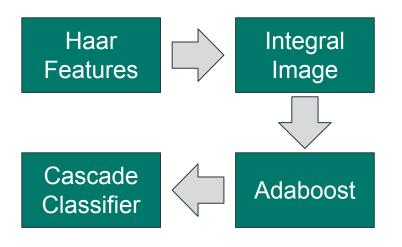
# REAL TIME FACE DETECTION WITH HAAR CASCADE ROBUST TO SKIN COLOR AND VARIED ILLUMINATION

William Chandra (13215052) Hansen (13214077)

Teddy Hadi Utama (13215086)

### Viola-Jones Framework: Haar Cascade

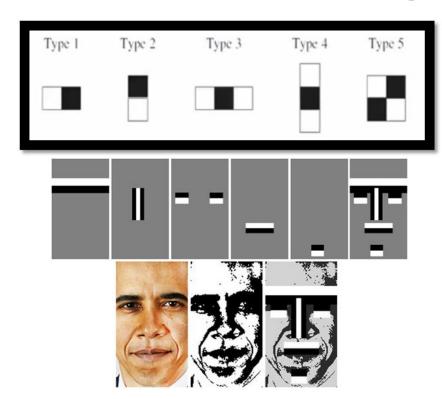
#### Background Theory - Viola Jones Algorithm

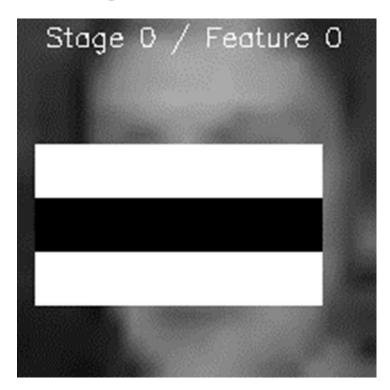


Viola Jones Algorithm **advantages** on **face detection**:

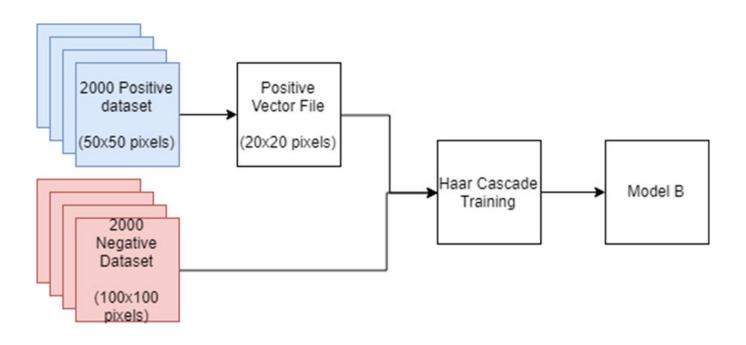
- 1. Fast feature computation
- 2. Enable feature scaling
- 3. Fixed computational cost
- 4. High detection rate and real time applicable

#### Haar Features & Integral Image





#### Training Custom Viola-Jones Model



## Experiment: Check the speed and accuracy of trained model

10 Multiple Faces (Total 27)

30 Single Face

39 No Face







#### Statistical Test

#### "MODEL A"

Default model from internet: haarcascade\_frontalface\_default.xml[ref]

- · Windows resolution: 24x24
  - · Number of stages: 25
- Max Number of Features/stages: 190
  - Number dataset: ~10,000 images

#### "MODEL B"

Trained Model

- Windows resolution: 20x20
  - Number of stages: 15
- Max Number of Features/stages: 20
  - Number dataset: ~2,000 images

- False Positive : Number of non-face misclassified as face
- False Negative : Number of face misclassified as non-face
- Execution Program : Duration taken to process

Here we compare two Haar Cascade model:

- 1. **Model A**: Default model from Internet
- Model B: Trained Model

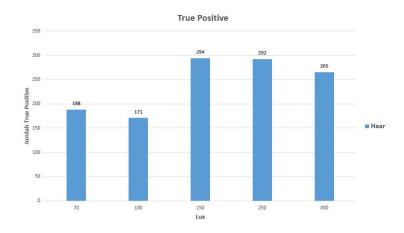
#### 640x480 Pixels Result

Photos type	Model A			Model B		
	False Positive	False Negative	Process duration(s)	False Positive	False Negative	Process Duration(s)
Single face(30)	8	4	4.056	8	10	1.52
Multiple faces(27)	6	3	1.98	3	3	0.9776
No faces(39)	7		4.506	22	12	1.67

