Number Plate Detection Using Parallel Processing

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Abstract

This mini-project demonstrates a number plate detection system using OpenCV and parallel processing techniques. The system captures video frames, detects number plates, and processes the images in parallel to improve performance.

1 Introduction

Number plate detection is a crucial task in many applications such as traffic monitoring and automated toll collection. This project utilizes OpenCV for image processing and EasyOCR for optical character recognition (OCR). To enhance the efficiency of the system, parallel processing is employed using Python's multiprocessing module.

2 System Overview

The system captures video frames from a webcam, detects number plates in the frames, and processes the images to extract the number plates. The processing is done in parallel to ensure real-time performance.

3 Implementation

3.1 Dependencies

The following libraries are required for this project:

- cv2 (OpenCV)
- multiprocessing (Python standard library)
- easyocr

• matplotlib

while True:

• joblib

3.2 Number Plate Detection

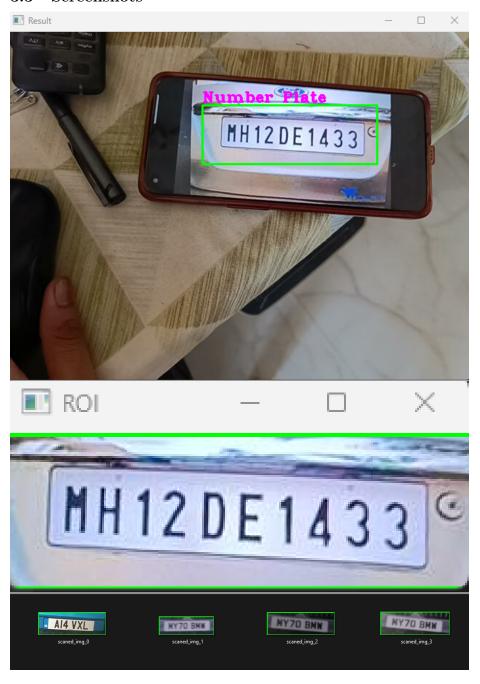
The number plate detection is implemented using OpenCV's Haar Cascade classifier. The video frames are captured from a webcam and processed in parallel.

Listing 1: Number Plate Detection with Parallel Processing import cv2 import multiprocessing as mp harcascade = "model/haarcascade_russian_plate_number.xml" cap = cv2. Video Capture (1)cap.set(3, 640) # widthcap.set(4, 480) # height $min_area = 500$ count = 0def detect_plates(frame_queue, result_queue): plate_cascade = cv2. Cascade Classifier (harcascade) while True: img = frame_queue.get() if img is None: break img_gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY) plates = plate_cascade.detectMultiScale(img_gray, 1.1, 4) for (x, y, w, h) in plates: area = w * hif area > min_area: cv2.rectangle(img, (x, y), (x + w, y + h), (0, 255, 0), 2) $cv2.putText(img, "Number-Plate", (x, y - 5), cv2.FONT.HERSHEY_COimg_roi = img[y: y + h, x: x + w]$ result_queue.put(img_roi) result_queue.put(img) def process_frame(frame_queue):

success, img = cap.read()

```
if not success:
            frame_queue.put(None)
        frame_queue.put(img)
def main():
    global count
    frame_queue = mp. Queue (maxsize=10)
    result_queue = mp. Queue ( maxsize=10)
   \# Start the plate detection process
    detection_process = mp. Process(target=detect_plates, args=(frame_queue, resu
    detection_process.start()
   \# Start capturing frames
    frame_capture_process = mp. Process(target=process_frame, args=(frame_queue,)
    frame_capture_process.start()
   while True:
        img = result_queue.get()
        if img is None:
            break
        cv2.imshow("Result", img)
        if cv2.waitKey(1) & 0xFF = ord('s'):
            img_roi = result_queue.get()
            cv2.imwrite("plates/scanned_img_" + str(count) + ".jpg", img_roi)
            cv2.rectangle(img, (0, 200), (640, 300), (0, 255, 0), cv2.FILLED)
            cv2.putText(img, "Plate-Saved", (150, 265), cv2.FONT_HERSHEY_COMPLEX
            cv2.imshow("Results", img)
            cv2.waitKey(500)
            count += 1
    frame_capture_process.join()
    detection_process.join()
    cap.release()
    cv2.destroyAllWindows()
if _-name_- = "_-main_-":
   main()
```

3.3 Screenshots



3.4 Text Recognition on Number Plates

To recognize the text on the detected number plates, EasyOCR is used. The OCR processing is also performed in parallel.

