

Design and implementation of a face detection and location technique to crop faces from human images DB

Michal Szabo
Aleix Granero Mol

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- AdaBoost
- Cascade

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Introduction

Images database



Detect and
crop faces



Cropped faces database



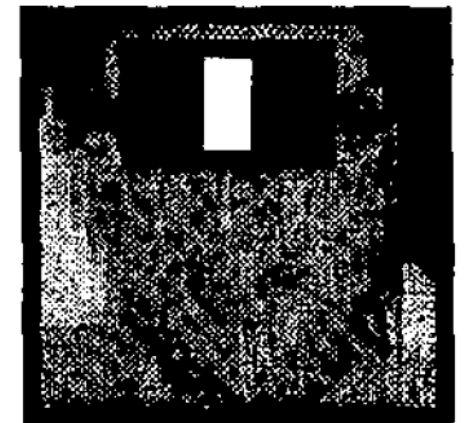
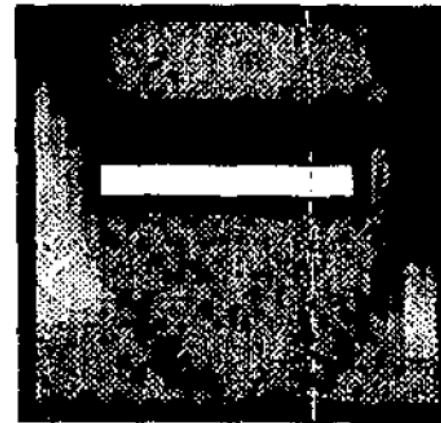
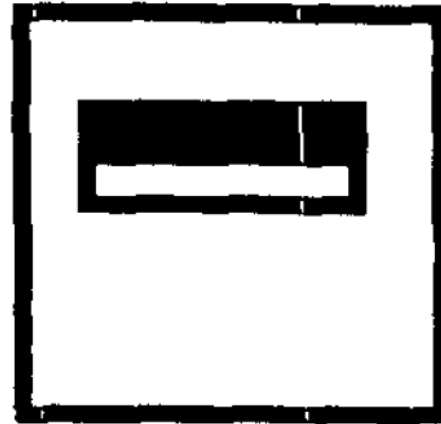
Viola-Jones Algorithm

- The Viola-Jones object detection framework is an object detection framework which was proposed in 2001 by Paul Viola and Michael Jones.



Haar Features

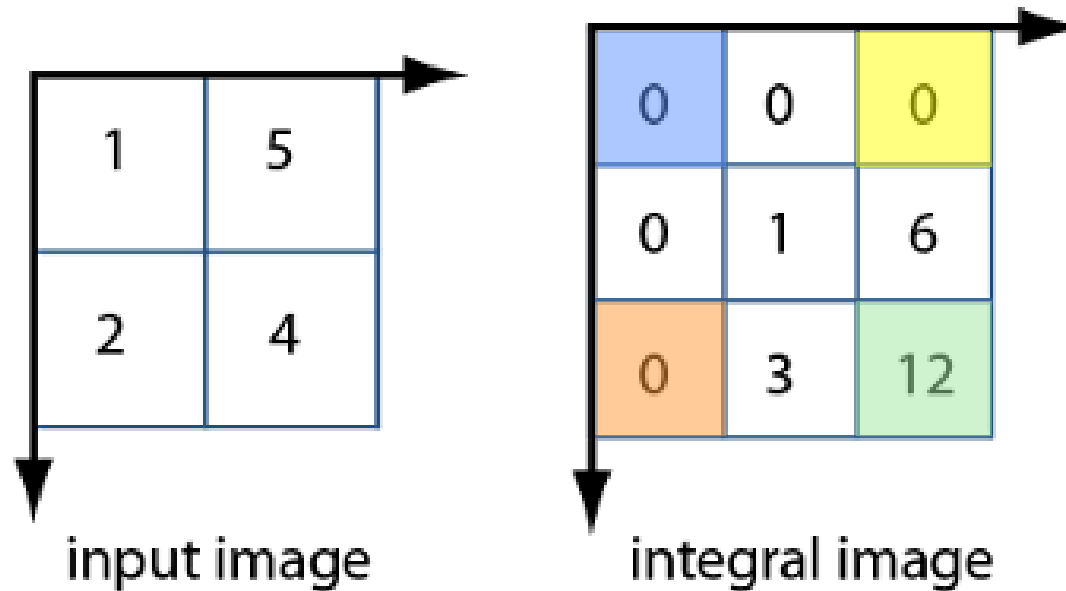
- Difference of the sum of pixels of areas inside the rectangle



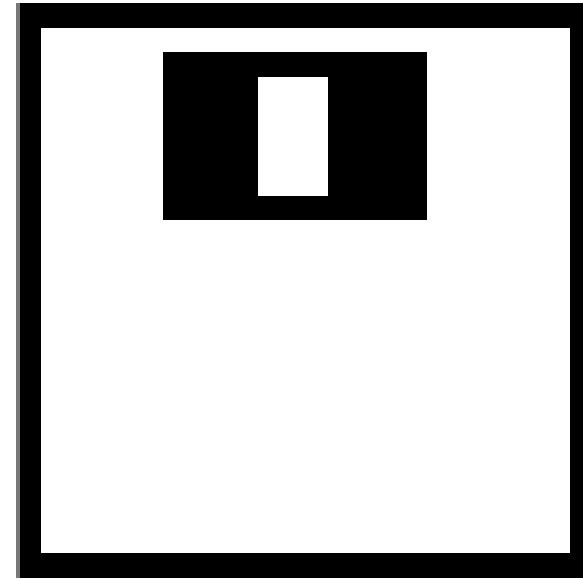
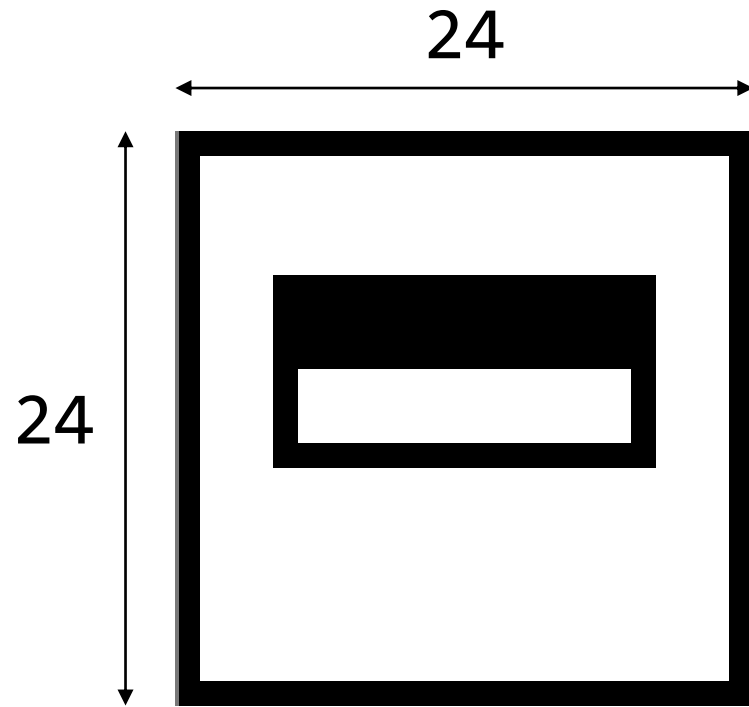
Integral Image

