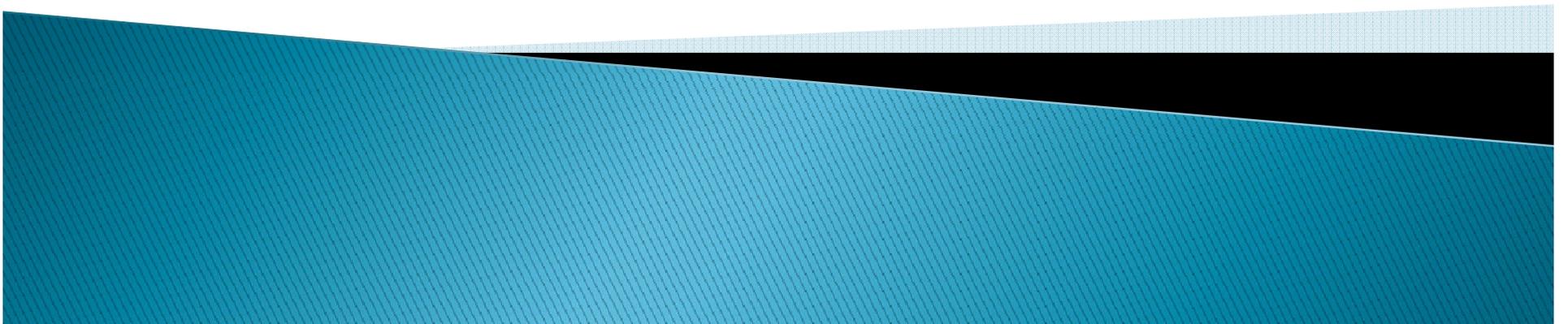


License Plate Recognition

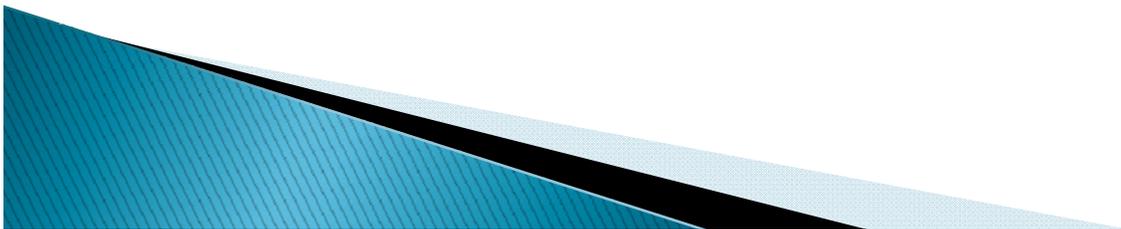
Brian Forbis



Commercial and Government Uses

A license plate locator can be used for:

- ▶ CCTV cameras at intersections and parking lots
- ▶ Toll roads to bill users of the road
- ▶ Dashboard mounted cameras on police cars to identify drivers
- ▶ Plate blurring on database images to preserve privacy



The Algorithm

- ▶ Image Pre-Processing
- ▶ Horizontal Line Scanning
- ▶ Platerun Extending/Merging
- ▶ Conditional Classifiers
- ▶ Extraction of Features
- ▶ Neural Network Classifiers

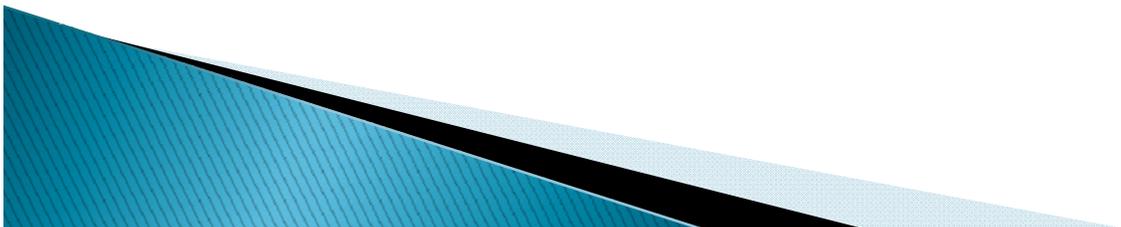


Image Pre-Processing

- ▶ Use a horizontal Sobel filter to filter out horizontal lines.

$$\begin{matrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{matrix}$$

- ▶ Use Otsu's method to get a binary thresholding of the image.
- ▶ Use a morphological opening with a disc shape of radius 1 structuring element

