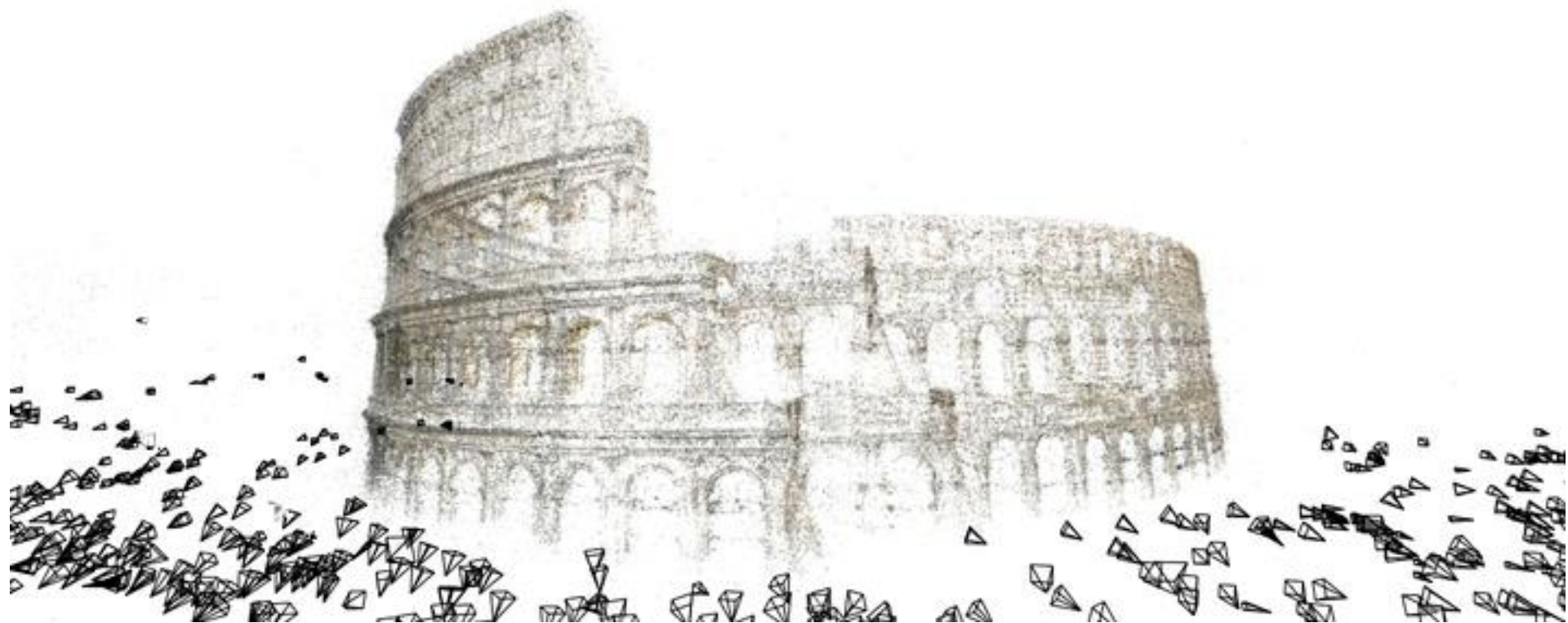


Structure from Motion



Frank Dellaert, Fall 2011

Building Rome in a Day
Agarwal et al

Outline

- Motivation/Visualization
- Feature Extraction
- Matching
- Optimization

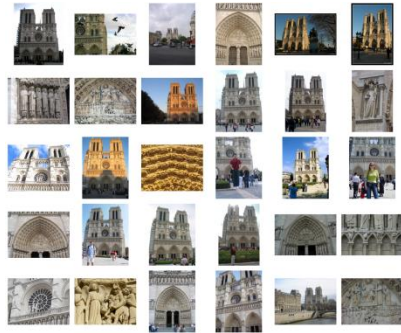
Motivation

- Photo Tourism
- Photosynth
- Multi-view stereo
- Building Rome in a Day
- Rome on a Cloudless Day

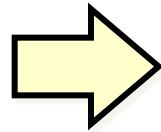
**See also CVPR 2010 Short Course:
Scene Reconstruction from Community Photo Collections**

Photo Tourism

Noah Snavely, Steven M. Seitz, Richard Szeliski,
[Photo tourism: Exploring photo collections in 3D," ACM Transactions on Graphics \(SIGGRAPH Proceedings\), 25\(3\), 2006, 835-846.](#)



