

## Efficiency and Accuracy Improvements to the Viola-Jones Object Detection Algorithm Using ML Methods

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#### Presentation Overview

- Introduce the face detection problem and its importance
- Overview of the original (2001) Viola-Jones algorithm
  - Outline the operation
  - Discuss the original results
- Provide examples of Viola-Jones in applications outside of face detection
- Highlight improvements made to the original algorithm
  - Redundancy reduction
  - Evolutionary pre-selection of features
  - Application of composite features
- Compare Viola-Jones performance to that of other modern algorithms
- Conclusion and wrap up with Q/A

#### Face Detection

- The first step of all other facial analysis algorithms
  - Face recognition, facial feature recognition, face scanning
- Recognition has several applications
  - Preventing retail crime, smart advertising, finding missing persons, aiding forensic investigation, protecting schools from potential threats, diagnosis of various diseases
- Easy for humans, not so easy for computers



Figure 1: Facial Recognition Graphic [2]





- Proposed in 2001 as an object detection algorithm, mainly proposed for use as a face detection algorithm
- Looks at regions of pixels and compares the sum of their values to find edges and other defining features
- 15 times greater frame rate than competing algorithms in 2001
- Four main components of the algorithm
  - Haar-like rectangular features
  - Concept of the integral image
  - Boosting with AdaBoost
  - Attentional cascade of classifiers



Figure 2: Images from Viola-Jones Paper, 2001 [3]

#### Haar-like Rectangular Features

#### The Concept of the Integral Image



Figure 3: Examples of Haar-like Rectangular Features



Figure 4: Example of Computing the Integral Image

#### Boosting with AdaBoost

#### The Attentional Cascade

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Further

**Processing** 



Figure 6: Degenerate Decision Tree (Cascaded Classifiers) [3]

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Figure 5: Various Rectangular Features [5]

#### Results of Viola-Jones, 2001

| False detections     |       |       |       |       |         |        |       |
|----------------------|-------|-------|-------|-------|---------|--------|-------|
| Detector             | 10    | 31    | 50    | 65    | 78      | 95     | 167   |
| Viola-Jones          | 76.1% | 88.4% | 91.4% | 92.0% | 92.1%   | 92.9%  | 93.9% |
| Viola-Jones (voting) | 81.1% | 89.7% | 92.1% | 93.1% | 93.1%   | 93.2 % | 93.7% |
| Rowley-Baluja-Kanade | 83.2% | 86.0% | -     | -     | -       | 89.2%  | 90.1% |
| Schneiderman-Kanade  | -     | -     | -     | 94.4% | -       | -      | -     |
| Roth-Yang-Ahuja      | -     | -     | -     | -     | (94.8%) | -      | -     |

Figure 7: Table Showing Detection Rates of Original Viola-Jones Algorithm Compared Against Other Algorithms of the Time [3]

#### Applications Outside of Face Detection

- Detection and tracking of locomotive activity of animals in wildlife video
- Emotion recognition to determine success in the learning environment
- Drowsiness detection to improve braincomputer interfaces, prevent auto accidents
- Vehicle counting system for traffic monitoring and surveillance
- Hand gesture recognition



Figure 8: Face Detection on Lions Using Viola-Jones [8]



# Improvements to the Algorithm



#### Redundancy Reduction

- Multiple similarly sized, similarly located windows detect the same face
- Accuracy, precision, and recall of the algorithm are improved
- True positive, true negative, false positive, false negative rates improved

|    | Comparison     |                    |                                   |  |  |  |
|----|----------------|--------------------|-----------------------------------|--|--|--|
| No | Test           | Proposed<br>Method | Traditional<br>Viola-Jones<br>[2] |  |  |  |
| 1  | True Positive  | 740                | 520                               |  |  |  |
| 2  | True Negative  | 25                 | 0                                 |  |  |  |
| 3  | False Negative | 100                | 155                               |  |  |  |
| 4  | False Positive | 35                 | 225                               |  |  |  |
| 5  | Accuracy       | 85%                | 77%                               |  |  |  |
| 6  | Precision      | 95%                | 71%                               |  |  |  |
| 7  | Recall         | 88%                | 81%                               |  |  |  |



Figure 9: Table of Reduced Redundancy Results [6]

Figure 10: Example Image Showing Redundancy (left) and Removal of Redundancy (right) [6]

