

Motion Estimation for Self-Driving Cars With a Generalized Camera

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Introduction

- **Objective**: Ego-motion estimation for a self-driving car equipped with a multi-camera system, i.e. a generalized camera.
- **Grobi** Our self-driving car with a generalized camera made up of 4 fisheye cameras with minimal/non- overlapping field-of-views.



• Sample images from the generalized camera.



- Main contributions:
 - New formulation of the generalized epipolar essential matrix from combining the generalized epipolar constraint (GEC) and Ackermann motion model.
 - Analytical 2-point minimal solution for ego-motion estimation with metric scale.
 - Investigation of and practical solution to the degenerate case of straight motion.

Generalized Camera Model

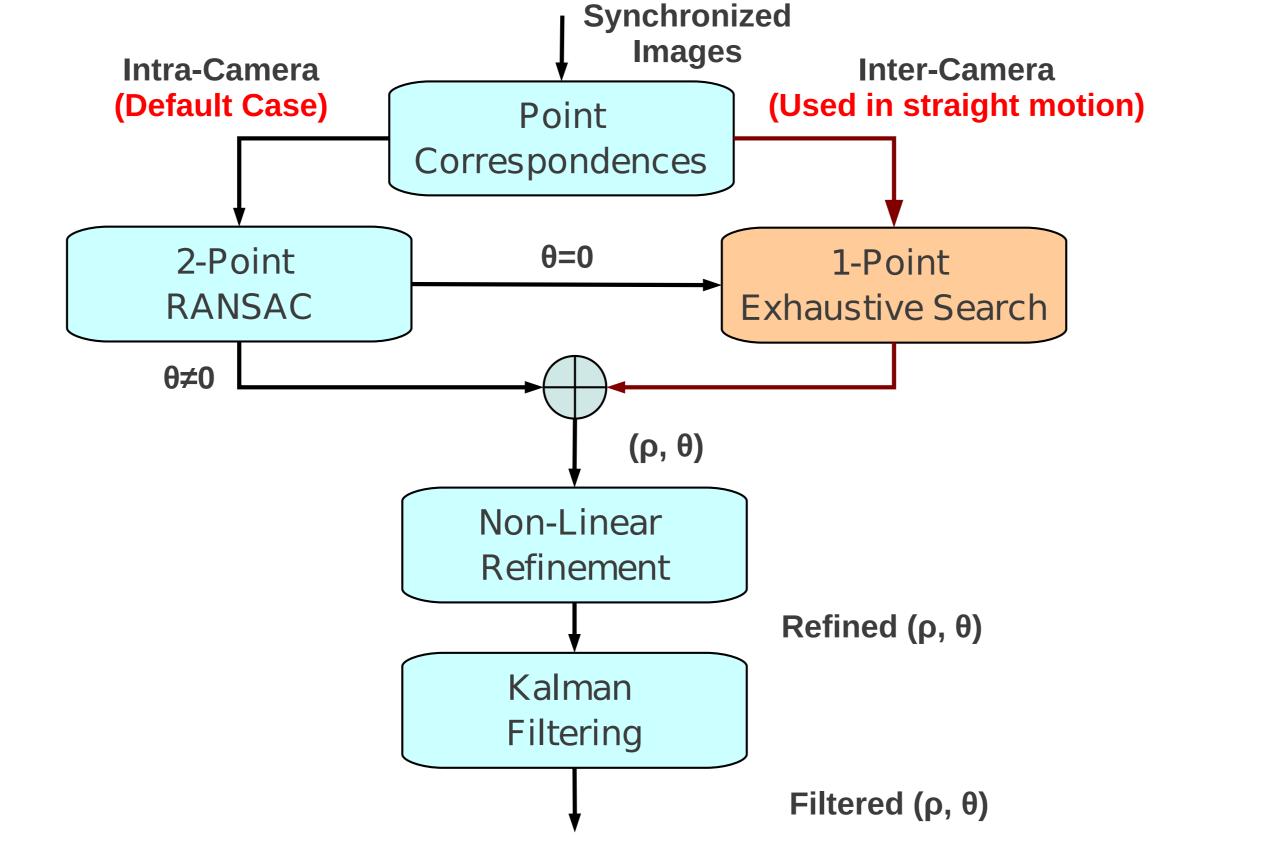
• Generalized epipolar constraint defined in [1].

