A Clever Elimination Strategy for Efficient Minimal Solvers

Motivation

- New strategy for solving minimal problems
- Do more computation in an off-line stage and less in an on-line stage
- Many minimal problems in computer vision lead to coupled sets of linear and polynomial equations where image measurements enter the linear equations only
- New strategy
 - . Eliminate all unknowns which do not appear in the linear equa**tions** - do this only once in the pre-processing step (offline)
 - 2. Extend solutions to the other unknowns
- Can be generalized to fully non-linear systems by monomial lifting
- New constraints on the fundamental matrix of partially calibrated cameras

Basic algebraic geometry

Consider a system of *m* polynomial equations in *n* unknowns $X = \{x_1, \ldots, x_n\}$

$$F = \{f_1(x_1, \dots, x_n) = 0, \dots, f_m(x_1, \dots, x_n) = 0\}$$
 (1)

Assume that the set $F = \{f_1, \ldots, f_m\}$ generates a zero dimensional **ideal**

$$I = \left\{ \sum_{i=1}^{m} g_i f_i, |g_1, \dots, g_m \in \mathbb{C}[X] \right\} \subset \mathbb{C}[X],$$
(2)

Gröbner basis *G* in Lex order - a special basis of the ideal $I = \langle G \rangle$ which contains polynomial in one variable

Elimination theory - classical algorithmic approach to eliminating some variables between polynomials of several variables

The *j*th elimination ideal of *I* is $I_j = I \cap \mathbb{C}[x_{j+1}, \dots, x_n]$ - we eliminate the first j variables to obtain I_j

Theorem: Let G be a Gröbner basis for I in the lex order. Then G ($\mathbb{C}[x_{j+1},\ldots,x_n]$ is a Gröbner basis for the $j^t h$ elimination ideal I_j .

Example: Let $I = \langle x^3 + xy + 1, x^2y^2, y^4 \rangle$. Its Gröbner basis in the lex order is $y^2, x^3 + xy + 1$. Therefore $I_x = \langle y^2 \rangle$

Standard elimination strategy

Partitioned *F* into two subsets:

 $F_L = \{ f_i \in F \mid deg(f_i) = 1 \},\$ (3)

 $F_N = \{ f_i \in F \mid deg(f_i) > 1 \}.$

 F_L - linear polynomials from F

 F_N - polynomials of higher degrees

SOTA Gröbner basis solvers - variables are mostly eliminated in the online phase **Online:**

- 1. Rewrite the linear equations F_L in the unknowns X_L as $MX_L = 0$, where M is a coefficient matrix.
- 2. Compute a null space basis N of M, re-parametrize the unknowns $X_L = N Y$
- 3. Substitute $X_L = \mathbb{N}Y$ into the non-linear equations F_N .
- 4. Solve the system $F_N(Y \cup (X \setminus X_L)) = 0$ (e.g. by the automatic generator [5]).



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Clever elimination strategy

Divide the *n* unknowns *X* into two subsets:

$$X_L = \{ x_i \in X \mid x_i \text{ appears in some } f \in F_L \}$$
(5)
$$X_N = X \setminus X_L.$$
(6)

 X_L - unknowns that appear in linear equations

 X_N - unknowns that appear in equations of higher degree only Notation:

 $|F_L| = m_L, |F_N| = m_N, |X_L| = n_L, |X_N| = n_N, m = m_L + m_N, n = n_L + n_N.$

Idea: Eliminate variables X_N from F_N in the pre-processing (offline) step

Offline:

1. Let $I = \langle F_N \rangle$ and consider the elimination ideal $I_{X_N} = I \cap \mathbb{C}[X_L]$.

2. Compute the generators G of I_{X_N} containing only unknowns in X_L . **Online:**

3. Rewrite the linear equations F_L in the unknowns X_L as $MX_L = 0$.

- 4. Compute null space basis N of M and re-parametrize the unknowns
- $X_L = \mathbb{N}Y$. For rank $\mathbb{M} = m_L$, Y contains $k = n_L m_L$ new unknowns.
- 5. Substitute $X_L = \mathbb{N}Y$ into the generators G of elimination ideal I_{X_N} .
- 6. Solve the new system of polynomial equations G(Y) = 0 in k un-
- knowns (e.g. by the automatic generator [5]).
- 7. Back-substitute to recover $X_L = \mathbb{N}Y$.
- 8. Extend partial solutions for X_L to solutions for X.

f+E+f relative pose problem

Problem of estimating relative pose and the common unknown focal length of two cameras from six image point correspondences:

Epipolar constraints

$$\mathbf{x}_i^{\top} \mathbf{F} \, \mathbf{x}_i' = 0 \tag{7}$$

for six image point correspondences $\mathbf{x}_i, \mathbf{x}'_i, i = 1, \dots, 6$. Rewrite in a matrix form

$$If = 0, \tag{8}$$

where M is a 6×9 coefficient matrix and f is a vector of 9 elements of F. For six image correspondences F can be parametrized by two unknowns as

$$\mathbf{F} = x \,\mathbf{F}_1 + y \,\mathbf{F}_2 + \mathbf{F}_3,\tag{9}$$

where F_1, F_2, F_3 are matrices from the three-dimensional null space of M and x and *y* are new unknowns.

