



# UAS-Based LiDAR Mapping

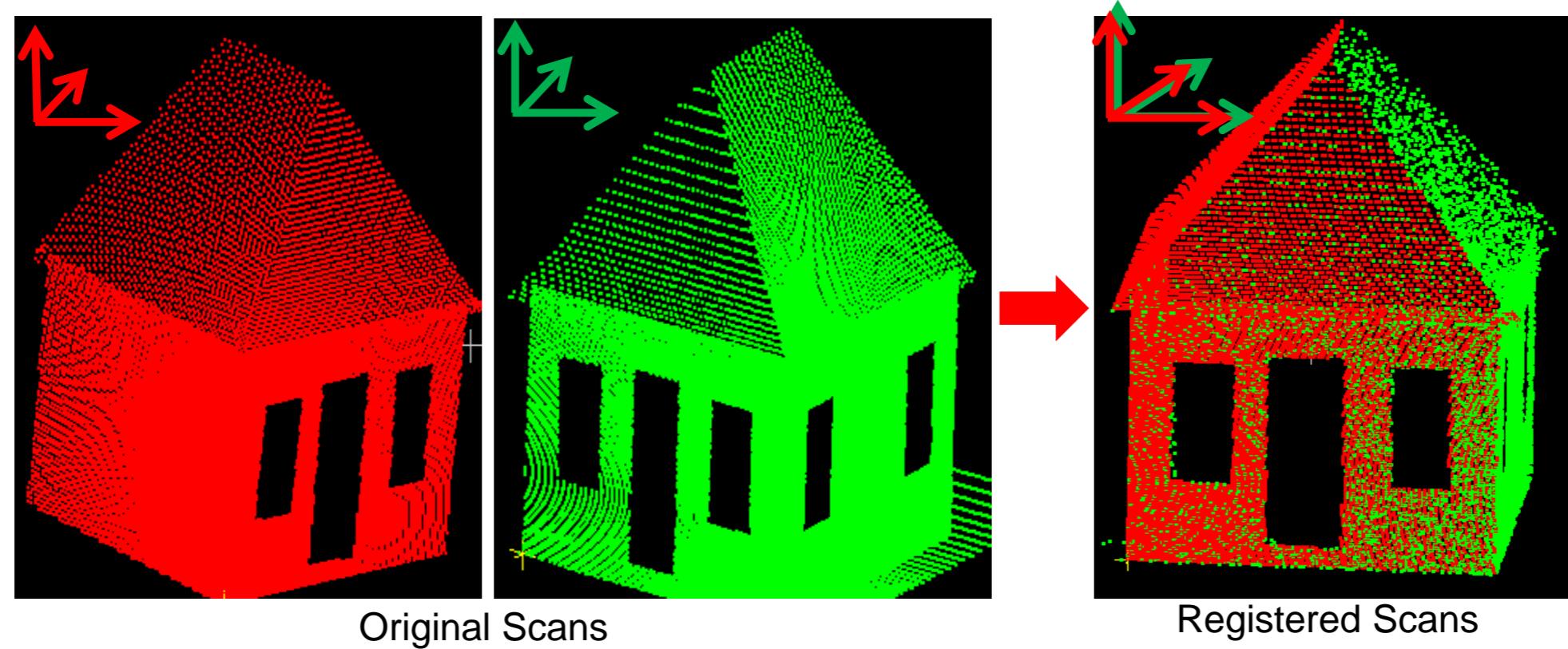
Video I



# Point Cloud Registration

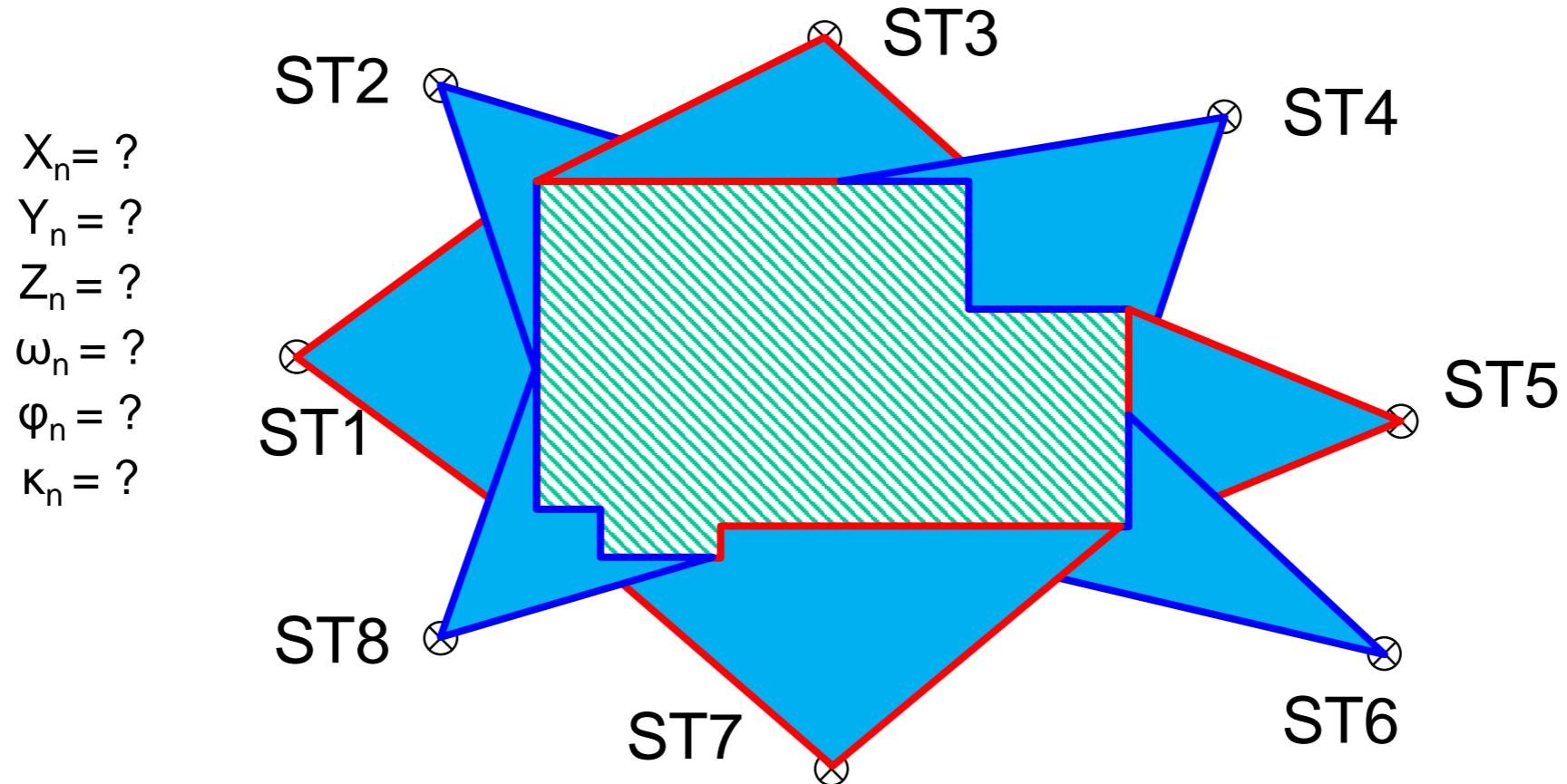
# Introduction: Registration of Static Scans

- Complex surfaces (or objects) require multiple scans with overlap for a full 3D model:
  - The separate point clouds must be registered to a common reference frame.

