



Final Project Panorama Stitching

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SIFT Basic process

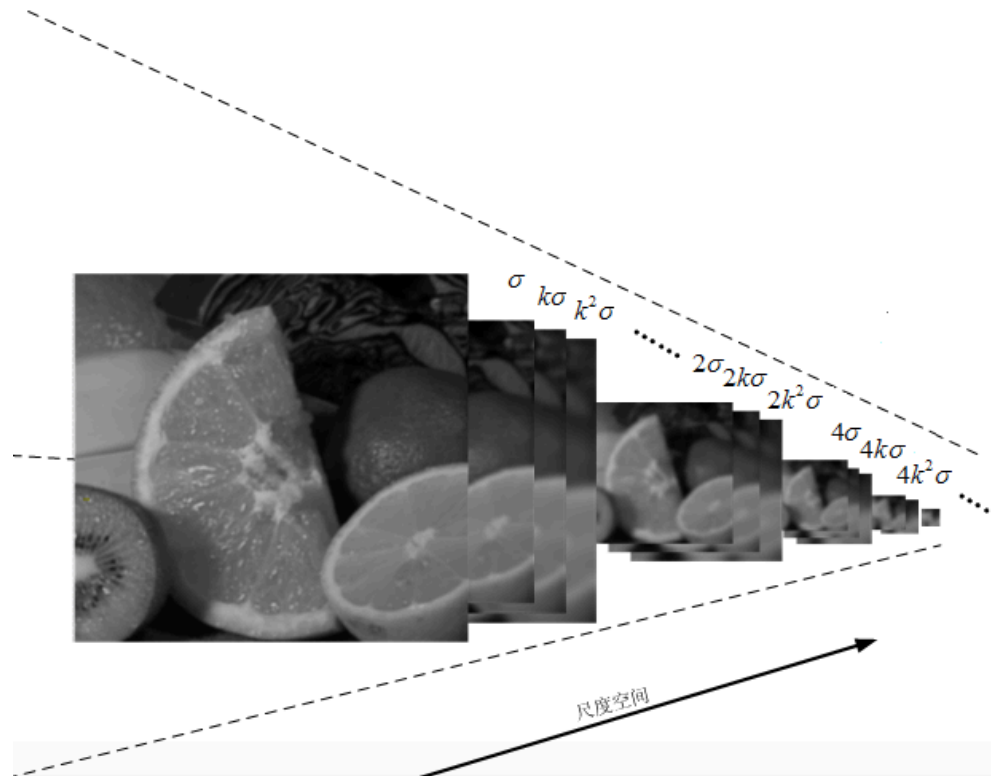
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SIFT — Scale Space & Image Pyramids

