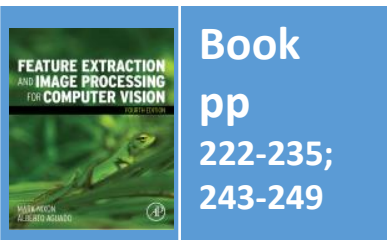


Lecture 8 Finding Shapes

COMP3204 & COMP6223 Computer Vision

How can we group points to find shapes?



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Feature extraction by thresholding



(a) image



(b) low threshold



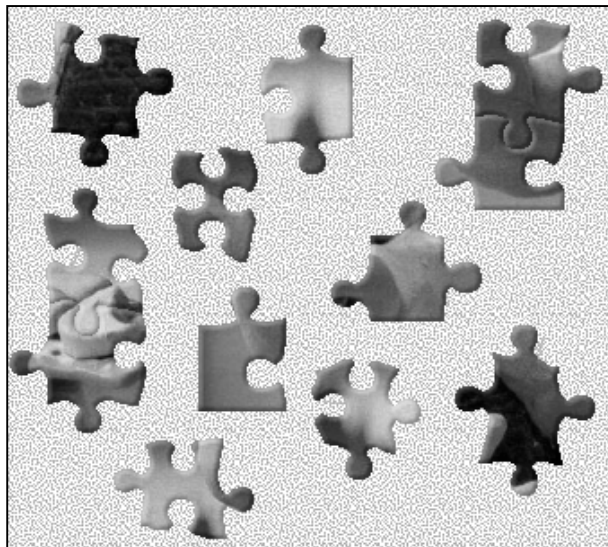
(c) high threshold

- Conclusion: we need **shape**!



Template Matching

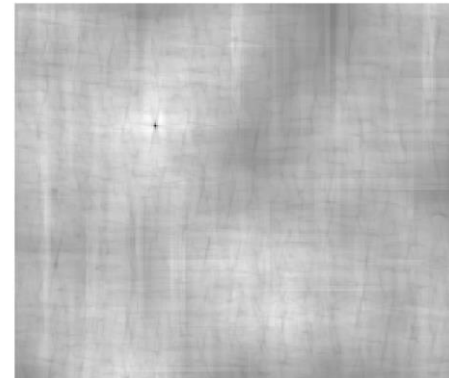
- Intuitively **simple**
- **Correlation** and convolution
- Implementation via **Fourier**
- Relationship with matched filter, viz: **optimality**



