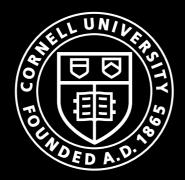
Scene Chronology Kevin Matzen and Noah Snavely

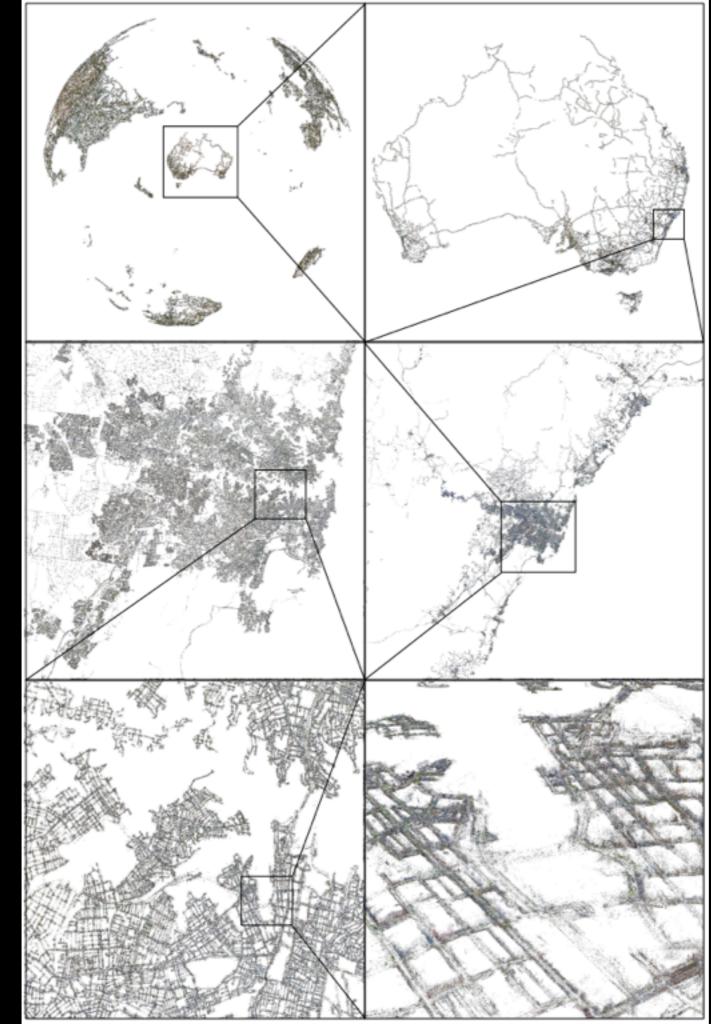


Cornell University



Model of Dubrovnik from 1,000s of Internet photos

[Agarwal 2009]

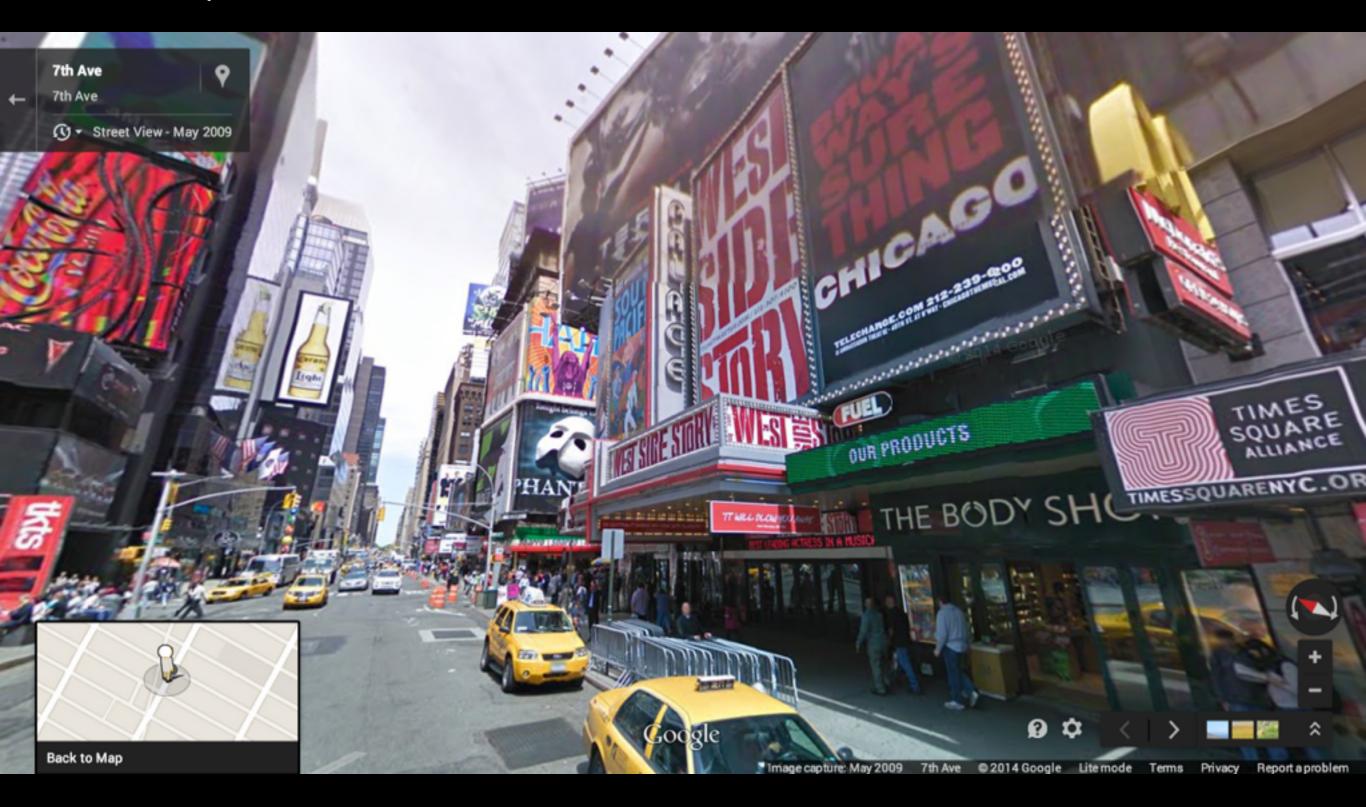


Planet-scale reconstruction [Klingner, et. al. ICCV 2011]