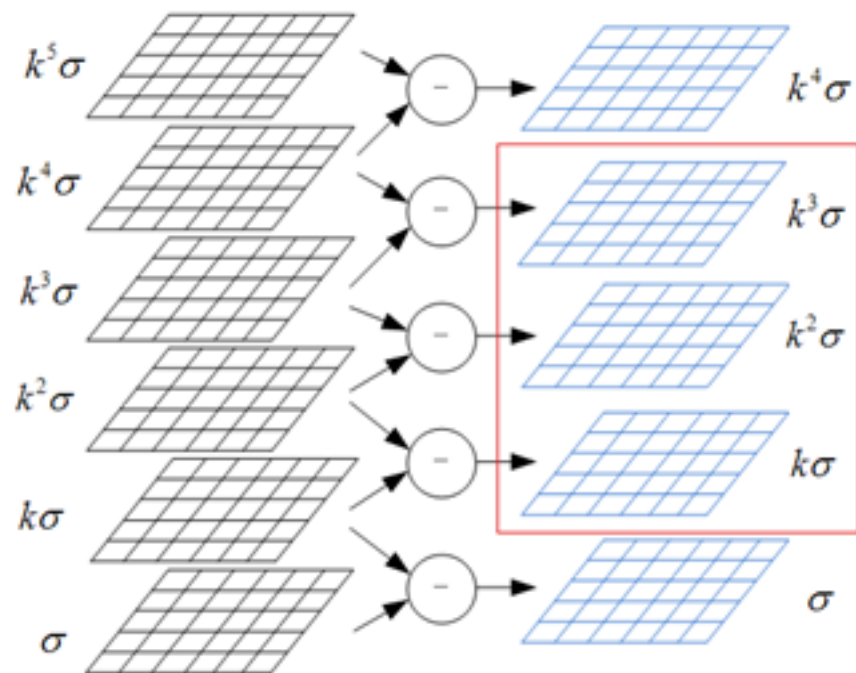
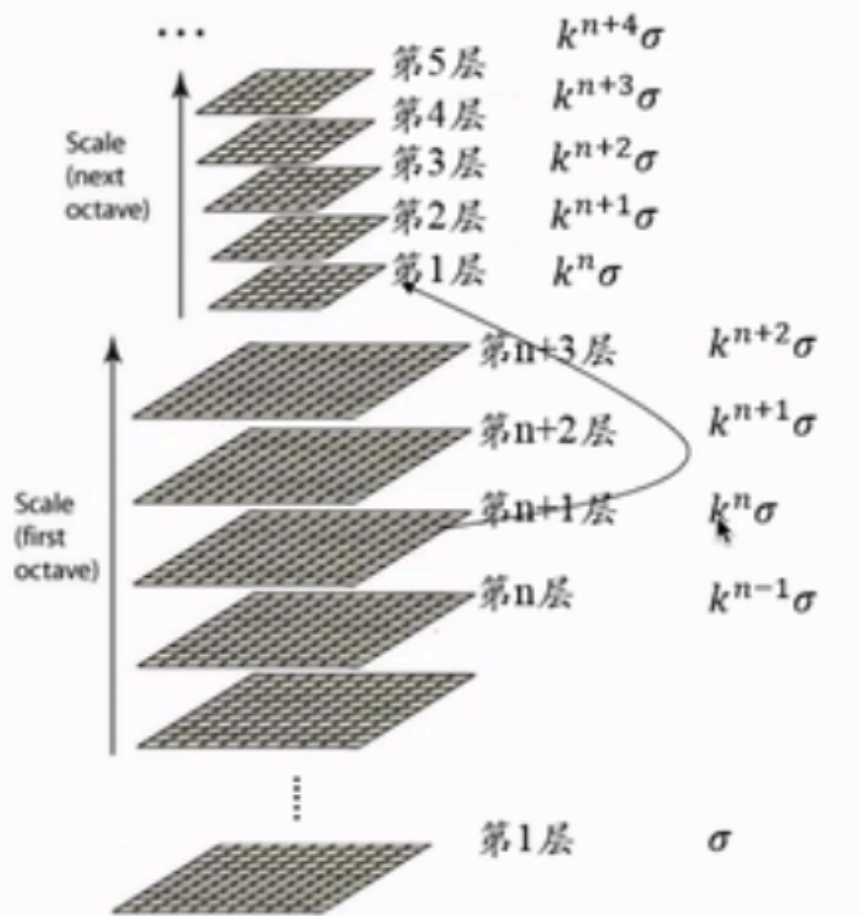
The Gaussian pyramid imitates different scales of the image -- "the depth of the image"

The problem solved by the scale space of the image is how to describe the image at all scales.



