

LUND UNIVERSITY

COMPUTER VISION

FMAN95

Assignment 4

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2 Robust Homography Estimation and Stitching

2.1 E1

By looking at some arbitrary point which has a projection in both cameras we have

$$x_1 = P_1 \mathbb{X} = \begin{bmatrix} A_1 & t_1 \end{bmatrix} \mathbb{X} = A_1 X + t \quad \Rightarrow \quad X = A_1^{-1}(x_1 - t_1) \quad (1)$$

$$x_2 = \dots = A_2 X + t \quad \Rightarrow \quad X = A_2^{-1}(x_2 - t_2) \quad (2)$$

so

$$A_1^{-1}(x_1 - t_1) = A_2^{-1}(x_2 - t_2) \quad \Rightarrow \quad x_2 = A_2 A_1^{-1}(x_1 - t_1) + t_2 \quad (3)$$

and with the same center, i.e. $t_1 = t_2 = 0$, we finally get

$$x_2 = A_2 A_1^{-1} x_1 \quad \Rightarrow \quad H = A_2 A_1^{-1} \quad (4)$$

2.2 E2

3x3 matrix, scale arbitrary \Rightarrow 8 degrees of freedom.

3 equations from each point, one extra unknown λ per equation, so $3N$ equations and $8 + N$ unknowns $\Rightarrow N \geq 4$ points

10% incorrect correspondence and 98% probability of outlier free sample set, so

$$P[\text{Inliers}] = 0.9 \Rightarrow P[n \text{ set outliers}] = (1 - 0.9^4)^n \leq (1 - 0.98) \Rightarrow$$

$$n \log(1 - 0.9^4) \leq \log(0.02) \Rightarrow n \geq \frac{\log(0.02)}{\log(1 - 0.9^4)} \approx 3.66 \Rightarrow 4 \text{ iterations}$$

2.3 CE1

947 SIFT features for image A and 865 for B, 204 matches and 151 inliers.

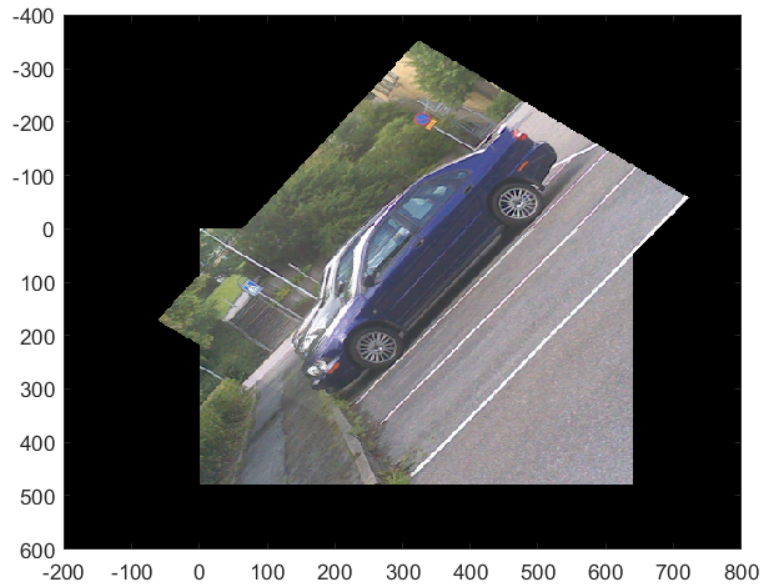


Figure 1: Panorama

3 Robust Essential Matrix Estimation

3.1 E3

R has 3 dofs, t has 3 dofs, scale arbitrary \Rightarrow E has 5 dofs. 8 points in algorithm and with the same restrictions as in E2 we need

$$n \geq \frac{\log(0.02)}{\log(1 - 0.9^8)} \approx 6.95 \Rightarrow 7 \text{ iterations}$$

