

Computer Vision I: Variational Methods

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Departments of Informatics & Mathematics

Technical University of Munich

Exercises



Marvin Eisenberger



Emanuel Laude

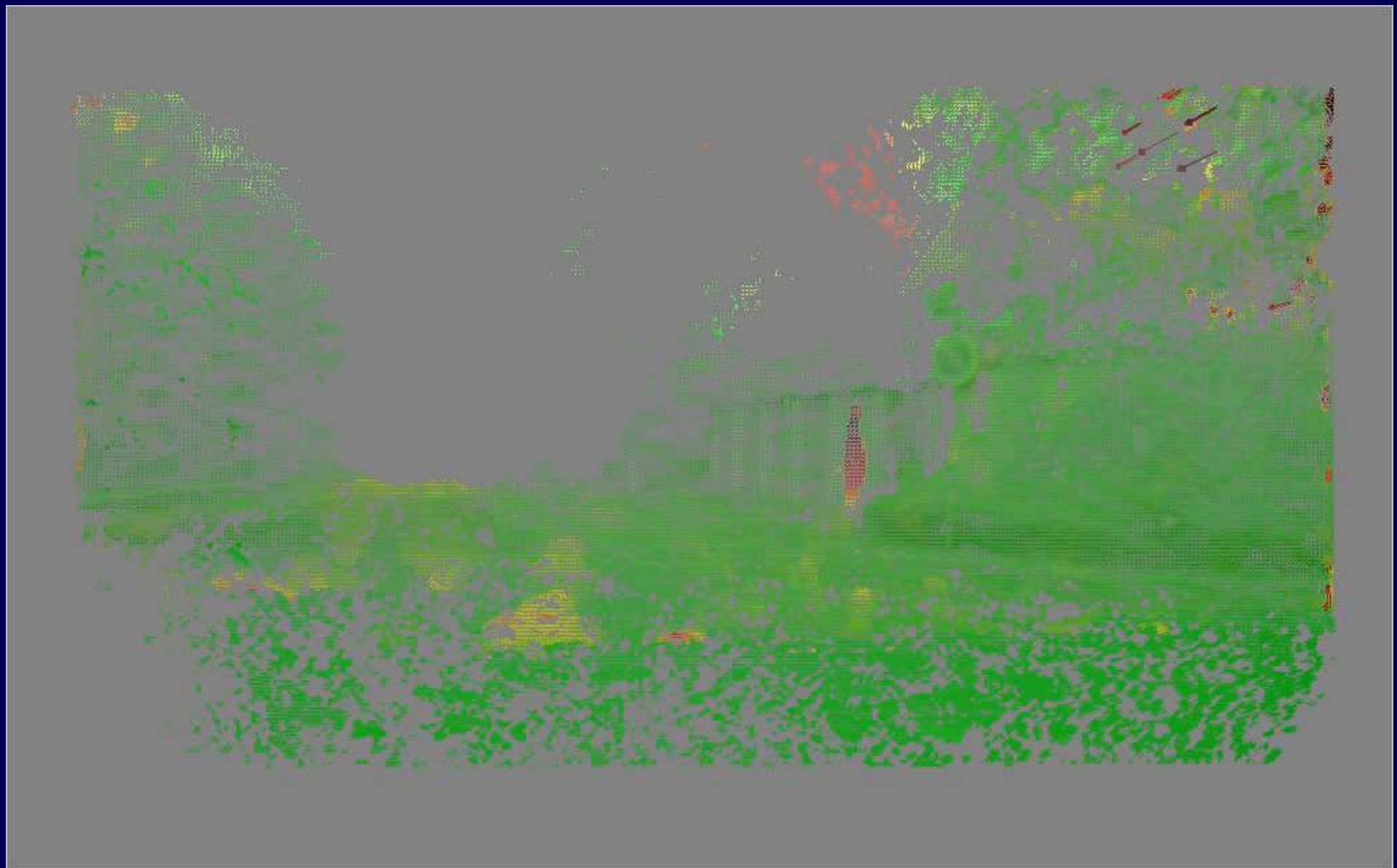


Original left image





Variational Scene Flow



Wedel et al. IJCV '11, Wedel & Cremers, Springer 2011



8 – 14 September 2018 | Munich, Germany

3300 delegates!

Technische
Universität
München

TUM



TUM Chair of Computer Vision & AI



Prof. Dr. Laura Leal-Taixé



Prof. Dr. Daniel Cremers



Sabine Wagner



Dr. habil. Rudolph Triebel



Dr. Tao Wu



Quirin Lohr



John Chiotellis



Patrick Wenzel



Nikolaus Demmel



Patrick Dendorfer



Marvin Eisenberger



Thomas Frerix



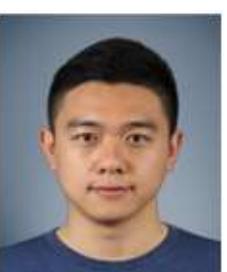
Vladimir Golkov



Björn Häfner



Caner Hazırbaş



Nan Yang



Qadeer Khan



Zorah Lähner



Emanuel Laude



Lingni Ma



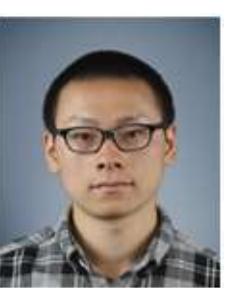
Robert Maier



Tim Meinhardt



Thomas Möllenhoff



Zhenzhang Ye



David Schubert



Yuesong Shen



Christiane Sommer



Lukas von Stumberg



Vladyslav Usenko



Matthias Vestner



Rui Wang



Qunjie Zhou



Spatially Dense 3D Reconstruction



infinite-dimensional optimization

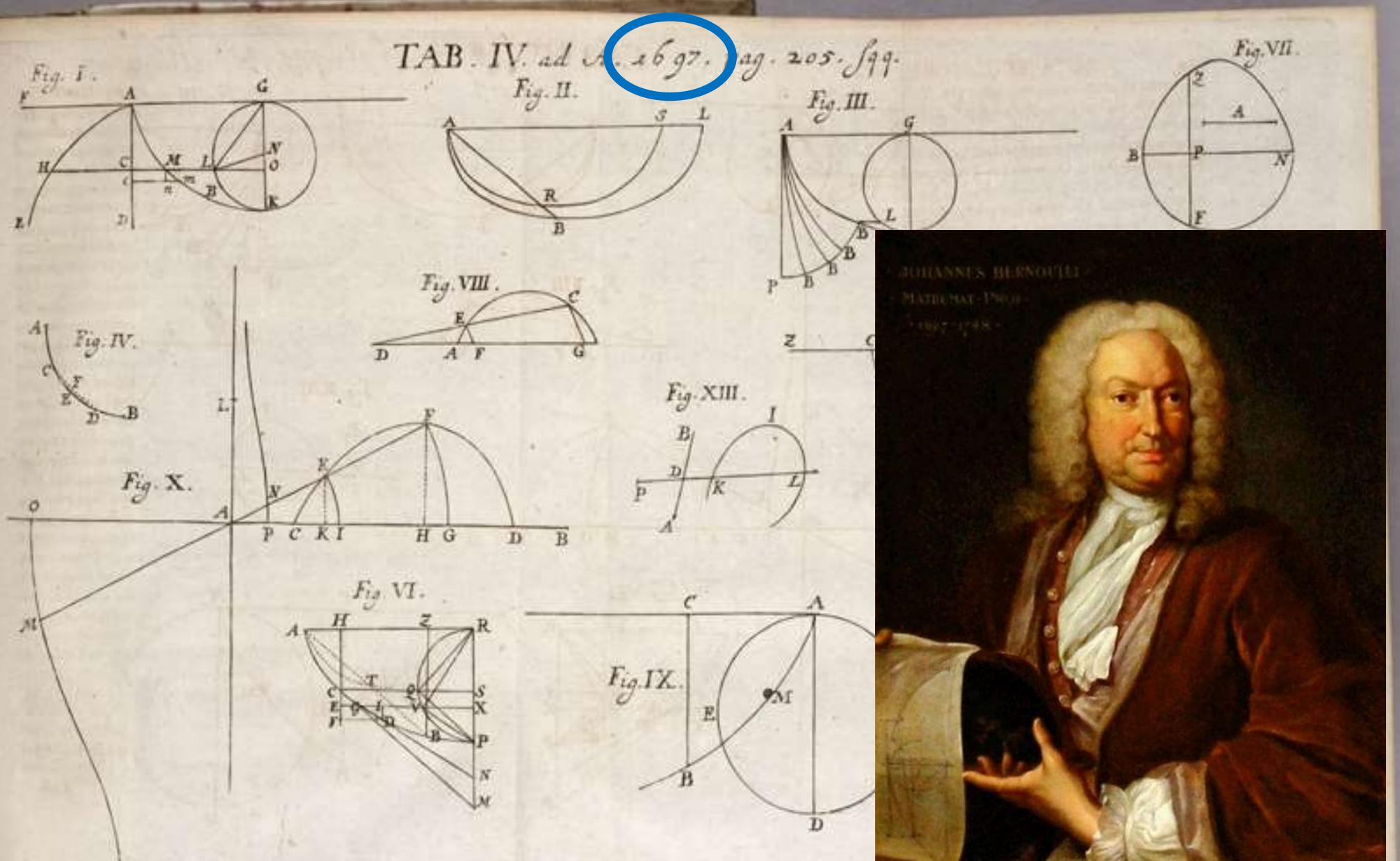


Which path is the fastest?





Bernoulli & The Brachistochrone



Johann Bernoulli (1667-1748)



Image segmentation:

*Geman, Geman '84, Blake, Zisserman '87, Kass et al. '88,
Mumford, Shah '89, Caselles et al. '95, Kichenassamy et al. '95,
Paragios, Deriche '99, Chan, Vese '01, Tsai et al. '01, ...*

Multiview stereo reconstruction:

Non-convex energies

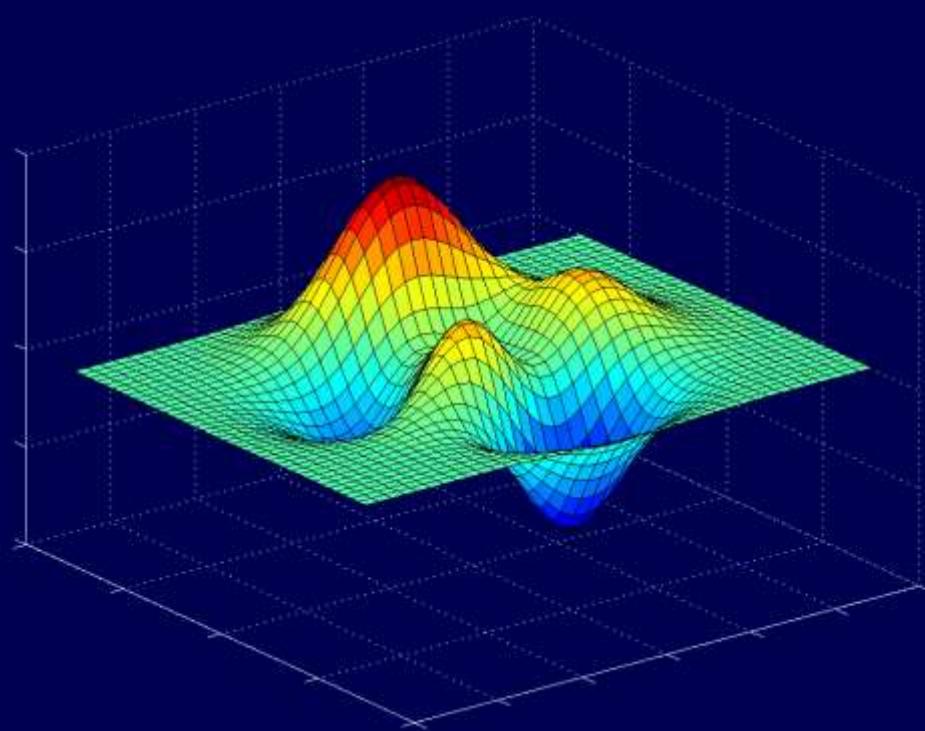
*Faugeras, Keriven '98, Duan et al. '04, Yezzi, Soatto '03,
Seitz et al. '06, Hernandez et al. '07, Labatut et al. '07, ...*

Optical flow estimation:

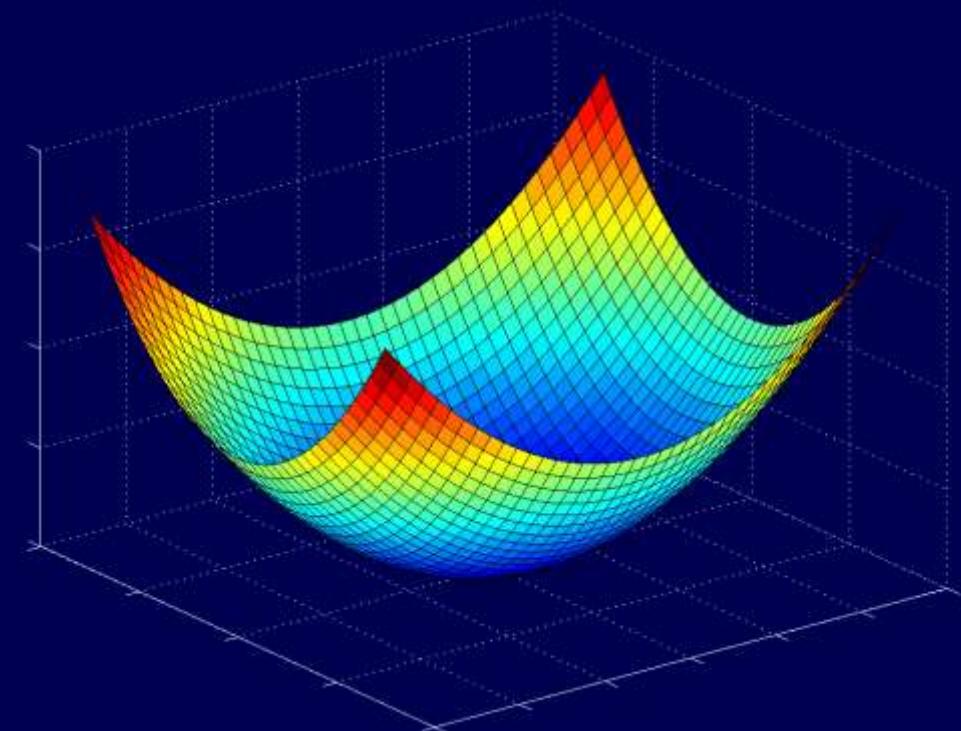
*Horn, Schunck '81, Nagel, Enkelmann '86, Black, Anandan '93,
Alvarez et al. '99, Brox et al. '04, Baker et al. '07, Zach et al. '07,
Sun et al. '08, Wedel et al. '09, ...*



Optimization and Convexity



Non-convex energy



Convex energy



Inverse Problems: Denoising



image $f : \Omega \rightarrow \mathbb{R}^3$, $\Omega \subset \mathbb{R}^2$

denoised image $u_{den} : \Omega \rightarrow \mathbb{R}^3$

$$u_{den} = \arg \min_u \int_{\Omega} (u - f)^2 dx + \lambda \int_{\Omega} |\nabla u| dx$$

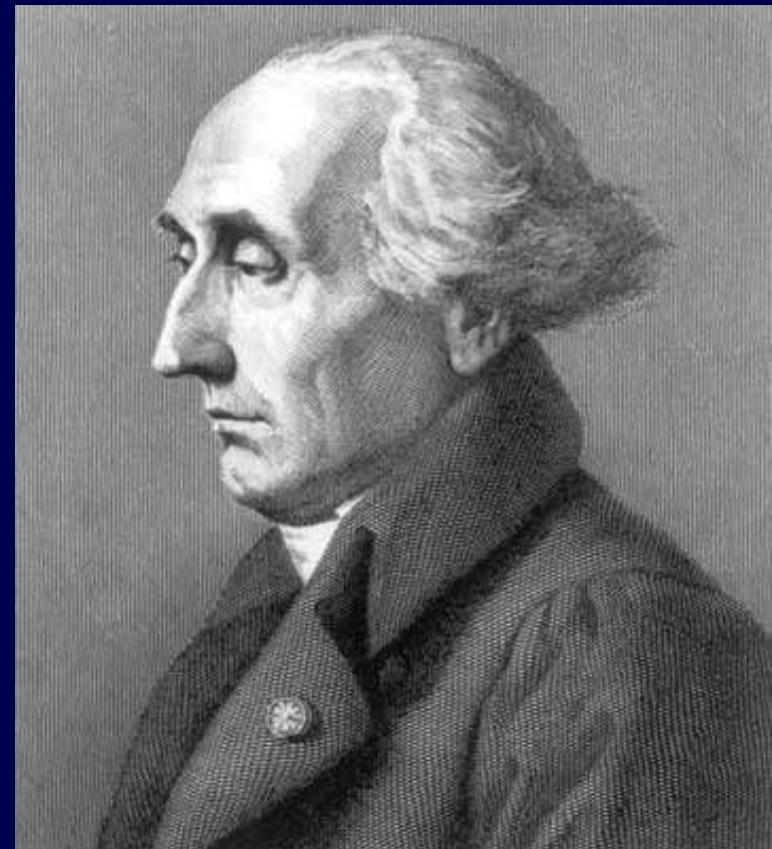
Rudin, Osher, Fatemi 1992, Goldlücke, Strekalovskiy, Cremers 2012



Variational Methods



Leonhard Euler
(1703–1783)



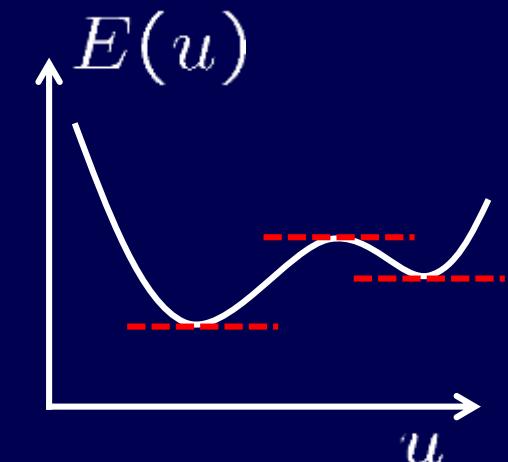
Joseph-Louis Lagrange
(1736 – 1813)



$$E(u) = \int_{\Omega} (u - f)^2 + \lambda |\nabla u| dx = \int_{\Omega} \mathcal{L}(u, u') dx$$

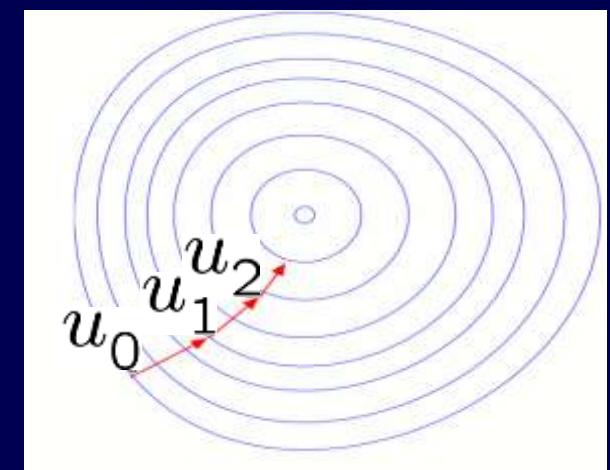
Euler-Lagrange equation as necessary condition:

$$\frac{dE}{du} = \frac{\partial \mathcal{L}}{\partial u} - \frac{d}{dx} \left(\frac{\partial \mathcal{L}}{\partial u'} \right) = 0$$



Gradient descent:

$$\frac{\partial u}{\partial t} = - \frac{dE}{du}$$





Overview



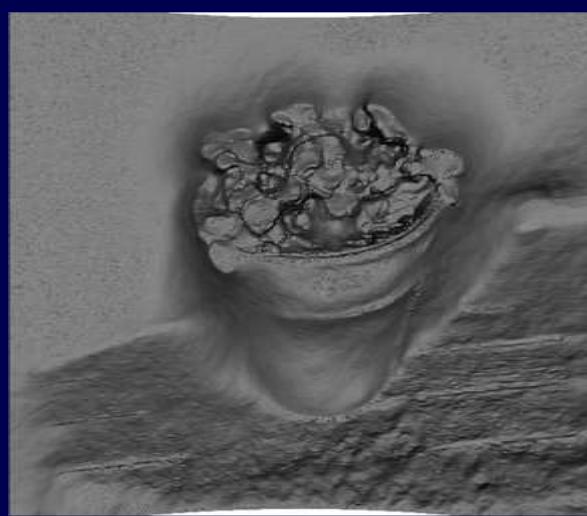
Multiview reconstruction



Super-res.textures



Stereo reconstruction



Realtime dense geometry



RGB-D cameras



Reconstruction on the fly



Overview



Multiview reconstruction



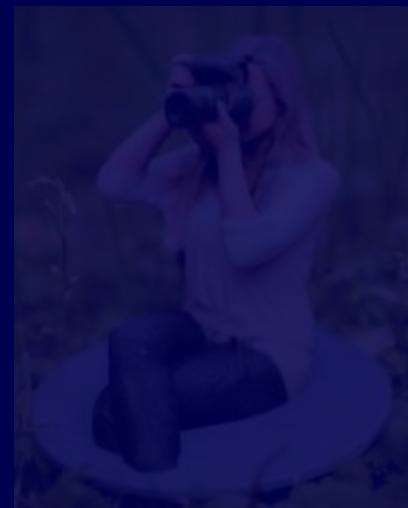
Super-res.textures



Stereo reconstruction



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RGB-D cameras



Reconstruction on the fly



3D Reconstruction from Multiple Views

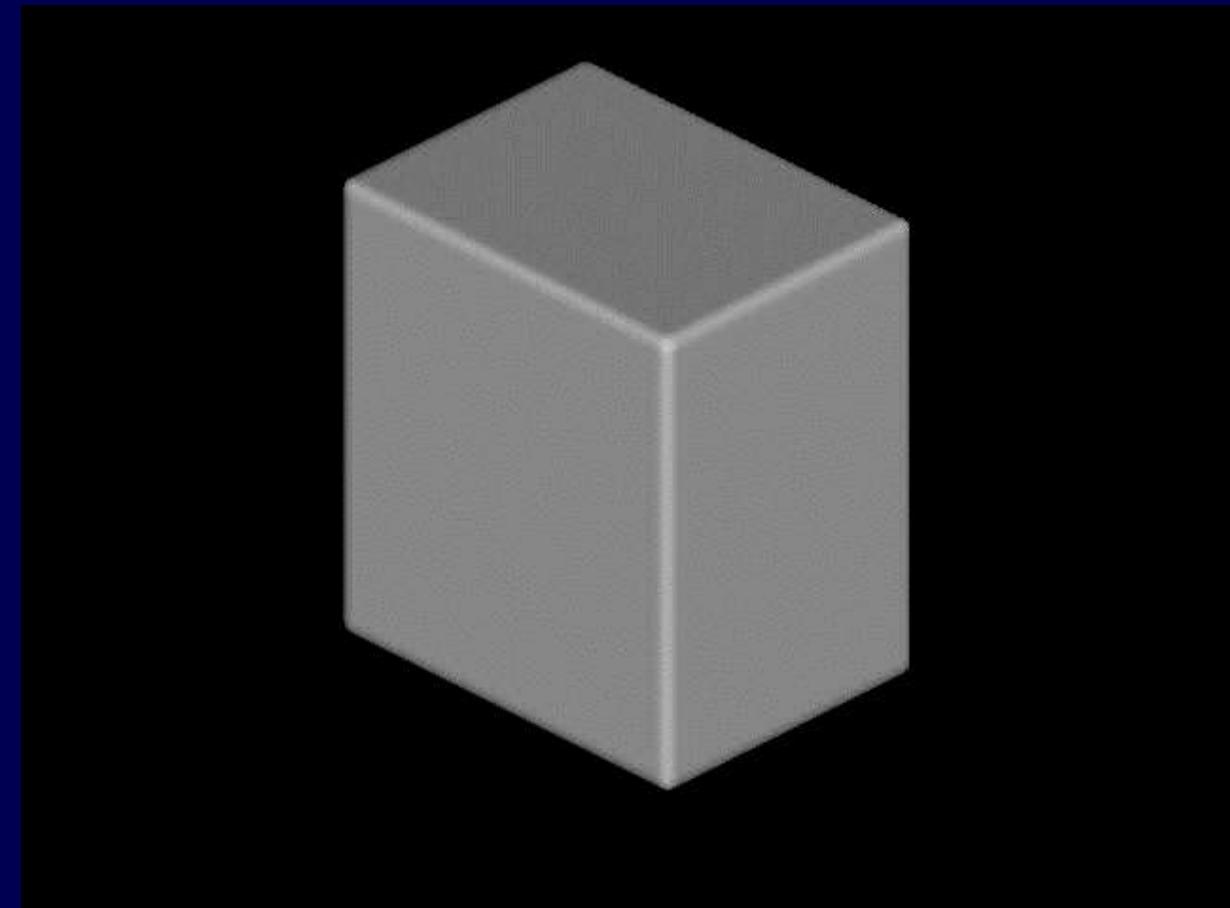


Kolev, Klodt, Brox, Cremers, Int. J. of Computer Vision '09:

Theorem: Globally optimal surfaces can be computed by convex optimization.



Evolution to Global Optimum



Kolev, Klodt, Brox, Cremers, Int. J. of Computer Vision '09:

Theorem: Globally optimal surfaces can be computed by convex optimization.



Reconstruction of Fine-scale Structures



Image data courtesy of Yasutaka Furukawa.



