** Alcantarilla, Pablo Fernández, Adrien Bartoli, and Andrew J. Davison. "KAZE features." European Conference on Computer Vision. Springer Berlin Heidelberg, 2012.

ISL

안재원

NOET

• Feature & Feature detectors

KAZE Feature

OpenCV

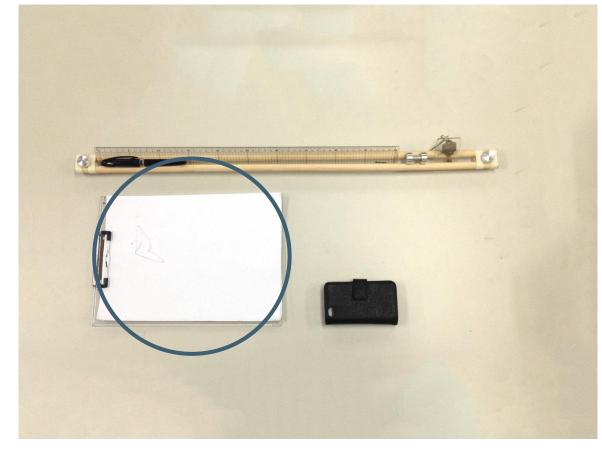
Example Video

Intro



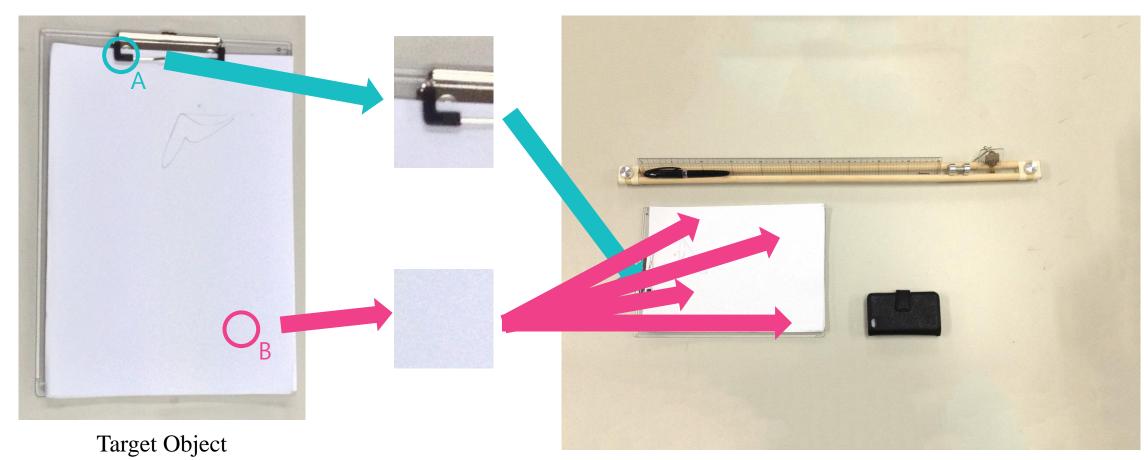
Target Object





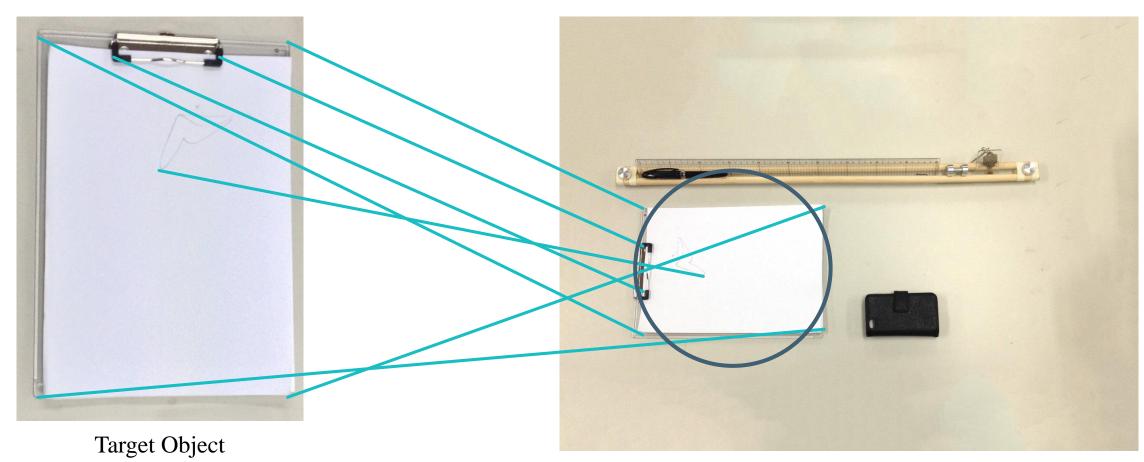
Input Image

Intro



Input Image

Intro



Input Image

0

Feature & Feature detectors

Features

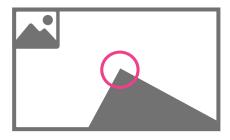
Edge



- Canny
- Sobel
- Prewitt

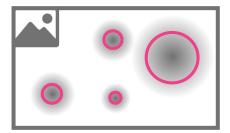


Corner



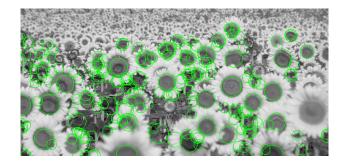
- Harris
- Shi & Tomasi
- SUSAN
- FAST





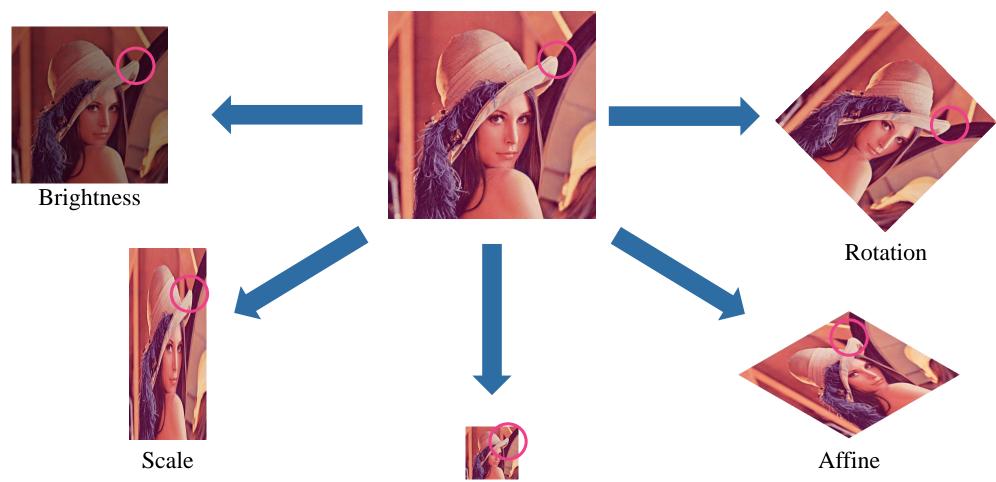
- LoG
- DoG







Good Feature

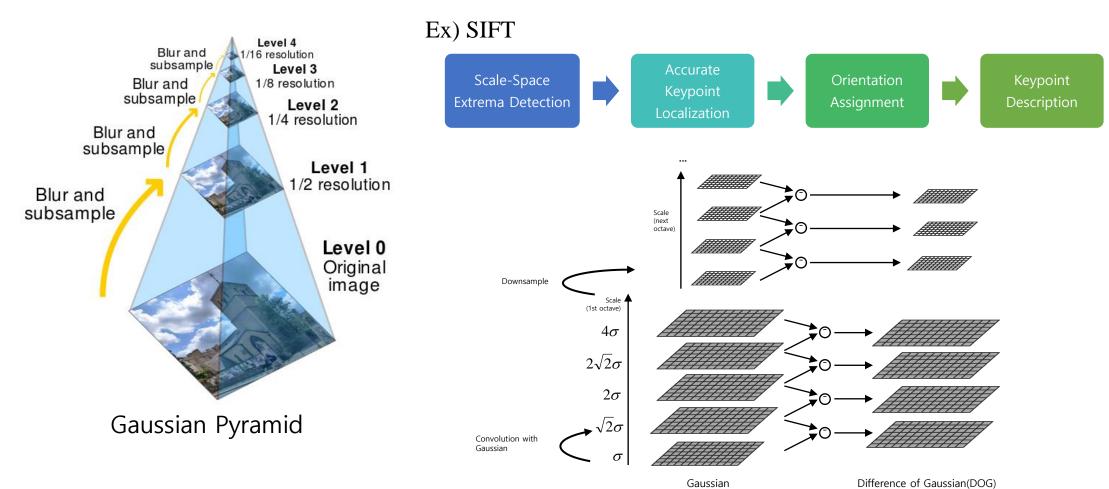


Scale



Intro

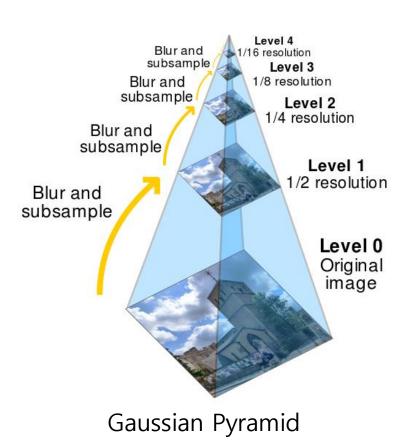
