# Stratified Autocalibration of Cameras with Euclidean Image Plane

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# 3D reconstruction from uncalibrated images



A reconstruction only up to a projective ambiguity can be obtained.

# Euclidean Image Plane Assumption

A camera with square pixels *i.e.* zero skew and unit aspect ratio is said to have a Euclidean Image Plane (EIP) [Heyden and Åström, 1997].

- Most modern cameras have (very close to) square pixels
- Not fully exploited in stratified autocalibration

Assumption	Methods
Constant intrinsics	[Pollefeys and Van Gool, 1999] [Chandraker et al., 2010] [Adlakha et al., 2019]
Constant intrinsics + zero skew	[Habed et al., 2012] [Wu et al., 2013]

# Contributions

Assuming a moving camera with EIP and constant intrinsic parameters,

- formulation of a new quartic polynomial in the plane at infinity,  $\pi_{\infty}$ , that is obtained for each image pair  $\longrightarrow$  affine reconstruction
- a stratified autocalibration method that can be used with 3 or more images

Experiments show that our method performs more reliably than existing ones.

# Background

• The perspective  $3 \times 4$  projection matrices (uncalibrated) are of the form,

$$P_i = [H_i | e_i], \quad i = 1, 2, ..., n$$

where  $H_i$  is the homography of the reference plane and  $e_i$  is the epipole.

• The inter-image homography induced by  $\pi_\infty$  is given as [Habed et al., 2012],

$$\mathsf{H}_{\infty ij} = \mathsf{H}_{j}\mathsf{H}_{i}^{*} - \mathsf{H}_{j}[\pi_{\infty}]_{\times} \mathsf{H}_{i}^{\mathsf{T}}[\mathsf{e}_{i}]_{\times}^{\mathsf{T}} - \mathsf{e}_{j}\pi_{\infty}^{\mathsf{T}}\mathsf{H}_{i}^{*}$$

where H<sup>\*</sup> is the adjoint matrix of H and  $[\pi]_{\times}$  is the skew-symmetric matrix associated with vector  $\pi$ . As such, H<sub> $\infty ij$ </sub> is linear in  $\pi_{\infty}$ .

# Modulus constraint

- For constant intrinsic parameters,  $H_{\infty ij}$  is a conjugate rotation
- Eigenvalues of  $H_{\infty ij}$  thus have equal moduli

Characteristic polynomial of  $H_{\infty ij}$ ,

$$\det(\mathsf{H}_{\infty 1j} - \lambda \mathsf{H}_{\infty 1i}) = -\det(\mathsf{H}_{\infty 1i})\lambda^3 + \operatorname{tr}(\mathsf{H}_{\infty ij})\lambda^2 - \operatorname{tr}(\mathsf{H}_{\infty ji})\lambda + \det(\mathsf{H}_{\infty 1j}) = 0$$

**Modulus constraint:** necessary condition on  $\pi_{\infty}$  for  $H_{\infty ij}$  to satisfy the eigenvalue property [Pollefeys and Van Gool, 1999],

$$m_{ij}(\pi_{\infty}) = \det(\mathsf{H}_{\infty 1i})\operatorname{tr}(\mathsf{H}_{\infty ji})^3 - \det(\mathsf{H}_{\infty 1j})\operatorname{tr}(\mathsf{H}_{\infty ij})^3 = 0$$
 for all  $i \neq j$ 

 $m_{ij}$  is a quartic polynomial in  $\pi_{\infty}$ .

