THE UNIVERSITY of NORTH CAROLINA at CHAPEL HILL

ISMAR 2016
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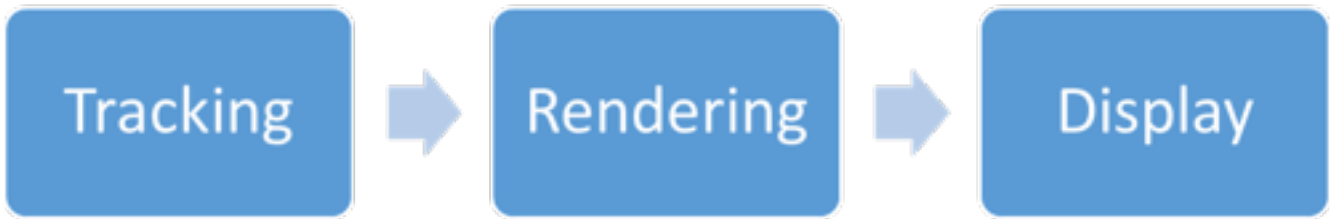
STEVENS INSTITUTE of TECHNOLOGY
THE INNOVATION UNIVERSITY

Towards kHz 6-DoF Visual Tracking Using an Egocentric Cluster of Rolling Shutter Cameras

Akash Bapat[1], Enrique Dunn[1,2] & Jan-Michael Frahm[1],
UNC Chapel Hill, USA[1],
Stevens Institute of Technology, USA[2]

AR/VR System Components



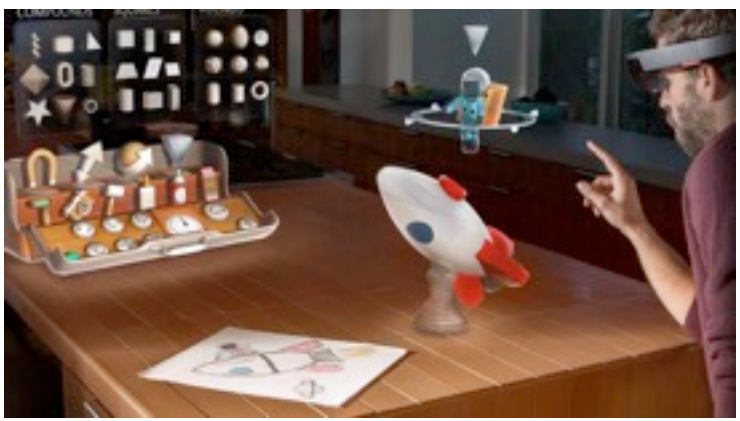


Image source : http://static3.gamespot.com/uploads/original/1179/11799911/2786939-hololens7.jpg

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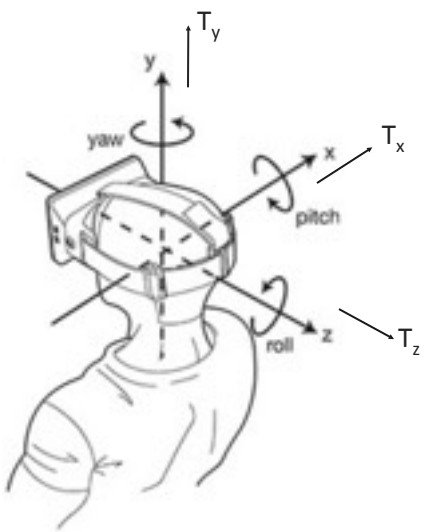
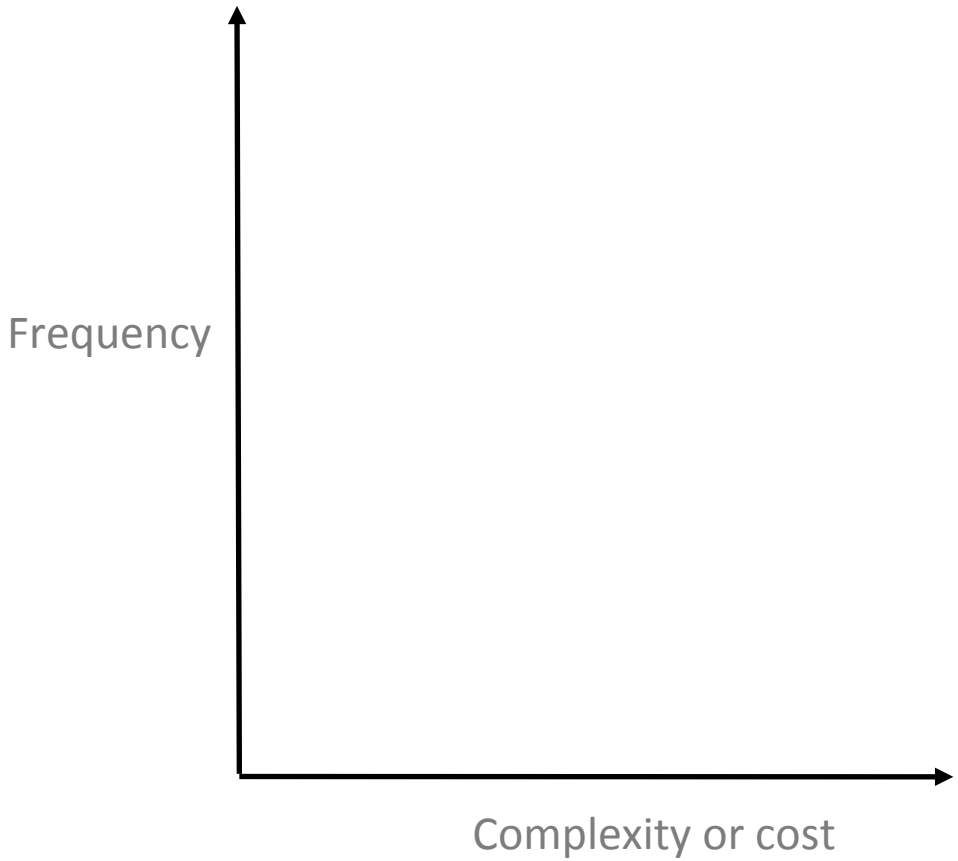


Find 6-DoF head pose[1]

1. LaValle etal, Head Tracking for Oculus Rift

Tracking Systems



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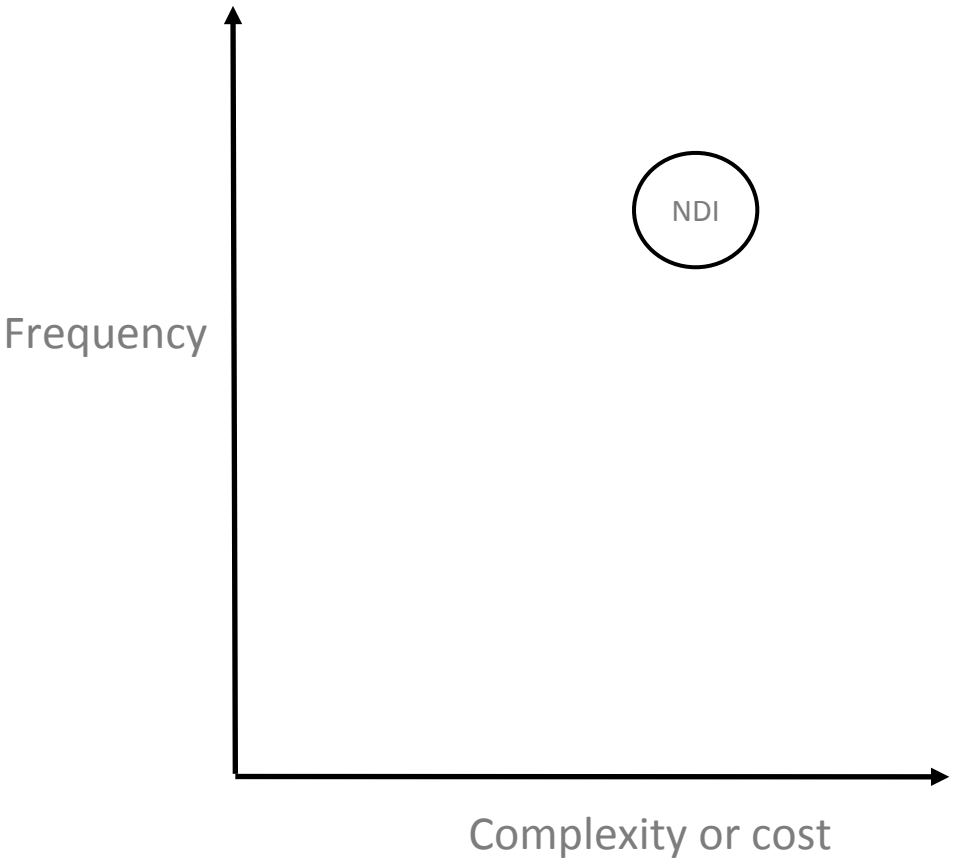

NDI Optotrak Certus[1]

- Tracking frequency ≈ 5kHz
- State of the art

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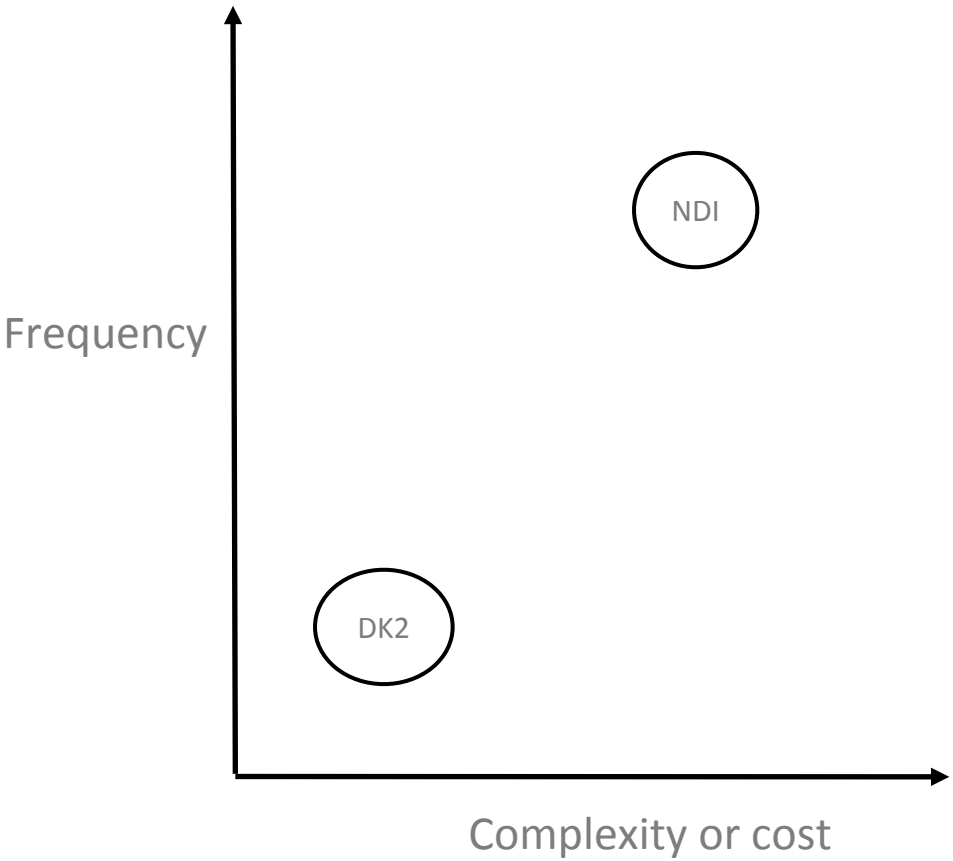
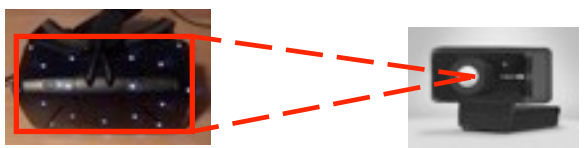


Oculus Rift DK2

Orientation tracking
IMU at 1000 Hz [4,5]



Positional tracking [2,3]
Camera fps = 60 fps

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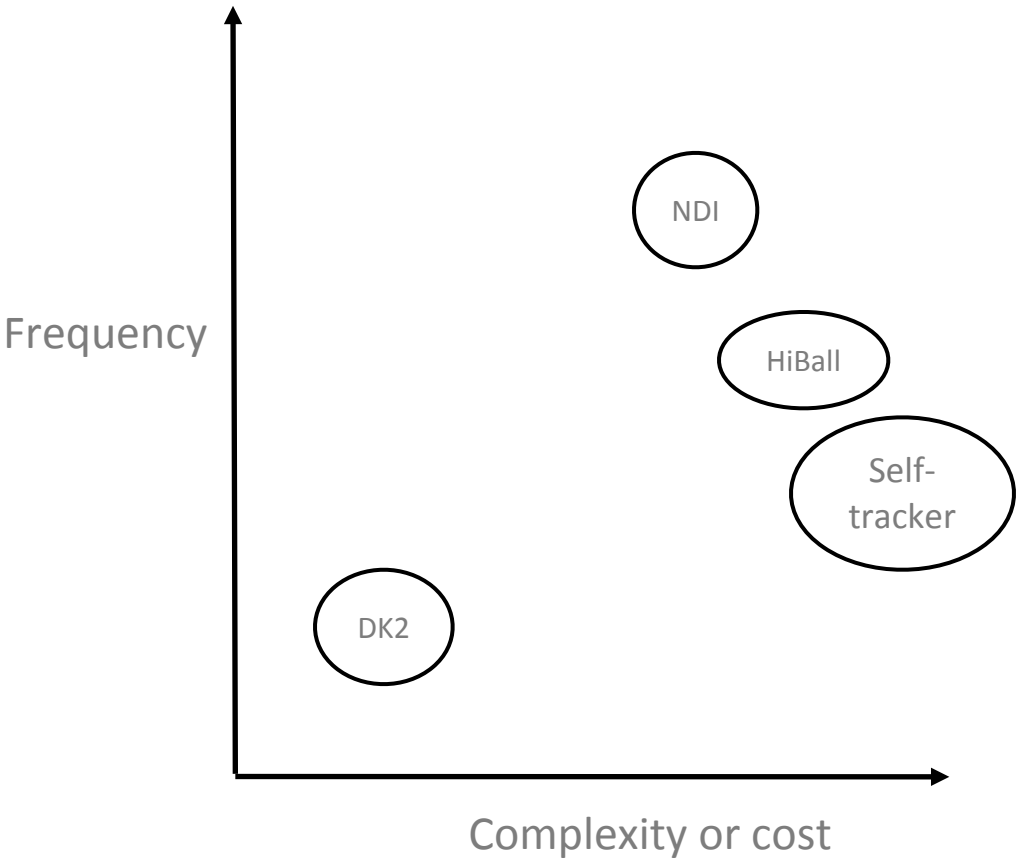

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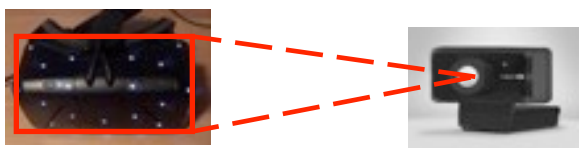

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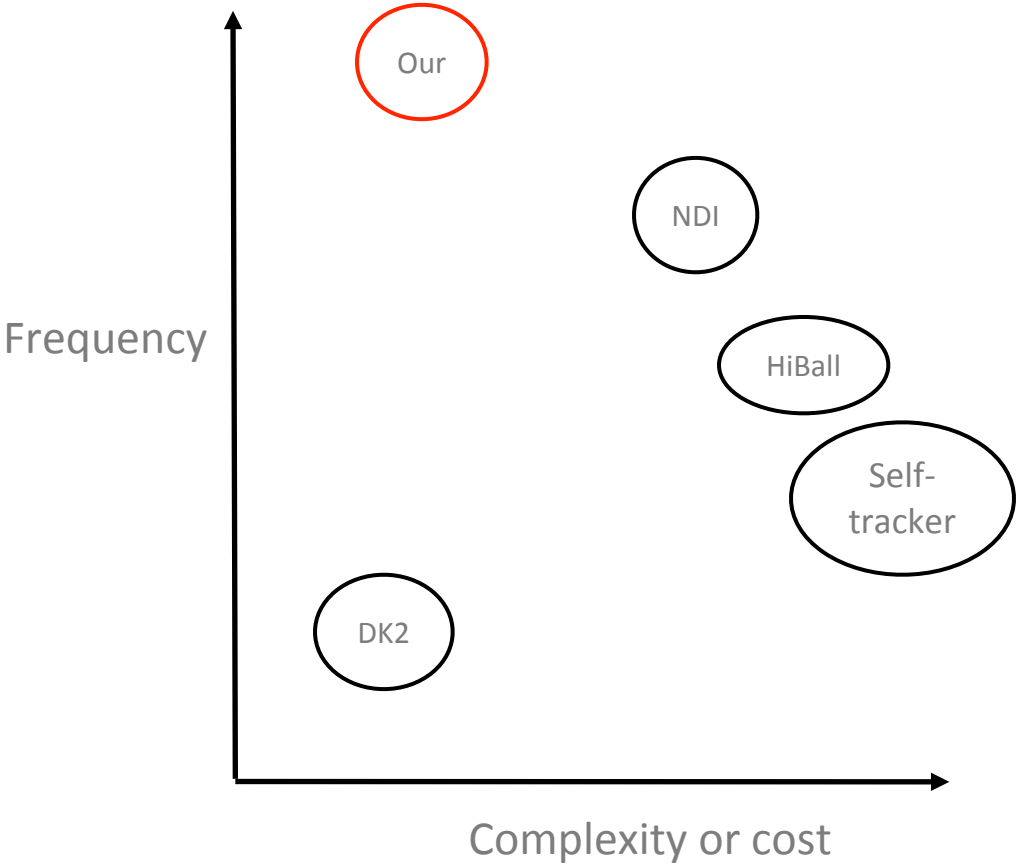

