

# Towards Asynchronous SLAM with Event Cameras

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# Research @ Vision For Robotics Lab



## Robust Perception for Robotics:

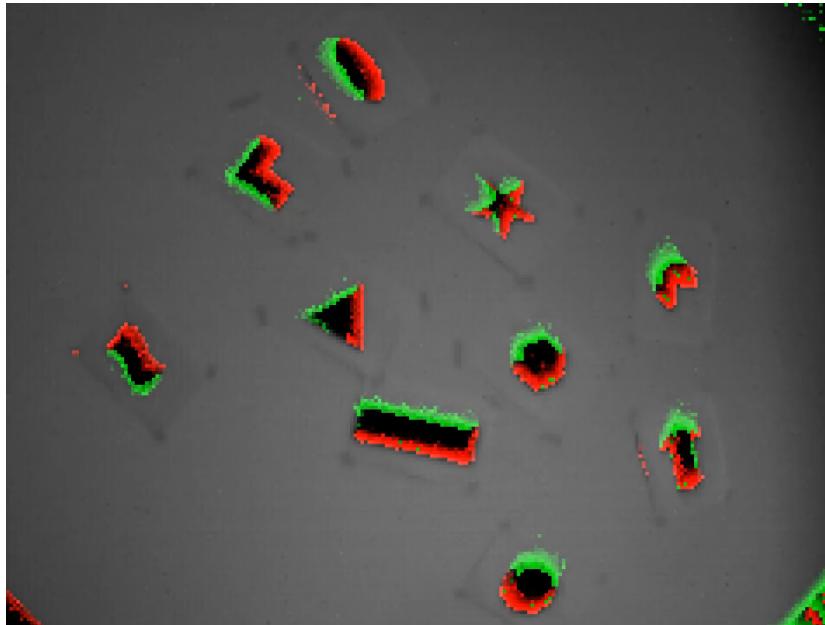
- Multi-Agent Visual SLAM
- Vision-based Navigation & Manipulation
- Viewpoint-tolerant Place Recognition



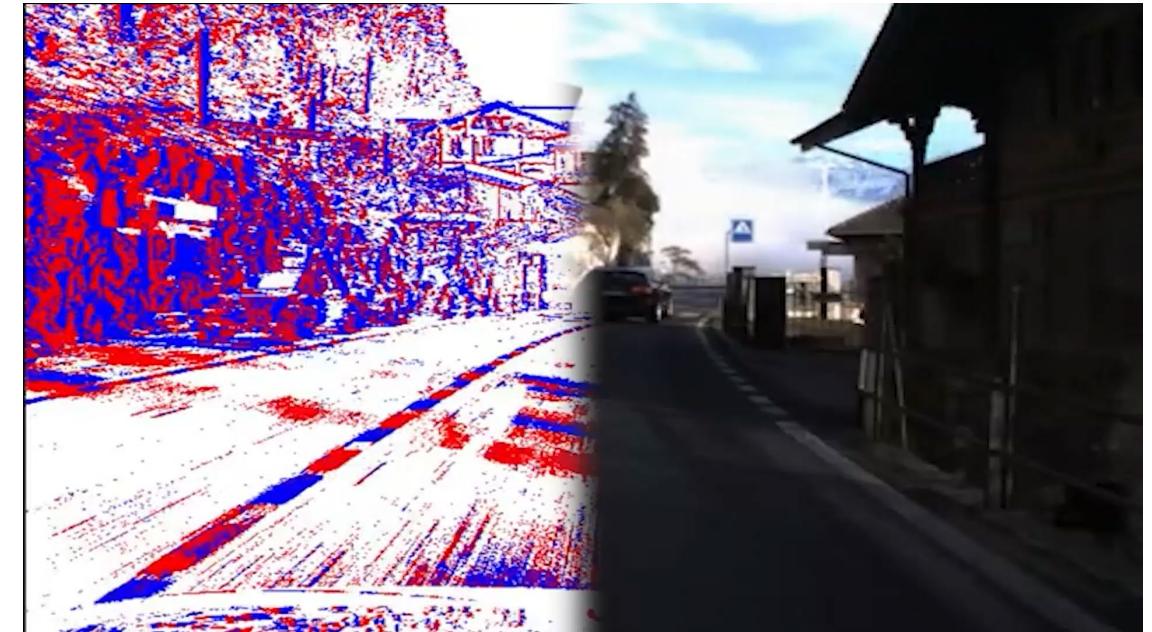
L'Agout photogrammetric 3D model

# Event Cameras in Robotics Application

Dataset\* [Mueggler et al., IJRR'17]



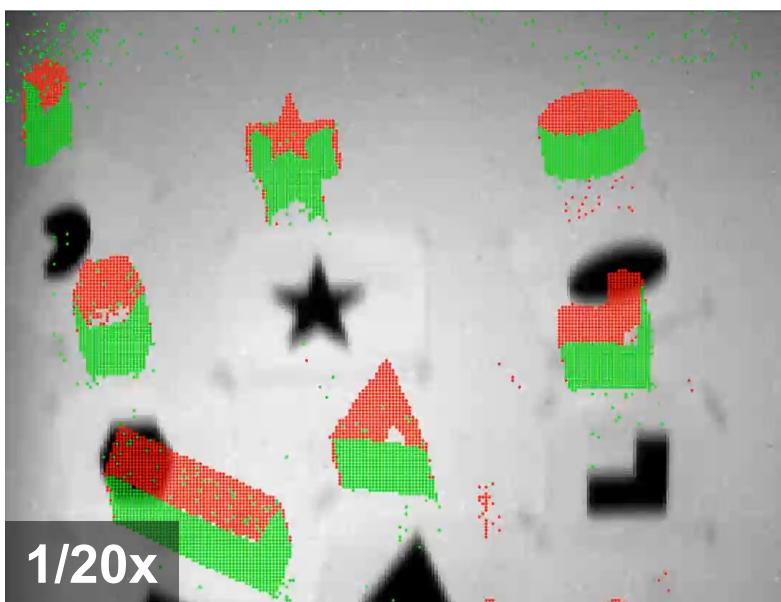
Dataset [Gehrig et al., RAL'21]



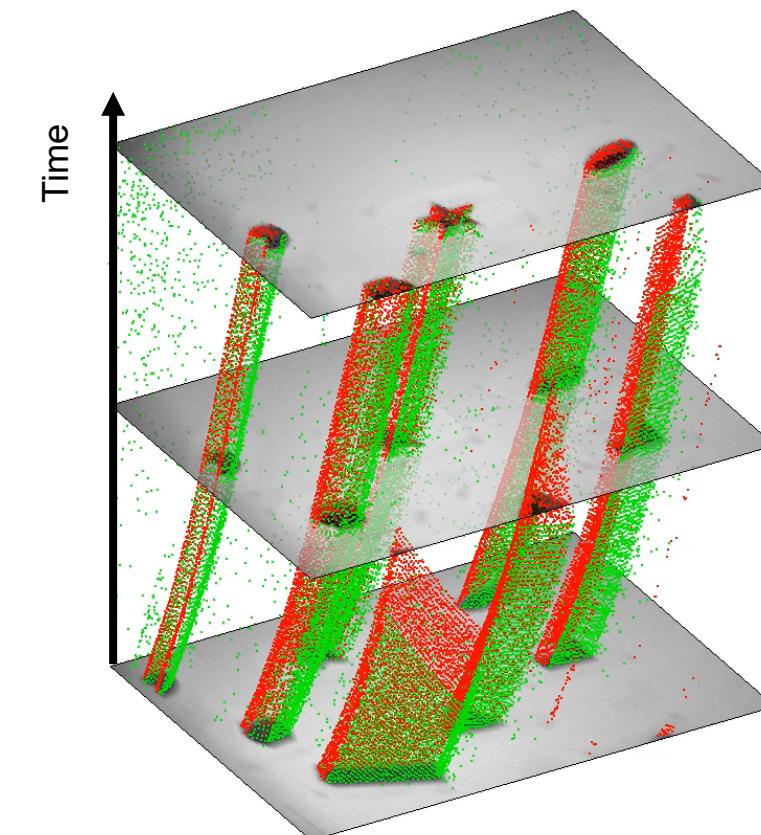
- Sensor: **Fast Perception, HDR capabilities, Low Power**
- Output: **Asynchronous and Sparse**

# Asynchronous and Sparse Event Stream

Events  
Image plane

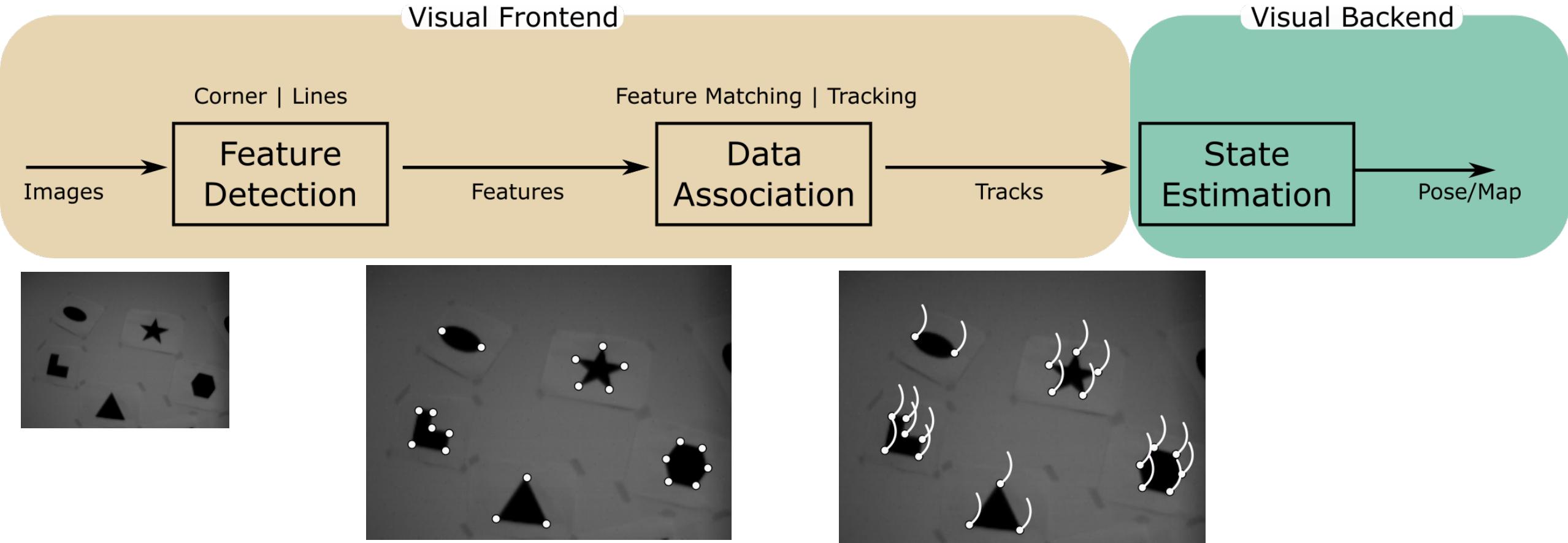


Event Stream  
Spatio-temporal space

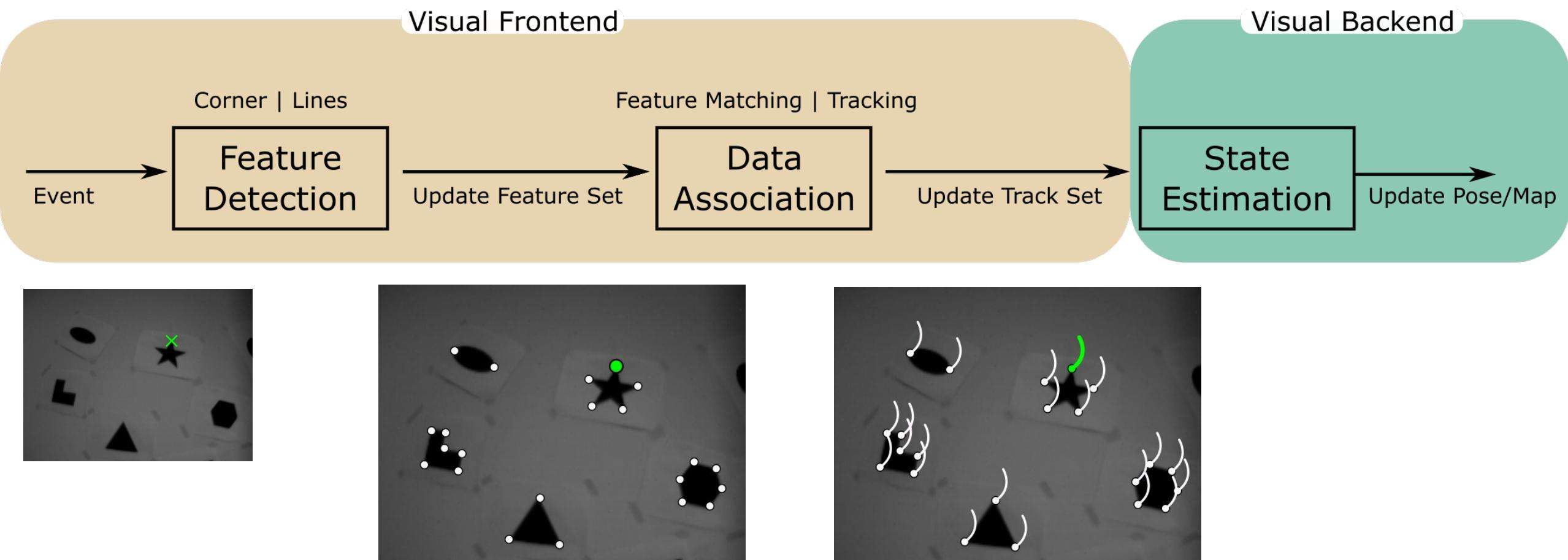


- No notion of **time-discretization**
- Only intensity **changes** are **captured**

# Traditional Visual SLAM Pipeline

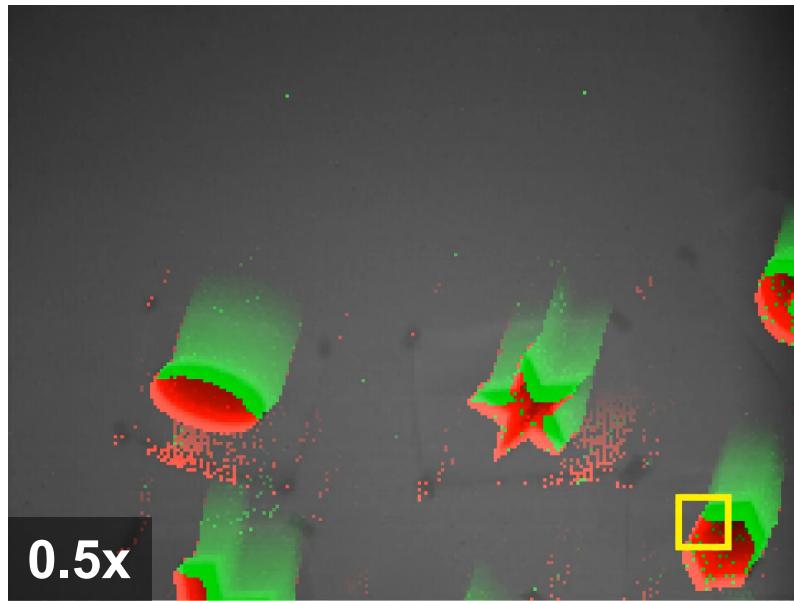


# Asynchronous Event-Driven SLAM Pipeline

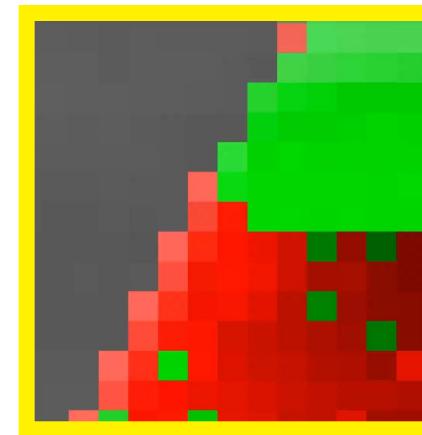


# Asynchronous Corner Detection

Events

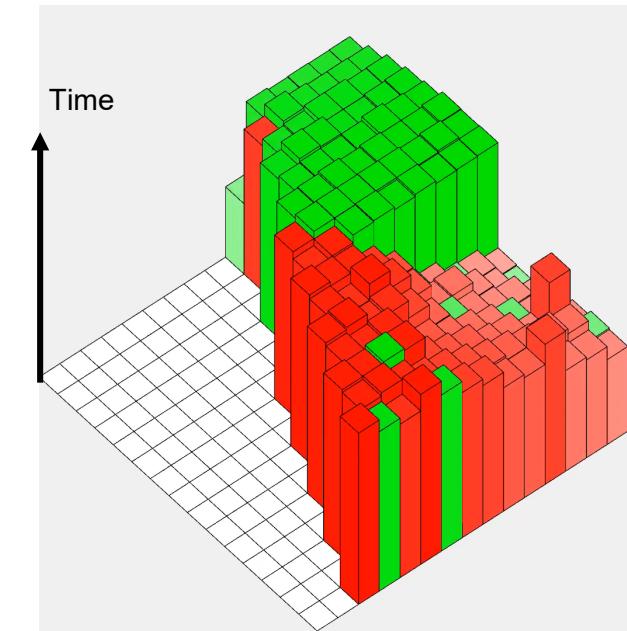


Corner Close-up



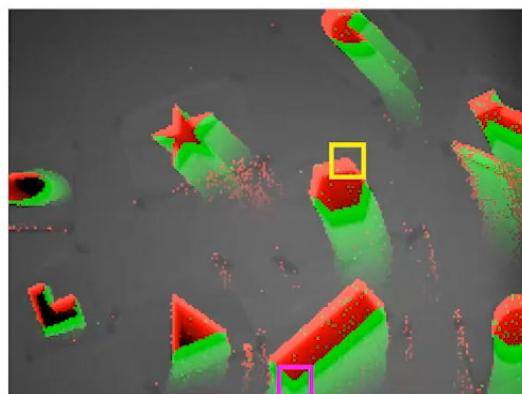
Surface of Active Events

(stores the timestamp of the last event in each location)

