Step1: Install OpenCV

!pip install opencv-python-headless

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-headless) (1.26.4)

Step 2 : Import Libraries

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
from google.colab import files
from io import BytesIO
from PIL import Image
# Function to display an image using matplotlib
def display_image(ing, title="Image"):
plt.imshow(cv2.cvtColor(ing, cv2.COLOR_BGR2RGB))
plt.title(title)
plt.axis('off')
plt.show()
# Function to display two images side by side
def display_images(ing1, ing2, title1="Image 1", title2="Image 2"):
plt.subplot(1, 2, 1)
plt.imshow(cv2.cvtColor(ing1, cv2.COLOR_BGR2RGB))
plt.title(title1)
plt.axis('off')
plt.subplot(1, 2, 2)
plt.imshow(cv2.cvtColor(ing2, cv2.COLOR_BGR2RGB))
plt.title(title2)
plt.simshow(cv2.cvtColor(ing2, cv2.COLOR_BGR2RGB))
plt.title(title2)
plt.show()
```

Step 3: Load an Image

```
# Upload an image
uploaded = files.upload()
# Convert to OpenCV format
image_path = next(iter(uploaded)) # Get the image file name
image = Image.open(BytesIO(uploaded[image_path]))
image = cv2.cvtColor(np.array(image), cv2.COLOR_RGB2BGR)
display_image(image, "Original Image")
```

Choose Files | Iorin id.png

lorin_id.png(image/png) - 375308 bytes, last modified: 9/16/2024 - 100% done



Exercise 1: Scaling and Rotation

```
# Scaling

def scale_image(img, scale_factor):
    height, width = img.shape(:2)
    scaled_img = cv2.resize(img,
    (int(width * scale_factor), int(height * scale_factor)), interpolation=cv2.INTER_LINEAR)
    return scaled_img
    """

scale_image(): This function scales the image by a given factor.

The cv2.resize() function takes the original dimensions of the image,
    multiplies them by the scale_factor, and resizes the image accordingly.

INTER_LINEAR is a common interpolation method for resizing.
    """

# Rotate

def rotate_image(img, angle):
    height, width = img.shape(:2)
    center = (width // 2, height // 2)
    matrix = cv2.getRotationMatrix2D(center, angle, 1.0)
    rotated_img = cv2.warpAffine(img, matrix, (width, height))
    return rotated_img
    """

rotate_image(): Rotates the image around its center. cv2.getRotationMatrix2D() creates
```

```
a transformation matrix for rotation, and cv2.warpAffine() applies this transformation.

The angle parameter controls the degree of rotation.

"""

# Scale image by 0.5

scaled_image = scale_image(image, 0.5)

display_image(scaled_image, "scaled Image (50%)")

# Rotate image by 45 degrees

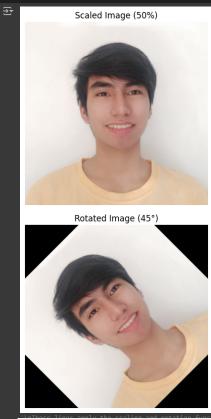
rotated_image = rotate_image(image, 45)

display_image(rotated_image, "Rotated Image (45°)")

"""

These lines apply the scaling and rotation functions to the uploaded image and display the results.

"""
```



Exercise 2: Blurring Techniques

```
# Gaussian Blur
gaussian_blur = cv2.GaussianBlur(image, (5, 5), 0)
display_image(gaussian_blur, "Gaussian Blur (5x5)")

"""

cv2.GaussianBlur(): Applies a Gaussian blur to the image, which smooths it by averaging
the pixel values in a 5x5 kernel (a small matrix). This is useful for reducing noise in an image.

"""

# Median Blur
median_blur = cv2.medianBlur(image, 5)
display_image(median_blur, "Median Blur (5x5)")

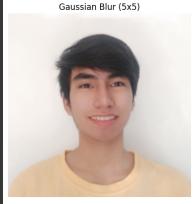
"""

cv2.medianBlur(): Applies a median blur, which replaces each pixel's value with the median value of its neighbors in a 5x5 kernel. This method is particularly effective in removing salt-and-pepper noise.

