

Geometry and Algebra in Computer Vision & Robotics

T o m a s P a j d l a

with contributions from

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Czech Technical University Prague

Center for Machine Perception pajdla





Geometry of Vision & Robotics Group

Tomas Pajdla

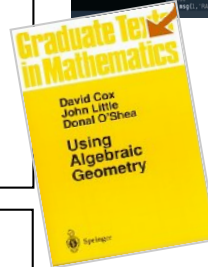
Assistant professor

Center for Machine Perception, Dpt. of Cybernetics, FEE, CTU in Prague

pajdla@cmp.felk.cvut.cz

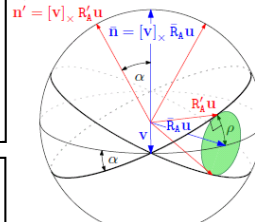
Research & applications

- 3D reconstruction
- Photogrammetry
- Robotics & manipulation
- Algebra, geometry, optimization



Teaching (PhD, MSc, BSc)

Geometry of Computer Vision & Robotics

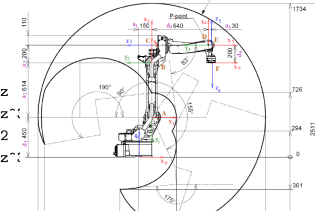


Macaulay2 program over the finite field $\mathbb{Z}/13$

```

R = ZZ/13[x,y,z, MonomialOrder=>GrevLex];
I = ideal(8*x^2*y^2 + 5*x*y^3 + 3*x^3*z,
          x^5 + 2*y^3*z^2 + 13*y^2*z^2,
          8*x^3 + 12*y^3 + x*z^2,
          7*x^2*y^4 + 18*x*y^3*z^2 + y^3*z^2);
G = gens gb I

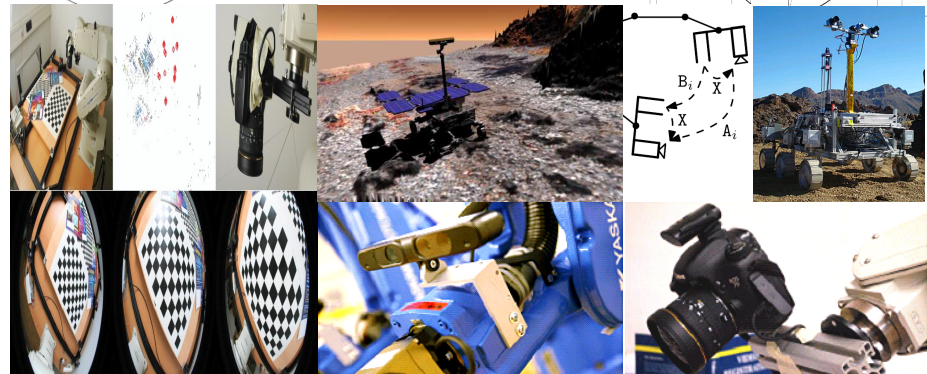
```



Research funding



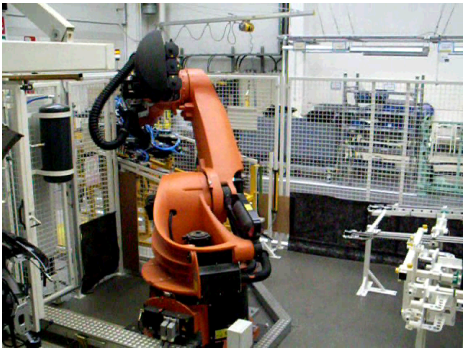
Industry collaboration



www.neovision.cz



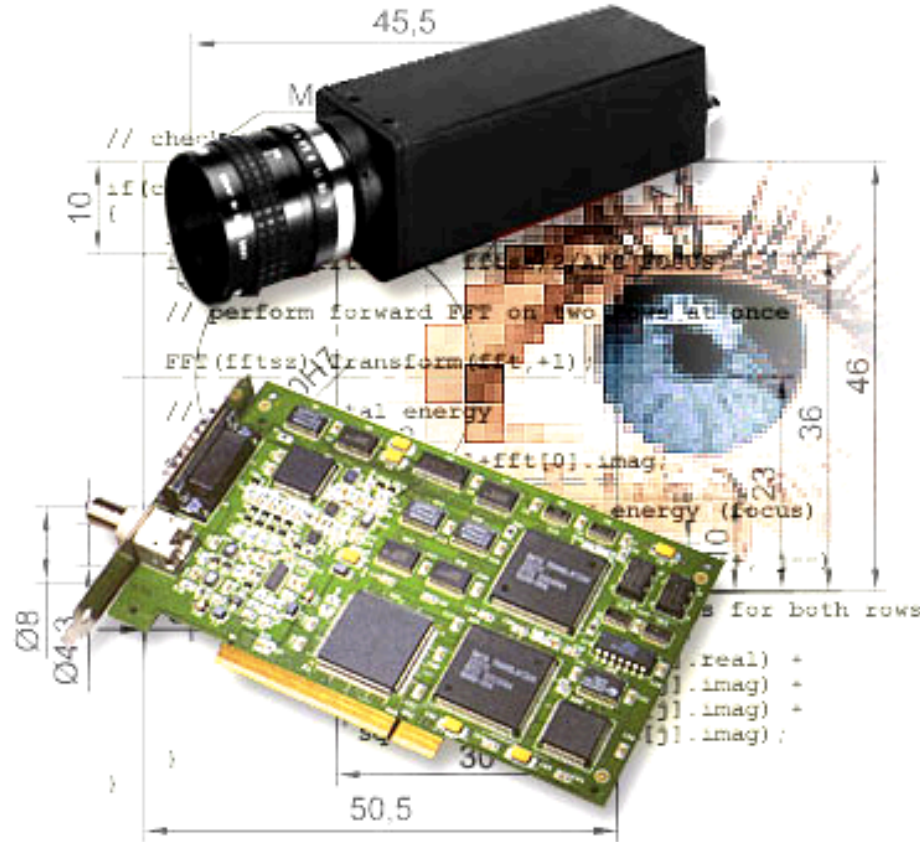
SERVIS robotics



AUTOMOTIVE robotics

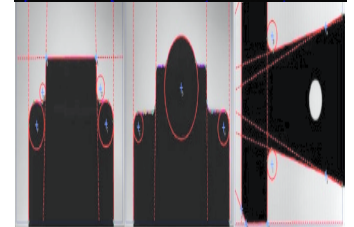
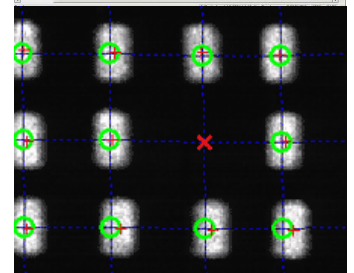
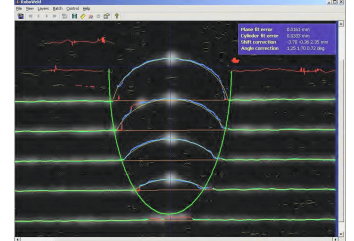
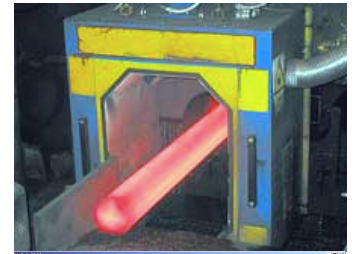


RETAIL monitoring



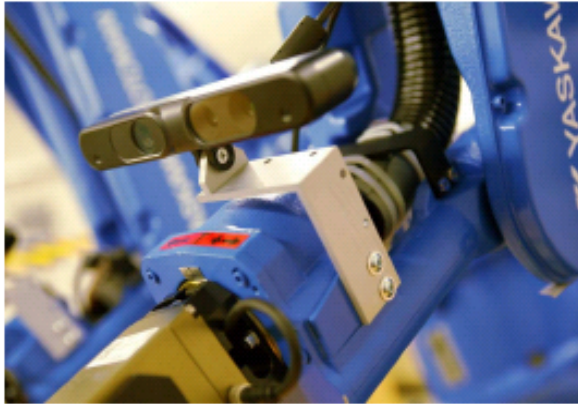
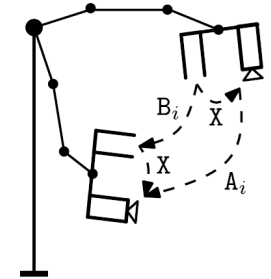
NEOVISION
Industrial Vision Systems

Machine vision and image processing
technology for industry and medicine

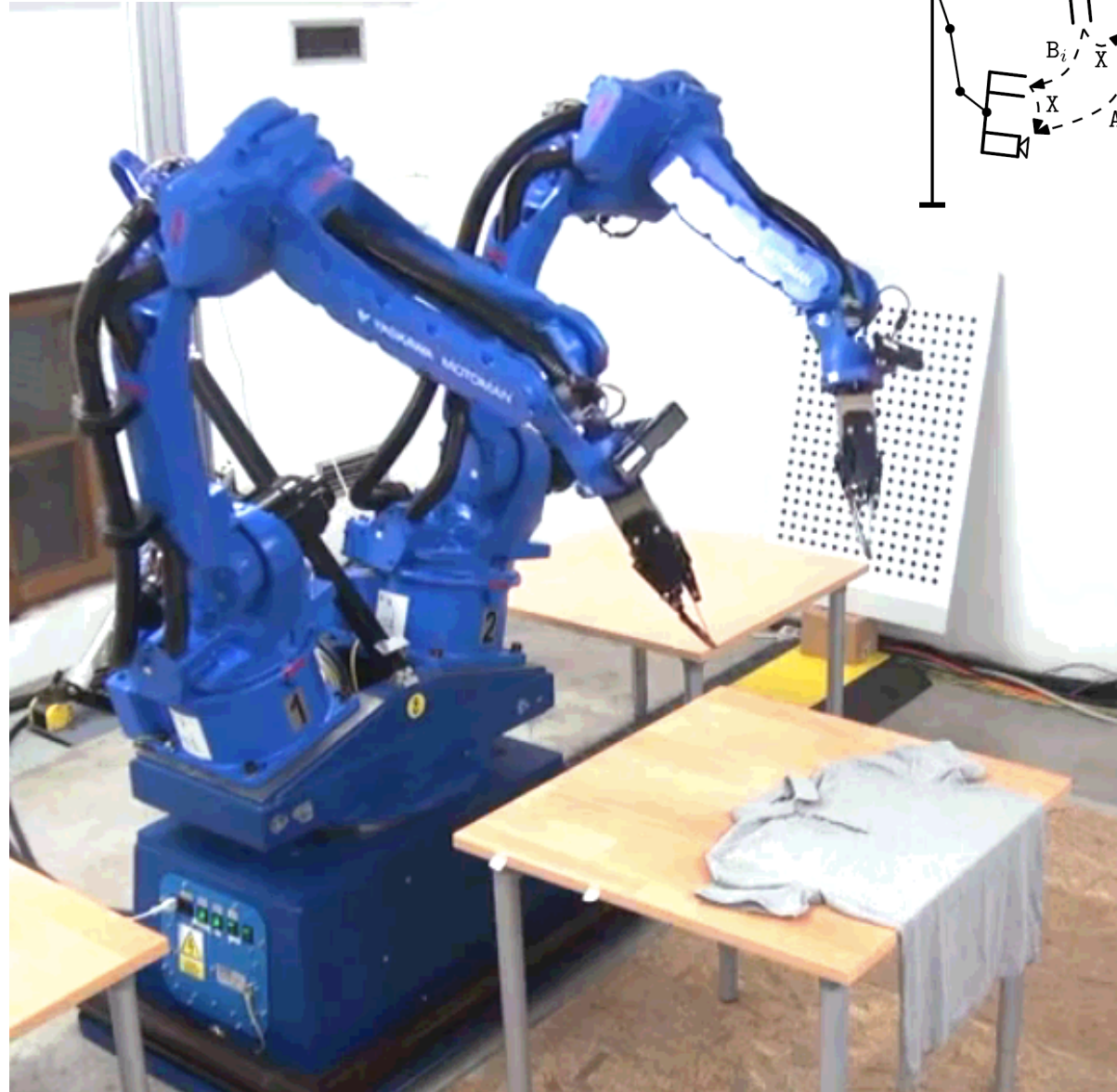




3D robotics

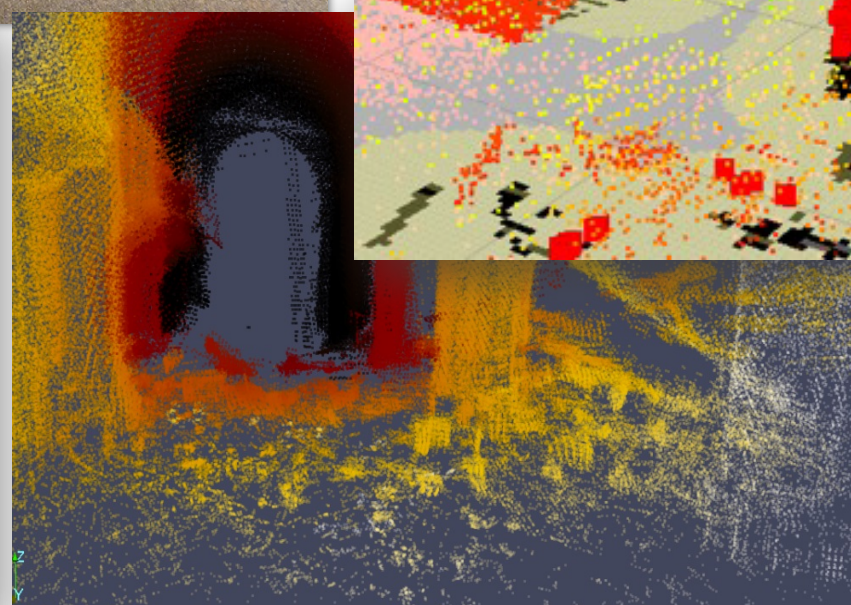
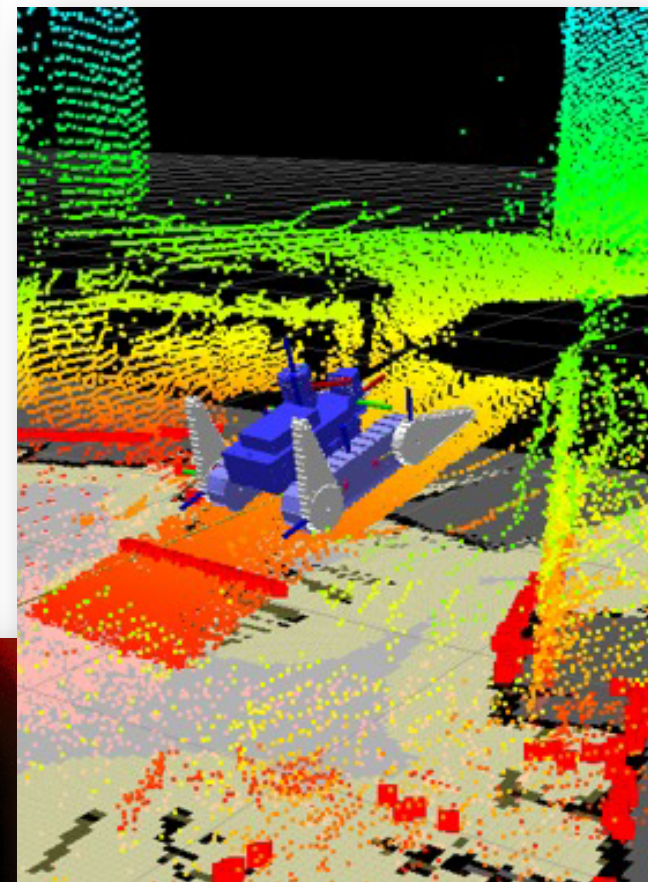
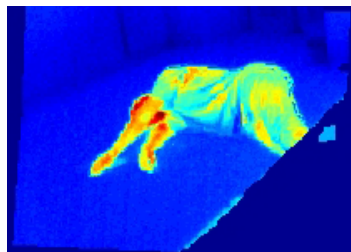
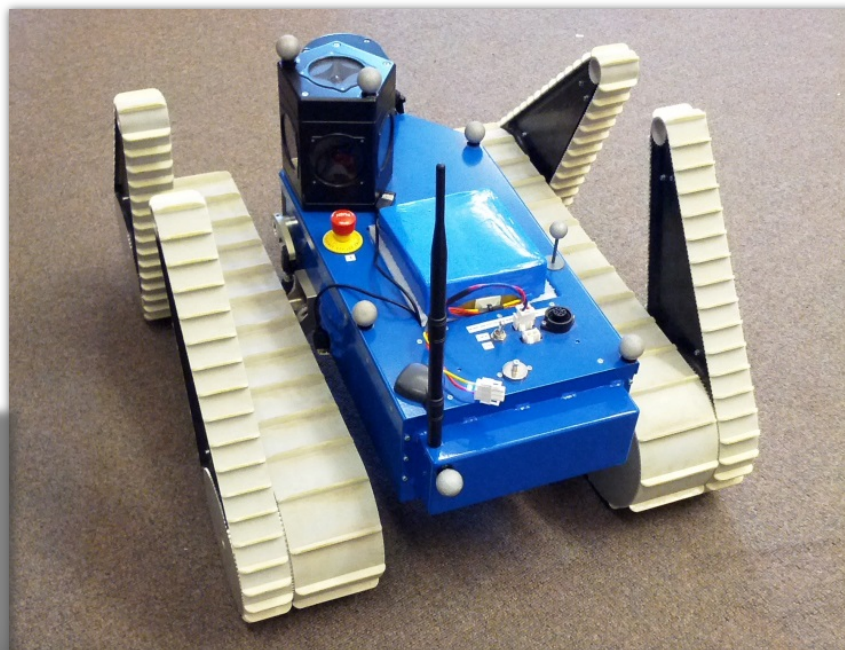
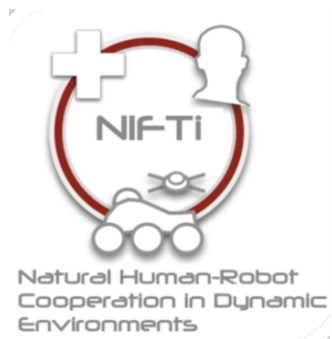


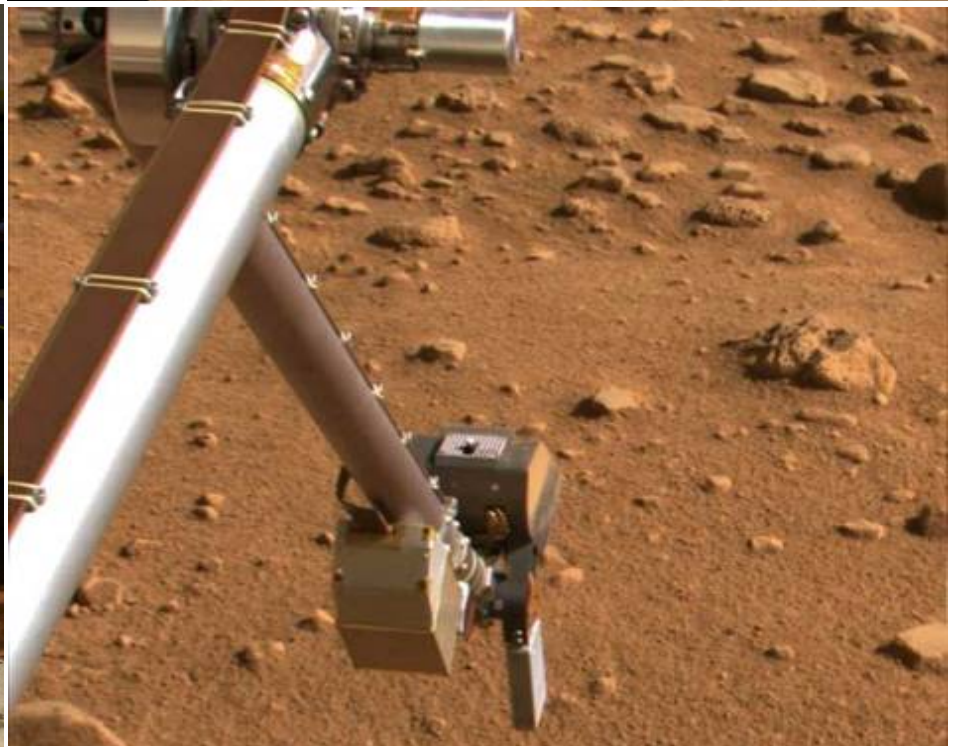
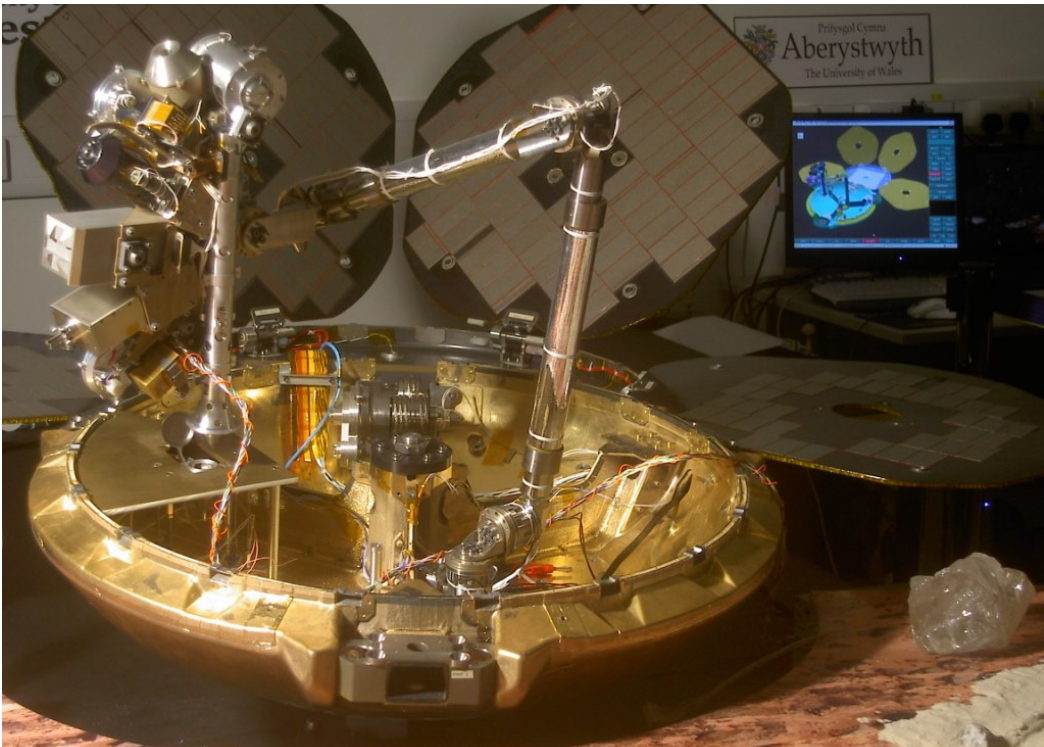
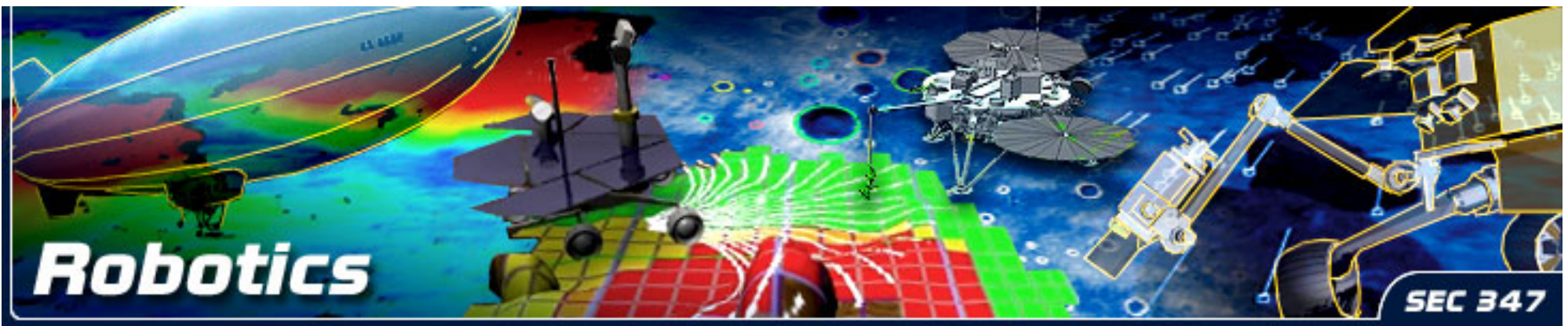
(a)



