

Applications of Basic DIRECT LINEAR TRANSFORM (DLT)

↓

from Geometric entities in the PROJECTIVE PLANE, you want to ESTIMATE that defined up to a scale factor.
to take into account and solve a linear system

how SVD can be used to estimate geometric entities... → a bunch of points in the plane

To estimate a CONIC, you need some points as the unknown

of the $\boxed{x^T C x = 0}$ $x := \text{homogeneous coord } \mathbb{R}^3$
 \uparrow
 $\boxed{3 \times 3} \rightarrow \text{but } C = C^T \text{ symm}$

in principle

C has 9 DOF BUT

being quadratic eq. $C = C^T \rightarrow$ so $9 - 3 = 6$ DOF

and since comic up to scaling factor → 5 DOF

this arguments holds for geometric entity to reconstruct

We need 5 points...

$(x_i, y_i, w_i) \quad i = 1, \dots, 5$ to define C of comic

as just equation of general comics

$$ax^2 + bxy + cx + dy + ey^2 = Q \quad \text{CARTESIAN}$$

in homogeneous also w in the equation

Knowing 5 points $\in C$

and so, the incidence relation! → $\forall x_i$: you write $x_i^T C x_i = Q$

→ linear system in which we reconstruct C

$$\left\{ \begin{matrix} [x \ y \ w] \begin{bmatrix} \dots \dots \end{bmatrix} \begin{bmatrix} x \\ y \\ w \end{bmatrix} = 0 \\ \vdots \\ \vdots \end{matrix} \right.$$

Solve the system of equations as homogeneous System



E2C_conic.m

```

1  %% Playing with conics in homogeneous coordinates
2  % Image Analysis and Computer Vision
3  % Politecnico di Milano
4  %
5  % Credits: Prof. Giacomo Boracchi
6  % Edits: Luca Magri.
7
8
9  %% Preparation
10 % First, we create an empty image and display it (will be our "canvas")
11 figure(1)
12 im=zeros(500,500);
13 imshow(im, []);
14 hold on;
15
16 %%
17 % Data input: enter five points (then enter)
18 disp('click 5 points, then enter');
19 [x, y]=getpts;<-- to CLICK 5 times on the image, 5 points selected
20 scatter(x,y,100,'filled');
21
22 %%
23 % x and y are now column vectors.
24 { x as we check
25 y to fit the points, we solve x, y are vectors of point  $x_1, x_2 \dots x_5$ 
26  $Ax = 0$  homogeneous eq  $X = (a, b, c, d, e, f)$  unknowns!  $y_1, y_2 \dots y_5$ 
27 %% Estimation of conic parameters
28 % using the technique shown in "Multiple View Geometry", page 9.
29 % this is equivalent to requiring that the conic passes through all the six
30 % points system of equations with final columns of 1 (because as EUCLIDEAN)
31 A = [x.^2 x.*y y.^2 x y ones(size(x))]; ← because as EUCLIDEAN
32 matrix build by stacking the various points coefficients (x,y)
33 % A is 5x6: we find the parameter vector (containing [a b c d e f]) as the
34 % right null space of A. 6 unknowns, 5 eq. I want Null(A)
35 % This returns a vector N such that A*N=0. Note that this expresses the
36 % fact that the conic passes through all the points we inserted.
37 N = null(A);
38 % alternatively you can use the svd
39 [~,~,V] = svd(A); // factorization SVD of A
40 N = V(:,end); ← last columns V is the smallest singular value column, which is
41 %% a basis of Null space
42 % Let's assign the values to variables and build the 3x3 conic matrix
43 % (which is symmetrical) ) rearrange
44 cc = N(:, 1);
45 % change the name of variables
46 [a b c d e f] = deal(cc(1),cc(2),cc(3),cc(4),cc(5),cc(6));
47 % here is the matrix of the conic: off-diagonal elements must be divided
48 % by two off diagonal elements divided by 2
49 C=[a b/2 d/2; b/2 c e/2; d/2 e/2 f]
50
51 % Remark: since
52 % when the pair ↑ Condividi Sposta Elimina Dettagli ...
53 % the system ac
54 % however, thes.

```



```

E2C_conic.m
51 % Remark: since the right null space has dimension one,
52 % when the points are taken in general position
53 % the system admits an infinite number of solutions
54 % however, these can be expressed as lambda * n, where n
55 % lambda \in R and n = null(A).
56 % thus, they all corresponds to the same conic.
57 % Draw the conic → CONIC is the set of POINT s.t  $x^T C x = 0$ 
58 % We pick every pixel, and compute the incidence relation.
59 im=zeros(500,500);           ← take all image pixel and I compute
60 for i=1:500
61   for j=1:500
62     im(i,j)=[j i 1]*C*[j i 1]'; % this is an algebraic error
63   end
64 end   where if it ~0 error we call it the BORDER
65
66 figure; imagesc(im); colormap("spring"); title('Algebraic error');
67 % keep in mind that for geometric problem the best is to deal with "geometric error"
68 %%
69 % Let's draw black for pixels less than zero, white for pixels greater than
70 % 0.          set black pixels with error < 0
71 figure(1)
72 imshow(im > 0);
73 scatter(x,y, 100, 'filled');
74
75 %% What happen when A is not full rank? / when we take 4 points, we get multiple solutions!
76 % Instead of taking 5 points take just 4
77 % N contains in each column a generator of the null space
78 N = null(A(1 : 4, :));
79
80 % then, any vector living in the null space give rise to a conic
81 % passing through those 4 points. Since the nullspace has dimension two
82 % we can find two conic passing through the four points
83
84 %%
85 % Let's assign the values to variables and build the 3x3 conic matrix
86 % (which is symmetrical)
87 % Let's start with a first solution, the first vector of the nullspace
88 cc = N(:, 1);
89 % cc = N(:, 1) + N(:, 2); % another example, any linear combination is fine
90 [a, b, c, d, e, f]=deal(cc(1),cc(2),cc(3),cc(4),cc(5),cc(6));
91 % here is the matrix of the conic
92 C1_four =[a b/2 d/2; b/2 c e/2; d/2 e/2 f]
93 % Let's consider a second solution, the second vector of the nullspace
94 cc = N(:, 2);
95 % cc = N(:, 1) + 2*N(:, 2); % another example, any linear combination is fine
96 [a, b, c, d, e, f]=deal(cc(1),cc(2),cc(3),cc(4),cc(5),cc(6));
97 % here is the matrix of the conic
98 C2_four =[a b/2 d/2; b/2 c e/2; d/2 e/2 f]
99 ↑ I get 2 conics
100
101 % Draw the conic
102 % We pick every
103 im_four=zeros(5
104 for i=1:500

```

↑

take all image pixel and I compute

↓

where if it ~0 error we call it the BORDER

set black pixels with error < 0

taking (x_i, y_i) $i=1..4$ MULTIPLE SOLUTIONS
5 unknowns but 4 equations → the dimension of the kernel increase by one dimension!

Now I'll take last 2 columns of V in the svd,
get the vectorspace of dimension 2 (2 solutions)
and any linear combination of this two is a vector solution of that one conic passing through that PENCIL OF CONICS)

↑

I get 2 conics

Condividi Sposta Elimina Dettagli Altro

```

E2C_conic.m
...
100
101 % Draw the conic
102 % We pick every pixel, and compute the incidence relation.
103 im_four=zeros(500,500);
104 for i=1:500
105     for j=1:500
106         im_four1(i,j)=[j i 1]*C1_four*[j i 1]'; //first nullspace
107         im_four2(i,j)=[j i 1]*C2_four*[j i 1]'; //second
108     end
109 end
110
111 figure(2)
112 subplot(1,2,1)
113 imshow(im_four1>0);
114 hold on;
115 scatter(x(1 : 4), y(1 : 4), 100, 'filled','cyan');
116 scatter(x(5), y(5), 100, 'filled','red');
117 title('1st nullspace col ')
118 subplot(1,2,2)
119 imshow(im_four2>0);
120 hold on
121 scatter(x(1 : 4), y(1 : 4), 100, 'filled','cyan');
122 scatter(x(5), y(5), 100, 'filled','red');
123 title('2n nullspace col ')
124 hold off
125
126
127 %% Any linear combination of N1 and N2 is a solution
128 % we can draw a pencil of conics
129 for lambda = [10, 200, 3000, 100000]    we take linear combination
130     of the two  $C_1 + \lambda C_2 = C'$ 
131
132 cc = 1*N(:,1)+ lambda*N(:,2);
133 [a, b, c, d, e, f]=deal(cc(1),cc(2),cc(3),cc(4),cc(5),cc(6));
134 C3_four =[a b/2 d/2; b/2 c e/2; d/2 e/2 f];
135 % We pick every pixel, and compute the incidence relation.
136 im_four=zeros(500,500);
137 for i=1:500
138     for j=1:500
139         im_four3(i,j)=[j i 1]*C3_four*[j i 1]';
140     end
141 end
142 figure(3)
143 imshow(im_four3>0);
144 hold on;
145 scatter(x(1 : 4), y(1 : 4), 100, 'filled','cyan');
146 scatter(x(5), y(5), 100, 'filled','red');
147 title('linear combination of nullspace cols ')
148 pause;
149
150 %% Tangent line
151 % Let's draw the tangent line
152 % coordinates i
153 % the line tangent to C at the time x is defined as t = C*x

```

drawing each conics, any
of them pass through
the 4 points!

