

Automated registration of 3D-range with 2D-color images: an overview

44th Annual Conference on Information Sciences and Systems
Invited Session:
3D Data Acquisition and Analysis

March 19th 2010

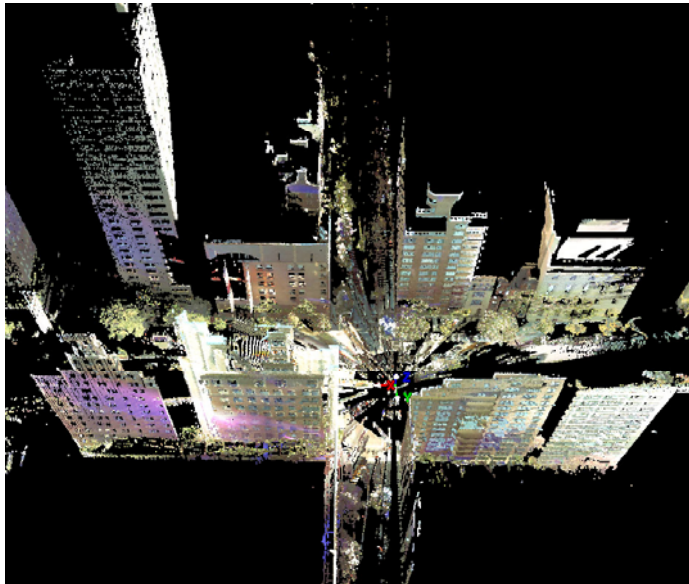


Ioannis Stamos
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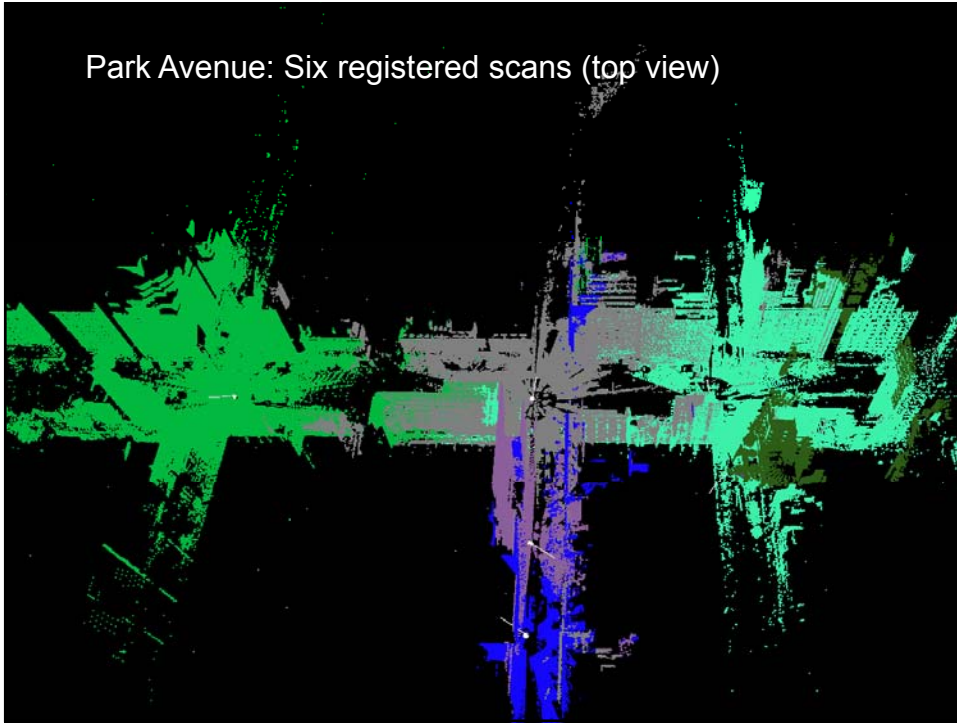


<http://www.cs.hunter.cuny.edu/~ioannis>

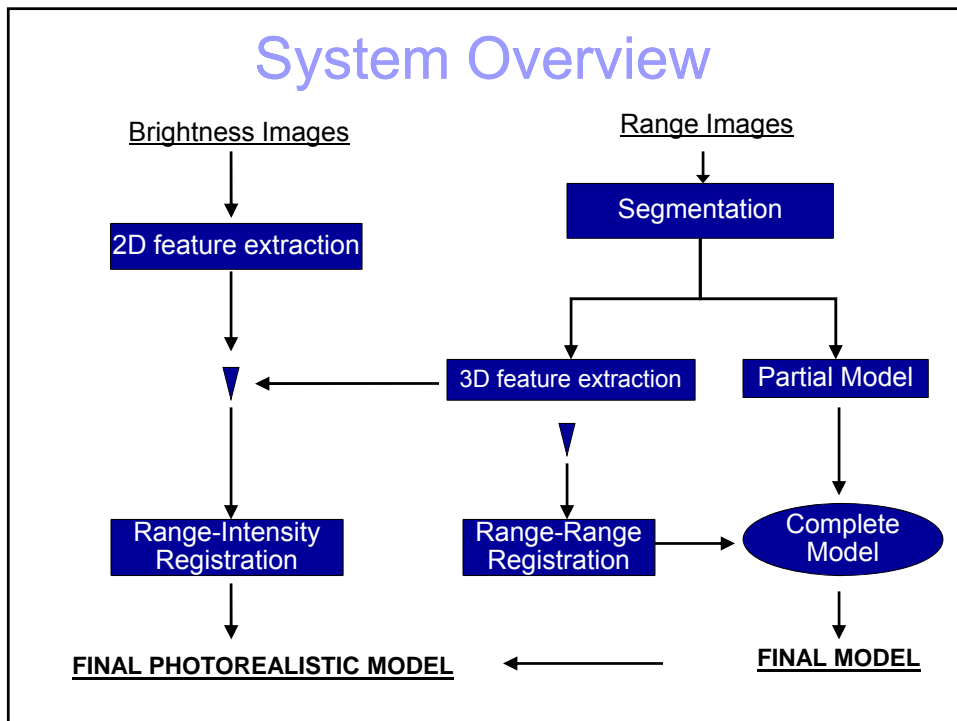
Data Acquisition, Leica Scan Station 2, Park Avenue and 70th Street, NY



Park Avenue: Six registered scans (top view)

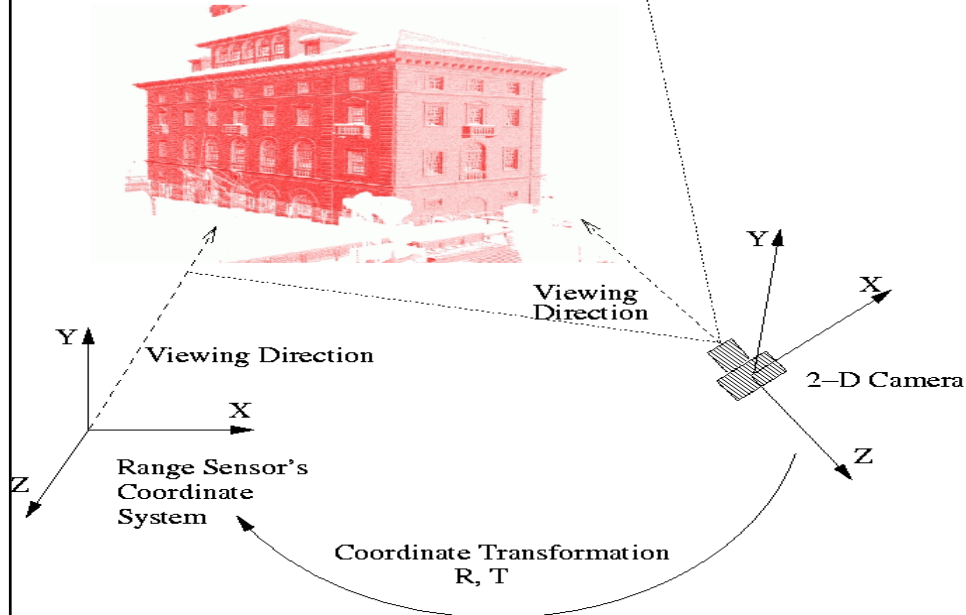


System Overview



Range-Intensity Registration

3-D depth map of the scene



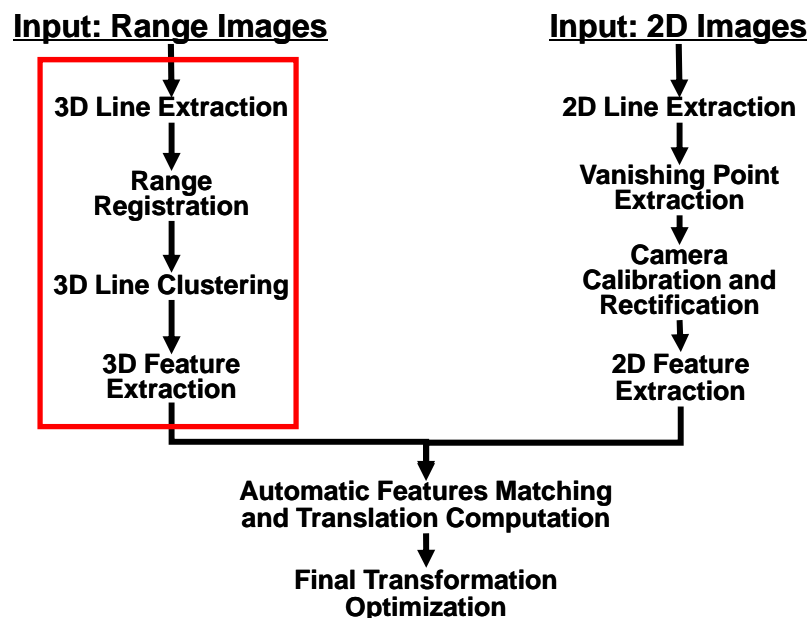
Fixed relative positioning - limitations