image



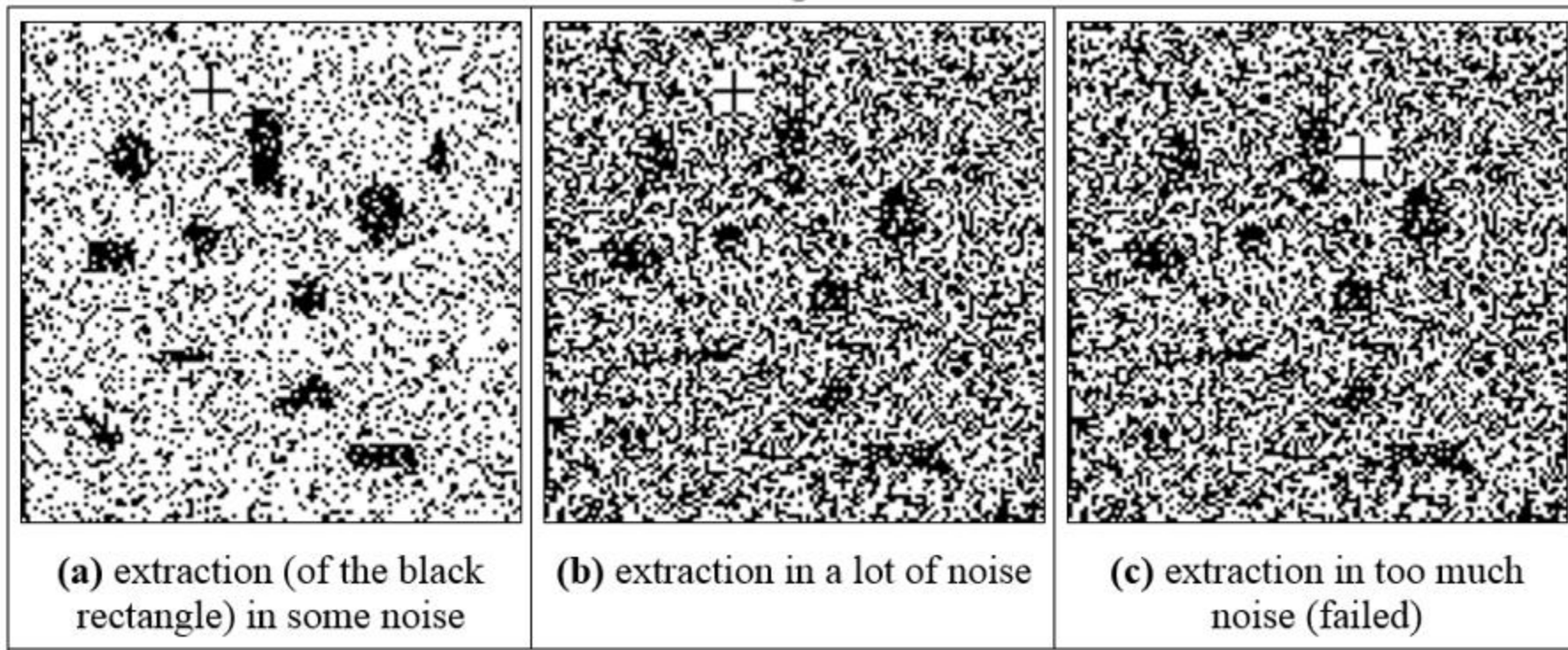
template



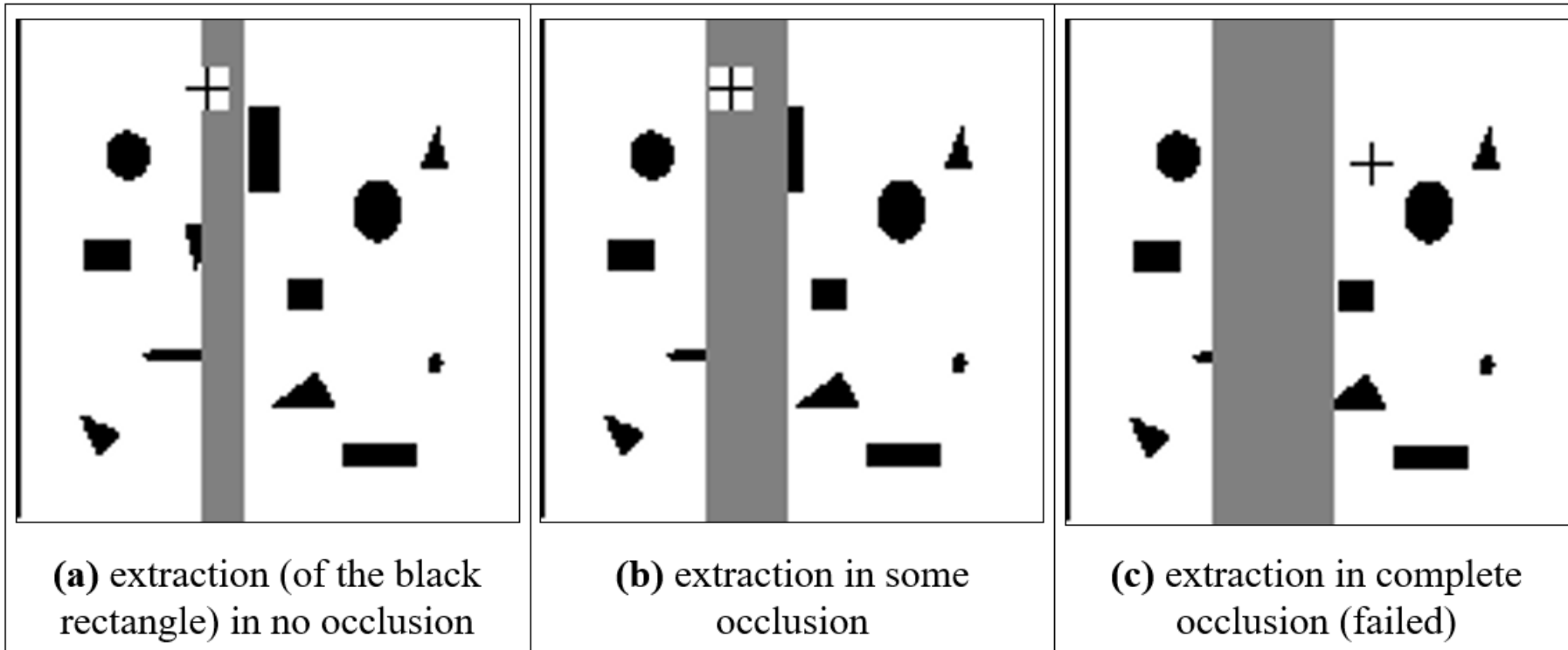
accumulator space