...

Condividi
Sposta
Elimina
Dettagli
Altro



E2C_conic.m

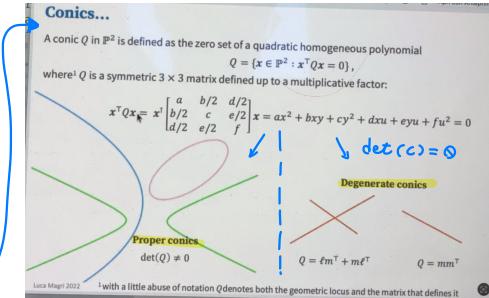
```

125
126
127 % Any linear combination of N1 and N2 is a solution
128 % we can draw a pencil of conics
129 for lambda = [10, 200, 3000, 100000]
130
131 cc = 1*N(:,1)+ lambda*N(:,2);
132 [a, b, c, d, e, f]=deal(cc(1),cc(2),cc(3),cc(4),cc(5),cc(6));
133 C3_four =[a b/2 d/2; b/2 c e/2; d/2 e/2 f];
134 % We pick every pixel, and compute the incidence relation.
135 im_four=zeros(500,500);
136 for i=1:500
137     for j=1:500
138         im_four3(i,j)=[j i 1]*C3_four*[j i 1]';
139     end
140 end
141 figure(3)
142 imshow(im_four3>0);
143 hold on;
144 scatter(x(1 : 4), y(1 : 4), 100, 'filled','cyan');
145 scatter(x(5), y(5), 100, 'filled','red');
146 title('linear combination of nullspace cols ')
147 pause;
148 end
149
150 %% Tangent lines
151 % Let's draw the tangents to our points: the tangent line in homogeneous
152 % coordinates is l
153 % the line tangent to C at the line x is defined as l = C*x
154 set of lines tangent to conics (*)  

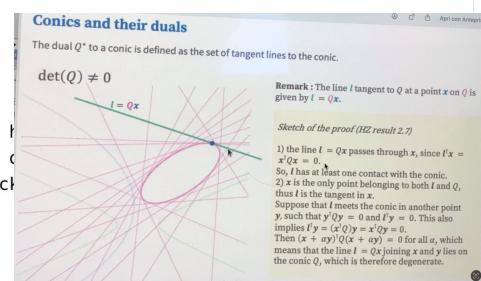
155 % this is the line tangent to the first point
156 l1 = C * [x(1); y(1); 1];
157 % here I compute all the tangent lines simultaneously
158 % stacking all the vectors in the columns of the matrix l
159 l=C*[x y ones(size(x))'
160 % in order to draw it we find its angle,
161 % then plot a line of length 200 around it.
162 % In order to write less lines of code, I'm doing a bit of
163 % matrices and vectors here: theta will be a column vector
164 % with a single plot command I'll plot all the lines. Check
165 % documentation of plot to understand what's happening.
166 theta=atan2(-l(1,:),l(2,:));
167 theta = theta';
168 figure(1), hold on
169 plot([x-cos(theta)*100 x+cos(theta)*100]',...
170 [y-sin(theta)*100 y+sin(theta)*100]', 'b-', 'LineWidth',5);
171
172 %%
173 % Let's compute the dual conic and check that all the tangent lines lie
174 % computed prev
175 % precision iss
176 % precision iss
177 Cstar=inv(C);
178 l(:,1)'*Cstar*l
179 1/./. 2/./. 3/./.

```

(*) tangent lines.
 point \leftrightarrow line in dual plane,
 which satisfy the dual conic
 equation (quadratic relation)
 when NON DEGENERATE
 so given C , x is easy
 to find the tangent line



When conic C is
 NON singular, the
 tangent line is easy
 to compute by Cx



easy to prove by
 simple linear algebra





E2C_conic.m

```

125
126
127 % Any linear combination of N1 and N2 is a solution
128 % we can draw a pencil of conics
129 for lambda = [10, 200, 3000, 100000]
130
131 cc = 1*N(:,1)+ lambda*N(:,2);
132 [a, b, c, d, e, f]=deal(cc(1),cc(2),cc(3),cc(4),cc(5),cc(6));
133 C3_four =[a b/2 d/2; b/2 c e/2; d/2 e/2 f];
134 % We pick every pixel, and compute the incidence relation.
135 im_four=zeros(500,500);
136 for i=1:500
137     for j=1:500
138         im_four3(i,j)=[j i 1]*C3_four*[j i 1]';
139     end
140 end
141 figure(3)
142 imshow(im_four3>0);
143 hold on;
144 scatter(x(1 : 4), y(1 : 4), 100, 'filled','cyan');
145 scatter(x(5), y(5), 100, 'filled','red');
146 title('linear combination of nullspace cols ')
147 pause;
148 end
149
150 %% Tangent lines
151 % Let's draw the tangents to our points: the tangent line in homogeneous
152 % coordinates is l
153 % the line tangent to C at the line x is defined as l = C*x
154
155 % this is the line tangent to the first point
156 l1 = C * [x(1); y(1); 1]; homogeneous equation, hom coord with w=1,
157 % here I compute all the tangent lines simultaneously euclidean coordinate
158 % stacking all the vectors in the columns of the matrix l
159 l=C*[x y ones(size(x))]' ↑
160 %  

161 % in order to draw it we find its angle,
162 % then plot a line of length 200 around it.
163 % In order to write less lines of code, I'm doing a bit of hacks with
164 % matrices and vectors here: theta will be a column vector of angles, and
165 % with a single plot command I'll plot all the lines. Check out the
166 % documentation of plot to understand what's happening.
167 theta=atan2(-l(1,:),l(2,:)); not always = C-1
168 theta = theta';
169 figure(1), hold on
170 plot([x-cos(theta)*100 x+cos(theta)*100]',... /
171 [y-sin(theta)*100 y+sin(theta)*100]', 'b-', 'LineWidth',5); C* = C-1 simply
172 %<-- dual conic and check that all lines pass through it C* before!
173 % Let's compute the dual conic and check that all the tangent lines are
174 % computed prev
175 % precision iss
176 % precision iss
177 Cstar=inv(C);
178 l(:,1)'*Cstar*l
179 l(:,1)'*Cstar*l

```

(being NON SINGULAR)

then to check if lines satisfy dual conics

MUST be $e^T C^ e \approx 0 \forall e$*

↑ computed

before!

Condividi
Sposta
Elimina
Dettagli
Altri

← so we just use SVD, solve linear system as our problem...

as homogeneous problem

⇒ A kind of TRANSFORMATION

between images we know is PROJECTIVITY
(HOMOGRAPHY)



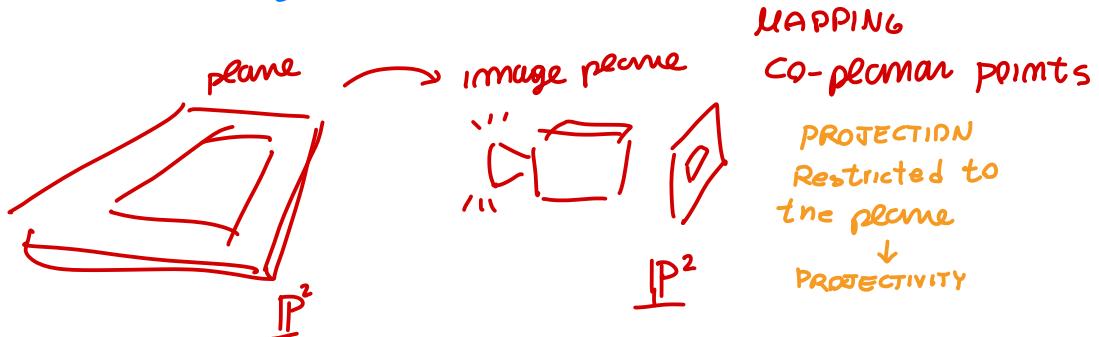
≈ which is as the
points in the scene 3D, then I take a
picture and a PROJECTION MAPPING
↓

projectivity is defined in a projective space
with same dimensions

PLANAR PROJECTIVE := 3×3 H invertible
mapping planar point into other

IN SPACE PROJECTIVITY instead we map $\mathbb{P}^d \rightarrow \mathbb{P}^d$ by $d \times d$
s.t. $\det(H_d) \neq 0$ MATRIX