Input photographs



**Scene
reconstruction**

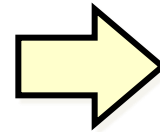
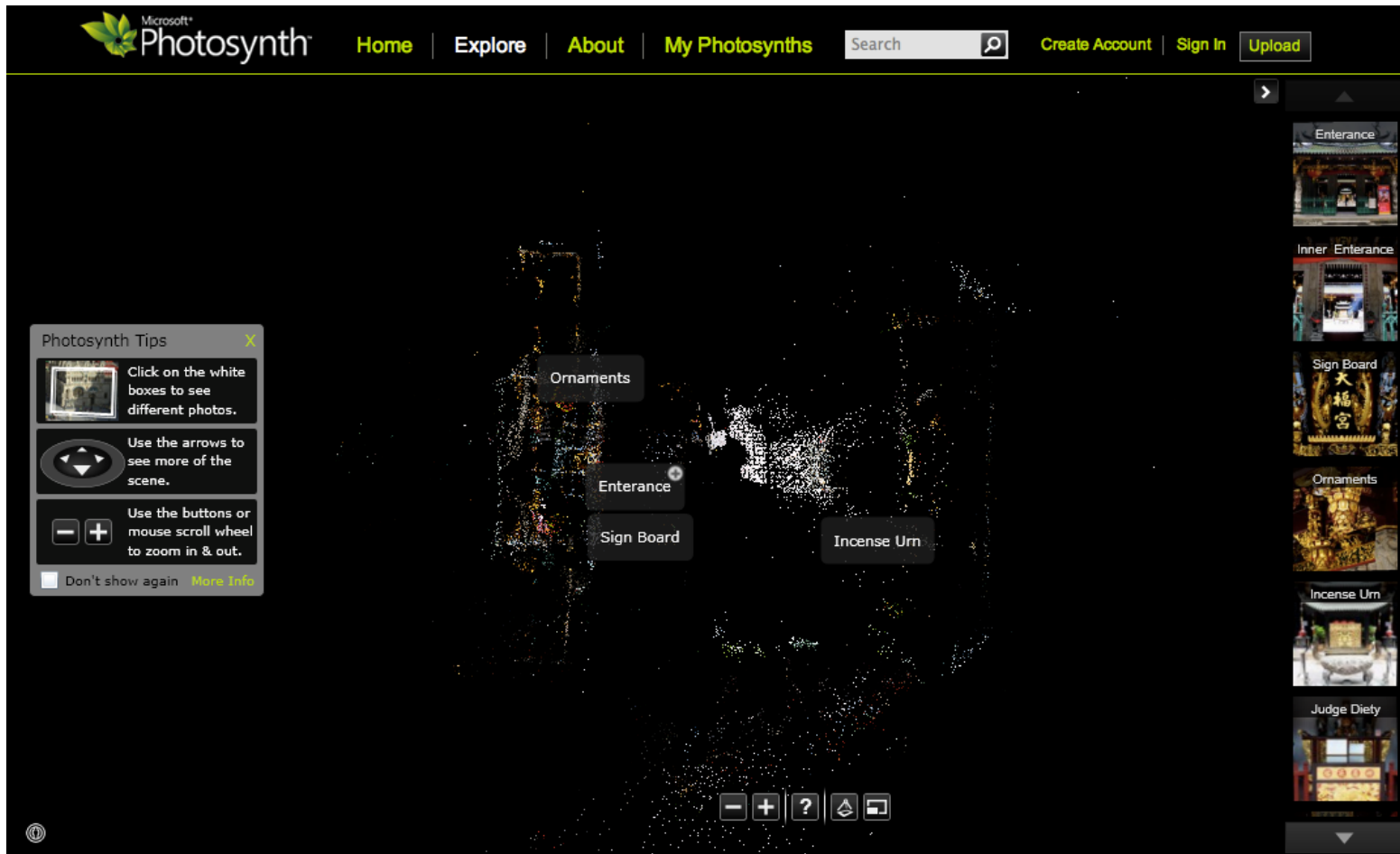


Photo Explorer

<http://phototour.cs.washington.edu/>

Photosynth

photosynth.net



- <http://photosynth.net/view.aspx?cid=29aa8616-a43a-43e4-9d6e-b8ad9b50483e>

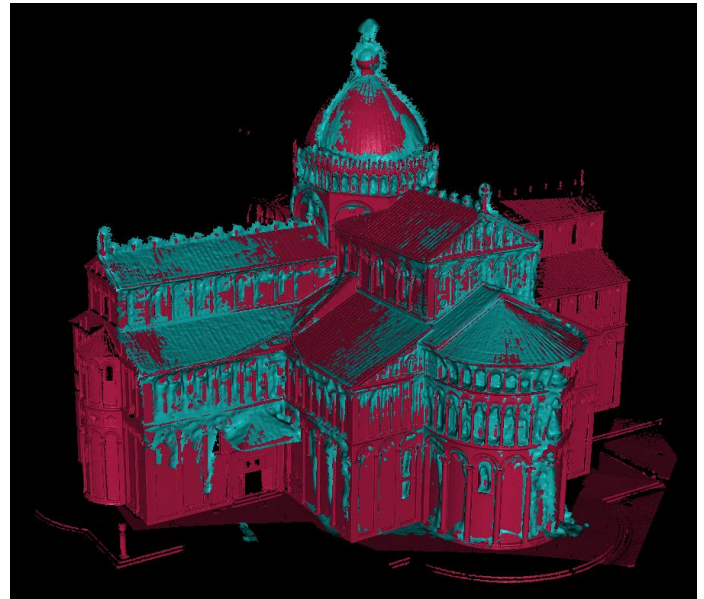
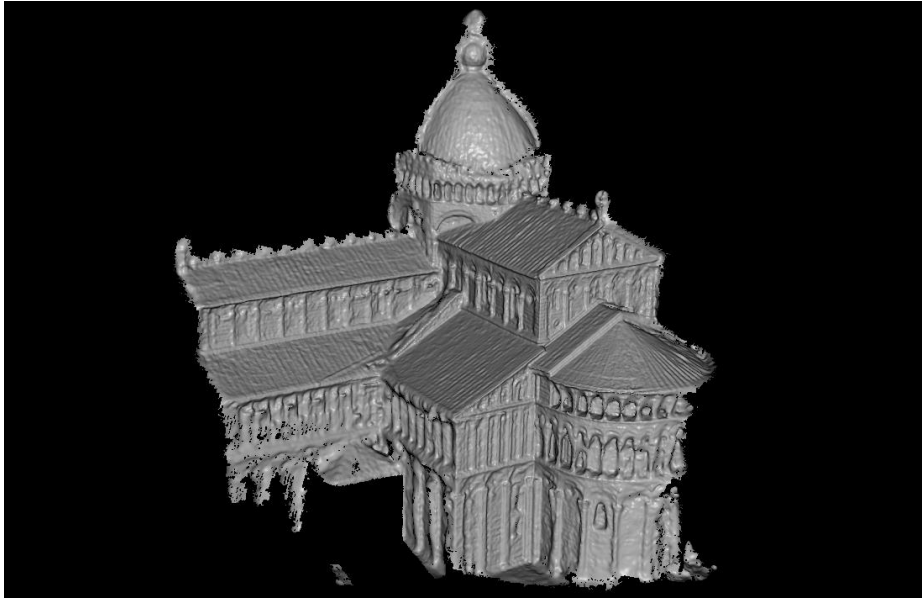
Multi-view Stereo

Multi-View Stereo for Community Photo Collections

Michael Goesele, Noah Snavely, Brian Curless, Hugues Hoppe, and Steven M. Seitz
ICCV 2007