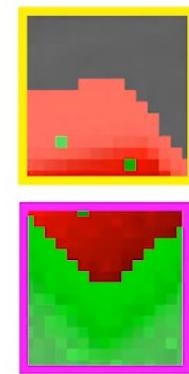


# Asynchronous Corner Detection

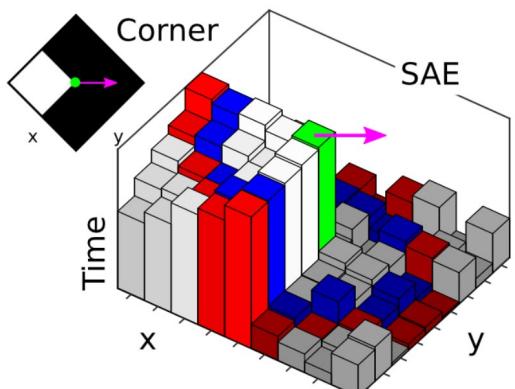
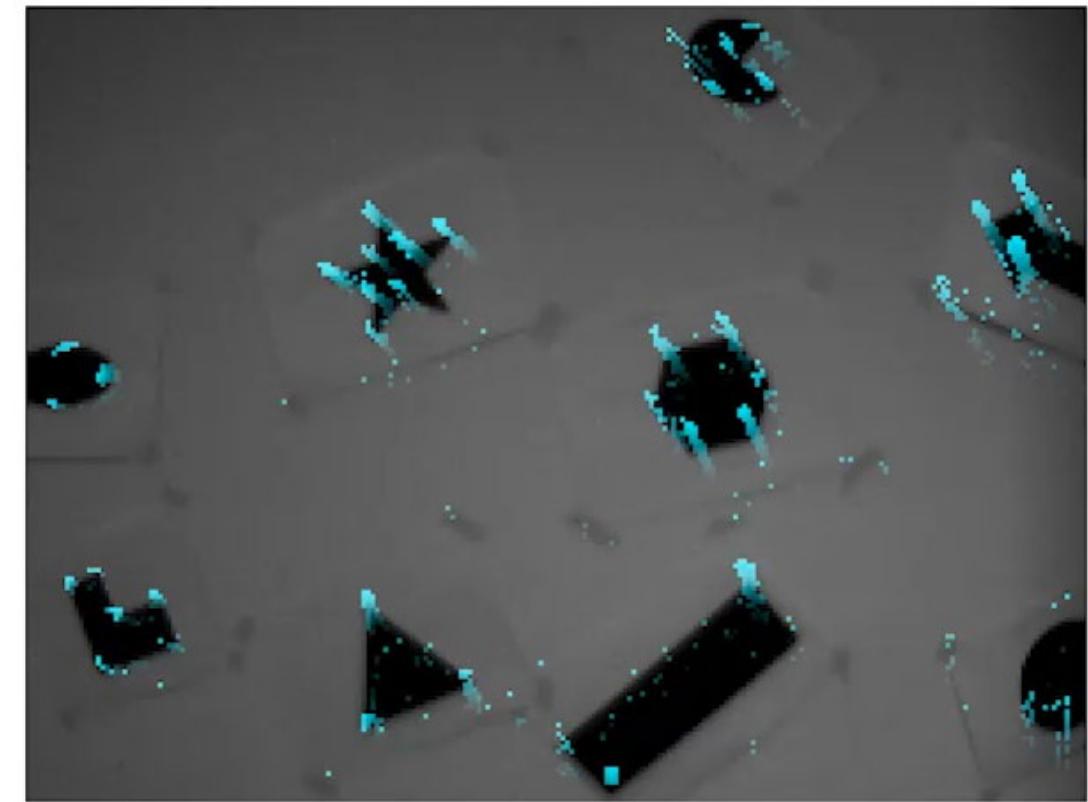
Events



Close-up



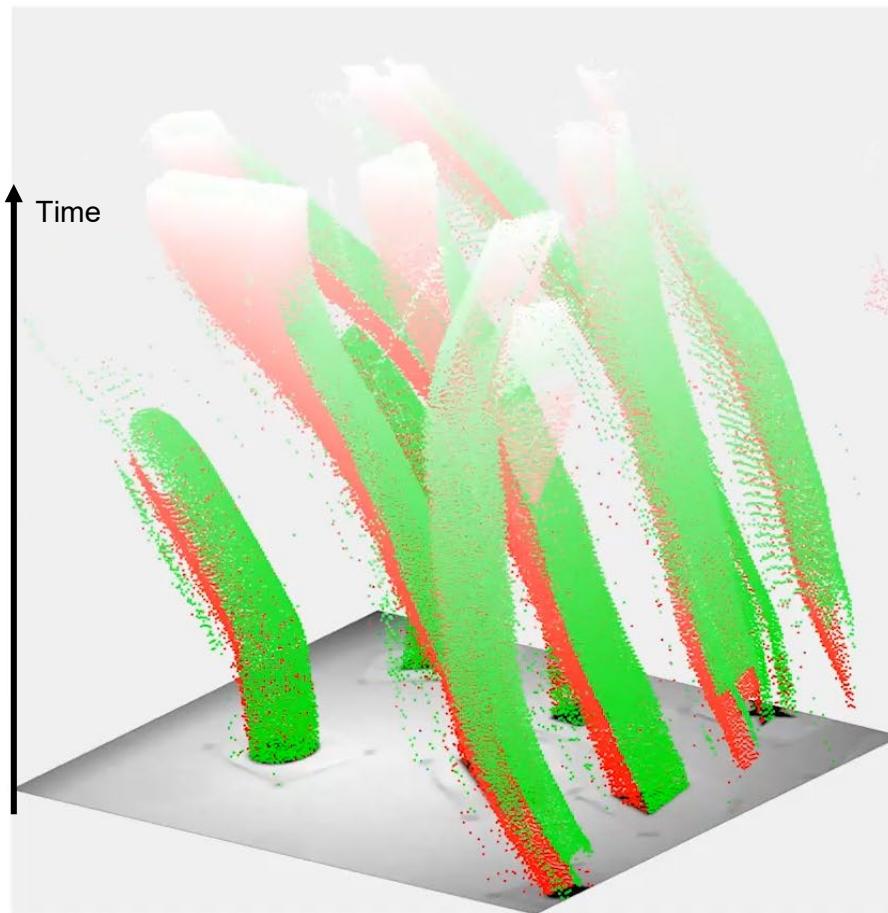
Corner Events



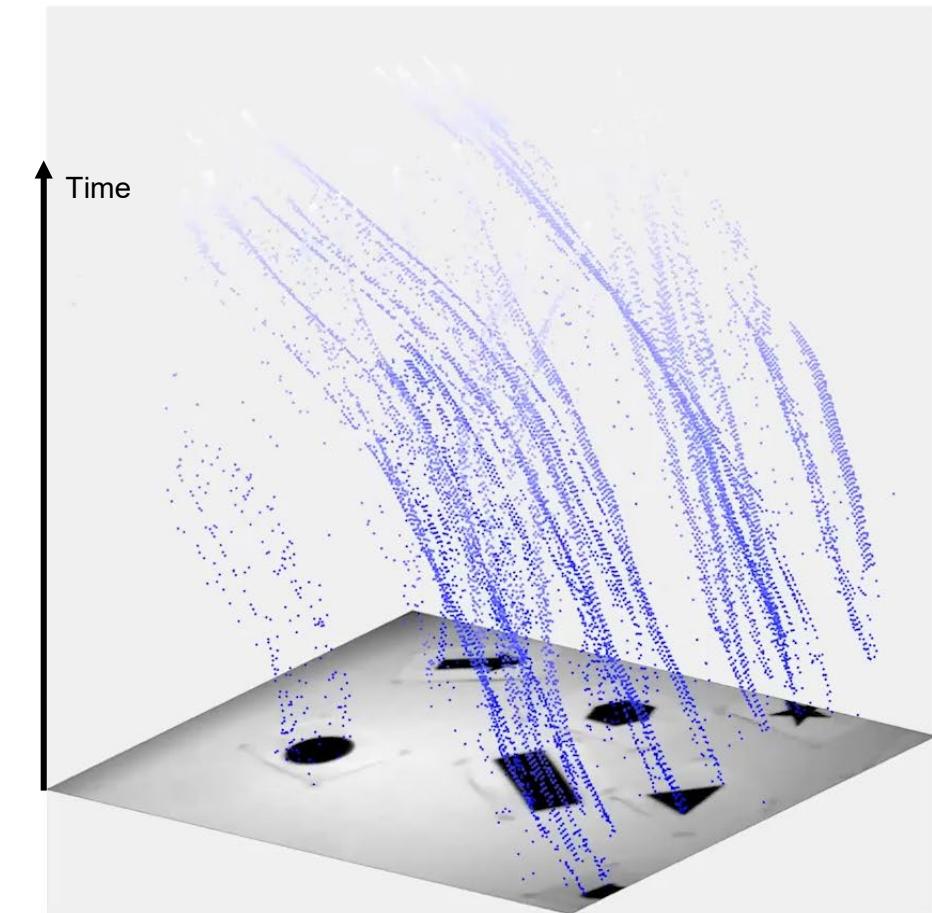
Local Surface of Events  
(centered at the newest event)

