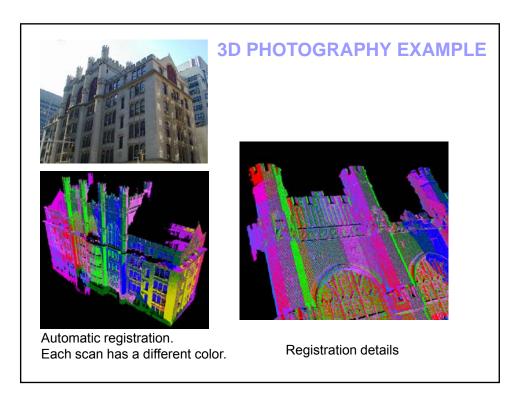
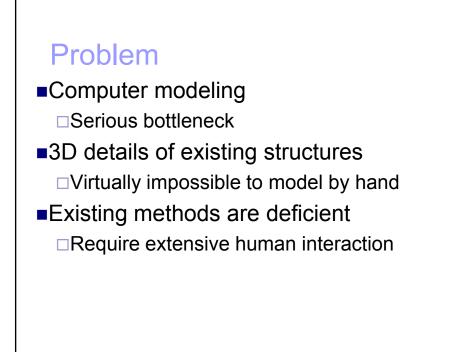
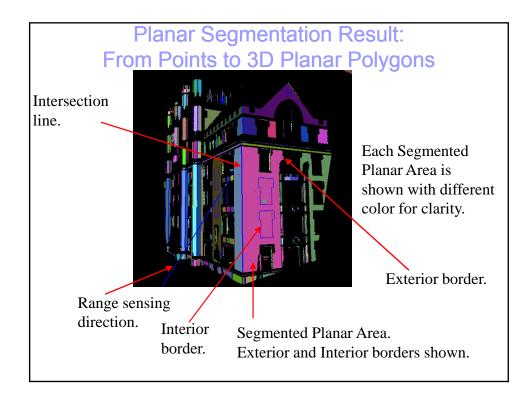
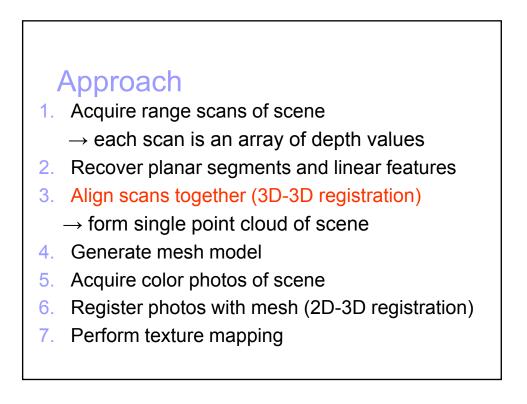
## 3D Pipeline Segmentation (planar) Registration (using lines)





