



# TEMPORALLY COHERENT 4D RECONSTRUCTION OF COMPLEX DYNAMIC SCENES

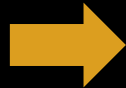
*Armin Mustafa, Hansung Kim, Jean-Yves Guillemaut, Adrian Hilton*

# PROBLEM STATEMENT

- Reconstruct complex dynamic scenes.
- Multi-view, wide-baseline and moving handheld cameras.
- Unknown background, structure and segmentation.



# PROBLEMS IN EXISTING METHODS



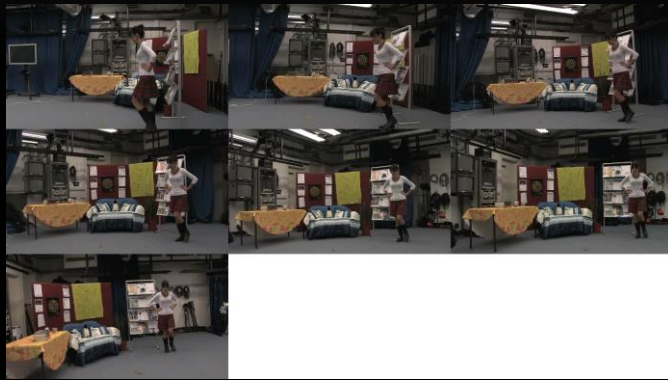
Segmentation



Depth map

- Requires accurate segmentation of foreground
- Known background and structure

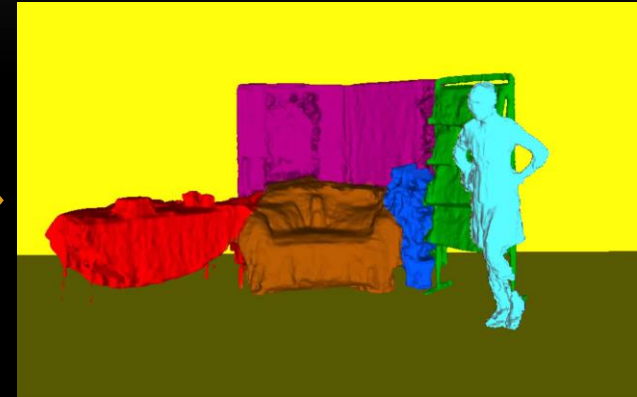
# OVERVIEW



- No prior
- Moving cameras



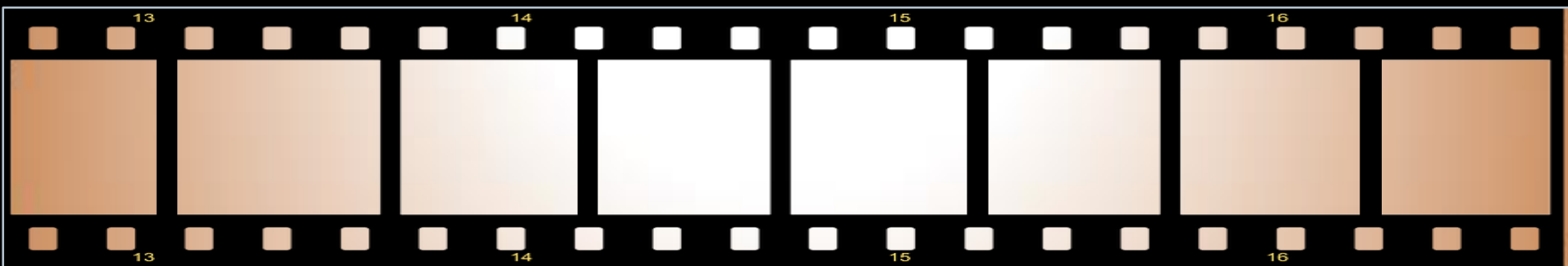
- Salient object identification
- Temporal coherence



4D scene reconstruction and segmentation

# FRAMEWORK

Multi-view  
data

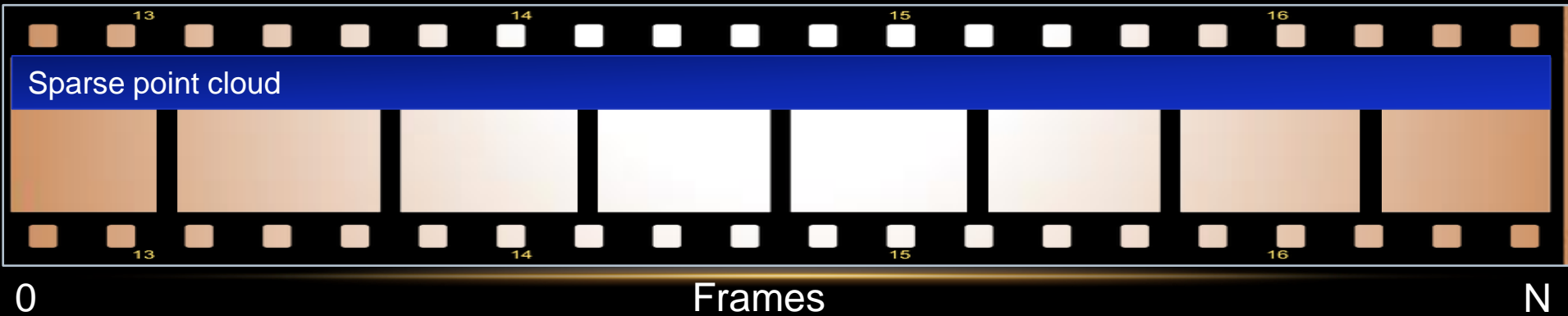
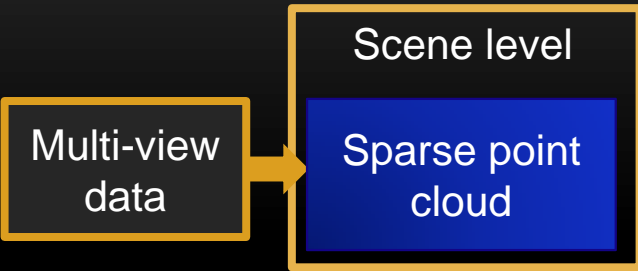


0

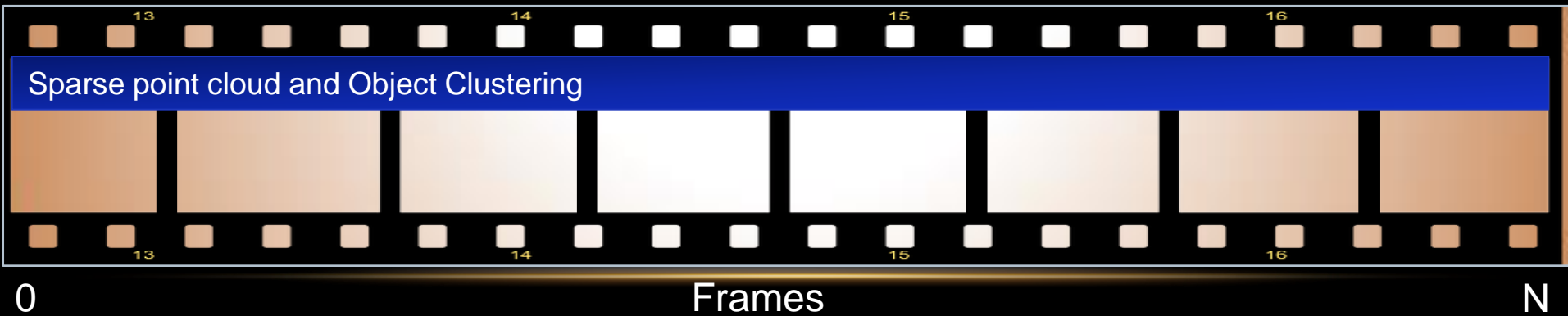
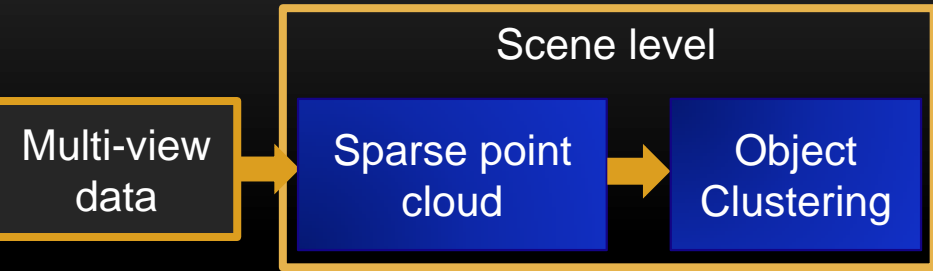
Frames