- Represents the cumulative sum of a corresponding input pixel with all pixels above and left of the input pixel

$$= d - c - b - a = 12$$

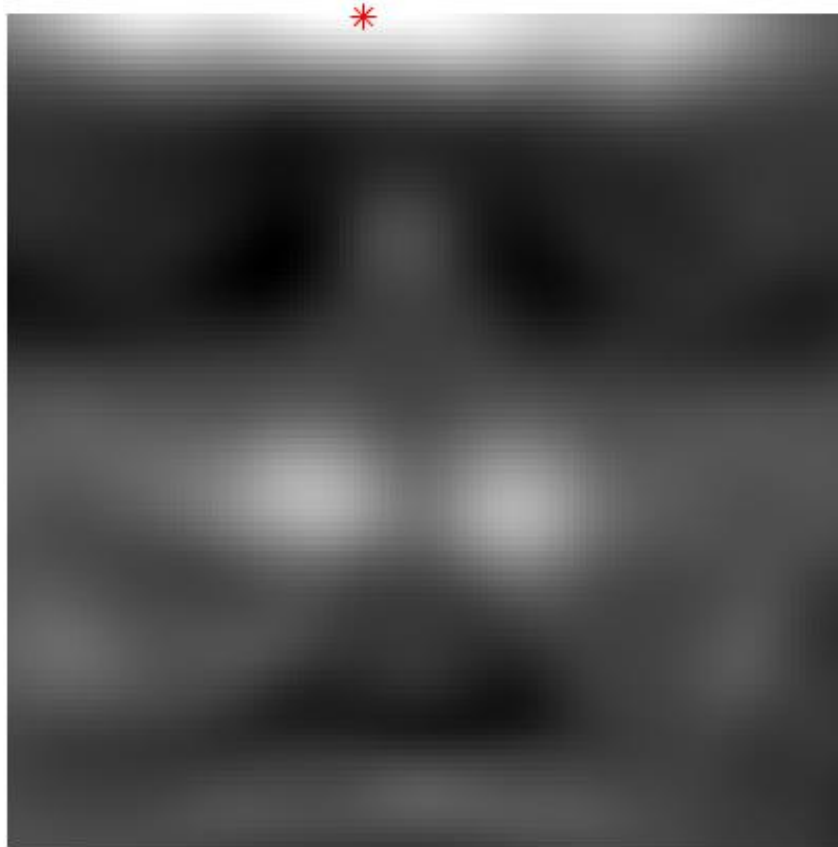


First Results

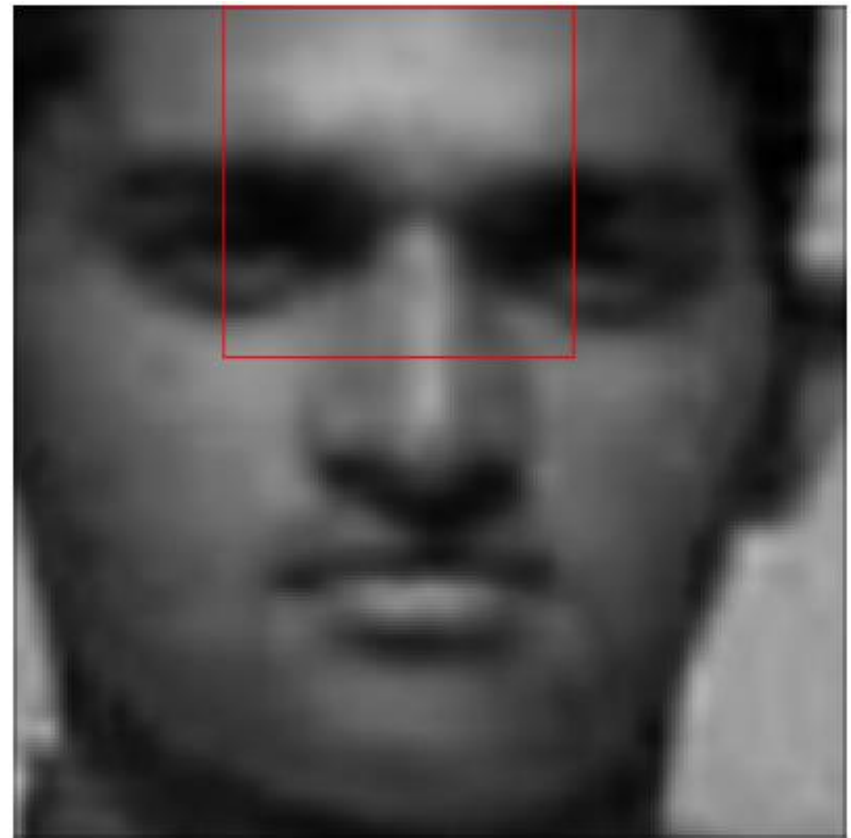


First Results

Intensity Map

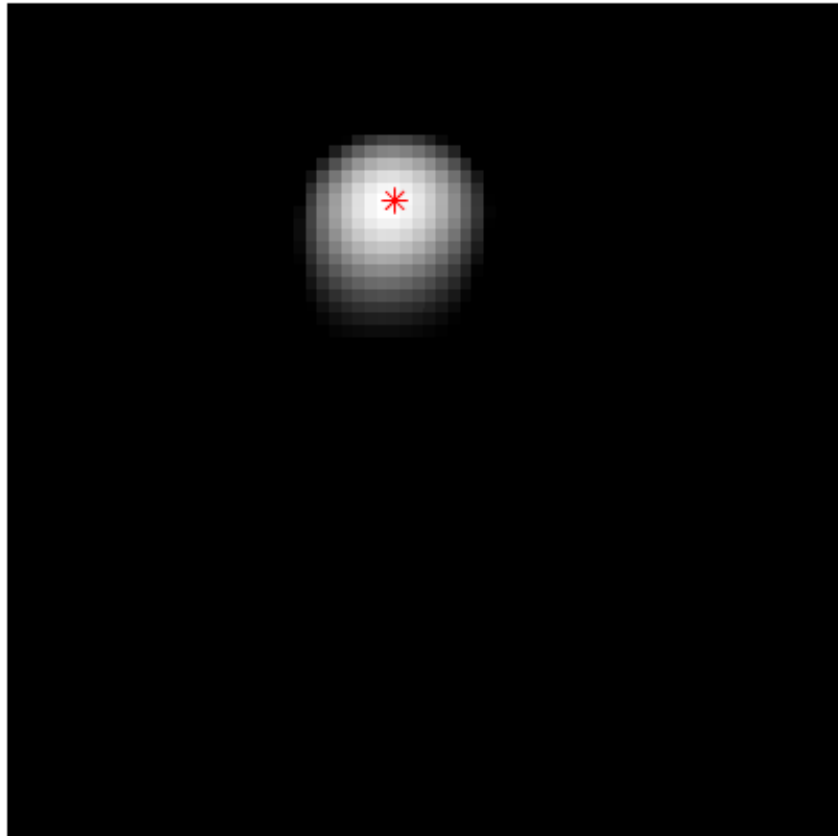


Detection

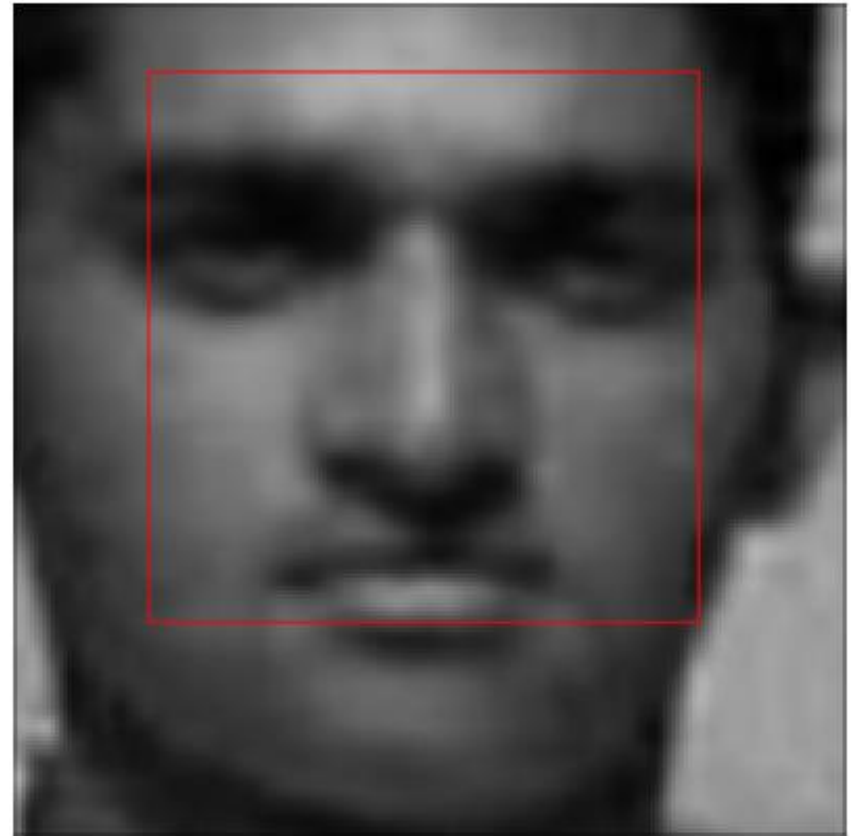


First Results

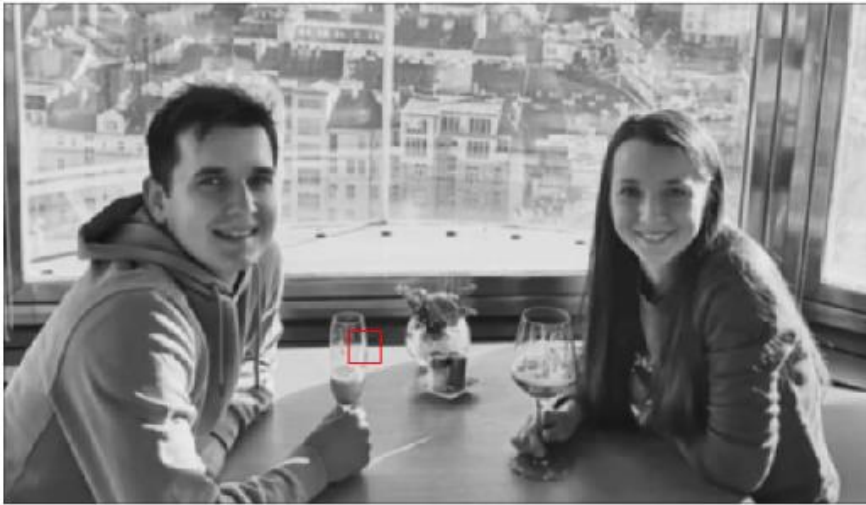
Intensity Map