- If points lay on same plane



← that has algebraic interpretation...

from camera MATRIX

$$P = \begin{bmatrix} P_1 & P_2 & P_3 & P_4 \end{bmatrix}$$

↑
3x4 camera MATRIX

then supposing
to take a point
on the scene
(which say on same plane)

↓
mapping [X Y Z] → to image

the PROJECTION MATRIX...

assuming the points are CO-PLANAR, then
belongs to $Z=0$ plane! so you get

$X \rightarrow x$ mapping to image plane

$$\begin{bmatrix} X & Y & 0 & 1 \end{bmatrix} \underset{Z=0}{\Rightarrow} \underbrace{\begin{bmatrix} P_1 & P_2 & P_4 \end{bmatrix}}_{3 \times 3 \text{ transformation}} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix} :$$

projection mapping
restricted to plane becomes
 3×3 MATRIX (invertible)

→ then PROJECTION Restriction is HOMOGRAPHY
maintain 3×3 matrix

then solve
the system
using SVD
to estimate
coefficients
↪

Let's estimate HOMOGRAPHY as

- PROJECTIVE entity and the DOF (8) of it!
each point provide 2 constraint NOT 9 because of scale factor
- AT Least 4 points to define $H \Rightarrow$ I can define H by constraints definition

estimate homography coefficients by svd

Homography estimation via DLT

Direct Linear Transform

Estimate an homography via DLT References: Hartley Zisserman, Multiview Geometry. Chapter 4

Image Analysis and Computer Vision Politecnico di Milano

Luca Magri for comments and suggestions please send an email to luca.magri@polimi.it

Image Analysis and Computer Vision Politecnico di Milano

Contents

- Let's take two picture of a planar scene
- Homography transformation
- Identify four pairs of corresponding points
- Direct linear transform
- Showing the mapping on the points
- Rendering the results
- Degenerate configurations - collinearity
- compute the homography
- let's show the result

Let's take two picture of a planar scene

```
{ im1 = imread('E3_data/ella1.jpg');
im2 = imread('E3_data/ella2.jpg');
im1 = imresize(im1, 0.6);
im2 = imresize(im2, 0.6);
figure;
subplot(1,2,1);
imshow(im1);
subplot(1,2,2);
imshow(im2);
```

from an almost planar scene → I want to identify corresponding points



Homography transformation

the image coordinates that lie on the same plane of these two images are related by a **homography**: Without loss of generality, we can assume that the point on the plane have coordinates $M = [x, y, 1]$, and are mapped via P_1 and P_2 to $m_1 = P_1^*M$ and $m_2 = P_2^*M$, where $P_i = K_i[R_i, t_i]$ and $R_i = [R_{i1}, R_{i2}, R_{i3}]$. Having $z=0$ implies that the coordinates $m_i = K_i[R_{i1}, R_{i2}, t_i] [x, y, 1]$ are determined by a 3×3 non singular transformation H_i . The relation between $\{m_1 \leftrightarrow m_2\}$ is hence given by $H = H_1^{-1}H_2$. Our aim is to determine such H .

Identify four pairs of corresponding points

```
m1 = nan(3,4);
m2 = nan(3,4);

figure;
imshow(im1);
hold on;
[x, y] = gptpts();
m1(:,1) = [x(1); y(1); 1];
m1(:,2) = [x(2); y(2); 1];
m1(:,3) = [x(3); y(3); 1];
m1(:,4) = [x(4); y(4); 1];

FNT_SZ = 20;
plot(m1(1,1), m1(2,1),'o', 'Color', 'r');
plot(m1(1,2), m1(2,2),'o', 'Color', 'r');
plot(m1(1,3), m1(2,3),'o', 'Color', 'r');
plot(m1(1,4), m1(2,4),'o', 'Color', 'r');
text(m1(1,1), m1(2,1), 'a', 'FontSize', FNT_SZ, 'Color', 'r');
```

```

text(m1(1,2), m1(2,2), 'b', 'FontSize', FNT_SZ, 'Color', 'r');
text(m1(1,3), m1(2,3), 'c', 'FontSize', FNT_SZ, 'Color', 'r');
text(m1(1,4), m1(2,4), 'd', 'FontSize', FNT_SZ, 'Color', 'r');

```



```

figure;
imshow(im2);
hold on;
[x, y] = getpts();
m2(:,1) = [x(1); y(1); 1];
m2(:,2) = [x(2); y(2); 1];
m2(:,3) = [x(3); y(3); 1];
m2(:,4) = [x(4); y(4); 1];

FNT_SZ = 20;
plot(m2(1,1), m2(2,1),'o', 'Color','r');
plot(m2(1,2), m2(2,2),'o', 'Color', 'r');
plot(m2(1,3), m2(2,3),'o', 'Color', 'r');
plot(m2(1,4), m2(2,4), 'o', 'Color', 'r');
text(m2(1,1), m2(2,1), 'a!!!', 'FontSize', FNT_SZ, 'Color', 'r');
text(m2(1,2), m2(2,2), 'b!!!', 'FontSize', FNT_SZ, 'Color', 'r');
text(m2(1,3), m2(2,3), 'c!!!', 'FontSize', FNT_SZ, 'Color', 'r');
text(m2(1,4), m2(2,4), 'd!!!', 'FontSize', FNT_SZ, 'Color', 'r');

```

, select SAME points on the second image
 as imgs taken of same person scene
 We get PAIRS of corresponding points
 in m1 and m2
 (handy selected, they can
 be automatically chosen)





↳ to build the equation for homography

Direct linear transform

Hartley & Zisserman Alg 4.2 page 109 in 2nd edition (idea behind)

We want to estimate the homography $H = [h1^t; h2^t; h3^t]^t$ that maps each $m_i = [x_i; y_i; w_i]$ in $m'_i = [x'_i; y'_i; w'_i]$. Since m_i and m'_i are in homogenous coordinates, m_i and m'_i represents the same points if exists a lambda different from zero such that $Hm_i = \lambda m'_i$. This condition can be expressed using the cross product as $m'_i \times Hm_i = 0$. This equation be written as $y'_i h_3 + x'_i h_2 - w'_i h_1 = 0$. So for each pairs of corresponding points we have three equations. But only two of the three are linear independent. So each correspondence provide 2 constraints. Four points are enough to constraints the 8 d.o.f of H ($= 3 \times 3 - 1$ to account for the scale).

```
% Build the matrix of the linear homogeneous system
% A vec(H) = 0
num_points = 4;
A = zeros(2*num_points, 9);
% the rows are the number of independent constraints
% the columns are the number of unknown (9 for a 3x3 matrix)
for i = 1:num_points
    p1 = m1(:, i);
    p2 = m2(:, i);
    A(2*i-1, :) = [zeros(1, 3), p1'*p2(3), -p1'*p2(2)];
    A(2*i, :) = [p2(3)*p1', zeros(1, 3), -p1'*p2(1)];
end

% now we have to solve an homogeneous system: svd!
[u, s, v] = svd(A);    ↗ solve system tally vector of last singular value
% let's have a look at how well the system is conditioned
s = diag(s);           ↓
% rough estimate of the nullspace dimension
nullspace_dimension = sum(s < eps * s(1) * 1e3);
if (nullspace_dimension > 1)
    fprintf('Nullspace is too large to accomodate a single solution...\n');
end
h = v(:, end);
H = reshape(h, 3, 3)';
H = H./norm(H);
% let's factorize this code into a function dlt_homography

% save('ella_homo.mat', 'H', 'm1', 'm2');
```

} buried coefficient matrix

↳ buried system coefficient matrix to solve the projective

Showing the mapping on the points

```
Hm1 = H*m1;
% dehomogenize points
Hm1 = Hm1./repmat(Hm1(3, :), 3, 1);
figure;
imshow(im2);    ↗ divide by last coord to dehomogenize
hold all;       ↗ to directly works well
scatter(m2(1, :), m2(2, :), MRK_SZ, 'ro', 'filled');   ↗ we can try to use it...
scatter(Hm1(1, :), Hm1(2, :), MRK_SZ, 'y+', 'LineWidth', 4);    ↗ I plot im CARTESIAN
MRK_SZ = 300;
scattered(m2(1, :), m2(2, :), MRK_SZ, 'ro', 'filled');
scattered(Hm1(1, :), Hm1(2, :), MRK_SZ, 'y+', 'LineWidth', 4);    ↗ they are mapped in a good way!
```

