



CNN for CV

AI for CV Group
2019

A large, irregular blue ink splash or watercolor blotch serves as the background for the text. The splash is centered and has a textured, painterly appearance with various shades of blue and some white highlights. It is surrounded by smaller, scattered blue ink droplets and splatters on a white background.

Week 2. What Is CV

Outline:

I. Convolution & Applications (Low Level)

- A. Low Level Image Processing (Part II)
- B. Applications

II. Feature Points (Mid Level)

- C. What Is A Feature Point
- D. Corner Point
- E. SIFT

III. Classical CV Procedure (High Level)

- F. Classification In Classical CV Procedure

I. Conv & App

(Low Level)

I. Convolution & Applications

A. Low Level Image Processing

A1. 1-dim convolution

I. Convolution & Applications

A. Low Level Image Processing

A2. 2-dim convolution

I. Convolution & Applications

A. Low Level Image Processing

A3. Image convolution

I. Convolution & Applications

Q.As for convolution

1. Do I have to turn over the kernel?

$$y(n) = \sum_{i=-\infty}^{\infty} x(i)h(n-i) = x(n) * h(n)$$

I. Convolution & Applications

Q.As for convolution

2. What can it do?

- First-order derivative
- Second-order derivative

I. Convolution & Applications

Q.As for convolution

2. What can it do?

- Response of first & second order derivative

- Conclusions

I. Convolution & Applications

Q.As for convolution

2. What can it do?

- Gaussian Kernel
- Acceleration

I. Convolution & Applications

Q.As for convolution

3. Applications

- Image Sharpening: Laplacian
- Edge Detection: Sobel
- Image Blurring: Median/Gaussian
- Reading: [Bilateral Filtering](#)

II. Feature Points

(Mid Level)



II. Feature Points

Why we need feature points?

What is a feature point?

What is a good feature point?

What is the form of a feature point?

How to get a feature point?

How many types of feature points?

Any applications?

.....

II. Feature Points

Why we need feature points?

CV → stitching/classification/reconstruction...



look for pixels/objects representation

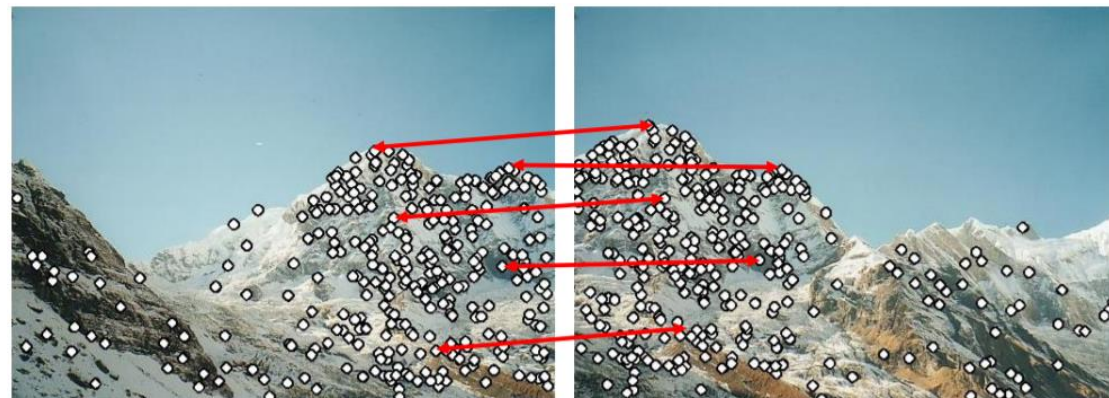
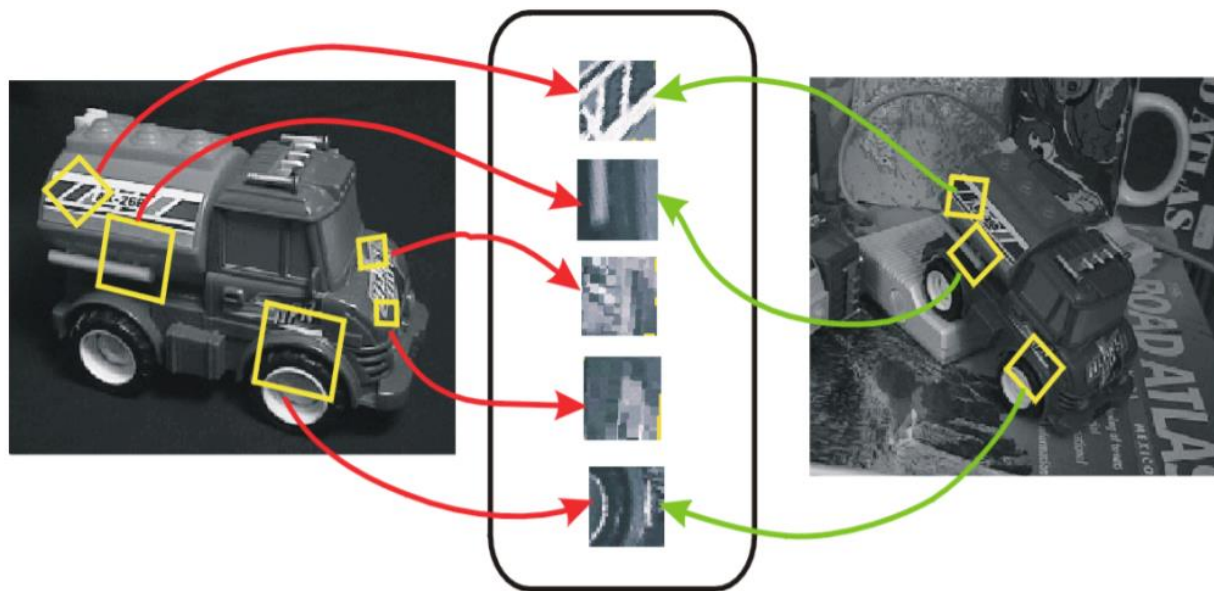