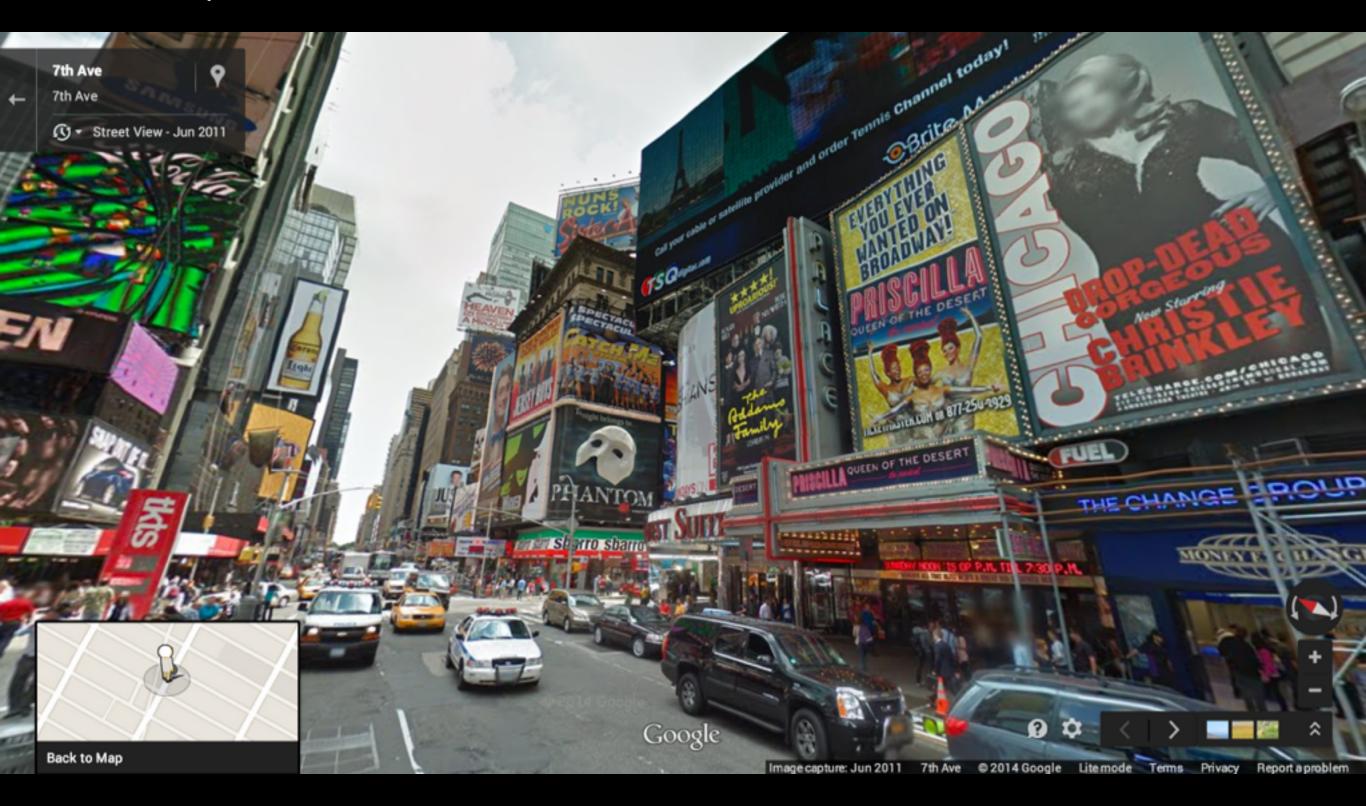
Are we done with 3D modeling?

- Standard assumption: static scenes
- We know the world is constantly changing

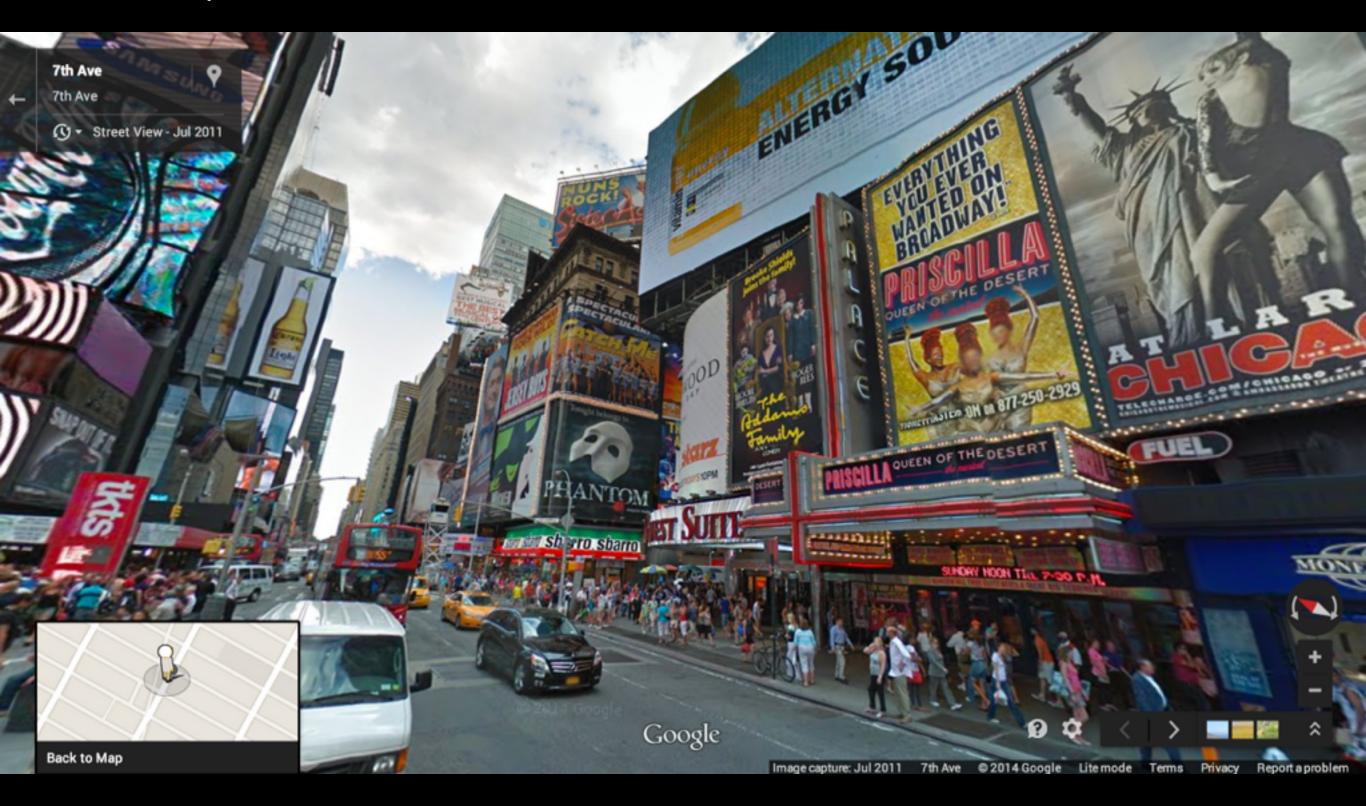
A key ongoing challenge in 3D modeling will be reconstructing **dynamic** scenes







