

Homework 2

Computer Vision 2017 Spring

2017.4.27

Feature Detection

- find features correspondences/compute homography matrix.
- SIFT –Scale Invariant Feature Detection
 - detect key points in the image and describe the points as 128-dimensional features.
- Check Ch.6 、 7 for more details of SIFT.

OpenCV



- Highly optimized C++ implementation of many CV algorithms
- Cross platform
- Many bindings (Python, Matlab, Java . . .)

<http://opencv.org>

Install

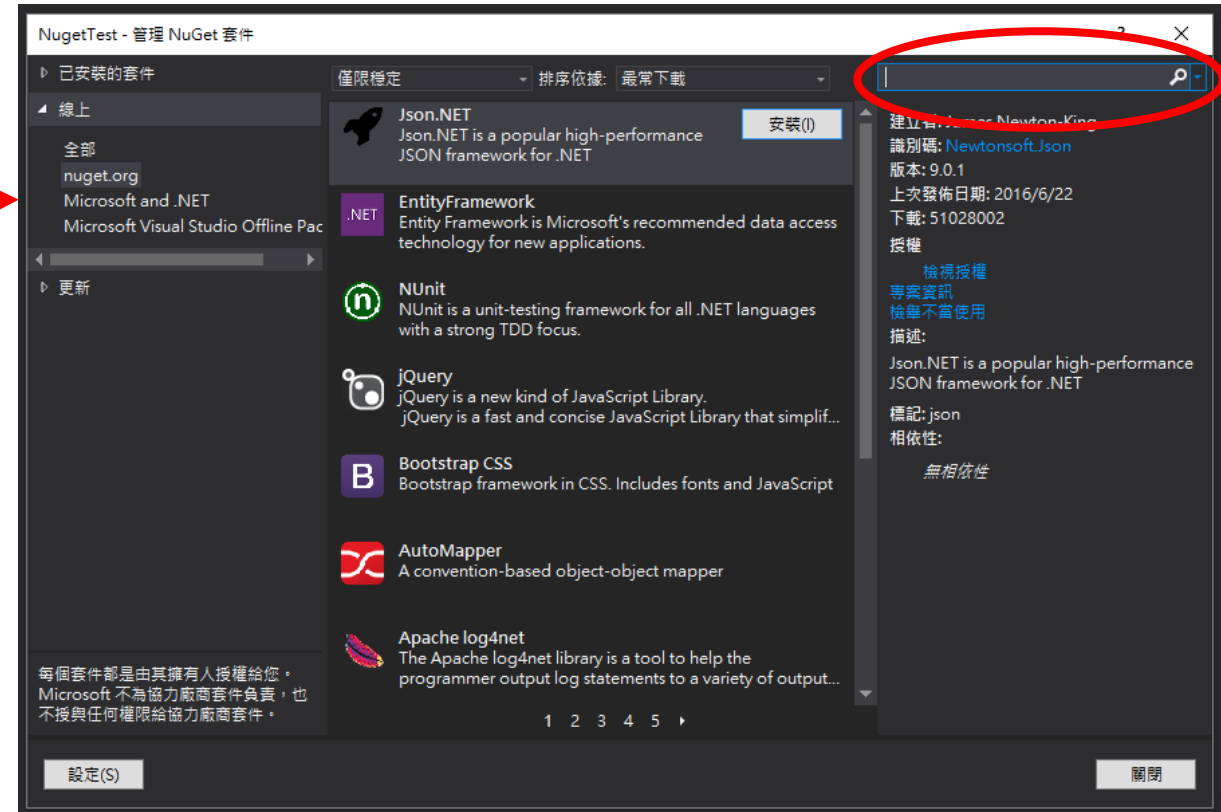
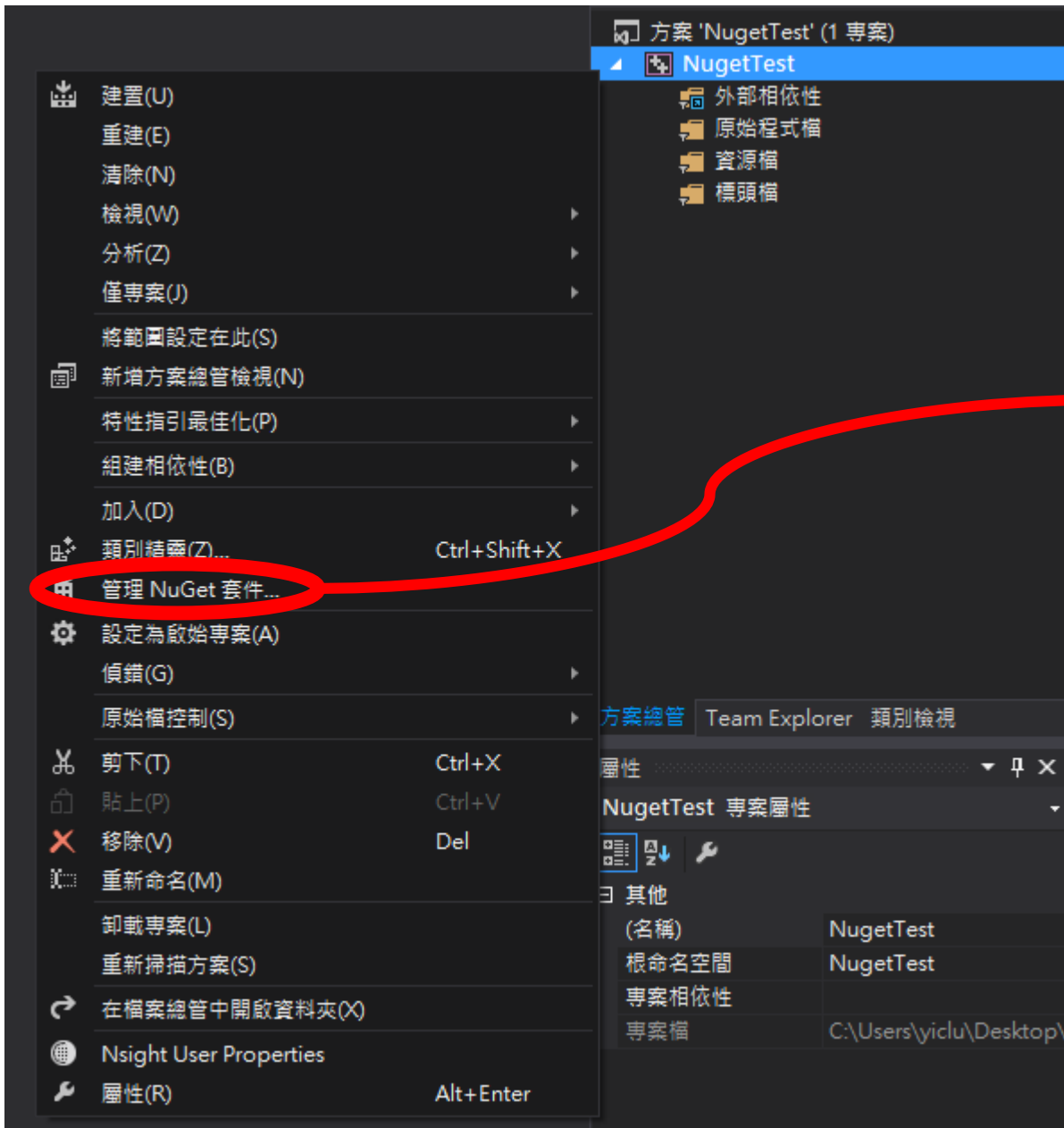
- VS 2013 、 VS 2015