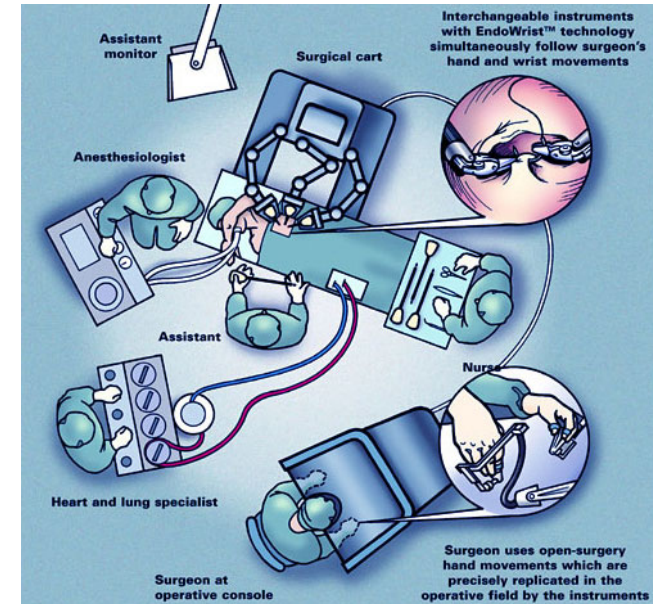
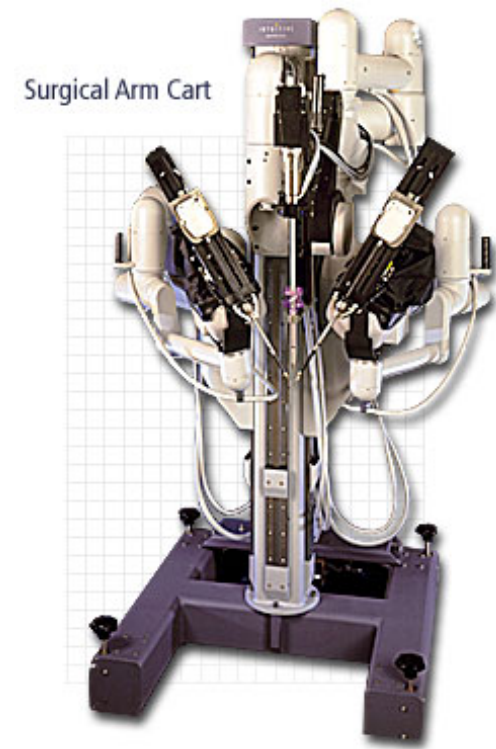
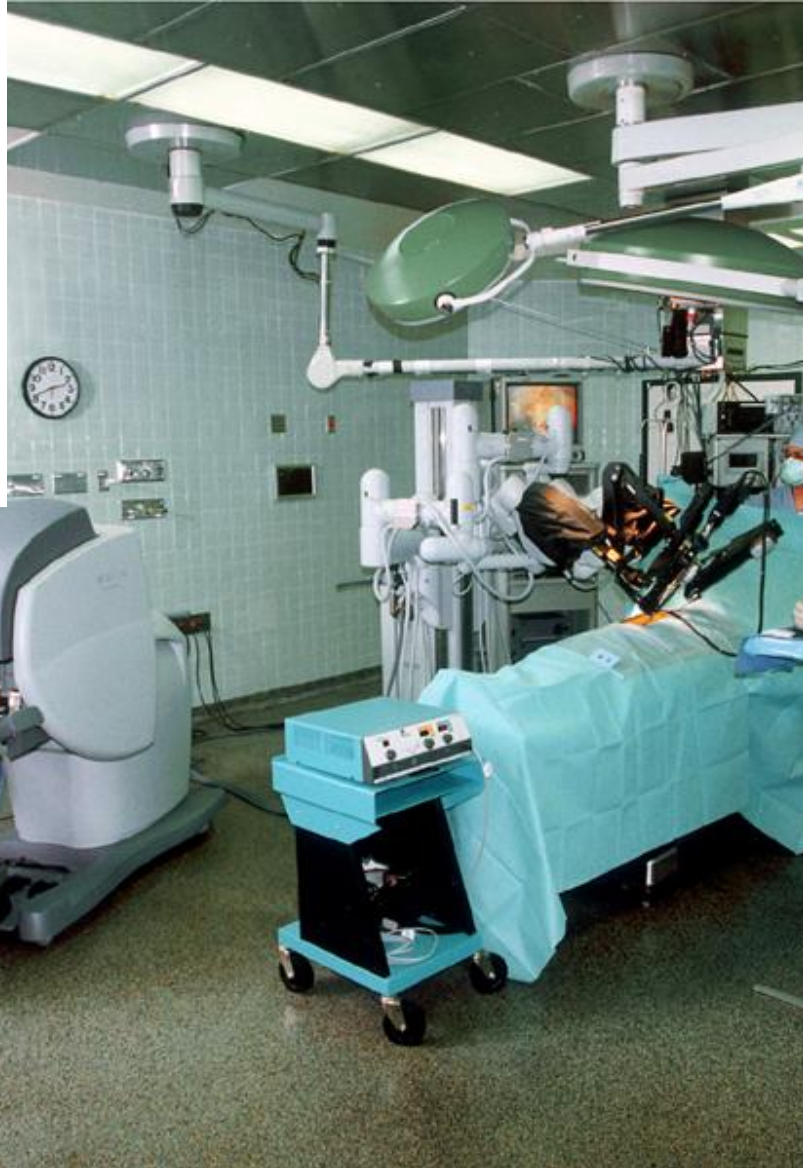
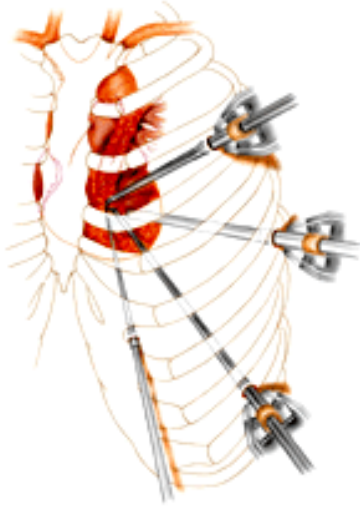


Intelligent Robotics





Precision for robotic surgery



<http://www.cts.usc.edu/rsi-article-robotputsuscatforefront.html>

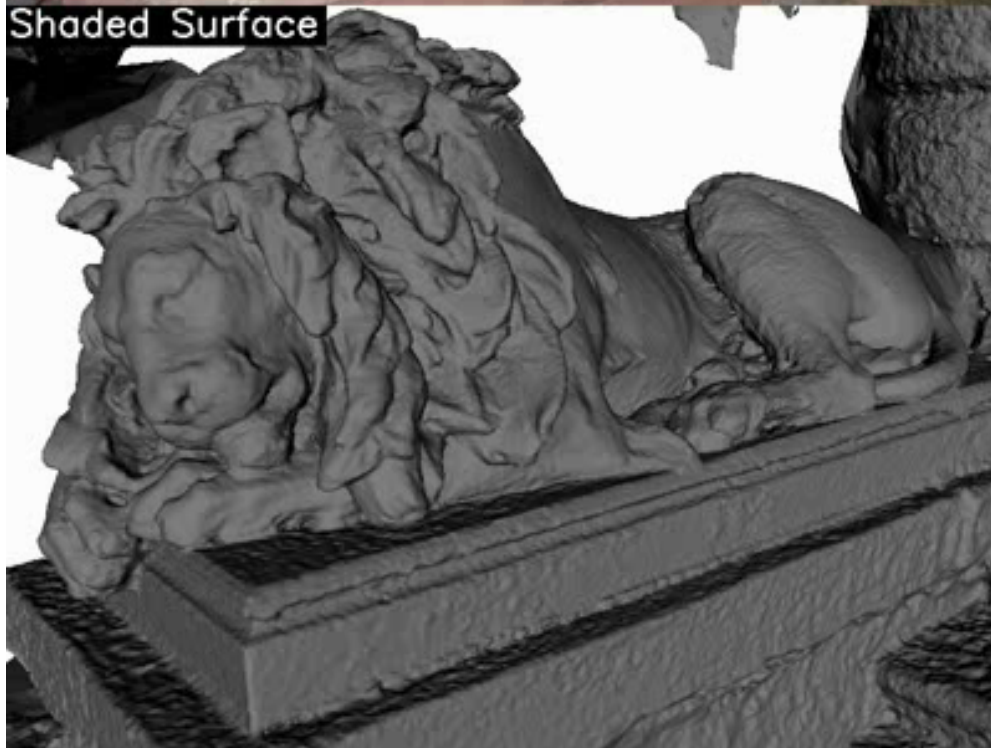
Original Image



3D Points



Shaded Surface



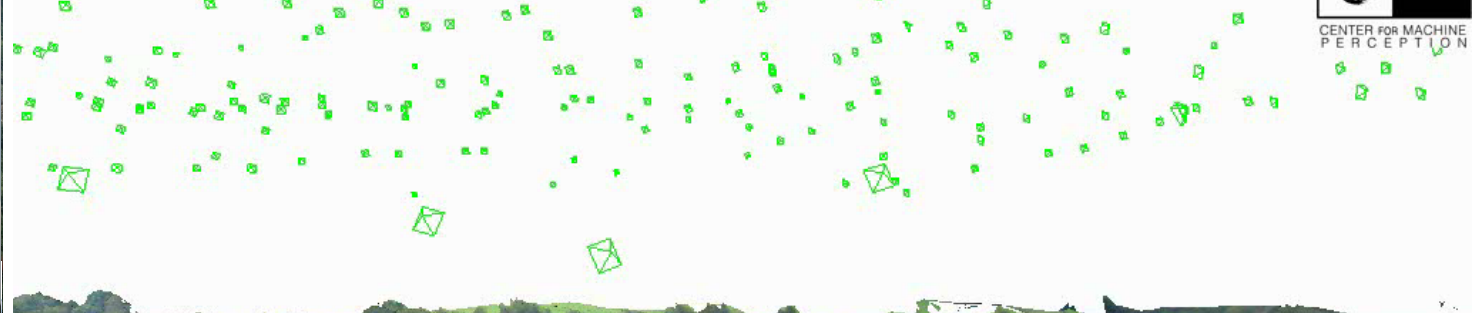
Textured Surface





2009©AirshipClub.com

Automatic 3D Reconstruction of Sternberg data-set.
Sternberg data-set: 324 (3056 x 2296) images.



(c) 12.4.2011, Michal Jancosek, jancom1@cmp.felk.cvut.cz





ASTRIUM

AN EADS COMPANY



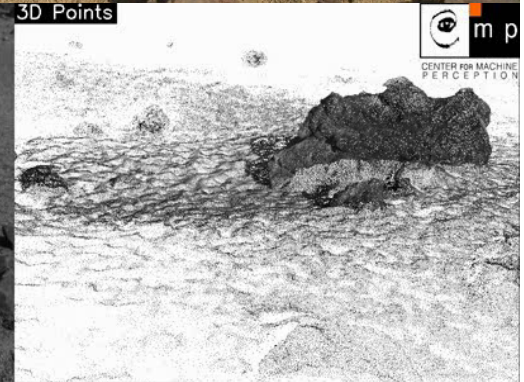
PRoVisG Field Trials Tenerife



Original Image



Shaded Surface



3D Points



Textured Surface

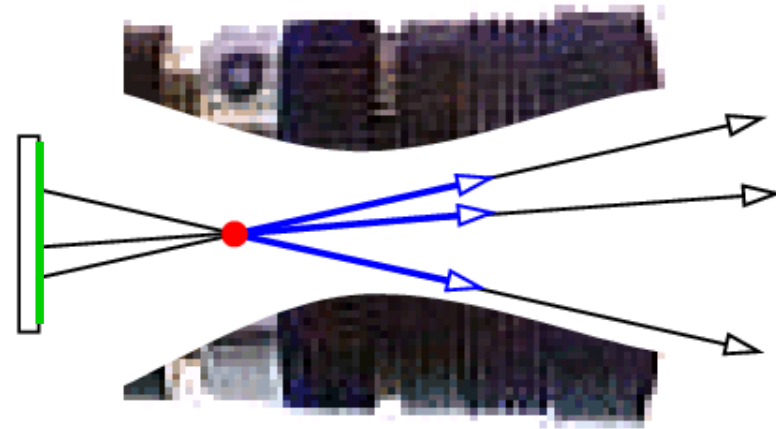


CENTER FOR MACHINE PERCEPTION

Perspective camera model

T. Pajdla. Elements of Geometry for Computer Vision
<http://cmp.felk.cvut.cz/~pajdla/gvg/GVG-2014-Lecture.pdf>

Camera



Digital cameras



Image projection model

1. Light extends along straight **rays**
2. Projection **center**
3. Projection **plane**

Image



pixel

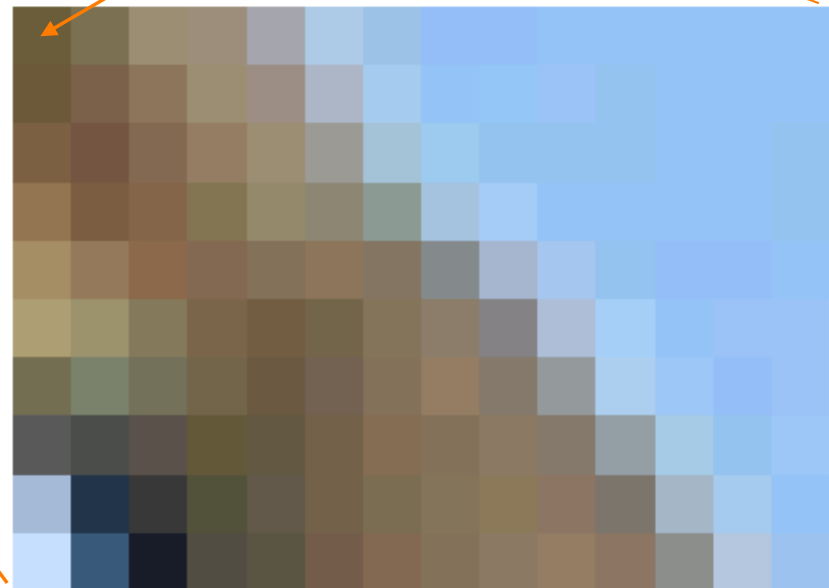
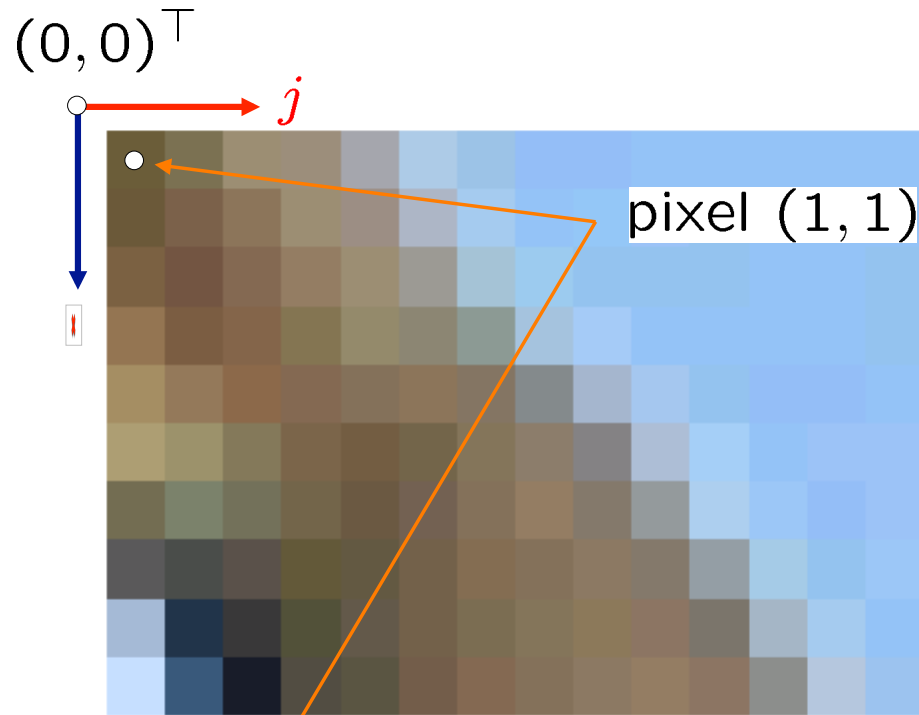


Image is a $m \times n \times 3$ matrix in Matlab



```
>>im = imread('karluv-most.jpg');  
>>imagesc(im(1:10,1:14,:));  
>>axis image
```

Indexing

j R ← RGB

```
>>im(3,4,1)
```

1 ... R

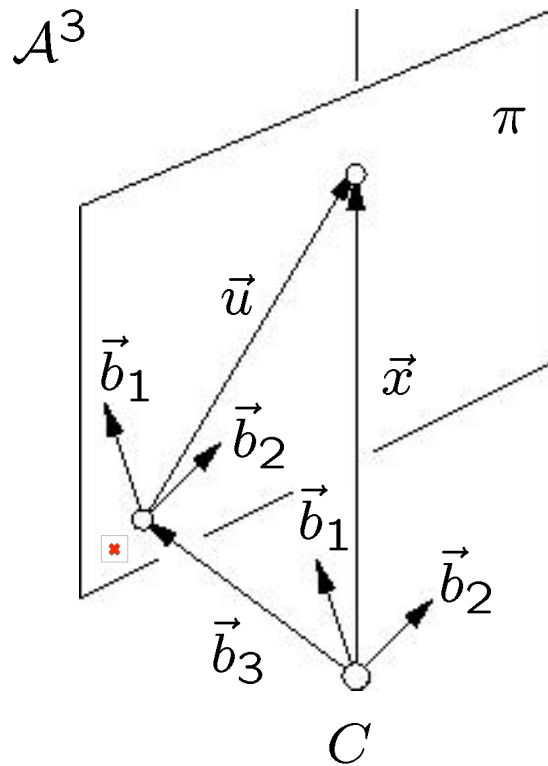
```
ans = 144
```

2 ... G

3 ... B

	1	2	3	4	5	6	7	8	9	10	11	12	13	14
1	107	126	155	154	161	172	155	145	149	151	150	150	150	149
2	107	121	139	156	154	174	166	146	149	152	150	148	149	151
3	125	116	130	144	159	157	167	159	149	150	148	150	151	151
4	148	125	128	132	147	140	140	167	160	146	148	151	150	150
5	164	149	136	128	131	137	130	133	160	164	150	151	149	151
6	168	154	134	122	114	118	134	137	131	173	165	148	152	153
7	112	127	116	112	106	116	130	144	129	145	170	152	147	152
8	89	76	91	98	102	115	128	131	139	130	148	167	151	152
9	164	37	62	85	98	115	127	131	139	139	125	165	164	149
10	194	62	27	84	95	112	128	131	137	150	139	142	176	158

Direction vector of projection ray



Miracle: The coordinates of the direction vector of a projection ray can be constructed by adding “1” to image coordinates:

We measure in image

$$\vec{u} = u \vec{b}_1 + v \vec{b}_2 \sim \mathbf{u}_{(\vec{b}_1, \vec{b}_2)} = \begin{pmatrix} u \\ v \end{pmatrix}$$