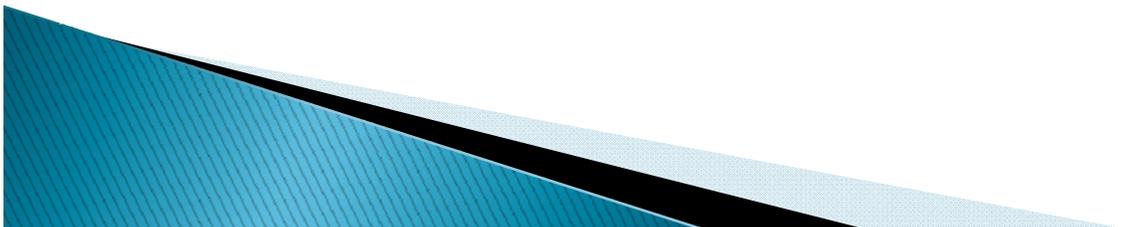


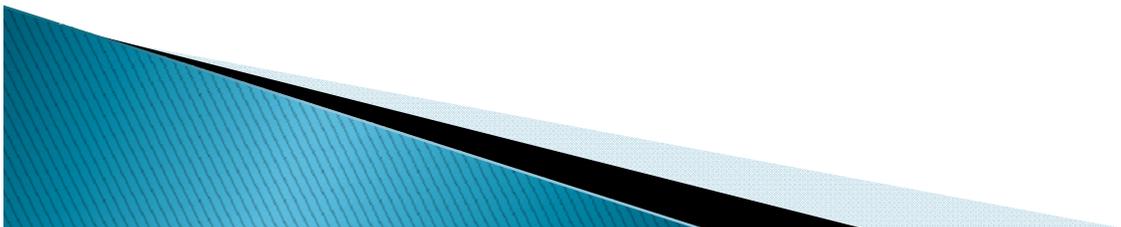
Image Pre-Processing cont.



Original
Image

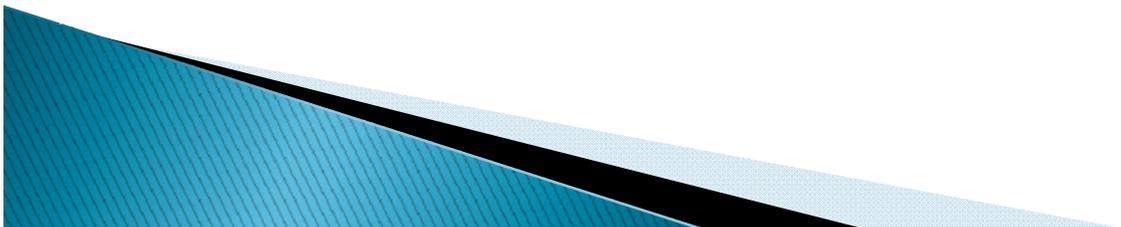


Binary Image after: filter,
threshold, imopen



Horizontal Line Scanning

- ▶ Algorithm 1: Horizontal Line Scan for Columns
- ▶ Data: row_y of BW
- ▶ Result: row of numbers depicting size of plateruns
- ▶ Initialize a vector Acc [width of BW]
- ▶ Initialize a matrix [size of BW]
- ▶ foreach row_{bw}
- ▶ $Acc = row_{bw}$
- ▶ foreach col_{row}
- ▶ if $Acc[col] == 1$, *then*
- ▶ $Acc[col] = Accs[row-1, col] + 1$;
- ▶ end
- ▶ $Accs[row, :] = Acc$;
- ▶ end

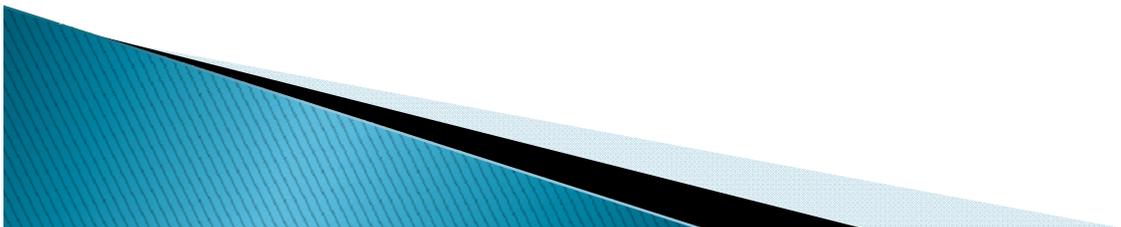


Horizontal Line Scanning, cont.

