

Moving object detection, tracking and following using an omnidirectional camera on a mobile robot

Ivan Marković¹, François Chaumette², Ivan Petrović¹

¹University of Zagreb, Faculty of Electrical Engineering and Computing, Croatia

²Inria Rennes-Bretagne Atlantique, Rennes, France

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University of Zagreb
Faculty of Electrical Engineering
and Computing



Centre of Research Excellence
for Advanced Cooperative Systems

- ① Introduction
- ② Detecting motion
 - Image formation
 - Detection of moving objects on the sphere
- ③ Tracking on the unit sphere
 - Recursive Bayesian tracking
 - von Mises-Fisher filter
- ④ Following via visual servoing
- ⑤ Conclusion

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- widely utilized sensor in navigation, SLAM, visual odometry

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- widely utilized sensor in navigation, SLAM, visual odometry
- additional cue in sensor fusion
- robust moving object detection and tracking

① Introduction

② Detecting motion

Image formation

Detection of moving objects on the sphere

③ Tracking on the unit sphere

Recursive Bayesian tracking

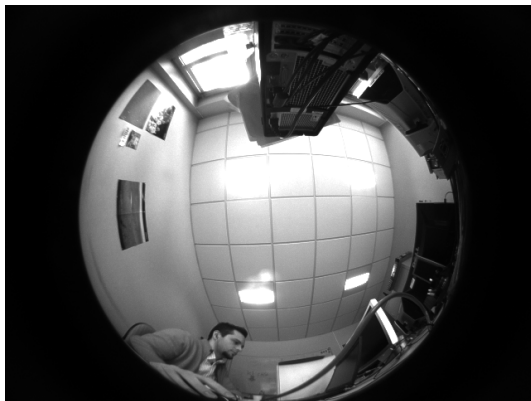
von Mises-Fisher filter

④ Following via visual servoing

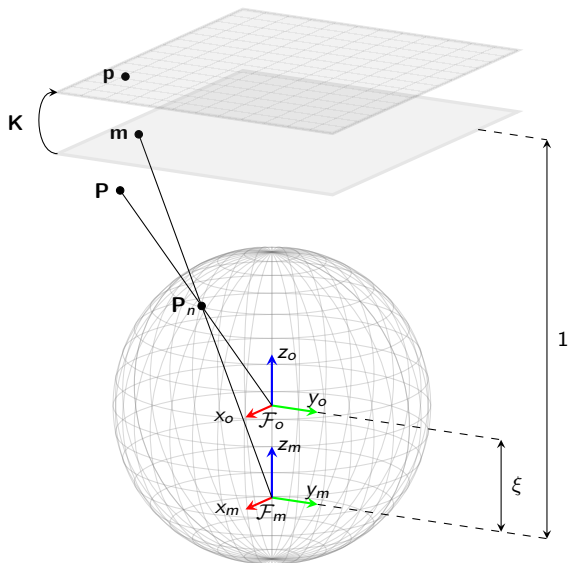
⑤ Conclusion

Omnidirectional image

- spherical projection model
[Geyer and Daniilidis, 2000, Barreto and Araújo, 2001]
- unifies image formation in central catadioptric systems and (in practice) fish-eye lenses [Ying and Hu, 2004]



Spherical projection model



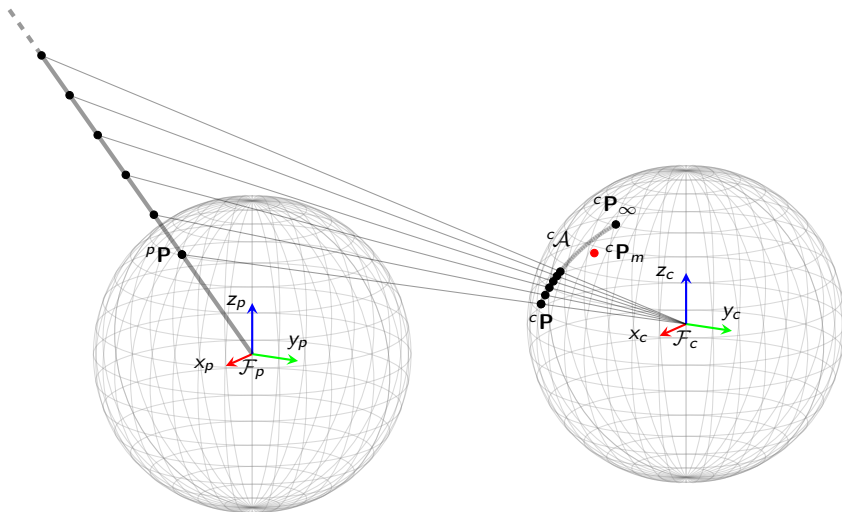
- method proposed in [Mei and Rives, 2007]
- from pixel to a point on the sphere