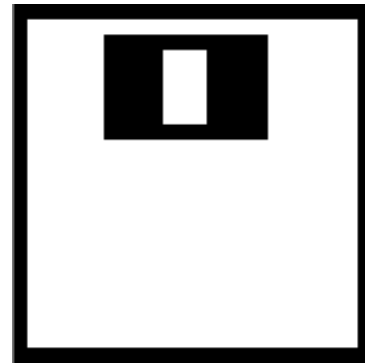
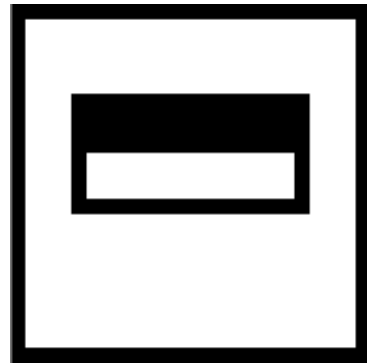
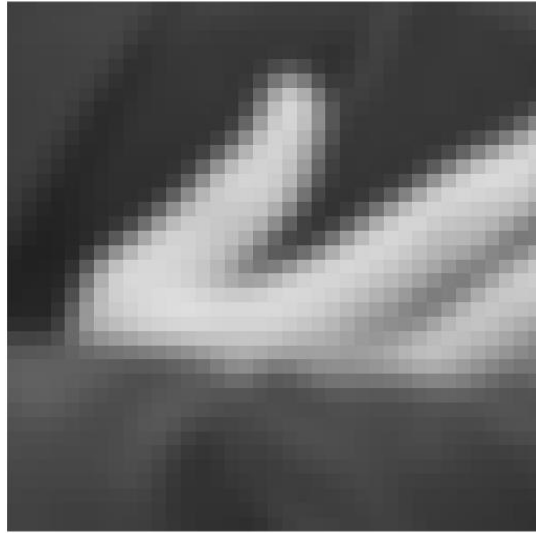
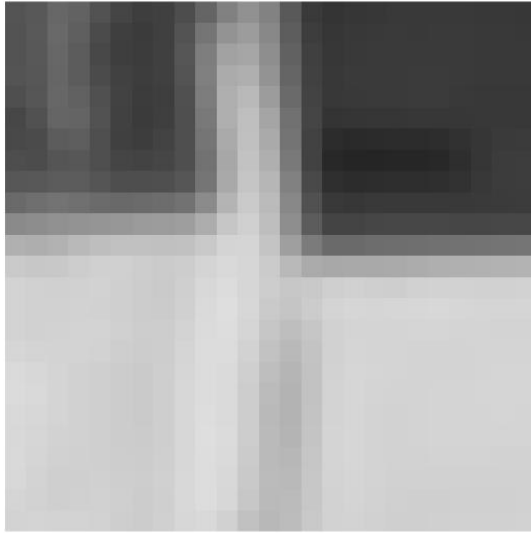
Detection



First Results



First Results



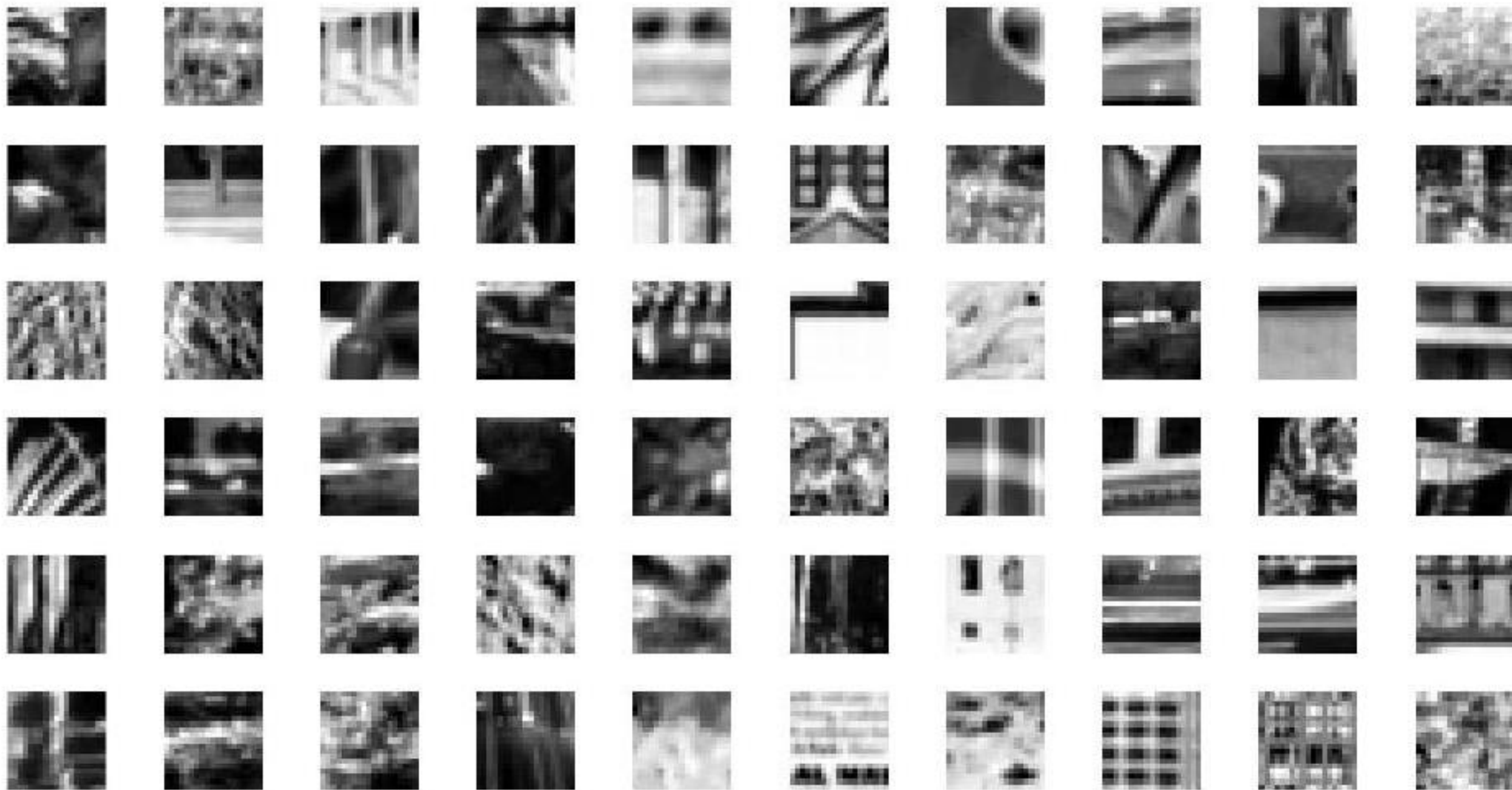
Training our own Haar features

- Dataset:
 - 2429 face images
 - 4547 non-face images
 - All images already in grayscale with size 19x19 pixels
- Calculate Haar feature values
- AdaBoost