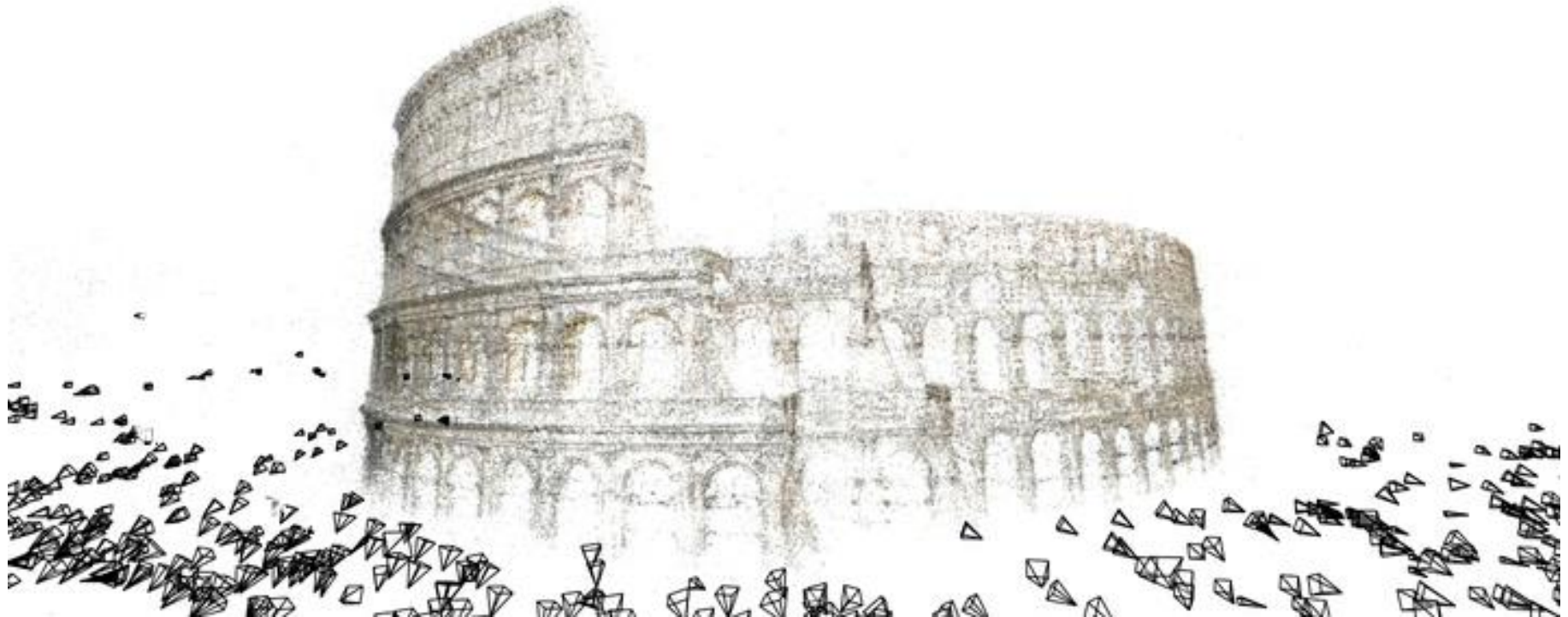
Multi-view Stereo



Compared with Laser-Scanner

Building Rome in a Day

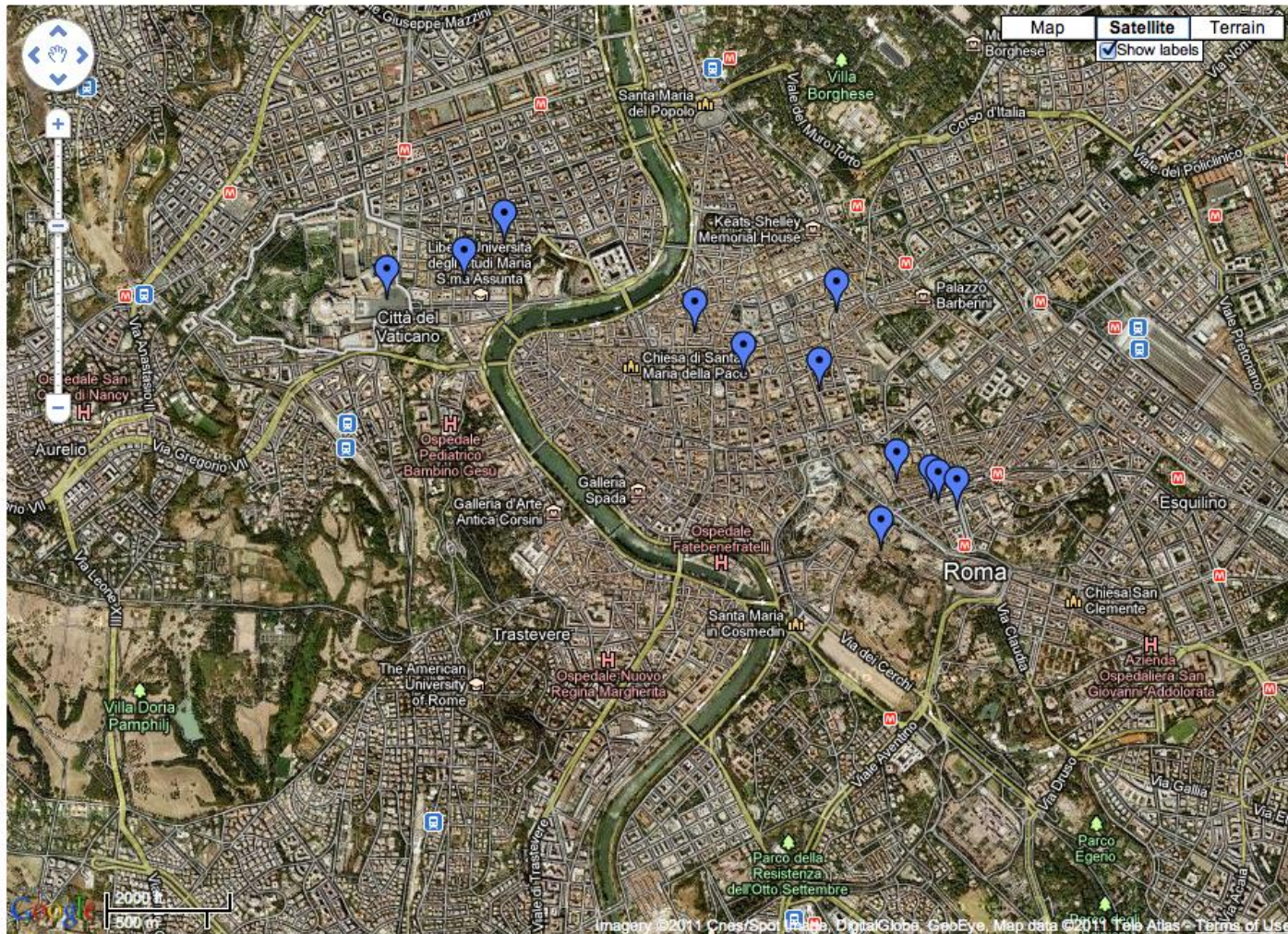
Building Rome in a Day Sameer Agarwal, Noah Snavely, Ian Simon, Steven M. Seitz and Richard Szeliski International Conference on Computer Vision, 2009, Kyoto, Japan.