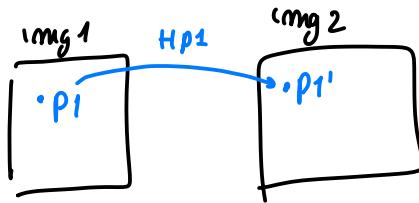
↑

Homography through 4 points is easy to estimate!

to build the homography

$$p_1 \rightarrow p_1'$$

$$H\pi \approx \pi'$$



almost true... equality holds in PROJECTIVE plane

\rightarrow then, $\exists \lambda: \lambda x^T = Hx$ ↪ the equality
Holds in PROJ Space

We says x' and Hx are PARALLEL to each other in space

meaning

$$\underbrace{x' \times Hx}_{\uparrow} \quad \left\{ \begin{array}{l} \propto \text{ if parallel} \\ \perp \text{ if perpendicular} \end{array} \right.$$

cross product represent
the area of the parallelogram
by two vector

We need to use equations to provide Algebraic
constraints in H

↳ doing matrix you
find the equations...

$$\left\{ \begin{array}{l} y_i^T H_3^T x_i - w_i^T H_2^T x_i = 0 \\ w_i^T H_1^T x_i - x_i^T H_3^T x_i = 0 \\ x_i^T H_2^T x_i - y_i^T H_1^T x_i = 0 \end{array} \right.$$

then only 2 of this one
important, last one is just
linear combination

↪



Q Noise by design!

Rendering the results

```
t = maketform('projective', H');  
J = imtransform(im1,t);
```

```
figure;  
subplot(1,3,1);  
imshow(im1);  
title('im1')  
subplot(1,3,2);  
imshow(J);  
title('transformed im1')  
subplot(1,3,3);  
imshow(im2);  
title('im2')
```

) what if I map all the pixels?
apply homography to all pixels of first image

reconstruct
same view point!



Degenerate configurations collinearity

now let's repeat the same procedure but this time we select 4 collinear points

```

m1 = nan(3,4);
m2 = nan(3,4);

% select points on the first image
figure;
imshow(im1);
hold on;
[x, y] = getpts();
m1(:,1) = [x(1); y(1); 1];
m1(:,2) = [x(2); y(2); 1];
m1(:,3) = [x(3); y(3); 1];
m1(:,4) = [x(4); y(4); 1];

FNT_SZ = 20;
plot(m1(1,1), m1(2,1),'o', 'Color', 'r');
plot(m1(1,2), m1(2,2),'o', 'Color', 'r');
plot(m1(1,3), m1(2,3),'o', 'Color', 'r');
plot(m1(1,4), m1(2,4),'o', 'Color', 'r');
text(m1(1,1), m1(2,1), 'a', 'FontSize', FNT_SZ, 'Color', 'r');
text(m1(1,2), m1(2,2), 'b', 'FontSize', FNT_SZ, 'Color', 'r');
text(m1(1,3), m1(2,3), 'c', 'FontSize', FNT_SZ, 'Color', 'r');
text(m1(1,4), m1(2,4), 'd', 'FontSize', FNT_SZ, 'Color', 'r');

```

what if we take COLINEAR points?

when the 4 points are in the same line,

degenerate configuration NOT in cell
plane! we don't reconstruct all
the plane → nullspace too big!

$\sim \hat{H}$ behaves bad on

remain pixels



select points on the second image

```
figure;
imshow(im2);
hold on;
[x, y] = getpts();
m2(:,1) = [x(1); y(1); 1];
m2(:,2) = [x(2); y(2); 1];
m2(:,3) = [x(3); y(3); 1];
m2(:,4) = [x(4); y(4); 1];

FNT_SZ = 20;
plot(m2(1,1), m2(2,1), 'o', 'Color', 'r');
plot(m2(1,2), m2(2,2), 'o', 'Color', 'r');
plot(m2(1,3), m2(2,3), 'o', 'Color', 'r');
plot(m2(1,4), m2(2,4), 'o', 'Color', 'r');
text(m2(1,1), m2(2,1), 'a!!!', 'FontSize', FNT_SZ, 'Color', 'r');
text(m2(1,2), m2(2,2), 'b!!!', 'FontSize', FNT_SZ, 'Color', 'r');
text(m2(1,3), m2(2,3), 'c!!!', 'FontSize', FNT_SZ, 'Color', 'r');
text(m2(1,4), m2(2,4), 'd!!!', 'FontSize', FNT_SZ, 'Color', 'r');
```






compute the homography

```
[H,A_collinear] = dlt_homography(m1,m2);

Hm1 = H*m1;
% dehomogenize points
Hm1 = Hm1./repmat(Hm1(3,:),3,1);
figure;
imshow(im2);
hold all;
MRK_SZ = 300;
scatter(m2(1,:),m2(2,:),MRK_SZ,'ro','filled');
scatter(Hm1(1,:),Hm1(2,:),MRK_SZ,'y');

% the condition number of A is much larger than 1, the matrix is sensitive to the inverse calculation.
% let's see what happen to the nullspace of A when points are collinear
cond(A)
cond(A_collinear);
[~,s_collinear,~]=svd(A_collinear);
s_collinear = diag(s_collinear);
figure;
bar([s(end-1),s_collinear(end-1),s(end),s_collinear(end)])
```

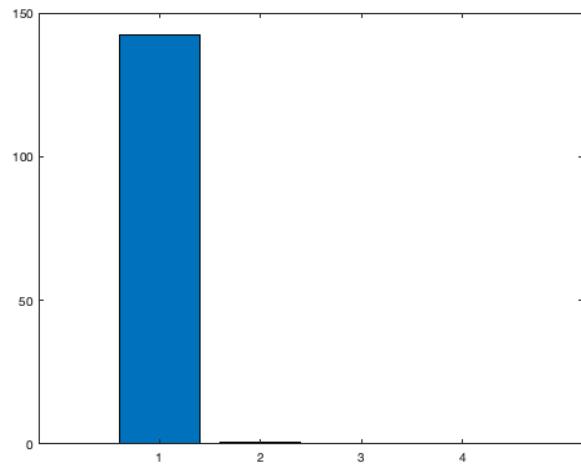
→ plotting the singular values of
SVD decomposition, as we can see,
the system is BAD CONDITIONED..

Big Nullspace!

the transformation compute
well points on line but
bad outside



I get a useless transformation =>



let's show the result

```
t = maketform('projective', H');
J = imtransform(im1,t);
```

```
figure;
subplot(1,2,1);
imshow(im2);
subplot(1,2,2);
imshow(J);
```

Warning: The condition number of A is 1604493683560.505.



Published with MATLAB® R2021b

Always do computation,
BUT we should understand geometry
to know when computation makes sense

seen homography transformation...

Image warping

We now have Matlab function that takes
 $(\text{image}, H) \rightarrow \text{output} := \text{transform image}$

Render an image transformed via an homography References: Andrea Fusiello, Visione Computazionale | Chapter 10 on image mosaicing

$\text{im} = \text{imwarp}(\text{im}, H)$

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Luca Magri for comments and suggestions please send an email to luca.magri@polimi.it

$\text{im} = \text{imwarp}(\text{im}, H)$

Contents

- overlay the transformed im1 over im2
- a bigger canvas
- determine the dimension of the canvas
- render the novel image using bilinear interpolation

```
clear;
close all;
clc;
```

