

15-381

**ARTIFICIAL
INTELLIGENCE**

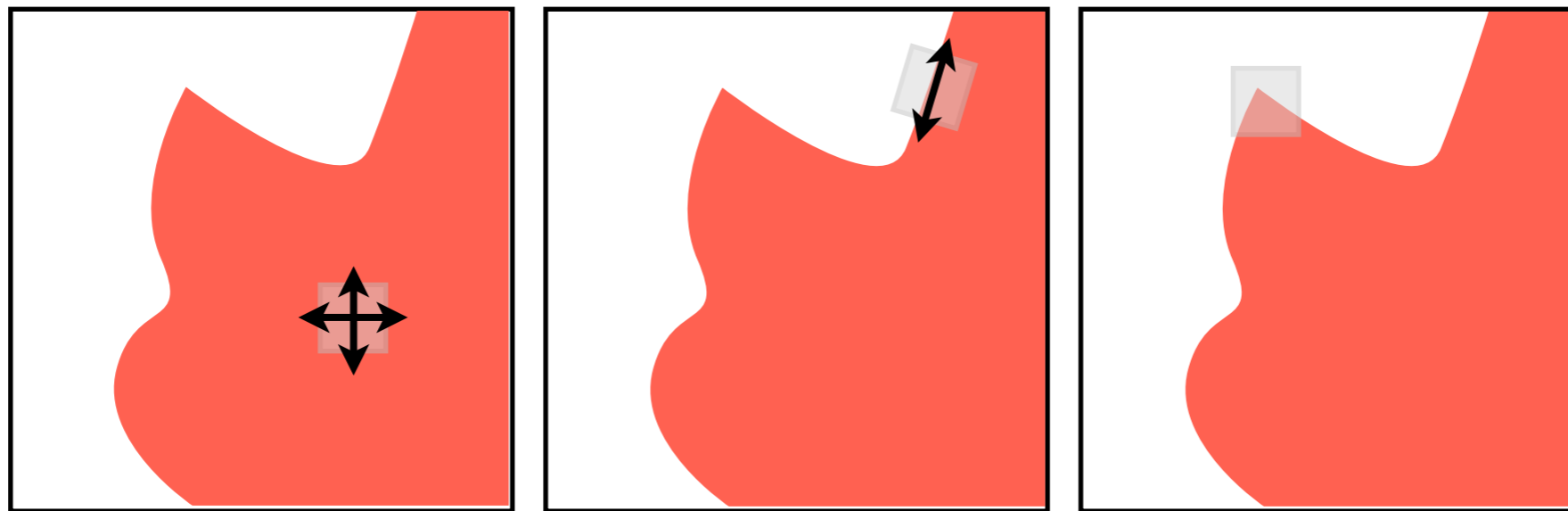
**LECTURE 16:
VISION IV: RECOGNITION**

FALL 2010

Several slides taken from: Fei-Fei Li, Josef Sivic, Kristen Grauman

CORNERNESS

- **IDEA:** SHIFTING PATCH IN ANY DIRECTION SHOULD PRODUCE LARGE CHANGE

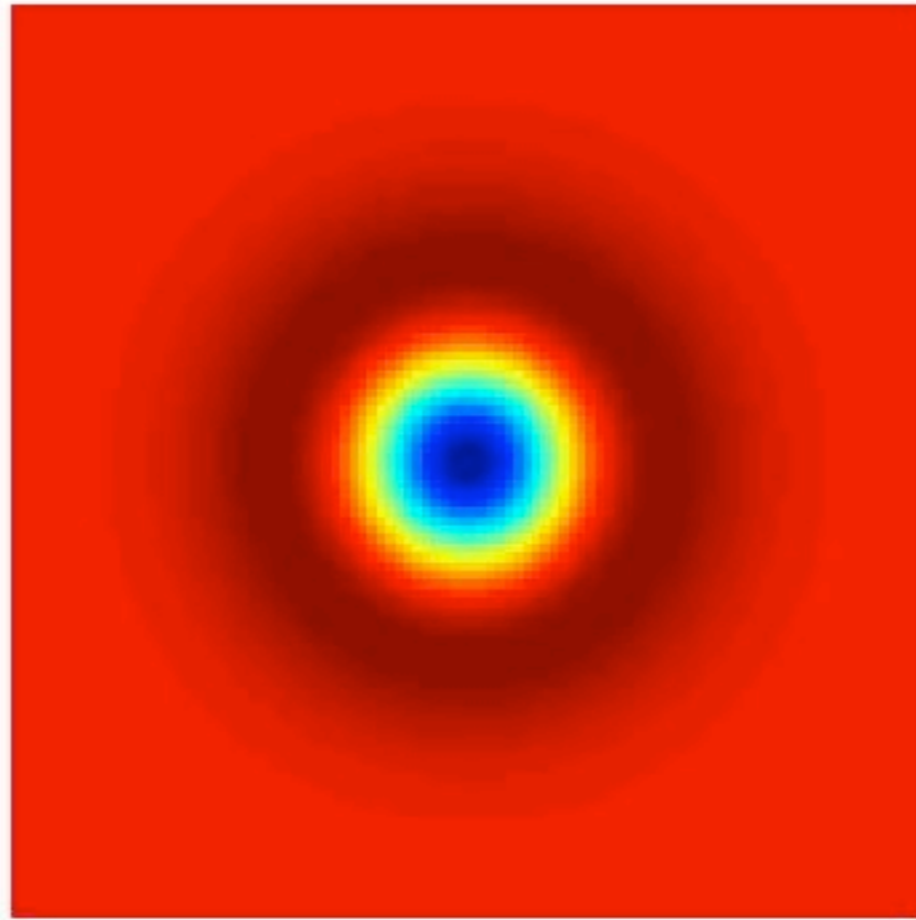


Flat areas
No change in
all directions

Edges
No change
along edge

Corners
Change in
all direction

DIFFERENCE OF GAUSSIAN



$$D(x, y, \sigma) = (G(x, y, k\sigma) - G(x, y, \sigma)) * I(x, y)$$

$$D(x, y, \sigma) = L(x, y, k\sigma) - L(x, y, \sigma)$$

FAST APPROXIMATION TO THE LAPLACIAN OF GAUSSIAN

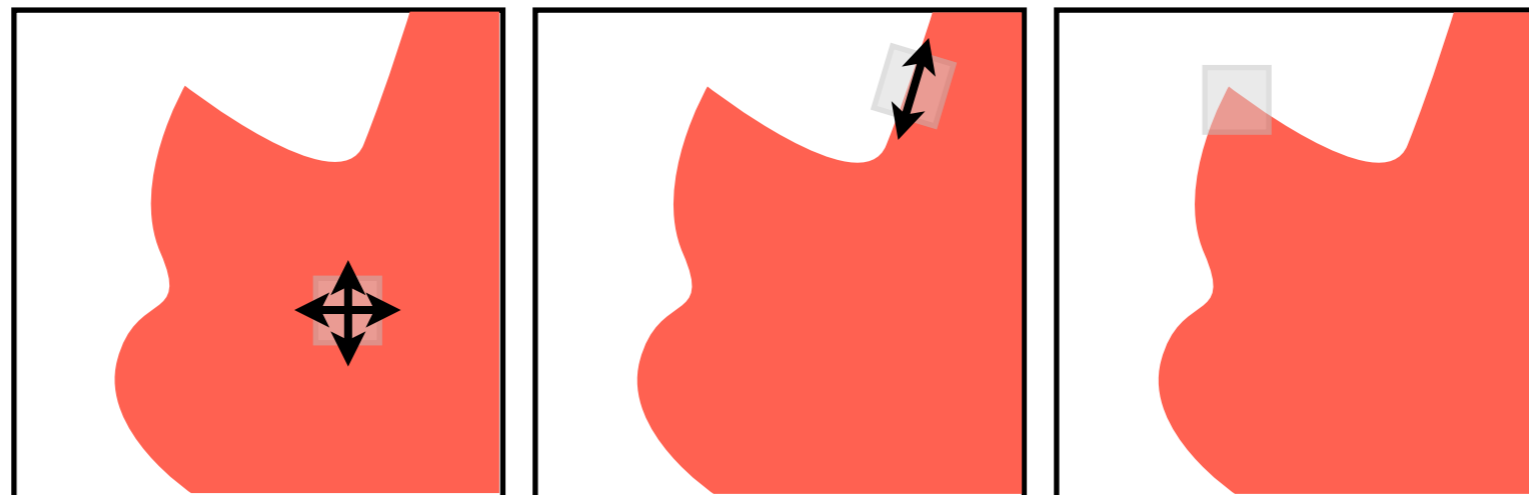
EDGE ELIMINATION

Hessian Matrix

Change in D
(computer by taking adjacent pixel differences)

$$\mathbf{H} = \begin{bmatrix} D_{xx} & D_{xy} \\ D_{xy} & D_{yy} \end{bmatrix}$$

The eigenvalues of \mathbf{H} are proportional to the principal curvature of D



Flat areas

Edges

Corners

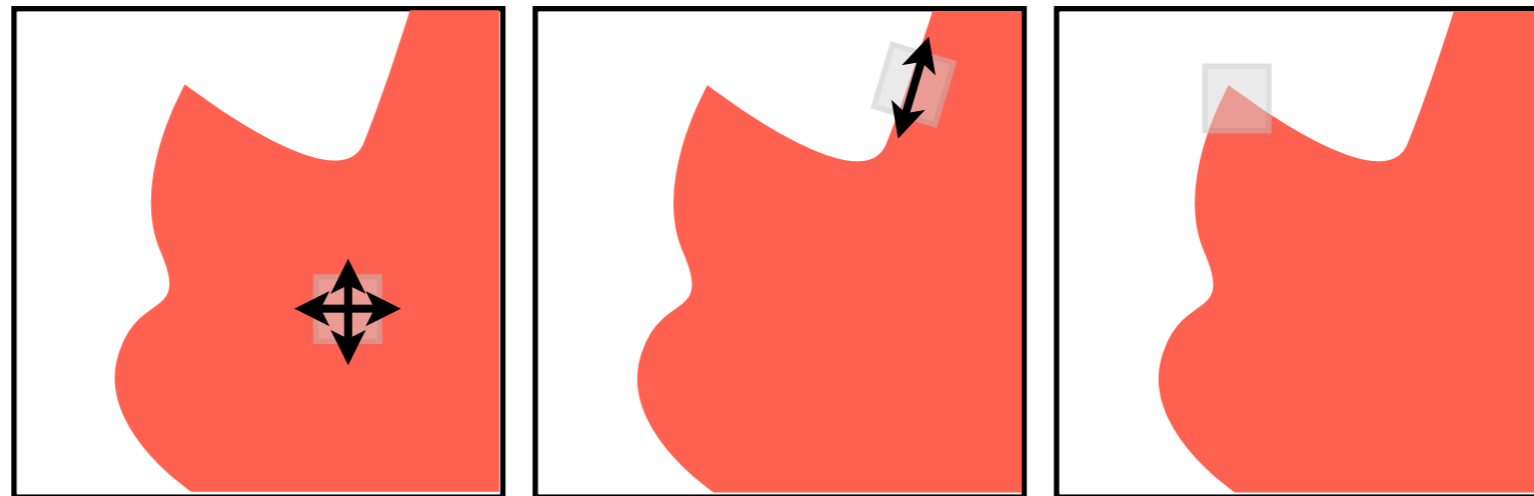
CORNERNESS

HARRIS-STEPHENS CORNER

Hessian Matrix

$$\mathbf{H} = \begin{bmatrix} D_{xx} & D_{xy} \\ D_{xy} & D_{yy} \end{bmatrix}$$