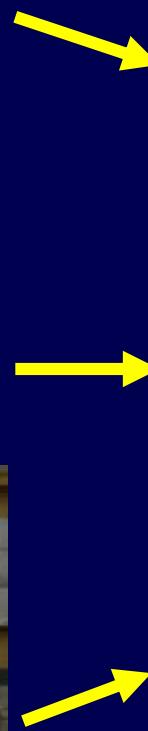
Reconstructing the Niobids Statues



Kolev, Cremers, ECCV '08, PAMI '11



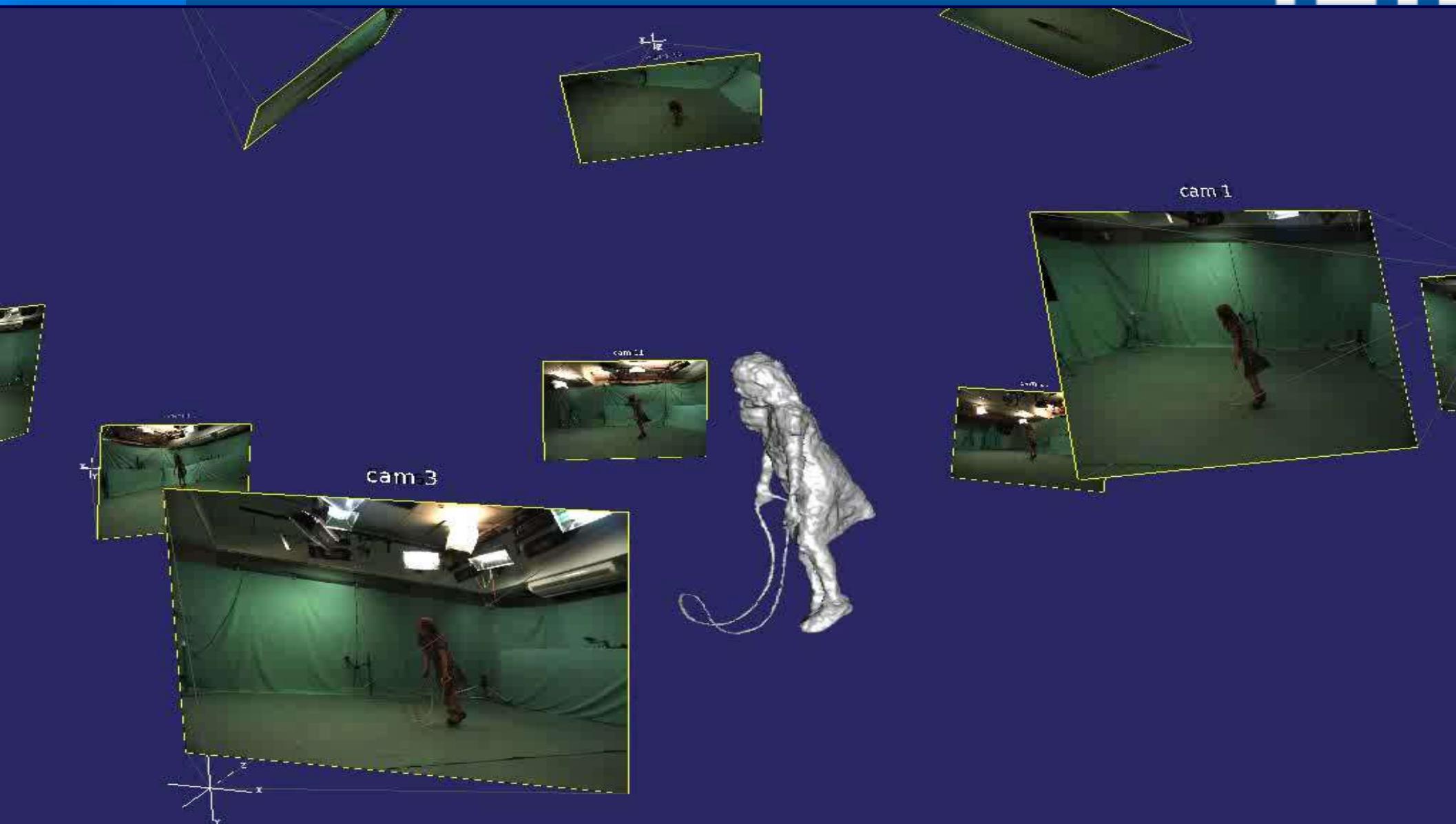
Multiview Reconstruction



Kolev, Cremers, ECCV '08, PAMI '11



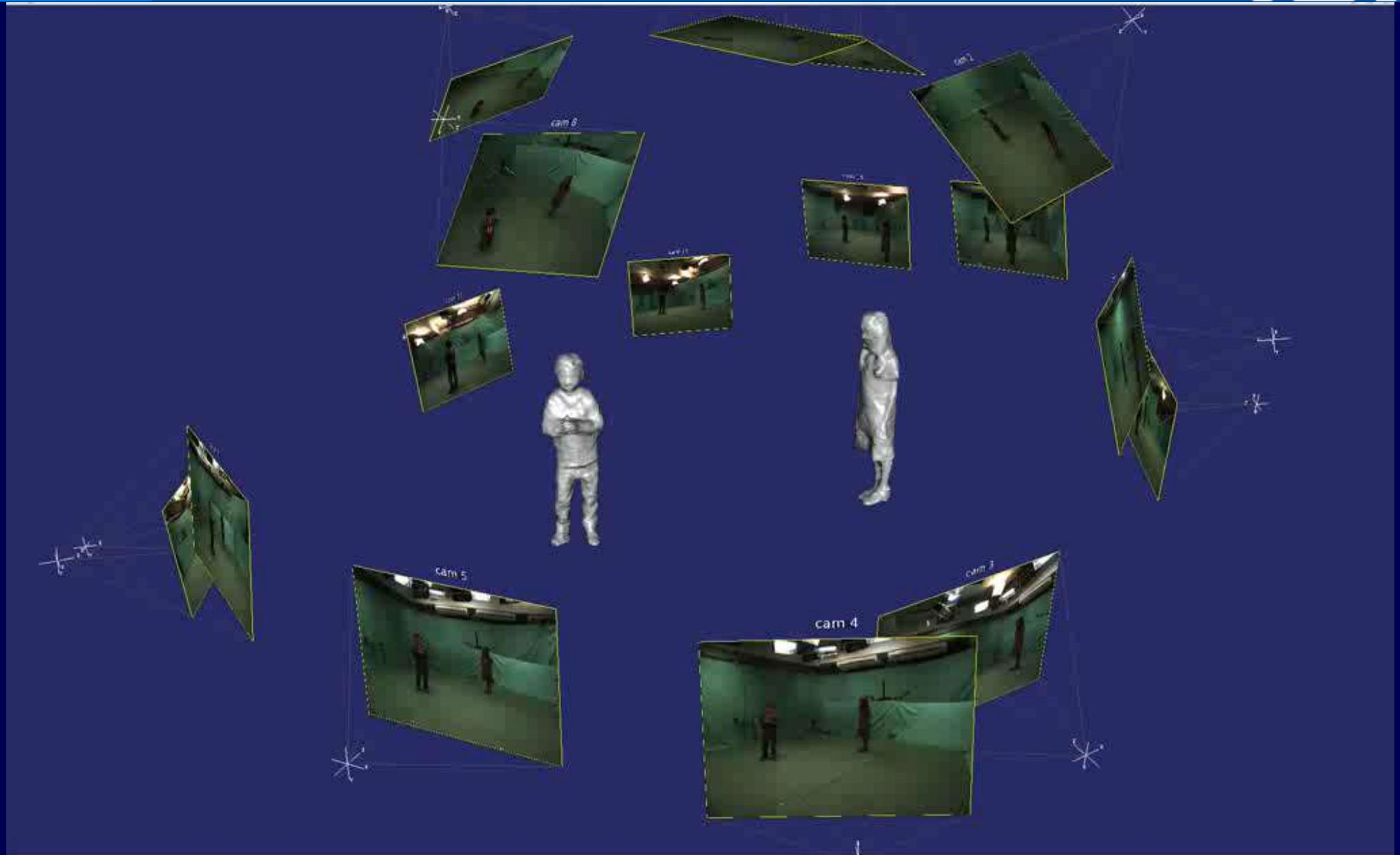
Reconstructing Dynamic Scenes



Oswald, Stühmer, Cremers, ECCV '14



Action Reconstruction



Oswald, Stühmer, Cremers, ECCV '14



Single View Reconstruction



Can we recover geometry from a single image?

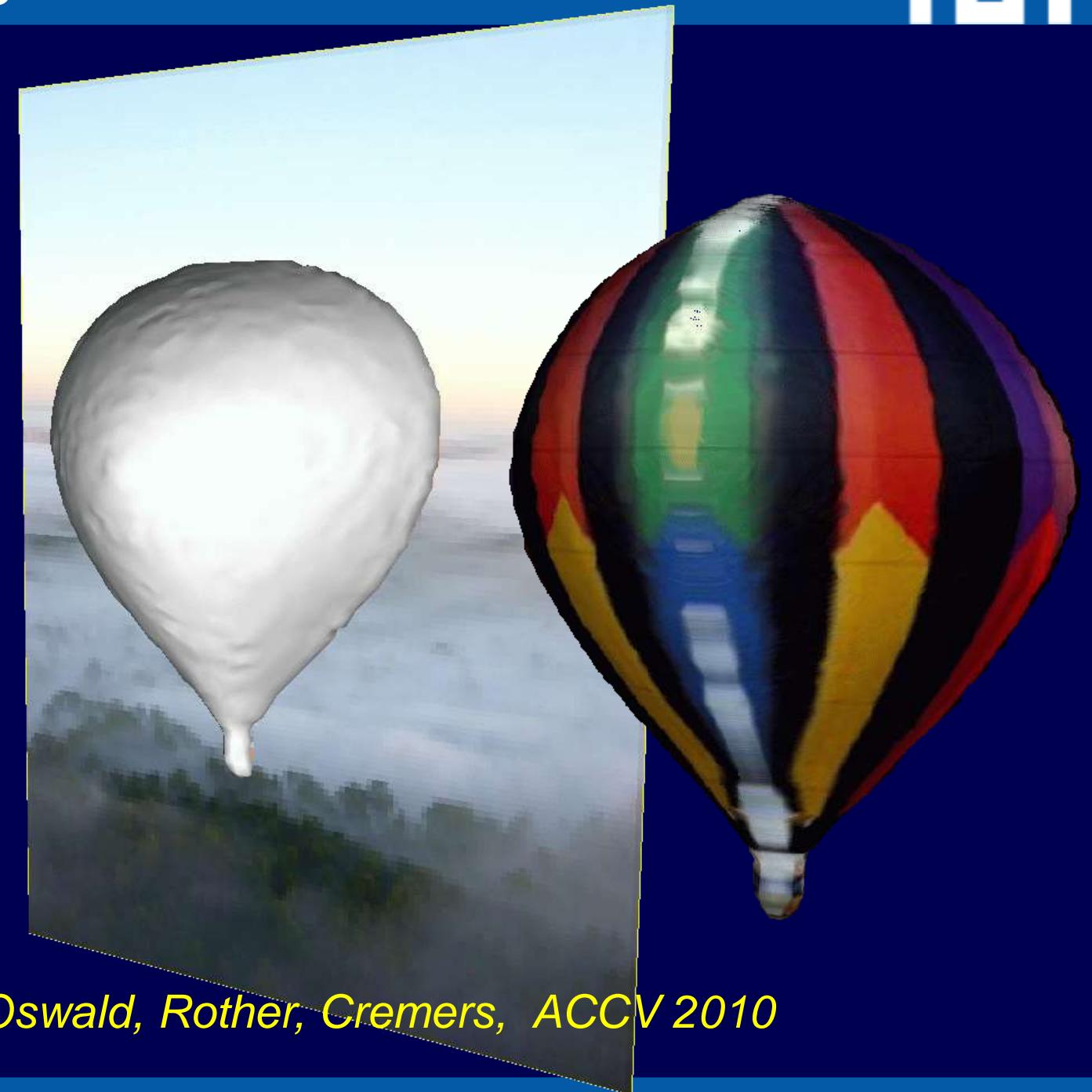
Yes: Shape-from-shading, shape-from-focus, shape from symmetry,...