We are given: - two images im1 and im2 - the homography between im1 and im2 our aim is to synthesize a novel image obtained by superimposing im2 and $H(\text{im1})$. Let's use the backward mapping.

```
im1 = imread("E3_data/ella1.jpg");
im2 = imread("E3_data/ella2.jpg");
im1 = imresize(im1, 0.6);
im2 = imresize(im2, 0.6);
data = load('E3_data/ella_homo.mat');
m1 = data.m1;
m2 = data.m2;
H = data.H;
```

overlay the transformed im1 over im2

let's use bilinear interpolation

```
canvas = im2;
lambda = 0.7;
for i=1:size(canvas,1)
    for j = 1:size(canvas,2)
        q =[j;i];
        % apply the inverse transformation
        p = inv(H)*q; % column is x and row is y
        p = p./p(3); % cartesian coordinate in the plane of the first image
        p = p; % cartesian ( $x_1, x_2, y_1, y_2$ ) → to decide color of pixel, we
        % if the preimage p lies inside the first image → bilinear combination
        if(round(p(1))>0 && round(p(1))<=size(im1,2) && round(p(2))>0 && round(p(2))<=size(im1,1))
            for ch = 1:3
                source_value = im1(round(p(2)),round(p(1)),ch);
                % apply bilinear interpolation
                if(p(1)>1 && p(1)<size(im1,2) && p(2)>1 && p(2)<size(im1,1))
                    y1 = floor(p(2));
                    y2 = ceil(p(2));
                    % interpolate along the x axis
                    x1 = floor(p(1));
                    x2 = ceil(p(1));
                    % at y1
                    v1 = (p(1)-x1)*im1(y1,x1,ch) + (x2-p(1))*im1(y1,x2,ch);
                    % at y2
                    v2 = (p(1)-x1)*im1(y2,x1,ch) + (x2-p(1))*im1(y2,x2,ch);
                    % interpolate along the y axis
                    source_value = (p(2)-y1).*v1 + (y2-p(2)).*v2;
                end
                canvas(i,j,ch) = (1-lambda).*canvas(i,j,ch) + lambda.*source_value;
            end
        end
    end
end
```

" make tform "

under this transformation,

to do this on our own it
is easy to do on our own

from an image we know how to deform it,
we want to paint new image distorted
I want to take each pixel and translate it
and do for each pixel, translate
the whole image

from a slice of the image, apply transform
 $P + H$ multiply by 2 cell pixels
as a translation with unknown values in the
image ... FORWARD MAPPING is feasible BUT NOT good...

Usually BACKWARD MAPPING is used.

from target image, I move pixel by $P' = \frac{1}{2}P$,

we check by inverting if inside preply

$x = 2 \rightarrow x' = 1$ back project pixels to source
 $x = 4 \rightarrow x' = 2$ to find if properly map... start
from destination to source

→ to decide color of pixel, we

bilinear combination

SOURCE !

backward mapping
to avoid any holes!

Possible because
perspective transform
(certainly invertible)

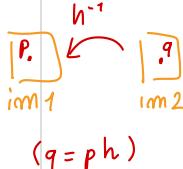


Image warping

Render an image transformed via an homography References: Andrea Fusiello, Visione Computazionale. Chapter 10 on image mosaicing

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Contents

- overlay the transformed im1 over im2
 - a bigger canvas
 - determine the dimension of the canvas
 - render the novel image using bilinear interpolation

```
clear;  
close all;  
clc;
```

We are given: - two images $im1$ and $im2$ - the homography between $im1$ and $im2$ our aim is to synthetize a novel image obtained by superimposing $im2$ and $H(im1)$. Let's use the backward mapping.

```

im1 = imread("E3_data/ella1.jpg");
im2 = imread("E3_data/ella2.jpg");
im1 = imresize(im1, 0.6);
im2 = imresize(im2, 0.6);
data = load('E3_data/ella_homo.mat');
m1 = data.m1;
m2 = data.m2;
H = data.H;

```

overlay the transformed im1 over im2

let's use bilinear interpolation

```

canvas = im2;
lambda = 0.7;
for i=1:size(canvas,1)
    for j = 1:size(canvas,2)
        q =[j;i;1];
        % apply the inverse transformation
        p = inv(H)*q; % column is x and row is y
        p = p./p(3); % cartesian coordinate in the plane of the first image
        p = p';
        % if the preimage p lies inside the first image
        if(round(p(1))>0 && round(p(1))<=size(im1,2) && round(p(2))>0 && round(p(3))>0)
            for ch = 1:3
                source_value = im1(round(p(2)),round(p(1)),ch);
                % apply bilinear interpolation
                if(p(1)>1 && p(1)<size(im1,2) && p(2)>1 && p(2)<size(im1,1))
                    y1 = floor(p(2));
                    y2 = ceil(p(2));
                    % interpolate along the x axis
                    x1 = floor(p(1));
                    x2 = ceil(p(1));
                    % at y1
                    v1 = (p(1)-x1)*im1(y1,x1,ch) + (x2-p(1))*im1(y1,x2,ch);
                    % at y2
                    v2 = (p(1)-x1)*im1(y2,x1,ch) + (x2-p(1))*im1(y2,x2,ch);
                    % interpolate along the y axis
                    source_value = (p(2)-y1).*v1 + (y2-p(2)).*v2;
                end
                canvas(i,j,ch) = (1-lambda).*canvas(i,j,ch) + lambda.*source_value;
            end
        end
    end
end

```

$\rightarrow p$, we get the value of
we take bilinear combination

*bilinear
interpolate*

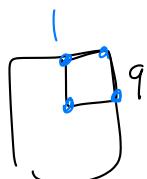
by definition of
bilinear interpolation

blending the colors, weightings linear combinations.

$q \rightarrow p$, we get the value of the pixel coordinate...

We take bimean combination of colors

sum up the colors,
such that



the color
influence most
are the closest ones

We get by the
intensity by the
area of neighbor points

{ by definition of
bilinear interpolation } ↳

```
figure;
imshow(canvas);
% still some misalignment are visible
```



New image overlayed!



some
misalignments due
to random
points
selection

We are
losing
INFORMATION
↓
Just printing
in the 2nd image

I want overlayed image and
see both with full resolution

a bigger canvas

first we must create a canvas to display our result. The canvas must be large enough to contain both im2 and im1. let's see where the corners of im1 are mapped

```
tl = H*[0;0;1]; % where top-left is mapped
tl = tl./tl(3);
tr = H*[size(im1,2);0;1]; % where top-right is mapped
tr = tr./tr(3);
bl = H*[0;size(im1,1);1]; % where bottom-left is mapped
bl = bl./bl(3);
br = H*[size(im1,2);size(im1,1);1];
br = br./br(3); % where bottom-right is mapped

% here we display the perimeter of im1 via H
figure;
imshow(canvas);
hold all;
plot(tl(1),tl(2),'ro','MarkerSize',10,'MarkerFaceColor','r');
plot(tr(1),tr(2),'ro','MarkerSize',10,'MarkerFaceColor','r');
```

→ to put
one outside
other, we can
just use
bigger canvas!

I check where points
mapped and draw compute
bounding box

← but image reference system needs
to be moved!
↓ handle mappings
to draw bigger picture

```

plot(bl(1),bl(2),'ro','MarkerSize',10,'MarkerFaceColor','r');
plot(br(1),br(2),'ro','MarkerSize',10,'MarkerFaceColor','r');
line([tl(1),tr(1)],[tl(2),tr(2)],'Color','r','LineWidth',3);
line([tr(1),br(1)],[tr(2),br(2)],'Color','r','LineWidth',3);
line([br(1),bl(1)],[br(2),bl(2)],'Color','r','LineWidth',3);
line([bl(1),tl(1)],[bl(2),tl(2)],'Color','r','LineWidth',3);

```



determine the dimension of the canvas

```

w_max = max([size(im2,2),tl(1),tr(1),bl(1),br(1)]);
w_min = min([size(im2,2),tl(1),tr(1),bl(1),br(1)]);
w = ceil(w_max-w_min);
h_max = max([size(im2,1),tl(2),tr(2),bl(2),br(2)]);
h_min = min([size(im2,1),tl(2),tr(2),bl(2),br(2)]);
h = ceil(h_max-h_min);
big_canvas = uint8(255.*ones(h,w,3));
big_canvas(-h_min+1:-h_min+size(im2,1), -w_min+1:-w_min+size(im2,2),:) = im2;
figure;
imshow(big_canvas);

```

Warning: Integer operands are required for colon operator when used as index.
Warning: Integer operands are required for colon operator when used as index.



render the novel image using bilinear interpolation

```
for i=1:size(big_canvas,1)
    for j = 1:size(big_canvas,2)
        % bring [j;i;1] from the reference system of the big_canvas to the
        % reference system of im2
        q =[j+w_min;i+h_min;1]; expression in parallel reference system
        % apply the inverse transformation
        p = inv(H)*q; % column is x and row is y
        p = p./p(3); % cartesian coordinate in the plane of the first image
        % if the preimage p lies inside the first image
        if(round(p(1))>0 && round(p(1))<=size(im1,2) && round(p(2))>0 && round(p(2))<=size(im1,1))
            for ch = 1:3
                source_value = im1(round(p(2)),round(p(1)),ch);
```

```

% apply bilinear interpolation
if(p(1)>1 && p(1)<size(im1,2) && p(2)>1 && p(2)<size(im1,1))
    y1 = floor(p(2));
    y2 = ceil(p(2));
    % interpolate along the x axis
    x1 = floor(p(1));
    x2 = ceil(p(1));
    % at y1
    v1 = (p(1)-x1)*im1(y1,x1,ch) + (x2-p(1))*im1(y1,x2,ch);
    % at y2
    v2 = (p(1)-x1)*im1(y2,x1,ch) + (x2-p(1))*im1(y2,x2,ch);
    % interpolate along the y axis
    source_value = (p(2)-y1).*v1 + (y2-p(2)).*v2;
end
big_canvas(i,j,ch) = (1-lambda).*big_canvas(i,j,ch) + lambda.*source_value;
end % end for channels
end % end if inside the image
end % end for j
end % end for i