KAZE Feature is multiscale 2D feature detection and description algorithm in nonlinear scale space.





Intro

KAZE Feature is multiscale 2D feature detection and description algorithm in nonlinear scale space.













- Gaussian blurring

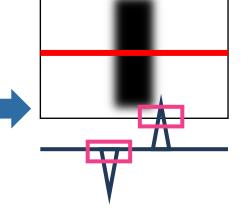


Smooth details

Reducing localization accuracy & distinctiveness

Ex) Canny Edge

- Canny Enhancer
 - Smoothing
 - Finding gradients
 - Estimate edge strength & orientaion
- Non-Max Suppression
 - Choose local maxima



	Normal				
	20	60	80	60	20
	20	70	80	60	20
	20	X	80	X	20
	20	{}	-30 -	50	20
	20	60	80	60	20

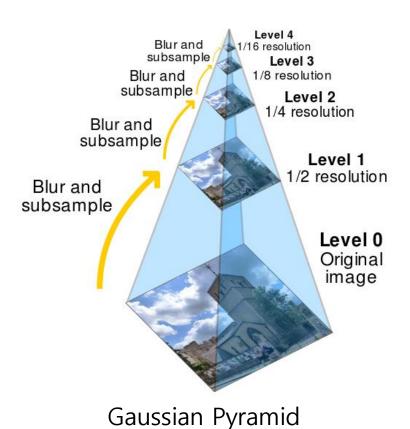
- Hysteresis Threshold
 - Double thresholding





Intro

KAZE Feature is multiscale 2D feature detection and description algorithm in nonlinear scale space.













- Adaptive blurring(nonlinear diffusion filtering)

















Nonlinear Diffusion Filtering

※ Perona, Pietro, and Jitendra Malik. "Scale-space and edge detection using anisotropic diffusion." IEEE Transactions on pattern analysis and machine intelligence 12.7 (1990): 629-639.

- Perona and Malik diffusion equation

Scale parameter
$$t = \frac{1}{2}\sigma^2$$

$$\frac{\partial L}{\partial t} = div(c(x, y, t) \cdot \nabla L)$$
Scalar function of image position

$$c(x, y, t) = g(|\nabla L_{\sigma}(x, y, t)|)$$



k = 5.10



k = 12.75



k = 38.25



k = 89.25

$$g_{1} = \exp\left(-\frac{\left|\nabla L_{\sigma}\right|^{2}}{k^{2}}\right)$$

$$g_{2} = \frac{1}{1 + \frac{\left|\nabla L_{\sigma}\right|^{2}}{k^{2}}}$$
Contrast factor(Level of diffusion)

$$g_{3} = \begin{cases} 1 & , \left| \nabla L_{\sigma} \right|^{2} = 0 \\ 1 - \exp \left(-\frac{3.315}{\left(\left| \nabla L_{\sigma} \right| / k \right)^{8}} \right) & , \left| \nabla L_{\sigma} \right|^{2} > 0 \end{cases}$$



