

On the Two-View Geometry of



Unsynchronized Cameras

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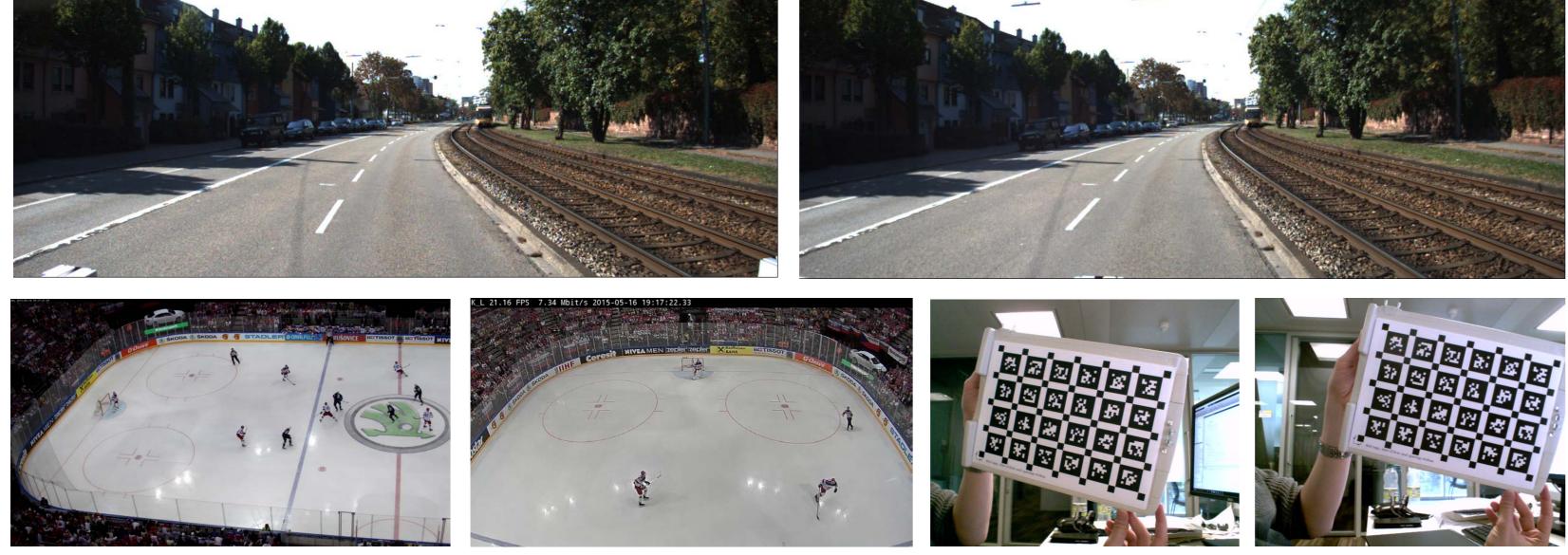
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WHAT?

- Minimal solvers for two-view geometry with a time shift
- Iterative algorithm for robust space-time camera calibration
- Synchronizes sequences several seconds apart
- Achieves sub-frame synchronization

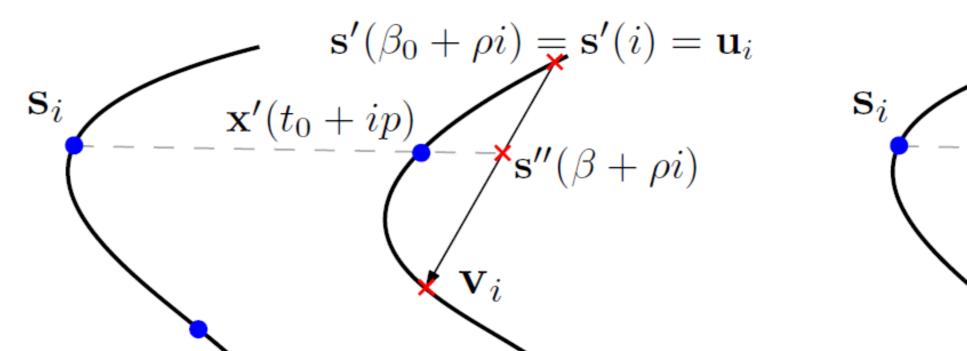
WHY?