```

```

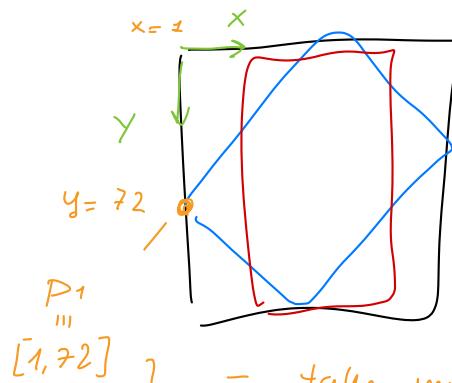
figure;
imshow(big_canvas);

```

↳ there is an image size problem... so we need to crop to get the pre-image!



the



↳ I take $im1$ and I see where lay! but H is defined between $im2 \rightarrow im2$
 you want to reference with offset... ↴

so I need to subtract offset vector...

then I need to cook @ color with bilinear interpolation



the PROBLEM is because homography is not defined on Big canvas BUT on 2nd image!

↑
 I want to express pixel coord in 2nd image ref

We can draw the
full image with
a bigger canvas →



Stiching

Estimate an homography to stitch together two images acquired by a rotating camera: - Normalization of the points - Direct Linear Transform - Transfer error minimization

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Contents

- load images
 - get correspondences
 - Preconditioning
 - compute the homography via DLT
 - compute the errors
 - iterative minimization
 - compute the errors

load images

```
clc;  
close all;  
clear;
```

another possible application is
IMAGE STITCHING,
rotating camera keeping axis fixed!
ROTATION around camera

ROTATION around camera

```
im1 = imread("E3_data/pragaA.jpeg");
im2 = imread("E3_data/pragaB.jpeg");
im1 = imresize(im1,0.6);
im2 = imresize(im2,0.6);
```

```
%data = load('E3_data/ella_homo.mat');  
%H0 = data.H;
```

get correspondences

```

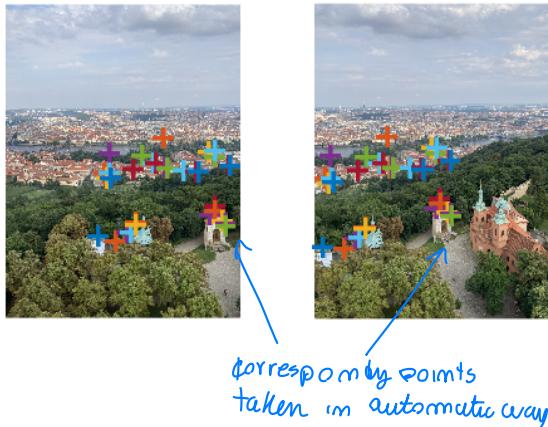
do_load_correspondences = 1;

if(do_load_correspondences)
    data = load("E3_data/praga_matches.mat");
    m1 = double(data.m1);
    m2 = double(data.m2);

    % add a little bit of noise
    sigma = 0;
    m1(1:2,:) = m1(1:2,:) + sigma.*randn(2,size(m1,2));
    m2(1:2,:) = m2(1:2,:) + sigma.*randn(2,size(m2,2));
end

figure;
subplot(1,2,1);
imshow(im1);
hold all;
for i=1:size(m1,2)
    plot(m1(1,i),m1(2,i),'+','MarkerSize',15,'LineWidth',4);
end
subplot(1,2,2);
imshow(im2);
hold all;
for i=1:size(m2,2)
    plot(m2(1,i),m2(2,i),'+','MarkerSize',15,'LineWidth',4);
end

```



I want to get a MOSAIC of the two images, as when we take PANORAMA photos...
do this with HOMOGRAPHY

Preconditioning

to improve numerical stability data should be normalized in order to have zero mean and mean distance to the center equal to $\sqrt{2}$. let's perform this transformation T_i that translate the points and rescale them Reference: Hartley, Zisserman

```
% let's encapsulate this in a function normalize2D points
% and let's use it on m2
num_points = size(m1, 2);
m1 = m1./repmat(m1(3, :), 3, 1);

centroid = mean(m1, 2);
% distance from the centroid
scale = mean(sqrt(sum((m1-repmat(centroid, 1, num_points)).^2, 1)));

T1 = diag([sqrt(2)/scale; sqrt(2)/scale; 1]);
T1(1:2, 3) = -sqrt(2)*centroid(1:2)/scale;
m1_cond = T1*m1;

% let's encode this preconditioning step in a function
[m2_cond, T2] = precondition(m2);
```

in homogeneous coord points are not very balanced!

random points can have very different order, this cause matrix having points of $10^4, 10^{-1} \dots$ when solving DLT with so different orders, can be solved by PRE-CONDITIONING

compute the homography via DLT

```
H0 = dlt_homography(m1_cond, m2_cond);

% Once we get the homography we have to get rid of the normalization
% process by applying the inverse transformations
H0 = T2\H0*T1;
H0 = H0./norm(H0);

lambda = 0.5;
[mosaic_big, offset_x, offset_y] = image_mosaic_big(im1, im2, H0, lambda);
```

Warning: Integer operands are required for colon operator when used as index.
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 Warning: Integer operands are required for colon operator when used as index.

compute the errors

```
%H0(3,3) = 1e-9*randn(1,1);

% we consider the symmetric transfer error
hml = H0*m1;
hml = hml./repmat(hml(3, :), 3, 1);
invhml2 = inv(H0)*m2;
invhml2 = invhml2./repmat(invhml2(3, :), 3, 1);
transfer_error = 0;
for i = 1:size(m1, 2)
  d1 = norm(m1(:, i) - invhml2(:, i));
  d2 = norm(m2(:, i) - hml(:, i));
  transfer_error = transfer_error + (d1 + d2);
end
disp(transfer_error)
```



we apply geometry, transforming
im1, im2 to make all points
around origin in circle of radius 2!

I can do it!

↓
HARTLEY NORMALIZATION

change picture shape, putting all
in a disc → all same magnitude!

having imy1, imy2 \Rightarrow estimate H? BUT imbalance coordinates!

affine transformation easy to invert, done
for both ... homography done on this planes

Preconditioning

to improve numerical stability data should be normalized in order to have zero mean and mean distance to the center equal to $\sqrt{2}$. let's perform this transformation via a planar transformation T_1 that translates the points and rescale them Reference: Hartley, Zisserman

```
% let's encapsulate this in a function normalize2D points
% and let's use it on m2
num_points = size(m1, 2);
m1 = m1./repmat(m1(3, :), 3, 1);

centroid = mean(m1, 2);
% distance from the centroid
scale = mean(sqrt(sum((m1-repmat(centroid, 1, num_points)).^2, 1)));

$$\begin{bmatrix} \sqrt{2} & t \\ 0 & 1 \end{bmatrix}$$
 as east = affine transform TRANSLATION & SCALING!
T1 = diag([sqrt(2)/scale; sqrt(2)/scale; 1]);
T1(1:2, 3) = -sqrt(2)*centroid(1:2)/scale;
m1_cond = T1*m1;

% let's encode this preconditioning step in a function
[m2_cond, T2] = precondition(m2);
```

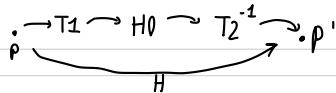
than final homography
 $\tilde{H} \cdot N^{-1}$ inverse, avoid
 numerical instability
 ↗ good practice to normalize
 { put point to get
 better condition numbers }!

compute the homography via DLT

```
H0 = dlt_homography(m1_cond, m2_cond);

% Once we get the homography we have to get rid of the normalization
% process by applying the inverse transformations
H0 = T2\H0*T1; ← final homography as
H0 = H0./norm(H0);

lambda = 0.5;
[mosaic_big, offset_x, offset_y] = image_mosaic_big(im1, im2, H0, lambda);
```