N

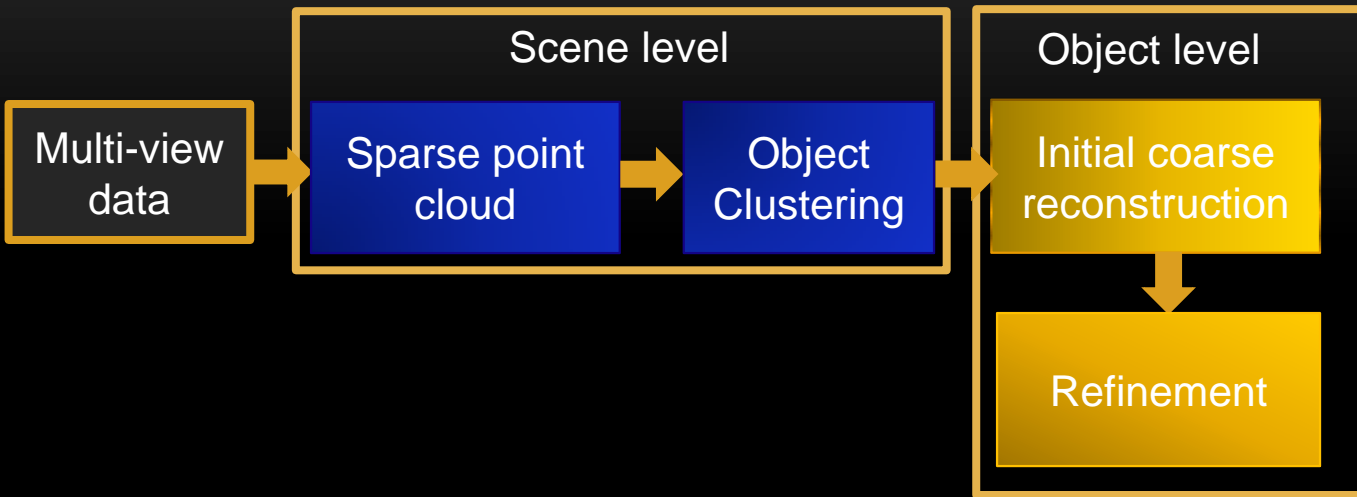
# FRAMEWORK



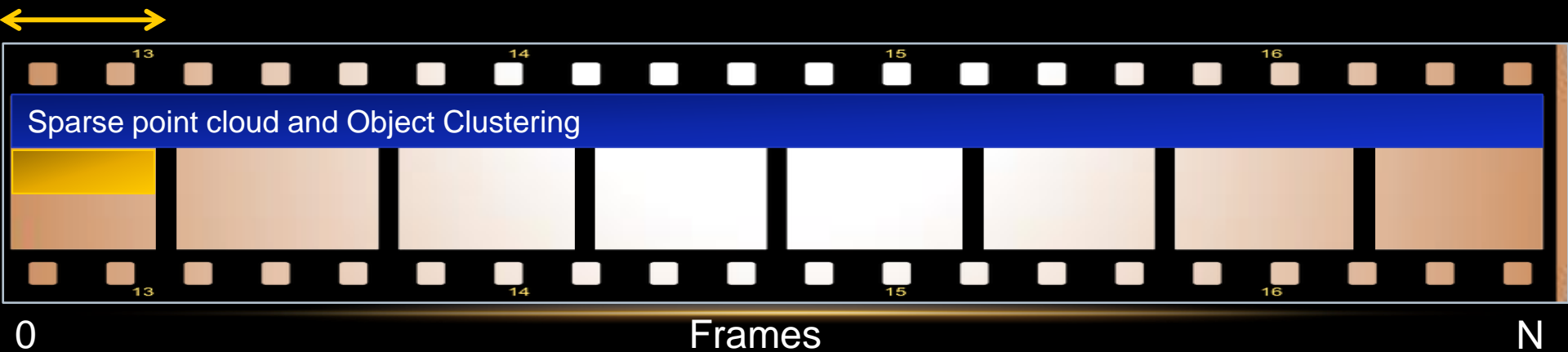
# FRAMEWORK



# FRAMEWORK

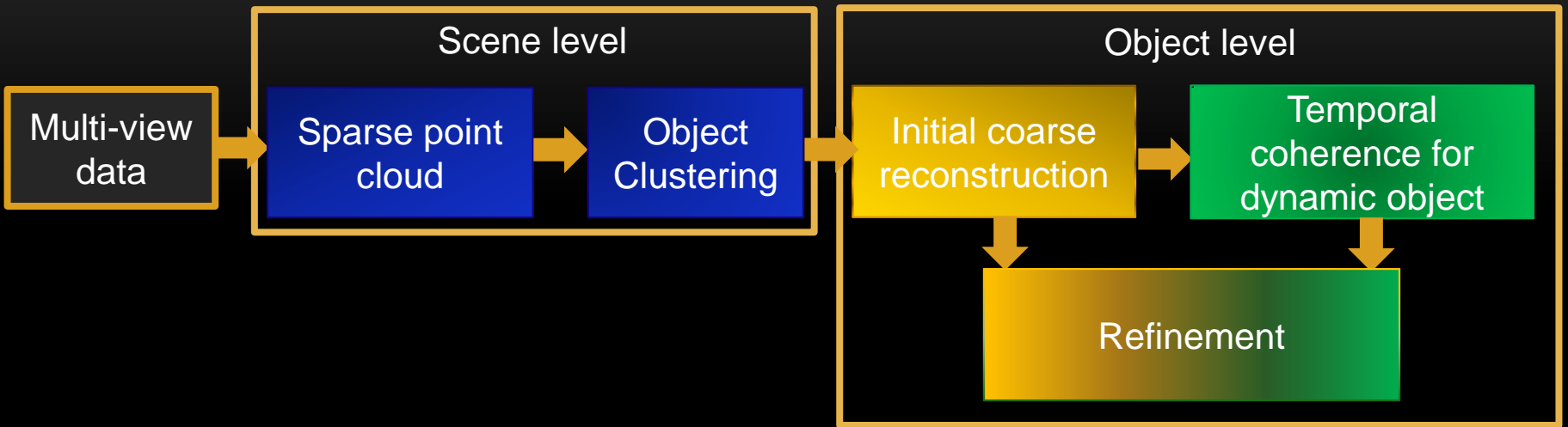


Initial coarse reconstruction and refinement

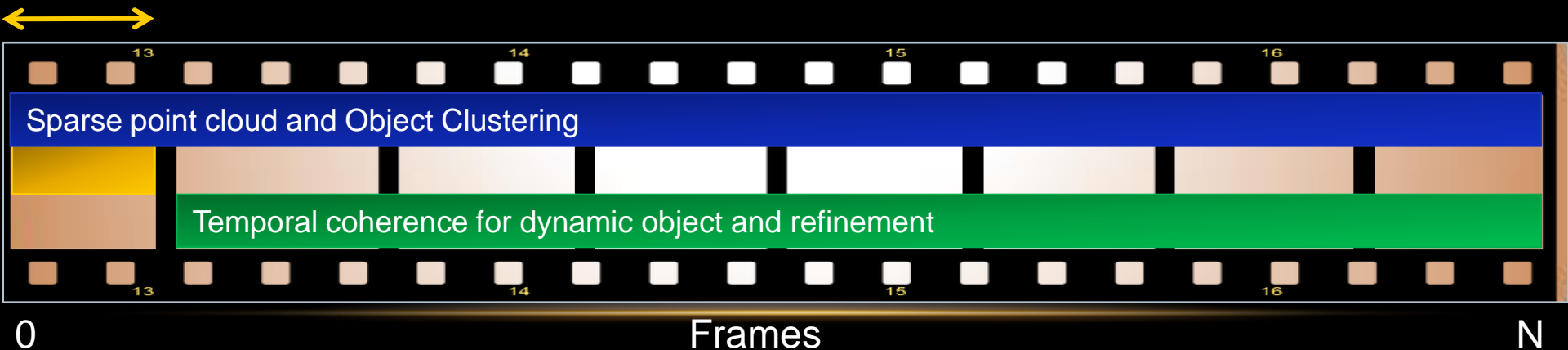




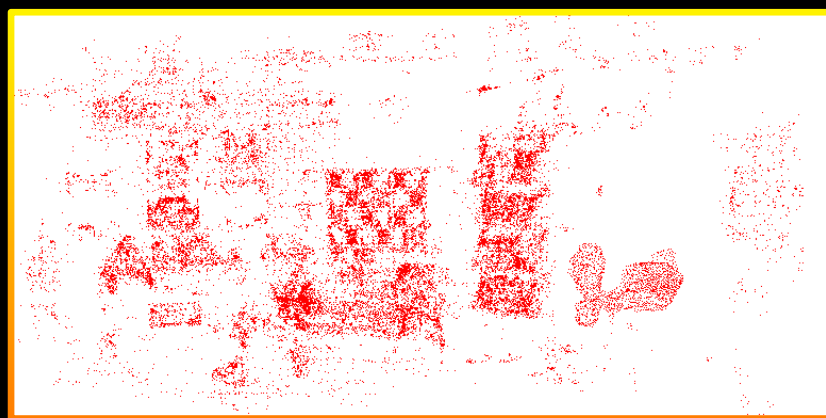
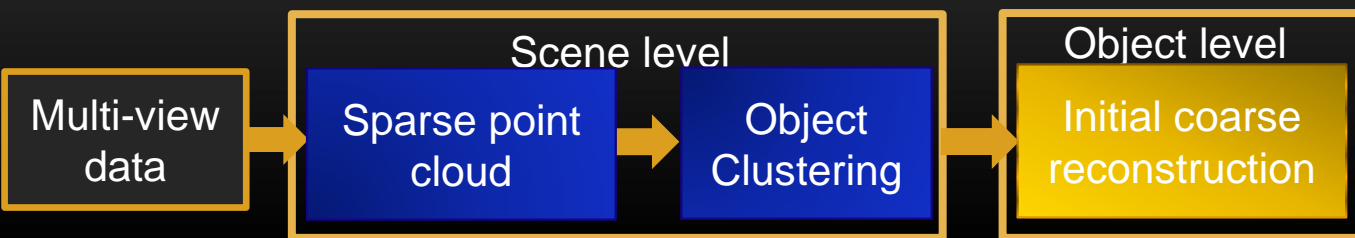
# FRAMEWORK



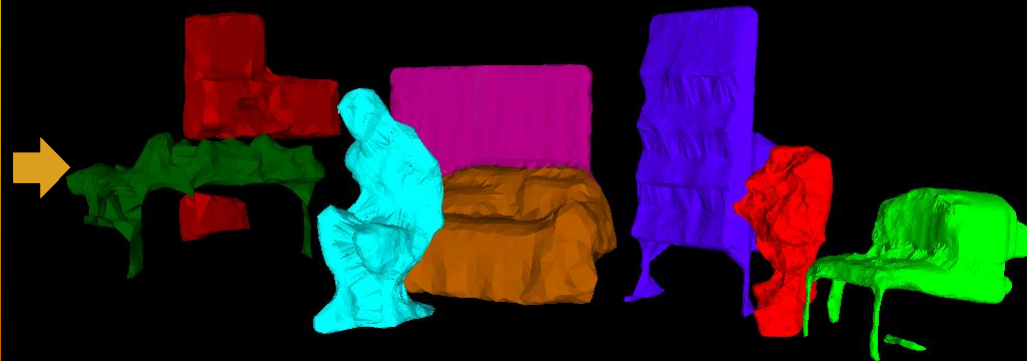
Initial coarse reconstruction and refinement