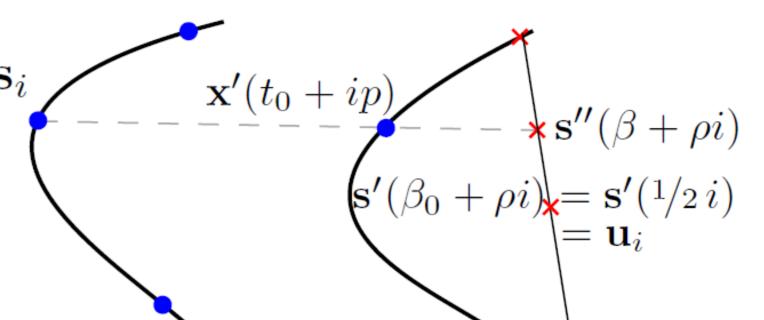




Linearized trajectory

• To solve for F or H and time shift β we assume linear trajectory





- Multi-camera systems perform 3D reconstruction, tracking, localization etc.
- They require precise time and space calibration

Existing methods

• Require special content

• Sensitive to outliers and initialization • Robust

• Efficient • Exhaustive and time-consuming

Our method

- Works with general data
- $\mathbf{x}(t)$ $\mathbf{x}(t)$ $\mathbf{s}'(\beta + \rho i) \approx \mathbf{s}'(\beta_0 + \rho i) + (\beta - \beta_0)\mathbf{v} = \mathbf{s}''(\beta + \rho i)$ $\mathbf{v} = \mathbf{s}'_{j_0+d} - \mathbf{s}'_{j_0}$ **Epipolar geometry** $\mathbf{s}'(\beta + \rho i)^{\top} \mathbf{F} \mathbf{s}_i = 0 \implies (\mathbf{u}_i + \beta \mathbf{v}_i)^{\top} \mathbf{F} \mathbf{s}_i = 0$ Homography
 - $\mathbf{H}\mathbf{s}_i = \lambda_i \left(\mathbf{u}_i + \beta \mathbf{v}_i\right)$