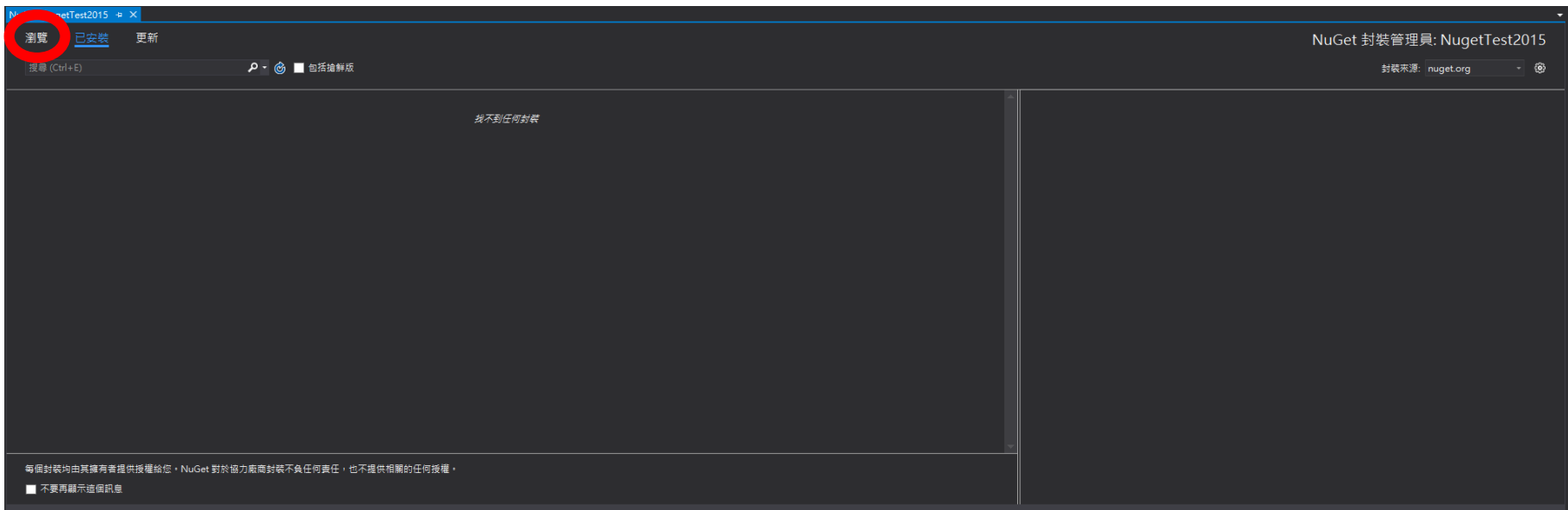
1 , Open your visual Studio

2 , New a Empty Project

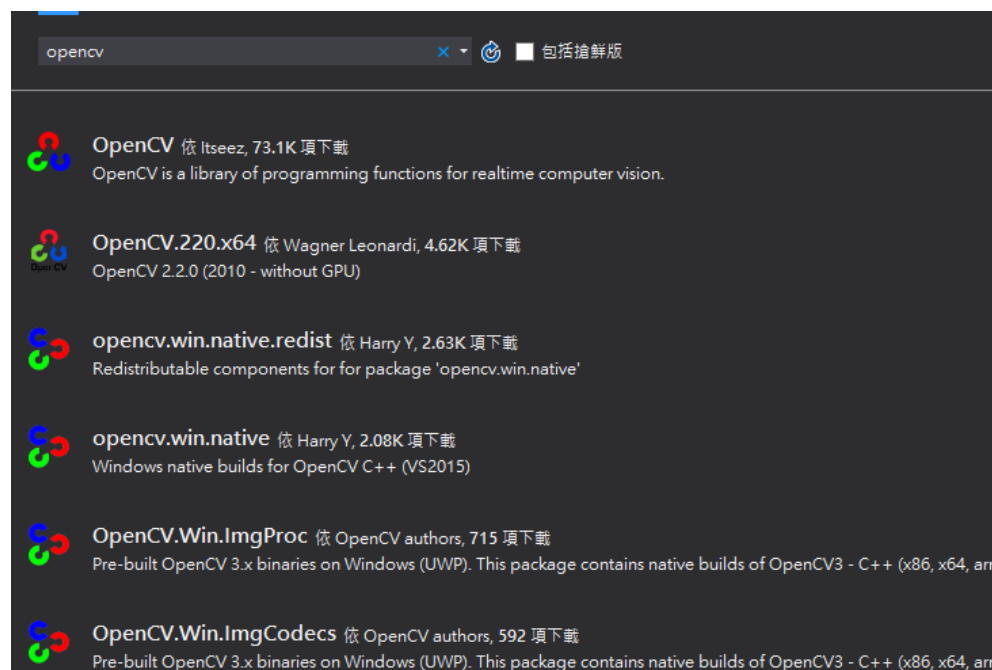
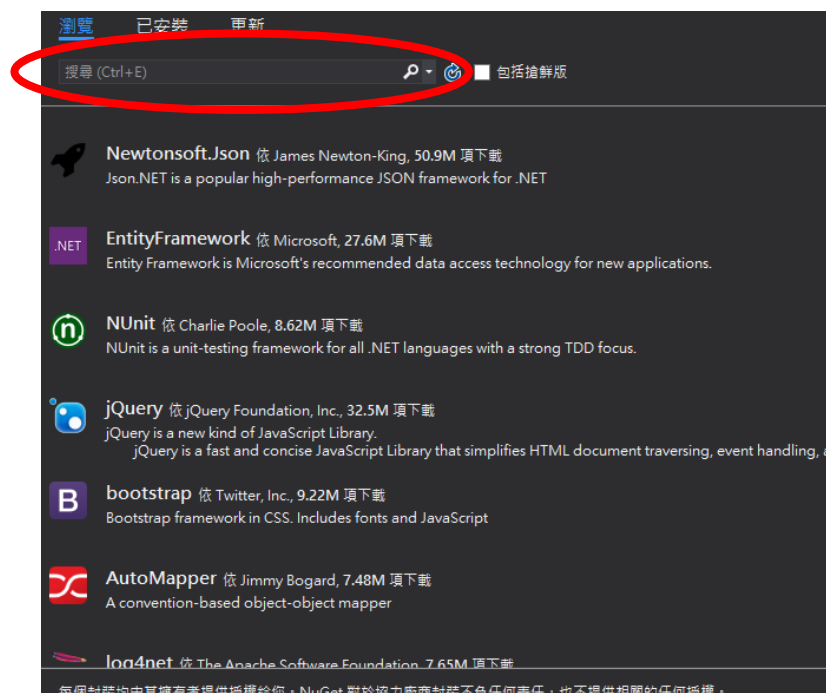
3 , Install OpenCV