Eigenvalues of \mathbf{H} : α, β

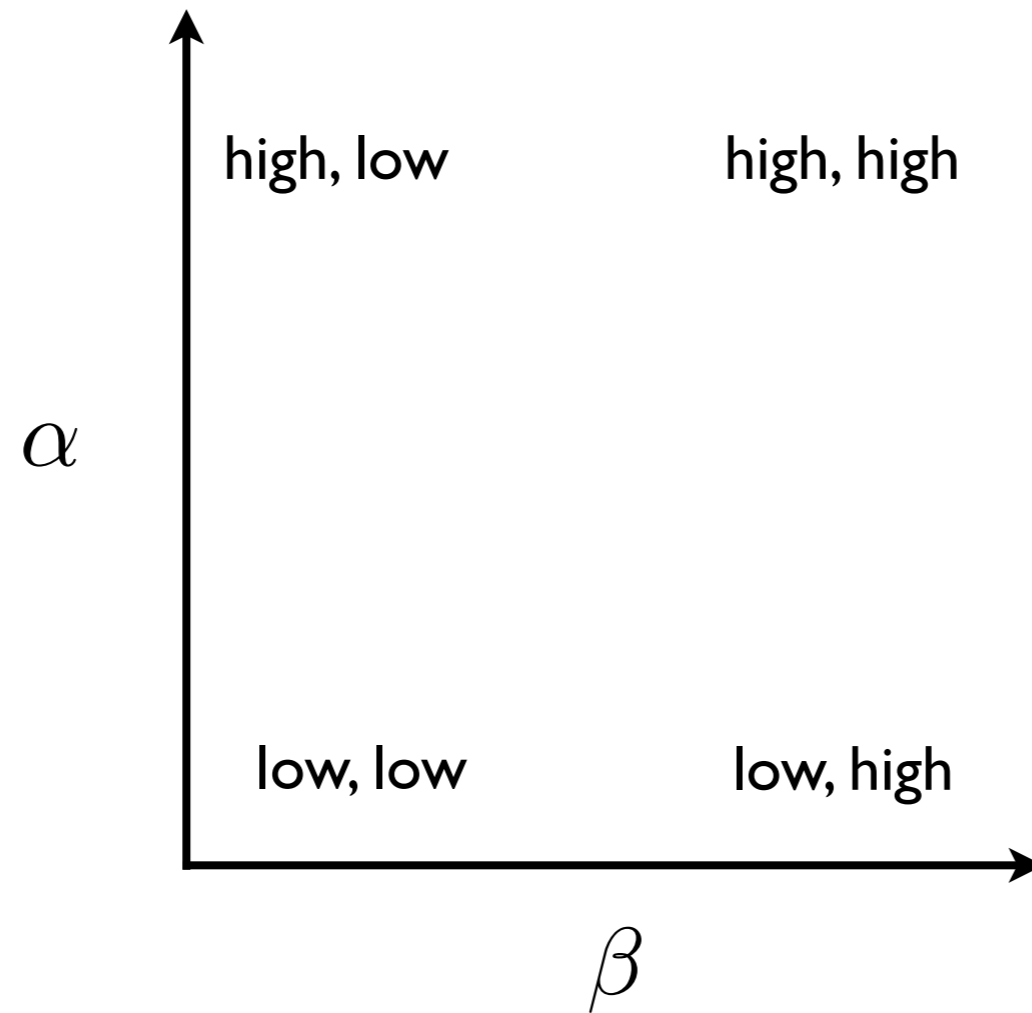


two small
eigenvalues

one large
one small
eigenvalues

two large
eigenvalues

SELECTING CORNERS



“A COMBINED CORNER AND EDGE DETECTOR”

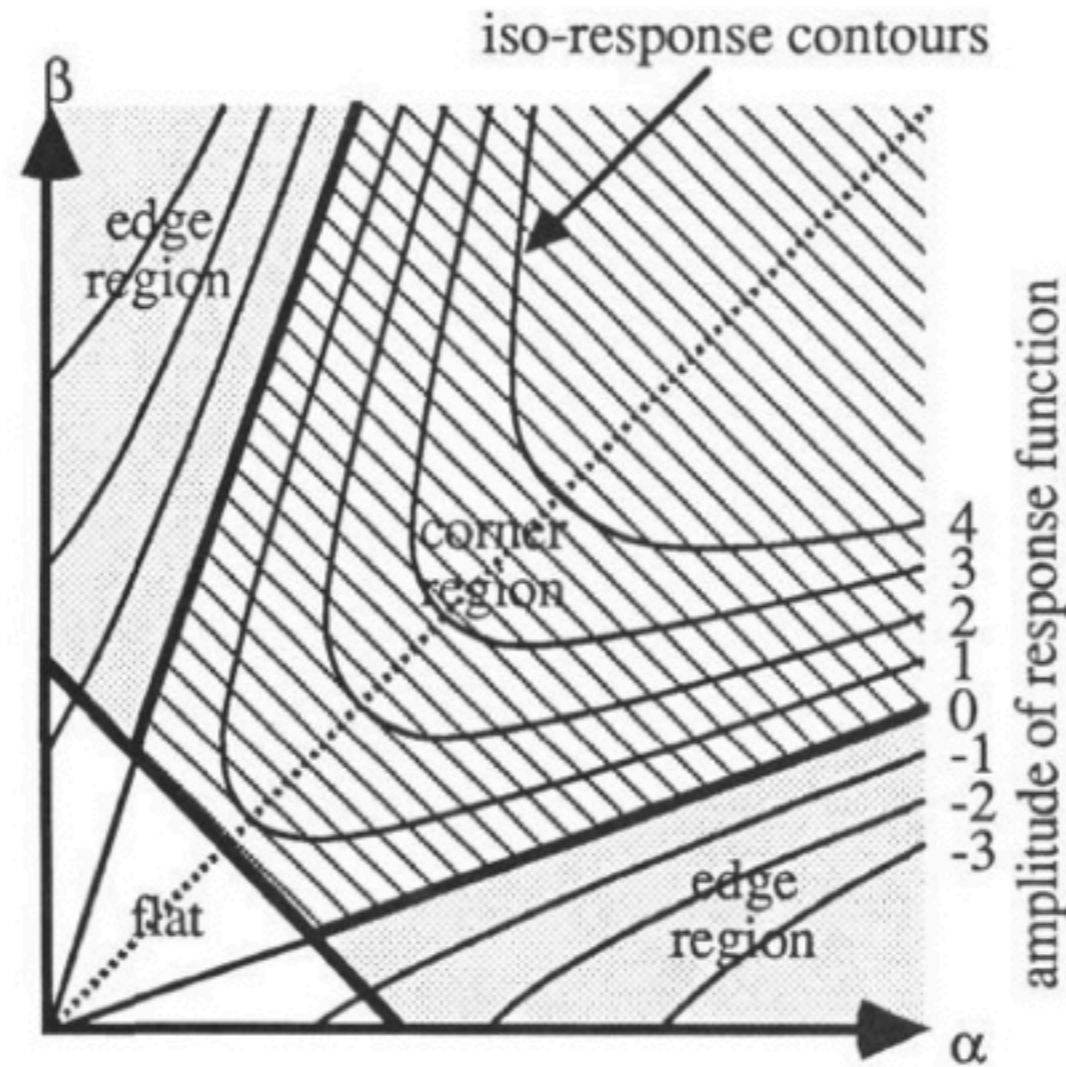


Figure 5. Auto-correlation principal curvature space—
heavy lines give corner/edge/flat classification,
fine lines are equi-response contours.