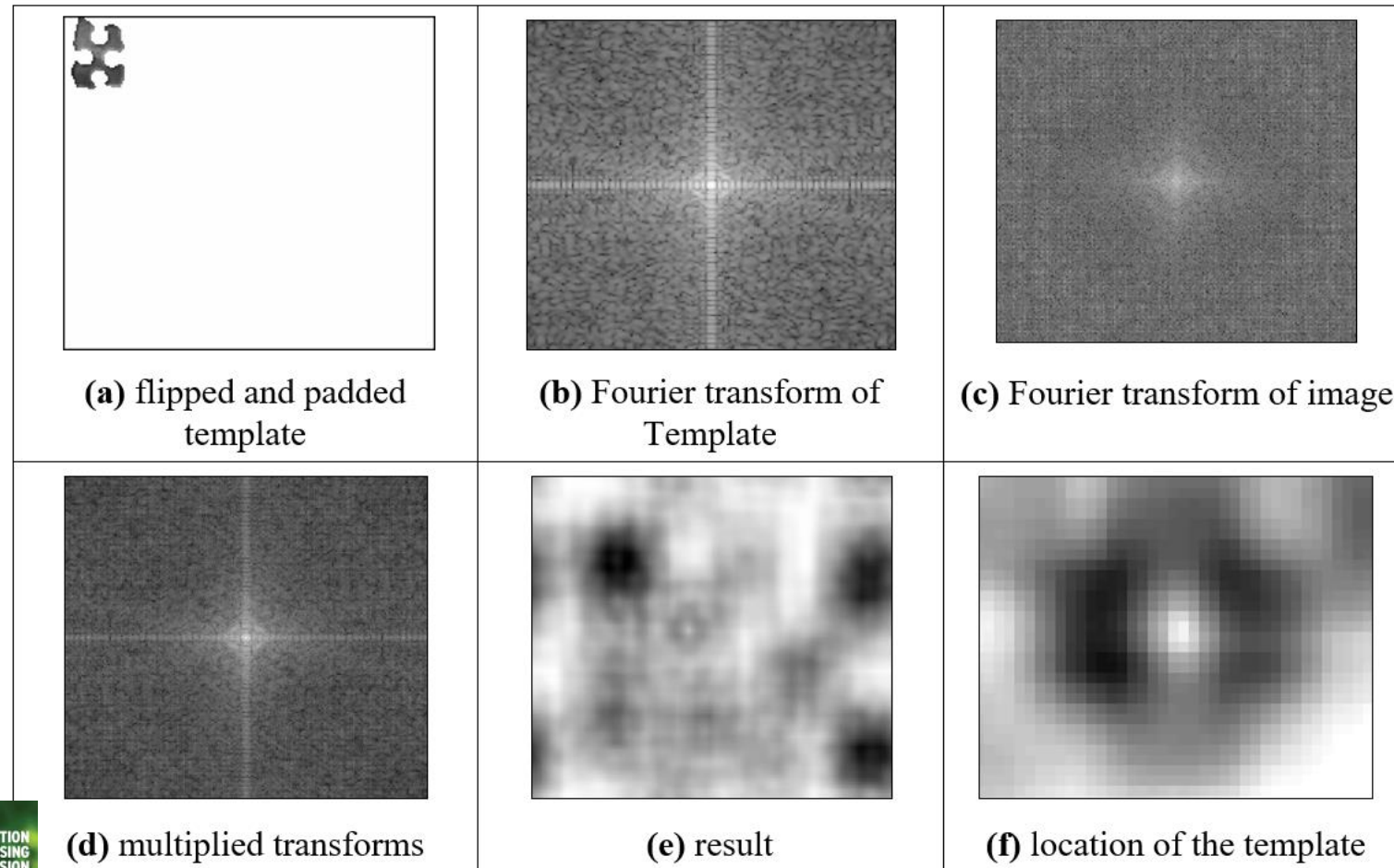
Template matching in noisy images



Template matching in occluded images



Encore, Monsieur Fourier!