AOS Scheme

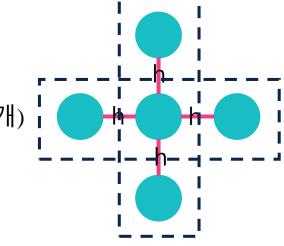
**Weickert, Joachim, BM Ter Haar Romeny, and Max A. Viergever. "Efficient and reliable schemes for nonlinear diffusion filtering." IEEE transactions on image processing 7.3 (1998): 398-410.

 $L^{i+1} = \left(I - \tau \sum_{l=1}^{m} A_{l}(L^{i})\right)^{-1} L^{i}$

- Discretization of the diffusion equation

$$\frac{\partial L}{\partial t} = div(c(x, y, t) \cdot \nabla L)$$

$$\frac{L^{i+1} - L^{i}}{\tau} = \sum_{l=1}^{m} A_{l}(L^{i})L^{i+1}$$
Coordinate $\mathfrak{P} \hookrightarrow (x, y, 27 \mathbb{H})$



- Thomas algorithm

$$\mathbf{A} \cdot L^{i+1} = L^i$$

LU-decomposition

$$LU \cdot L^{i+1} = L^{i}$$

$$L \cdot y = L^{i}$$

$$U \cdot L^{i+1} = y$$

Tridiagonal matrix

Triciagonal matrix
$$\begin{pmatrix} a_1 & b_1 & & & \\ c_1 & a_2 & b_2 & & & \\ & c_2 & \ddots & \ddots & \\ & & \ddots & \ddots & b_{n-1} \\ & & & c_{n-1} & a_n \end{pmatrix}$$

$$-\frac{c_i + c_j}{2h^2} \qquad (j \in \mathbf{N}(i)),$$

$$\sum_{n \in N(i)} \frac{c_i + c_n}{2h^2} \qquad (j = i),$$

$$0 \qquad (else)$$



Computation of the Nonlinear Scale Space

Image Input



Gaussian filtering (Reduce noise)



Image gradient histogram for k



Get Multiscale space image







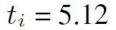


k = 38.25 k = 89.25

 $L^{i+1} = \left(I - \tau \sum_{l=1}^{m} A_{l}(L^{i})\right)^{-1} L^{i}$

70% Percentile 의 값을 Contrast parameter k로 사용한다.







 $t_i = 20.48$



 $t_i = 81.92$



$$t_i = 130.04$$



$$t_i = 206.42$$





Feature Detection

**Brown, Matthew, and David G. Lowe. "Invariant Features from Interest Point Groups." BMVC. No. s 1. 2002.

XLindeberg, Tony. "Feature detection with automatic scale selection." *International journal of computer vision* 30.2 (1998): 79-116.

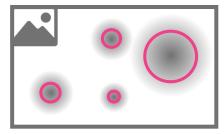
- Hessian Matrix

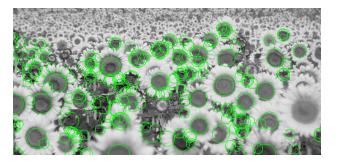
$$H = egin{bmatrix} L_{xx} & L_{xy} \ L_{xy} & L_{yy} \end{bmatrix}$$