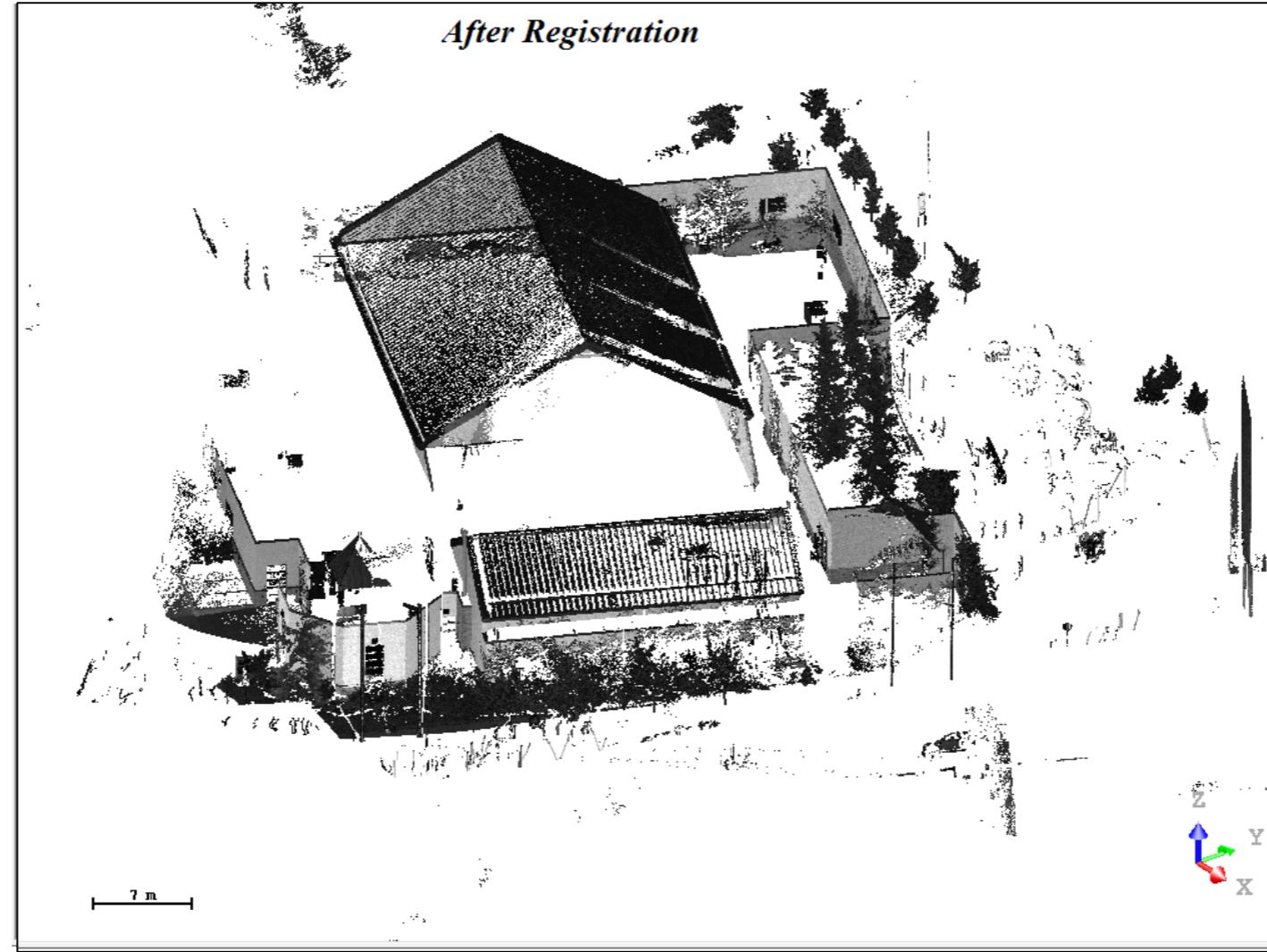


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# Introduction: Registration of Static Scans



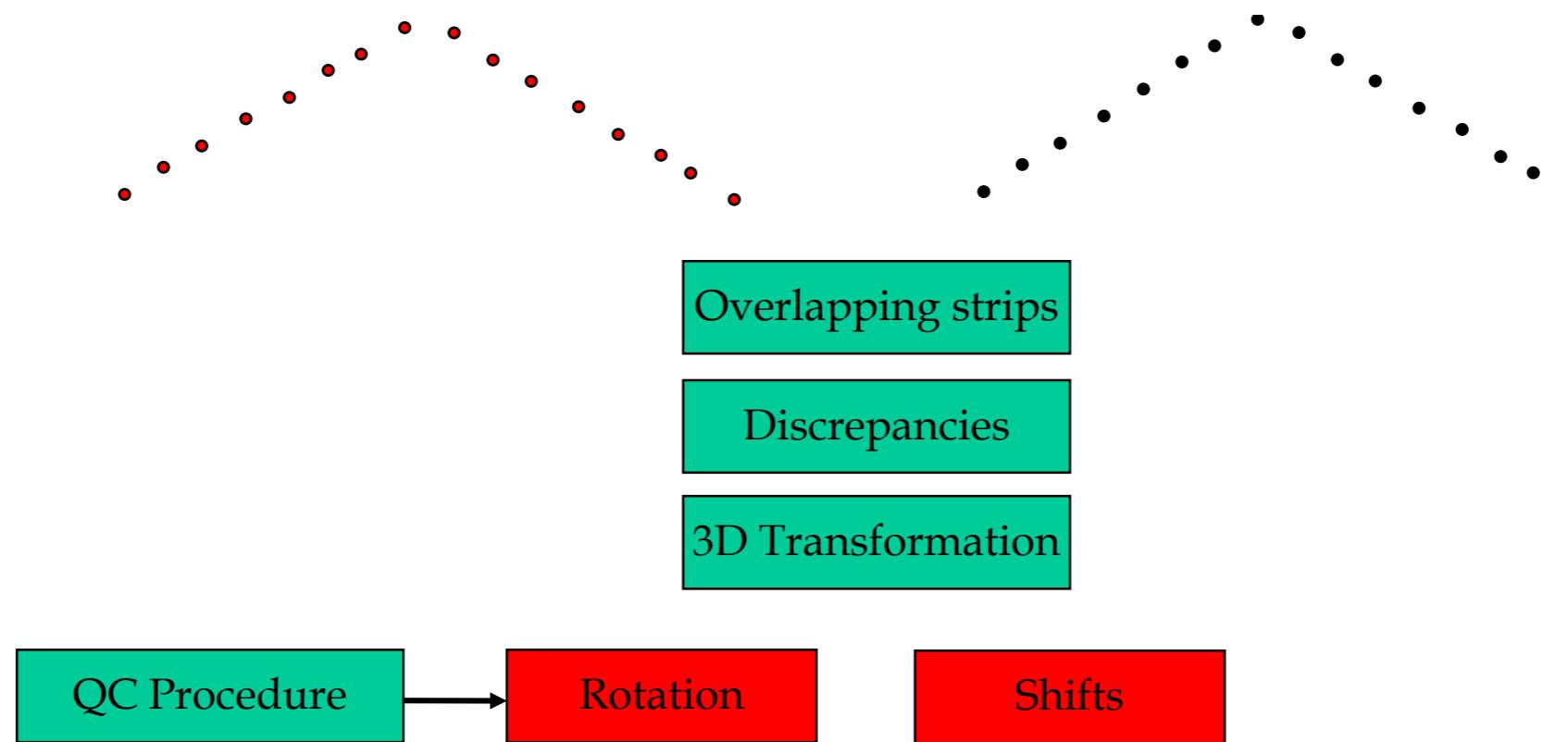
# Introduction: Registration of Mobile Scans