# Asynchronous Corner Detection

Stream of Events

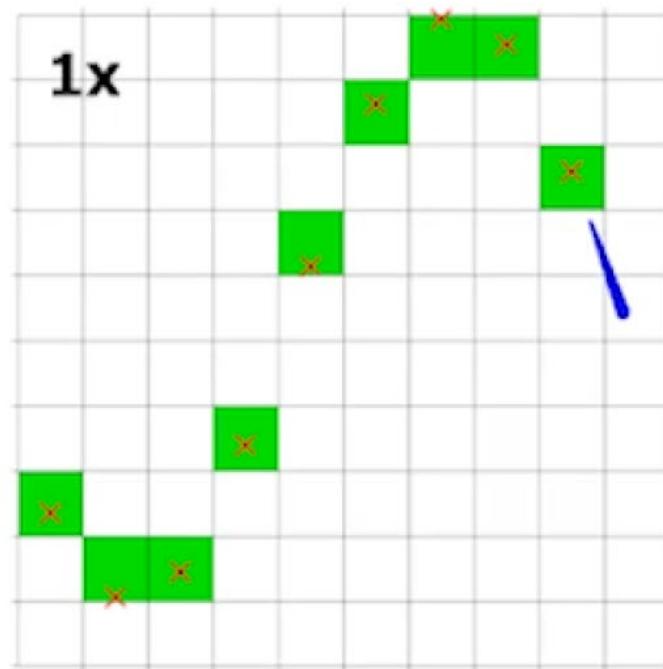


Stream of Corner Events

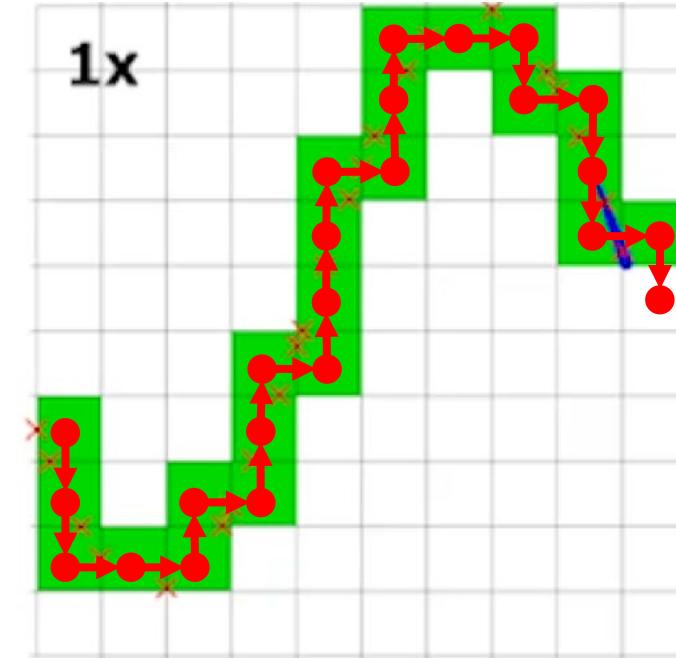


# Asynchronous Corner Detection

Frame-based detection

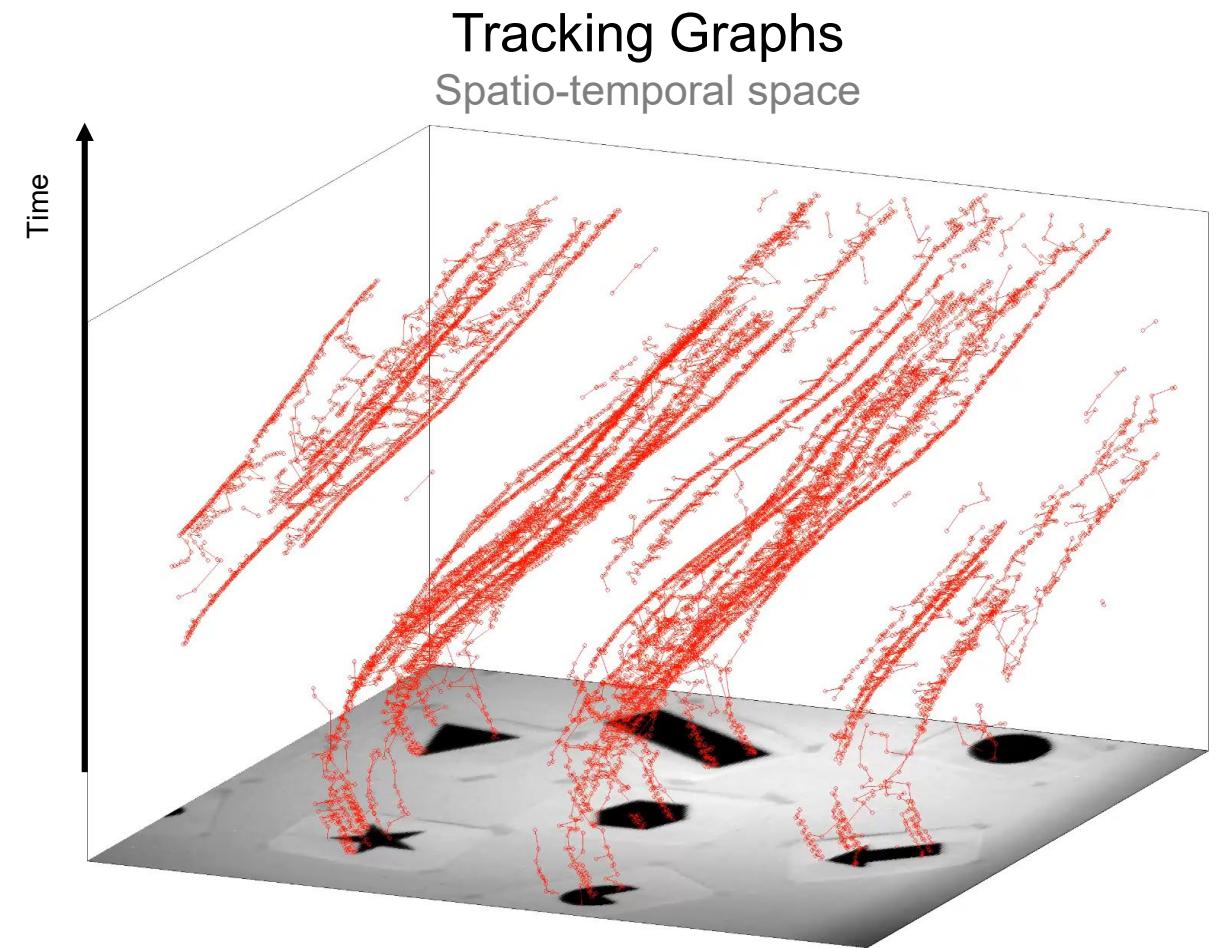
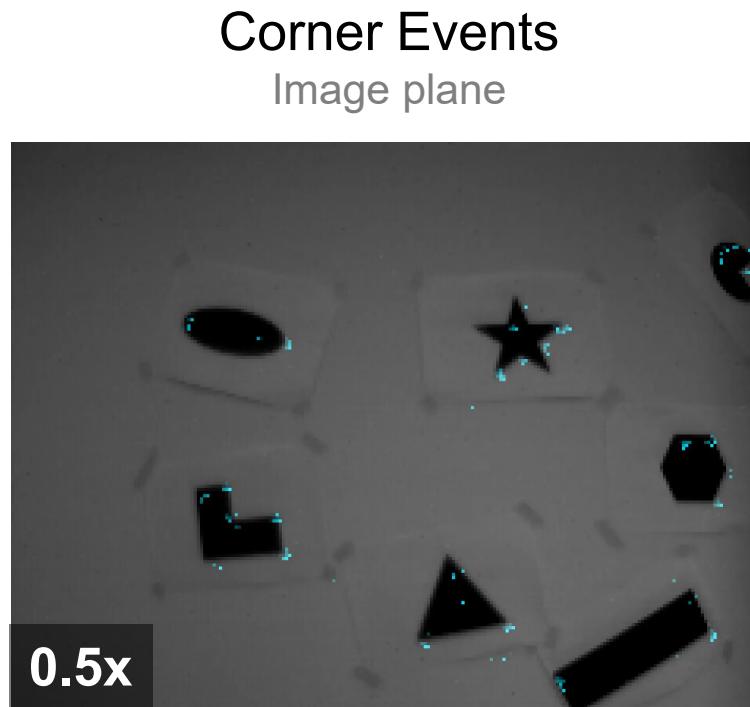


Event-driven detection



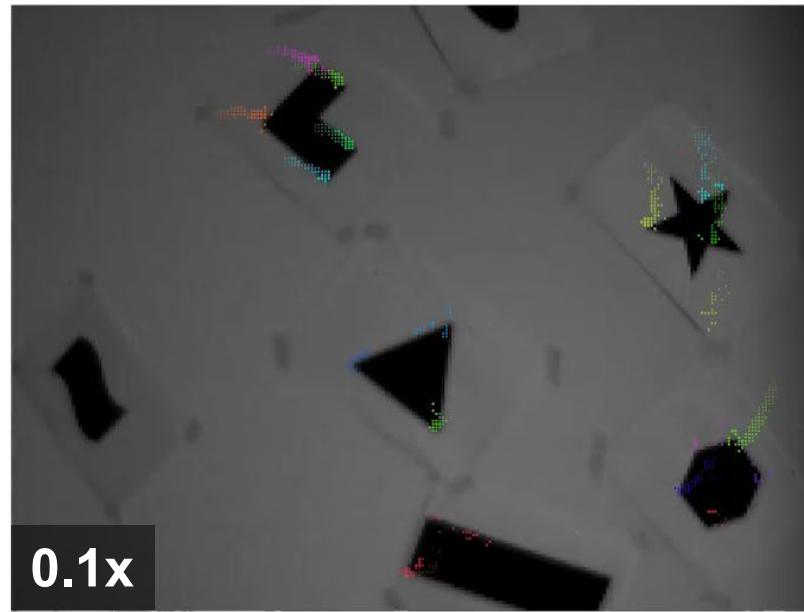
Feature projection   Feature detection   Feature pixel-detection

