

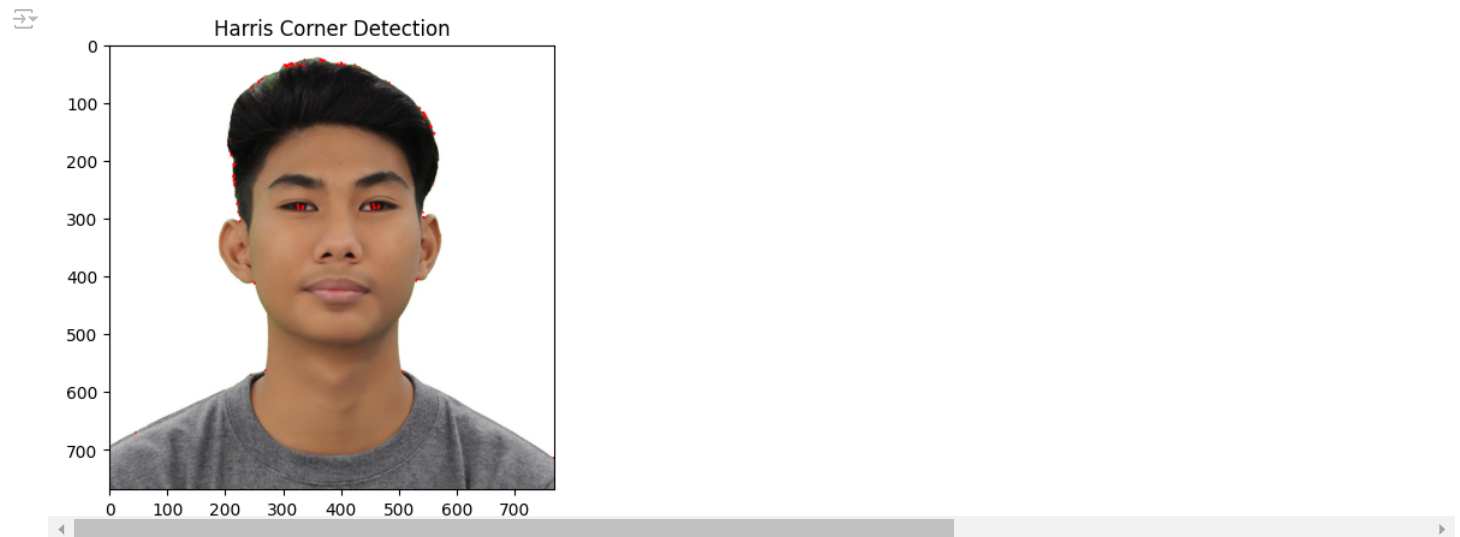
## ✓ Download Libraries

```
!pip install opencv-python opencv-python-headless
```

```
Requirement already satisfied: opencv-python in /usr/local/lib/python3.10/dist-packages (4.10.0.84)  
Requirement already satisfied: opencv-python-headless in /usr/local/lib/python3.10/dist-packages (4.10.0.84)  
Requirement already satisfied: numpy>=1.21.2 in /usr/local/lib/python3.10/dist-packages (from opencv-python) (1.26.4)
```