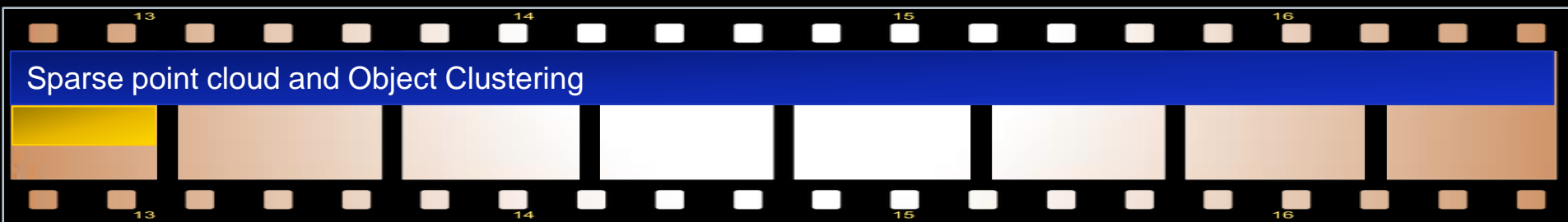
# FRAMEWORK



Sparse point cloud



Object clustering



0

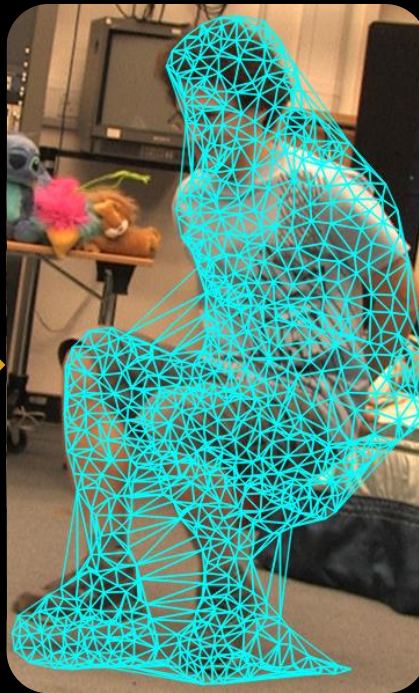
Frames

N

# INITIAL COARSE RECONSTRUCTION



Sparse Cluster

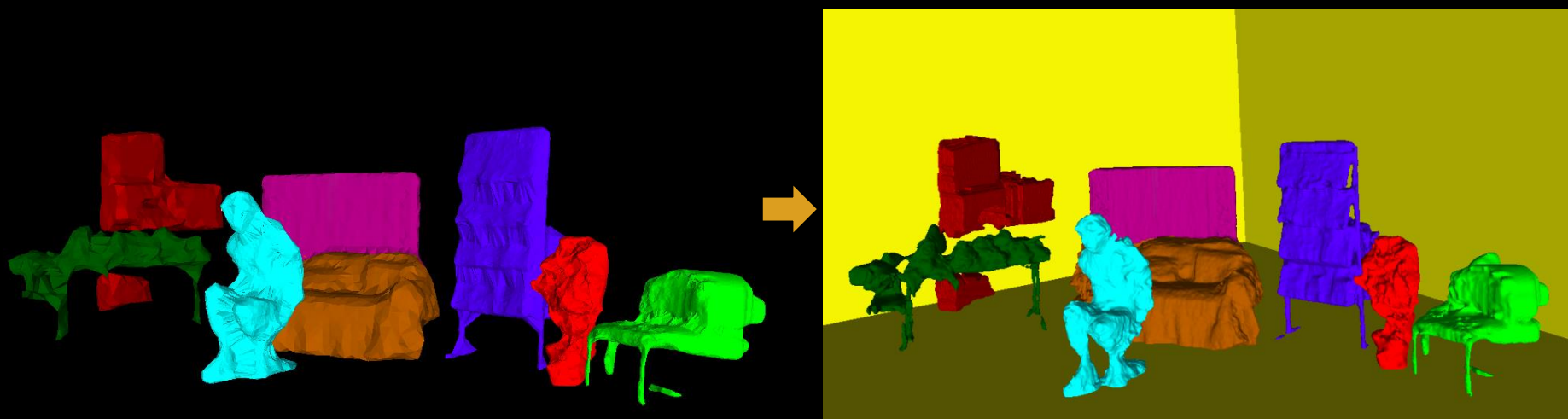
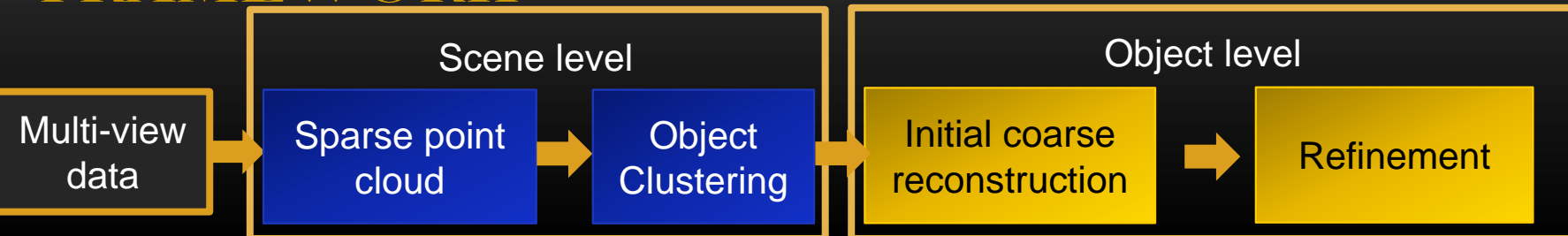


Triangulation



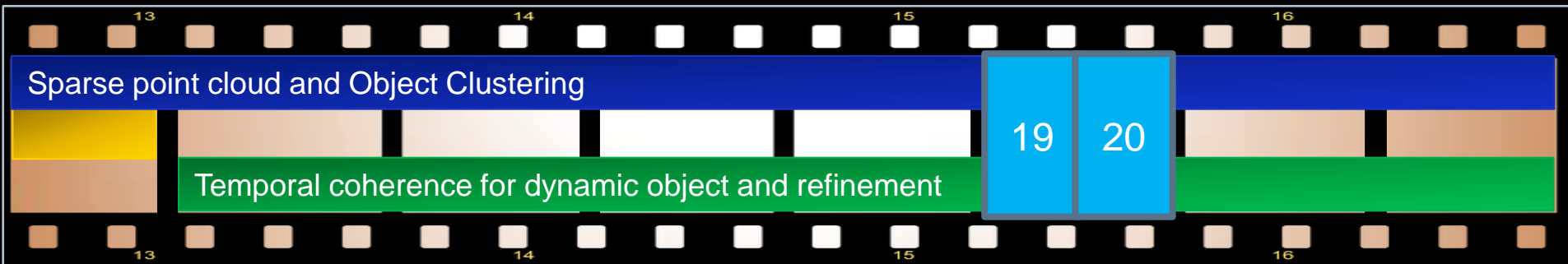
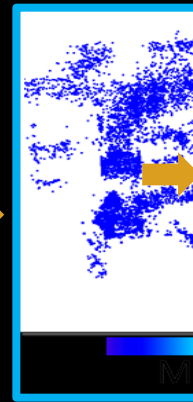
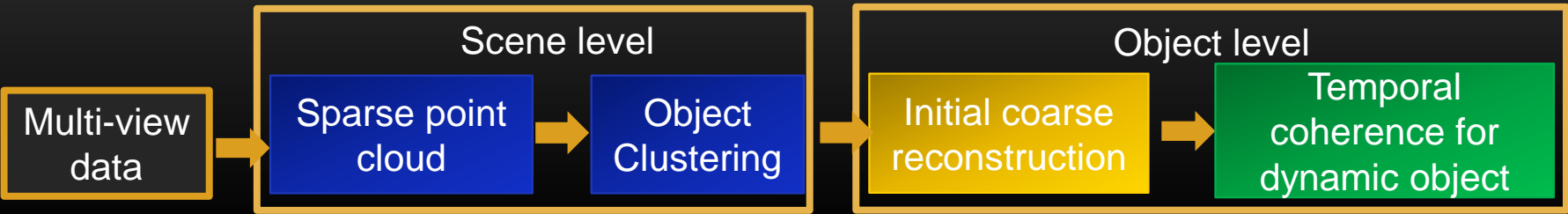
Initial coarse reconstruction