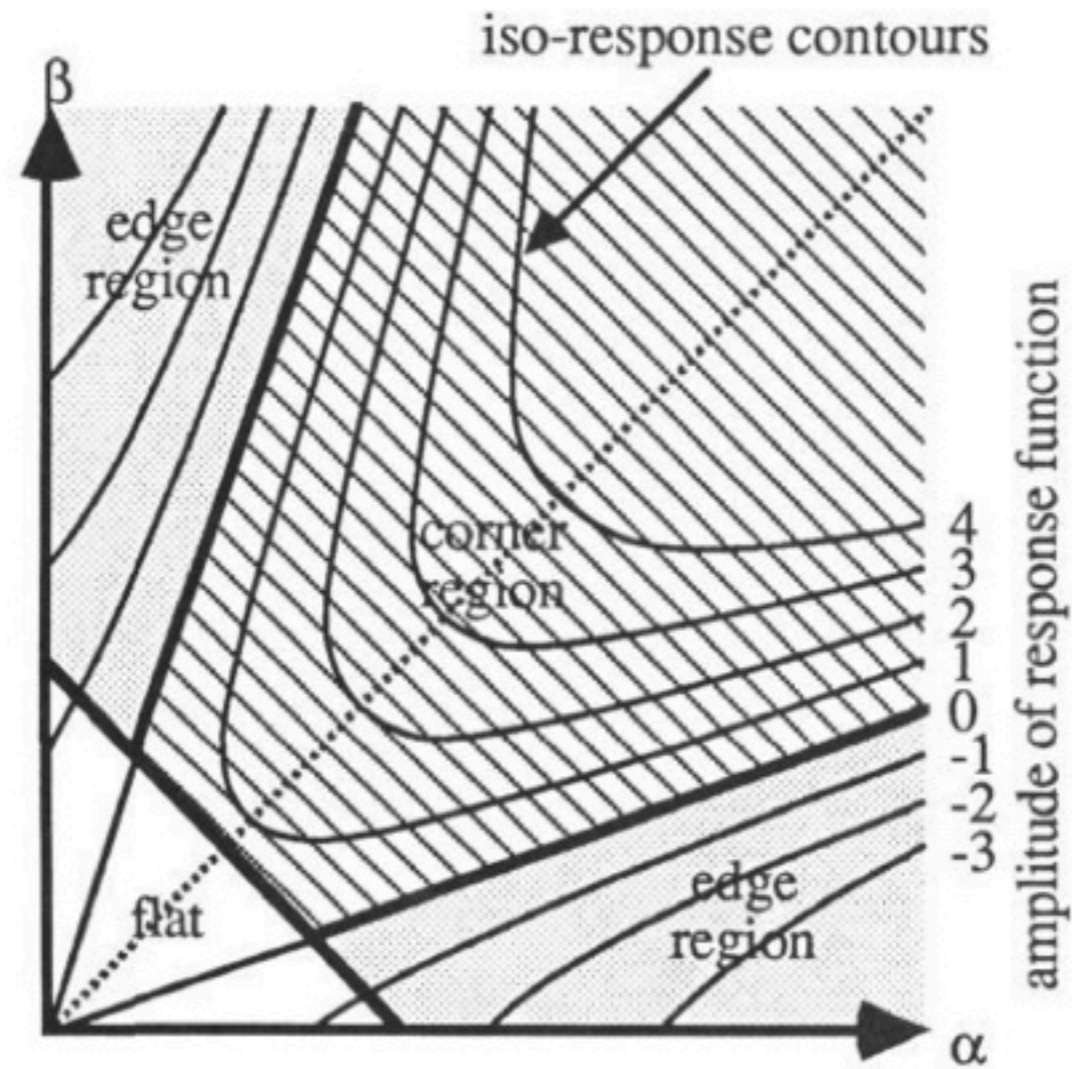
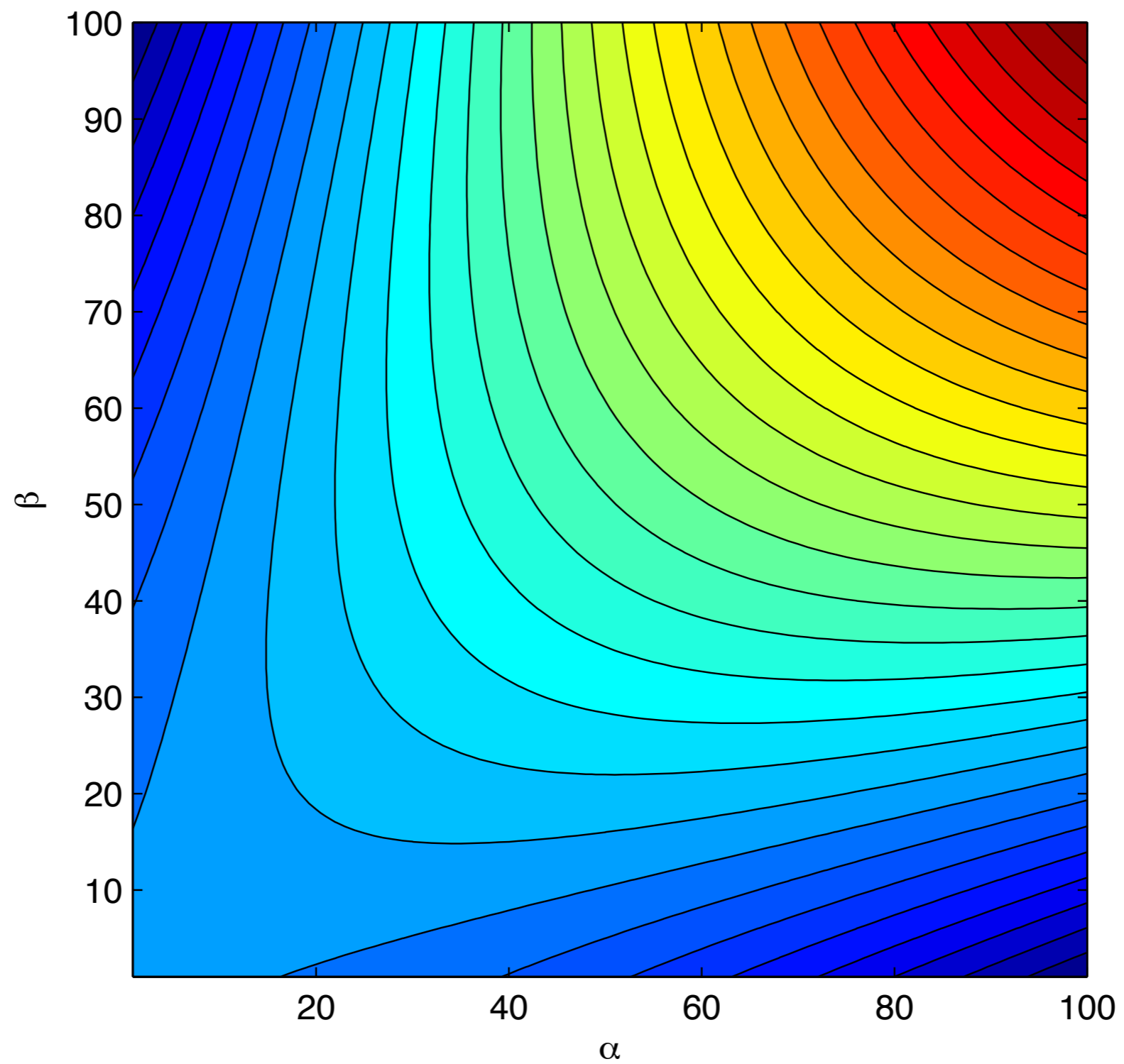


Figure 5. Auto-correlation principal curvature space-
heavy lines give corner/edge/flat classification,
fine lines are equi-response contours.

How do we get a response surface like that?

$$\alpha\beta - (\alpha + \beta)^2$$



CORNERNESS

HARRIS-STEPHENS CORNER

Hessian Matrix

$$\mathbf{H} = \begin{bmatrix} D_{xx} & D_{xy} \\ D_{xy} & D_{yy} \end{bmatrix}$$

Eigen values of \mathbf{H} : α, β

$$\text{Tr}(\mathbf{H}) = D_{xx} + D_{yy} = \alpha + \beta$$

$$\text{Det}(\mathbf{H}) = D_{xx}D_{yy} - D_{xy}^2 = \alpha\beta$$

$$M_c = \alpha\beta - \kappa(\alpha + \beta)^2 = \text{det}(\mathbf{H}) - \kappa\text{trace}^2(\mathbf{H})$$

$$M_c > \text{threshold}$$

$$\kappa = 0.04$$

$$\kappa = 0.15$$

HARRIS CORNERS

- ESTIMATE DIFFERENCE OF GAUSSIAN AT APPROPRIATE SCALE
- CONSTRUCT HESSIAN MATRIX FOR EACH PIXEL
- ESTIMATE M FOR EACH PIXEL
- CHOOSE LOCAL MAXIMA AS CORNERS

RECOGNITION:
IDENTIFICATION OF SOMETHING
PREVIOUSLY SEEN

BAG OF WORDS MODEL

Of all the sensory impressions proceeding to the brain, the visual experience is the dominant ones. Our perception of the world is based essentially on the information which the brain receives from the eyes. It is thought that the retina is the first point to which the visual information is sent. The cerebral cortex which the visual information reaches after the discarding of the irrelevant information that behind the scenes of the brain there is a complex course of events. The visual information along their path through the visual cortex, Hubel and Wiesel were able to demonstrate that the visual information falling on the retina undergoes a step-by-step analysis in a system of nerve cells stored in columns. In this system each cell has its specific function and is responsible for a specific detail in the pattern of the retinal image.

**sensory, brain,
visual, perception,
retinal, cerebral cortex,
eye, cell, optical
nerve, image
Hubel, Wiesel**

China is forecasting a trade surplus of \$90bn (£51bn) to \$100bn this year, a threefold increase on 2004's \$32bn. The Ministry said the surplus would be a 20% jump in exports to \$100bn and a rise in imports to \$10bn to annoy the US. China's exports to the US are too high and the value of China's exports to the US country are too high. China's demand for US goods is too high. China increased the value of the dollar by 2.1% within a narrow band. China has made it clear that it will take its time and carefully before allowing the yuan to rise in value.

