

lect 12 Face Detection

Face detection & Recognition.

↓
only find the face

↓
know who it is

a mature technology

Challenge for face detection.

- Sliding window should cover all possible locations and say "True" or "False".

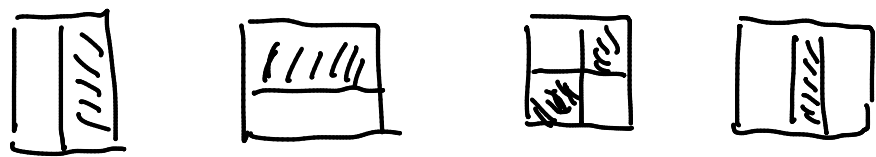
But faces are rare. Say, in a image with 1 Million pixels, only at most 10 faces are common.

- Time efficiency. Spend less time on non-face areas
- False positive / False alarm rate should $< 10^{-6}$

Viola/Jones Face Detection.

- use integral image for fast feature evaluation
- use boosting method for feature selection
- additional cascade strategy for fast rejection

Feature detector.



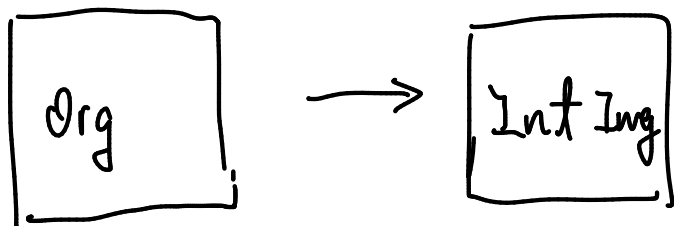
Weak Filters
with ensemble method

$$\square = -1 \quad \square = 1$$

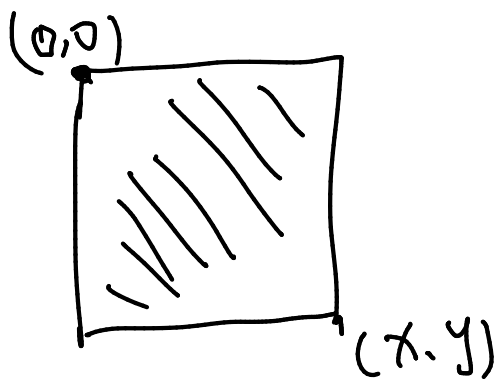
So the value = \sum white area - \sum black area

If image is totally noise, the summation will be 0

We choose these "rectangle" like detectors, because we could make use of integral image

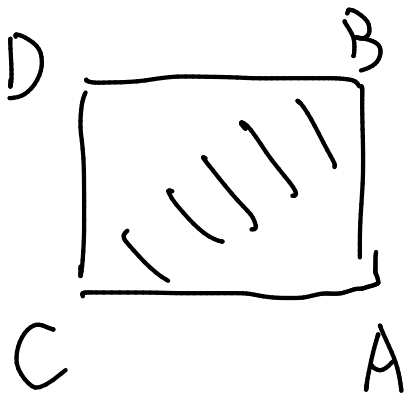


pixel in integral image say (x, y) ,
is the summation from $(0, 0)$ to (x, y)



$$\text{Val}(x, y) = \sum_{x, y} \text{img}(x, y)$$

the operation could be paralleled. First sum row,
then sum col.



$$\text{sum} = A - B - C + D$$

So, we could have a efficient filter computer But we still could not deploy thousands of filters of different sizes.
 ↓

Boosting / Ensemble classifier.

Weak learners \Rightarrow Strong ensemble learner.

\rightarrow for feature filter i , it's response is h_i

$$h_i(x) = \begin{cases} 1 & \text{if } p_i f_i(x) > p_i \theta \\ 0 & \text{else} \end{cases}$$

\uparrow window \uparrow filter result \uparrow parity \nwarrow threshold

\rightarrow Ensemble Classification function \swarrow half vote

$$C(x) = \begin{cases} 1 & \text{if } \sum_{i=1}^T d_i h_i(x) > \frac{1}{2} \sum_{i=1}^T d_i \\ 0 & \text{else} \end{cases}$$

\uparrow majority vote \uparrow learnt weight

Training Stage.