# Infinite Cayley Transform

For two cameras i and j with constant intrinsic parameters, the Infinite Cayley Transform (ICT) [Wu et al., 2013, Habed et al., 2012] is,

$$\mathsf{Q}_{\infty ij} = \lambda_j \mathsf{H}_{\infty ij} - \lambda_i \mathsf{H}_{\infty ji}$$

#### Properties

- $Q_{\infty ij}$  is similar to a skew-symmetric matrix
- $\operatorname{tr}(\mathbf{Q}_{\infty ij}^*) > 0$ 
  - $\rightarrow$  combined with the modulus constraint, form necessary and sufficient conditions for  $Q_{\infty ij}$  to be similar to a skew-symmetric matrix [Wu et al., 2013]
- Assuming zero skew, the coordinates of the principal point (u,v) can be expressed as [Habed et al., 2012],

$$u = (\mathbf{Q}_{\infty ij})_{11}/(\mathbf{Q}_{\infty ij})_{31}, \quad v = (\mathbf{Q}_{\infty ij})_{22}/(\mathbf{Q}_{\infty ij})_{32}$$

# New EIP polynomial constraint

Given a  $3 \times 3$  matrix B, we define the matrix operator  $\Phi$  as,

 $\Phi(\mathsf{B}) = (\mathsf{B}^* \circ \mathsf{B})_{31} + (\mathsf{B}^* \circ \mathsf{B})_{32}$ 

where  $\circ$  is the Hadamard (elementwise) product.

**ICT property:** Consider two cameras *i* and *j* with EIP and constant intrinsics,

$$\Phi(\mathbf{Q}_{\infty ij}) = 0$$

A quartic polynomial constraint on  $\pi_\infty$  can be derived using this property.

## New EIP polynomial constraint

We observe that  $\Phi(\mathsf{Q}_{\infty ij})$  expands as,

$$\Phi(\mathbf{Q}_{\infty ij}) = a_{ij}(\pi_{\infty})\lambda_j^3 - b_{ij}(\pi_{\infty})\lambda_i\lambda_j^2 + b_{ji}(\pi_{\infty})\lambda_i^2\lambda_j - a_{ji}(\pi_{\infty})\lambda_i^3 = 0$$

where the coefficients  $a_{ij}$  and  $b_{ij}$  are cubic polynomials in  $\pi_{\infty}$ .

**Key result:**  $\lambda_j^3 a_{ij}(\pi) = \lambda_i^3 a_{ji}(\pi)$  if the modulus constraint is satisfied. Thus,  $p_{ij}(\pi_{\infty}) = -b_{ij}(\pi_{\infty}) \operatorname{tr}(\mathsf{H}_{\infty ji}) + b_{ji}(\pi_{\infty}) \operatorname{tr}(\mathsf{H}_{\infty ij}) = 0$ 

 $p_{ij}$  is a new quartic polynomial in  $\pi_{\infty}$ : the EIP polynomial.

# Polynomial inequality constraints

- Chirality (camera centers):  $c_i(\pi_{\infty}) > 0, \quad i = 1, ..., n$ , where  $c_i(\pi_{\infty}) = \det(\mathsf{H}_{\infty 1i})$
- ICT property:  $q_{ij}(\pi_{\infty}) = \operatorname{tr}(\mathbf{Q}^*_{\infty ij}) > 0, \quad i = 1, \dots, n-1,$  $j = i+1, \dots, n$
- Principal point bounds: for an image-centered  $2\overline{u} \times 2\overline{v}$  image,

$$u_{ij}(\pi_{\infty}) = \overline{u}^{2}(\mathbf{Q}_{\infty ij})_{31}^{2} - (\mathbf{Q}_{\infty ij})_{11}^{2} \ge 0, \quad i = 1, \dots, n-1,$$
  
$$v_{ij}(\pi_{\infty}) = \overline{v}^{2}(\mathbf{Q}_{\infty ij})_{32}^{2} - (\mathbf{Q}_{\infty ij})_{22}^{2} \ge 0, \quad j = i+1, \dots, n$$

