

# Documentation for Exercises No. 2: Feature Extraction and Object Detection

## 1. Introduction

This notebook demonstrates the process of feature extraction and object detection using different feature detection methods, such as SIFT (Scale-Invariant Feature Transform), SURF (Speeded-Up Robust Features), and ORB (Oriented FAST and Rotated BRIEF). The steps include loading images, extracting features, matching features between images, and aligning images based on feature matching.

### Key Tasks:

1. **SIFT Feature Extraction**
2. **SURF Feature Extraction**
3. **ORB Feature Extraction**
4. **Feature Matching**
5. **Applications of Feature Matching (Homography and Image Alignment)**
6. **Combining Different Feature Extraction Methods**

## 2. Importing Necessary Libraries

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
from google.colab import files
```

These libraries are required for image processing (using OpenCV), numerical operations (using NumPy), and visualization (using Matplotlib).

## 3. Loading Images

```
image1 = cv2.imread('/iIMAGE2.jpg')
image2 = cv2.imread('/IMAGE3.jpg')
```

We load two images (image1 and image2) from the file system. The images are then verified for successful loading.

### Verification:

```
if image1 is None:
    print("Error loading image1!")
else:
    print(f"Image1 loaded with shape: {image1.shape}")
```

```

if image2 is None:
    print("Error loading image2!")
else:
    print(f"Image2 loaded with shape: {image2.shape}")

```

This code ensures that both images are loaded correctly. If an image fails to load, it prints an error message.

## 4. Converting Images to Grayscale

```

gray1 = cv2.cvtColor(image1, cv2.COLOR_BGR2GRAY)
gray2 = cv2.cvtColor(image2, cv2.COLOR_BGR2GRAY)

```

Both images are converted to grayscale to simplify feature detection. This step is necessary because feature extraction algorithms generally work with grayscale images.

## 5. Task 1: SIFT Feature Extraction

```

sift = cv2.SIFT_create()
kp1_sift, des1_sift = sift.detectAndCompute(gray1, None)
kp2_sift, des2_sift = sift.detectAndCompute(gray2, None)

```

SIFT (Scale-Invariant Feature Transform) is used to detect keypoints and compute descriptors for the images.

### Displaying the Keypoints:

```

image_matches_sift = cv2.drawMatches(image1, kp1_sift, image2, kp2_sift,
None, None, flags=cv2.DrawMatchesFlags_NOT_DRAW_SINGLE_POINTS)
plt.imshow(cv2.cvtColor(image_matches_sift, cv2.COLOR_BGR2RGB))
plt.title("SIFT Feature Extraction")
plt.axis('off')
plt.show()
cv2.imwrite('/content/4B-BERCADES-EXER2-sift-keypoints.jpg',
image_matches_sift)

```

This code visualizes the keypoints detected by SIFT and saves the image for future use.

## 6. Task 2: SURF Feature Extraction

SURF (Speeded-Up Robust Features) is another feature detection algorithm, but it is non-free and may not be available in all OpenCV installations. We attempt to initialize the SURF detector:

```

try:
    surf = cv2.xfeatures2d.SURF_create()
    print("SURF is ready to use!")
except:

```

```
print("SURF is still not available.")
```

If SURF is available, we proceed with feature extraction:

```
keypoints, descriptors = surf.detectAndCompute(gray1, None)
image_with_keypoints = cv2.drawKeypoints(image1, keypoints, None,
flags=cv2.DRAW_MATCHES_FLAGS_DRAW_RICH_KEYPOINTS)
plt.imshow(cv2.cvtColor(image_with_keypoints, cv2.COLOR_BGR2RGB))
plt.title("SURF Keypoints")
plt.axis('off')
plt.show()
```

### **Note:**

SURF requires non-free modules, which may be disabled in certain OpenCV builds.

## **7. Task 3: ORB Feature Extraction**

ORB (Oriented FAST and Rotated BRIEF) is a fast and efficient feature detector and descriptor.