#### Feature Pre-Selection

- Artificial evolutionary process to pre-select features for the classifiers, reduce training time and overfitting
- Incremental size changes for features, duplicates are skipped so that only unique classifiers are formed
- Only aiming for cascades with fewer stages than Viola-Jones
- Obtain 5 cascades with better performance, less stages, less features



Figure 11: Incremental Feature Size in Evolutionary Process [5]

| Cascade                    | Evaluated | Recall              | Precision     | False Pos. Rate     | Stages           | Features   |
|----------------------------|-----------|---------------------|---------------|---------------------|------------------|------------|
| Control                    | 1         | 0.9398              | 0.0925        | 0.3931              | 12               | 68         |
| Evolved                    | 120       | 0.9297±0.0022       | 0.1044±0.0033 | 0.3867±0.0141       | 11.33±0.07       | 76.46±1.05 |
| + Perf.                    | 26        | $0.9487 \pm 0.0011$ | 0.1054±0.0027 | $0.3493 \pm 0.0090$ | 11.35±0.15       | 75.46±1.79 |
| + Perf., - Stages          | 13        | 0.9501±0.0015       | 0.1019±0.0030 | $0.3605 \pm 0.0097$ | $10.85 \pm 0.19$ | 71.54±2.32 |
| + Perf., - Stages, - Feat. | 5         | 0.9493±0.0023       | 0.0988±0.0003 | $0.3692 \pm 0.0005$ | 10.2±0.2         | 64±0       |

Figure 12: Table of Results When Features are Pre-Selected [5]

### Application of Composite Features

- Alternative weak learner to the simple rectangular features
- Drastically decreases the number of false positives detected
- Slight decrease in frame rate results from more complex features

| rable 5 comprehensive comparison of the two |                                 |  |   |  |  |
|---|---------------------------------|--|---|--|--|
| algorithms                                  |                                 |  |   |  |  |
| Total<br>face<br>number                     | Total<br>missing<br>count       | Total<br>error<br>count                      | Total<br>missing<br>rate  | Gross<br>error<br>rate   |  |
| 1372  | 189                             | 21   | 0.138   | 0.015  |  |
| 1372  | 75                              | 1  | 0.055   | 0.0007   |  |
|   | Total<br>face<br>number<br>1372 | algorithTotalfacemissingnumbercount189137275 | algorithms   Total Total   Total Total   face missing   number count   189 21   1372 75 1 | algorithmsTotalTotalTotalTotalTotalTotalfacemissingerrornumbercountcount189210.13813727510.055 |  |

Table 3 Comprehensive comparison of the two

Figure 13: Table Comparing Results with and without Composite Features [7]



Figure 14: Simple Features vs. Composite Features [9]



## **Comparison to other Algorithms**



### Viola-Jones Compared with Other Algorithms

- Other algorithms have been written in recent years using:
  - Neural networks
  - SMQT features
  - Support Vector Machines
- Neural network-based face detectors outperform Viola-Jones in terms of detection rates
- Viola-Jones outperforms NNbased detectors in terms of time complexity and memory consumption



SMQT – Successive Mean Quantization Transform

Figure 15: SMQT Feature Examples [1]

#### Viola-Jones Compared with Other Algorithms

| Detector            | Time   |      | Memory           |  |
|---------------------|--------|------|------------------|--|
|                     | GFLOPS | FPS  | consumption (GB) |  |
| Viola-Jones [18]    | 0.6    | 60.0 | 0.1              |  |
| HeadHunter DPM [20] | 5.0    | 1    | 2.0              |  |
| SSD[6]              | 45.8   | 13.3 | 0.7              |  |
| Faster R-CNN [5]    | 223.9  | 5.8  | 2.1              |  |
| R-FCN 50 [3]        | 132.1  | 6.0  | 2.4              |  |
| R-FCN 101 [3]       | 186.6  | 4.7  | 3.1              |  |
| PVANET [13]         | 40.1   | 9.0  | 2.6              |  |
| Local RCNN [19]     | 1206.8 | 0.5  | 2.1              |  |
| Yolo 9000 [16]      | 34.90  | 19.2 | 2.1              |  |

Figure 16: Table Showing Superior Time Complexity, Memory Consumption of Viola-Jones Compared to Competing Algorithms [4]

## Thank you! Questions?

#### References

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