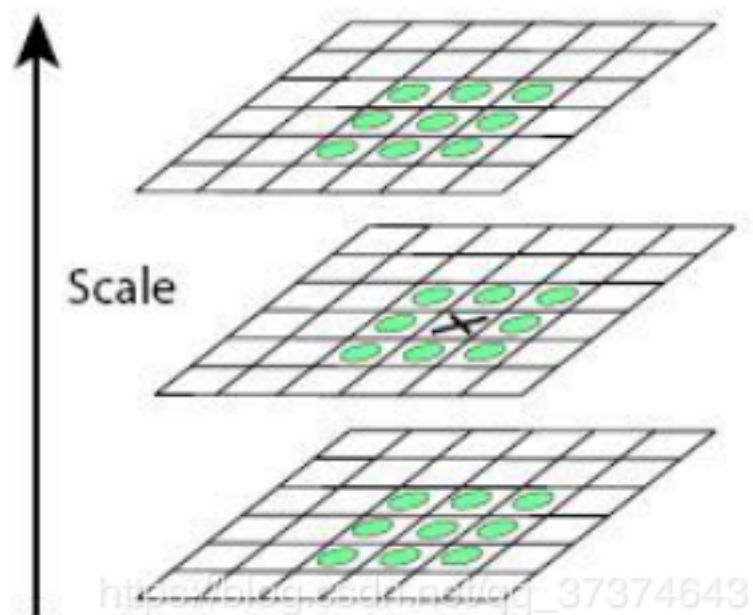


SIFT — Scale Space & Image Pyramids





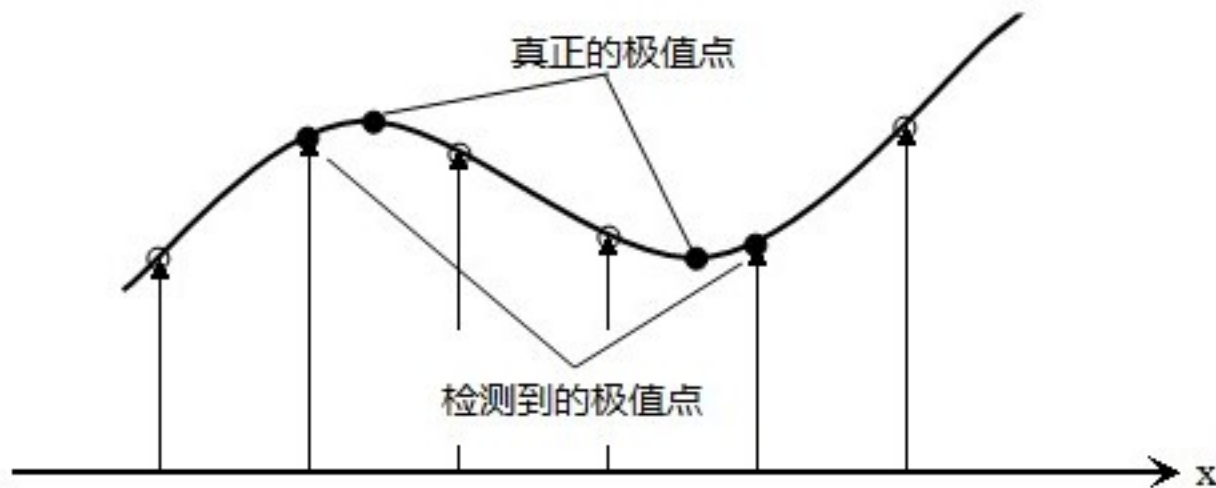
SIFT — Localizing Extrema





SIFT — Localizing Extrema

Sub-pixel interpolation



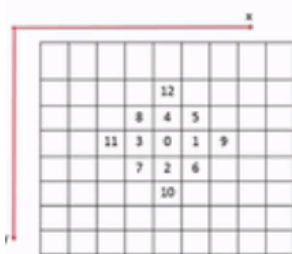


SIFT — Localizing Extrema

Sub-pixel interpolation

$$D(\Delta x, \Delta y, \Delta \sigma) = D(x, y, \sigma) + \begin{bmatrix} \frac{\partial D}{\partial x} & \frac{\partial D}{\partial y} & \frac{\partial D}{\partial \sigma} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \\ \Delta \sigma \end{bmatrix} + \frac{1}{2} \begin{bmatrix} \Delta x & \Delta y & \Delta \sigma \end{bmatrix} \begin{bmatrix} \frac{\partial^2 D}{\partial x^2} & \frac{\partial^2 D}{\partial x \partial y} & \frac{\partial^2 D}{\partial x \partial \sigma} \\ \frac{\partial^2 D}{\partial y \partial x} & \frac{\partial^2 D}{\partial y^2} & \frac{\partial^2 D}{\partial y \partial \sigma} \\ \frac{\partial^2 D}{\partial \sigma \partial x} & \frac{\partial^2 D}{\partial \sigma \partial y} & \frac{\partial^2 D}{\partial \sigma^2} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \\ \Delta \sigma \end{bmatrix}$$

附：有限差分求导法



$$\left(\frac{\partial f}{\partial x} \right) = \frac{f_1 - f_3}{2h} \quad (1)$$

$$\left(\frac{\partial f}{\partial y} \right) = \frac{f_5 - f_7}{2h} \quad (2)$$

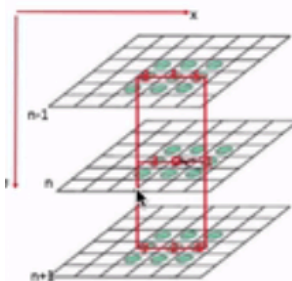
$$\left(\frac{\partial^2 f}{\partial x^2} \right) = \frac{f_1 + f_3 - 2f_0}{h^2} \quad (3)$$

$$\left(\frac{\partial^2 f}{\partial y^2} \right) = \frac{f_5 + f_7 - 2f_0}{h^2} \quad (4)$$

$$\left(\frac{\partial^2 f}{\partial x \partial y} \right) = \frac{(f_1 + f_5) - (f_3 + f_7)}{4h^2} \quad (5)$$

$$D(x) = D + \frac{\partial D^T}{\partial x} \Delta x + \frac{1}{2} \Delta x^T \frac{\partial^2 D^T}{\partial x^2} \Delta x$$

