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FINAL ESSAY

DIGITAL IMAGE PROCESSING

HO CHI MINH CITY, 2024
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Advised by **Dr. Trinh Hung Cuong**

HO CHI MINH CITY, 2024

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INTRODUCTION TO DIGITAL IMAGE PROCESSING

This report outlines the methodology and results for two programming tasks related to image and video processing: detecting traffic signs in a video and detecting digits in an image. The tasks were approached using techniques such as color thresholding, contour detection, and drawing bounding boxes around detected objects, all implemented with OpenCV.

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CHAPTER 1. METHODOLOGY OF SOLVING TASKS

1.1 Related Background Knowledge and Methods

To detect traffic signs effectively in video frames, the solution employs a combination of digital image processing techniques and methods:

1.1.1 Color Filtering in HSV Color Space

- What is HSV:
 - The HSV (Hue, Saturation, Value) color model is more suitable than RGB for detecting specific colors like red, blue, and yellow because it separates chromatic information (color) from intensity.
 - **Hue** represents the type of color (red, blue, etc.).
 - Saturation measures the intensity of the color.
 - Value measures brightness.
- How HSV is used here:
 - Traffic signs are usually in standard red, blue, or yellow colors. The program converts each frame from **BGR** (default OpenCV format) to **HSV** using cv2.cvtColor() and creates binary masks for these colors:

```
lower_red = np.array([150, 50, 50]) # Lower bound of red
upper_red = np.array([180, 255, 255]) # Upper bound of red
# Single blue range
lower_blue = np.array([90, 120, 100]) # Lower bound of blue
upper_blue = np.array([130, 255, 255]) # Upper bound of blue
# Yellow range
lower_yellow = np.array([0, 120, 0]) # Lower bound of yellow
upper_yellow = np.array([30, 255, 255]) # Upper bound of yellow
```

Image 1.1 Color Filtering in HSV Color Space

• A binary mask for each color is created using cv2.inRange(). Pixels within the range are set to white (255), while others are black (0).

1.1.2 Morphological Operations

- **Purpose**: After creating the binary masks, noise can appear as small white regions in the image that do not correspond to traffic signs.
- Steps Taken:
 - Opening (Noise Removal): Removes small white regions caused by noise. This is achieved by first eroding the mask and then dilating it.
 - Closing (Filling Gaps): Closes small gaps within the white regions, ensuring smoother detection.
- Code Example:

```
# Noise removal (Morphological operations)
kernel = np.ones((3, 3), np.uint8)
combined_mask = cv2.morphologyEx(combined_mask, cv2.MORPH_OPEN, kernel)
combined_mask = cv2.morphologyEx(combined_mask, cv2.MORPH_CLOSE, kernel)
```

Image 1.2 Noise removal (Morphological operations)

1.1.3 Contour Detection

- Contours are continuous curves joining all the points on the boundary of a connected white region in the binary mask.
- Key Step:
 - Use cv2.findContours() to detect contours in the combined mask.
 - Filter the contours by area (cv2.contourArea()) to eliminate small regions that are unlikely to be traffic signs.

• Aspect Ratio Check:

■ The bounding box around the contour is extracted using cv2.boundingRect(). Only contours with an aspect ratio close to 1:1 (square) are considered valid traffic signs.

1.1.4 Template Matching

- What is Template Matching?
 - Compares a region of interest (ROI) extracted from the frame with preloaded template images of traffic signs.
 - Uses cv2.matchTemplate() to calculate a similarity score between the ROI and each template.
 - The template with the highest similarity score (above a threshold of 0.55) is considered the best match.

• Steps for Template Matching:

- Resize the ROI and template to the same size.
- Convert both images to grayscale.
- Use cv2.matchTemplate() with a normalized correlation coefficient (cv2.TM_CCOEFF_NORMED) to calculate the similarity.
- Similarity threshold: 0.55.

1.1.5 Write the result and display:

- Draw the rectangle and the detected sign name on the frame.
- Write the output video using cv2.VideoWriter().

1.2 Source Code Presentation

1.2.1 Import libraries

```
import cv2
import numpy as np
import os
```

Image 1.3 Import libraries

- cv2: OpenCV library, used for image and video processing.
- numpy: A Python library for numerical computing, used to handle image data as matrices.