Listing 2: Text Recognition Using EasyOCR

```
!pip install easyocr
import matplotlib.pyplot as plt
import cv2
import easyocr
from IPython.display import Image
from joblib import Parallel, delayed
img_path = "scaned_img_0.jpg"
Image (img_path)
image = cv2.imread(img_path)
# Convert to grayscale
gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
# Apply bilateral filter to reduce noise while keeping edges sharp
filtered = cv2.bilateralFilter(gray, 11, 17, 17)
# Apply thresholding
_, thresh = cv2.threshold(filtered, 0, 255, cv2.THRESH.BINARY + cv2.THRESH.OTSU)
# Display the preprocessed image
plt.imshow(cv2.cvtColor(thresh, cv2.COLOR_BGR2RGB))
plt.title('Preprocessed - Image')
plt.show()
reader = easyorr.Reader(['en'])
# Define a function to process OCR on a single image
def process_ocr(img):
    # Read the text from the image
    output = reader.readtext(img)
    return output
# Perform OCR in parallel
output = Parallel(n_jobs=-1)(delayed(process_ocr)(img_path) for _ in range(1))
output
cord = output[-1][0]
```

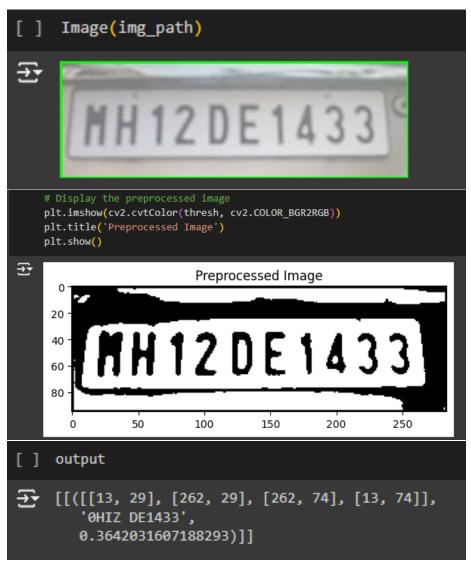
```
cord
cord , _ , _ = output[-1][0]

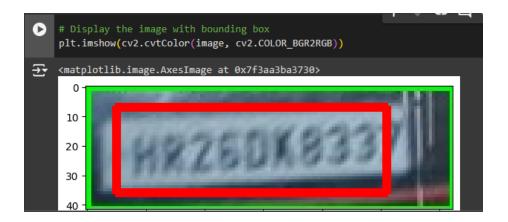
# Find the bounding box coordinates
x_values = [coord[0] for coord in cord]
y_values = [coord[1] for coord in cord]
x_min = min(x_values)
y_min = min(y_values)
x_max = max(x_values)
y_max = max(y_values)

# Display the bounding box on the image
image = cv2.imread(img_path)
cv2.rectangle(image, (x_min, y_min), (x_max, y_max), (0, 0, 255), 2)

# Display the image with bounding box
plt.imshow(cv2.cvtColor(image, cv2.COLOR_BGR2RGB))
```

3.5 Screenshots of Text Recognition





4 Parallel Processing Method

In this project, parallel processing is utilized to enhance the performance of number plate detection. The Python multiprocessing module is used to achieve this. The main idea is to separate the tasks of frame capturing and plate detection into different processes that run concurrently.

4.1 Why Parallel Processing?

In a sequential processing approach, the main program captures a frame from the video stream, processes it to detect number plates, and then displays the result. This sequence of operations can create a bottleneck, especially when real-time performance is required. By using parallel processing, we can overlap the operations of capturing frames and detecting plates, thus improving the throughput and responsiveness of the system.

4.2 Implementation Details

The implementation involves creating two separate processes:

- 1. **Frame Capturing Process:** Continuously captures frames from the webcam and puts them into a shared queue.
- 2. **Plate Detection Process:** Continuously reads frames from the shared queue, processes them to detect number plates, and puts the results into another shared queue.