Derivation, substitution, iterative process to get new expression



$$\left(\frac{\partial f}{\partial \sigma} \right) = \frac{f_1 - f_6}{2h} \quad (6)$$

$$\left(\frac{\partial^2 f}{\partial \sigma^2} \right) = \frac{f_1 + f_6 - 2f_0}{h^2} \quad (7)$$

$$\left(\frac{\partial^2 f}{\partial x \partial \sigma} \right) = \frac{(f_1 + f_5) - (f_3 + f_7)}{4h^2} \quad (8)$$



SIFT — Localizing Extrema

Edge effect removal

A flat DoG response peak tends to have a larger main curvature across the edge and a smaller main curvature in the direction of the vertical edge.

$$H(x, y) = \begin{bmatrix} D_{xx}(x, y) & D_{xy}(x, y) \\ D_{xy}(x, y) & D_{yy}(x, y) \end{bmatrix}$$

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

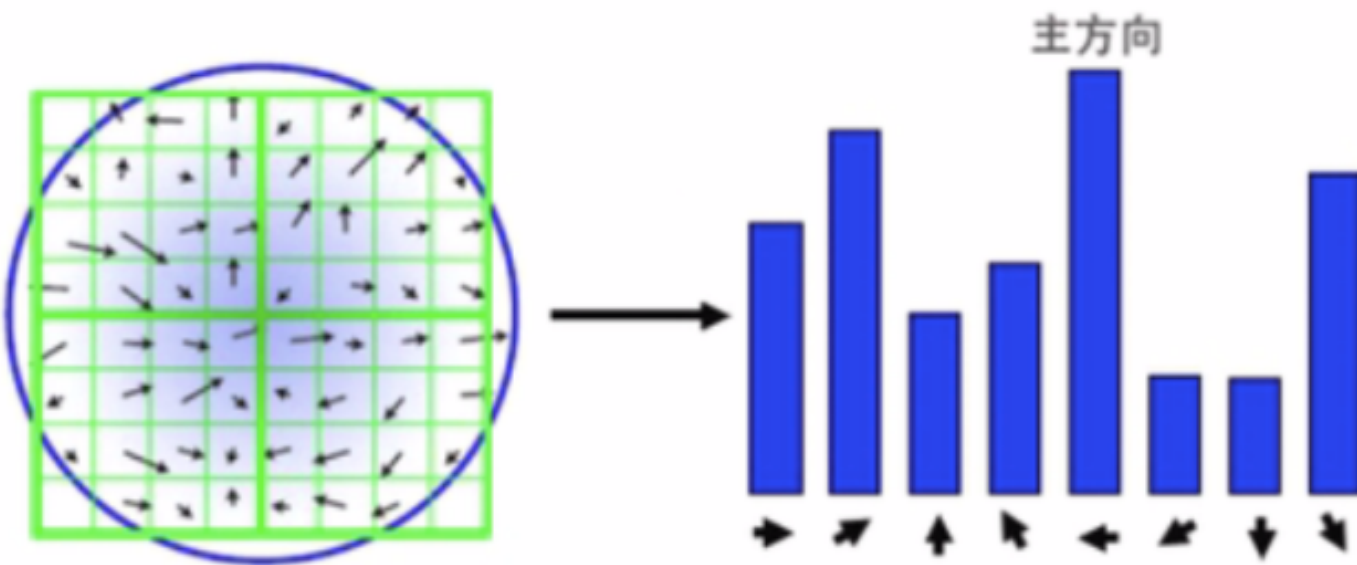
$$Det(H) = D_{xx}D_{yy} - (D_{xy})^2 = \alpha\beta$$

$$\frac{Tr(H)^2}{Det(H)} < \frac{(\gamma + 1)^2}{\gamma}$$

$$\frac{Tr(H)^2}{Det(H)} = \frac{(\alpha + \beta)^2}{\alpha\beta} = \frac{(\gamma + 1)^2}{\gamma}$$



