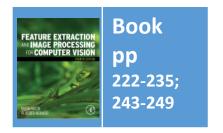
## Lecture 8 Finding Shapes

COMP3204 & COMP6223 Computer Vision

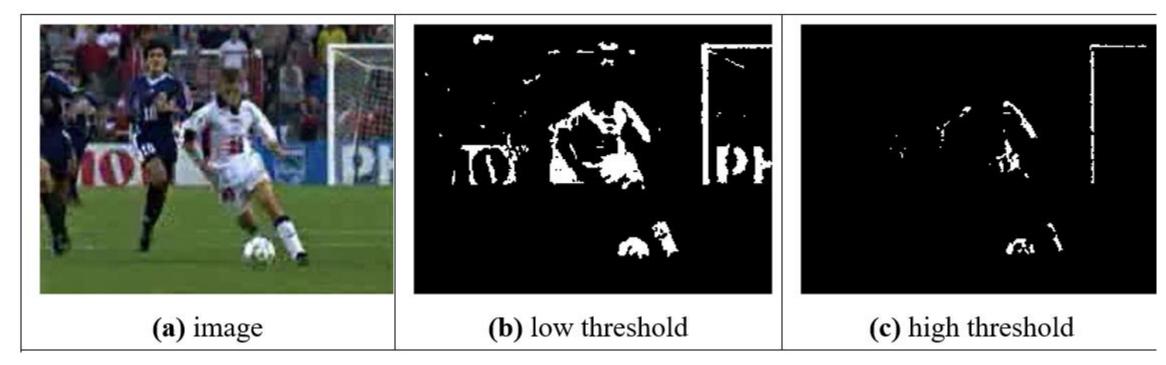
#### How can we group points to find shapes?



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#### Feature extraction by thresholding



Conclusion: we need shape!



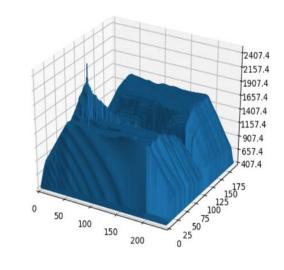
#### Template Matching

- Intuitively simple
- Correlation and convolution
- Implementation via Fourier
- Relationship with matched filter, viz: optimality







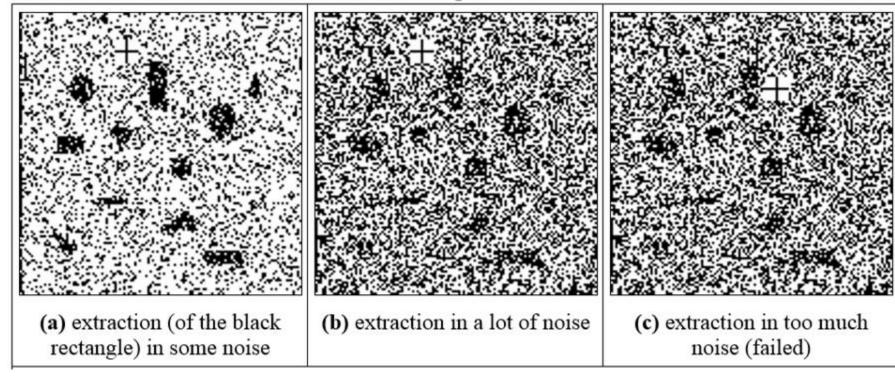


image

template

accumulator space

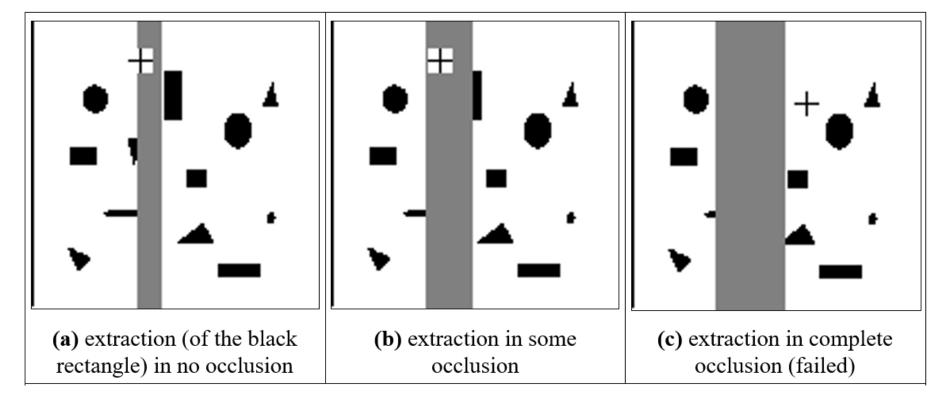
## Template matching in noisy images







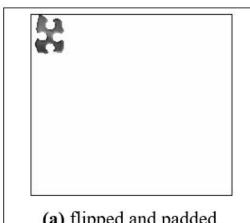
### Template matching in occluded images



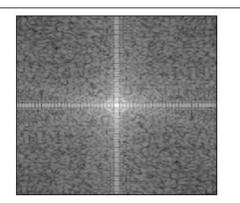




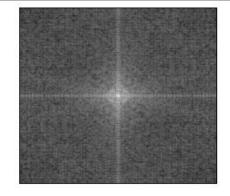
#### Encore, Monsieur Fourier!