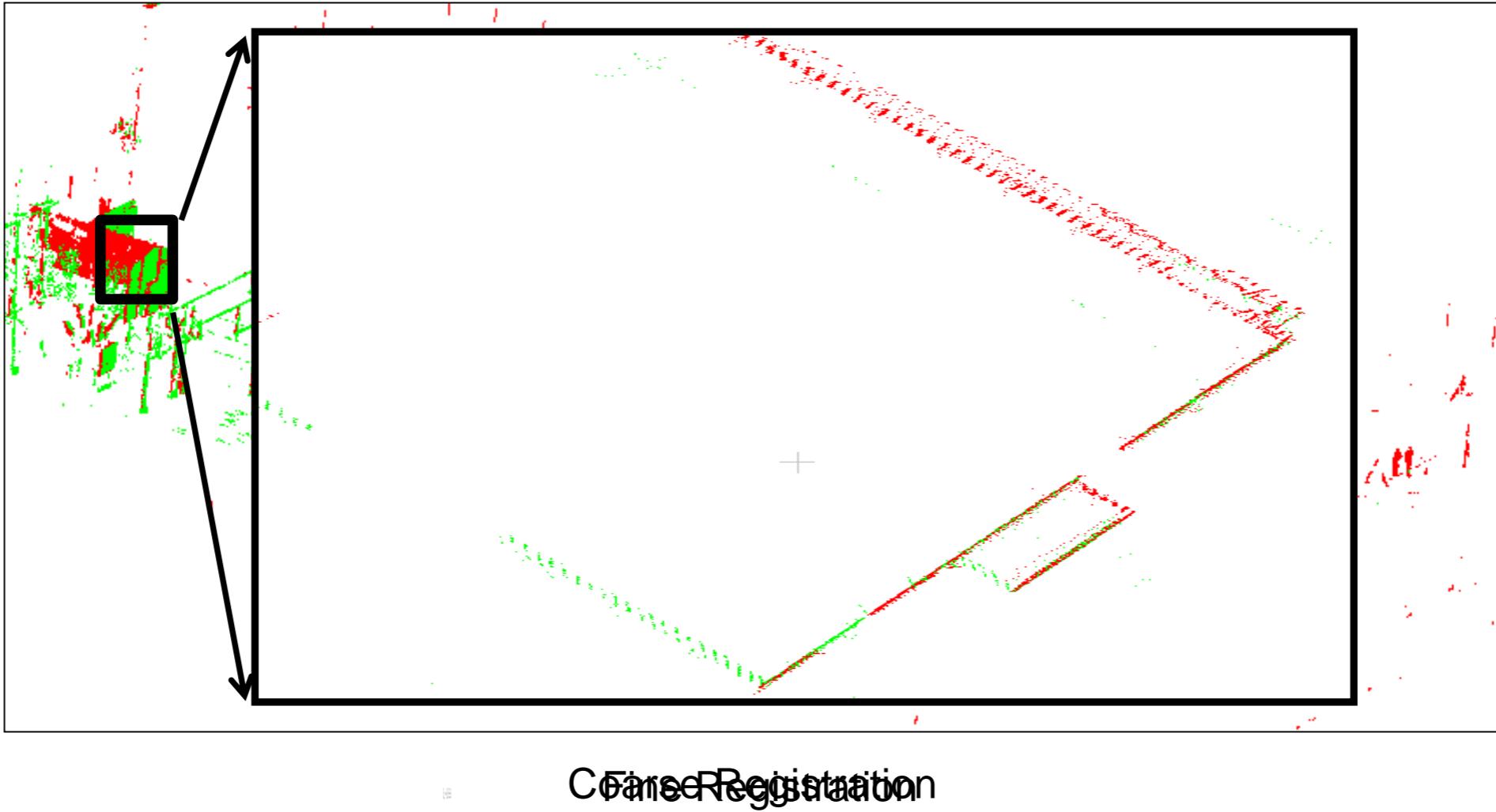
- For LiDAR data, which has been captured by a mobile system – whether terrestrial or airborne, the point cloud will be given relative to a unified coordinate system.
  - Defined by the onboard GNSS/INS unit
- **Registration is not necessary for this type of data.**
- **However, in some situations, systematic errors in the data acquisition system will lead to discrepancies among overlapping point clouds.**
- **Therefore, we might need to register overlapping mobile LiDAR data.**
  - To ensure the alignment of the different datasets, and
  - To evaluate the quality of the system performance (QC).

# Introduction: Registration of Mobile Scans

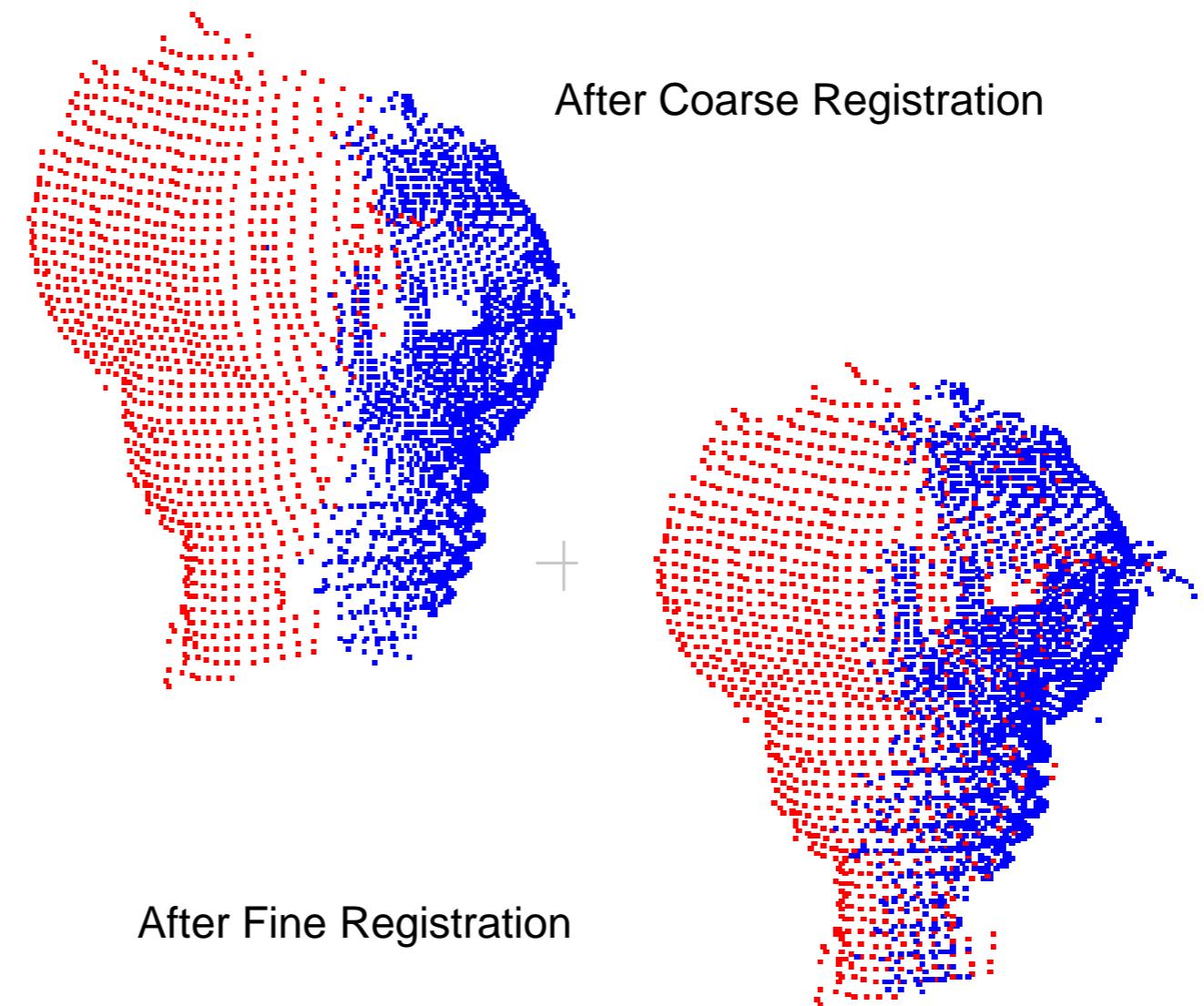
- We are interested in evaluating the registration transformation parameters between point clouds from multiple flight lines/drive runs.
  - These registration transformation parameters are used as quality control measures.