```
orb = cv2.ORB_create()
kp1_orb, des1_orb = orb.detectAndCompute(gray1, None)
kp2_orb, des2_orb = orb.detectAndCompute(gray2, None)
```

ORB keypoints are detected and displayed as follows:

```
image_matches_orb = cv2.drawMatches(image1, kp1_orb, image2, kp2_orb, None,
None, flags=cv2.DrawMatchesFlags_NOT_DRAW_SINGLE_POINTS)
plt.imshow(cv2.cvtColor(image_matches_orb, cv2.COLOR_BGR2RGB))
plt.title("ORB Feature Extraction")
plt.axis('off')
plt.show()
cv2.imwrite('/content/4B-BERCADES-EXER2-orb-keypoints.jpg',
image_matches_orb)
```

## **8. Task 4: Feature Matching**

### **SIFT Matching**

```
bf_sift = cv2.BFMatcher(cv2.NORM_L2, crossCheck=True)
matches_sift = bf_sift.match(des1_sift, des2_sift)
matches_sift = sorted(matches_sift, key=lambda x: x.distance)
```

### **ORB Matching**

```
bf_orb = cv2.BFMatcher(cv2.NORM_HAMMING, crossCheck=True)
matches_orb = bf_orb.match(des1_orb, des2_orb)
matches_orb = sorted(matches_orb, key=lambda x: x.distance)
```

We use the brute-force matcher for both SIFT and ORB descriptors and sort the matches by distance to find the best ones.

## Displaying the First 10 Matches for SIFT and ORB

```
image_matches_sift_combined = cv2.drawMatches(image1, kp1_sift, image2,
kp2_sift, matches_sift[:10], None,
flags=cv2.DrawMatchesFlags_NOT_DRAW_SINGLE_POINTS)
image_matches_orb_combined = cv2.drawMatches(image1, kp1_orb, image2,
kp2_orb, matches_orb[:10], None,
flags=cv2.DrawMatchesFlags_NOT_DRAW_SINGLE_POINTS)
```

We display the top 10 matches for both SIFT and ORB in side-by-side images.

## 9. Task 5: Applications of Feature Matching

### Homography and Image Alignment Using ORB Matches

```
pts1_orb = np.float32([kp1_orb[m.queryIdx].pt for m in
matches_orb]).reshape(-1, 1, 2)
pts2_orb = np.float32([kp2_orb[m.trainIdx].pt for m in
matches_orb]).reshape(-1, 1, 2)

H_orb, mask_orb = cv2.findHomography(pts1_orb, pts2_orb, cv2.RANSAC, 5.0)
aligned_image_orb = cv2.warpPerspective(image1, H_orb, (image2.shape[1],
image2.shape[0]))
```

This code uses RANSAC to compute the homography matrix and then warps the first image to align with the second image.

### Displaying the Aligned Image

```
plt.imshow(cv2.cvtColor(aligned_image_orb, cv2.COLOR_BGR2RGB))
plt.title("Aligned Image using Homography (ORB)")
plt.axis('off')
plt.show()
```

## 10. Task 6: Combining Feature Extraction Methods

In this task, SIFT and ORB feature extraction methods are applied, and their matches are displayed and aligned using homography.

```
# Repeat SIFT and ORB matching and homography steps
```

We visualize the results of SIFT and ORB matches and aligned images for further analysis.

## 11. Conclusion

In this exercise, we explored three feature extraction methods: **SIFT**, **SURF**, and **ORB**, applied feature matching, and demonstrated practical applications such as image alignment through homography. The methods are useful for image stitching, object recognition, and other computer vision tasks.

## 12. Image Download

The results are saved and made available for download:

```
cv2.imwrite('/content/4B-BERCADES-EXER2-sift-matches.jpg',
image_matches_sift_combined)
cv2.imwrite('/content/4B-BERCADES-EXER2-orb-matches.jpg',
image_matches_orb_combined)
cv2.imwrite('/content/4B-BERCADES-EXER2-aligned-orb.jpg', aligned_image_orb)
files.download('/content/4B-BERCADES-EXER2-sift-matches.jpg')
files.download('/content/4B-BERCADES-EXER2-orb-matches.jpg')
files.download('/content/4B-BERCADES-EXER2-aligned-orb.jpg')
```

This code allows users to download the output images of matches and aligned images after processing.

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This documentation provides a step-by-step guide through the feature extraction, matching, and application processes in object detection using OpenCV.