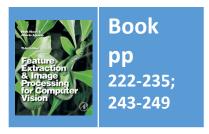
### Lecture 8 Finding Shapes

COMP3204 & COMP6223 Computer Vision

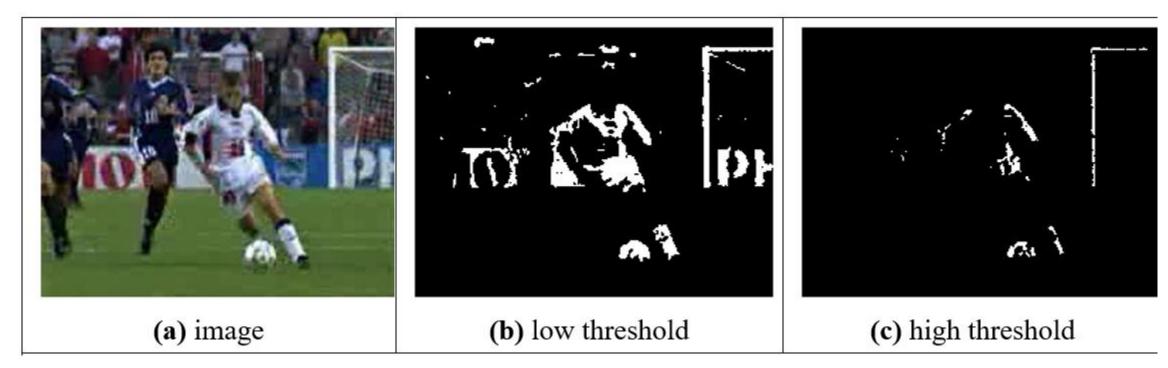
How can we group points to find shapes?



Department of Electronics and Computer Science



#### Feature extraction by thresholding

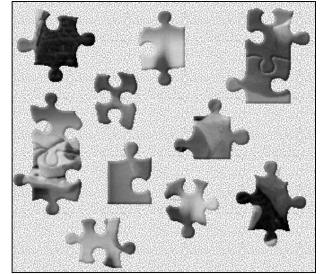


Conclusion: we need shape!



#### Template Matching

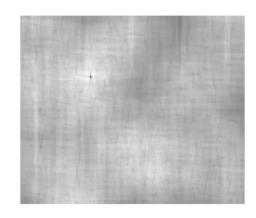
- Intuitively simple
- Correlation and convolution
- Implementation via Fourier
- Relationship with matched filter, viz: optimality