• os: A library to work with the file system, used to read the traffic sign templates from a folder.

1.2.2 Variables and main data structures

1.1.1.1 SIGN TEMPLATES

```
# Dictionary to store template images
SIGN_TEMPLATES = {}
```

Image 1.4 Dictionary to store template images

1.1.1.2 SIGN NAME MAPPING

```
# Dictionary to map template sign names to new names

SIGN_NAME_MAPPING = {
    "camnguocchieu": "Wrong way",
    "wrongway": "Wrong way",
    "huongdi1": "Keep right",
    "keepright": "Keep right",
    "no_left": "No left turn",
    "camdungxe": "No parking",
    "noparking": "No parking",
    "children": "Children",
    "slow": "Slow",
    "nostopandparking": "No stopping and parking",
    "camdungxedonha":"No parking",
    "camdungxedonha":"No parking",
    "camdungxe":"No parking",
}
```

Image 1.5 Dictionary to map template sign names to new names

- A dictionary to map the original sign names (from the sign_templates) to more descriptive or user-friendly names.
- **Purpose**: Customize the names of detected signs for display.

1.2.3 load templates() function

```
def load_templates(template_dir='sign_templates'):
    """Load template images from directory"""
    for filename in os.listdir(template_dir):
        if filename.endswith(('.png', '.jpg')):
            sign_name = os.path.splitext(filename)[0]
            template_path = os.path.join(template_dir, filename)
            template = cv2.imread(template_path)
            SIGN_TEMPLATES[sign_name] = template
```

Image 1.6 The load templates function

Functionality:

• Load all traffic sign templates from the folder sign_templates into the SIGN_TEMPLATES dictionary.

Steps:

- 1. **Iterate through files**: Use os.listdir(template dir) to list all files in the directory.
- 2. Filter image files: Only process files with extensions .png or .jpg.
- 3. Load images: Use cv2.imread(template path) to load each template image.
- 4. **Store templates**: Save each template into the dictionary SIGN_TEMPLATES with the filename (minus the extension) as the key.

1.2.4 match template() function

```
def match_template(roi, template, target_size=(100, 100)):
    """Compare ROI with template using template matching"""
    # Resize both images to same size
    roi_resized = cv2.resize(roi, target_size)
    template_resized = cv2.resize(template, target_size)

# Convert both to grayscale
    roi_gray = cv2.cvtColor(roi_resized, cv2.COLOR_BGR2GRAY)
    template_gray = cv2.cvtColor(template_resized, cv2.COLOR_BGR2GRAY)

# Template matching
    result = cv2.matchTemplate(roi_gray, template_gray, cv2.TM_CCOEFF_NORMED)
    return np.max(result)
```

Image 1.7 The match template function

Functionality:

• Compare a region of interest (ROI) in the frame with a traffic sign template using **template matching**.

Steps:

- 1. **Resize images:** Normalize the ROI and template to a consistent size (100x100) using cv2.resize.
- 2. **Convert to grayscale:** Simplify the images by converting them to grayscale using cv2.cvtColor.
- 3. **Perform template matching:** Use cv2.matchTemplate to calculate a similarity score between the ROI and the template.
- 4. **Return maximum score:** Extract the highest similarity score using np.max(result).