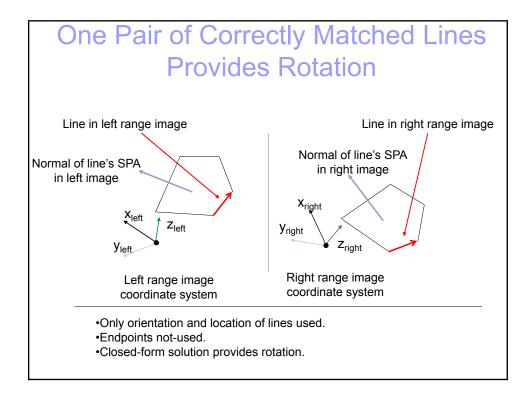


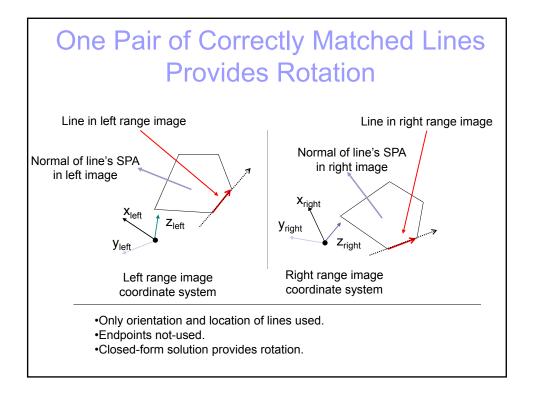


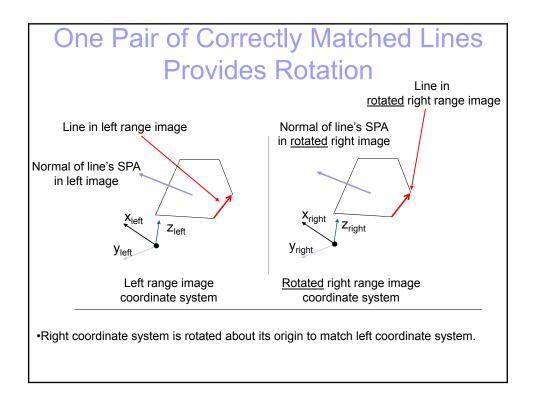


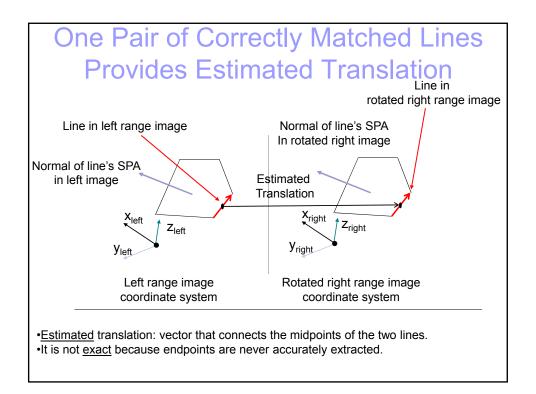
# Feature-based range-range registration

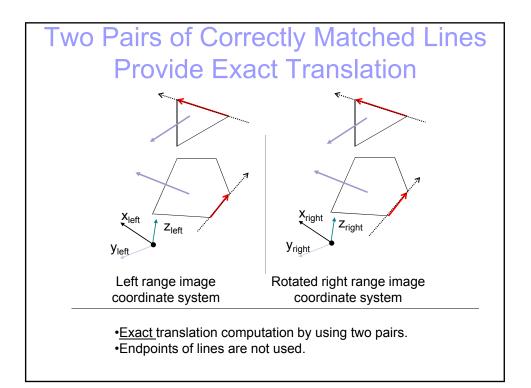
- Pairwise registration between two scans.
- Automated method uses linear features.
- Features extracted at boundaries of SPAs.
- Two correctly matched lines between scans provide solution.
- If scan A contains N lines and scan B contains M lines
  Need to consider O(N<sup>2</sup>M<sup>2</sup>) pairs.
  - □ For each pair verification of registration needed.
- Naïve method is time consuming.
- Two efficient novel algorithms developed.
- Problems also induced by scene symmetry.
- User-interface for smart user interaction developed.





