

# Fast 3D Model Generation in Urban Environments

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IEEE MFI 2001

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Machine Vision Seminar 2011

# Outline

- ⦿ Overview
- ⦿ Goals and Objectives
- ⦿ Approach
- ⦿ Ground-based Modeling

## Overview

# Why do we need 3D Models?

- ⦿ Urban Planning
- ⦿ Virtual Reality
- ⦿ Simulation
- ⦿ Special Effects
- ⦿ Car Navigation

## Overview

# Known Techniques (relative to the paper)

- ⦿ remote sensing
- ⦿ using lines extracted from merged camera images
- ⦿ 3D laser scanners
- ⦿ 2D laser scanners on mobile robot

# Overview Challenges?

- ⌚ acquisition is difficult and time-consuming.
- ⌚ significant manual intervention.
- ⌚ data acquired in a stop-and-go fashion.
- ⌚ impossible to monitor changes over time.

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- ⦿ Ground-based Modeling

# Goals and Objectives

reconstruct 3D city models for  
virtual walk-throughs and fly-throughs

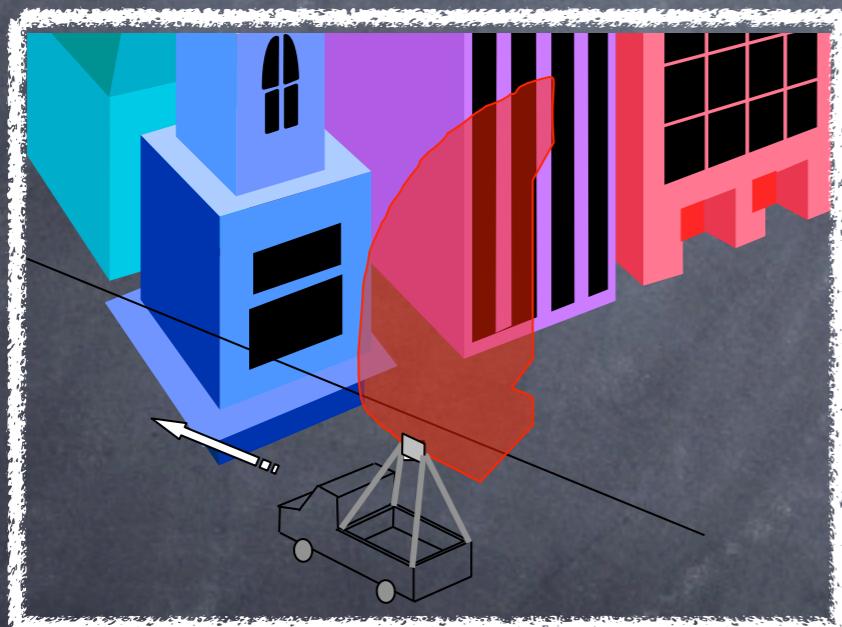
- ⦿ automated
- ⦿ photorealistic
- ⦿ fast
- ⦿ scalable

# Outline

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- ⦿ Approach
- ⦿ Ground-based Modeling

# Approach

ground-based modeling  
building facades



airborne modeling  
rooftops & terrain



registration  
+  
fusion

3D city model

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- ⦿ Approach
- ⦿ Ground-based Modeling

# Ground-based Modeling

- ⦿ data-acquisition system
- ⦿ scan matching
- ⦿ path computation
- ⦿ point cloud generation
- ⦿ mesh generation
- ⦿ texture mapping

# Ground-based Modeling Data-acquisition system

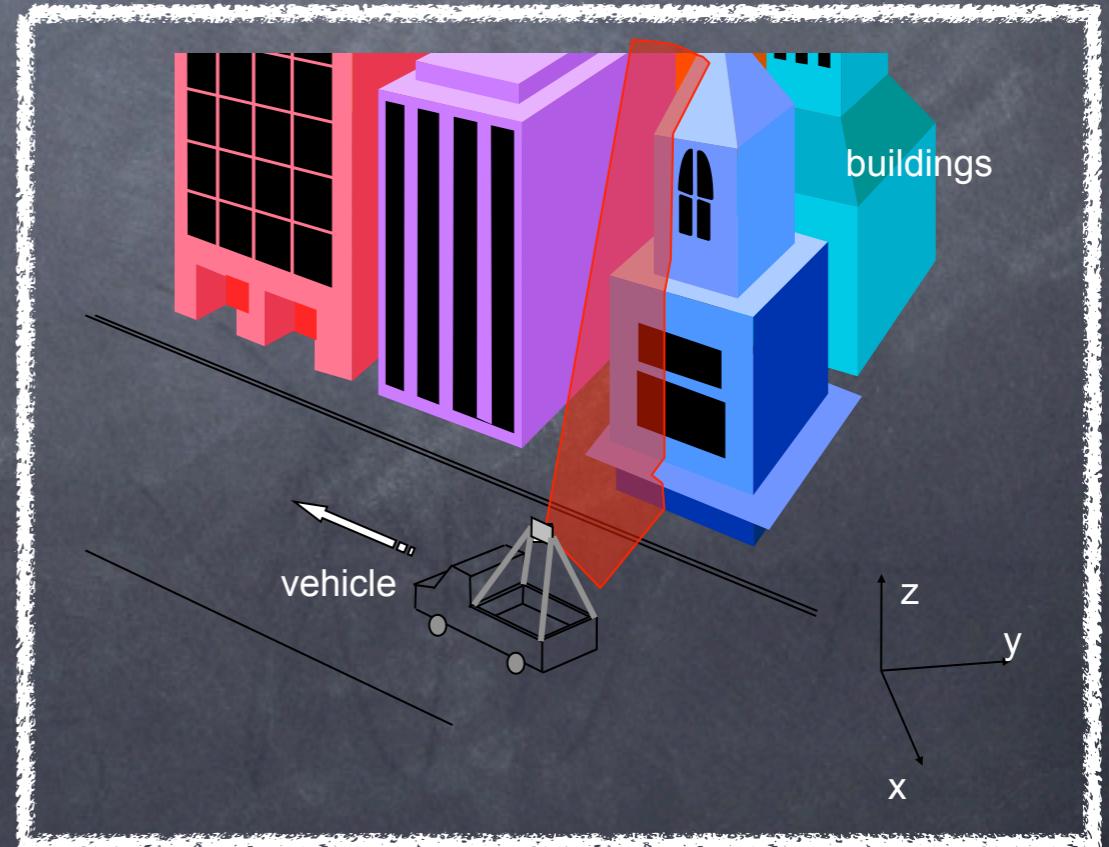
- ⦿ pickup truck
- ⦿ 2 fast 2D laser scanners
- ⦿ sync'd digital camera
- ⦿ heading sensor
- ⦿ true ground speed sensor



# Ground-based Modeling Drive-by Scanning

continuous data acquisition  
from ground-level while driving

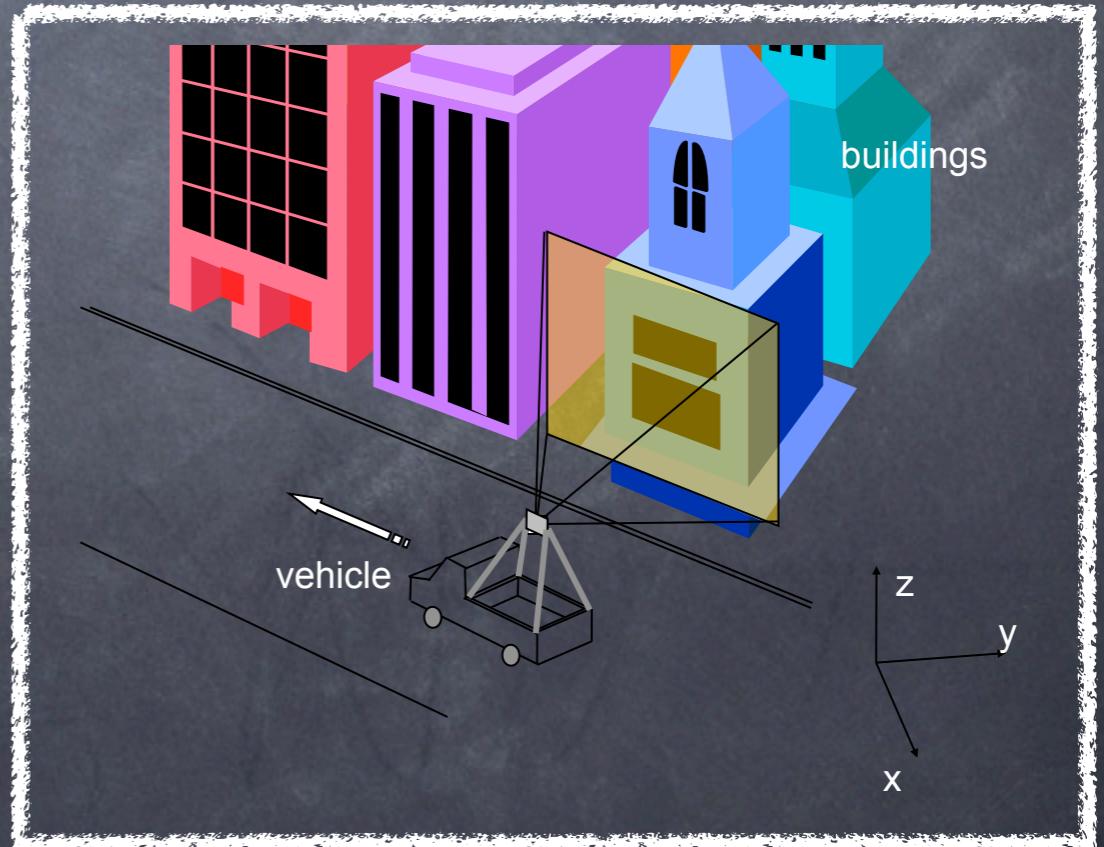
vertical 2D laser scanner  
to acquire the geometry:  
shape of complete  
building facades



# Ground-based Modeling Drive-by Scanning

continuous data acquisition  
from ground-level while driving

synchronized digital  
camera to acquire the  
texture

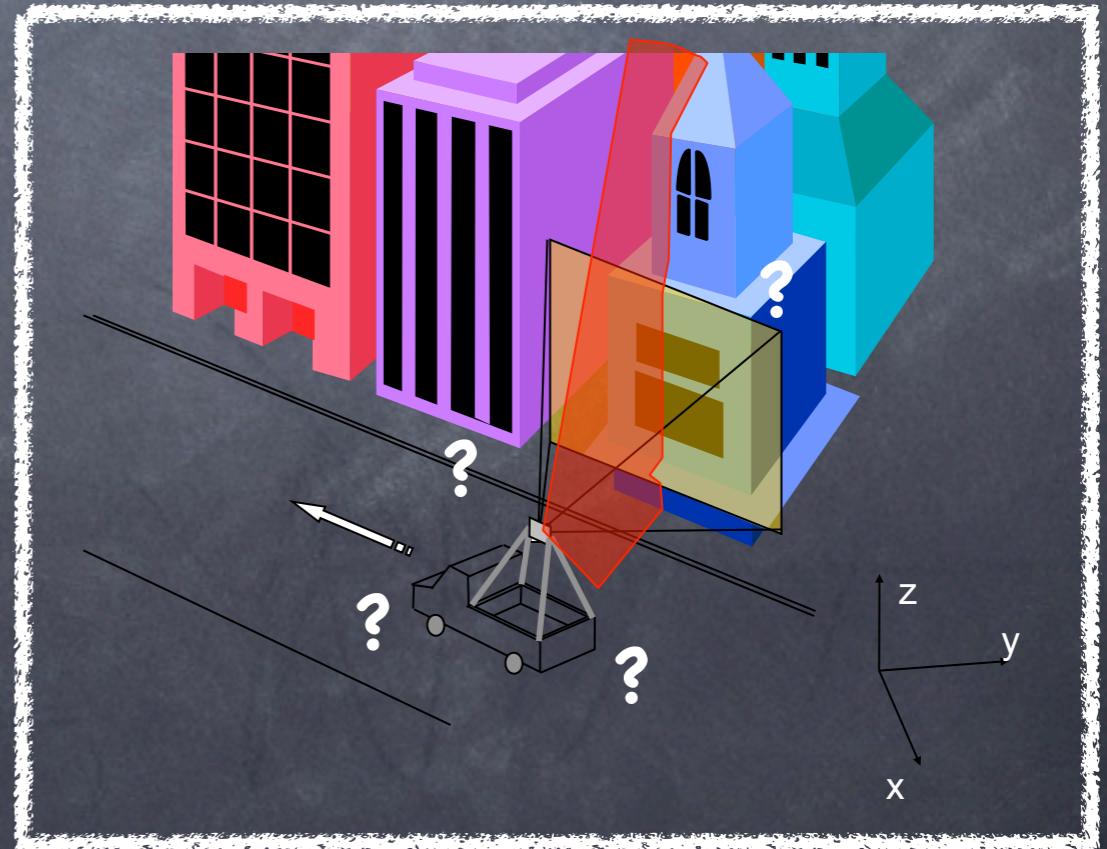


# Ground-based Modeling Drive-by Scanning

continuous data acquisition  
from ground-level while driving

problem: localization?

