Medical Image Processing for Diagnostic Applications

Iterative Closest Point Algorithm – Theory

Online Course – Unit 70 Andreas Maier, Joachim Hornegger, Eva Kollorz, Frank Schebesch Pattern Recognition Lab (CS 5)













Topics

Iterative Closest Point (ICP)

Theory

Point-to-Point Error Metric

Point-to-Plane Error Metric

Summary

Take Home Messages

Further Readings







Algorithm Outline

Given: Two point sets $P = \{\mathbf{p}_i\}, Q = \{\mathbf{q}_i\}, i = 1, ..., N$, where $\mathbf{p}_i, \mathbf{q}_i$ are 3×1 column vectors

Wanted: Best transformation **T** between these two point sets, consisting of

- rotation matrix R,
- translation t







Algorithm 1: Iterative closest point [5]

Input: Two point clouds: P, Q

Output: Transformation T, which aligns P and Q

```
1 T \leftarrow T_0;
  2 while not converged do
              for i \leftarrow 1 to N do
                     \mathbf{c}_i \leftarrow \text{GetClosestPointInQ}(\mathbf{T} \cdot \mathbf{p}_i);
                      if \|\mathbf{T} \cdot \mathbf{p}_i - \mathbf{c}_i\| \leq \theta_{max} then
                            \omega_i \leftarrow 1;
                      else
                              \omega_i \leftarrow 0;
                      end
              end
10
            \mathbf{T} \leftarrow \underset{i=1}{\operatorname{arg\,min}} \sum_{i=1}^{N} \omega_{i} \|\mathbf{T} \cdot \mathbf{p}_{i} - \mathbf{c}_{i}\|^{2};
11
```

12 end







Point-to-Point Error Metric

→ minimizes the Euclidean distance between selected point pairs.

The optimization problem can be solved by:

- a singular value decomposition (SVD) based method [1],
- a quaternion method [3],
- orthonormal matrices [4], or
- calculation based on dual quaternions [6].







Optimization function of the ICP:

$$\varepsilon = \sum_{i=1}^{N} \| (\mathbf{R}\mathbf{p}_i + \mathbf{t}) - \mathbf{q}_i \|^2,$$

where

$$\mathbf{q}_i = \underset{\mathbf{q}_X \in Q}{\operatorname{arg min}} d(\mathbf{q}_X, \mathbf{Rp}_i)$$

Decouple translation and rotation:

 \rightarrow computation of the center points $(\bar{\bf p},\bar{\bf q})$ of both point sets for translation

$$\left\{ egin{aligned} P': & \mathbf{p}_i' = \mathbf{p}_i - \mathbf{ar{p}}, \ Q': & \mathbf{q}_i' = \mathbf{q}_i - \mathbf{ar{q}} \end{aligned}
ight\} \qquad \Rightarrow \qquad \mathcal{E} = \sum_{j=1}^N \left\| \mathbf{R} \mathbf{p}_i' - \mathbf{q}_i' \right\|^2$$

Rrotation matrixttranslation vectord(x,y)Euclidean distancebetween two points x and y $Q = \{\mathbf{q}_i\}$ 3-D point sets $P = \{\mathbf{p}_i\}$ 3-D point sets

i = 1, ..., N number of points







Optimization function of the ICP:

$$\varepsilon = \sum_{i=1}^{N} \left\| \mathbf{R} \mathbf{p}_{i}' - \mathbf{q}_{i}' \right\|^{2}$$

Two simplifications:

- Index i matches the closest points of both point clouds.
- Both point clouds contain the same number of points.

The rotation matrix **R** is computed via help of the matrix **H**:

$$\mathbf{H} = \sum_{i=1}^{N} \mathbf{p}_{i}' \mathbf{q}_{i}' = \mathbf{U} \mathbf{\Sigma} \mathbf{V}^{\mathsf{T}}, \qquad \mathbf{R} = \mathbf{V} \mathbf{U}^{\mathsf{T}}.$$

The decoupled translation vector is recovered by

$$t = \bar{q} - R\bar{p}$$
.







