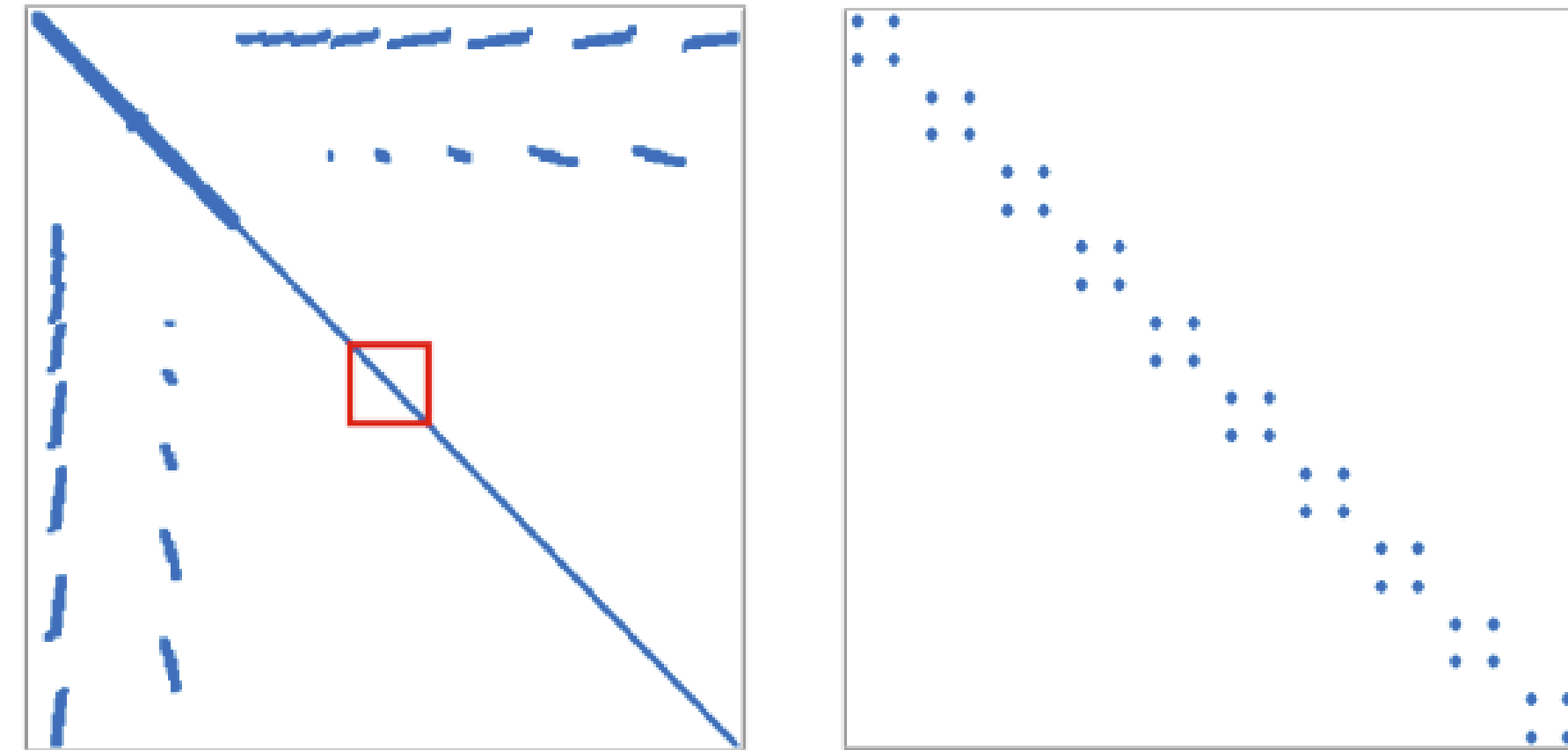




Summary

We tackle the problem of **simultaneous refining the camera orientations and scene map** for a **purely rotating event camera**. We formulate the problem as **regularized non-linear least squares (NLLS)** optimization. The loss function is defined using the **linearized event generation model** in the camera orientations and the panoramic gradient map of the scene.