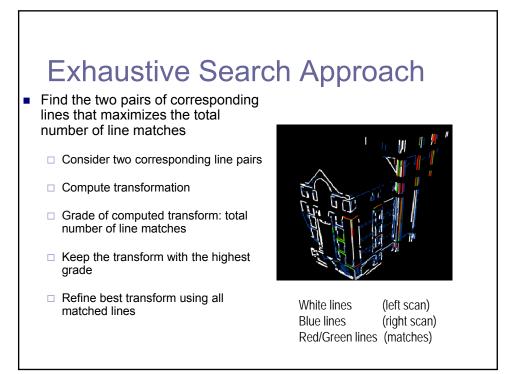


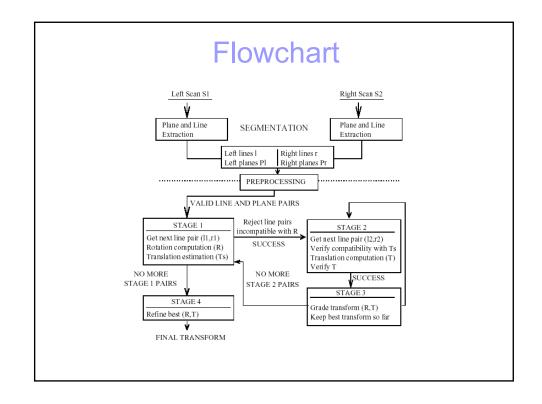


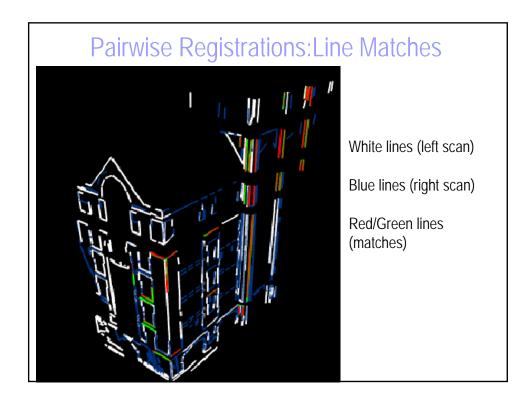


#### First efficient algorithm

- Problems to tackle:
  - □ Noise: Lines and normals do not match exactly.
  - □ Search space is large.
  - □ Verification of correct match expensive.
- Search for correct pairs of matched lines.
  Search first for one pair.
  - Proceed to search for second pair.
  - Grade each computed transform: # of matches.
  - □ Keep the transform with the highest grade.
  - □ At the end refine best transform using all lines.