- 2D sensor is limited by 3D scanner
 - 3D scanning slow/laborious process
 - 2D image capture fast/flexible
- 2D sensor cannot dynamically adjust parameters on site
- Cannot handle historical photographs

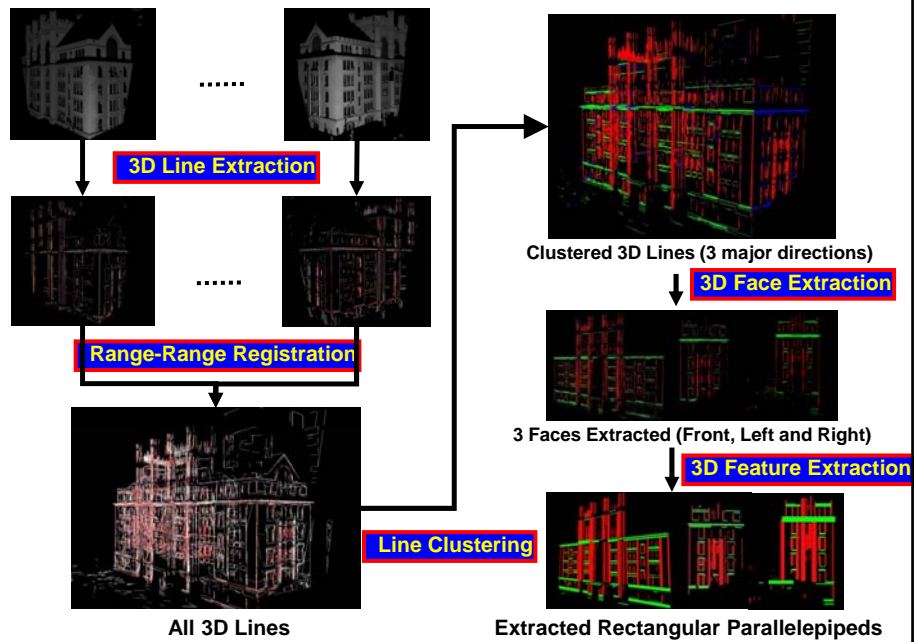
Automated methods

- **One 2D image – Untextured 3D model**
 - Ikeuchi 03 (edges – reflectance image)
 - Troccoli & Allen 04 (shadows)
 - Stamos & Allen 01 (rectangles)
 - Liu, Stamos 05 (parallelepipeds)
 - Liu, Stamos 07 (lines)
- **One 2D image - Textured 3D model**
 - Yang, Stewart etal. 07
 - Schindler (regularities)
- **Multiple 2D images (SfM/stereo) – 3D model**
 - Liu, Stamos 06, Zhao etal. 05
- **Aerial 2D images – 3D model**
 - Ding 08, Fisher 09, Neumann 09, Kaminsky 09.

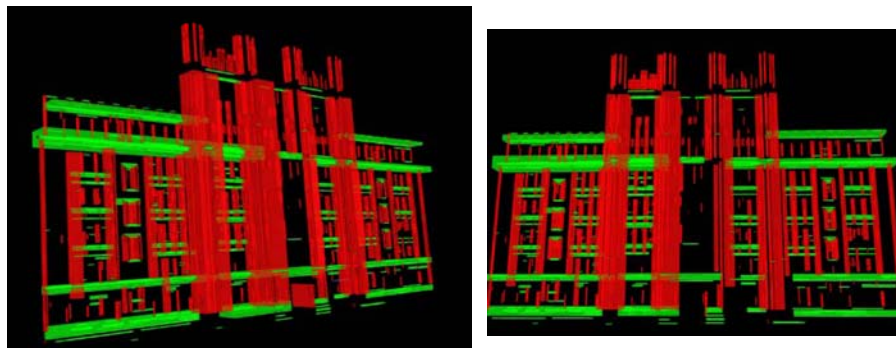
System Outline (Liu, Stamos)



3D feature extraction, Thomas Hunter Building



Clustering lines into vertical/horizontal parallelepipeds



System Outline (Liu, Stamos)

Input: Range Images

3D Line Extraction

Range
Registration

3D Line Clustering

3D Feature
Extraction

Input: 2D Images

2D Line Extraction

Vanishing Points
Extraction

Camera
Calibration and
Rectification

2D Feature
Extraction

Automatic Features Matching
and Translation Computation

Final Transformation
Optimization

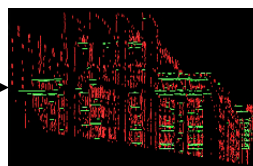
2D FEATURE EXTRACTION



2D Line
Extraction



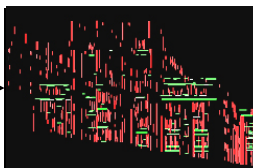
VPs
Extraction



Rectified 2D
Lines

Camera
Calibration and
Rectification

Matching between VPs and 3D
scene major directions



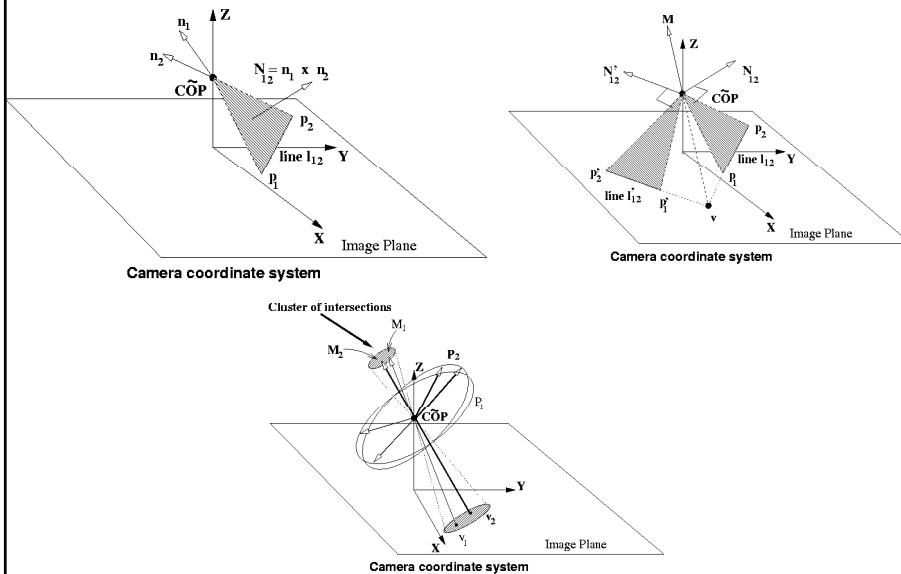
Extracted
Rectangles

Rotation between
camera and 3D
scene

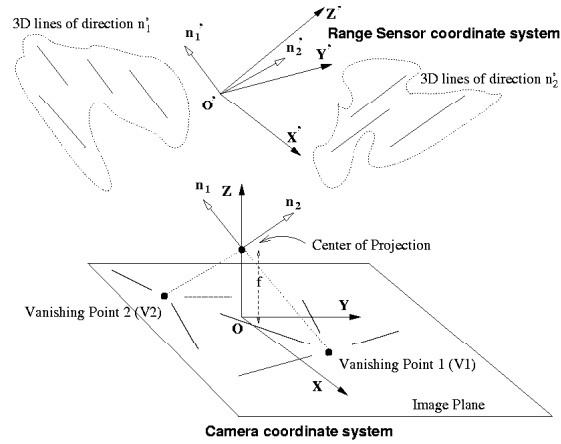
Vanishing Points (2D line clustering)