Local minimum: D > 0 and $L_{xx} > 0$

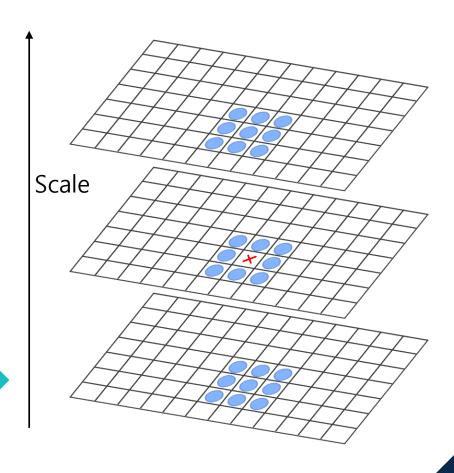
Local maximum: D > 0 and $L_{xx} < 0$

Blob





$$L_{Hessian} = \sigma^2 (L_{xx} L_{yy} - L_{xy}^2)$$





Feature Description

- Finding the Dominant Orientation

Find Circular area of radius 6σ



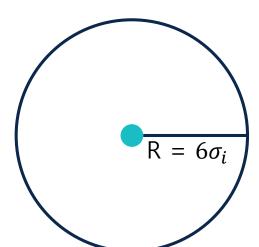
Get Gaussian weight

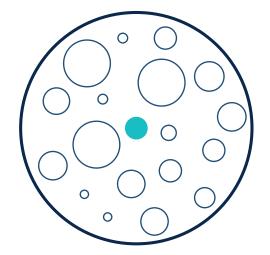


Sliding area & To compare value

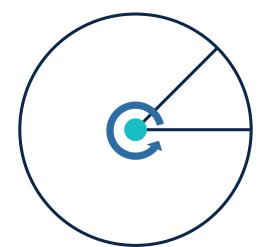


Finding the dominant orientation

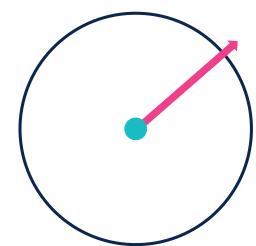




$$G = e^{-\frac{dx^2 + dy^2}{2\sigma_i}}$$



$$\left(\sum GL_{x}\right)^{2} + \left(\sum GL_{y}\right)^{2}$$



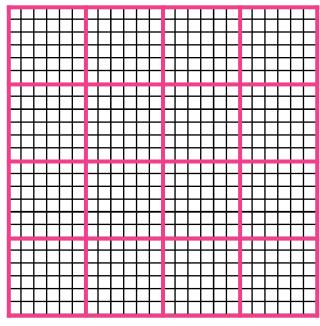
$$O = \tan^{-1} \left(\frac{\sum GL_y}{\sum GL_x} \right)$$



Feature Description

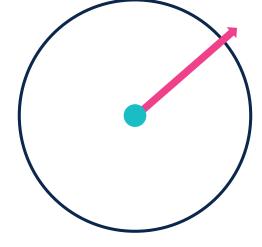
- Building the Descriptor

$$L_x L_y$$



 $24\sigma \times 24\sigma$

- This grid is divided into 4x4 sub-regions.
- Opverlap of 2σ

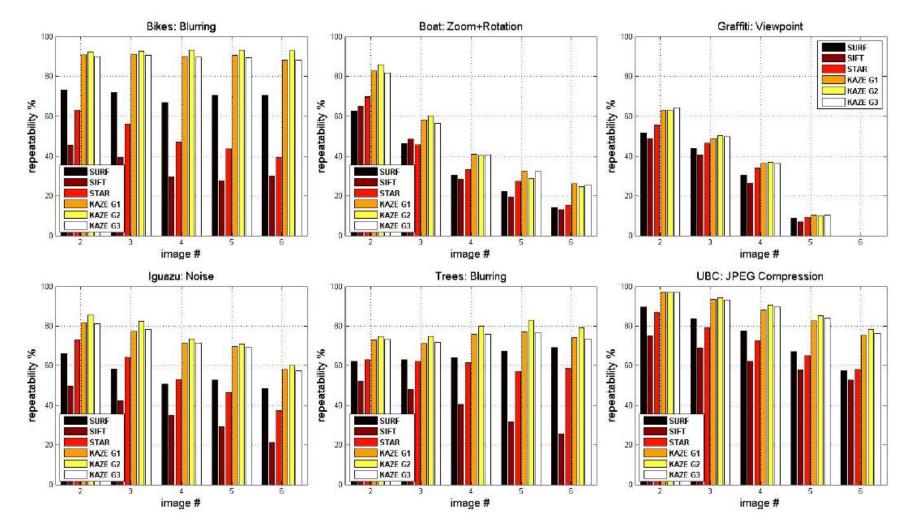


- Descriptor vector

$$d_{v} = \left[G_{c} \sum GL_{x}, G_{c} \sum GL_{y}, G_{c} \sum |GL_{x}|, G_{c} \sum |GL_{y}|\right]$$