"""
```



∓



Median Blur (5x5)

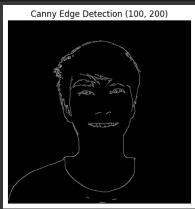


'\ncv2.medianBlur(): Applies a median blur, which replaces each pixel's value with the\nmedian value of its neighbors in a 5x5 kernel. This method is particularly effective in\nremoving

Exercise 3: Edge Detection using Canny

```
# Canny Edge Detection
edges = cv2.Canny(image, 100, 200)
display_image(edges, "Canny Edge Detection (100, 200)")

"""
cv2.Canny(): Detects edges in the image by calculating the gradient (rate of intensity change)
between pixels. The two threshold values (100 and 200) define the edges'
sensitivity. Lower thresholds detect more edges, while higher thresholds detect only the
most prominent edges.
"""
```



'\ncv2.Canny(): Detects edges in the image by calculating the gradient (rate of intensity change)\nbetween pixels. The two threshold values (100 and 200) define the edges \nsensitivity lower thresholds datest energy adds while higher thresholds datest only without considering adds in

Exercise 4: Basic Image Processor (Interactive)

```
def process_image(img, action):
   if action == 'scale':
        return scale_image(img, 0.5)
   elif action == 'rotate':
        return rotate_image(img, 45)
   elif action == 'gaussian_blur':
        return cv2.GaussianBlur(img, (5, 5), 0)
   elif action == 'median_blur':
        return cv2.medianBlur(img, 5)
   elif action == 'canny':
        return cv2.Caunny(img, 100, 200)
```

```
else:
    return img

"""

process_image(): This function allows users to specify an image transformation (scaling, rotation, blurring, or edge detection). Depending on the action passed, it will apply the corresponding image processing technique and return the processed image.

"""

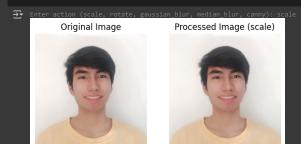
action = input("Enter action (scale, rotate, gaussian_blur, median_blur, canny): ")

processed_image = process_image(image, action)

display_images(image, processed_image, "Original Image", f"Processed Image ({action})")

"""

This allows users to enter their desired transformation interactively (via the input() function). It processes the image and displays both the original and transformed versions side by side.
```



"\nThis allows users to enter their desired transformation interactively (via the\ninput() function). It processes the image and displays both the original and transformed\nversions sid

Exercise 5: Comparison of Filtering Techniques

```
# Applying Gaussian, Median, and Bilateral filters
gaussian_blur = cv2.GaussianBlur(image, (5, 5), 0)
median_blur = cv2.medianBlur(image, 5)
bilateral_filter = cv2.bilateralFilter(image, 9, 75, 75)

"""
cv2.bilateralFilter(): This filter smooths the image while keeping edges sharp, unlike
Gaussian or median filters. It's useful for reducing noise while preserving details.

"""

# Display the results for comparison
plt.figure(figsize=(10, 5))
plt.subplot(1, 3, 1)
plt.imshow(cv2.cvtColor(gaussian_blur, cv2.COLOR_BGR2RGB))
plt.title("Gaussian Blur")

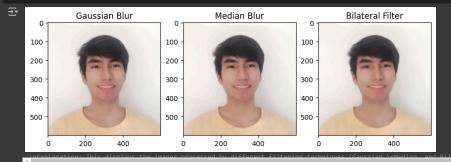
plt.subplot(1, 3, 2)
plt.imshow(cv2.cvtColor(median_blur, cv2.COLOR_BGR2RGB))
plt.title("Median Blur")

plt.subplot(1, 3, 3)
plt.imshow(cv2.cvtColor(bilateral_filter, cv2.COLOR_BGR2RGB))
plt.title("Bilateral Filter")

plt.show()

"""

Explanation: This displays the images processed by different filtering techniques (Gaussian, Median, and Bilateral) side by side for comparison.
```

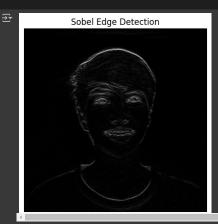


Sobel Edge Detection

```
def sobel_edge_detection(img):
    #convert to gratscale
    gray = cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
    #Sobel edge detection in the x direction
    sobelx = cv2.Sobel(gray,cv2.CV_64F,0,1,ksize=5)
    #Sobel edge detection in the y direction
    sobely = cv2.Sobel(gray,cv2.CV_64F,0,1,ksize=5)
    #Combine the two gradients
    sobel_combined=cv2.magnitude(sobelx,sobely)
```

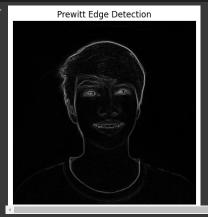
```
return sobel_combined

#Apply Sobel edge detection to the uploaded image sobel_edges = sobel_edge_detection(image) plt.imshow(sobel_edges,cmap='gray') plt.title('Sobel Edge Detection') plt.axis('off') plt.sxis('off')
```



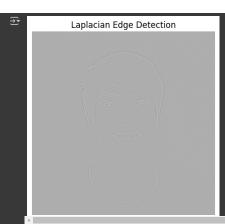
Prewitt Edge Detection