Problem sparsity (Normal Eqs)



$$\begin{pmatrix} A_{11} & A_{12} \\ A_{12}^T & A_{22} \end{pmatrix} \begin{pmatrix} \Delta P_{\alpha}^* \\ \Delta P_{\beta}^* \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$$

Quantitative Results

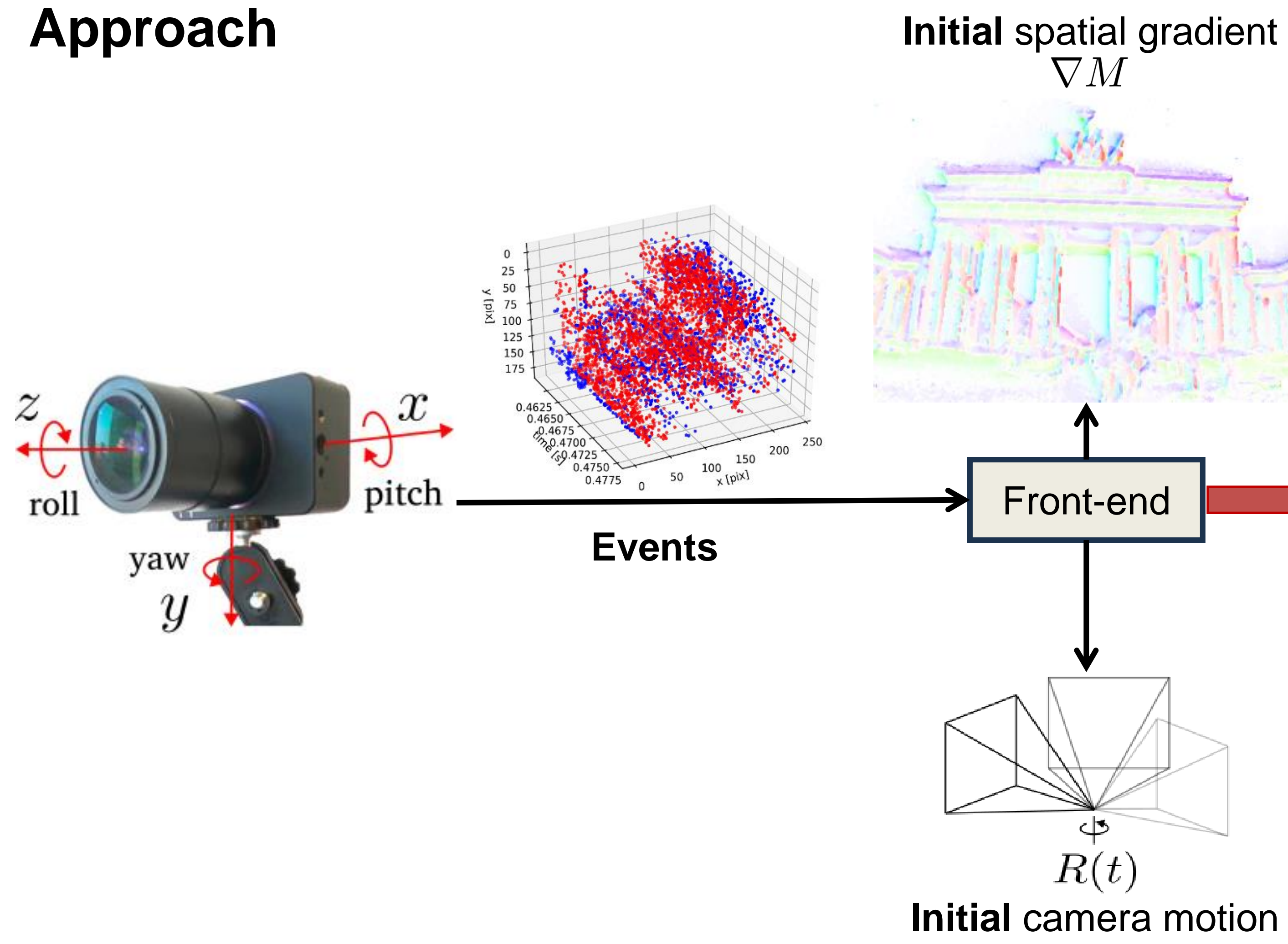
- Rotation RMSE [deg]

