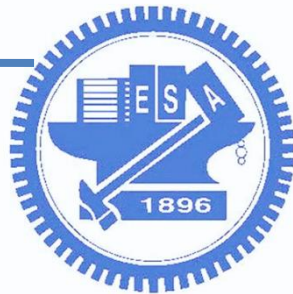


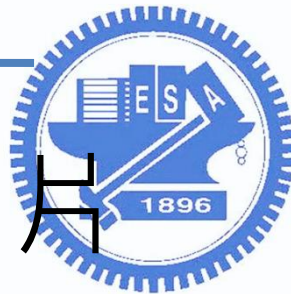
6. Facial camera

National Chiao Tung University



Outline

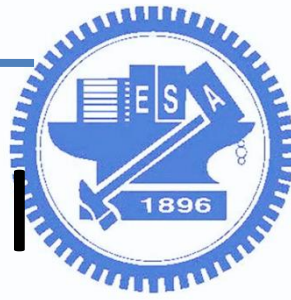
- 網路攝影機
 - 觀看Raspberry Pi Camera的圖片
 - 影像辨識 (opencv)
 - 圖片旋轉, 裁切, 縮放
 - 人臉識別



觀看Raspberry Pi Camera的圖片

- `python -m SimpleHTTPServer 8000`
- `winscp`
- `vnc`

- Create Wi-Fi hotspot on PI
- View image in terminal



Create Wi-Fi hotspot on Pi

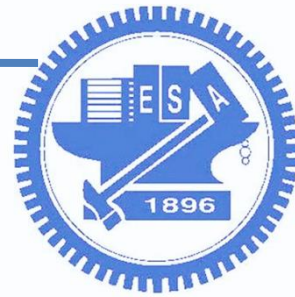
- ❑ `wget https://raw.githubusercontent.com/raspberrypi-tw/sh/master/dual_mode.sh`
- ❑ `chmod +x dual_mode.sh`
- ❑ `sudo ./dual_mode.sh on` # it will install related packages and reboot
- ❑ `sudo ./dual_mode.sh off` # it will reboot

```
(COM8) [80x24]
連線(C) 編輯(E) 檢視(V) 視窗(W) 選項(O) 說明(H)
pi@raspberrypi:~/Pictures$ sudo ./dual_mode.sh on
Check Pi 3 OK
Dual mode on...

Input a number from [3] to [252]...> 4

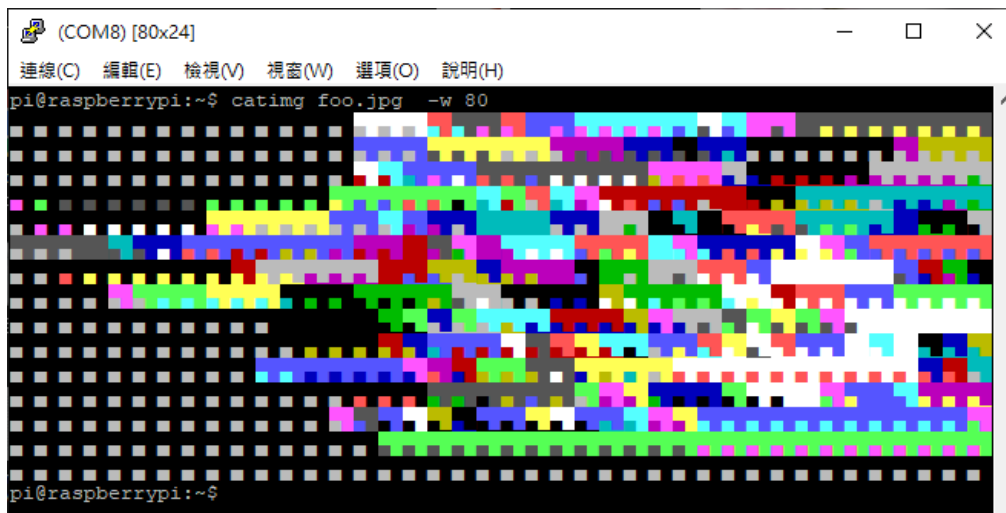
=====
SSID: [RPi-4]
PSK:  [1234567890]
=====
After connect to [RPi-4], you can SSH 'pi@192.168.[4].1' to your Pi

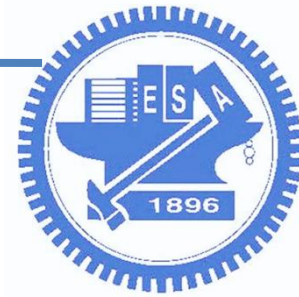
Confirm the setting? [yes/no] > yes
```



View image in terminal

- Insanely fast image printing in your terminal
 - `git clone https://github.com/posva/cating`
 - `cd cating/`
 - `cmake .`
 - `sudo make install`
- Usage: `cating xxxxxxxx.jpg -w 80`