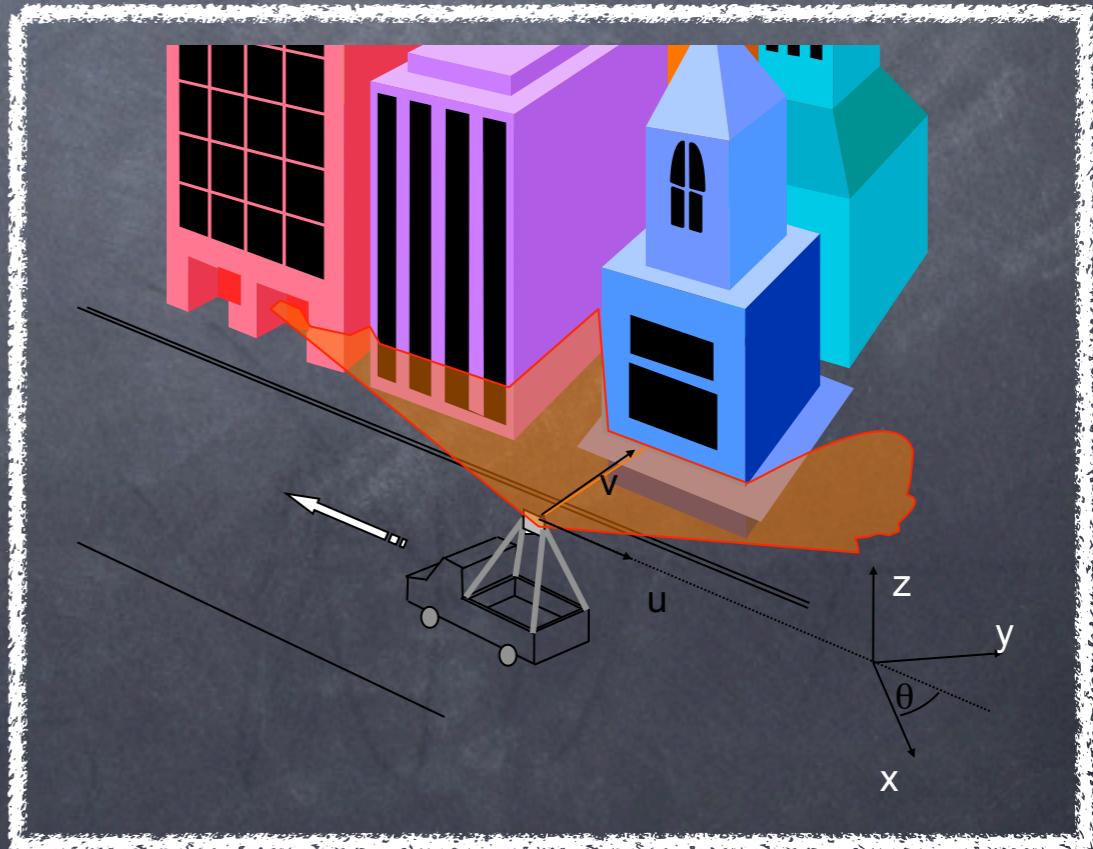
position and orientation of  
truck and its sensor unit  
need to be accurately  
determined



# Ground-based Modeling Drive-by Scanning

continuous data acquisition  
from ground-level while driving

horizontal 2D laser  
scanner to localize the  
truck: pose estimation



# Ground-based Modeling Drive-by Scanning

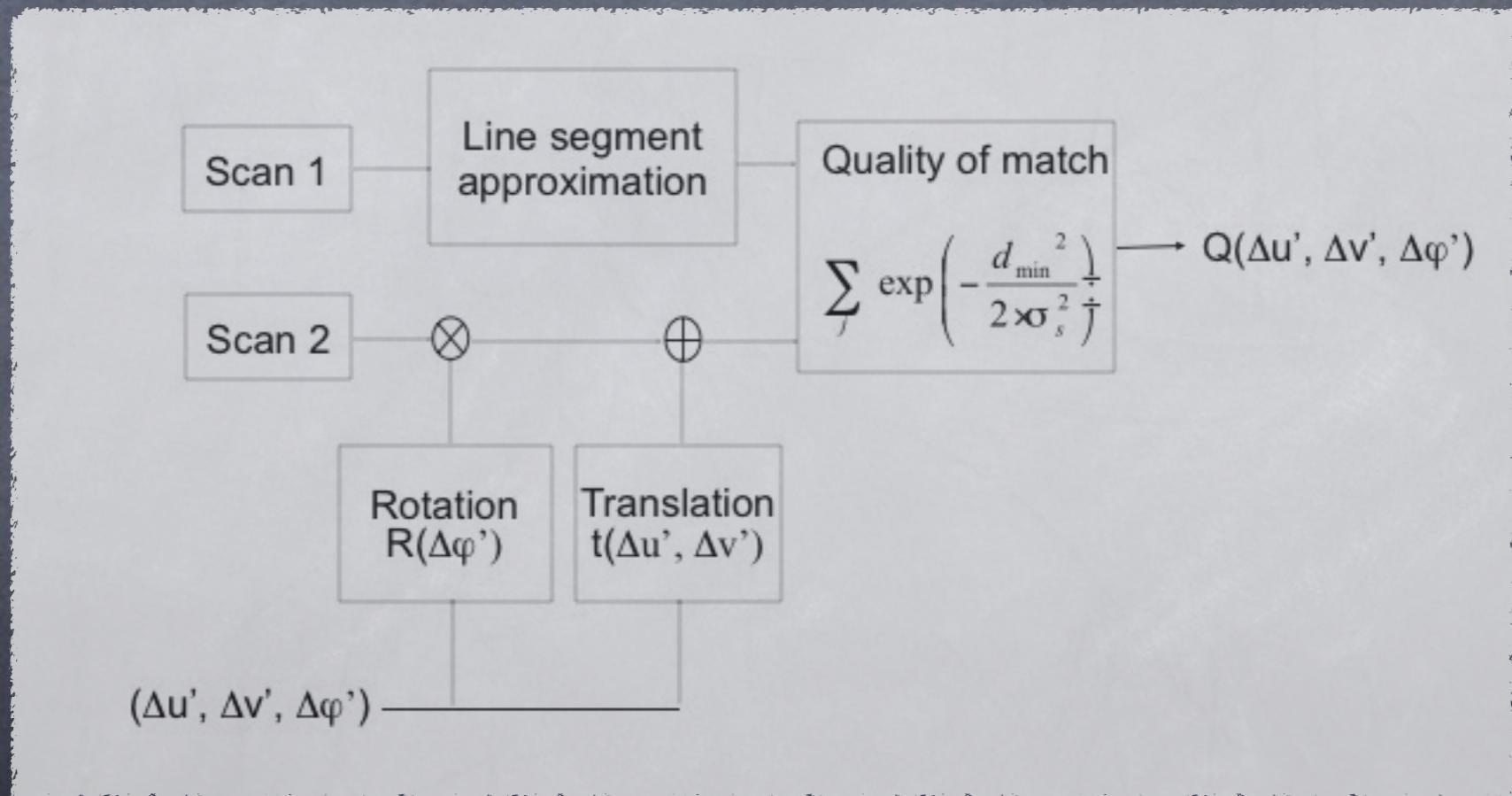
continuous data acquisition  
from ground-level while driving

- scanners and cameras are sync'd by trigger signals
- heading sensor is used to determine orientation
- TGSS provides non-contact speed measurements.

# Ground-based Modeling

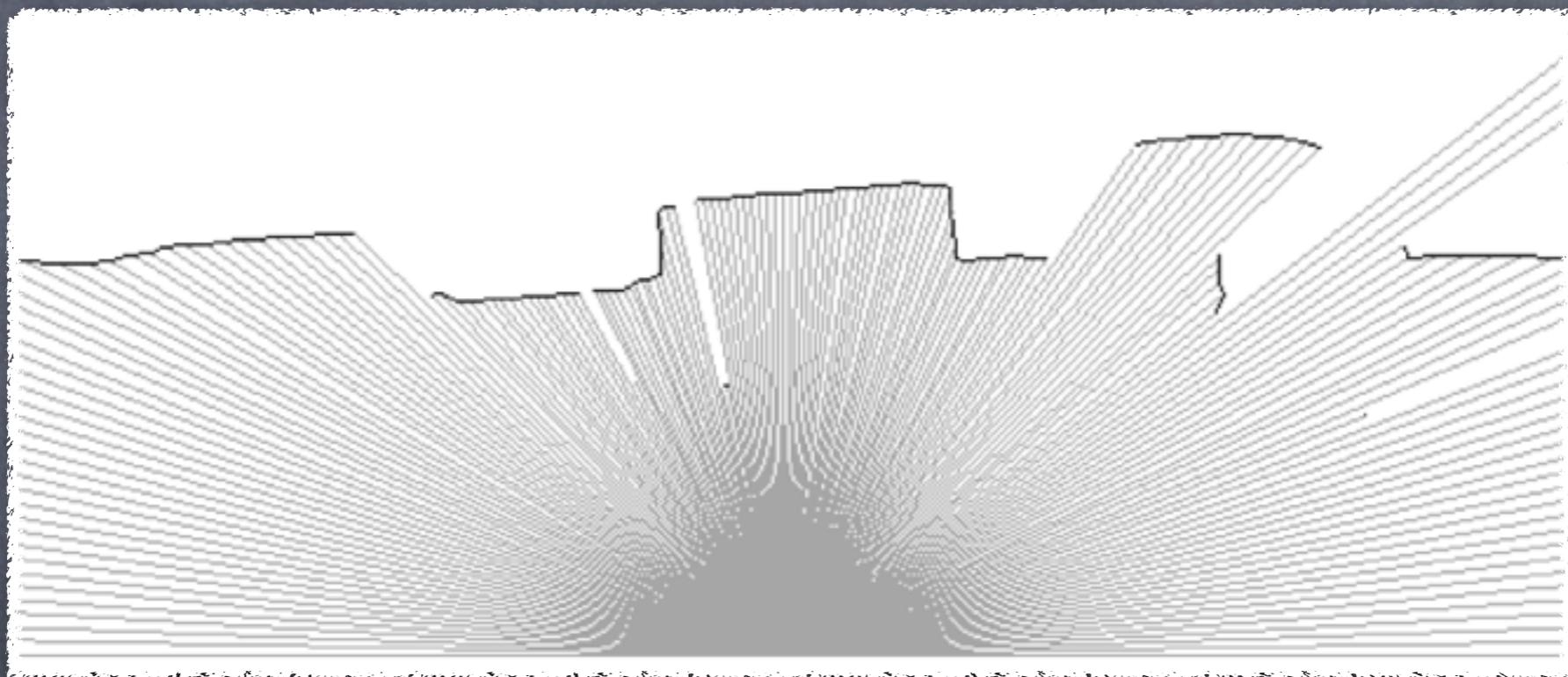
- ⦿ data-acquisition system
- ⦿ scan matching
- ⦿ path computation
- ⦿ point cloud generation
- ⦿ mesh generation
- ⦿ texture mapping