1.2.5 detect_traffic_signs() function

```
lower_red = np.array([150, 50, 50]) # Lower bound of red
upper_red = np.array([180, 255, 255]) # Upper bound of red
         lower_blue = np.array([98, 120, 100]) # Lower bound of blue upper_blue = np.array([130, 255, 255]) # Upper bound of blue
         lower_yellow = np.array([0, 120, 0]) # Lower bound of yellow
upper_yellow = np.array([30, 255, 255]) # Upper bound of yellow
         # Create red, blue, and yellow masks
red_mask = cv2.inRange(hsv, lower_red, upper_red)
blue_mask = cv2.inRange(hsv, lower_blue, upper_blue)
yellow_mask = cv2.inRange(hsv, lower_yellow, upper_yellow)
         # Combine the red, blue, and yellow masks
combined_mask = cv2.bitwise_or(red_mask, blue_mask)
combined_mask = cv2.bitwise_or(combined_mask, yellow_mask)
         # Noise removal (Morphological operations)
kernel = np.ones((3, 3), np.uint8)
combined_mask = cv2.morphologyEx(combined_mask, cv2.MORPH_OPEN, kernel)
combined_mask = cv2.morphologyEx(combined_mask, cv2.MORPH_CLOSE, kernel)
         # Find contours

contours, _ = cv2.findContours(combined_mask, cv2.RETR_EXTERNAL, cv2.CHAIN_APPROX_SIMPLE)
         detected_signs = []
                   if cv2.contourArea(contour) > 500:
    x, y, w, h = cv2.boundingRect(contour)
    aspect_ratio = float(w) / h
# Get color type (Red, Blue, or Yellow)
roi_hsv = cv2.cvtColor(roi, cv2.CDLOR_BGR2HSV)
red_pixels = cv2.countNonZero(cv2.inRange(roi_hsv, lower_red, upper_red))
blue_pixels = cv2.countNonZero(cv2.inRange(roi_hsv, lower_blue, upper_blue))
yellow_pixels = cv2.countNonZero(cv2.inRange(roi_hsv,
lower_yellow, upper_yellow))
                                   # Determine the color type
if red_pixels > max(blue_pixels, yellow_pixels):
    color_type = "Red"
elif blue_pixels > max(red_pixels, yellow_pixels):
    color_type = "Blue"
else:
                                             color_type = "Yellow"
                                   # Template matching
best_match = None
best_score = 0
                                    for sign_name, template in SIGN_TEMPLATES.items():
                                             score = match_template(roi, template)
if score > best_score and score > 0.55:
                                                     best_score = score
best_match = sign_name
                                   if best_match:
    # Map the sign name to a custom name
    new_sign_name = SIGN_NAME_MAPPING.get(best_match, best_match) # If no
                                             sign_info = f" new_sign_name) ( color_type ) "
detected_signs.append((x, y, w, h, sign_info))
         return detected_signs
```

Image 1.8 The detect_traffic_signs function

Functionality:

• Detect traffic signs in a single frame of the video.

Steps:

1. **Convert frame to HSV color space**: Use cv2.cvtColor to convert the frame from RGB to HSV, which simplifies color-based segmentation.

2. Define color ranges:

- a) Create ranges for red, blue, and yellow in HSV color space.
- b) These ranges help identify traffic signs based on their dominant colors.

3. Create masks:

- a) Use cv2.inRange to create binary masks for red, blue, and yellow areas.
- b) Combine the masks using cv2.bitwise_or to get a single mask for all traffic signs.

4. Noise removal:

a) Use morphological operations (cv2.morphologyEx) to clean up the mask by removing noise and filling gaps.

5. Find contours:

- a) Extract the shapes in the mask using cv2.findContours.
- b) Filter contours based on area (>500 pixels) and aspect ratio (close to square).

6. Template matching:

- a) For each ROI (Region of Interest) corresponding to a detected contour:
 - i. Compare it with all templates in SIGN_TEMPLATES using match template.
 - ii. Keep track of the best matching template if the score exceeds 0.55.

7. Color classification:

a) Check the dominant color (red, blue, or yellow) in the ROI by counting the number of pixels that match each color range.

8. Store results:

a) Append the detected sign's position and name to a list.

1.2.6 process video() function

```
. . .
def process_video(video_path):
   load_templates()
   cap = cv2.VideoCapture(video_path)
   frame_width = int(cap.get(cv2.CAP_PROP_FRAME_WIDTH))
   frame_height = int(cap.get(cv2.CAP_PROP_FRAME_HEIGHT))
   fps = int(cap.get(cv2.CAP_PROP_FPS))
   out = cv2.VideoWriter('output2.avi',
                         cv2.VideoWriter_fourcc(*'XVID'),
                         fps, (frame_width, frame_height))
   while cap.isOpened():
        ret, frame = cap.read()
       if not ret:
            break
        signs = detect_traffic_signs(frame)
        for (x, y, w, h, sign_name) in signs:
            cv2.rectangle(frame, (x, y), (x+w, y+h), (0, 255, 0), 2)
            cv2.putText(frame, sign_name, (x, y-10),
                       cv2.FONT_HERSHEY_SIMPLEX, 0.9, (0, 0, 255), 2)
        cv2.imshow("Processed Video", frame)
        out.write(frame)
        if cv2.waitKey(1) & 0xFF == ord('q'):
            break
    cap.release()
    out.release()
    cv2.destroyAllWindows()
```

Image 1.9 The detect traffic signs function

Functionality:

• Process the input video to detect traffic signs in each frame, annotate them, and save the result as a new video.