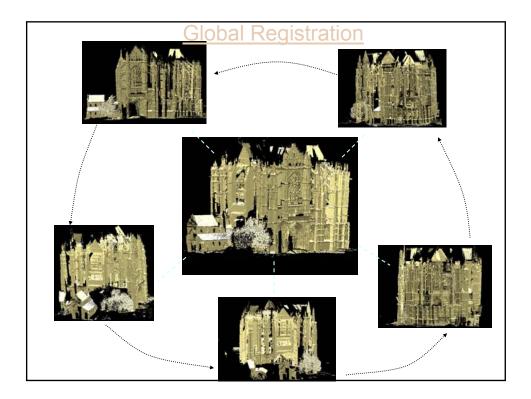


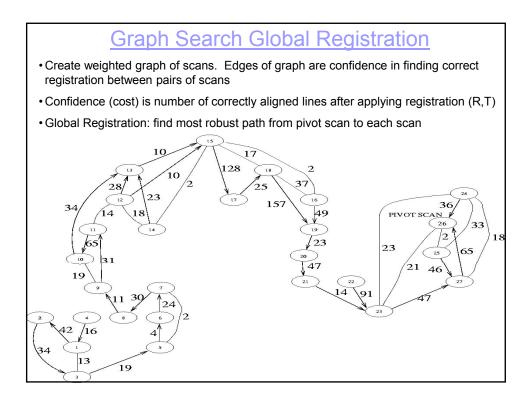


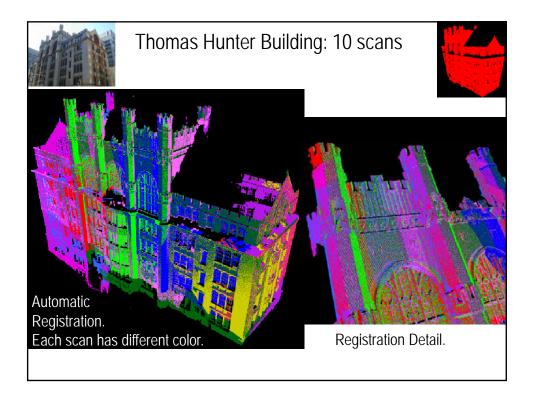


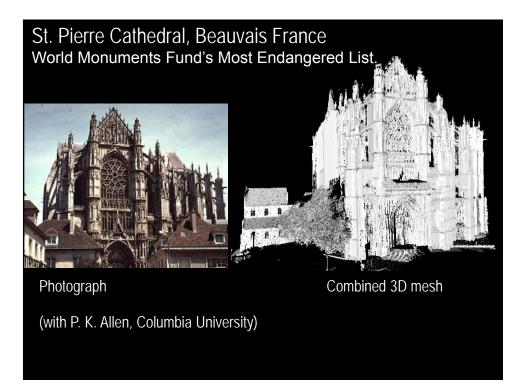
Results
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Pair	Line Pairs	Pre (%)	S2 % (#)	S3 % (#)	Matches	t (sec)	Error
1	$301 \times 303$	16	1.7(1555)	0.38 (346)	35	15	10.99mm
2	$303 \times 290$	17	2.8(2429)	0.84(735)	25	29	6.28mm
3	$290 \times 317$	21	2.8(2572)	1.88 (1728)	36	52	2.77mm
4	$317 \times 180$	19	3.4(1955)	1.15(656)	28	21	14.96mm
5	$211 \times 180$	21	4.6 (1759)	2.1 (802)	31	19	9.26mm
6	$180 \times 274$	17	2.6(1306)	0.34(168)	22	9	11.42mm
7	$114 \times 274$	19	1.6(507)	2.2 (894)	33	6	-5.61mm
8	$274 \times 138$	16	1.8(667)	1.5(557)	31	5	- 3.08mm
9	$114 \times 138$	18	2.7(423)	3.8(593)	32	4	- 3.94mm
10	$138 \times 247$	18	2.3(791)	1.3(429)	20	5	-1.36mm
		Cathedra	l - Results	(average erre	or 17.3mm	ι)	
1	$406 \times 464$	7	0.9(1650)	0.3(615)	42	39	9.37mm
2	$464 \times 269$	7	0.7(888)	0.3(443)	34	16	16.9mm
3	$406 \times 269$	11	0.7 (794)	0.1 (104)	13	9	56.08 mm
4	$151 \times 406$	21	1.1(668)	0.8(480)	16	7	5.34mm
5	$269 \times 387$	11	0.7(702)	0.4(369)	19	9	15.8mm
6	$326 \times 197$	10	0.9(597)	0.1 (49)	24	4	11.68mm
7	$197 \times 143$	15	1.0 (290)	0.3 (82)	30	3	6.44mm
8	$143 \times 194$	16	1.9(520)	0.1 (31)	11	3	29.24mm
9	$194 \times 356$	15	2.0(1429)	0.1 (93)	19	11	30.82mm