<http://grail.cs.washington.edu/rome/>

Rome on a Cloudless Day

Jan-Michael Frahm, Pierre Georgel, David Gallup, Tim Johnson, Rahul Raguram, Changchang Wu, Yi-Hung Jen, Enrique Dunn, Brian Clipp, Svetlana Lazebnik, Marc Pollefeys, *ECCV 2010*

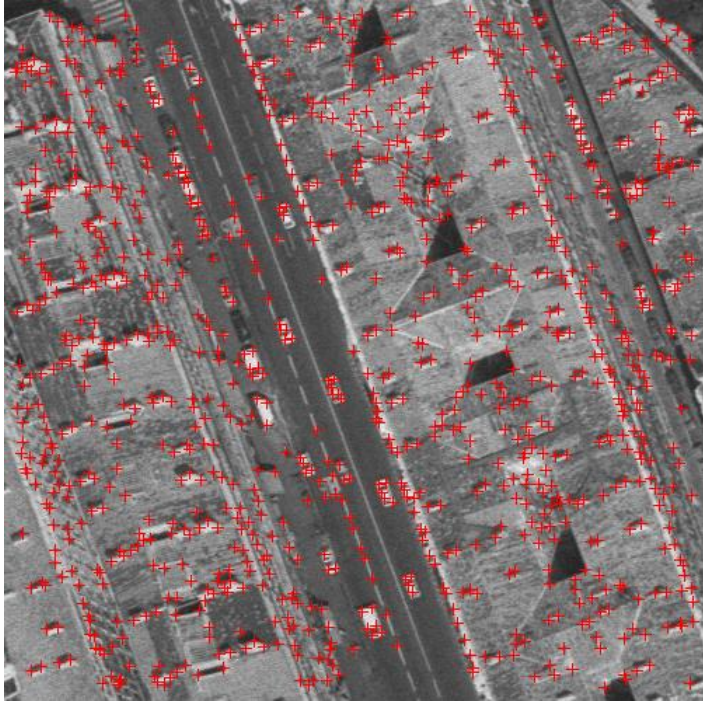


http://www.cs.unc.edu/~jmf/rome_on_a_cloudless_day/

Outline

- Motivation/Visualization
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Interest points



Geometric features

➡ repeatable under transformations

➡ 2D characteristics of the signal

high informational content

Comparison of different detectors
[Schmid98]