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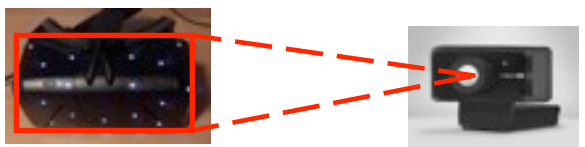

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Rolling Shutter

Rolling shutter capture[1]

1. http://jasmcole.com/2014/10/12/rolling-shutters/

Rolling Shutter

- Row-by-row acquisition of linescan snapshots at slightly different times

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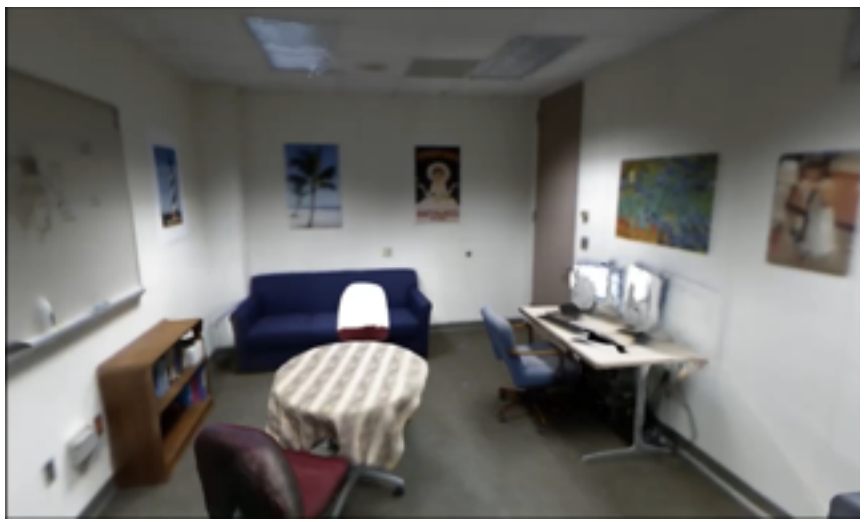
Rolling shutter capture[1]

- Stream of row-images



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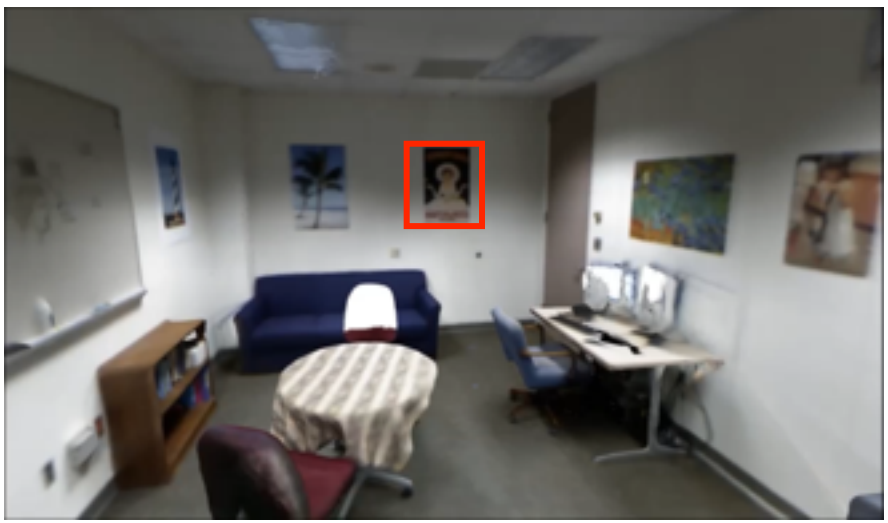

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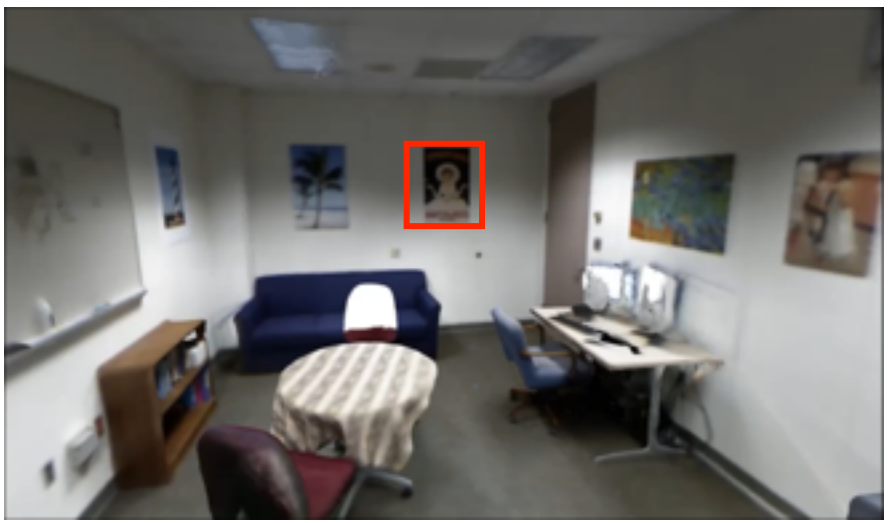

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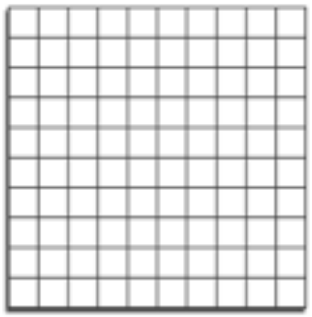
Stream of Rows

- Frequency of row-samples
  - F = FPS * Height
    = 120 * 720 > 80kHz

Rolling shutter[1]



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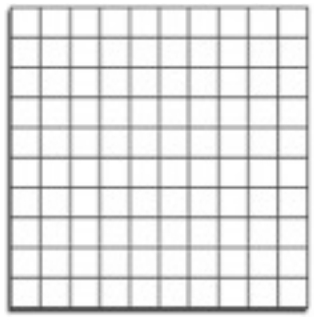

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Our Tracker

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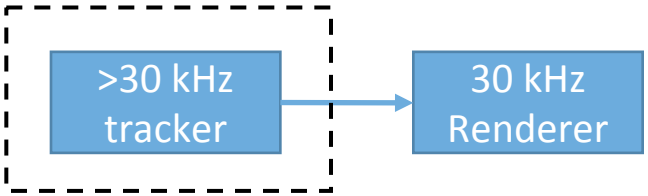
Our Tracker

- Enabling component for rendering
  1. Zheng et al., 2014
  2. Lincoln et al., 2016



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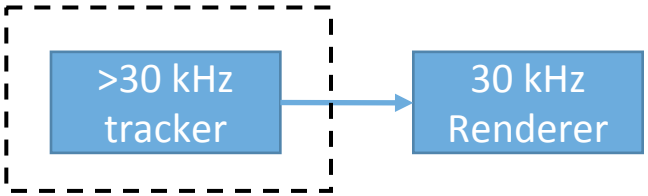

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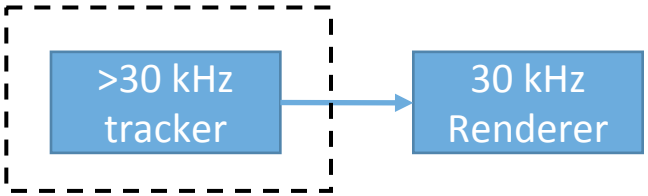

- Use commodity cameras

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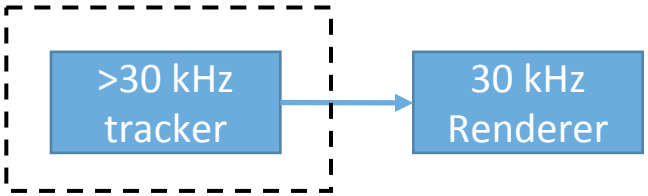

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GoPro[1]

- Tracking frequency
  - Up to 80 kHz
- Break frame-rate barrier
  - Process each row of image

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Related Work

Artifact visualisation[1]

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Related Work

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- Removing rolling shutter
  - Forssen et al., CVPR 2010
  - Track points using KLT tracker, estimate rotation
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  - Multi-view stereo: Saurer et al., ICCV 2013
  - Adapt plane sweep stereo for rolling shutter

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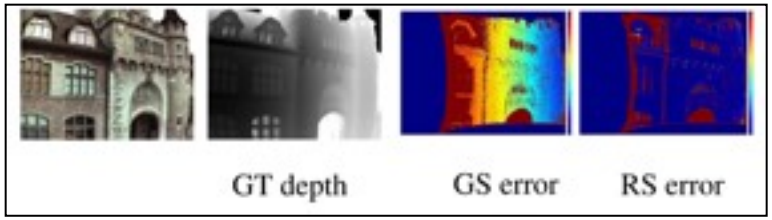

- Velocity estimation
  - Ait-Aider et al., ICVS 2006
  - Solve for pose and velocity using bundle adjustment
  - Use 2D-3D correspondence, similar to camera calibration

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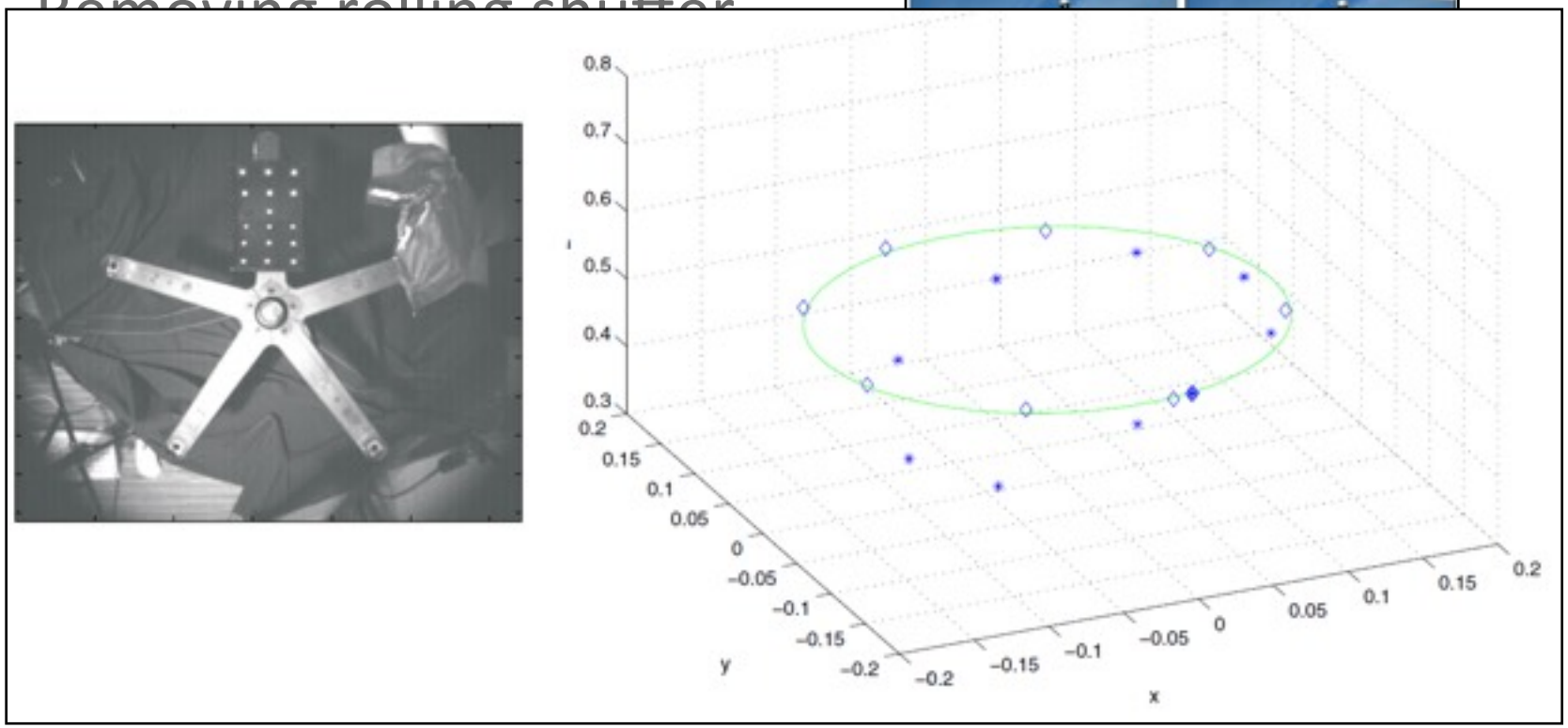

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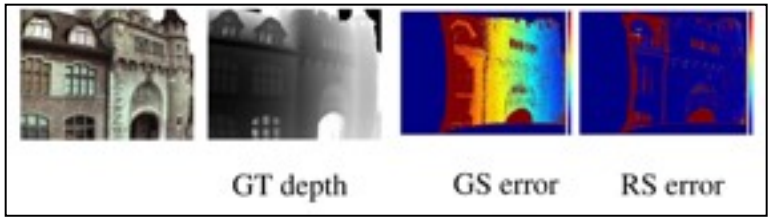

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Approach Overview

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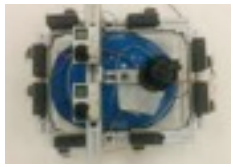
Camera Cluster



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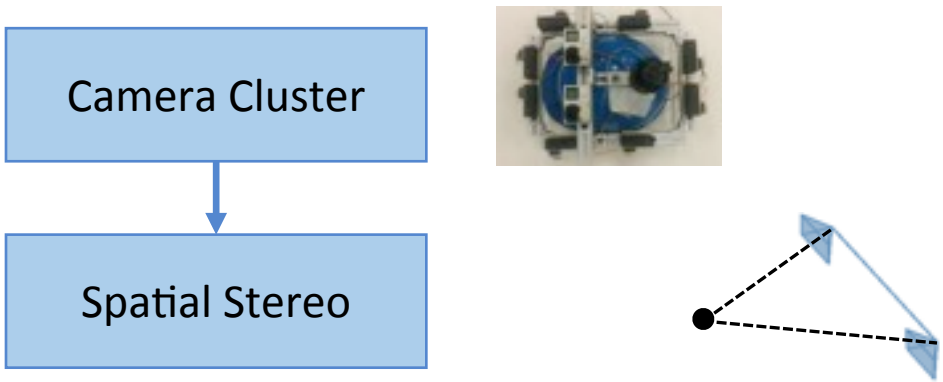

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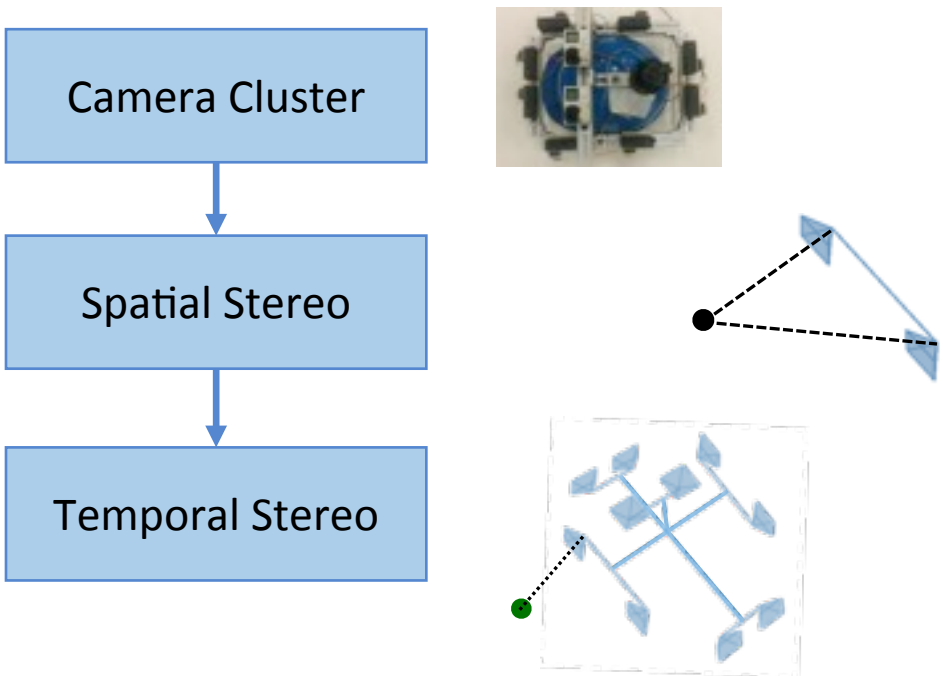