feature points

II. Feature Points

What is a feature point?

Represents of objects/pixels



II. Feature Points

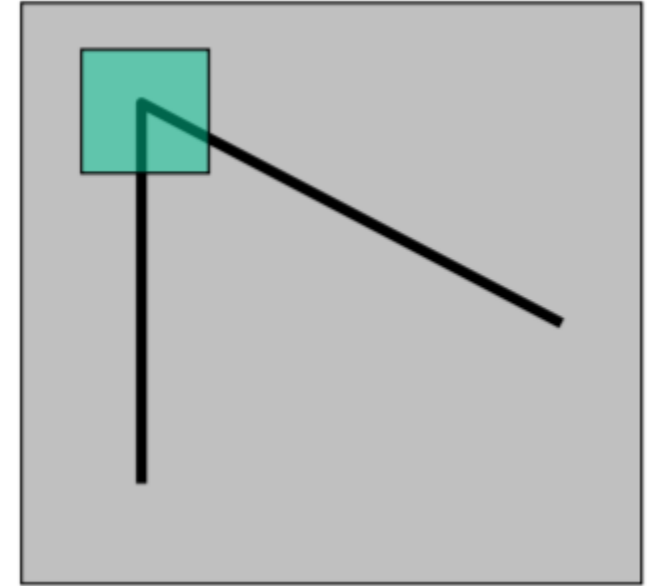
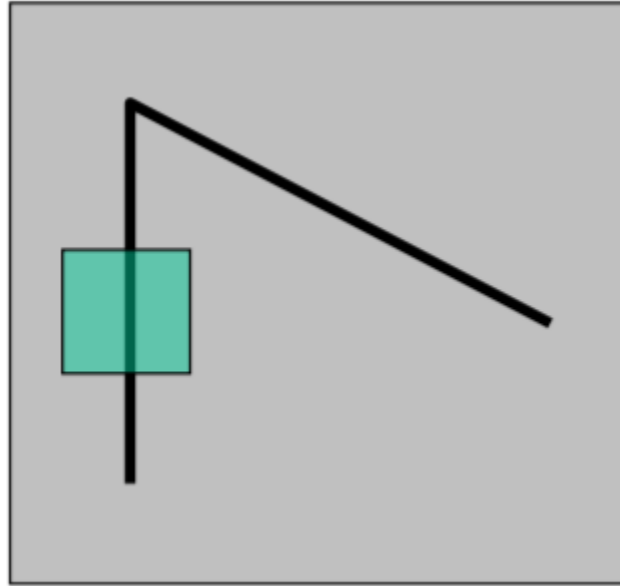
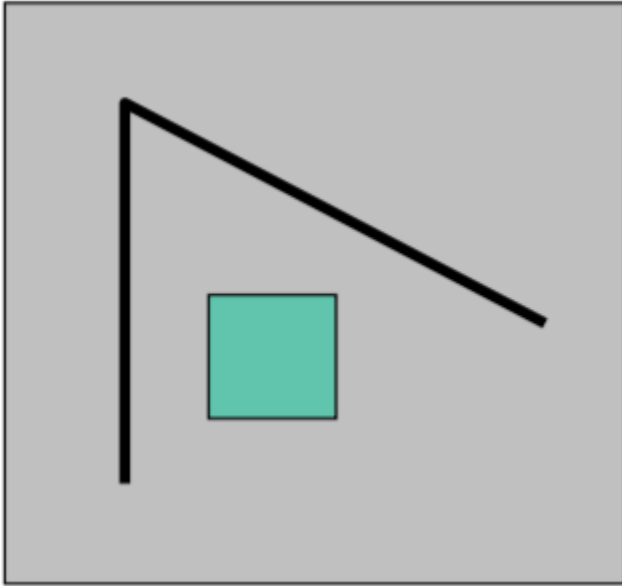
What is a good feature point?

- Very informational

II. Feature Points

What is a good feature point?