Coordinate system with origin C

$$\beta = (\vec{b}_1, \vec{b}_2, \vec{b}_3)$$

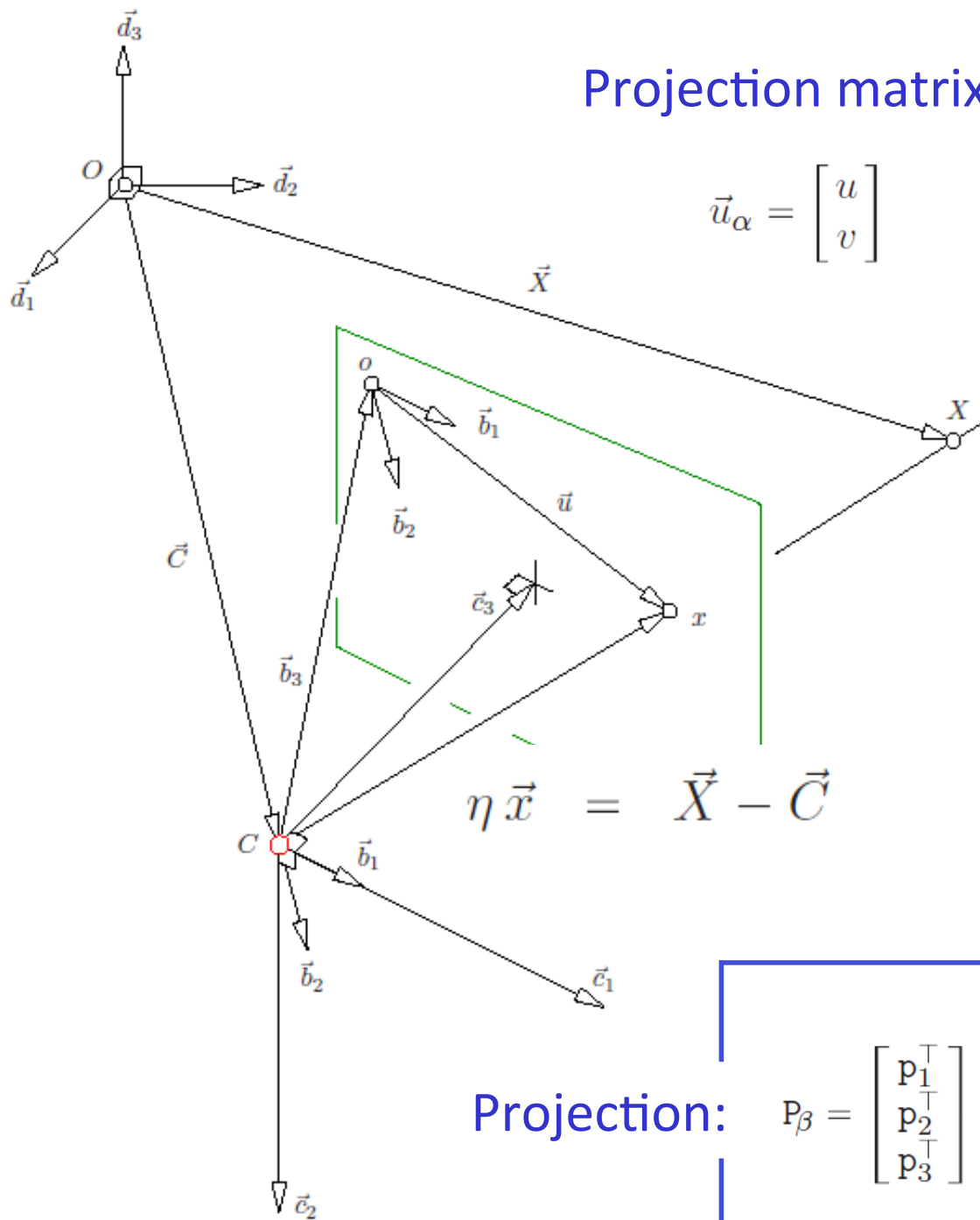
$$S = (C, \beta)$$

$$\vec{b}_3 = \varphi(C, o)$$

Triangle equality

$$\vec{x} = \vec{u} + \vec{b}_3$$

$$\vec{x} = u \vec{b}_1 + v \vec{b}_2 + 1 \vec{b}_3 \sim \mathbf{x}_\beta = \begin{pmatrix} u \\ v \\ 1 \end{pmatrix}$$



Projection matrix:

$$\vec{u}_\alpha = \begin{bmatrix} u \\ v \end{bmatrix}$$

$$\eta \vec{x}_\beta = \vec{X}_\beta - \vec{C}_\beta$$

$$\eta \begin{bmatrix} \vec{u}_\alpha \\ 1 \end{bmatrix} = \vec{X}_\beta - \vec{C}_\beta$$

$$\vec{y}_\beta = \mathbf{A} \vec{y}_\delta$$

$$\eta \begin{bmatrix} \vec{u}_\alpha \\ 1 \end{bmatrix} = \mathbf{A} (\vec{X}_\delta - \vec{C}_\delta)$$

$$\eta \begin{bmatrix} \vec{u}_\alpha \\ 1 \end{bmatrix} = \mathbf{A} \left[\mathbf{I} \mid -\vec{C}_\delta \right] \begin{bmatrix} \vec{X}_\delta \\ 1 \end{bmatrix}$$

$$\eta \begin{bmatrix} \vec{u}_\alpha \\ 1 \end{bmatrix} = \mathbf{P}_\beta \begin{bmatrix} \vec{X}_\delta \\ 1 \end{bmatrix}$$

$$\eta \vec{x}_\beta = \mathbf{P}_\beta \begin{bmatrix} \vec{X}_\delta \\ 1 \end{bmatrix}$$

Projection:

$$\mathbf{P}_\beta = \begin{bmatrix} p_1^\top \\ p_2^\top \\ p_3^\top \end{bmatrix}$$

$$\text{and } \mathbf{x} = \begin{bmatrix} \vec{X}_\delta \\ 1 \end{bmatrix}$$

$$\vec{u}_\alpha = \begin{bmatrix} \frac{p_1^\top \mathbf{x}}{p_3^\top \mathbf{x}} \\ \frac{p_2^\top \mathbf{x}}{p_3^\top \mathbf{x}} \end{bmatrix}$$

Figure 6.2: Coordinate systems of perspective camera.

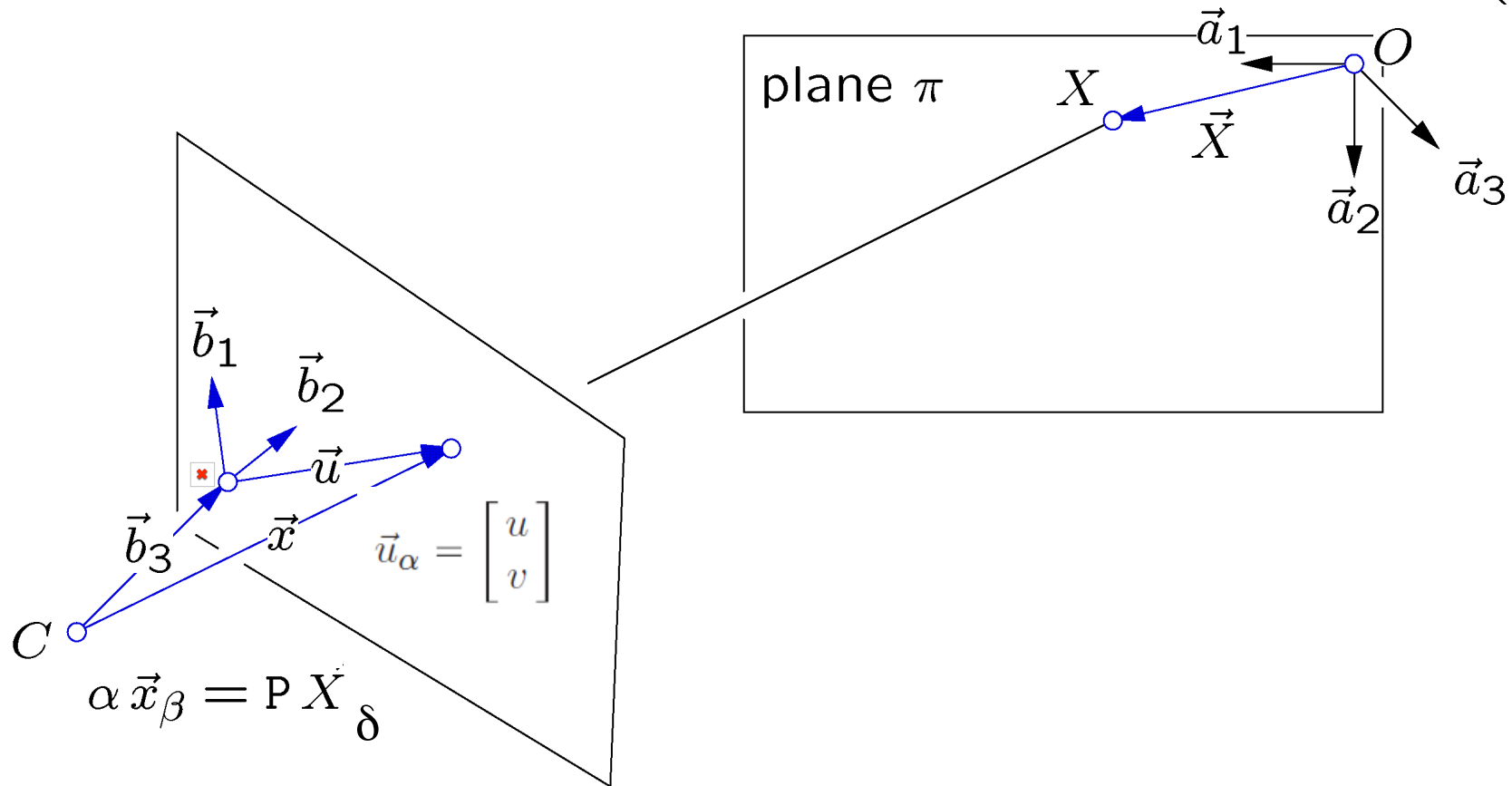
Homography

Camera observes a plane

World coordinate system

$$\delta = (\vec{a}_1, \vec{a}_2, \vec{a}_3)$$

$$W = (O, \delta)$$



$$\beta = (\vec{b}_1, \vec{b}_2, \vec{b}_3)$$

$$S = (C, \beta)$$

Camera coordinate system

$$\zeta \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \zeta \vec{x}_\beta = P \begin{bmatrix} \vec{X}_\delta \\ 1 \end{bmatrix} = \begin{bmatrix} p_1 & p_2 & p_3 & p_4 \end{bmatrix} \begin{bmatrix} x \\ y \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} p_1 & p_2 & p_4 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = H \vec{y}_\tau$$

COMPUTING THE HOMOGRAPHY
from 4 correspondences

Computing the homography

$\exists H \in \mathbb{R}^{3 \times 3}$, $\text{rank } H = 3$, so that $\forall (u, v) \overset{\text{corr}}{\leftrightarrow} (u', v') \exists \alpha \in \mathbb{R}$:

$$\alpha \begin{pmatrix} u' \\ v' \\ 1 \end{pmatrix} = H \begin{pmatrix} u \\ v \\ 1 \end{pmatrix}$$

Introduce symbols for rows of homography H

$$H = \begin{pmatrix} \mathbf{h}_1^\top \\ \mathbf{h}_2^\top \\ \mathbf{h}_3^\top \end{pmatrix}$$

and rewrite the above matrix equation as

$$\begin{aligned} \alpha u' &= \mathbf{h}_1^\top \mathbf{x} \\ \alpha v' &= \mathbf{h}_2^\top \mathbf{x} \\ \alpha &= \mathbf{h}_3^\top \mathbf{x} \end{aligned}$$

Computing the homography

Eliminate α from the first two equations using the third one

$$\begin{aligned}(\mathbf{h}_3^\top \mathbf{x}) u' &= \mathbf{h}_1^\top \mathbf{x} \\ (\mathbf{h}_3^\top \mathbf{x}) v' &= \mathbf{h}_2^\top \mathbf{x}\end{aligned}$$

move all to the left hand side and reshape it using $\mathbf{x}^\top \mathbf{y} = \mathbf{y}^\top \mathbf{x}$

$$\begin{aligned}\mathbf{x}^\top \mathbf{h}_1 - (u' \mathbf{x}^\top) \mathbf{h}_3 &= 0 \\ \mathbf{x}^\top \mathbf{h}_2 - (v' \mathbf{x}^\top) \mathbf{h}_3 &= 0\end{aligned}$$

Introduce notation

$$\mathbf{h} = \left(\mathbf{h}_1^\top \quad \mathbf{h}_2^\top \quad \mathbf{h}_3^\top \right)^\top$$

and express the above two equations in a matrix form

$$\begin{pmatrix} u & v & 1 & 0 & 0 & 0 & -u'u & -u'v & -u' \\ 0 & 0 & 0 & u & v & 1 & -v'u & -v'v & -v' \end{pmatrix} \mathbf{h} = 0$$

Computing the homography

Notice that A can be written in the form

$$A = \begin{pmatrix} u_1 & v_1 & 1 & 0 & 0 & 0 & -u'_1 u_1 & -u'_1 v_1 & -u'_1 \\ u_2 & v_2 & 1 & 0 & 0 & 0 & -u'_2 u_2 & -u'_2 v_2 & -u'_2 \\ & & & & \vdots & & & & \\ 0 & 0 & 0 & u_1 & v_1 & 1 & -v'_1 u_1 & -v'_1 v_1 & -v'_1 \\ 0 & 0 & 0 & u_2 & v_2 & 1 & -v'_2 u_2 & -v'_2 v_2 & -v'_2 \\ & & & & \vdots & & & & \end{pmatrix}$$

which can be rewritten more concisely as

$$A = \begin{pmatrix} \mathbf{x}_1^\top & \mathbf{0} & -u'_1 \mathbf{x}_1^\top \\ \mathbf{x}_2^\top & \mathbf{0} & -u'_2 \mathbf{x}_2^\top \\ & \vdots & \\ \mathbf{0} & \mathbf{x}_1^\top & -v'_1 \mathbf{x}_1^\top \\ \mathbf{0} & \mathbf{x}_2^\top & -v'_2 \mathbf{x}_2^\top \\ & \vdots & \end{pmatrix}$$

Computing the homography from 4 points on 2 lines in Matlab

```
% 4 points
```

```
>>x = [0 0 1;1 0 1;0 1 1;1 1 1]';
```

```
>>y = [1 1 1;1 0 1;0 1 1;0 0 1]';
```

```
% the 2-line algorithm
```

```
>>A = [[x' zeros(size(x')) [-y(1,:)'*ones(1,3)].*(x')]
        [zeros(size(x')) x' [-y(2,:)'*ones(1,3)].*(x')]];
```

```
>>H = reshape(null(A),3,3)';
```

```
% verification
```

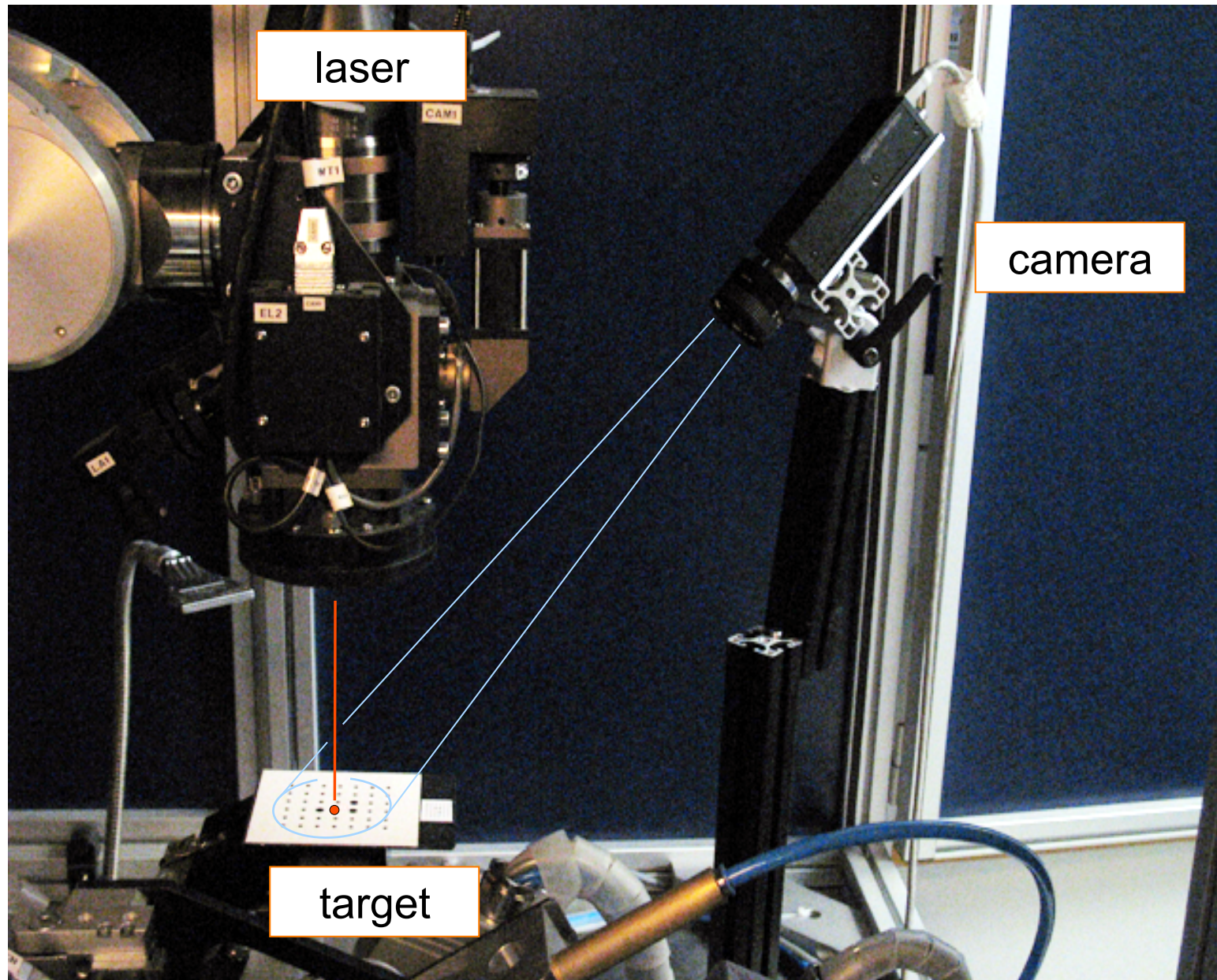
```
>> e = y - (H*x)./[ [1;1;1]*(H(3,:)*x)]
```

```
e =
```

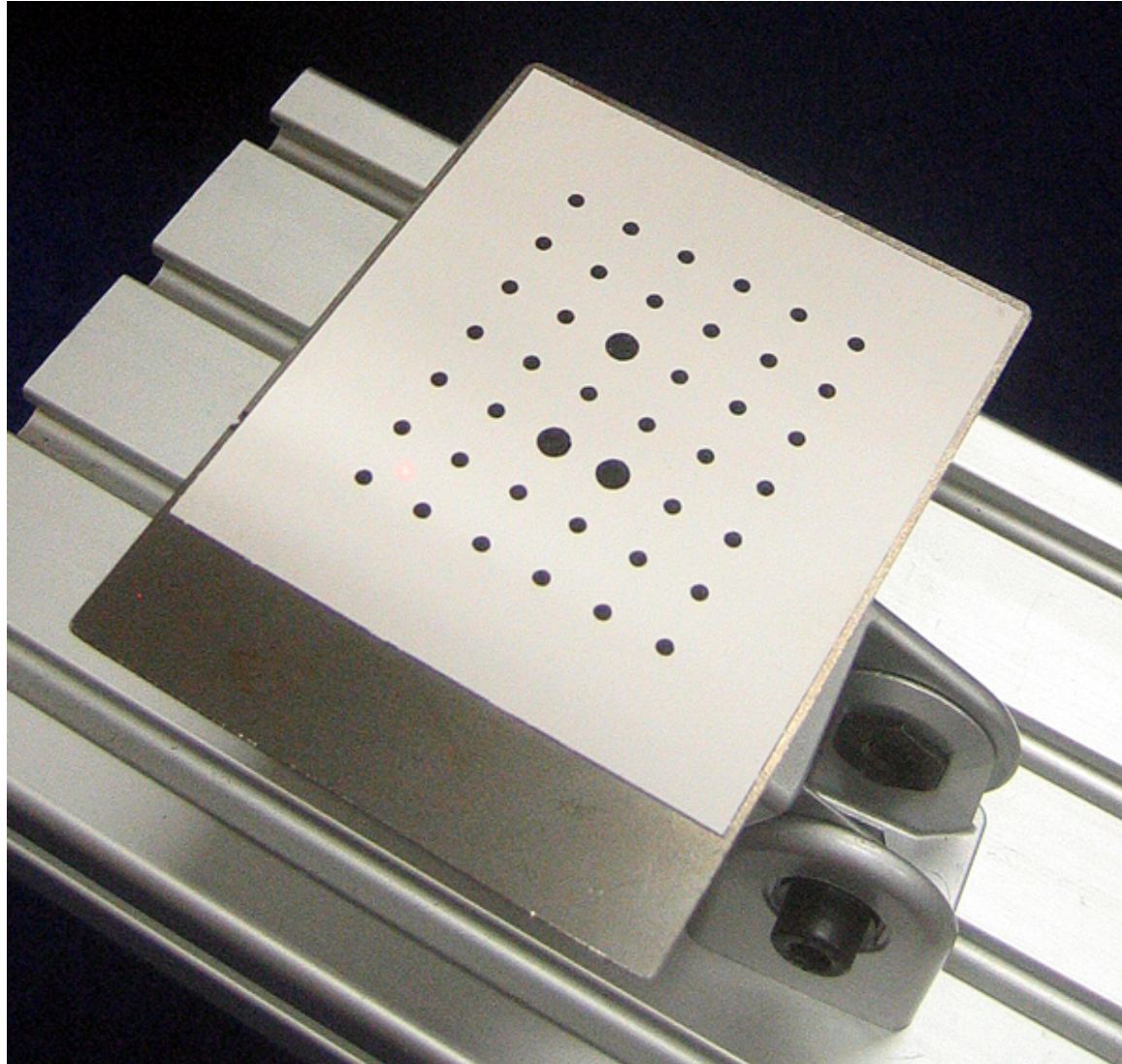
```
1.0e-015 *
```

```
0    0.0481   -0.2220   -0.4441
0    0.2220    0.0961   -0.2220
0         0         0         0
```