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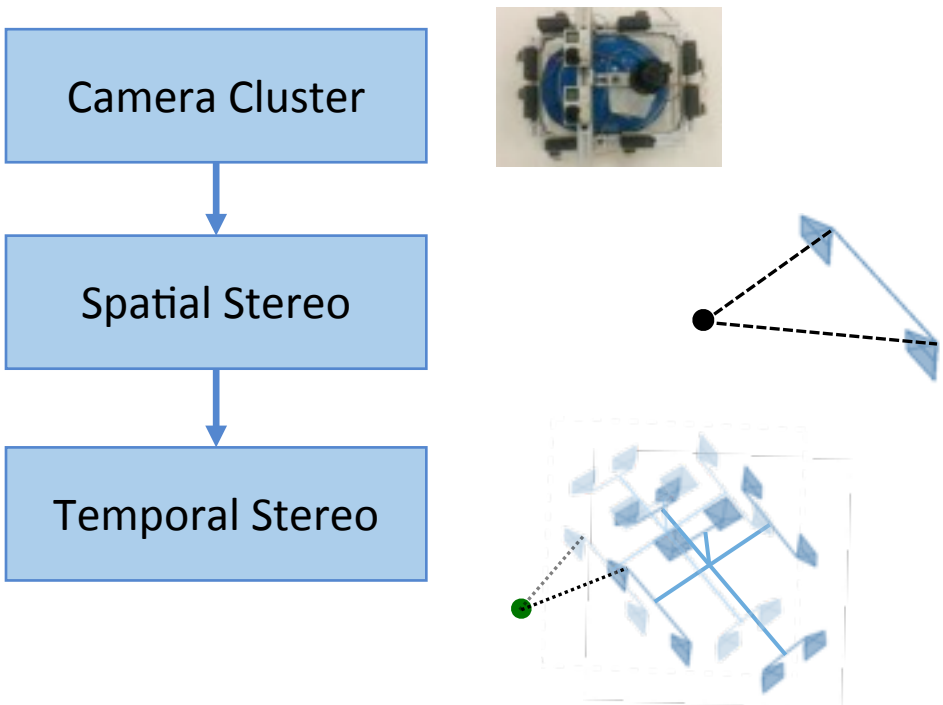

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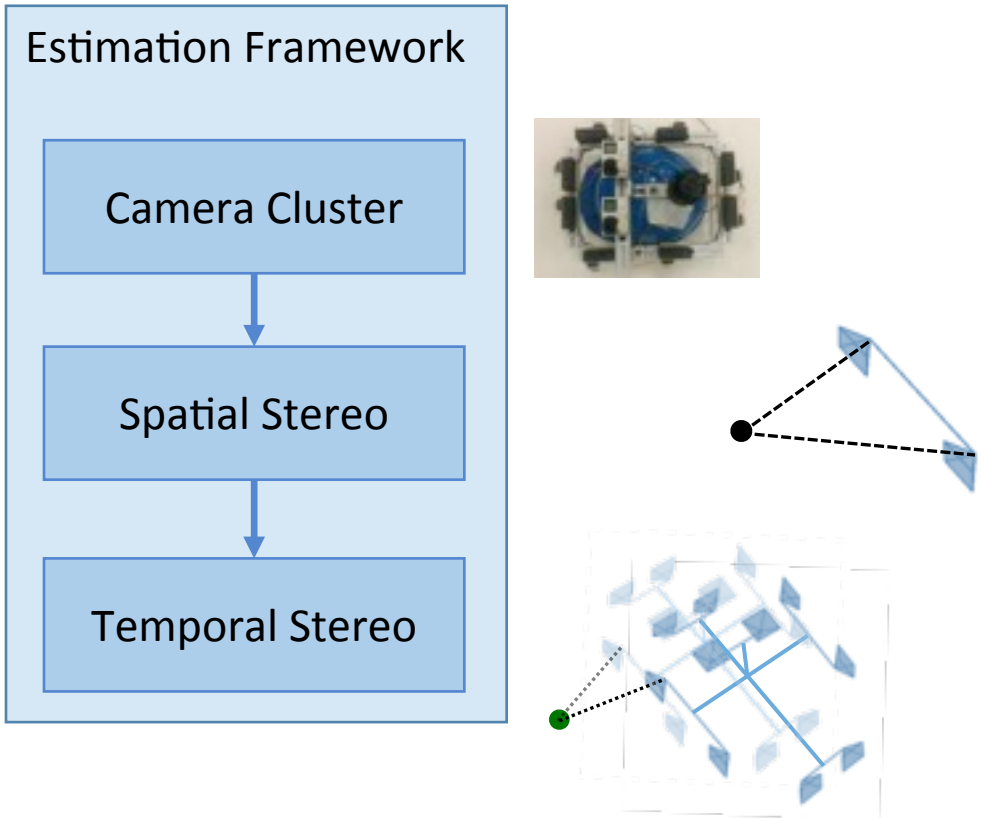

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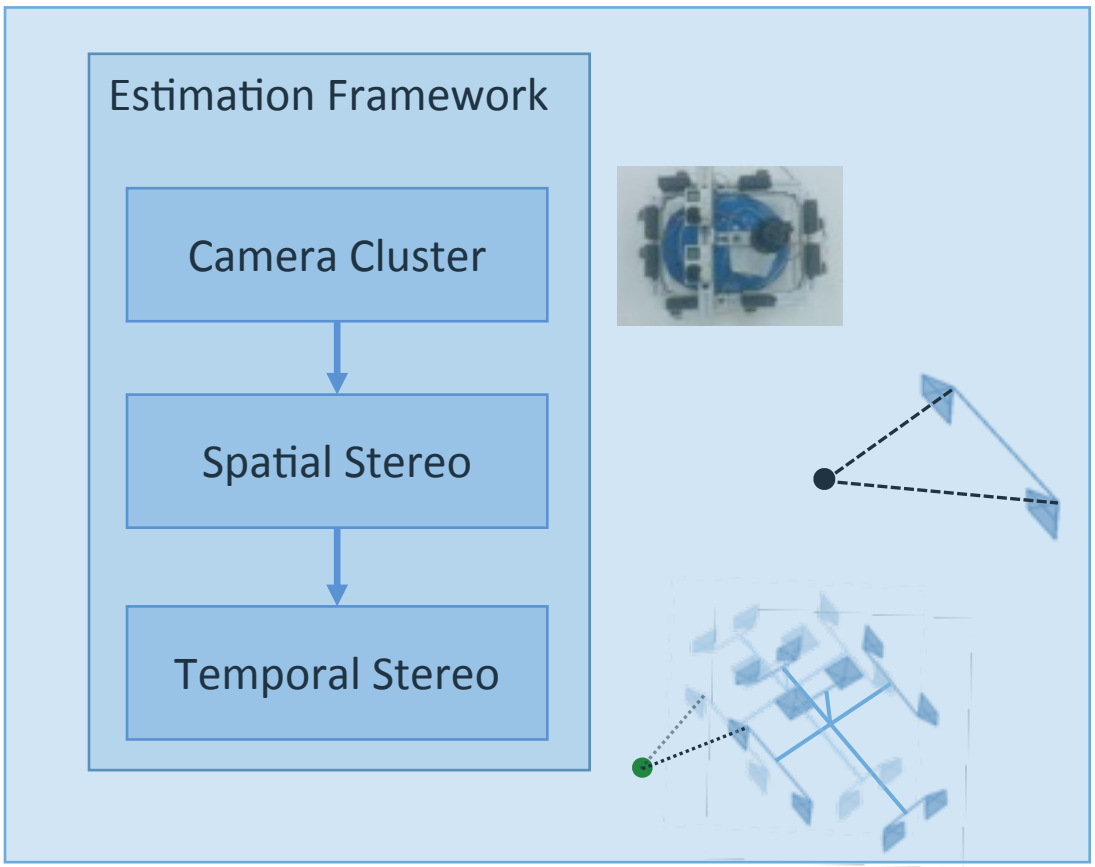

Our approach

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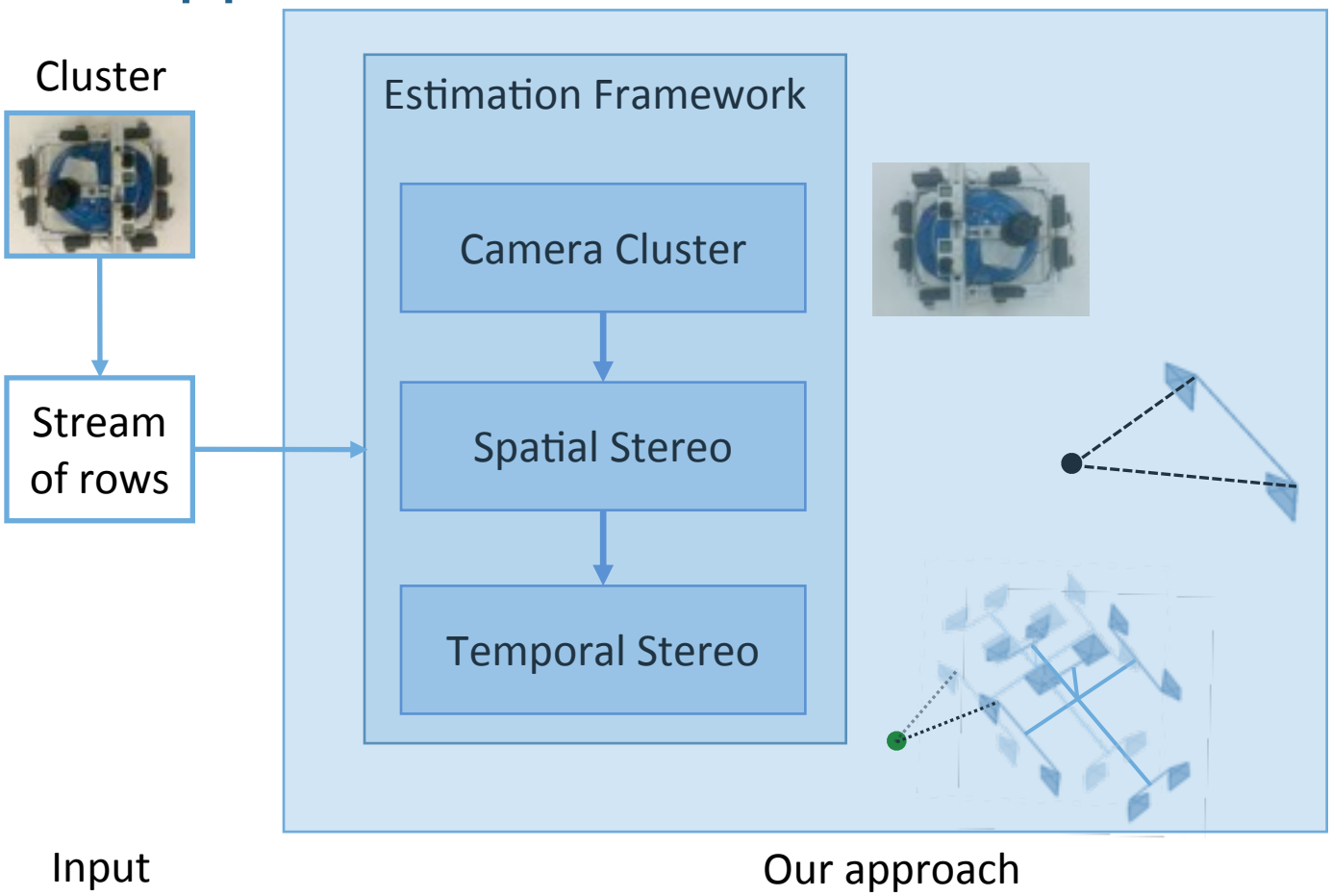

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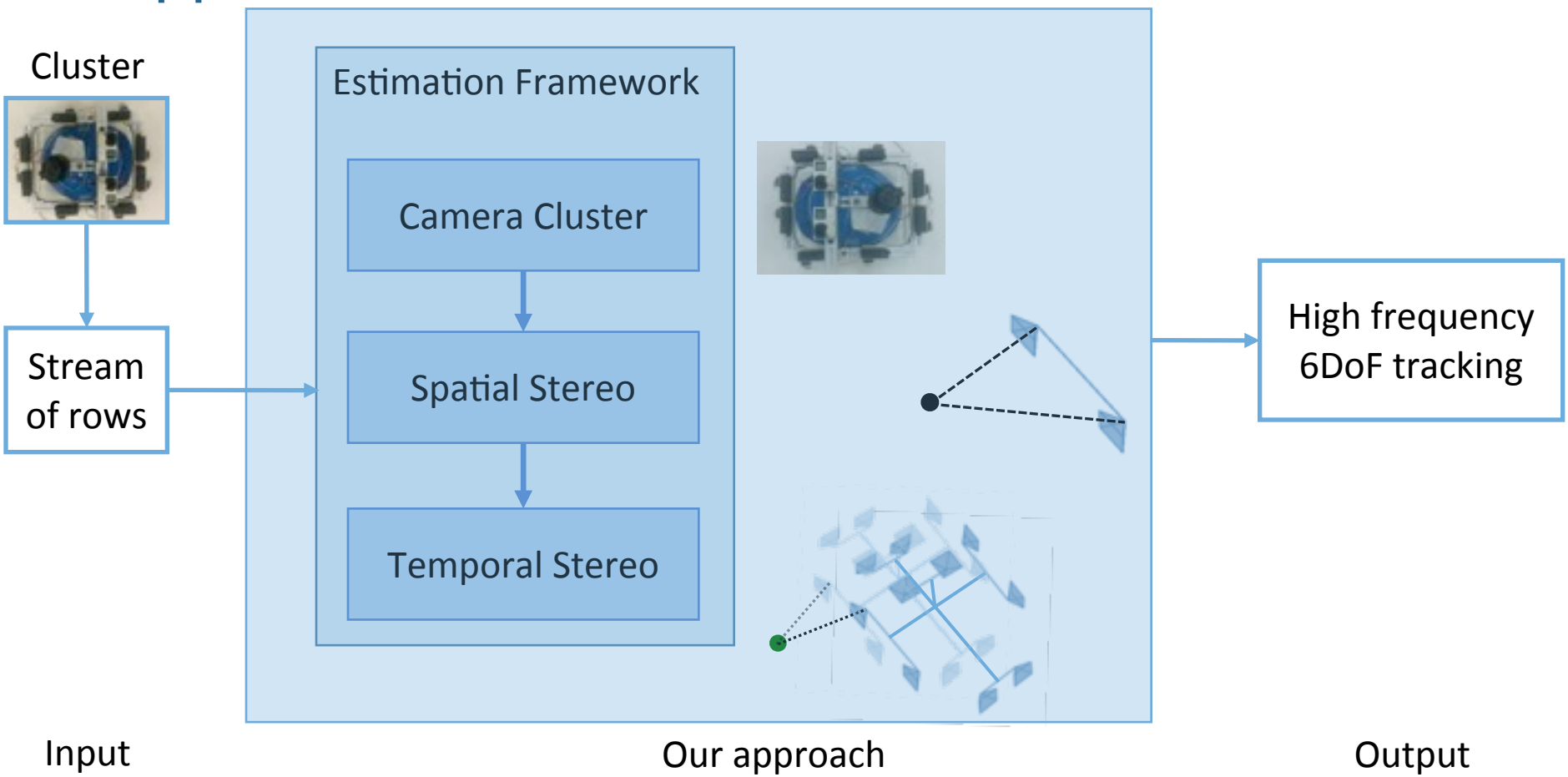

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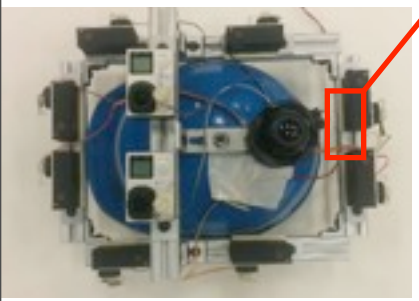

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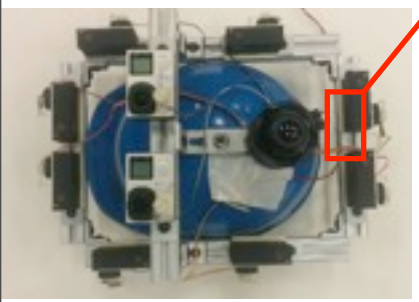

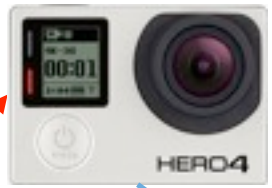

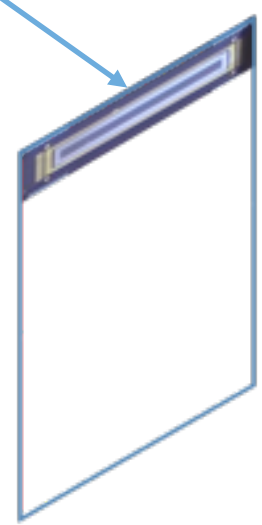

Dynamic sensor with periodic movement[2]

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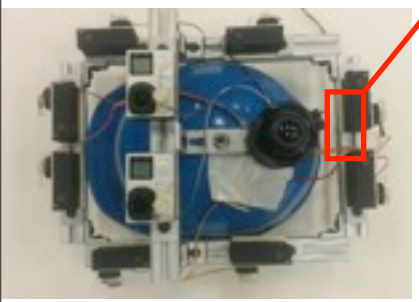

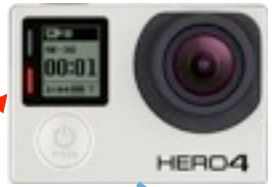

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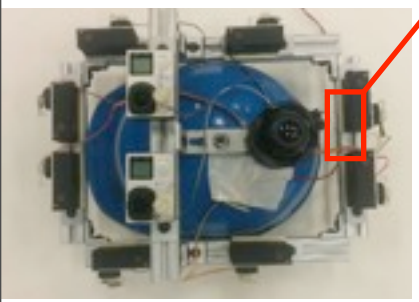