```
def prewitt_edge_detection(img):
    #convert to grayscale
    gray = cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
    #prewitt operator kernels for x and y directions
    kernelx = np.array([[1, 0, -1], [1, 0, -1], [1, 0, -1]], dtype=int)
    kernely = np.array([[1, 1, 1], [0, 0, 0], [-1, -1, -1]], dtype=int)
    # Applying the Prewitt operator
    prewittx = cv2.filter2D(gray, cv2.CV_64F, kernelx)
    prewitty = cv2.filter2D(gray, cv2.CV_64F, kernely)
    # Combine the x and y gradients by converting to floating point
    prewitt_combined = cv2.magnitude(prewittx, prewitty)
    return prewitt_combined
# Apply Prewitt edge detection to the uploaded image
    prewitt_edges = prewitt_edge_detection(image)
    plt.imshow(prewitt_edges, cmap='gray')
    plt.title("Prewitt Edge Detection")
    plt.sxis('off')
    plt.show()
```



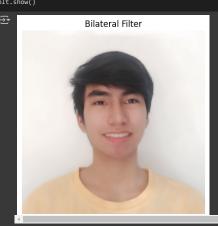
Laplacian Edge Detection

```
def laplacian_edge_detection(img):
    # Convert to grayscale
    gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    # Apply Laplacian operator
    laplacian = cv2.Laplacian(gray, cv2.CV_64F)
    return laplacian
# Apply Laplacian edge detection to the uploaded image
laplacian_edges = laplacian_edge_detection(image)
plt.imshow(laplacian_edges, cmap='gray')
plt.title("Laplacian Edge Detection")
plt.axis('off')
plt.show()
```



Bilateral Filter

```
def bilateral_blur(img):
   bilateral = cv2.bilateralFilter(img, 9, 75, 75)
   return bilateral
# Apply Bilateral filter to the uploaded image
   bilateral_blurred = bilateral_blur(image)
   plt.imshow(cv2.cvtColor(bilateral_blurred, cv2.COLOR_BGR2RGB))
   plt.title("Bilateral Filter")
   plt.axis('off')
   plt.show()
```



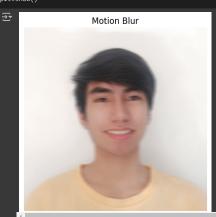
Box Filter

```
def box_blur(img):
    box = cv2.boxFilter(img, -1, (5, 5))
    return box
# Apply Box filter to the uploaded image
box_blurred = box_blur(image)
plt.imshow(cv2.cvtColor(box_blurred, cv2.COLOR_BGR2RGB))
plt.title("Box Filter")
plt.axis('off')
plt.show()
```



Motion Blur

```
# Create motion blur kernel (size 15x15)
kernel_size = 15
kernel = np.zeros((kernel_size, kernel_size))
kernel[int((kernel_size - 1) / 2), :] = np.ones(kernel_size)
kernel = kernel / kernel_size
# Apply motion blur
motion_blurred = cv2.filter2D(img, -1, kernel)
return motion_blurred
# Apply Motion blur to the uploaded image
motion_blurred = motion_blur(image)
plt.imshow(cv2.cvtColor(motion_blurred, cv2.COLOR_BGR2RGB))
plt.title("Motion Blur")
plt.axis('off')
```



Unsharp Masking (Sharpening)

```
def unsharp_mask(img):
    # Create a Gaussian blur version of the image
    blurred = cv2.GaussianBlur(img, (9, 9), 10.0)
    # Sharpen by adding the difference between the original and the blurred image
    sharpened = cv2.addWeighted(img, 1.5, blurred, -0.5, 0)
    return sharpened
# Apply Unsharp Masking to the uploaded image
sharpened_image = unsharp_mask(image)
plt.imshow(cv2.cvtColor(sharpened_image, cv2.COLOR_BGR2RGB))
plt.title("Unsharp Mask (Sharpening)")
plt.axis('off')
plt.show()
```



All Output

```
actions = ['scale', 'rotate', 'gaussian_blur', 'median_blur', 'canny', 'sobel', 'laplacian', 'prewitt', 'bilateral_blur', 'box_blur', 'motion_blur', 'unsharp_mask']

plt.figure(figsize=(15, 10))

for i, action in enumerate(actions):
    processed_image = process_image(image, action)

# Ensure the processed image is in the correct format for display
    if processed_image.dtype != 'uint8': # Convert to uint8 if not already
        processed_image = cv2.convert5caleAbs(processed_image)

# Display the processed image
plt.subplot(3, 4, i + 1)
plt.imshow(cv2.cvtColor(processed_image, cv2.ColOR_BGR2RGB))
plt.title(f"Processed Image ((action))")
plt.axis('off')

plt.show()
```

Processed Image (scale)



Processed Image (canny)



Processed Image (bilateral_blur)



Processed Image (rotate)



Processed Image (sobel)



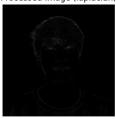
Processed Image (box_blur)



Processed Image (gaussian_blur)



Processed Image (laplacian)



Processed Image (motion_blur)



Processed Image (median_blur)



Processed Image (prewitt)



Processed Image (unsharp_mask)