3.2 CE2

Number of inliers : 1466

RMS(Im1) = 0.5666

RMS(Im2) = 0.6083

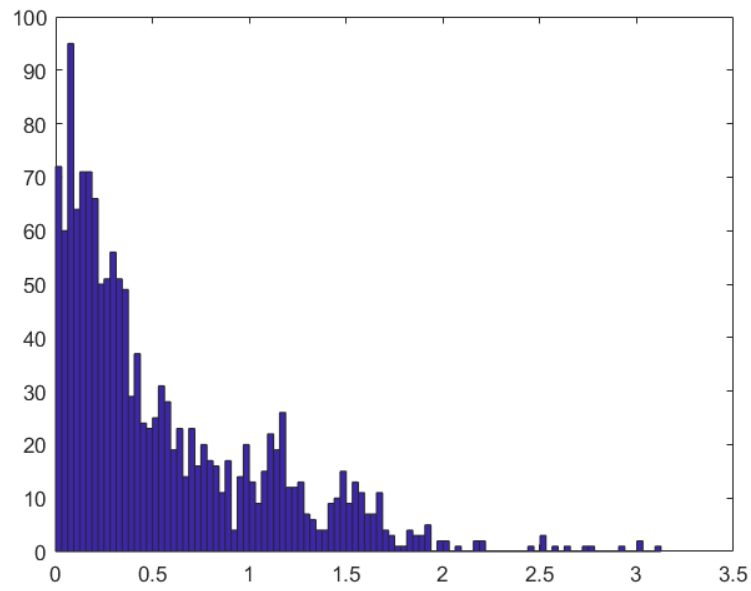


Figure 2: Reprojection errors Image 1

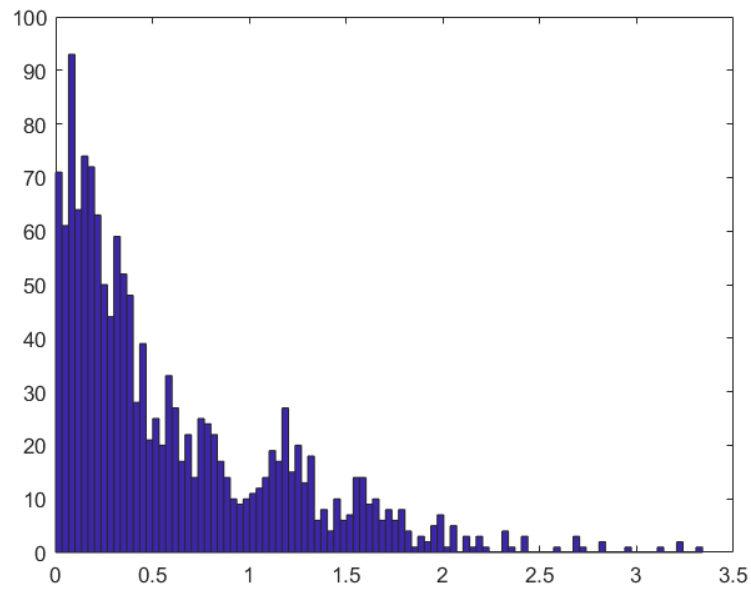


Figure 3: Reprojection errors Image 2

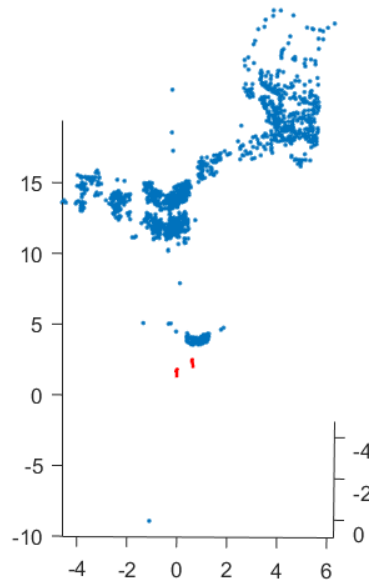


Figure 4: Reconstruction

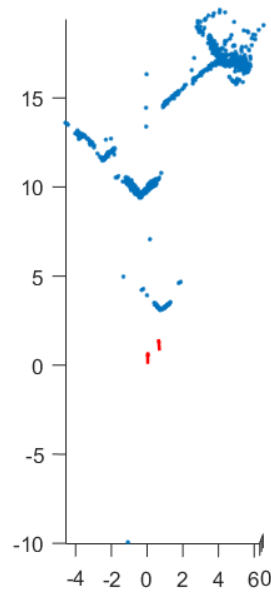


Figure 5: Reconstruction as seen from above

4 Calibrated Structure form Motion an Local Optimization

4.1 CE3

$$\text{RMS}(\text{Im1}) = 0.5666 \qquad \text{RMS}(\text{Im2}) = 0.5318 \qquad (5)$$