# Introduction: Coarse Vs. Fine Registration



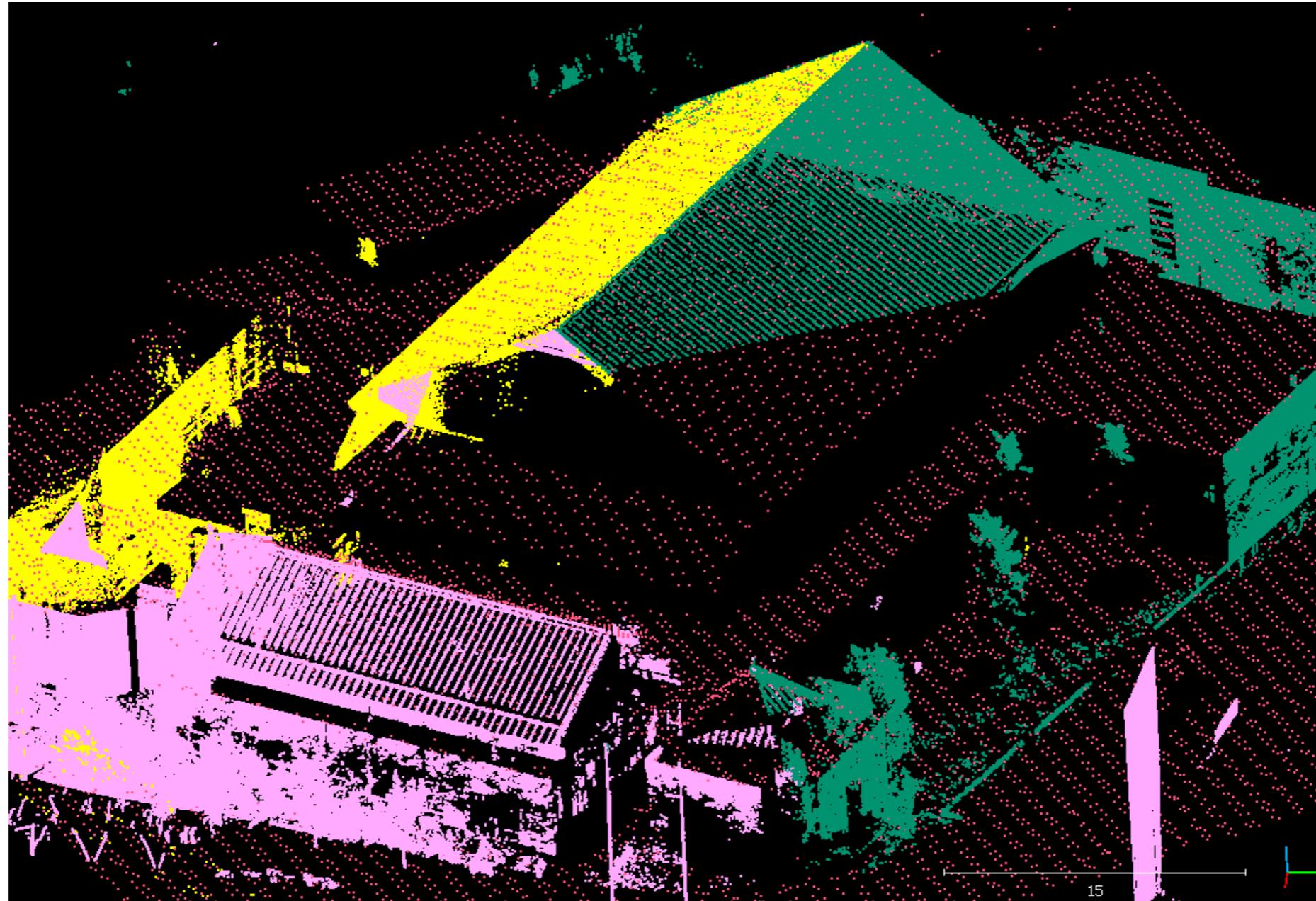
# Introduction: Coarse Vs. Fine Registration



# Introduction: Coarse Vs. Fine Registration

- For the registration of static scans, we need to address both coarse and fine registration.
- For the registration of directly-georeferenced mobile scans, we only need to address the fine registration.
  - The point clouds are already generated in a common reference frame defined by the onboard GNSS/INS unit.
  - The impact of erroneous system parameters on the alignment of point cloud is relative small.

# Introduction: TLS & ALS Registration



# Registration Paradigm Elements

## Registration Primitives

- Features that will be identified in the individual scans, e.g., Points, Lines, and Planar features

## Transformation Parameters

- Transformation parameters that describe the relationship between the reference frames of the different scans