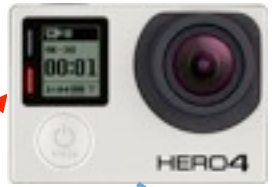

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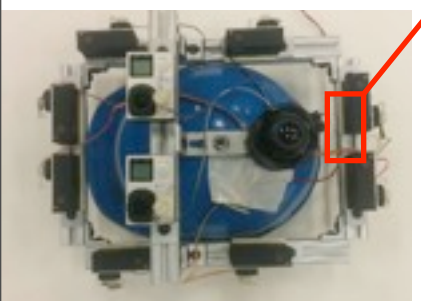
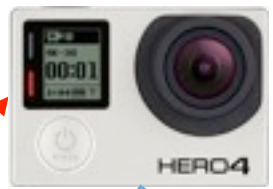
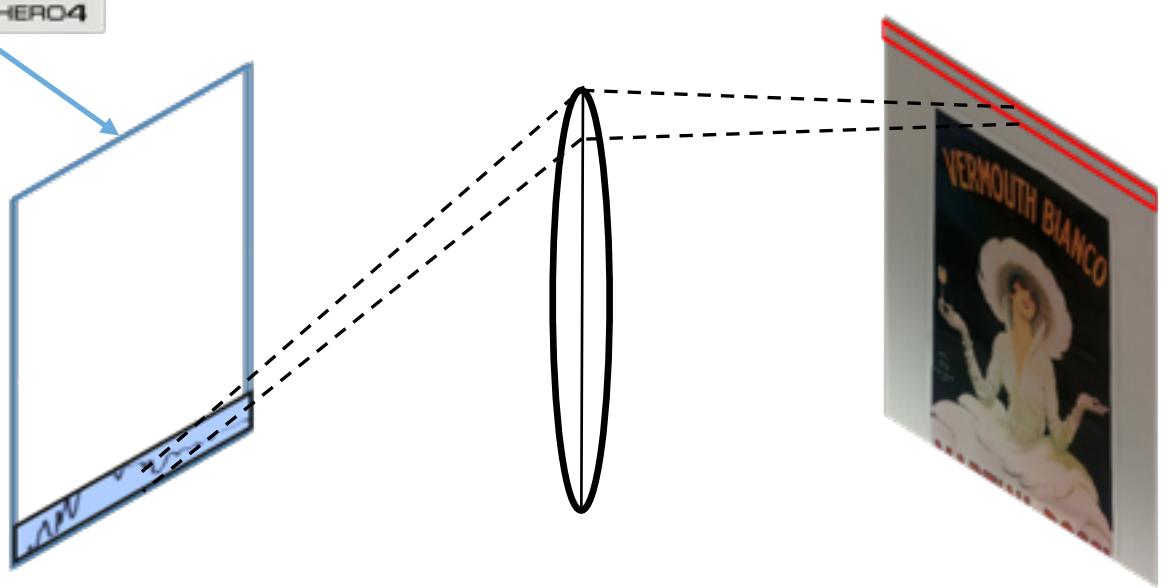




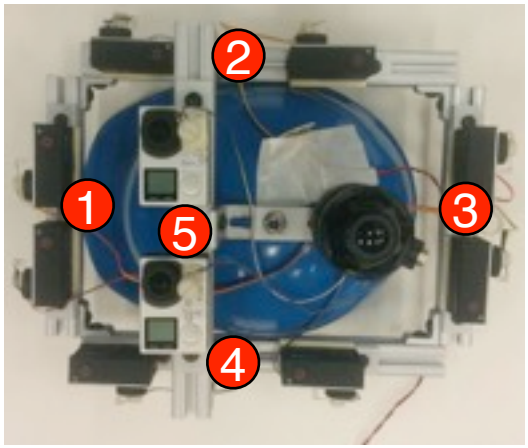


Dynamic sensor with periodic movement[2]

Small vertical FoV

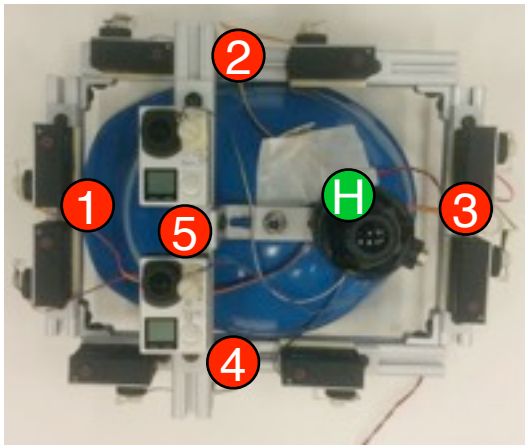
1. http://shop.gopro.com/hero4/hero4-black/CHDHX-401.html
2. http://users.erols.com/njastro/faas/image001/CCD2.jpg

Our Cluster



Our Cluster

- N stereo-pairs of rolling shutter camera, N=5
  - Precalibrated intrinsic and extrinsics
  - Temporal sync
  - Known history of motion
- H Hi-Ball for ground truth



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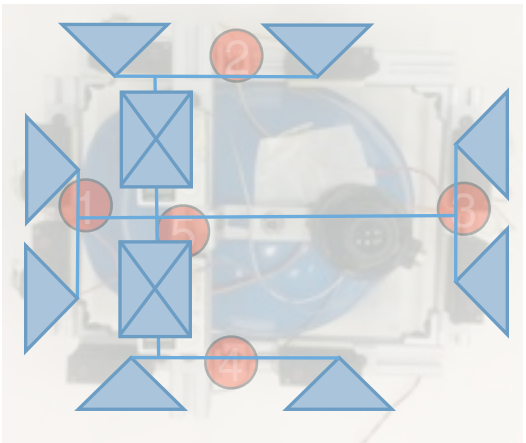


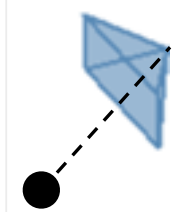
Geometry of cluster

Motion Model

$$X^{t+1}_{cam_i} = M_{cam_i} V^t$$

3D point

$X^t_{cam_i}$



3D point X

Motion Model

3D point

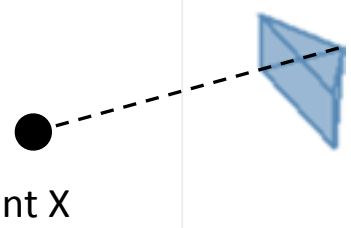
$$X_{cam_i}^{t+1} = M_{cam_i} X_{cam_i}^{t}$$

Motion Model

$$X_{cam_i}^{t+1} = M_{cam_i} X_{cam_i}^{t}$$

3D point

Motion of camera



3D point X

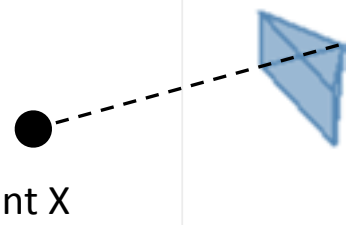
Motion Model

Moved 3D point

3D point

$$X_{cam_i}^{t+1} = M_{cam_i} X_{cam_i}^{t}$$

Motion of camera



3D point X

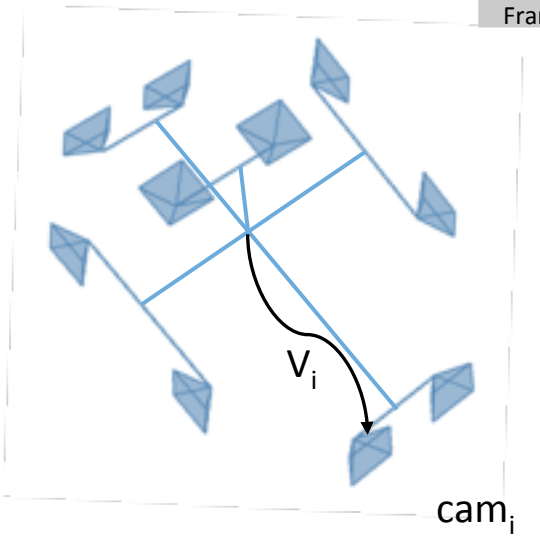
Motion Model

Moved 3D point

3D point

$$X^{t+1}_{cam_i} = M_{cam_i} X^{t}_{cam_i}$$

$$X^{t+1}_{cam_i} = V_i \, M_{cluster} \, V_i^{-1} \, X^{t}_{cam_i}$$



RS cluster

Linearized Motion Model

$$X_{cam_i}^{t+1} = M_{cam_i} X_{cam_i}^{t}$$

$$X_{cam_i}^{t+1} = V_i M_{cluster} V_i^{-1} X_{cam_i}^{t}$$

$$M_{cluster} = \begin{bmatrix} R(\theta_x, \theta_y, \theta_z) & T \\ 0 & 1 \end{bmatrix}$$

Linearized Motion Model

$$X_{cam_i}^{t+1} = M_{cam_i} X_{cam_i}^{t}$$

$$X_{cam_i}^{t+1} = V_i \, M_{cluster} \, V_i^{-1} \, X_{cam_i}^{t}$$

$$M_{cluster} = \begin{bmatrix} R(\theta_x, \theta_y, \theta_z) & T \\ 0 & 1 \end{bmatrix} \Rightarrow dM = \begin{bmatrix} 1 & -\theta_z & \theta_y & \delta T_x \\ \theta_z & 1 & -\theta_x & \delta T_y \\ -\theta_y & \theta_x & 1 & \delta T_z \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Linearized Motion Model

$$X_{cam_i}^{t+1} = M_{cam_i} X_{cam_i}^{t}$$

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Linearized Motion Model

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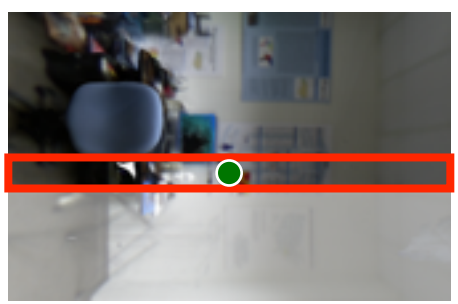
Motion Estimation

$$X_{cam_i}^{t+1} = V_i \; M_{cluster} \; V_i^{-1} \; X_{cam_i}^{t}$$

$$X_{cam_i}^{t} = [0 \quad y \quad d \quad 1]^T$$



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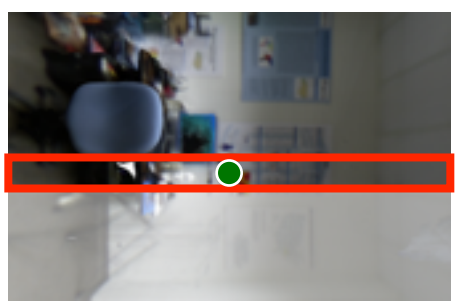
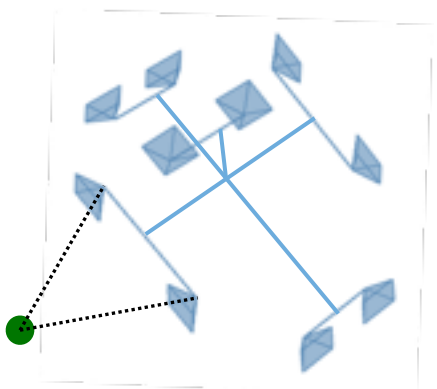
$$X_{cam_i}^{t+1} = V_i \; M_{cluster} \; V_i^{-1} \; X_{cam_i}^{t}$$

$$X_{cam_i}^{t} = [0 \quad y \quad d \quad 1]^T$$

Stereo in space

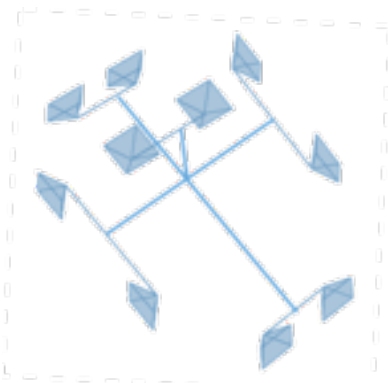




Motion Estimation

$$X_{cam_i}^{t+1} = [\Delta x \quad y + \Delta y \quad d + \Delta d \quad 1]^T$$

$$X_{cam_i}^{t+1} = V_i \; M_{cluster} \; V_i^{-1} \; X_{cam_i}^{t}$$



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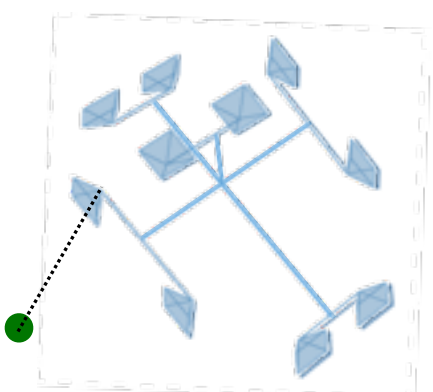

Motion Estimation

$$X_{cam_i}^{t+1} = [\Delta x \quad y + \Delta y \quad d + \Delta d \quad 1]^T$$

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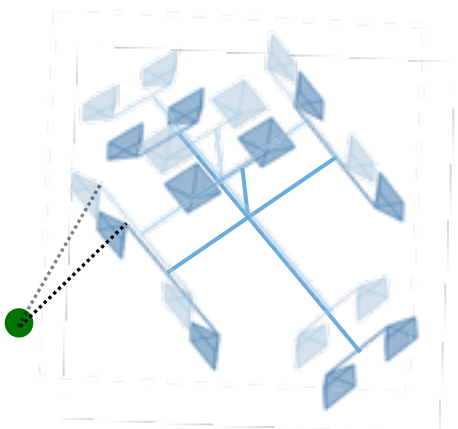
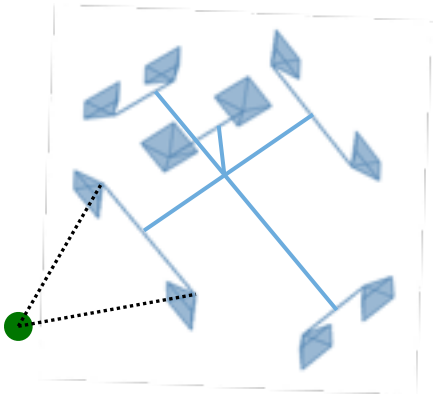


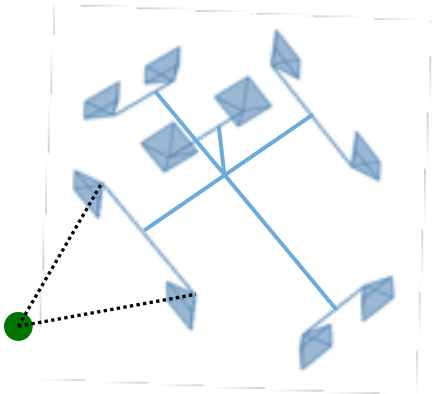
Stereo Estimation



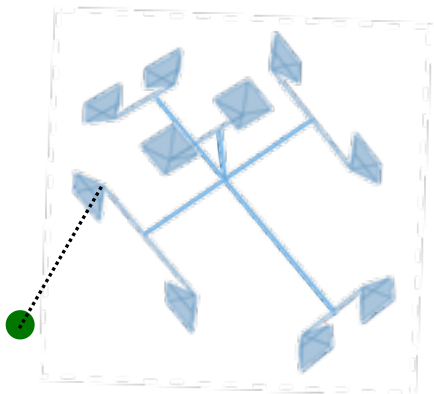
Cluster
Spatial Stereo
Temporal Stereo
Estimation Framework

Stereo Estimation

- Spatial stereo s_d
  - Use stereo-pair of cameras
  - Measure pixel disparity



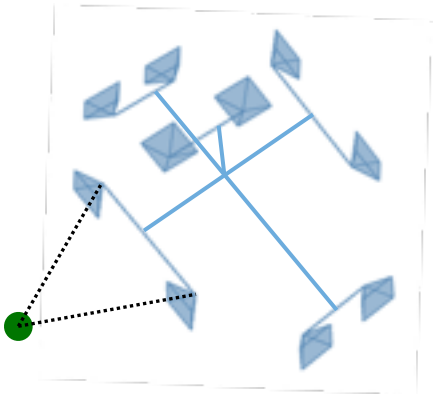
- Temporal stereo s_t
  - Measure small shifts in pixel
  - Compare row at time t and t-k



t1 ⟶ t2

Stereo Estimation

- Spatial stereo s_d
  - Use stereo-pair of cameras
  - Measure pixel disparity



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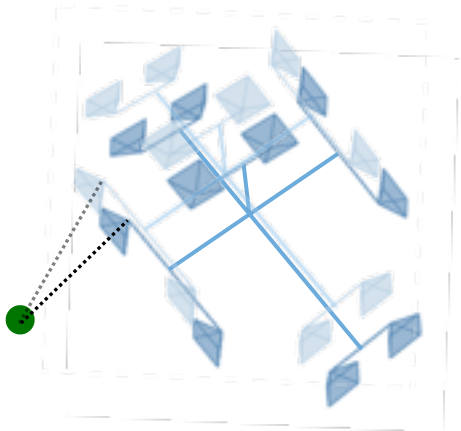
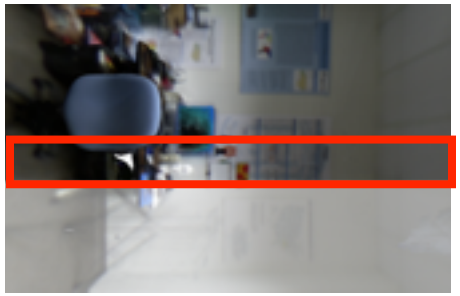
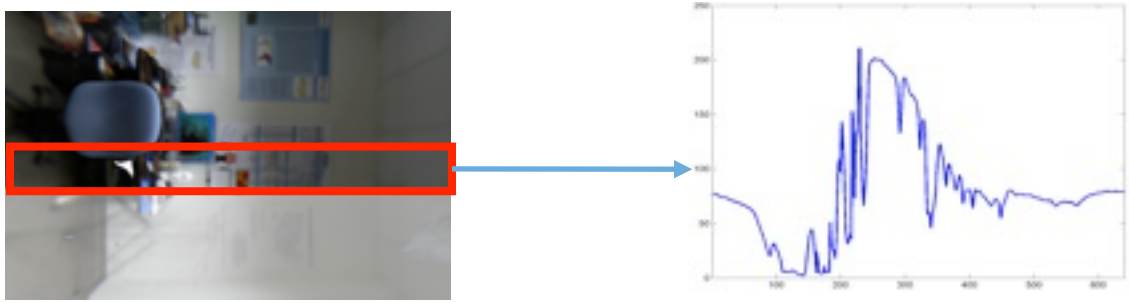


t1 → t2

Binary Row Descriptor

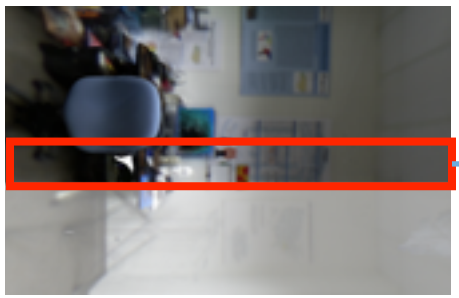


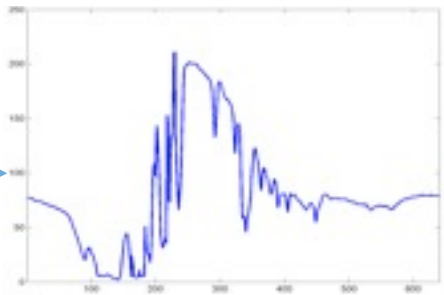
Binary Row Descriptor



Visualization of row-image (296^{th})