Warning: Integer operands are required for colon operator when used as index.
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compute the errors \rightarrow to check how well homography fit the data we use squared difference between points and projected

```
% we consider the symmetric transfer error
hml = H0*m1;
hml = hml./repmat(hml(3, :), 3, 1);
invhm2 = inv(H0)*m2;
invhm2 = invhm2./repmat(invhm2(3, :), 3, 1);
transfer_error = 0;
for i = 1:size(m1, 2)
    d1 = norm(m1(:, i) - invhm2(:, i));
    d2 = norm(m2(:, i) - hml(:, i));
    transfer_error = transfer_error + (d1 + d2);
end
disp(transfer_error)
```

113.5701

- we get error as squared algebraic error

```

cmap = brewermap(size(m1,2), 'Paired');
figure;
imshow(mosaic_big);
hold all;
for i=1:size(m1,2)
    plot(m2(1,i)+offset_x,m2(2,i)+offset_y,'o','MarkerSize',25,'Linewidth',4,'Color',cmap(i,:));
    plot(hm1(1,i)+offset_x,hm1(2,i)+offset_y,'+', 'MarkerSize',25,'Linewidth',4,'Color',cmap(i,:));
end

```

I'm fitting problem
 ↴ ↴
 ALGEBRAIC GEOMETRIC
 in the bending of
 time, Algebraic error
 change a lot, while
 geometric NOT!

if ↓
 in HOMOGRAPHY
 we minimize Algebraic
 error we get error that
 depends on how homography
 lay in space (NOT invariant
 under geometric transform)
 ↑
 after you compute \tilde{H} , estimate
 then you feed \tilde{H} to optimizer
 function NON LINEAR which
 minimize geometric error
 ↴ IMPROVE a lot!



① DCT to get \tilde{H}
 minimize Algebraic
 error when there are
 correspondances

iterative minimization

```

MaxIterations = 1e5;
FunctionTol = 1e-8;
StepTol = 1e-16;

options = optimoptions('lsqnonlin','Algorithm','levenberg-marquardt',...
    'MaxIterations',MaxIterations,...
    'StepTolerance', StepTol, 'FunctionTolerance', FunctionTol,'Display','iter','Diagnostics','on');

fobj = @(h) get_transfer_errors(h,m1,m2);
h0 = H0(:);
h = lsqnonlin(fobj,h0,[],[],options);
H = reshape(h,3,3);
H = H./norm(H);

```

Diagnostic Information

Number of variables: 9

Functions
 Objective:

$\oplus(h) \text{get_transfer_errors}(h, m1, m2)$

solved from
 a starting point ~
 as the estimate \tilde{H}

we want to minimize a geometric error!
 (NOT algebraic)

② from \tilde{H} , we
 get geometric
 error which is
 more significative
 (more informative)

we wanna minimize
 that, NON LINEAR.
 that require solver!

from \tilde{H} guess + definition of reprojection error

feed both to NON LINEAR
SOLVER, that find a
better solution H^*

```
Gradient:          finite-differencing
```

```
Number of lower bound constraints:      0  
Number of upper bound constraints:      0
```

```
Algorithm selected  
levenberg-marquardt
```

```
End diagnostic information
```

Iteration	Func-count	Residual	First-Order optimality	Lambda	Norm of step
0	10	391.935	4.38e+11	0.01	
1	41				5.70508e-17

```
Local minimum possible.  
lsqnonlin stopped because the relative size of the current step is less than  
the value of the step size tolerance.
```

```
lambda = 0.5;  
[mosaic_big,offset_x,offset_y] = image_mosaic_big(im1,im2,H,lambda);
```

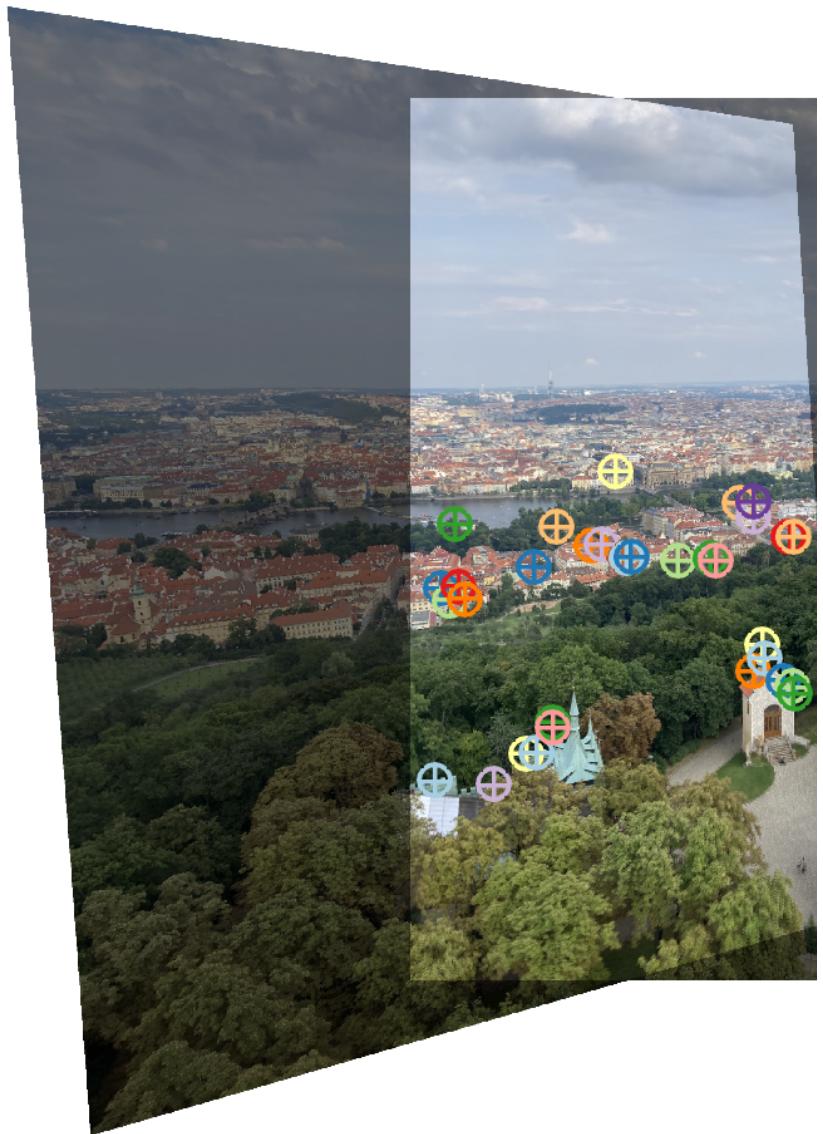
```
Warning: Integer operands are required for colon operator when used as index.  
Warning: Integer operands are required for colon operator when used as index.  
Warning: Integer operands are required for colon operator when used as index.  
Warning: Integer operands are required for colon operator when used as index.  
Warning: Integer operands are required for colon operator when used as index.  
Warning: Integer operands are required for colon operator when used as index.
```

compute the errors

we consider the symmetric transfer error

```
hml = H*m1;  
hml = hml./repmat(hml(3,:),3,1);  
invhm2 = inv(H)*m2;  
invhm2 = invhm2./repmat(invhm2(3,:),3,1);  
transfer_error = 0;  
for i = 1:size(m1,2)  
    d1 = norm(m1(:,i)-invhm2(:,i));  
    d2 = norm(m2(:,i)-hml(:,i));  
    transfer_error = transfer_error + (d1 + d2);  
end  
disp(transfer_error)
```