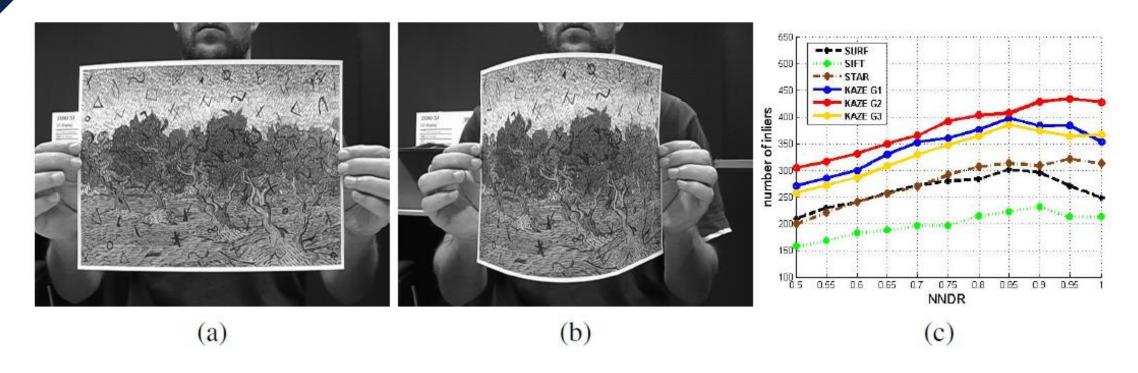


Experimental Result





Experimental Result





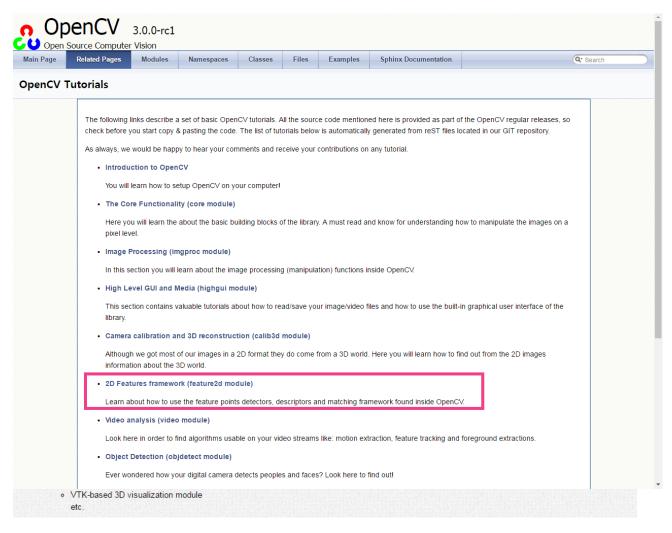
Experimental Result

KAZE	UBC 1	Trees 6
Nonlinear Scale Space	1.14	1.53
Feature Detection	0.68	0.93
Feature Description	0.38	0.20
Total Time	2.20	2.66
SURF	0.89	0.63
SIFT	2.66	2.77
STAR	0.25	0.32
Image Resolution	800×640	1000×700
Number of Keypoints	1463	765



OpenCV

OpenCV 3.x



- features2d.hpp
- KAZE/AKAZE 2종류가 추가됨.