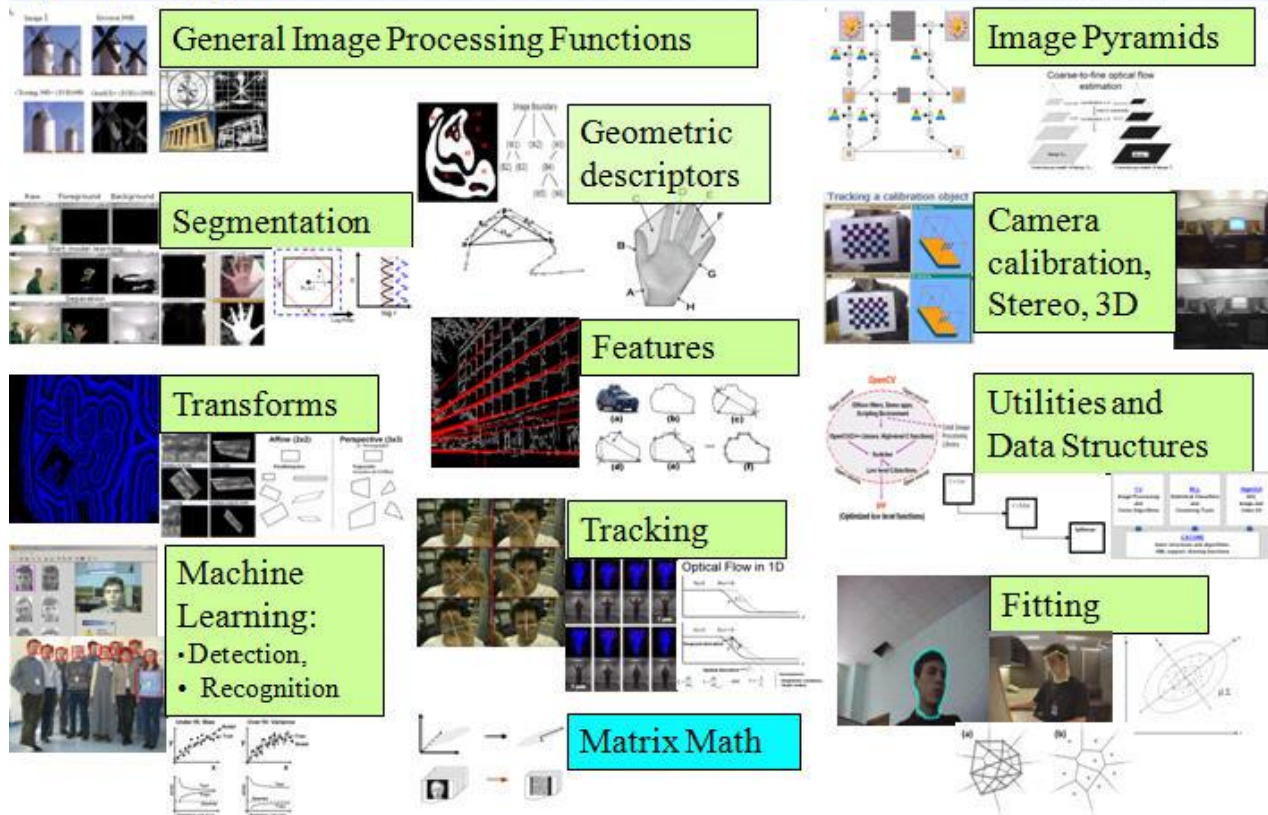
opencv

Open Source Computer Vision Library

OpenCV Overview: > 500 functions

opencv.willowgarage.com

Robot support



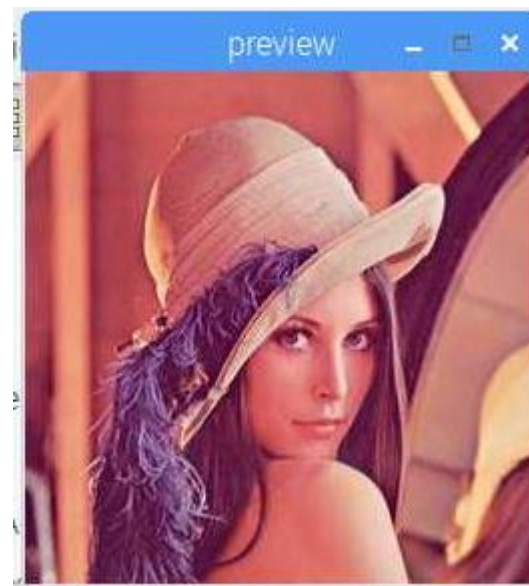
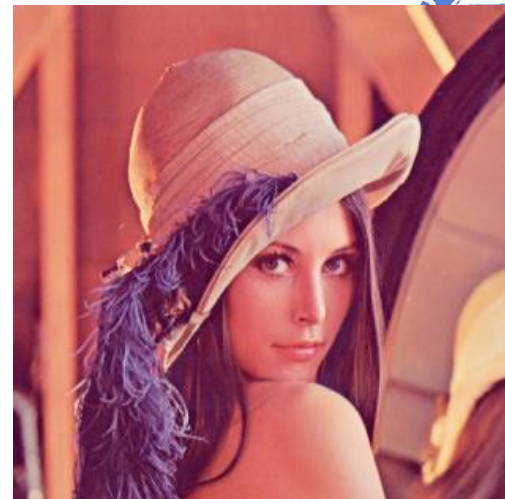


preview

□ Sample code

```
import cv2
import numpy as np
img = cv2.imread('lena256rgb.jpg')

cv2.imshow('preview', img)
cv2.waitKey(0)
cv2.destroyAllWindows()
```

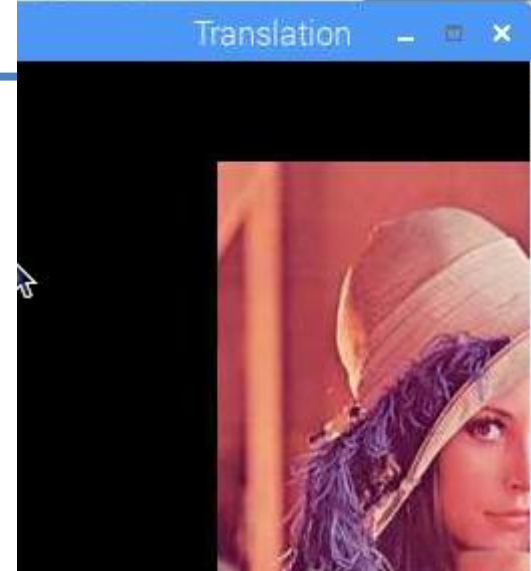


translation

- Applies an affine transformation to an image.

```
import cv2
import numpy as np
img = cv2.imread('lena256rgb.jpg')
rows, cols = img.shape[:2]
M = np.float32([ [1,0,100], [0,1,50] ])
translation = cv2.warpAffine(img, M, (cols, rows))
cv2.imshow('Translation', translation)
cv2.waitKey(0)

cv2.destroyAllWindows()
```



The function `warpAffine` transforms the source image using the specified matrix:

$$\text{dst}(x, y) = \text{src}(M_{11}x + M_{12}y + M_{13}, M_{21}x + M_{22}y + M_{23})$$

rotation

- Calculates an affine matrix of 2D rotation.