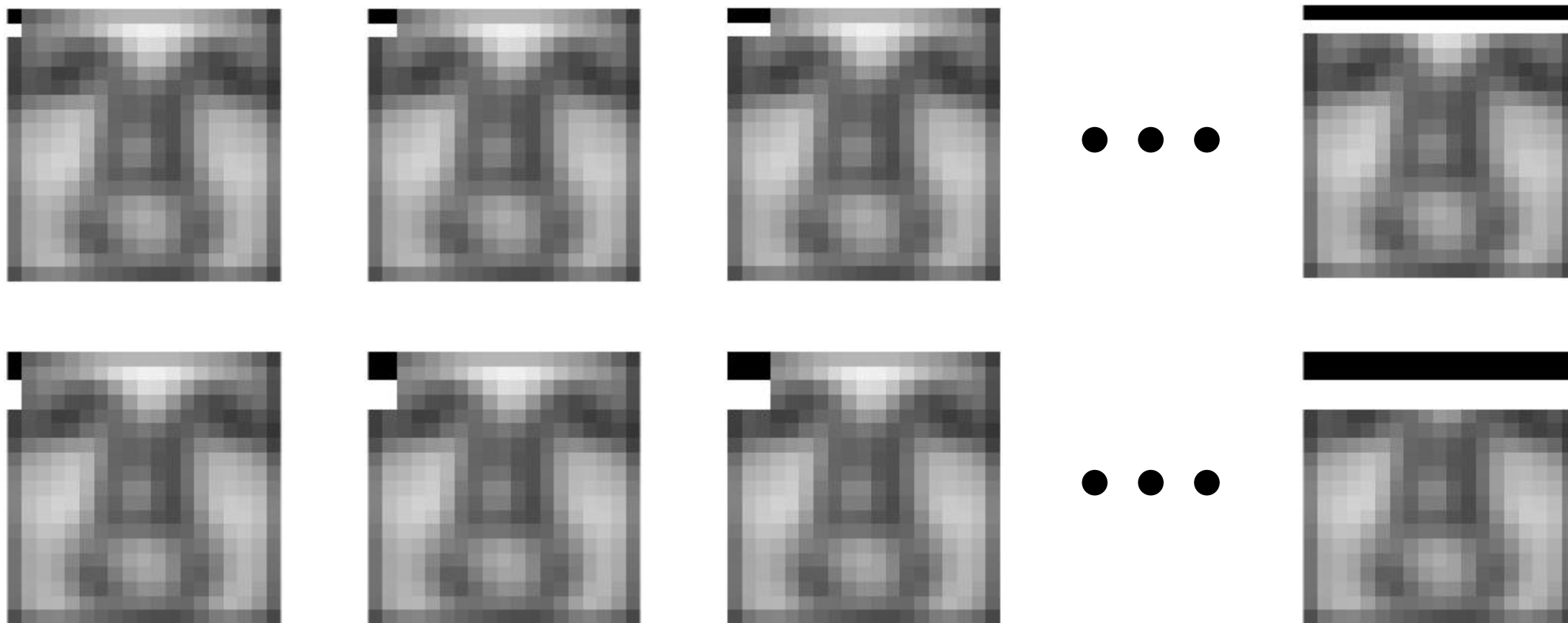
Face Images



Non-Face Images



Calculating Haar feature values



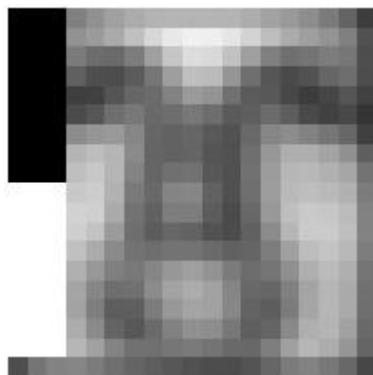
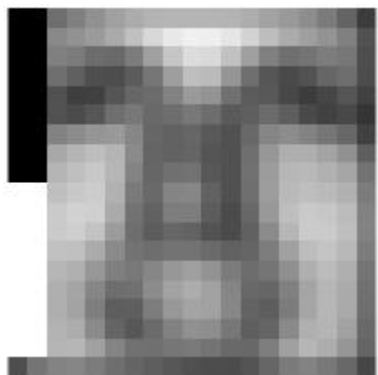
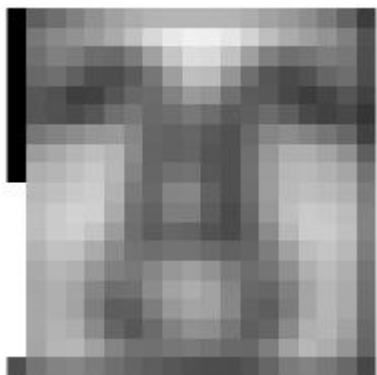
Calculating Haar feature values

⋮

⋮

⋮

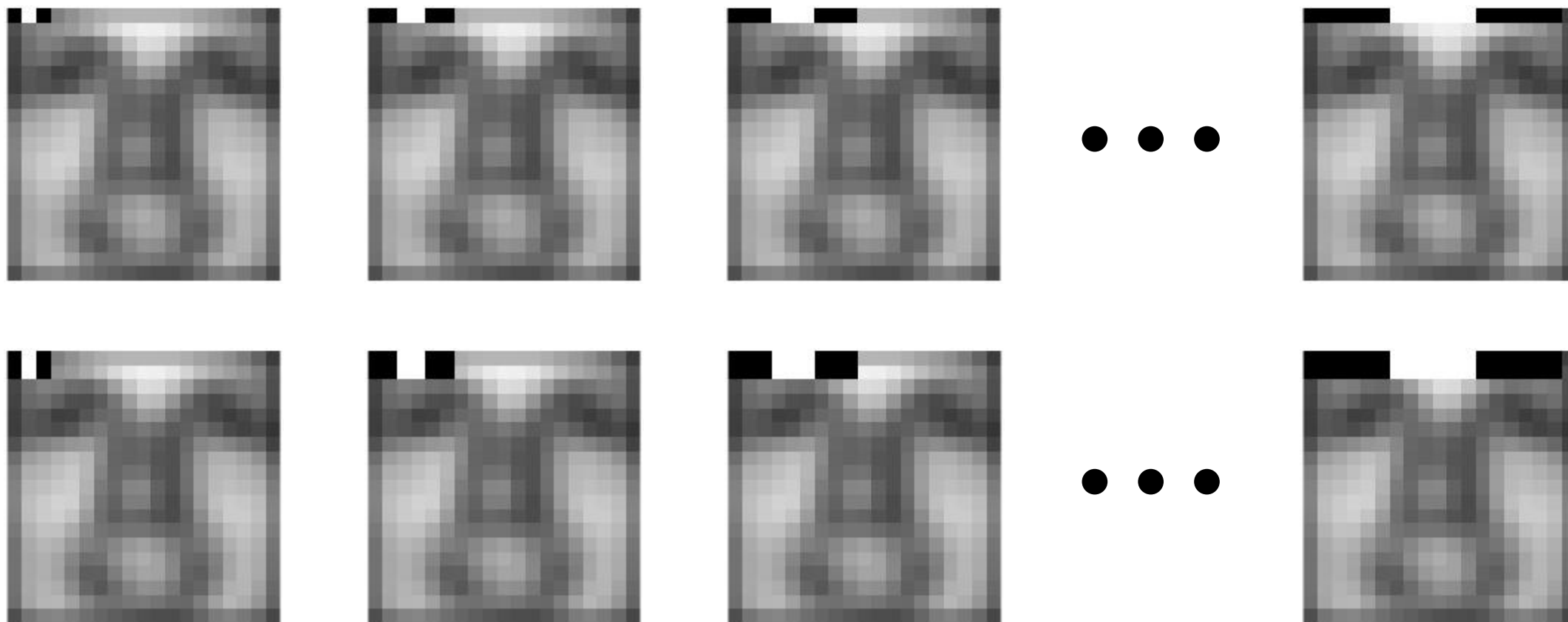
⋮



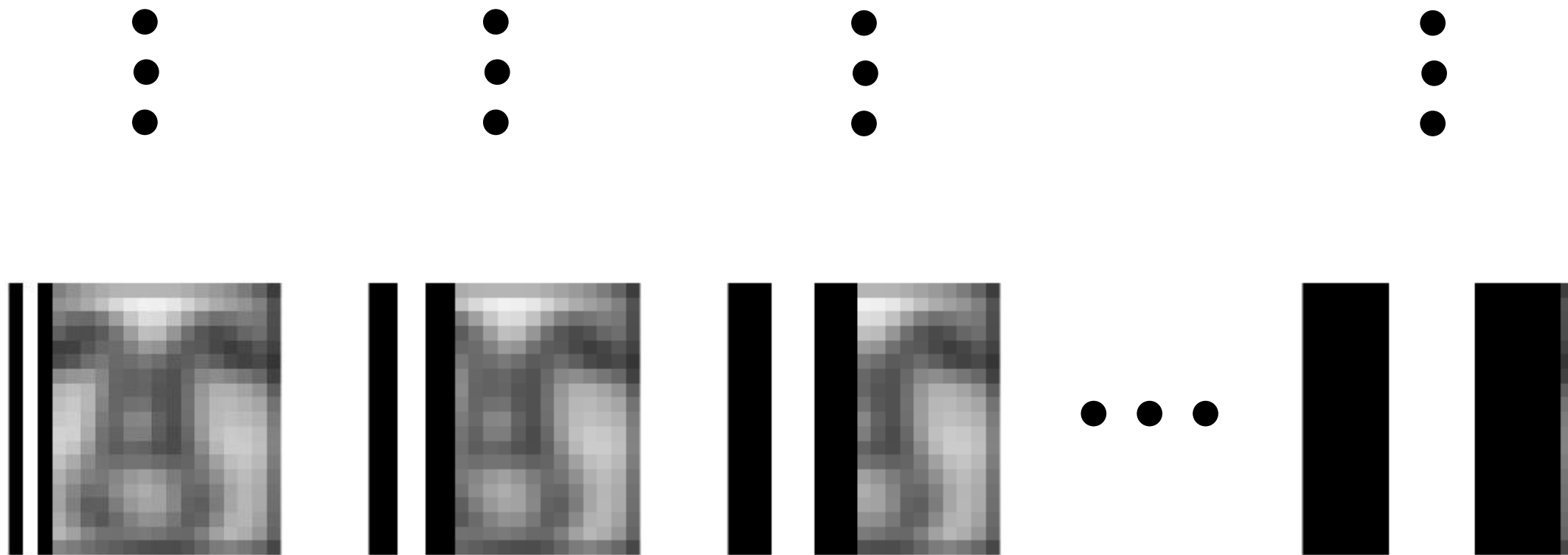
...