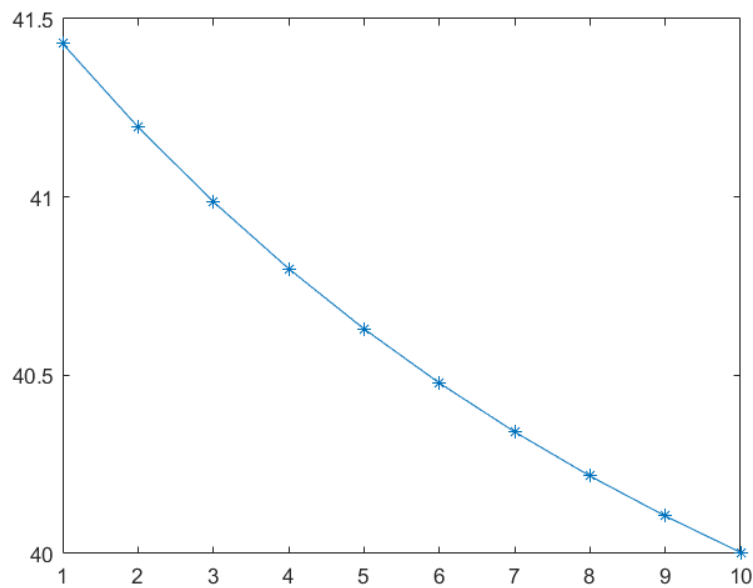


Figure 6: Objective value for 10 iterations

4.2 CE4

$$\text{RMS}(\text{Im1}) = 0.1725 \qquad \text{RMS}(\text{Im2}) = 0.1582 \qquad (6)$$

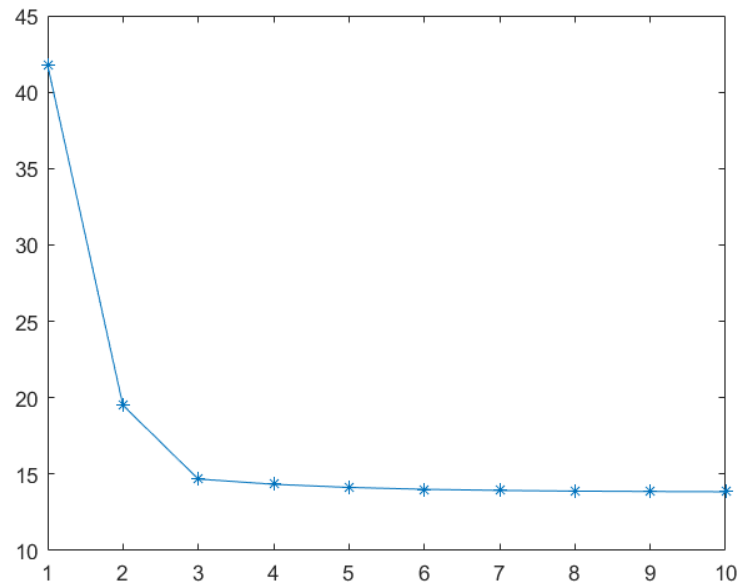


Figure 7: Objective value for 10 iterations, clearly a lot better than the method in CE3