**China, trade,
surplus, commerce,
exports, imports, US,
yuan, bank, domestic,
foreign, increase,
trade, value**



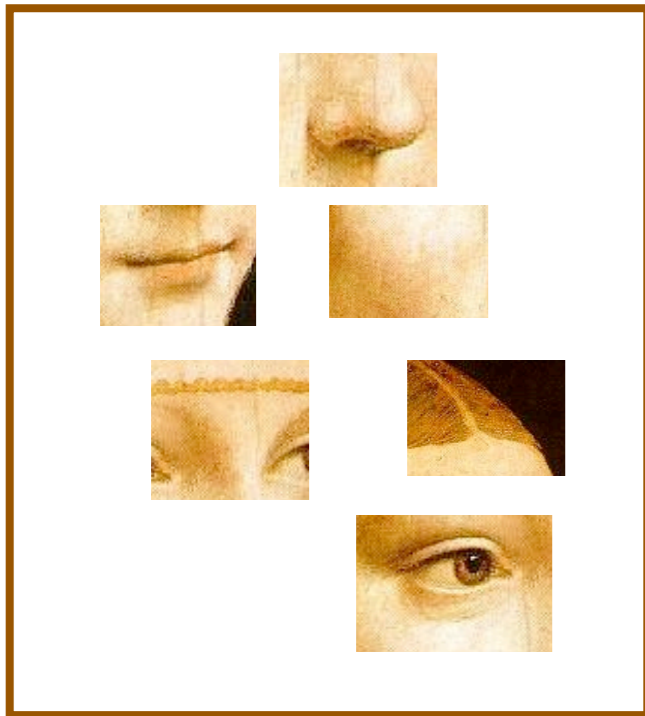
OBJECT



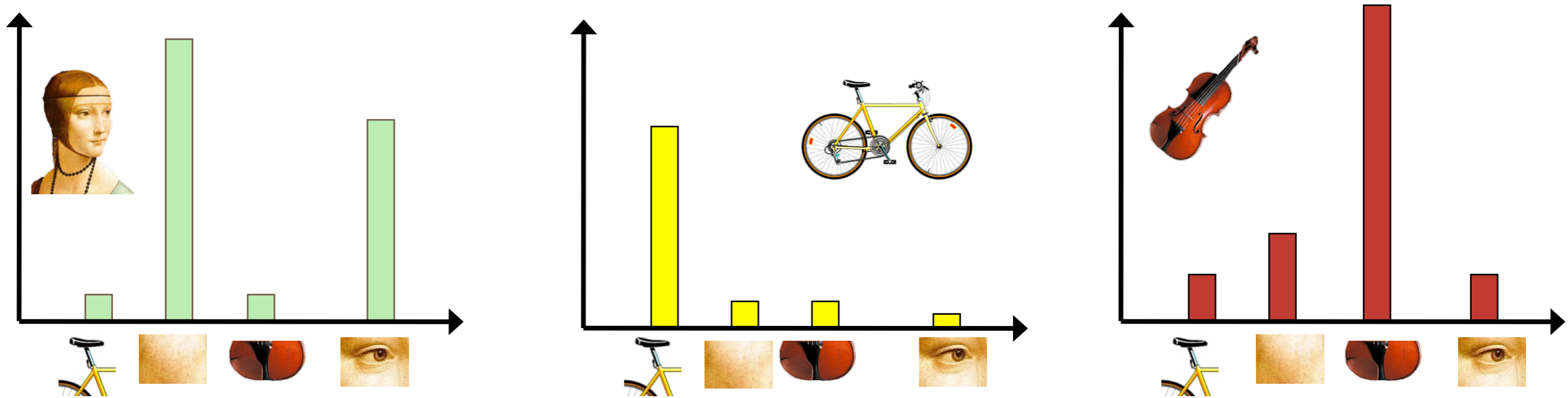
BAG OF WORDS

BoW: INDEPENDENT FEATURES

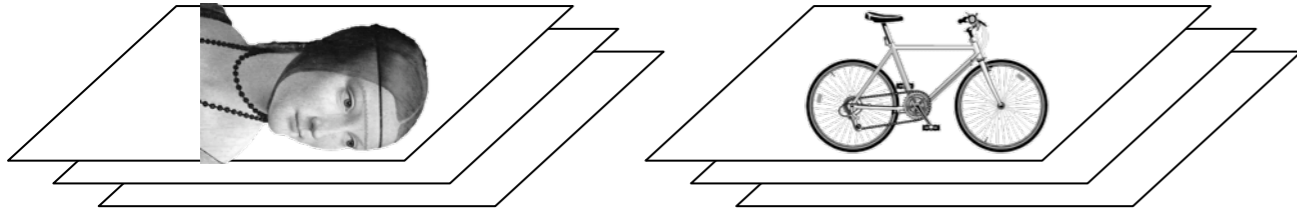
- Independent features



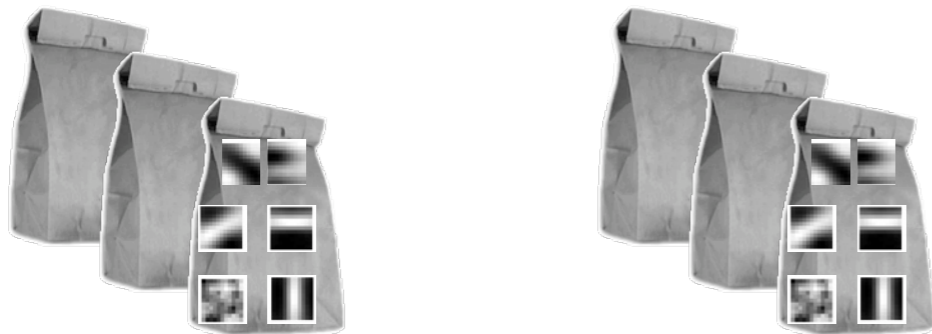
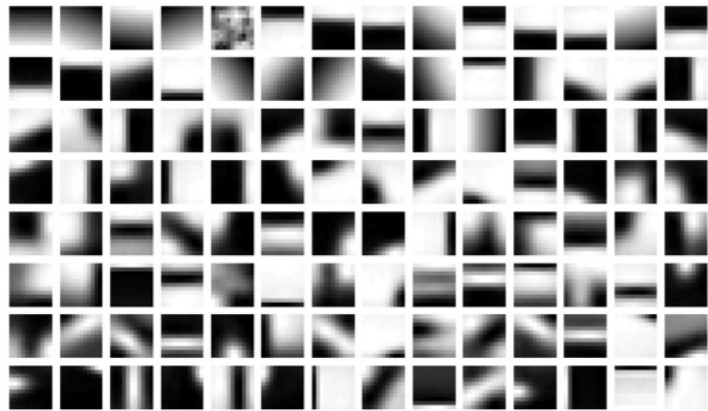
HISTOGRAM OF FEATURES



learning

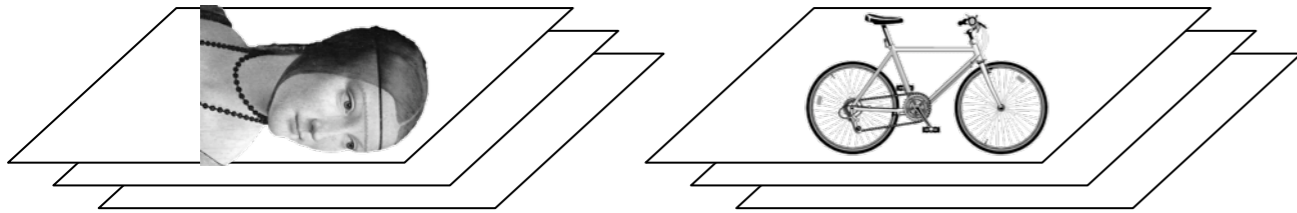


codewords dictionary



corpus representation

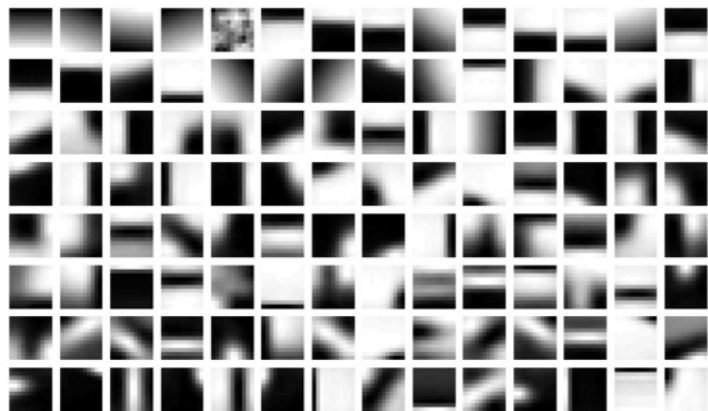
learning



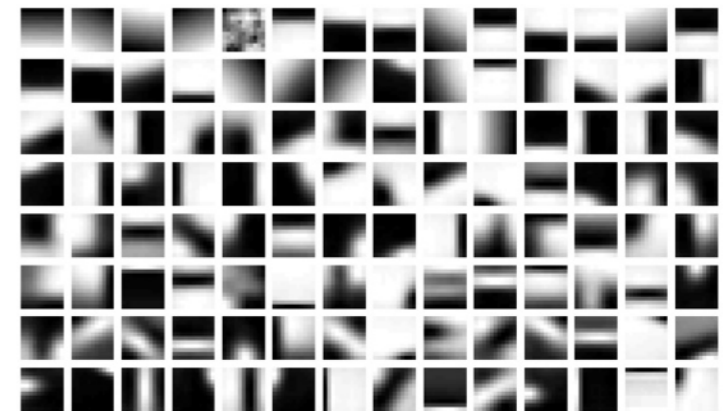
recognition



codewords dictionary



codewords dictionary



corpus representation

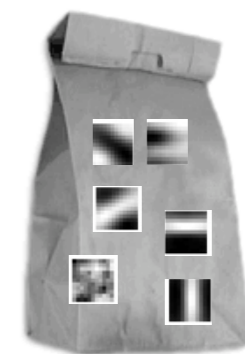
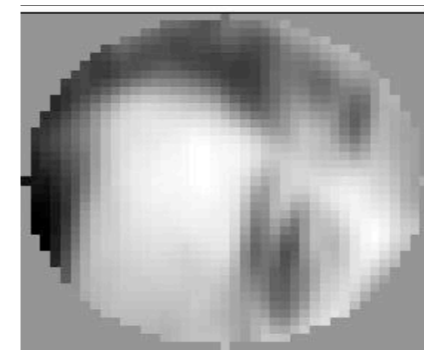
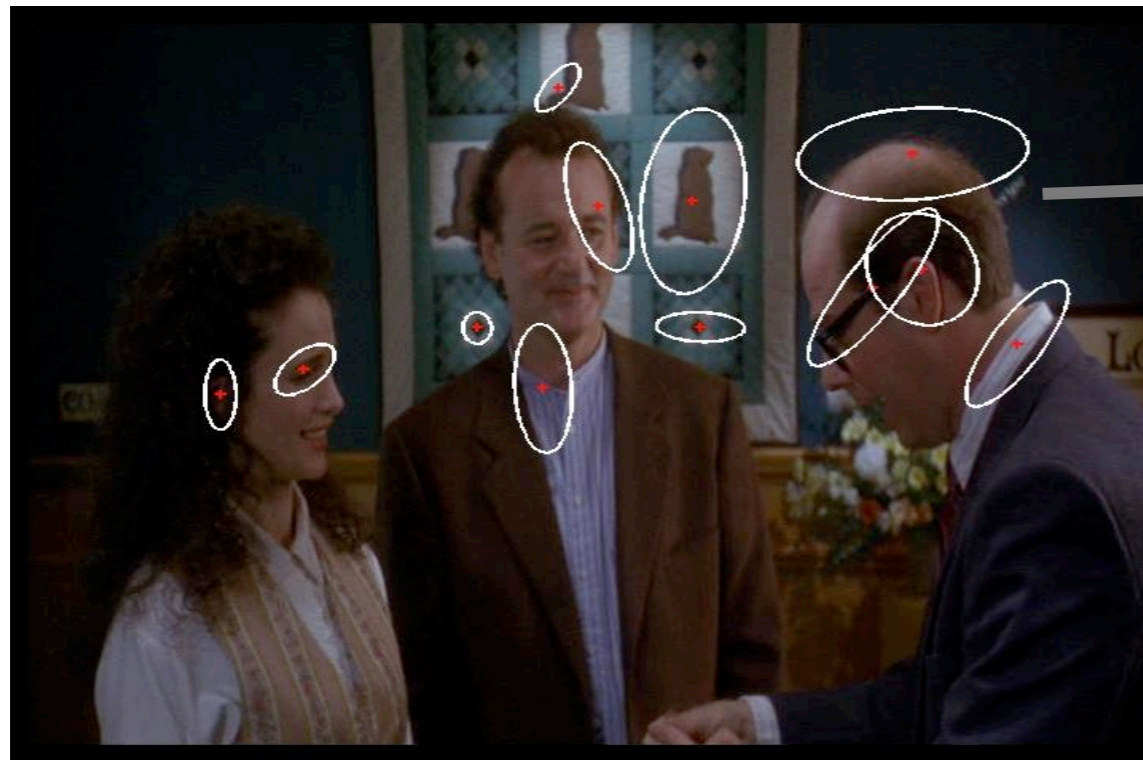


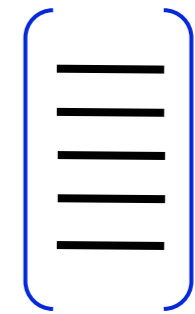
image representation

BUILDING A CODEBOOK

FEATURE DETECTION & REPRESENTATION



Normalize patch



Compute SIFT
descriptor

Detect patches

[Mikojaczyk and Schmid '02]