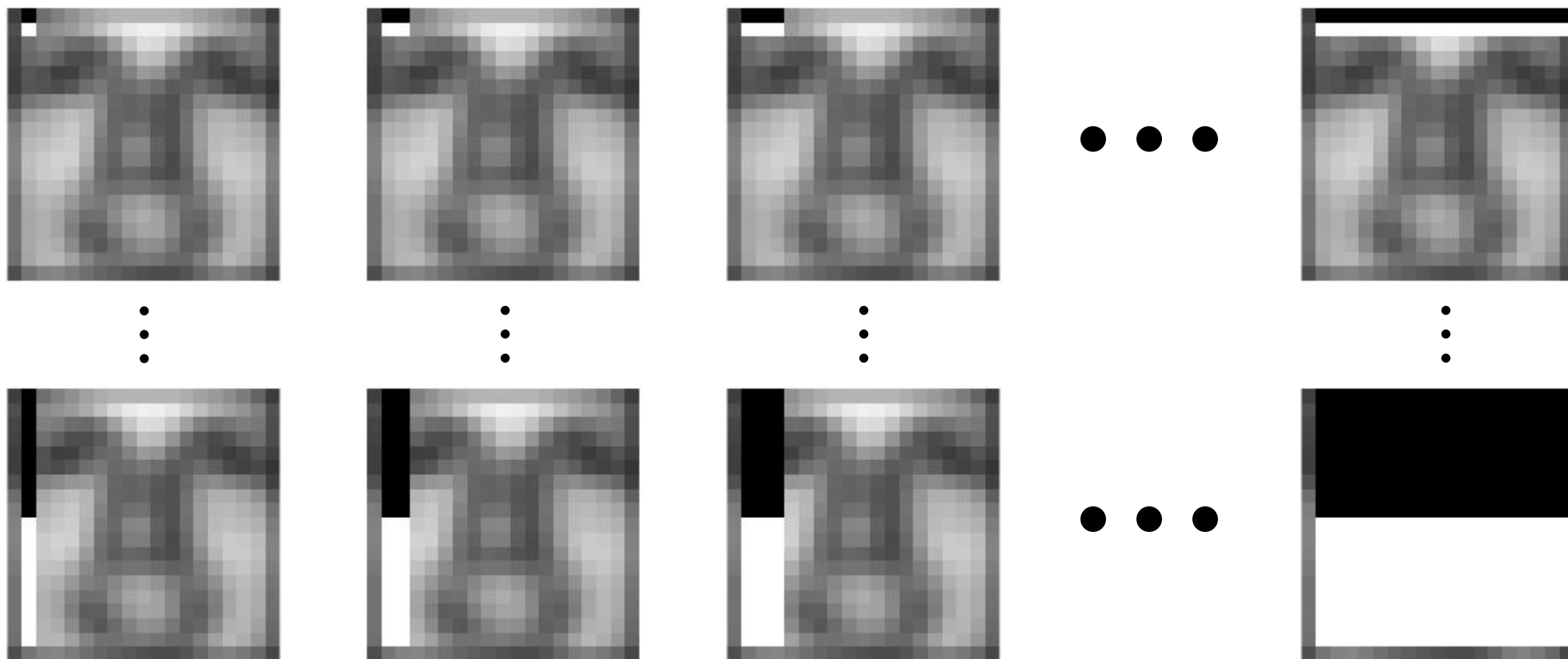
Calculating Haar feature values



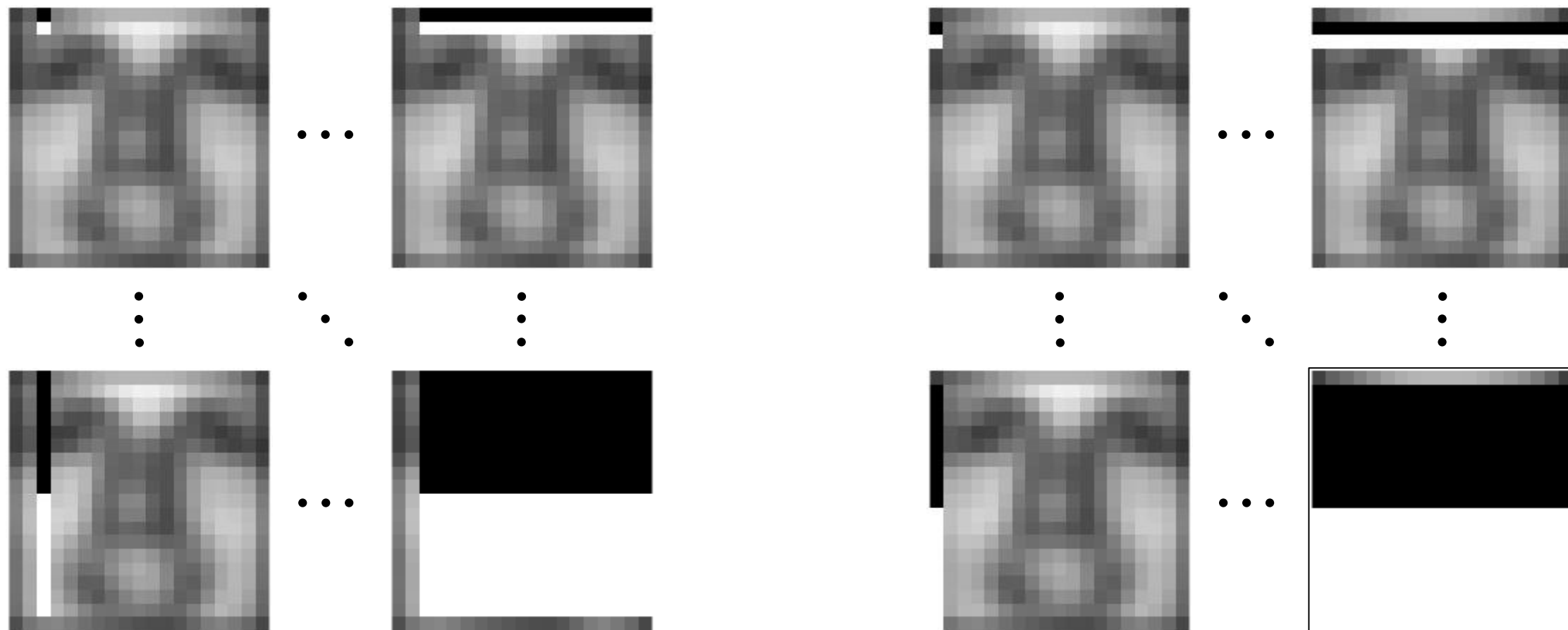
Calculating Haar feature values



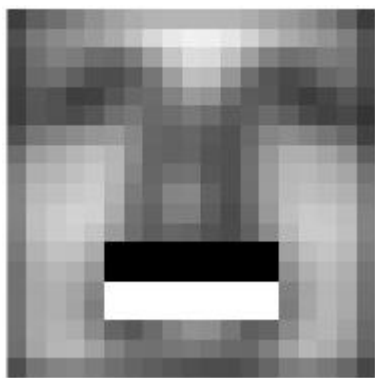
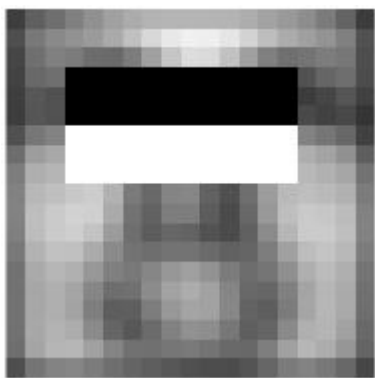
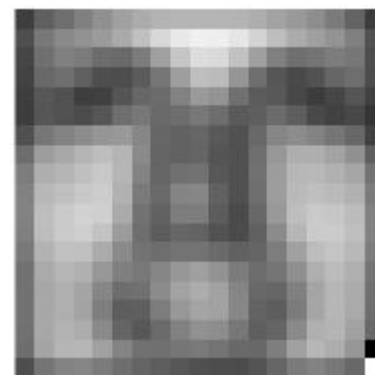
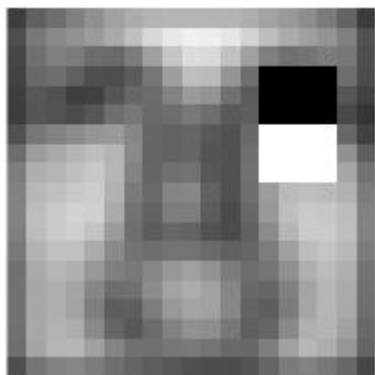
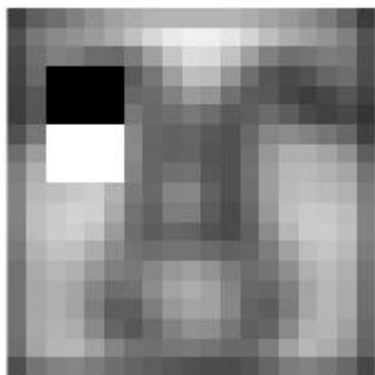
Calculating Haar feature values



Calculating Haar feature values

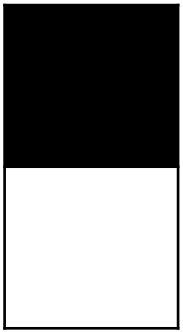


Calculating Haar feature values



Types of Haar features

Type 1



Type 2



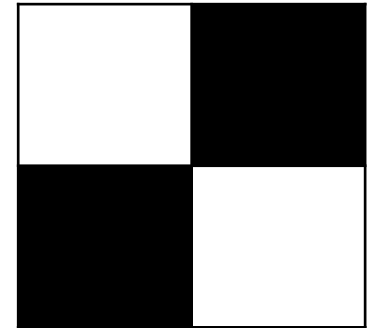
Type 3



Type 4



Type 5



AdaBoost

1.Step: Sorting

fval	sign
2	1
4	1
-1	1
1	-1
-3	-1
-5	-1

Sorting

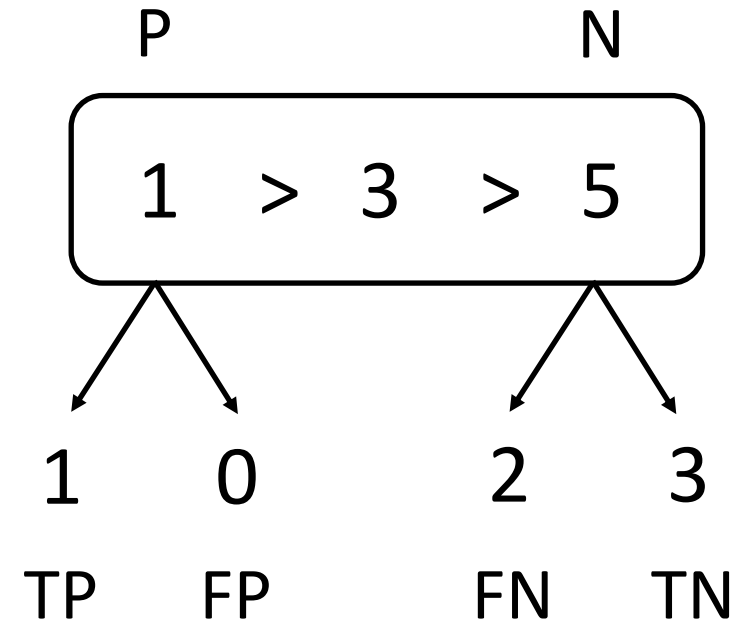


fval	sign
4	1
2	1
1	-1
-1	1
-3	-1
-5	-1

2.Step: Threshold

fval	sign
4	1
2	1
1	-1
-1	1
-3	-1
-5	-1

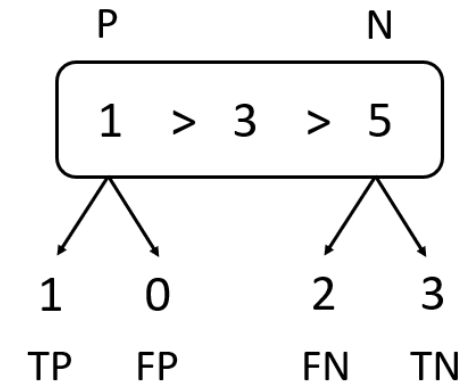
➤ $(4+2)/2 = 3$
➤ 1.5
➤ 0
➤ -2
➤ -4



Gini impurity

Gini impurity for a leaf = $1 - (\text{the probability of face})^2 - (\text{the probability of no face})^2$