# Scan Matching Flowchart



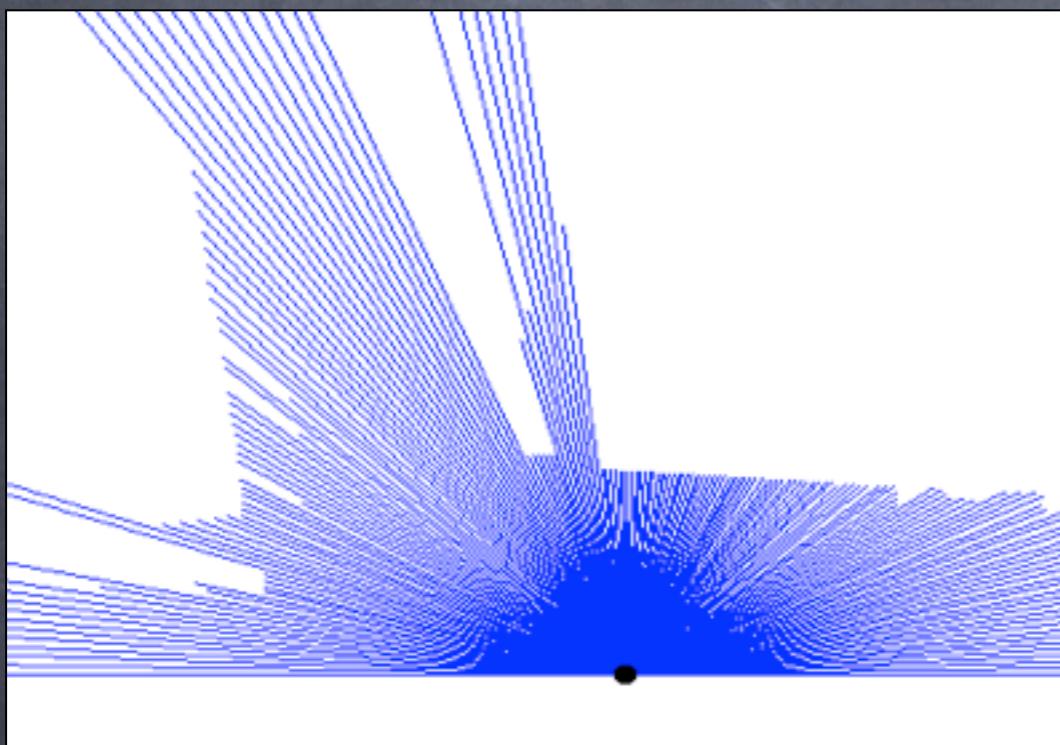
# Scan Matching

## Line Segment Approximation of Reference Scan

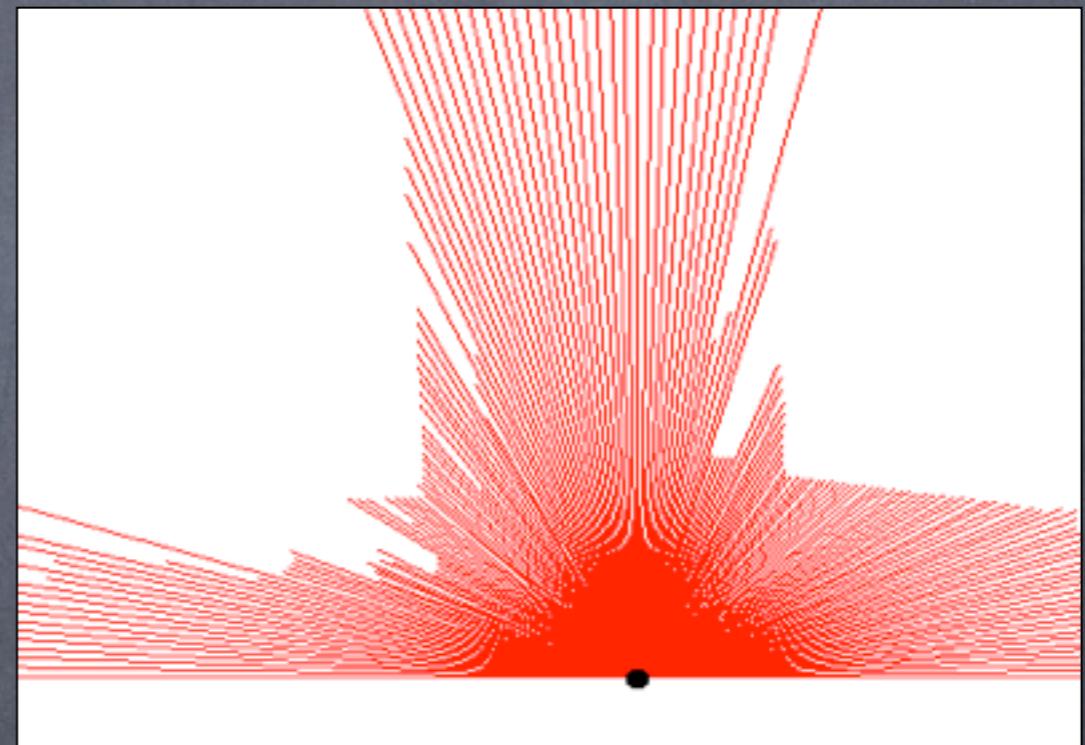


connect successive points to form a line; if the difference between their depth values does not exceed a depth dependent threshold