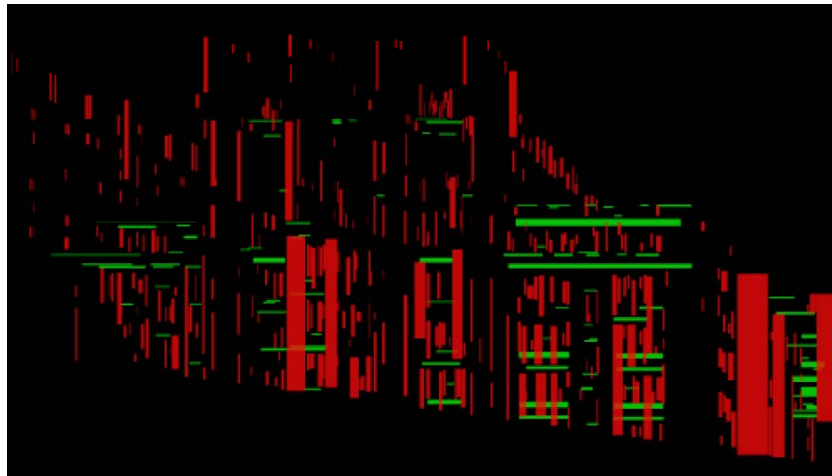
Computing Vanishing Points

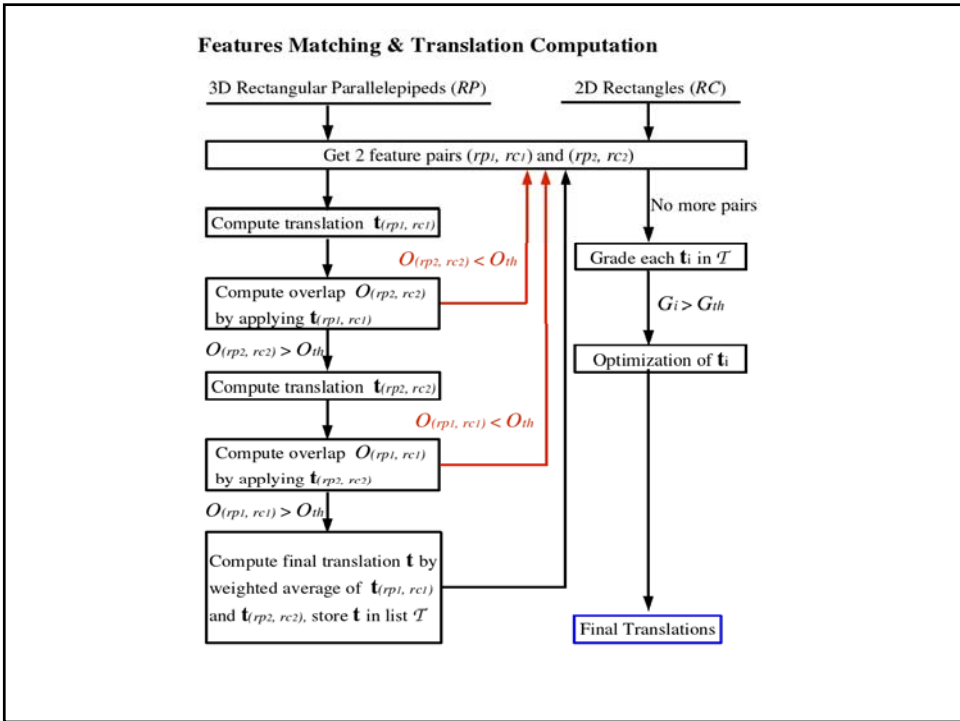
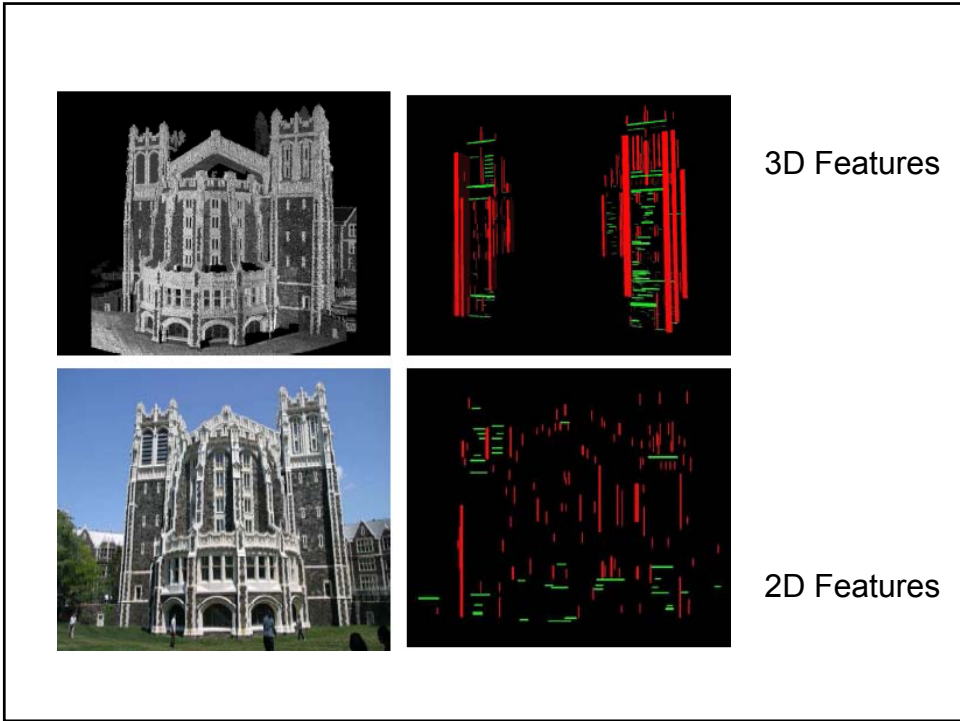


Solving for the rotation

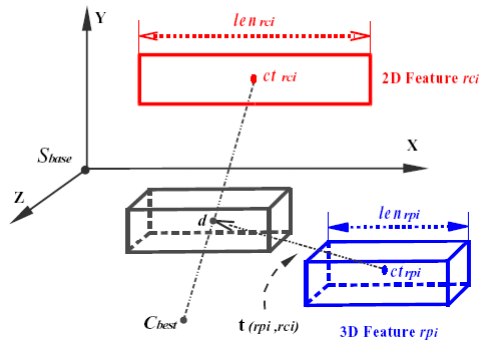


Clustering 2D lines from 2D image into horizontal/vertical rectangles

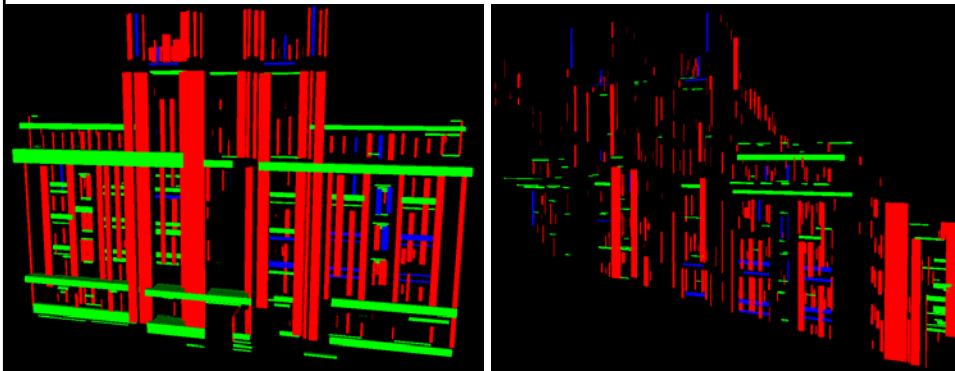




Translation computation from a matched 3D-2D pair



Matching 3D parallelepipeds with 2D rectangles



3D parallelepipeds

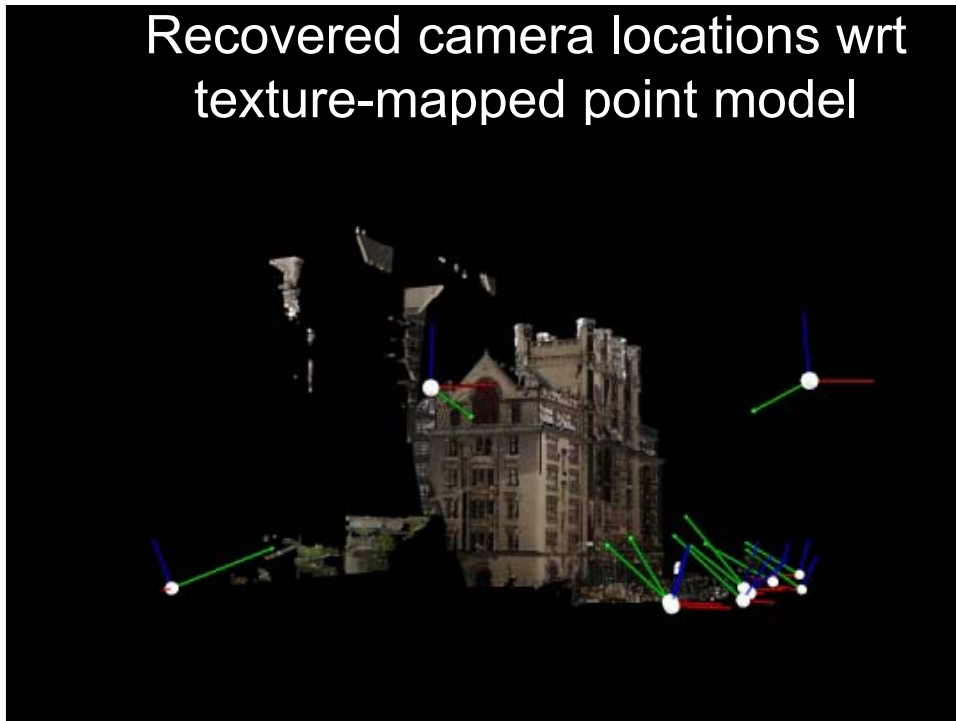
2D rectangles

Matches shown in blue

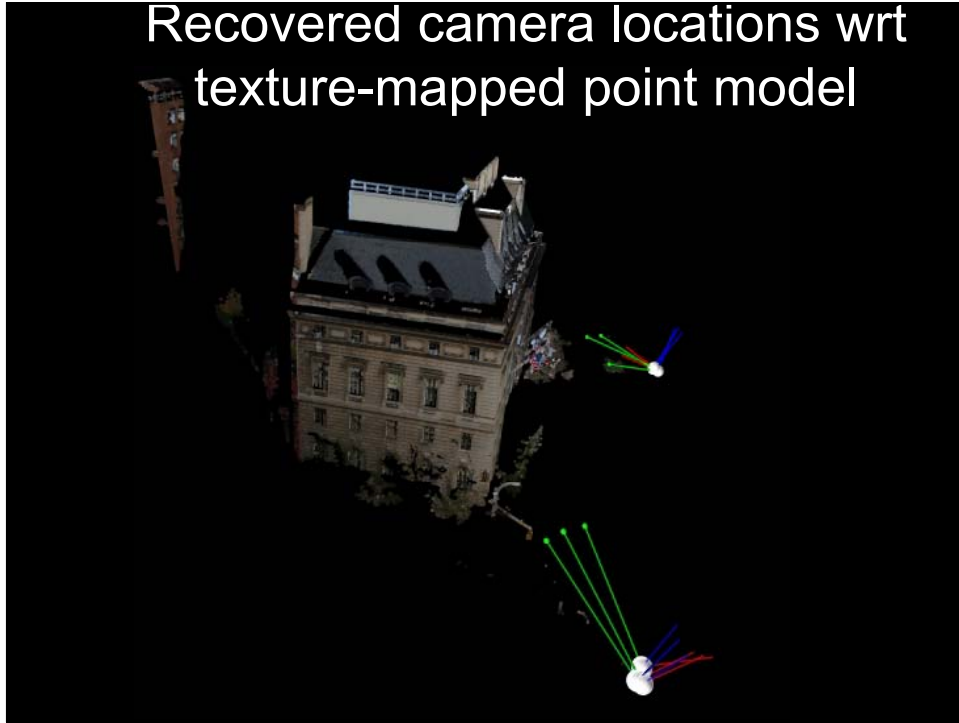
The performance on Thomas Hunter Building