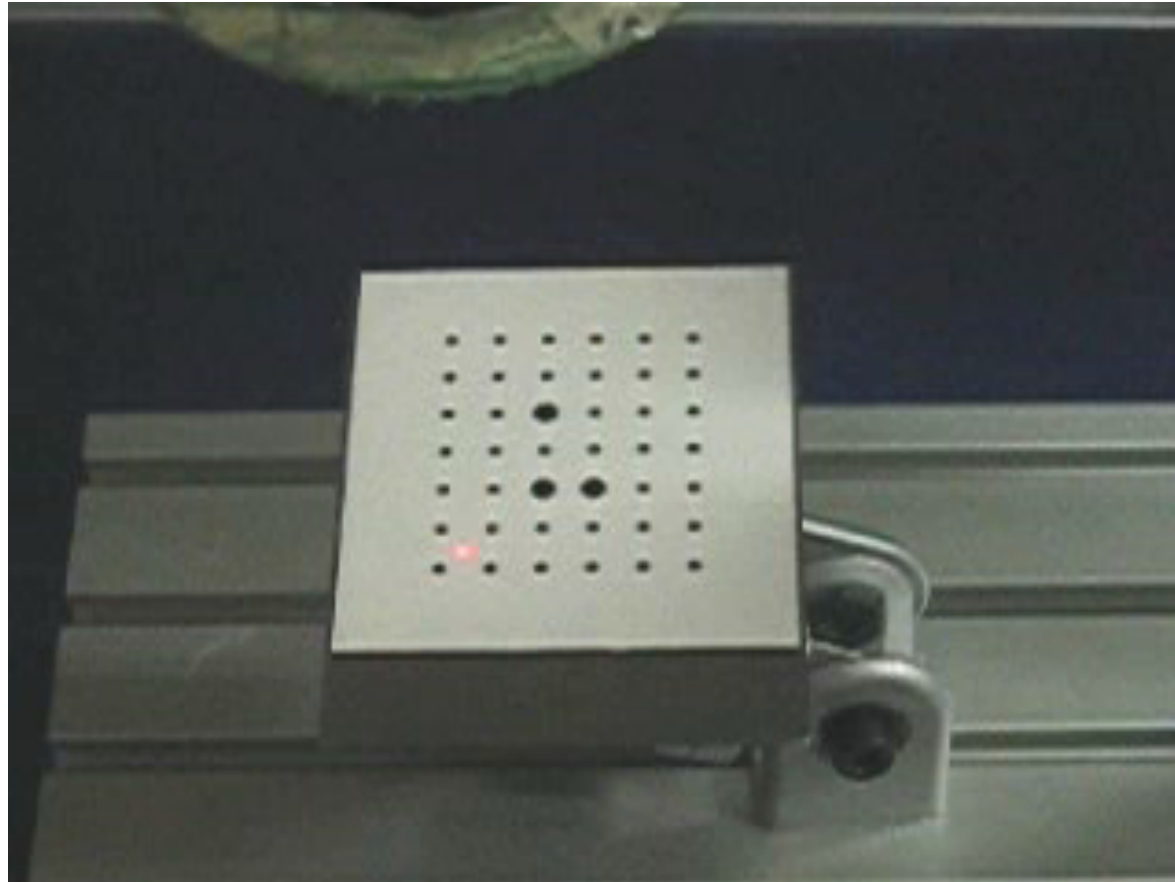
Setup



Calibration target



Laser spot



Laser spot in the camera view

1 pxl \sim 0.1 mm, i.e. resolution 0.1 pxl \sim 0.01 mm

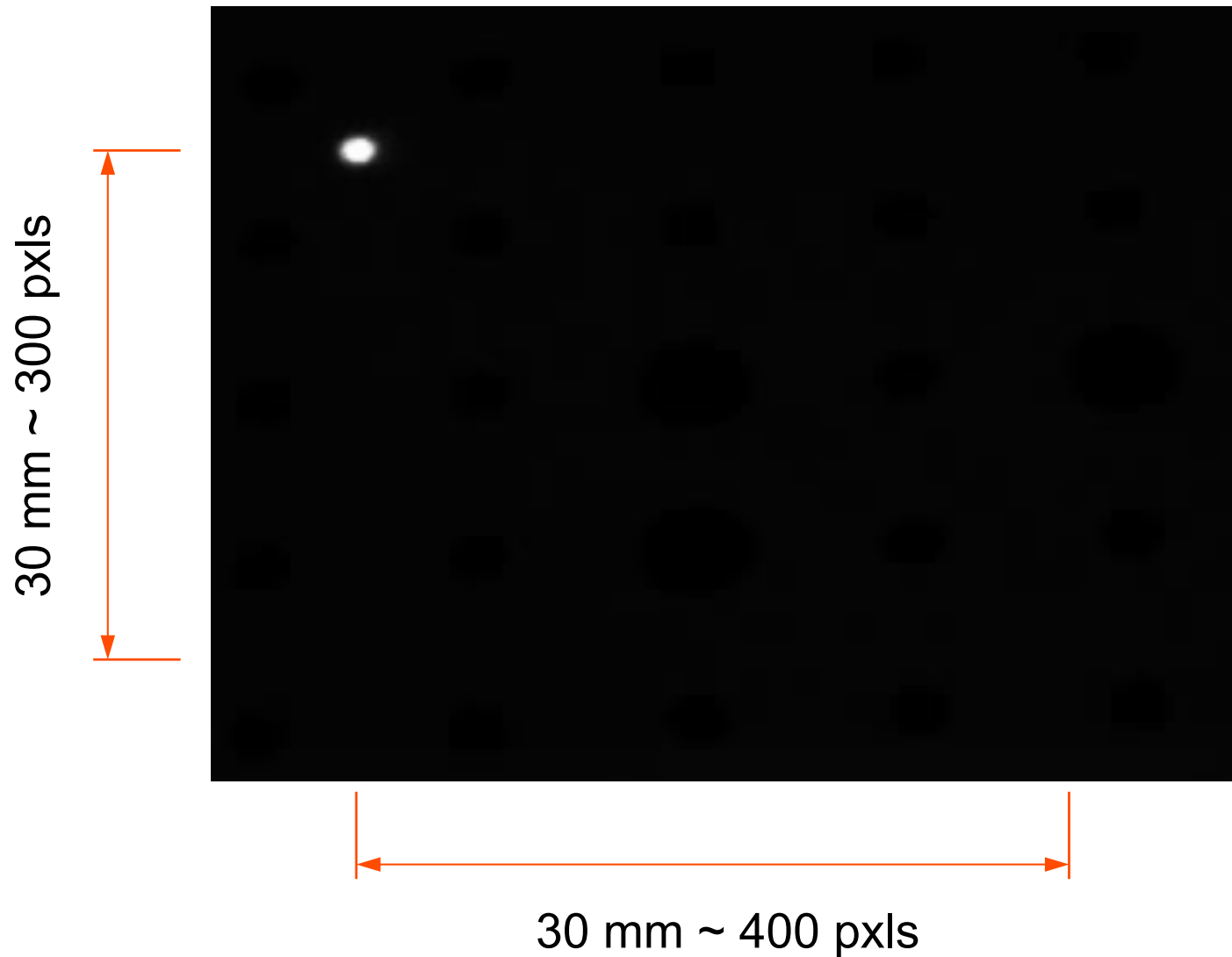
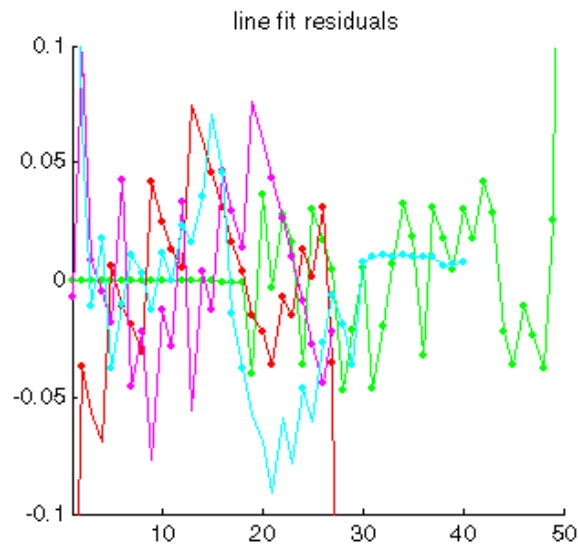
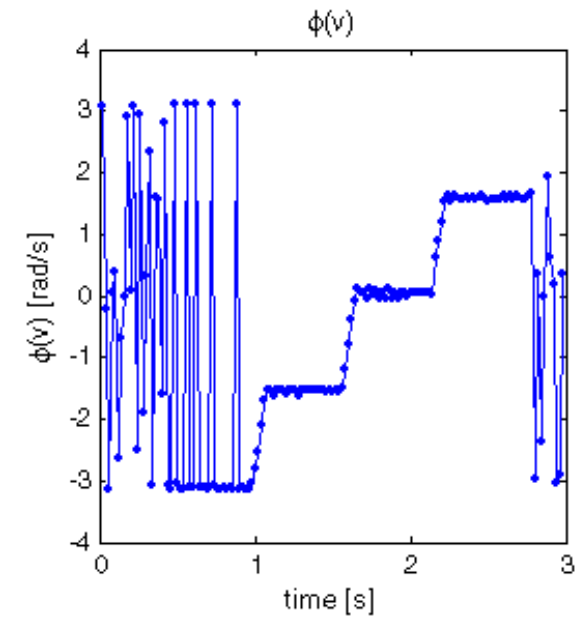
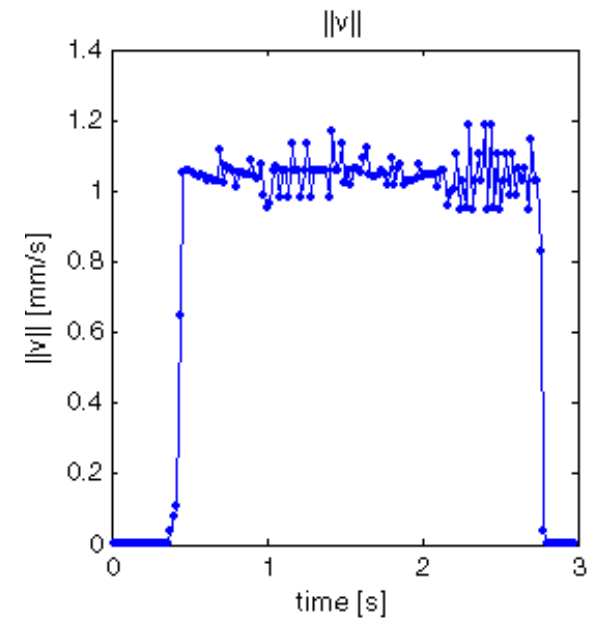
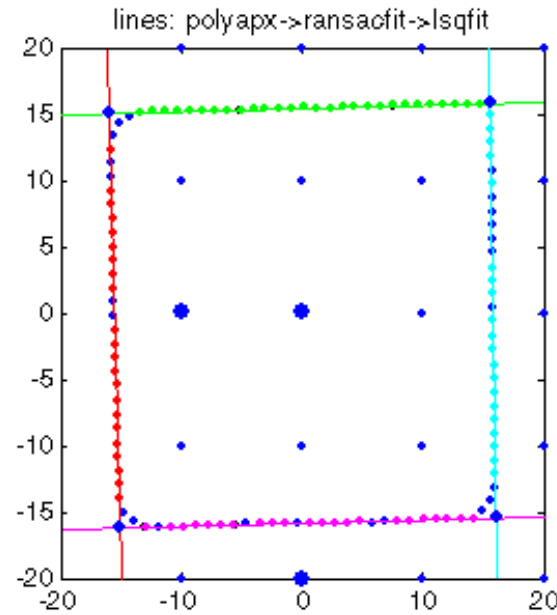
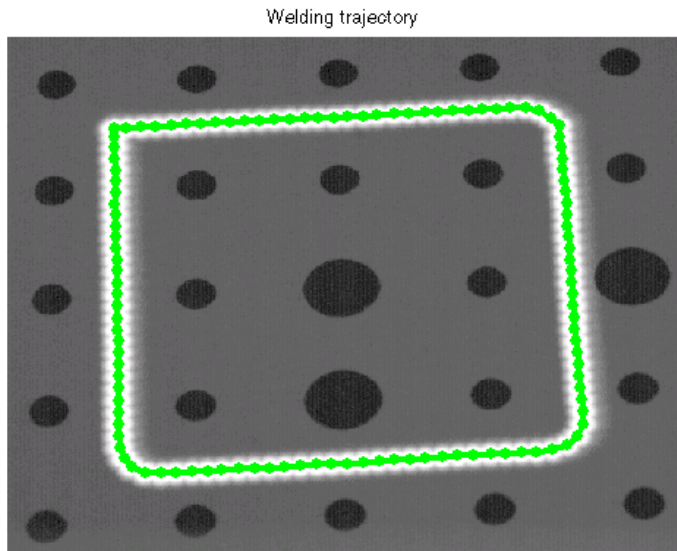


abb-sq-02:

VAR speeddata speed:= [15,15,0,0];

CONST zonedata zn := [FALSE, 3.0, 3.0, 3.0, 0.30, 3.0, 0.30];



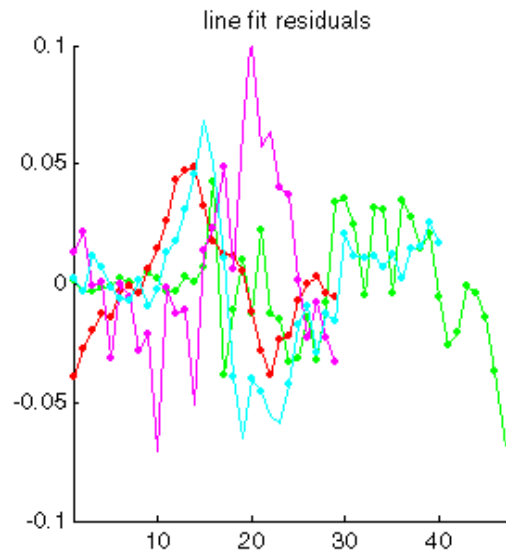
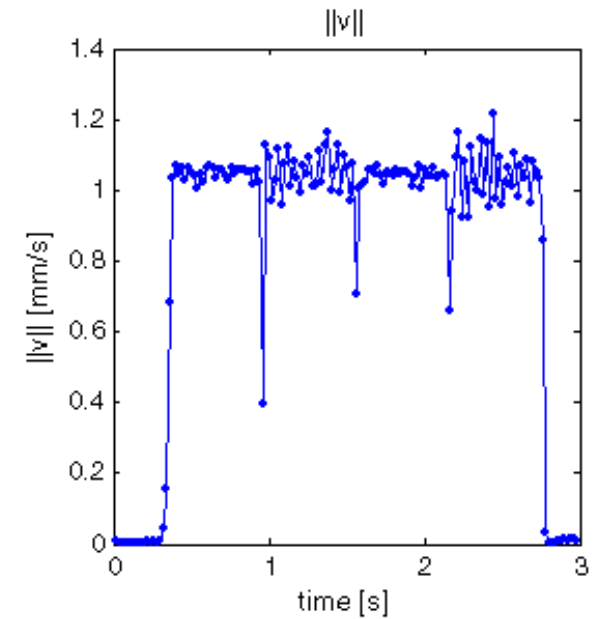
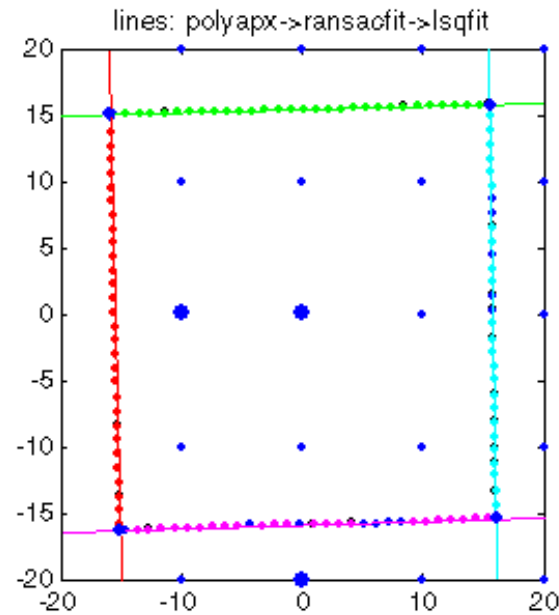
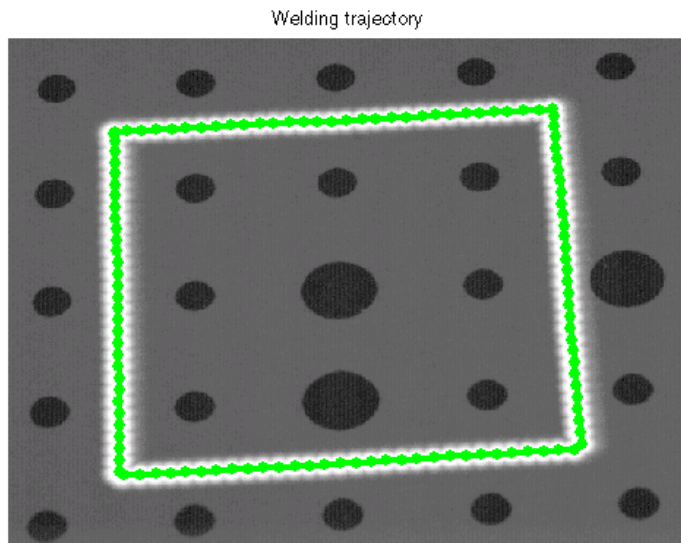
4 - corners

	segment 1	2	3	4	5	6
D [mm]	31.31	31.16	31.26	31.49	44.22	44.32
dD [mm]	0.01	-0.15	-0.05	0.18	-0.05	0.05
dphi [deg]	-0.12	-0.49	0.38	0.23	-0.07	NaN

abb-sq-03:

VAR speeddata speed:= [15,15,0,0];

CONST zonedata zn := [FALSE, 0.1, 0.1, 0.1, 0.01, 0.1, 0.01];



4 - corners

	segment 1	2	3	4	5	6
D [mm]	31.36	31.19	31.20	31.48	44.18	44.36
dD [mm]	0.06	-0.12	-0.11	0.17	-0.09	0.09
dphi [deg]	0.12	-0.65	0.35	0.19	-0.10	NaN

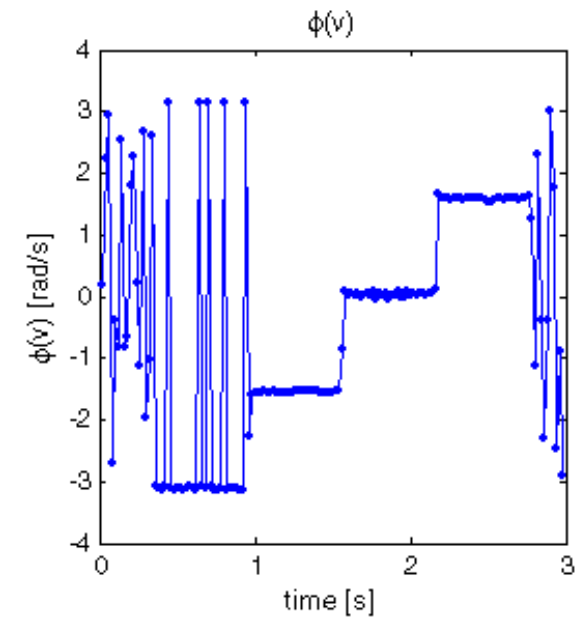
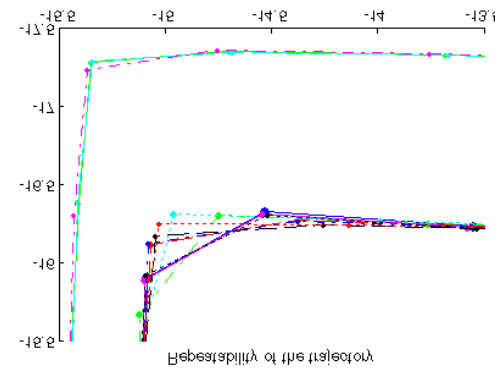
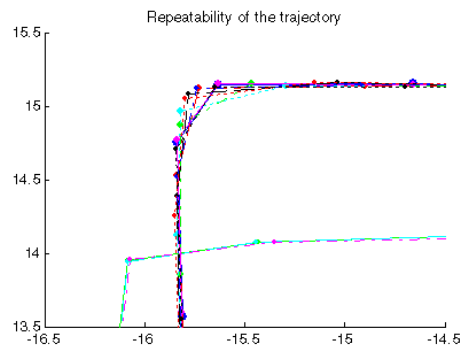
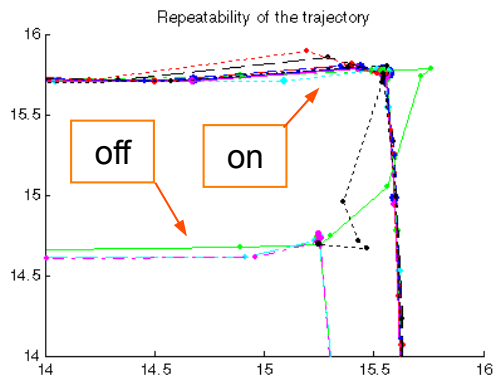
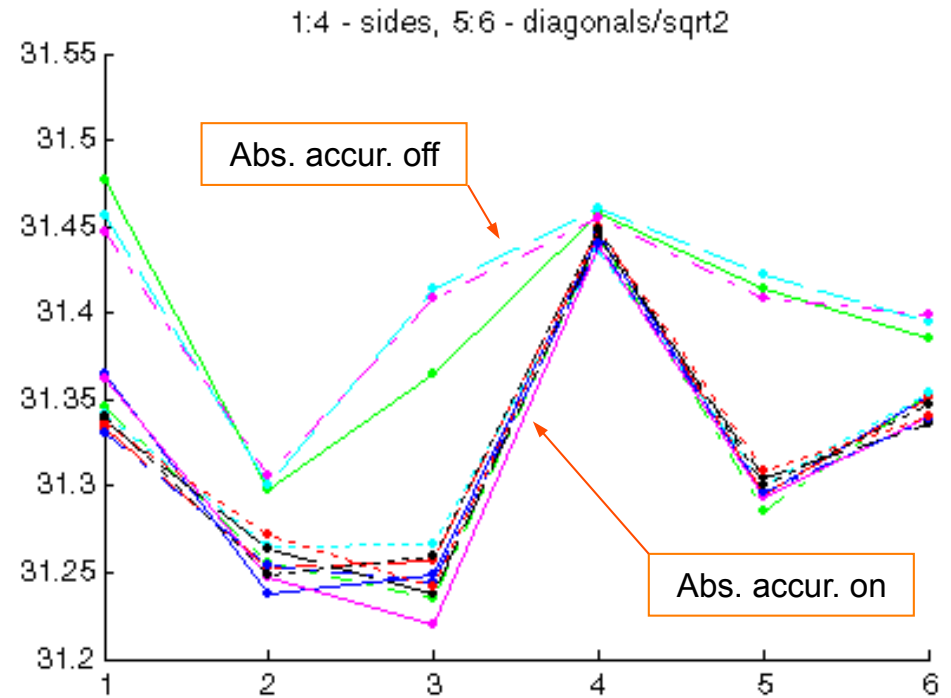
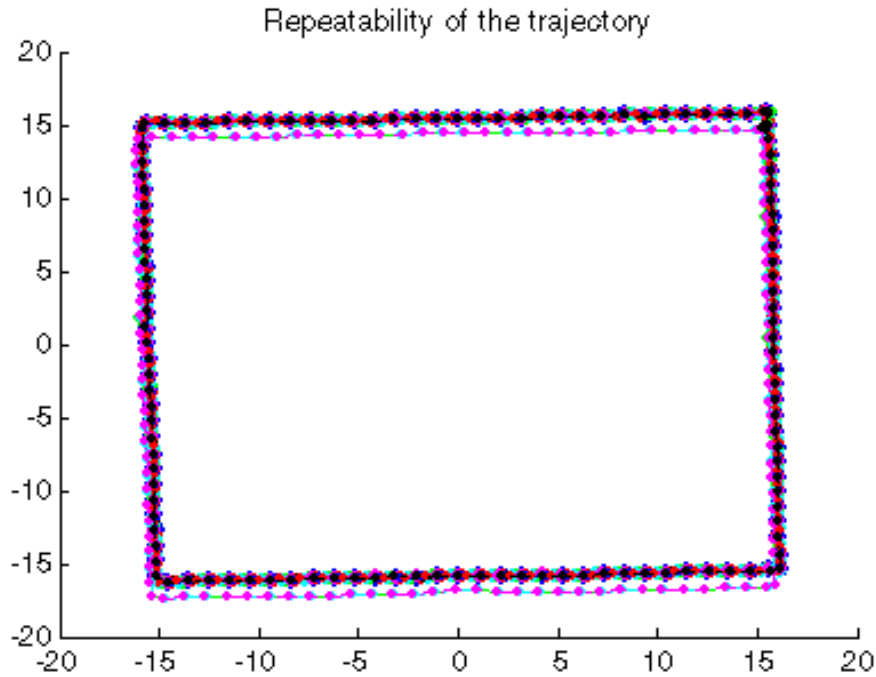


abb-sq-03:

VAR speeddata speed:= [15,15,0,0];

CONST zonedata zn := [FALSE, 0.1, 0.1, 0.1, 0.01, 0.1, 0.01];



ptak.felk.cvut.cz/sfmservice/



Logged in as **pajdla**

[Log out](#)

Admin Menu

- Datasets
- Jobs
- Users
- Browse
- Disk Usage
- Sanity Check

User Menu

- Datasets
- Jobs
- Browse
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- Documentation
 - SfM Tutorial
 - Browser Interface
 - CLI Interface
- Examples

CMP SfM Web Service

[Like](#) 11 [Send](#)

What?

CMP SfM Web Service provides a remote access to the 3D reconstruction systems developed in Center for Machine Perception, FEE, CTU Prague. Our service is available for research purposes only and access is granted on email request to Tomas Pajdla <pajdla@cmp.felk.cvut.cz>. Any commercial use of the service and/or the obtained results is prohibited.

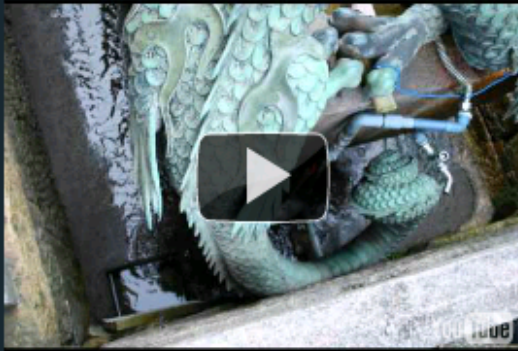
Why?

We provide the access to the service to our partners and to people in the Computer Vision community to make it easier to use our codes. There is no need to install any code on a client's computer and all the computations are performed on our dedicated computing cluster. Further, it makes it easier to compare the results of different methods to ours based on the same data.

News

- 31/05/11 - New web interface (version Prague).
- 17/03/11 - You can now use ► icon in 'Datasets table' to run SfM (No XML file needed).

Input Images



Output Model



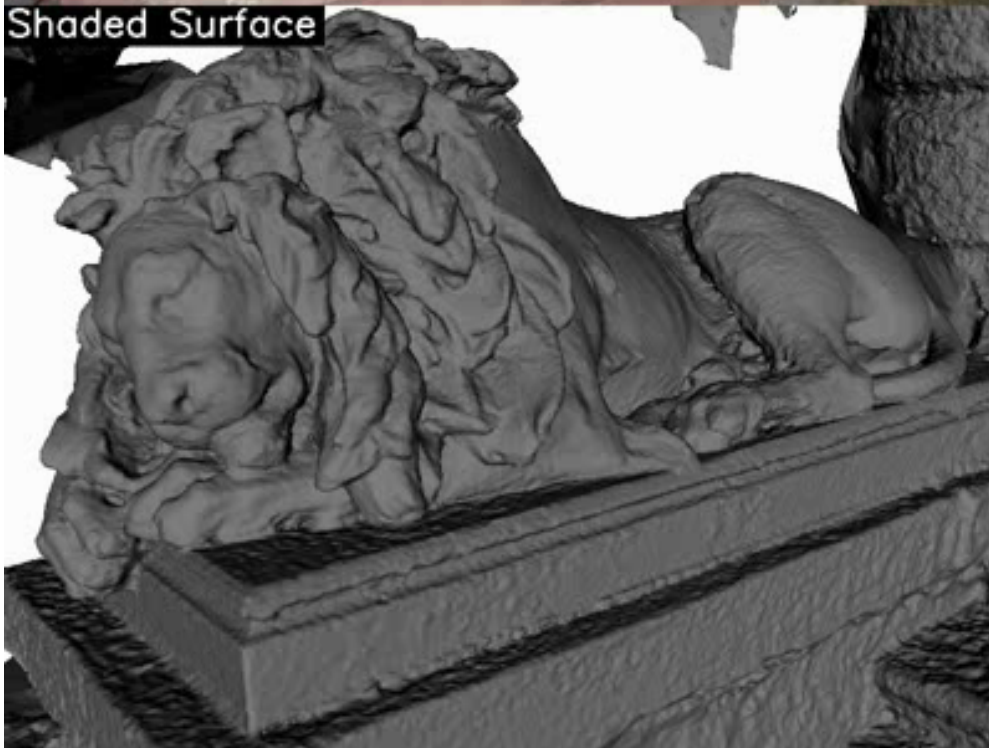
Original Image



3D Points



Shaded Surface



Textured Surface



IMAGE No. 1

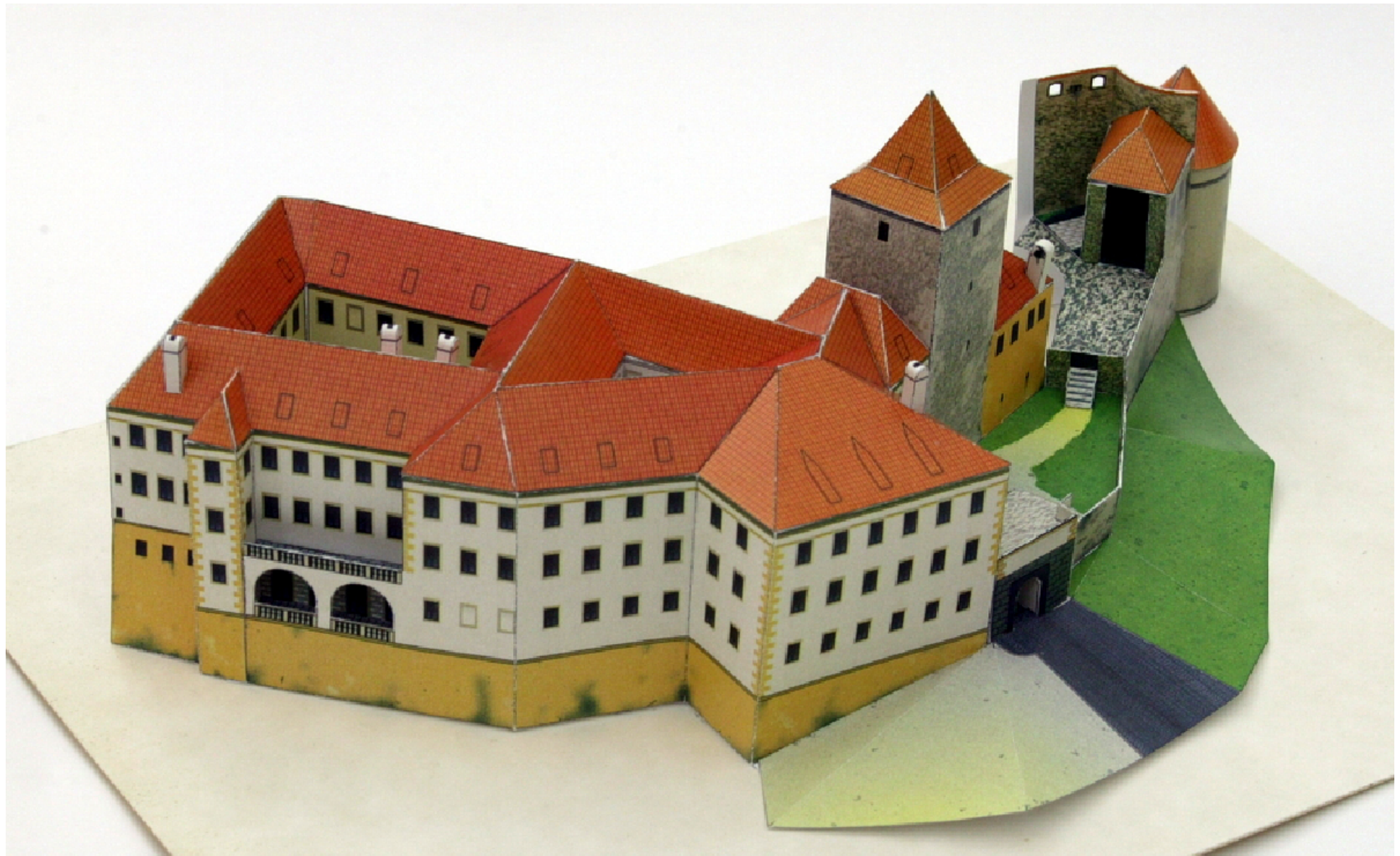
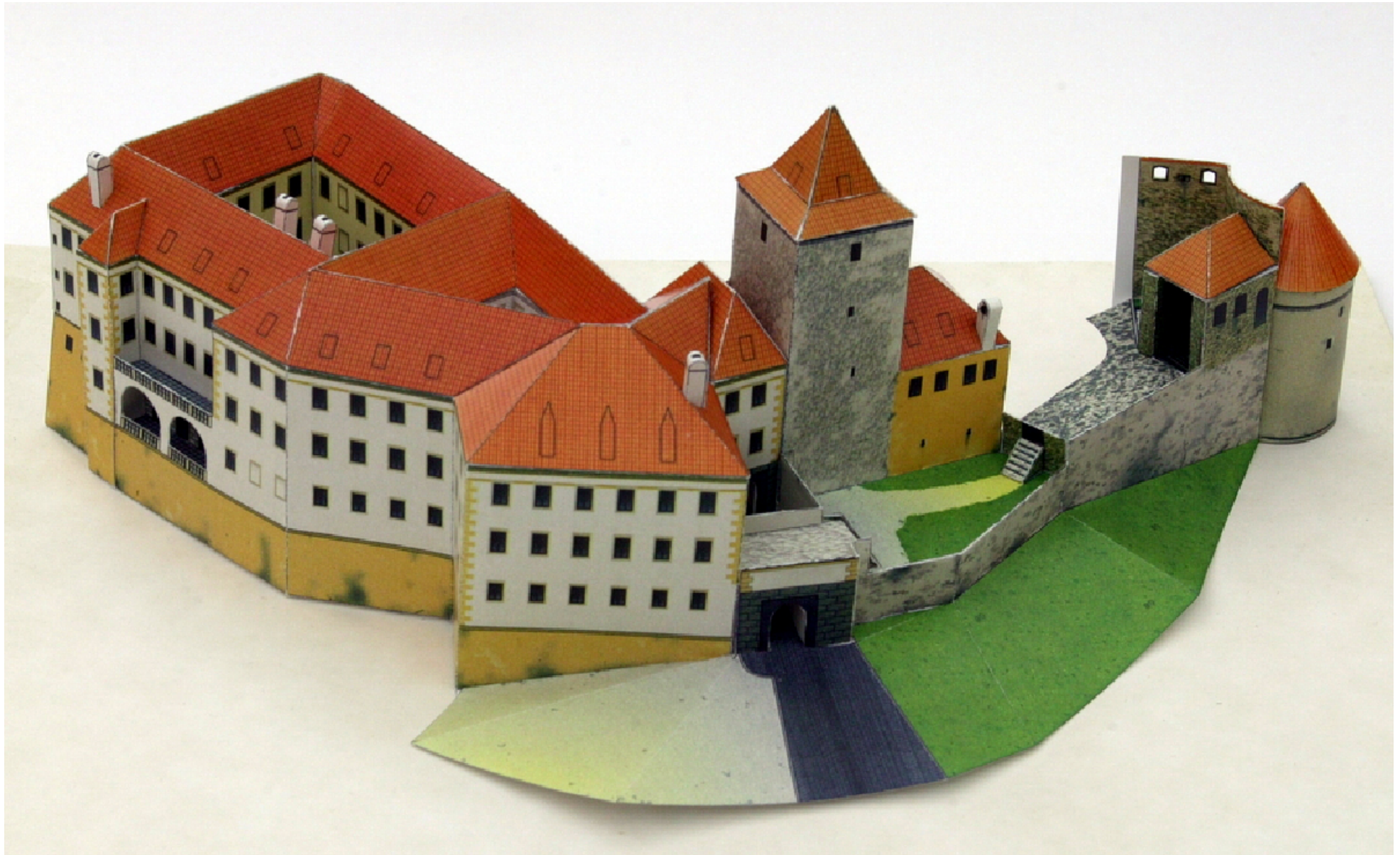
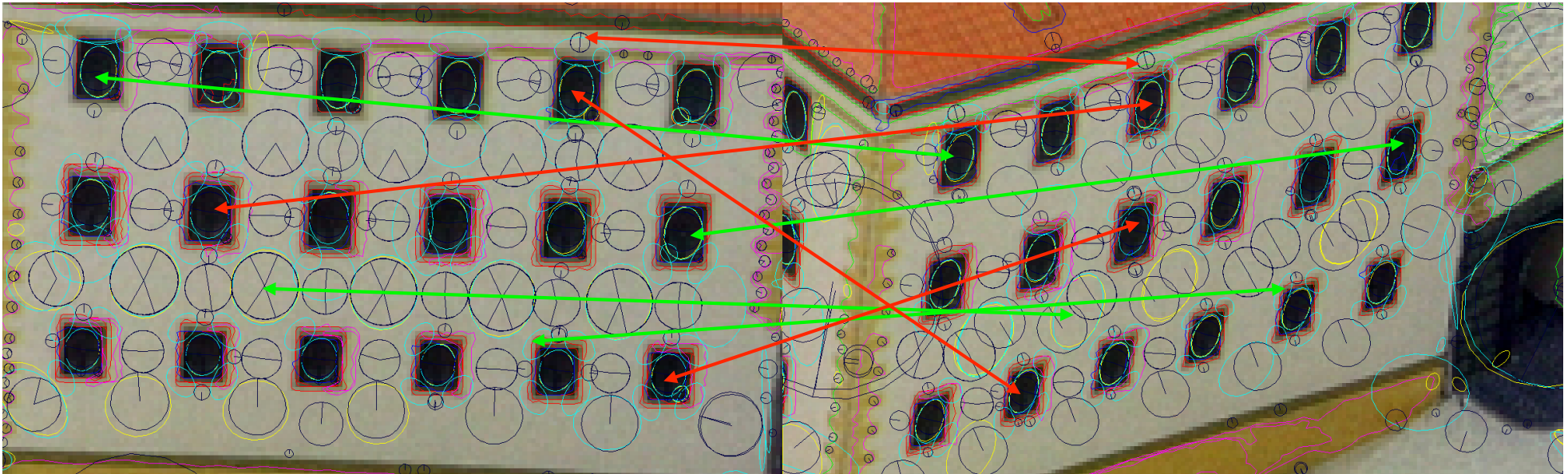