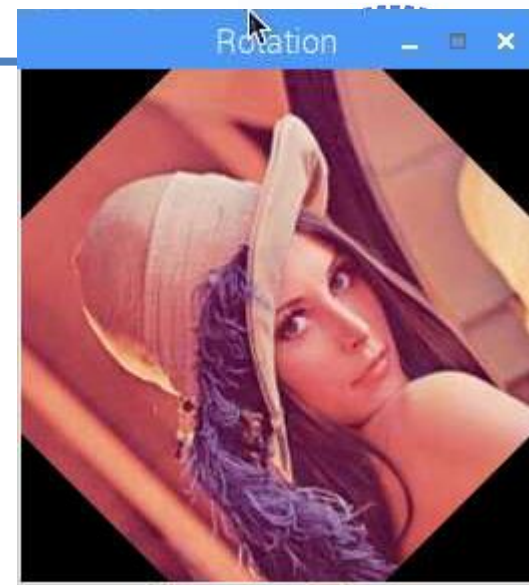
```
import cv2
import numpy as np

img = cv2.imread("lena256rgb.jpg")
rows, cols = img.shape[:2]

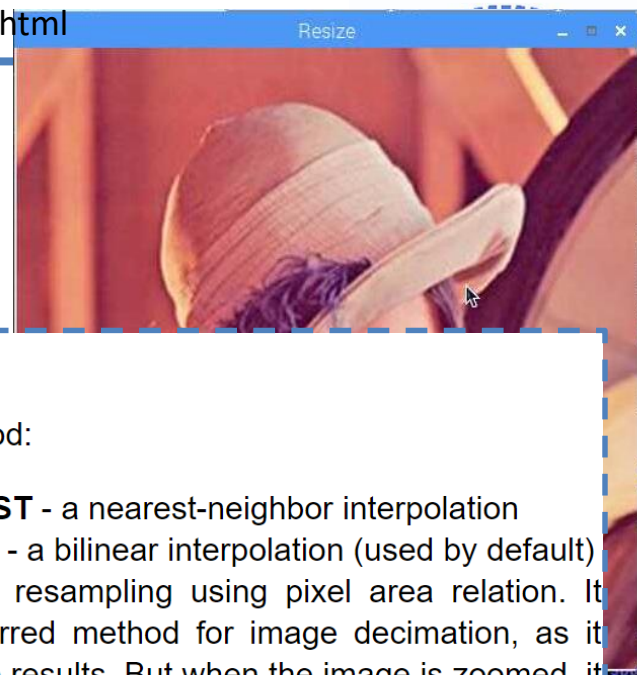
M = cv2.getRotationMatrix2D((cols/2, rows/2), 45, 1)
rotation = cv2.warpAffine(img, M, (cols, rows))

cv2.imshow('Rotation', rotation)
cv2.waitKey(0)

cv2.destroyAllWindows()
```



resize



- Resizes an image.

```
import cv2
import numpy as np

img = cv2.imread("lena256rgb.jpg")
rows, cols = img.shape[:2]

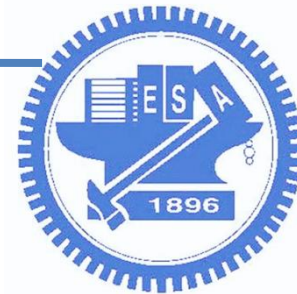
resize = cv2.resize(img, (2*rows, 2*cols), interpolation =
cv2.INTER_CUBIC)
cv2.imshow('Resize', resize)
cv2.waitKey(0)

cv2.destroyAllWindows()
```

- interpolation –

interpolation method:

- **INTER_NEAREST** - a nearest-neighbor interpolation
- **INTER_LINEAR** - a bilinear interpolation (used by default)
- **INTER_AREA** - resampling using pixel area relation. It may be a preferred method for image decimation, as it gives moire'-free results. But when the image is zoomed, it is similar to the **INTER_NEAREST** method.
- **INTER_CUBIC** - a bicubic interpolation over 4x4 pixel neighborhood
- **INTER_LANCZOS4** - a Lanczos interpolation over 8x8 pixel neighborhood



crop

□ Sample code

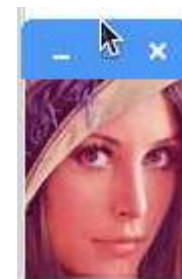
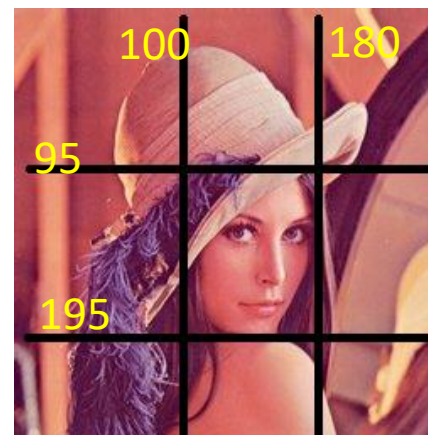
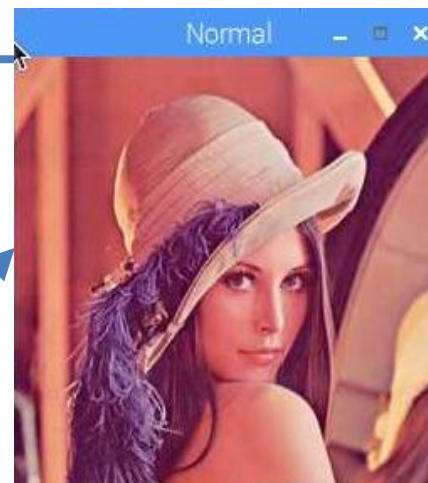
```
import cv2
import numpy as np

img = cv2.imread("lena256rgb.jpg")
cv2.imshow("Normal", img)
cv2.waitKey(0)
```

```
face = img[95:195, 100:180]
cv2.imshow("Face", face)
cv2.waitKey(0)
```

```
body = img[20:, 35:210]
cv2.imshow("Body", body)
cv2.waitKey(0)
```

```
cv2.destroyAllWindows()
```





Install opencv

□ Command

- **sudo apt-get install python-opencv**
- Download sample code and unzip it (6_face_detection.zip)
- Load module: **sudo modprobe bcm2835-v4l2**

□ Two sample code

- Analyze image
 - `python image_face_detect.py your_img haarcascade_frontalface_default.xml`
- Analyze camera
 - `python camera_face_detect.py haarcascade_frontalface_default.xml`