# Scan Matching Pose Estimation



$t = t_0$

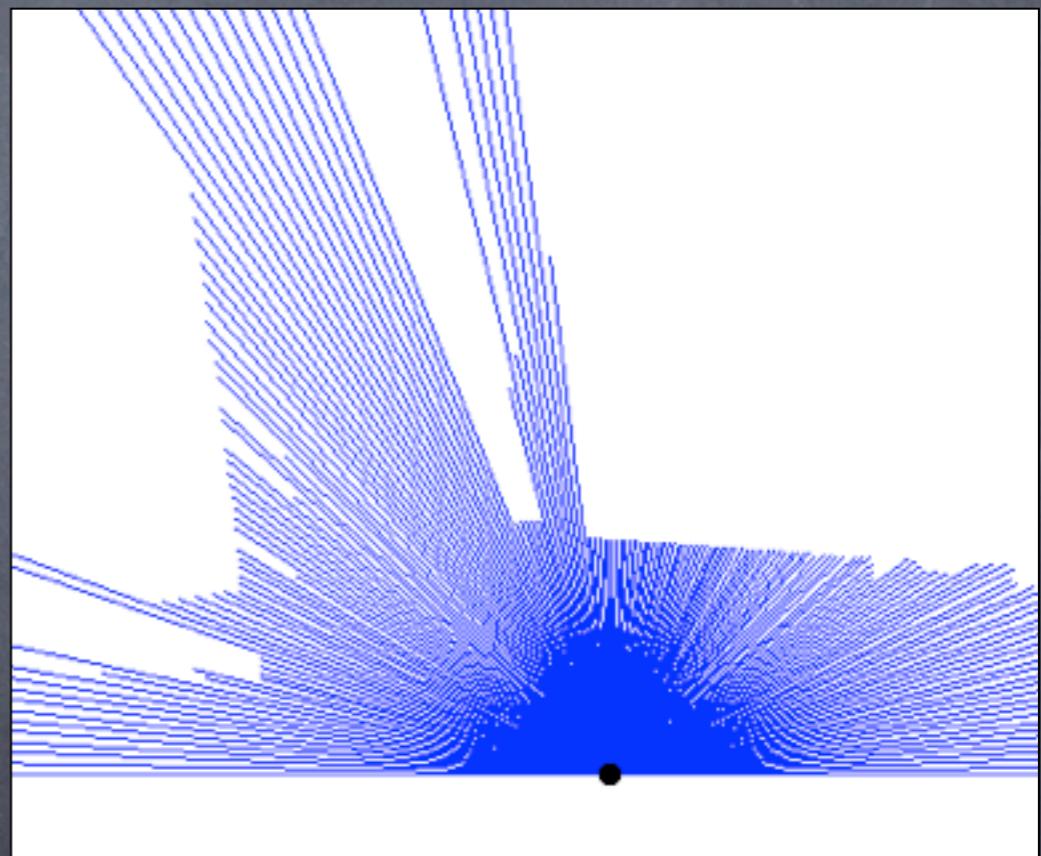


$t = t_1$

horizontal scans at two different time instances

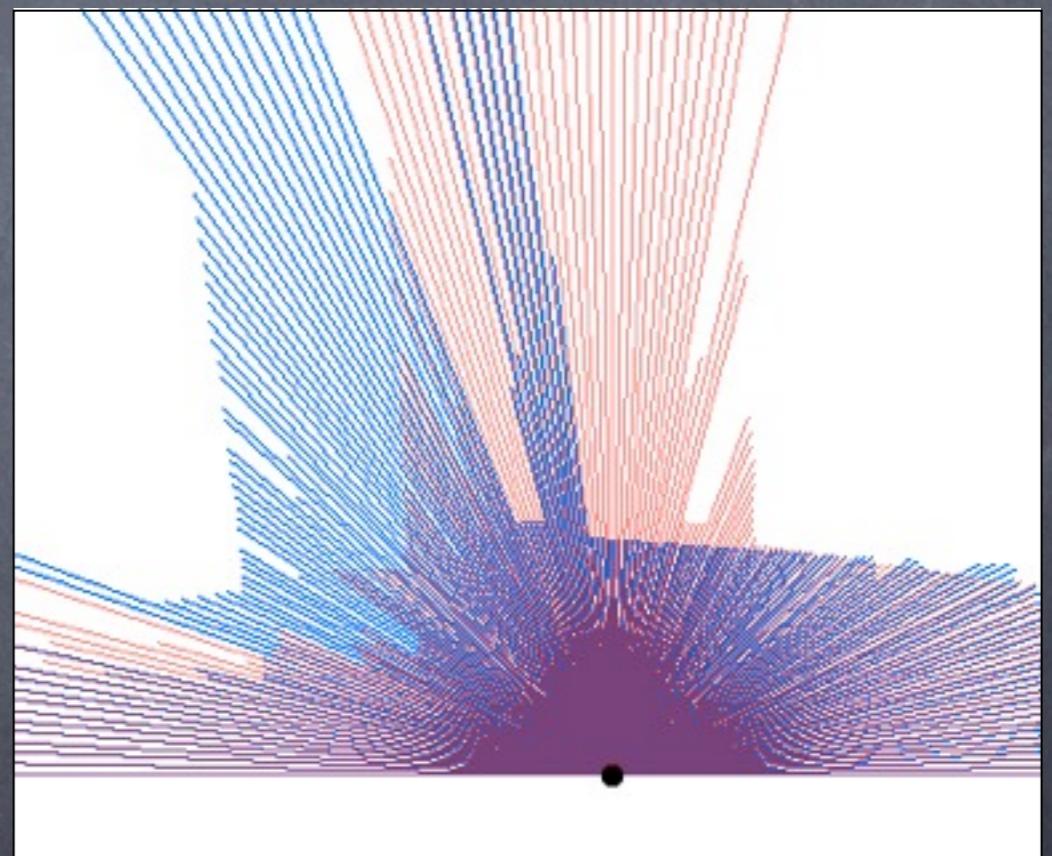
# Scan Matching Pose Estimation

- reference scan:  $t = t_0$



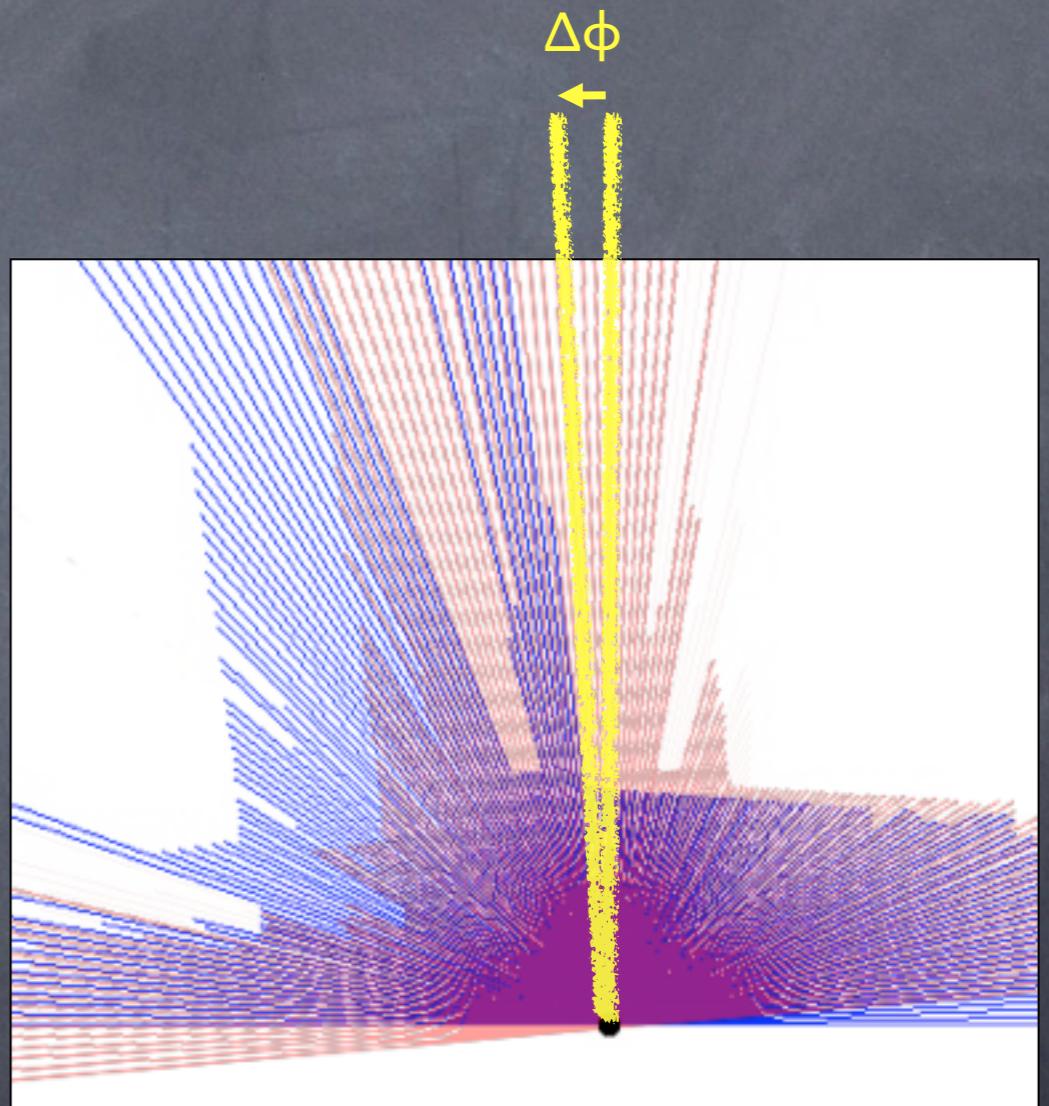
# Scan Matching Pose Estimation

- reference scan:  $t = t_0$
- second scan:  $t = t_1$



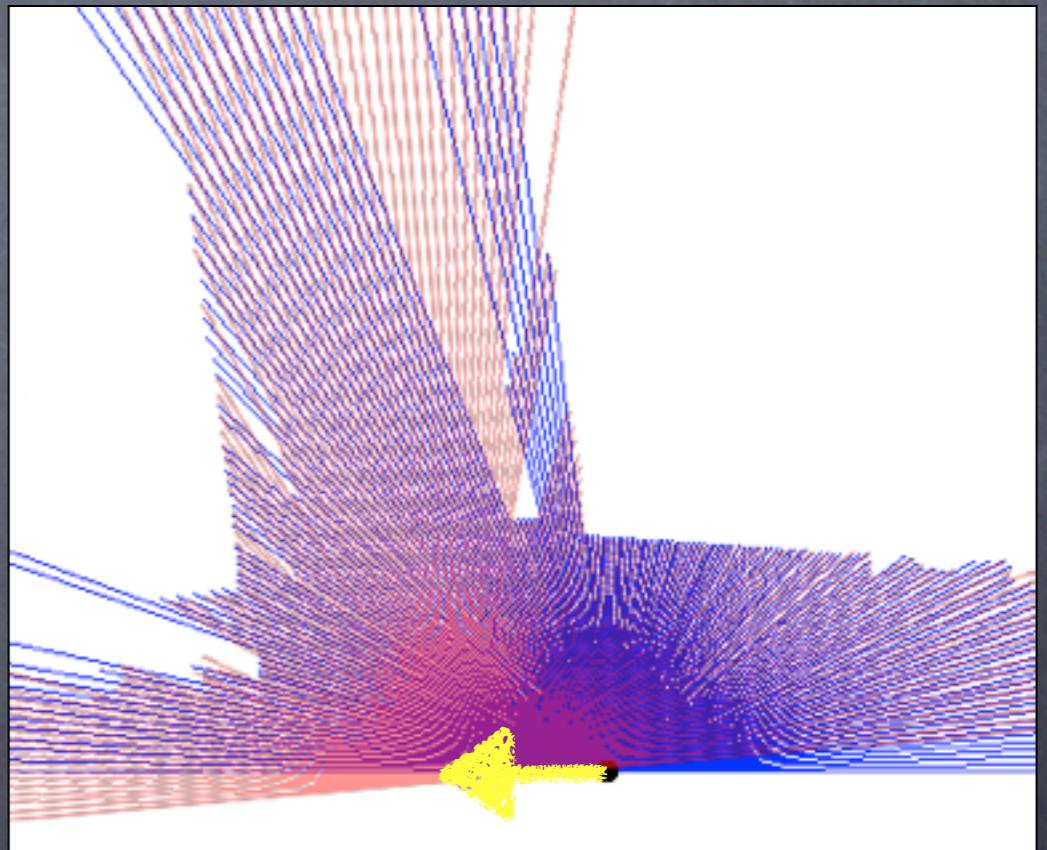
# Scan Matching Pose Estimation

- reference scan:  $t = t_0$
- second scan:  $t = t_1$
- rotate by  $\Delta\phi$



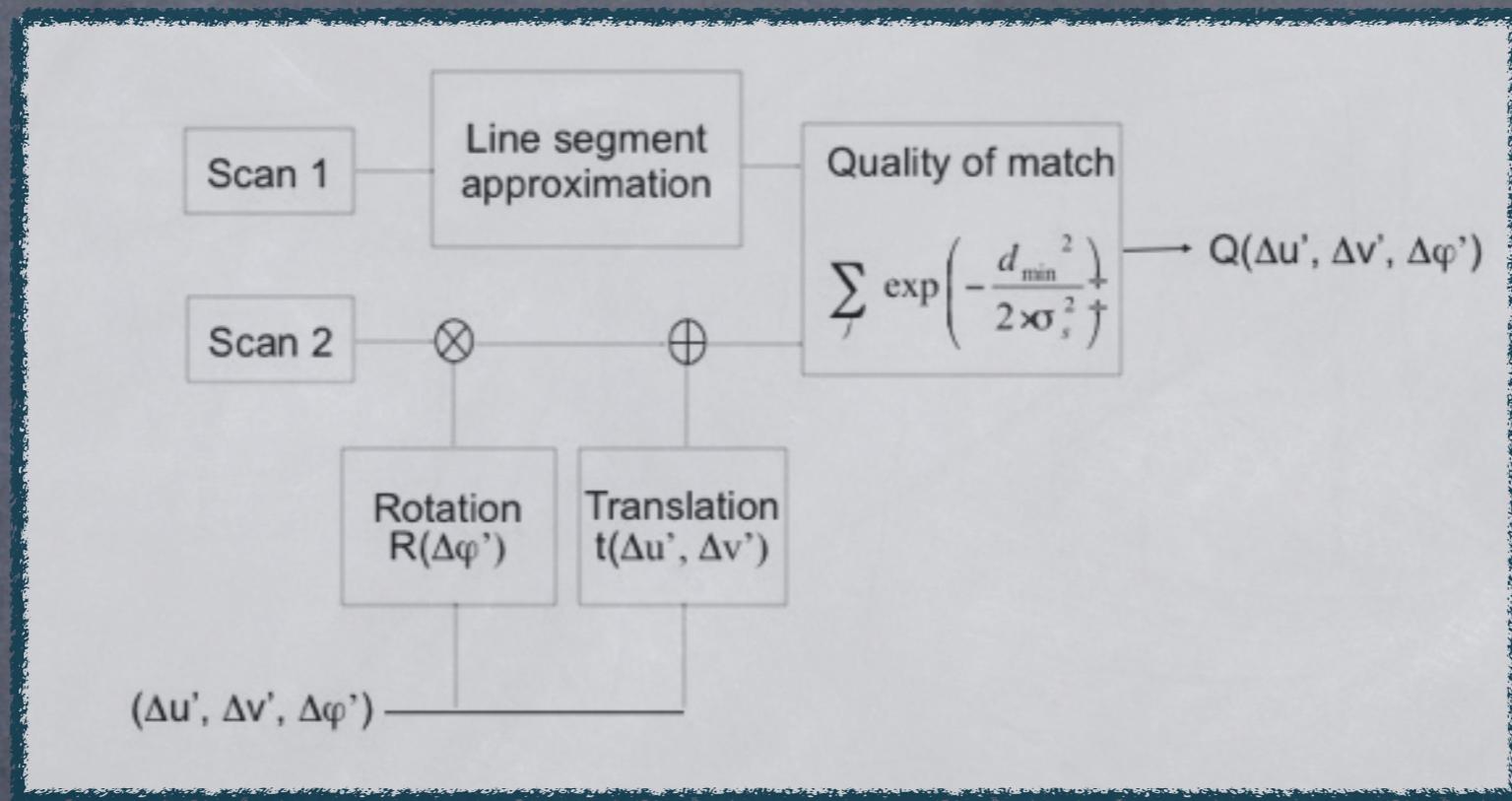
# Scan Matching Pose Estimation

- reference scan:  $t = t_0$
- second scan:  $t = t_1$
- rotate by  $\Delta\phi$
- translate  $(\Delta u, \Delta v)$