IMAGE No. 2

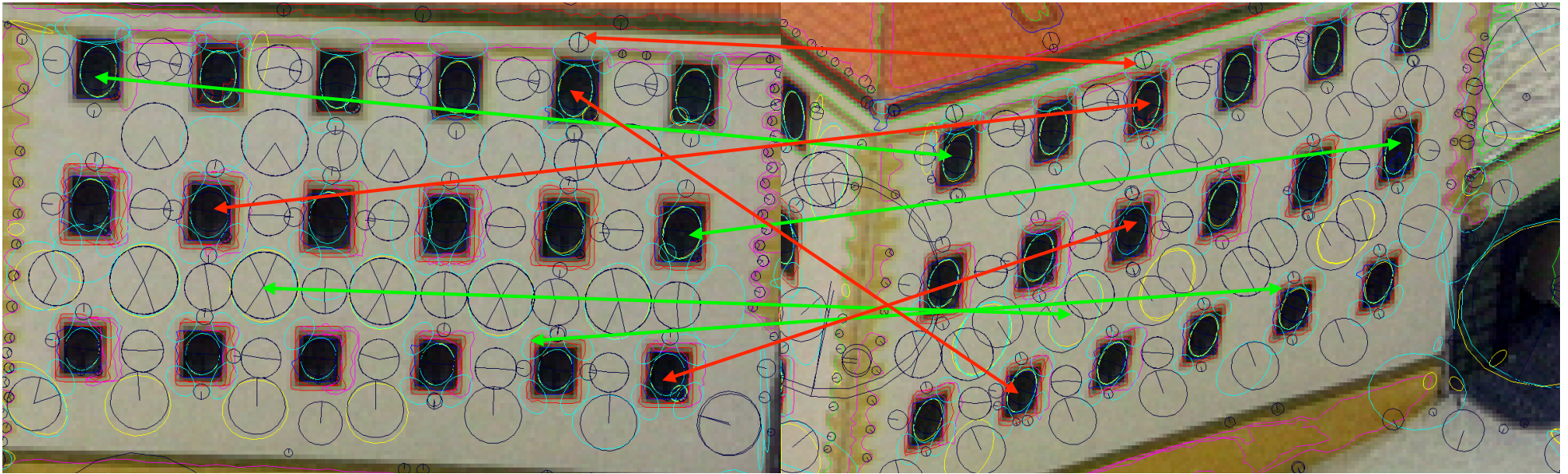


SIMILAR FEATURES FORM MATCHES

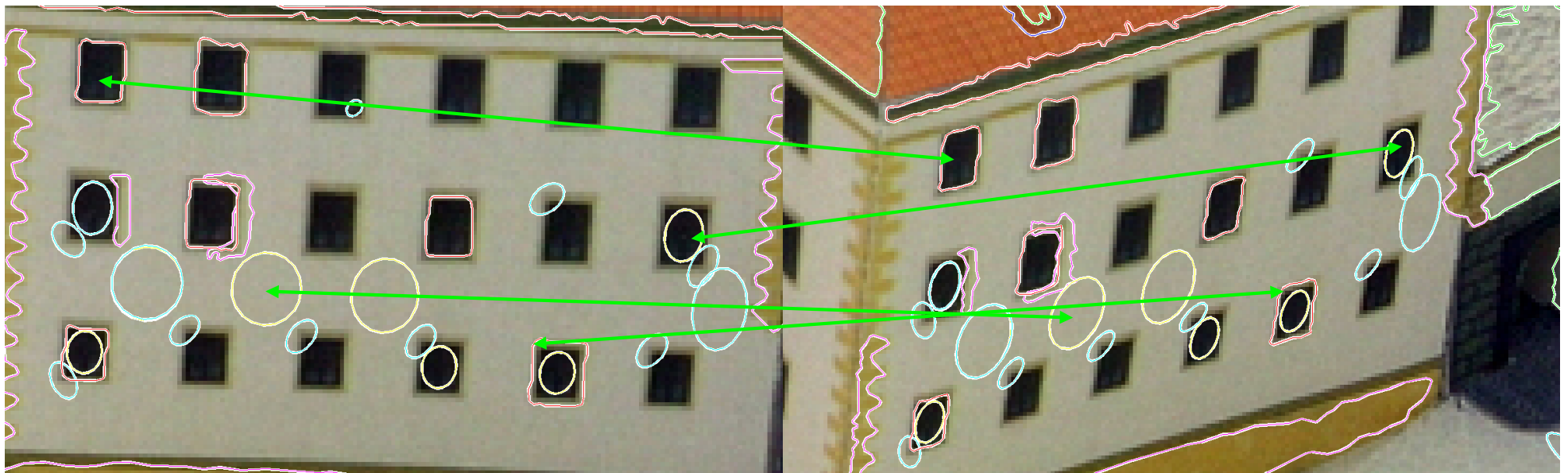


Some feature are similar but do not really match

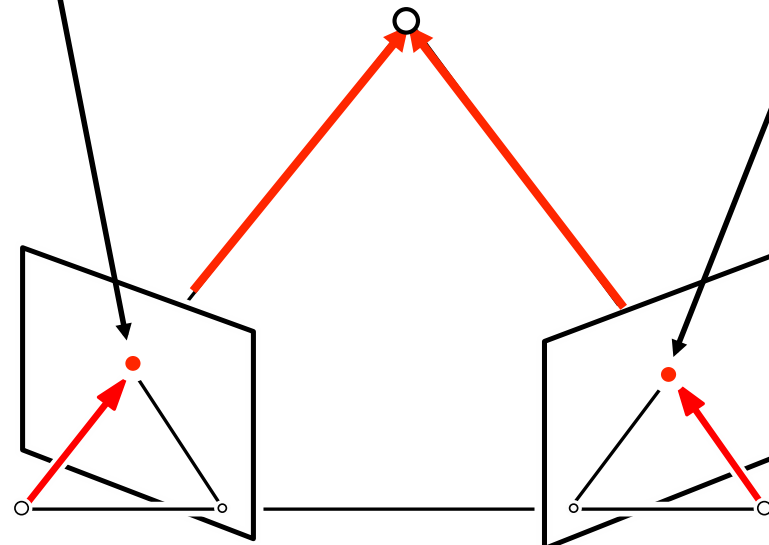
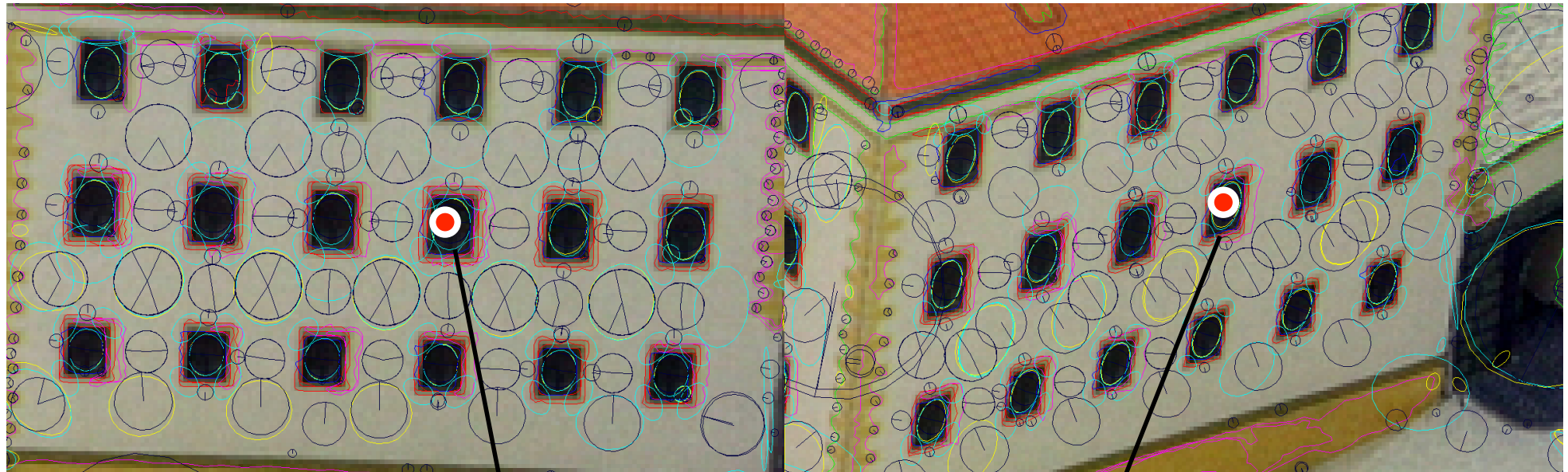
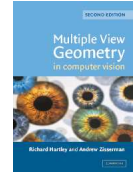
NOT ALL MATCHES ARE CORRECT



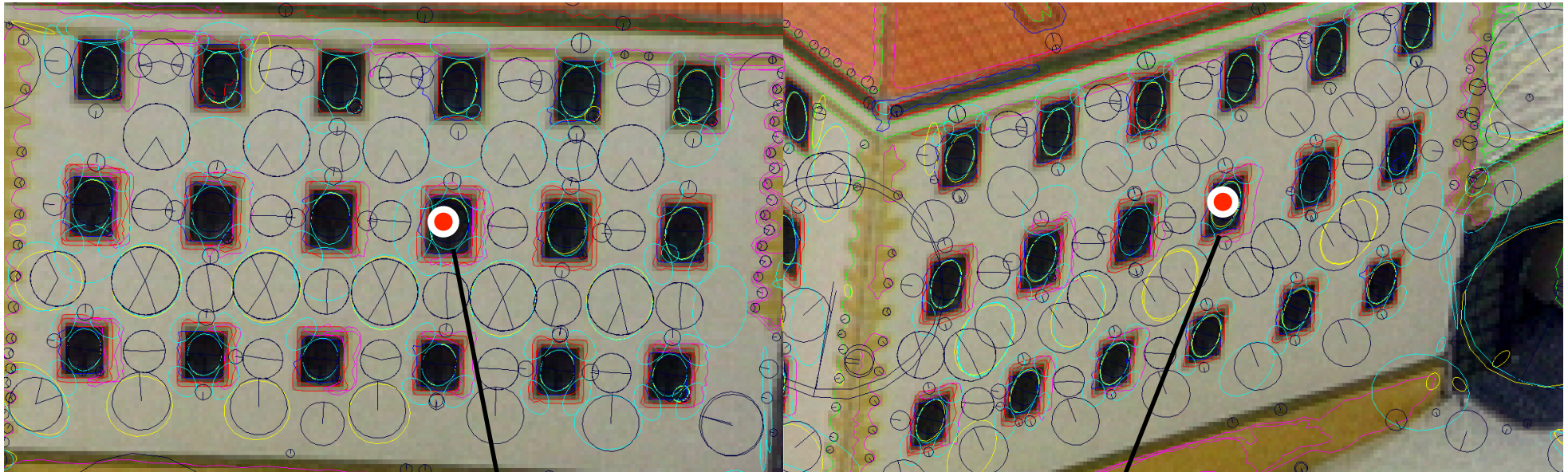
BUT SOME ARE



EPIPOLAR CONSTRAINT



EPIPOLAR CONSTR \rightarrow algebraic equation



$$\mathbf{x}_2^T \mathbf{F} \mathbf{x}_1 = 0$$

HOW TO GET ONE F?



F can be computed from 5 good matches

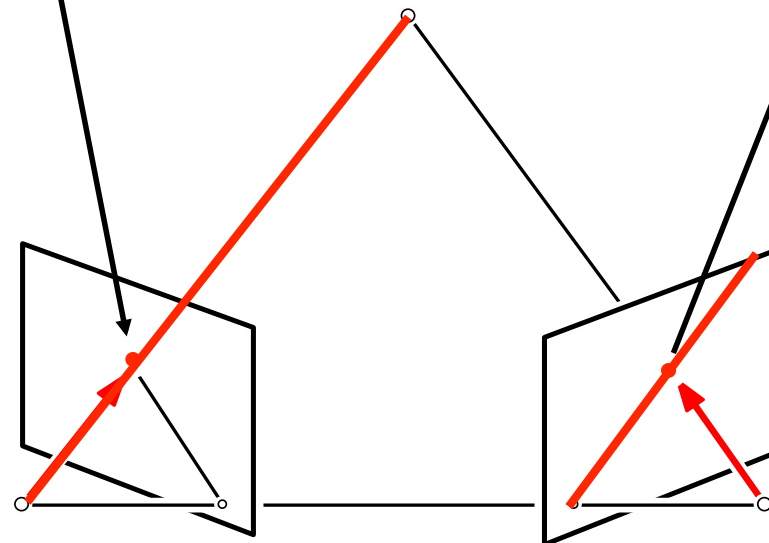
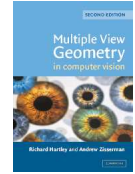
$$\mathbf{x}_2^T \mathbf{F} \mathbf{x}_1 = 0$$

HOW TO GET THE BEST F?

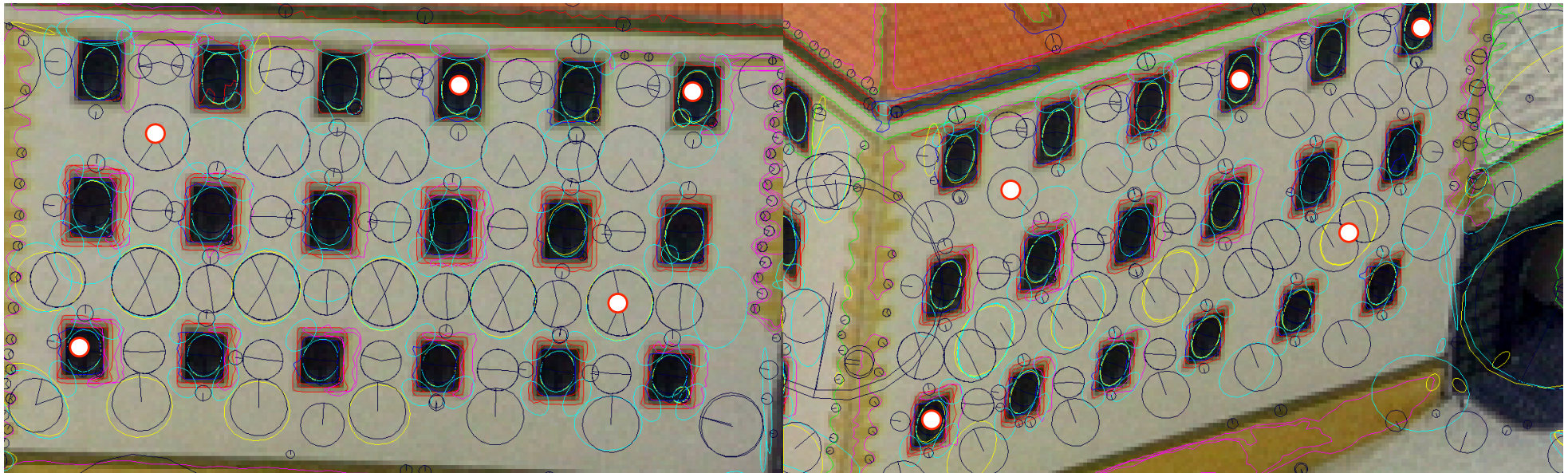
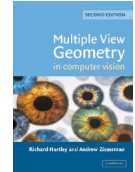


The best F is consistent with the highest number of matches

CONSISTENCY WITH F



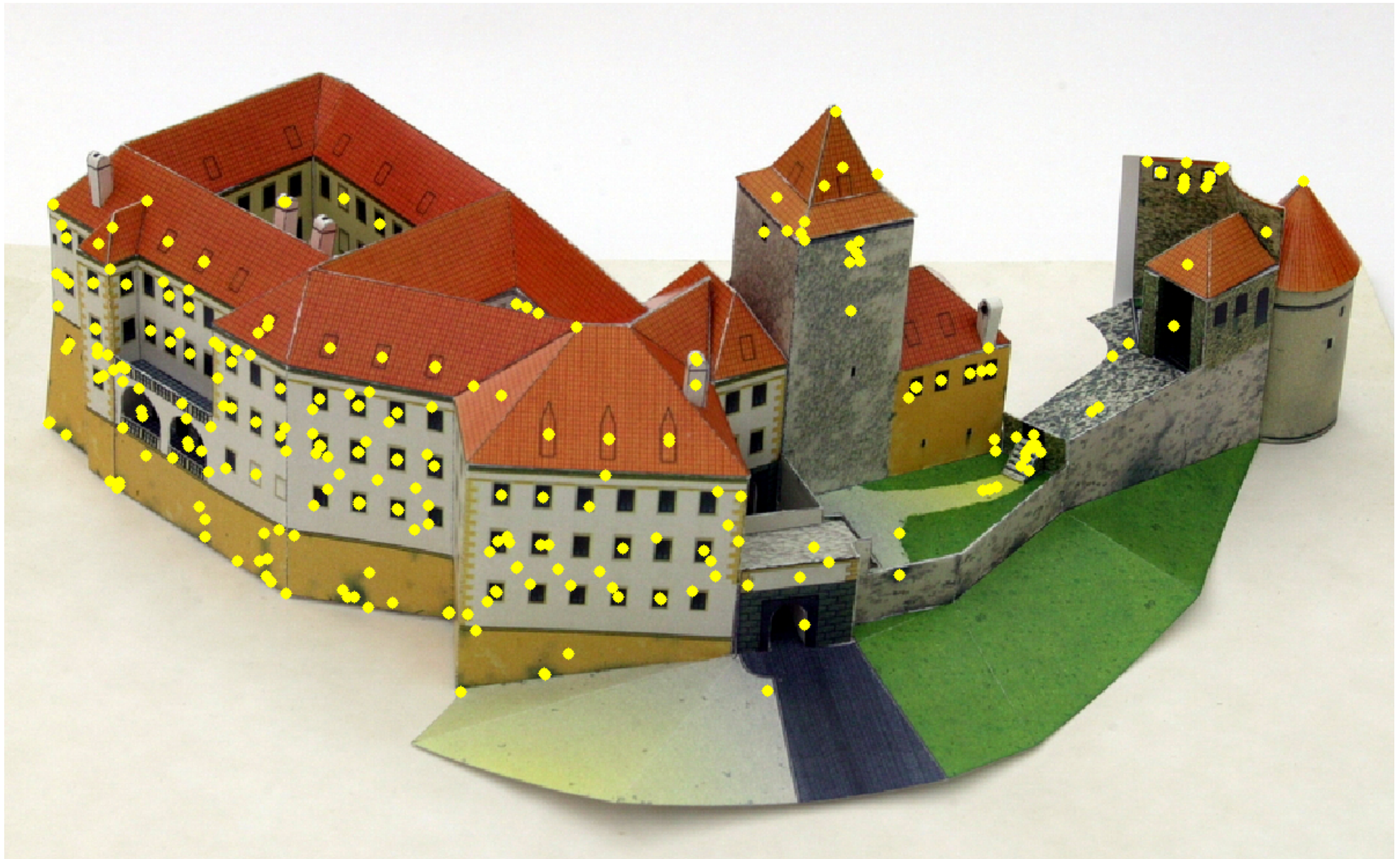
FINDING THE BEST F

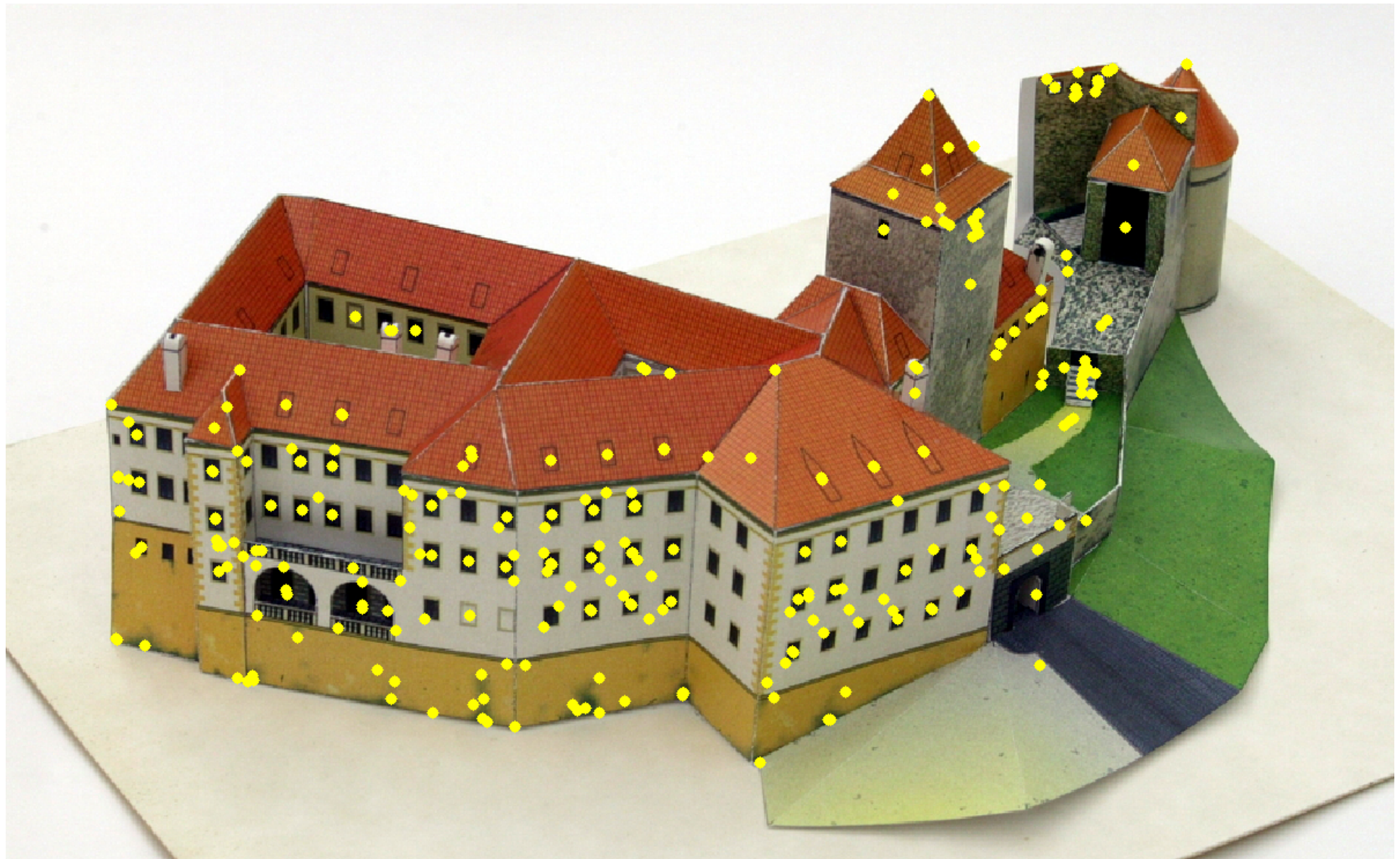


RANSAC (RANDOM SAMPLING CONSENSUS)

1. Generate random 5-tuples of matches
2. Compute F by solving $\mathbf{x}_2^T F \mathbf{x}_1 = 0$ (not so trivial)
3. Count the number of good matches