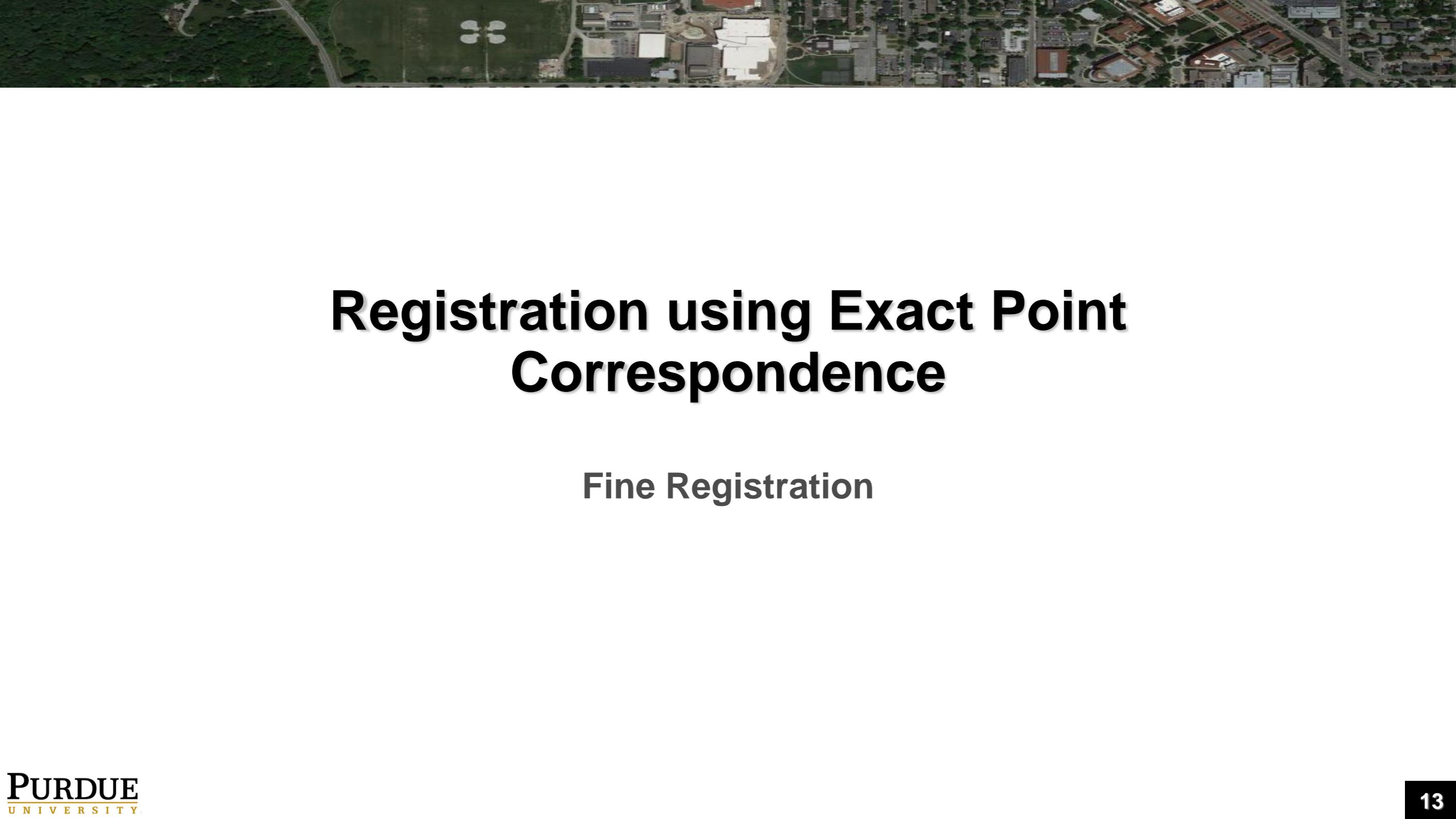
## Similarity Measure

- Describes the coincidence of conjugate primitives after registering different scans to a common reference frame

## Matching Strategy

- Controlling framework which is used for manipulating primitives, transformation parameters, and similarity measure

*(Habib A. & Al-Ruzouq R. , 2004)*



# Registration using Exact Point Correspondence

## Fine Registration

# Registration Targets

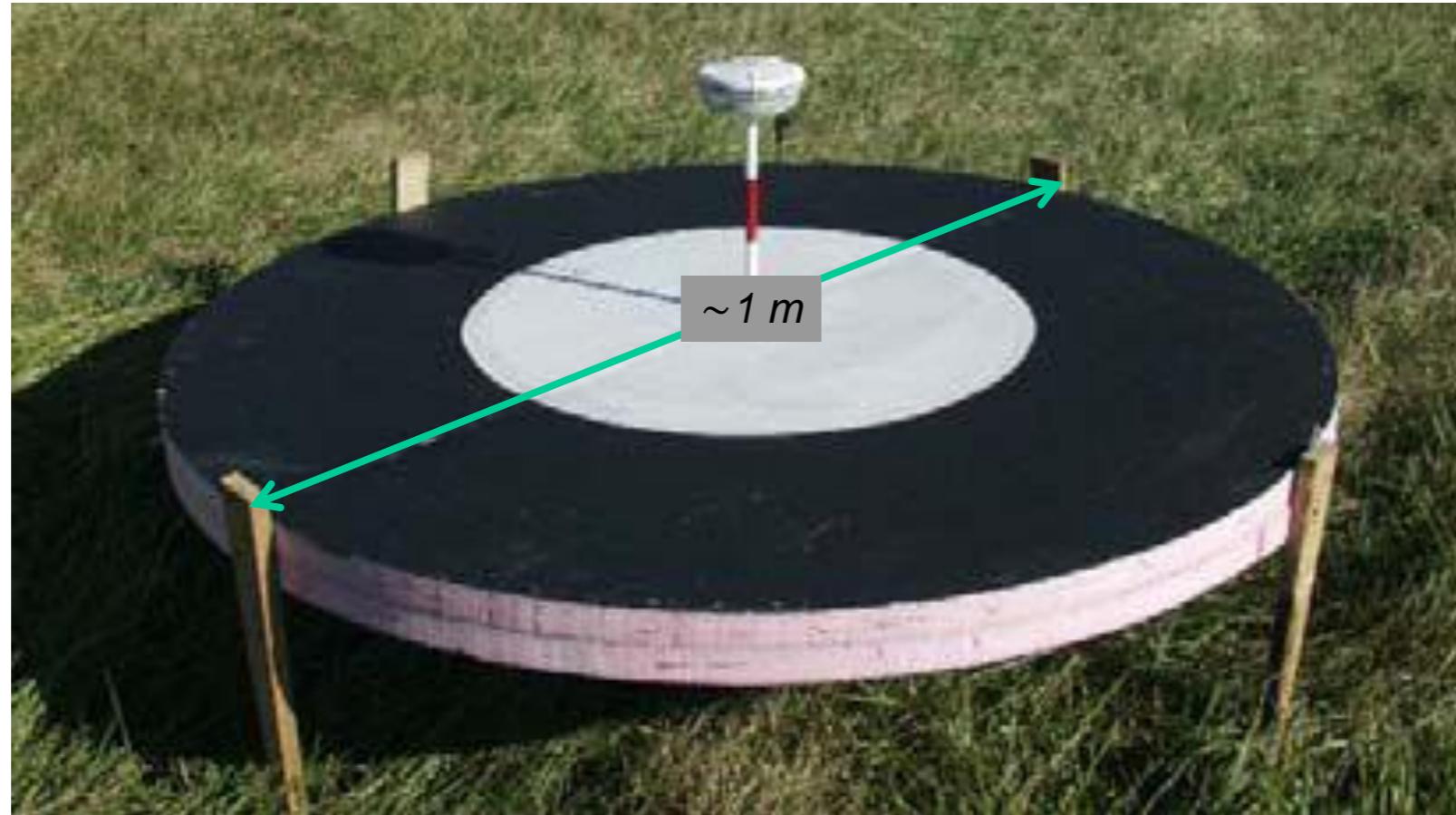
- Registration is performed using corresponding points, which could be signalized targets).