# Naïve Asynchronous Corner Tracking

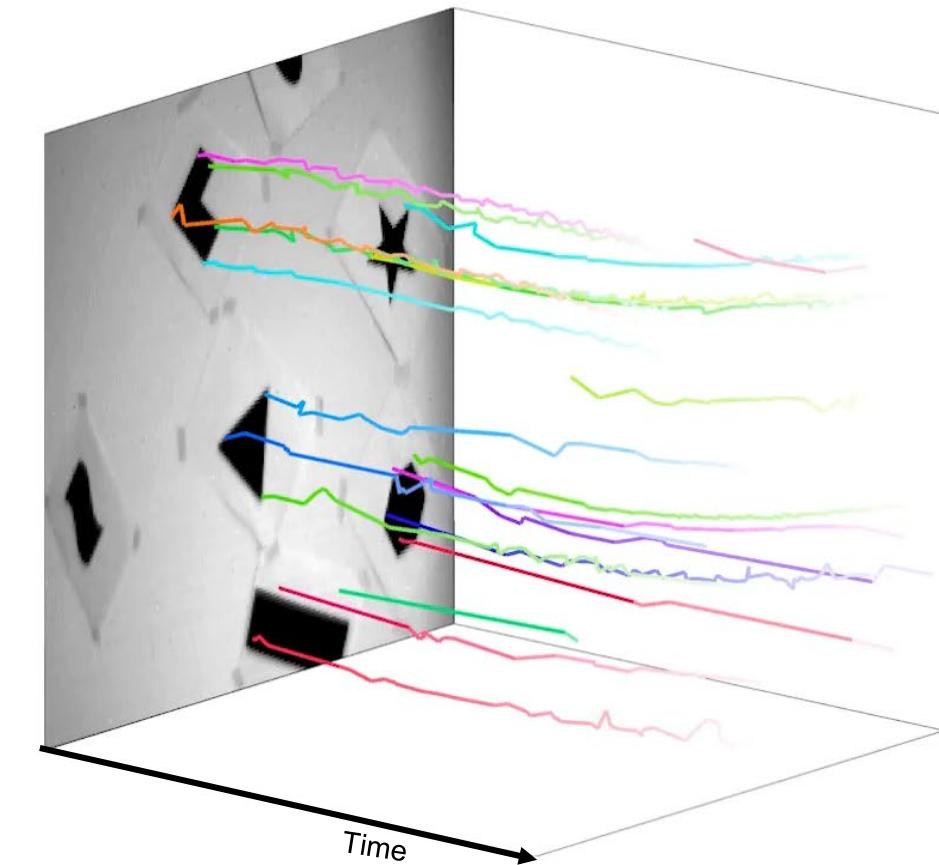


# Naïve Asynchronous Corner Tracking

Corner Tracking  
Image plane

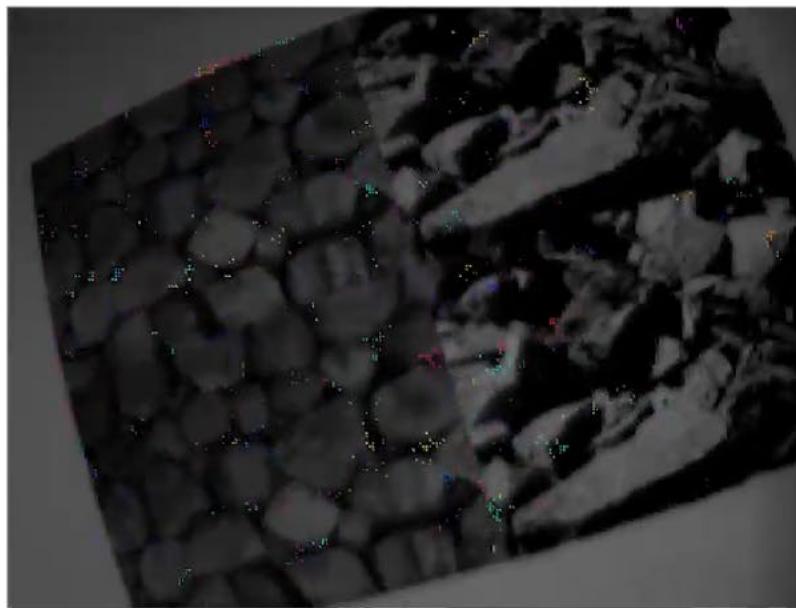


Corner Tracking  
Spatio-temporal space

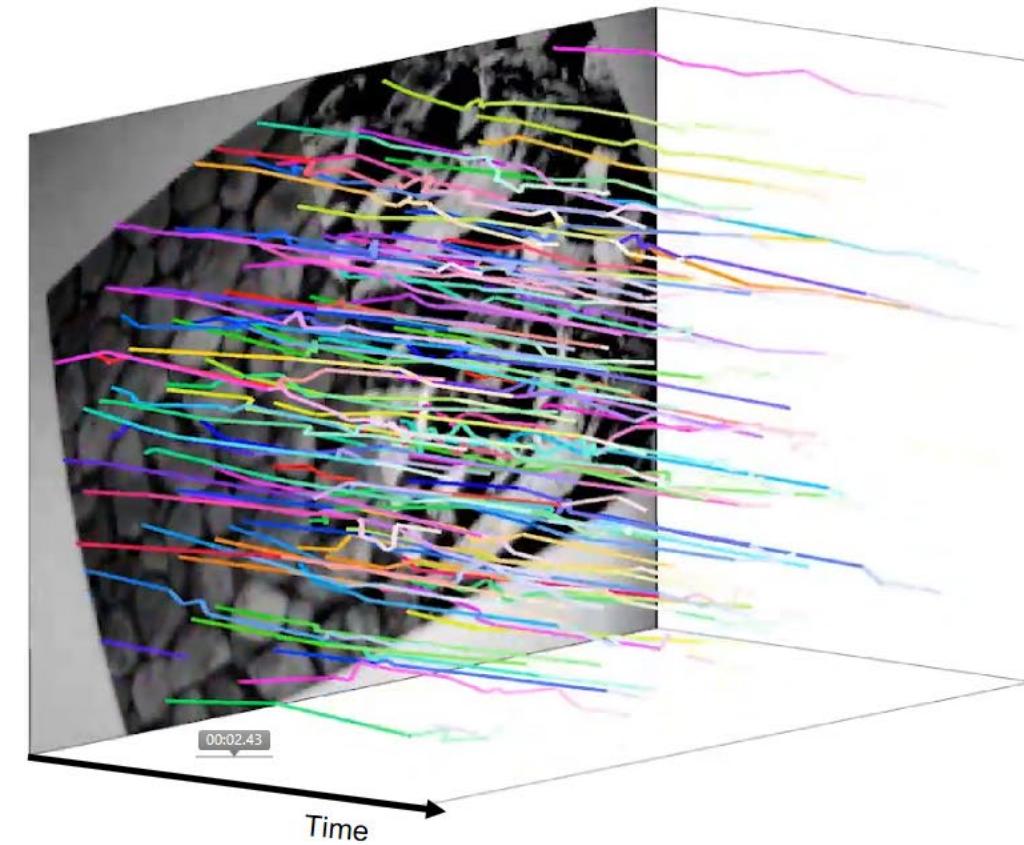


# Naïve Asynchronous Corner Tracking

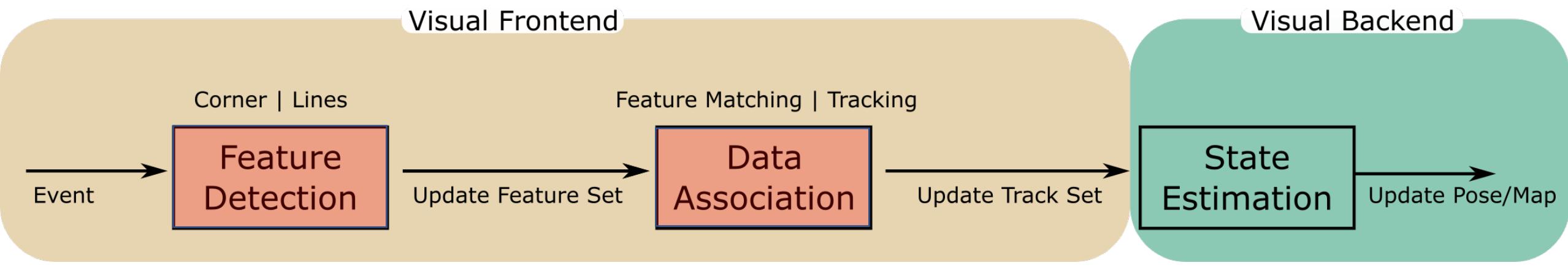
Corner Tracking  
Image plane



Corner Tracking  
Spatio-temporal space

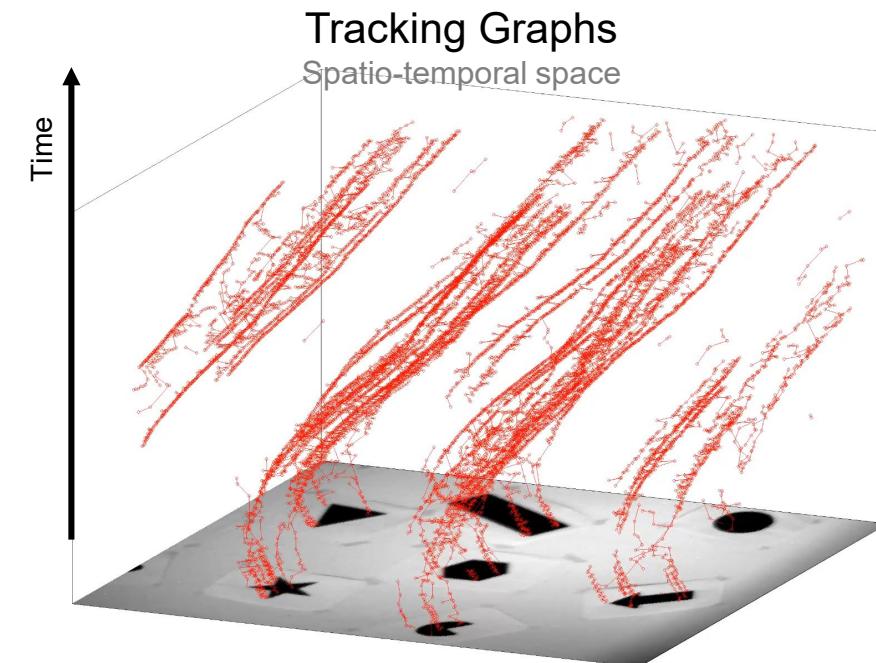


# Asynchronous Event-Driven SLAM Pipeline



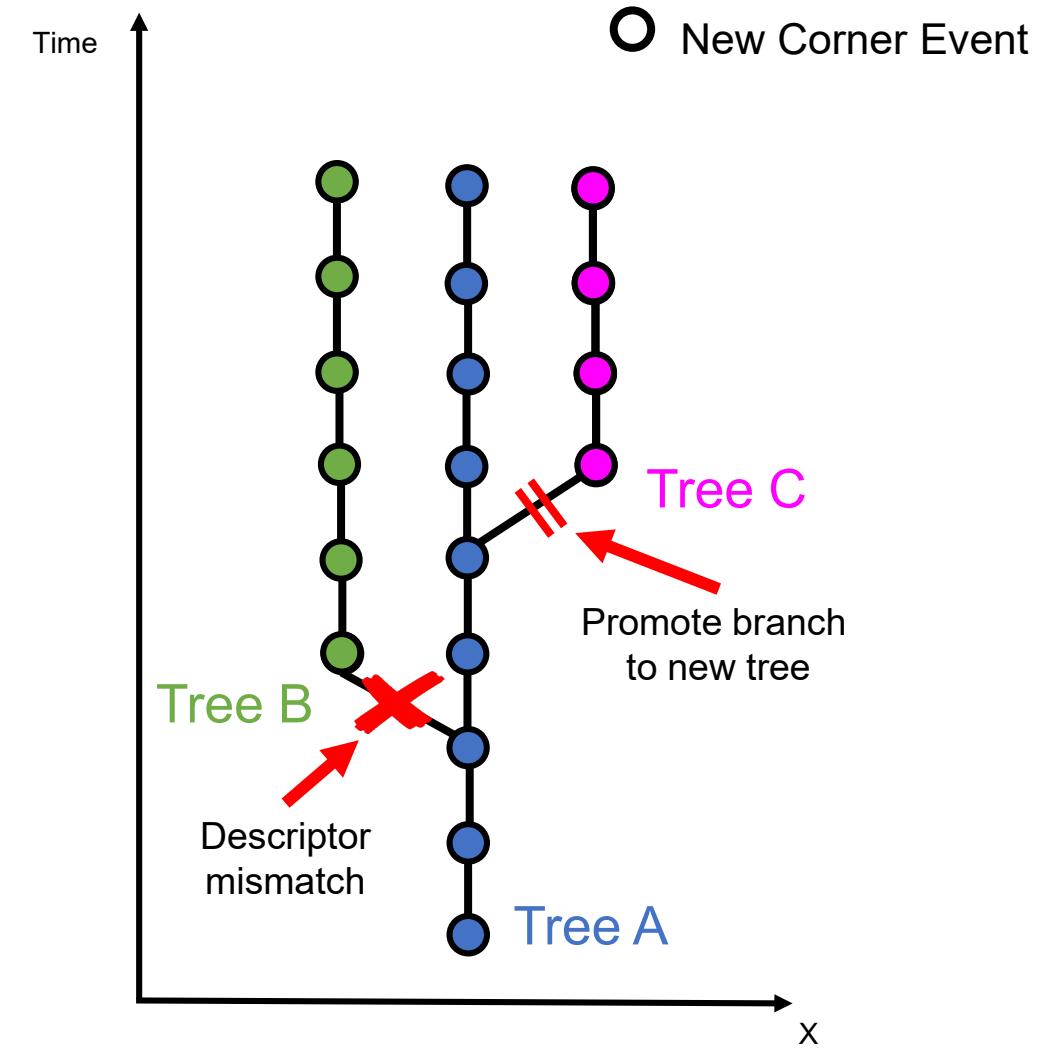
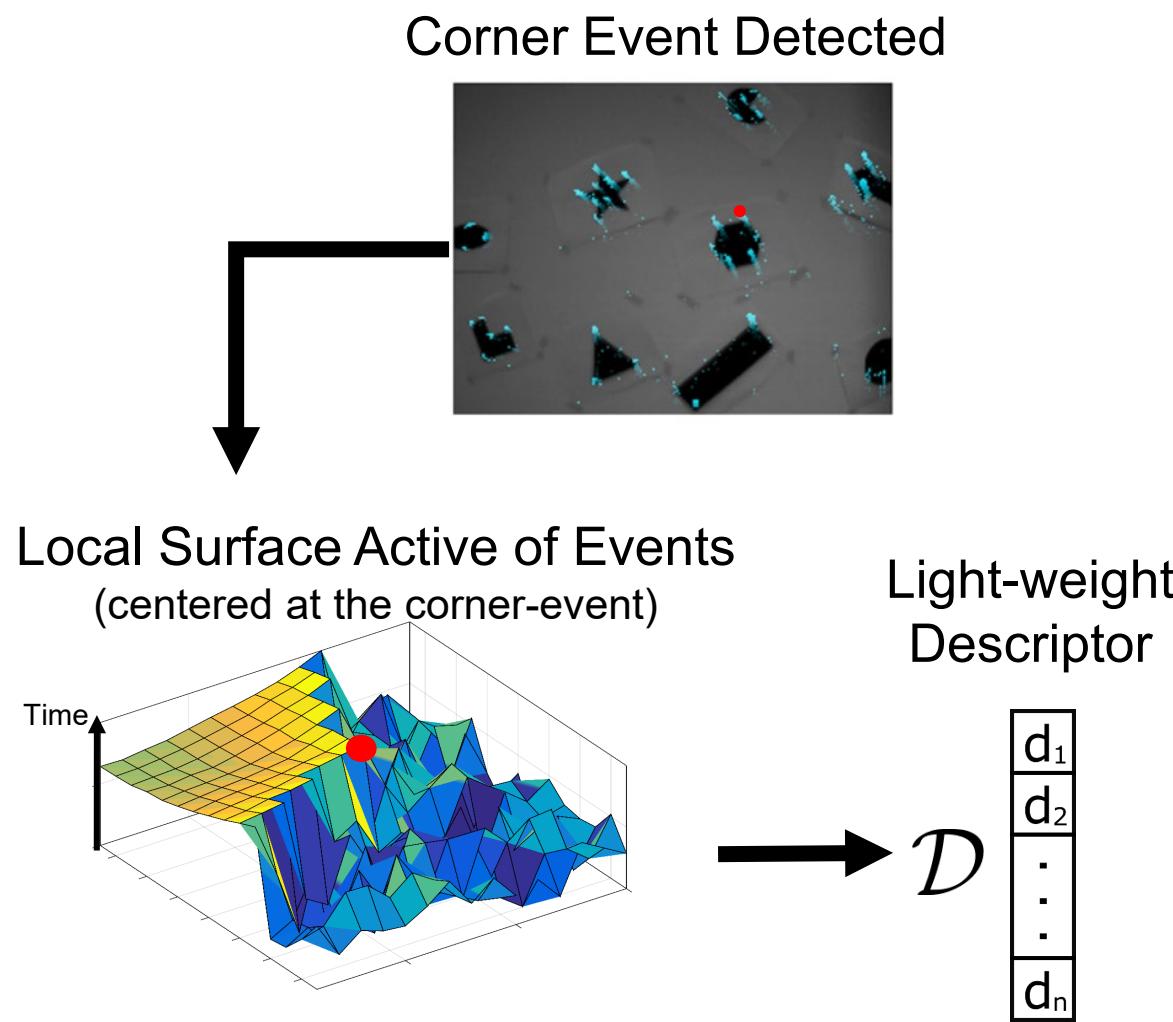
- Event Corner Detection
- Naïve Event Corner Association
- Offline Corner Tracks Retrieval