Facial detection (python)

```
python camera_face_detect.py haarcascade_frontalface_default.xml
```

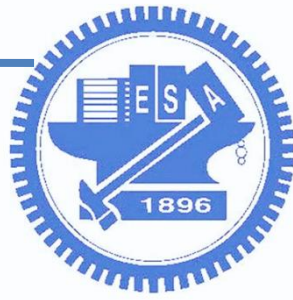
```
import sys
import cv2

cascPath = sys.argv[1]
faceCascade = cv2.CascadeClassifier(cascPath)

if cv2.__version__.startswith('2'):
    PROP_FRAME_WIDTH = cv2.cv.CV_CAP_PROP_FRAME_WIDTH
    PROP_FRAME_HEIGHT = cv2.cv.CV_CAP_PROP_FRAME_HEIGHT
    HAAR_FLAGS = cv2.cv.CV_HAAR_SCALE_IMAGE

....

cap = cv2.VideoCapture(0)
cap.set(PROP_FRAME_WIDTH, 320)
cap.set(PROP_FRAME_HEIGHT, 240)
```

Facial detection (python)

```
while True:
    # Capture frame-by-frame
    ret, frame = cap.read()

    gray = cv2.cvtColor(frame, cv2.COLOR_BGR2GRAY)

    faces = faceCascade.detectMultiScale(
        gray,
        scaleFactor=1.1,
        minNeighbors=5,
        minSize=(30, 30),
        flags=HAAR_FLAGS
    )
```




Facial detection (python)

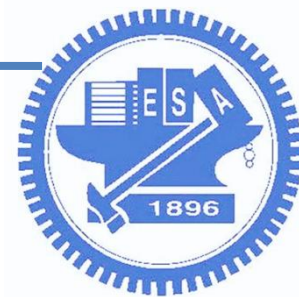
```
print "Found {0} faces!".format(len(faces))

# Draw a rectangle around the faces
for (x, y, w, h) in faces:
    cv2.rectangle(frame, (x, y), (x+w, y+h), (0, 255, 0), 2)

# Display the resulting frame
cv2.imshow("preview", frame)

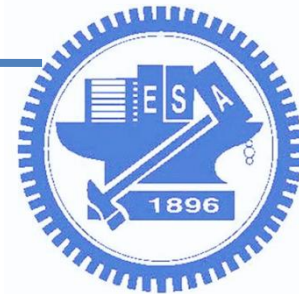
if cv2.waitKey(1) & 0xFF == ord("q"):
    break

# When everything is done, release the capture
cap.release()
cv2.destroyAllWindows()
```



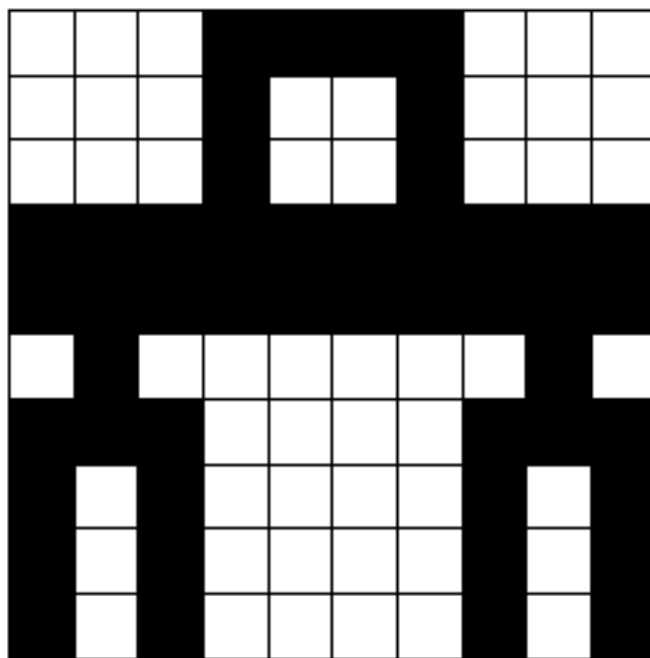
Cascade Classification

- Haar Feature-based Cascade Classifier for Object Detection
 - ▣ The object detector described below has been initially proposed by Paul Viola [Viola01] and improved by Rainer Lienhart [Lienhart02].
 - ▣ a classifier is trained with a few hundred sample views of a particular object (i.e., a face or a car), called positive examples
 - Output 1: the region is likely to show the object (i.e., face/car)
 - Output 0: otherwise

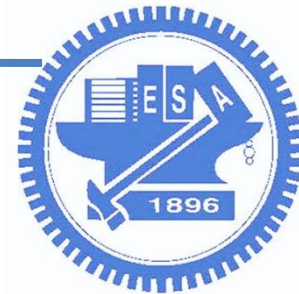


Bitmap images

- Example: black-and-white image

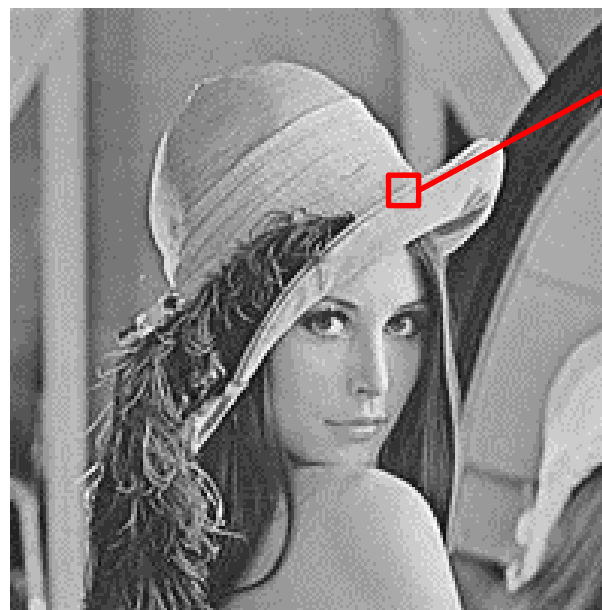


| | | | | | | | | | |
|---|---|---|---|---|---|---|---|---|---|
| 0 | 0 | 0 | 1 | 1 | 1 | 1 | 0 | 0 | 0 |
| 0 | 0 | 0 | 1 | 0 | 0 | 1 | 0 | 0 | 0 |
| 0 | 0 | 0 | 1 | 0 | 0 | 1 | 0 | 0 | 0 |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 |
| 1 | 1 | 1 | 0 | 0 | 0 | 0 | 1 | 1 | 1 |
| 1 | 0 | 1 | 0 | 0 | 0 | 0 | 1 | 0 | 1 |
| 1 | 0 | 1 | 0 | 0 | 0 | 0 | 1 | 0 | 1 |
| 1 | 0 | 1 | 0 | 0 | 0 | 0 | 1 | 0 | 1 |