[Mata, Chum, Urban & Pajdla, '02]

[Sivic & Zisserman, '03]

SIFT DESCRIPTOR

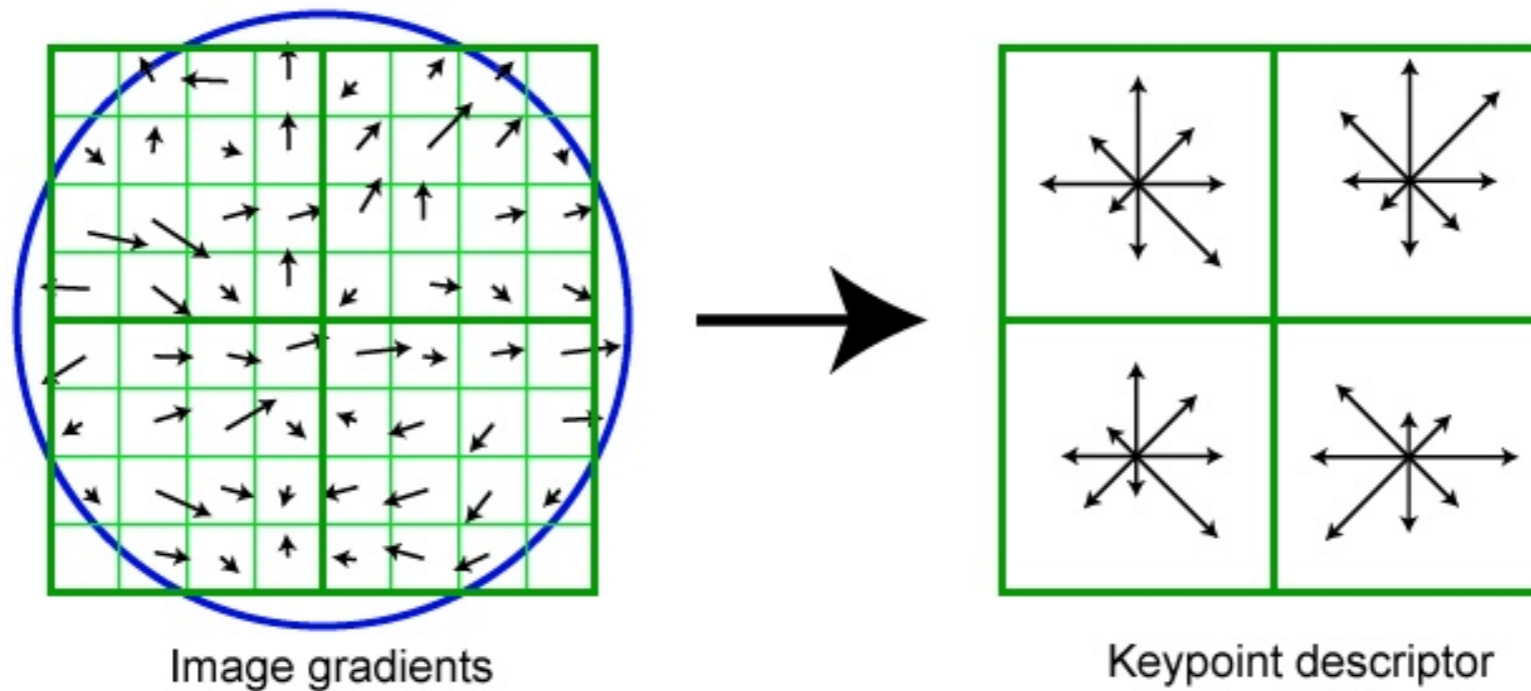


Image gradients

Keypoint descriptor

8x8

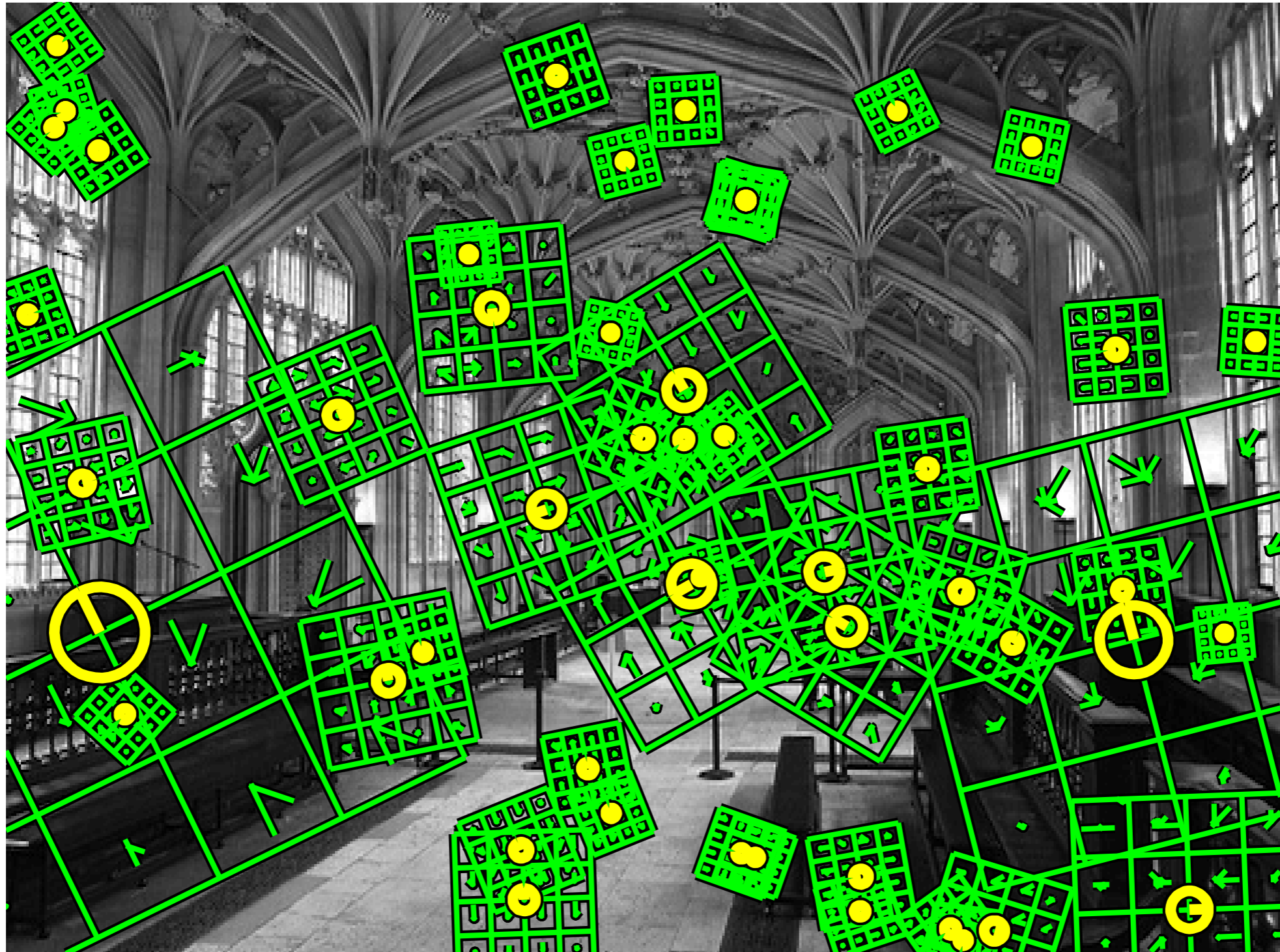
2x2

In practice: **16x16**

4x4

(4 x 4) cells x 8 orientations = 128 length descriptor

SIFT

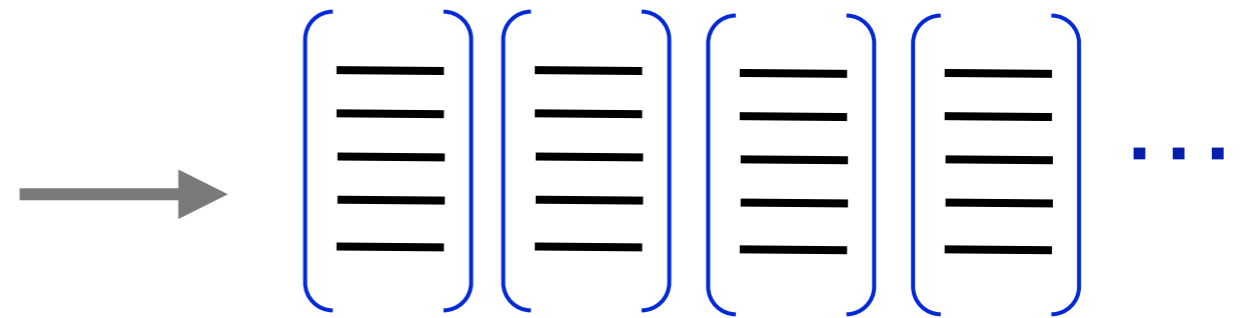
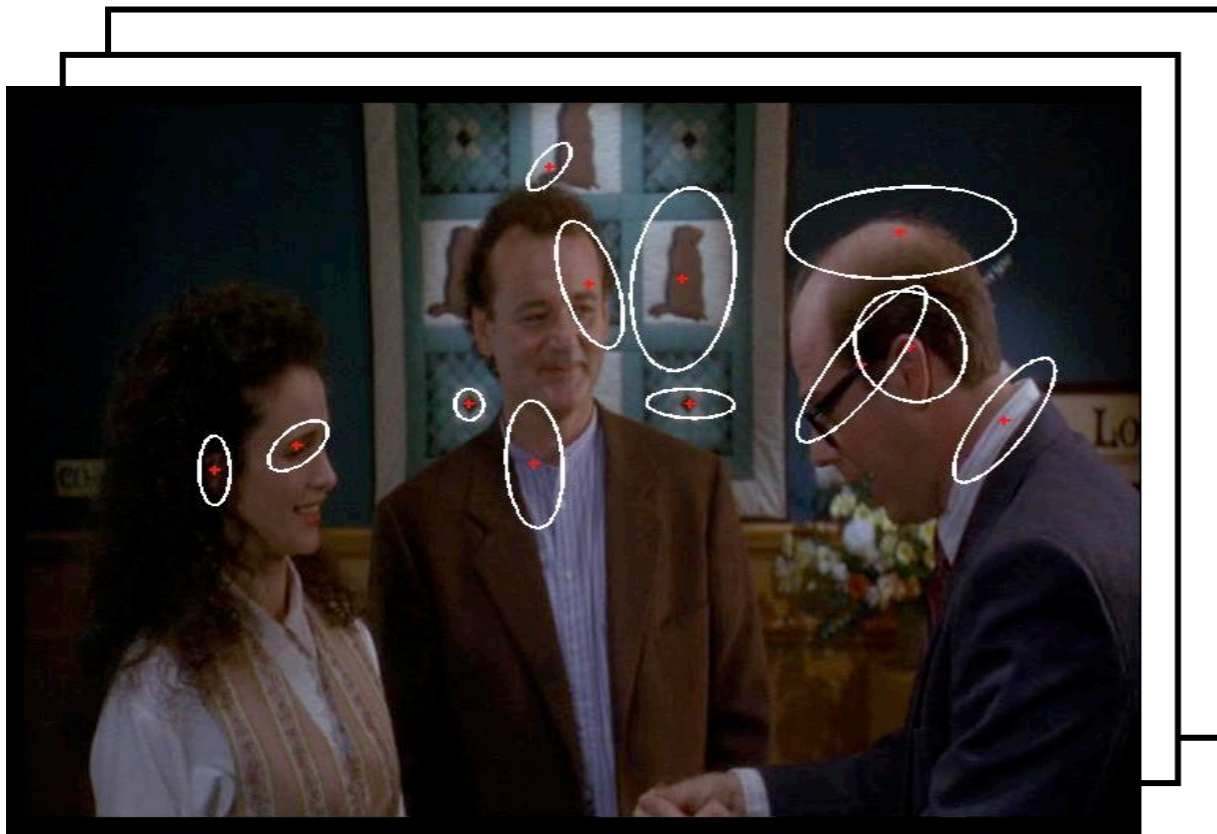