5 Appendix

Listing 1: main.m

```
1 %% CE1
2 % load images and setup vl
3 im_A = imread('a.jpg');
4 im_B = imread('b.jpg');
5 % im_A = imread('nils1.jpg');
6 % im_B = imread('nils2.jpg');
7 % im_A = imread('tavla1.jpg');
8 % im_B = imread('tavla2.jpg');
9
10 run vl_setup.m;
11
12 %% compute sift shit
13
14 [fA dA] = vl_sift( single(rgb2gray(im_A)) );
15 [fB dB] = vl_sift( single(rgb2gray(im_B)) );
16
17 %% plots
18 figure(1)
19 imagesc(im_A);
20 hold on
21 vl_plotframe(fA);
22 hold off
23 axis equal
24
25 figure(2)
26 imagesc(im_B);
27 vl_plotframe(fB);
28 hold off
29 axis equal
30
31 %% compute matches
32 matches = vl_ubcmatch(dA,dB);
33
34 % xA = fA(1:2, matches(1,:));
35 % xB = fB(1:2, matches(1,:));
36
37 xA = [fA(1,matches (1 ,:)); fA(2,matches (1 ,:))];
38 xB = [fB(1,matches (2 ,:)); fB(2,matches (2 ,:))];
39
```

```

40 perm = randperm(size(matches ,2));
41
42 %% plot lines
43
44 figure(3);
45 imagesc ([im_A im_B]);
46 hold on;
47 plot([xA(1,perm (1:10)); xB(1,perm (1:10))+ size(im_A ,2)], ...
48      [xA(2,perm (1:10)); xB(2,perm (1:10))] , '-');
49 axis equal
50 hold off;
51
52
53 %% compute H
54 cp = [];
55 for i=1:5000
56     % Pick 4 random points
57     rand = randperm(length(xA), 4);
58     x = [xA(:,rand); ones(1,4)];
59     y = [xB(:,rand); ones(1,4)];
60
61     % Compute estimated H
62     Me = zeros(12, 13);
63     for j = 1:length(x)
64         Me(3*j-2:3*j,:) = ...
65             [x(:,j)' zeros(1,6) zeros(1,j-1) y(1,j) zeros(1,4-j);
66              zeros(1,3) x(:,j)' zeros(1,3) zeros(1,j-1) y(2,j) zeros(1,4-j);
67              zeros(1,6) x(:,j)' zeros(1,j-1) y(3,j) zeros(1,4-j)];
68     end
69
70     [U,S,V] = svd(Me);
71     v = V(1:9, end);
72     He = reshape(v, [3 3])';
73
74     % Compute estimated points and find those with error < 5 pixels
75     xBe = He*[xA ; ones(1, (length(xA)))];
76     xBeFlat = pflat(xBe);
77     err = sqrt((xB(1,:)-xBeFlat(1,:)).^2 + (xB(2,:)-xBeFlat(2,:)).^2);
78     cps = find(err<5);
79
80     % Save the best estimate of the loop
81     if (length(cps) > length(cp))
82         cp = cps;

```

```

83         Hest = He;
84     end
85
86 end
87
88 Hest = Hest./Hest(end, end);
89
90 length(cp)
91
92 %%
93
94 % Transform image to comon coordinate system
95 Htform = projective2d(Hest');
96 Rout = imref2d ( size(im_A) ,[ -200 800] ,[ -400 600]);
97 % Rout = imref2d ( size(im_A) ,[ -1500 3800] ,[ -1400 4800]);
98 [Atransf] = imwarp(im_A,Htform , 'OutputView',Rout );
99 Idtform = projective2d(eye(3));
100 [Btransf] = imwarp(im_B, Idtform , 'OutputView',Rout );
101 AB = Btransf;
102 AB( Btransf < Atransf ) = Atransf( Btransf < Atransf );
103 figure(4)
104 imagesc( Rout . XWorldLimits , Rout . YWorldLimits ,AB );
105
106
107
108
109
110
111
112 %% Ce2
113
114 load("compEx2data.mat");
115 im1 = imread("im1.jpg");
116 im2 = imread("im2.jpg");
117 x1 = [x{1}; ones(1, length(x{1}))];
118 x2 = [x{2}; ones(1, length(x{1}))];
119
120 cp = [];
121 l1 = [];
122 l2 = [];
123 E = [];
124 F = [];
125