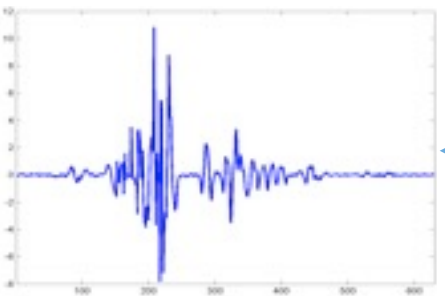
Binary Row Descriptor





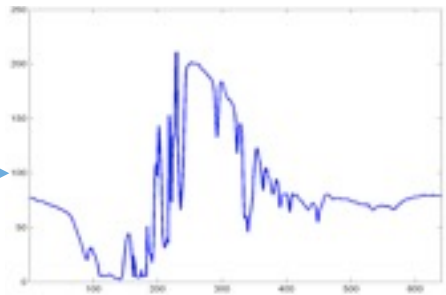
Visualization of row-image (296th)

$$\frac{d^2G(x,0,\sigma)}{dx^2} * row$$



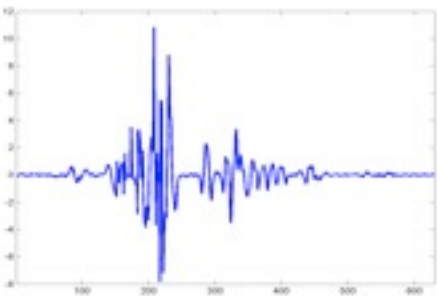
2nd derivative

Binary Row Descriptor



Visualization of row-image (296th)

Sign



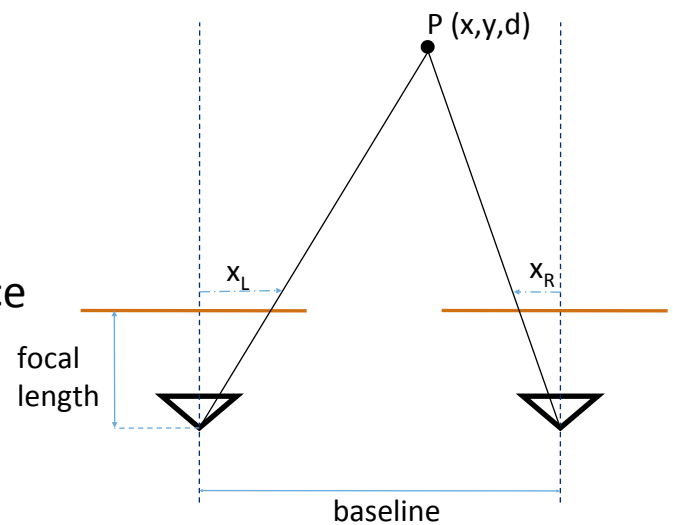
2nd derivative

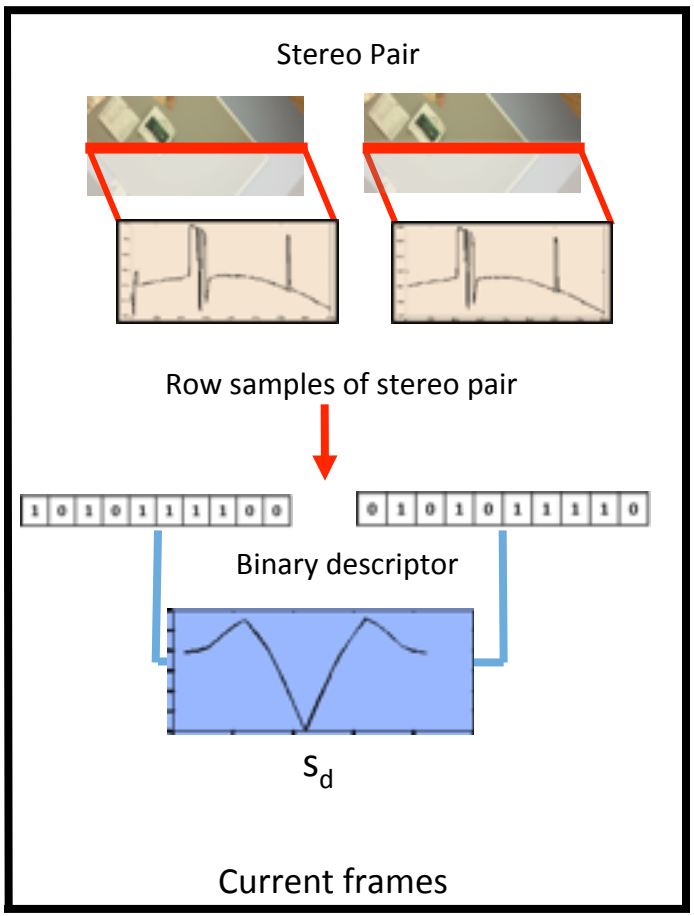
$$\frac{d^2G(x,0,\sigma)}{dx^2} * row$$

Spatial Stereo : Measure Disparity

- Compare binary descriptor of rows of stereo cameras
- Fast hamming cost matching
- $depth = \frac{focal * baseline}{disparity}$

Stereo in space





Temporal Stereo: Measure Shift s_t

- Different regions in space are captured at different timestamps
- Stereo in time
  - Need snaps of same space but at different timestamps

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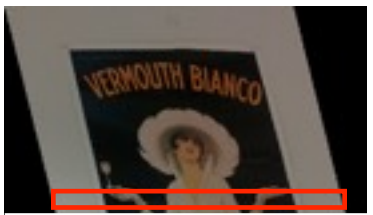
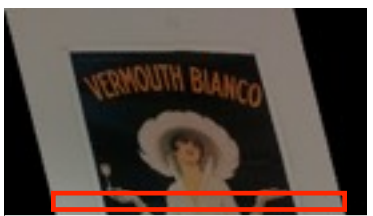


Current time

Temporal Stereo: Measure Shift s_t

- Different regions in space are captured at different timestamps
- Stereo in time
  - Need snaps of same space but at different timestamps

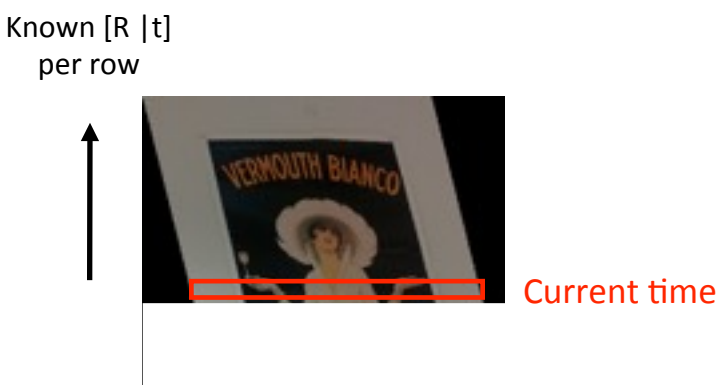
Previous Frame



Current time

Temporal Stereo: Measure Shift s_t

- Different regions in space are captured at different timestamps
- Stereo in time
  - Need snaps of same space but at different timestamps



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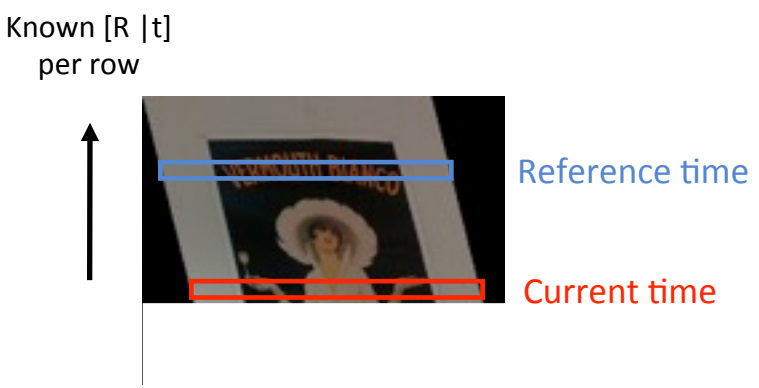

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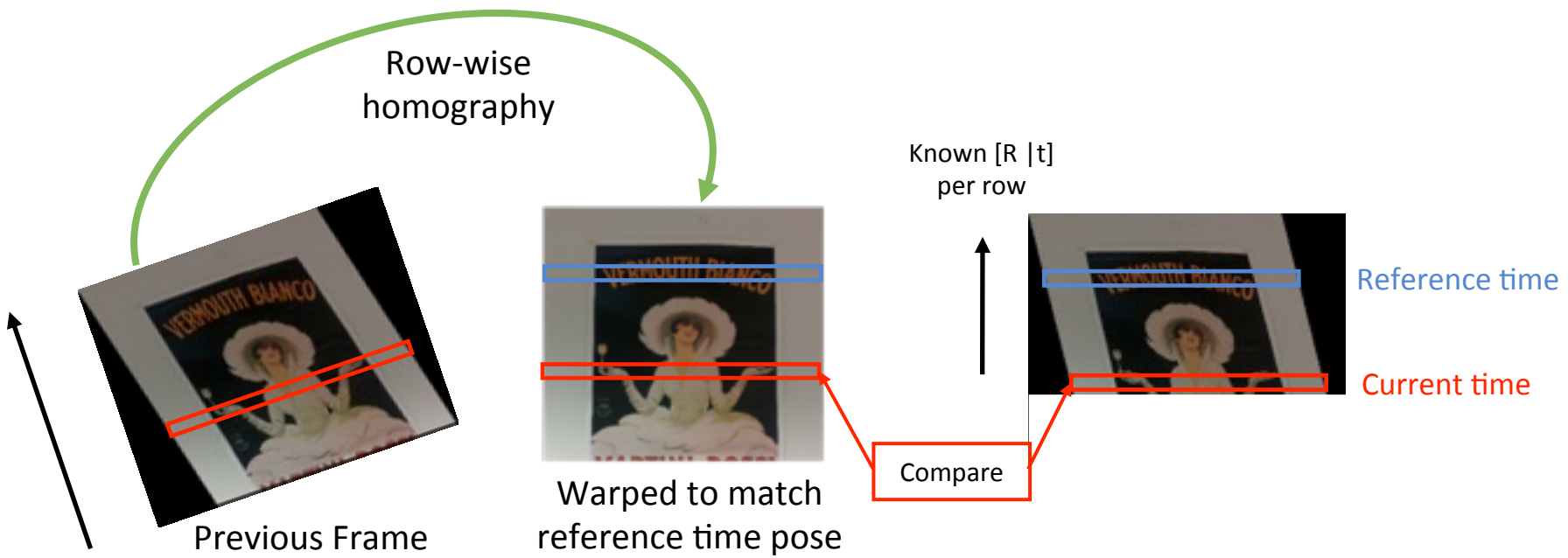
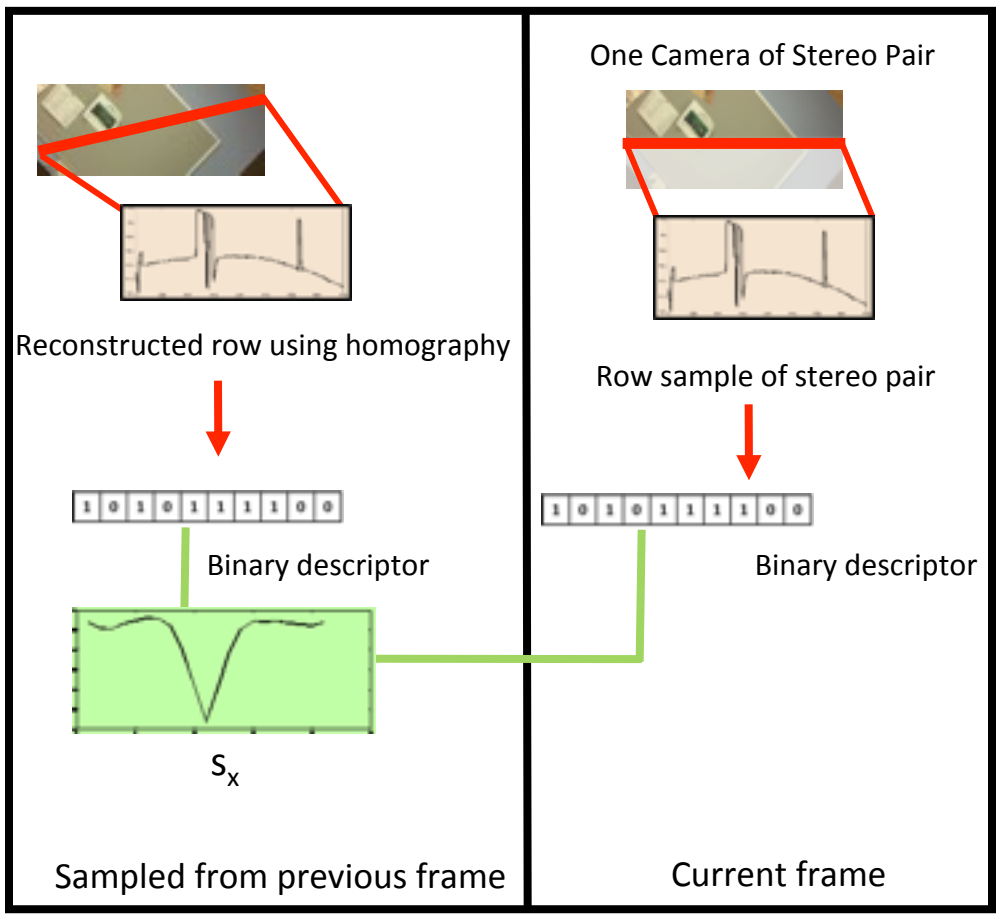
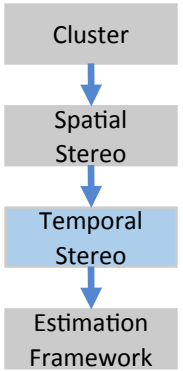


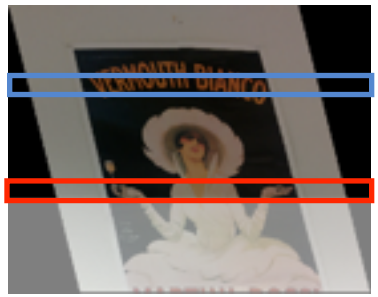
Temporal Stereo: Measure Shift s_t

- Reconstruct row using per- row homography
- Use binary descriptors and hamming distance