The rank constraint

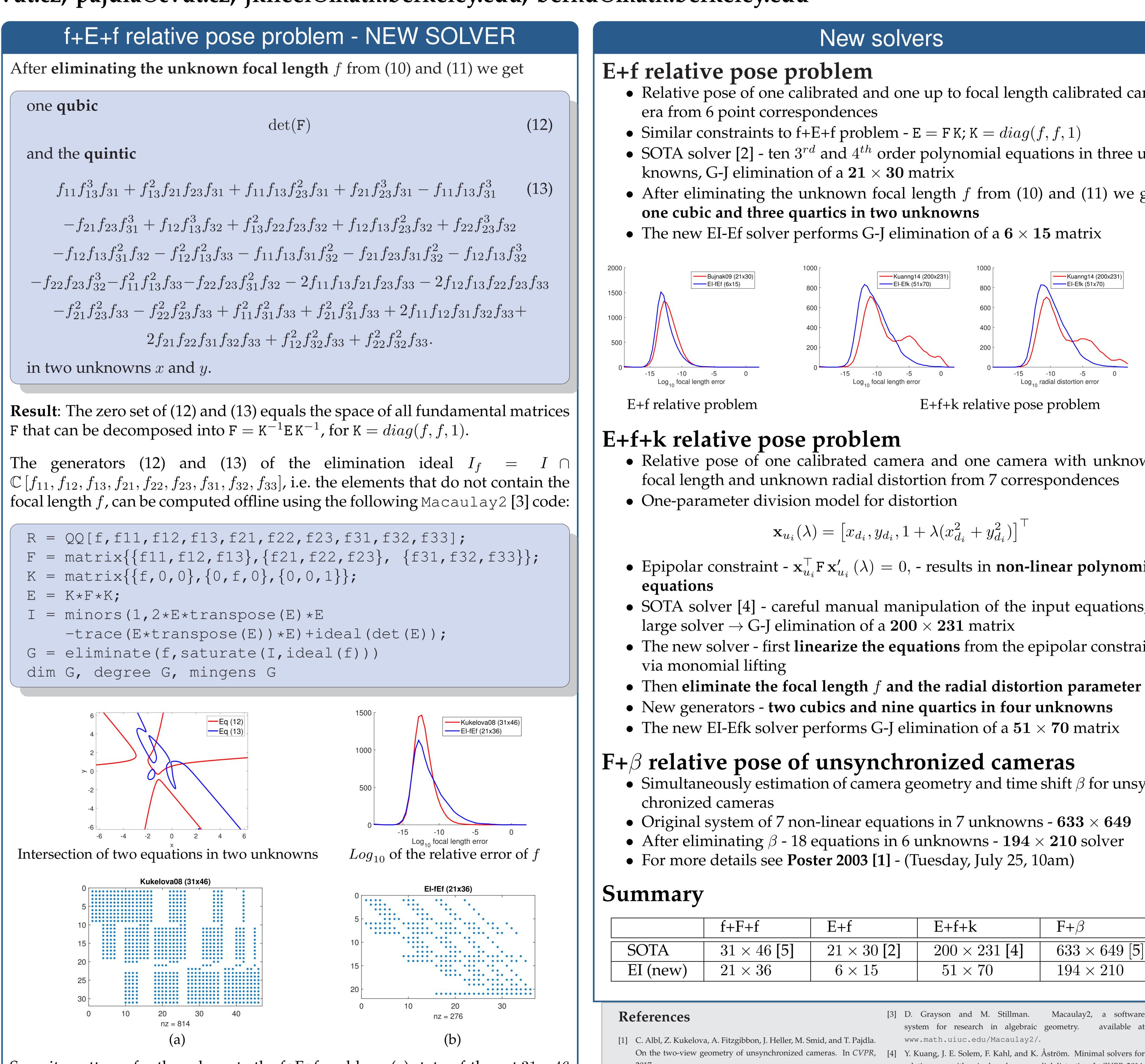
$$\det(\mathbf{F}) = 0 \tag{10}$$

The trace constraint

$$2 \mathbf{F} \mathbf{Q} \mathbf{F}^{\top} \mathbf{Q} \mathbf{F} - trace(\mathbf{F} \mathbf{Q} \mathbf{F}^{\top} \mathbf{Q}) \mathbf{F} = \mathbf{0}, \qquad (11)$$

where $Q = KK = diag(f^2, f^2, 1)$.

All SOTA solvers solve these ten third- and fifth-order polynomial equations (10) and (11) in three unknowns x, y and $w = 1/f^2$



Sparsity patterns for the solvers to the f+F+f problem: (a) state-of-the-art 31×46 Kukelova08 [5] solver (b) the new 21×36 EI-fEf solver.



- Relative pose of one calibrated and one up to focal length calibrated cam-
- SOTA solver [2] ten 3^{rd} and 4^{th} order polynomial equations in three un-
- After eliminating the unknown focal length *f* from (10) and (11) we get

- Relative pose of one calibrated camera and one camera with unknown

$$\mathbf{x}_{u_i}(\lambda) = \left[x_{d_i}, y_{d_i}, 1 + \lambda (x_{d_i}^2 + y_{d_i}^2) \right]^\top$$

- Epipolar constraint $\mathbf{x}_{u_i}^{\top} \mathbf{F} \mathbf{x}_{u_i}' (\lambda) = 0$, results in **non-linear polynomial**
- SOTA solver [4] careful manual manipulation of the input equations, a
- The new solver first **linearize the equations** from the epipolar constraint
- Then eliminate the focal length f and the radial distortion parameter λ

- Simultaneously estimation of camera geometry and time shift β for unsyn-

	f+F+f	E+f	E+f+k	$F+\beta$
SOTA	31 × 46 [5]	21 × 30 [2]	200×231 [4]	633×649 [5]
EI (new)	21×36	6×15	51×70	194×210

- Macaulay2, a software

M. Bujnak, Z. Kukelova, and T. Pajdla. 3D reconstruction from image collections with a single known focal length. In *ICCV*, 2009.

relative pose with a single unknown radial distortion. In CVPR, 2014 [5] Z. Kukelova, M. Bujnak, and T. Pajdla. Automatic generator of min-

imal problem solvers. In *ECCV*, 2008.