Solution: Fixed-volume silhouette-consistent minimal surface.

$$\min_S |S| \quad \text{s.t. } \text{Vol}(S) = V_0, \pi(S) = S_0$$



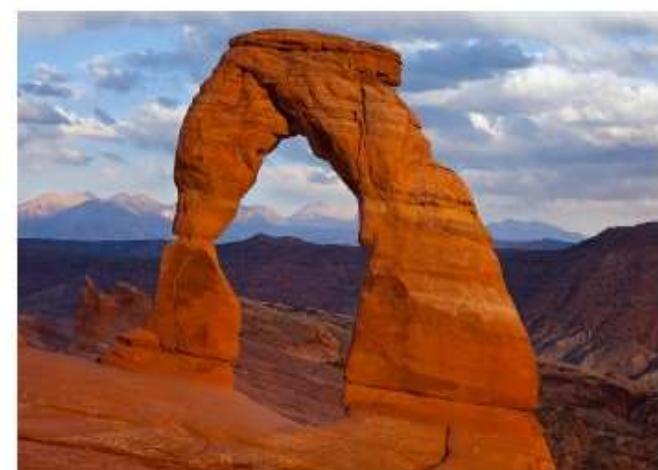
Single View Reconstruction



Toeppe, Oswald, Rother, Cremers, ACCV 2010



Single View Reconstruction



Input



Reconstruction



+30% volume



+40% volume

Reconstruction computed in fractions of a second on GPU

Toeppe et al. ACCV 2010, Oswald et al. CVPR 2012



Single View Reconstruction



Toeppe, Oswald, Rother, Cremers, ACCV 2010



Modifying the Material Properties





Single View Reconstruction



Toeppe, Oswald, Rother, Cremers, ACCV 2010



Single View Reconstruction



*Toeppe, Oswald, Rother, Cremers, ACCV 2010**

** Best Paper Honorable Mention*

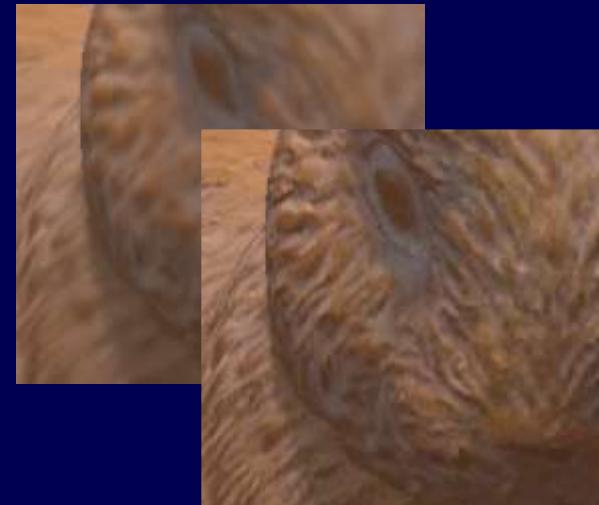
In collaboration with Microsoft Research



Overview



Multiview reconstruction



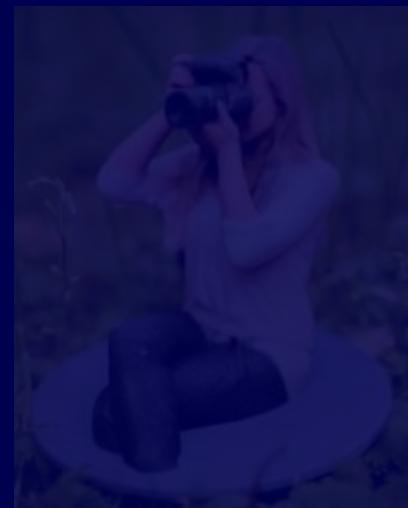
Super-res.textures



Stereo reconstruction



Realtime dense geometry



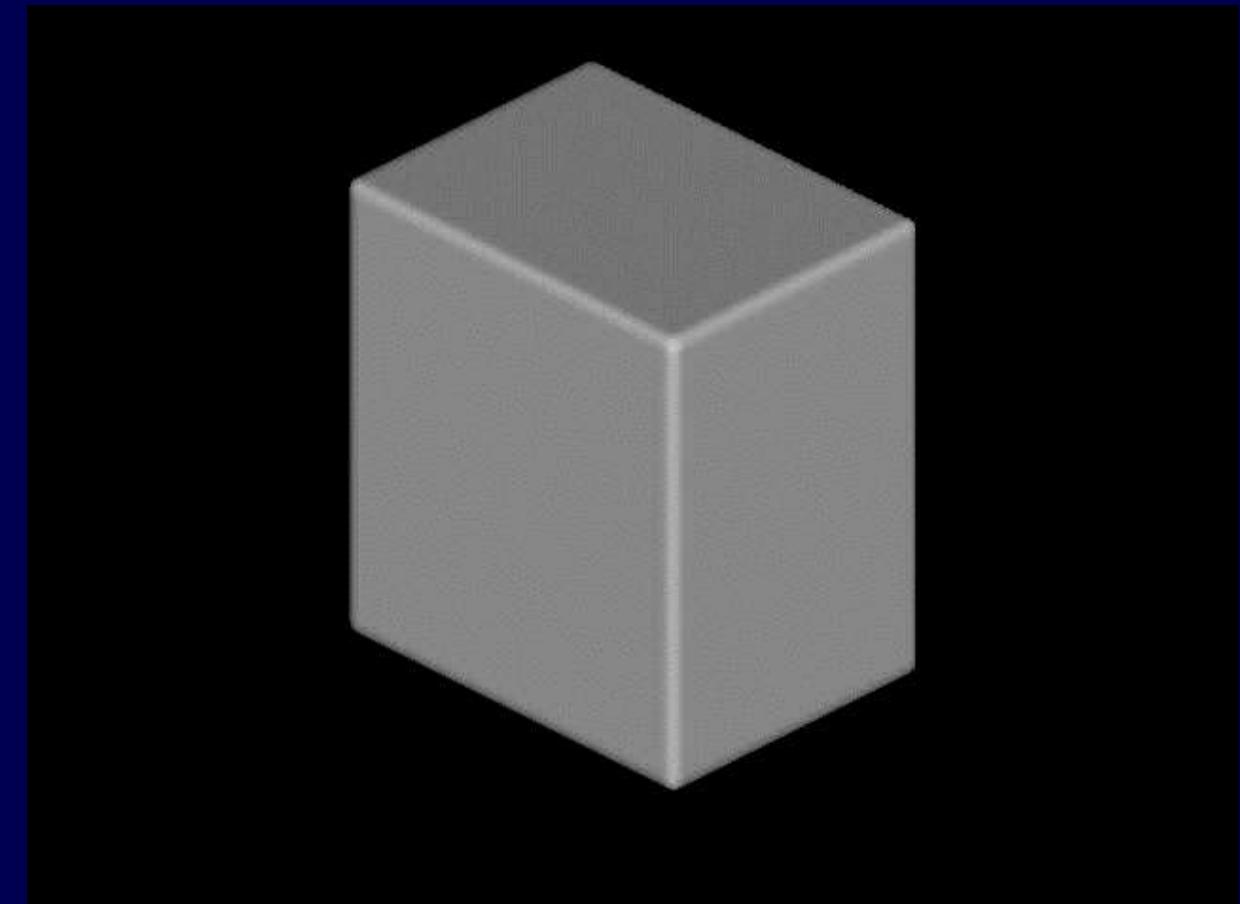
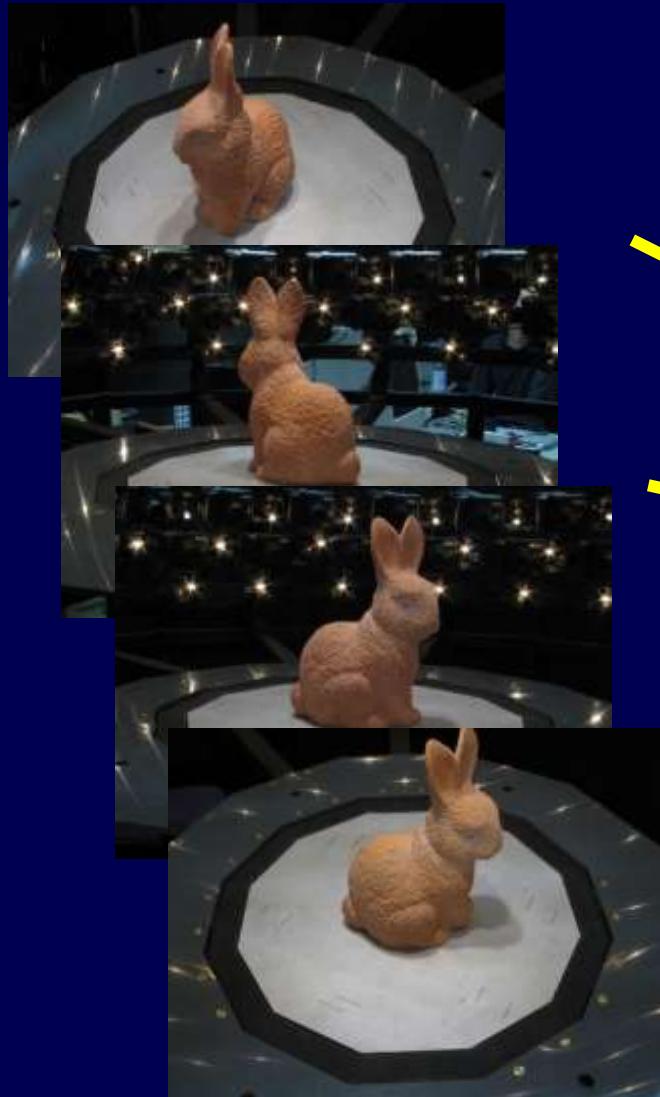
RGB-D cameras



Reconstruction on the fly



Evolution to Global Optimum



Kolev, Klodt, Brox, Cremers, IJCV 2009



Super-Resolution Texture Map



Given all images $\mathcal{I}_i : \Omega_i \rightarrow \mathbb{R}^3$, determine the surface color $T : S \rightarrow \mathbb{R}^3$

$$\min_T \sum_{i=1}^n \int_{\Omega_i} \left(b * (T \circ \beta_i) - \mathcal{I}_i \right)^2 dx + \lambda \int_S \|\nabla_S T\| ds$$