# FRAMEWORK



Frames

# FRAMEWORK

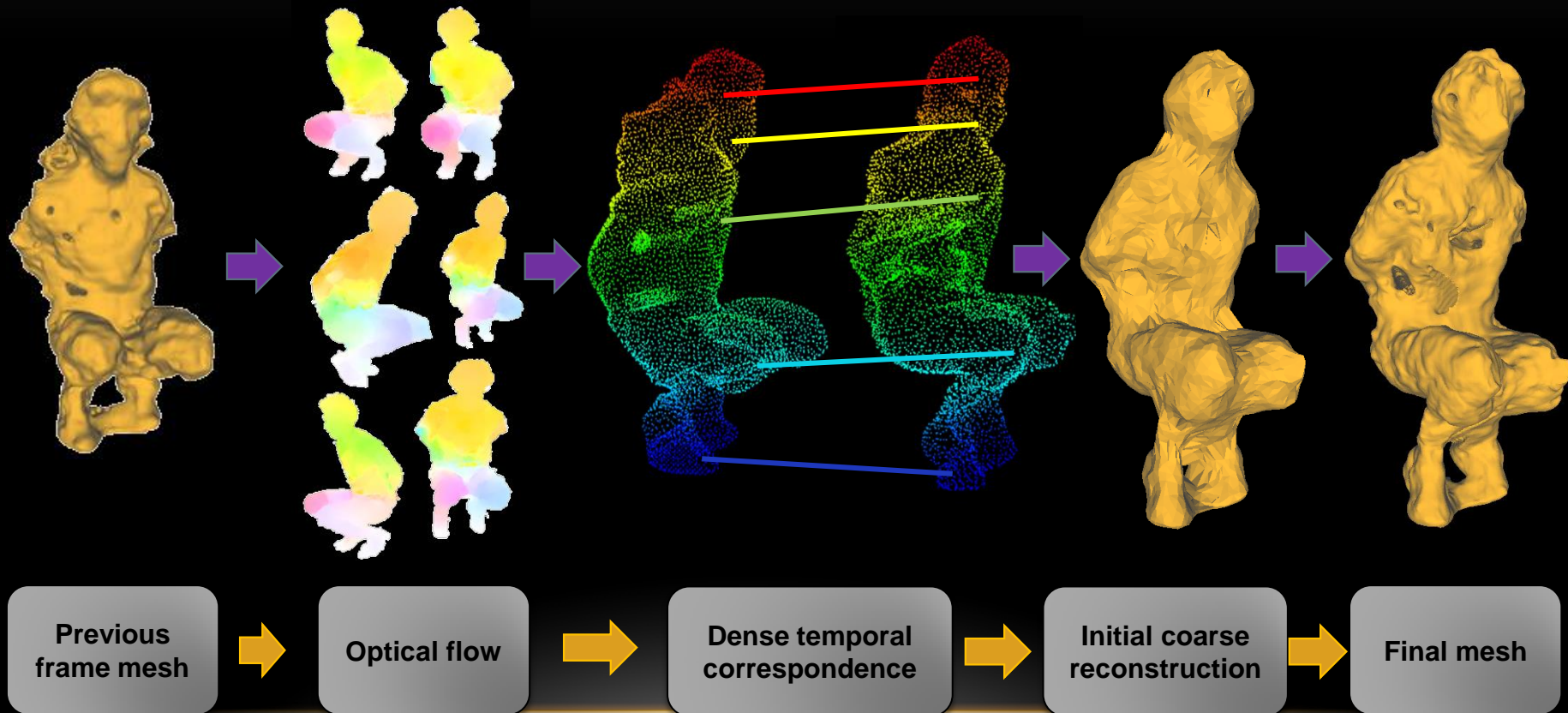


0      Frames      N

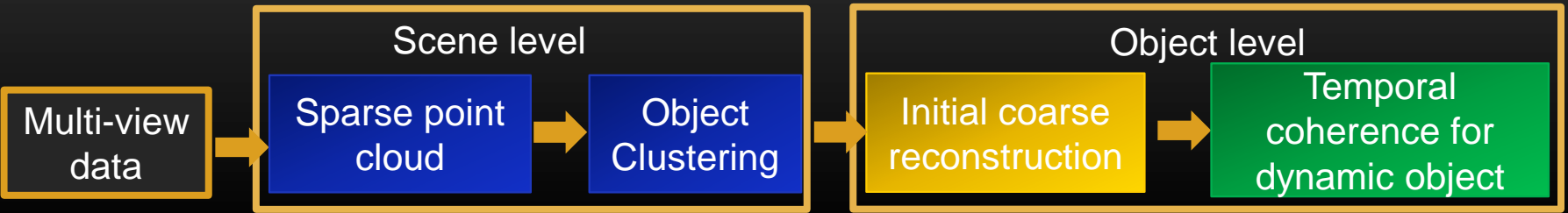


# TEMPORAL COHERENCE :

Sparse to dense reconstruction and refinement:



# FRAMEWORK



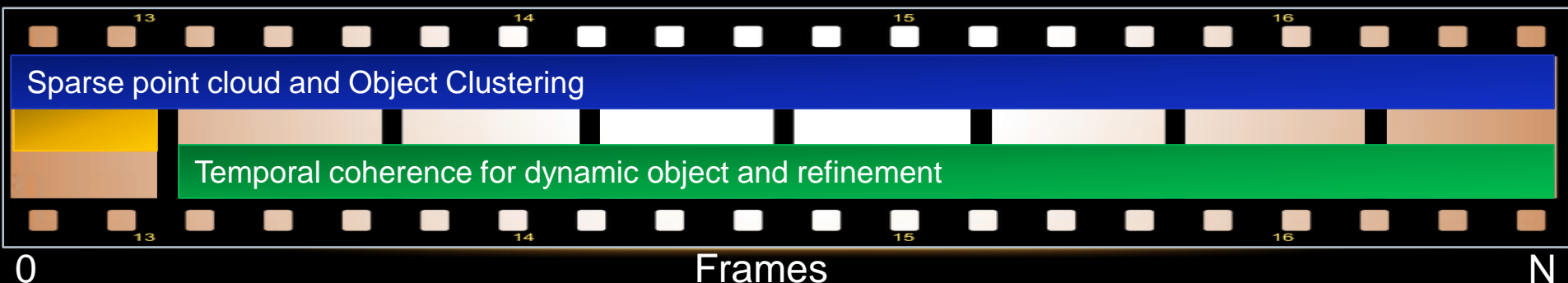
Multi-view video



Frame 1

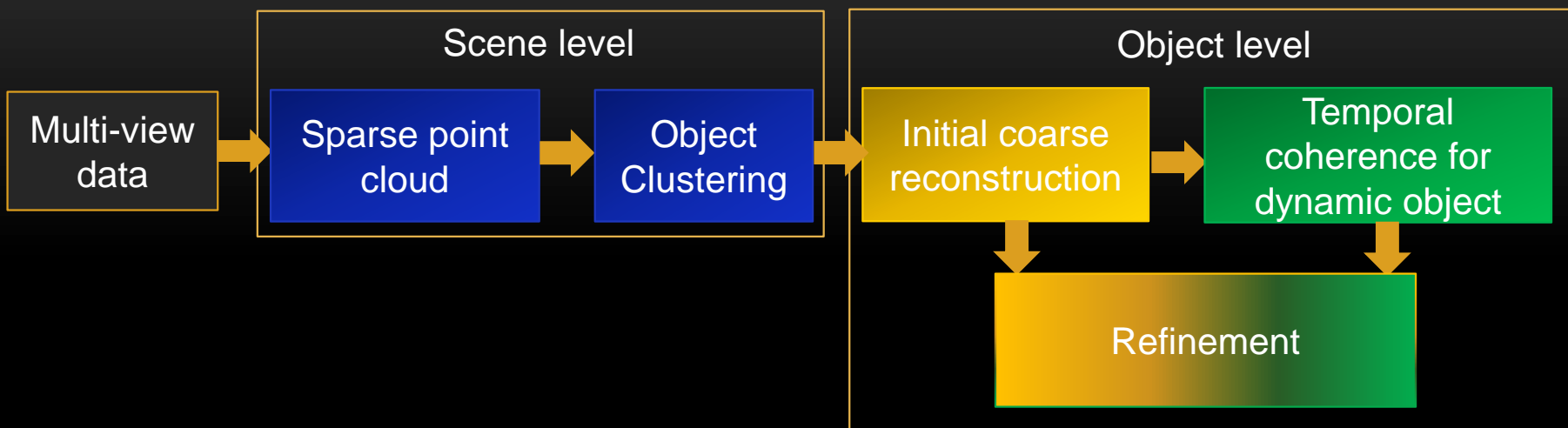


Temporal coherence

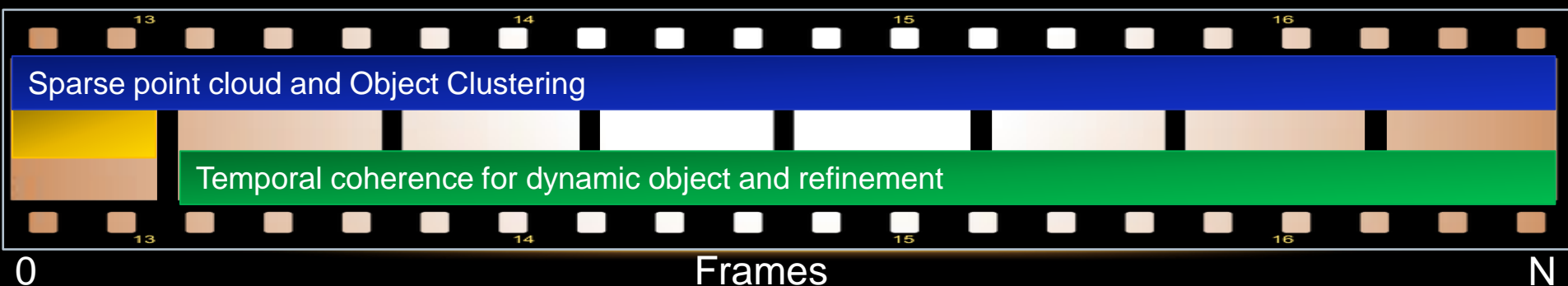


Frames

# FRAMEWORK



- Joint segmentation and reconstruction
- Optimized based on graph cuts



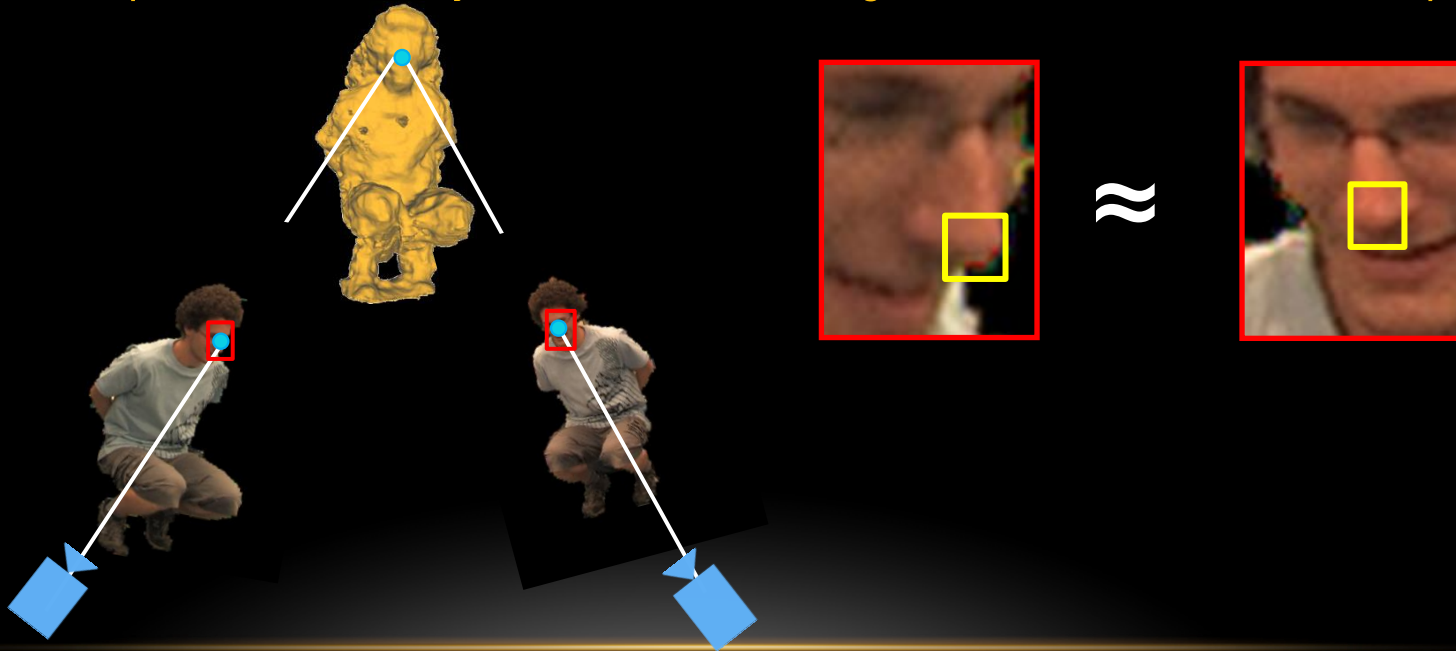


# REFINEMENT: SHAPE

$$E(l,d) = \alpha E_{\text{data}}(d) + \beta E_{\text{smooth}}(l) + \gamma E_{\text{color}}(l) + \eta E_{\text{contrast}}(l,d)$$

where  $l$  is the label and  $d$  is the depth

- Error tolerant photo-consistency is combined with edge information to refine the depth.*

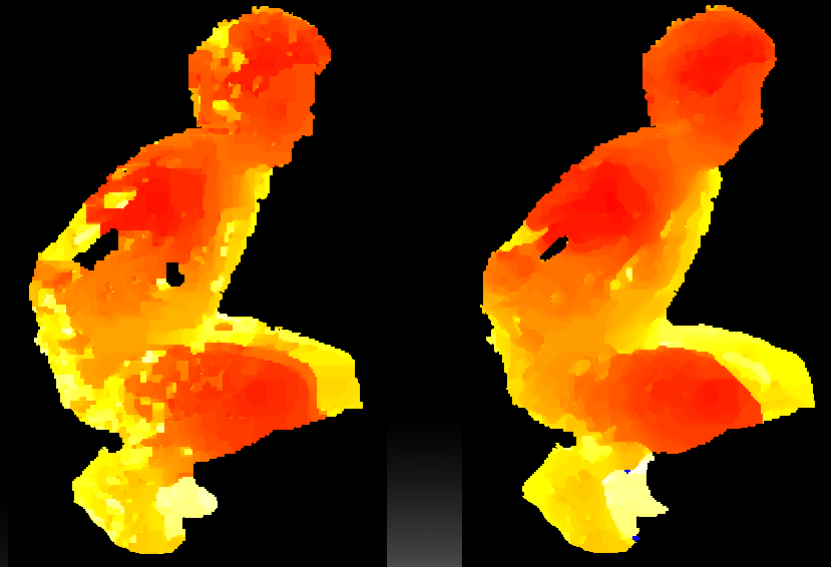


# REFINEMENT: SHAPE

$$E(l,d) = \alpha E_{\text{data}}(d) + \beta E_{\text{smooth}}(l) + \gamma E_{\text{color}}(l) + \eta E_{\text{contrast}}(l,d)$$

where  $l$  is the label and  $d$  is the depth

- *Smoothness is to ensure consistency of depth between neighbouring pixels.*



# REFINEMENT: SEGMENTATION

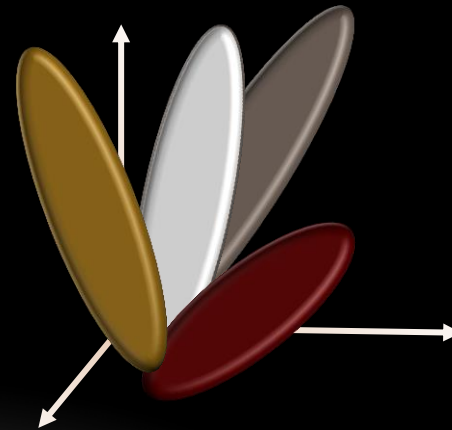