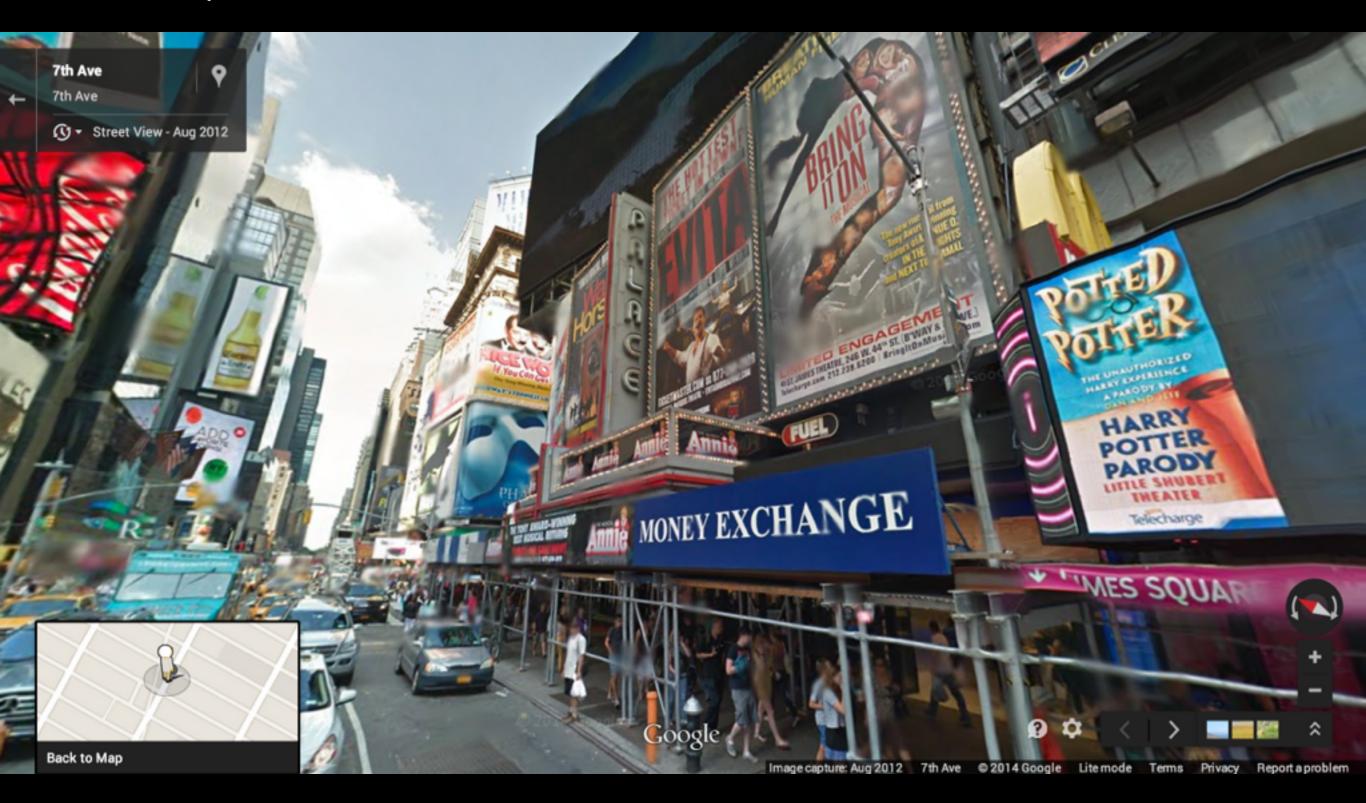
June 2011



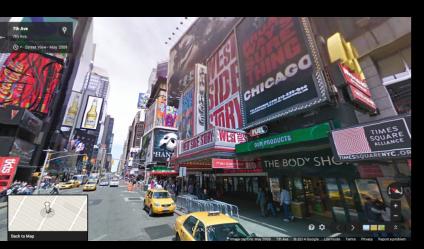
July 2011

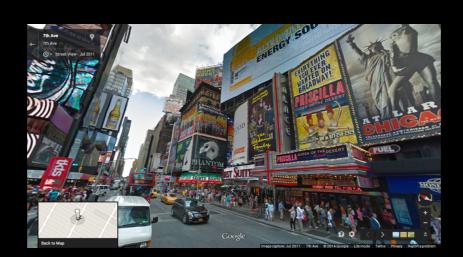


August 2011



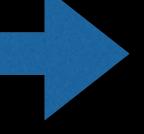
August 2012











Goal

- Rich reconstructions
- Capturing fine-grained temporal structure
 - On the order of weeks or days
- Enabling visualizations where we can dial back to any point in time



Street View?

Structured, high spatial coverage

Low temporal resolution





Street View? Structured, high spatial coverage

Webcams? High temporal resolution Low viewpoint variation

Low temporal resolution







Street View?

Structured, high spatial coverage

Low

temporal resolution

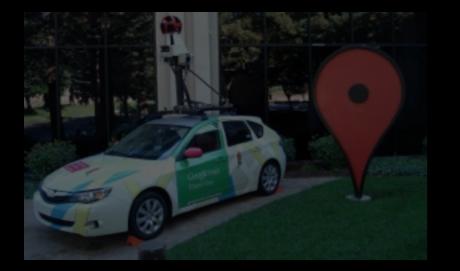
High temporal resolution Low

Webcams?

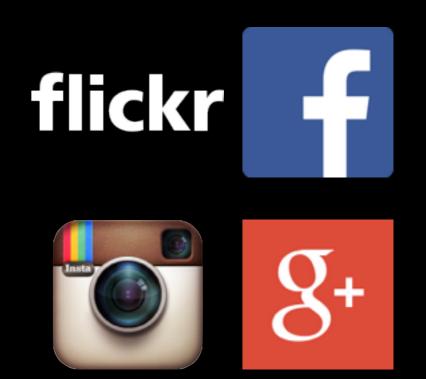
viewpoint variation

User-generated content? Massive collections

Unstructured







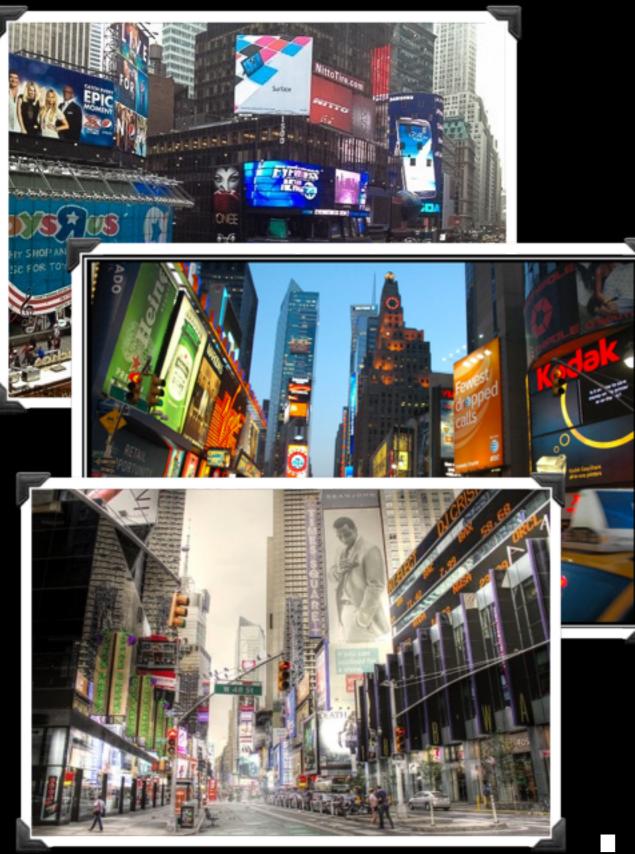
Street View?