FP	CM	G	OP (%)	RDT (%)	TIME (sec)
117 x 100	25	14	91.23	0.21	42
34 x 54	9	13	95.67	0.49	2
117 x 55	28	17	93.45	0.43	11
117 x 145	8	20	92.34	0.04	85
117 x 106	15	9	91.23	0.12	44
117 x 44	19	11	92.05	0.37	4
117 x 74	7	32	94.56	0.08	21
117 x 113	20	18	87.14	0.15	55
35 x 24	1	16	98.40	0.10	0.6
117 x 74	14	16	93.78	0.17	18
35 x 59	9	5	89.31	0.43	1.5

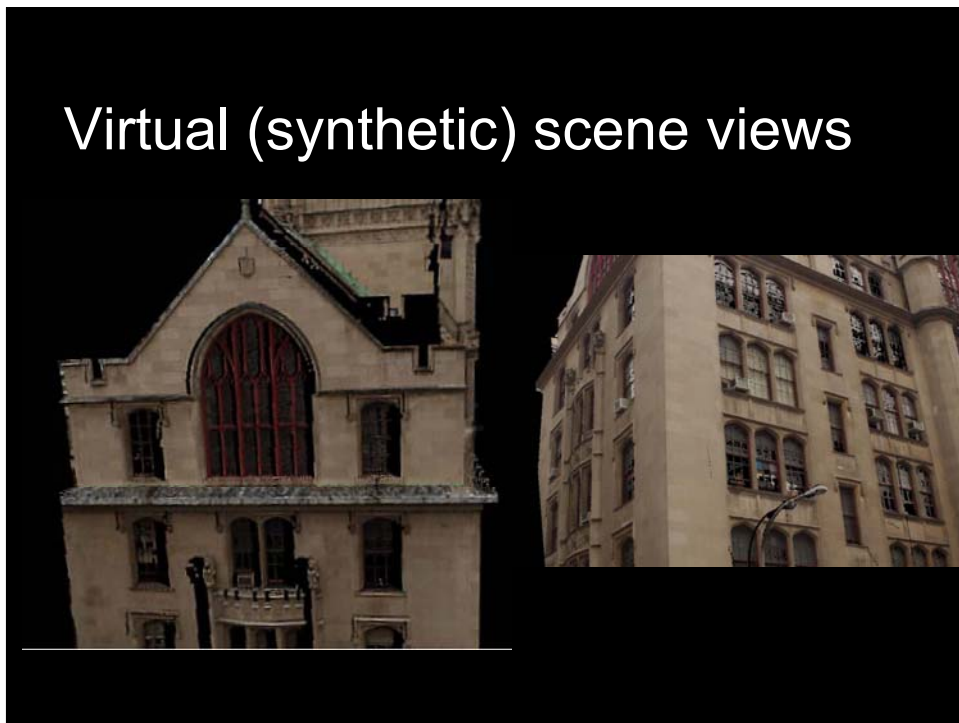
Recovered camera locations wrt texture-mapped point model



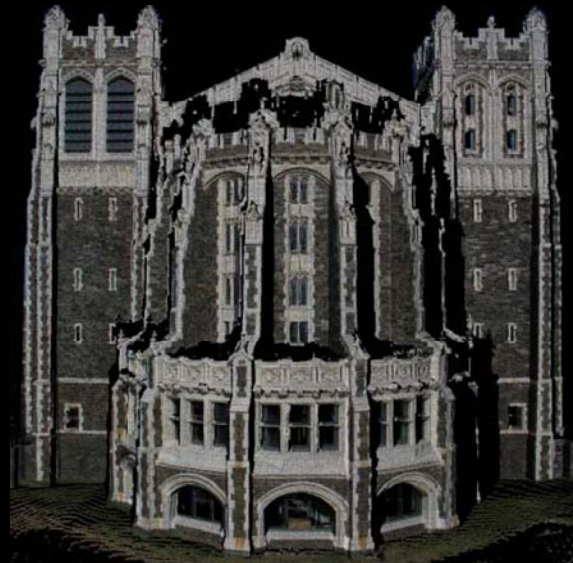
Recovered camera locations wrt
texture-mapped point model



Virtual (synthetic) scene views



Virtual (synthetic) scene views



Virtual (synthetic) scene views



Discussion

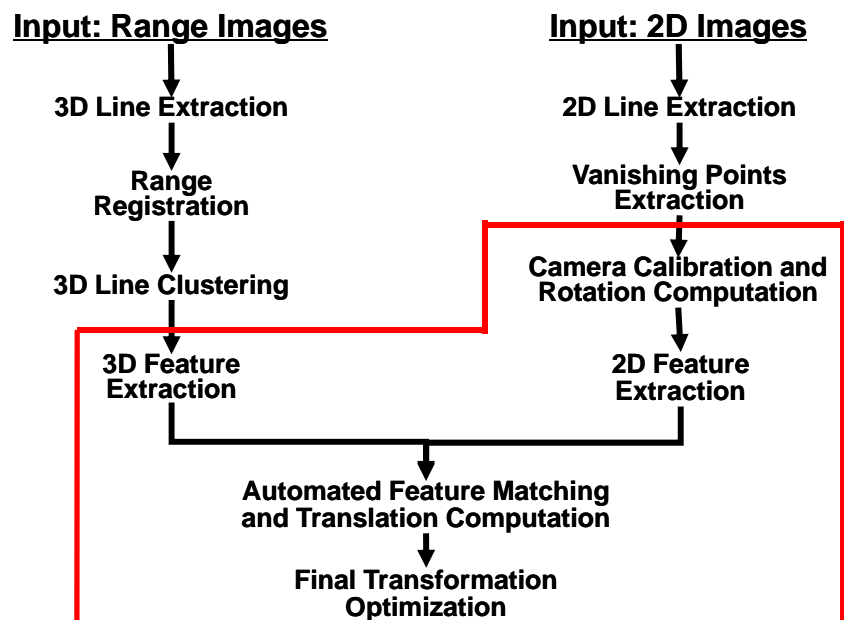
■ Advantages

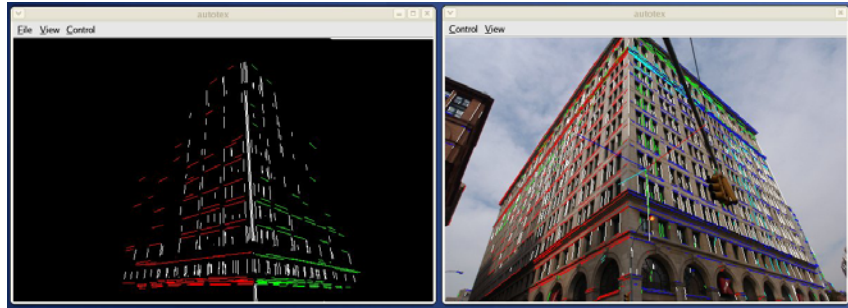
- Accurate
- Limited user interaction
 - 2D plane to 3D plane matching

■ Limitations

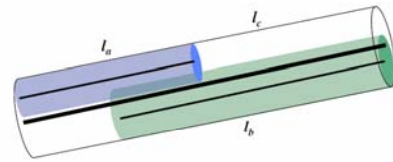
- Requires major facades
- Higher-order features decrease generality
- Non-interactive speed

System II Outline (lines)