SIFT



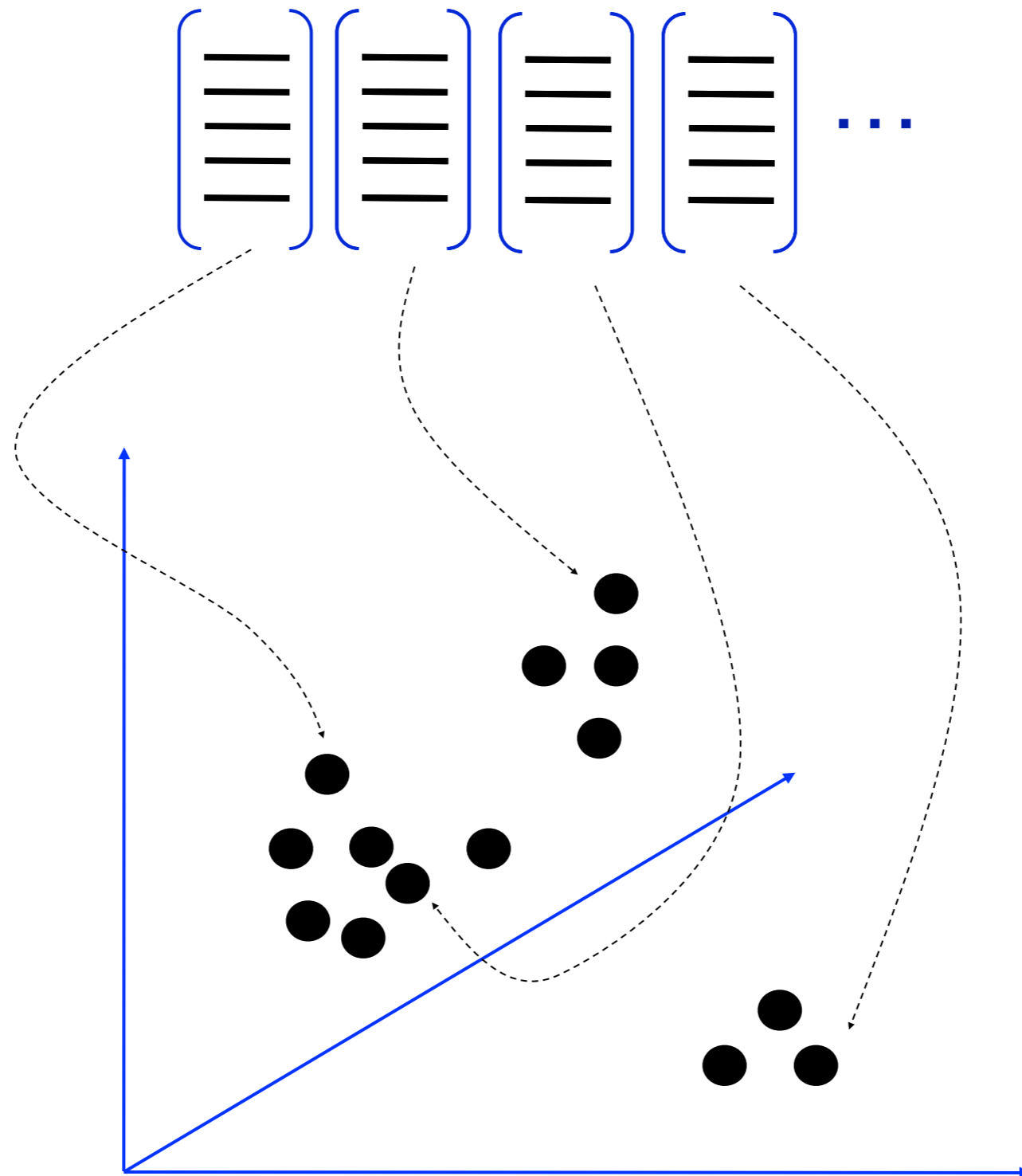
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FEATURE DETECTION & REPRESENTATION