$(\Delta u, \Delta v)$

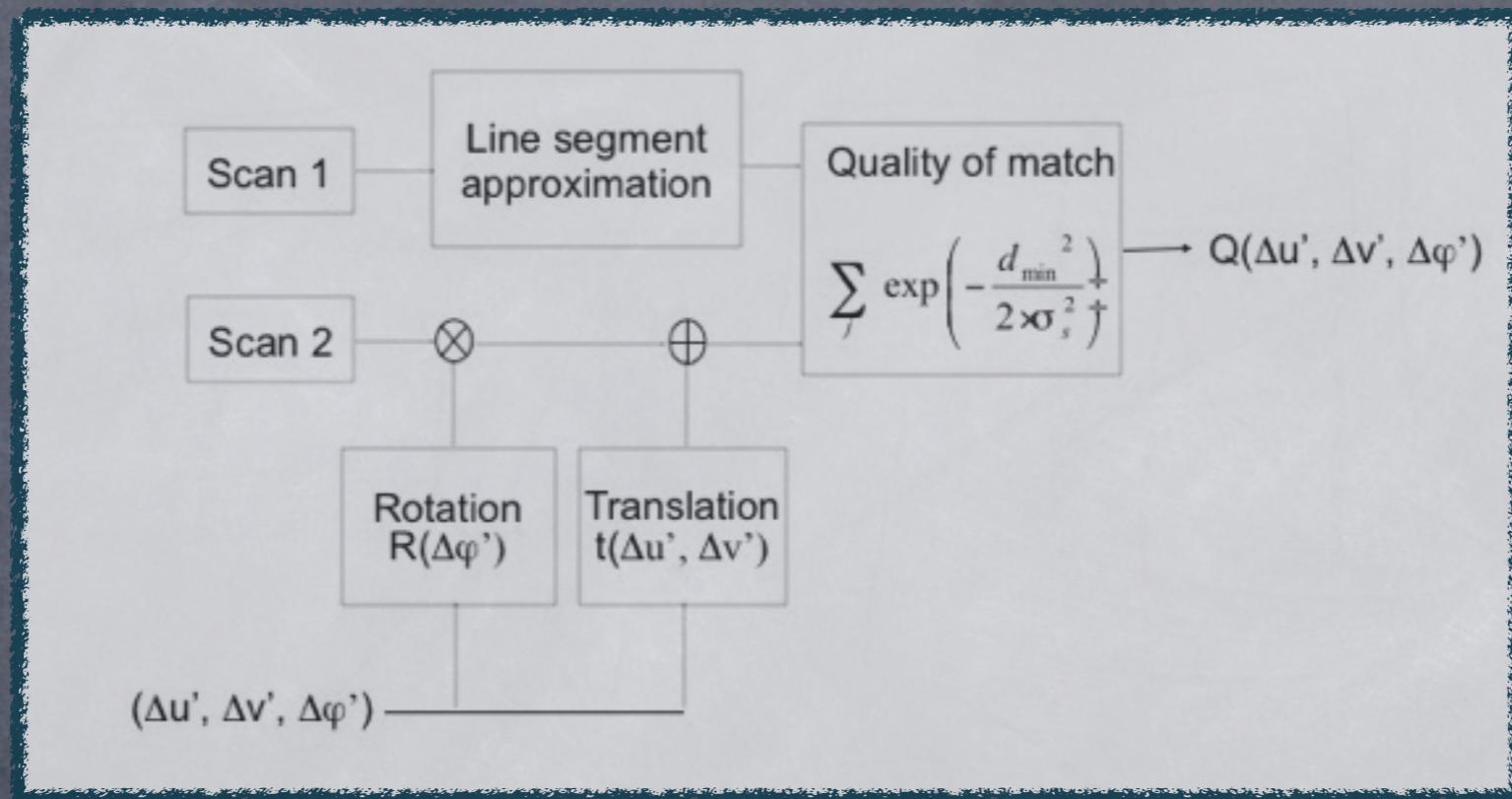
# Scan Matching Post Processing



transform the point of second scan in coordinate system of reference scan.

$$\vec{p'}_j (\Delta u, \Delta v, \Delta \varphi) = R(\Delta \varphi) \cdot \vec{p}_j + \vec{t} .$$

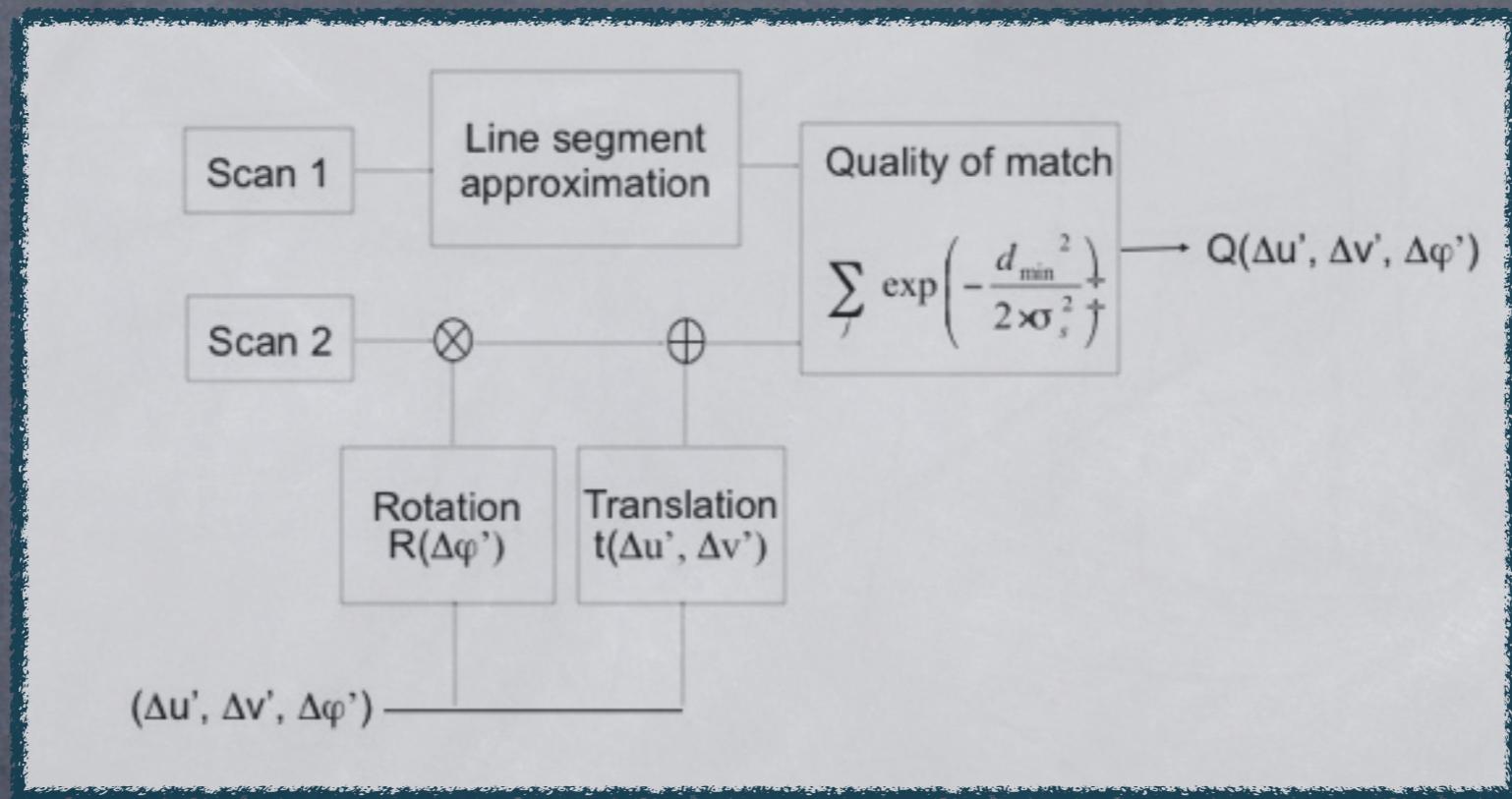
# Scan Matching Post Processing



find the closest line segment  
corresponding to transformed  
points in the reference scan

$$d_{\min}(\vec{p'}_j(\Delta u, \Delta v, \Delta \varphi)) = \min_i \{ d(\vec{p'}_j, l_i) \}.$$

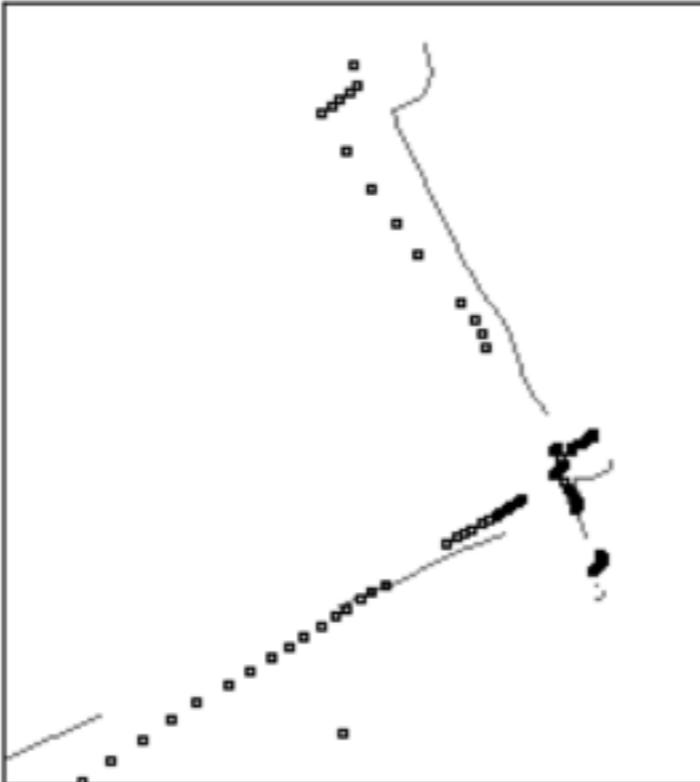
# Scan Matching Post Processing



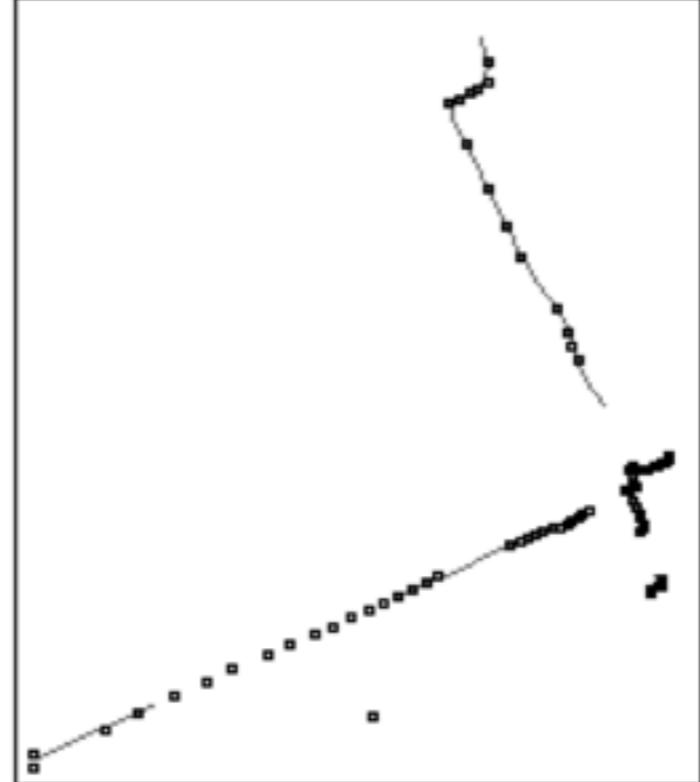
maximize the quality of  
alignment function

$$Q(\Delta u, \Delta v, \Delta \varphi) = \sum_j \exp\left(-\frac{d_{\min}(\vec{p}_j'(\Delta u, \Delta v, \Delta \varphi))^2}{2 \cdot \sigma_s^2}\right)$$

# Scan Matching Results?



before match



after match

# Ground-based Modeling

- ⦿ data-acquisition system
- ⦿ scan matching
- ⦿ path computation
- ⦿ point cloud generation
- ⦿ mesh generation
- ⦿ texture mapping

# Path Computation

- ⦿ a global pose estimate is needed to generate 3D point clouds from vertical scans.
- ⦿ problem?
  - ⦿ no sensor used to provide a global pose estimate.  
(how about using GPS?)
- ⦿ solution?
  - ⦿ compute traversed path by successively adding relative pose estimates in the local coordinate system.  
(previously determined from horizontal scan matching)
  - ⦿ use TGSS and heading sensor for consistency checks.