➡ Harris
detector

Harris detector

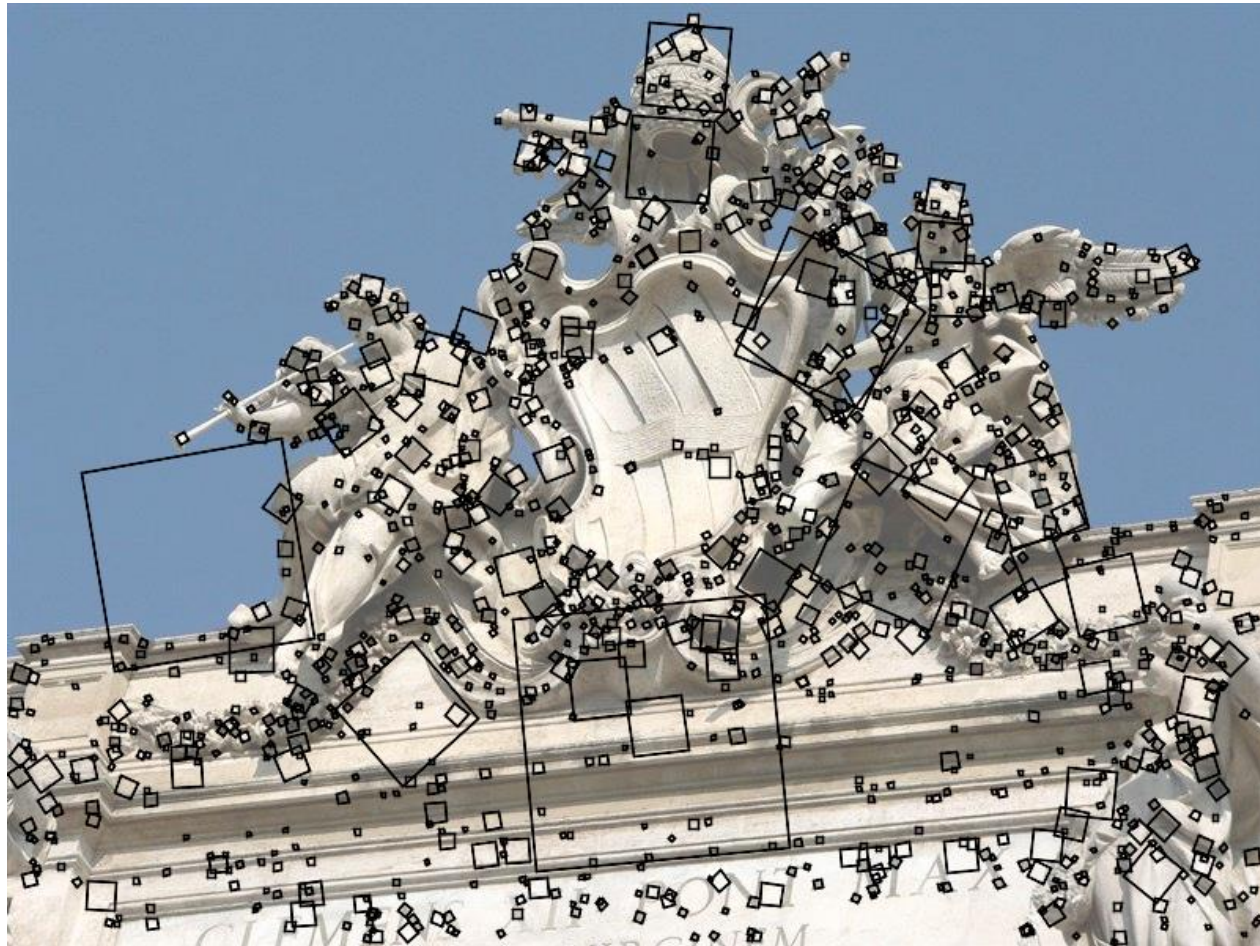
Based on the idea of auto-correlation



Important difference in all directions => interest point

Feature detection

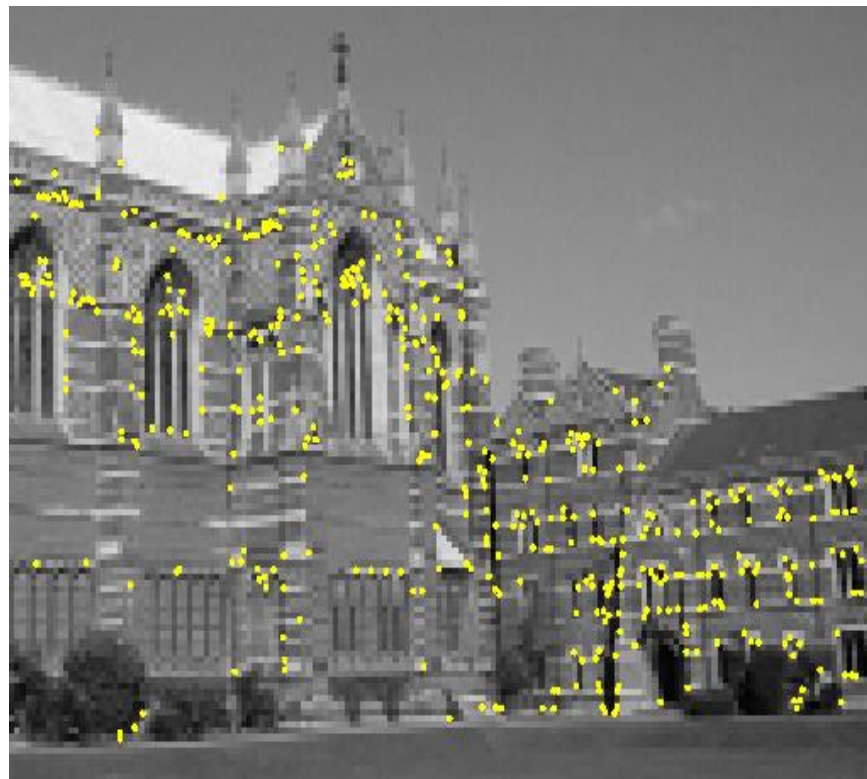
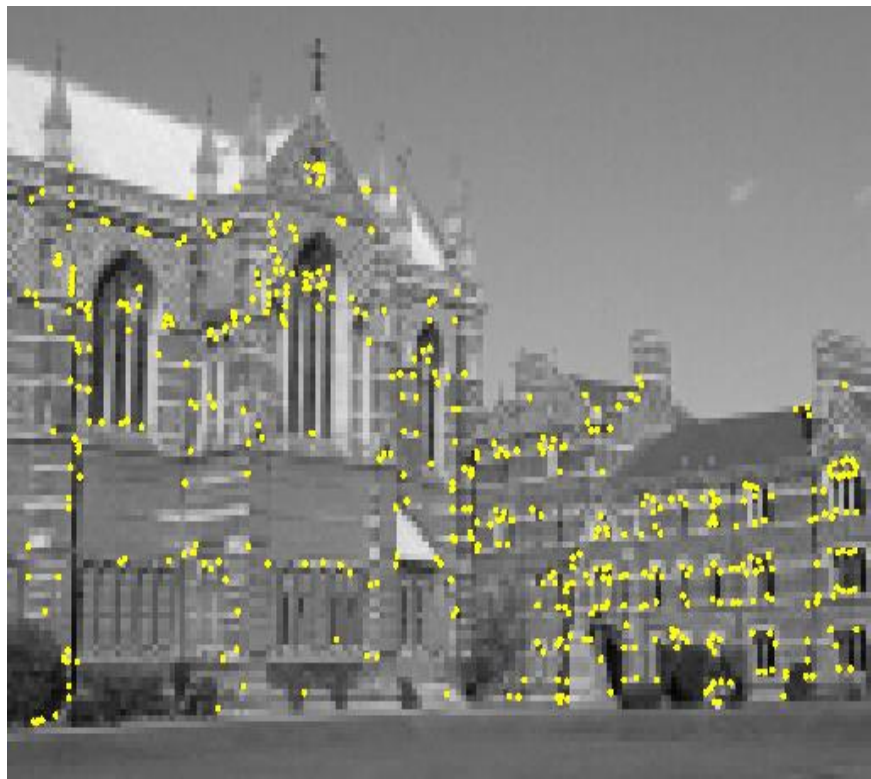
Detect features using SIFT [Lowe, IJCV 2004]