Steps:

1. **Load templates**: Call load_templates() to load traffic sign templates into memory.

2. Read input video:

- a) Use cv2.VideoCapture to load the video.
- b) Extract video properties such as frame width, height, and FPS.

3. Prepare output video:

a) Create a cv2.VideoWriter object to save the processed frames into a new video.

4. Process frames:

- a) For each frame:
 - i. Call detect traffic signs() to detect traffic signs.
 - ii. Draw bounding boxes and labels for each detected sign using cv2.rectangle and cv2.putText.
 - iii. Save the annotated frame to the output video.

5. Release resources:

a) Close the video input and output streams and destroy all OpenCV windows.

1.2.7 Directory Structure

```
# Create directory structure
if not os.path.exists('sign_templates'):
    os.makedirs('sign_templates')
```

Image 1.10 Create directory structure

Ensure that the sign templates folder exists to store traffic sign template images.

1.2.8 Example Usage

```
# Process video
process_video('video2.mp4')
```

Image 1.11 Process video

- Call the process video() function with the input video (video1 or video2).
- The script processes the video frame by frame, detects traffic signs, and saves the output as output2.avi.

CHAPTER 2. TASK RESULT

2.1 Task result of video 1:



Image 2.1 Output the frame at 5 seconds mark in video 1



Image 2.2 Output the frame at 7 seconds mark in video 1



Image 2.3 Output the frame at 18 seconds mark in video 1



Image 2.4 Output the frame at 25 seconds mark in video 1



Image 2.5 Output the frame at 36 seconds mark in video 1



Image 2.6 Output the frame at 42 seconds mark in video 1

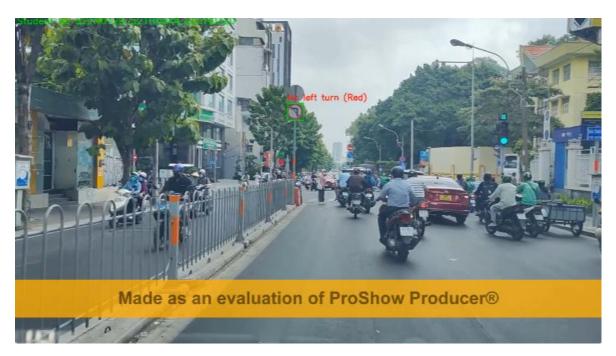


Image 2.7 Output the frame at 46 seconds mark in video 1



Image 2.8 Output the frame at 57 seconds mark in video 1

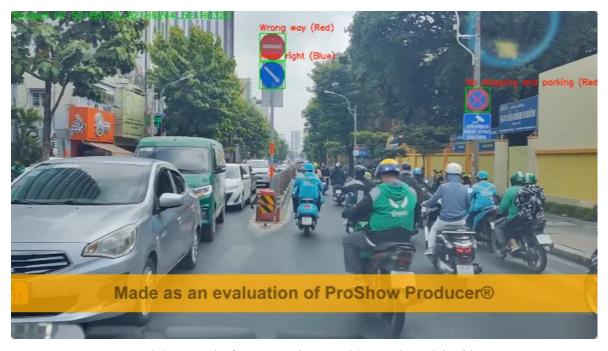


Image 2.9 Output the frame at 1 minutes and 9 seconds mark in video 1



Image 2.10 Output the frame at 1 minutes and 26 seconds mark in video 1



Image 2.11 Output the frame at 1 minutes and 46 seconds mark in video 1

2.2 Task result of video 2:



Image 2.12 Output the frame at 4 seconds mark in video 2



Image 2.13 Output the frame at 23 seconds mark in video 2



Image 2.14 Output the frame at 1 minutes and 4 seconds mark in video $2\,$

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