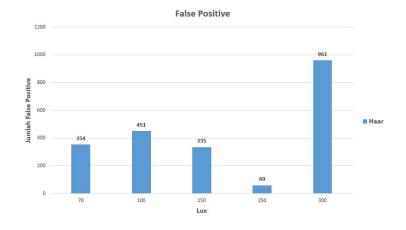
- Model B yields <u>2,5 times faster detection</u> <u>process</u> than Model A
- Model B has a lot worse False Positive on negative test set than Model A
- Model B has more false negative on single pictures than Model A

Note: <u>Trained model B</u> has much <u>fast detection</u> and needed <u>to reduced</u> in <u>false positive</u>

#### Implementation on Larger Dataset

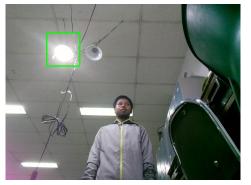


- Datasets are taken with varieties of brightness: 70 lux, 100 Lux, 150 Lux, 250 Lux, and 300 Lux
- 2250 Sample Images



 Measured the number of True Positive and False Positive with the varying brightness

#### False Positive





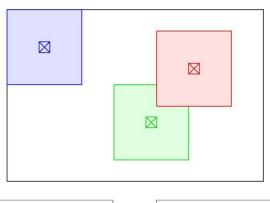


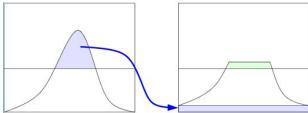
- Illumination problem
- Differentiating the **color** between human skin and other object

Therefore needed an algorithm that able to solve illumination problem and skin detection

## Haar Cascade with CLAHE as Preprocessing Algorithm

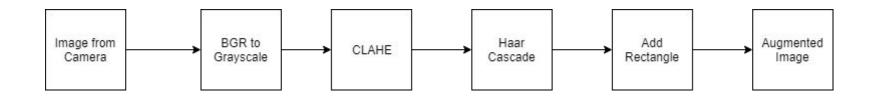
#### Background Theory - CLAHE





- Transforming each pixel with a transformation function derived from a neighbourhood region
- Each pixel is transformed based on the histogram of a square surrounding the pixel
- The transformation function is proportional to the cumulative distribution function (CDF) of pixel values in the neighbourhood
- CLAHE limits the amplification by clipping the histogram at a predefined value before computing the CDF

## Haar Cascade Diagram with Preprocessing using CLAHE



CLAHE do **preprocessing** for every images before acting as input for Haar Cascade Algorithm

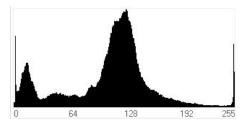
#### CLAHE Result on Grayscale Image

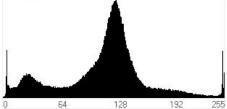
**Original Image** 



**CLAHE'd Image** 



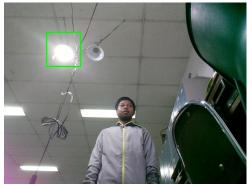




- Better image contrast
- Segmentation in image become clearer
- Reduce the wide illumination variation on the image

#### Face Detection Result

Haar Cascade

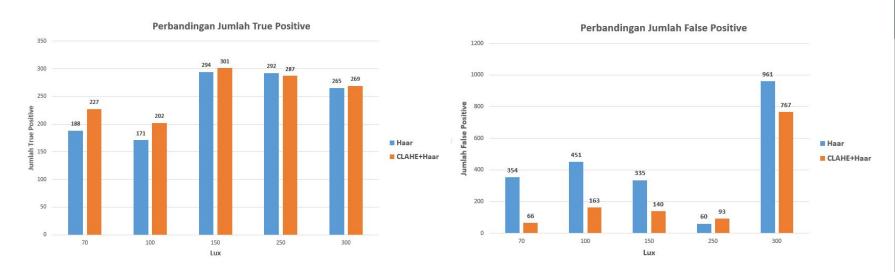


Haar Cascade + CLAHE



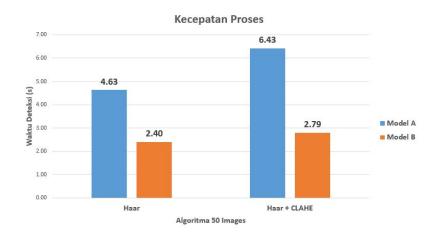
- Lamp no more detected as face (reduce false positive)
- Face is detected and increasing the true positive
- Illumination is no more a problem for the system

#### Statistical Evaluation



- Overall True Positive is increased, especially on the low brightness situation (70 lux and 100 lux)
- Overall False Positive decreased significantly

#### Algorithm Time Performance



- CLAHE didn't affect significantly on the overall system performance with 0.39 s difference in processing 50 images using model B
- Still applicable for real time system