- Very informational



II. Feature Points

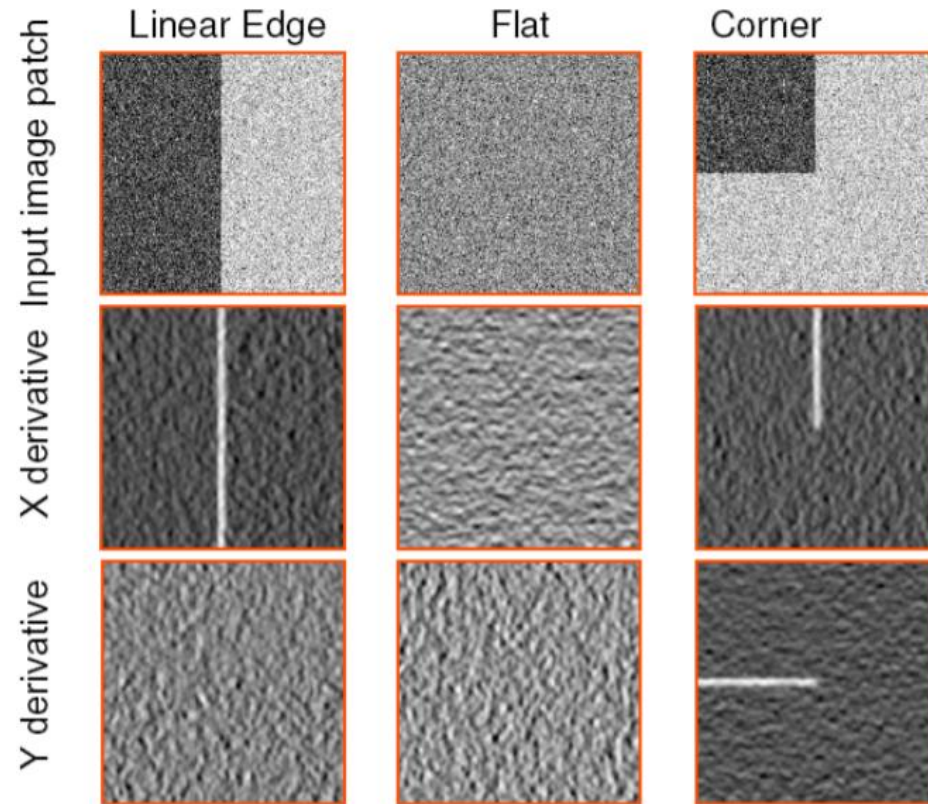
What is a good feature point?

- Very informational (Harris Corner Detector)

II. Feature Points

What is a good feature point?

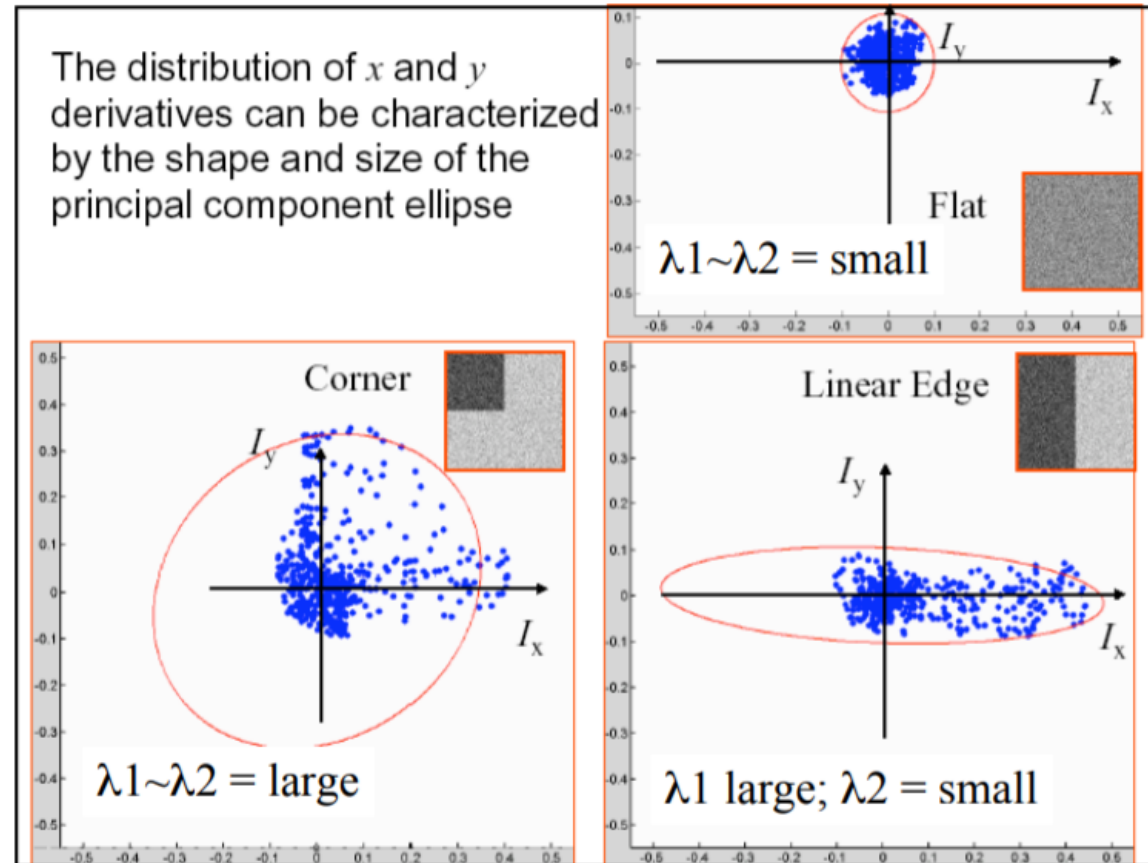
- Very informational (Harris Corner Detector)