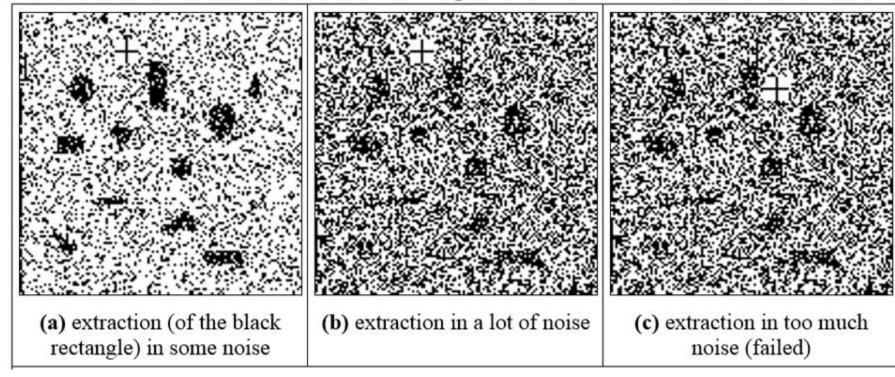
template







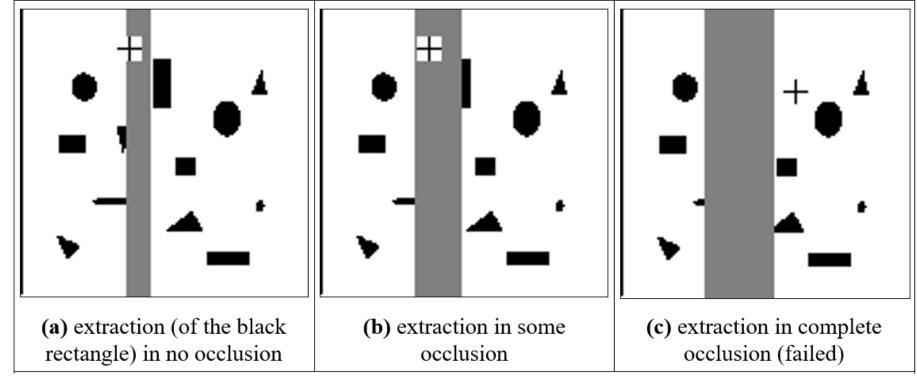
#### Template matching in noisy images





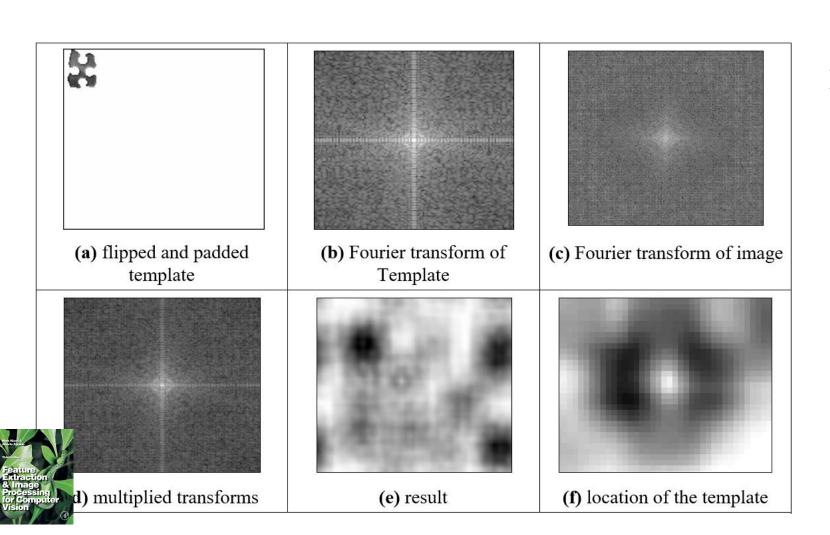


#### Template matching in occluded images





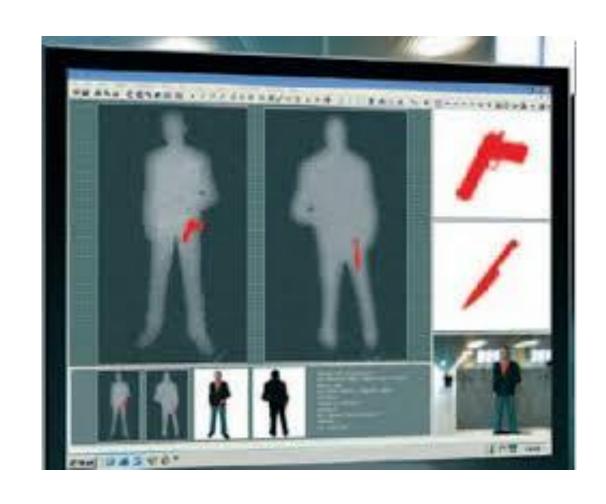
#### Encore, Monsieur Fourier!



$$\mathbf{P} \otimes \mathbf{T} = F^{-1} \Big( F(\mathbf{P}) \times \Big( F(\mathbf{T}) \Big)^C \Big)$$
$$= \sum_{i \in \mathbf{P}} \sum_{j \in \mathbf{P}} \mathbf{P}_{i,j} \mathbf{T}_{i+n,j+m}$$