OpenCV



OpenCV 3.x

```
| □#include ⟨opencv2/features2d.hpp⟩
    #include opencv2/imgcodecs.hpp>
     #include opencv2/opencv.hpp>
     #include opencv2/highgui.hpp>
     #include <vector>
     #include <iostream>
 8 ⊟using namespace std;
    using namespace cv;
     const float inlier threshold = 2.5f; // Distance threshold to
     const float nn_match_ratio = 0.8f; // Nearest neighbor match
14 □ int main(void)
          Mat img1 = imread("graf1.png", IMREAD_GRAYSCALE);
          Mat img2 = imread("graf3.png", IMREAD GRAYSCALE);
18
          Mat homography;
          FileStorage fs("H1to3p.xml", FileStorage::READ);
fs.getFirstTopLevelNode() >> homography;
21
          vector(KeyPoint) kpts1, kpts2;
          Mat desc1, desc2;
26
          Ptr\AKAZE\> akaze = AKAZE::create();
akaze-\detectAndCompute(img1, noArray(), kpts1, desc1);
akaze-\detectAndCompute(img2, noArray(), kpts2, desc2);
```

```
BFMatcher matcher(NORM_HAMMING);
vector(vector(DMatch)) nn_matches;
matcher.knnMatch(desc1, desc2, nn_matches, 2);
   vector(KeyPoint) matched1, matched2, inliers1, inliers2;
vector(DMatch) good matches;
for(size_t i = 0; i < nn_matches.size(); i++) {
    DMatch first = nn_matches[i][0];
    float dist1 = nn_matches[i][0] distance;
    float dist2 = nn_matches[i][1].distance;</pre>
                 if(dist1 < nn_match_ratio * dist2) {
  matched1.push_back(kpts1[first.queryIdx]);
  matched2.push_back(kpts2[first.trainIdx]);</pre>
    Infor(unsigned i = 0; i < matched1.size(); i++) {
    Mat col = Mat::ones(3, 1, 0, 64F);
    col.at(double)(0) = matched1[i].pt.x;
    col.at(double)(1) = matched1[i].pt.y;</pre>
                 \begin{array}{ll} \text{col} &= \text{homography} \, \star \, \text{col}; \\ \text{col} & /= \text{col.at}\langle \text{double}\rangle(2); \\ \text{double dist} &= & & \text{sqrt}(\text{pom}(\text{col.at}\langle \text{double}\rangle(0) - \text{matched2}[i].pt.x, 2) + \\ \text{pom}(\text{col.at}\langle \text{double}\rangle(1) - \text{matched2}[i].pt.y, 2)); \\ \end{array} 
                 if(dist \langle inlier_threshold) {
  int new_i = static_cast(int)\(inliers1.size());
  inliers1.push_back(matched1[i]);
  inliers2.push_back(matched2[i]);
  good_matches.push_back(DMatch(new_i, new_i, 0));
}
Mat res;
drawMatches(img1, inliers1, img2, inliers2, good_matches, res);
namedWindow("input1", WINDOW_AUTOSIZE);
namedWindow("input2", WINDOW_AUTOSIZE);
      imshow("input1", img1);
imshow("input2", img2);
       imshow("res.png", res);
```

- detectAndCompute(input, mask, key-point, descriptor)

OpenCV



OpenCV 3.x

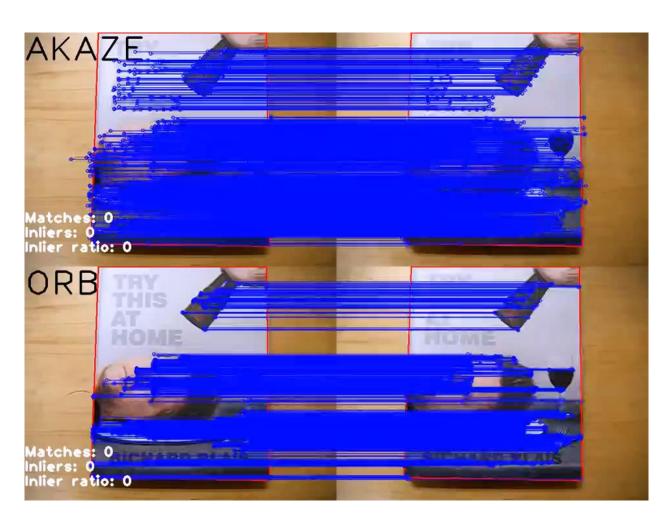


OA

Example Video

A-Kaze vs ORB Feature

Accelerated-KAZE



- Matches 626
- Inlier ratio 0.58
- Keypoints 1117

- Matches 504
- Inlier ratio 0.56
- Keypoints 1112
- faster

Q&A