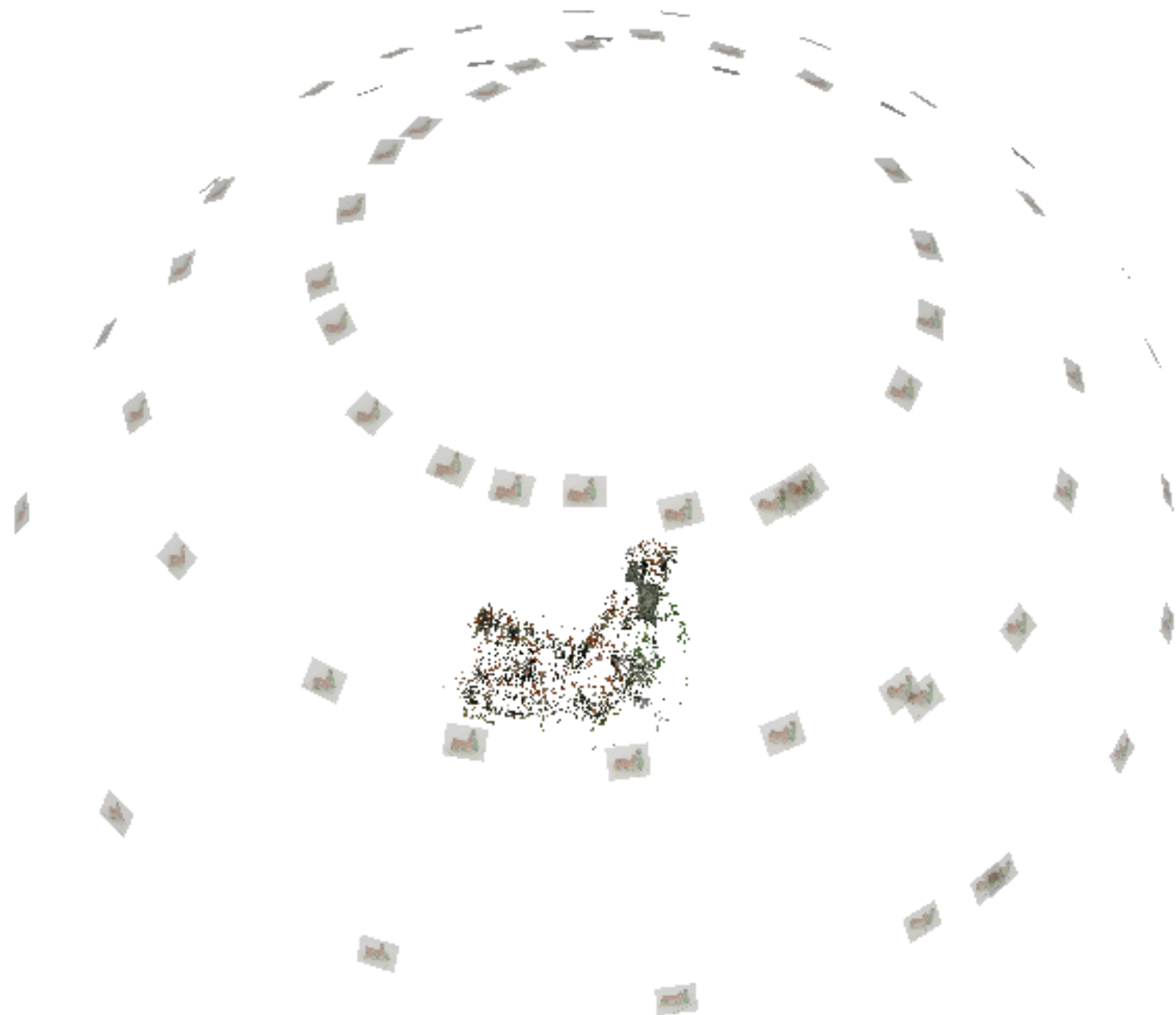
Return the largest set of good matches





INTEGRATED FROM MANY CAMERAS



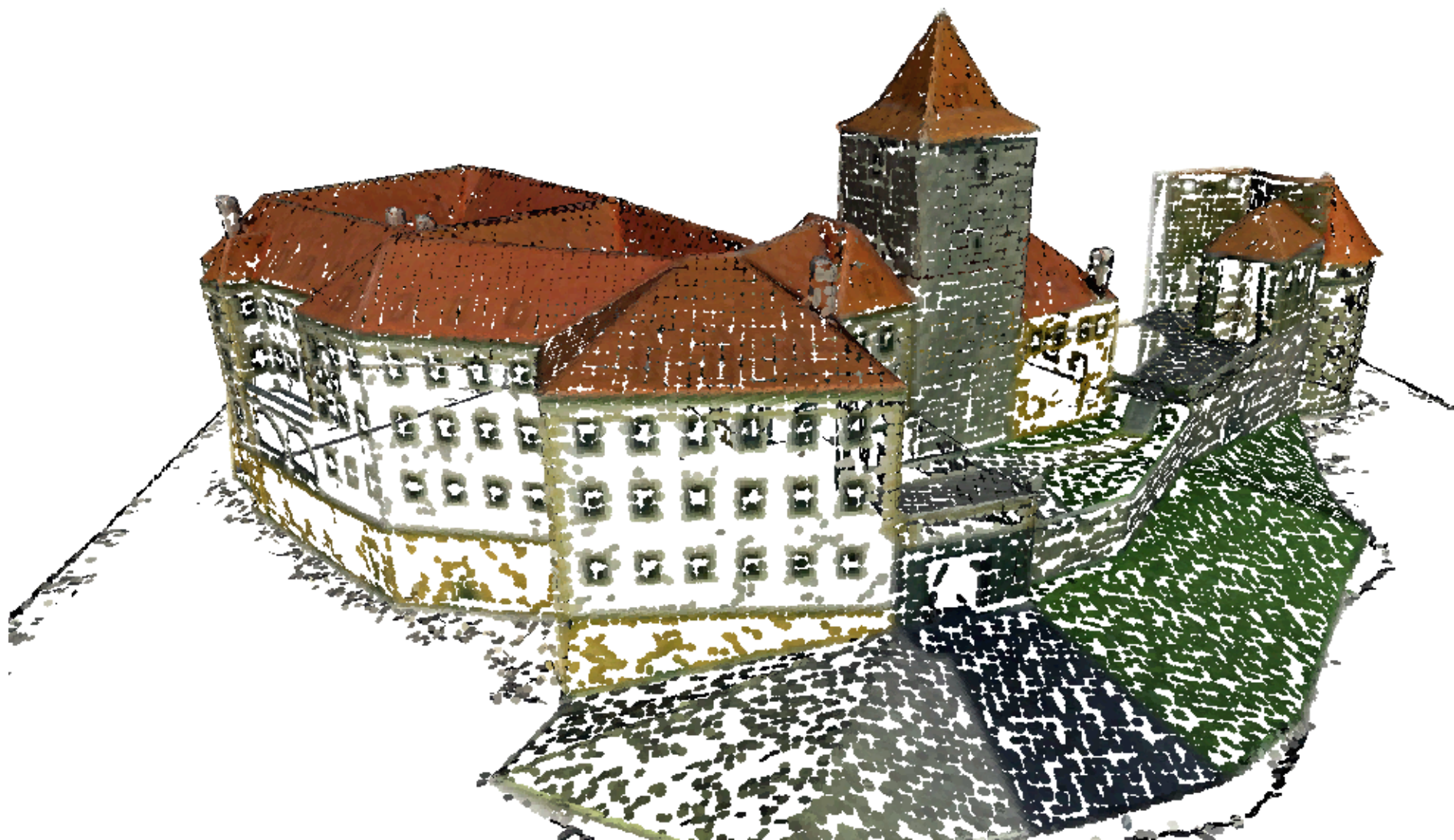




RESULT

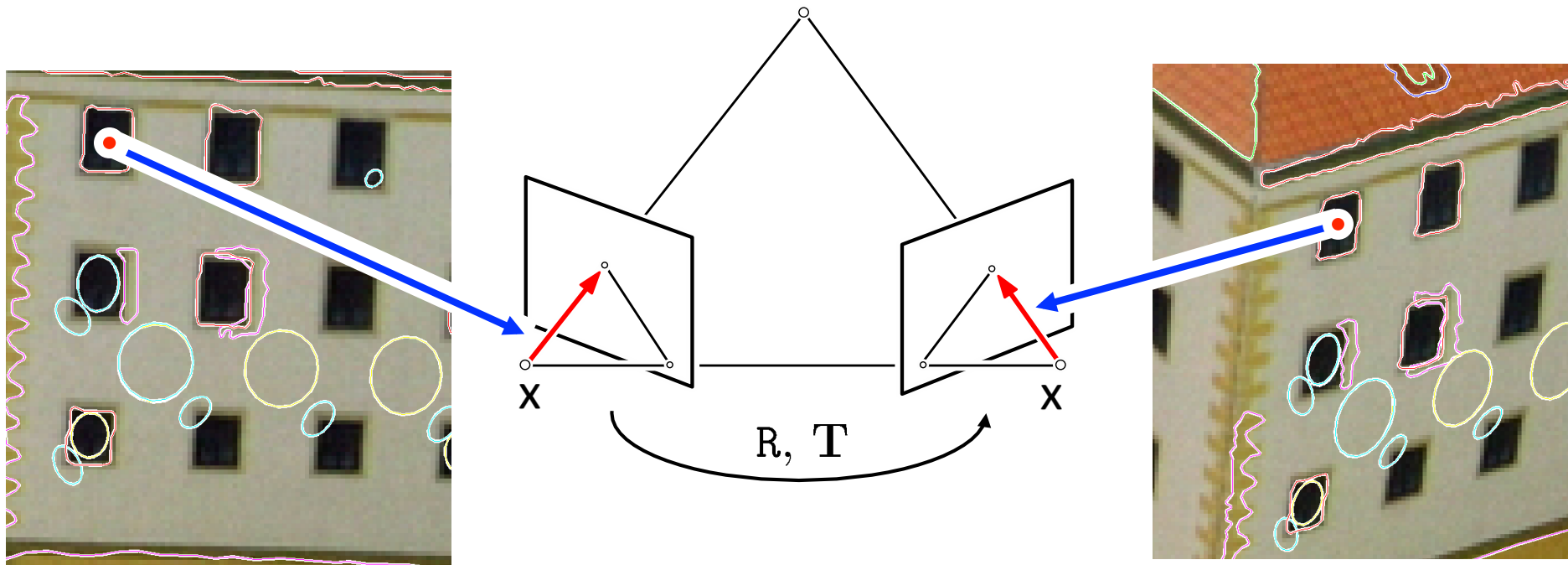


Dense 3D by fish scales of R. Sara et al.



Dense 3D by fish scales of R. Sara et al.

RELATIVE CAMERA POSE PROBLEM



$$\mathbf{x}_2^\top \mathbf{F} \mathbf{x}_1 = 0$$

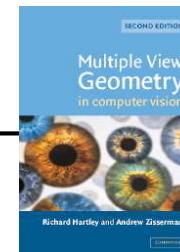
$$\det \mathbf{F} = 0$$

$$2\mathbf{F}\mathbf{F}^\top\mathbf{F} - \text{trace}(\mathbf{F}\mathbf{F}^\top)\mathbf{F} = 0$$

Algebraic equations

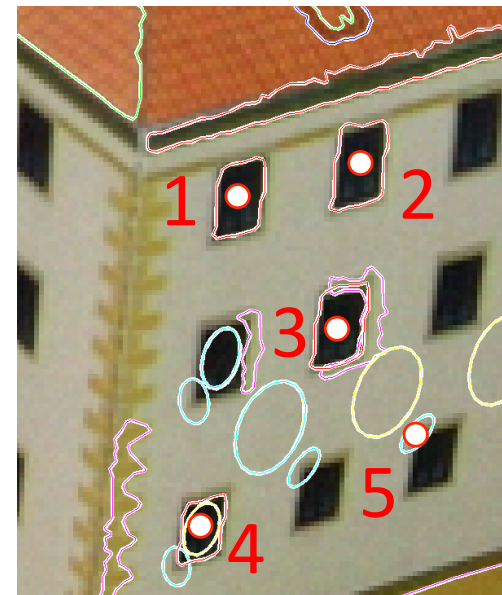
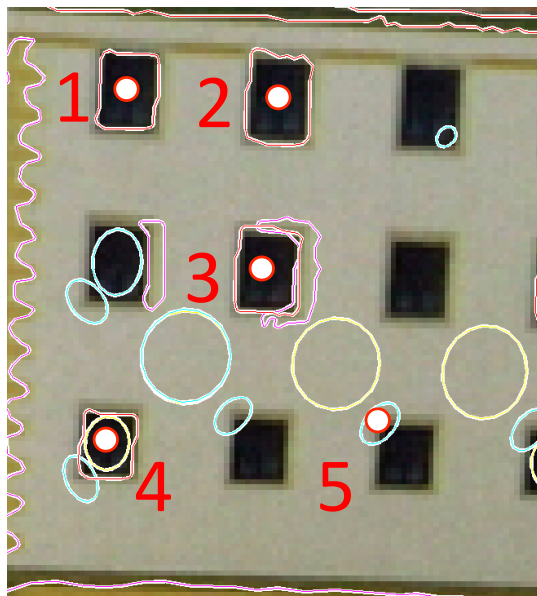
solve \longrightarrow

\mathbf{F}



$\mathbf{R}, \begin{bmatrix} \mathbf{T} \\ \|\mathbf{T}\| \end{bmatrix}$

MINIMAL PROBLEMS & RANSAC



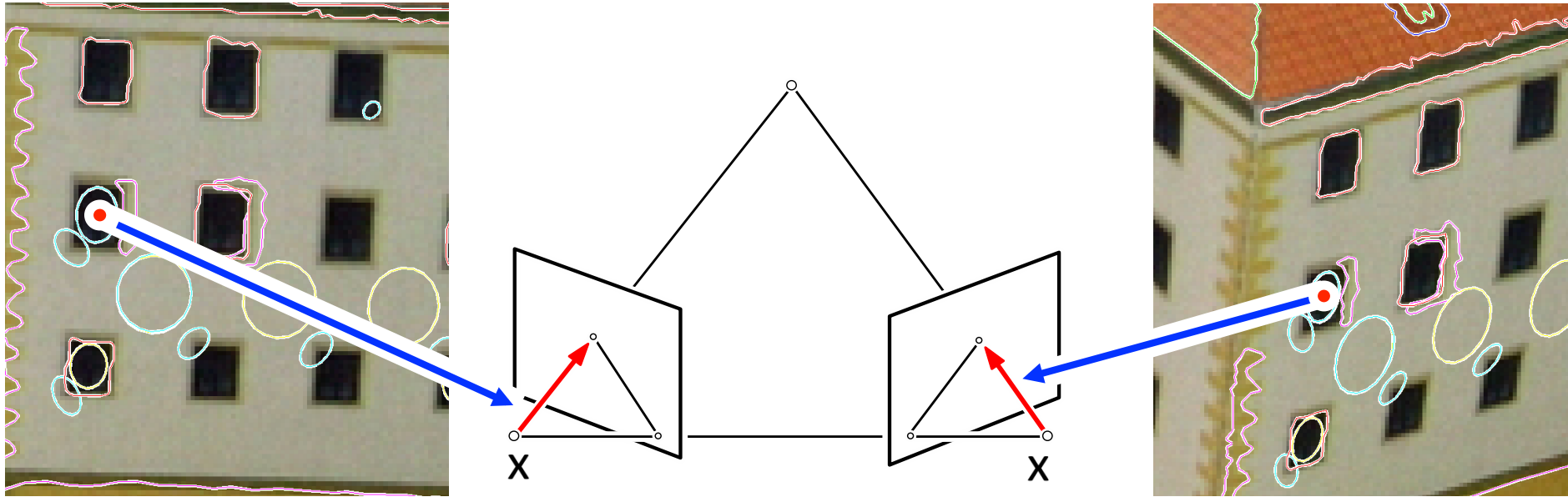
5 matches are necessary and sufficient

RANSAC: find F to maximize the # of good matches

select 5 matches \rightarrow compute F \rightarrow record the support




FORMULATION



$$\left. \begin{aligned} \mathbf{x}_2^\top \mathbf{F} \mathbf{x}_1 &= 0 \\ \det \mathbf{F} &= 0 \\ 2 \mathbf{F} \mathbf{F}^\top \mathbf{F} - \text{trace}(\mathbf{F} \mathbf{F}^\top) \mathbf{F} &= 0 \end{aligned} \right\} \text{Algebraic equations}$$

UNKNOWN S

$$F = \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix}$$


9 unknowns but only 8 have to be found (F up to scale)

→ we need at least 8 independent equations

EQUATIONS

$$\det F = 0$$

1 equation, degree 3

$$2FF^T F - \text{trace}(FF^T)F = 0$$

9 equations, degree 3

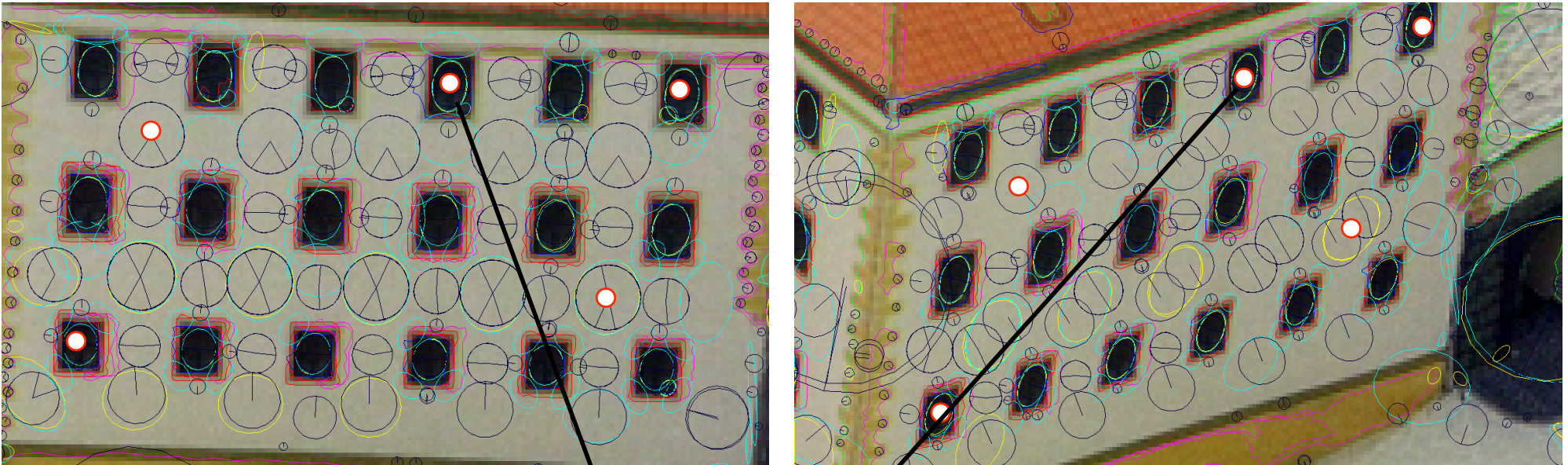


10 equations but only 3 “independent”

$$8 = 3 + 5$$

→ 5 more equations needed

5 EQUATIONS from image points



$$\mathbf{x}_2^T \mathbf{F} \mathbf{x}_1 = 0$$

5 linear equations:

$$\bullet f_{11} + \bullet f_{12} + \bullet f_{13} + \bullet f_{21} + \bullet f_{22} + \bullet f_{23} + \bullet f_{13} + \bullet f_{23} + \bullet f_{31} + \bullet f_{32} + \bullet f_{33} = 0$$

ELIMINATING UNKNOWNNS

5 linear equations:

$$\bullet f_{11} + \bullet f_{12} + \bullet f_{13} + \bullet f_{21} + \bullet f_{22} + \bullet f_{23} + \bullet f_{13} + \bullet f_{23} + \bullet f_{31} + \bullet f_{32} + \bullet f_{33} = 0$$

can be written in a matrix form

$$\underbrace{\begin{bmatrix} \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \end{bmatrix}}_A \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ \vdots \\ f_{33} \end{bmatrix} = 0$$

ELIMINATING UNKNOWNNS

$$\underbrace{\begin{bmatrix} \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \end{bmatrix}}_A \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ \vdots \\ f_{33} \end{bmatrix} = 0$$

A 5×9 matrix \rightarrow it has a 4 dimensional nullspace

$$A N = 0$$

known

$$N = \boxed{x} N_1 + \boxed{y} N_2 + \boxed{z} N_3 + \boxed{w} N_4$$

new unknowns

ELIMINATING UNKNOWNNS

$$F \sim x N_1 + y N_2 + z N_3 + w N_4 \quad \dots 4 \text{ unknowns}$$

F is up to scale \rightarrow choose a representative by setting $w = 1$

$$F := x N_1 + y N_2 + z N_3 + N_4$$



3 unknowns x, y, z

\downarrow substitute

$$\left. \begin{array}{l} \det F = 0 \\ 2 F F^T F - \text{trace}(F F^T) F = 0 \end{array} \right\} \begin{array}{l} 10 \text{ 3}^{\text{rd}} \text{ order equations in} \\ 3 \text{ unknowns} \end{array}$$

SOLVING IT

$$\left. \begin{array}{l} \det F = 0 \\ 2FF^T F - \text{trace}(FF^T)F = 0 \end{array} \right\} \begin{array}{l} 10 \text{ 3}^{\text{rd}} \text{ order equations in} \\ 3 \text{ unknowns} \end{array}$$

← 10 x 20 matrix

$X = 0$

M

$$X = [x^3, x^2y, x^2z, xy^2, xyz, xz^2, y^3, yz^2, z^3, x^2, xy, xz, y^2, yz, z^2, x, y, z, 1]^T$$

SOLVING IT

$$A_t = \begin{bmatrix} \boxed{-B} \\ 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

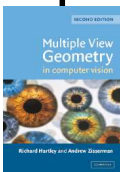
SOLUTIONS

$$[x, y, z] = \text{eig}(A_t)$$

... up to 10 solutions

back-substitution

$$F := x N_1 + y N_2 + z N_3 + N_4$$



R, T ... camera relative motion

ALGORITHM

1. Construct matrix M

2. Build matrix A_t

```
B = inv( M(:,1:10) ) * M(:,11:20);
```

```
A_t = zeros(10);
```

```
A_t(1:6,:) = -B([1:6], :);
```

```
A_t(7,1) = 1;
```

```
A_t(8,2) = 1;
```

```
A_t(9,3) = 1;
```

```
A_t(10,7) = 1;
```

3. Compute eigenvalues

```
[x, y, z] = eig(A_t)
```

4. Recover camera relative motion R, T

ALGORITHM

1. Not trivial to find
2. Simple & fast

SOLVING ALGEBRAIC EQUATIONS

The previous procedure was a particular case of a general technique for solving systems of algebraic equations.

HISTORY



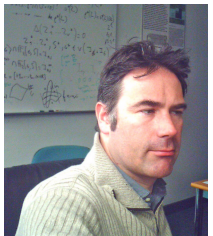
1888 *David Hilbert: Finiteness theorem*
Every ideal has a finite generating set



1965 *Bruno Buchberger: Groebner bases*
Computational procedure for solving systems
of polynomial equations
(Extremely simple: 20 lines of Maple code!)



1998 *Hans Stetter: Multiplication matrix*
A stable numerical procedure via eigenvectors



1999 *Jean-Charles Faugere: F4 algorithm*
An efficient computational tool for cryptography

SOLVING ALGEBRAIC EQUATIONS

1 equation, 1 variable → companion matrix → eigenvalues

$$f(x) = x^3 + 4x^2 + x - 6 = -6 + 1x + 4x^2 + 1x^3$$

$$M_x = \begin{bmatrix} 0 & 0 & 6 \\ 1 & 0 & -1 \\ 0 & 1 & -4 \end{bmatrix}$$

... a simple rule

```
>> e=eig(M_x)
```

$$e = \begin{bmatrix} 1 \\ -2 \\ -3 \end{bmatrix}$$

$$x_1 = 1, x_2 = -2, x_3 = -3$$

It works when eig works, i.e. order 100 in Matlab is often OK.