Structured, high spatial coverage

Webcams? High temporal resolution User-generated content? Massive collections

Low temporal resolution Low viewpoint variation

Unstructured







Challenges

- Noisy timestamps
- Irregular spatial and temporal sampling
- Large scale photo collections
 - Need to sample 4D space of photos

Related Work



[Schindler CVPR 2010]

192 1928

[Pollard CVPR 2007] [Taneja ICCV 2011] [Taneja CVPR 2013] [Ulusoy ECCV 2014] ...

Related Work



Our scale differs significantly. Let's rethink the whole approach.

[Schindler CVPR 2010]

192 1928

[Pollard CVPR 2007] [Taneja ICCV 2011] [Taneja CVPR 2013]

Contributions

New approach for reconstructing chronology of a scene from millions of unstructured photos

- New method for robustly estimating per-point time intervals for a 3D reconstruction
- New plane-time representation and segmentation algorithm
- Scalable to millions of photos

Outline

- Representation for 4D scenes
- 4D reconstruction approach
- Results

Possible 4D Representations

- 3D points augmented with time
 - Too low level
- Temporally coherent sets of 3D points
 - Basis for spatio-temporal segmentation









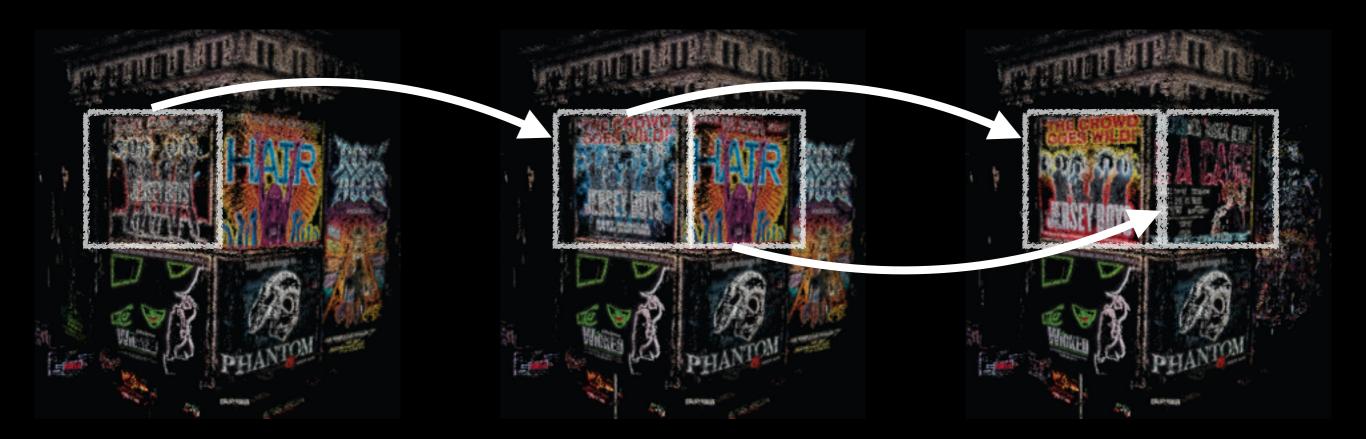








Time



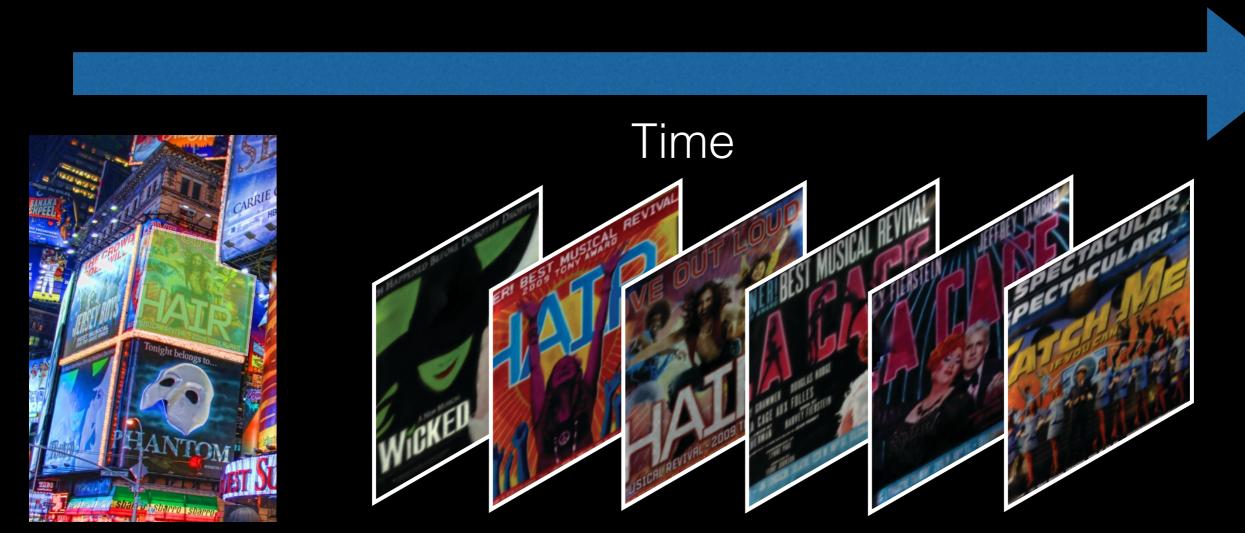


Time











"Times Square" fickr



"Times Square" fickr



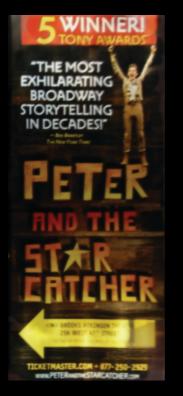
"Times Square" flickr

Approach



"Times Square" flickr









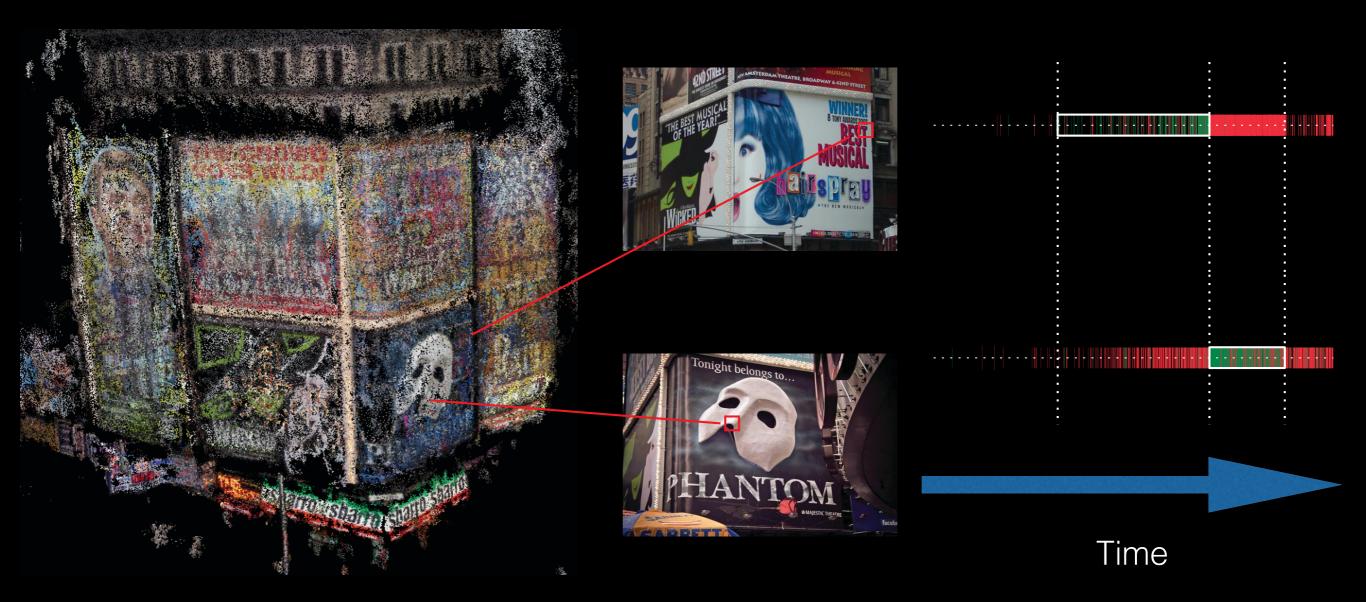
EXCLUSIVELY AT #Discovery Times Square Expose 2 BLOCKS SOUTH 44TH BETWEE





Approach





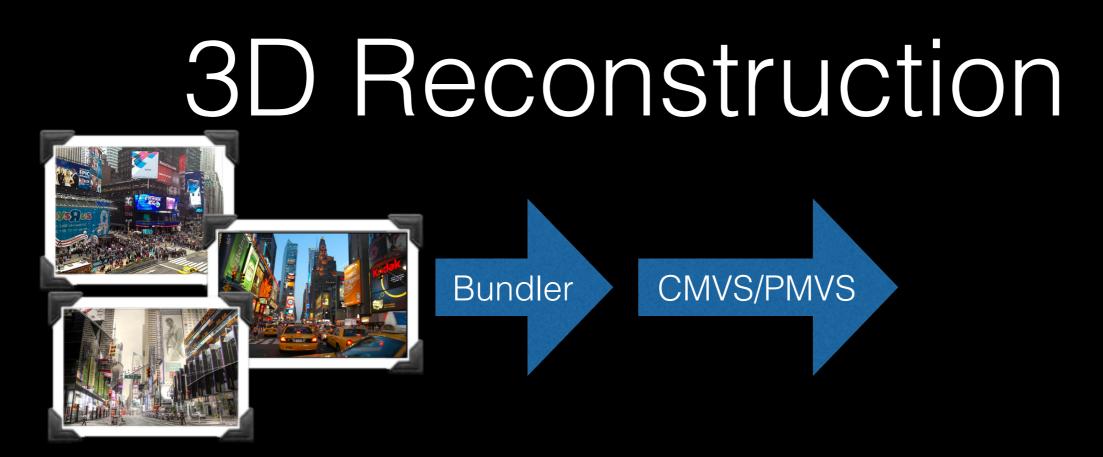
Point-based 4D Reconstruction

3D Reconstruction



Input: Millions of photos

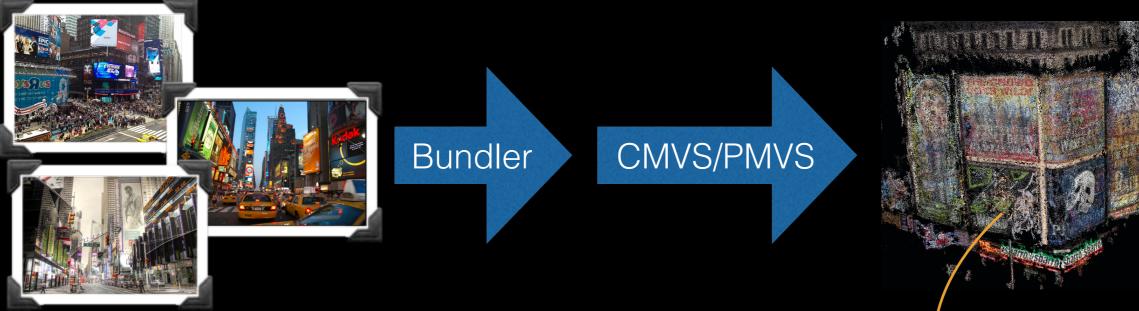
[Agarwal ICCV 2009] [Furukawa PAMI 2010] [Furukawa CVPR 2010]



Input: Millions of photos

[Agarwal ICCV 2009] [Furukawa PAMI 2010] [Furukawa CVPR 2010]

3D Reconstruction



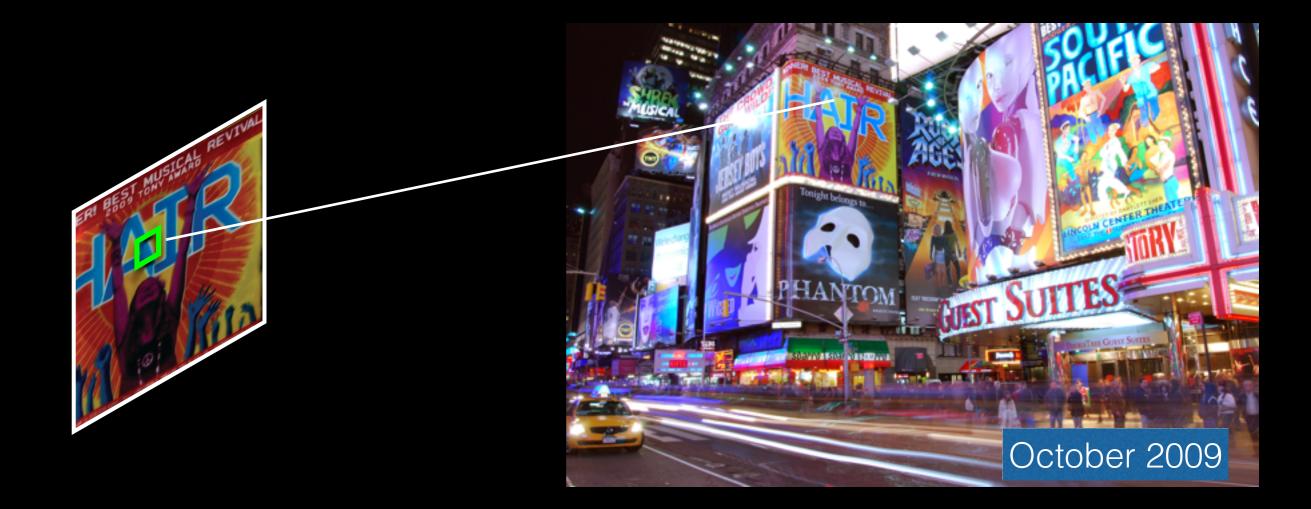
Input: Millions of photos

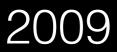
Output:
(1) Millions of patches with position, normal, and appearance
(2) 100,000s registered photos 3D
(3) Visibility information

