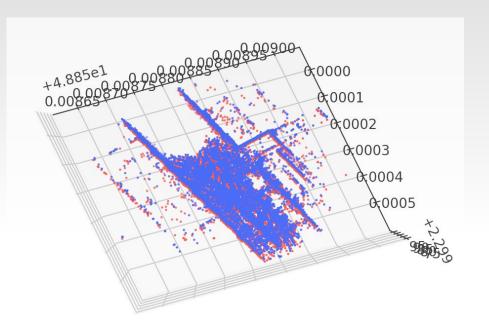
# **Point Clouds Registration**

Solve the point-set matching problem using ICP algorithm.

#### The Problem Declaration

- Input: two point clouds X, Y
- Target: find a map function
- Challenge: no extra info.



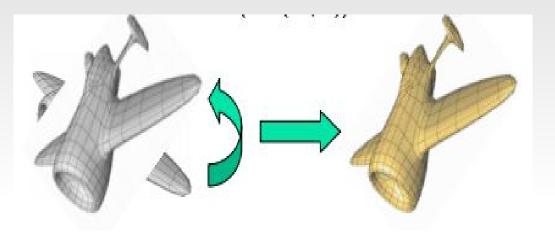
Blue:Point Cloud X Red:Point Cloud Y

Solution: ICP

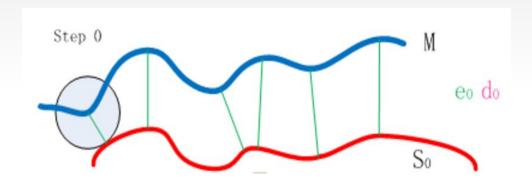
Result: find map Y = map(X)

Output: a space transformation map matrix

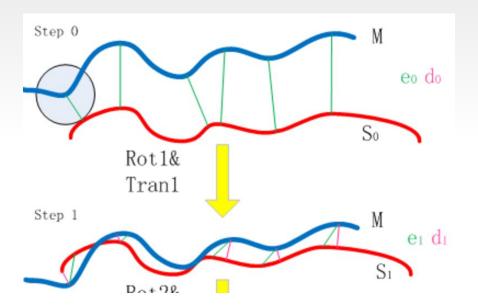
Feature: two major action



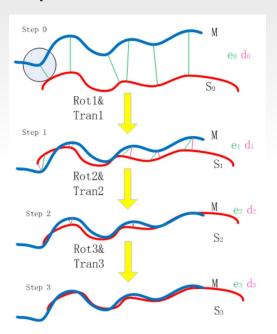
1. Find the closest match points in two point sets using inital R and T.



2. Update R and T to minimize error



3. Repeat previous steps untill R and T are converged.



#### **Implementation Details**

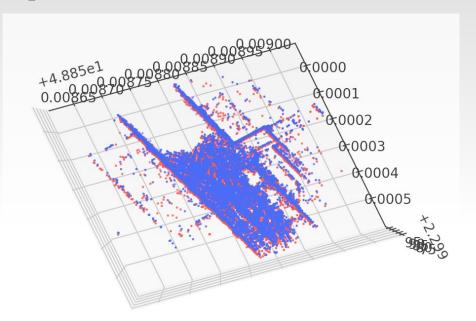
#### Our method contains following steps:

- 1. Find the closest point
- 2. Calculate the alignment
- 3. Apply the alignment
- 4. Iterate to reduce the error
- 5. Improvement
- 6. Result and analysis

#### 1. Find the closest points

Using KD-Tree

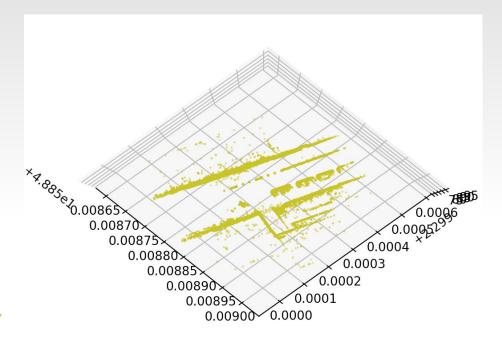
Blue:point Cloud A Red:Point Cloud B



#### 1. Find the closest points

4. Then we get a new point set Y.  $(Y = \{y_i\})$  for i=1...N\_a)

And Y is an injection from set A to set B.



Point Cloud Y

From point set A to point set Y, we need to compute the registration.