$$E(l,d) = \alpha E_{\text{data}}(d) + \beta E_{\text{smooth}}(l) + \gamma E_{\text{color}}(l) + \eta E_{\text{contrast}}(l,d)$$

where  $l$  is the label and  $d$  is the depth

- Color and contrast information combined with geodesic star-convexity is used to refine segmentation.*



Background and Foreground

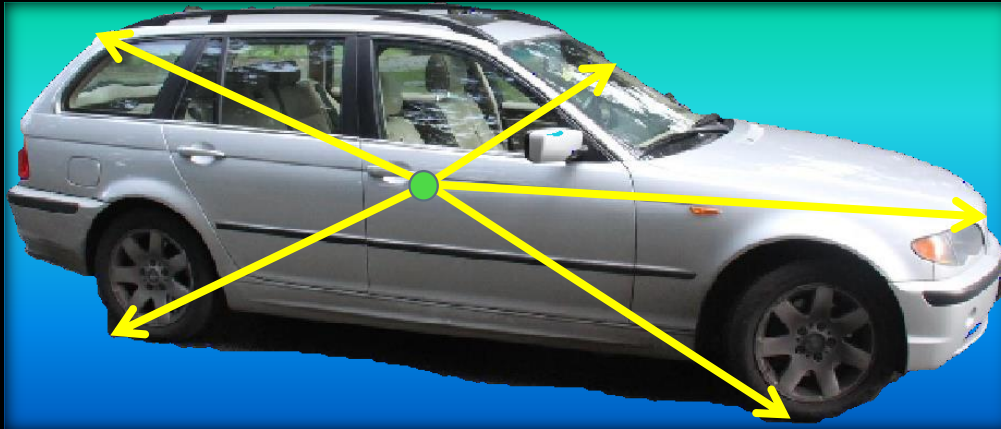


GMM models

# REFINEMENT: SEGMENTATION

## Geodesic star convexity(GSC):

- Shape constraints improves segmentation



Star convex object

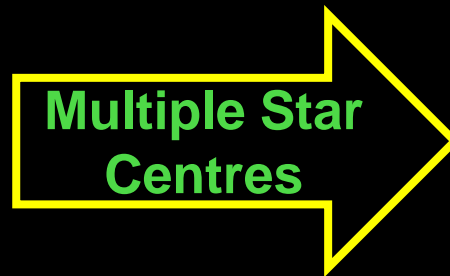


Not star convex

# REFINEMENT: SEGMENTATION

## Geodesic star convexity:

- Geodesic distances instead of Euclidean



# REFINEMENT: SEGMENTATION

$$E(l,d) = \alpha E_{\text{data}}(d) + \beta E_{\text{smooth}}(l) + \gamma E_{\text{color}}(l) + \eta E_{\text{contrast}}(l,d)$$

where  $l$  is the label and  $d$  is the depth

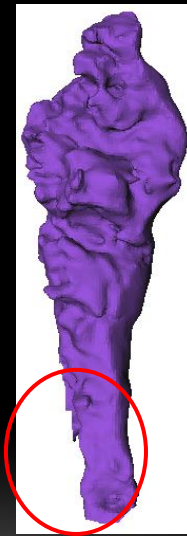
Subject to GSC

- *Geodesic star-convexity to refine segmentation automatically.*

With GSC



Without GSC



# REFINEMENT: SEGMENTATION

Geodesic star convexity:



Input



No constraint



Star convex



Geodesic star convex



# REFINEMENT:

## Temporal coherence:

$$E(l,d) = \alpha E_{\text{data}}(d) + \beta E_{\text{smooth}}(l) + \gamma E_{\text{color}}(l) + \eta E_{\text{contrast}}(l,d)$$



Input



No temporal coherence



Temporal coherence



# REFINEMENT:

## Temporal coherence:

$$E(l,d) = \alpha E_{\text{data}}(d) + \beta E_{\text{smooth}}(l) + \gamma E_{\text{color}}(l) + \eta E_{\text{contrast}}(l,d)$$



Input



No temporal coherence



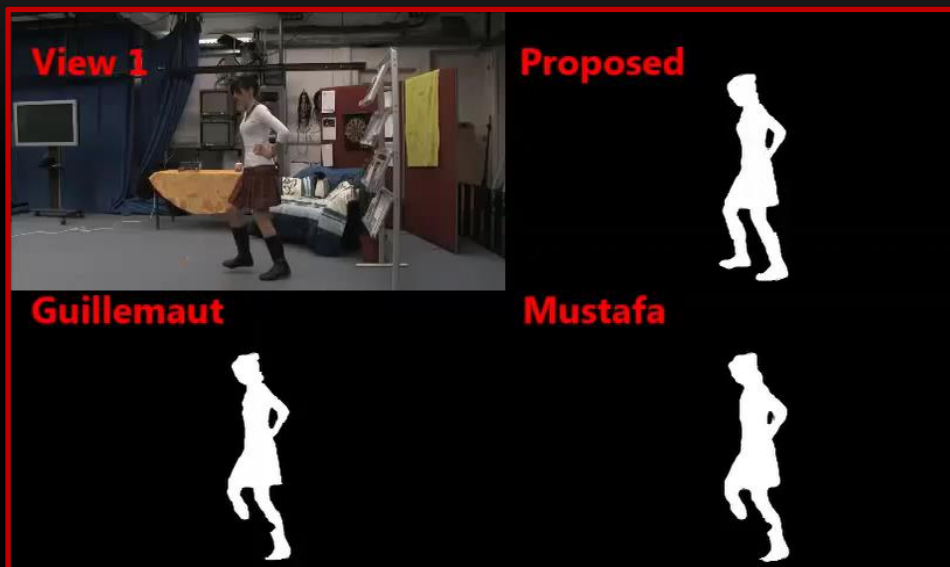
Temporal coherence

# RESULTS

Method	No Priors	Temporal coherence	Joint refinement (Segmentation)
Furukawa PAMI 2010	✓	✗	✗
Guillemaut 3DV 2012	✗	✓	✓
Mustafa ICCV 2015	✓	✗	✓
Proposed	✓	✓	✓

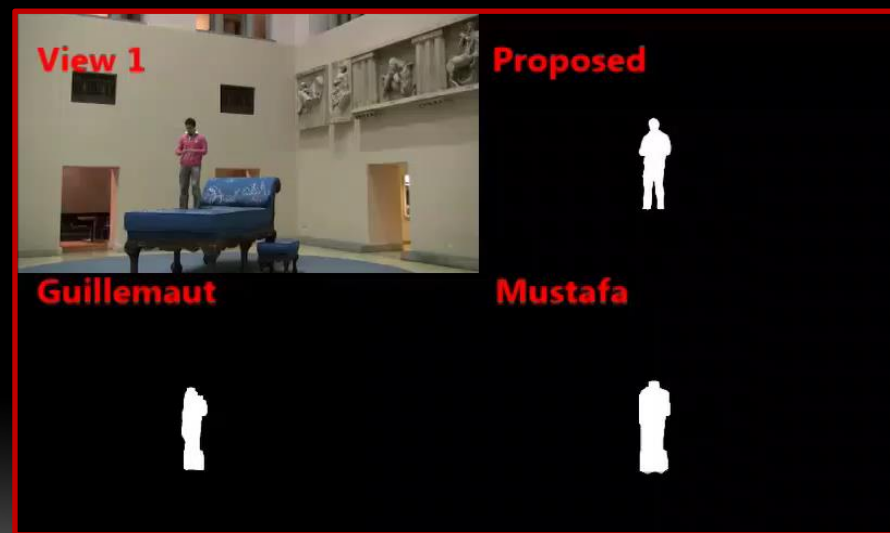
A. **Mustafa**, H. Kim, J-Y. Guillemaut and A. Hilton General Dynamic Scene Reconstruction from Multiple View Video. ICCV 2015  
J-Y. **Guillemaut** and A. Hilton Space-time joint multi-layer segmentation and depth estimation. 3DIMPVT 2012  
Y. **Furukawa** and J. Ponce Accurate, Dense and Robust Multi-View Stereopsis. PAMI 2010