```

```

126 iter = 1000;
127 for i = 1:iter
128     % get five random points
129     randsel = randperm(length(x1), 5);
130     x1r = x1(:, randsel);
131     x2r = x2(:, randsel);
132
133     % Calibrate with K
134     x1rn = K\x1r;
135     x2rn = K\x2r;
136
137     % Calculate essential matrix estimation
138     Ei = fivepoint_solver(x1rn, x2rn);
139
140     for j = 1:length(Ei)
141
142         %compute F
143         Fe = K'\Ei{j}/K;
144         Fj = Fe./Fe(3,3);
145
146         % compute lines
147         l1j = Fj'*x2;
148         l1j = l1j./sqrt(repmat(l1j(1,:).^2 + l1j(2,:).^2, [3 1])); %
            normalize
149
150         l2j = Fj*x1;
151         l2j = l2j./sqrt(repmat(l2j(1,:).^2 + l2j(2,:).^2, [3 1])); %
            normalize
152
153         % compute distances to lines
154         dist1 = abs(sum(l1j.*x1));
155         dist2 = abs(sum(l2j.*x2));
156
157         % compute inliers
158         cp1 = find(dist1 < 5);
159         cp2 = find(dist2 < 5);
160
161         % compute matching inliers for both images
162         [cpj usch] = intersect(cp1, cp2);
163
164         % save the best results
165         if (length(cp) < length(cpj))
166             cp = cpj;

```

```

167         l1 = l1j;
168         l2 = l2j;
169         E = Ei{j};
170         F = Fj;
171     end
172 end
173 end
174
175 disp('number of matching inliers:')
176 length(cp)
177
178 %% Plot
179
180 figure(5)
181 imagesc(im1)
182 hold on
183 plot(x1(1,cp), x1(2,cp), 'y*', 'Markersize', 10)
184 rital(l1(:,cp))
185 hold off
186 axis equal
187
188 figure(6)
189 imagesc(im2)
190 hold on
191 plot(x2(1,cp), x2(2,cp), 'y*', 'Markersize', 10)
192 rital(l2(:,cp))
193 hold off
194 axis equal
195
196 %% compute cameras and triangulate
197
198 x1n = K\x1;
199 x2n = K\x2;
200
201 [U S V] = svd(E);
202
203 u3 = U(:,end);
204 W = [0 -1 0 ; 1 0 0 ; 0 0 1];
205
206 P1 = [eye(3) [0;0;0]];
207 % CAmEra solutions
208 P2 = {[U*W*V' u3], [U*W*V' -u3], [U*W'*V' u3], [U*W'*V' -u3]};
209

```

```
210 in_front = 0;
211 bestP2 = 0;
212 bestX = 0;
213
214 for i=1:4
215
216     % Triangulate
217     X = [];
218     for j = 1:length(x1n)
219         Me = [P1, x1n(:,j) zeros(3,1) ; ...
220             P2{i} zeros(3,1) x2n(:,j)];
221         [Ut, St, Vt] = svd(Me);
222         X(:,j) = pflat(Vt(1:4,end));
223     end
224
225     % Check number of points in front of camera
226     temp = sum(P1(3,:)*X > 0 & P2{i}(3,:)*X > 0);
227     if temp >= in_front
228         bestP2 = P2{i};
229         bestX = X;
230         in_front = temp;
231         i
232     end
233 end
234
235 P2f = K*bestP2;
236 X = bestX;
237 x1proj = pflat(K*P1*X);
238 x2proj = pflat(P2f*X);
239
240 %% plots projections of 3d and 3d model with cameras'
241
242 figure(7)
243 imagesc(im1)
244 hold on
245 plot(x1(1,cp), x1(2,cp), 'r*', 'Markersize', 5)
246 plot(x1proj(1,cp), x1proj(2,cp), 'b+', 'Markersize', 5)
247 hold off
248 axis equal
249
250 figure(8)
251 imagesc(im2)
252 hold on
```

```

253 plot(x2(1,:), x2(2,:), 'r*', 'Markersize', 5)
254 plot(x2proj(1,:), x2proj(2,:), 'b+', 'Markersize', 5)
255 hold off
256 axis equal
257 %%
258 figure(22)
259 plot3(X(1,cp), X(2,cp), X(3,cp), 'r.')
260 hold on
261 plotcams({P1, P2f})
262 hold off
263 axis equal
264
265
266 %% Plot histograms of errors
267
268 figure(9)
269 hist(sum(abs(x1(:,cp)-x1proj(:,cp))), 100);
270
271 figure(10)
272 hist(sum(abs(x2(:,cp)-x2proj(:,cp))), 100);
273
274 %% Compute rms errors
275 x1rms = mean(sqrt(sum((x1(:,cp)-x1proj(:,cp)).^2)))
276 x2rms = mean(sqrt(sum((x2(:,cp)-x2proj(:,cp)).^2)))
277
278
279
280 %% CE3
281
282 P = {K*P1 P2f};
283 U = X(:,cp);
284 u = {x1(:, cp) x2(:, cp)};
285
286 iter = 10;
287 gammak = 1;
288 rnorms = zeros(1, 10);
289 [r,J] = LinearizeReprojErr(P,U,u);
290 for i = 1:iter
291     lastr = r;
292     while 1
293         deltav = -gammak*J'*r;
294         [Pnew , Unew ] = update_solution(deltav ,P,U);
295         [r,J] = LinearizeReprojErr(Pnew,Unew,u);