# Path Computation

$$\begin{aligned}x_{i+1} &= x_i + \Delta u_i \cdot \cos(\theta_i + \Delta\varphi_i) - \Delta v_i \cdot \sin(\theta_i + \Delta\varphi_i) \\y_{i+1} &= y_i + \Delta u_i \cdot \sin(\theta_i + \Delta\varphi_i) + \Delta v_i \cdot \cos(\theta_i + \Delta\varphi_i) \\ \theta_{i+1} &= \theta_i + \Delta\varphi_i\end{aligned}$$

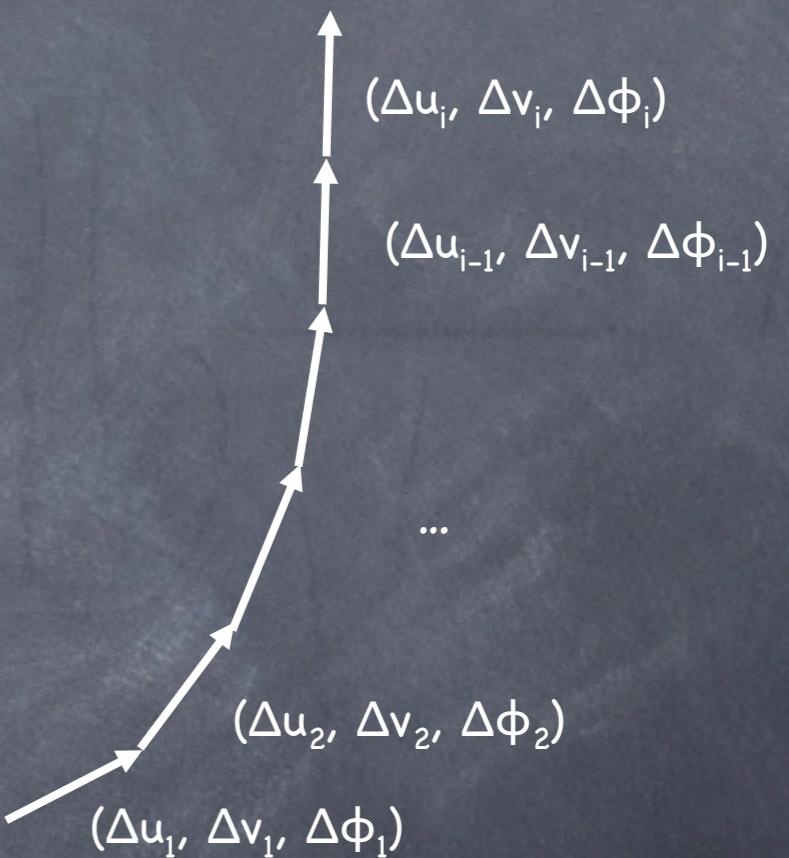
consistency checks :-

If there is a significant change in the computed displacement vector, we accept it as valid if both

- successive vectors change similarly
- the speedometer indicates a speed change

If there is a significant change in the orientation angle, we accept the measurement if both

- succeeding angles change similarly
- the heading sensor indicates an angle change



concatenate  
steps to path

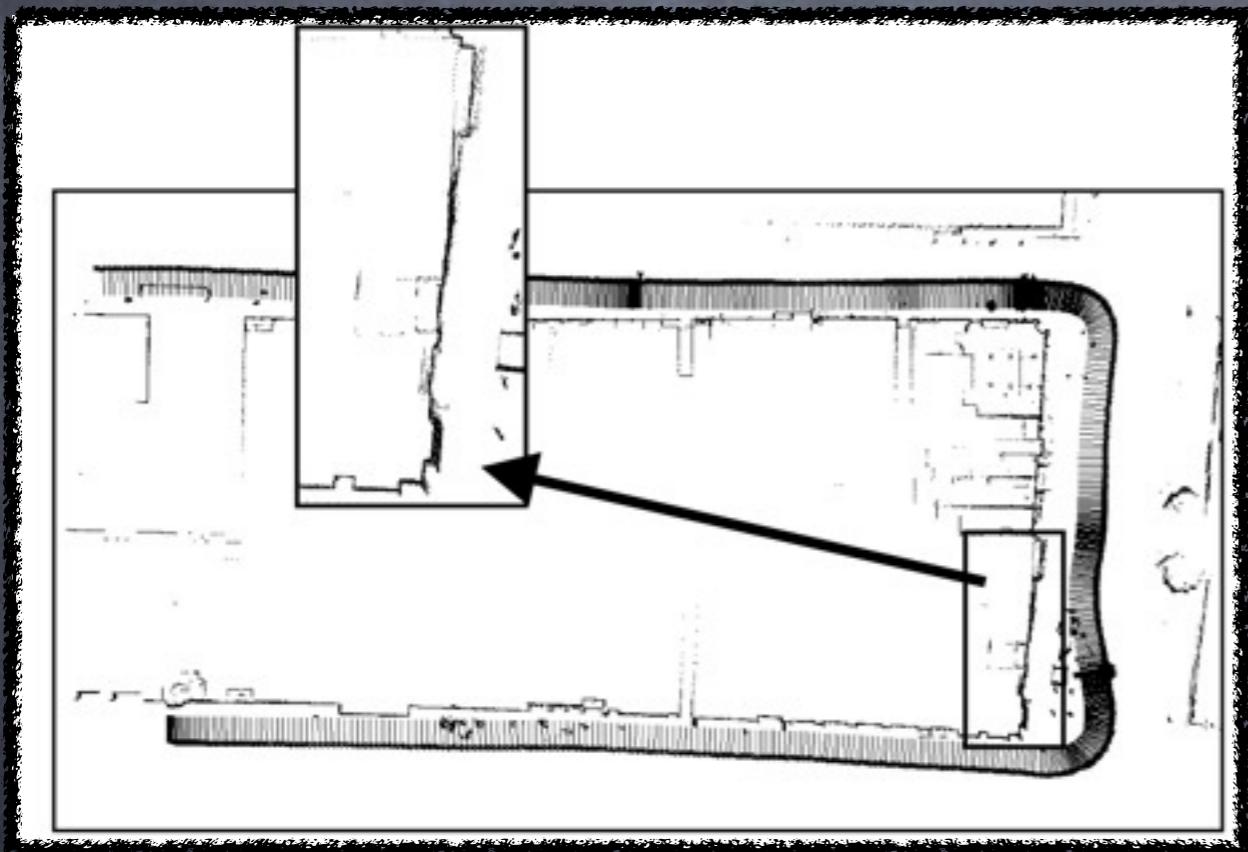
# Path Computation Tradeoff

- errors in the estimate accumulate in each iteration
- subsample the scan by a large factor to recover the path in few steps?

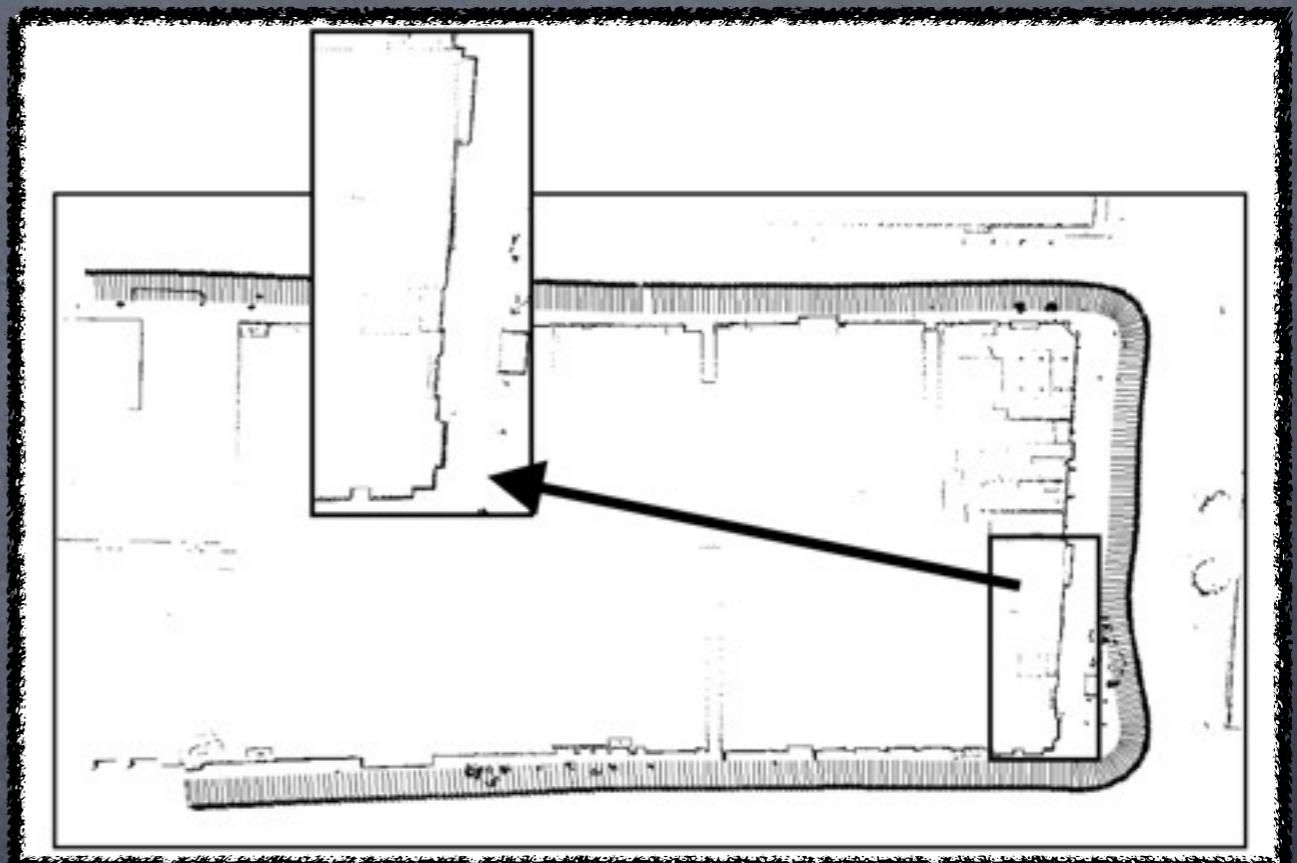
or

- subsample the scans by a small factor to maximize overlap for accurate match?
- solution? use an adaptive subsampling factor.