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

No sliding of
templates
here;

cost is Fourier
Transform plus
multiplication

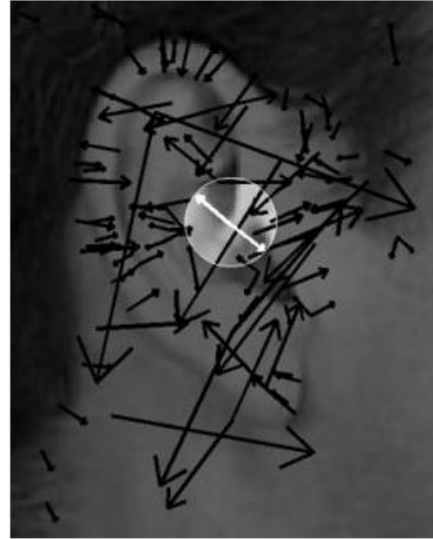
Applying template matching



Applying SIFT in ear biometrics



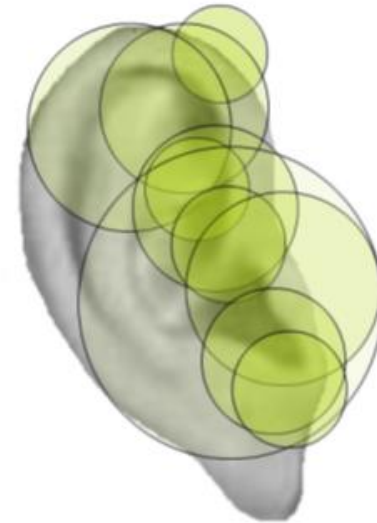
(a) detected SIFT
points



(b) one feature



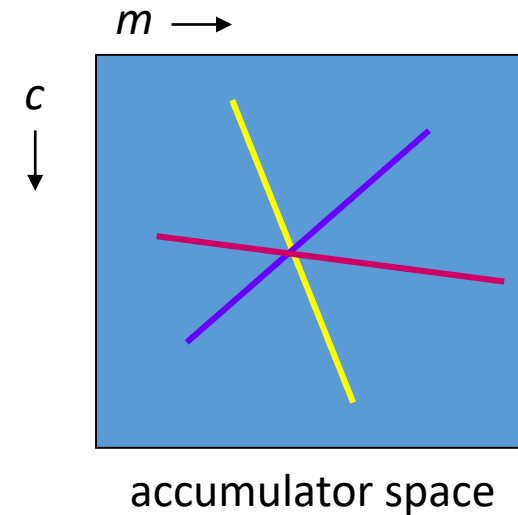
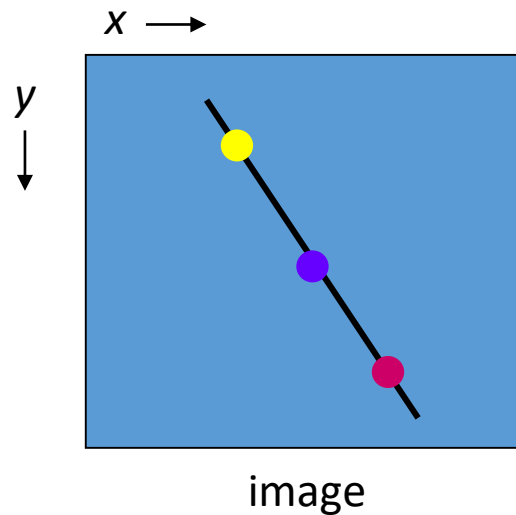
(c) same feature as (b)
in a different ear