SOLVING ALGEBRAIC EQUATIONS

m equations, n variables

$$\begin{aligned} f_1(x, y) &= 25xy - 15x - 20y + 12 \\ f_2(x, y) &= x^2 + y^2 - 1 \end{aligned}$$

→ Groebner basis

(a set of polynomials with the same solutions but easier to solve)

$$g_1(x, y) = 125y^3 - 75y^2 + 27 - 45y$$

$$\begin{aligned} g_2(x, y) &= 25xy - 15x - 20y + 12 \\ g_3(x, y) &= x^2 + y^2 - 1 \end{aligned}$$

SOLVING ALGEBRAIC EQUATIONS

m equations, n variables

$$f_1(x, y) = 25xy - 15x - 20y + 12$$

$$f_2(x, y) = x^2 + y^2 - 1$$

→ Groebner basis → generalized
companion
matrix

$$M_{x+y} = \begin{bmatrix} 0 & 125 & 0 & 125 \\ -60 & 100 & 125 & 75 \\ -63 & 45 & 175 & 45 \\ 65 & 100 & -125 & 75 \end{bmatrix}$$

SOLVING ALGEBRAIC EQUATIONS

m equations, n variables

$$f_1(x, y) = 25xy - 15x - 20y + 12$$

$$f_2(x, y) = x^2 + y^2 - 1$$

→ Groebner basis → generalized companion matrix → eigenvectors

$$v \sim \begin{bmatrix} 1 & 1 & 1 & 1 \\ -\frac{4}{5} & \frac{4}{5} & \frac{4}{5} & \frac{4}{5} \\ \frac{3}{5} & -\frac{3}{5} & \frac{3}{5} & \frac{3}{5} \\ \frac{16}{25} & \frac{16}{25} & \frac{16}{25} & \frac{16}{25} \end{bmatrix}$$

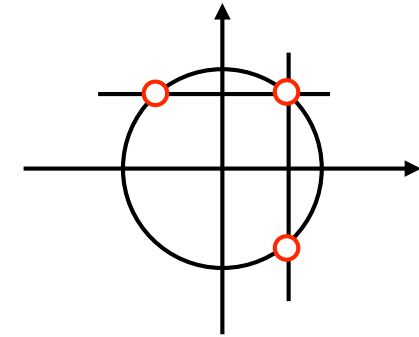
$$x_1 = -\frac{4}{5}, \quad y_1 = \frac{3}{5}$$

SOLVING ALGEBRAIC EQUATIONS

m equations, n variables

$$f_1(x, y) = 25xy - 15x - 20y + 12$$

$$f_2(x, y) = x^2 + y^2 - 1$$



→ Groebner basis → generalized companion matrix → eigenvectors

$$v \sim \begin{bmatrix} 1 & 1 & 1 & 1 \\ -\frac{4}{5} & \frac{4}{5} & \frac{4}{5} & \frac{4}{5} \\ \frac{3}{5} & -\frac{3}{5} & \frac{3}{5} & \frac{3}{5} \\ \frac{16}{25} & \frac{16}{25} & \frac{16}{25} & \frac{16}{25} \end{bmatrix}$$

$$\begin{aligned} x_1 &= -\frac{4}{5}, & y_1 &= \frac{3}{5} \\ x_2 &= \frac{4}{5}, & y_2 &= -\frac{3}{5} \\ x_3 &= \frac{4}{5}, & y_3 &= \frac{3}{5} \end{aligned}$$

THE DIFFICULT PART

Equations \rightarrow Groebner basis



no simple rule:

1. NP-complete in m, n (\sim 3-coloring of graphs)
(takes very long time to compute)
2. EXPSPACE-complete problem
(needs huge space to remember intermediate results)

GB COMPUTATION ALGORITHMS

Manipulation ... polynomial multiplication & pseudo-division

$$f_1 = 25xy - 15x - 20y + 12$$

$$f_2 = x^2 + y^2 - 1$$



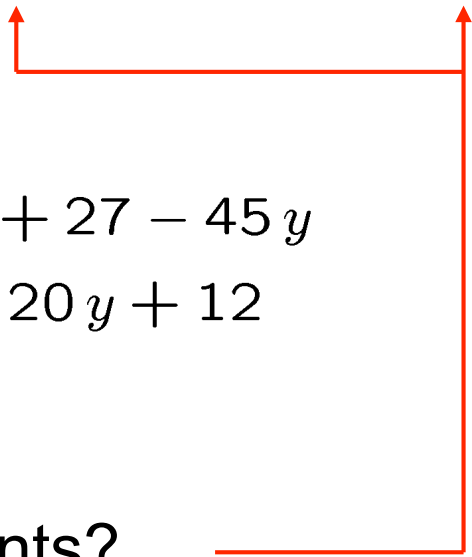
$$? \quad g_1 = (-5y - 3)f_1 + (125x - 10)f_2$$

$$g_1 = 125y^3 - 75y^2 + 27 - 45y$$

$$g_2 = 25xy - 15x - 20y + 12$$

$$g_3 = x^2 + y^2 - 1$$

How to find coefficients?



GB COMPUTATION ALGORITHMS

1. “Standard” (Hironaka 1964) and Groebner (Burchberger 1965) bases
2. 1965: Buchberger’s algorithm
 - a generalization of the Gauss-Jordan elimination
 - extremely simple: 7 (+ 13 for the rem division) lines of code
 - not efficient
 - not good for numerical approximations
3. 1999 (2005): F4 (F5) algorithm (J.-C. Faugere)
 - more efficient, more robust, more complex

COMPUTING GB MAY BE VERY HARD

Example:

4 polynomials, 3 variables, degree ≤ 6 , small integer coeffs

$$f_1 = 8x^2y^2 + 5xy^3 + 3x^3z + x^2yz$$

$$f_2 = x^5 + 2y^3z^2 + 13y^2z^3 + 5yz^4$$

$$f_3 = 8x^3 + 12y^3 + xz^2 + 3$$

$$f_4 = 7x^2y^4 + 18xy^3z^2 + y^3z^3$$

have extremely simple Groebner basis

$$g_1 = x$$

$$g_2 = y^3 + \frac{1}{4}$$

$$g_3 = z^2$$

HOWEVER

when computed by the Buchberger's algorithm over the rational numbers w.r.t. the grevlex ordering $x > y > z$, the following polynomial appears during the computation:

$$y^3 - 1735906504290451290764747182 \dots$$



~ 80,000 digits

COMPUTATION

Macaulay2 program over the rational field \mathbb{Q}

```
R = QQ[x,y,z, MonomialOrder=>GRevLex];  
I = ideal(8*x^2*y^2 + 5*x*y^3 + 3*x^3*z + x^2*y*z,  
          x^5 + 2*y^3*z^2 + 13*y^2*z^3 + 5*y*z^4,  
          8*x^3 + 12*y^3 + x*z^2 + 3,  
          7*x^2*y^4 + 18*x*y^3*z^2 + y^3*z^3)  
G = gens gb I
```

will run very long.

The problem is in remembering very long coefficients.

COMPUTATION

Macaulay2 program over the finite field $\mathbb{Z}/13$

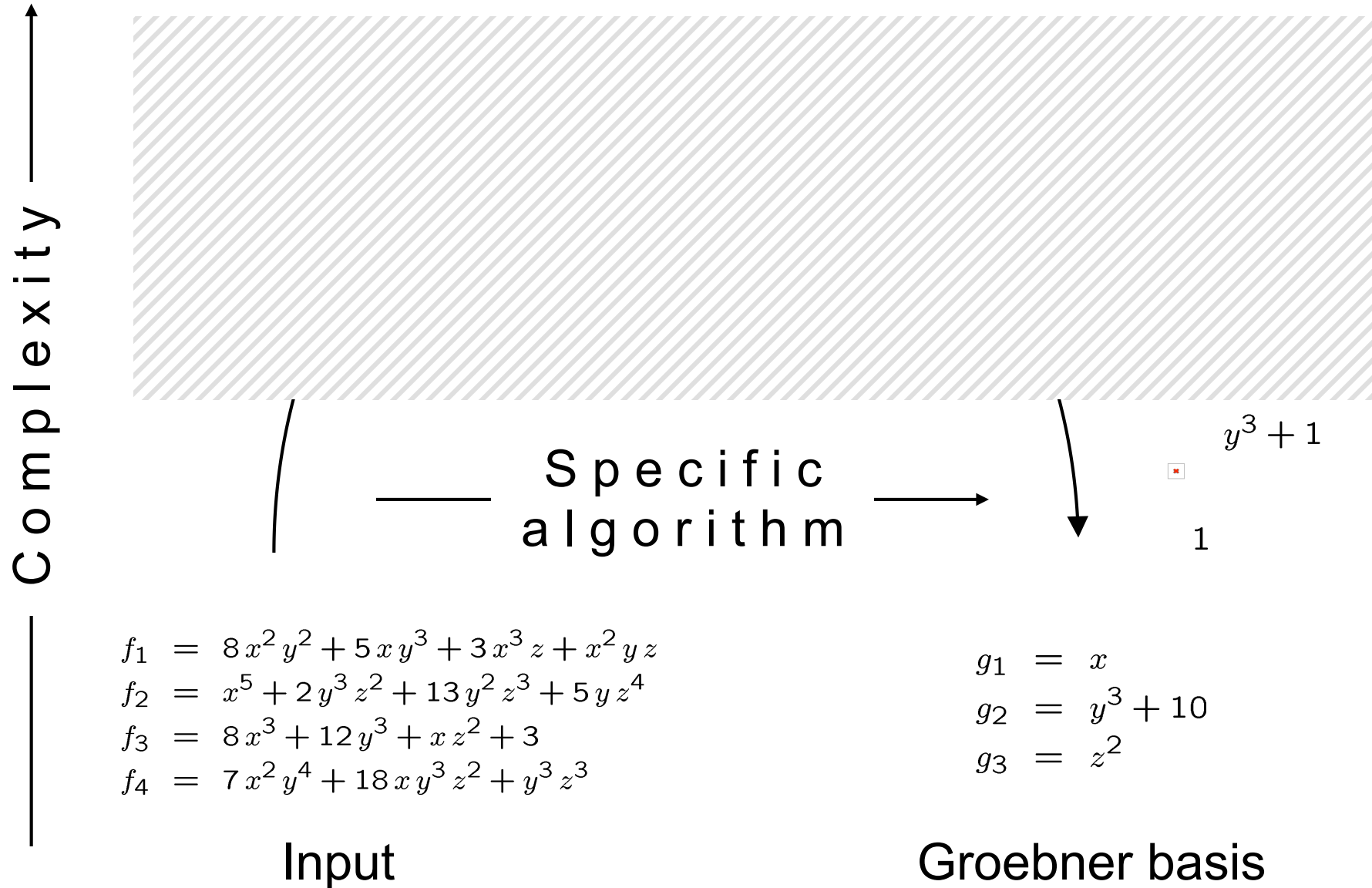
```
R = ZZ/13[x,y,z, MonomialOrder=>GRevLex];
I = ideal(8*x^2*y^2 + 5*x*y^3 + 3*x^3*z + x^2*y*z,
          x^5 + 2*y^3*z^2 + 13*y^2*z^3 + 5*y*z^4,
          8*x^3 + 12*y^3 + x*z^2 + 3,
          7*x^2*y^4 + 18*x*y^3*z^2 + y^3*z^3)
G = gens gb I
```

returns a very similar basis in the fraction of the second

$$\begin{array}{l} g_1 = x \\ g_2 = y^3 + 10 \\ g_3 = z^2 \end{array} \quad \left(\begin{array}{l} g_1 = x \\ g_2 = y^3 + \frac{1}{4} \\ g_3 = z^2 \end{array} \right)$$

WHY?

General algorithms can construct all GBs but often generate many complicated polynomials



SPECIFIC GB CONSTRUCTION ALG'S

1. Find a short path towards the GB which is independent from the actual coefficients, implement it efficiently.
2. Use floating-point arithmetics to do the manipulations to avoid huge coefficients.

SHORT PATH TO GB

1. Restructure & reformulate the problem to reduce the number of variables and the degree of monomials.
2. Use a computer algebra system (Macaulay2) to compute the Groebner basis in a finite field (fast!) for random coefficients and remember the path:

```
R = ZZ/P[x,y,z, MonomialOrder=>GRevLex];
```

↑

“lucky” prime number (always exists)

try $P = 1, 2, 3, 5, 7, \dots, 30011, 30013, 30029, \dots$

until the result stabilizes (always does)

FLOATING POINT ARITHMETICS

Strictly speaking, polynomial manipulations must be done in exact arithmetics

$$f_1 = 25xy + 25x + 12$$

$$f_2 = x^2y + x^2 + 3$$

$$\begin{aligned} x f_1 - 25 f_2 &= x(25xy + 25x + 12) - 25(x^2y + x^2 + 3) \\ &= 12x - 75 \end{aligned}$$

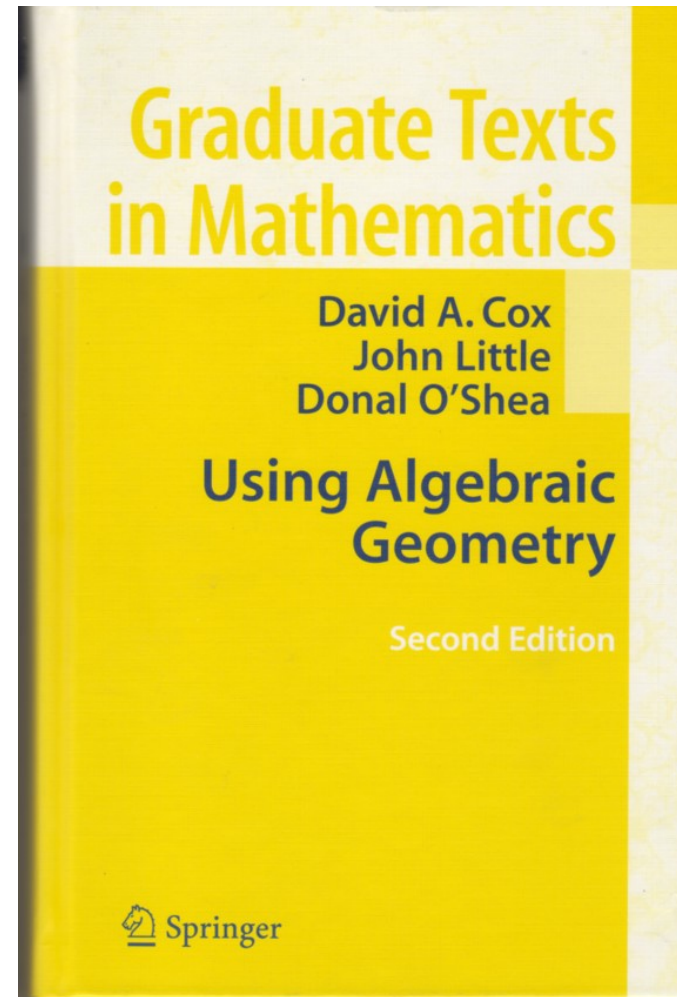
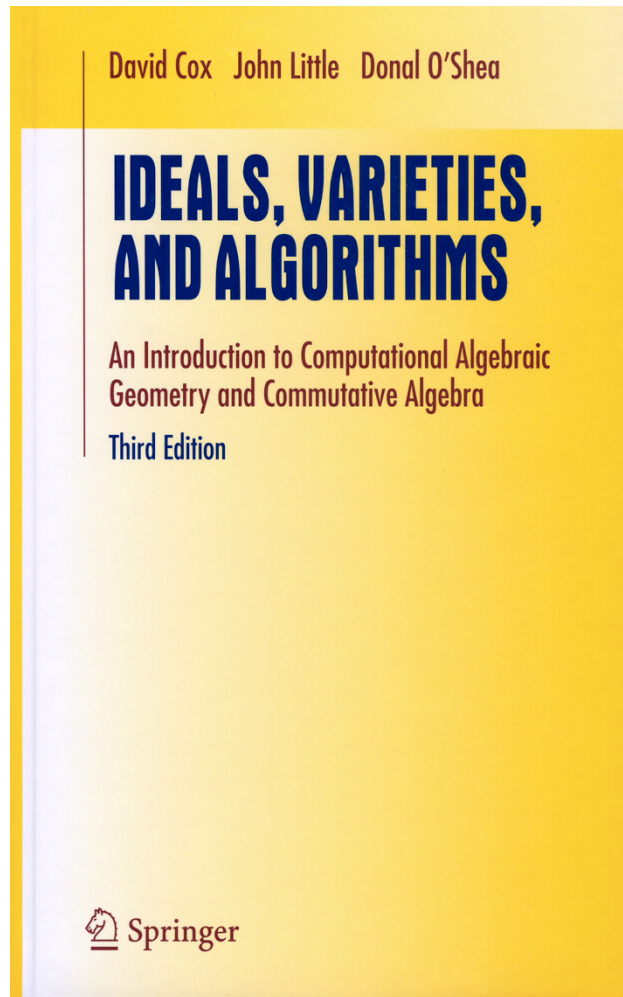
Double canceling  may fail when rounding occurs

in floating point arithmetics

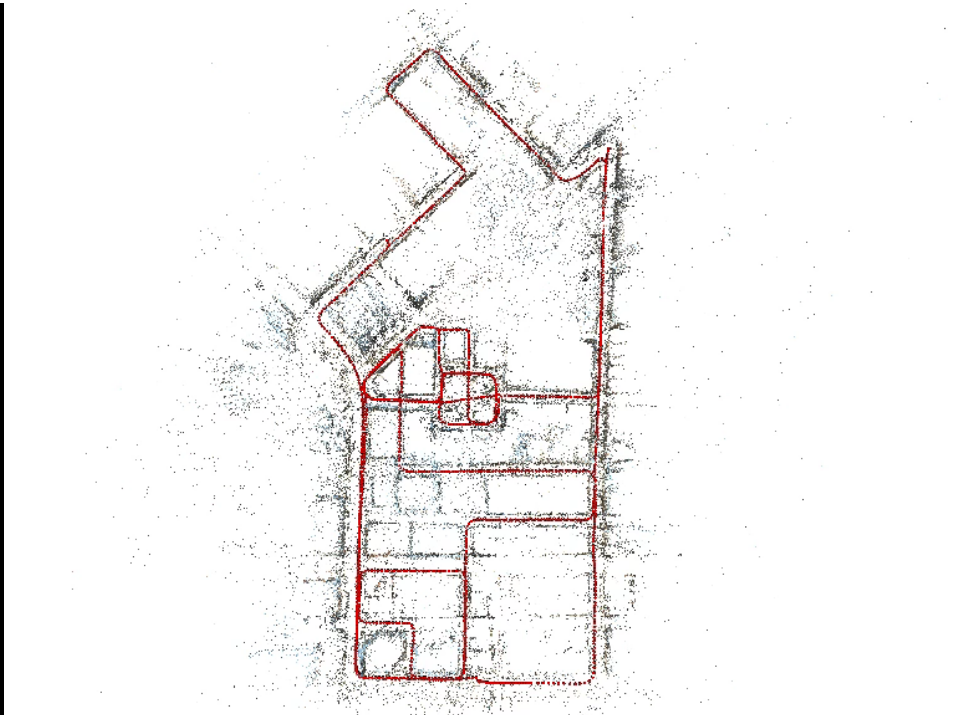
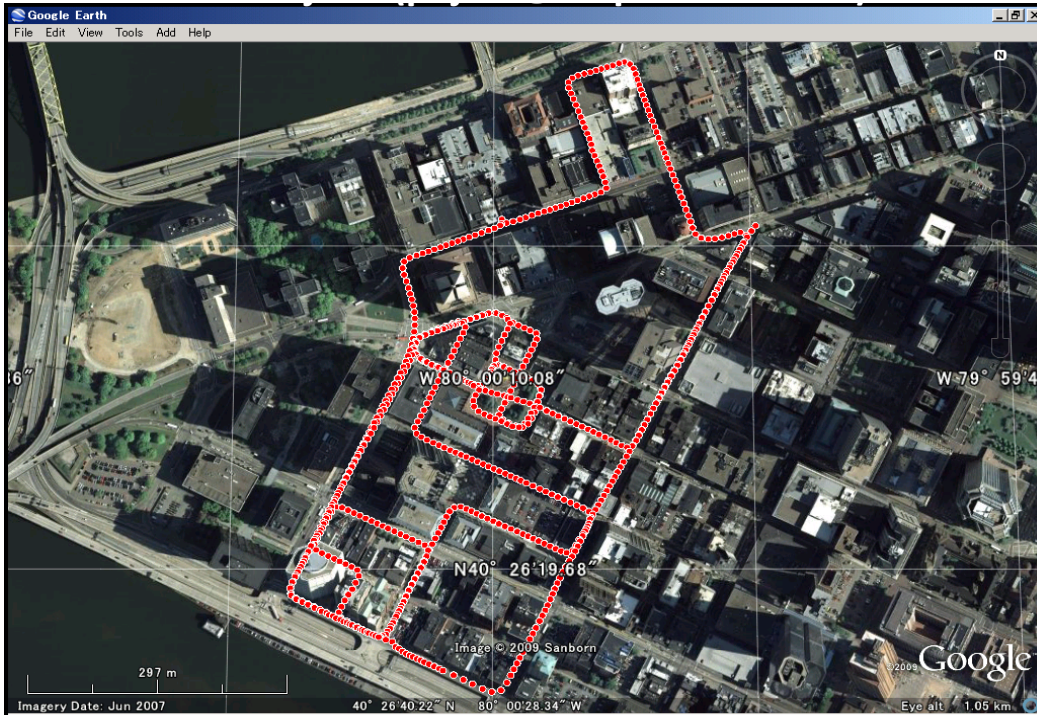
FLOATING POINT ARITHMETICS

For some computer vision problems rounding does not destroy the result.

All manipulations have to be done with care and checked on typical data in randomized experiments.



From Google Street View to 3D City Models



Welcome to CMP SfM Web Service

ptak.felk.cvut.cz/sfmservice/

CMP SfM Web Service

User login

Username:
 Password:

What?
 CMP SfM Web Service provides a remote access to the 3D reconstruction systems developed in Center for Machine Perception. The access to the system is granted on request by email to Tomas Pajda <pajda@cmp.felk.cvut.cz>.

Why?
 We provide the access to the service to our partners and to people in the Computer Vision community to make it easier to use our codes. There is no need to install any code on a client's computer and all the computations are performed on our dedicated computing cluster. Further, it makes it easier to compare the results of different methods to ours based on the same data.

News

- 17/03/11 - You can now use icon in 'Datasets table' to run SfM (No XML file needed).
- 06/01/11 - New web interface (version Tokyo).

Menu

- SfM Web Service
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Input Images

Output Model

Multi-View Stereo : Automatic 3.

3D Reconstruction of Tocnik data-set.
 -set: 96 (1528 x 1148) images.



PRoVisG Mars 3D Challenge

User login

Username:
 Password:

Menu

- Introduction
- Registration
- Forum
- Challenges
- Calibration
- Submit Results
- Evaluation
- Organizers
- Participants

Welcome to the PRoVisG Mars 3D Challenge!

The Mars 3D Challenge aims at testing and improving the state of the art in visual odometry and 3D terrain reconstruction in planetary exploration.

The task of the challenge is to reconstruct depth, camera trajectory and 3D map of Mars landscape observed by the Mars Exploration Rovers (MER). Results of the challenge will help to develop technology for future planetary mission such as ESA ExoMars.

The challenge consists of three stages. The results of the challenge will be evaluated by the PRoVisG Project consortium including the JPL/NASA, operating the MERs. Results of the challenge will be presented at a ICV 2011 workshop "CV/VEIM - Computer Vision in Vehicle Technology: From Earth to Mars" and published in a follow-up journal paper. The winners of the challenge will be awarded a travel to and a visit of a JPL/ESA space laboratory.

Register to the challenge and contribute to understanding the Universe!

All results welcome: it's possible to compete in stage 2 even if you didn't submit stage 1 results!

News

- 15/6/2011 - Stage 2 Challenge Released
- 9/3/2011 - Additional info at MDDC Forum
- 1/3/2011 - Stage 1 Challenge Released