Bitmap images

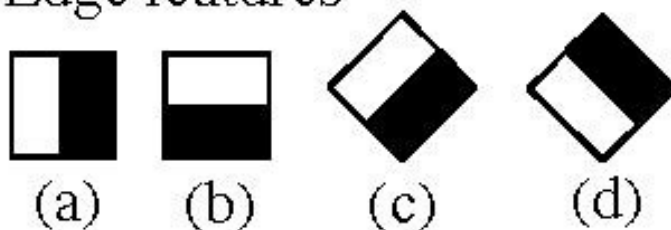
- Example: grayscale picture
 - 8 bits per pixel
 - This pixel depth allows 256 different intensities



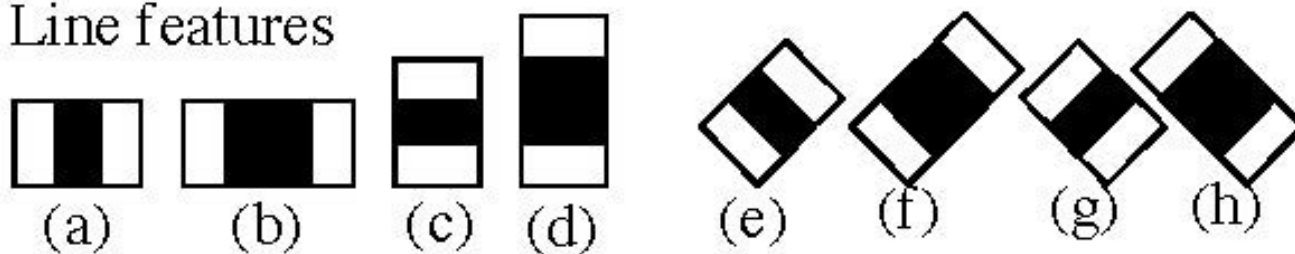
| | | | | |
|-----|-----|-----|-----|-----|
| 154 | 108 | 198 | 216 | 52 |
| 61 | 168 | 148 | 52 | 45 |
| 72 | 80 | 55 | 134 | 39 |
| 89 | 129 | 232 | 204 | 155 |
| 156 | 99 | 118 | 125 | 83 |

Haar-Like Features

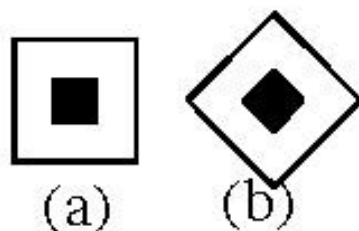
1. Edge features



2. Line features

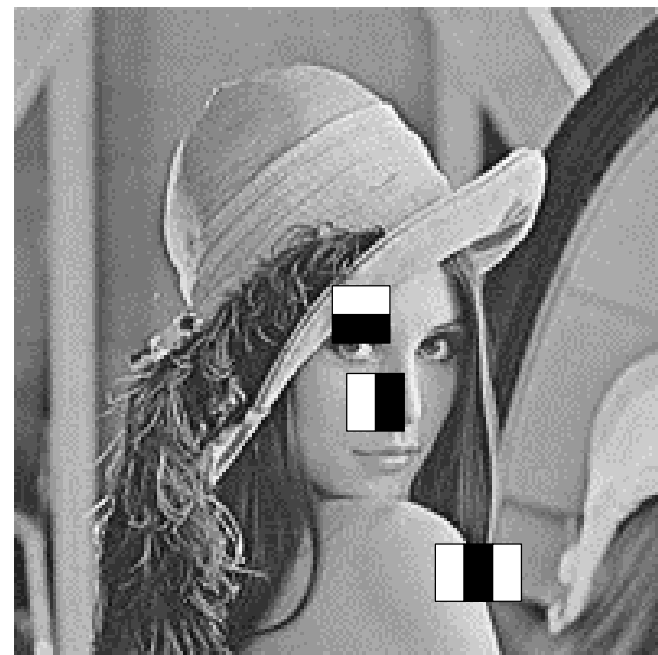


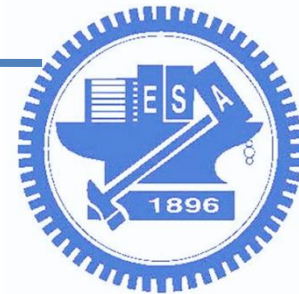
3. Center-surround features



Find features

- Pick a scale (ex: 24x24 pixels) for the feature
- Slide it across the image
- Compute the average pixel values under the white area and the black area
- If the difference between the areas is above some threshold, the feature matches





Find features

1. Calculate the average of white/black pixel
2. Calculate the difference

| | | | |
|---|---|---|---|
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |

image

| | | | |
|---|---|---|---|
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |

Edge feature

$$\Delta = \text{black} - \text{white} = 1$$

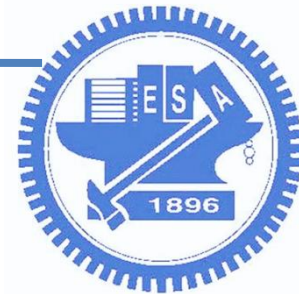
| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.2 | 0.6 | 0.8 |
| 0.2 | 0.3 | 0.8 | 0.6 |
| 0.2 | 0.1 | 0.6 | 0.8 |
| 0.2 | 0.1 | 0.8 | 0.9 |

image

| | | | |
|---|---|---|---|
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |
| 0 | 0 | 1 | 1 |