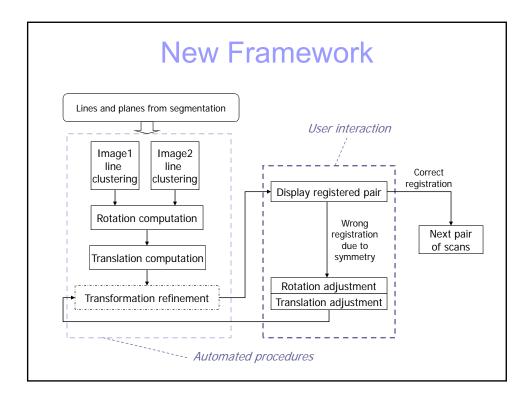


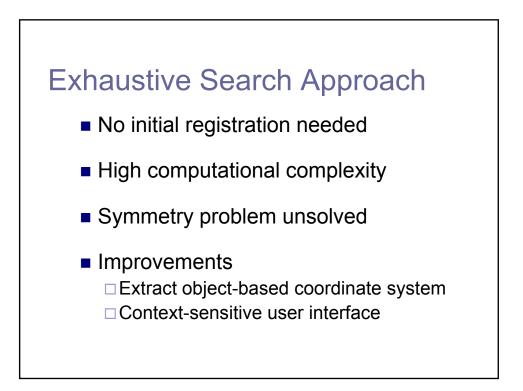


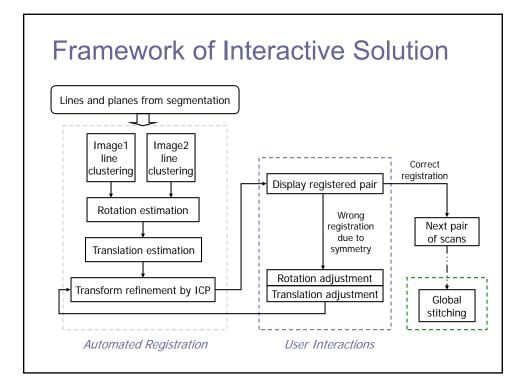


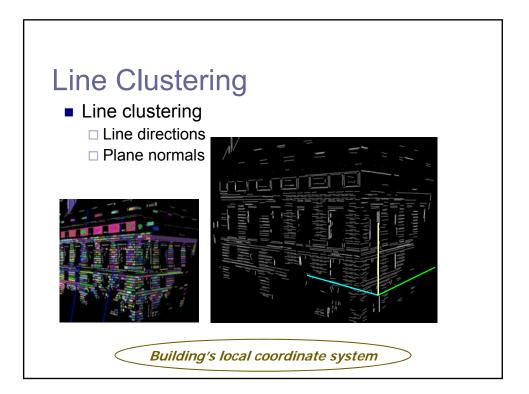


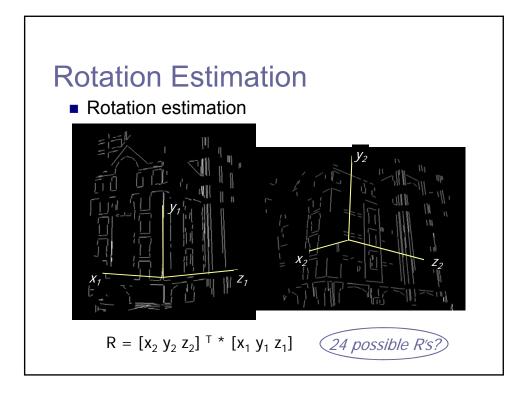


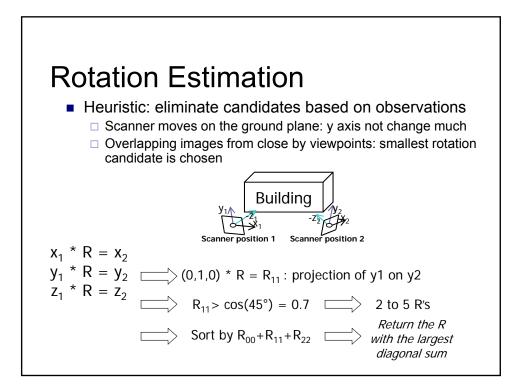






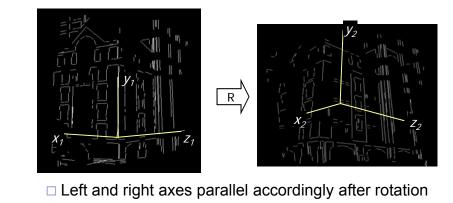