- Initially, each training example are equally treated
 - For each boosting round:
 - Find the weak learner achieves lowest weighted training error
 - Raise the misclassified training examples' importance
- Intuition, we find out which weak learner that can do well on difficult examples
- Final classifier is a combination of all weighted weak learners
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Boosting :

Pros :

- ① Testing is fast
- ② Easy to implement
- ③ flexible of weak learner choosing

Con :

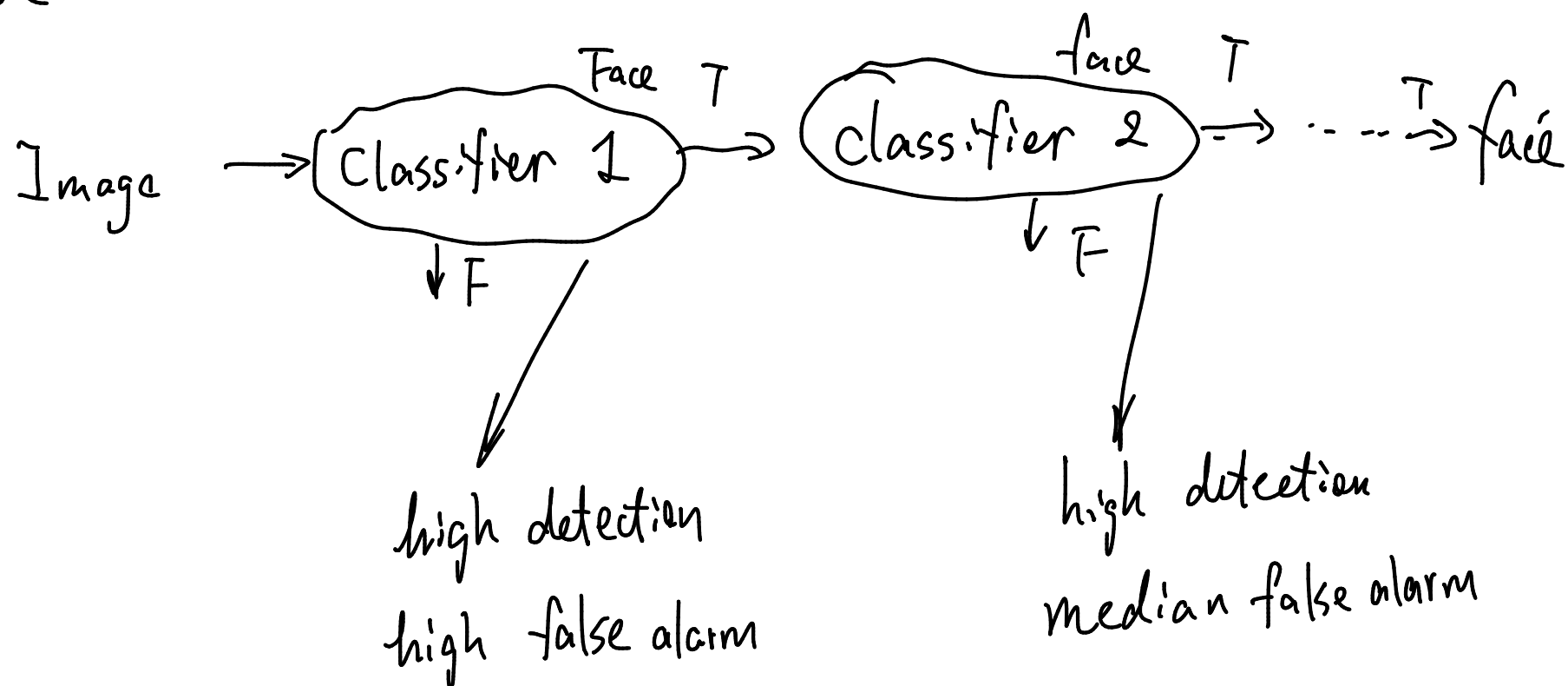
- ① Training slow
- ② needs lots of samples

ROC (Receiver Operating Characteristic) Curve.

a 200-feature detect would have 95% Detection / $\frac{1}{14084}$ False alarm.

The false alarm rate is high, 'ecouse we may apply the filter millions of time. We need more!

We have another filter could reject non-faces.



We can apply more and more complicated filters.

Train cascade

- Set a target detection/false alarm rate
- Keep adding features, until the target rate is met
- Add stage if false alarm rate is not met

Training set

say 1% face, 99% non-face