$$\begin{array}{ll} \text{P:} & \text{N:} \\ = 1 - \left(\frac{1}{1+0}\right)^2 - \left(\frac{0}{1+0}\right)^2 & = 1 - \left(\frac{2}{2+3}\right)^2 - \left(\frac{3}{2+3}\right)^2 \\ = 1 - 1 - 0 & = 1 - 0.16 - 0.36 \\ = 0 & = 0.48 \end{array}$$



Total Gini impurity = weighted average of Gini impurities for the leaves

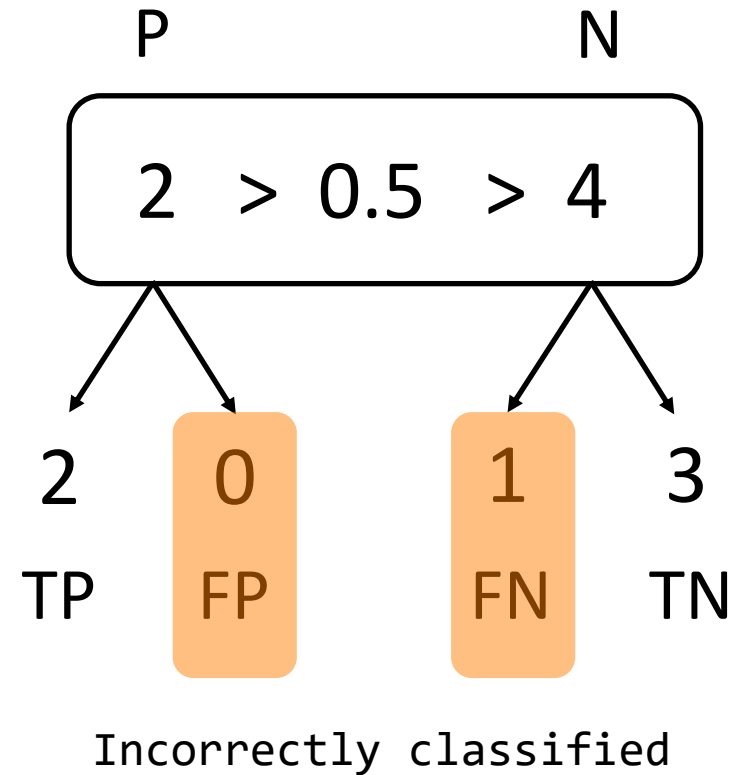
$$= \left(\frac{1}{1+5}\right) \cdot 0 + \left(\frac{5}{1+5}\right) \cdot 0.48 = 0 + 0.4 = \mathbf{0.4}$$

Determining threshold

fval	sign
4	1
2	1
1	-1
-1	1
-3	-1
-5	-1



3 \Rightarrow 0.4
0.5 \Rightarrow 0.25
0 \Rightarrow 0.44
-2 \Rightarrow 0.25
-4 \Rightarrow 0.4



3.Step: Calculating New Sample weights

fval	sign	weight
4	1	1/6
2	1	1/6
1	-1	1/6
-1	1	1/6
-3	-1	1/6
-5	-1	1/6

Total error = $\frac{\text{sum of weights of incorrectly classified samples}}{6} = \frac{1}{6} = 0.16\bar{6}$

$$\text{Amount of say} = \frac{1}{2} \log \left(\frac{1 - \text{Total error}}{\text{Total error}} \right) = \frac{1}{2} \log \left(\frac{1 - 1/6}{1/6} \right) = 0.805$$

