



## Lab 6

### SIFT Detector and Features Matching with OpenCV

#### Learning Objectives:

- Understand how SIFT works and why it is useful.
- Detect keypoints and compute descriptors using SIFT.
- Perform feature matching between two images..
- Use SIFT for applications like object recognition and image stitching

## Introduction

Feature detection and matching are fundamental techniques in computer vision used to identify corresponding points in different images. These techniques enable tasks such as **object recognition**, **image stitching**, **3D reconstruction**, **augmented reality**, and **autonomous navigation**. Among the most powerful feature detection methods is the **Scale-Invariant Feature Transform (SIFT)** algorithm, which extracts unique keypoints and descriptors that remain consistent across different scales, rotations, and lighting conditions.

### 1 Understanding SIFT (Scale-Invariant Feature Transform)

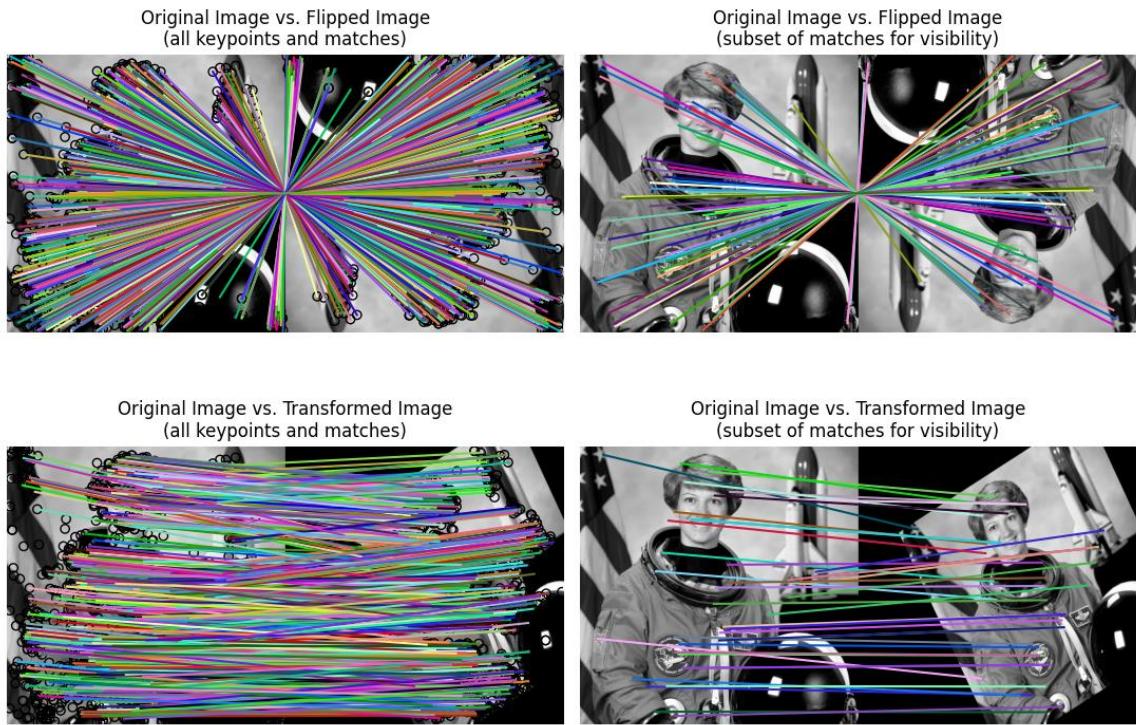
#### 📌 Why Is SIFT Important?

SIFT is a widely used method for detecting and describing local features in images. Unlike traditional edge detection methods, SIFT keypoints are:

- **Scale-invariant:** They remain consistent regardless of the image size.
- **Rotation-invariant:** They can be detected in different orientations.
- **Illumination-robust:** They work well under varying lighting conditions.

#### 📌 How Does SIFT Work?

- **Scale-space extrema detection:** Identifies keypoints at different scales.
- **Keypoint localization:** Filters out low-contrast keypoints.
- **Orientation assignment:** Ensures keypoints are rotation-invariant.
- **Descriptor computation:** Creates a feature vector for each keypoint.



## 2 Detecting SIFT Features

### 🛠 Task: Extract SIFT Keypoints from an Image

```
import cv2
import numpy as np

# Load an image
gray = cv2.imread('image.jpg', cv2.IMREAD_GRAYSCALE)

# Initialize SIFT detector
sift = cv2.SIFT_create()

# Detect keypoints and compute descriptors
keypoints, descriptors = sift.detectAndCompute(gray, None)

# Draw keypoints on the image
img_keypoints = cv2.drawKeypoints(gray, keypoints, None,
flags=cv2.DRAW_MATCHES_FLAGS_DRAW_RICH_KEYPOINTS)

cv2.imshow('SIFT Keypoints', img_keypoints)
cv2.waitKey(0)
cv2.destroyAllWindows()
```