II. Feature Points

What is a good feature point?

- Very informational (Harris Corner Detector)



II. Feature Points

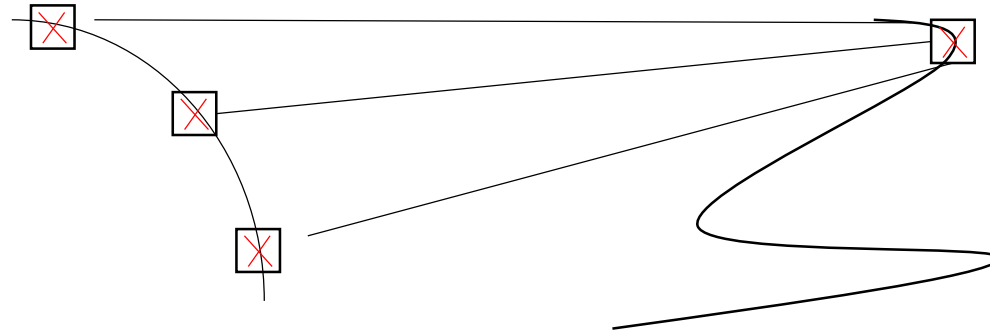
What is a good feature point?

- Rotation/Brightness resistance (Harris Corner Detector)

II. Feature Points

What is a good feature point?

- Scale resistance (Harris Corner Detector)



A + B + C → Good feature point

II. Feature Points

What is the form of a feature point?

Physical in location

Abstract in formation (usually a vector)



Feature Descriptor

II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

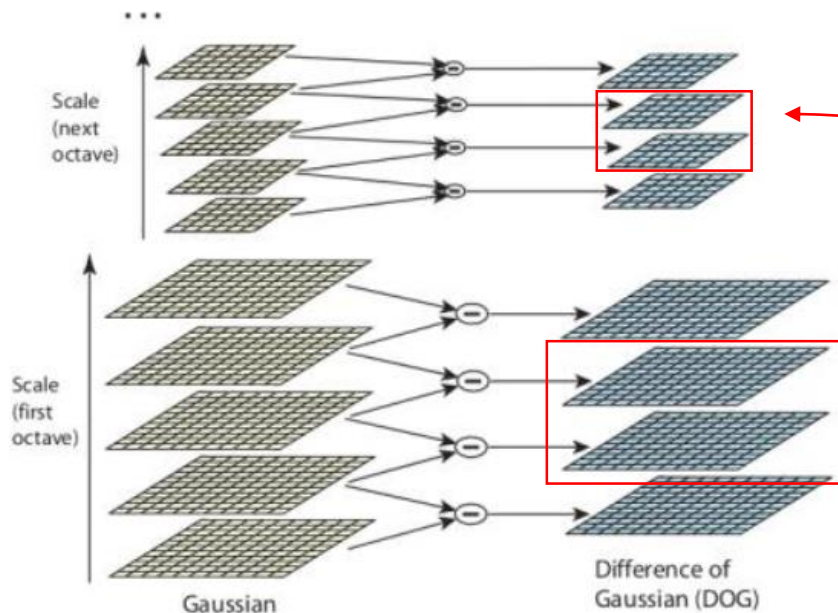
1. Generate Scale-space: DoG
2. Scale-space Extrema Detection
3. Accurate Keypoint Localization
4. Eliminating Edge Responses
5. Orientation Assignment
6. Keypoint Descriptor