#### Things to Improve

Haar Cascade

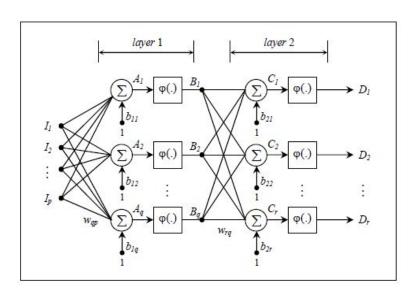
Haar Cascade + CLAHE



- CLAHE able to reduce the
   illumination effect on the images
   but sometimes produce
   enhancement on objects that
   recognized as face by Haar
   Cascades Algorithm
- Algorithm that able to
   differentiate human skin color
   and non human object is needed

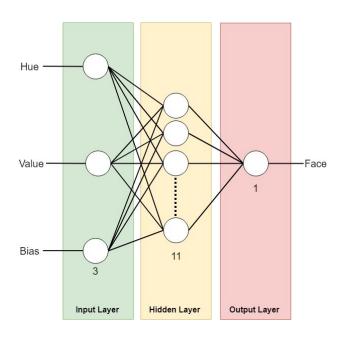
Haar Cascade Post
Processing Neural Network
for Face Skin Detection
with CLAHE as
Preprocessing Algorithm

#### Artificial Neural Network



- System inspired by biological neural network
- System learn the task by consider example
- Signal travel from input layer,
   hidden layer and output layer

#### Skin Detection (Neural Network)



#### Input parameter Skin Detection:

- Hue
- Value

#### Output parameter Skin Detection:

Face/No-Face

### **Back Propagation**

Epoch	Error propagation	
100	246.15046	
200	85.17840	
300	85.13202	
400	84.82097	
500	84.77503	
False Negative	269/2464	

Output	Weight	
Hidden Neuron	Face	
1	-0.56987	
2	0.286753	
3	-2.23135	
4	-0.0465	
5	0.464328	
6	1.576555	
7	-0.24542	
8	-1.52262	
9	-1.08785	
10	1.165909	
11	-0.42567	

Input Weight								
Hidden Neuron	Hue	Value	Bias					
1	-1.37757	1.019475	0.092551					
2	1.657802	0.85319	0.066129					
3	-9.29431	-0.48865	-0.21385					
4	-0.70679	0.062445	-0.20377					
5	0.205661	0.095307	-0.11653					
6	0.853825	0.785654	0.067265					
7	-2.90769	-0.69242	0.108759					
8	12.86952	7.257907	0.188112					
9	0.836774	1.457793	0.16153					
10	-0.93046	-0.17844	0.008187					
11	0.772516	0.178698	0.049118					

### Fixing Image Using Neural Network







Haar Cascade

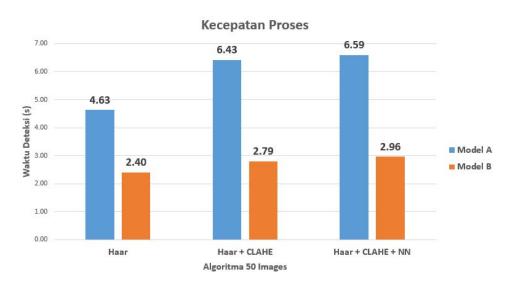
Haar Cascade + CLAHE

Haar Cascade + CLAHE + NN

#### Comparison: Algorithm False Positive



#### Comparison: Algorithm Performance Time



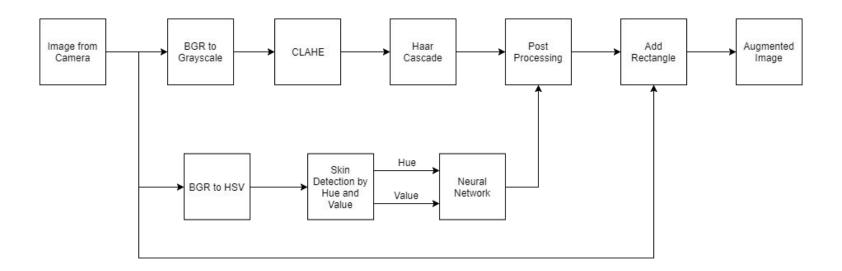
#### On model B:

Haar + CLAHE + NN only **0.56s slower** than Haar Cascade alone in processing 50 images.

Thus still applicable for real time face detection

### Real Time Implementation

### System Block Diagram



#### Live Detection



- Laptop
- Camera
- Python with OpenCV library

## DEMO

## Questions?

### THANK YOU