## 3 Understanding Feature Matching

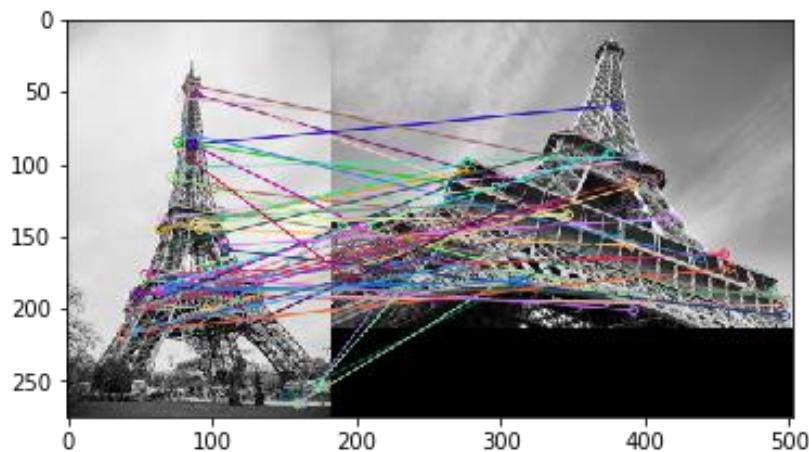
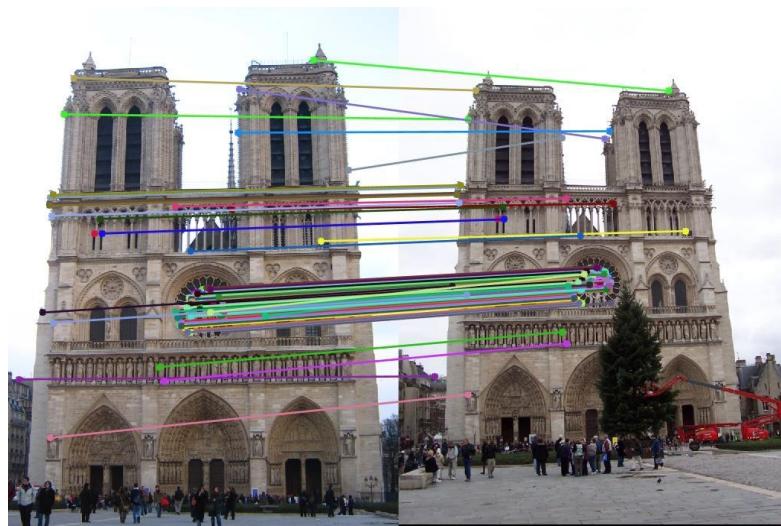
### 📌 Why Is Feature Matching Important?

Feature matching allows us to compare images and recognize patterns despite changes in **scale, rotation, illumination, and perspective**. It enables:

- **Object recognition:** Identifying objects in different images.
- **Image stitching:** Combining multiple images to create panoramas.
- **3D reconstruction:** Estimating depth and structure from multiple images.
- **Augmented reality (AR):** Detecting real-world surfaces to overlay virtual content.
- **Autonomous navigation:** Enabling robots and self-driving cars to recognize landmarks and objects.

### 📌 How Does Feature Matching Work?

1. **Feature Detection:** Identify keypoints in an image (e.g., edges, corners, textures).
2. **Descriptor Computation:** Generate unique numerical representations of each keypoint.
3. **Feature Matching:** Compare descriptors between two images to find corresponding features.
4. **Filtering Matches:** Apply techniques like Lowe's ratio test to remove weak matches.



## 4 Feature Matching with SIFT

### Task: Match Features Between Two Images

```
# Load two images
gray1 = cv2.imread('image1.jpg', cv2.IMREAD_GRAYSCALE)
gray2 = cv2.imread('image2.jpg', cv2.IMREAD_GRAYSCALE)

# Detect and compute features
sift = cv2.SIFT_create()
kp1, des1 = sift.detectAndCompute(gray1, None)
kp2, des2 = sift.detectAndCompute(gray2, None)

# Use BFMatcher for feature matching
bf = cv2.BFMatcher()
matches = bf.knnMatch(des1, des2, k=2)

# Apply ratio test (Lowe's ratio test)
good_matches = []
for m, n in matches:
    if m.distance < 0.75 * n.distance:
        good_matches.append(m)

# Draw matches
img_matches = cv2.drawMatches(gray1, kp1, gray2, kp2, good_matches, None,
flags=cv2.DrawMatchesFlags_NOT_DRAW_SINGLE_POINTS)

cv2.imshow('SIFT Feature Matching', img_matches)
cv2.waitKey(0)
cv2.destroyAllWindows()
```

## Exercise 1: Extract SIFT Keypoints from an Image

### **Objective:**

Extract and visualize SIFT keypoints from an image to understand how local features are detected.

Instructions:

1. *Load an image and convert it to grayscale.*
2. *Use SIFT to detect keypoints and compute descriptors.*
3. *Draw the detected keypoints on the image and display it.*

## Exercise 2: Match Features Between Two Images

### **Objective:**

Match features between two images using SIFT and Brute Force Matcher to find correspondences.

Instructions:

1. *Load two images and convert them to grayscale.*
2. *Detect and compute SIFT features in both images.*
3. *Use Brute Force Matcher (BFMatcher) to find feature correspondences.*
4. *Filter matches using Lowe's ratio test.*
5. *Draw the matching keypoints between the two images and display them.*

### **References:**

- [OpenCV](#)
- [Good Article](#)

- [Youtube](#)