$$\mathbf{m} = \mathbf{K}^{-1}\mathbf{p}$$

$$\mathbf{P}_n = \begin{bmatrix} \frac{\xi + \sqrt{1 + (1 - \xi^2)(x^2 + y^2)}}{x^2 + y^2 + 1} x \\ \frac{\xi + \sqrt{1 + (1 - \xi^2)(x^2 + y^2)}}{x^2 + y^2 + 1} y \\ \frac{\xi + \sqrt{1 + (1 - \xi^2)(x^2 + y^2)}}{x^2 + y^2 + 1} - \xi \end{bmatrix}$$

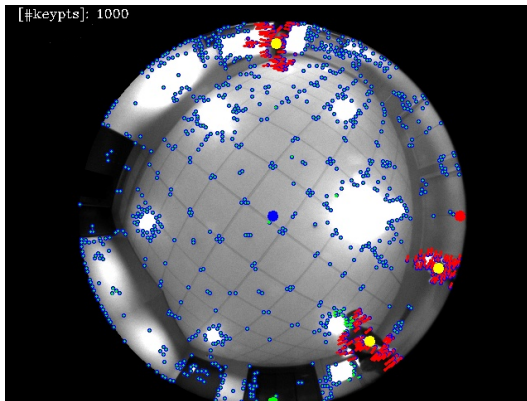
- optical flow is a confluence of camera motion, independent object motion, and the 3D structure of the scene [Palmer, 1999]
- corner detection and pyramidal Lucas-Kanade algorithm (sparse optical flow) [Bouguet, 2000]
- discrimination based on odometry (no depth information)

Camera displacement

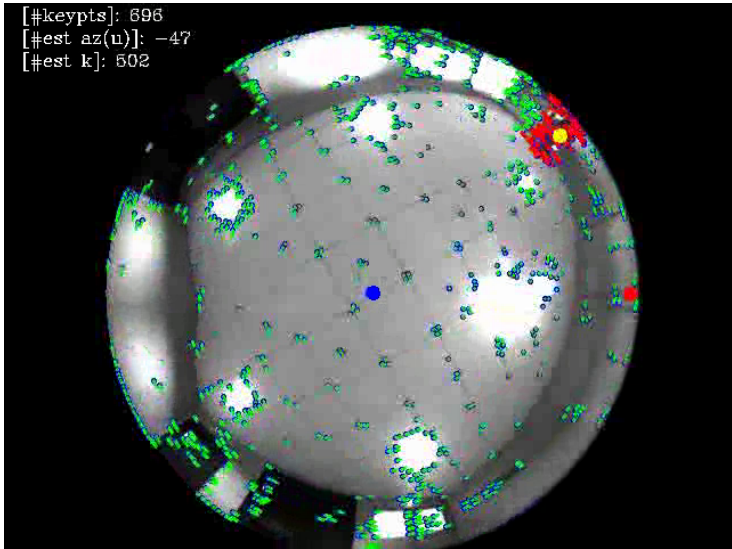


Flow vector clustering

- disjoint-set union find algorithm
- each cluster yields a point on the sphere



[#keypts]: 696
[#est az(u)]: -47
[#est k]: 502



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- estimate $p(\mathbf{x}_k | \mathbf{z}_{1:k})$ —sequence of prediction-correction steps

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- **prediction** via

$$p(\mathbf{x}_k|\mathbf{z}_{1:k-1}) = \int p(\mathbf{x}_k|\mathbf{x}_{k-1})p(\mathbf{x}_{k-1}|\mathbf{z}_{1:k-1}) d\mathbf{x}_{k-1}$$

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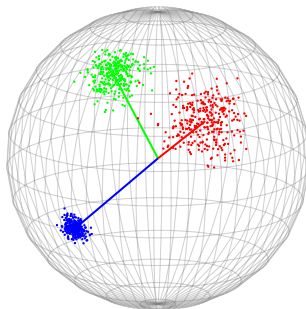
- **correction** via Bayes rule

$$p(\mathbf{x}_k|\mathbf{z}_{1:k}) = \frac{p(\mathbf{z}_k|\mathbf{x}_k)p(\mathbf{x}_k|\mathbf{z}_{1:k-1})}{p(\mathbf{z}_k|\mathbf{z}_{1:k-1})}$$