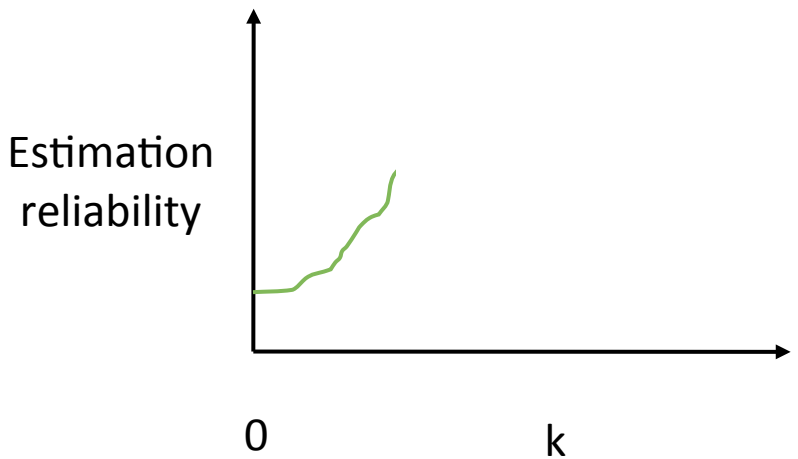


Adaptive Reference



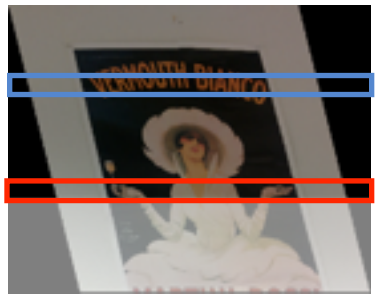


Current frame



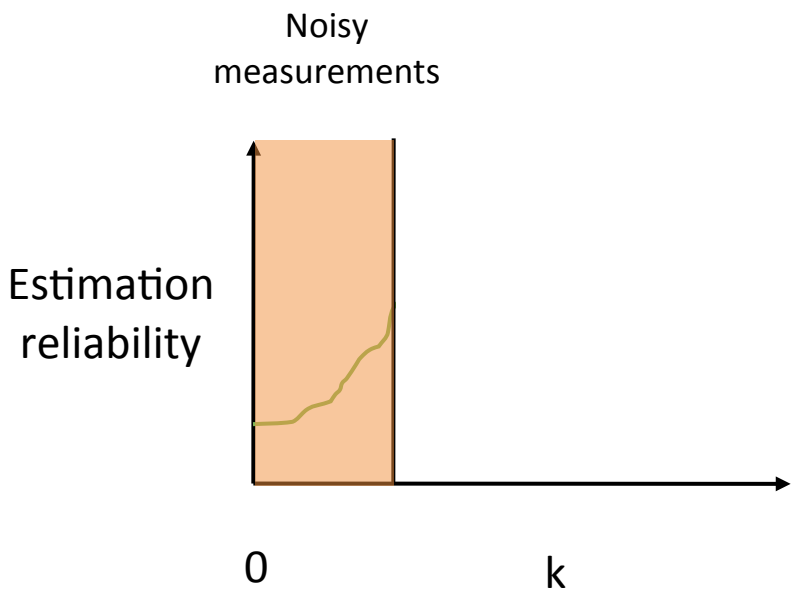
Adaptive Reference

- Leave sufficient motion
  - Row-to-row motion
  - Interpolation & pixel measurement noise



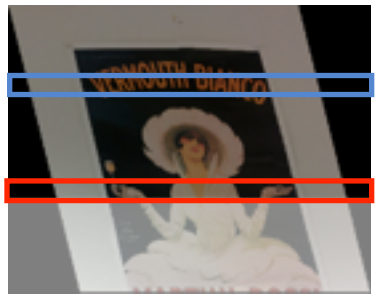
Current frame

- Satisfy small-motion assumption
  - Reference should not be far away



Adaptive Reference

- Leave sufficient motion
  - Row-to-row motion
  - Interpolation & pixel measurement noise



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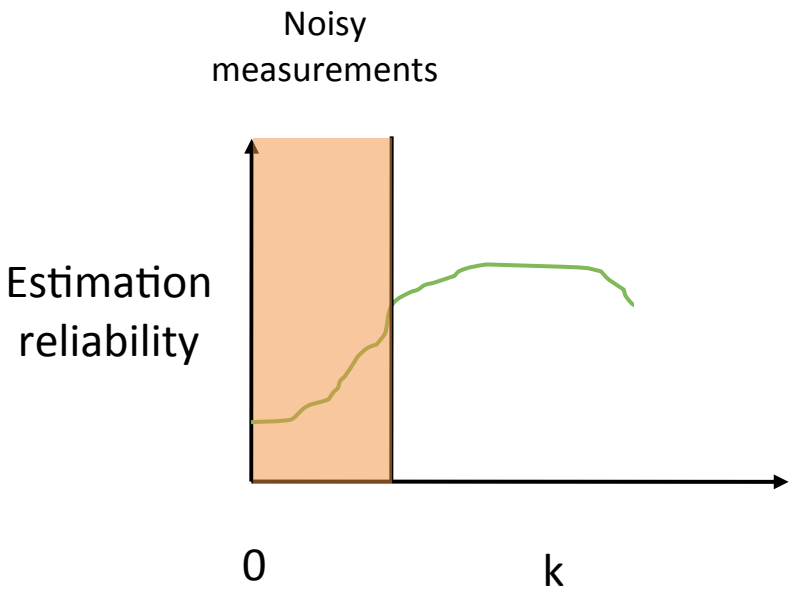

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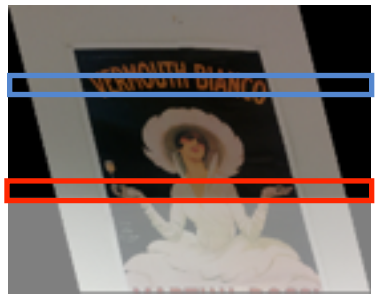

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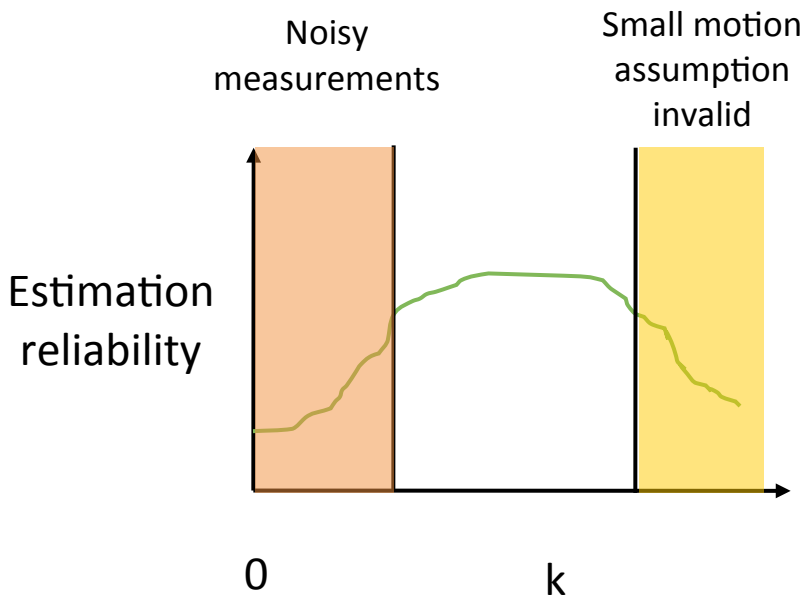
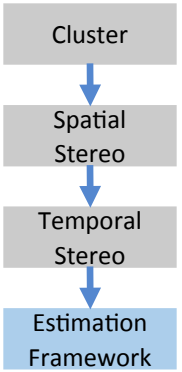


Confidence Scores

High confidence

Low Confidence



1. Quality of minimum: c_{PKR}
   - Is the valley unique?
   - Use Peak ratio (PKR)[1]

2. Temporal consistency : c_t
   - Consistent shifts in time
   - Penalize sudden changes

3. $c_{i,t} = c_{PKR} * c_t$

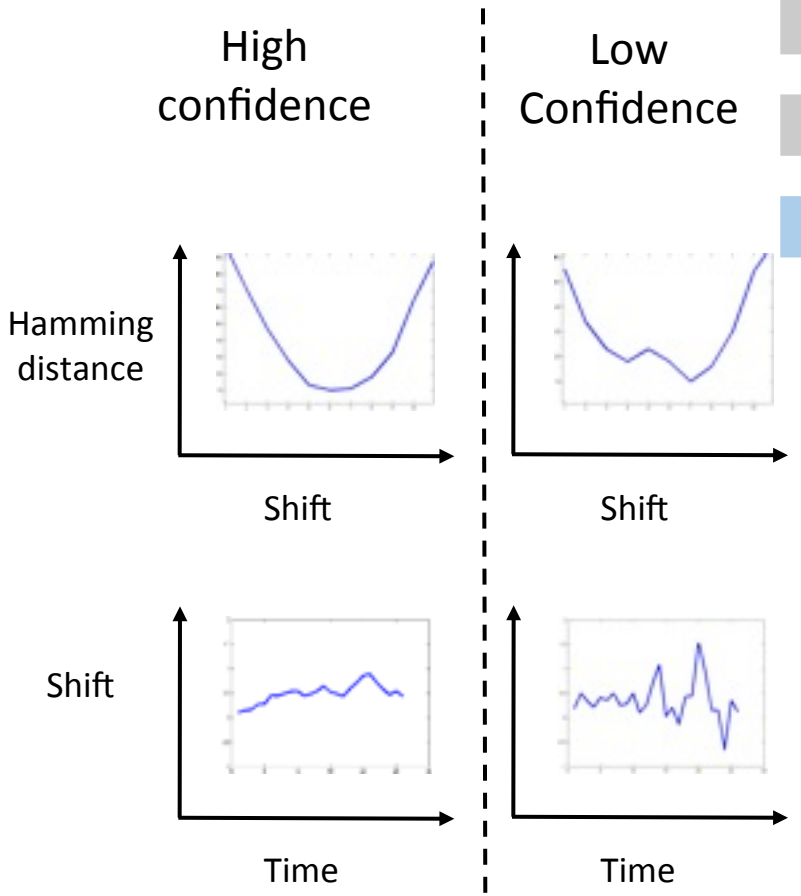
1. X. Hu and P. Mordohai. A quantitative evaluation of confidence measures for stereo vision, PAMI 2012

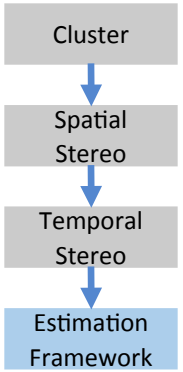
Confidence Scores

1. Quality of minimum: c_{PKR}
   - Is the valley unique?
   - Use Peak ratio (PKR)[1]

2. Temporal consistency : c_t
   - Consistent shifts in time
   - Penalize sudden changes

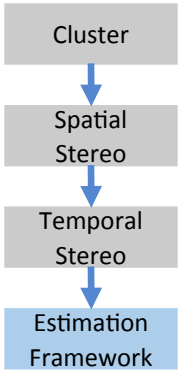
3. $c_{i,t} = c_{PKR} * c_t$





$$\mathbf{C}(t) = \begin{bmatrix} c_{1,t} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & c_{n,t} \end{bmatrix}$$

1. X. Hu and P. Mordohai. A quantitative evaluation of confidence measures for stereo vision, PAMI 2012



Motion Estimation

$$X_{cam_i}^{t+1} = [\Delta x \quad y + \Delta y \quad d + \Delta d \quad 1]^T$$

Stereo in time

$$X_{cam_i}^{t+1} = V_i \; M_{cluster} \; V_i^{-1} \; X_{cam_i}^{t}$$

$$X_{cam_i}^{t} = [0 \quad y \quad d \quad 1]^T$$

Stereo in space

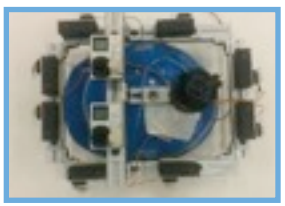
Weighted Linear System

- $X_{cam_i}^{t+1} = V_i \, M_{cluster} \, V_i^{-1} X_{cam_i}^{t}$
  - One equation from each camera
  - Use more cameras for robustness

$$C\,A \begin{bmatrix} \delta T_x \\ \delta T_y \\ \delta T_z \\ \theta_x \\ \theta_y \\ \theta_z \end{bmatrix} = C\;B$$

Approach Summary

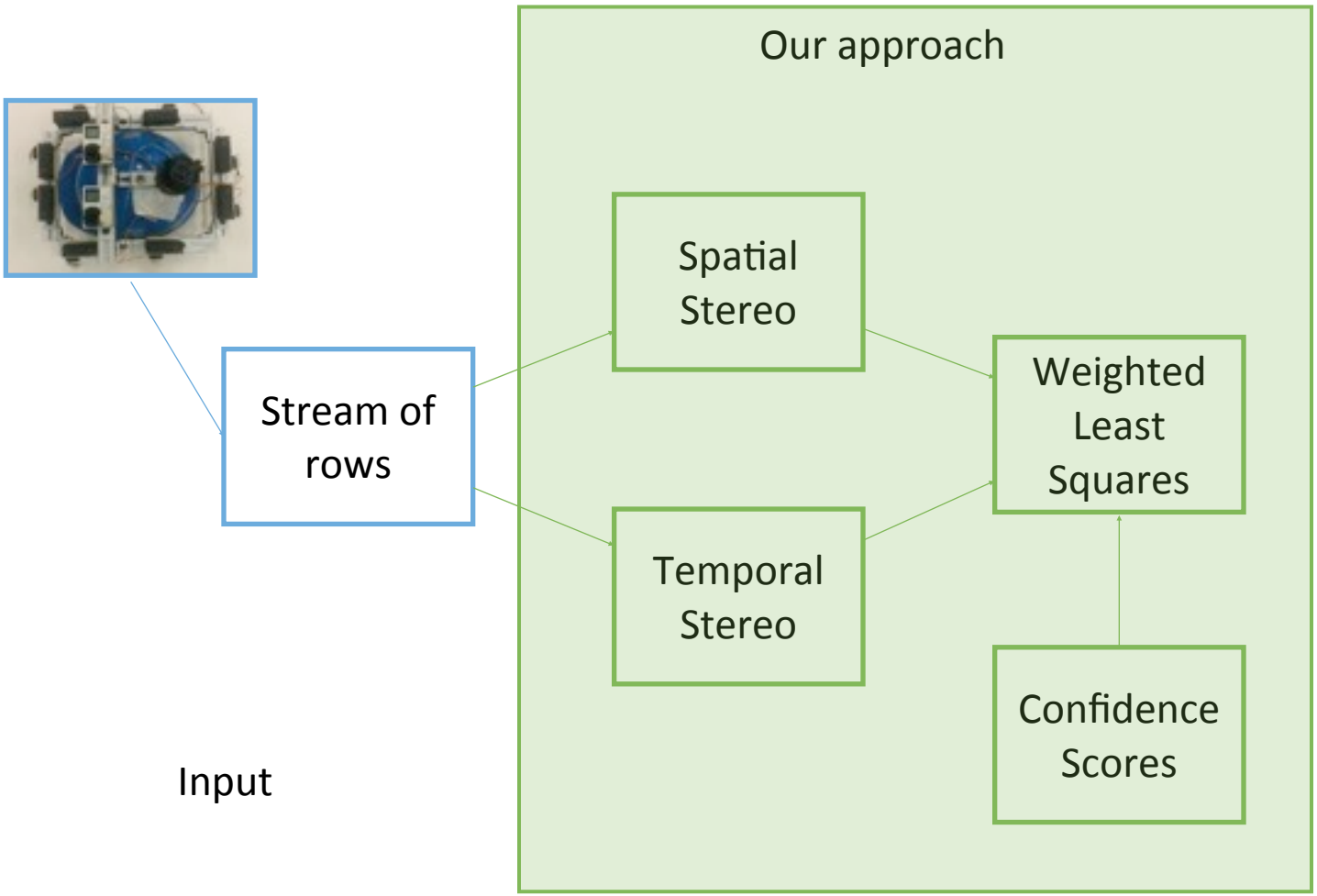
Approach Summary



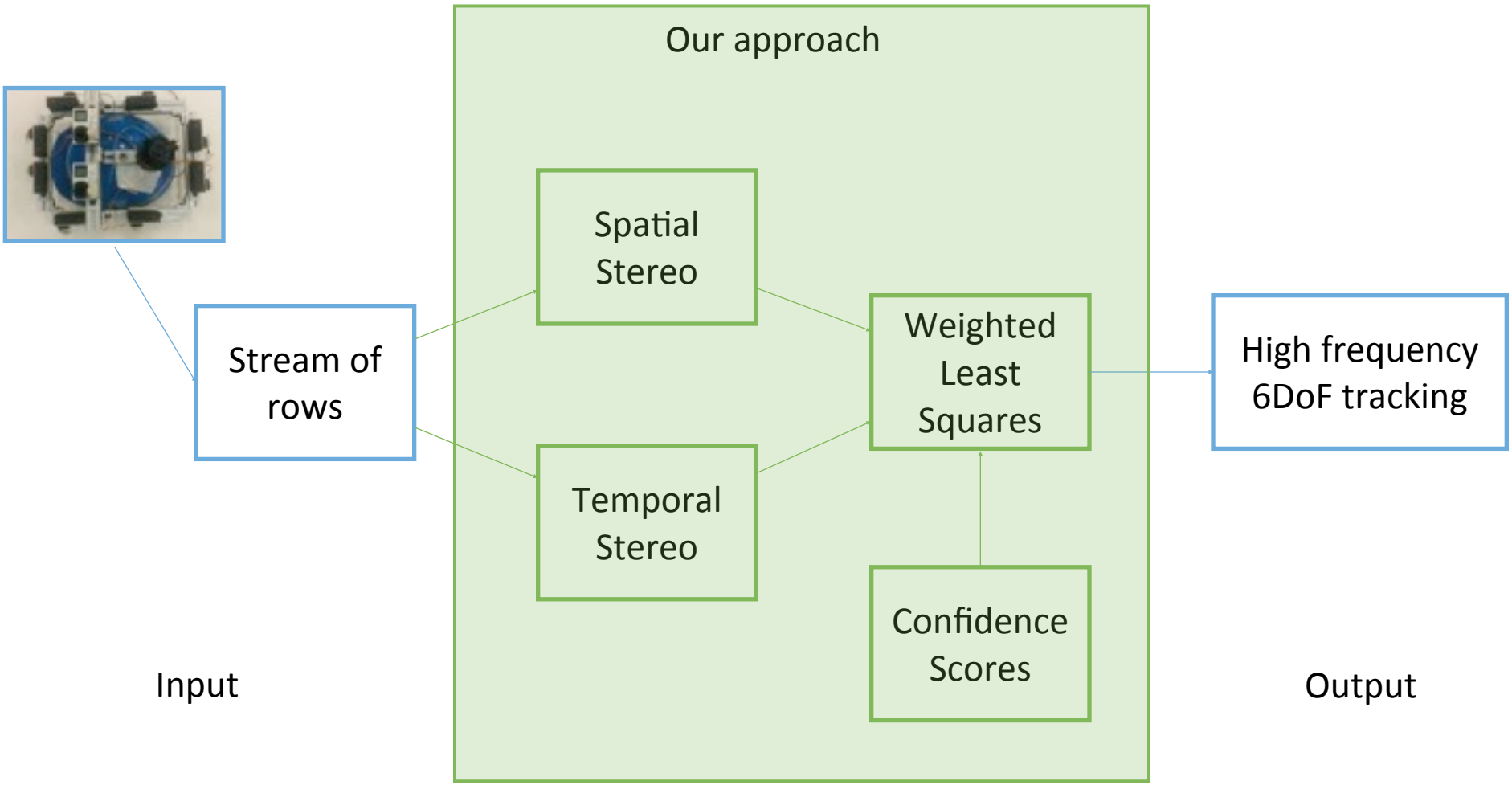
Stream of rows

Input

Approach Summary



Approach Summary



Simulator

- Developed in OpenGL+Qt and Unity3D
  - Support for Hi-Ball tracker data
  - Motion Blur