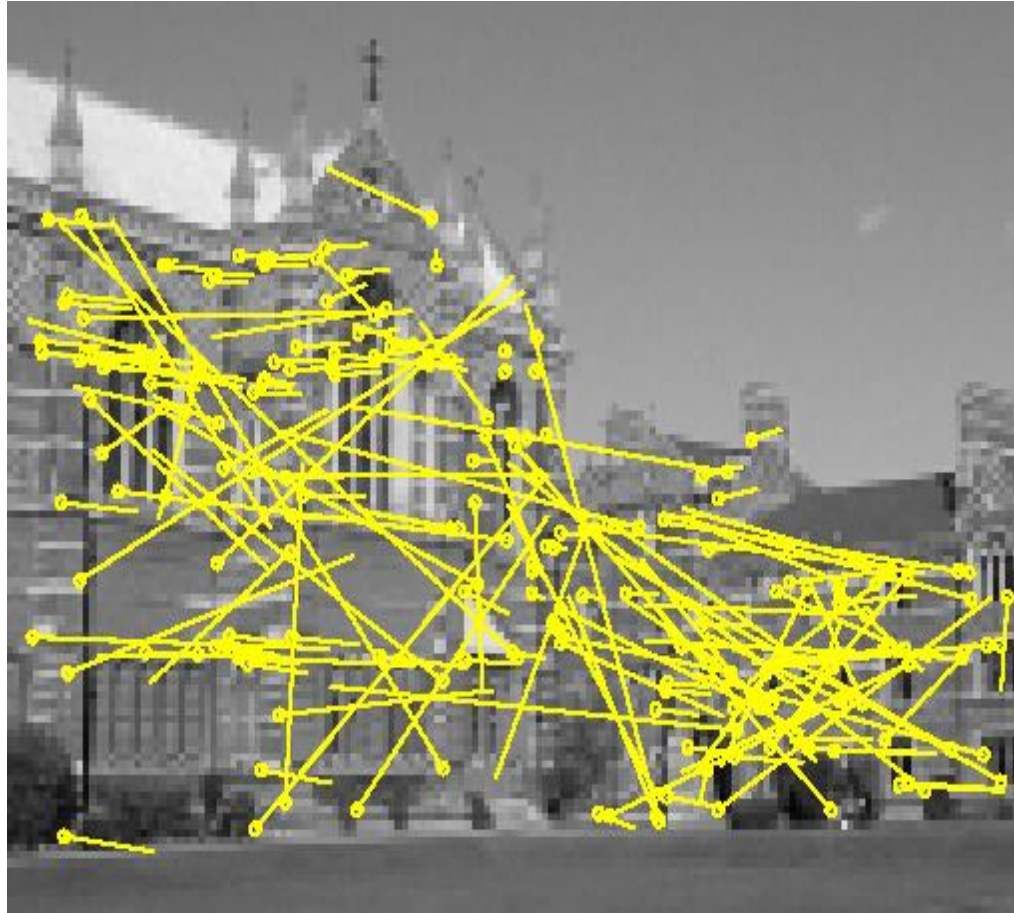
Outline

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Feature Matching !



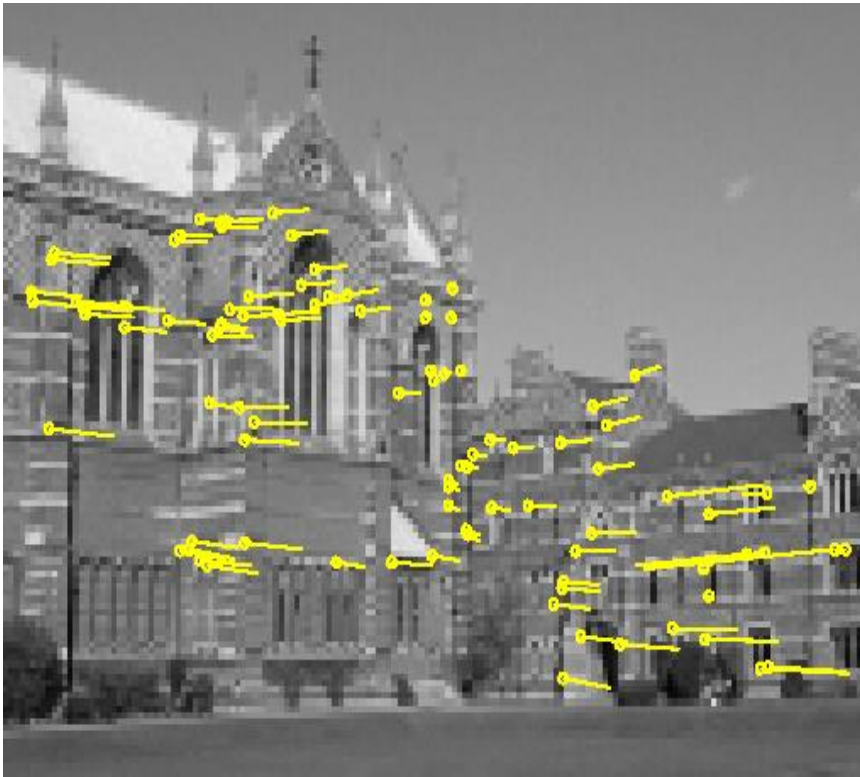
Cross-correlation matching



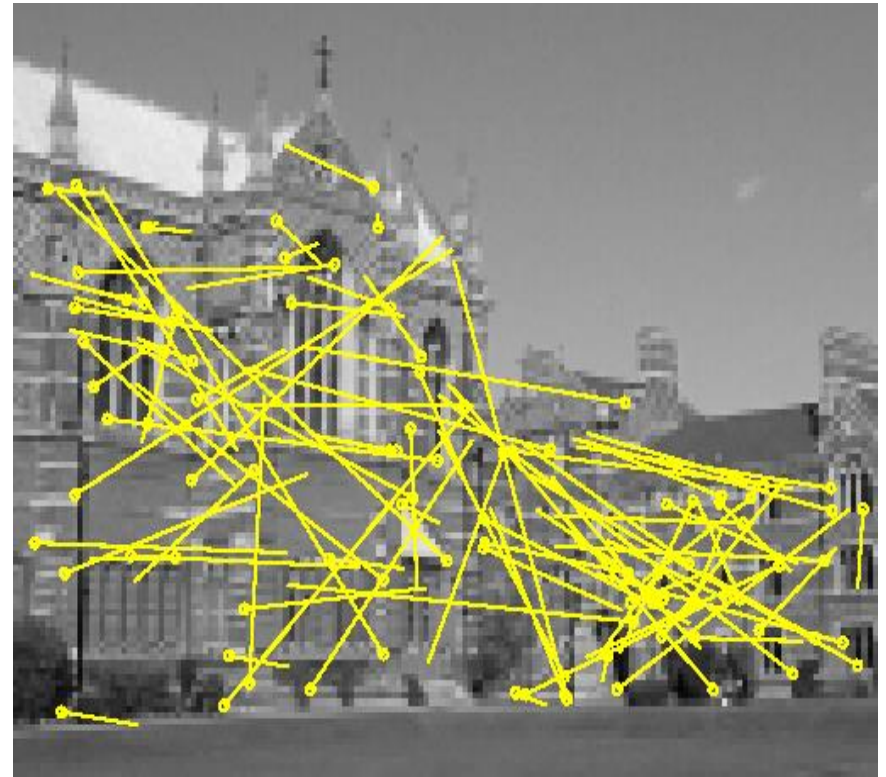
Initial matches (188 pairs)