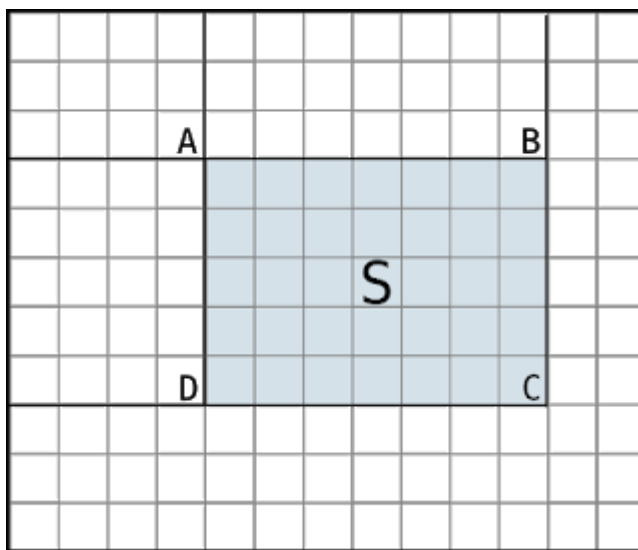
Edge feature

$$\Delta = \frac{0.6 + 0.8 + \dots}{8} - \frac{0.1 + 0.2 + \dots}{8} \\ = 0.7375 - 0.175 = 0.56$$

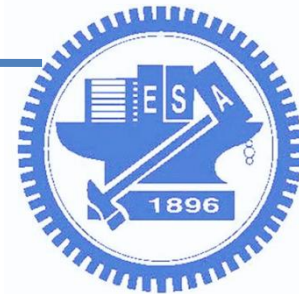


Integral Image

- a quick and effective way of calculating the sum of values (pixel values) of a rectangular subset of a grid
- It can also be used for calculating the average intensity within a given image.



$$\text{Sum} = \text{Value}(C) - \text{Value}(B) - \text{Value}(D) + \text{Value}(A)$$



Integral Image

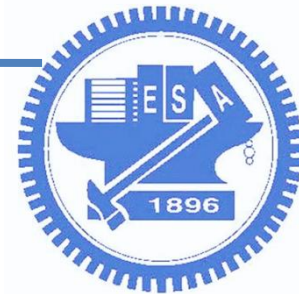
| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.2 | 0.6 | 0.8 |
| 0.2 | 0.3 | 0.8 | 0.6 |
| 0.2 | 0.1 | 0.6 | 0.8 |
| 0.2 | 0.1 | 0.8 | 0.9 |

Original image



| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.3 | 0.9 | 0.8 |
| 0.3 | 0.8 | 2.2 | 3.6 |
| 0.5 | 1.1 | 3.1 | 5.3 |
| 0.7 | 1.4 | 4.2 | 7.3 |

integral image

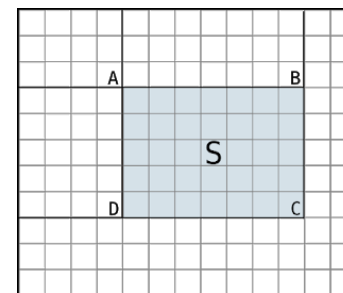


Integral Image

| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.2 | 0.6 | 0.8 |
| 0.2 | 0.3 | 0.8 | 0.6 |
| 0.2 | 0.1 | 0.6 | 0.8 |
| 0.2 | 0.1 | 0.8 | 0.9 |



| | | | |
|-------------------------|-----|-------------------------|-----|
| 0.1 _A | 0.3 | 0.9 _B | 0.8 |
| 0.3 | 0.8 | 2.2 | 3.6 |
| 0.5 _D | 1.1 | 3.1 _C | 5.3 |
| 0.7 | 1.4 | 4.2 | 7.3 |



Calculate the are summation

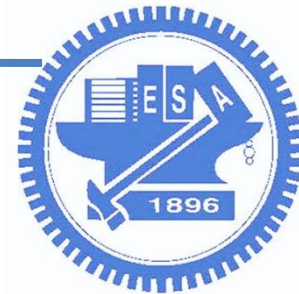
| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.3 | 0.9 | 0.8 |
| 0.3 | 0.8 | 2.2 | 3.6 |
| 0.5 | 1.1 | 3.1 | 5.3 |
| 0.7 | 1.4 | 4.2 | 7.3 |

| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.3 | 0.9 | 0.8 |
| 0.3 | 0.8 | 2.2 | 3.6 |
| 0.5 | 1.1 | 3.1 | 5.3 |
| 0.7 | 1.4 | 4.2 | 7.3 |

| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.3 | 0.9 | 0.8 |
| 0.3 | 0.8 | 2.2 | 3.6 |
| 0.5 | 1.1 | 3.1 | 5.3 |
| 0.7 | 1.4 | 4.2 | 7.3 |

| | | | |
|-----|-----|-----|-----|
| 0.1 | 0.3 | 0.9 | 0.8 |
| 0.3 | 0.8 | 2.2 | 3.6 |
| 0.5 | 1.1 | 3.1 | 5.3 |
| 0.7 | 1.4 | 4.2 | 7.3 |

$$\text{Sum} = \text{Value}(C) - \text{Value}(B) + \text{Value}(A) - \text{Value}(D)$$



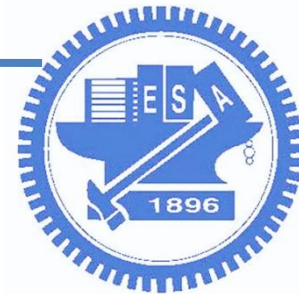
Discussion

- How to calculate the area summation by integral image?
 - 1. Write down the value of integral image
 - 2. $\text{Sum} = \text{Value}(C) - \text{Value}(B) + \text{Value}(A) - \text{Value}(D) = \text{?????}$

| | | | | |
|-----|-----|-----|-----|-----|
| 0.1 | 0.2 | 0.3 | 0.5 | 0.8 |
| 0.2 | 0.2 | 0.3 | 0.7 | 0.9 |
| 0.1 | 0.2 | 0.4 | 0.7 | 0.9 |
| 0.3 | 0.1 | 0.2 | 0.8 | 0.2 |
| 0.1 | 0.2 | 0.4 | 0.6 | 0.1 |