Simulator

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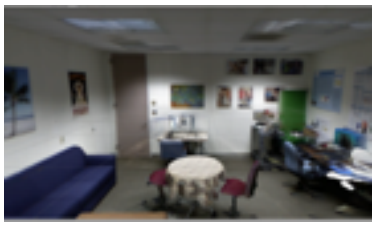
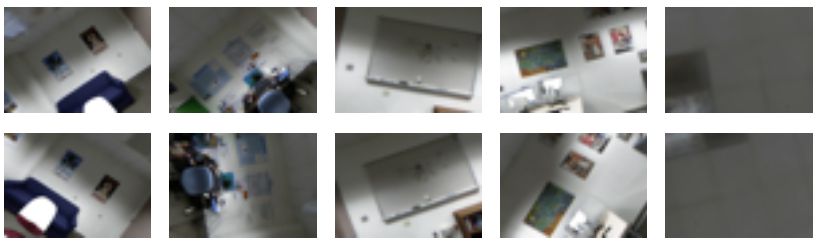
  - Support for Hi-Ball tracker data
  - Motion Blur

Experiments

- Camera specs:
  - 640 × 480 pixels
  - FoV = 60°
  - Stereo pairs = 10
  - 120 fps

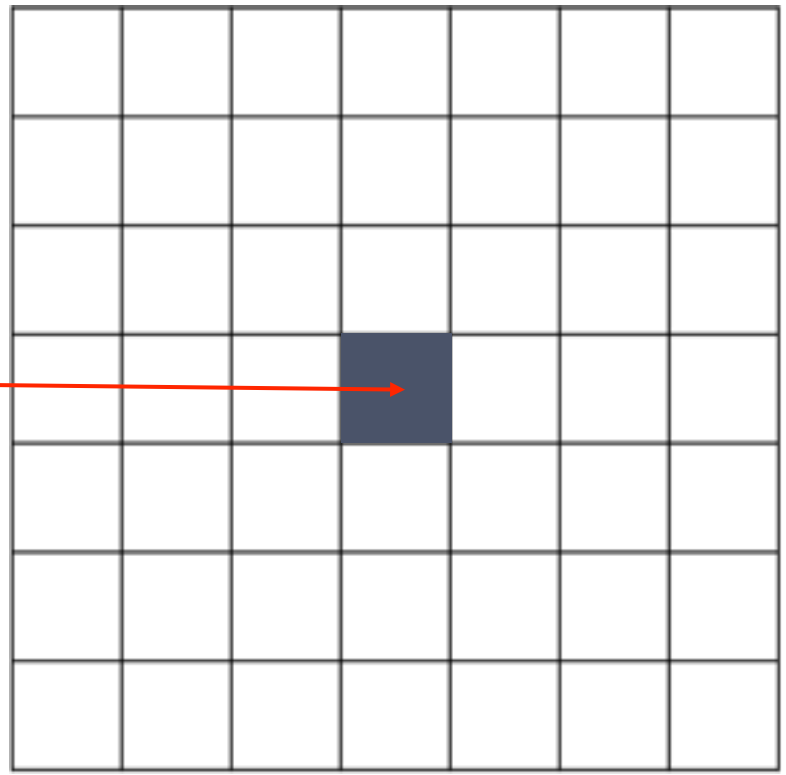




- Quantify errors in terms of display pixel errors
  - Display specs : HTC Vive
    - 120° FoV
    - 1080 × 1200 pixels per eye
    - Point 1m in front

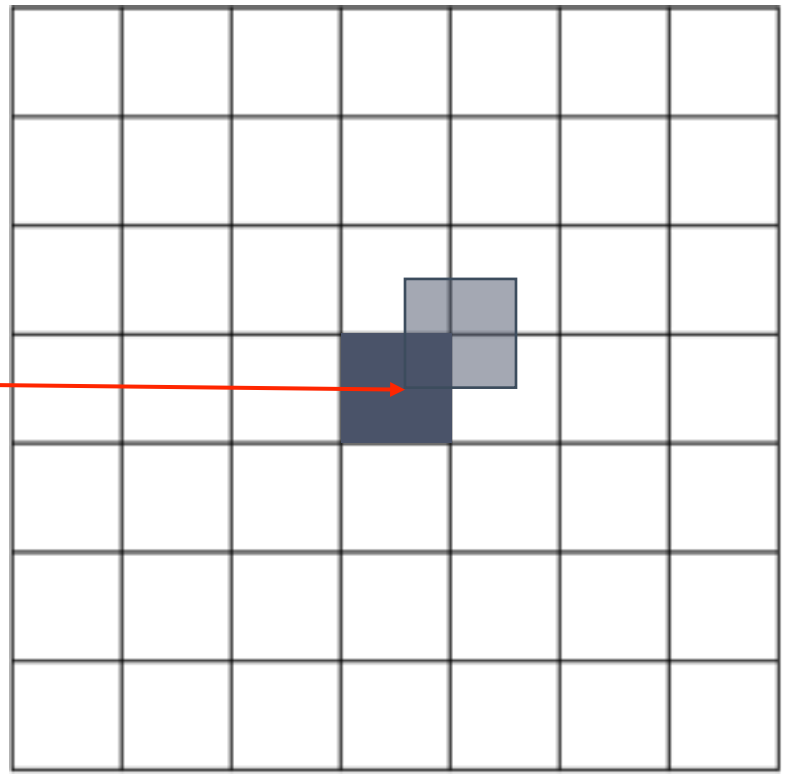
1.Mingsong Duo et al, Exploring high-level plane primitives for indoor 3d reconstruction with a hand-held rgb-d camera, ACCV 2012

