK_GAUSSIAN`.

Exercise 2: Sparse Optical Flow with Lucas-Kanade Method

Objective:

Track a set of selected points (features) between consecutive frames.

Method Overview:

The Lucas-Kanade method tracks points by solving motion equations over a small neighborhood.

Steps:

1. Detect good features to track in the first frame:

```
prev_pts = cv.goodFeaturesToTrack(prev_gray, maxCorners=100, qualityLevel=0.3,
minDistance=7)
```

2. Compute the optical flow for these points:

```
next_pts, status, err = cv.calcOpticalFlowPyrLK(prev_gray, next_gray, prev_pts,
None, winSize=(21,21), maxLevel=3,
          criteria=(cv.TERM_CRITERIA_EPS |
cv.TERM_CRITERIA_COUNT, 30, 0.01))
```

3. Draw the tracked points:

- Draw lines or points to show how features have moved.

Lucas-Kanade Function Parameters Overview:

- `winSize`: Size of the search window.
- `maxLevel`: Number of pyramid levels.
- `criteria`: Termination criteria (max iterations or minimum movement).
- `flags`: Control behavior (e.g., use initial flow, get minimum eigenvalues).

Key Differences Recap:

Dense Optical Flow (Farneback)	Sparse Optical Flow (Lucas-Kanade)
Motion estimated for every pixel.	Motion estimated for selected points only.
More computationally expensive.	More efficient for tracking.
Useful for global motion analysis.	Useful for tracking specific features (e.g., corners, edges).

Additional reference:

1. https://docs.opencv.org/3.4/d4/dee/tutorial_optical_flow.html
2. https://www.youtube.com/watch?v=WrlH5hHv0gE&t=237s&ab_channel=NicolaiNielsen-ComputerVision%26AI
3. https://www.youtube.com/watch?v=hfXMw2dQO4E&ab_channel=NicolaiNielsen-ComputerVision%26AI

Coarse-to-fine Estimation with Median Filter