## ✓ Exercise 1. Harris Corner Detection

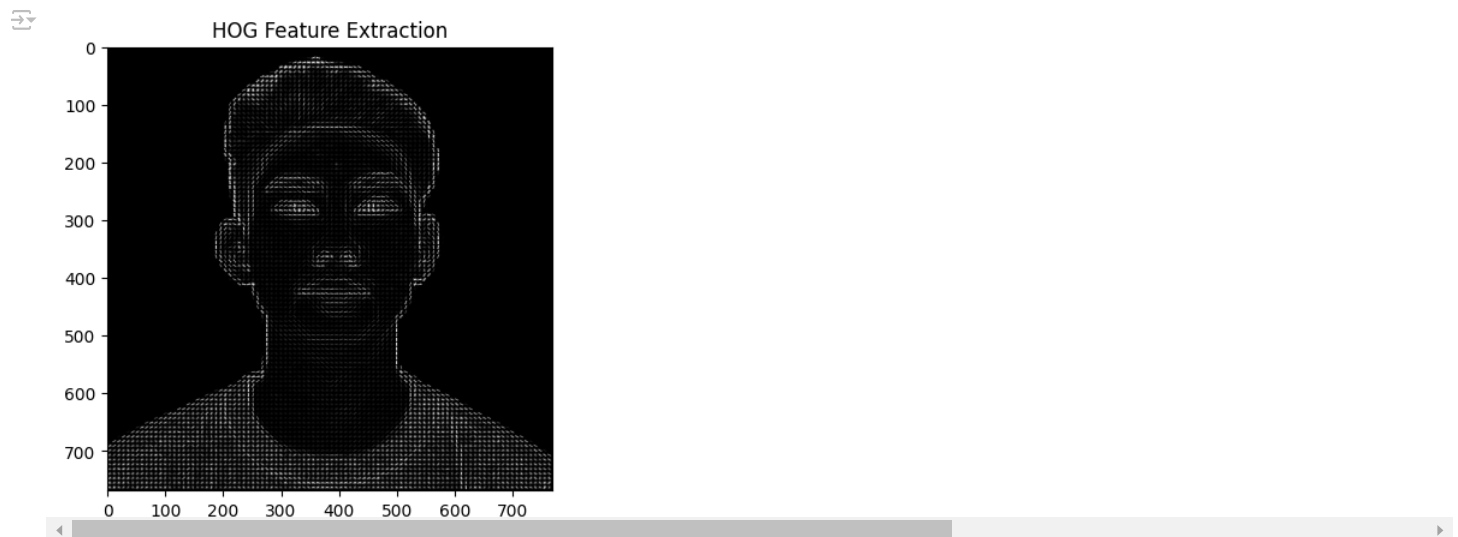
```
import cv2  
import numpy as np  
import matplotlib.pyplot as plt  
  
# Load the image  
image = cv2.imread('Pictureme.jpg')  
gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)  
  
# Harris Corner Detection  
gray = np.float32(gray)  
harris_corners = cv2.cornerHarris(gray, 2, 3, 0.04)  
  
# Dilate corner image for better visualization  
harris_corners = cv2.dilate(harris_corners, None)  
  
# Mark the corners in the original image  
image[harris_corners > 0.01 * harris_corners.max()] = [0, 0, 255]  
  
# Display the image  
plt.imshow(cv2.cvtColor(image, cv2.COLOR_BGR2RGB))  
plt.title('Harris Corner Detection')  
plt.show()
```



## ✓ Exercise 2. HOG (Histogram of Oriented Gradients) Feature Extraction

```
from skimage.feature import hog  
from skimage import data, exposure  
  
# Load an image  
image = cv2.imread('Pictureme.jpg')  
gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)  
  
# HOG feature extraction  
hog_features, hog_image = hog(gray, orientations=9, pixels_per_cell=(8, 8),  
                               cells_per_block=(2, 2), visualize=True, block_norm='L2-Hys')  
  
# Enhance the HOG image for better visualization  
hog_image_rescaled = exposure.rescale_intensity(hog_image, in_range=(0, 10))  
  
# Display the image  
plt.imshow(hog_image_rescaled, cmap='gray')
```

```
plt.title('HOG Feature Extraction')
plt.show()
```