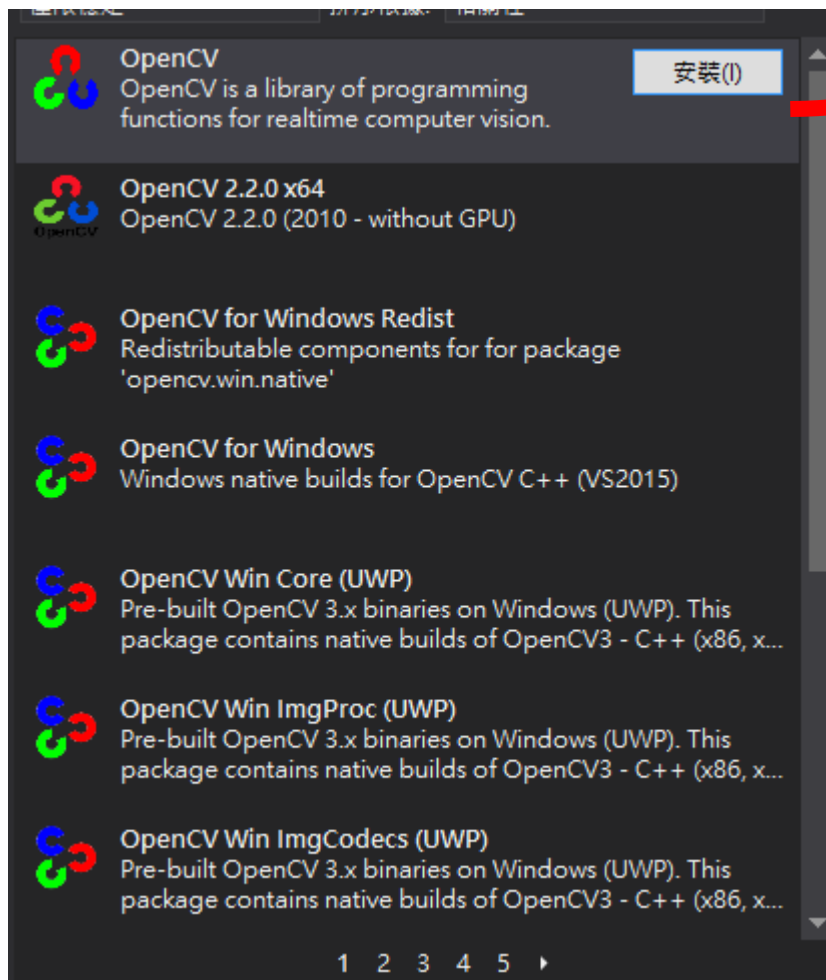
VS 2013



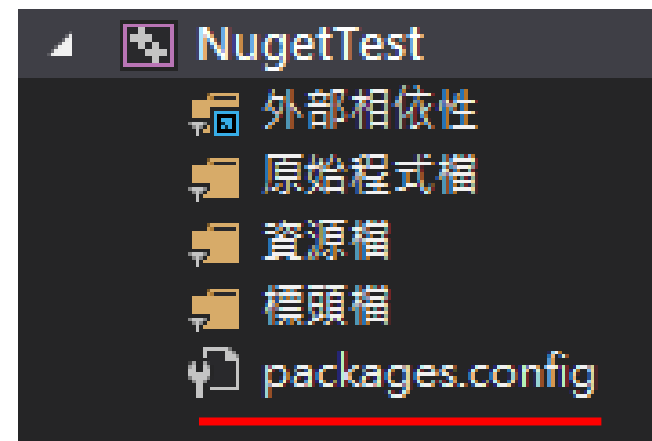


VS 2015





VS 2013
VS 2015



SIFT in OpenCV

```
//#include <opencv/highgui.h> // For VS2013  
#include <opencv2/highgui/highgui.hpp> // For VS2015  
#include <opencv2/nonfree/features2d.hpp>
```


SIFT in OpenCV

```
// SIFT feature detector and feature extractor
SiftFeatureDetector detector;
SiftDescriptorExtractor extractor;

// Feature detection
vector<KeyPoint> keypoints;
detector.detect(Image, keypoints);
cout << "Keypoints' number = " << keypoints.size() << endl;

// Feature descriptor computation
Mat descriptor;
extractor.compute(Image, keypoints, descriptor);
cout << "Descriptor's size = " << descriptor.size() << endl;

// Feature display on image
Mat feature;
drawKeypoints(Image, keypoints, feature, Scalar(255, 255, 255), DrawMatchesFlags::DRAW_RICH_KEYPOINTS);
imwrite("result.bmp", feature);
```