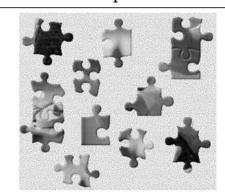
(a) flipped and padded template



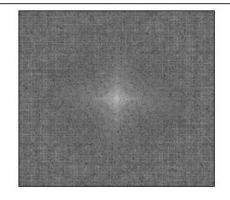
**(b)** Fourier transform of template



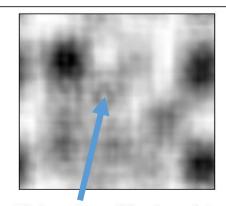
(c) multiplied transforms



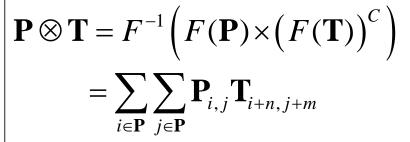
(d) image



(e) Fourier transform of image



(f) location of the template

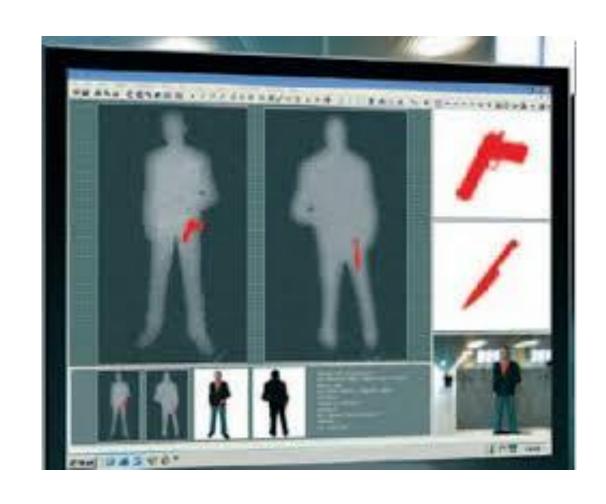


No sliding of templates here;

cost is Fourier Transform plus multiplication



## Applying template matching



### Applying SIFT in ear biometrics



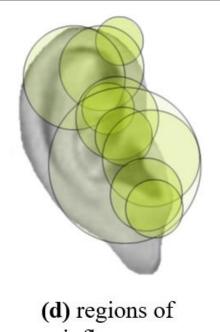
(a) detected SIFT points



(b) one feature



(c) same feature as (b) in a different ear

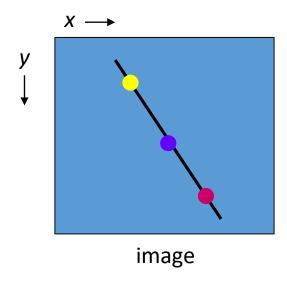


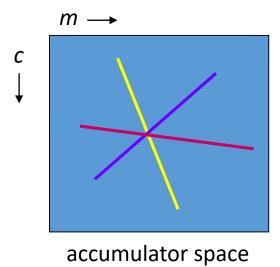
influence



### Hough Transform

- Performance equivalent to template matching, but faster
- A line is points x,y gradient m intercept c  $y = m \times x + c$
- and is points m,c gradient -x intercept y  $c = -x \times m + y$







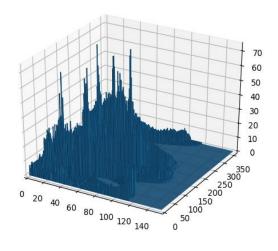


In maths it's the principle of duality

## Applying the Hough transform for lines







image

detected lines

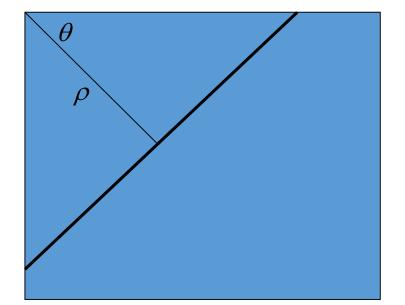
accumulator space





#### Hough Transform for Lines ... problems

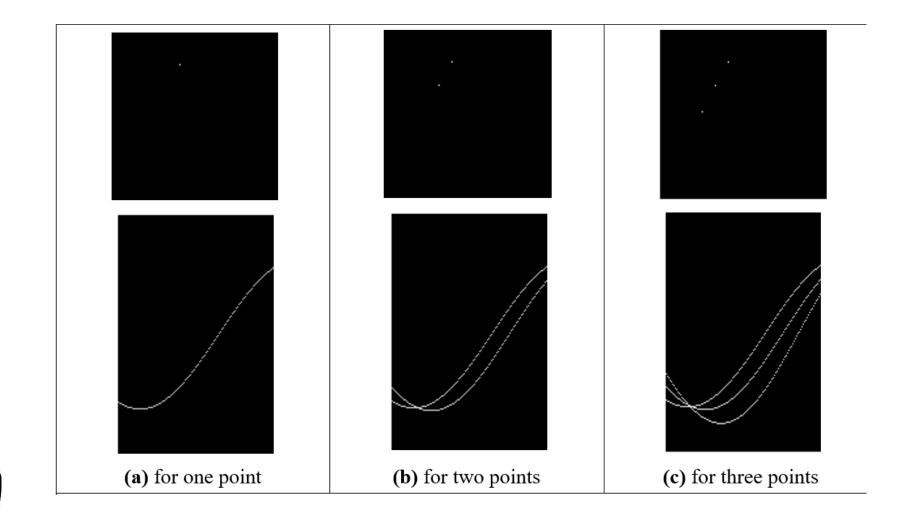
- *m,c* tend to infinity
- Change the parameterisation
- Use foot of normal  $\rho = x \cos \theta + y \sin \theta$
- Gives polar HT for lines







# Images and the accumulator space of the polar Hough transform





## Applying the Hough transform