113.5701

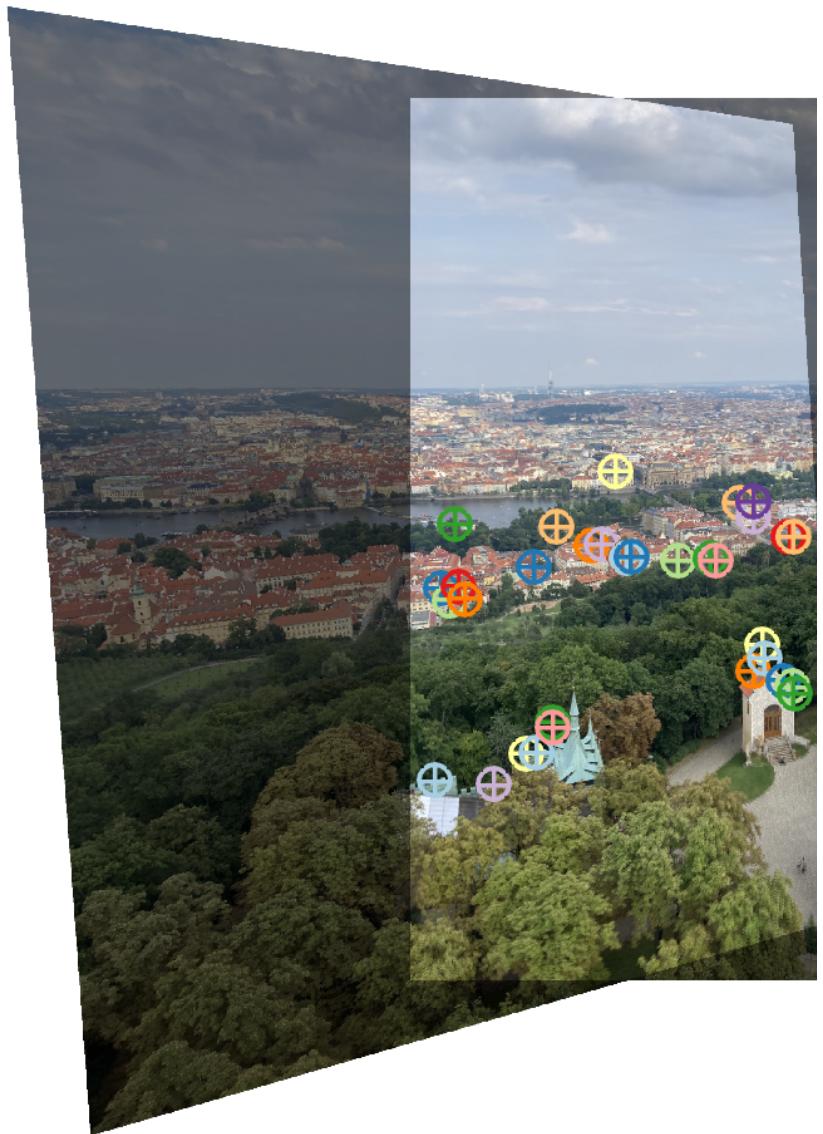


→ im new image, error get decreased!

```
cmap = brewermap(size(m1,2),'Paired');
figure;
imshow(mosaic_big);
hold all;
for i=1:size(m1,2)
    plot(m2(1,i)+offset_x,m2(2,i)+offset_y,'o','MarkerSize',25,'Linewidth',4,'Color',cmap(i,:));
    plot(hml(1,i)+offset_x,hml(2,i)+offset_y,'+', 'MarkerSize',25,'Linewidth',4,'Color',cmap(i,:));
end

H./H(3,3)
% Affine homographies have their third row fixed to
% [0, 0, 1] and arise when the image displacements come from a small
% image region or a large focal length lens is being used. When
% estimating affine homographies, the non-linear refinement is omitted: Non-linear refinement is unnecessary
% for them because in that case, the algebraic error equals the
% geometric one.
```

```
ans =
1.0e+03 *
0.0014    0.0001   -1.1085
0.0002    0.0013   -0.2522
0.0000    0.0000    0.0010
```



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IF we have multiple images we can combine it
to get full scene mosaic

Rectification

Rectification .\W//.\W//.\W//.\W//.\W//.\W//.\W//.\W//.\W//.\W//.\W//.\W//.\W//.

Image Analysis and Computer Vision Politecnico di Milano

Luca Magri for comments and suggestions please send an email to luca.magri@polimi.it

W°//W°//W°//W°//W°//W°//W°//W°//W°//W°//W°//W°//W°//

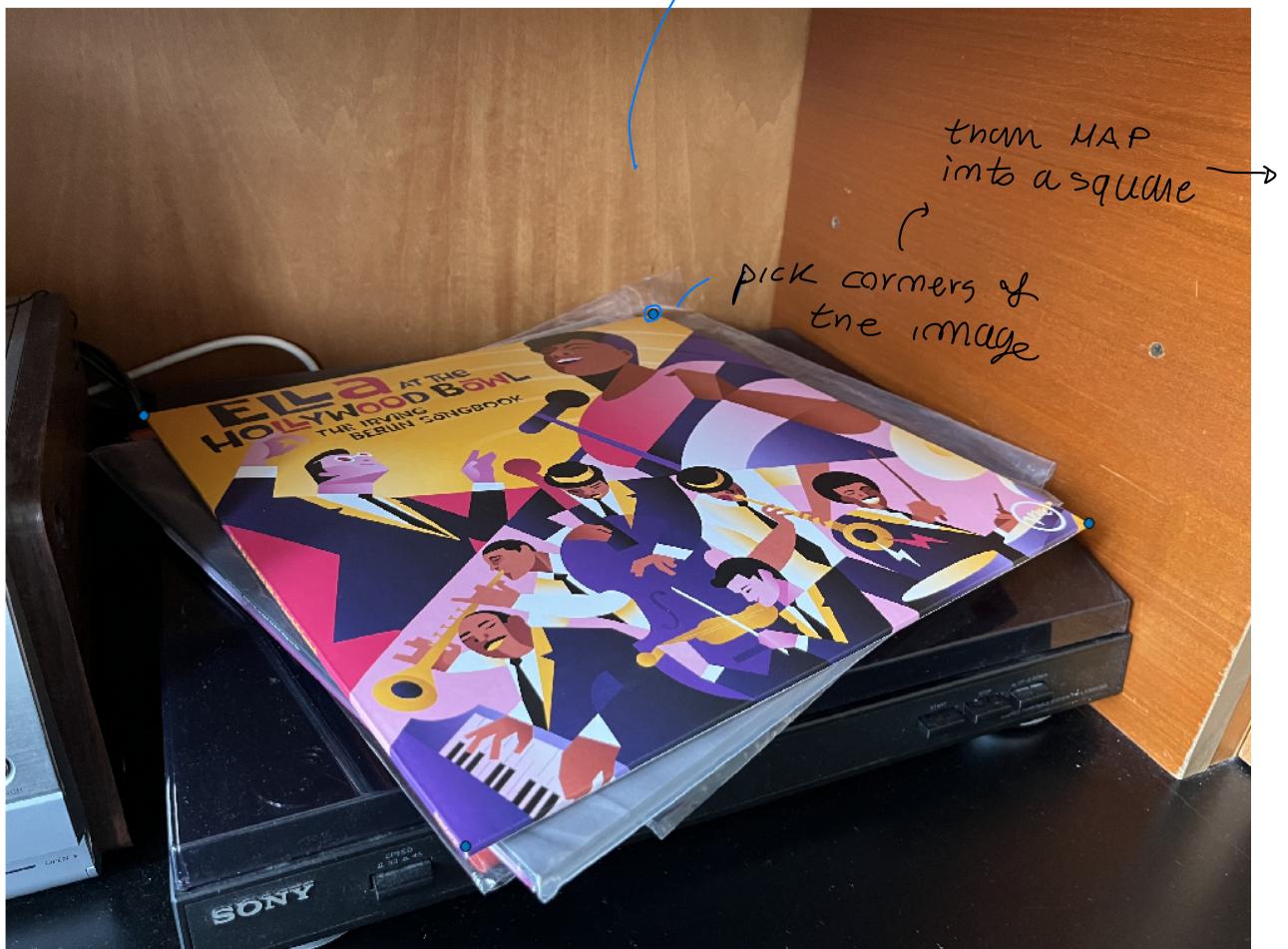
```
clear;
clc;
close all;
```

```
im = imread('E3_data/giradischi.jpg');
figure;
imshow(im);

a = drawpoint();
b = drawpoint();
c = drawpoint();
d = drawpoint();
```

Another application

how to have nice
pictures of it?



```

m1 = [a.Position;b.Position;c.Position;d.Position];
m1 = [m1'; ones(1,4)];
l = 500;
m2=[1 1 1 1;...
      1 1 1 1;...
      1 1 1 1];

```

positions of square
of dimension 500

```
[m1_normalized,T1] = precondition(m1);
[m2_normalized,T2] = precondition(m2);
H = dlt_homography(m1,m2);
%H = inv(T2)*H*T1;
H = H./norm(H);
```

} normalize all ..

```

lambda =1;
[mosaic] = image_mosaic_same(im,uint8(zeros(1,1,3)),H,lambda);
figure; imshow(mosaic)

```



↘ MAP image pick
 corners to a SQUARE!
 render
 the image
 as Rectified
 (this is what
 happens when
 you scan in the phone!)
 ↓
 RECTIFICATION
 as of parallel img

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NOTICE: homography is defined up to

scale factor

↳ to avoid numerical instability
 is better to have determinant

constrained

↳

It is a nice idea
 to normalize all
 by norm(MATRIX)... OR you get too big numbers

strange results
 can be caused by BAD numeric
 computation... NOT normalization,
 make small entries and
 divide by Frobenius Norm