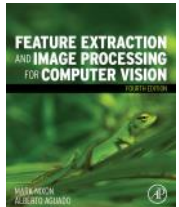
(d) regions of influence

Hough Transform

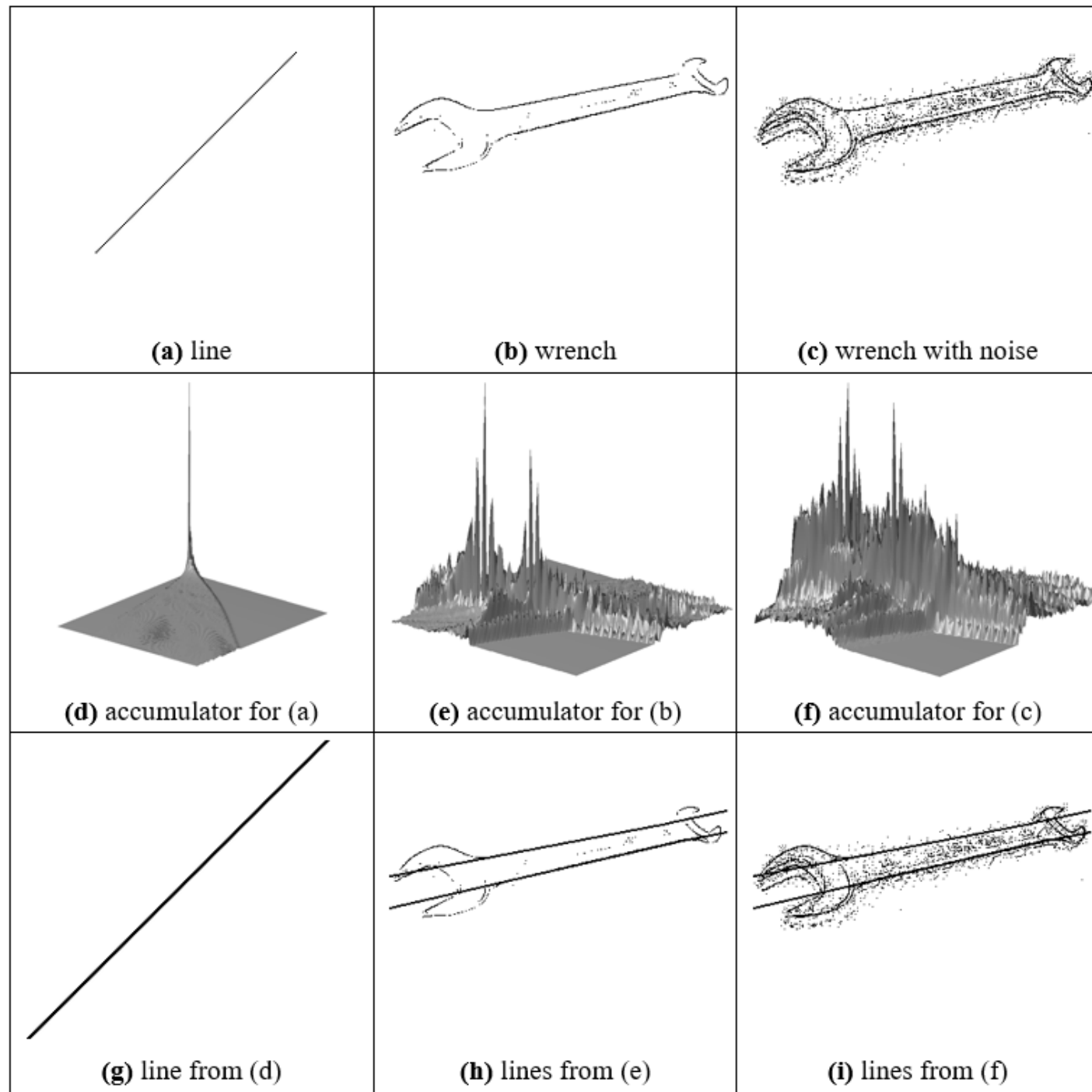
- **Performance** equivalent to template matching, but **faster**
- A line is points x,y gradient m intercept c $y = m \times x + c$
- **and** is points m,c gradient $-x$ intercept y $c = -x \times m + y$