- a distribution on the unit sphere

$$p(\mathbf{x}; \kappa, \boldsymbol{\mu}) = \frac{\kappa}{4\pi \sinh \kappa} \exp \left(\kappa \boldsymbol{\mu}^T \mathbf{x} \right),$$

where $\boldsymbol{\mu}$ is the mean direction vector and κ is the concentration parameter



- **prediction**—solve the integral

$$\boldsymbol{\mu}_{k|k-1} = \boldsymbol{\mu}_{k-1},$$

$$\kappa_{k|k-1} = A^{-1}(A(\kappa_{k-1})A(\kappa_Q)), \quad A(\kappa) = \frac{1}{\tanh \kappa} - \frac{1}{\kappa}$$

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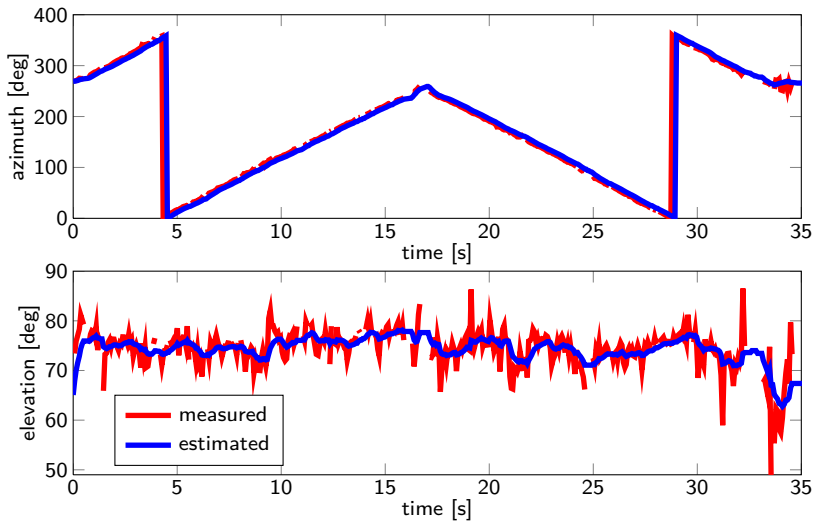
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- **correction**—calculate the Bayes rule

$$\boldsymbol{\mu}_{k|k} = \frac{\kappa_{k|k-1}\boldsymbol{\mu}_{k|k-1} + \kappa_R\mathbf{z}_k}{\kappa_{k|k}}$$

$$\kappa_{k|k} = \sqrt{\kappa_{k|k-1}^2 + \kappa_R^2 + 2\kappa_{k|k-1}\kappa_R(\boldsymbol{\mu}_{k|k-1} \cdot \mathbf{z}_k)}$$



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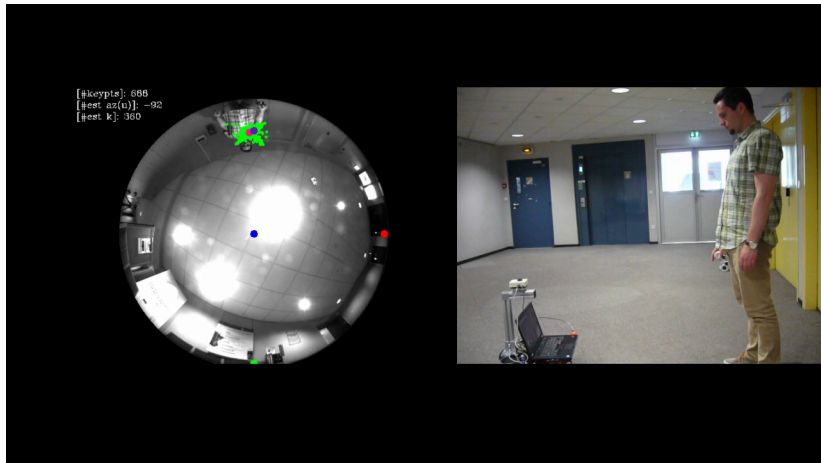
- cylindrical coordinates in the image
[Iwatsuki and Okiyama, 2005, Fomena, 2008]

$$\begin{bmatrix} \dot{\rho} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \frac{-\cos \theta}{P_z} & 0 \\ \frac{-\sin \theta}{\rho P_z} & -1 \end{bmatrix} \begin{bmatrix} v \\ \omega \end{bmatrix} = \mathbf{L}_s \begin{bmatrix} v \\ \omega \end{bmatrix}$$