| Front-end | playroom | bicycle | city | street | town | bay |
|----------------|----------|---------|------|--------|------|------|
| EKF-SMT | 5.86 | 1.47 | 1.69 | 3.44 | 4.32 | 2.50 |
| | 6.09 | 1.18 | 1.68 | 3.46 | 4.40 | 2.41 |
| CMax-GAE | 4.63 | 1.65 | — | — | 4.66 | — |
| | 4.42 | 1.50 | N/A | N/A | 4.53 | N/A |
| CMax- ω | 3.22 | 1.69 | 1.53 | 0.97 | 1.91 | 1.80 |
| | 2.86 | 0.92 | 0.97 | 0.74 | 0.86 | 1.41 |

- Squared Event-based Photometric error [$\cdot 10^6$]

| Front-end | playroom | bicycle | city | street | town | bay |
|----------------|----------|---------|------|--------|------|------|
| EKF-SMT | 0.35 | 0.52 | 2.62 | 1.82 | 1.88 | 2.26 |
| | 0.23 | 0.30 | 2.13 | 1.52 | 1.51 | 1.96 |
| CMax-GAE | 0.35 | 0.53 | — | — | 1.90 | — |
| | 0.19 | 0.31 | N/A | N/A | 1.54 | N/A |
| CMax- ω | 0.33 | 0.55 | 2.71 | 1.90 | 1.92 | 2.30 |
| | 0.15 | 0.30 | 1.98 | 1.34 | 1.43 | 1.83 |

Approach



EMBA back-end

Methodology:

- We start from the Linearized Event Generation Model (LEGM):

$$\Delta L \approx -\nabla L(\mathbf{x}_k, t_k) \cdot \mathbf{v} \Delta t_k = s_k C$$

- The final objective function with L^2 map regularization:

$$\min_{\{R_i\}, \nabla M} \sum_{k=1}^{N_{\text{events}}} \left(\underbrace{\nabla M \cdot \mathbf{v}(\{R_i\}) \Delta t_k - s_k C_{\text{th}}}_{\text{LEGM predicts } \Delta L} \right)^2 + \underbrace{\eta \|\nabla M\|^2}_{\text{Regularizer}}$$

Photometric error (NLLS)

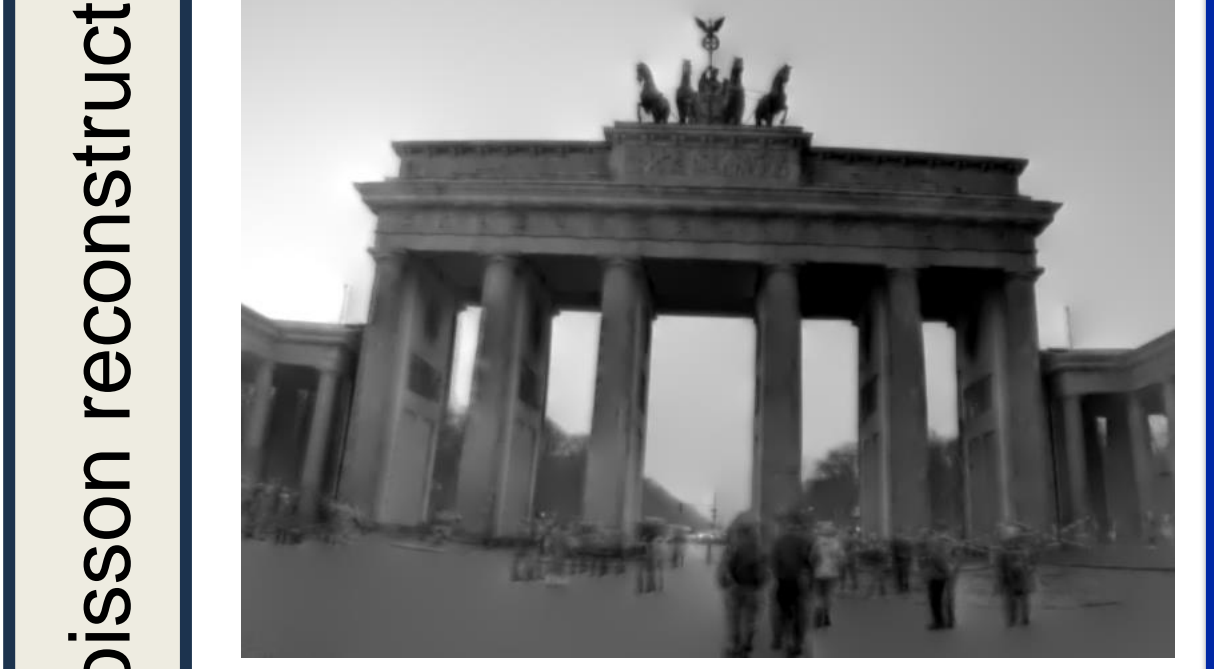
Key properties:

- Refines camera rotations and panoramic gradient map by minimizing the photometric error conveyed by all events.
- Leverages map regularization to improve iteration convergence.
- Exploits the sparsity of the Normal Equations for speed-up.
- Can be initialized by rotations estimated by various front-ends.
- Works well with various event camera resolutions, such as DAVIS, DVXplorer, and Prophesee's 1Megapixel camera.

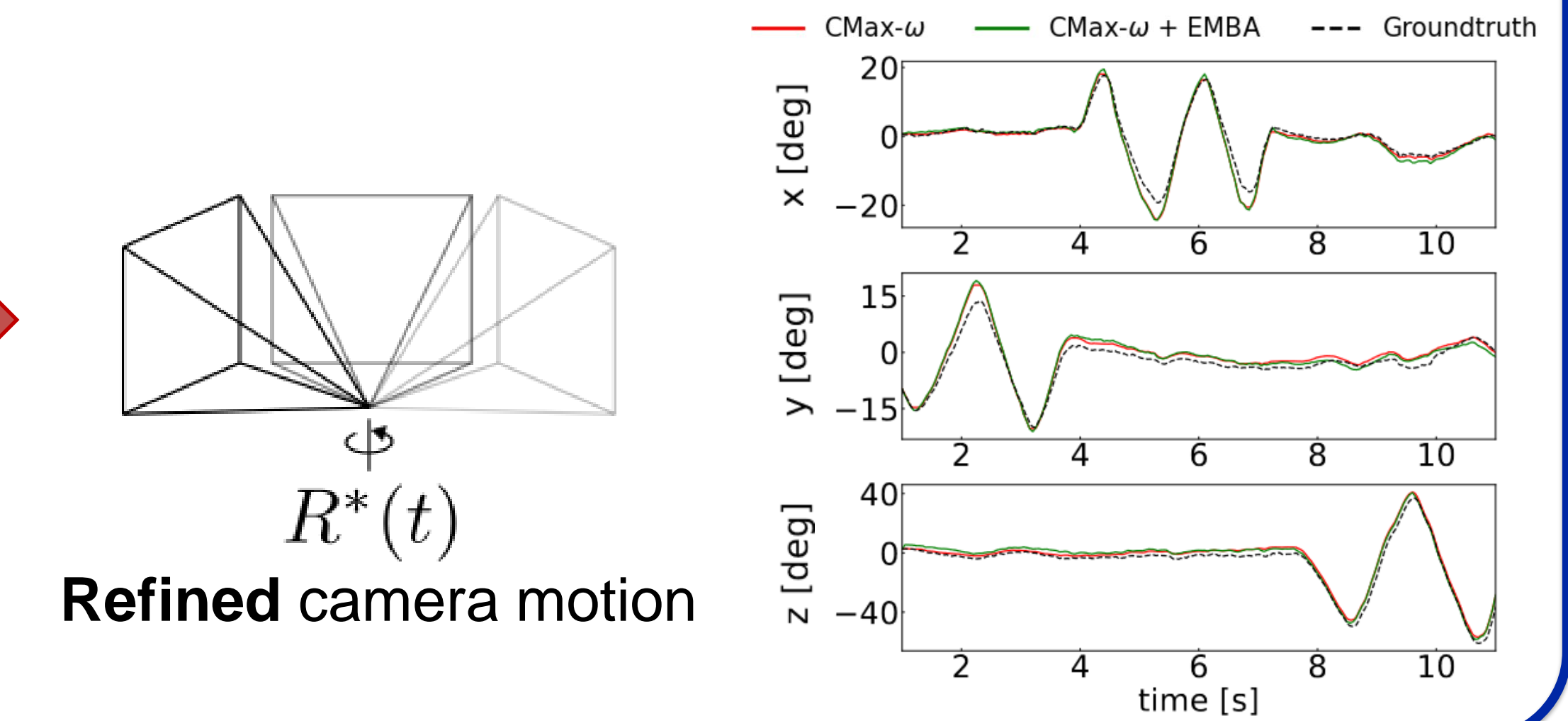