blur & downsample back-projection



* Best Paper
Award

Goldlücke, Cremers, ICCV '09, DAGM '09*, IJCV '13



Super-Resolution Texture Map



* Best Paper
Award

Goldlücke, Cremers, ICCV '09, DAGM '09*, IJCV '13



Super-Resolution Texture Map



Weighted average



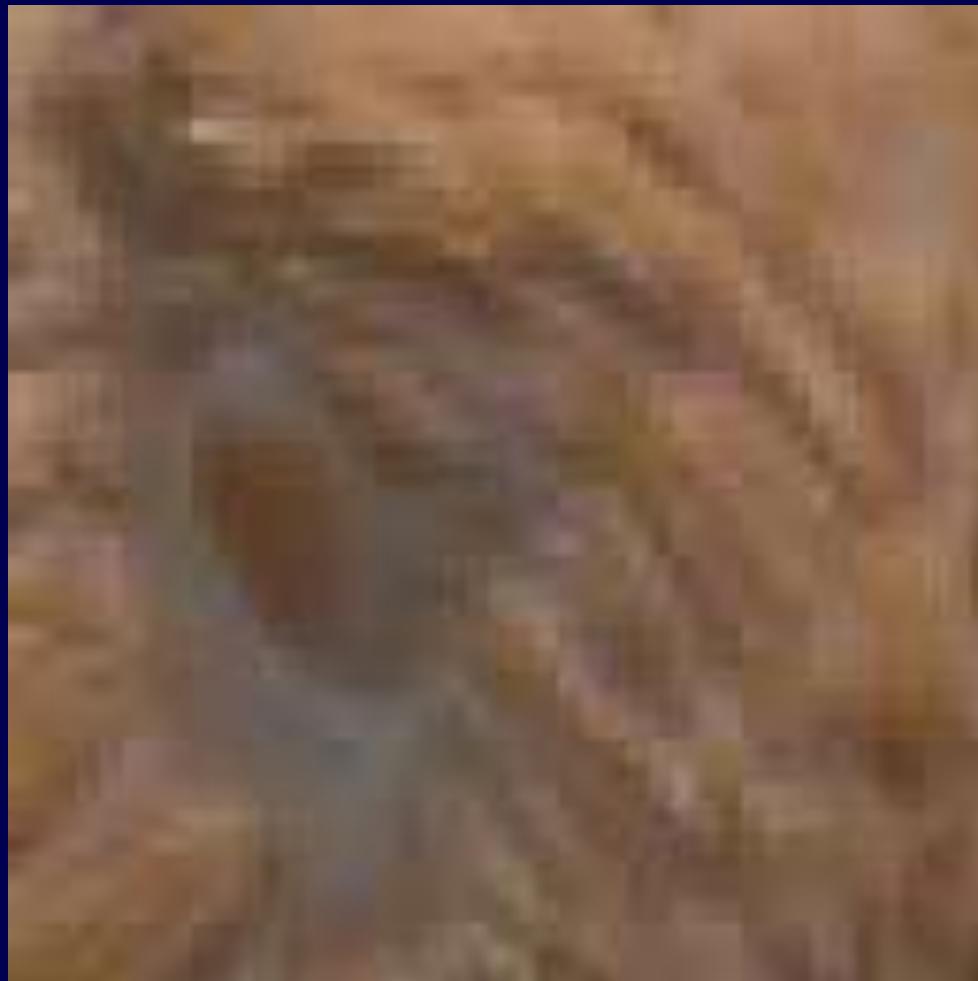
Super-resolution texture

Goldlücke, Cremers, ICCV '09, DAGM '09, IJCV '13*

* Best Paper
Award



Super-Resolution Texture Map



Closeup of input image



Super-resolution texture

Goldlücke, Cremers, ICCV '09, DAGM '09, IJCV '13*

* Best Paper
Award



Overview



Multiview reconstruction



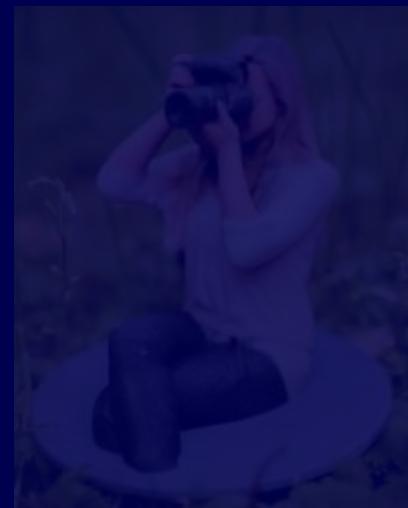
Super-res.textures



Stereo reconstruction



Realtime dense geometry



RGB-D cameras



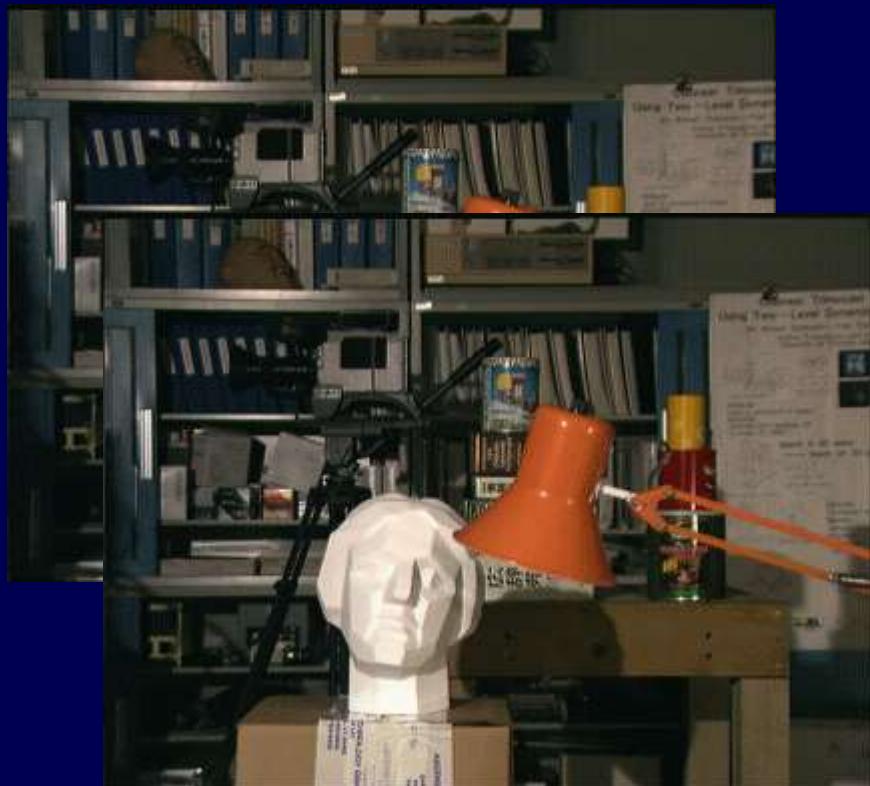
Reconstruction on the fly



From Binary to Multilabel Optimization



$$u : \Omega \rightarrow \Gamma = [\gamma_{min}, \gamma_{max}]$$



Example: Stereo Reconstruction

Pock, Schoenemann, Bischof, Cremers, Europ. Conf. on Computer Vision '08:

Theorem: Stereo reconstruction can be solved by convex optimization.



Evolution to Global Minimum

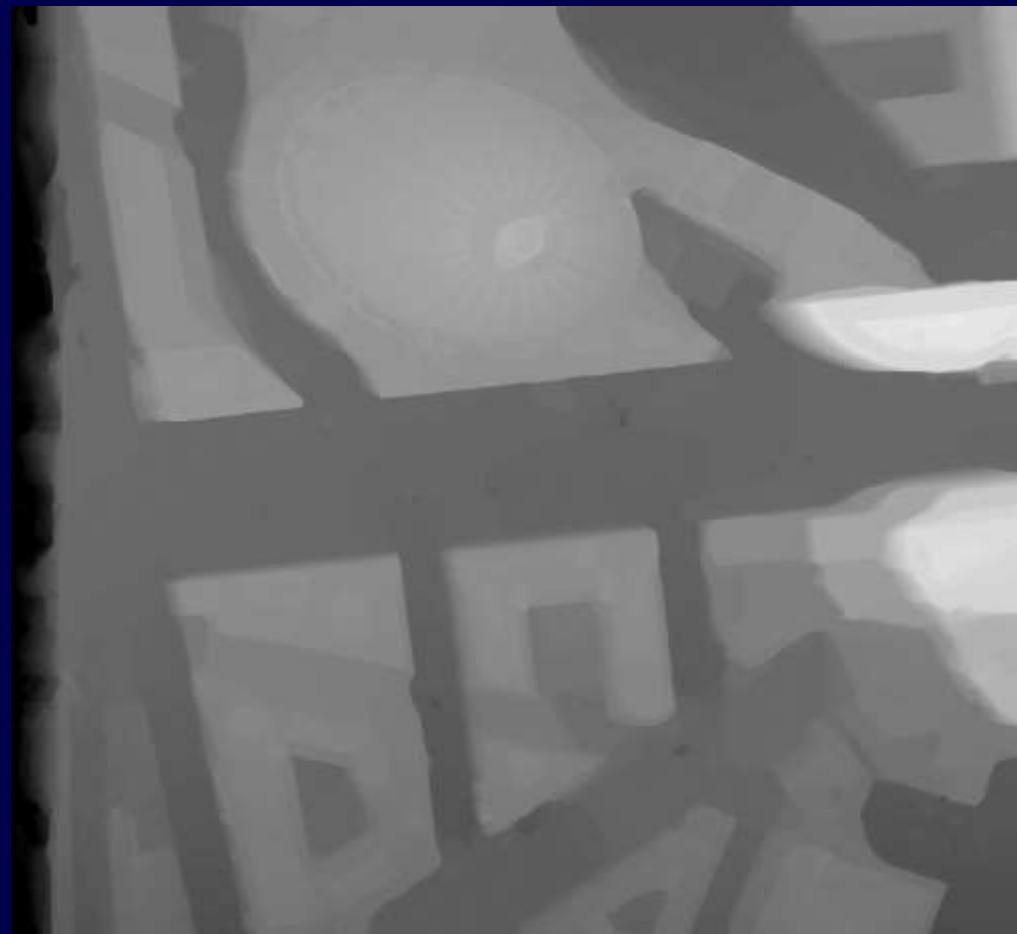




Reconstruction from Aerial Images



1/2 input images (6 Mpixel)
Courtesy of H. Hirschmüller



Depth reconstruction
77 seconds

Stangl, Souiai, Cremers, GCPR '13



Reconstruction from Aerial Images



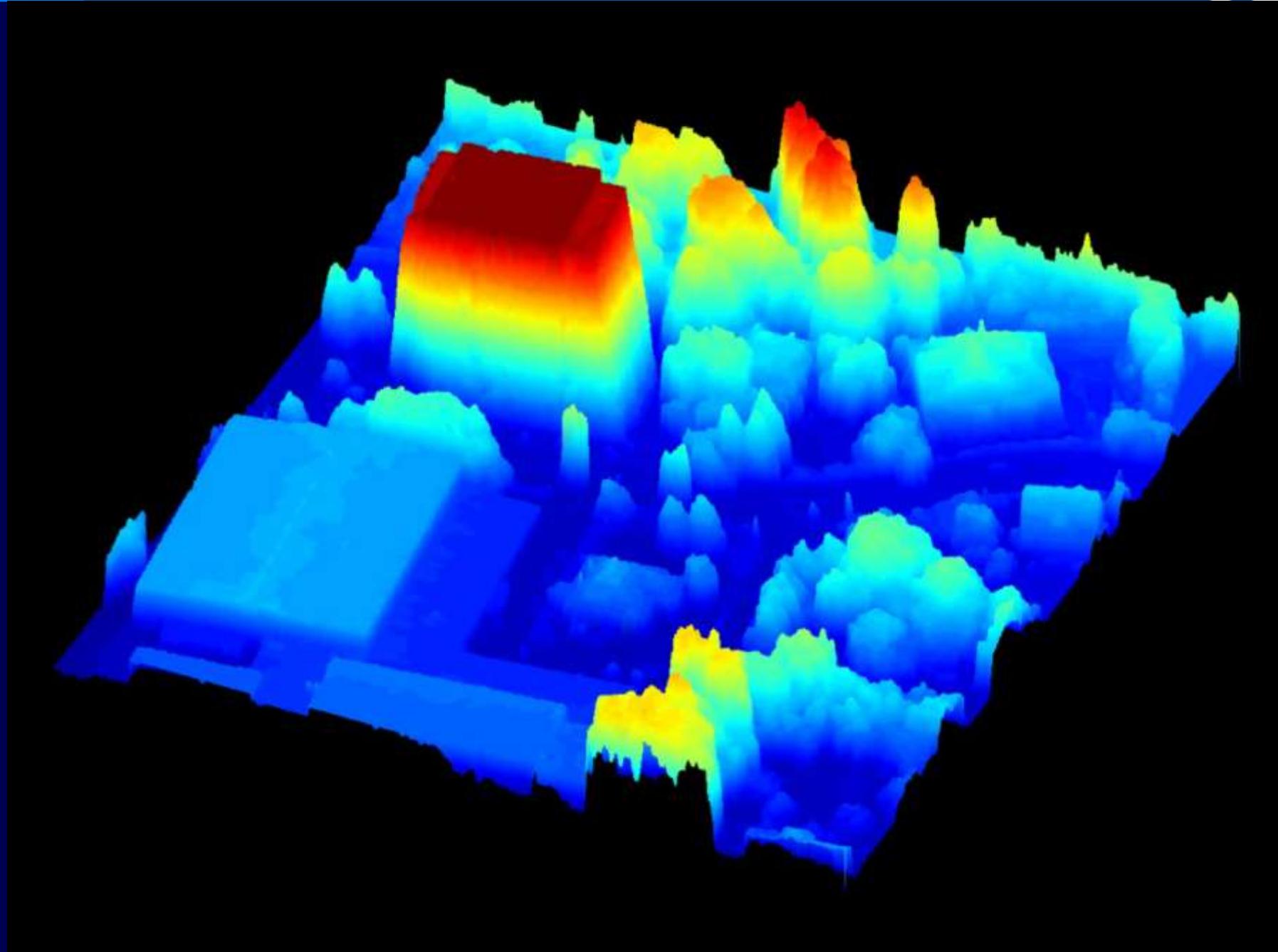
One of two input images
Courtesy of Microsoft