- ▶ The creation of Accs should look like:

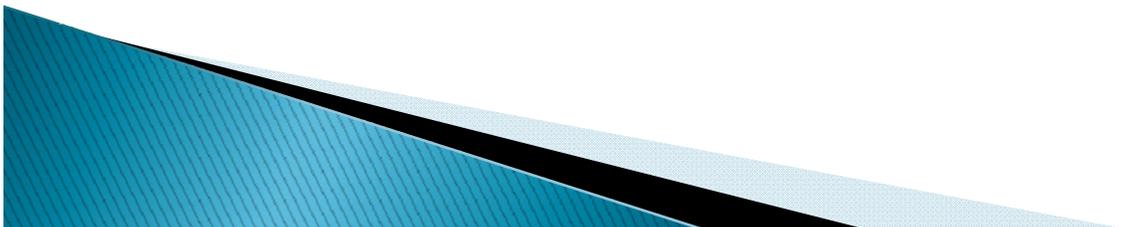
| Binary Image | | | | |
|--------------|---|---|---|---|
| 1 | 0 | 0 | 1 | 0 |
| 0 | 0 | 0 | 1 | 1 |
| 1 | 1 | 1 | 1 | 1 |
| 1 | 0 | 1 | 0 | 1 |
| 1 | 0 | 1 | 1 | 0 |
| (a) | | | | |

| Accs | | | | |
|------|---|---|---|---|
| 1 | 0 | 0 | 1 | 0 |
| 0 | 0 | 0 | 2 | 1 |
| 1 | 1 | 1 | 3 | 2 |
| 2 | 0 | 2 | 0 | 3 |
| 3 | 0 | 3 | 1 | 0 |
| (b) | | | | |



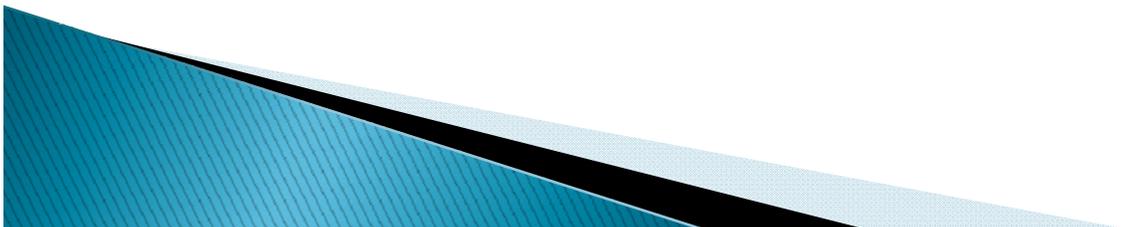
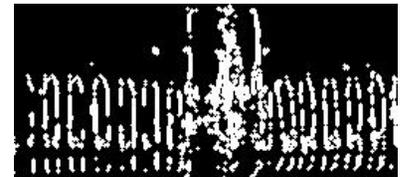
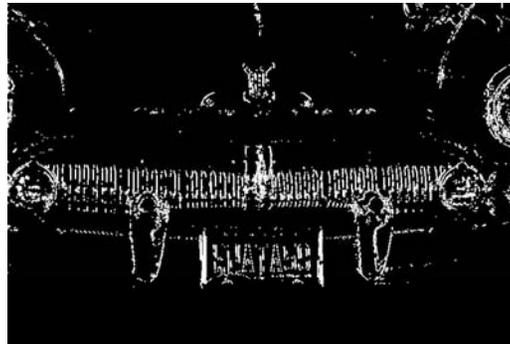
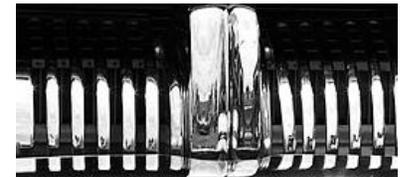
Platerun Extending/Merging

- ▶ Algorithm 2: Extend Plateruns
- ▶ Scan Accs for Plateruns and add to matrix $plateruns [r_1, c_1, r_2, c_2]$
- ▶ foreach $pr_{plateruns}$
 - while exists a platerun within λ_x
 - Scan horizontal and merge plateruns within λ_x
- ▶ end
- ▶ end