II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

1. Generate Scale-space: DoG



$$Octave = \lceil \log_2 \min(W, H) \rceil - S$$

$$\sigma_{o,s} = \sigma_0 k^{o+s/S}$$

$$k = 2^{\frac{1}{S}}, S = 2, 3$$

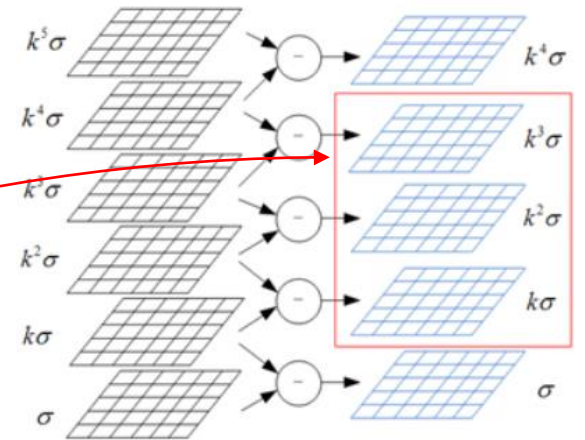
$$\sigma_0 = 1.6 \rightarrow \text{original}$$

$$= \sqrt{\sigma_0^2 - (2\sigma_n)^2} \rightarrow 1^{\text{st}} \text{ layer, } 1^{\text{st}} \text{ octave}$$

$$\sigma_n = 0.5$$

$$\# \text{ of scale images} = nOctave * (S + 3)$$

Use bilinear interpolation to get 2xImage as 1st layer in 1st octave



$$\text{第0组: } \sigma_0 \rightarrow 2^{1/3}\sigma_0 \rightarrow 2^{2/3}\sigma_0 \rightarrow 2^{3/3}\sigma_0 \rightarrow 2^{4/3}\sigma_0 \rightarrow 2^{5/3}\sigma_0$$

$$\text{第1组: } 2\sigma_0 \rightarrow 2 * 2^{1/3}\sigma_0 \rightarrow 2 * 2^{2/3}\sigma_0 \rightarrow 2 * 2^{3/3}\sigma_0 \rightarrow 2 * 2^{4/3}\sigma_0 \rightarrow 2 * 2^{5/3}\sigma_0$$

II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

1. Generate Scale-space: DoG

II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

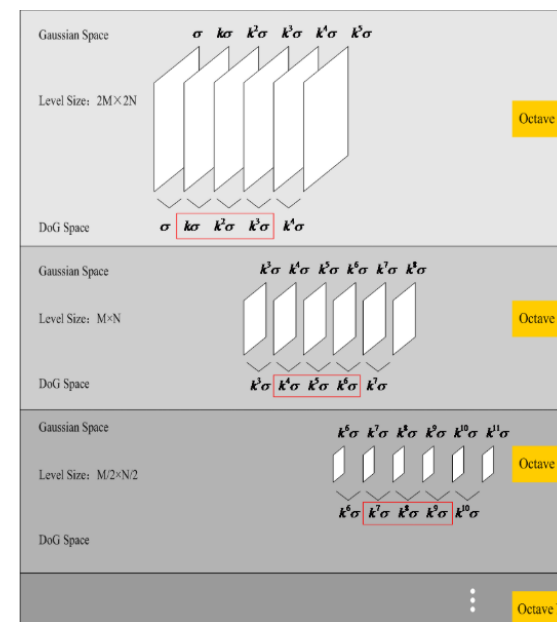
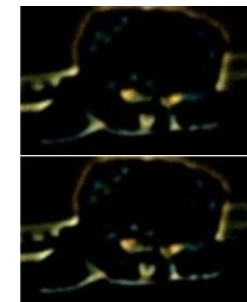
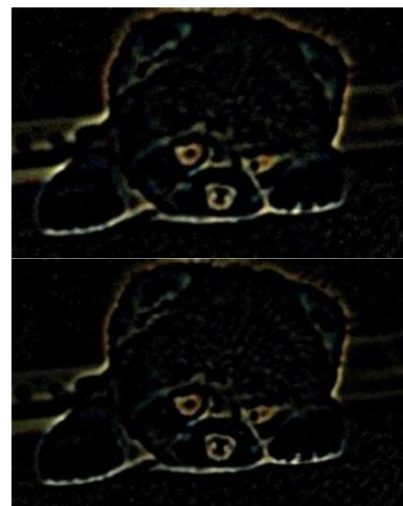
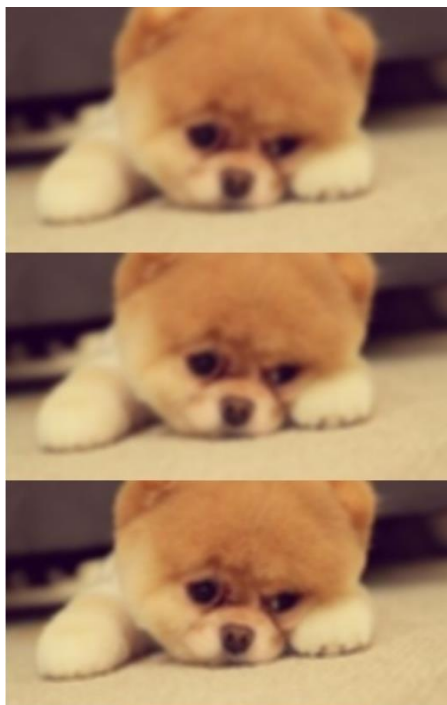
1. Generate Scale-space: DoG

II. Feature Points

How to get a feature point/descriptor?