3D patch with appearance and normal

[Agarwal ICCV 2009] [Furukawa PAMI 2010] [Furukawa CVPR 2010]





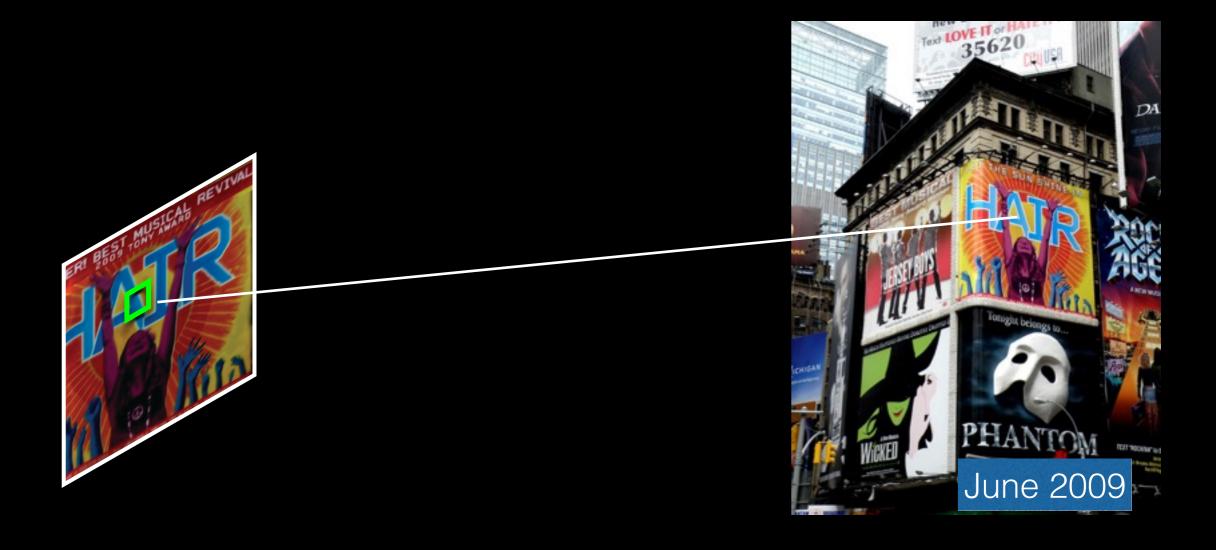


2014





2009



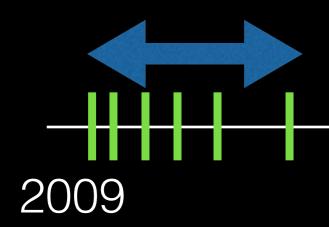




2014

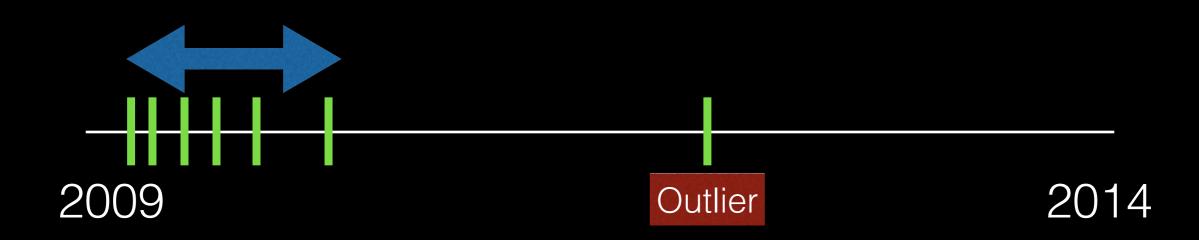


Is this visibility list enough?

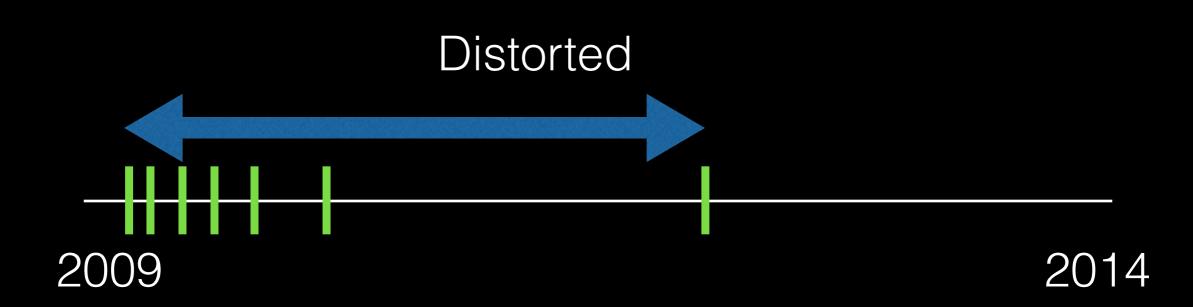




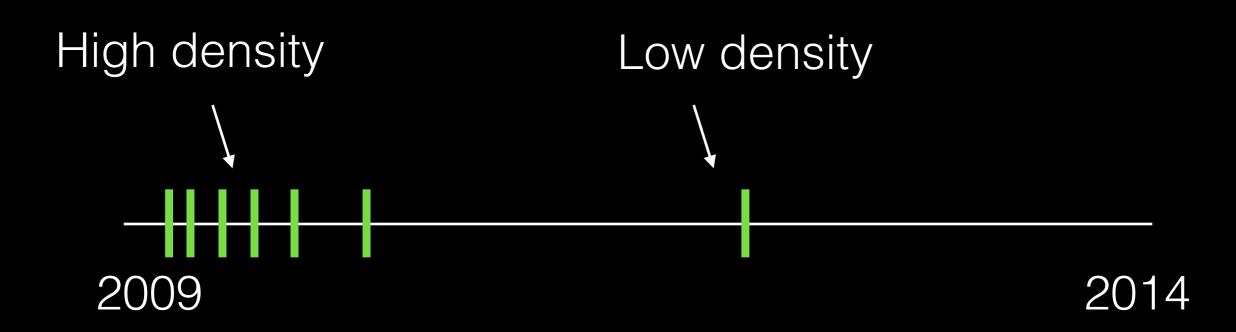
Is this visibility list enough?