MINIMAL SOLUTIONS

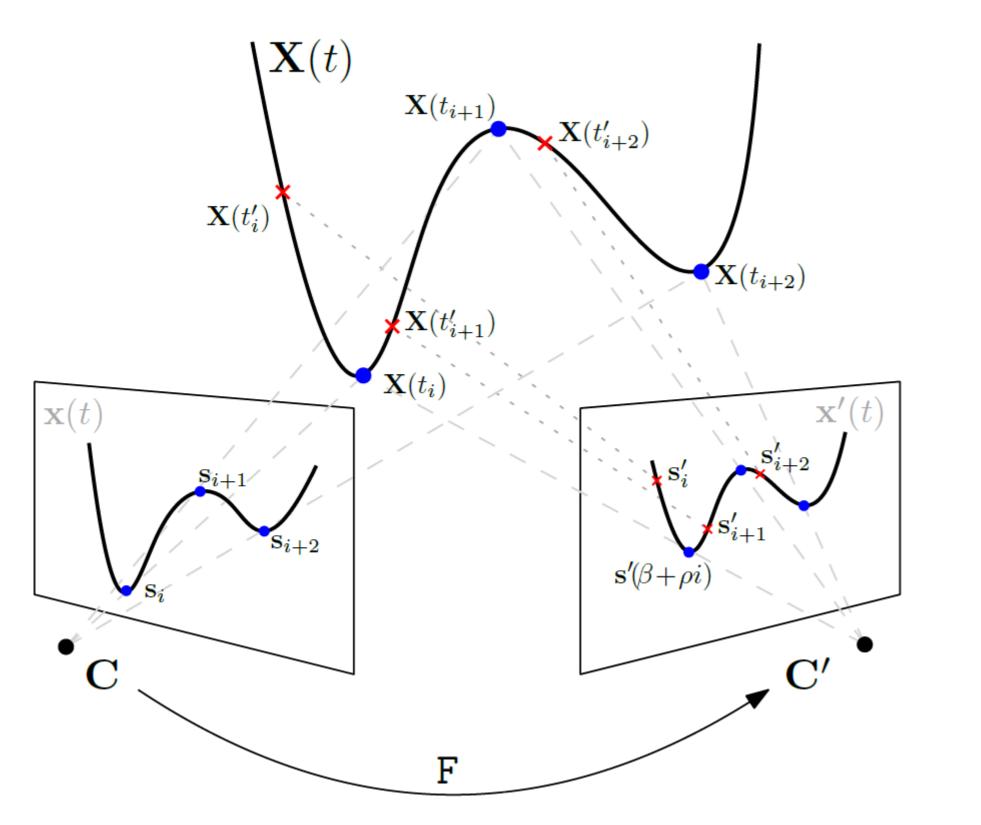
Epipolar geometry

•
$$(\mathbf{u}_i + \beta \mathbf{v}_i)^\top \mathbf{F} \mathbf{s}_i = 0$$
 can be written as $\mathbf{M} \mathbf{w} = \mathbf{0}$

HOW?

Image trajectory approximation

• Cameras (in a rig) sample a moving object with trajectory $\mathbf{X}(t'_i)$ (or moving camera rig+static object) as samples s_i and s'_j



• The object trajectory in the second image $\mathbf{x}'(t'_i)$ is approximated

 $\mathbf{w} = [f_{11}, f_{12}, f_{13}, f_{21}, f_{22}, f_{23}, f_{31}, f_{32}, f_{33}, \beta f_{11}, \beta f_{12}, \beta f_{13}, \beta f_{21}, \beta f_{22}, \beta f_{23}]$

- The minimal number of samples 8
- M is a 8x15 matrix of rank 8
- F defined up to scale, write w using the null space of M $\mathbf{w} = \mathbf{n}_0 + \sum_{i=1}^6 \alpha_i \mathbf{n}_i$
- Additionally the elements of w satisfy

 $\beta w_j = w_k \text{ for } (j,k) \in \{(1,10), ..., (6,15)\}$

- Quite complicated system of 7 polynomial equations in 7 unknowns (solver size 633x649)
- SOLUTION eliminate β first using elimination ideal method [1] and then solve by automatic generator [2]

• Solver size 194x210 with 16 solutions





Homography

- $[\mathbf{u}_i + \beta \mathbf{v}_i]_{\times} \operatorname{Hs}_i = \mathbf{0} 3$ equations 2 linearly independent
- The minimal number of samples 4.5 (5)
- Equations corresponding to the first and second row in a matrix form