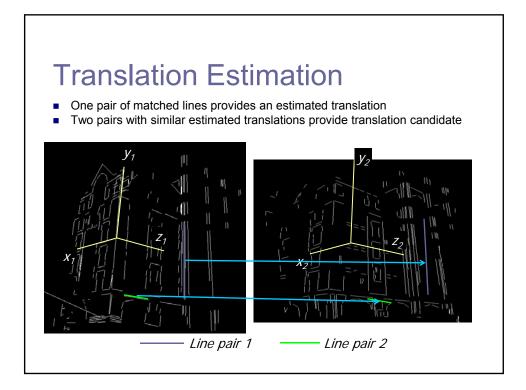


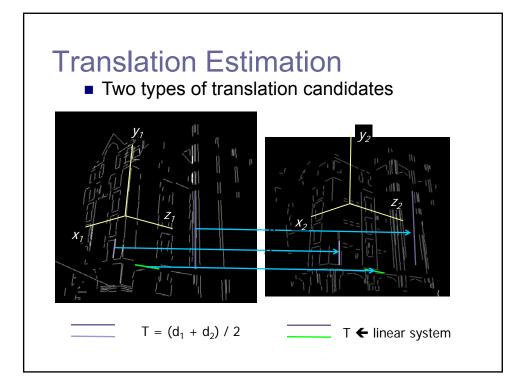
## **Translation Estimation**

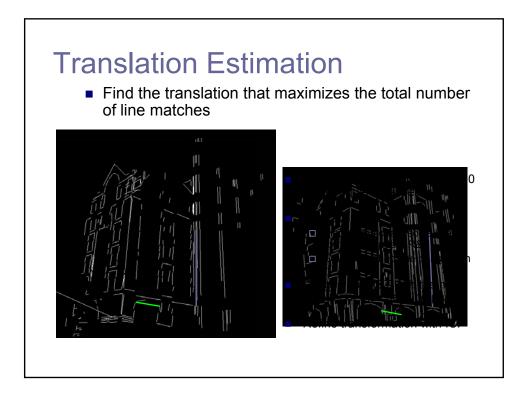
Translation estimation

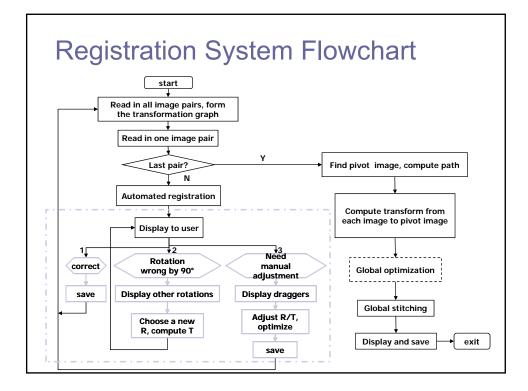


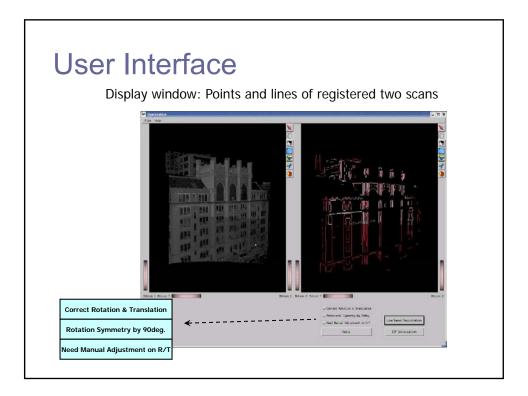
□ Pick robust line pairs to estimate translation