**Asynchronous Corner Detection and Tracking for Event Cameras in Real-Time**  
 [Alzugaray & Chli, RAL'18]



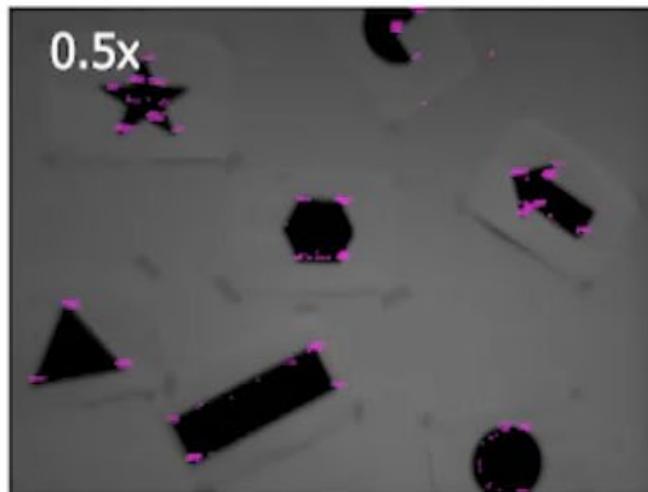
Unreliable Data Association

# Asynchronous Multi-hypothesis Corner Tracking

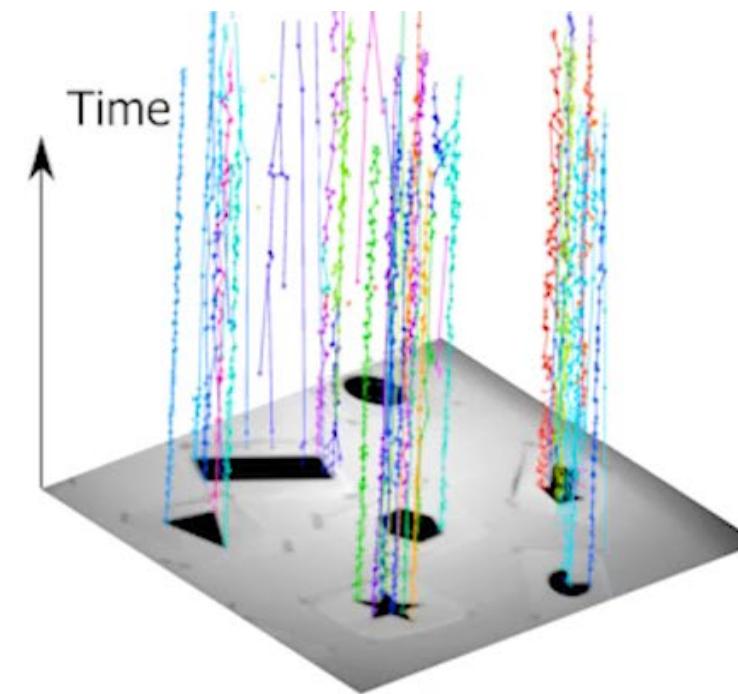


# Asynchronous Multi-hypothesis Corner Tracking

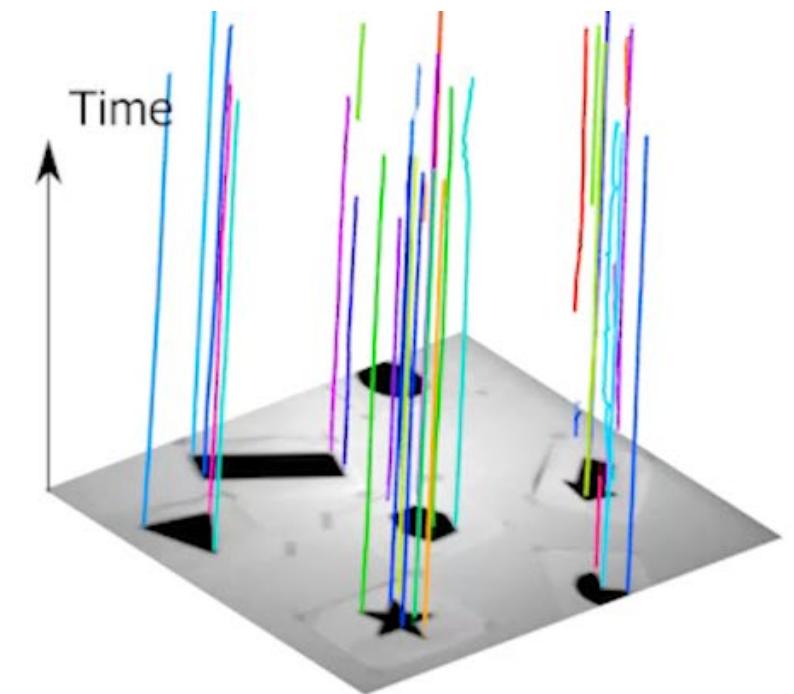
Corner Events



Tracking Graph  
Multi-hypothesis features

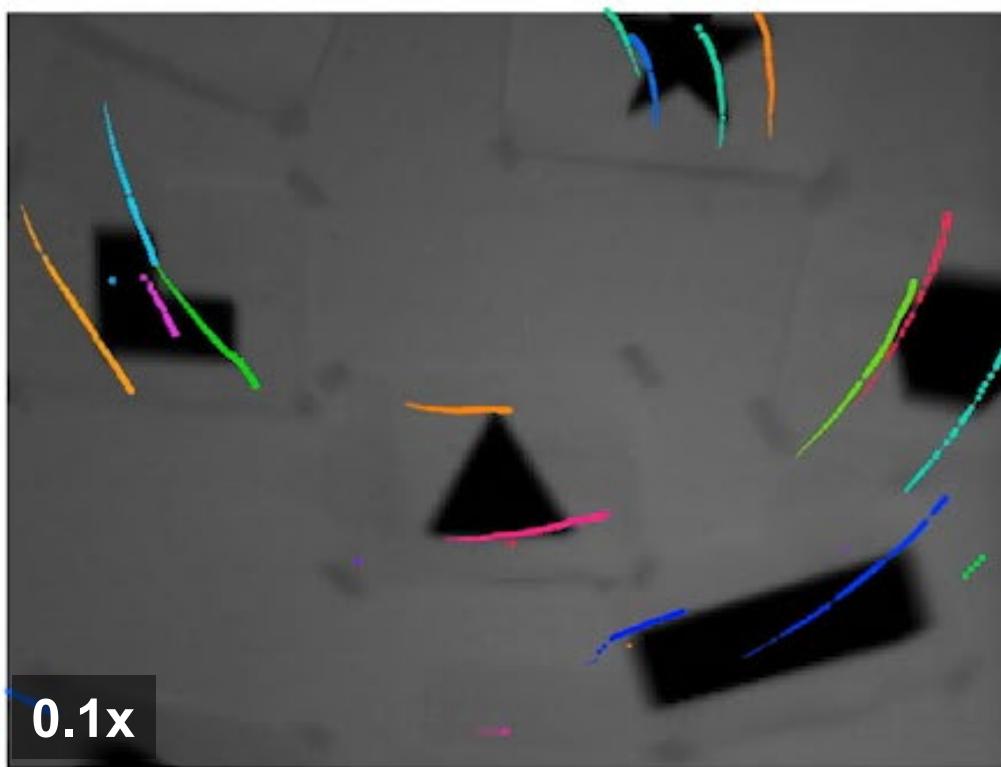


Feature Tracks  
Hypothesis selection

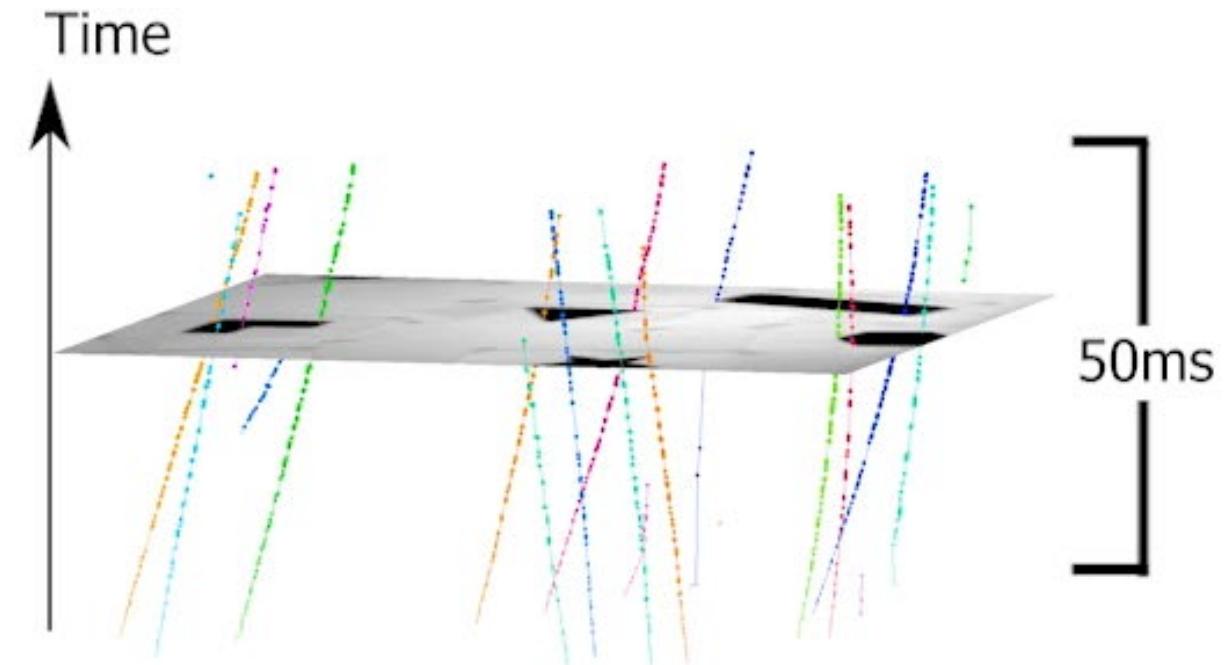


# Asynchronous Multi-hypothesis Corner Tracking

Corner Tracking  
Image plane

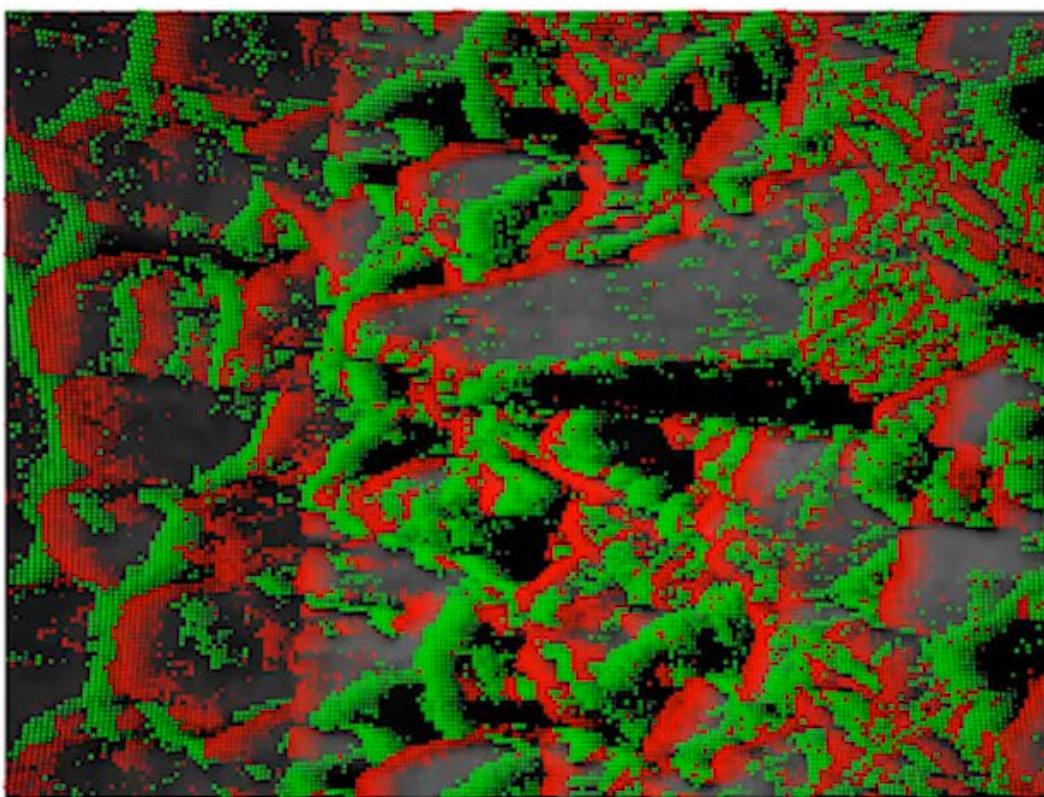


Corner Tracking  
Spatio-temporal space

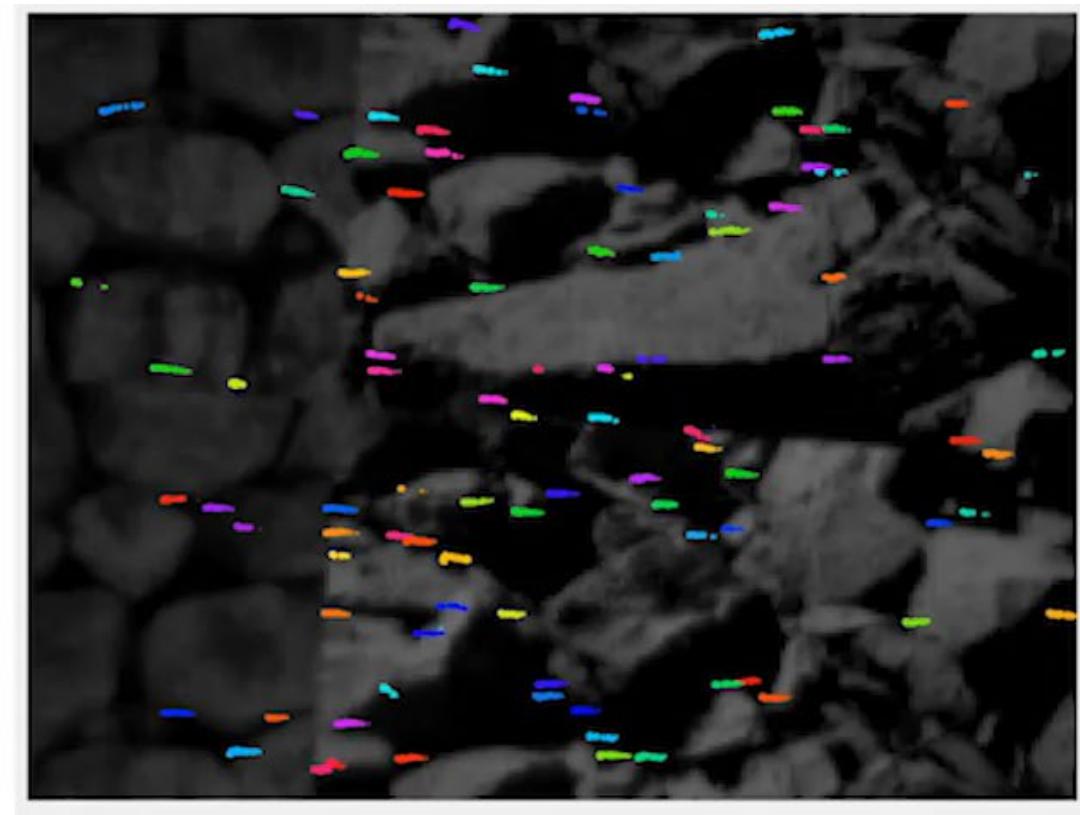


# Asynchronous Multi-hypothesis Corner Tracking

Events

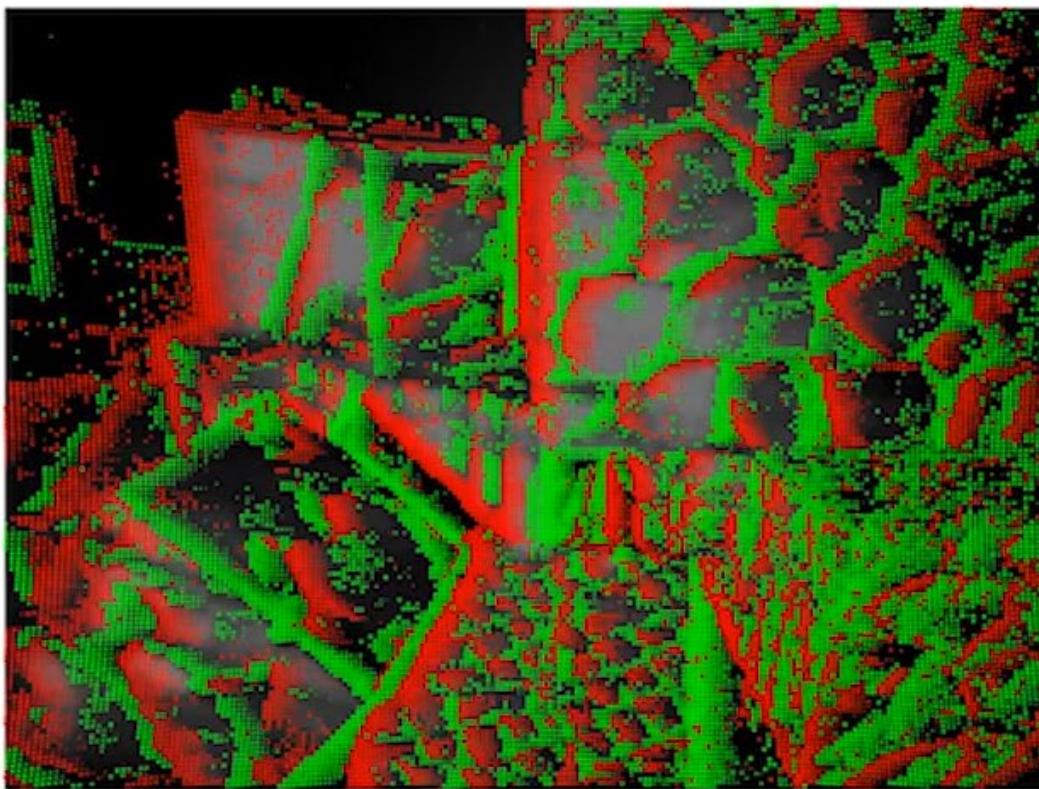


Corner Tracking



# Asynchronous Multi-hypothesis Corner Tracking

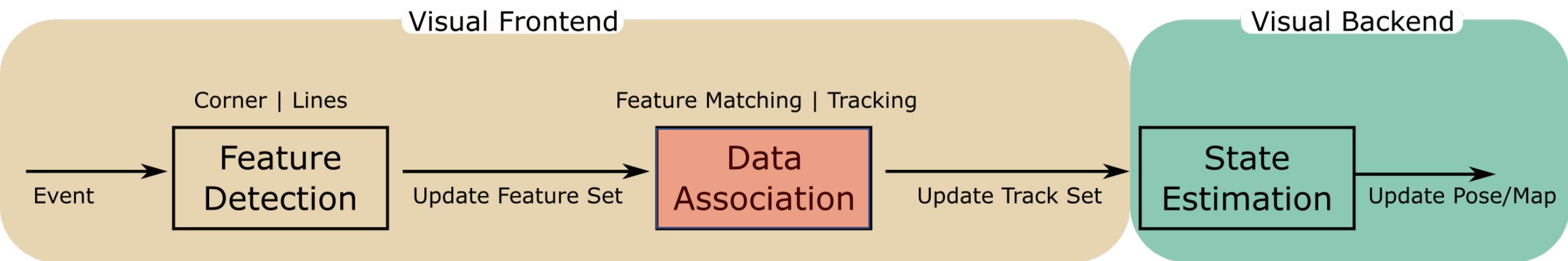
Events



Corner Tracking



# Asynchronous Event-Driven SLAM Pipeline

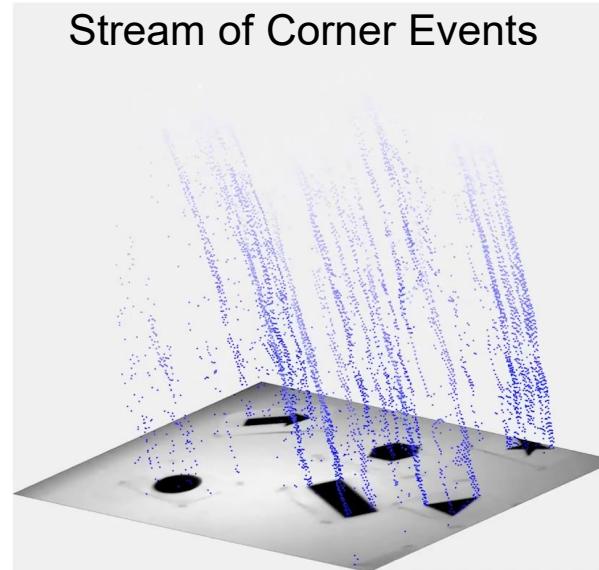


- Event Corner Detection
- Naïve Event Corner Association
- Offline Corner Tracks Retrieval

**Asynchronous Corner Detection and Tracking for Event Cameras in Real-Time**  
 [Alzugaray & Chli, RAL'18]

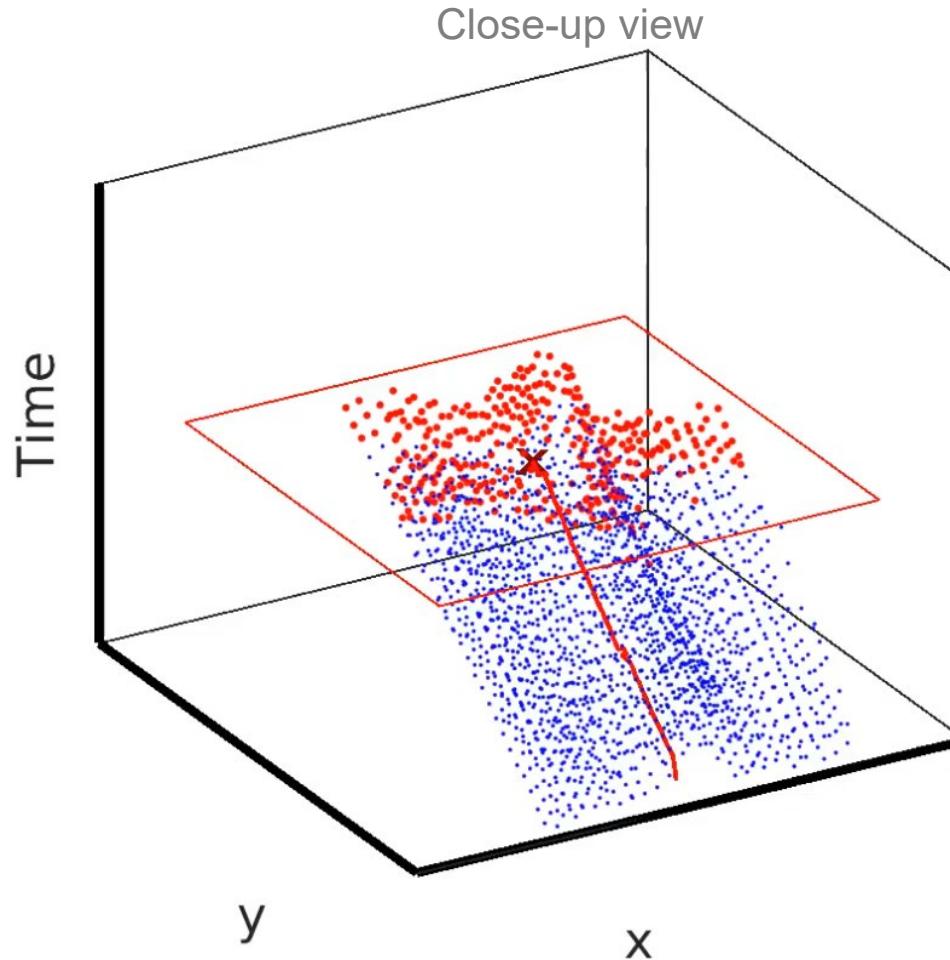
- Local Event Descriptor
- Multi-hypothesis Data Association
- Online Corner Track Retrieval