No sliding of templates here; cost is Fourier Transform plus multiplication

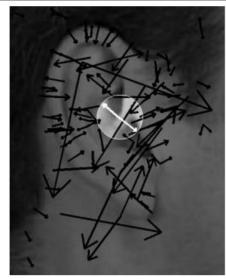
#### Applying template matching



#### Applying SIFT in ear biometrics



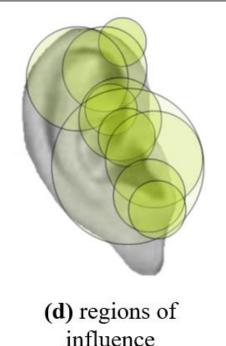
(a) detected SIFT points



(b) one feature



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

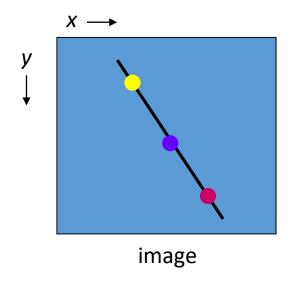


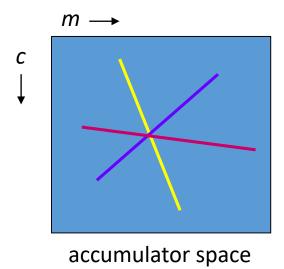
influence



#### 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  $c = -x \times m + y$

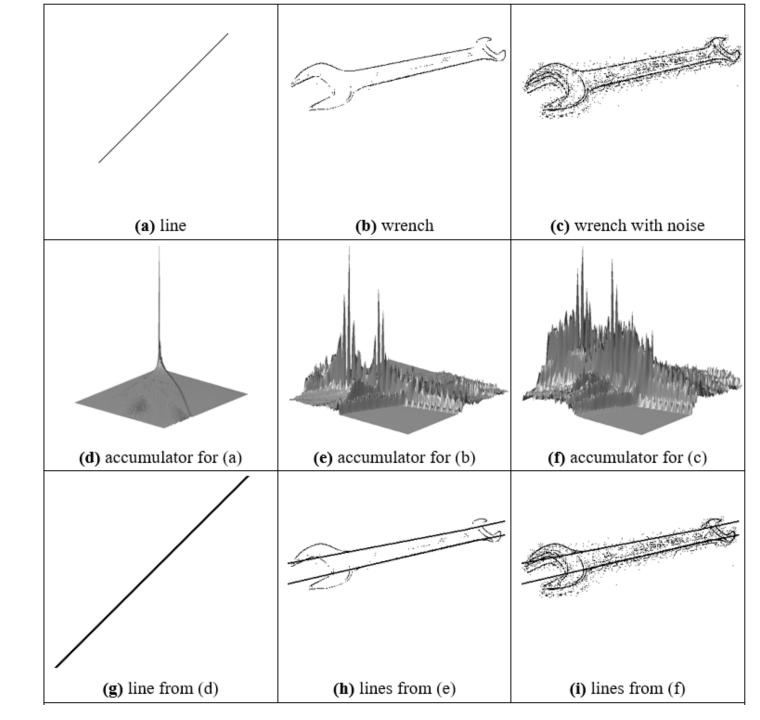






In maths it's the principle of duality

## Applying the Hough transform for lines





#### 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

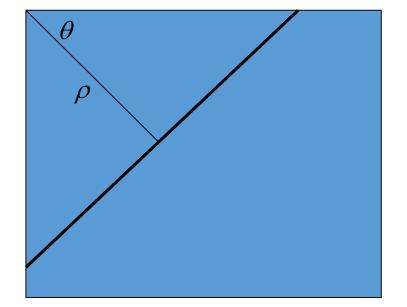
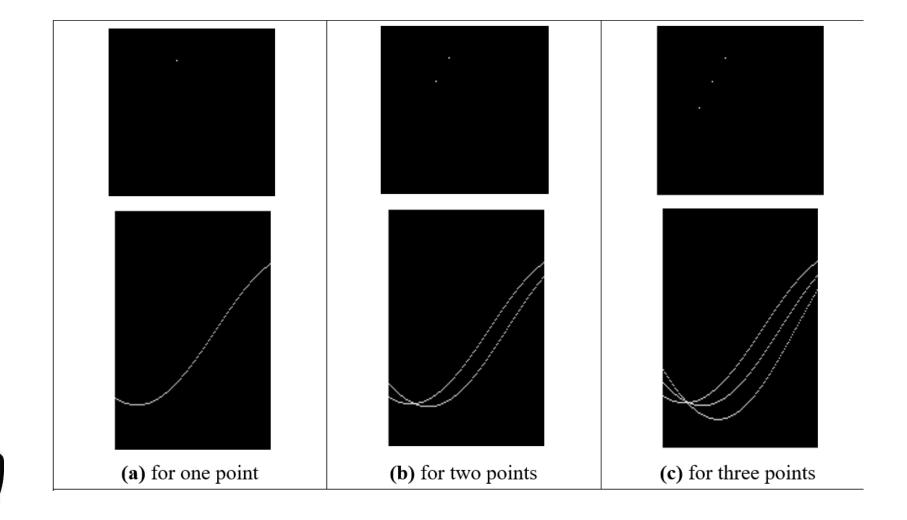




Image containing line

# Images and the accumulator space of the polar Hough transform





#### Applying the Hough transform