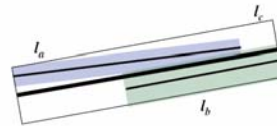




Feature Detection

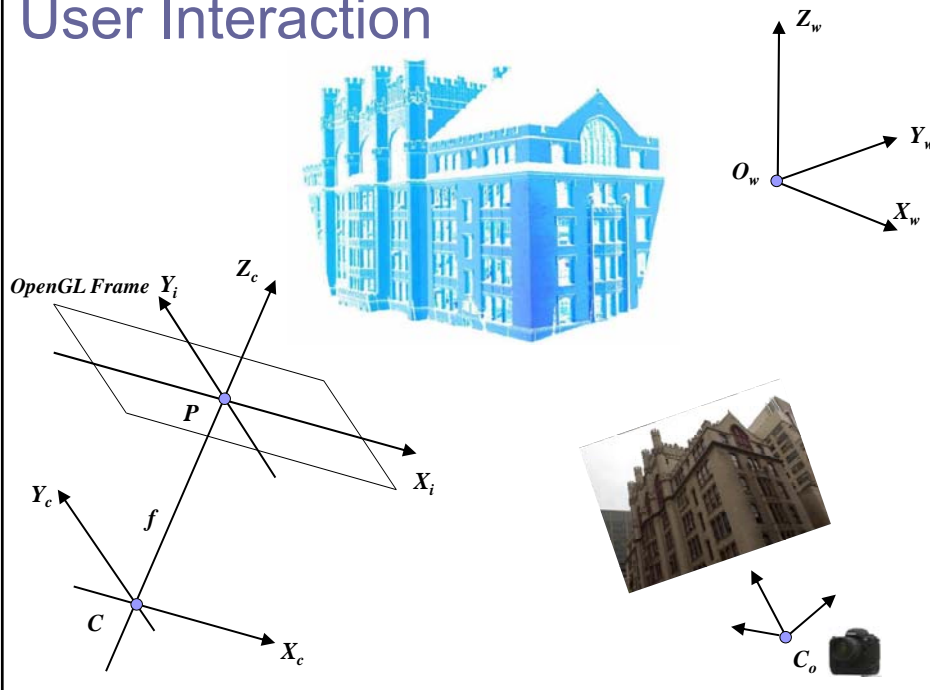


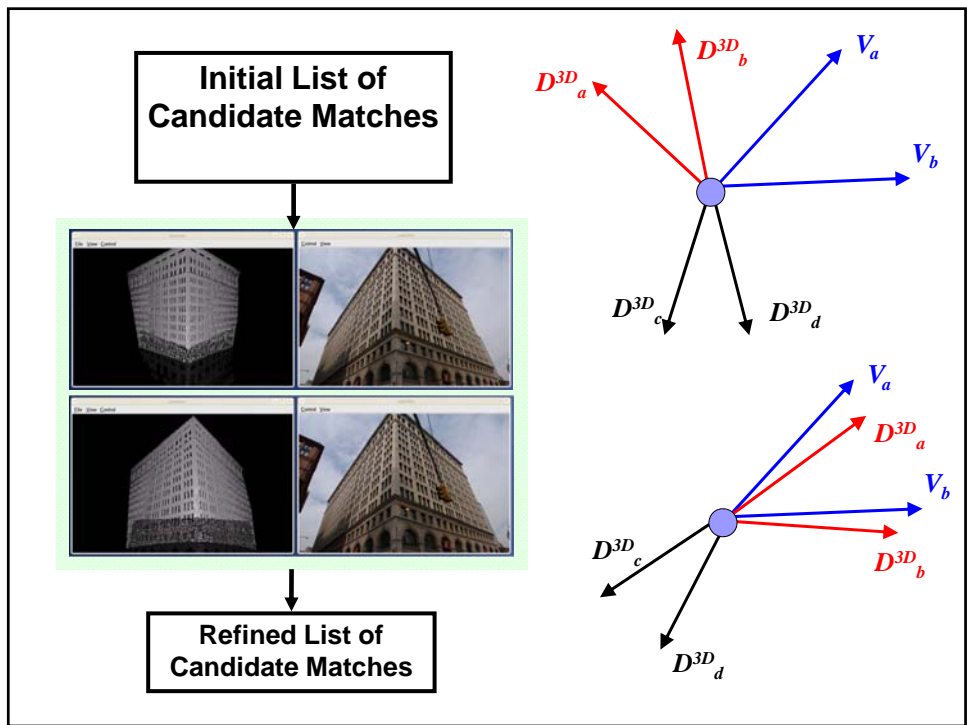
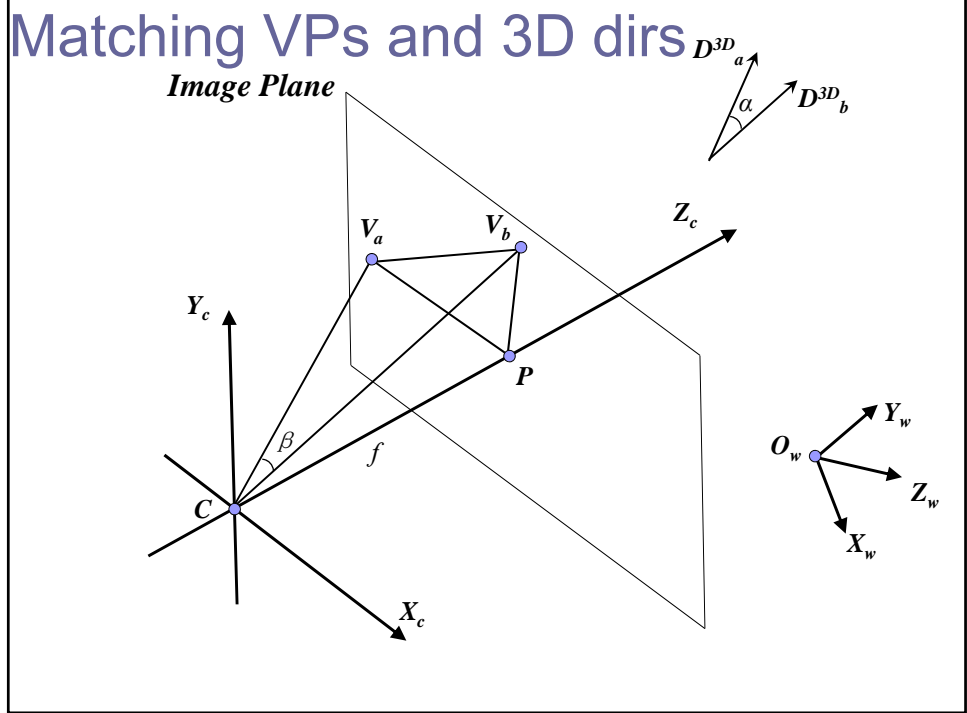
3D feature merging (l_a and l_b merged into l_c)



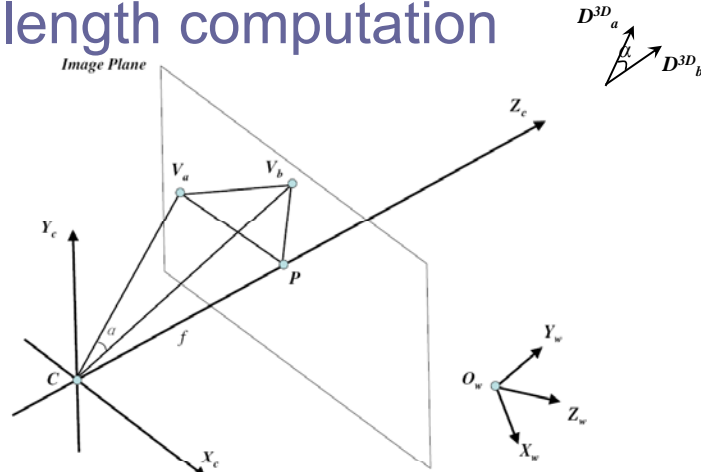
2D feature merging (l_a and l_b merged into l_c)

User Interaction





Focal length computation



$$|CV_a|^2 = |PV_a|^2 + f^2$$

$$|CV_b|^2 = |PV_b|^2 + f^2$$

$$|V_a V_b|^2 = |CV_a|^2 + |CV_b|^2 - 2 \cdot |CV_a| \cdot |CV_b| \cdot \cos \alpha$$

Automatic Feature Matching and Translation Computation

3D Line Feature

2D Line Feature

Get 2 pairs (l^{3D}_a, l^{2D}_a) and (l^{3D}_b, l^{2D}_b)

Compute Translation $C(l^{3D}_a, l^{2D}_a)$

Compute overlap $O(l^{3D}_b, l^{2D}_b)$ by applying $C(l^{3D}_a, l^{2D}_a)$ $O(l^{3D}_b, l^{2D}_b) < O_{th}$

$O(l^{3D}_b, l^{2D}_b) > O_{th}$
Compute Translation $C(l^{3D}_b, l^{2D}_b)$

Compute overlap $O(l^{3D}_a, l^{2D}_a)$ by applying $C(l^{3D}_b, l^{2D}_b)$ $O(l^{3D}_a, l^{2D}_a) < O_{th}$

$O(l^{3D}_a, l^{2D}_a) > O_{th}$

Candidate Trans C is the weighted average of $C(l^{3D}_a, l^{2D}_a)$ and $C(l^{3D}_b, l^{2D}_b)$. Store C in list T