e.g SIFT \rightarrow Scale-Invariant Feature Transform

1. Generate Scale-space: DoG

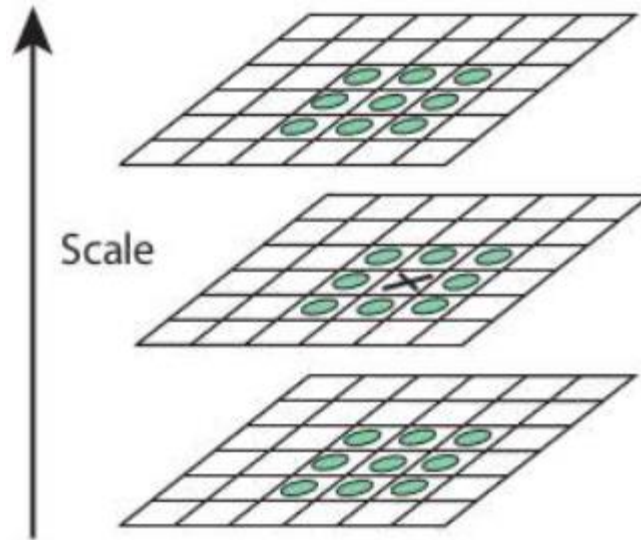


II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

2. Scale-space Extrema Detection (min & max)

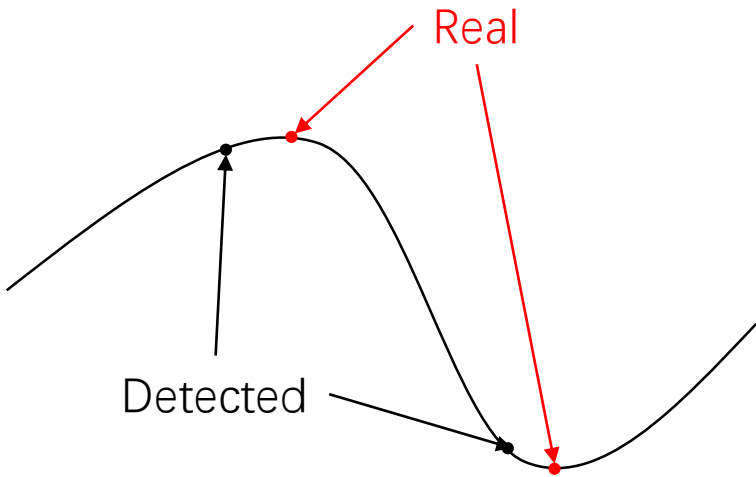


II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

3. Accurate Keypoint Localization



II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

4. Eliminating Edge Responses

Reason: DoG is too sensitive to edge, unstable

Theory: [Curvature & Hessian Matrix](#)

Method:

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

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

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

$$\alpha = r\beta$$

$$ratio = \frac{Tr(H)^2}{Det(H)} = \frac{(r+1)^2}{r}$$

$$let\ ratio < \frac{(r+1)^2}{r} \text{ when } r_0 = 10$$

II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

5. Orientation Assignment

$$r = 3 * 1.5\sigma$$

$$m(x, y) = \sqrt{(L(x + 1, y) - L(x - 1, y))^2 + (L(x, y + 1) - L(x, y - 1))^2}$$

$$\theta(x, y) = \arctan\left(\frac{L(x, y + 1) - L(x, y - 1)}{L(x + 1, y) - L(x - 1, y)}\right)$$

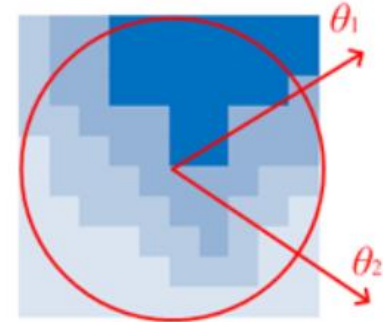
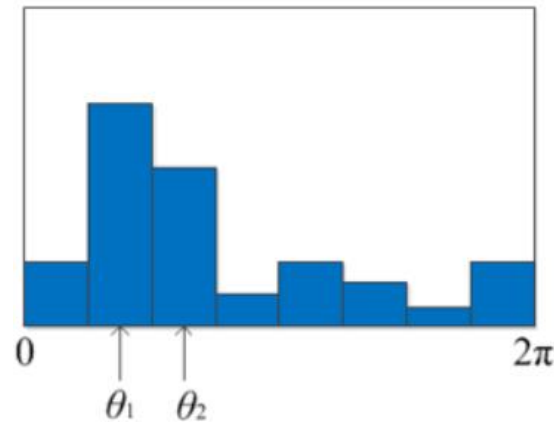
10bin, 0-360°

(we can also use interpolation here)