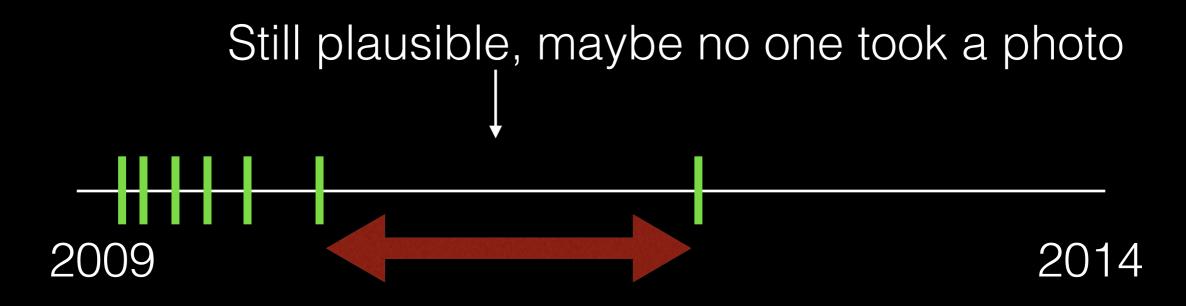
Is this visibility list enough? Min and max are not robust



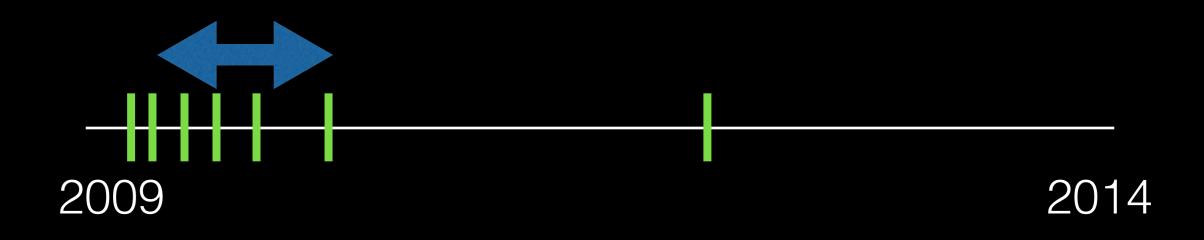
Is this visibility list enough? Min and max are not robust



Is this visibility list enough? Min and max are not robust Density might be misleading



Is this visibility list enough? Min and max are not robust Density might be misleading Trimming leads to systematic shrinkage



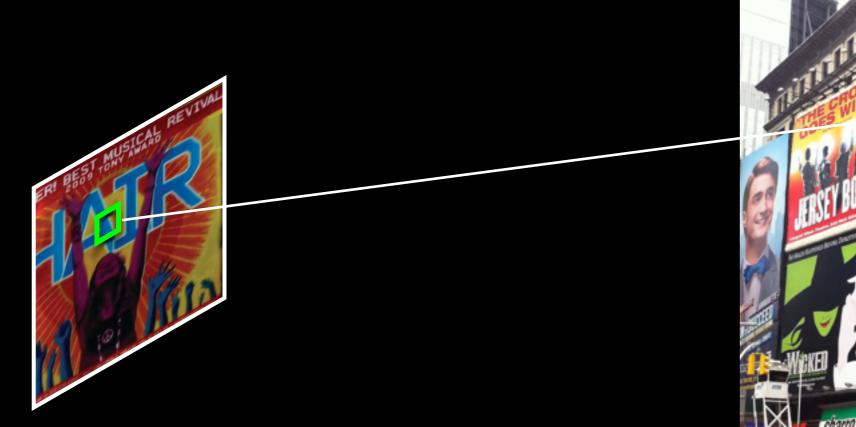
Is this visibility list enough? Min and max are not robust Density might be misleading Trimming leads to systematic shrinkage Can we make use of negative information?







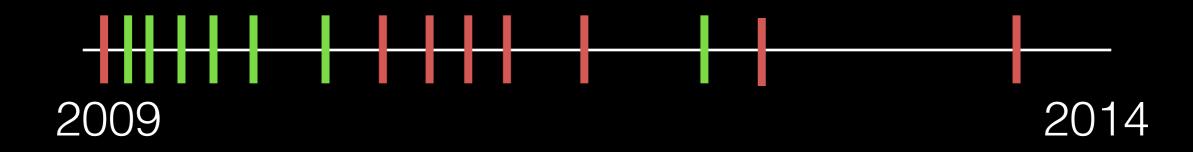






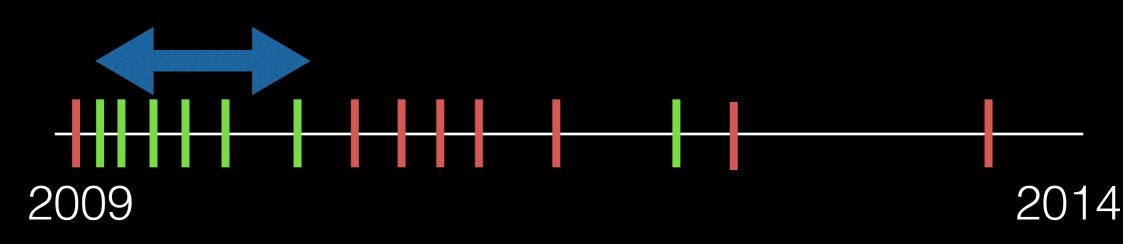








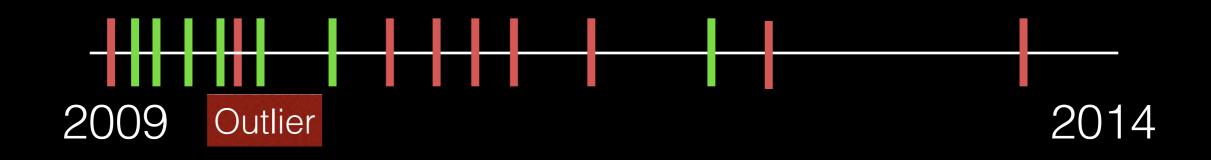
Both positive and negatives tell a part of the story