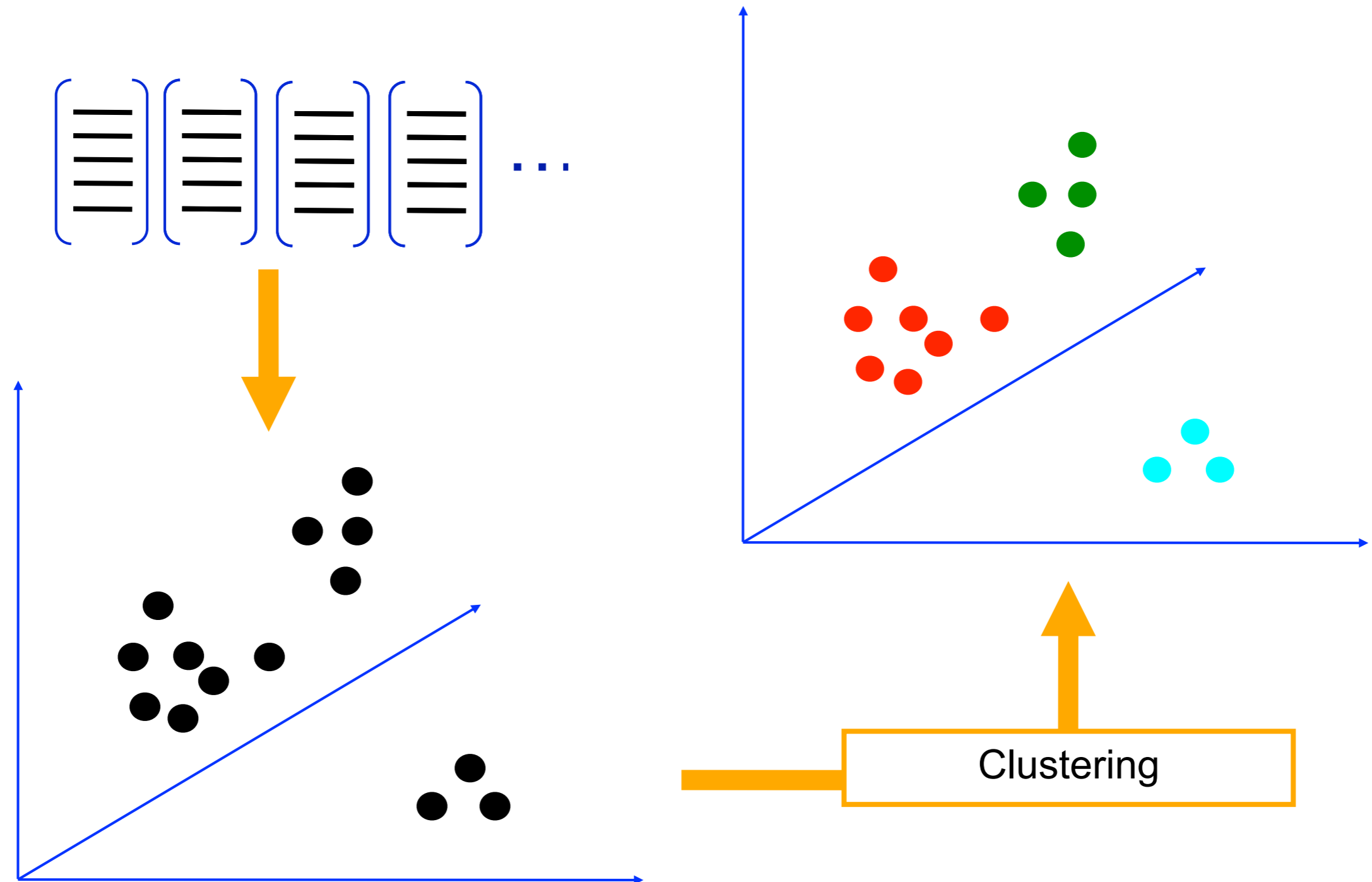
BUILDING A CODEBOOK

DICTIONARY FORMATION



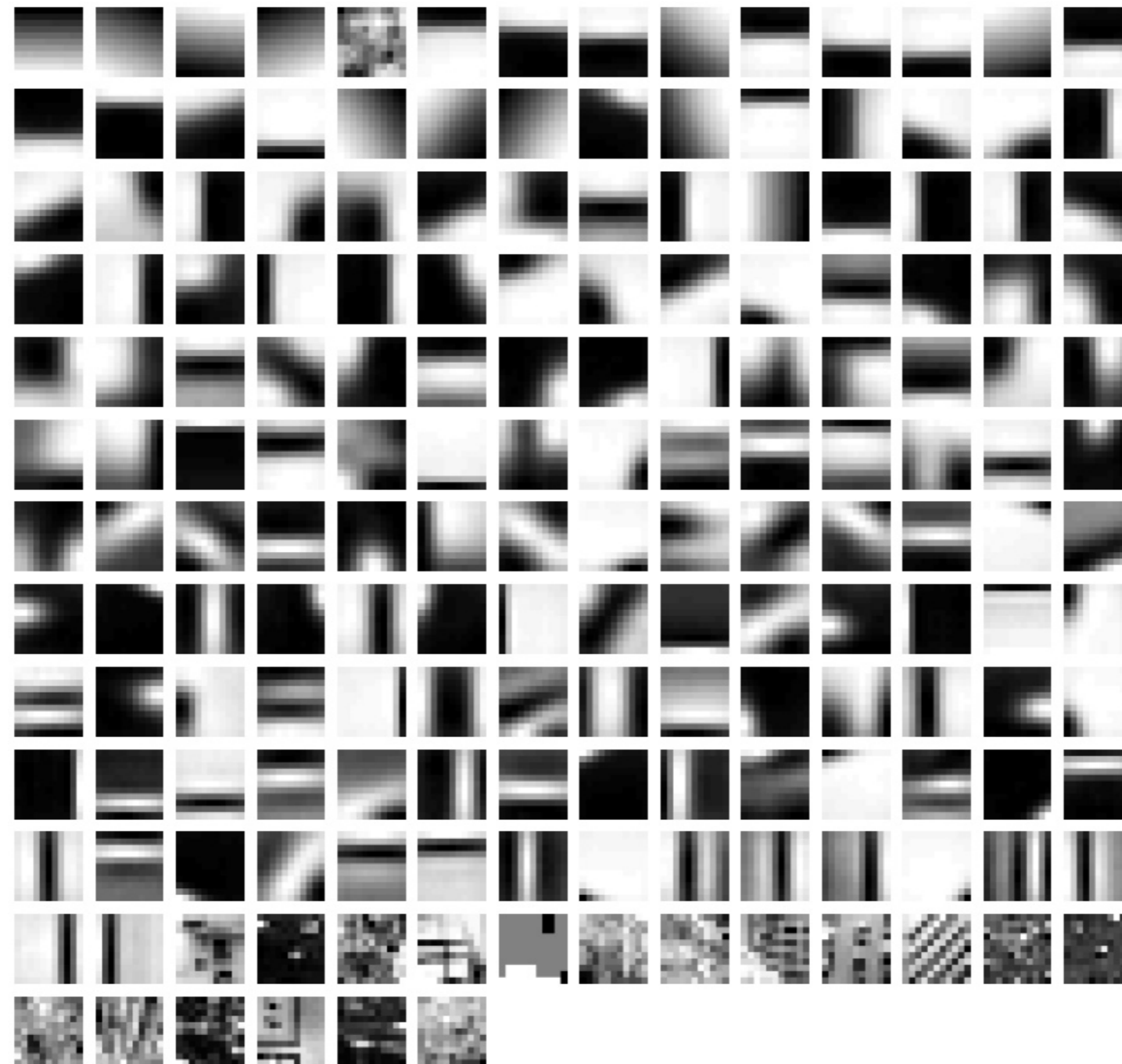
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DICTIONARY FORMATION



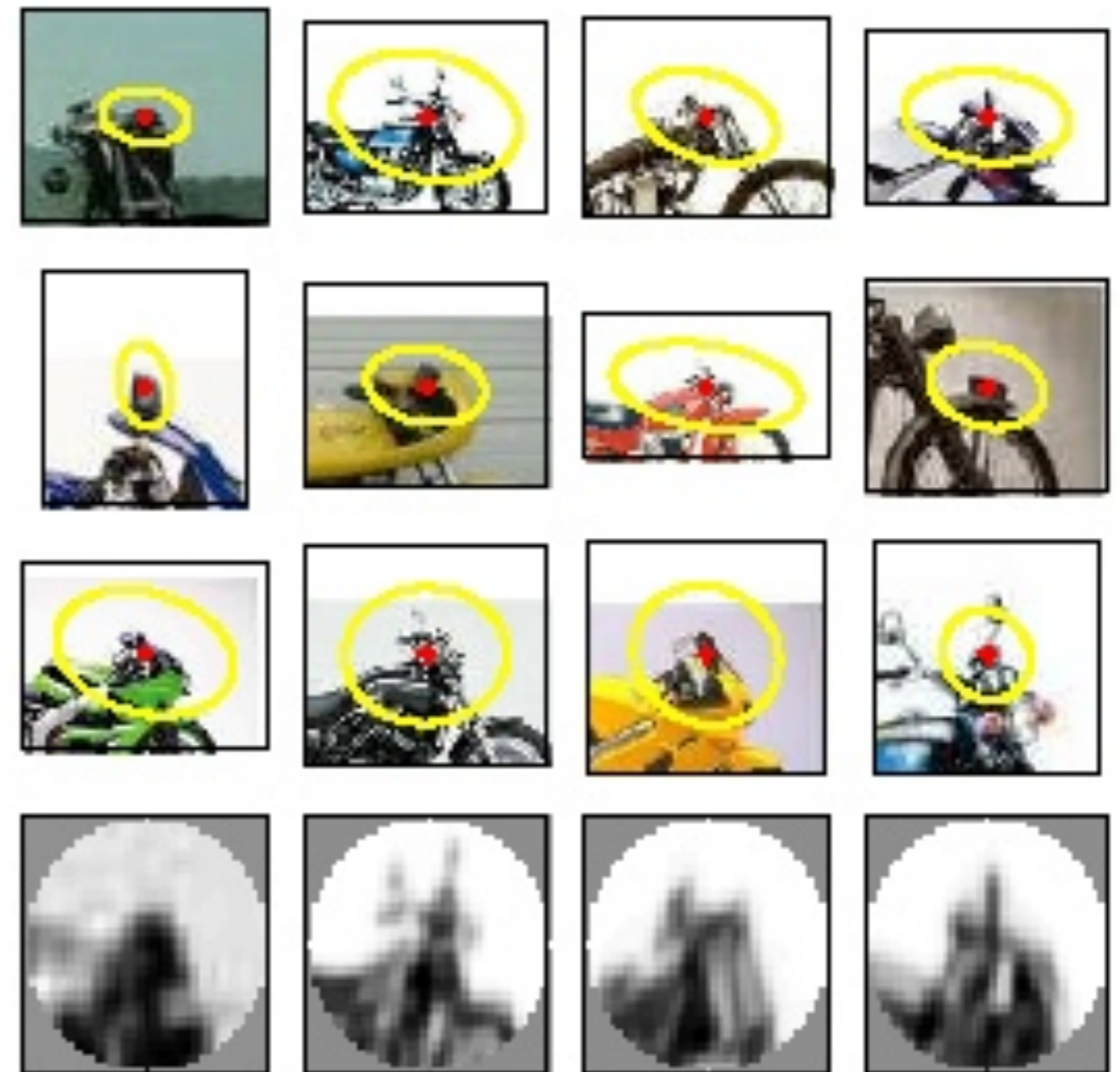
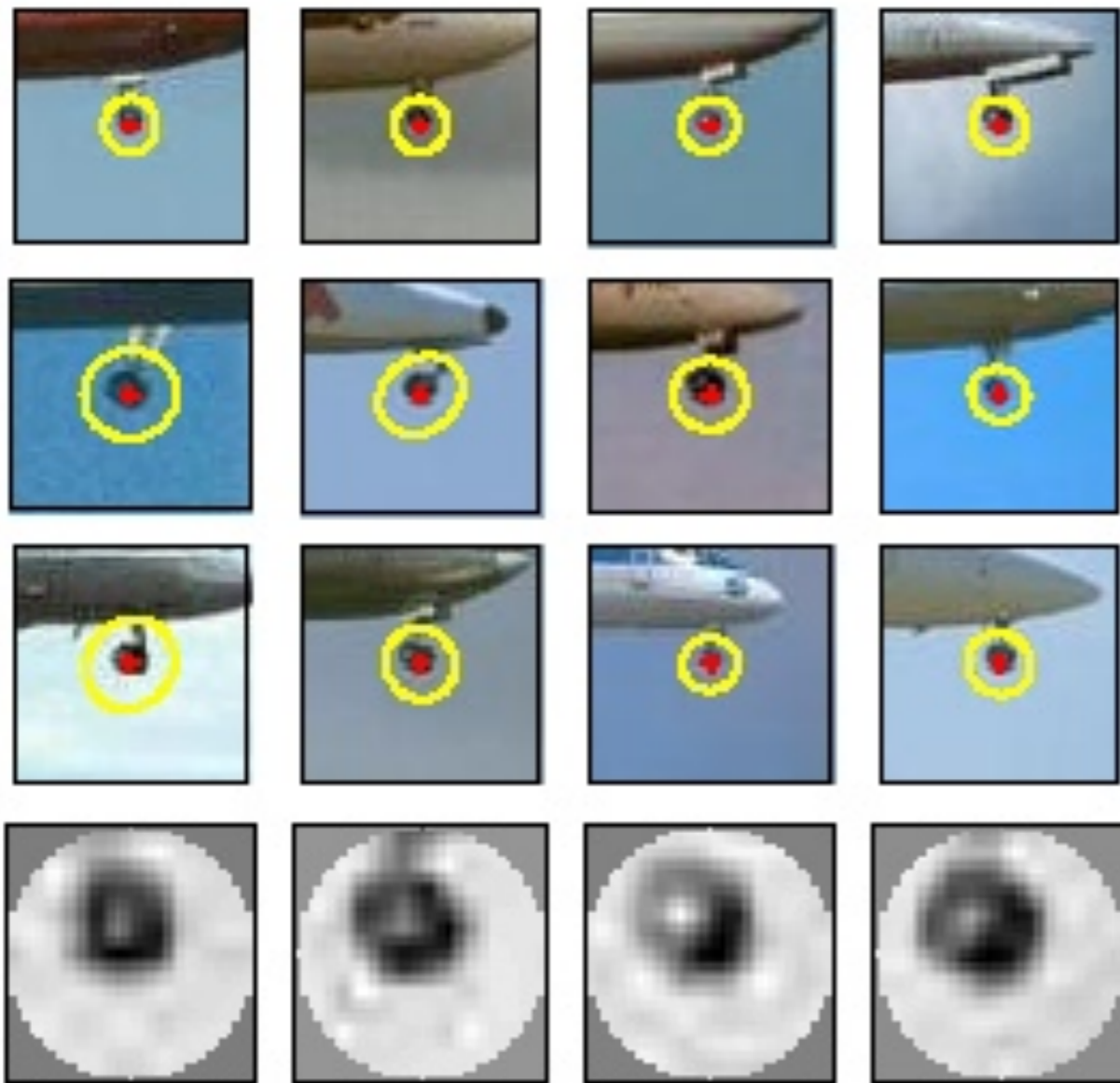
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EXAMPLE CODEBOOK

