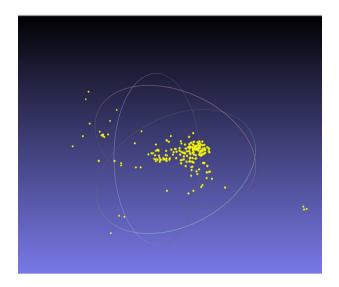
HUYNH Cong Lap – Travail Seul 11419778 – Matching and Reconstruction

In the last TP, we get: Distortion Matrix, Parameter intrinsic - Camera matrix (K) and parameters extrinsic (R|t). This information will be used in this TP, to reconstruct the matching point from 2D to 3D



Before doing this TP, we've already had corrected photo from last TP. This will give us precision.

Process in this TP: 3 main steps

1. Feature detection and description \rightarrow **2. Matching Feature between 2 images** \rightarrow **Find Good Matched** \rightarrow **3. Triangulation between 2 matched points (Hartley and Andrew Zisserman)** \rightarrow **Regain the Point in 3D.**

1. Feature detection and description

Feature Matching with FLANN

Use the <u>cv::FlannBasedMatcher</u> interface in order to perform a quick and efficient matching by using the <u>Clustering and Search in Multi-Dimensional Spaces</u> module

Use the **cv::FeatureDetector** interface in order to find interest points. Specifically:

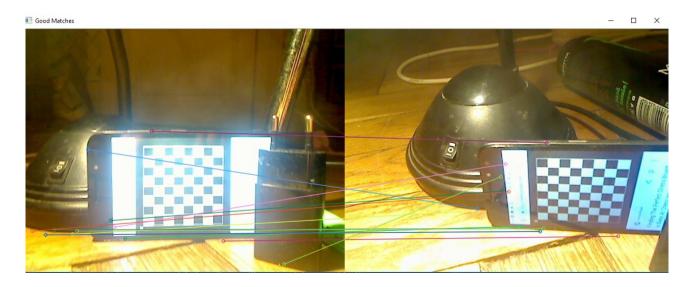
- Use the cv::xfeatures2d::SURF::detect to perform the detection process
- Use the function cv::drawKeypoints to draw the detected keypoints

```
//-- Step 1: Detect the keypoints using SURF Detector, compute the descriptors
int minHessian = 400;
Ptr<SURF> detector = SURF::create();
detector->setHessianThreshold(minHessian);
std::vector<KeyPoint> keypoints_1, keypoints_2;
Mat descriptors_1, descriptors_2;
detector->detectAndCompute(img_1, Mat(), keypoints_1, descriptors_1);
detector->detectAndCompute(img_2, Mat(), keypoints_2, descriptors_2);
```

2. Matching Feature between 2 images

```
//-- Step 2: Matching descriptor vectors using FLANN matcher
FlannBasedMatcher matcher;
std::vector< DMatch > matches;
matcher.match(descriptors_1, descriptors_2, matches);
double max dist = 0; double min dist = 100;
for (int i = 0; i < descriptors_1.rows; i++)</pre>
    double dist = matches[i].distance;
    if (dist < min_dist) min_dist = dist;</pre>
    if (dist > max dist) max dist = dist;
printf("-- Max dist : %f \n", max_dist);
printf("-- Min dist : %f \n", min dist);
std::vector< DMatch > good_matches;
for (int i = 0; i < descriptors_1.rows; i++)</pre>
    if (matches[i].distance <= max(2 * min dist, 0.02))</pre>
        good matches.push back(matches[i]);
//-- Draw only "good" matches
Mat img matches;
drawMatches(img_1, keypoints_1, img_2, keypoints_2,
    good_matches, img_matches, Scalar::all(-1), Scalar::all(-1),
    vector<char>(), DrawMatchesFlags::NOT DRAW SINGLE POINTS);
```

Result:



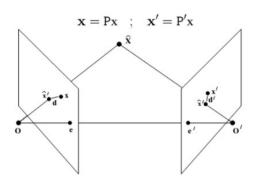
3. Triangulation between 2 matched points

Theory:

Modélisation géométrique de la reconstruction

Avec deux cameras

- Connaissant : P et P'
 Connaissant : x et x'
- Trouver X tel que :
 - x=PX
 - x'=P' X



In the last TP, we had intrinsic (K) of this camera and extrinsic (R|t) of every scene

I read it from output.yml . Then we find the 3x4 Projection Matrix (P).

```
Simply by: P = K*[R \mid t]
```

From that we can have P and P' from the formula above.

The next step is to find x and x'. And we already have them from Step 1 and Step 2 in this TP (Detection and Matching).

Now we can be able to find the real point in 3D. I tried cv::triangulatePoints, but somehow it calculates garbage. I was forced to implement a linear triangulation method manually, which returns a 4x1 matrix for the triangulated 3D point:

```
Mat triangulate_Linear_LS(Mat mat_P_l, Mat mat_P_r, Mat left_pixel, Mat right_pixel)
{
    // resolve equations ...
    solve(A,b,X,DECOMP_SVD);
    vconcat(X,W,X_homogeneous);
    return X_homogeneous;
}
```

The input parameters are two 3x4 camera projection matrices and a corresponding Left/Right pixel pair (x,y,w).

I repeat step 1, 2, 3 with every pair of images.

For exemple: image 1 vs image 2, 3, 4 ...

image 2 vs image 3, 4, 5...

image 3 vs image 4, 5, 6...

So the time of processing will be increase very fast when we increase the number of image

After Step 3:

I have a file (cloudpoint.txt) containing 3D coordinates (X,Y,Z) of every pair matched points.

```
File Edit Format View Help
4.0699 2.30154 -10.1286
2.40378 5.64405 4.83641
4.17577 2.59594 -11.4828
4.01704 2.43304 -11.5615
4.57265 4.93976 -1.15586
-3.29592 1.9202 -15.0906
4.59804 2.74857 -12.8135
3.25345 5.08457 1.41272
4.59502 5.04850 2.616849
4.41154 4.03417 0.690475
5.16315 4.39158 0.333646
3.13182 5.26016 3.25812
-3.51312 0.888522 -12.4141
-4.76596 0.749367 -11.428
-5.38878 0.430427 -11.8833
4.39438 3.94388 2.95738
1.76358 5.18663 5.61233
3.88131 3.4821 2.615
4.61892 3.4209 3.03063
4.68726 0.691153 -12.219
3.62072 3.57169 3.05481
-8.86602 5.55421 -0.265562
3.89721 3.98782 3.8865
5.6974 3.12329 -18.0524
-4.68525 0.683328 -12.2047
4.54196 2.8861 1.49186
2.99699 2.45548 -8.6471
6.42401 2.43967 -19.1687
-7.50625 -1.45543 0.599136
5.78164 5.10605 -0.3575694
-3.34312 2.00207 -14.6668
4.58923 2.87745 -12.4075
-2.49685 2.87745 -12.4075
-2.49685 2.87745 -12.675
-2.49685 2.8876 -15.9984
5.11269 2.94832 -13.5107
```

We can use **MeshLab** to visualize the result:

