Car Counting Using Morphological Operations

May 11, 2025

Abstract

This document demonstrates a car counting system using morphological image processing techniques. The pipeline includes image preprocessing, binarization, dilation, and connected component analysis to count vehicles in a traffic scene.

1 Libraries Used

- numpy: For numerical operations and array manipulation
- cv2 (OpenCV): For image processing and computer vision operations
- matplotlib.pyplot: For image visualization and plotting

2 Step-by-Step Process

2.1 Step 1: Import Libraries

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

2.2 Step 2: Download Image

Download the sample traffic image:

```
!wget https://raw.githubusercontent.com/AsadiAhmad/Car-Counter-
Morphology/main/Pictures/Cars.jpg -O Cars.jpg
```

2.3 Step 3: Load and Display Image

Load the image in grayscale and display:

```
cars = cv.imread("Cars.jpg", cv.IMREAD_GRAYSCALE)
plt.imshow(cars, cmap="gray")
```

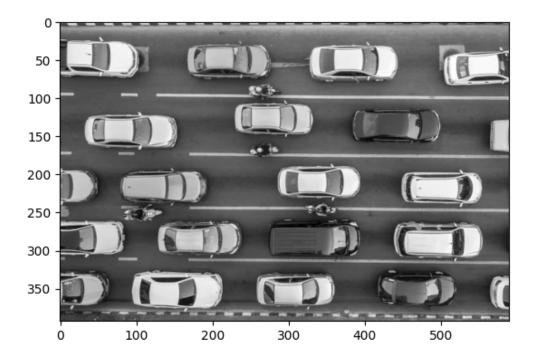


Figure 1: Original traffic image

2.4 Step 4: Crop Image

Remove unnecessary borders from the image:

```
height, width = cars.shape
cropped_cars = cars[15:height-15, :]
plt.imshow(cropped_cars, cmap="gray")
```

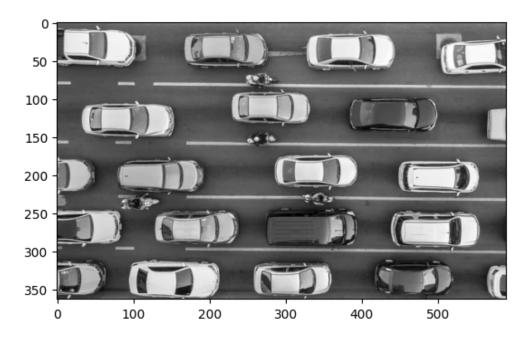


Figure 2: Cropped image removing top/bottom borders

2.5 Step 5: Blur Image

Apply Gaussian blur to reduce noise:

```
blurred_cars = cv.GaussianBlur(cropped_cars, (51, 51), 0)
plt.imshow(blurred_cars, cmap="gray")
```

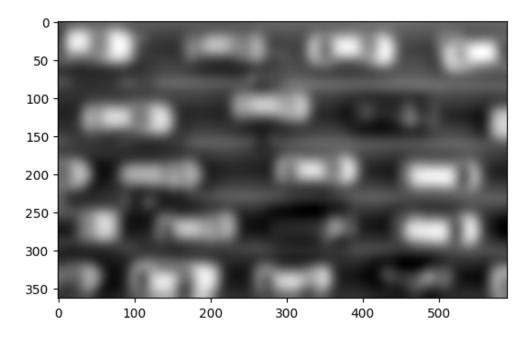


Figure 3: Image after Gaussian blur $(51 \times 51 \text{ kernel})$

2.6 Step 6: Binarization

Convert to binary image for morphological processing:

```
bin_cars = np.where(blurred_cars > 127, 255, 0)
plt.imshow(bin_cars, cmap="gray")
scaled_cars = bin_cars.astype(np.uint8)
```

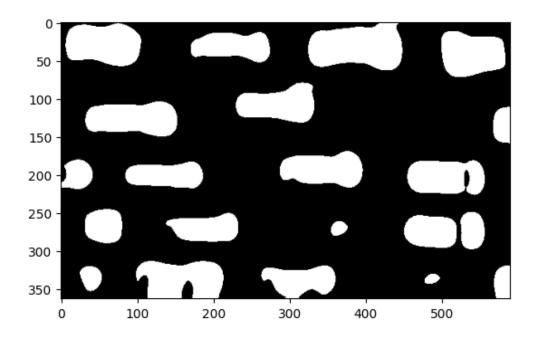


Figure 4: Binarized image (threshold = 127)

2.7 Step 7: Dilation for Connecting Car Parts

Connect car components using dilation:

```
kernel = np.ones((3, 3), np.uint8)
car_dilate = cv.dilate(scaled_cars, kernel, iterations=6)
plt.imshow(car_dilate, cmap="gray")
```

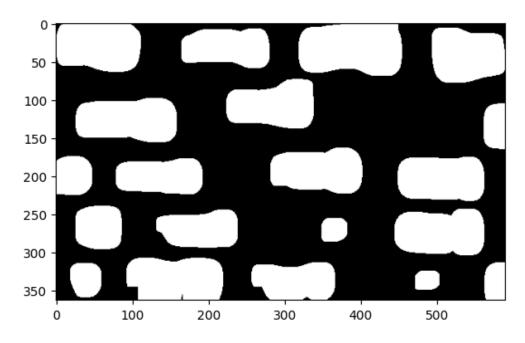


Figure 5: Image after dilation (6 iterations)

2.8 Step 8: Counting Cars

Count connected components:

```
num_labels, labels = cv.connectedComponents(car_dilate)
car_count = num_labels - 1  # Subtract background
print(car_count)
```

Output:

20

2.9 Step 9: Calculating Accuracy

Calculate counting accuracy:

```
percent = (car_count/21)*100
print(f'{percent:.2f}%')
```

Output:

95.24%

3 Technical Explanations

3.1 Morphological Processing

• Cropping: Removes irrelevant image regions

• Gaussian Blur: Reduces noise while preserving edges

• Binarization: Simplifies image for morphological operations

• Dilation: Connects nearby car parts into single components

• Connected Components: Identifies and counts individual cars

3.2 Performance Notes

• 95.24% accuracy achieved on sample image

• Kernel size and iterations affect counting results

• Works best with clear separation between vehicles



https://github.com/AsadiAhmad/Car-Counter-Morphology