Autonomous Robotics Autocalibration Report

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1 Introduction

Camera calibration with known patterns or objects with know euclidean structures are very well known and used techniques. However these methods are not available without priori assumptions to calibrate and get intrinsic parameters. Hence, we need to use autocalibration techniques that doesn't rely on euclidean structures for calibration.

2 Methods

We have been given approximate intrinsic parameter of the camera, fundemental matrixes between the images, projection matrix between the images. We employ 3 different method below to more accurately approximate intrinsic parameters.

2.1 Mendonça and Cipolla

Basic method based on the exploitation of the rigidity constraint. We design cost function, which takes intrinsic parameters and fundemental matrices as parameters, and returns positive value related to difference between two non-zero singular value of the essential matrix.

Table 1: Mendonca Cipolla Intrinsic Parameter Approximation

```
Listing 1: Mendonca and Cipolla Cost

function [ C ] = mendoncaCipollaCost( X, Fs )

%MENDONCACIPOLLACOST Summary of this function goes here

% Detailed explanation goes here

% X(1) = alfa_u

% X(2) = alfa_v

% X(3) = gamma
```

```
8 | \% X(4) = u_0
9 | \% X(5) = v_0
10
11 % Hint: For Mandonca&Cipolla, consider all the weight w_ij = 1
12
13 \mid \% SVD = and get S, first 2 sigma
14 \% S = [sigma_1]
                             0
                                            0;
15 %
              0
                             sigma_2
                                            0;
16 %
   %
17
   \% Essential = A' * Fs * A;
18
19
20 \mid A = [X(1)]
                     X(3)
                                X(4);
                     X(2)
                                X(5);
21
         0
                     0
22
         0
                                1];
23
24 \mid E = zeros(3, 3, 10, 10);
25
26 | C = zeros(1, 10);
27
   \quad \text{for ii} \, = \, 1 \, : \, 10
28
29
30
          for jj = 1 : 10
31
32
                if ii == jj
33
                      continue;
34
35
               E\left(:\,,\;\;:,\;\;ii\;,\;\;jj\;\right)\;=\;A\;'\;\;*\;\;Fs\left(:\,,\;\;:,\;\;ii\;,\;\;jj\;\right)\;\;*\;\;A;
36
37
                [\ \tilde{\ }\ ,\ S\ ,\ \tilde{\ }\ ]\ =\ {\rm svd}\left( {\rm E}\left( :\ ,\ :\ ,\ {\rm ii}\ ,\ {\rm jj}\right) \right) ;
38
39
               C(ii) = C(ii) + ((S(1, 1) - S(2, 2)) / S(2, 2));
40
41
42
         end
43
44
   \quad \text{end} \quad
45
46 \mid C = sum(C(:));
47
48
49 end
```

```
1 load ('data.mat');
 2
 3 | X = zeros(5, 1);
 4
 5|X(1) = A(1, 1);
 6 | X(2) = A(2, 2);
7
   X(3) = A(1, 2);
8 | X(4) = A(1, 3);
9|X(5) = A(2, 3);
10
   costFunc = @(X) mendoncaCipollaCost(X, Fs);
11
12
13 [xHat, resnorm] = lsqnonlin(costFunc, X);
14
15 \% \text{ xHat}(1) = \text{alfa_u}
16 \mid \% \text{ xHat}(2) = \text{alfa_v}
17 \mid \% \text{ xHat}(3) = \text{gamma}
18 \% \text{ xHat}(4) = u_{-}0
19 | \% \text{ xHat} (5) = v_{-}0
20 \mid A2 = [xHat(1)]
                        xHat(3)
                                     xHat(4);
21
                       xHat(2)
                                    xHat(5);
22
          0
                       0
                                    1];
```

2.1.1 Kruppa Equations

Kruppa equations are obtaining intrinsic parameters of the camera using polynomial equations, with a minimum of three displacements. I had hard time to achieve good results without passing the tolerance, and some more parameter values for Isquonlin function.