Global constraints

Robust estimation of the fundamental matrix



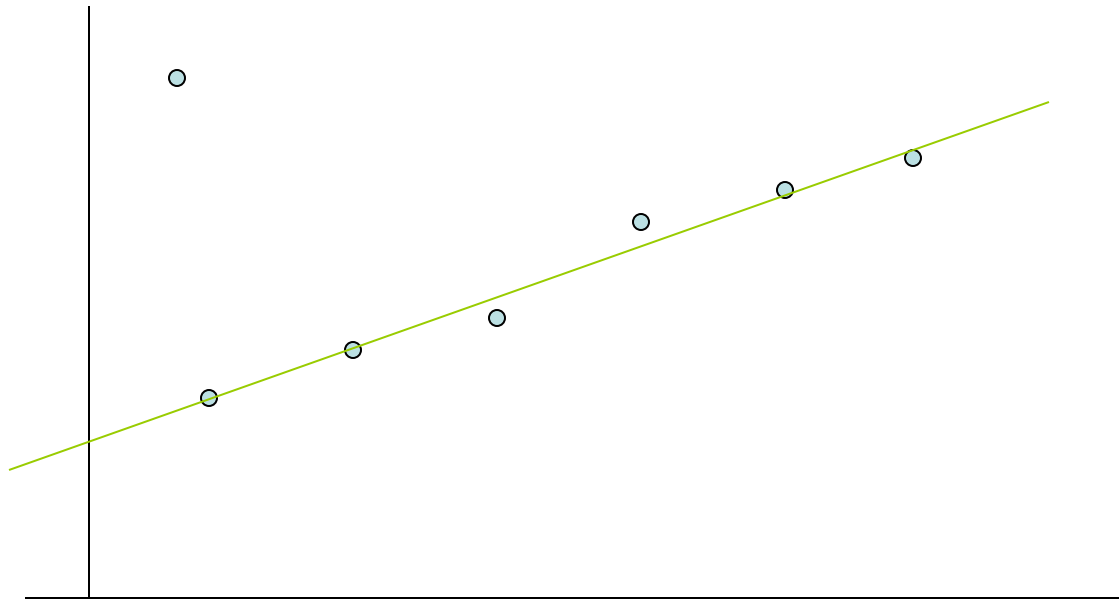
99 inliers



89
outliers

Simpler Example

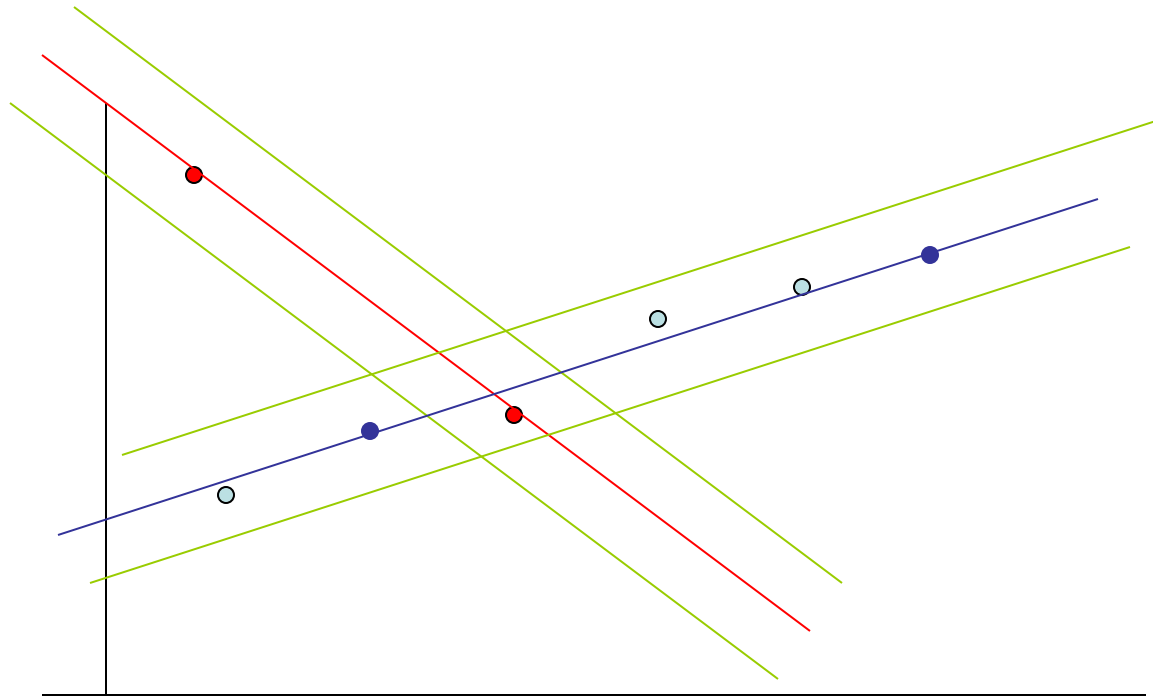
- Fitting a straight line



RANSAC

- Select 2 points at random
- Fit a line
- “Support” = number of inliers
- Line with most inliers wins

Why will this work ?



Outline

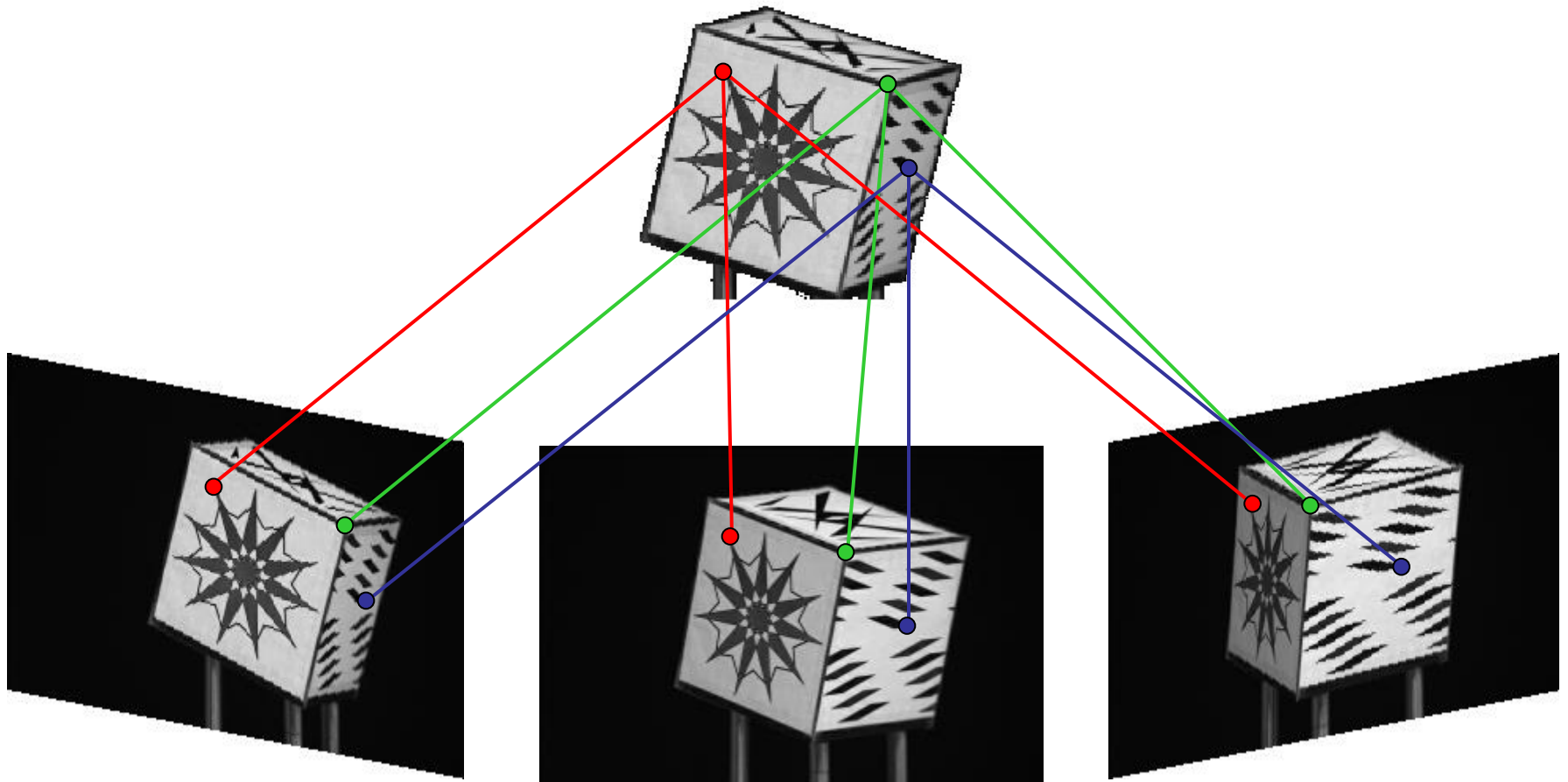
- Motivation/Visualization
- Feature Extraction
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- Optimization