```

```

296         if norm(r) >= norm(lastr)
297             gammak = .5* gammak;
298         else
299             P = Pnew;
300             U = Unew;
301             break;
302         end
303     end
304     rnorms(i) = norm(r);
305 end
306 %%
307 figure(11)
308 plot(linspace(1, iter, iter), rnorms, '*-')
309
310 Pnew{2} = Pnew{2}./Pnew{2}(end, end);
311 x1e = pflat(Pnew{1}*Unew);
312 x2e = pflat(Pnew{2}*Unew);
313
314
315 figure(12)
316 imagesc(im1)
317 hold on
318 plot(x1(1,:), x1(2,:), 'b*', 'Markersize', 5)
319 plot(x1e(1,:), x1e(2,:), 'r.', 'Markersize', 5)
320 hold off
321 axis equal
322
323 figure(13)
324 imagesc(im2)
325 hold on
326 plot(x2(1,:), x2(2,:), 'b*', 'Markersize', 5)
327 plot(x2e(1,:), x2e(2,:), 'r.', 'Markersize', 5)
328 hold off
329 axis equal
330
331 % RMS error
332 disp('RMS x1')
333 mean(sqrt(sum((x1(:, cp)-x1e).^2)))
334 disp('RMS x2')
335 mean(sqrt(sum((x2(:, cp)-x2e).^2)))
336
337 %% CE4
338 P = {K*P1 P2f};

```



```

339 U = X(:,cp);
340 u = {x1(:, cp) x2(:, cp)};
341
342 iter = 10;
343 lambda = 1;
344 rnorms = zeros(1, 10);
345 [r,J] = LinearizeReprojErr(P,U,u);
346 C = J'*J+lambda*speye(size(J ,2));
347 c = J'*r;
348 deltav = -C\c;
349 [Pnew , Unew ] = update_solution(deltav ,P,U);
350 for i = 1:iter
351     lastr = r;
352     C = J'*J+lambda*speye(size(J ,2));
353     c = J'*r;
354     deltav = -C\c;
355     [Pnew , Unew ] = update_solution(deltav ,Pnew,Unew);
356     [r,J] = LinearizeReprojErr(Pnew,Unew,u);
357     rnorms(i) = norm(r);
358 end
359
360 %% plots
361
362 figure(14)
363 plot(linspace(1, iter, iter), rnorms, '*-')
364 %%
365 x1e = pflat(Pnew{1}*Unew);
366 x2e = pflat(Pnew{2}*Unew);
367
368 figure(15)
369 imagesc(im1)
370 hold on
371 plot(x1(1,:), x1(2,:), 'b*', 'Markersize', 5)
372 plot(x1e(1,:), x1e(2,:), 'r.', 'Markersize', 5)
373 hold off
374 axis equal
375
376 figure(16)
377 imagesc(im2)
378 hold on
379 plot(x2(1,:), x2(2,:), 'b*', 'Markersize', 5)
380 plot(x2e(1,:), x2e(2,:), 'r.', 'Markersize', 5)
381 hold off

```

```
382 axis equal
383 %%
384 % RMS error
385 disp('RMS x1')
386 mean(sqrt(sum((x1(:, cp)-x1e).^2)))
387 disp('RMS x2')
388 mean(sqrt(sum((x2(:, cp)-x2e).^2)))
```