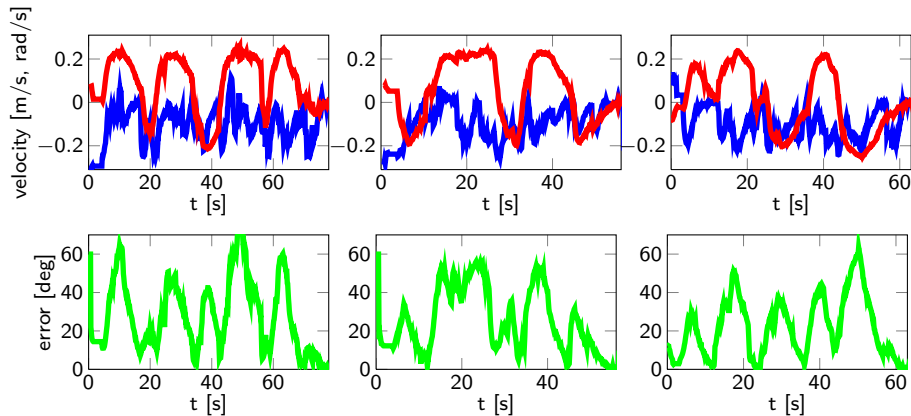
- control law

$$\begin{bmatrix} v \\ \omega \end{bmatrix} = -\lambda \hat{\mathbf{L}}_s^{-1} \begin{bmatrix} \rho - \rho^* \\ \theta - \theta^* \end{bmatrix}$$

Following via visual servoing



Experiments

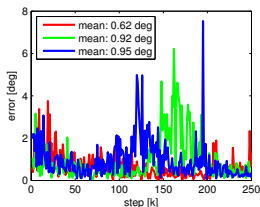
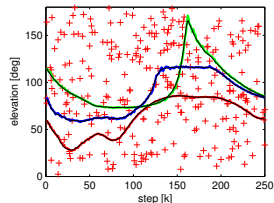
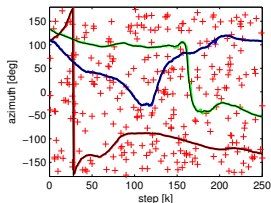
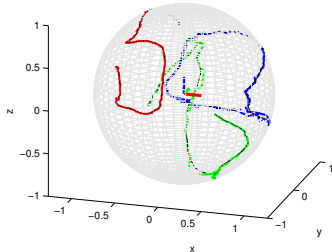


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- detection and tracking of moving objects in omnidirectional images
- spherical representation — von Mises-Fisher tracker
- following via visual servoing

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Thank you for your attention!



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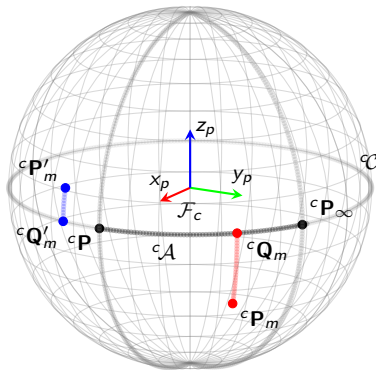


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Distance to great circle arc



Distance to great circle arc

- great circle arc defined by ${}^cP, {}^cP_\infty$
- find closest point to matched cP_m on the great circle

$$\begin{aligned}P' &= {}^cP_m - ({}^cP_m \cdot n)n, \quad n = {}^cP_m \times {}^cP_\infty \\ {}^cQ_m &= \frac{P'}{|P'|}\end{aligned}$$

- check if in the lune $({}^cP, {}^cP_\infty)$

$$\begin{aligned}({}^cP \times {}^cQ_m) \cdot ({}^cQ_m \times {}^cP_\infty) &> 0 \&\& \\ ({}^cP \times {}^cQ_m) \cdot ({}^cP \times {}^cP_\infty) &> 0\end{aligned}$$

- great circle distance

$$d(P, Q) = \arccos(P \cdot Q)$$