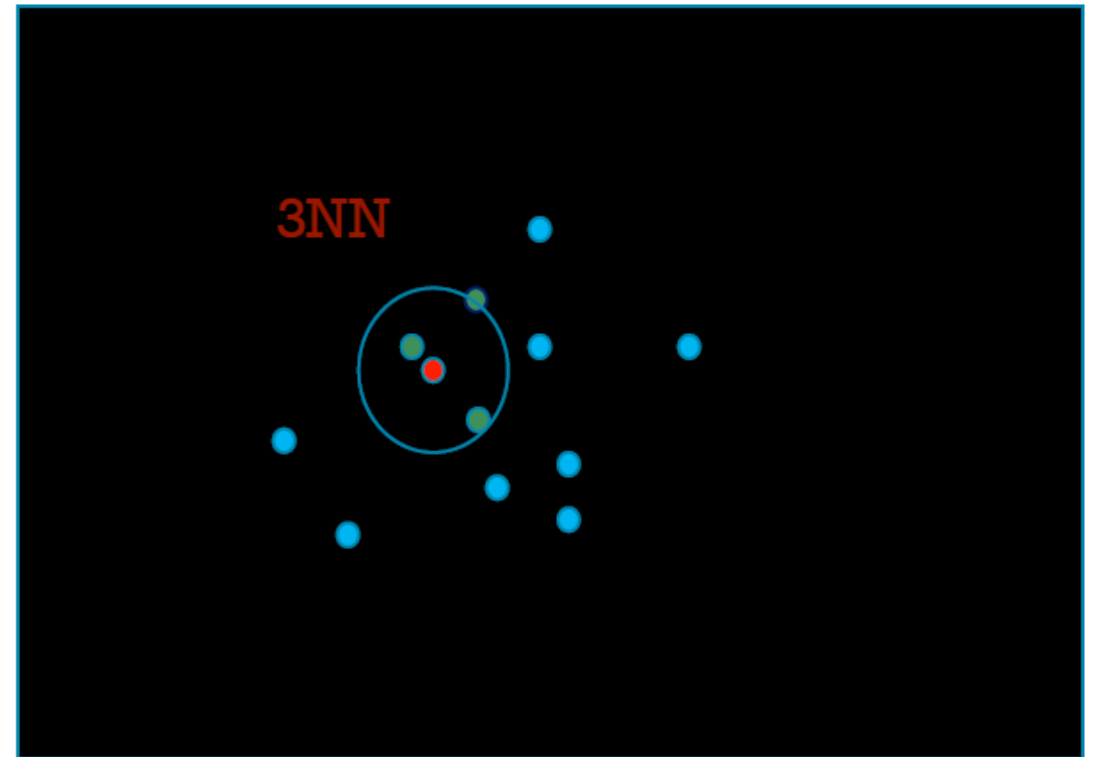
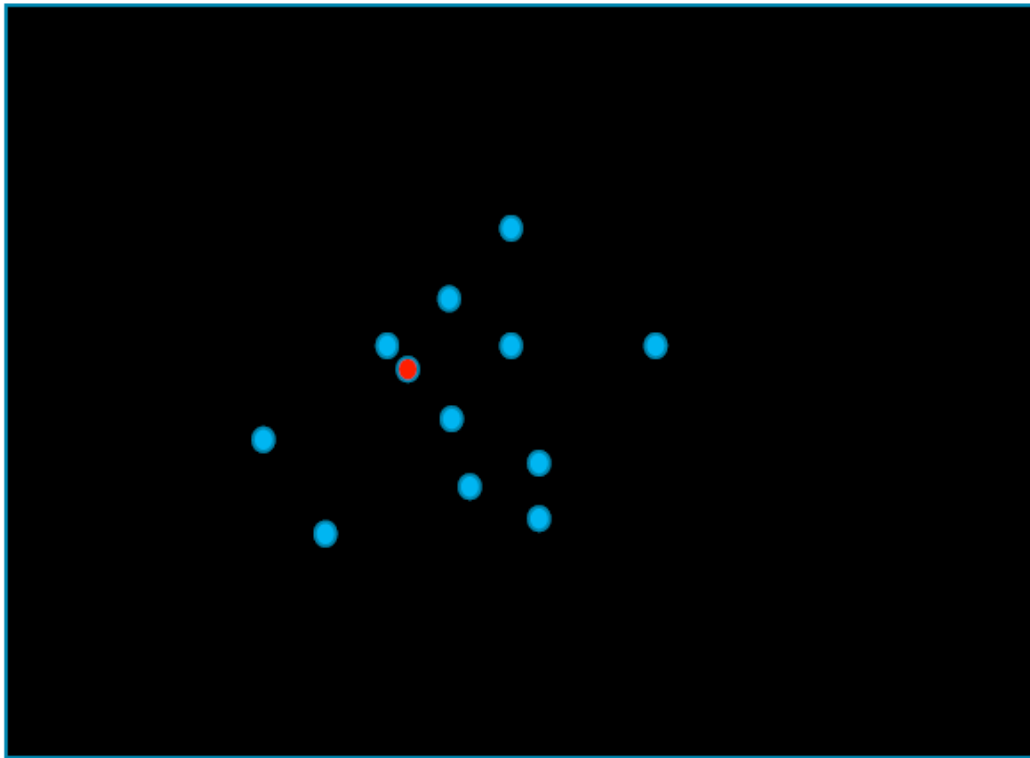
SIFT

- Result
 - Feature image output in result.bmp
 - Feature points stored in keypoints
 - Descriptors stored in descriptor



KNN

- K-Nearest Neighbor
 - Find the K closest neighbors to the target.