No more pairs

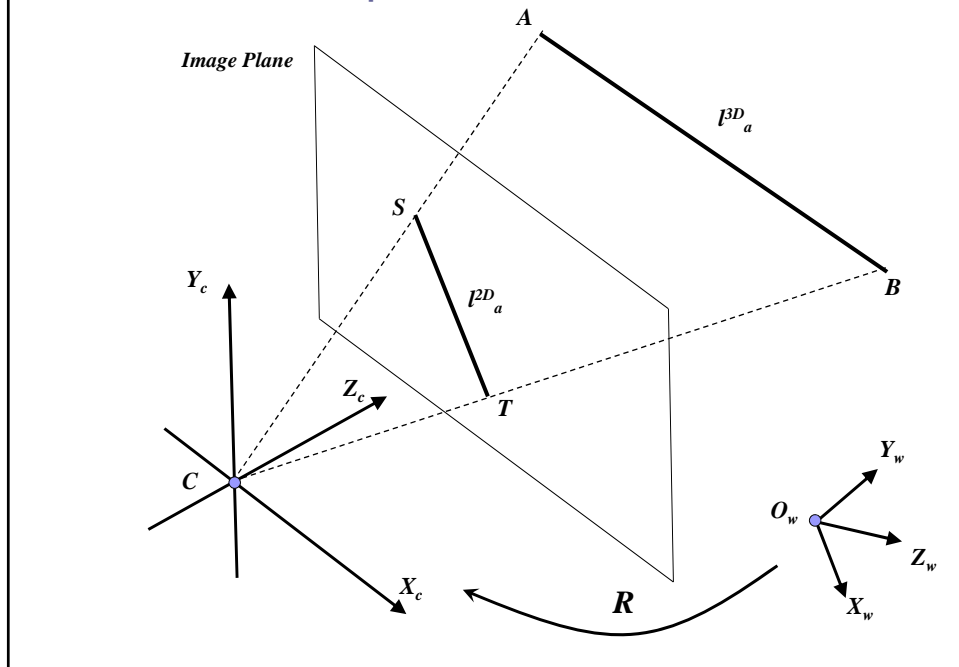
Grade each C_i in T

$G_i > G_{th}$

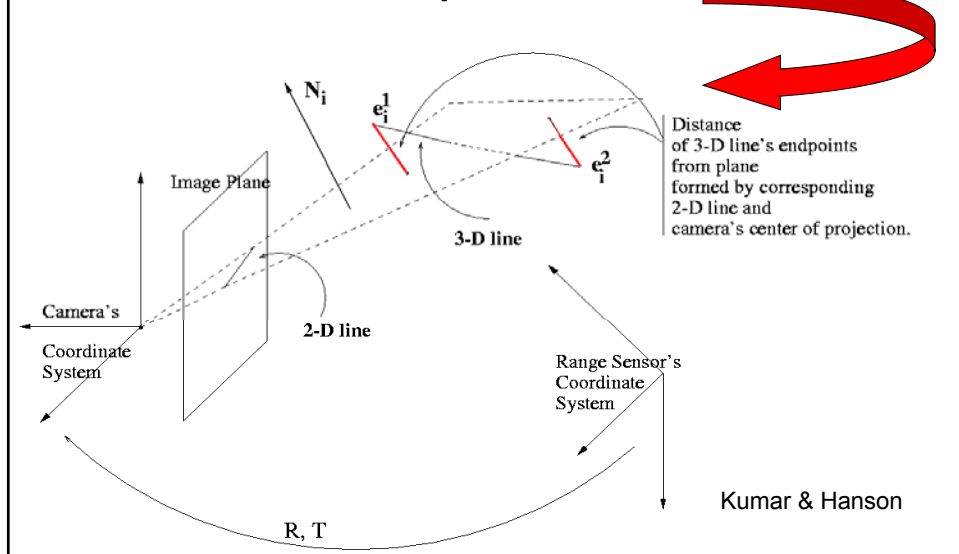
Optimization of C_i

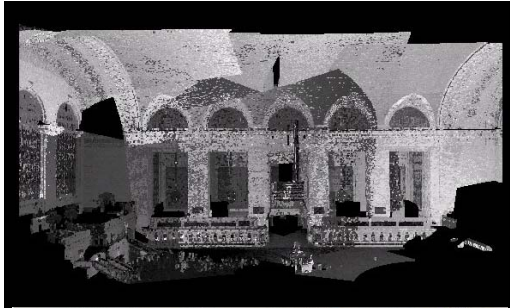
Final Translation

Translation computation: 3D-2D line match



Metric to be minimized for transform computation





GCT

3D scene

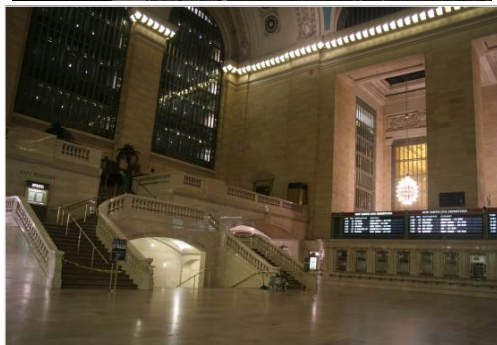


2D image

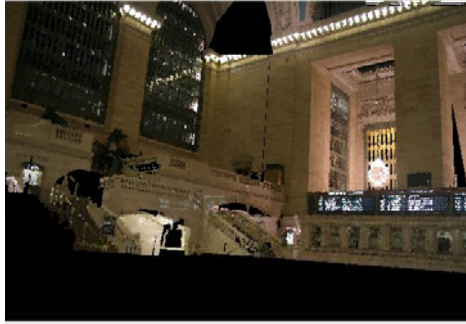


GCT

User interaction



2D image

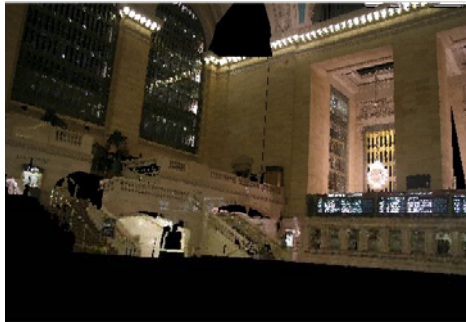


GCT

Text. mapped
3D model
(automated)



2D image



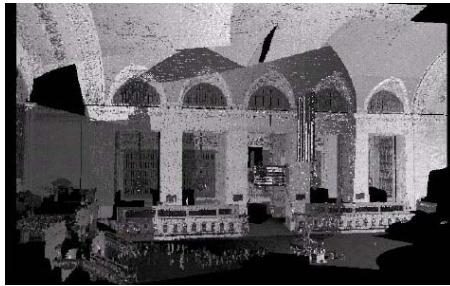
GCT

Text. mapped
3D model
(automated)



2D image
Corresponding
2D/proj. 3D lines

GCT



3D scene

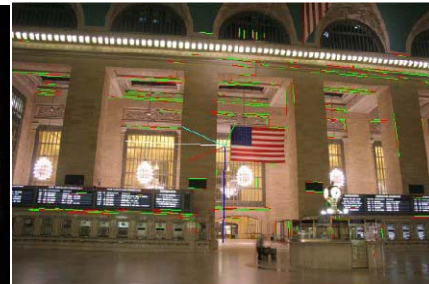


2D image

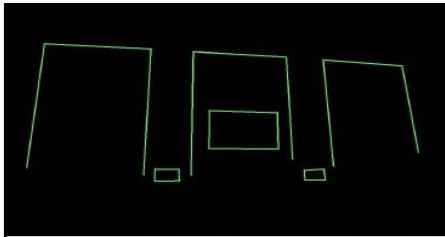
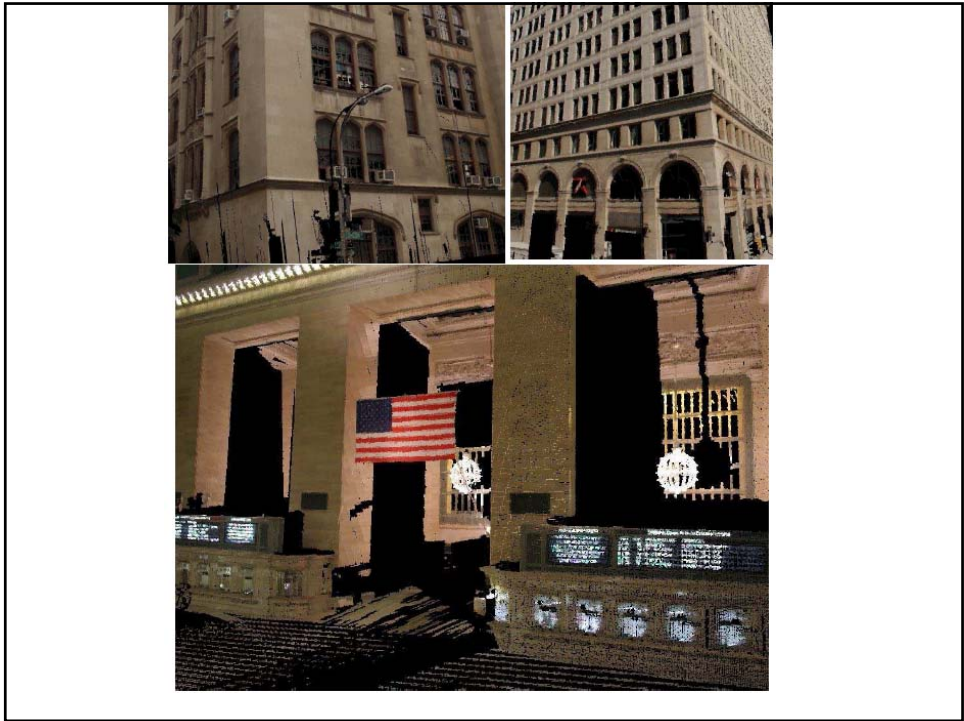
GCT