Derivation for H:

1. Expanding right-hand side of ε :

$$\varepsilon = \sum_{i=1}^{N} \left\| \mathbf{R} \mathbf{p}_i' - \mathbf{q}_i' \right\|^2 = \sum_{i=1}^{N} \left(\mathbf{R} \mathbf{p}_i' - \mathbf{q}_i' \right)^\mathsf{T} \left(\mathbf{R} \mathbf{p}_i' - \mathbf{q}_i' \right) = \sum_{i=1}^{N} \left(\mathbf{p}_i'^\mathsf{T} \mathbf{R}^\mathsf{T} \mathbf{R} \mathbf{p}_i' + \mathbf{q}_i'^\mathsf{T} \mathbf{q}_i' - 2 \mathbf{q}_i'^\mathsf{T} \mathbf{R} \mathbf{p}_i' \right)$$

2. Minimizing ε is equivalent to maximizing:

$$F = \sum_{i=1}^{N} \mathbf{q}_{i}^{\prime \mathsf{T}} \mathbf{R} \mathbf{p}_{i}^{\prime} = \operatorname{tr} \left(\sum_{i=1}^{N} \mathbf{R} \mathbf{p}_{i}^{\prime} \mathbf{q}_{i}^{\prime \mathsf{T}} \right) = \operatorname{tr} \left(\mathbf{R} \mathbf{H} \right)$$







Depending on the properties of the point clouds, the solution may not be unique:

- If P' is collinear
 - → infinitely many rotations and reflections.
- If P' is coplanar
 - \rightarrow two unique solutions, the desired rotation matrix and its reflection:
 - the reflection is given if $\det \mathbf{R} = -1$,
 - the correct rotation matrix is given by $\mathbf{R} = \mathbf{V}'\mathbf{U}^{\mathsf{T}}$,
 - \mathbf{V}' is constructed by flipping the sign of the *i*-th column of \mathbf{V} , where index *i* identifies the zero singular value.
- If P' is not coplanar
 - \rightarrow one unique solution for $\mathbf{R} = \mathbf{V}\mathbf{U}^{\mathsf{T}}$.







Optimization function of the ICP:

$$\varepsilon = \frac{1}{N} \sum_{i=1}^{N} \| (\mathbf{R}(\mathbf{q}_R) \mathbf{p}_i + \mathbf{q}_T) - \mathbf{q}_i \|^2,$$

where

$$\mathbf{q}_i = \underset{\mathbf{q}_X \in Q}{\operatorname{arg\,min}} d\left(\mathbf{q}_X, \mathbf{R}\mathbf{p}_i\right)$$

$R = R(q_R)$	rotation matrix generated
_	by a unit quaternion
$\mathbf{q}_R = \left[q_0 q_1 q_2 \underline{q}_3\right]^T$	unit quaternion (rotation)
$\mathbf{q}_T = \left[q_4 q_5 q_6^{}\right]^{T}$	unit quaternion (translation)
$\mathbf{q} = \left[\mathbf{q}_{B} \mathbf{q}_{T} ight]^{T}$	registration state vector
d(x,y)	Euclidean distance
	between two points x and y
$Q = \{\mathbf{q}_i\}$	3-D point sets
$P = \{\mathbf{p}_i\}$	3-D point sets
i = 1,, N	number of points







Optimization function of the ICP:

$$\varepsilon = \frac{1}{N} \sum_{i=1}^{N} \| (\mathbf{R}(\mathbf{q}_R) \mathbf{p}_i + \mathbf{q}_T) - \mathbf{q}_i \|^2,$$

where

$$\mathbf{q}_i = \underset{\mathbf{q}_X \in Q}{\operatorname{arg\,min}} d(\mathbf{q}_X, \mathbf{Rp}_i)$$

Rotation matrix generated by a unit quaternion:

$$\mathbf{R} = \begin{pmatrix} q_0^2 + q_1^2 - q_2^2 - q_3^2 & 2(q_1q_2 - q_0q_3) & 2(q_1q_3 - q_0q_2) \\ 2(q_1q_2 + q_0q_3) & q_0^2 + q_2^2 - q_1^2 - q_3^2 & 2(q_2q_3 - q_0q_1) \\ 2(q_1q_3 - q_0q_2) & 2(q_2q_3 + q_0q_1) & q_0^2 + q_3^2 - q_1^2 - q_2^2 \end{pmatrix}$$