Depth reconstruction



Reconstruction from Aerial Images

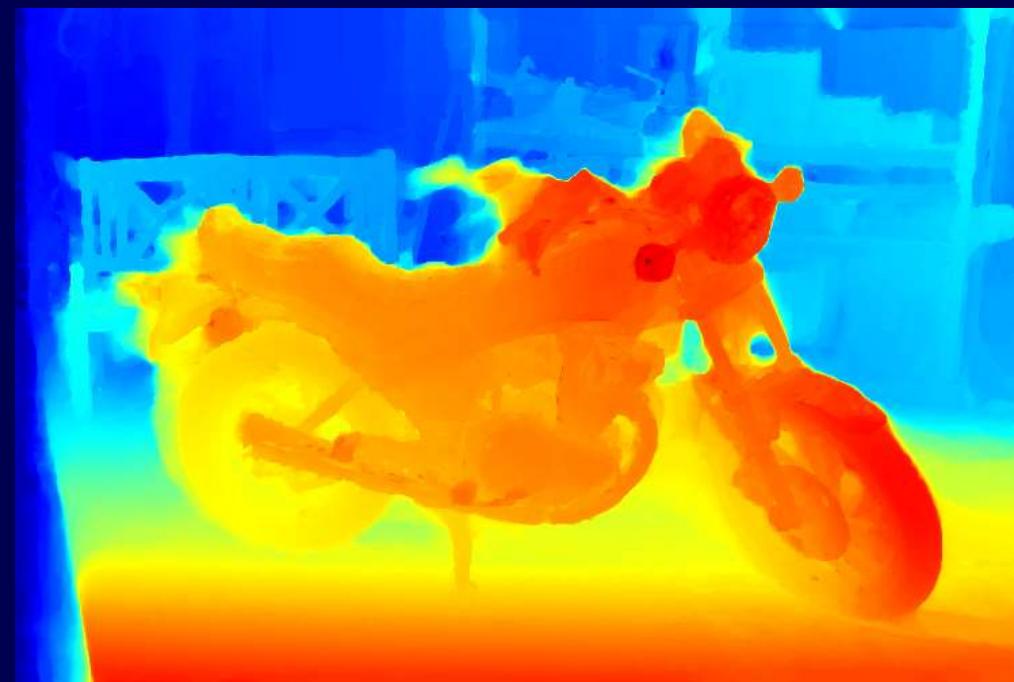




Highly accurate Stereo Reconstruction



Stereo input



Depth reconstruction

*Möllenhoff, Laude, Möller, Lellmann, Cremers, CVPR '16 **

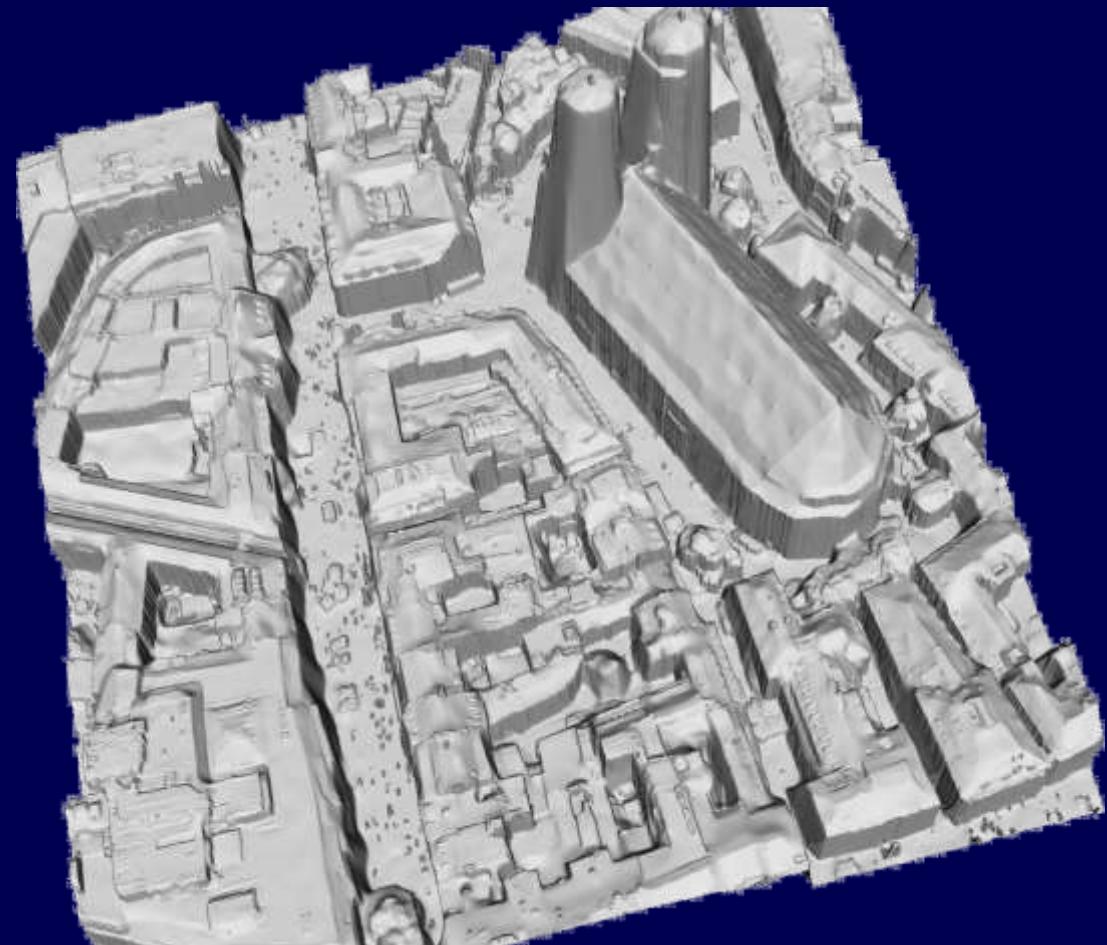
* Best Paper Honorable Mention



Munich from the Air



1/2 input images (1000x1000)



Depth reconstruction

Kuschk, Cremers, ICCV Big Data Workshop 2013



Overview



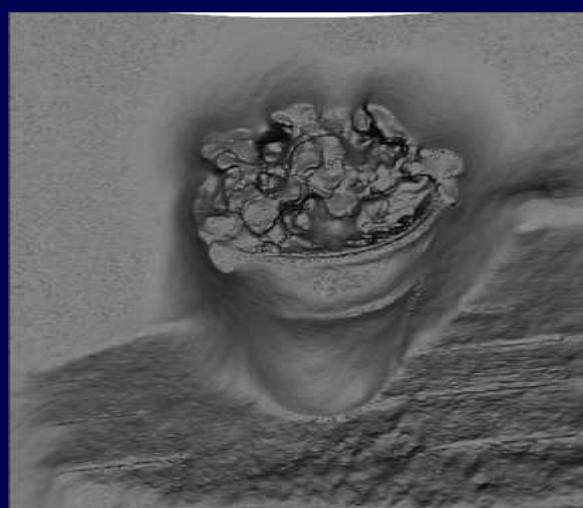
Multiview reconstruction



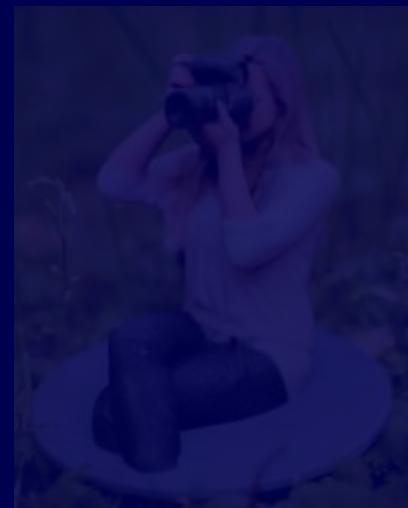
Super-res.textures



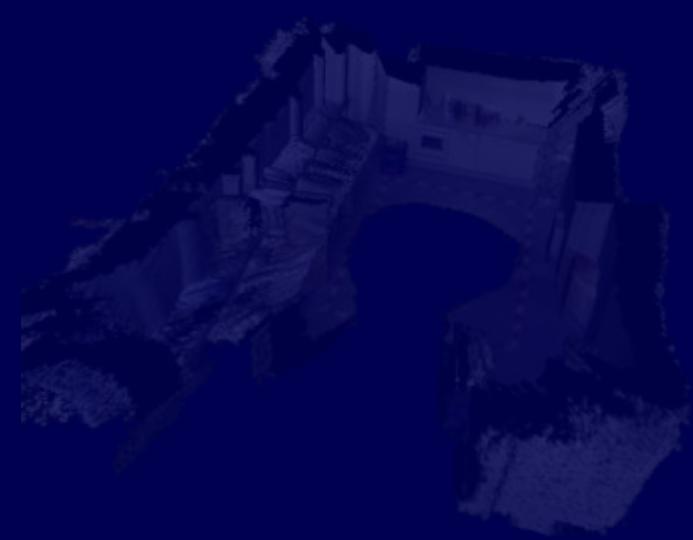
Stereo reconstruction



Realtime dense geometry



RGB-D cameras



Reconstruction on the fly



From Dense Flow to Dense Geometry



Input video



Optical flow field

$$\min_{u: \Omega \rightarrow \mathbb{R}^2} \int_{\Omega} |I_1(x) - I_2(x + u)| dx + J(u)$$

Horn & Schunck '81, Zach et al. DAGM '07, Wedel et al. ICCV '09



From Dense Flow to Dense Geometry



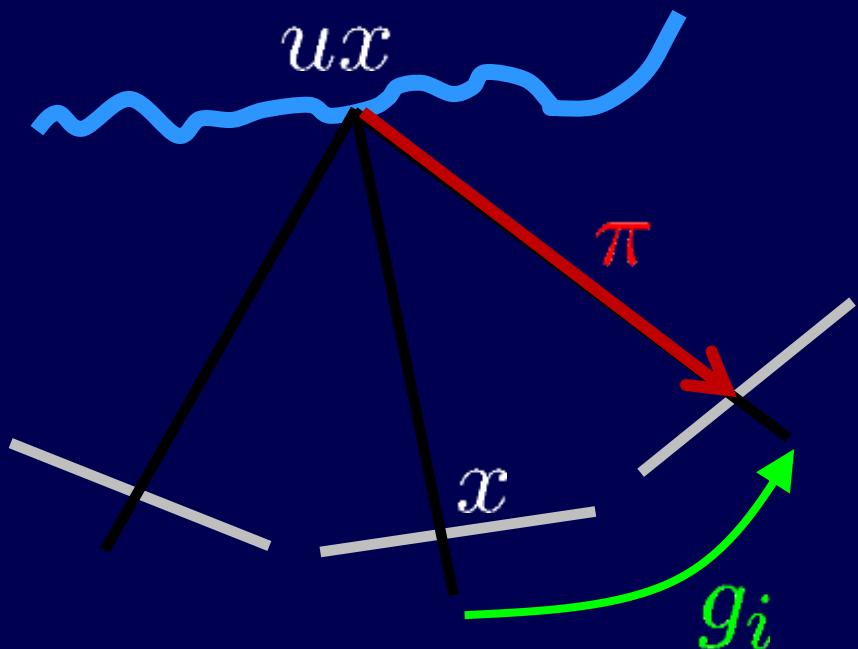
Input video

Optical flow field*

* 60 fps @ 640x480

$$\min_{u: \Omega \rightarrow \mathbb{R}^2} \int_{\Omega} |I_1(x) - I_2(x + u)| dx + J(u)$$

Horn & Schunck '81, Zach et al. DAGM '07, Wedel et al. ICCV '09



Brightness constancy:

$$I_0(x) \stackrel{!}{=} I_i(\pi(g_i(ux)))$$

$$\min_u \sum_i \int_{\Omega} |I_0(x) - I_i(\pi(g_i(u \cdot x)))| dx + \int_{\Omega} |\nabla u(x)| dx$$