 $M \mathbf{w} = \mathbf{0}$

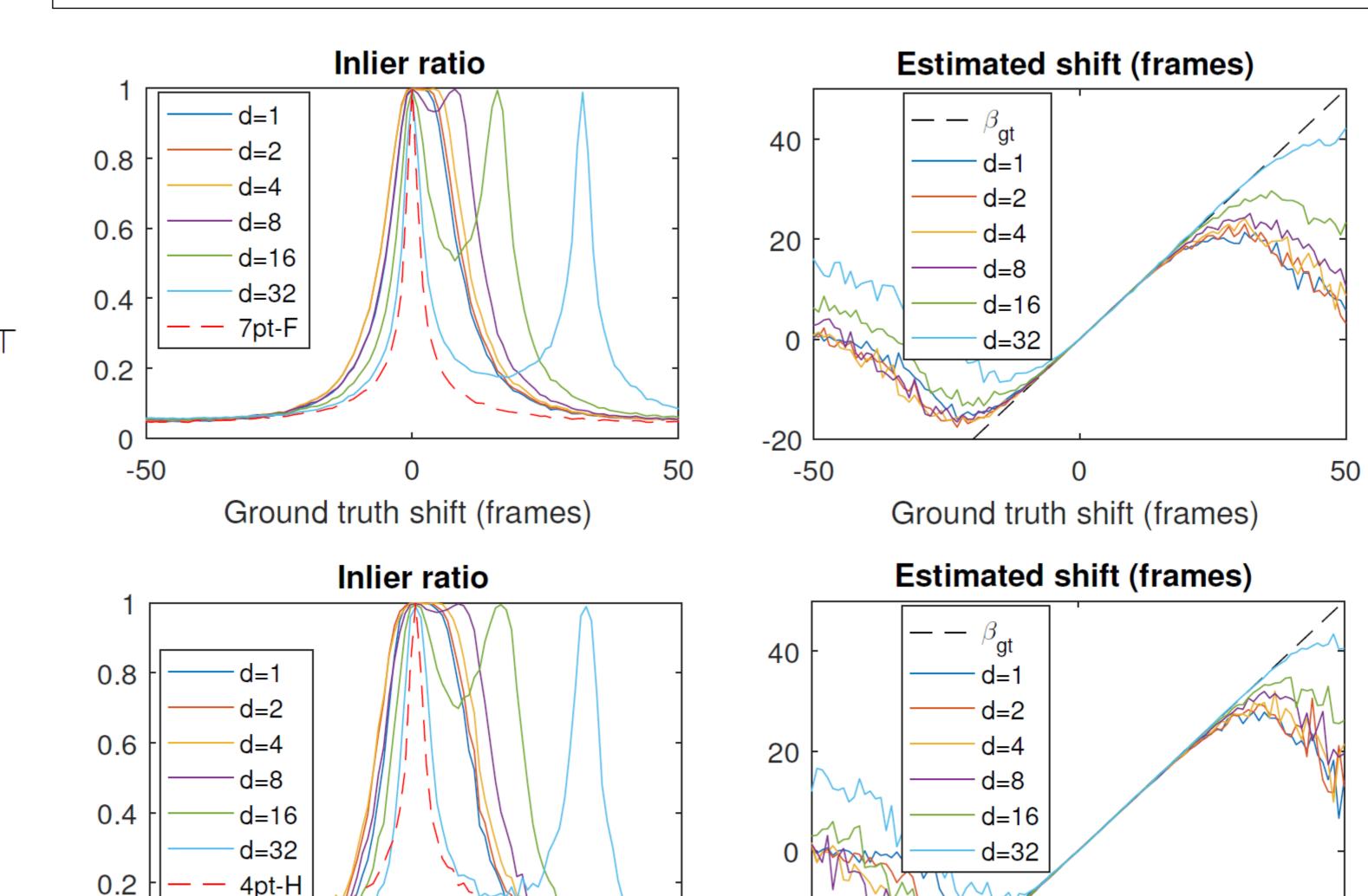
- $\mathbf{w} = [h_{11}, h_{12}, h_{13}, h_{21}, h_{22}, h_{23}, h_{31}, h_{32}, h_{33}, \beta h_{31}, \beta h_{32}, \beta h_{33}]^{\top}$
- M is a 9x12 matrix of rank 9 3dim null space

$$\mathbf{w} = \sum_{i=1}^{3} \gamma_i \mathbf{n}$$

- H is defined up to scale \implies we can fix $\gamma_3=1$
- The elements of w satisfy

 $w_{10} = \beta w_7, w_{11} = \beta w_8, w_{12} = \beta w_9$

- System of 3 quadratic equations in 3 unknowns and 6 monomials
- Solved by G-J elimination of the 3x6 coefficient matrix and