New Sample weight = Sample weight $\times e^{\text{Amount of say}}$

$$= \frac{1}{6} \cdot e^{0.805} = \mathbf{0.373}$$

Calculating New Sample weights

fval	sign	weight
4	1	0.166
2	1	0.166
1	-1	0.166
-1	1	0.373
-3	-1	0.166
-5	-1	0.166
		1.206

$$\times \frac{1}{1.206}$$

fval	sign	weight
4	1	0.138
2	1	0.138
1	-1	0.138
-1	1	0.309
-3	-1	0.138
-5	-1	0.138
		1

4.Step: New feature values

fval	sign	weight	csum
4	1	0.138	0.138
2	1	0.138	0.276
1	-1	0.138	0.415
-1	1	0.309	0.723
-3	-1	0.138	0.862
-5	-1	0.138	1

Random number
between 0-1

0.1067

0.5265

0.7749

0.3998

0.8173

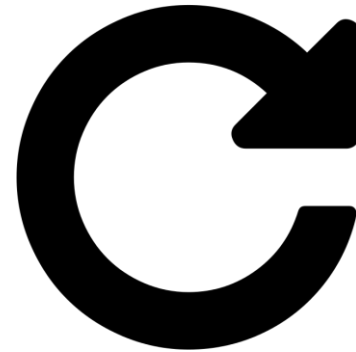
0.6032

fval	sign
4	1
-1	1
-3	-1
2	1
-3	-1
-1	1

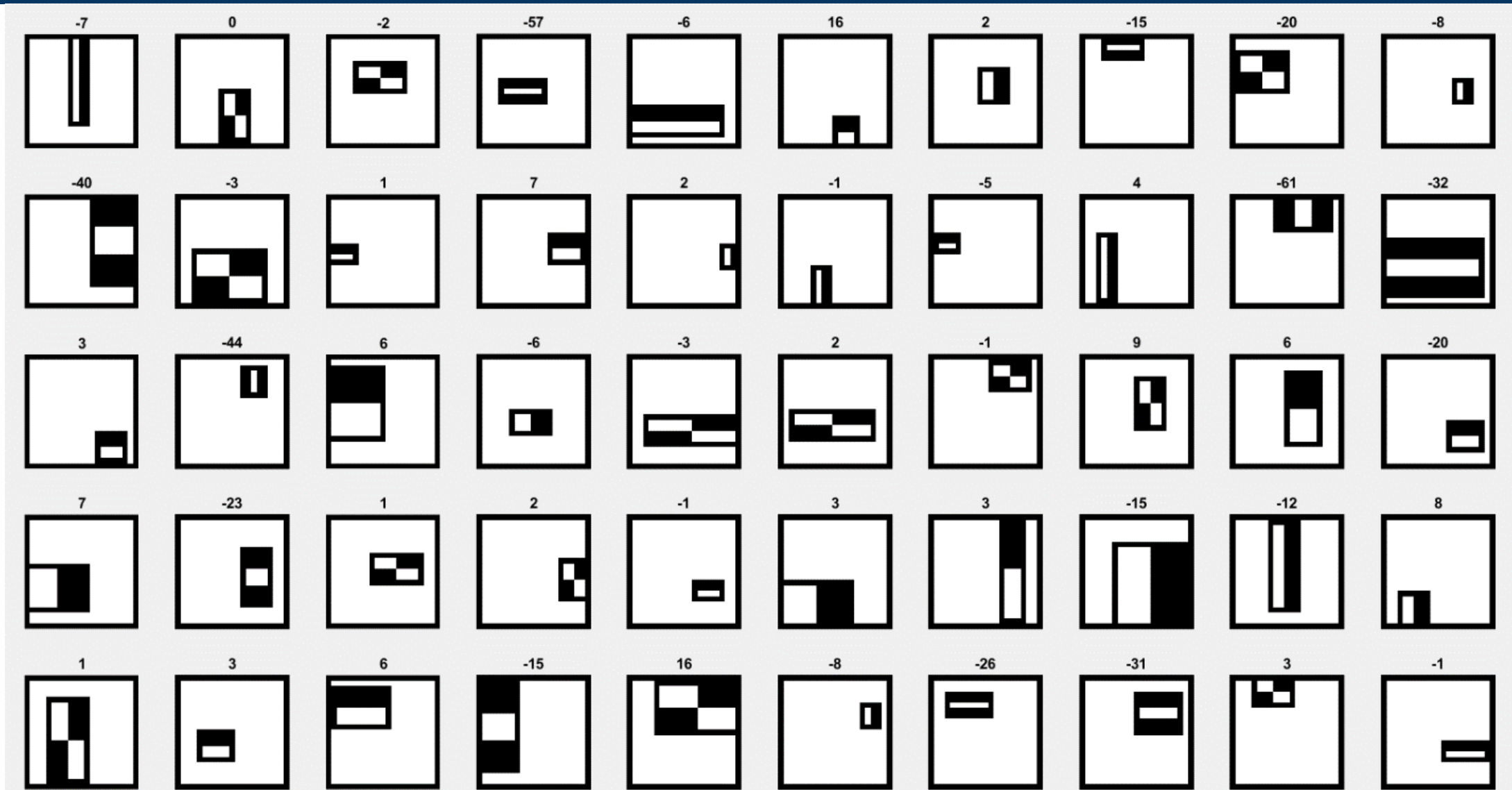
5.Step: Repeat

fval	sign
4	1
-1	1
-3	-1
2	1
-3	-1
-1	1

Repeat for a number
of iterations



Results from AdaBoost (50 itr)



Final Results

Applied Haar features

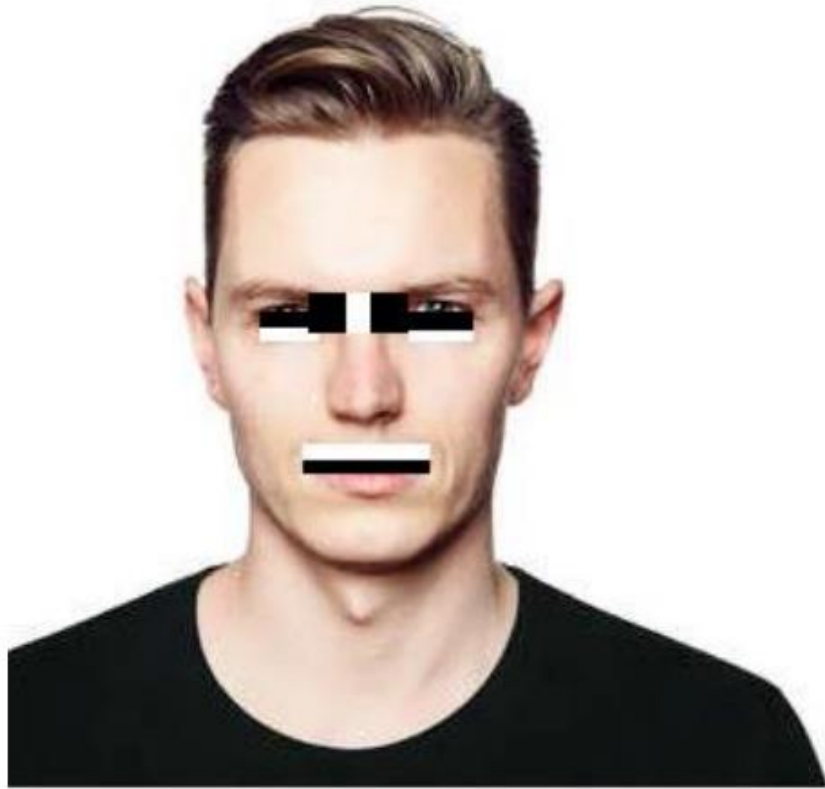
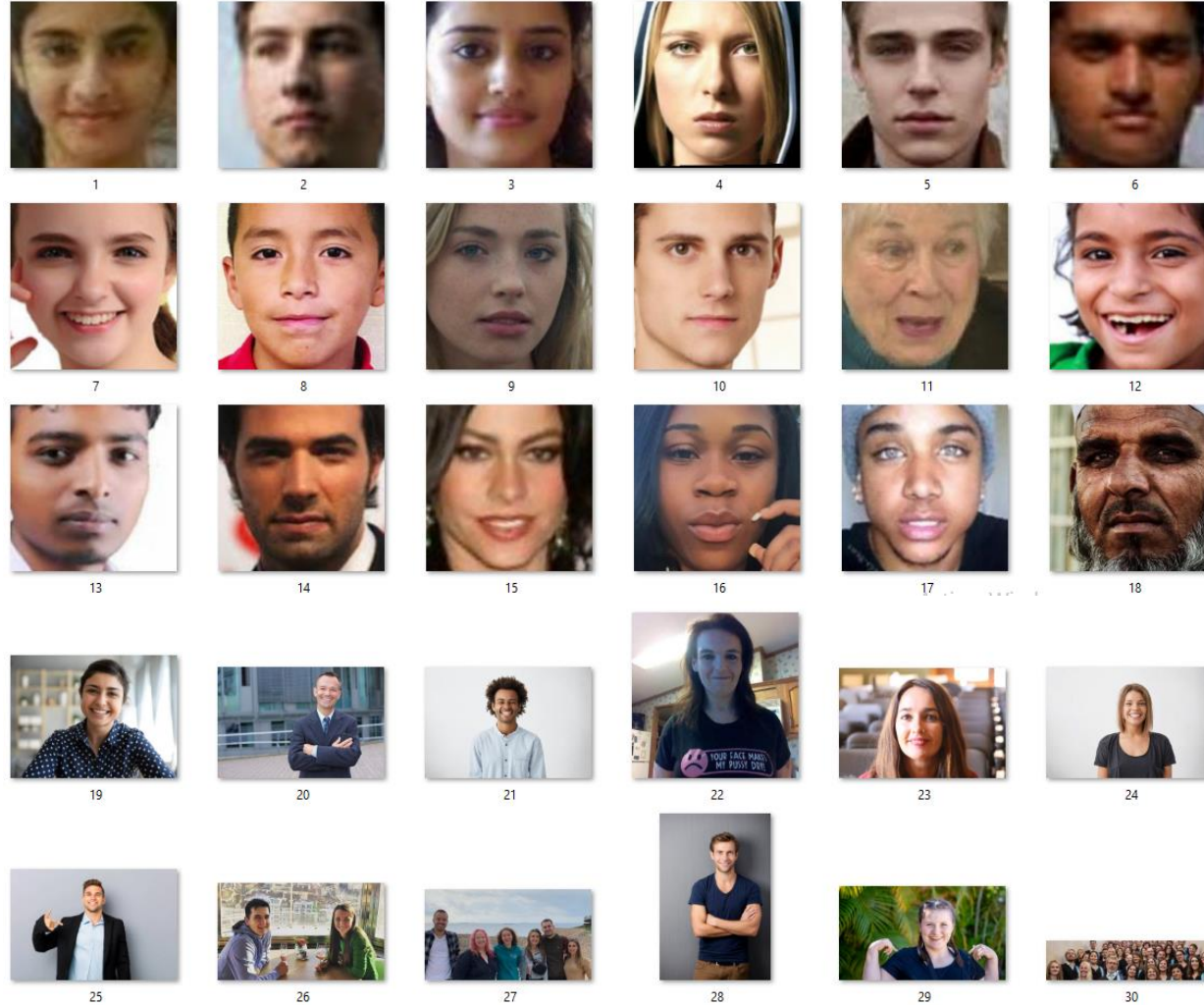