$$\mathbf{l'}^T \begin{bmatrix} E & R \\ R & 0 \end{bmatrix} \mathbf{l} = 0 \tag{1}$$

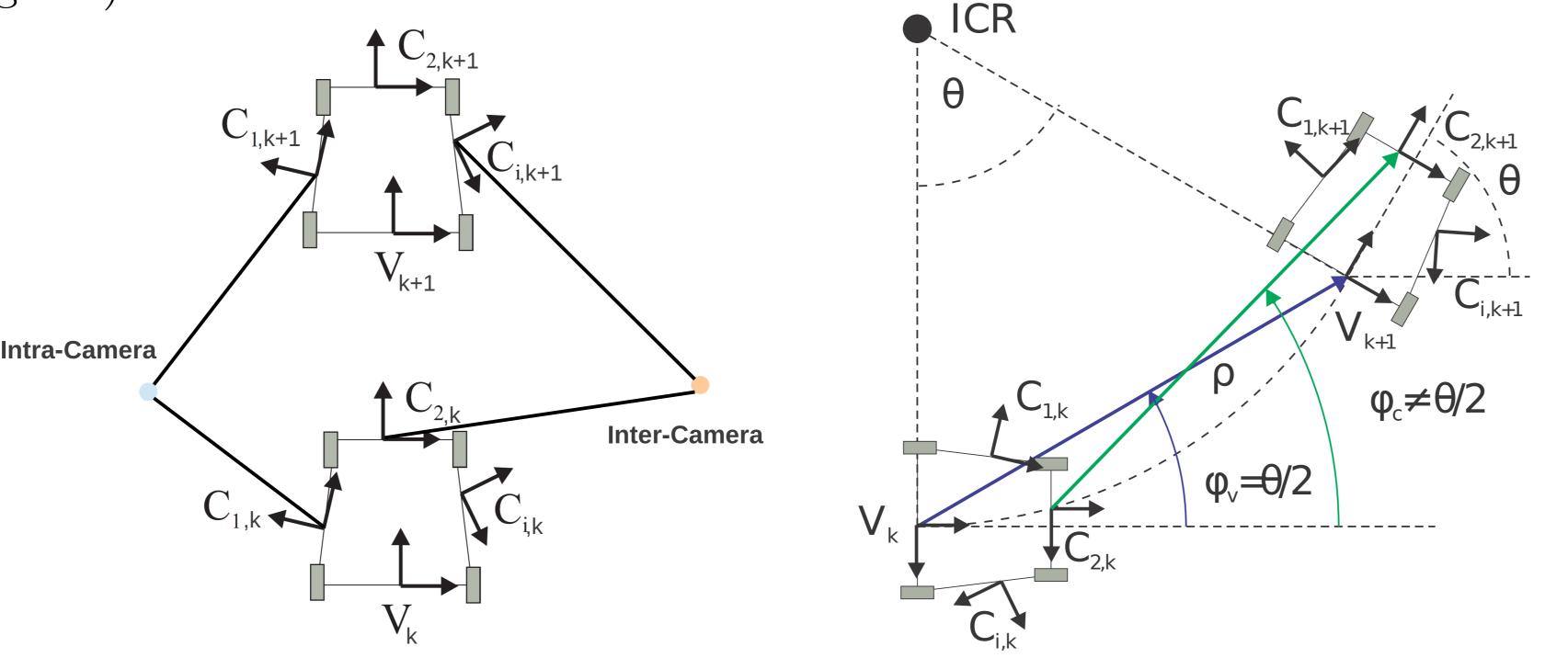
where E_{GC} is the generalized essential matrix. $l \leftrightarrow l'$ are point correspondences represented as 6-Vector Plücker lines. $E = \lfloor t \rfloor_{\mathbf{x}} R$ is the conventional essential matrix. R and t are the rotation and translation of the relative motion.

Motion Estimation

• System overview for motion estimation with the generalized camera on a car.



- Inter-camera and intra-camera point correspondences (left diagram).
- Relation between generalized camera and Ackermann motion (right diagram).