# Estimating the plane at infinity

**Homogenized polynomials:** for a polynomial f of degree d in  $\pi$ ,

 ${}^{h}f(\pi,\pi_{4}) = \pi_{4}^{d}f(\pi/\pi_{4})$ 

#### Polynomial optimization problem:

$$\min_{\pi, \pi_{4}} \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} {}^{h} m_{ij}^{2}(\pi, \pi_{4}) + {}^{h} p_{ij}^{2}(\pi, \pi_{4})$$
s.t.
$$\stackrel{h}{c_{i}(\pi, \pi_{4}) > 0,} \qquad i = 1, \dots, n,$$

$$\stackrel{h}{u_{ij}(\pi, \pi_{4}) > 0,} \qquad i = 1, \dots, n-1, \quad j = i+1, \dots, n,$$

$$\stackrel{h}{u_{ij}(\pi, \pi_{4}) \ge 0,} {}^{h} v_{ij}(\pi, \pi_{4}) \ge 0, \qquad i = 1, \dots, n-1, \quad j = i+1, \dots, n,$$

$$\stackrel{h}{c_{1}(\pi, \pi_{4})^{h} c_{n}(\pi, \pi_{4})} + \frac{1}{n-1} \sum_{i=1}^{n-1} {}^{h} c_{i}(\pi, \pi_{4})^{h} c_{i+1}(\pi, \pi_{4}) = 1$$

Solved using Lasserre's hierarchy [Lasserre, 2008, Henrion et al., 2009].

# Stratified autocalibration algorithm

Given a projective reconstruction,

- 1. Estimate  $\pi_\infty$  by solving the polynomial optimization problem using Lasserre's hierarchy
- 2. Refine the estimated  $\pi_\infty$  using the normalized cost:

$$\pi_{\infty} = \arg\min_{\pi} \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \frac{m_{ij}^2(\pi) + p_{ij}^2(\pi)}{(c_i(\pi)c_j(\pi))^4}$$

3. Compute the intrinsic parameters linearly

# **Experimental results**



EIP\* our algorithm

**EIP** EIP\* - inequality constraints

GO-Stratified QUARCH\*M GO-DAQ [Chandraker et al., 2010] [Adlakha et al., 2019] [Chandraker et al., 2007]

# Experimental results

Sequence	Method	$\Delta f(\%)$	$\Delta uv(\%)$	$\Delta\gamma$	Time (s)
fountain-P11	EIP	0.08	0.25	1.06	0.59
	GO-Stratified QUARCH*M	$\begin{array}{c} 0.10 \\ 0.05 \end{array}$	$\begin{array}{c} 0.19 \\ 0.23 \end{array}$	$\begin{array}{c} 1.08 \\ 1.05 \end{array}$	$\begin{array}{c} 302.90 \\ 2.44 \end{array}$
	GO-DAQ	0.36	1.26	0.01	1.49
Herz-Jesu-P8	EIP	0.55	2.84	3.98	0.57
	GO-Stratified QUARCH*M	$\begin{array}{c} 43.86\\ 0.88 \end{array}$	$\begin{array}{c} 31.13\\ 3.11 \end{array}$	$\begin{array}{c} 157.31 \\ 2.03 \end{array}$	$243.18 \\ 1.26$
	GO-DAQ	1.43	1.27	0.05	1.53
City hall Leuven	EIP	0.78	0.72	2.80	0.56
	GO-Stratified QUARCH*M GO-DAQ	$7.09 \\ 2.94 \\ 9.93$	$10.10 \\ 6.70 \\ 7.68$	$25.85 \\ 5.81 \\ 9.70$	$169.21 \\ 1.02 \\ 1.38$

# **Experimental results**





Golden Statue



Eglise du Dome



Alcatraz Water Tower





Arbre aux Serpents<sup>1</sup>

<sup>1</sup>L'Arbre aux Serpents de Niki de Saint Phalle ©Musées d'Angers/ ©2017 Niki Charitable Art Foundation. Image courtesy Renato Saleri.

## Summary

- Formulated a new quartic polynomial constraint on  $\pi_\infty$  assuming a moving camera with EIP and constant intrinsic parameters
- Our stratified autocalibration method relies on polynomial optimization and can be used with 3 or more images
- Experiments showed that our method performs more reliably than existing ones, especially for short sequences