# Path Computation Results

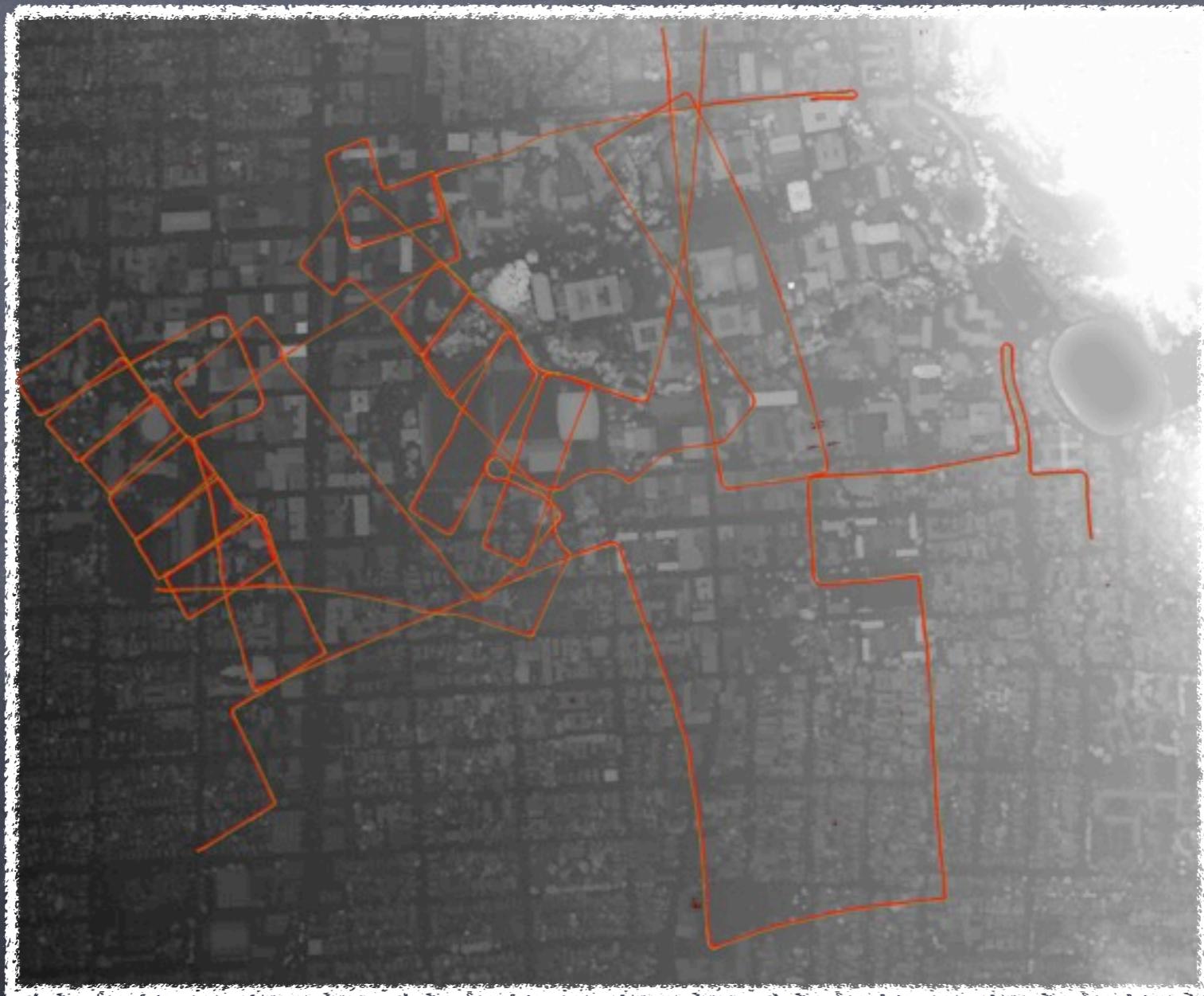


path computed using a  
fixed subsampling factor



path computed using  
adaptive subsampling

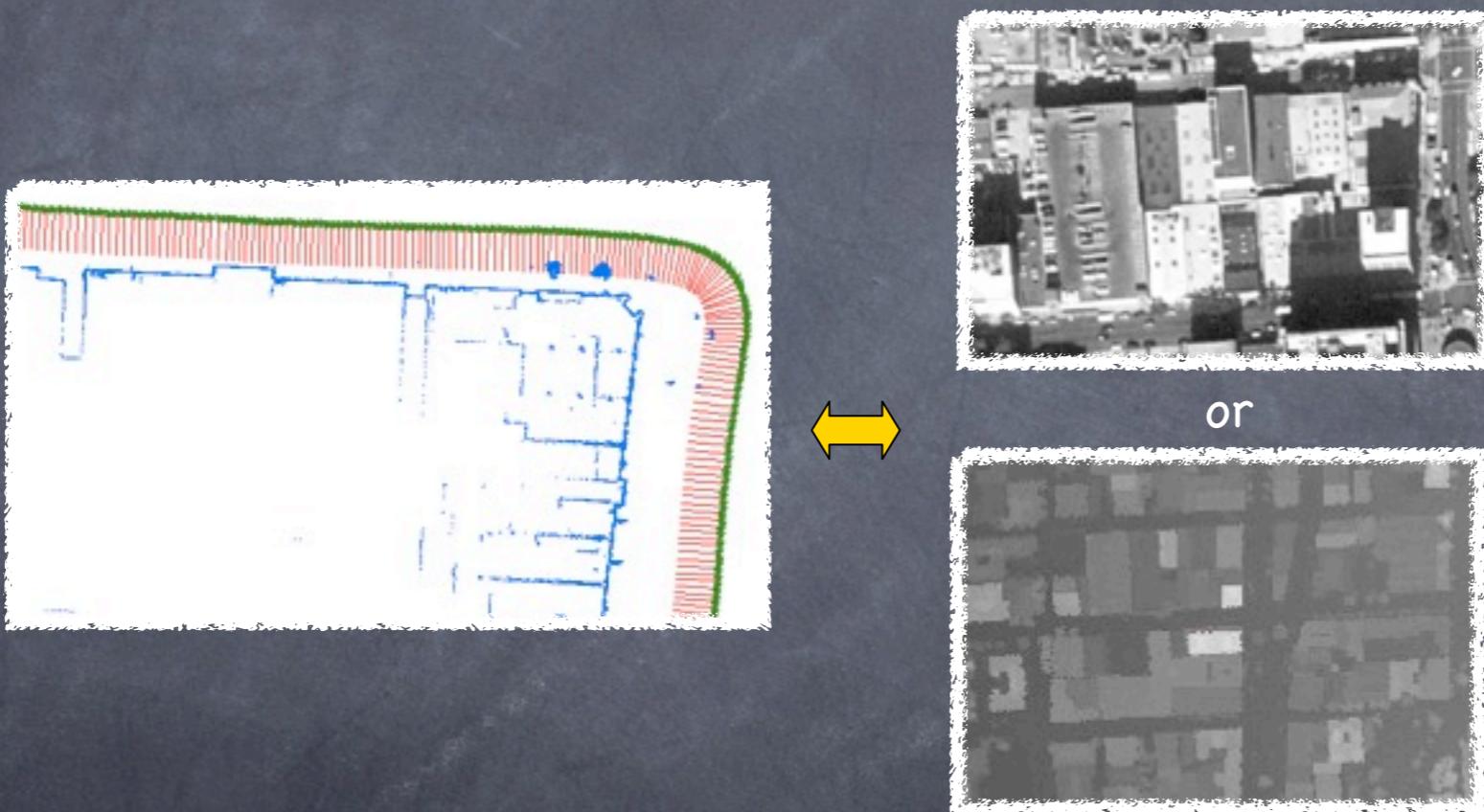
(beyond the paper)  
Problem?



needs global correction

(beyond the paper)

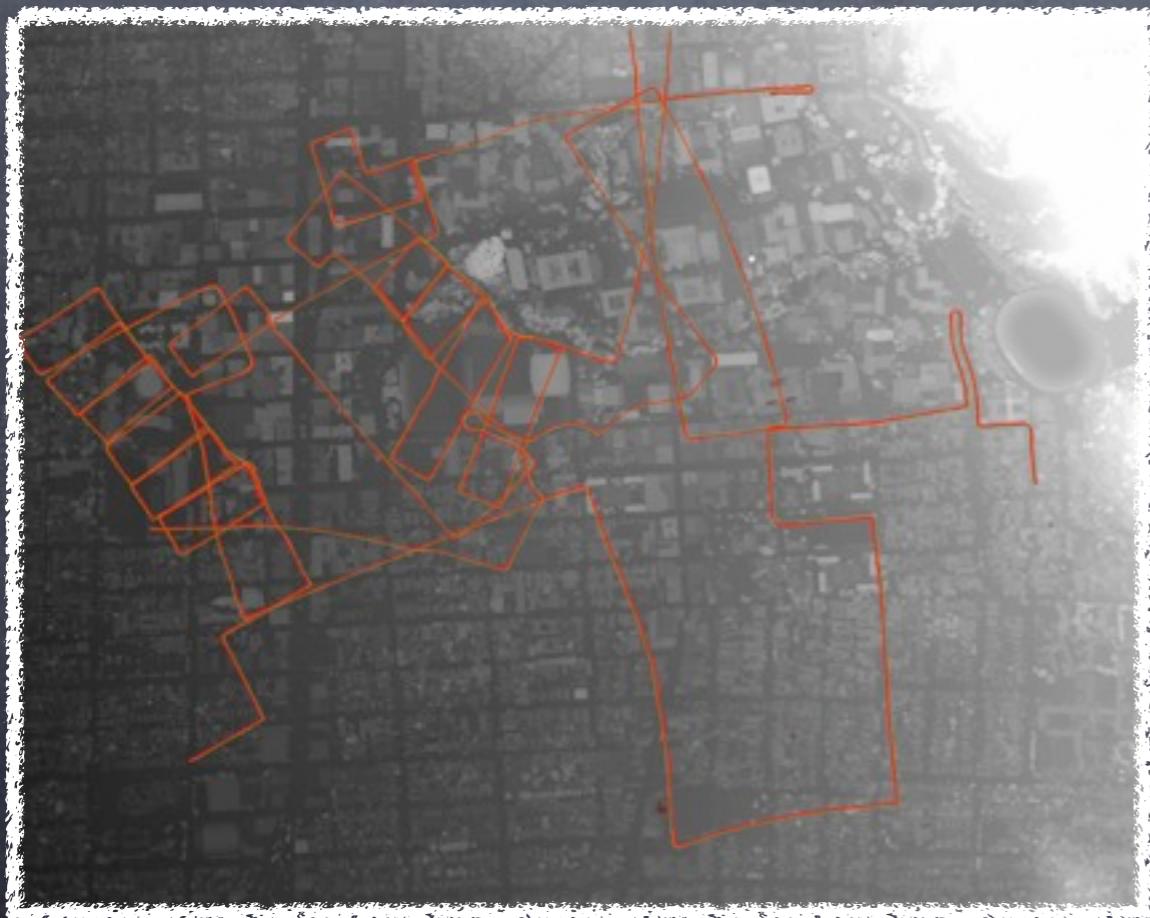
# Global Correction with Monte Carlo Localization



register ground-based laser scans with  
edge map from airborne-images or digital  
surface models (DSM)

(beyond the paper)