Display Pixel Error



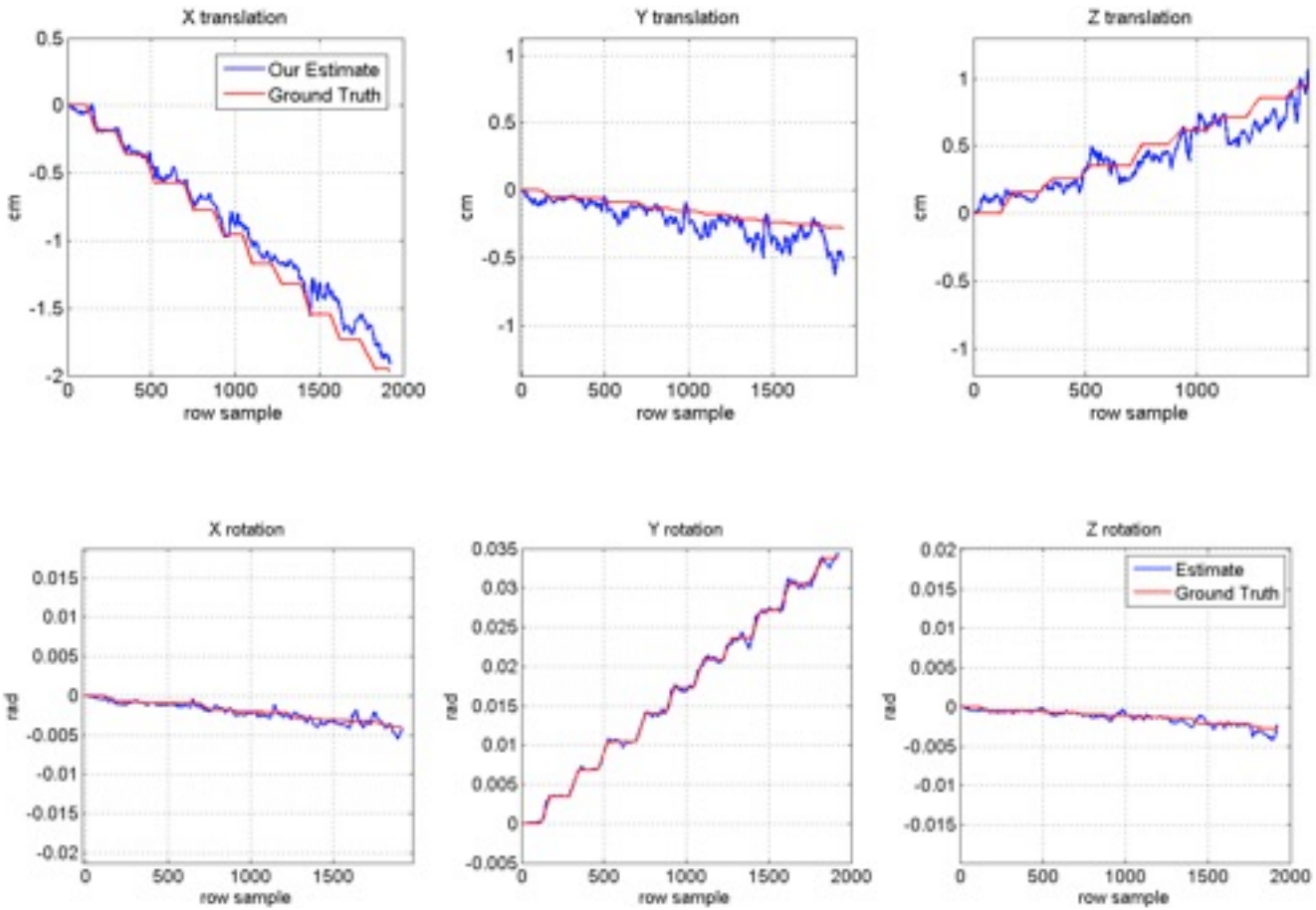
Pixels shown in HMD

Display Pixel Error



Pixels shown in HMD

Experiments : Real motion[1] in Synthetic Room

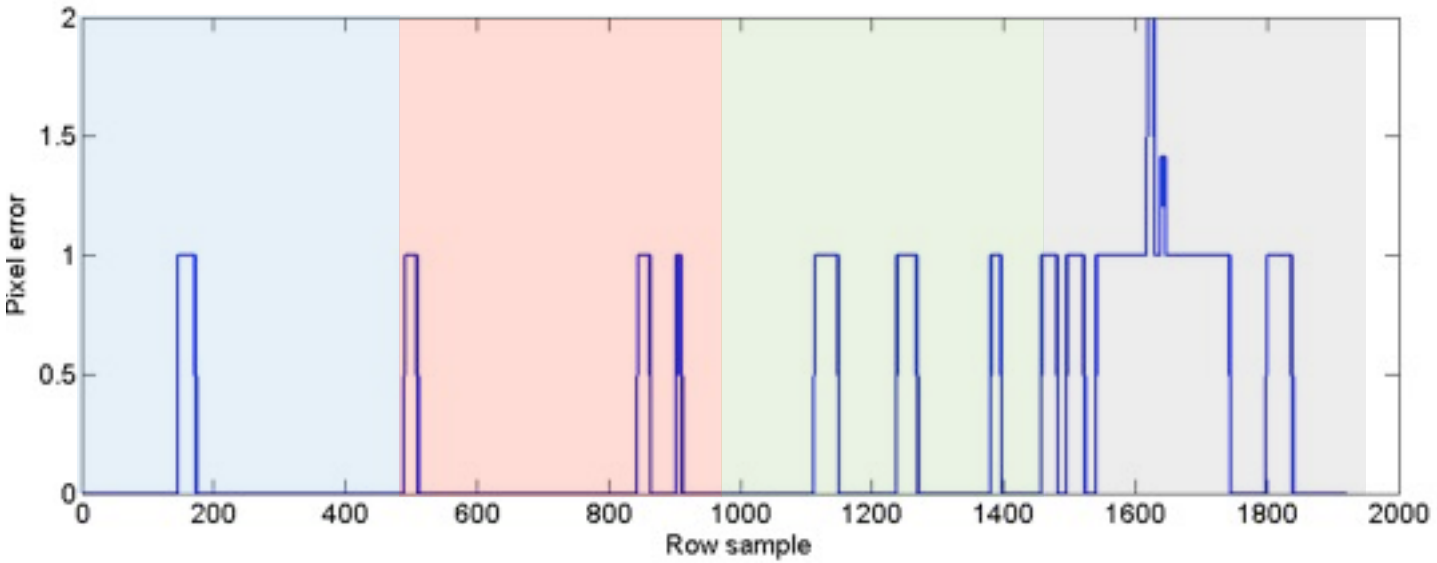


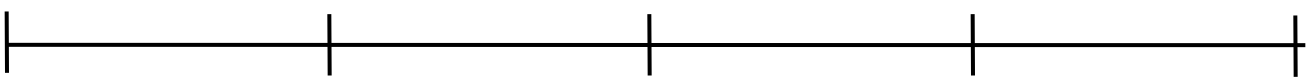
f= 56.7kHz

1. Hi-Ball tracker data

Experiments : Real motion[1] in Synthetic Room

f= 56.7kHz

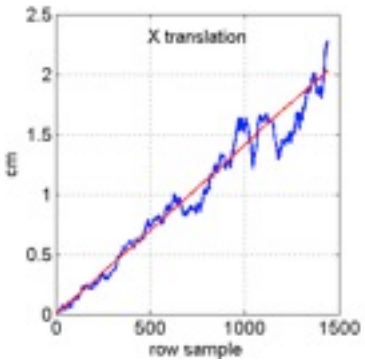


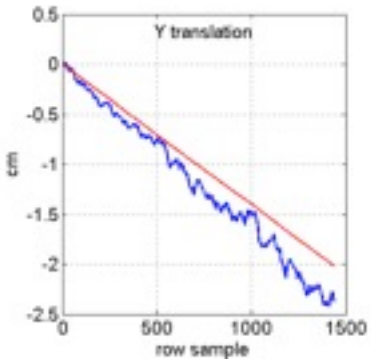


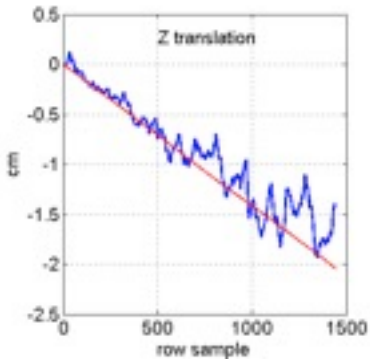
Frame i i+1 i+2 i+3 i+4 …

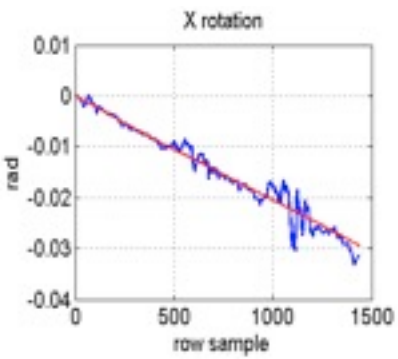
1. Hi-Ball tracker data

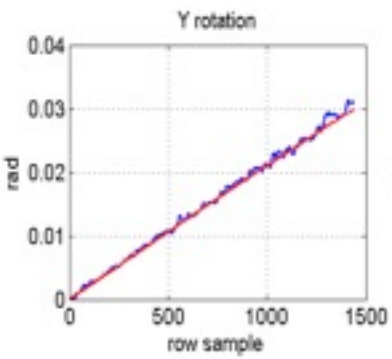
Experiments: Synthetic Large Motion

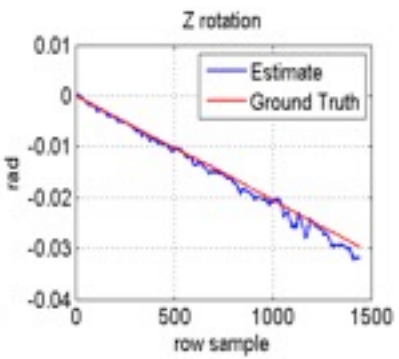








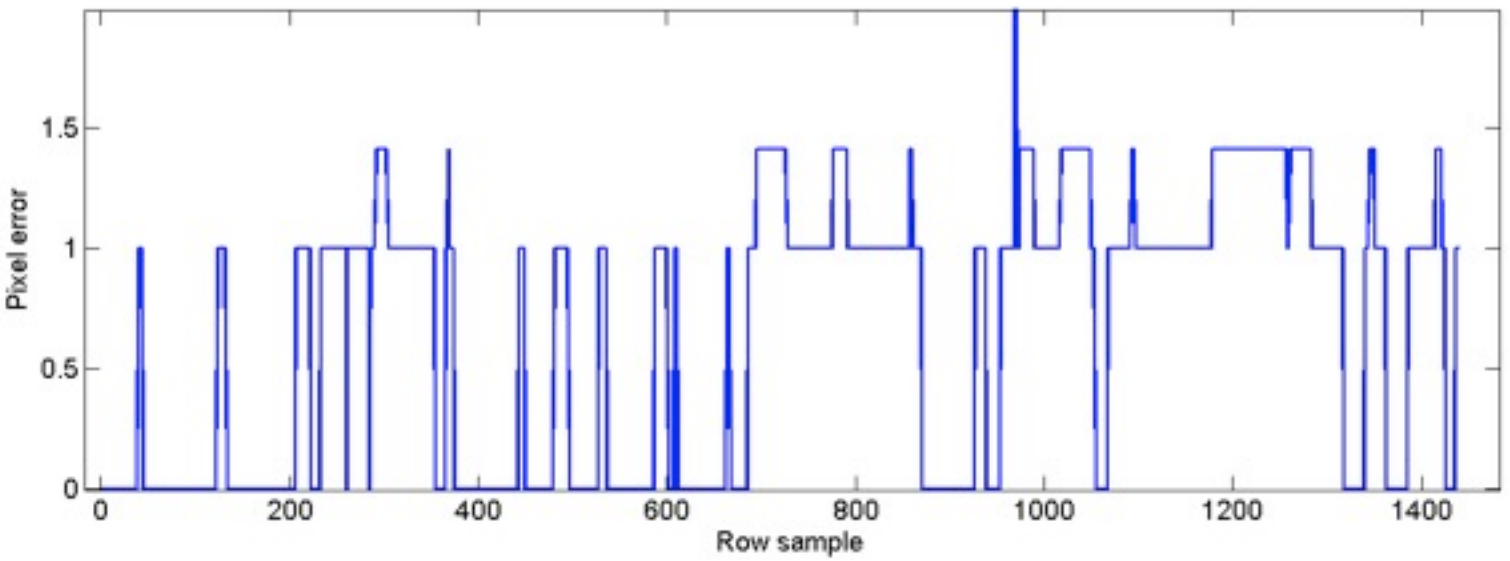




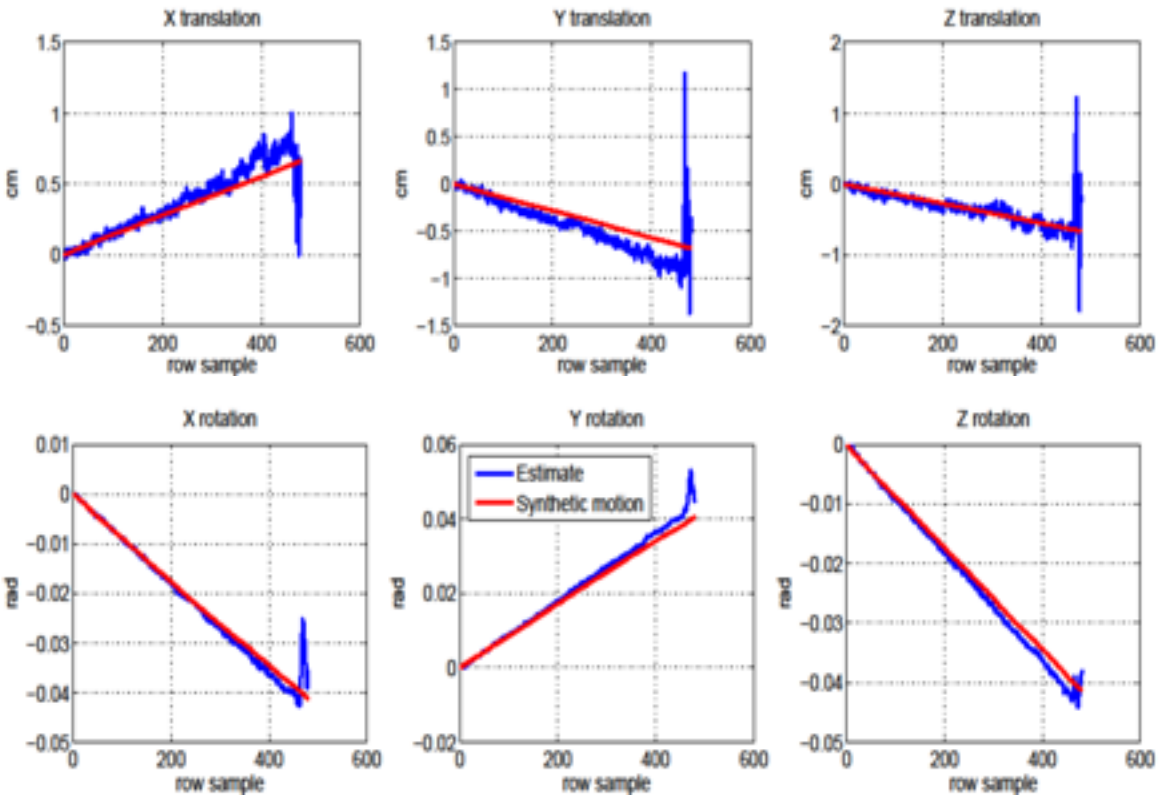
f= 56.7kHz
v= 1.4 m/s
ω= 120 deg/s

Experiments: Synthetic Large Motion

f= 56.7kHz



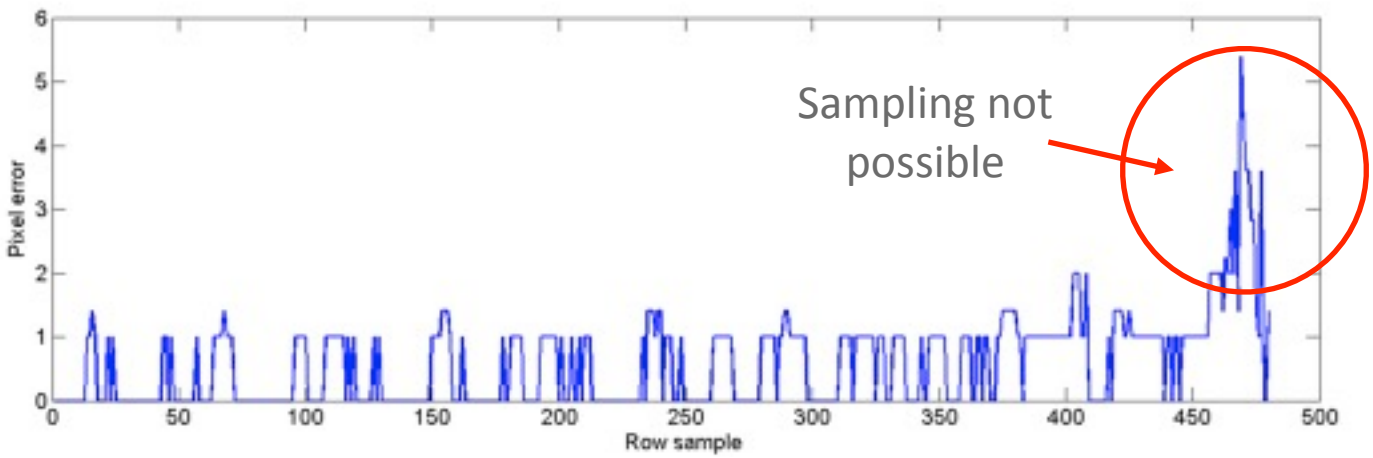
Experiments : Synthetic Extreme Motion

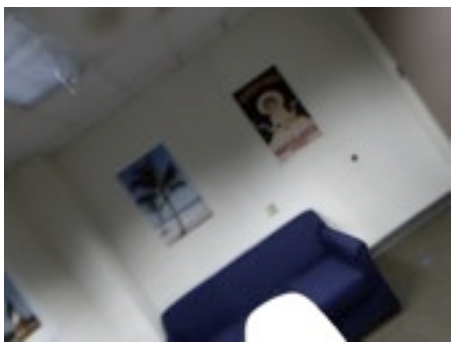


f= 56.7kHz
v= 1.4 m/s
ω= 500 deg/s

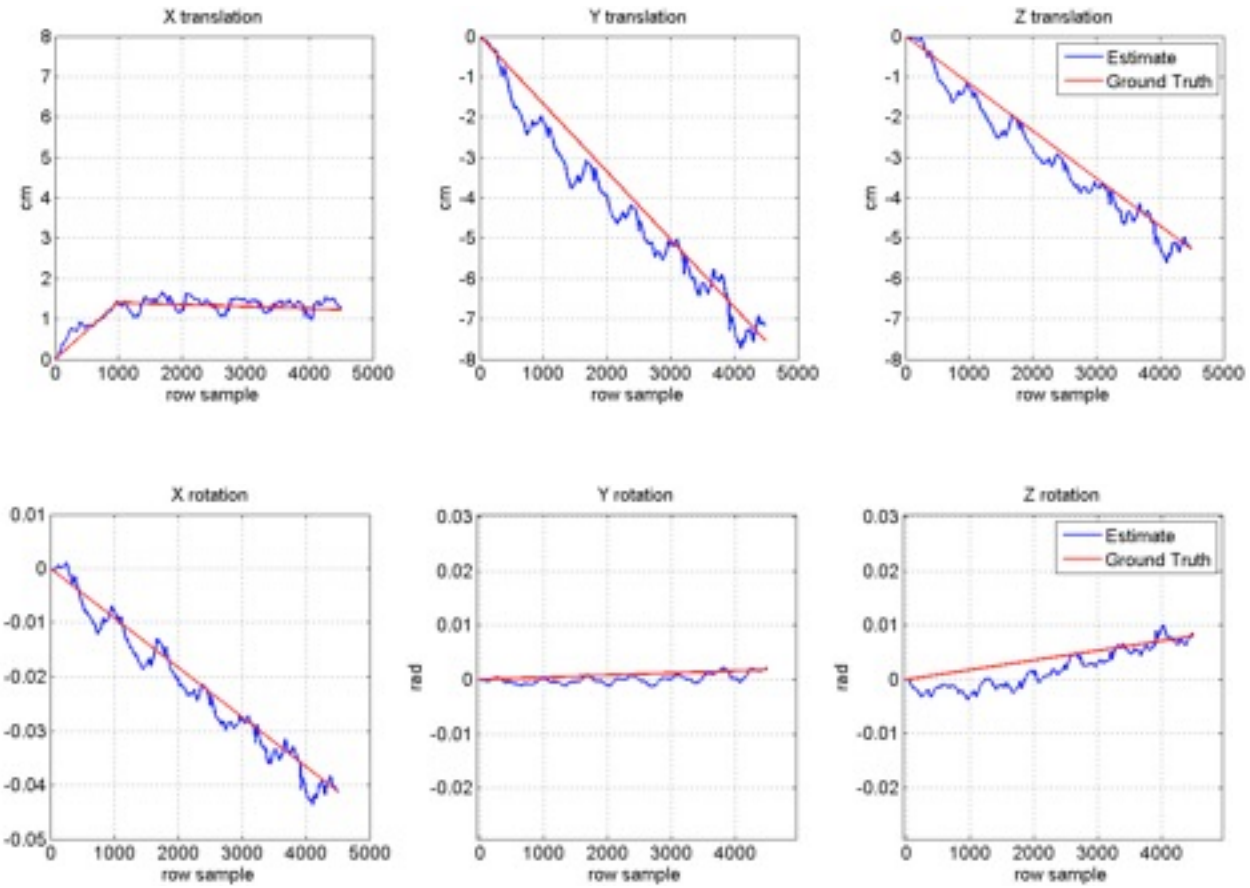
Experiments : Synthetic Extreme Motion

f= 56.7kHz
v= 1.4 m/s
ω= 500 deg/s





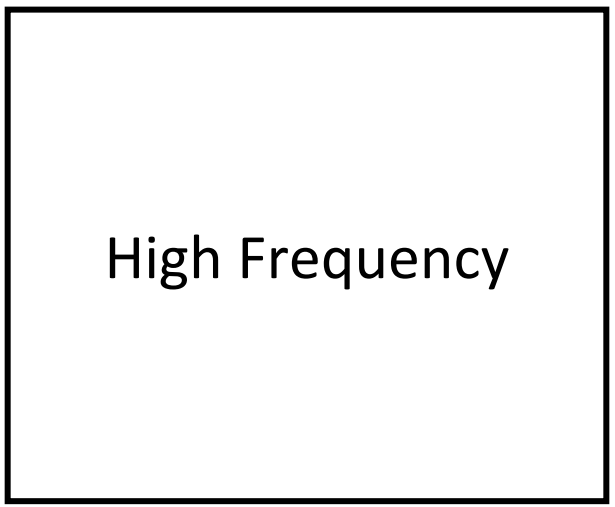
Results for Real Imagery



f= 80.4kHz

Conclusion

- High frequency visual tracker
  - Up to 80 kHz
  - Off-the-shelf cameras



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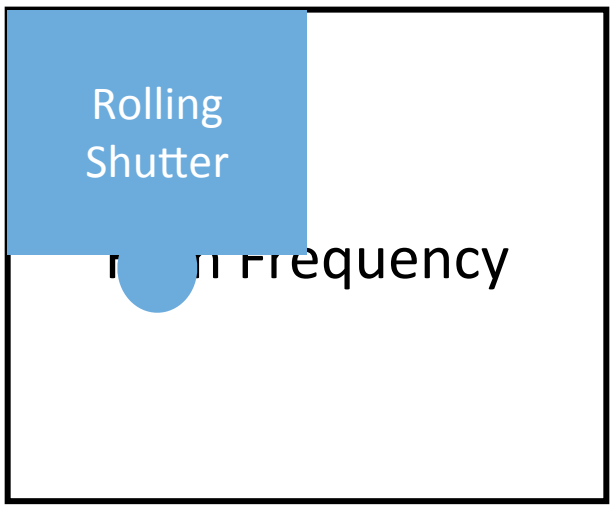

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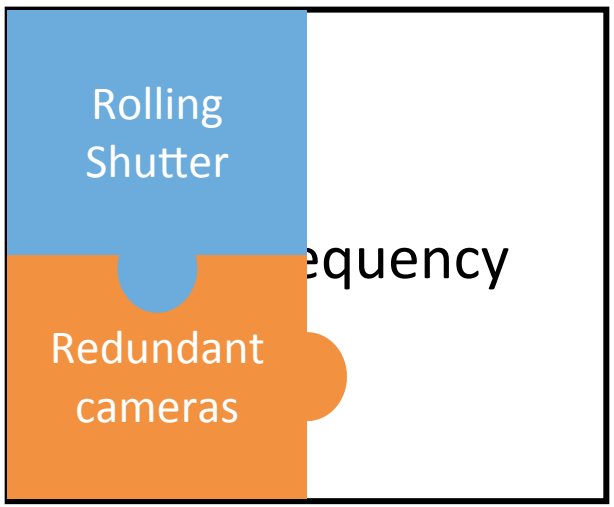

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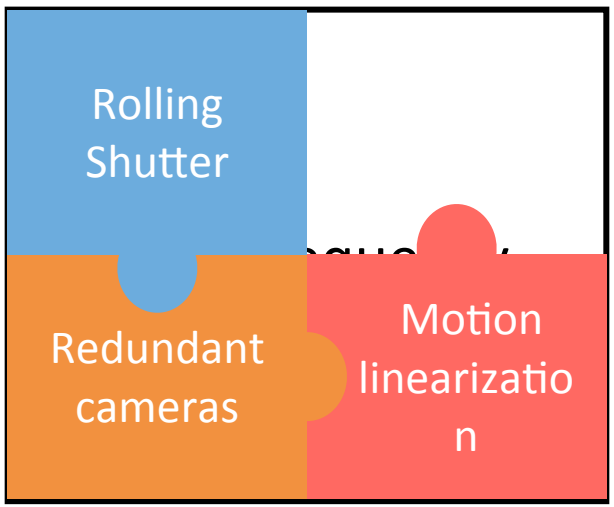

Conclusion

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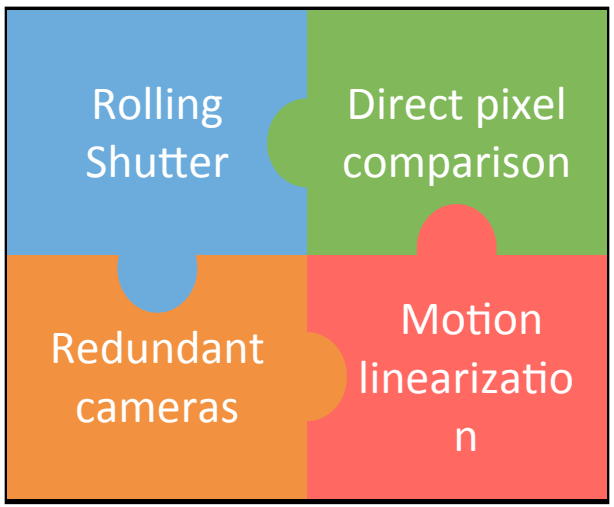
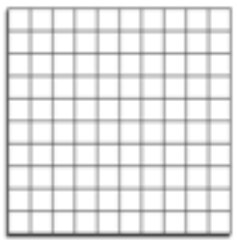
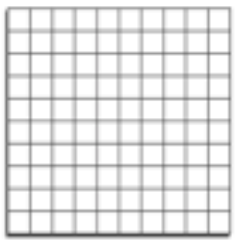


Thank you!

Contact: akash@cs.unc.edu

Camera tracking : The spectrum

Global Shutter	Rolling Shutter as 1-D sensor	1-D sensor
		 1-D camera[1]
+ No distortion + Drift correction – High cost – Low fps ≈ 200Hz – Higher noise	– Distortion + Drift correction + Cheap + High fps, 56 kHz + Low noise	+ No distortion – No drift correction + High Cost + High fps, till 87 kHz – Higher noise