2 Problems !

Correspondence

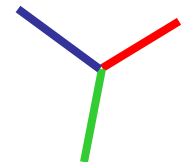
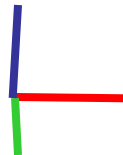
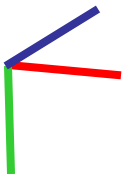
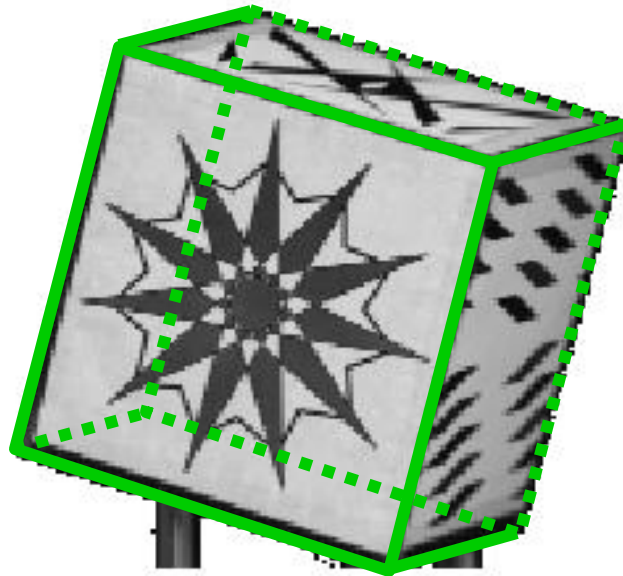
Optimization

A Correspondence Problem



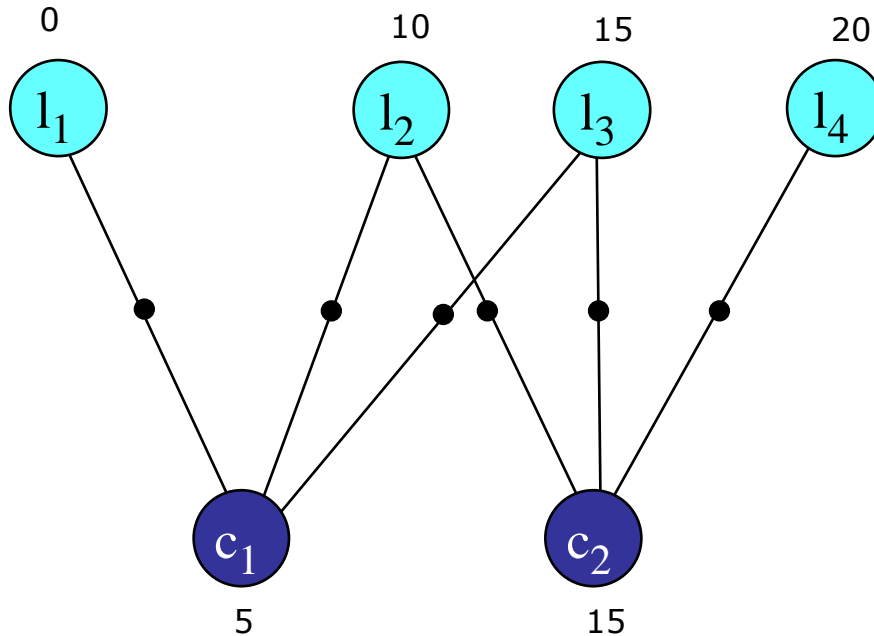
An Optimization Problem

- Find the **most likely** structure and motion Θ



Very simple example

- 1-dimensional



$$u = l - c$$

$$X = [c_1 \ c_2 \ l_1 \ l_2 \ l_3 \ l_4]'$$

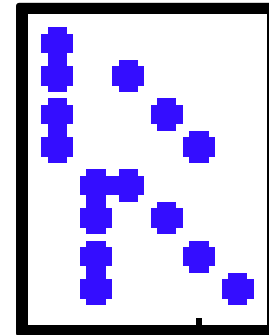
Sparse Jacobian

A =

1	0	0	0	0	0
-1	0	1	0	0	0
-1	0	0	1	0	0
-1	0	0	0	1	0
0	-1	1	0	0	0
0	-1	0	1	0	0
0	-1	0	0	1	0
0	-1	0	0	0	1

b =

5
-5
5
10
-15
-5
0
5



$A^*A = \text{inv}(\text{Sigma})$

4	0	-1	-1	-1	0
0	4	-1	-1	-1	-1
-1	-1	2	0	0	0
-1	-1	0	2	0	0
-1	-1	0	0	2	0
0	-1	0	0	0	1

$(A^*A) \backslash A^*b =$

5.0000
15.0000
0.0000
10.0000
15.0000
20.0000