# RESULTS - SEGMENTATION:



Dance dataset

Magician dataset



# RESULTS -RECONSTRUCTION:

Juggler dataset



Initial  
reconstruction



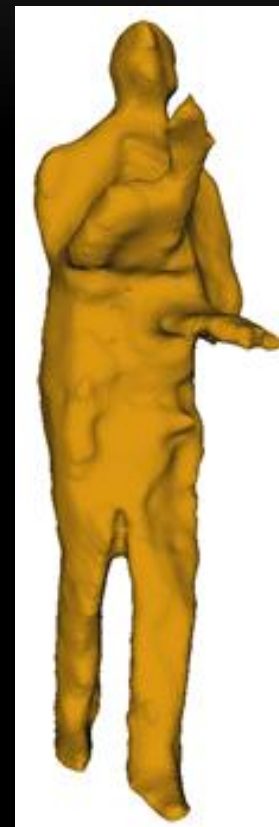
Furukawa



Guillemaut



Mustafa



Proposed

# RESULTS -RECONSTRUCTION:



**Proposed**



**Guillemaut**



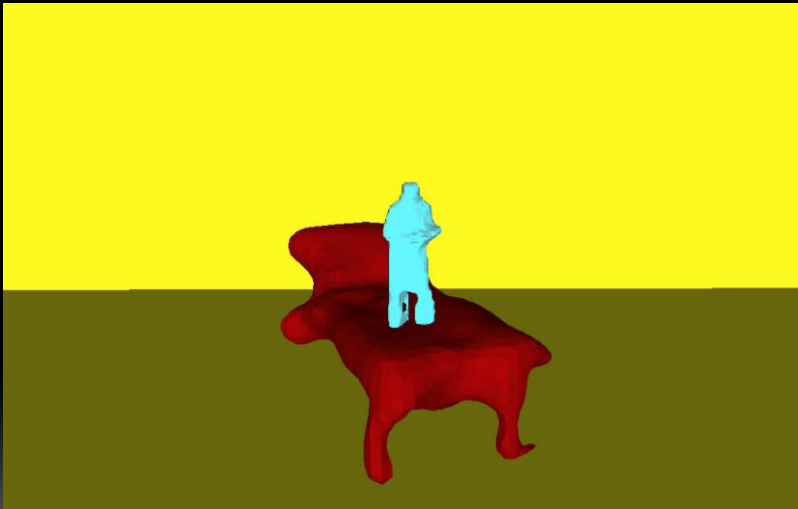
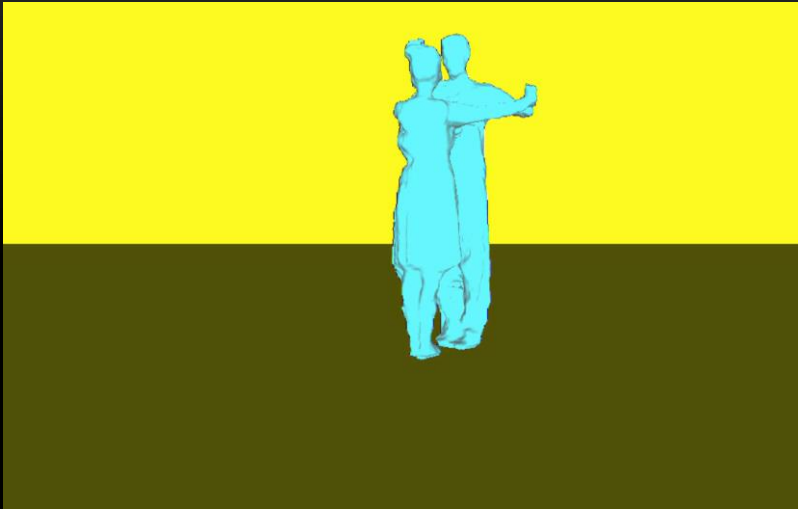
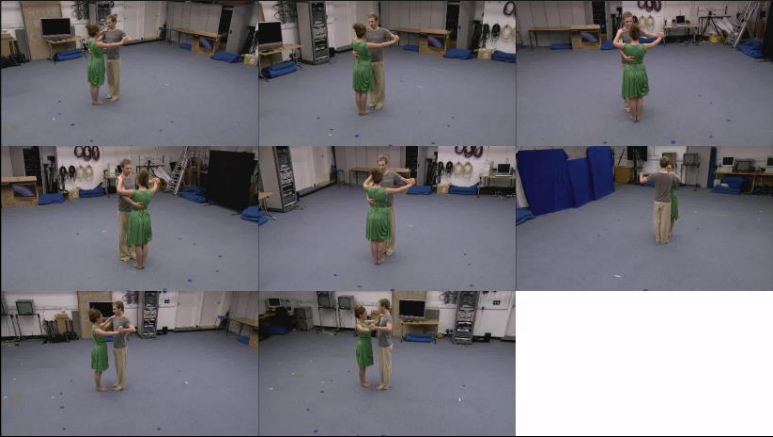
**Mustafa**

Dance dataset

Magician dataset

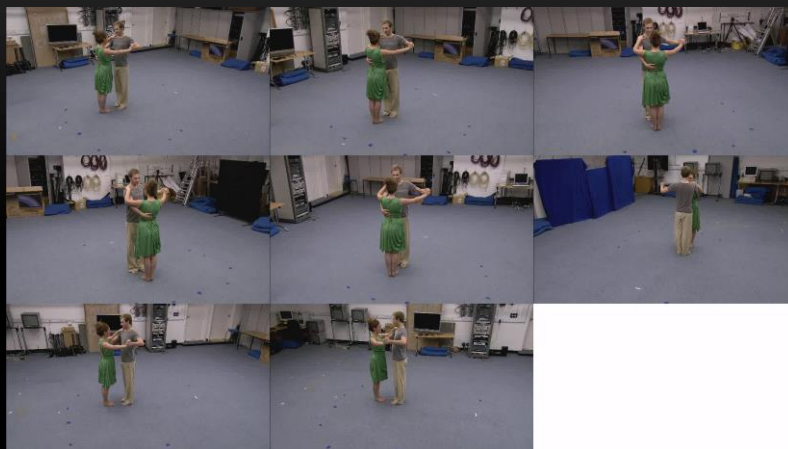


# RESULTS - 4D RECONSTRUCTION:



Input

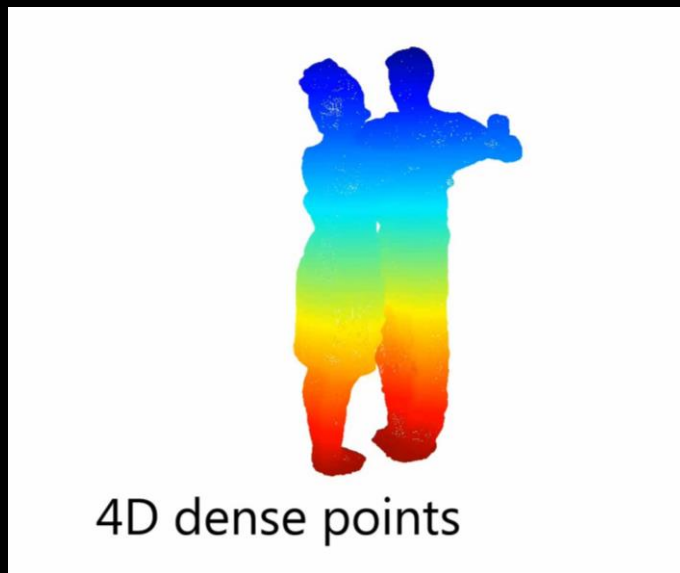
# RESULTS - TEMPORAL COHERENCE :



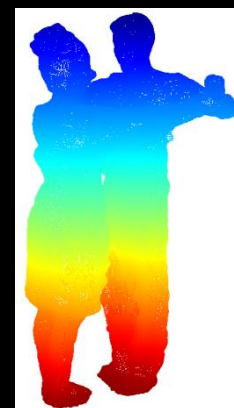
Frame 1



2D dense correspondence



4D dense points



Frame 1



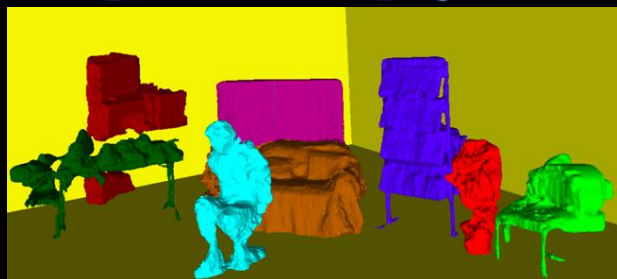
## RESULTS - COMPUTATION TIME:

<b>Dataset</b>	<b>Furukawa (s)</b>	<b>Guillemaut (s)</b>	<b>Mustafa (s)</b>	<b>Ours (s)</b>
Dance1	326	493	295	254
Magician	311	608	377	325
Odzemok	381	598	394	363
Office	339	533	347	291
Juggler	394	634	411	378
Dance2	312	432	323	278



# CONCLUSIONS

- An automatic framework for temporally coherent 4D reconstruction.
- Sparse to dense temporal coherence to improve quality.
- Joint segmentation and reconstruction refinement using GSC.



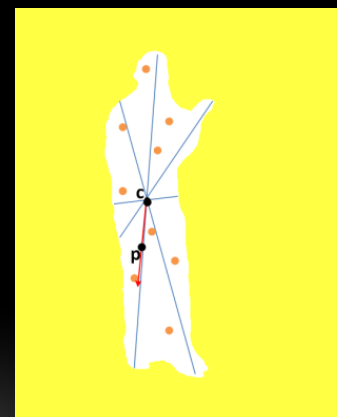
Input



W/O temporal



With temporal



Without geodesic



With geodesic



# FUTURE WORK

- Extending 4D reconstruction to single view video.
- Joint semantic segmentation using recognition.
- Handle crowded dynamic scenes

# THANK YOU!

Temporally coherent 4D reconstruction of complex dynamic scenes

Armin Mustafa, Hansung Kim, Jean-Yves Guillemaut, Adrian Hilton

<http://cvssp.org/projects/4d/4DRecon/>

## Poster number : 12