*Example of targets used in the registration of terrestrial laser scans (photos courtesy of leica-geosystems)*

# Registration Targets

- Targets for ALS registration



*Example of targets that could be used in the registration of airborne laser scans (photo courtesy of Csanyi & Toth, 2007)*



# Registration: Point-Based Mathematical Model

$$r_{a_2}^{S_2} = r_{S_1}^{S_2} + SR_{S_1}^{S_2}r_{a_1}^{S_1}$$

- $a_1$  and  $a_2$  are corresponding points in scans  $S_1$  and  $S_2$ , respectively,
- $r_{a_2}^{S_2}$  is the coordinate of  $a_2$  relative to the reference frame of scan  $S_2$ ,
- $r_{a_1}^{S_1}$  is the coordinate of  $a_1$  relative to the reference frame of scan  $S_1$ ,
- $r_{S_1}^{S_2}$  is the shift between the reference frames of the two scans (relative to the reference frame of scan  $S_2$ ),
- $R_{S_1}^{S_2}$  is the rotation matrix between the reference frames of the two scans, and
- S is the scale factor.

For LiDAR data, the scale **might not** be necessary.



# **Point-Based Registration without Exact Point-to-Point Correspondence**

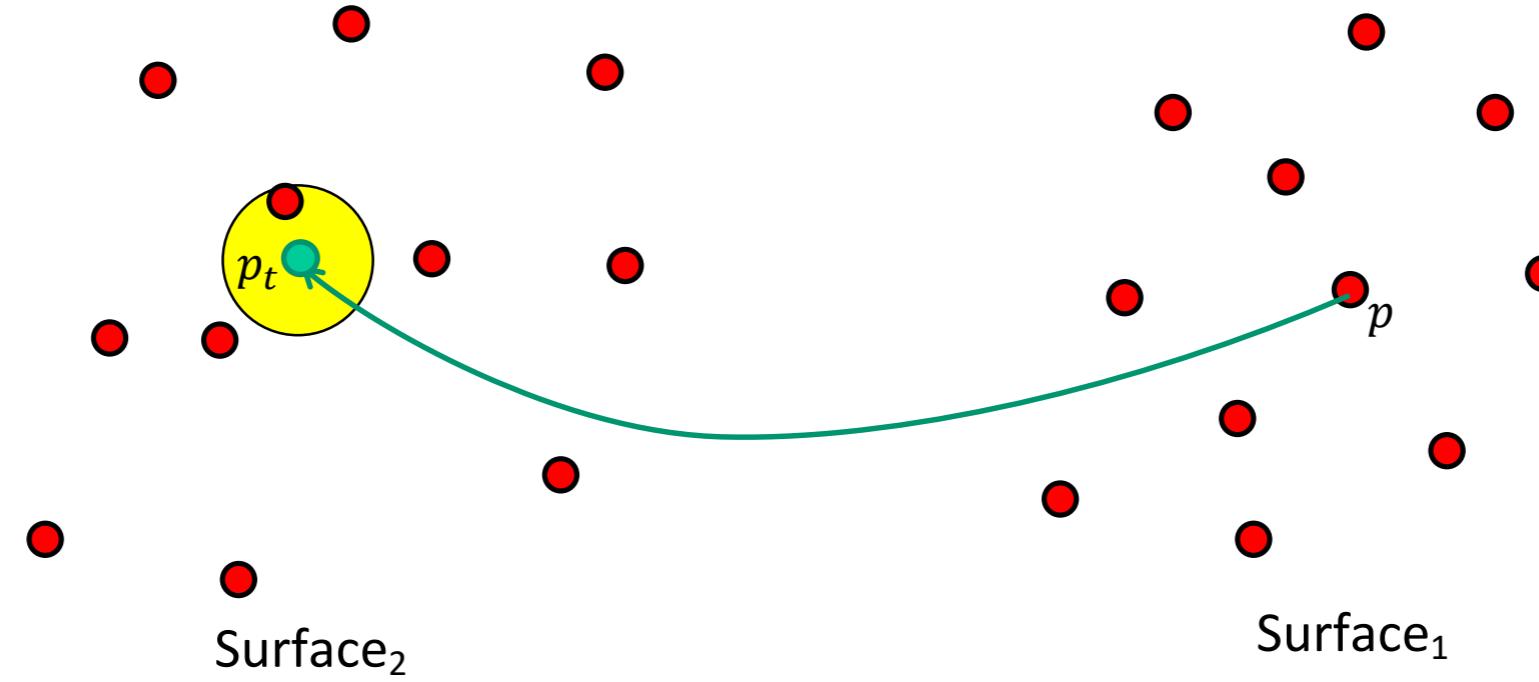
Fine Registration

# Registration: Automated Point-based Strategies

- **Commonly Adopted Point-Based Registration Methodologies without Exact Point-to-Point Correspondence:**
  - The Iterative Closest Point “ICP” (*Besl and Mckay, 1992*)
  - The Iterative Closest Patch “ICPatch” (*Habib et al., 2010*)
  - The Iterative Closest Projected Point “ICPP” (*Al-Durgham et al., 2011*)
- **Point-based registration methodologies require good initial approximations of the transformation parameters, which could be established through **manual interaction**.**

# Registration: Automated Point-based Strategies

- Iterative Closest Point (ICP):

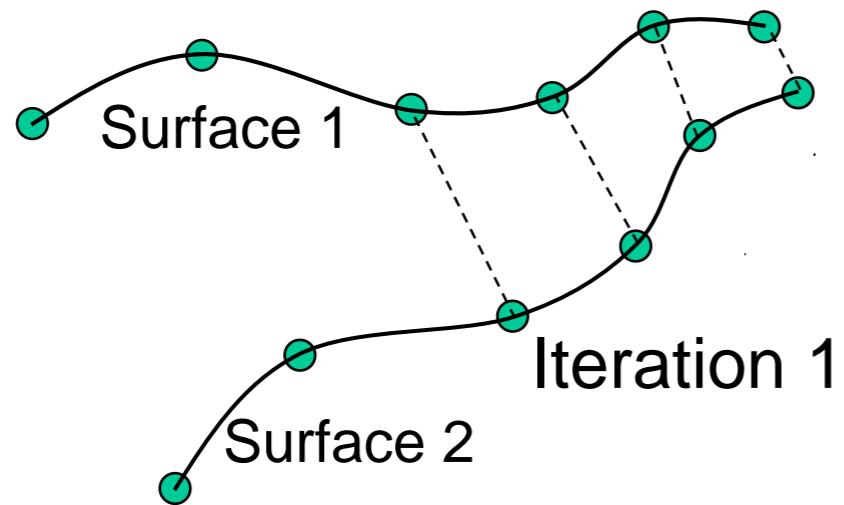


Note: ICP assumes point-to-point correspondence which is invalid among LiDAR points due to the irregular sampling nature of the points.

# Registration: Automated Point-based Strategies

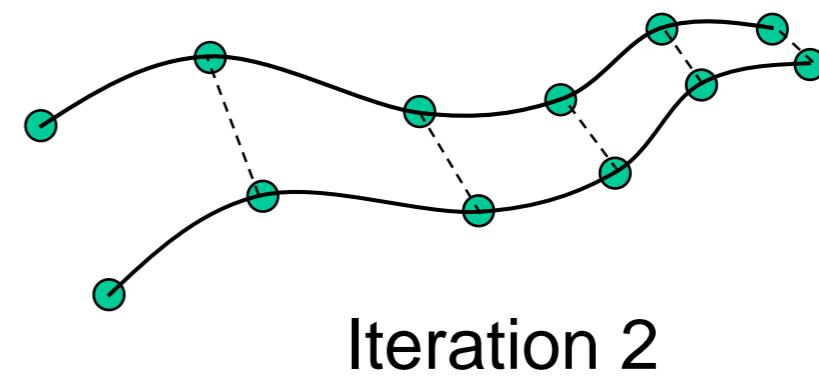
- **Iterative Closest Point (ICP):**

- Besl and McKay (1992)
  - Points as primitives
  - 3D rigid body transformation
  - Minimizes Euclidian distances
  - Performed iteratively



**Problem:**

- Exact point correspondences cannot be guaranteed when dealing with irregular point clouds.