• Ackermann motion is parameterized by the relative yaw angle θ and scale ρ of the relative translation.

$$R = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad t = \rho \begin{bmatrix} \cos \varphi_v \\ \sin \varphi_v \\ 0 \end{bmatrix}$$
 (2)

• Putting Equation 2 into E_{GC} from Equation 1, we get the generalized essential matrix with Ackermann motion.

$$E_{GC} = \begin{bmatrix} 0 & 0 & \rho \sin\frac{\theta}{2} & \cos\theta & -\sin\theta & 0\\ 0 & 0 & -\rho \cos\frac{\theta}{2} & \sin\theta & \cos\theta & 0\\ \rho \sin\frac{\theta}{2} & \rho \cos\frac{\theta}{2} & 0 & 0 & 0\\ \cos\theta & -\sin\theta & 0 & 0 & 0 & 0\\ \sin\theta & \cos\theta & 0 & 0 & 0 & 0\\ 0 & 0 & 1 & 0 & 0 & 0 \end{bmatrix}$$
(3)

2-Point Minimal Solution

• Putting Equation 3 into Equation 1, we get the generalized epipolar constraint with Ackermann motion.

$$a\cos\theta + b\sin\theta + c\rho\cos\frac{\theta}{2} + d\rho\sin\frac{\theta}{2} + e = 0$$
 (4)

a, b, c, d and e are coefficients computed from the Plücker line.

- 2-point correspondences are needed to solve for the 2 unknowns θ and ρ from Equation 4.
- This leads to the minimal solution of 3-degree polynomial with $\gamma = \sin^2 \frac{\theta}{2}$ where the yaw angle θ can be solved in closed-form.

$$A\gamma^3 + B\gamma^2 + C\gamma + D = 0 \tag{5}$$

A, B, C and D are the coefficients computed from the Plücker line correspondence $l \leftrightarrow l'$.

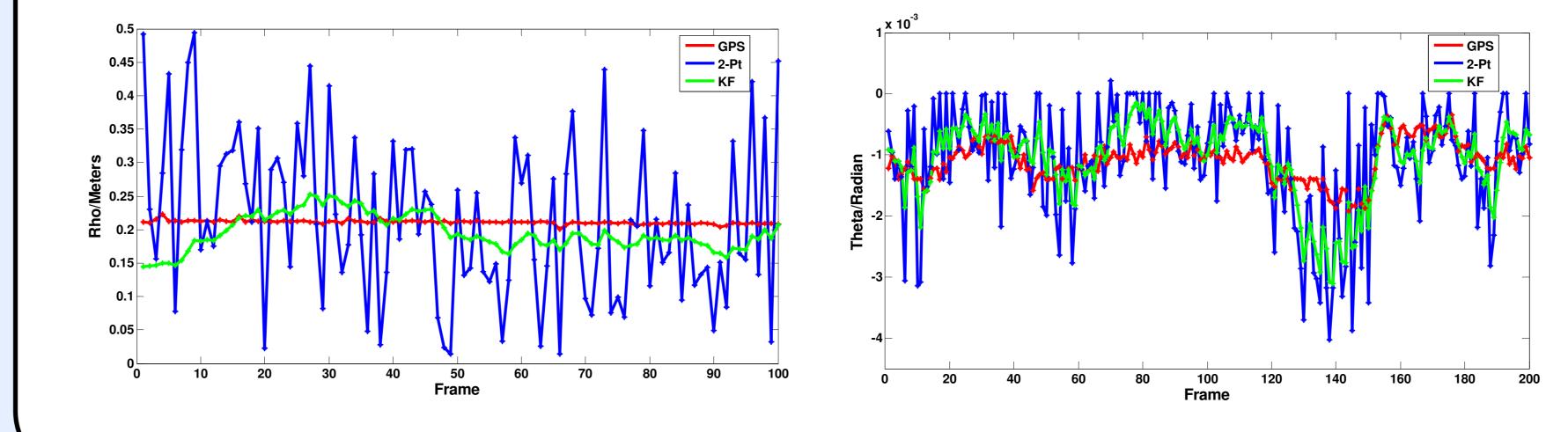
• The metric scale ρ can be solved by back-substitution.

Degenerated Case and Its Solution

- More intra-camera than inter-camera correspondences \Rightarrow intra-camera correspondences as the default case.
- Metric scale ρ cannot be uniquely computed when the car goes straight, i.e. $\theta = 0$ (when using intra-camera tracks only).
- **Solution**: Retrieve the metric scale with 1 additional inter-camera correspondence.