θ_2 at least 80% of θ_1 ,

main direction

We get (x, y, σ, θ) for each feature point

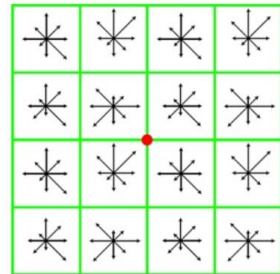
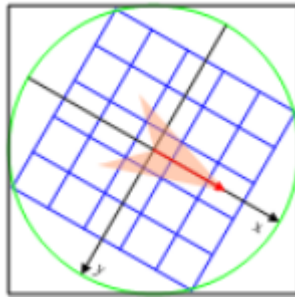
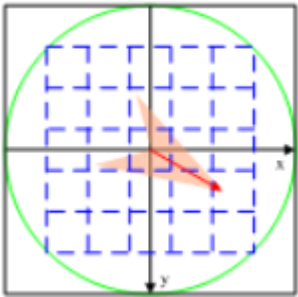


II. Feature Points

How to get a feature point/descriptor?

e.g SIFT \rightarrow Scale-Invariant Feature Transform

6. Keypoint Descriptor



For each keypoint, we have a $4 \times 4 \times 8 = 128$ dimension vector as its descriptor

II. Feature Points

How to get a feature point/descriptor?

e.g SIFT → Scale-Invariant Feature Transform

6. Keypoint Descriptor



Features:

- a. Almost the most accurate one
- b. Scale, brightness, rotation friendly
- c. Computing consuming and slow

II. Feature Points

How many types of feature points?

Besides SIFT, we have:

FAST

SURF

BRIEF

HoG

Orb

...

Reading assignment

FAST: faster version of SIFT

HoG: detection

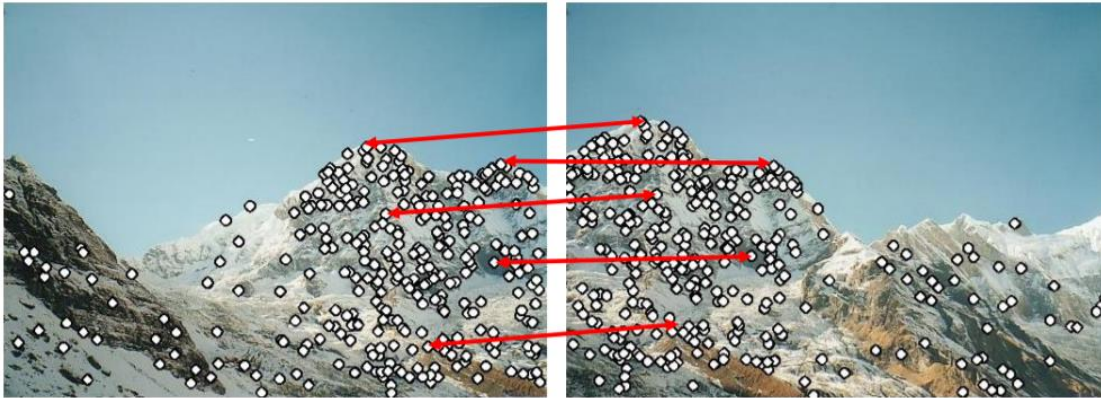
Orb: SLAM guys' friend



II. Feature Points

Any applications?

- Image Stitching



Pipeline:

1. Find feature points in each image
2. Use **RANSAC** to find keypoint matches
3. Use **homography matrix** to get transferring info
4. Merge two images

This is an additional coding assignment.



II. Feature Points

Any applications?