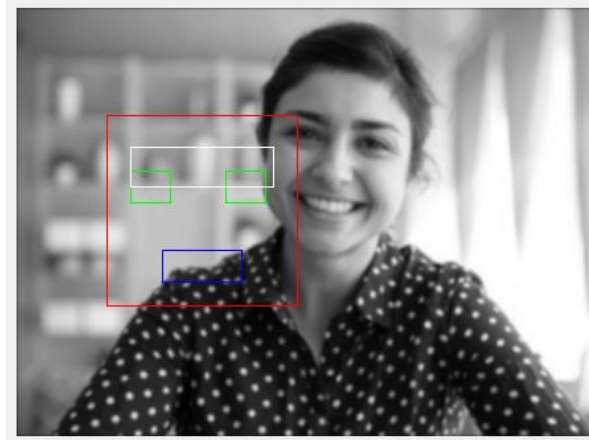
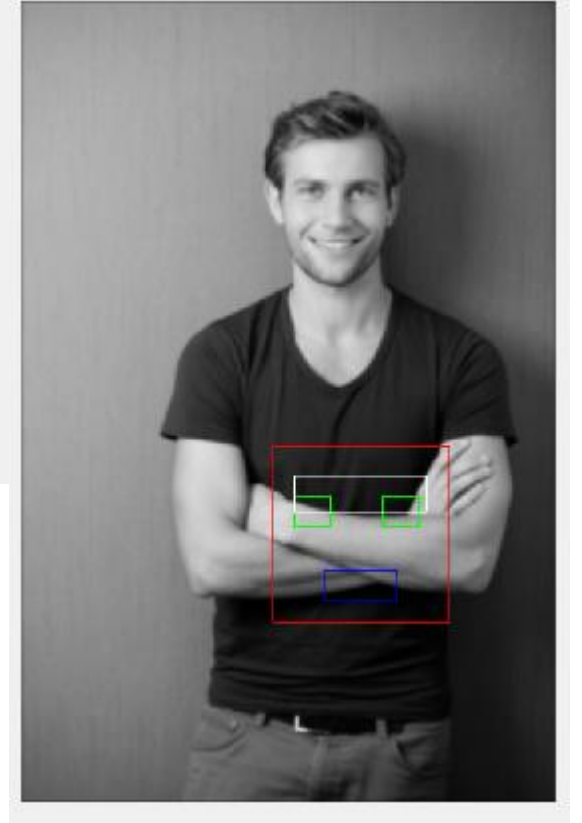
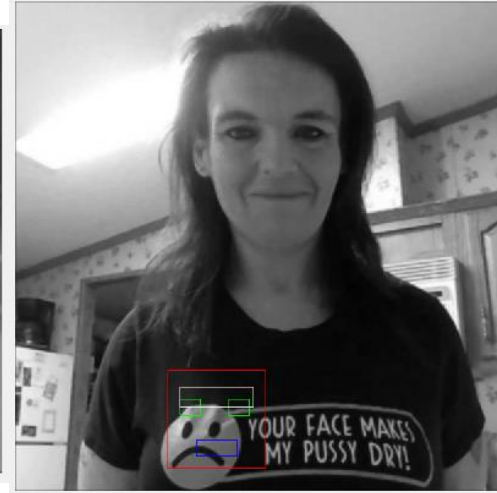
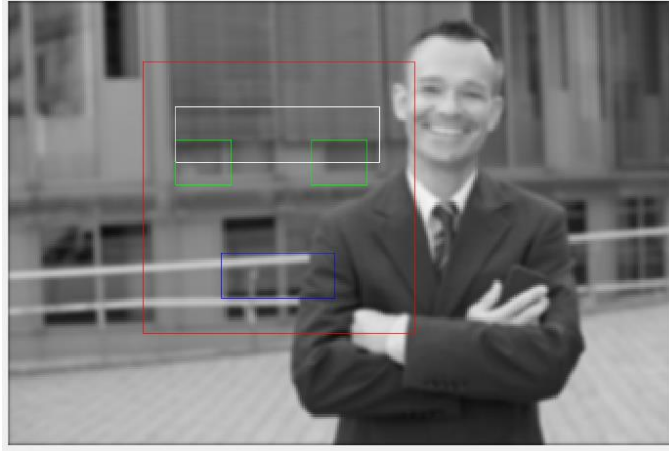
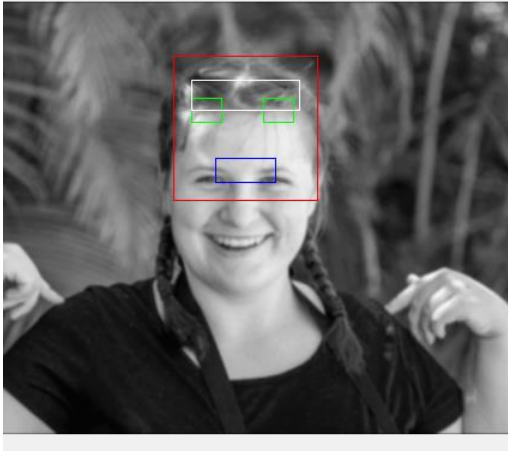
Image database



Final results



Final results



Comparison of results



Things we tried

- Optimizing the initial size of the Haar feature
- Different combinations of Haar features
- Changing threshold
- Cascade
- Calculating other types of Haar features using AdaBoost

We want to try:

- To apply exactly the features obtained from AdaBoost with their calculated threshold

Conclusions

- It's **HARD** and it takes a lot of time to try different features manually.
- Threshold is very important and is different for each type and size of Haar feature.
- Almost impossible to get the perfect cascade of classifiers that works on complicated images manually.
- **STUDY MACHINE LEARNING**

%% Thank you for attention

% If you have any questions
% don't hesitate to ask