RANSAC

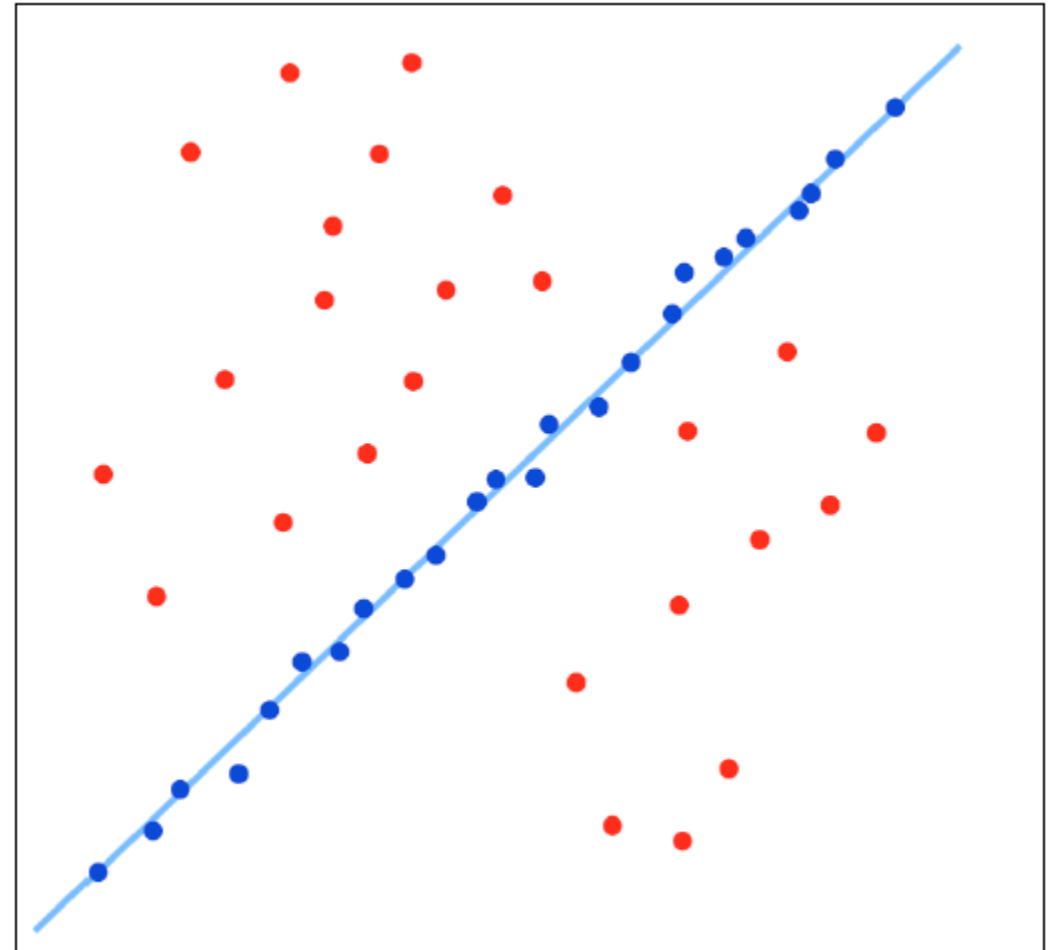
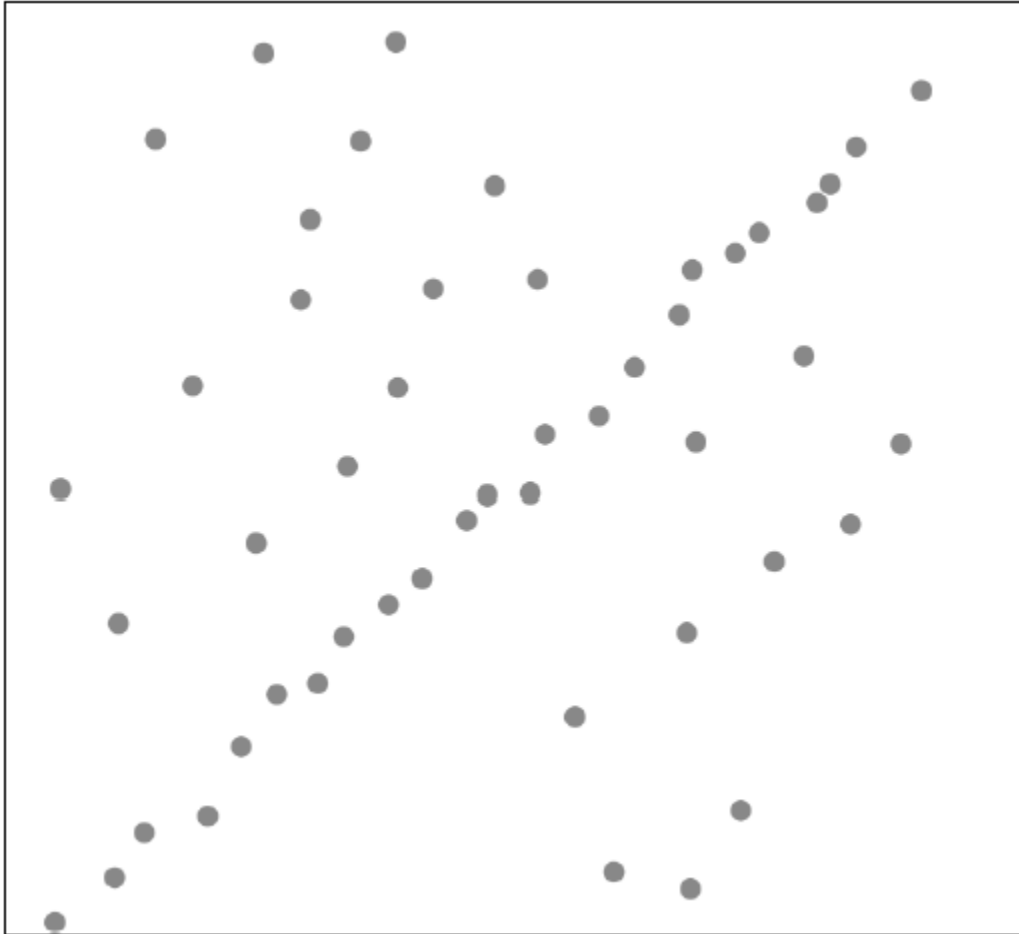
Random Sample Consensus

Input : M data points;

1. Randomly select N data points as inliers S . ($N \ll M$)
2. Fit a model \mathcal{M} to S .
3. Test all data points against \mathcal{M} , add the points consistent with \mathcal{M} to S , which is called a consensus set.
4. If $|S|$ is larger than ever, mark \mathcal{M} as the best estimated model \mathcal{M}^* .
5. If some stopping criterion is satisfied, end
6. Else go to step 1.