**ACE: An Efficient Asynchronous Corner Tracker for Event Cameras**  
 [Alzugaray & Chli, 3DV'18]



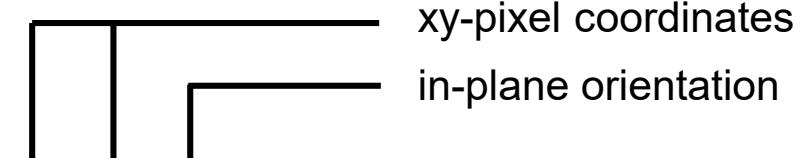
# Asynchronous Tracking of Events

Feature Track in the Event Stream



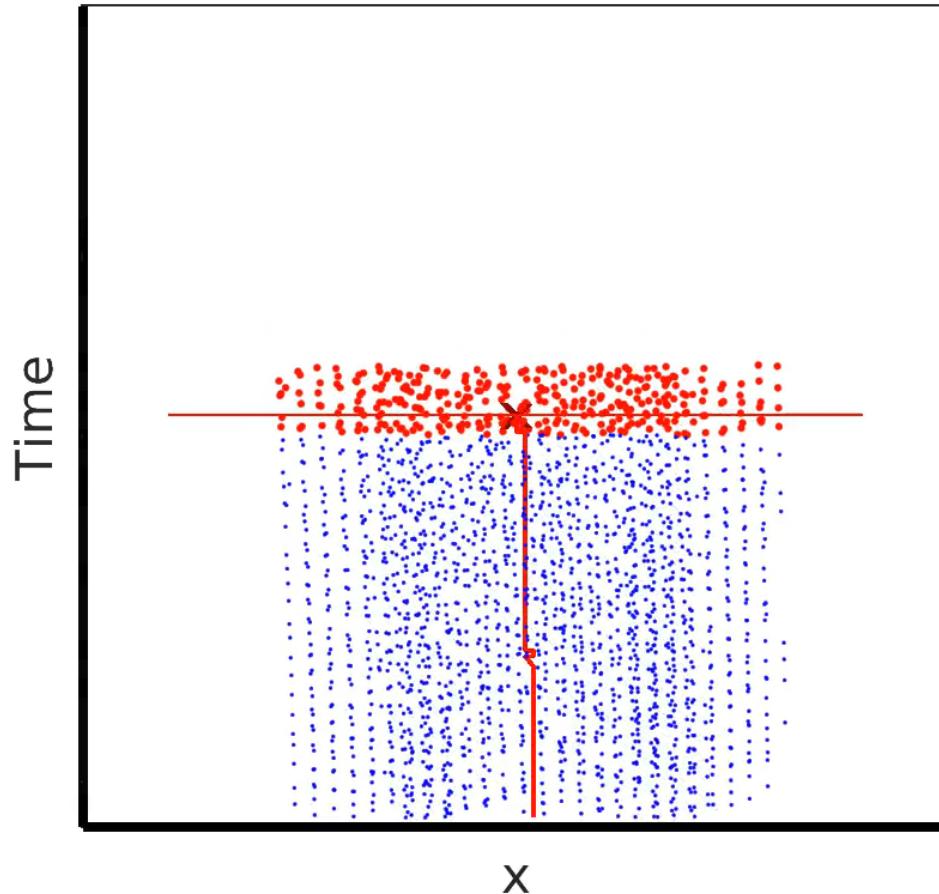
Feature  $\mathcal{F}$

- State  $\mathbf{x} = \{x, y, \theta\}$
- Window of Latest Events  $\mathcal{E}$ 
  - Fixed number of events



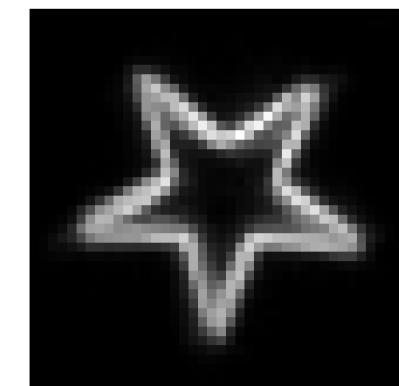
# Asynchronous Tracking of Events

Feature Track in the Event Stream  
Close-up view

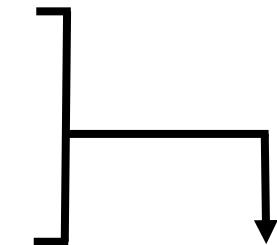
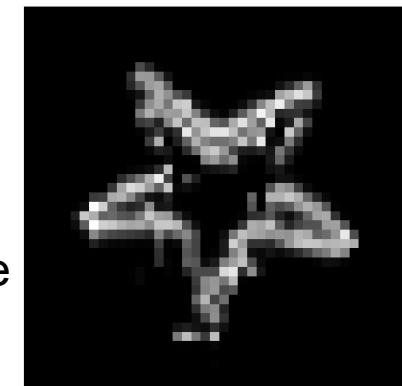


Feature  $\mathcal{F}$

- State  $\mathbf{x} = \{x, y, \theta\}$
- Window of Latest Events  $\mathcal{E}$ 
  - Fixed number of events
- Template  $\mathcal{T}$

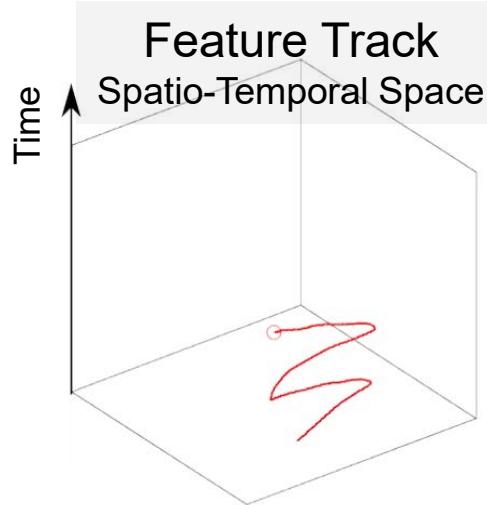


Alignment Score  
 $f$

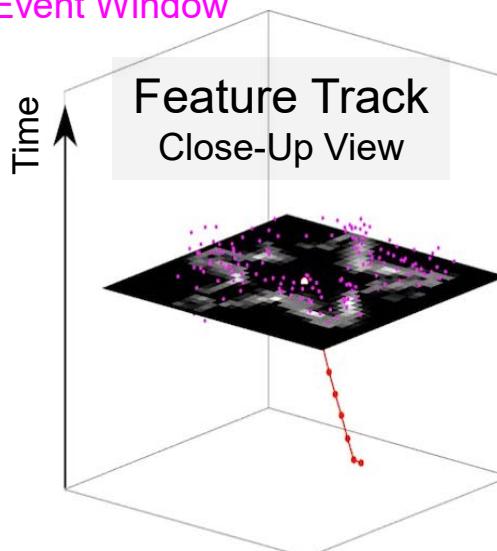


Model  $\mathcal{M}(\mathbf{x}, \mathcal{E})$

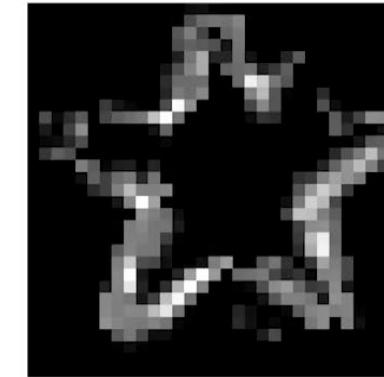
# Feature Tracking as Optimization Problem



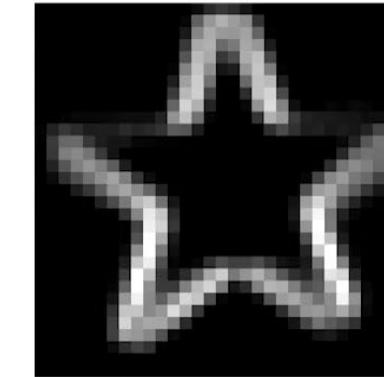
— Feature Track  
● Event Window



Model  $\mathcal{M}(\mathbf{x}, \mathcal{E})$



Template  $\mathcal{T}$



Alignment Score Function

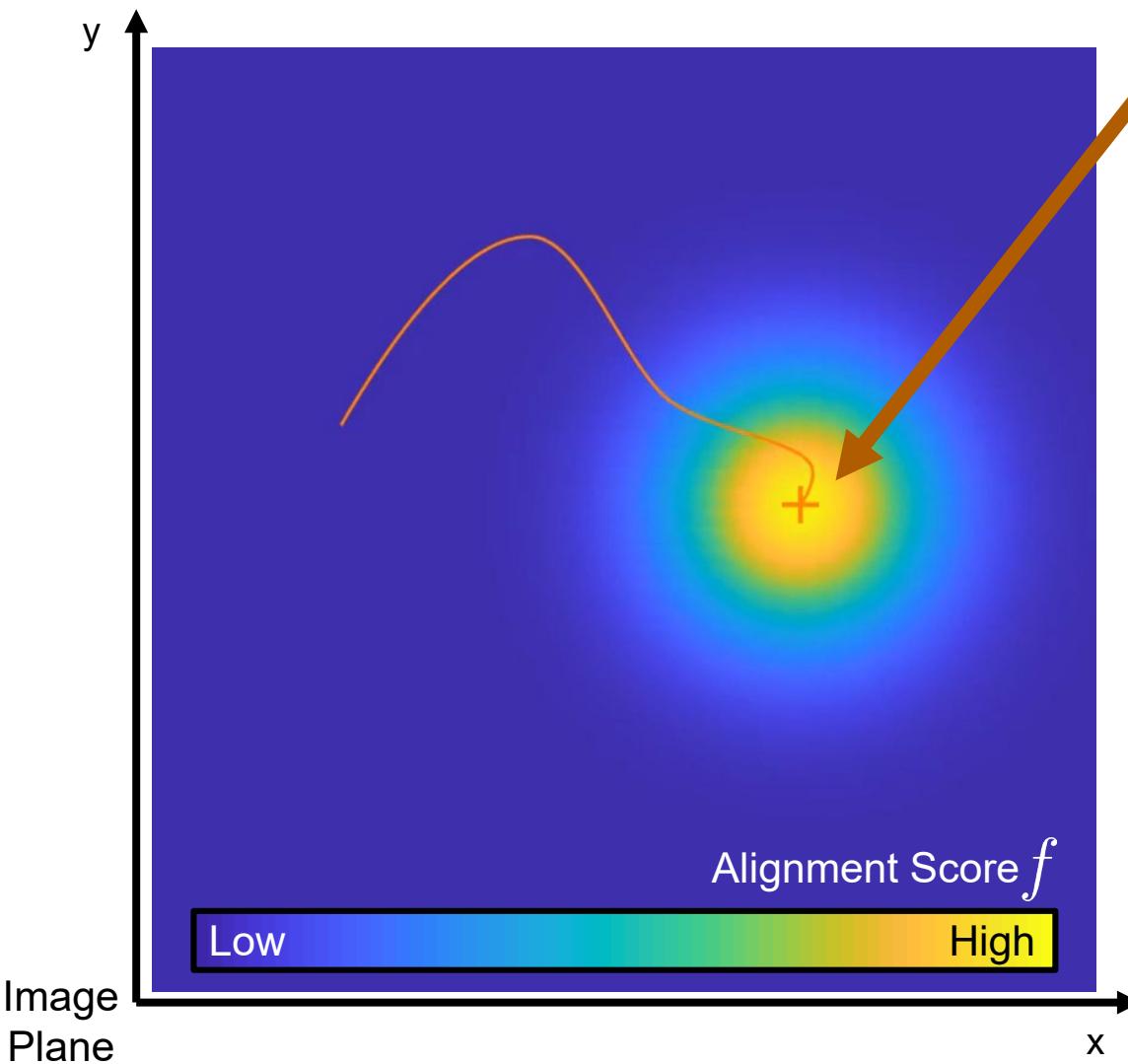
$$\mathbf{x}^* = \arg \max_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathcal{E}, \mathcal{T})$$

Optimal Feature State

Up to Millions of Events per second!

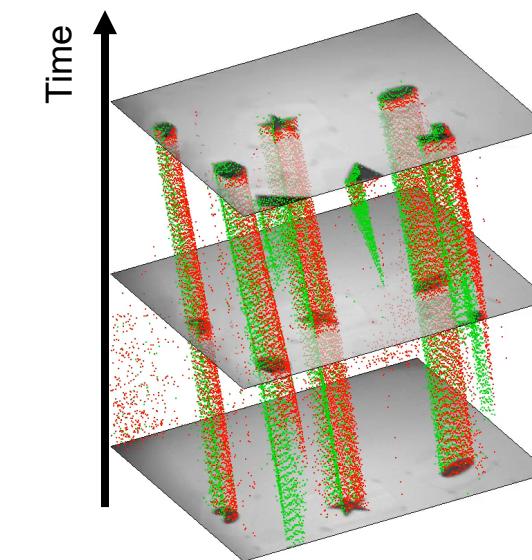
Template  
Window of latest events

# Feature Tracking with Asynchronous Hypotheses Evaluation



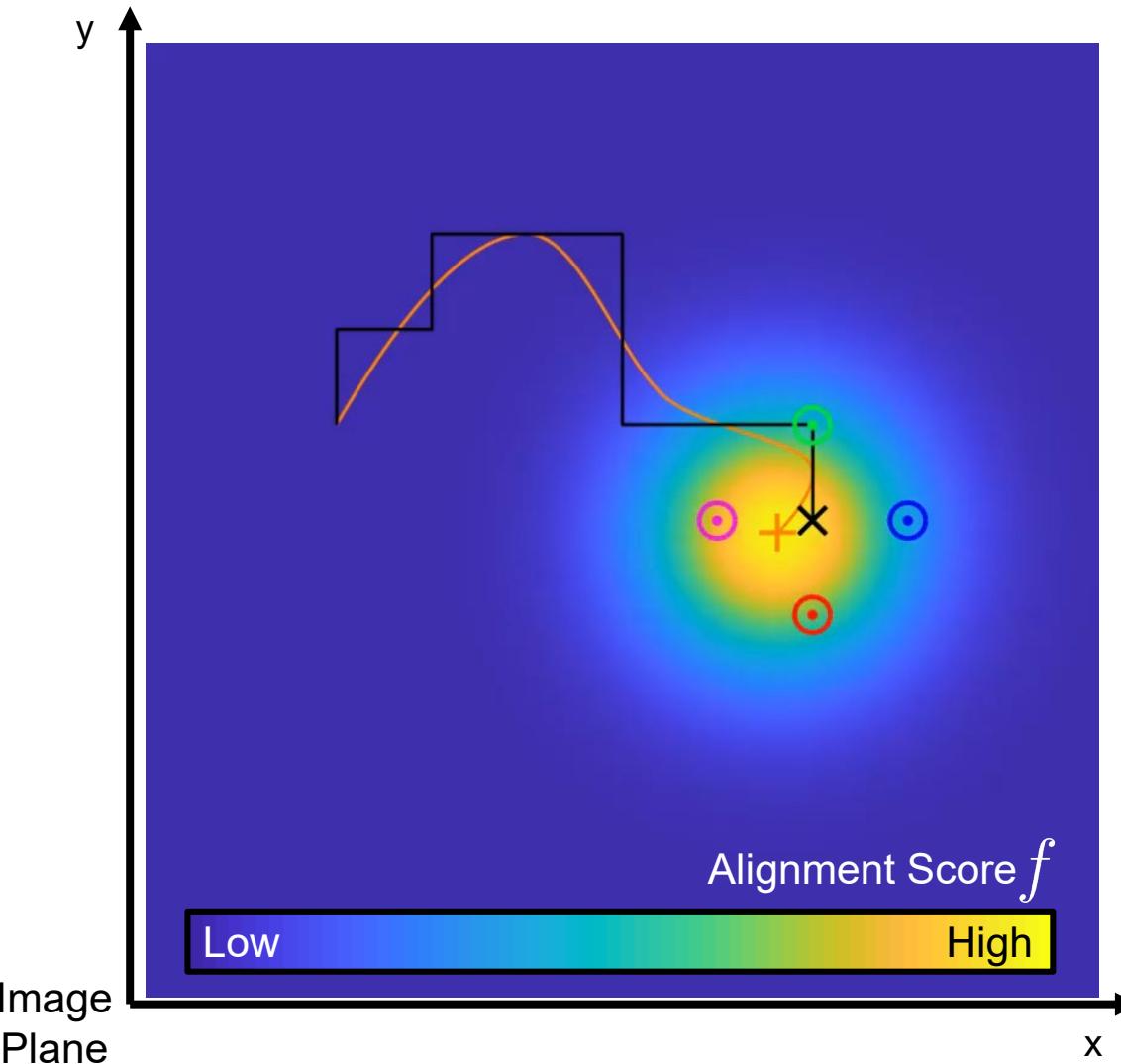
+ Optimal State (Continuous)

$$\mathbf{x}^{(k+1)} = \arg \max_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathcal{E}^{(k+1)}, \mathcal{T}^{(k+1)})$$