Texture mapped
3D model



Corresponding
2D/proj. 3D lines



Evaluation

Selected 3D lines from 3D cloud

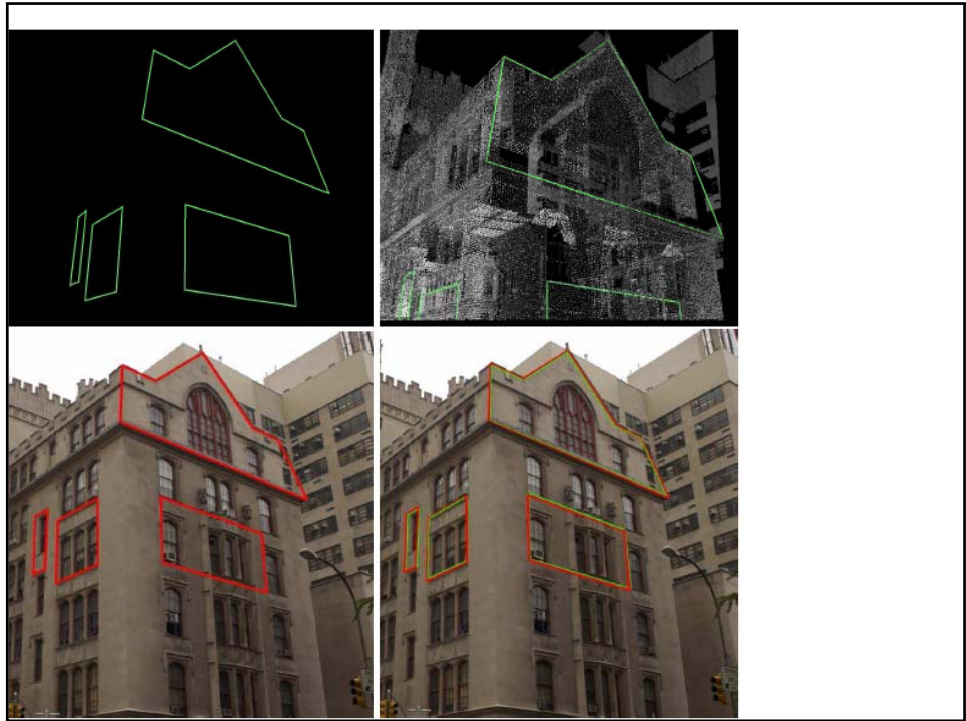


Selected 2D lines

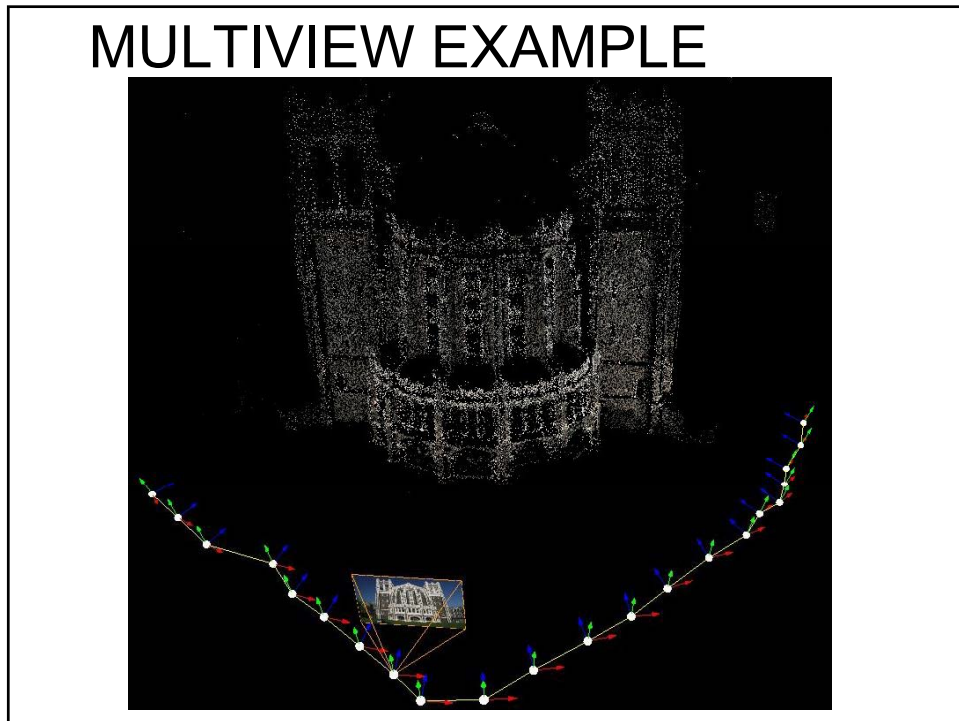
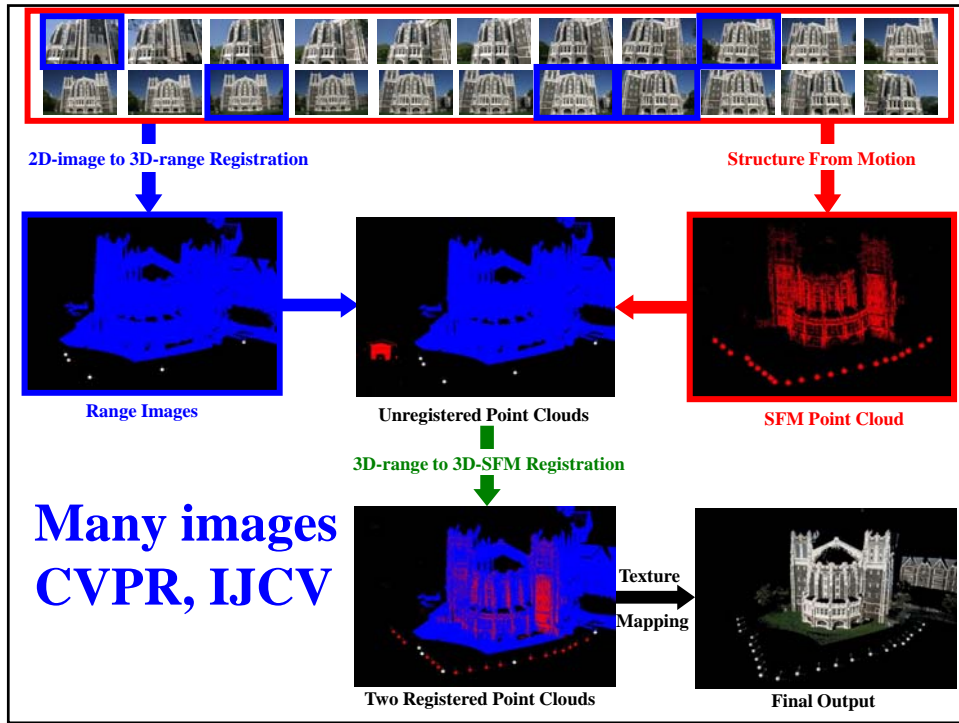


Projection of 3D lines on 2D img

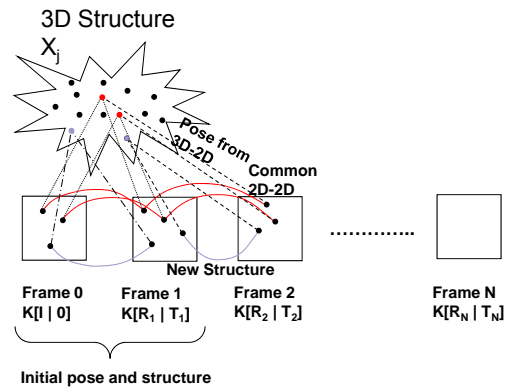
Calculate error



	3D lines	2D lines	focal (init)	focal (comp)	Matches	Error (pixels)
	F3D	F2D	Fi	Fr	M	E
Thomas Hunter	672	412	3065.83	3072.42	119	1.9872
	583	345	3065.83	3075.34	103	2.0121
	409	390	3065.83	3071.90	112	2.1029
	392	230	3065.83	3069.45	93	1.8752
	321	312	3065.83	3073.23	187	1.6523
	456	387	3065.83	3072.12	134	1.3892
	402	390	3065.83	3071.29	94	1.8973
	390	219	3065.83	3069.22	87	1.9653
	592	539	3065.83	3071.90	212	1.2393
	390	416	3065.83	3061.39	145	1.4203
	271	392	3065.83	3073.38	123	1.9153
	430	456	3065.83	3076.19	209	1.0872
390	549	3065.83	3063.56	115	1.6847	
Astor Place	438	789	1185.03	1165.65	114	1.4328
	421	654	1185.03	1175.89	83	1.5832
	389	520	1185.03	1172.90	88	1.2348
	402	432	1185.03	1179.34	101	1.5932
	389	598	1185.03	1172.90	91	1.6932
	435	621	1185.03	1169.39	156	1.5120
	419	535	1185.03	1178.17	182	1.7684
Grand Central Terminal	543	245	2805.81	2833.45	63	1.9574
	390	190	2805.81	2839.93	50	2.2383
	493	231	2805.81	2812.24	63	2.4892
	301	189	2805.81	2829.39	58	1.9432



POSE AND STRUCTURE RECOVERY



- Compute $[R_i | T_i]$ using 6-point RANSAC
- Sequential updating to recover pose and structure

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STRUCTURE AND MOTION REFINEMENT

- Bundle adjustment

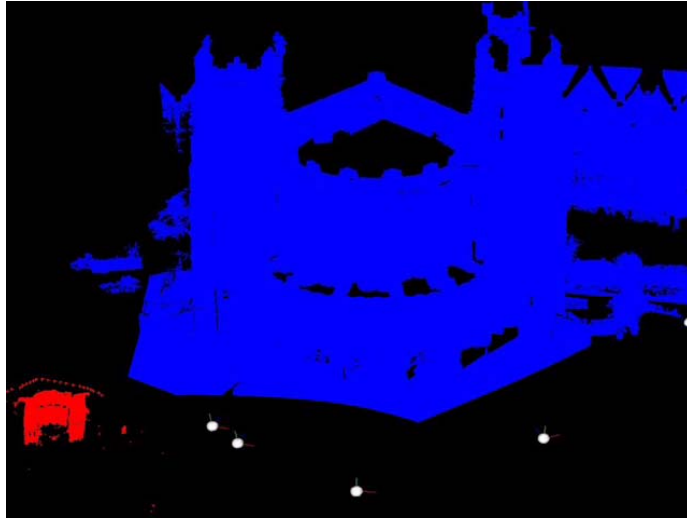
- Minimize reprojection error:

$$\min_{\hat{P}_k, \hat{M}_i} \sum_{k=1}^m \sum_{i=1}^n D(m_{ki}, P_k M_i)^2$$

- Maximum Likelihood Estimation
(assume error is zero-mean Gaussian noise)
- Huge problem but can be solved efficiently
(use sparse bundle adjustment)