Note that you can re-estimate the models with the consensus sets.

RANSAC



Recover Homography

- Construct a linear system as: $p' = Hp$, where p' and p are correspondence points.
- Follow the Lecture 7 page 6~8. You may try Affine mappings(DOF=6) or Projective mappings(DOF=8).

Recover Homography

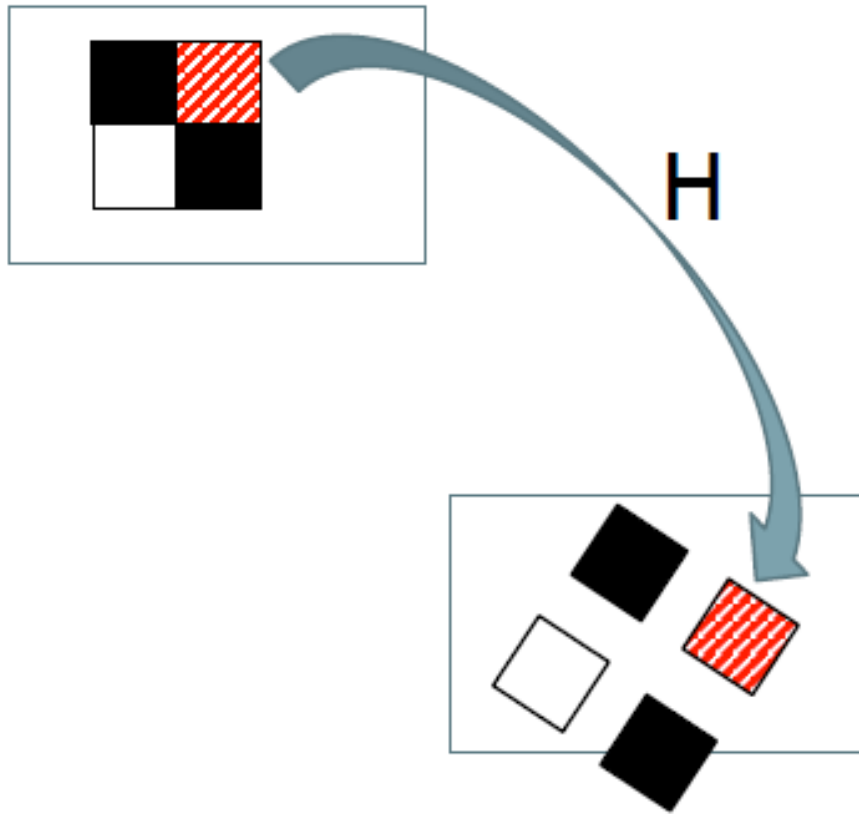
- Solve $Ax=0$
- `cv::eigen`
 - (用法可查詢http://docs.opencv.org/2.4/modules/core/doc/operations_on_arrays.html#eigen)

$$\begin{bmatrix} wx \\ wy \\ w \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix} * \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