Stuehmer, Gumhold, Cremers, DAGM '10



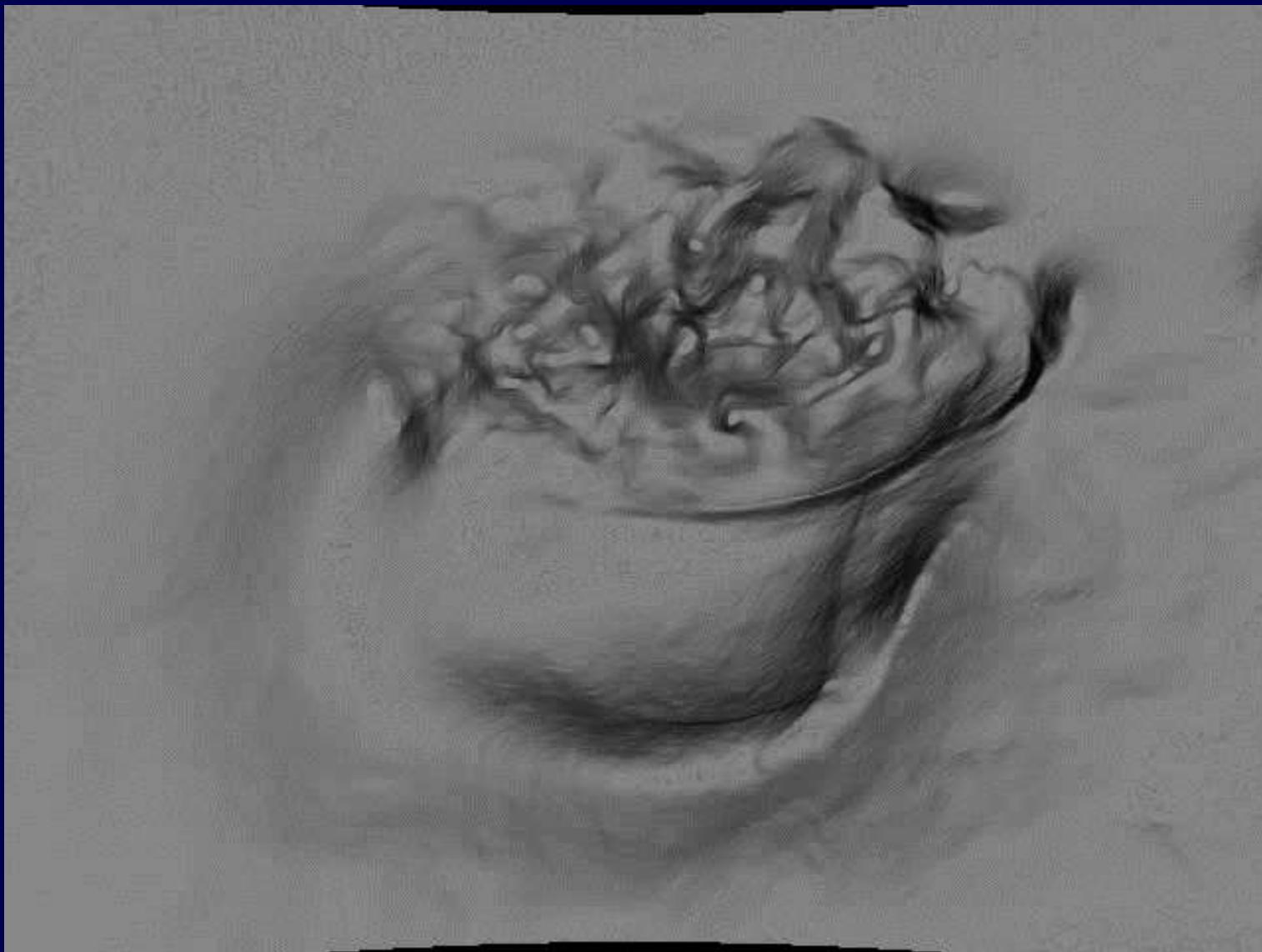
Dense geometry from hand-held camera



Stuehmer, Gumhold, Cremers, DAGM '10



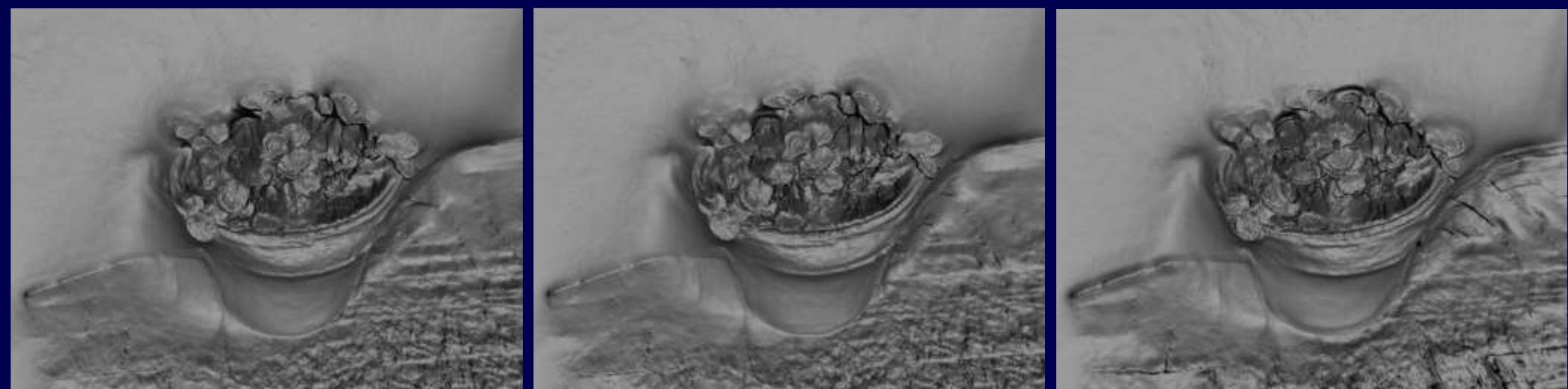
Dense geometry from hand-held camera



Stuehmer, Gumhold, Cremers, DAGM '10



Realtime Dense Reconstruction



16.0 fps

22.0 fps

41.1 fps

Stuehmer, Gumhold, Cremers, DAGM '10



Overview



Multiview reconstruction



Super-res.textures



Stereo reconstruction



Realtime dense geometry



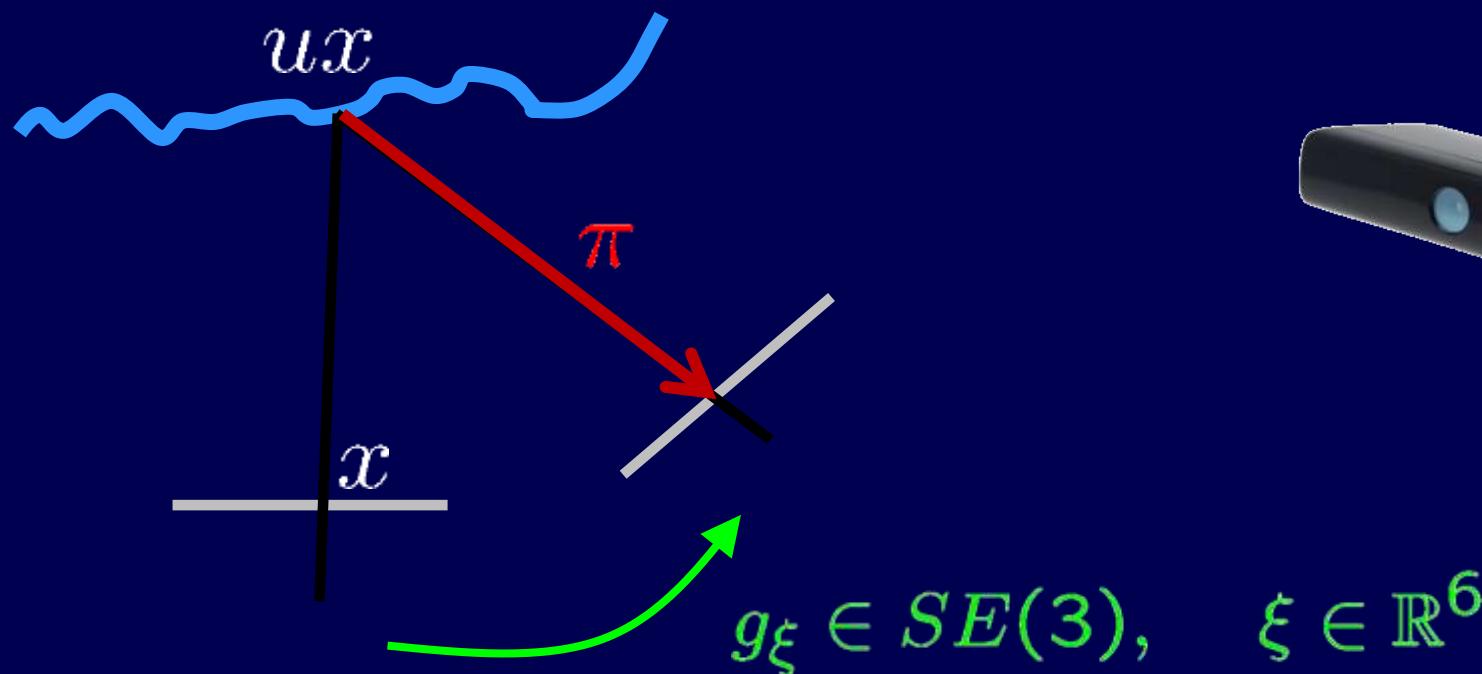
RGB-D cameras



Reconstruction on the fly



RGB-D Camera Tracking



Optimize dense photo-consistency:

$$\min_{\xi \in \mathbb{R}^6} \int_{\Omega} |I_0(x) - I_i(\pi(g_\xi(u \cdot x))| dx$$

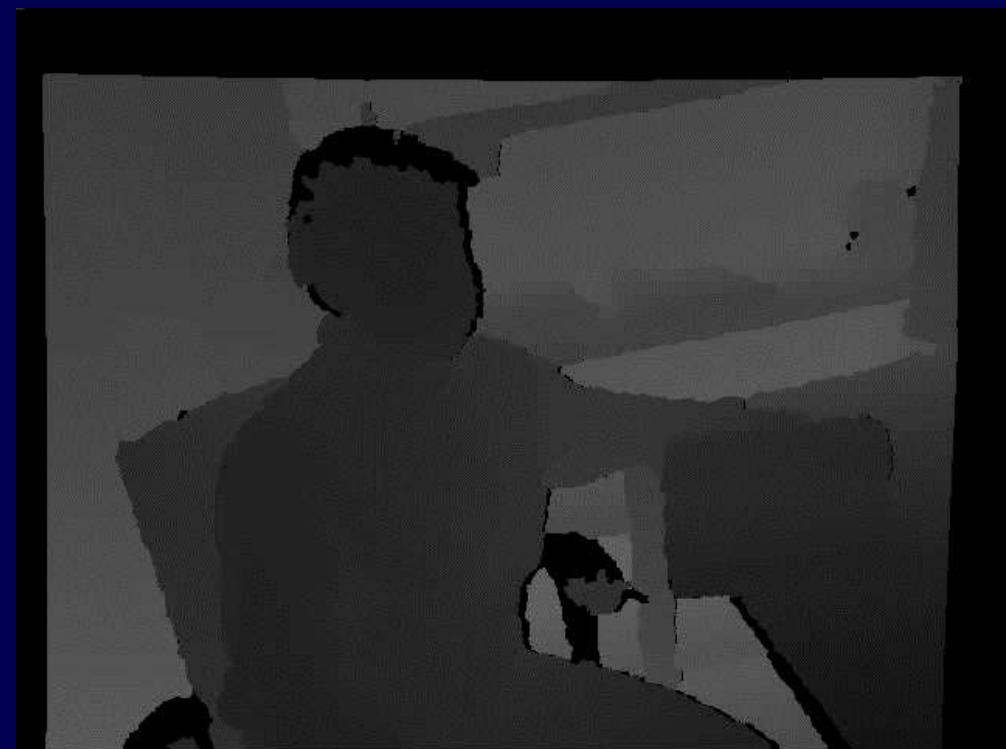
Steinbruecker et al. ICCV '11, Kerl et al., ICRA '13