# Monte-Carlo Localization



before MCL correction

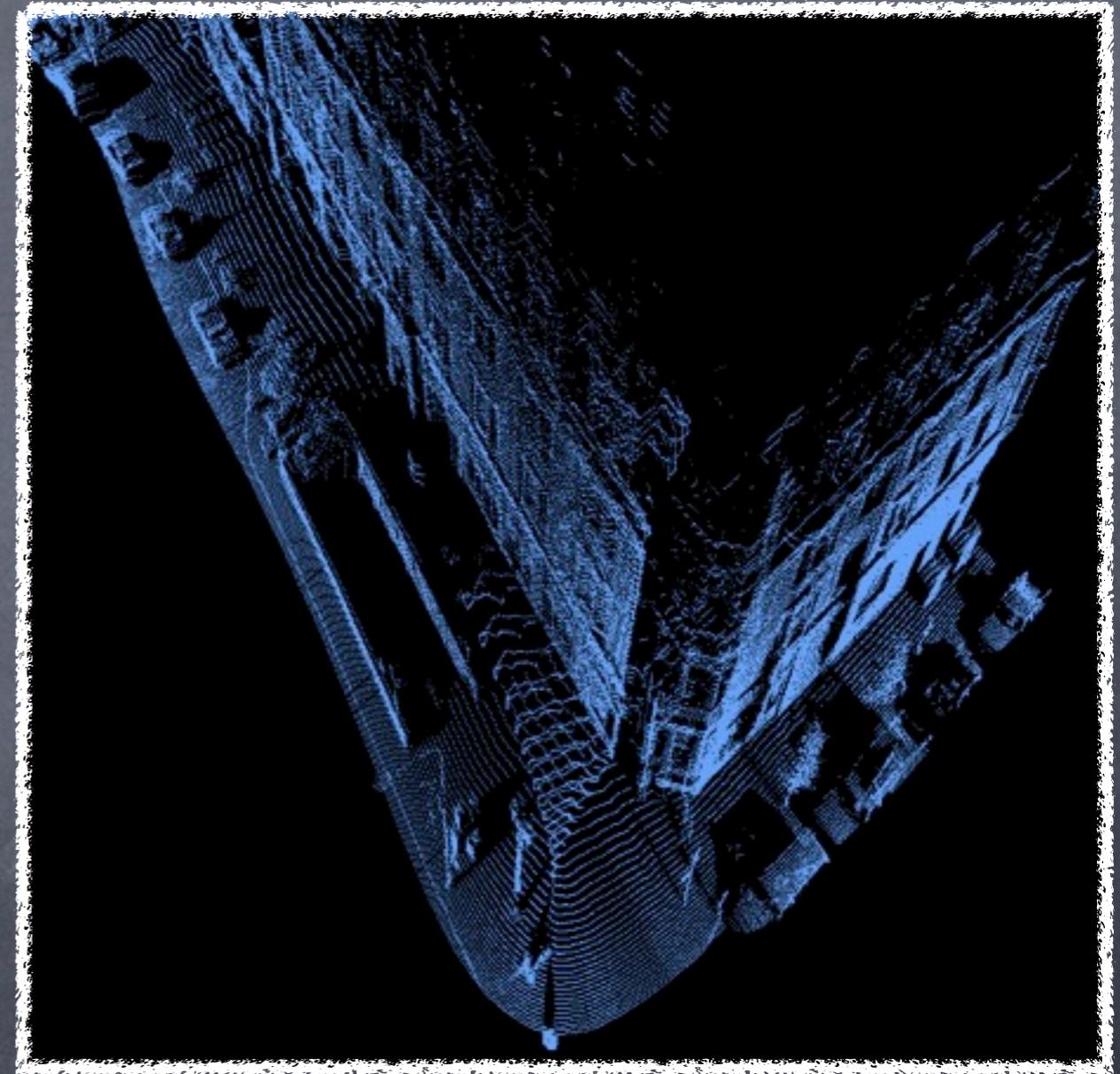
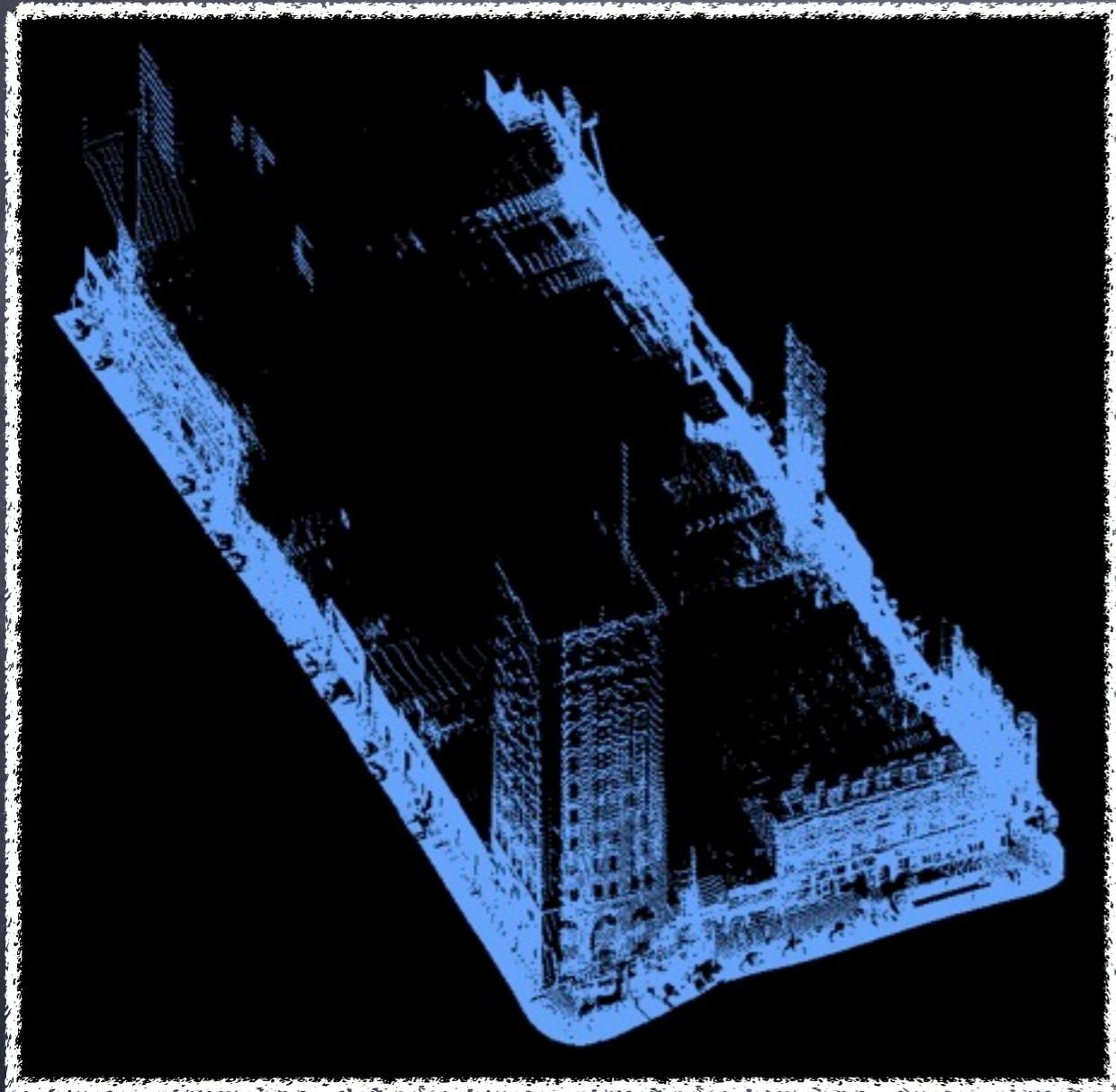


after MCL correction

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# Point Cloud Generation



stack the vertical scans at appropriate  
distances from each other to get point cloud.

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(beyond the paper)

# Mesh Generation

- structure vertical scan vertices as a 2D grid
- connect adjacent vertices if depth discontinuity < threshold



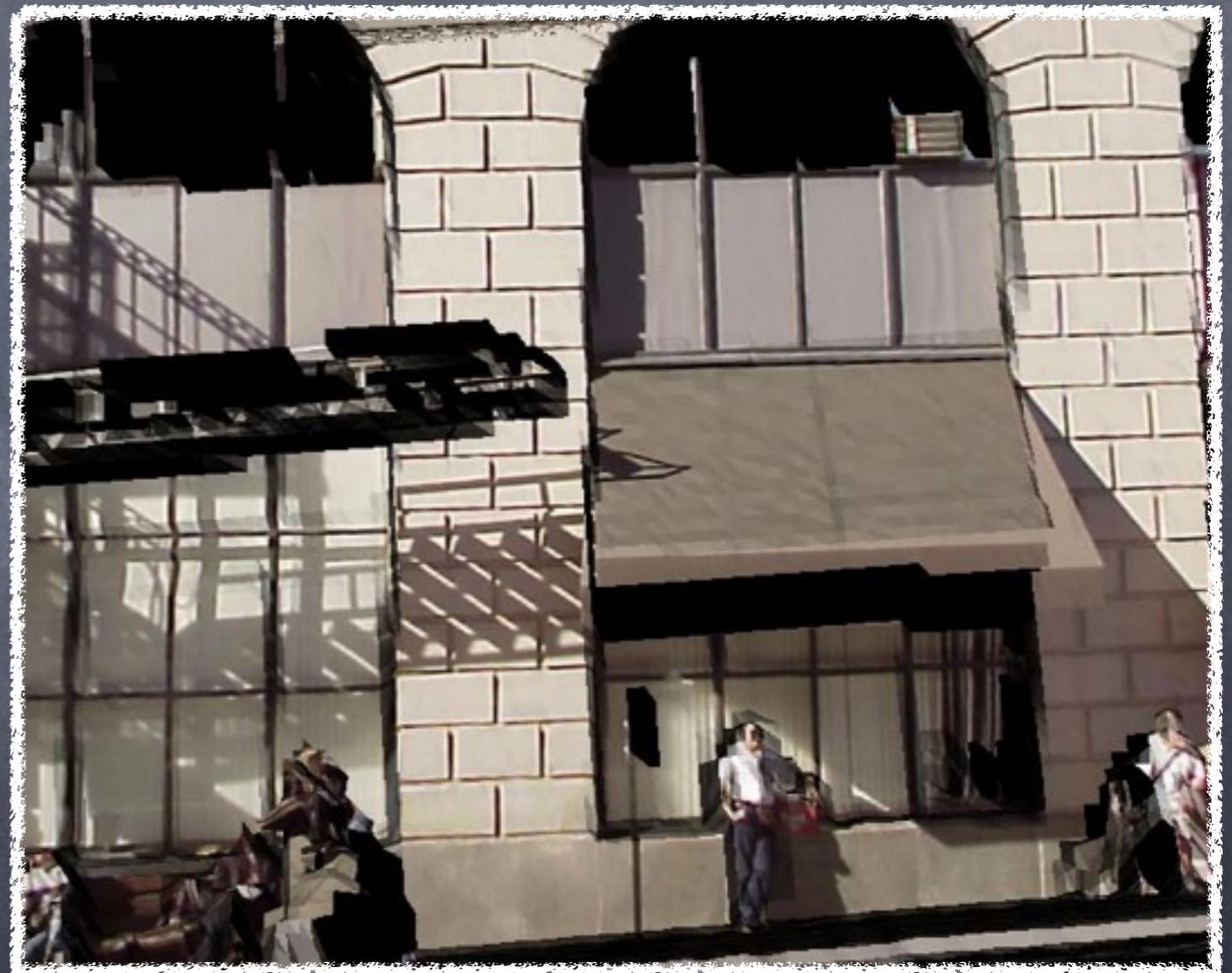
# Ground-based Modeling

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(beyond the paper)

# Texture Mapping

- assign 2D texture from digital camera images to corresponding 3D mesh triangles.



# References

- Fruh and Zakhor. Fast 3D model generation in urban environments. Multisensor Fusion and Integration for Intelligent Systems, 2001. MFI 2001. International Conference on (2002) pp. 165-170