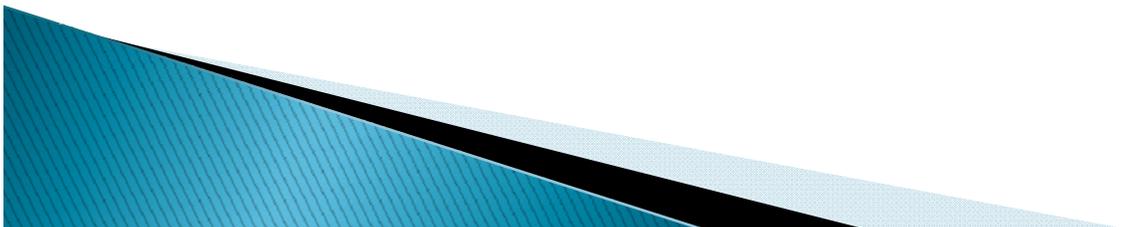
Example of Failed Platerun

The front grill in this image is picked up as a platerun. This is due to the fact that it consists of many vertical lines in that are close together.



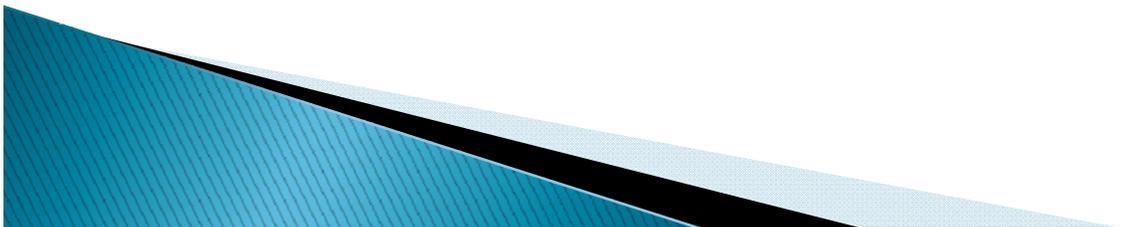
Conditional Classifiers

- ▶ Plateruns are thrown out if they pass any of the following classifiers:
- ▶ $w < 40$
 $h < 20$
 $w > 2/5$ width of full image
 $h > 1/4$ height of full image
 $h/w > 0.55$
 $h/w < 0.22$



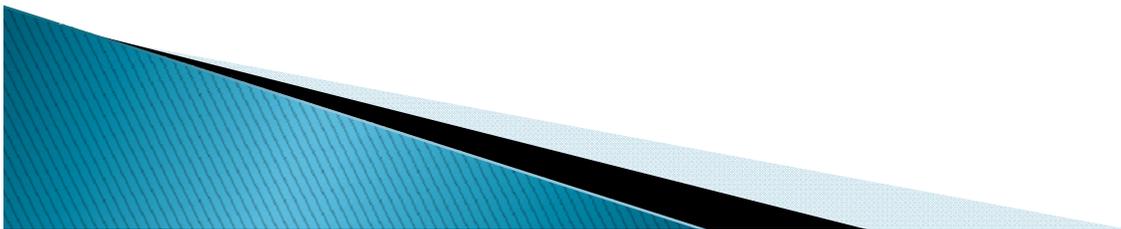
Feature Extraction

- ▶ Features are extracted from plateruns to create 80x1 feature vectors. The extracted features are:
 - ▶ Mean
 - ▶ Standard deviation
 - ▶ Skewness
 - ▶ 18-bin edge histogram
 - ▶ 9-D texture feature



Neural Network

- ▶ MATLAB's built-in neural network is used to classify the plate candidates.
- ▶ The neural network was trained with 232 extracted plates and 232 extracted nonplates using the normalized 80×1 feature vectors.



Results

- ▶ The results from running the program on 66 testing images are as follows:

| | |
|-----------------|-------|
| True Positive | 31 |
| False Positive | 19 |
| True Negatives | 44099 |
| False Negatives | 35 |
| Accuracy | 0.99 |
| Precision | 0.62 |

$$\text{accuracy} = \frac{\text{number of true positives} + \text{number of true negatives}}{\text{numbers of true positives} + \text{false positives} + \text{false negatives} + \text{true negatives}}$$

$$\text{precision} = \frac{\text{number of true positives}}{\text{number of true positives} + \text{false positives}}$$



Conclusion

- ▶ The high accuracy of this algorithm shows that with some modification, it could be used commercially. Speed can be improved by converting the code to a different language base such as C++.