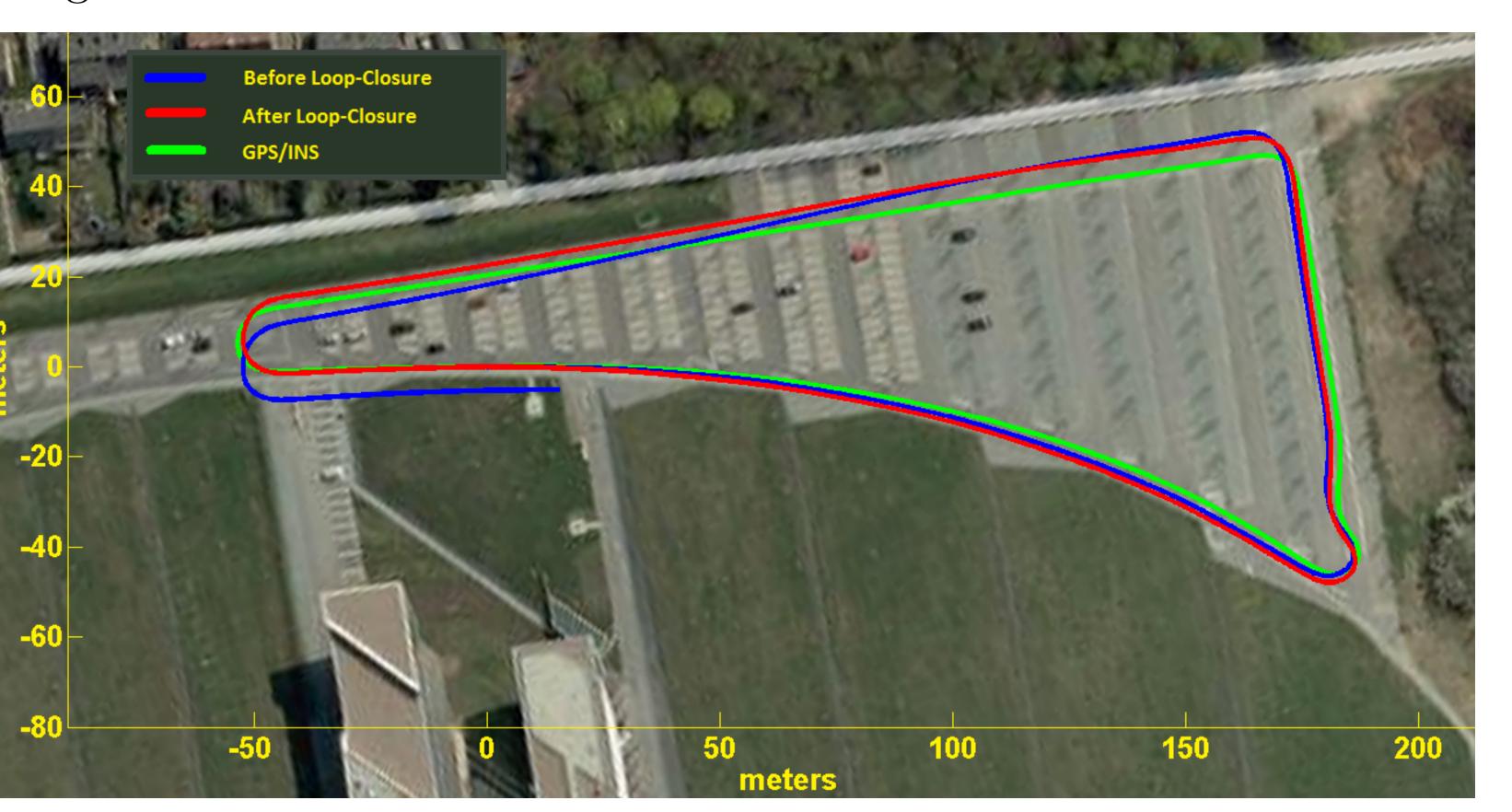
Kalman Filtering

- Two independent 1D Kalman filters with constant velocity prior to smooth out noisy estimates.
- Example of scales ρ (left) and yaw angles θ (right) between consecutive frames after Kalman filtering.



Results

- A total of 4×2500 images are used in the test of our algorithm.
- Additional 1 inter-camera correspondence used to compute scale for 79.9% when the car is moving straight.
- Trajectories before and after pose-graph loop-closure compared with GPS/ INS ground truth.



• Top view of trajectory and 3D map points after pose-graph loop-closure and full bundle adjustment compared with GPS/INS ground truth.



References / Acknowledgement

[1] R. Pless, Using many cameras as one, CVPR 2003.

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