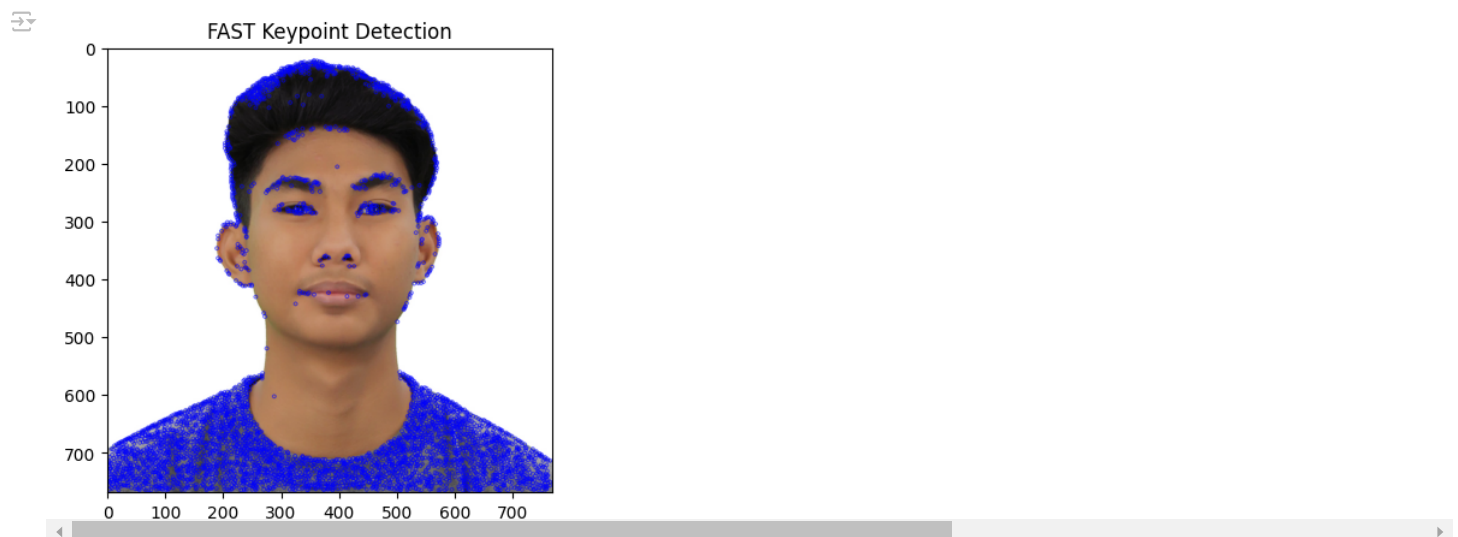
### Exercise 3. FAST Keypoint Detection

```
# Load an image
image = cv2.imread('Pictureme.jpg')
gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)

# FAST Keypoint Detector
fast = cv2.FastFeatureDetector_create()
keypoints = fast.detect(gray, None)

# Draw keypoints on the image
output_image = cv2.drawKeypoints(image, keypoints, None, color=(255, 0, 0))

# Display the image
plt.imshow(cv2.cvtColor(output_image, cv2.COLOR_BGR2RGB))
plt.title('FAST Keypoint Detection')
plt.show()
```