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ALIGNING RANGE AND SFM POINT CLOUDS



Range and SFM models



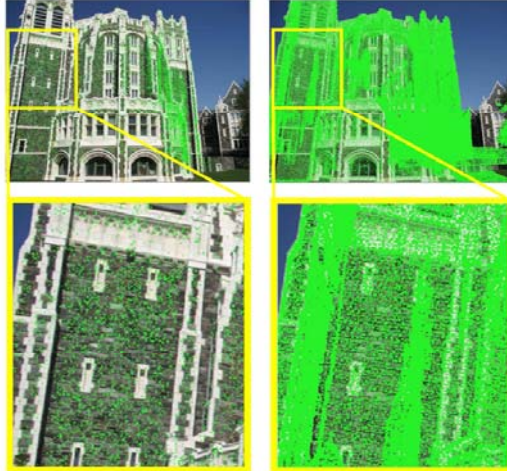
Range model



Range with overlaid SFM model (red)

3D-RANGE TO 3D-SFM REGISTRATION

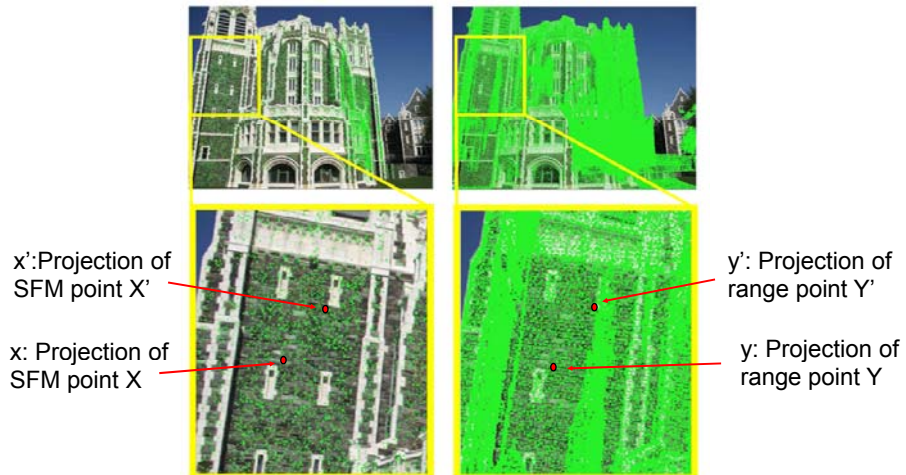
3D-SFM points projected onto image 3D-Range points projected on same image



Finding corresponding points between 3D-range and 3D-SFM models is possible on the 2D-image space

3D-RANGE TO 3D-SFM REGISTRATION

3D-SFM points projected onto image 3D-Range points projected on same image



(X, Y) and (X', Y') are candidate matches:
 y is the closest point to x and y' the closest point to x' in 2D-image space

3D-RANGE TO 3D-SFM REGISTRATION

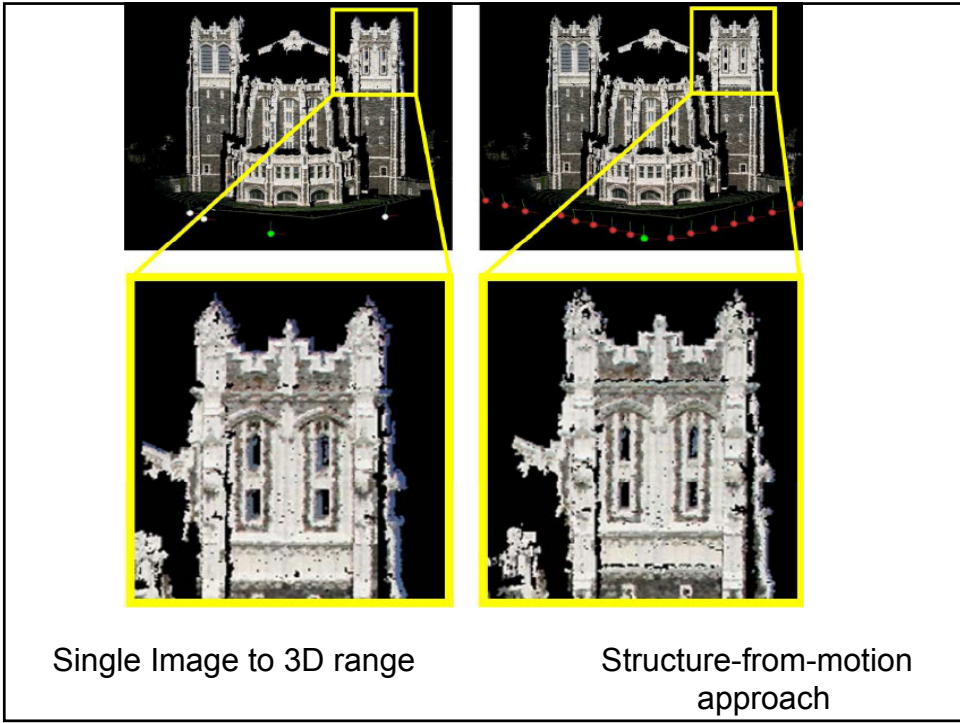
- Candidate matches (X, Y) and (X', Y') are compatible iff the following scale factors are equal:
 1. $s_0 = \|X - C^{sfm}\| / \|Y - C^{range}\|$
 2. $s_1 = \|X' - C^{sfm}\| / \|Y' - C^{range}\|$
 3. $s_2 = \|X' - X\| / \|Y' - Y\|$
 where C^{sfm} is the COP wrt the SFM model and C^{range} is the COP wrt the range model
- Confidence of $(X, Y) := \#$ compatible matches (X', Y') .
- Match (X, Y) with largest confidence among all 2D images provides the optimal scale factor s_{opt}
- List \mathcal{C} of robust matches :=

$$\{(X, Y) \text{ such that } s_0 \approx s_{opt}\} \cup \{(C^{sfm}, C^{range}) \text{ of images}\}$$
- Scale, rotation, and translation computed by minimizing

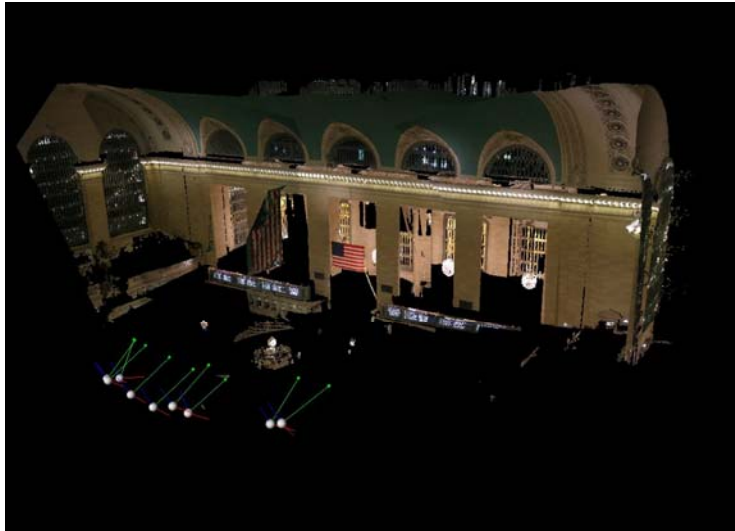
$$E = \sum_{(X, Y) \in \mathcal{C}} w \|sR \cdot Y + T - X\|^2$$

QUANTITATIVE RESULTS

	Shepard Hall		Great Hall
Number of points (M_{range})	12,483,568		13,234,532
Number of points (M_{sfm})	2,034	45,392	1,655
2D-images used	10	22	7
2D-to-3D registrations (Sec. 4)	10	5	3
No. of matches in \mathcal{C} (Sec. 6)	258	1632	156
Final optimization (Sec. 6)	8.65 s	19.20 s	3.18 s



GRAND CENTRAL (1)



GRAND CENTRAL (2)



CONCLUSIONS

- Integration of multiview geometry with range registration
 - 2D-to-3D registration is used for a subset of images that contain a sufficient number of linear features
 - Brings these images into alignment with the dense 3D-range model
 - Multiview geometry exploits 2D-point correspondences
 - Brings all images into alignment and produces sparse SFM model
 - 3D-range to 3D-SFM registration
 - Aligns all images with the dense 3D-range model
- Accurate texture mapping onto dense 3D-range data
- Limitations:
 - Need accurate vanishing points (3D-2D registration)
 - Need accurate 3D features
 - Need to handle symmetries

Credits

- Collaborators:

- Dr. Cecilia Chao Chen (Google, Inc.)
- Dr. Lingyun Liu (Google, Inc.)
- Dr. Siavash Zokai
- Dr. Gene Yu
- Prof. George Wolberg (CCNY)

- Funding:

- NSF CAREER, MRI, RI, MSC Awards
- PSC-CUNY Grants
- CUNY Institute of Software Design & Development

Thank you

<http://www.cs.hunter.cuny.edu/~ioannis>

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