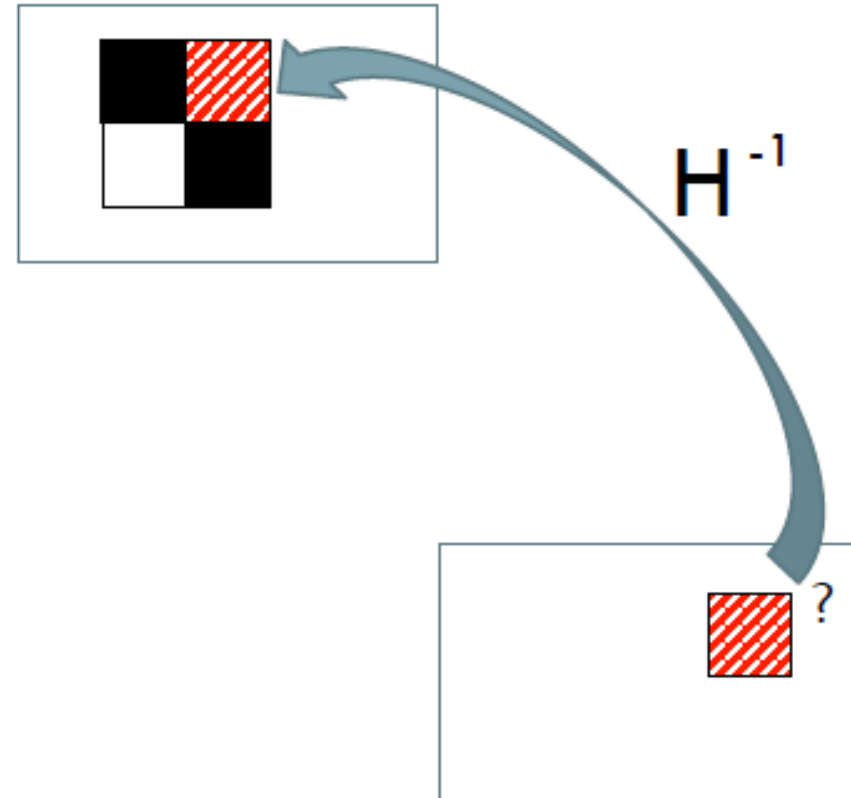
$$\begin{bmatrix} X_1 & Y_1 & 1 & 0 & 0 & 0 & -x_1X_1 & -x_1Y_1 & -x_1 \\ 0 & 0 & 0 & X_1 & Y_1 & 1 & -y_1X_1 & -y_1Y_1 & -y_1 \\ X_2 & Y_2 & 1 & 0 & 0 & 0 & -x_2X_2 & -x_2Y_2 & -x_2 \\ 0 & 0 & 0 & X_2 & Y_2 & 1 & -y_2X_2 & -y_2Y_2 & -y_2 \\ X_3 & Y_3 & 1 & 0 & 0 & 0 & -x_3X_3 & -x_3Y_3 & -x_3 \\ 0 & 0 & 0 & X_3 & Y_3 & 1 & -y_3X_3 & -y_3Y_3 & -y_3 \\ X_4 & Y_4 & 1 & 0 & 0 & 0 & -x_4X_4 & -x_4Y_4 & -x_4 \\ 0 & 0 & 0 & X_4 & Y_4 & 1 & -y_4X_4 & -y_4Y_4 & -y_4 \end{bmatrix} * \begin{bmatrix} A_{11} \\ A_{12} \\ A_{13} \\ A_{21} \\ A_{22} \\ A_{23} \\ A_{31} \\ A_{32} \\ A_{33} \end{bmatrix} = 0$$

Warp the Images

■ Forward



■ Backward



Warp the Images

- Warp the object image without the image background.
- In this assignment, you may cut the background by :

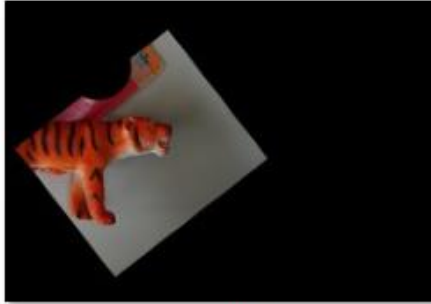
```
if ( pixel-color != black ){  
    warping;  
}  
else{  
    skip;  
}
```

Algorithm(for reference)

1. SIFT feature detection
2. For each feature point in the object image:
Find KNN points in the image (according to the descriptors)
3. Randomly select 4 (or more) points in the object image:
For all $k*k*k*k$ possible match sets:
Construct the Homography matrix
Compute inliers and outliers
If $> \text{Threshold}$, output
else repeat 3
4. Use the recovered Homography matrix to warp the image.

Requirements

- Input :



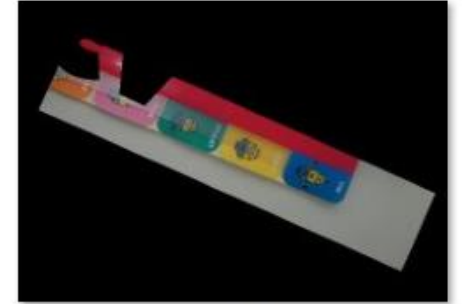
puzzle1



puzzle2



puzzle3



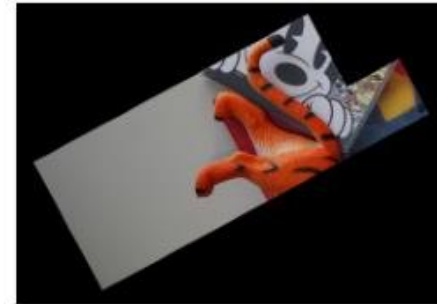
puzzle4



puzzle5



puzzle6



puzzle7



sample



target

Requirements

- Output :



Requirements

- 此次作業只可以使用image read/save, pixel set/get, matrix basic operation(+-* /, 反矩陣等)，內建eigenvalue及eigenvector函式，請勿使用其他功能太強大的函式(如有要使用，不確定的，再詢問助教)
- Homography matrix 要自己找，KNN、RANSAC、Warping都要自己寫，請勿呼叫內建函式直接完成
- 可使用C/C++以外的語言，但限制同上

Scores

- Find object in target image and warp it.
 - 1 object successful 75 %
 - 2 objects successful +10 %
 - Another testing dataset when demo +10 %
- Other interesting things (ex. Speed up.....) +15 % at most

Deadline

- 期限: 2017/05/18 (四) 11:59:59 pm
- 請將作業壓縮並以學號命名: ex 0987654-hw2.zip
 - 資料夾內包含:
 - 1. 學號-hw2-report.pdf
 - 2. code 或 完整專案
 - 3. Result image
- 上傳至e3
- Code需要加上對應流程註解
- Report包含程式流程說明 & 執行方式 & 自己多做了哪些功能&其他你想寫的...