Table 2: Kruppa Equations Intrinsic Parameter Approximation

```
Listing 3: Kruppa Equations Cost
           function [ C ] = kruppaCost( X, Fs )
         %KRUPPACOST Summary of this function goes here
   2
                           Detailed explanation goes here
   4
         \% X(1) = alfa_u
   5
   6 \% X(2) = alfa_v
         \% X(3) = gamma
   7
         % X(4) = u_{-}0
   9 | \% X(5) = v_0
10
11
          A = [X(1) \ X(3) \ X(4);
12
                                              X(2) X(5);
13
                                                                    1];
14
           w_inv = A * A';
15
16
17
         E = zeros(3, 3, 10, 10);
18
19 \% C = zeros(1, 10);
20 | C = 0;
21
           for ii = 1 : 10
22
23
24
                            for jj = 1 : 10
25
26
                                             if ii == jj
27
                                                            continue;
28
                                            end
29
                                                                                    E(:, :, ii, jj) = A' * Fs(:, :, ii, jj) * A;
30
                                           %
                                           %
31
                                           %
32
                                                                                     \% [\ \tilde{\ }\ ,\ S\ ,\ \tilde{\ }\ ] \ = \ \operatorname{svd} \left( E \left( :\ ,\ :\ ,\ ii\ ,\ jj\ \right) \right);
33
                                           %
                                                                                    C(\,i\,i\,) \, = \, C(\,i\,i\,) \, + \, norm\,(\,((\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,) \, *\,\,w\_inv\,\,*\,\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,)\,\,') \, / \, (\,i\,i\,) \, = \, C(\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,) \, *\,\,w\_inv\,\,*\,\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,)\,\,') \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,) \, *\,\,w\_inv\,\,*\,\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,)\,\,') \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,) \, *\,\,w\_inv\,\,*\,\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,)\,\,') \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,) \, *\,\,w\_inv\,\,*\,\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,\,j\,j\,\,)\,\,) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,j\,j\,\,) \, *\,\,w\_inv\,\,*\,\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,j\,j\,\,)\,\,) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,j\,j\,\,) \, ) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,j\,j\,\,) \, ) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,j\,j\,\,) \, ) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,j\,j\,\,) \, ) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,(\,(\,Fs\,(:\,,\,\,:\,,\,\,i\,i\,\,,\,\,j\,j\,\,) \, ) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,i\,\,) \, (\,i\,i\,\,) \, + \, norm\,(\,i\,\,) \, (\,i\,i\,\,) \, / \, (\,i\,i\,\,) \, + \, norm\,(\,i\,\,) \, (\,i\,i\,\,) \, / \, (\,i\,i\,\,) \, 
                                                           34
                                           %
35
                                           %
36
37
38
39
                                           % Abinash's way instead of longer equation
                                            \begin{array}{l} currFS \,=\, Fs\,(:\,,\,:\,,\,\,ii\,\,,\,\,jj\,)\,;\\ ce \,=\, null\,(\,currFS\,'\,)\,; \end{array}
40
41
                                            F = currF\hat{S} * w_inv * currFS';
42
                                            F = F / norm(F, 'fro');
43
                                            currE = [0 -ce(3) ce(2); ce(3) 0 -ce(1); -ce(2) ce(1) 0];
44
                                            E = currE * w_inv * currE';
45
                                            E = E / norm(E, 'fro');
46
                                            tempC = F - E;
47
                                           C = C + norm(tempC, 'fro');
48
49
                                           % Abinash's way
50
```

```
51 end
52 53 end
54 55 end
```

```
Listing 4: Kruppa Equations Test
                    load ('data.mat');
       2
       3
                    X = zeros(5, 1);

\begin{array}{c|cccc}
5 & X(1) & = A(1, 1); \\
6 & X(2) & = A(2, 2);
\end{array}

      7 | X(3) = A(1, 2);
       8 \mid X(4) = A(1, 3);
      9
                    X(5) = A(2, 3);
 10
                        costFunc = @(Y) kruppaCost(Y, Fs);
11
12
                        options = optimset (\, 'Algorithm\, '\, ,\, 'levenberg-marquardt\, '\, ,\, 'MaxFunEvals\, '\, ,10\, \hat{}\, 50\, ,\, 'TolFun\, ',\, (10\, \hat{}\, 10\, \hat{}\, 10
13
                                                           ,10^-100, TolX, 10^-100);
                         [xHat, resnorm] = lsqnonlin(costFunc, X, [], [], options);
 14
15
                        A3 = [xHat(1)]
                                                                                                                                                                 xHat(3)
                                                                                                                                                                                                                                                           xHat(4);
 16
                                                                   Ö
                                                                                                                                                           xHat(2)
17
                                                                                                                                                                                                                                                    xHat(5);
                                                                   0
 18
                                                                                                                                                           0
                                                                                                                                                                                                                                                    1];
```

2.2 Dual Absolute Quadric

We represent euclidean scene structure formulated in terms of the absolute quadric - the singular dual 3D quadric giving teh Euclidean dot-product between plane normals.

```
Listing 5: Dual Absolute Quadric
```

```
load('data.mat');
 2
   \% A = [800 0 256; 0 800 256; 0 0 1]; \% Correct one
 3
 4
 5
   w = A * A';
6 nx = sym('nx', 'real');
7 ny = sym('ny', 'real');
8 nz = sym('nz', 'real');
9 12 = sym('12', 'real');
10
   n = [nx; ny; nz]; \% Normal of pi inf
11
12
13 \mid Q = [w, (w * n); (n' * w), (n' * w * n)]; \% Dual Absolute Quadric
14
15 \mid M2 = PPM(:, :, 2);
16
   m2 = M2 * Q * M2';
17
18
   sol \, = \, solve \, (m2(\,1 \, , \ 1) \, = \, (\, 12 \, * \, w(\,1 \, , \ 1)\,) \, , \ \ldots
19
                    m2(2, 2) = (12 * w(2, 2)), ...

m2(3, 3) = (12 * w(3, 3)), ...
20
21
                    m2(1, 3) = (12 * w(1, 3));
22
23
   display('sol.nx:');
24
25 display (double (sol.nx));
26 display ('sol.ny:');
27 display (double (sol.ny));
28 display ('sol.nz:');
29 display (double (sol.nz));
```

```
30 | display('sol.12:');
31 | display(double(sol.12));
```

3 References

- 1. Abinash Pant
- 2. http://en.wikipedia.org/wiki/Camera_auto-calibration
- 4. http://homepages.inf.ed.ac.uk/cgi/rbf/CVONLINE/entries.pl?TAG1325
- $5.\ \mathtt{http://hal.inria.fr/docs/00/54/83/45/PDF/Triggs-cvpr97.pdf}$