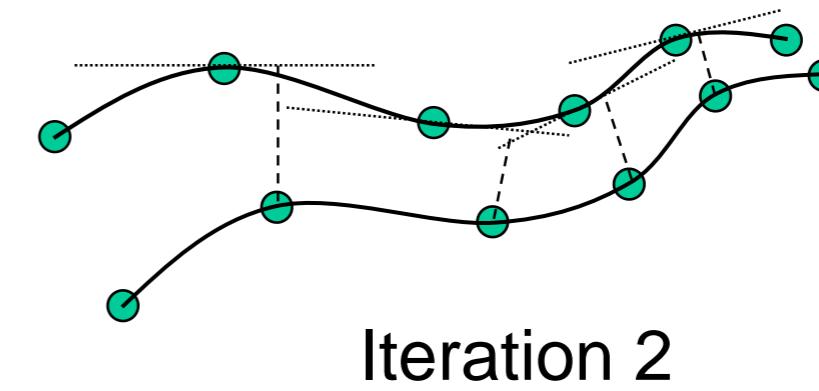
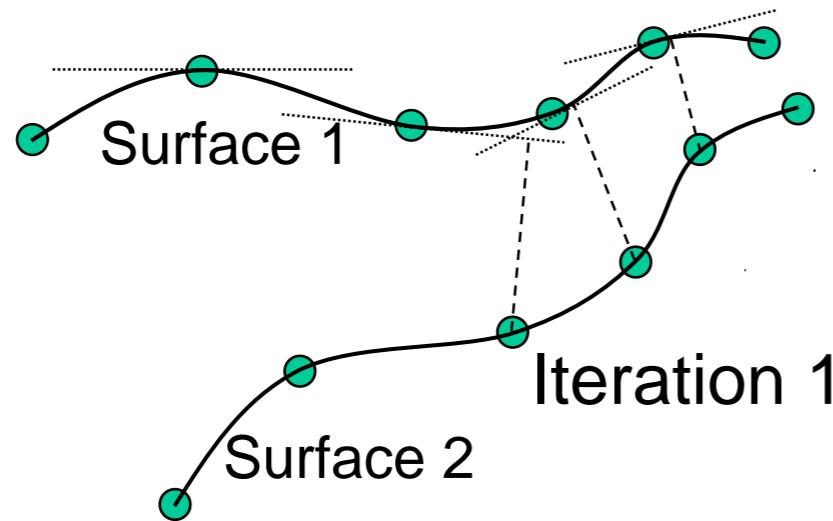
# Registration: Automated Point-based Strategies

- **Iterative Closest Point (ICP):**

- Chen and Medioni (1992)
  - Points and planes as primitives
  - Minimizes normal distances

Prerequisite:

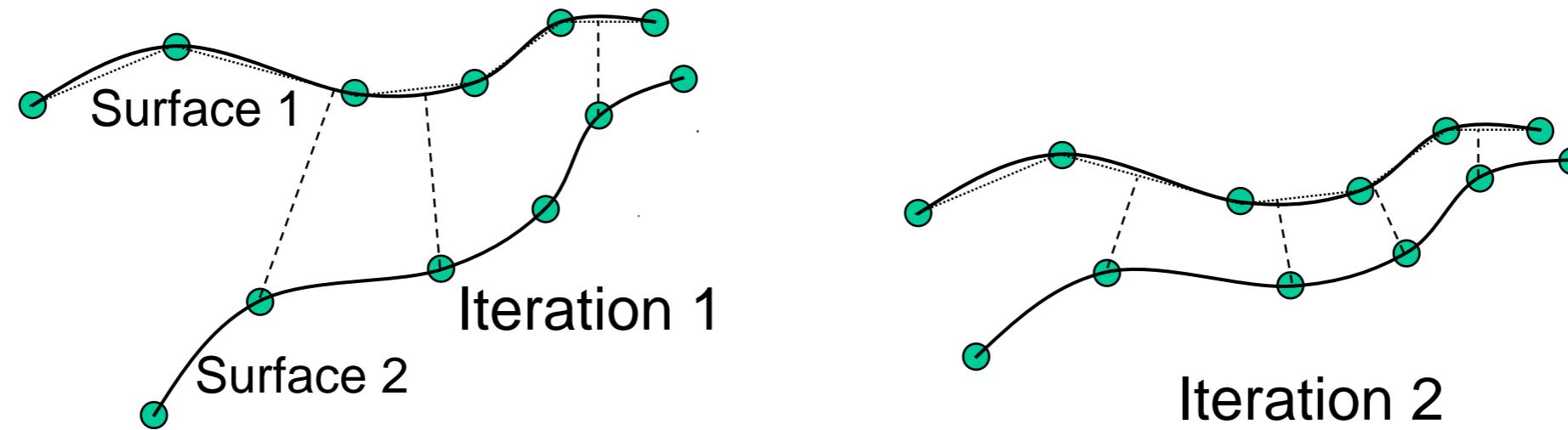
- The algorithm requires local plane fitting.



# Registration: Automated Point-based Strategies

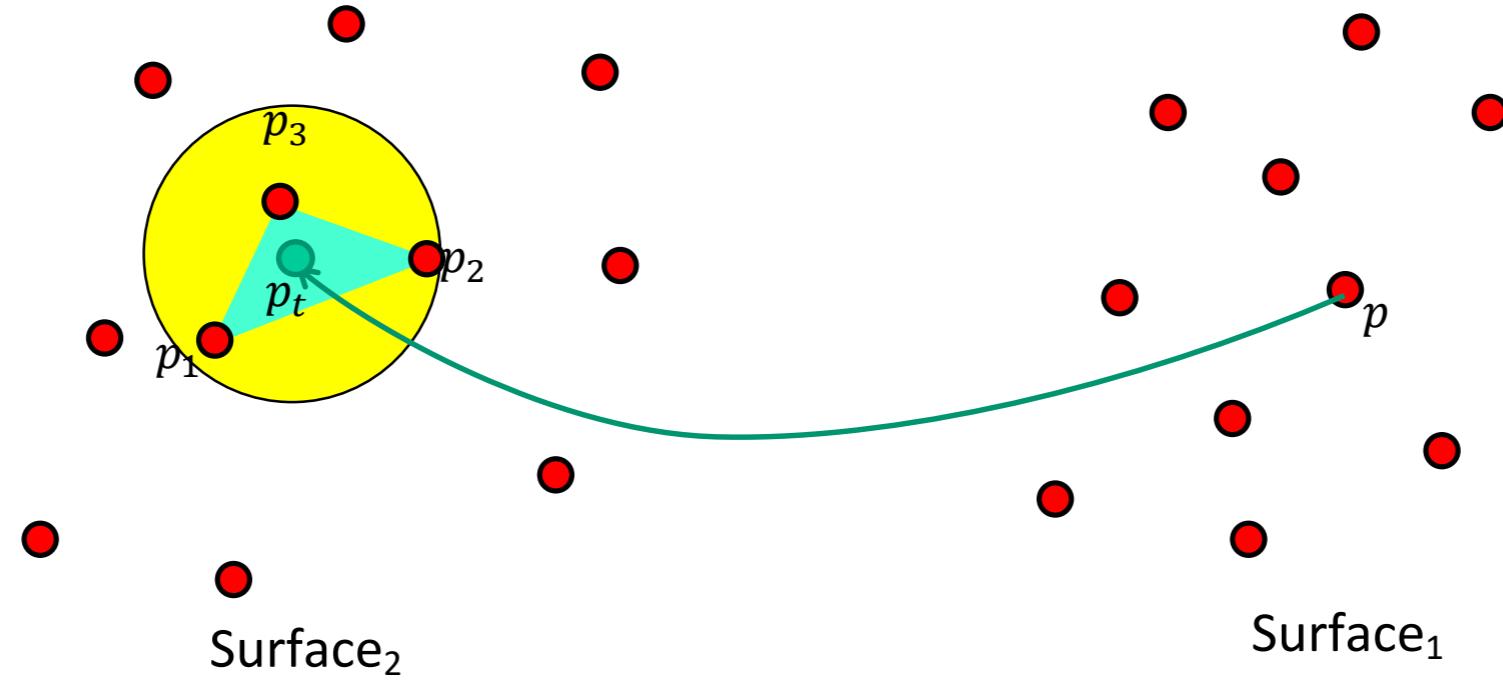
- **Iterative Closest Point (ICP): Variants**