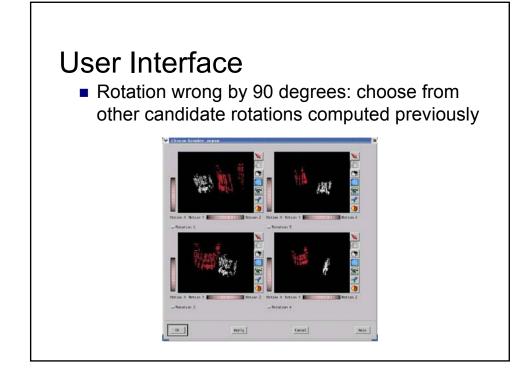


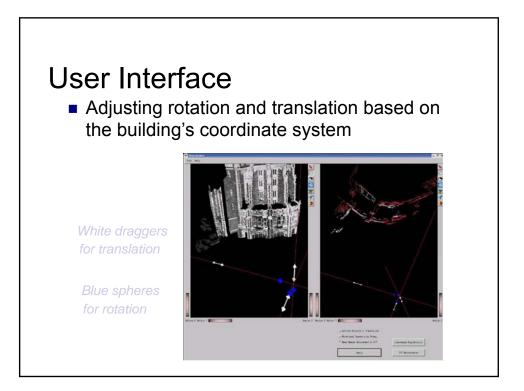


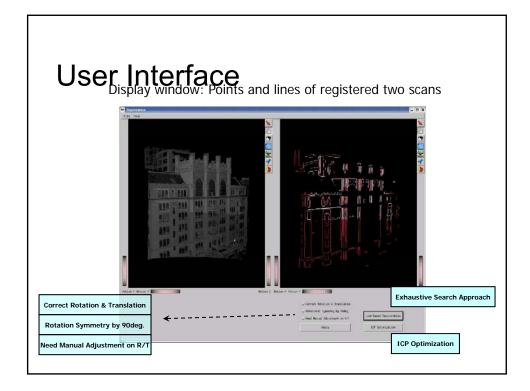


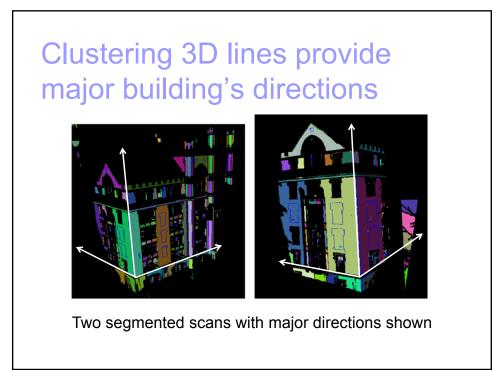


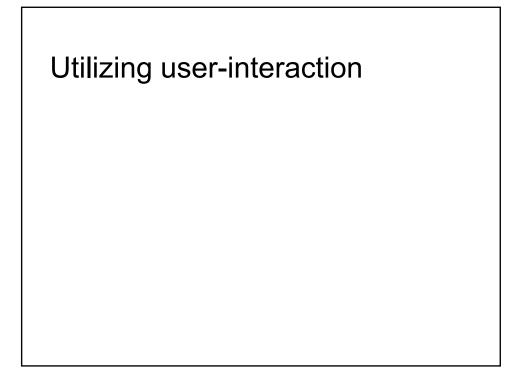


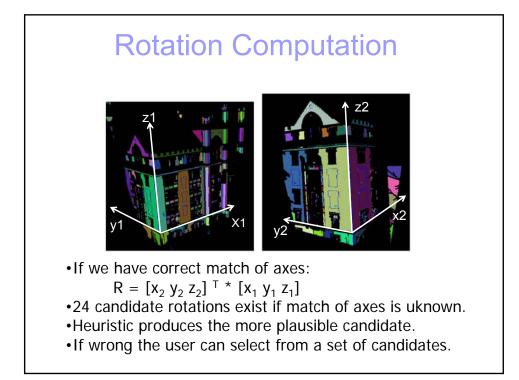


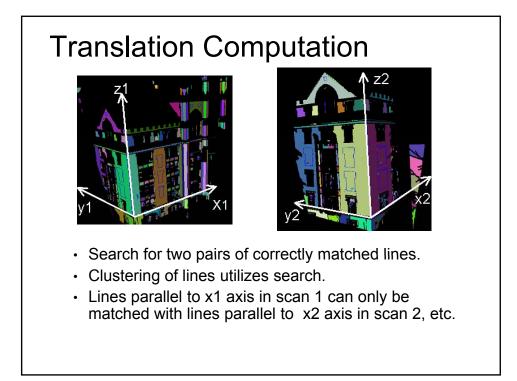


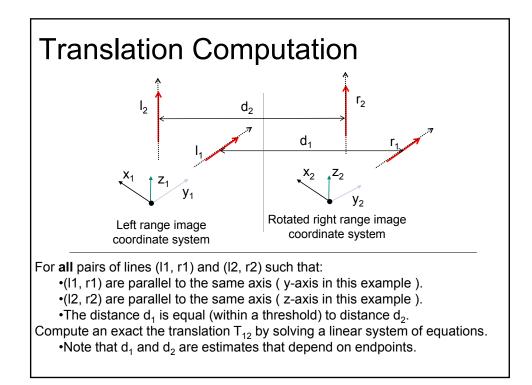






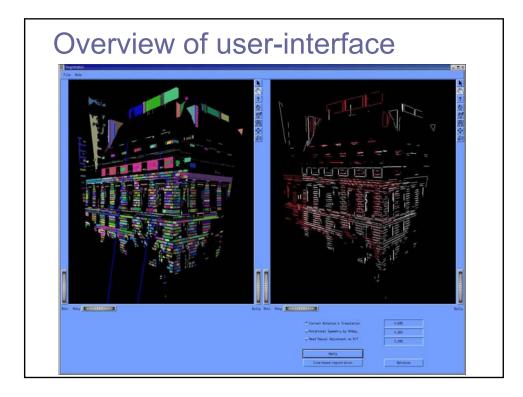


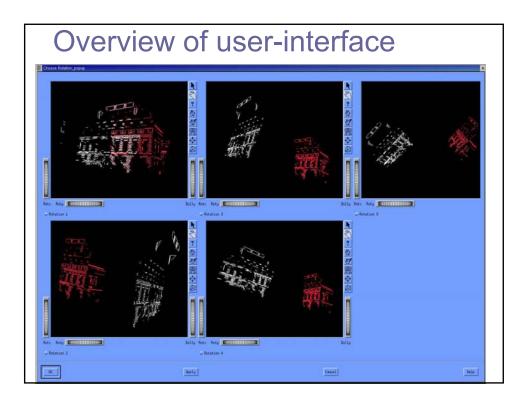


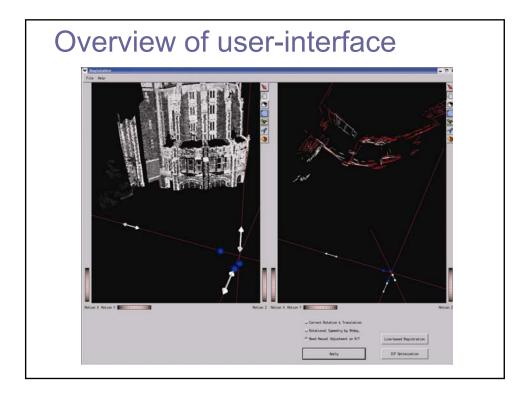




- Cluster all computed translations.
- Pick N most frequently appeared translations.
- For each of the N translations:
  - $\square$  Apply an optimization routine on R and T.
  - $\hfill\square$  Count matched line pairs with optimized T.
- Pick the T with the largest number of matches.







### **Iterative Closest Point Algorithm**



Before ICP



After ICP