### Exercise 4. Feature Matching using ORB and FLANN

```
# Load two images
img1 = cv2.imread('Pictureme.jpg', 0)
img2 = cv2.imread('Pictureme.jpg', 0)

# Initialize ORB detector
orb = cv2.ORB_create()
```

```
# Find keypoints and descriptors with ORB
kp1, des1 = orb.detectAndCompute(img1, None)
kp2, des2 = orb.detectAndCompute(img2, None)

# Initialize FLANN-based matcher
FLANN_INDEX_LSH = 6
index_params = dict(algorithm=FLANN_INDEX_LSH, table_number=6, key_size=12, multi_probe_level=1)
search_params = dict(checks=50)

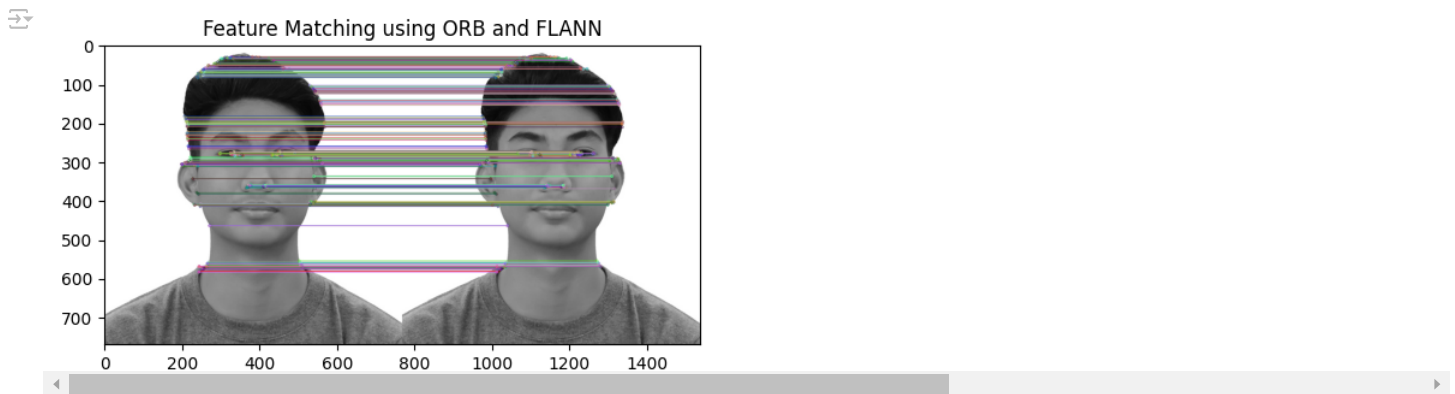
flann = cv2.FlannBasedMatcher(index_params, search_params)

# Match descriptors
matches = flann.knnMatch(des1, des2, k=2)

# Store all the good matches
good_matches = []
for m, n in matches:
    if m.distance < 0.75 * n.distance:
        good_matches.append(m)

# Draw matches
result = cv2.drawMatches(img1, kp1, img2, kp2, good_matches, None, flags=2)

# Display the matched features
plt.imshow(result)
plt.title('Feature Matching using ORB and FLANN')
plt.show()
```



## Exercise 5. Image Segmentation using Watershed Algorithm

```
import cv2
import numpy as np
import matplotlib.pyplot as plt

# Load an image
image = cv2.imread('Pictureme2.jpg')
gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)

# Apply threshold to convert the image to binary
_, thresh = cv2.threshold(gray, 0, 255, cv2.THRESH_BINARY_INV + cv2.THRESH_OTSU)

# Noise removal using Morphological operations
kernel = np.ones((3, 3), np.uint8)
opening = cv2.morphologyEx(thresh, cv2.MORPH_OPEN, kernel, iterations=2)

# Finding sure background area
sure_bg = cv2.dilate(opening, kernel, iterations=3)

# Finding sure foreground area
dist_transform = cv2.distanceTransform(opening, cv2.DIST_L2, 5)
_, sure_fg = cv2.threshold(dist_transform, 0.7 * dist_transform.max(), 255, 0)

# Finding unknown region
sure_fg = np.uint8(sure_fg)
unknown = cv2.subtract(sure_bg, sure_fg)

# Marker labeling
_, markers = cv2.connectedComponents(sure_fg)

# Add 1 to all labels so that sure background is not 0, but 1
markers = markers + 1

# Mark the unknown regions with 0
markers[unknown == 0] = 0
```

```
# Apply the Watershed algorithm
markers = cv2.watershed(image, markers)

# Mark the boundaries in red
image[markers == -1] = [255, 0, 0]

# Optional: Enhance boundaries for better visualization
boundary_img = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
boundary_img[markers == -1] = [255, 0, 0] # Set boundaries to red

# Display the segmented regions with boundaries in red
plt.figure(figsize=(10, 8))
plt.imshow(boundary_img)
plt.title('Watershed Segmentation')
plt.show()
```