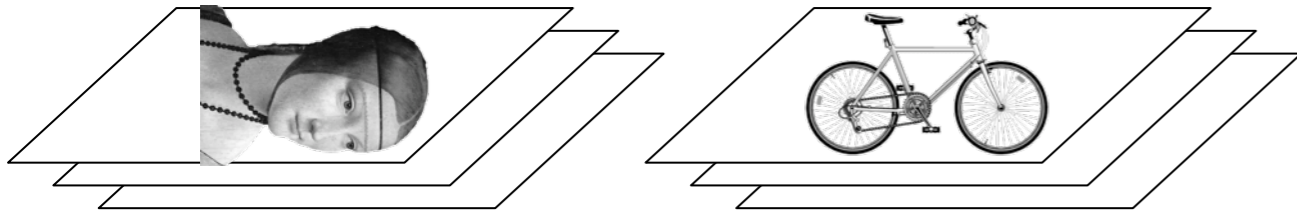


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EXAMPLE CODEBOOK



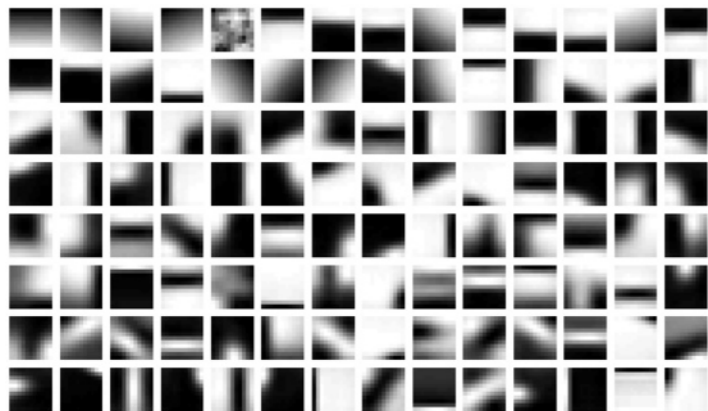
learning



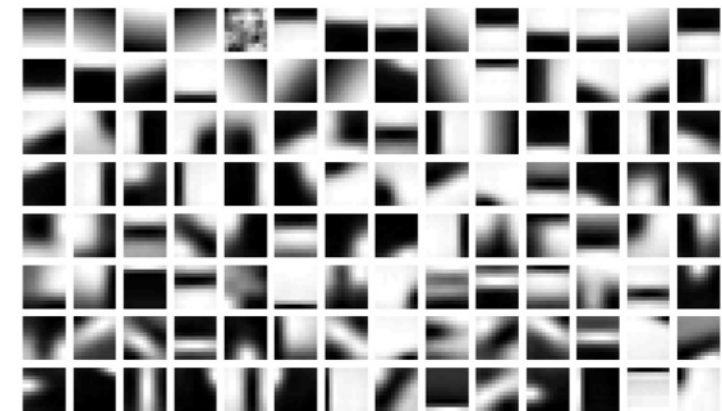
recognition



codewords dictionary



codewords dictionary



corpus representation

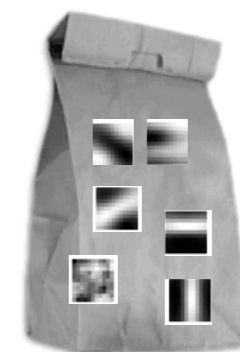
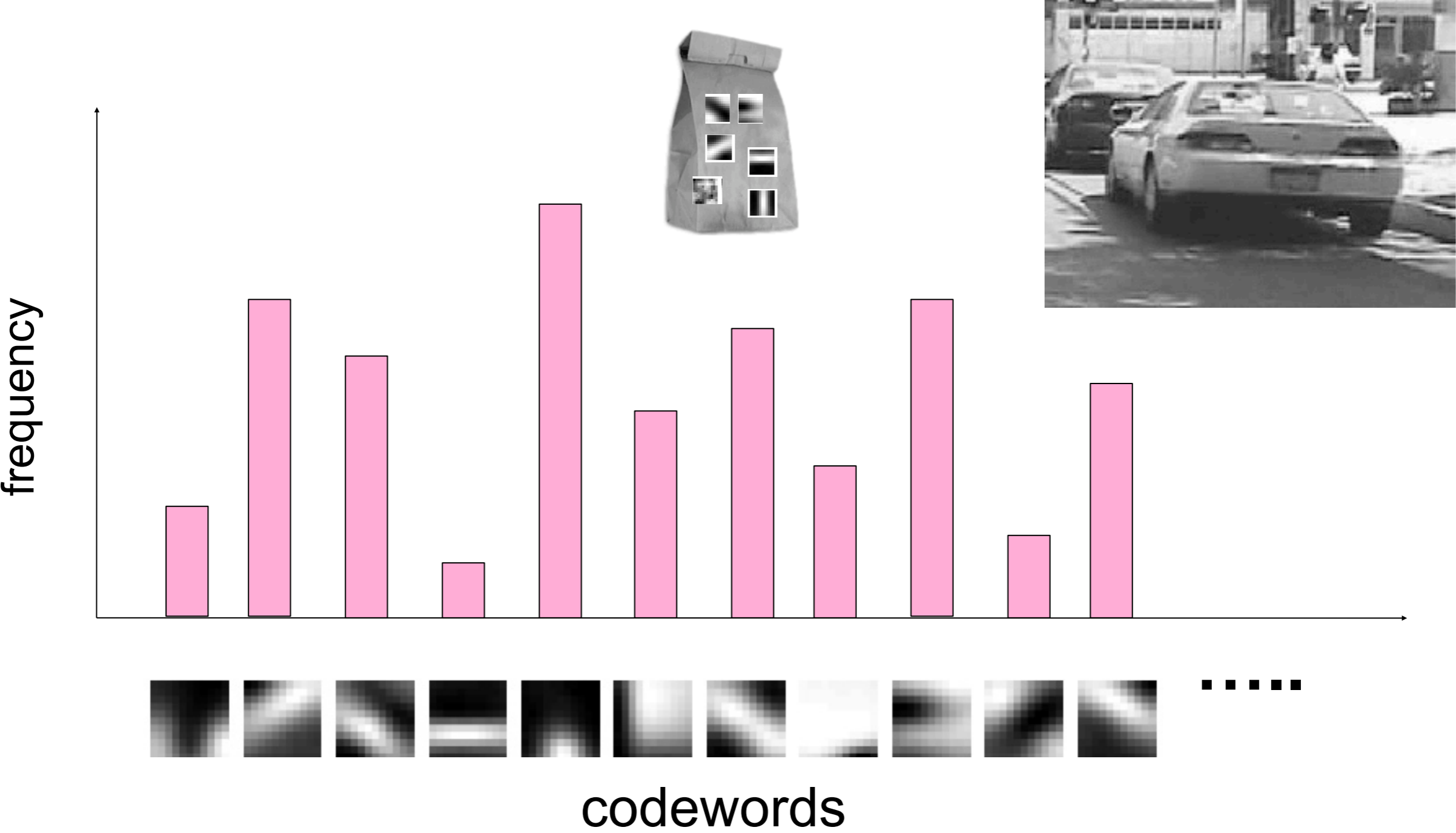


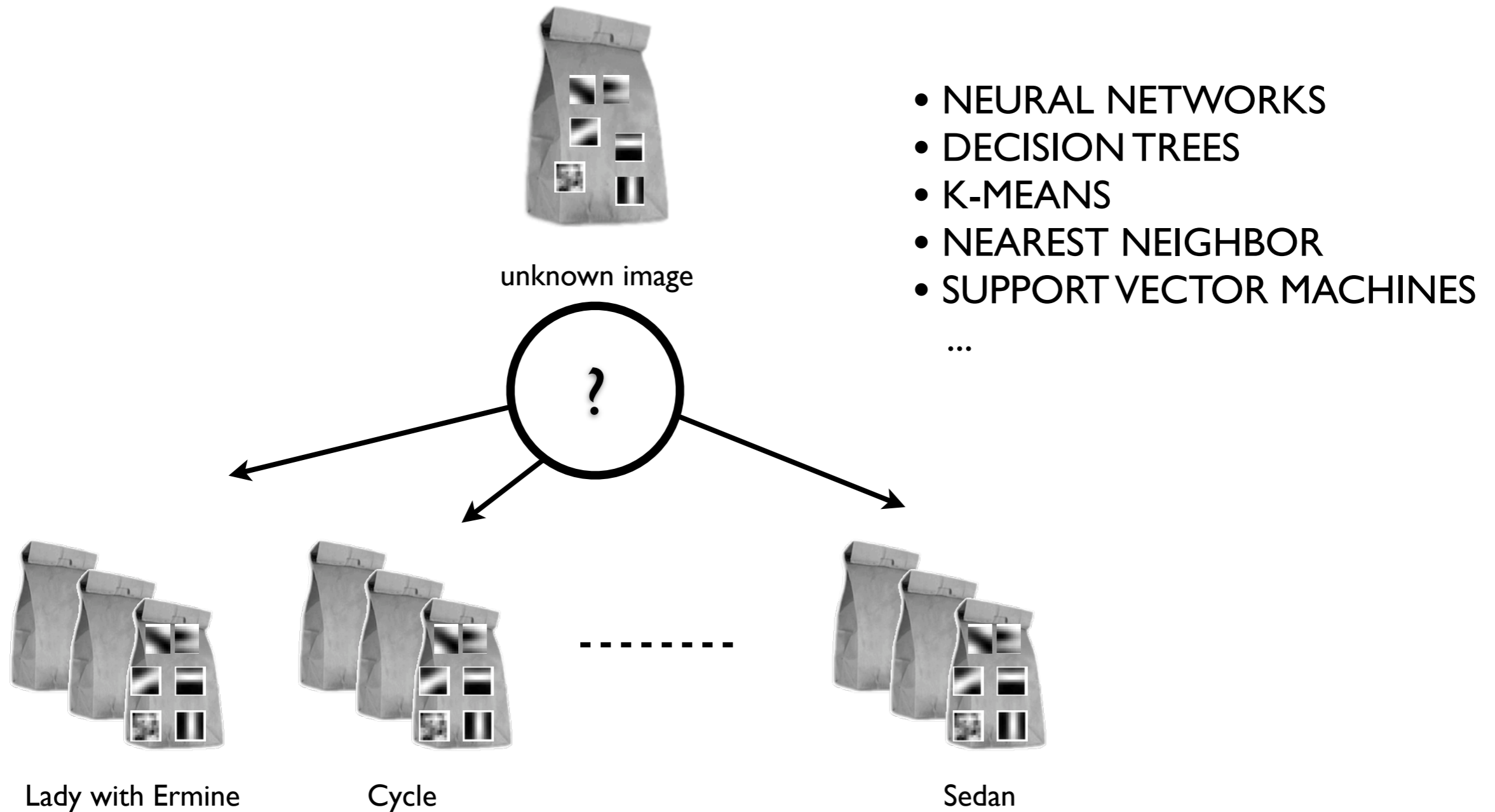
image representation

IMAGE REPRESENTATION

CODEWORDS



CLASSIFICATION



CODE

- SIFT implementation in MATLAB: <http://www.vlfeat.org/>
- Bag of words: <http://people.csail.mit.edu/fergus/iccv2005/bagwords.html>
- **Data:**
 - Image-net: <http://www.image-net.org/>