Realtime 3D Modeling



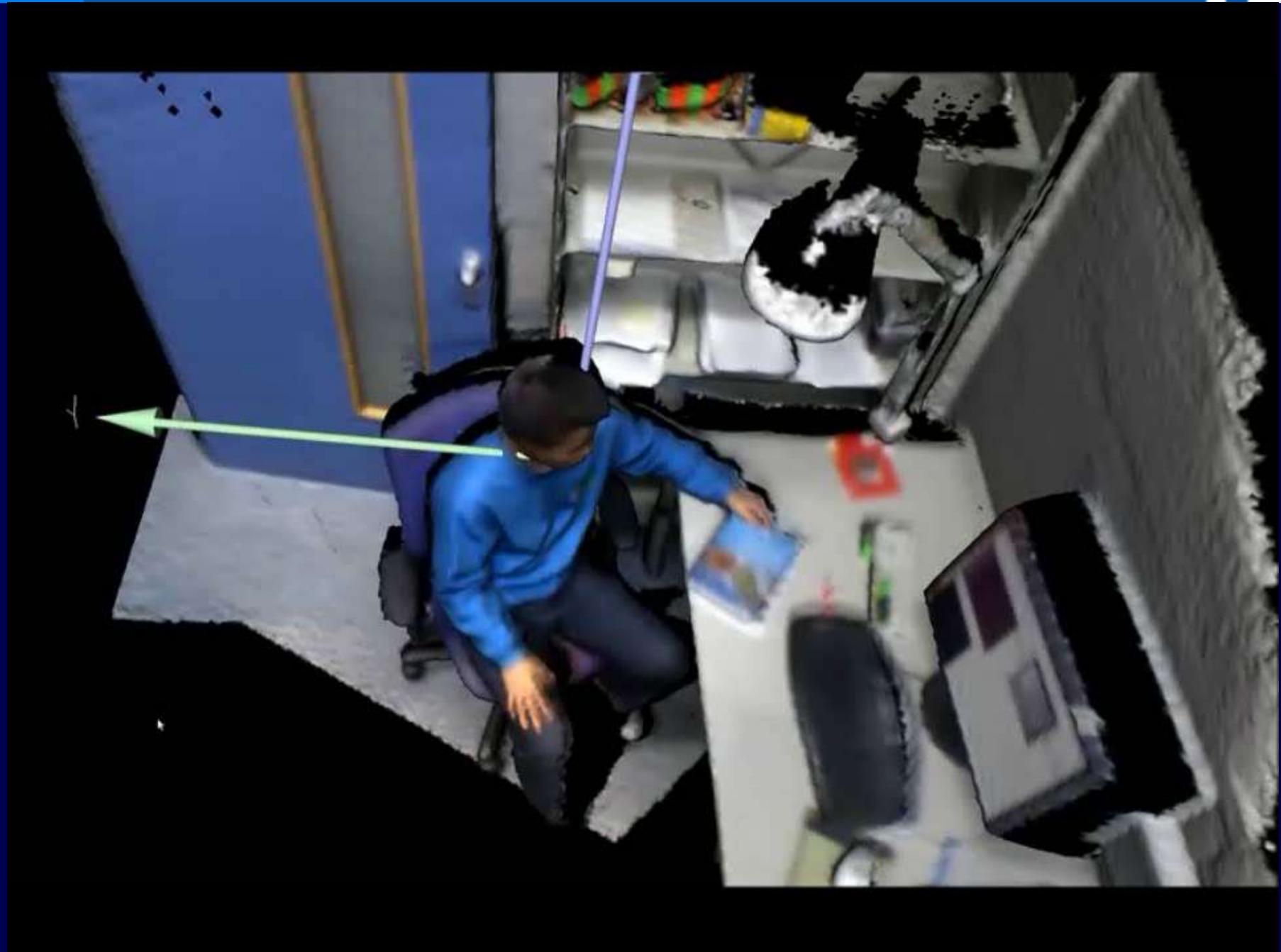
Color input



Depth input

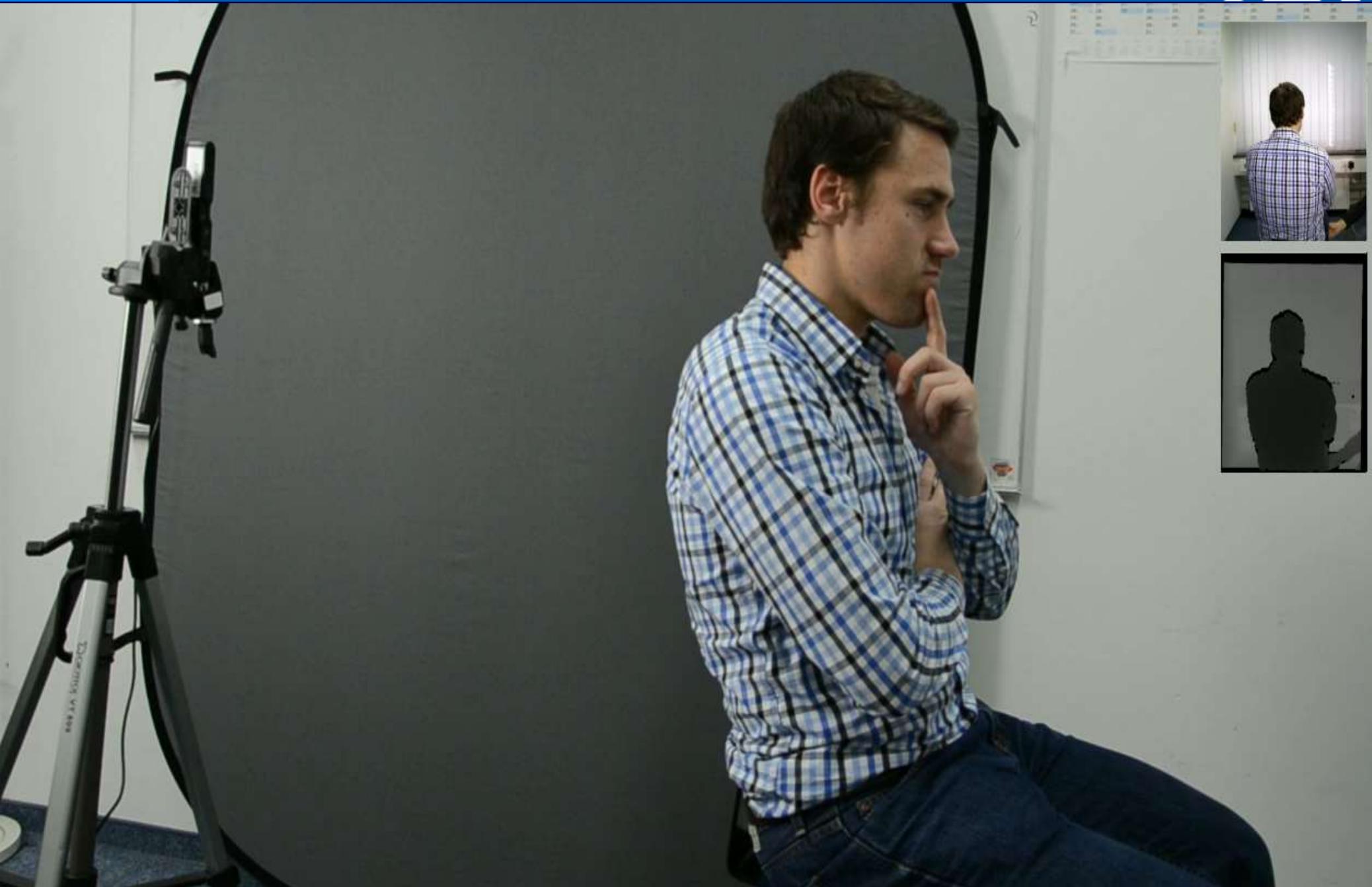


Realtime 3D Modeling





Realtime 3D Modeling





Realtime 3D Modeling





Realtime 3D Modeling





Realtime 3D Modeling





Realtime 3D Modeling





Realtime 3D Modeling





Full-Body Scanner





Overview



Multiview reconstruction



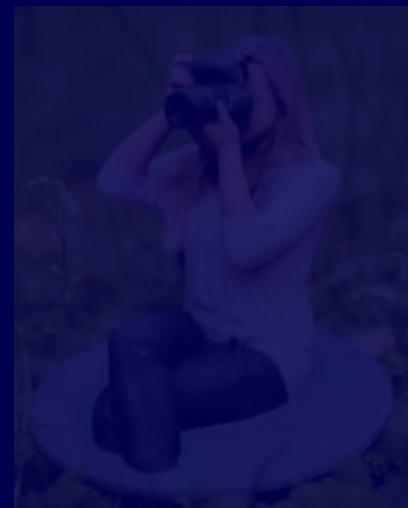
Super-res.textures



Stereo reconstruction



Realtime dense geometry



RGB-D cameras



Reconstruction on the fly



Reconstruction on the Fly



Bylow, Sturm, Kerl, Kahl, Cremers RSS '13



Large Scale: Loop Closure

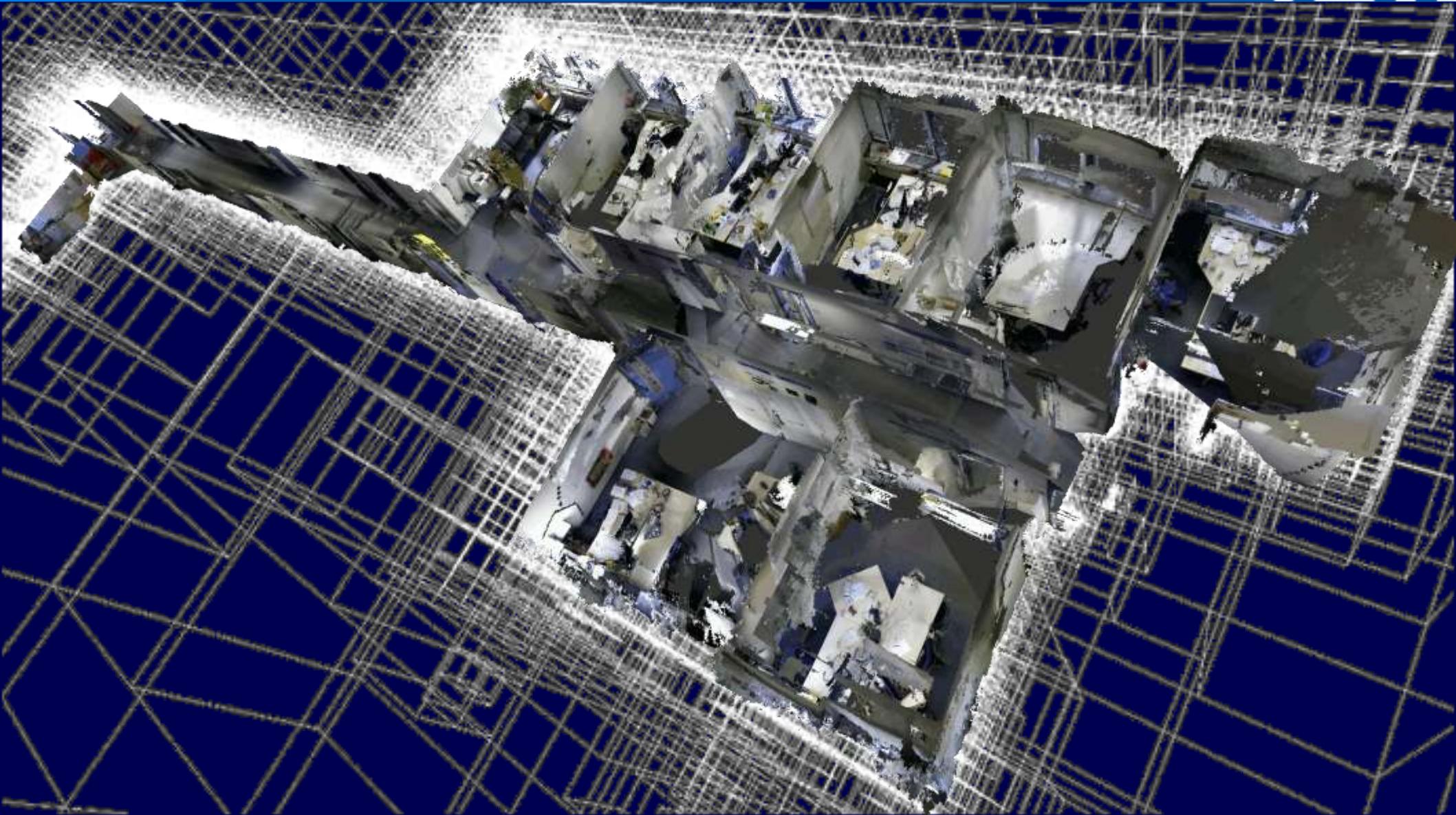


RGB-D dataset 'fr3/office'

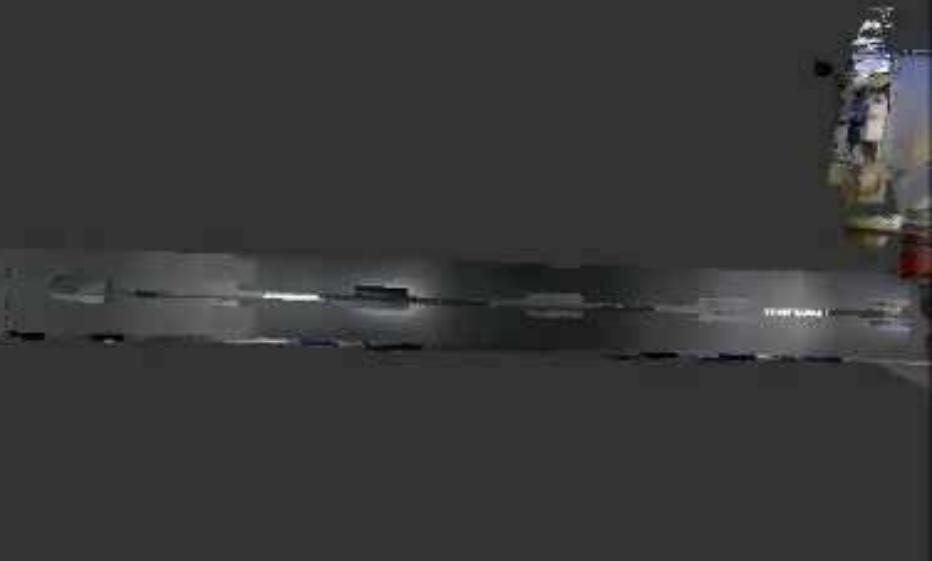
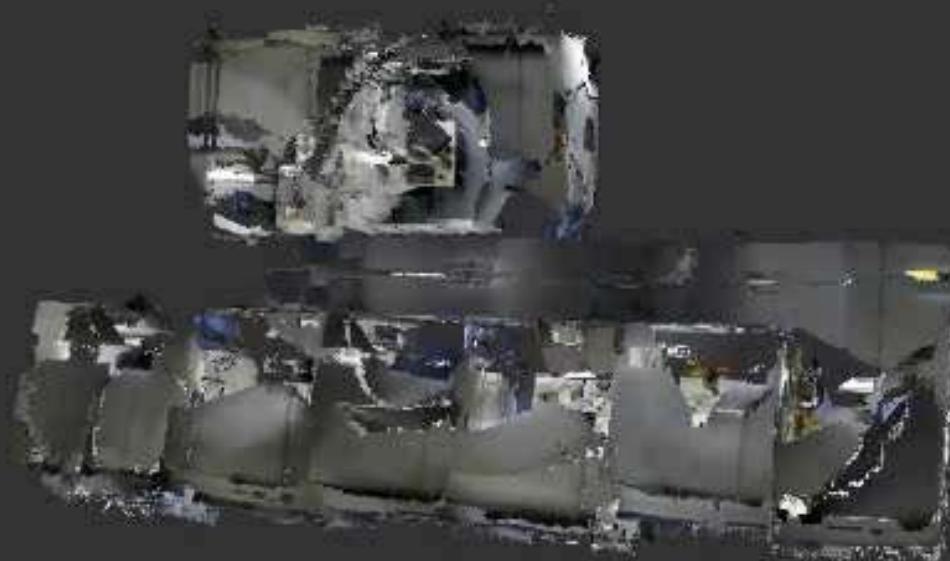
Kerl, Sturm, Cremers ICRA '13



Large Scale: Octrees



Steinbrücker, Kerl, Sturm, Cremers ICCV '13



Large-Scale Reconstruction

Steinbrücker, Kerl, Sturm, Cremers ICCV '13, ICRA '14



Summary



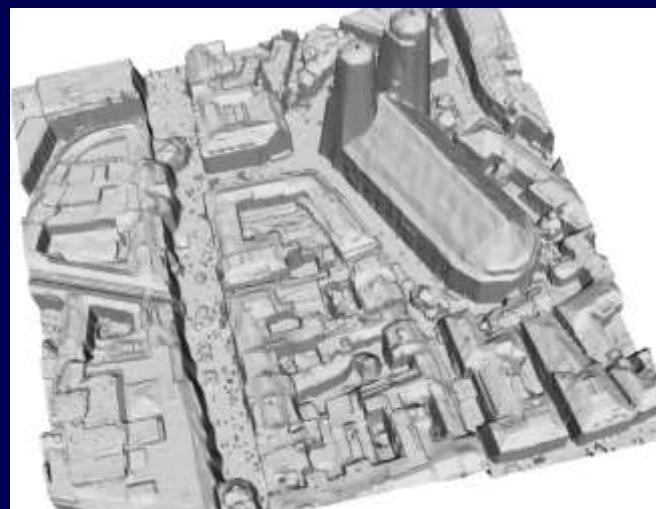
multiview reconstruction



super-res. textures



action reconstruction



stereo reconstruction



RGB-D modeling



3D on the fly