SIFT — Generating Orientations & Descriptors



Robustness requirements:

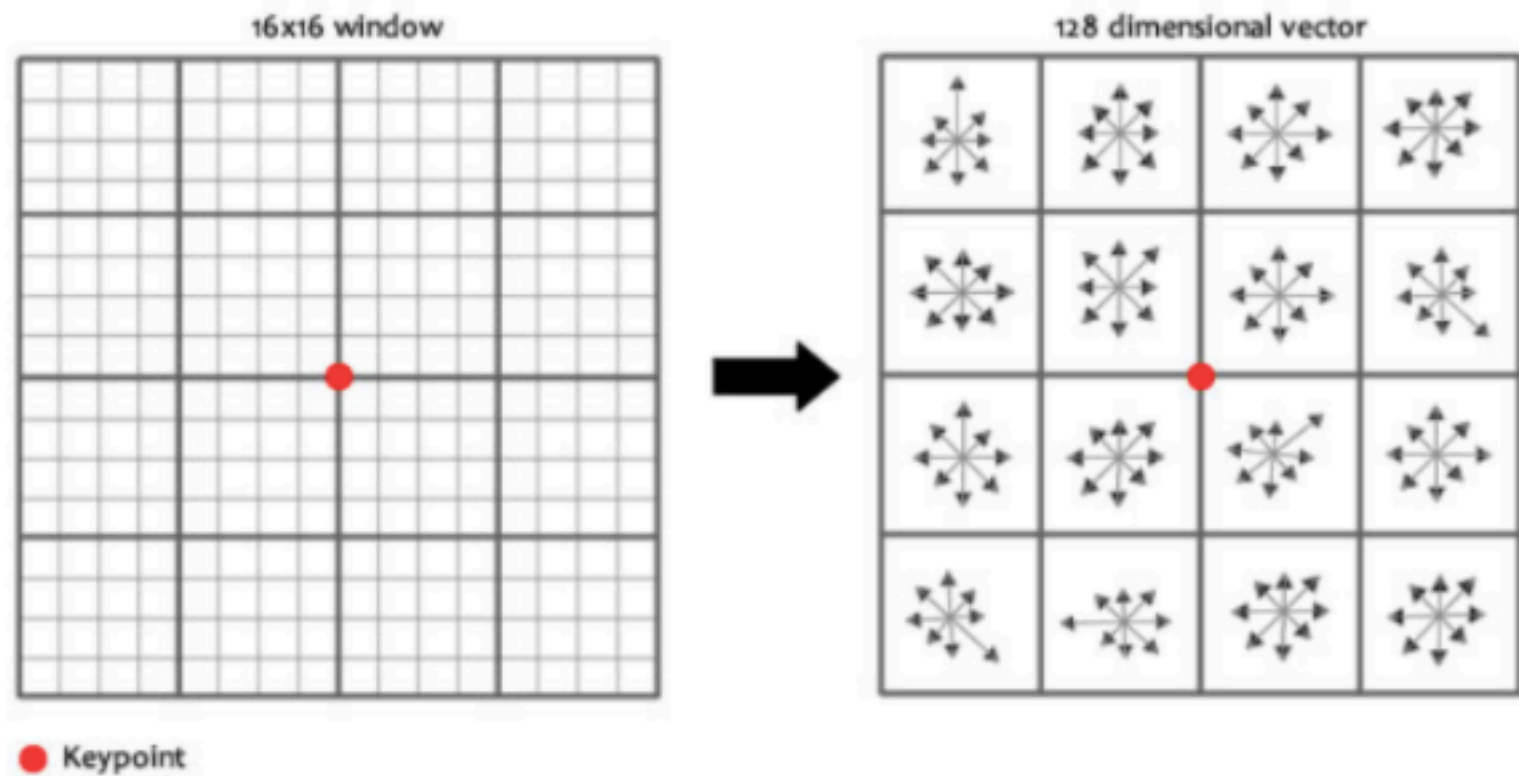
- Secondary direction
- Smoothing

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



SIFT — Generating Orientations & Descriptors

对每个关键点使用 4×4 共 16 个种子点来描述，这样一个关键点就可以产生 128 维的 SIFT 特征向量。



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SIFT — Result