Refined spatial gradient



Brightness panorama



Poisson reconstruction

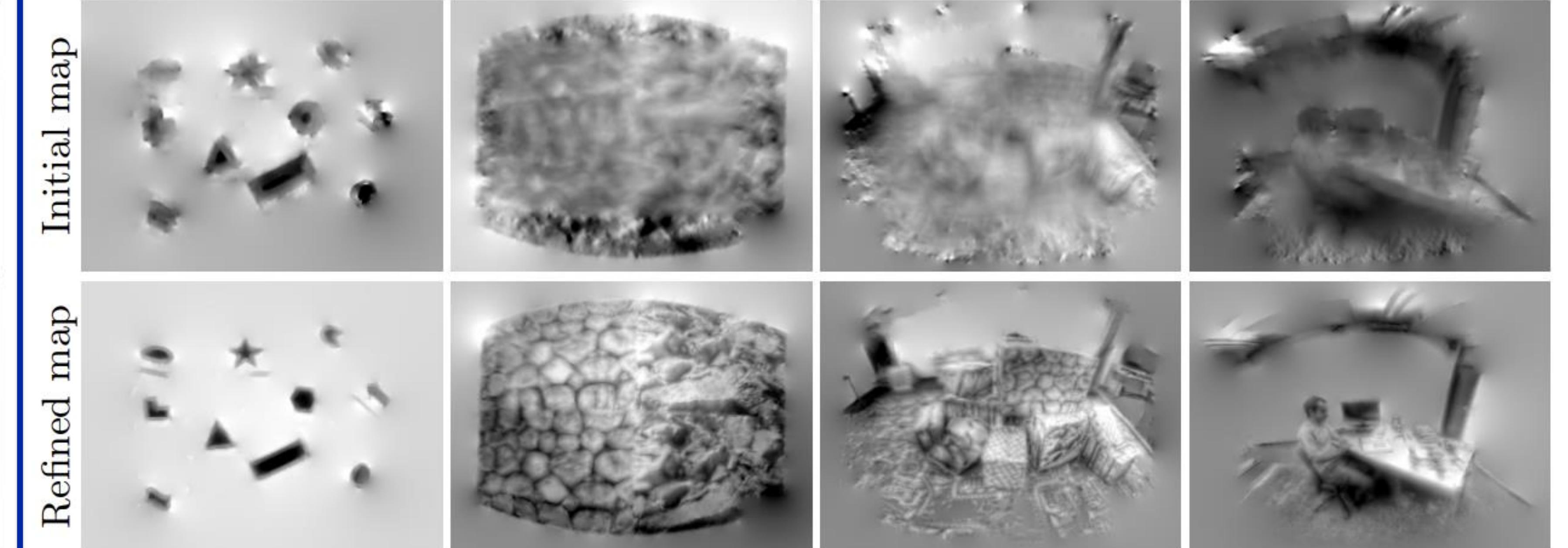


Qualitative Results

Synthetic data: Input 240 x 180 px. Output: 2048 x 1024 px (2K) map.



Real-world data: Input 240 x 180 px. Output: 1K map.



Prophesee's EVK4 data: Input 1280 x 720 px. Output: 4096 x 2048 px (4K) map.



iniVation's DVXplorer data: Input 640 x 480 px. Output: 4096 x 2048 px (4K) map.