- Points and triangular irregular network (TIN) patches
- 3D similarity transformation
- Coplanarity constraint and modified weight matrix
- Performed iteratively



# Registration: Automated Point-based Strategies

- Iterative Closest Patch (ICPatch):



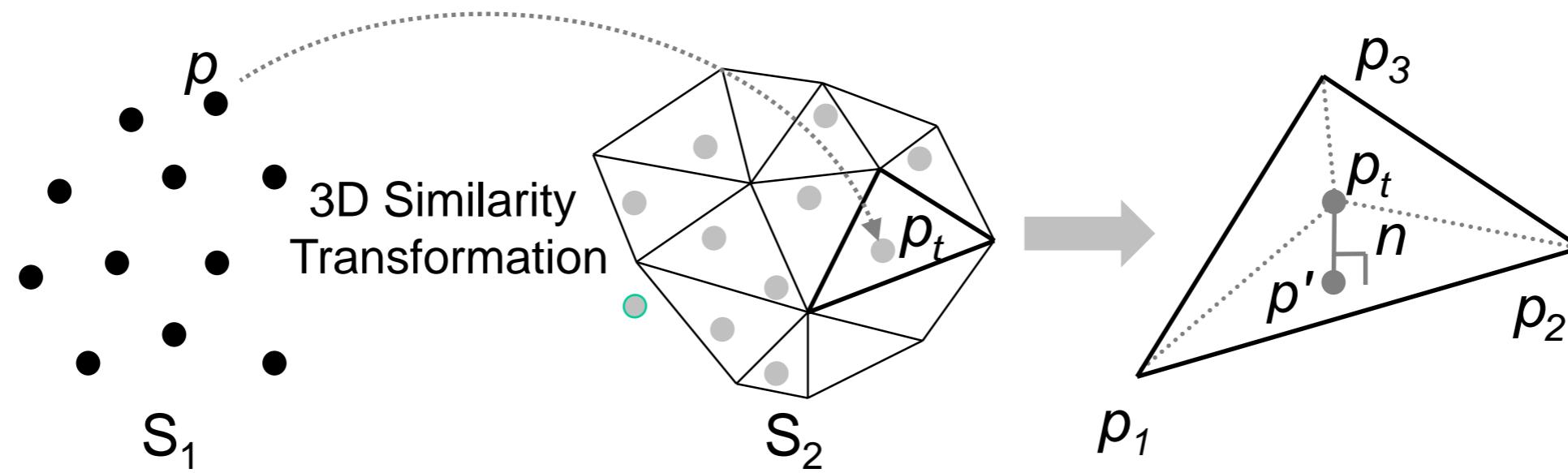
$$\mathbf{r}_p^{s_2} = \mathbf{r}_{s_1}^{s_2} + R_{s_1}^{s_2} \mathbf{r}_p^{s_1}$$

$$\mathbf{p}_t = \mathbf{r}_{s_1}^{s_2} + R_{s_1}^{s_2} \mathbf{p}$$

# Registration: Automated Point-based Strategies

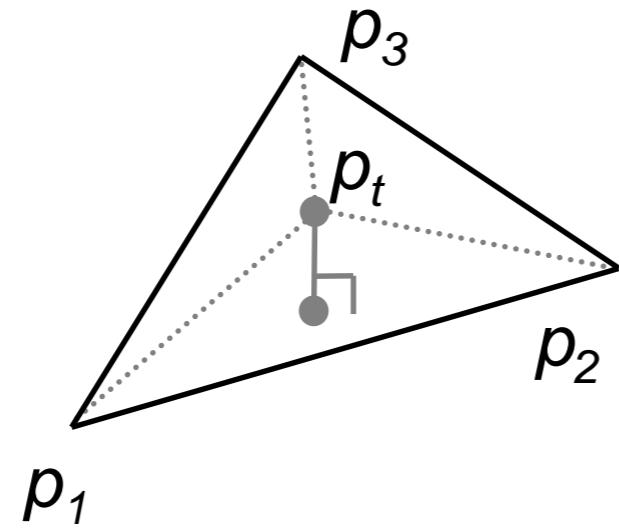
- Iterative Closest Patch (ICPatch):

- Conditions for valid conjugate point-patch pairs:
  - ✓ Triangular patch  $\Delta p_1 p_2 p_3$  must be the closest to the transformed point  $p_t$ , i.e.  $n=\min$ .
  - ✓ The normal distance,  $n$ , must be within a threshold.
  - ✓ The projection of  $p_t$ ,  $p'$ , must fall within  $\Delta p_1 p_2 p_3$ .



# Registration: Automated Point-based Strategies

- Iterative Closest Patch (ICPatch):
- Similarity Measure:  $p_1, p_2, p_3$  and  $p_t$  are assumed to be coplanar, i.e. the volume of the pyramid formed by vertices  $p_1, p_2, p_3$  and  $p_t$  should be zero.



$$\det \begin{bmatrix} X_{p_t} & Y_{p_t} & Z_{p_t} & 1 \\ X_{p_1} & Y_{p_1} & Z_{p_1} & 1 \\ X_{p_2} & Y_{p_2} & Z_{p_2} & 1 \\ X_{p_3} & Y_{p_3} & Z_{p_3} & 1 \end{bmatrix} = 0$$

$$\mathbf{p}_t = \mathbf{r}_{s_1}^{s_2} + \mathbf{R}_{s_1}^{s_2} \mathbf{p}$$

# Registration: Automated Point-based Strategies

- **Iterative Closest Projected Point (ICPP):**

- For a point  $p$  in  $S_1$ , find the *closest* three points in  $S_2$ .
- A match is established between a point in  $S_1$  and a triangle  $(p_1, p_2, p_3)$  in  $S_2$ .

The pair  $(p_t, p')$  is used for matching through the conventional ICP techniques, thus named the ICPP.

$$0 = r_2^m + R_2^m p' - (r_1^m + R_1^m p)$$

Conditions:

- $p_t \in \text{Convex}(p_1, p_2, p_3, p_e)$
- Compatible surface normals