- Classification

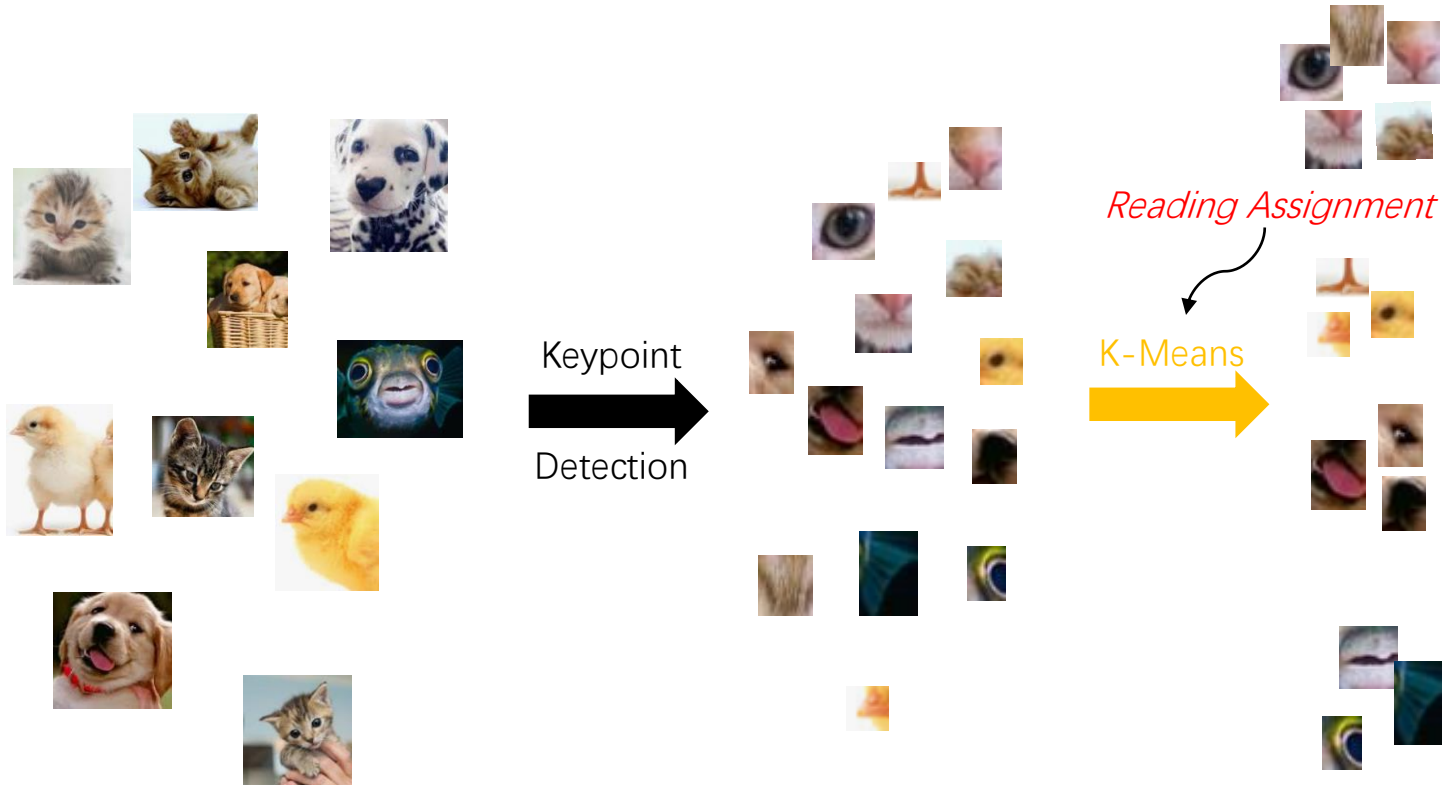
Original image → Get features → Classification

III. Classical CV Procedure

(High Level)

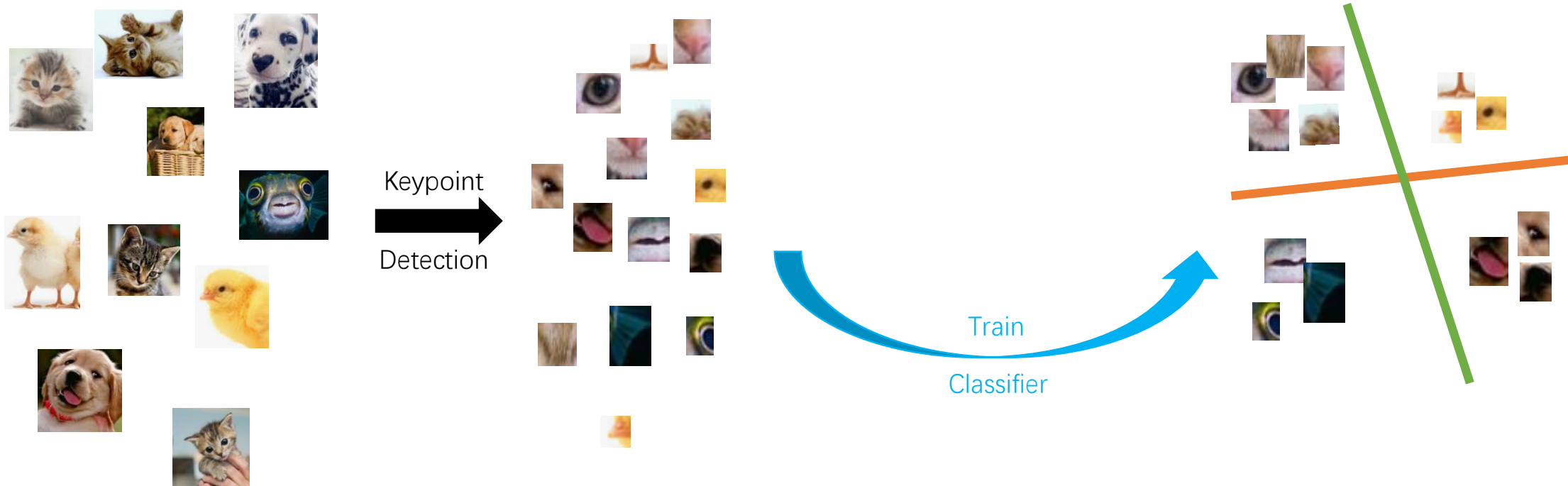
III. Classical CV Procedure

e.g Image Classification (**Bag of Words/BoW**)



III. Classical CV Procedure

e.g Image Classification ([ML](#))



III. Classical CV Procedure

e.g Image Classification (BoW/ML/DL)