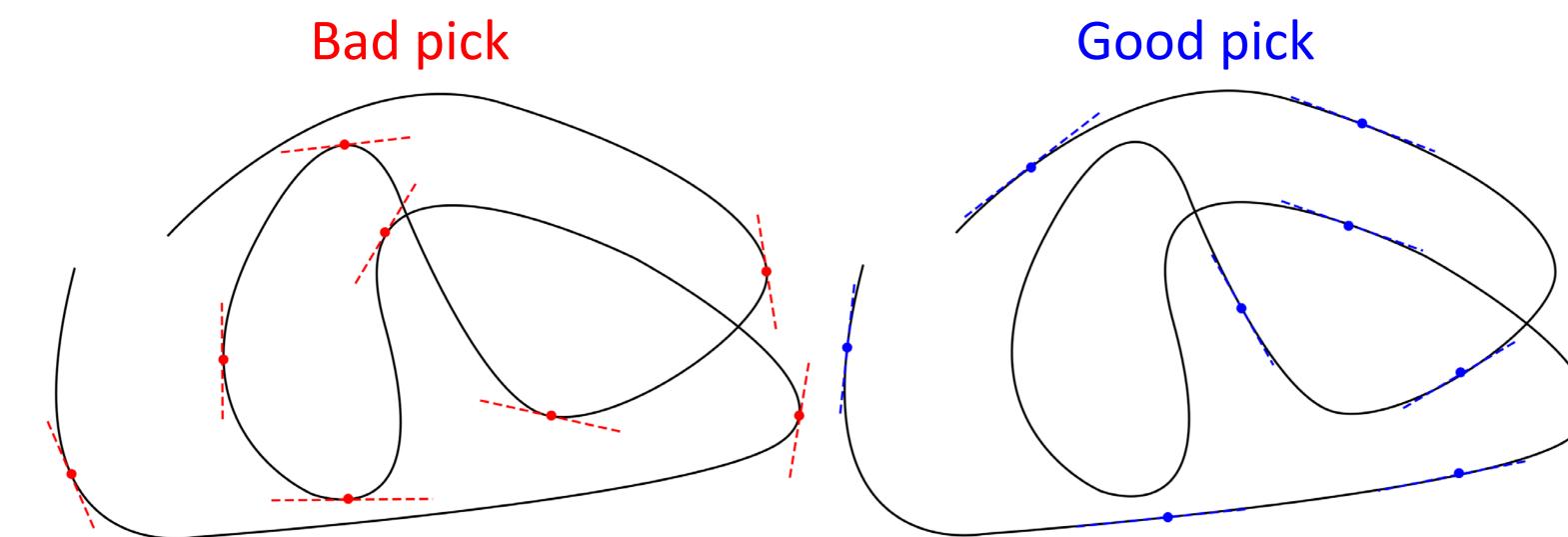
RESULTS

omography

computing eigenvalues of a 3x3 matrix

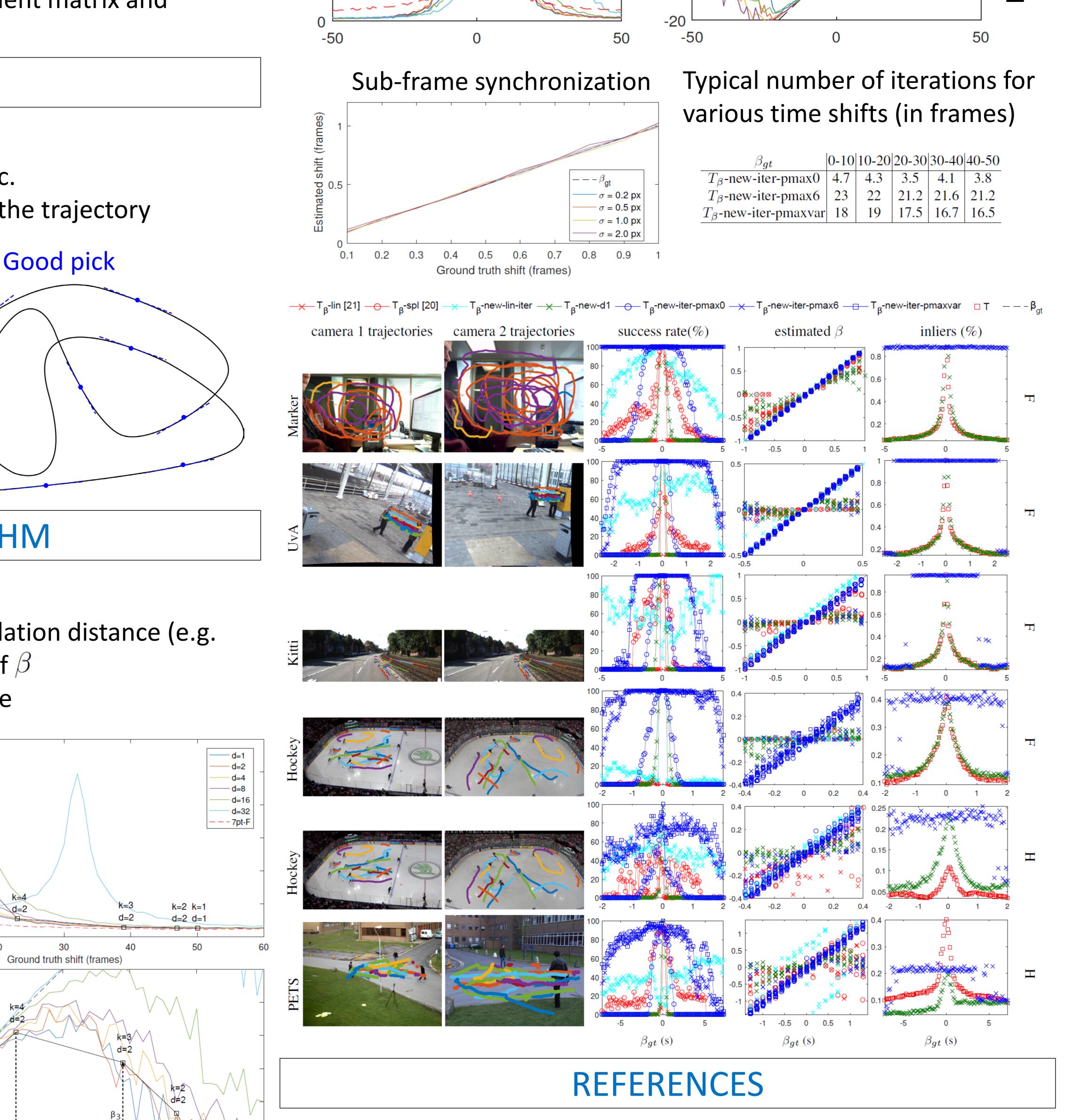
USING RANSAC

- RANSAC use has 2 reasons:
 - Outliers in the data, mismatches etc.
 - Outliers due to the linearization of the trajectory



ITERATIVE ALGORITHM

- d=32



- For larger time shifts (multiple frames)
- When time shift is large, increasing interpolation distance (e.g. next 2^k-th sample) improves the estimate of β
- Start with interpolation distance d = 1 frame
- LOOP

RANSAC F or H and β IF no improvement in inliers increase d (e.g. d = 2^k) ELSE

Correct the shift by β END IF END LOOP

 Algorithm stops when all interpolation distances up to 2^{kmax} have failed to provided more inliers at

[1] Z. Kukelova, J. Kileel, B. Sturmfels, and T. Pajdla. A clever elimination strategy for efficient minimal solvers.