Here's the code snippet showing the parallel processing setup:

```
Listing 3: Parallel Processing Setup
```

```
frame_queue = mp. Queue(maxsize=10)
result_queue = mp. Queue(maxsize=10)
```

```
# Start the plate detection process
detection_process = mp. Process(target=detect_plates, args=(frame_queue, result_q
detection_process.start()
# Start capturing frames
frame_capture_process = mp. Process(target=process_frame, args=(frame_queue,))
frame_capture_process.start()
# Main loop to display results
while True:
     img = result_queue.get()
     if img is None:
           break
     cv2.imshow("Result", img)
     if cv2.waitKey(1) & 0xFF = ord('s'):
           img_roi = result_queue.get()
            \begin{array}{l} {\rm cv2.imwrite}("\,{\rm plates/scanned\_img\_"} \, + \, {\bf str}({\rm count}) \, + \, ".jpg" \, , \, \, {\rm img\_roi}) \\ {\rm cv2.rectangle}({\rm img} \, , \, \, (0 \, , \, \, 200) \, , \, \, (640 \, , \, \, 300) \, , \, \, (0 \, , \, \, 255 \, , \, \, 0) \, , \, \, {\rm cv2.FILLED}) \\ \end{array} 
           cv2.putText(img, "Plate-Saved", (150, 265), cv2.FONT.HERSHEY.COMPLEX.SMA
           cv2.imshow("Results", img)
           cv2. waitKey (500)
           count += 1
frame_capture_process.join()
detection_process.join()
cap.release()
cv2.destroyAllWindows()
```

5 Comparison with Non-Parallel Method

5.1 Non-Parallel Method

In the non-parallel method, all operations (frame capturing, plate detection, and result display) are performed sequentially in a single process. Here is a simplified version of how the code would look without parallel processing:

Listing 4: Non-Parallel Method

```
cap = cv2.VideoCapture(1)
cap.set(3, 640)
cap.set(4, 480)
min_area = 500
count = 0
```

```
plate_cascade = cv2. CascadeClassifier("model/haarcascade_russian_plate_number.xn
while True:
    success, img = cap.read()
    if not success:
        break
    img_gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    plates = plate_cascade.detectMultiScale(img_gray, 1.1, 4)
    for (x, y, w, h) in plates:
        area = w * h
        if area > min_area:
            cv2.rectangle(img, (x, y), (x + w, y + h), (0, 255, 0), 2)
            {
m cv2.putText(img,\ "Number-Plate",\ (x,\ y-5),\ cv2.FONT\_HERSHEY\_COMPLECTION }
            img\_roi = img[y: y + h, x: x + w]
            cv2.imwrite("plates/scanned\_img\_" + \mathbf{str}(count) + ".jpg", img\_roi)
            count += 1
    cv2.imshow("Result", img)
    if cv2.waitKey(1) & 0xFF = ord('q'):
        break
cap.release()
```

5.2 Performance Comparison

cv2.destroyAllWindows()

- Sequential Processing: In the non-parallel approach, the frame rate can be significantly lower because each frame has to go through the complete process of capturing, detecting, and displaying sequentially. This can lead to delays and reduced real-time performance.
- Parallel Processing: By separating the frame capturing and plate detection into different processes, the system can handle more frames per second, leading to smoother and faster real-time detection. The overlapping of operations allows the system to utilize CPU resources more efficiently, reducing the idle time between operations.

5.3 Advantages of Parallel Processing

- Improved Throughput: More frames can be processed per second.
- Reduced Latency: Faster response time for real-time applications.
- Better Resource Utilization: Efficient use of multi-core processors.

5.4 Disadvantages of Parallel Processing

- Complexity: More complex code structure and debugging.
- Inter-Process Communication Overhead: Potential overhead due to data exchange between processes.

6 Conclusion

This project demonstrates the implementation of a number plate detection system using parallel processing to improve efficiency. By leveraging OpenCV for image processing and EasyOCR for text recognition, the system can effectively detect and recognize number plates in real-time. The parallel processing approach significantly enhances performance compared to a sequential method, making it suitable for applications requiring high throughput and low latency.