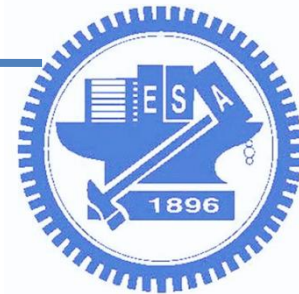
| | | | | |
|--|---|---|---|--|
| | | | | |
| | | | | |
| | ? | ? | ? | |
| | ! | ! | ! | |
| | | | | |



AdaBoost

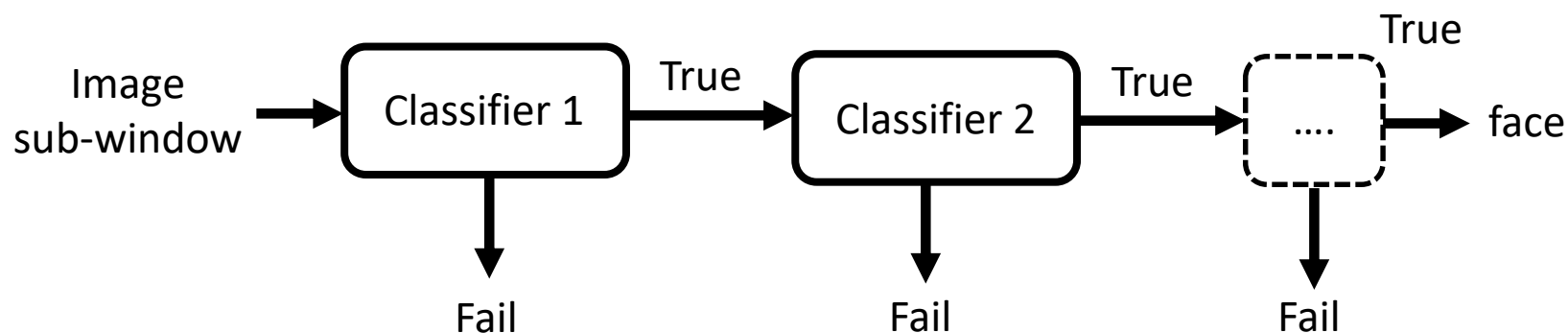
- Adaptive Boosting
 - Try out multiple weak classifiers over several rounds
 - Select the best weak classifier in each round and combining the best weak classifiers to create a strong classifier

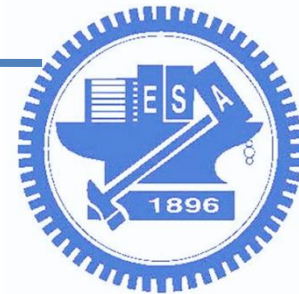
| Data point | Classifier 1 | Classifier 2 | Classifier 3 | ... |
|------------|--------------|--------------|--------------|-----|
| P_1 | Pass | Fail | Fail | ... |
| P_2 | Pass | Pass | Pass | ... |
| P_3 | Fail | Pass | Pass | ... |
| ... | ... | ... | ... | ... |



Cascades

- Haar cascades consists of a series of weak classifiers
 - barely better than 50% correct
 - If an area passes a single classifier, go to the next classifier; otherwise, area doesn't match



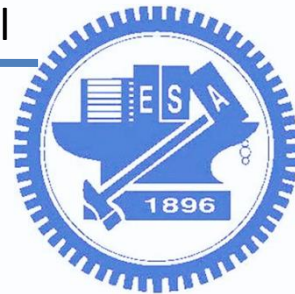


Recall the code

```
while True:
    # Capture frame-by-frame
    ret, frame = cap.read()

    gray = cv2.cvtColor(frame, cv2.COLOR_BGR2GRAY)

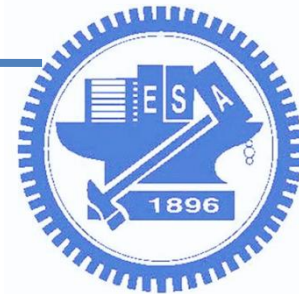
    faces = faceCascade.detectMultiScale(
        gray,
        scaleFactor=1.1,
        minNeighbors=5,
        minSize=(30, 30),
        flags=HAAR_FLAGS
    )
```