Both positive and negatives tell a part of the story

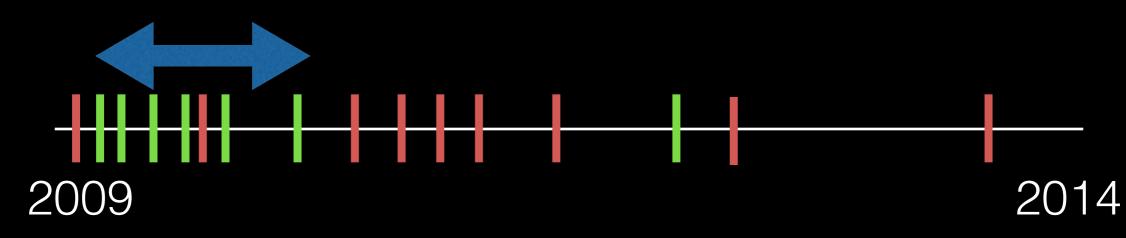
But there can be negative outliers too



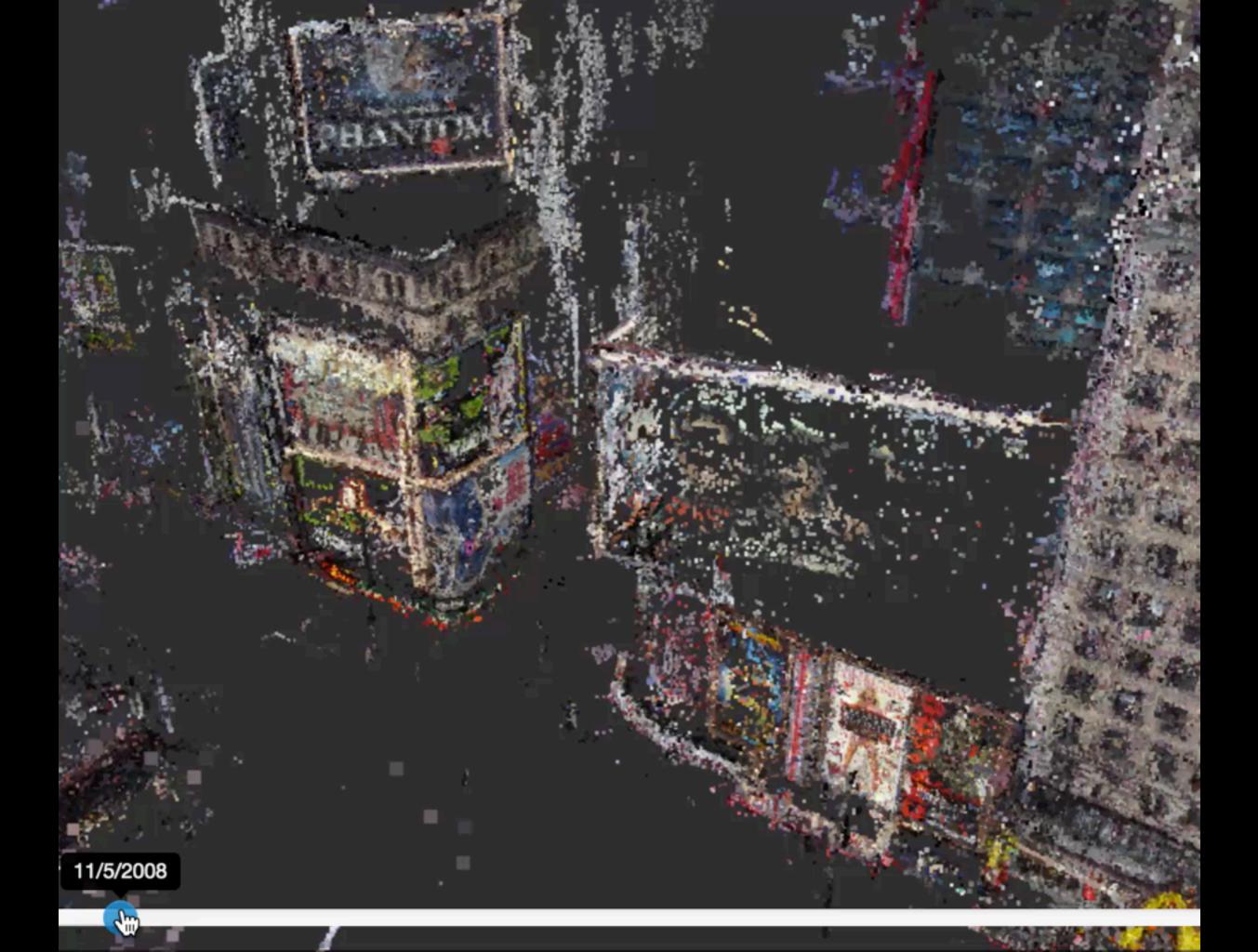
Estimating Intervals

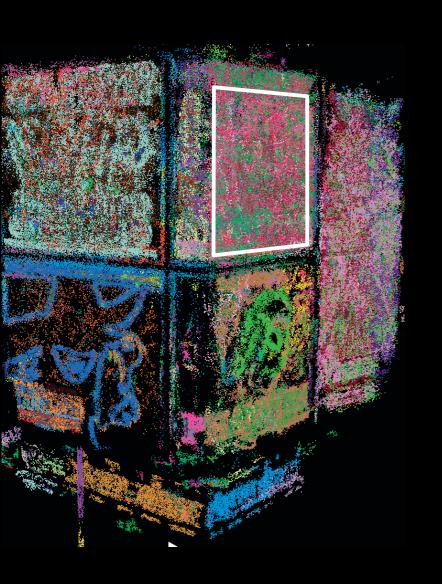


Idea: Use F₁ score to balance precision and recall



Point-Based Visualization







Time

Spatio-Temporal 4D Segmentation

Spatio-Temporal Graph

- Point-to-point affinity encodes
 - position distance
 - difference in normal
 - temporal interval overlap

RANSAC Segmentation

- Plane-Time RANSAC
 - Planar hypothesis with temporal extent
- Find inliers
- Removed induced subgraph
- Apply recursively to remaining connected components

Results

Times Square - Manhattan, NYC

HSBC 41

1.2 million photos250k registered13.5 million points17 billion observations

Times Square - Manhattan, NYC

November 2008



Akihabara - Tokyo

JUR CV 170k photos 14k registered 1.7 million points 400 million observations

Akihabara - Tokyo

September 2010



5Pointz - Queens, NYC



5Pointz - Queens, NYC



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Discovered Space-Time Elements



2 August 2013 to 8 December



7 August 2012 to 3 May



25 May 2010 to 13 May 2012



5 June 2010 to 29 April 2011



16 April 2008 to 8 February 2014



31 March '04 to 26 August '13



17 July 2005 to 5 May 2008 Timestamp Prediction





















Limitations

- Assume we can reconstruct the 3D scene
- Assumes enough redundancy to identify incorrect timestamps
- Some semantic elements are periodic
- Segmentation granularity dependent on thresholds

Conclusion

- Scalable system
- Unreliably timestamped photos to 4D reconstruction
- Two representations: (1) Point-based and (2) Semantic segmentation
- Future work
 - Moving beyond planes
 - More sophisticated timestamp prediction
 - Richer, more complete visualizations
 - Non-intersection constraints

Acknowledgements

- National Science Foundation
- Intel Science and Technology Center for Visual Computing
- Amazon Web Services in Education

Questions?

bit.ly/cornell-chrono