$$\mathbf{R} = \mathbf{R}(\mathbf{q}_R)$$

$$\mathbf{q}_R = [q_0 q_1 q_2 q_3]$$
 $\mathbf{q}_T = [q_4 q_5 q_6]^{\mathsf{T}}$
 $\mathbf{q} = [\mathbf{q}_R | \mathbf{q}_T]^{\mathsf{T}}$
 $\mathrm{d}(x, y)$

$$egin{aligned} Q &= \{\mathbf{q}_i\} \ P &= \{\mathbf{p}_i\} \ i &= 1,...,N \end{aligned}$$

rotation matrix generated by a unit quaternion $\mathbf{q}_R = [q_0 q_1 q_2 q_3]^T$ unit quaternion (rotation) unit quaternion (translation) registration state vector

Euclidean distance







Optimization function of the ICP:

$$\varepsilon = \frac{1}{N} \sum_{i=1}^{N} \left\| (\mathbf{R}(\mathbf{q}_R) \mathbf{p}_i + \mathbf{q}_T) - \mathbf{q}_i \right\|^2,$$

where

$$\mathbf{q}_i = \underset{\mathbf{q}_X \in Q}{\operatorname{arg\,min}} d(\mathbf{q}_X, \mathbf{Rp}_i)$$

Decouple translation and rotation:

 \rightarrow computation of the center points $(\bar{\mathbf{p}}, \bar{\mathbf{q}})$ of both point sets for translation

$$\left\{ egin{aligned} P': & \mathbf{p}_i' = \mathbf{p}_i - \mathbf{ar{p}}, \ Q': & \mathbf{q}_i' = \mathbf{q}_i - \mathbf{ar{q}} \end{aligned}
ight\} \qquad \Rightarrow \qquad \sum_{pq} = rac{1}{N} \sum_{i=1}^{N} \mathbf{p}_i' \mathbf{q}_i'^{\mathsf{T}}$$

$$\mathbf{R} = \mathbf{R}(\mathbf{q}_R)$$
 $\mathbf{q}_R = [q_0 q_1 q_2 q_3 q_4 q_5 q_6]^{\mathsf{T}}$
 $\mathbf{q} = [\mathbf{q}_R | \mathbf{q}_T]^{\mathsf{T}}$
 $\mathbf{d}(x, y)$

$$Q = \{\mathbf{q}_i\}$$
 $P = \{\mathbf{p}_i\}$
 $i = 1, ..., N$

rotation matrix generated by a unit quaternion $\mathbf{q}_R = [q_0 q_1 q_2 q_3]^{\mathsf{I}}$ unit quaternion (rotation) unit quaternion (translation) registration state vector Euclidean distance between two points x and y 3-D point sets 3-D point sets

number of points







Cross-covariance matrix:

$$\Sigma_{pq} = \frac{1}{N} \sum_{i=1}^{N} (\mathbf{p}_i - \bar{\mathbf{p}}) (\mathbf{q}_i - \bar{\mathbf{q}})^{\mathsf{T}} = \frac{1}{N} \sum_{i=1}^{N} \mathbf{p}_i' \mathbf{q}_i'^{\mathsf{T}},$$

- ightarrow Build anti-symmetric matrix **A** with $A_{ij} = \left(\Sigma_{pq} \Sigma_{pq}^{\mathsf{T}}\right)_{ij}$
- ightarrow Form column vector $\Delta = (A_{23}, A_{31}, A_{12})^{\mathsf{T}}$
- \rightarrow Form symmetric 4 \times 4 matrix $\mathbf{Q}(\Sigma_{pq})$:

$$\mathbf{Q}(\Sigma_{pq}) = \begin{pmatrix} \operatorname{tr}(\Sigma_{pq}) & \Delta^{\mathsf{T}} \\ \Delta & \Sigma_{pq} + \Sigma_{pq}^{\mathsf{T}} - \operatorname{tr}(\Sigma_{pq}) \mathbf{\textit{I}}_{3} \end{pmatrix}, \quad \mathbf{\textit{I}}_{3} \text{ is the 3} \times 3 \text{ identity matrix}$$