- 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

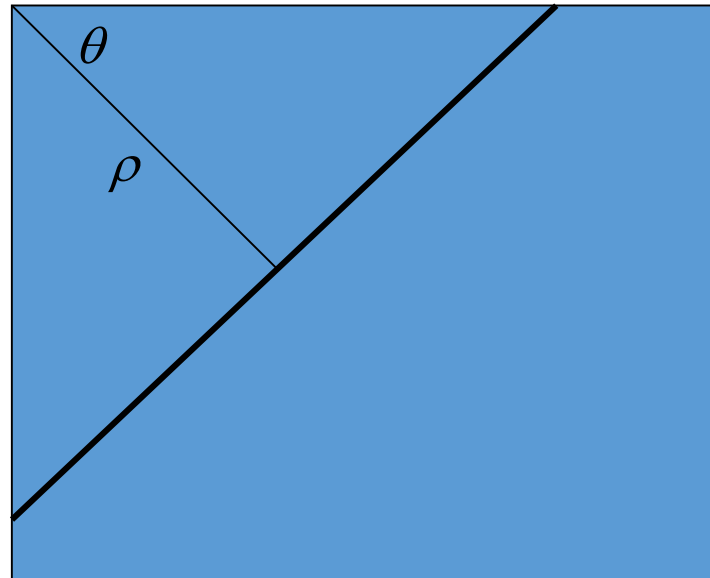
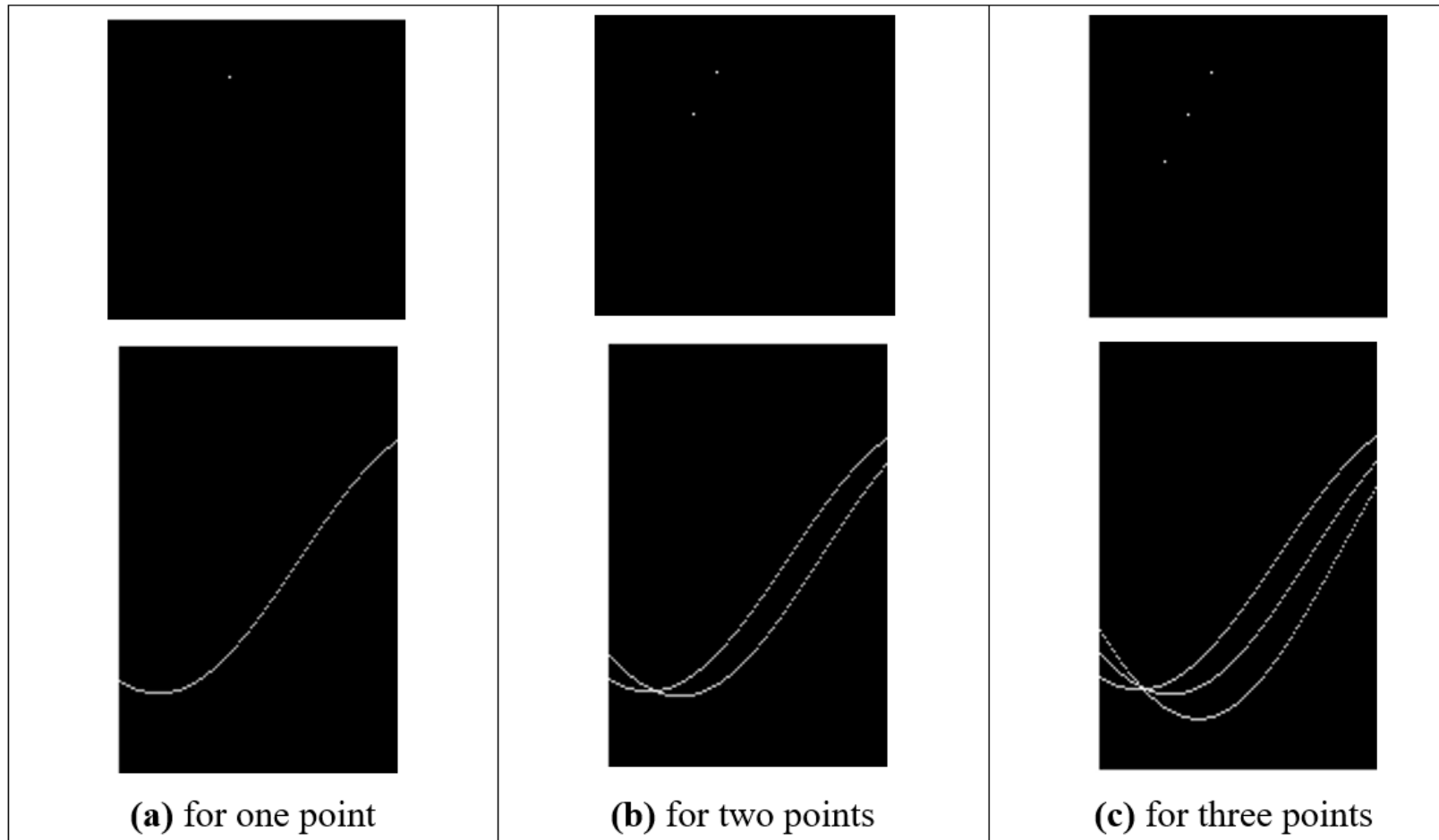


Image containing line



Images and the accumulator space of the polar Hough transform



Applying the Hough transform