Continuous stream of information

# Feature Tracking with Asynchronous Hypotheses Evaluation



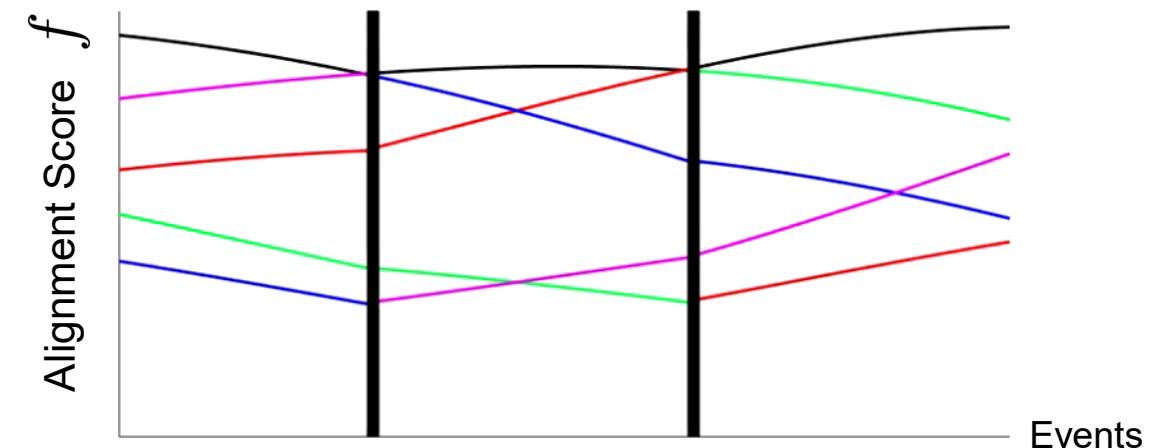
⊕ Optimal State (Continuous)

$$\mathbf{x}^{(k+1)} = \arg \max_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathcal{E}^{(k+1)}, \mathcal{T}^{(k+1)})$$

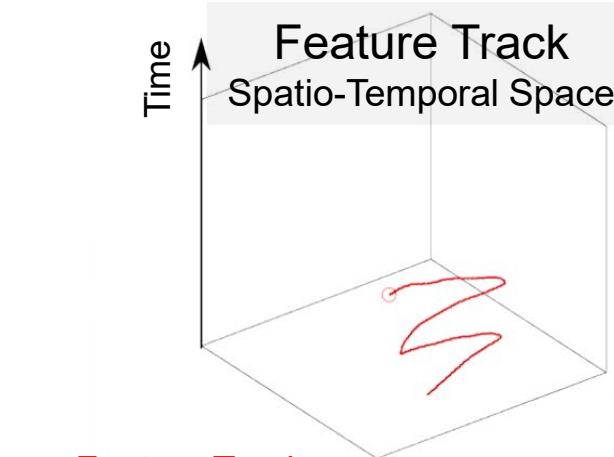
✗ Current Optimal State (Discrete)

$$\mathbf{x}^{(k+1)} = \arg \max_{\mathbf{x} \in \mathcal{H}(\mathbf{x}^{(k)}) \subset \mathcal{X}} f(\mathbf{x}, \mathcal{E}^{(k+1)}, \mathcal{T}^{(k+1)})$$

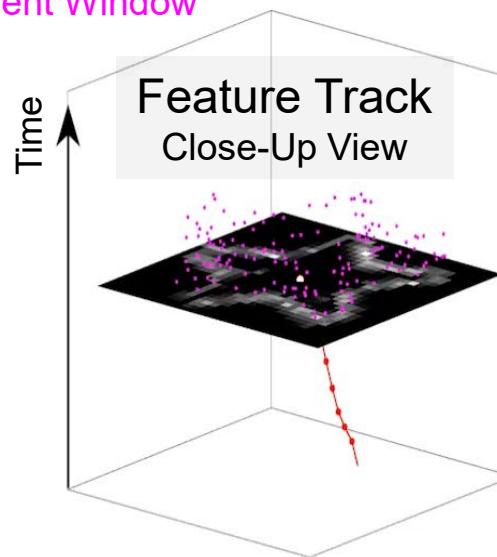
Set of hypothetical states: {Null, North, East, South, West}



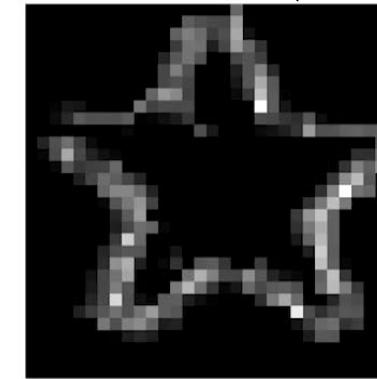
# Feature Tracking with Asynchronous Hypotheses Evaluation



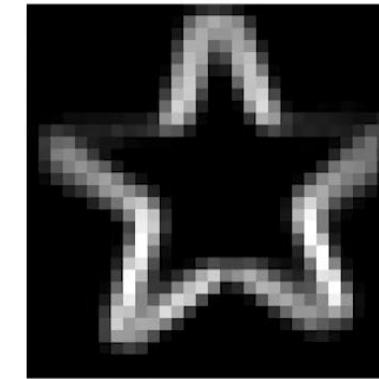
— Feature Track  
● Event Window



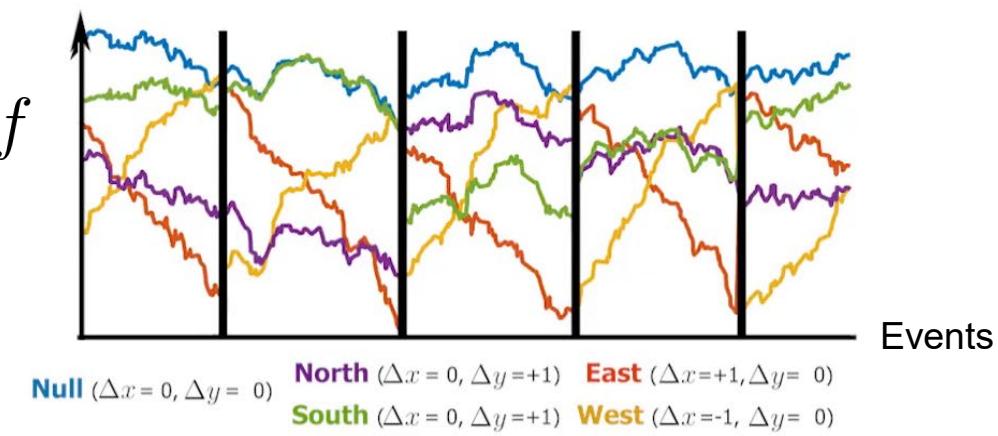
Model  $\mathcal{M}(\mathbf{x}, \mathcal{E})$



Template  $\mathcal{T}$



Alignment Score  $f$   
per hypothesis



# Feature Tracking with Incremental Alignment Score

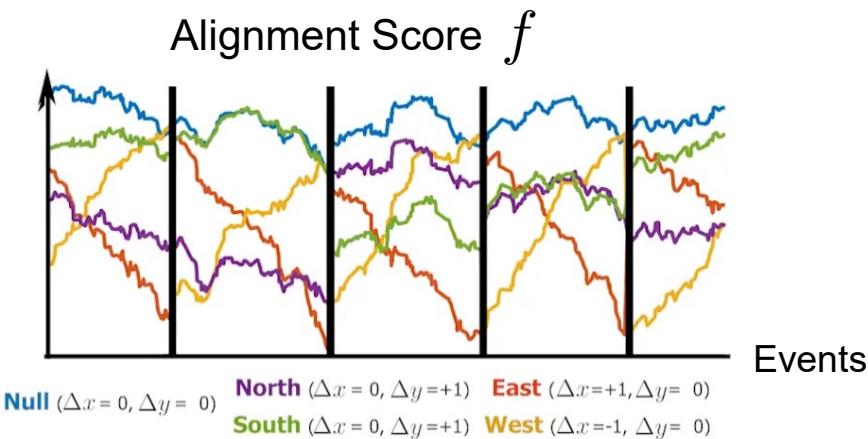
For hypothesis  $\mathbf{X}_h$

with alignment score  $f^{(k)} = f(\mathbf{x}_h, \mathcal{E}^{(k)}, \mathcal{T}^{(k)})$

New event is generated  $\mathbf{e}_{k+1}$

Update Event Window  $\mathcal{E}^{(k)} \rightarrow \mathcal{E}^{(k+1)}$   
 Update Template  $\mathcal{T}^{(k)} \rightarrow \mathcal{T}^{(k+1)}$

→ alignment score must be reevaluated  $f^{(k+1)}$   
 in [Alzugaray & Chli, 3DV'19]

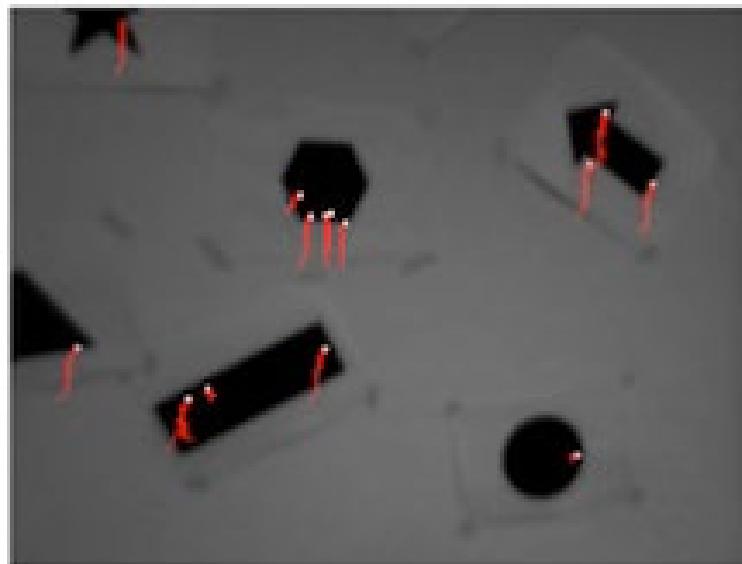


Incremental Alignment Score update:

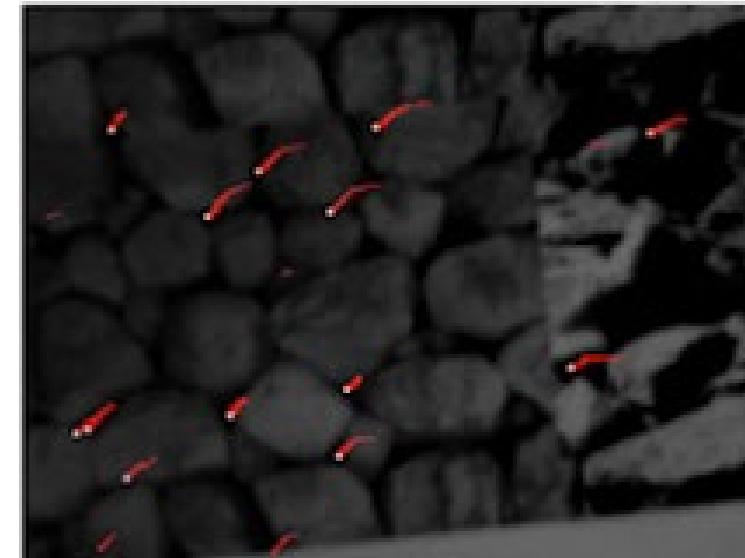
$$f^{(k+1)} = g(f^{(k)}, \mathbf{e}_{k+1})$$