- ullet Optimal rotation $oldsymbol{q}_R$: unit eigenvector corresponding to the maximum eigenvalue of matrix $oldsymbol{Q}(\Sigma_{pq})$
- Optimal translation vector $\mathbf{q}_T = \mathbf{\bar{q}} \mathbf{R}(\mathbf{q}_R)\mathbf{\bar{p}}$







Point-to-Plane Error Metric [1]

Optimization function of the ICP:

$$\varepsilon = \sum_{i=1}^{N} \left\| \left(\left(\mathbf{R} \mathbf{p}_i + \mathbf{t} \right) - \mathbf{q}'_i \right) \mathbf{n}_i \right\|^2,$$

where

$$\mathbf{q}_{j}' = \left\{ \mathbf{q} \mid \underset{\mathbf{q} \in s_{j}}{\operatorname{arg\,min}} \ \|\mathbf{R}\mathbf{p}_{i} + \mathbf{t} - \mathbf{q}\| \right\}$$

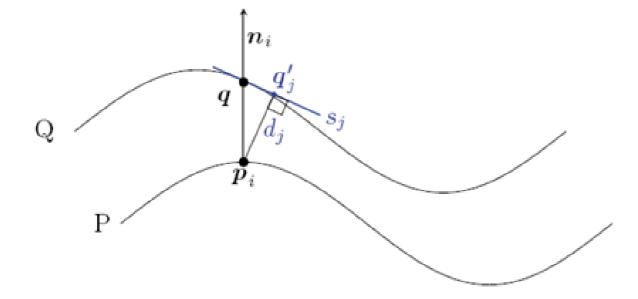


Figure 1: Image courtesy of Konrad Sickel [5]

R	rotation matrix
t	translation vector
d_j	distance between
-	the tangent plane s_i
	and the point \mathbf{p}_i
$Q = \{\mathbf{q}_i\}$	3-D point sets
$P = \{\mathbf{p}_i\}$	3-D point sets
s_j	tangent plane of Q at q
n_i	surface normal
i = 1,, N	number of points







Point-to-Plane Error Metric [1]

- More robust, accurate, and converges faster than the point-to-point error metric
- Utilizes surface normal as an additional input and allows that smooth or planar areas of the meshes slide over each other easily







Topics

Iterative Closest Point (ICP)

Theory

Point-to-Point Error Metric

Point-to-Plane Error Metric

Summary

Take Home Messages

Further Readings







Take Home Messages

- Several ways to solve the optimization problem of the ICP algorithm are known, two of which we have seen: using SVD and quaternions.
- Several error metrics can be used, we learned about point-to-point error metrics and point-to-plane metrics.







Further Readings

- [1] K. S. Arun, T. S. Huang, and S. D. Blostein. "Least-Squares Fitting of Two 3-D Point Sets". In: *IEEE Transactions on* Pattern Analysis and Machine Intelligence PAMI-9.5 (Sept. 1987), pp. 698-700. DOI: 10.1109/TPAMI.1987.4767965.
- [2] Paul J. Besl and Neil D. McKay. "A Method for Registration of 3-D Shapes". In: IEEE Transactions on Pattern Analysis and Machine Intelligence 14.2 (Feb. 1992), pp. 239–256. DOI: 10.1109/34.121791.
- [3] Berthold K. P. Horn. "Closed-form Solution of Absolute Orientation Using Unit Quaternions". In: Journal of the Optical Society of America A 4.4 (Apr. 1987), pp. 629–642. DOI: 10.1364/JOSAA.4.000629.
- [4] Berthold K. P. Horn, Hugh M. Hilden, and Shahriar Negahdaripour. "Closed-form Solution of Absolute Orientation Using Orthonormal Matrices". In: *Journal of the Optical Society of America A* 5.7 (July 1988), pp. 1127–1135. DOI: 10.1364/JOSAA.5.001127.
- [5] Konrad Sickel. "Computerized Automatic Modeling of Medical Prostheses". PhD Thesis. Erlangen: Pattern Recognition Lab, Friedrich-Alexander-Universität Erlangen-Nürnberg, Apr. 2013.
- [6] Michael W. Walker, Lejun Shao, and Richard A. Volz. "Estimating 3-D Location Parameters Using Dual Number Quaternions". In: CVGIP: image understanding 54.3 (Nov. 1991), pp. 358–367. DOI: 10.1016/1049-9660(91)90036-0.