Related parameters

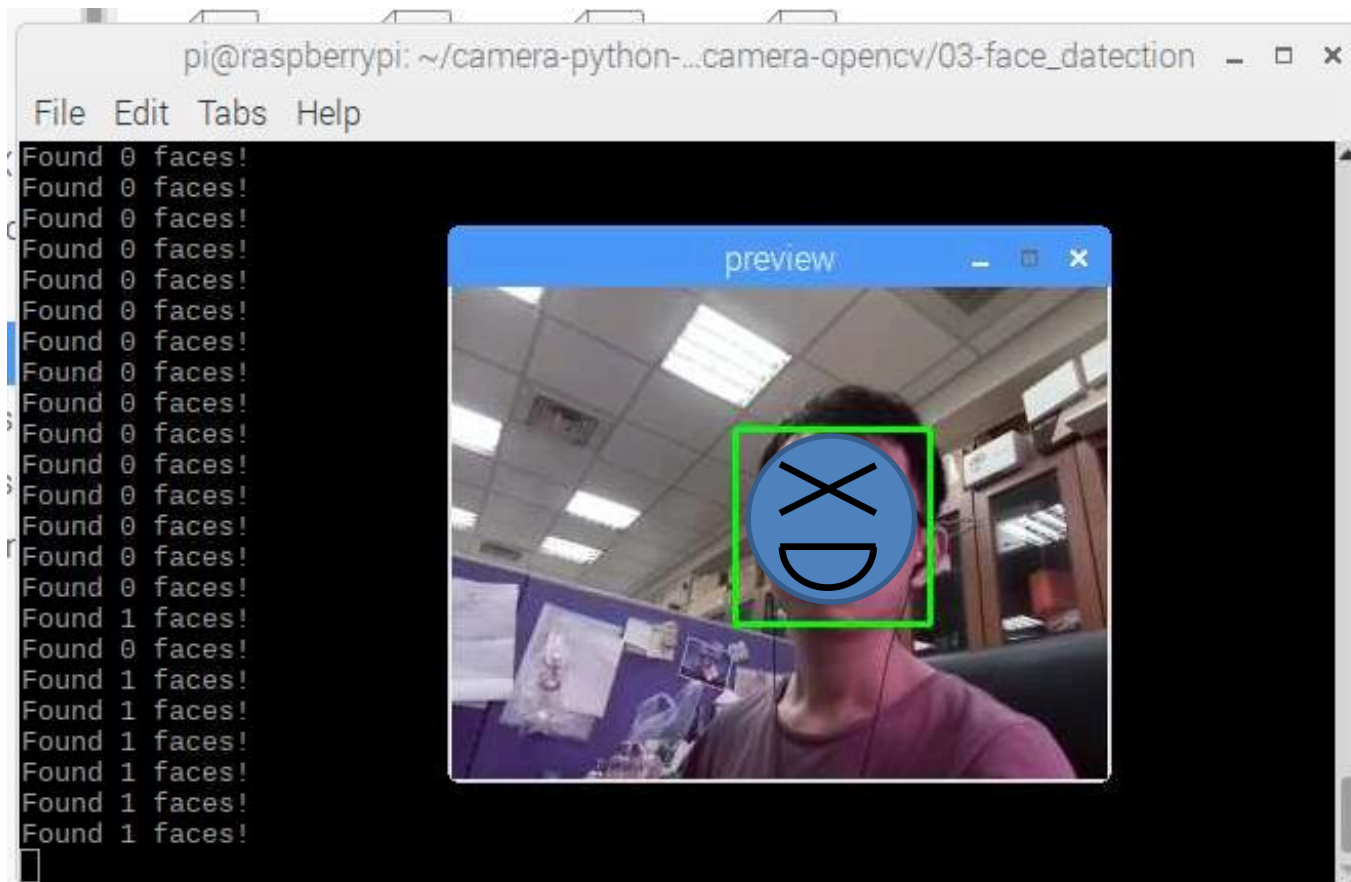
- **cascade** – Haar classifier cascade (OpenCV 1.x API only). It can be loaded from XML or YAML file using `Load()`. When the cascade is not needed anymore, release it using `cvReleaseHaarClassifierCascade(&cascade)`.
- **image** – Matrix of the type `CV_8U` containing an image where objects are detected.
- **objects** – Vector of rectangles where each rectangle contains the detected object.
- **scaleFactor** – Parameter specifying how much the image size is reduced at each image scale.
- **minNeighbors** – Parameter specifying how many neighbors each candidate rectangle should have to retain it.
- **flags** – Parameter with the same meaning for an old cascade as in the function `cvHaarDetectObjects`. It is not used for a new cascade.
- **minSize** – Minimum possible object size. Objects smaller than that are ignored.
- **maxSize** – Maximum possible object size. Objects larger than that are ignored.

Try to use different **parameters**, you will get different results.



Face detection on PI

- python camera_face_detect.py
haarcascade_frontalface_default.xml





Last week

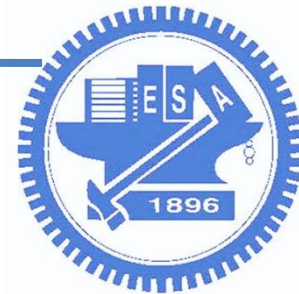
□ MJPG on PI



Hello Stream

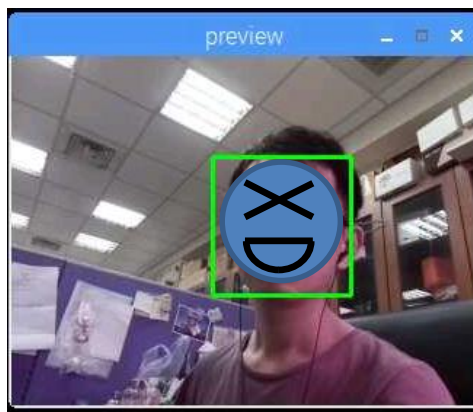


No stream? You might need: **`sudo modprobe bcm2835-v4l2`**



Quiz 1

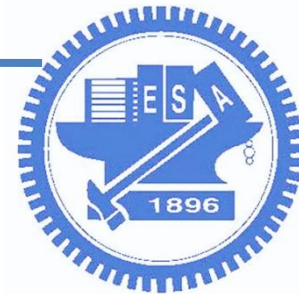
- Based on the sample code:
 - 1. MJPG on PI
 - 2. Facial detection
- Combine them together!



Hello Stream

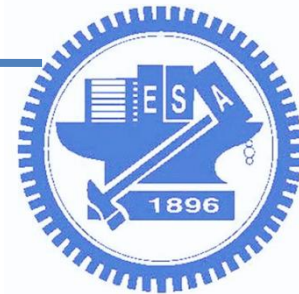


We can use browser to watch the facial detection results.



Summary

- Practice Lab (opencv example, facial detection)
- Write down the answer for discussion
 - **Discussion**
 - How to calculate the area summation by integral image?
- Write code for **Quiz 1**, then **demonstrate it to TAs**
 - **Quiz1: MJPG + facial detection**



Reference

□ Online resource

□ Facial Detection

■ <https://www.youtube.com/watch?v=sWTvK72-SPU>

□ Computer Vision - Haar-Features

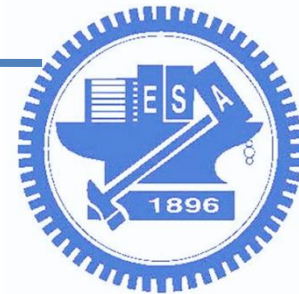
■ <https://www.youtube.com/watch?v=F5rysk51txQ>

□ Computer Vision - Integral Images

■ <https://www.youtube.com/watch?v=x41KFOFGnUE>

□ Recognition Part II: Face Detection via AdaBoost

■ https://courses.cs.washington.edu/courses/cse455/16wi/notes/15_FaceDetection.pdf



Reference

- [Viola01] Paul Viola and Michael J. Jones. Rapid Object Detection using a Boosted Cascade of Simple Features. IEEE CVPR, 2001.
 - http://research.microsoft.com/en-us/um/people/viola/Pubs/Detect/violaJones_CVPR2001.pdf
- [Lienhart02] Rainer Lienhart and Jochen Maydt. An Extended Set of Haar-like Features for Rapid Object Detection. IEEE ICIP, Vol. 1, pp. 900-903, Sep. 2002.
 - <http://www.multimedia-computing.de/mediawiki//images/5/52/MRL-TR-May02-revised-Dec02.pdf>