Compute the "center of point" μ(Y) of the measured point set Y

$$\vec{u}_Y = \frac{1}{N_Y} \sum_{i=1}^{N_Y} \vec{y}_i$$

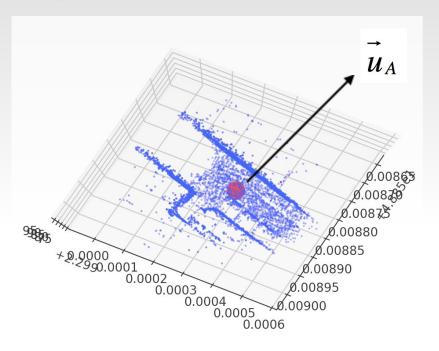
2. Compute the "center of mass"  $\mu(A)$  for the point set A

$$\vec{u}_A = \frac{1}{N_A} \sum_{i=1}^{N_A} \vec{a}_i$$

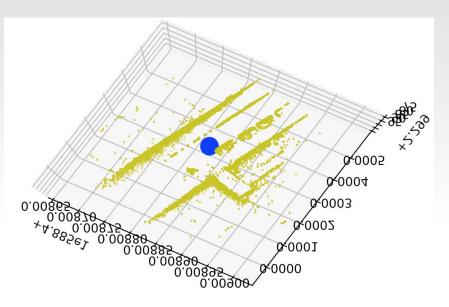
3. Compute cross-corvariance matrix of set Y and A is given by :

$$\sum_{AY} = \frac{1}{N_Y} \sum_{i=1}^{N_Y} \overrightarrow{y_i} [(\overrightarrow{y_i} - \overrightarrow{u_Y})(\overrightarrow{a_i} - \overrightarrow{u_A})^t]$$

The red dot is the center of mess for point set A



The blue dot is the center of mess for point set Y



4. The cyclic components of the anti-symmetric matrix :

$$A_{ij} = (\sum_{AY} - \sum_{AY}^{T})_{ij}$$

5. We use  $A_{ii}$  to form the column vector:

$$\triangle = \begin{bmatrix} A_{23} \\ A_{31} \\ A_{12} \end{bmatrix}$$

6. Then we get a matrix :  $Q(\Sigma_{AY})$ 

$$Q(\Sigma_{AY}) = \begin{bmatrix} tr(\Sigma_{AY}) & \Delta^T \\ \Delta & \sum_{AY} + \sum_{AY}^T - tr(\Sigma_{AY})I_3 \end{bmatrix}$$

7. The unit eigenvector of  $Q(\Sigma_{AY})$  is selected as optimal rotation vector :

$$\vec{q}_{Rotate} = \begin{bmatrix} q_0 & q_0 & q_2 & q_3 \end{bmatrix}^T$$

8. The optimal translation vector:

$$\vec{q}_{Translate} = \vec{u}_A - M(\vec{q}_{Rotate})\vec{u}_Y$$

The transformation matrix we get:

## 3. Apply the alignment

1. Before applying the alignment, the mean square error(MSE) between point set A and B is :

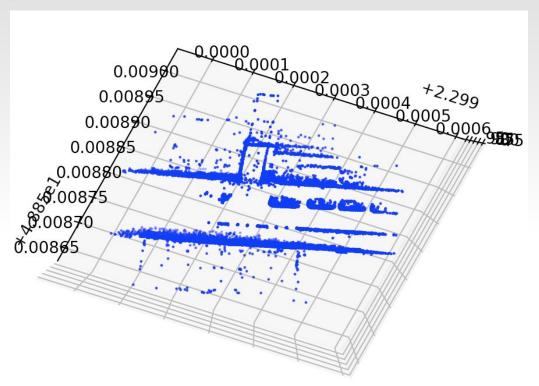
$$e_0 = \frac{1}{N_A} \sum_{i=1}^{N_A} ||a_{i0} - b_{i0}||$$

- 2. We transform the point set A using the obtained rotation and translation vectors into a new point set A 1.
- 3. After transformation, the MSE between point set A\_1 and B:

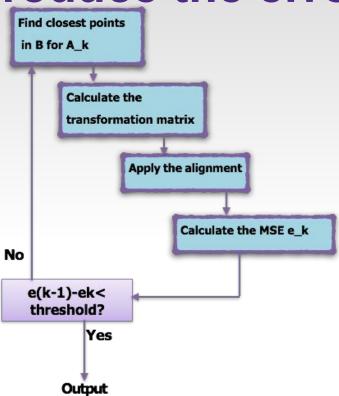
$$e_1 = \frac{1}{N_A} \sum_{i=1}^{N_A} ||a_{i1} - b_{i0}|| < e_0$$

## 3. Apply the alignment

Point set A\_1 after alignment.



#### 4. Iterate to reduce the error



#### 4. Iterate to reduce the error

. . . . . . . .

#### After 6 iterations, we get a more accurate result:

# Thanks for your attention!