# Asynchronous Multi-Hypothesis Tracking of Events

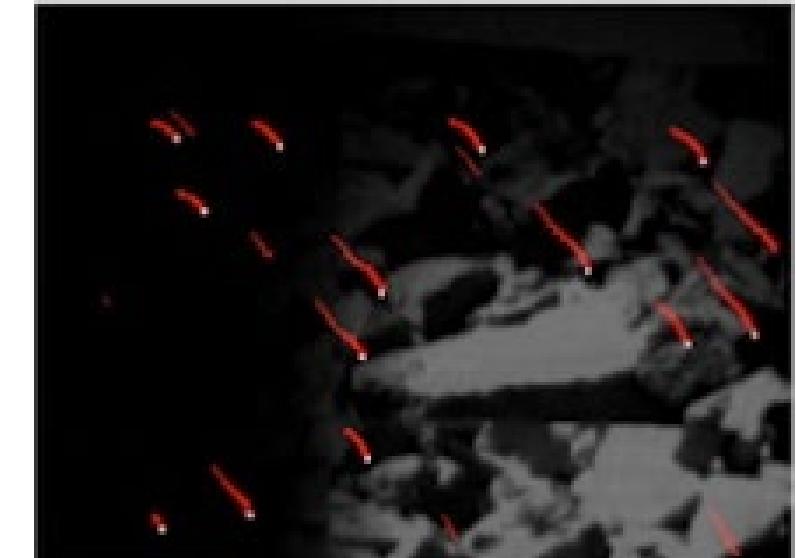
Simple Scene



High-Textured Scene

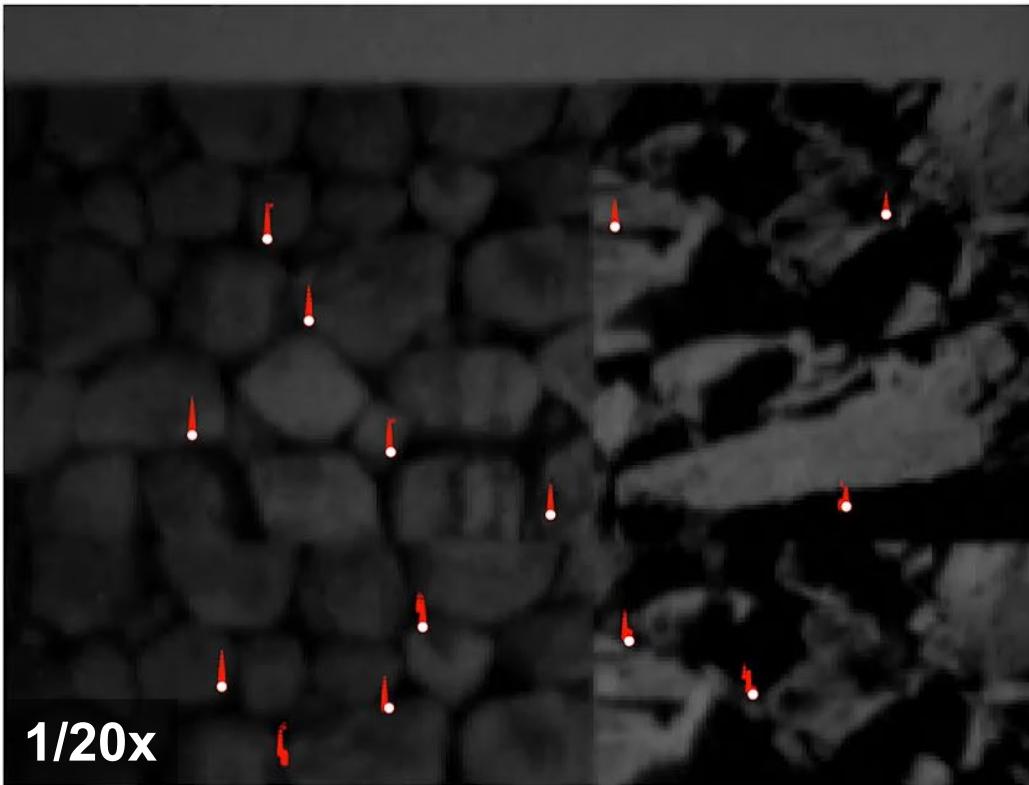


HDR Illumination

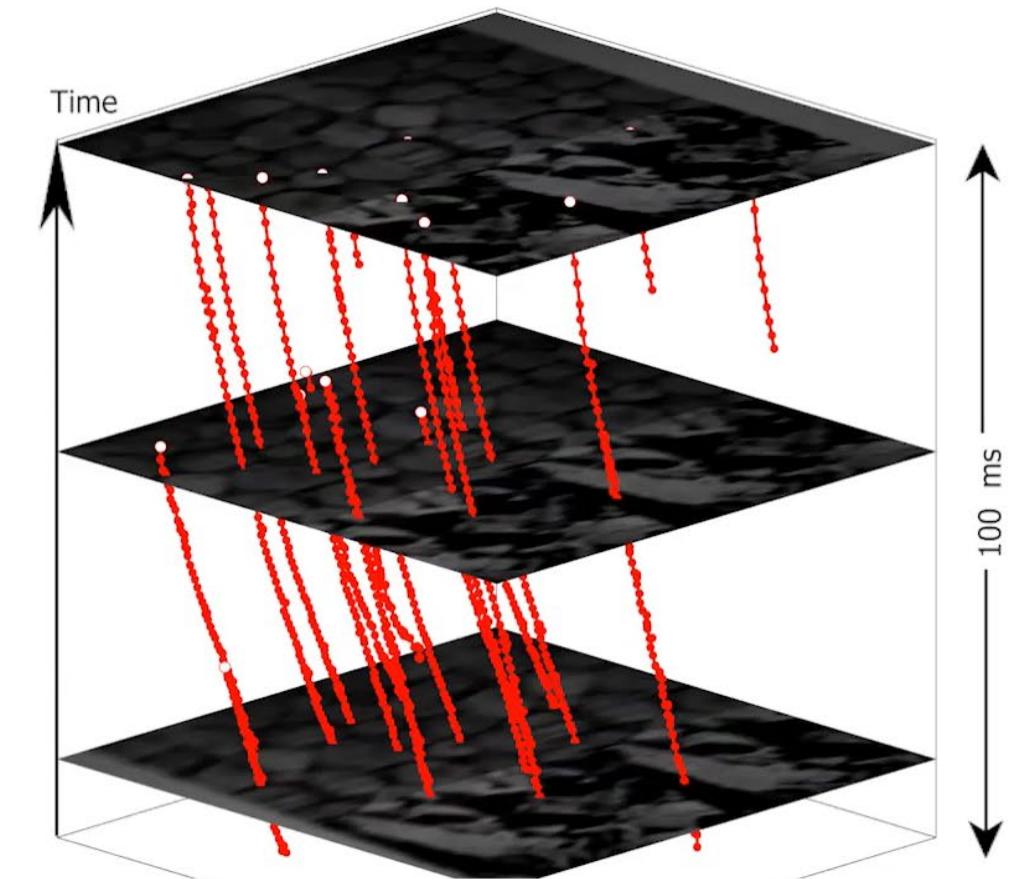


# Asynchronous Multi-Hypothesis Tracking of Events

Tracking under high-speed camera motion



**Feature Tracks**  
in image space

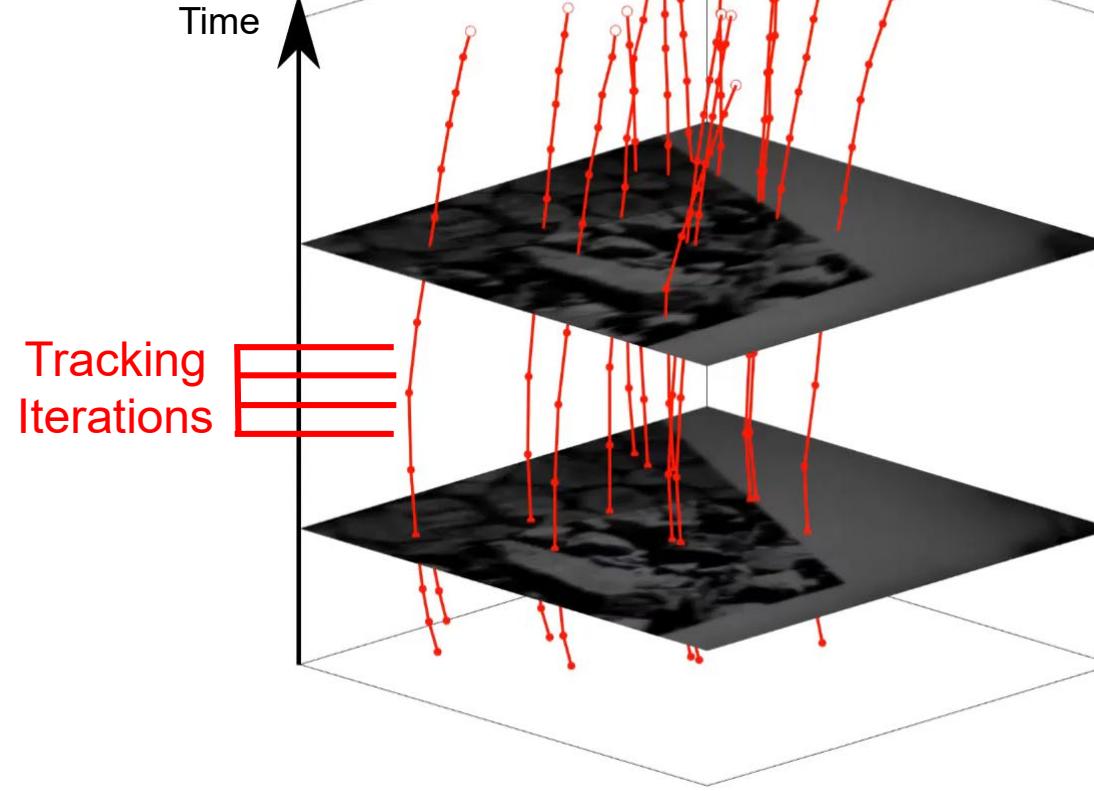


**Feature Tracks**  
in spatio-temporal space

# Asynchronous Event-Driven Feature Tracking

**Frame-like,  
Event-based Tracking**

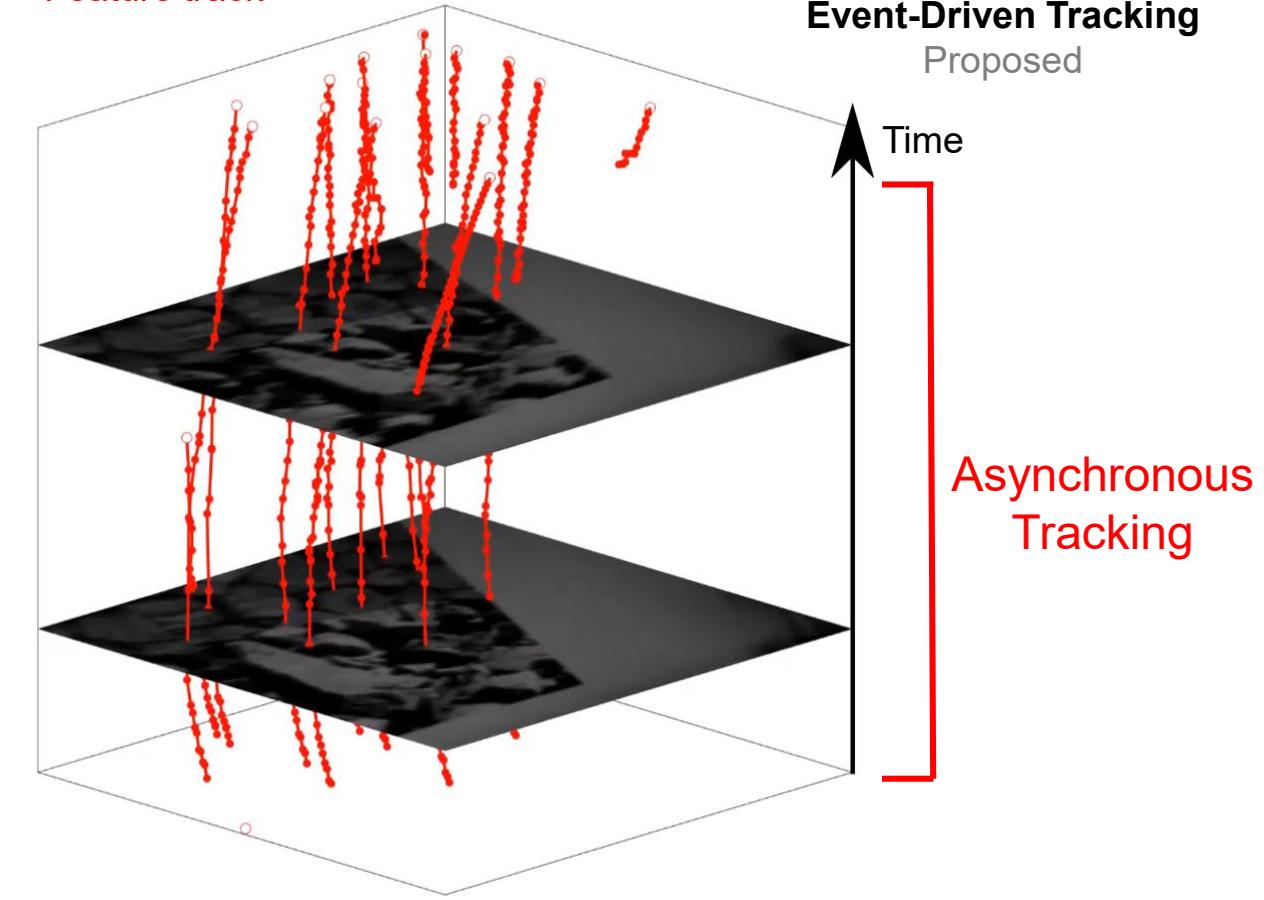
Zhu et al., ICRA'17



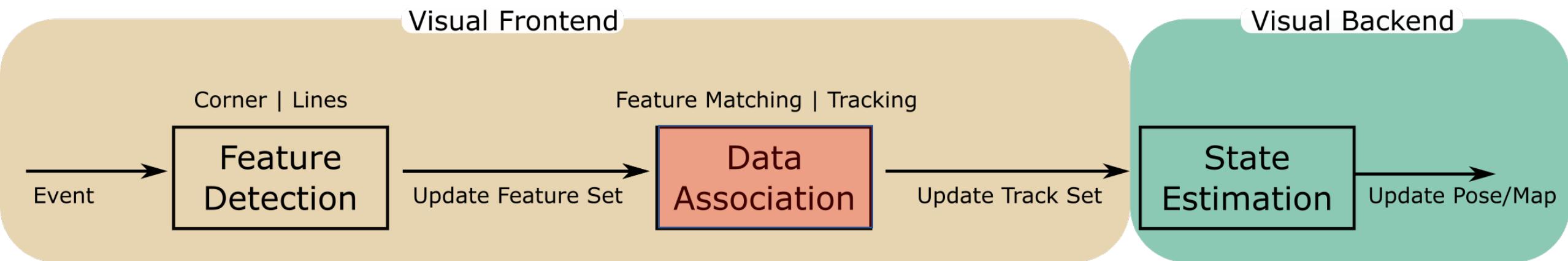
● Track update    — Feature track

**Asynchronous,  
Event-Driven Tracking**

Proposed



# Asynchronous Event-Driven SLAM Pipeline



- Event Corner Detection
- Naïve Event Corner Association
- Offline Corner Tracks Retrieval

**Asynchronous Corner Detection and Tracking for Event Cameras in Real-Time**  
[Alzugaray & Chli, RAL'18]

- Local Event Descriptor
- Multi-hypothesis Data Association
- Online Corner Track Retrieval

**ACE: An Efficient Asynchronous Corner Tracker for Event Cameras**  
[Alzugaray & Chli, 3DV'18]

- Hypothesis-based Optimization Framework
- Tracking directly on Raw Events
- Not Real-Time

**Asynchronous Multi-Hypothesis Tracking of Features with Event Cameras**  
[Alzugaray & Chli, 3DV'19]

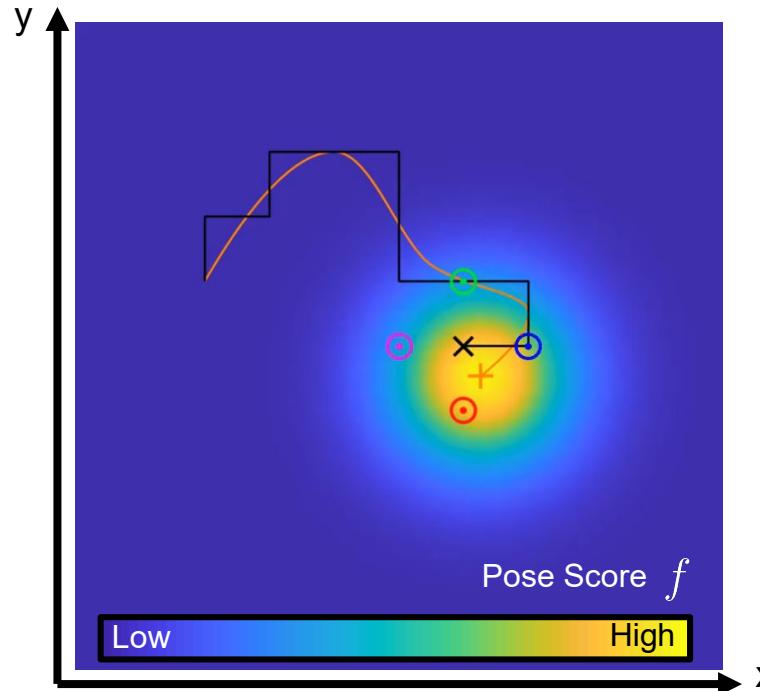
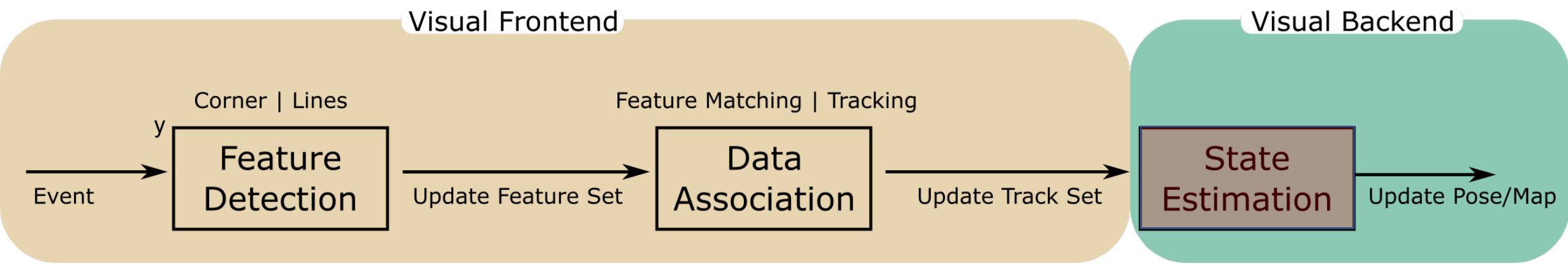
- Incremental Hypothesis-based Optimization
- Real-time capabilities.

**HASTE: multi-Hypothesis Asynchronous Speeded-up Tracking of Events**  
[Alzugaray & Chli, BMVC'20]

Publicly available:

[github.com/ialzugaray/arc\\_star\\_ros](https://github.com/ialzugaray/arc_star_ros)  
[github.com/ialzugaray/haste](https://github.com/ialzugaray/haste)

# Asynchronous Event-Driven SLAM Pipeline



- Hypothesis-based Optimization Framework
- Tracking directly on Raw Events
- Not Real-Time

**Asynchronous Multi-Hypothesis Tracking of Features with Event Cameras**  
[Alzugaray & Chli, 3DV'19]

- Incremental Hypothesis-based Optimization
- Real-time capabilities.

**HASTE: multi-Hypothesis Asynchronous Speeded-up Tracking of Events**  
[Alzugaray & Chli, BMVC'20]

# On Event-Driven Perception

- Natural to event cameras
  - Exploit Sparsity & Asynchronicity
- Reduce the number of assumptions
  - Motion-speed tuning / Event-window tuning
- Algorithms require careful design
  - Efficiency / Robustness / Scalability

