## ICP using SVD

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Define points sets:

$$P = \{p_1, p_2, \dots, p_n\}, \quad P' = \{p'_1, p'_2, \dots, p'_n\}$$
(1)

Points transform:

$$\forall i, p_i = Rp_i' + t \tag{2}$$

Define error:

$$e_i = p_i - (Rp_i' + t) \tag{3}$$

$$E = \frac{1}{N} \sum_{i=1}^{N} ||p_i - (Rp_i' + t)||^2$$
(4)

Constructing least squares problems:

$$\underset{R,t}{\operatorname{arg\,min}} \frac{1}{N} \sum_{i=1}^{N} ||p_i - (Rp_i' + t)||^2 \tag{5}$$

Solve for the displacement matrix t:

$$\frac{dE}{dt} = \frac{2}{N} \sum_{i=1}^{N} (Rp_i' + t - p_i) = \frac{2}{N} \sum_{i=1}^{N} t + \frac{2}{N} \sum_{i=1}^{N} Rp_i' - \frac{2}{N} \sum_{i=1}^{N} p_i = 0$$
 (6)

$$t + \frac{1}{N} \sum_{i=1}^{N} Rp_i' - \frac{1}{N} \sum_{i=1}^{N} p_i = 0$$
 (7)

Define centroid:

$$p = \frac{1}{N} \sum_{i=1}^{N} p_i, \quad p' = \frac{1}{N} \sum_{i=1}^{N} p'_i$$
 (8)

Thus:

$$t = p - Rp' \tag{9}$$

Substitute back to (4)

$$E = \frac{1}{N} \sum_{i=1}^{N} ||p_i - Rp_i' - p + Rp'||^2$$
(10)

$$= \frac{1}{N} \sum_{i=1}^{N} ||p_i - p - R(p_i' - p')||^2$$
(11)

Define decentroided point clouds:

$$q_i = p_i - p, \quad q'_i = p'_i - p'$$
 (12)

Thus:

$$E = \frac{1}{N} \sum_{i=1}^{N} ||q_i - Rq_i'||^2$$
(13)

$$\underset{R}{\operatorname{arg\,min}} \frac{1}{N} \sum_{i=1}^{N} ||q_i - Rq_i'||^2 \tag{14}$$

Solve for the rotation matrix R:

$$||q_i - Rq_i'||^2 = (q_i - Rq_i')^T (q_i - Rq_i') = q_i^T q_i - q_i^T Rq_i' - q_i'^T R^T q_i + R^T q_i' Rq_i'$$
(15)

Since:

$$q_i^T R q_i' : (3 \times 1)(3 \times 3)(1 \times 3) = (1 \times 1) \Rightarrow scalar \quad quantity$$
 (16)

$$q_i^T R^T q_i : (3 \times 1)(3 \times 3)(1 \times 3) = (1 \times 1) \Rightarrow scalar \quad quantity \tag{17}$$

Thus:

$$||q_i - Rq_i'||^2 = q_i^T q_i - 2q_i^T Rq_i' + R^T q_i' Rq_i'$$
(18)

Substitute back to (13, 14)

$$E = \frac{1}{N} \sum_{i=1}^{N} (q_i^T q_i - 2q_i^T R q_i' + R^T q_i' R q_i')$$
(19)

$$\underset{R}{\operatorname{arg\,min}} \frac{1}{N} \sum_{i=1}^{N} (q_i^T q_i - 2q_i^T R q_i' + R^T q_i' R q_i')$$
 (20)

$$= \arg\min_{R} \frac{1}{N} \sum_{i=1}^{N} (-2q_i^T R q_i')$$
 (21)

$$= \arg\max_{R} \frac{1}{N} \sum_{i=1}^{N} (q_i^T R q_i')$$
 (22)

The trace of a constant is known to be equal to the constant itself, and tr(AB) = tr(BA):

$$\frac{1}{N} \sum_{i=1}^{N} (q_i^T R q_i') = tr(\frac{1}{N} \sum_{i=1}^{N} (q_i^T R q_i'))$$
(23)

$$= \frac{1}{N} tr(R \sum_{i=1}^{N} (q_i' q_i^T))$$
 (24)

Define:

$$H = \sum_{i=1}^{N} (q_i q_i^{\prime T}) \quad or \quad H = \frac{1}{N} \sum_{i=1}^{N} (q_i q_i^{\prime T})$$
 (25)

Substitute back to (22)

$$\underset{R}{\arg\max} tr(RH^T) \tag{26}$$

SVD:

$$H = U\Sigma V^T \tag{27}$$

$$\underset{R}{\arg\max} tr(R(U\Sigma V^T)^T) = \underset{R}{\arg\max} tr(RV\Sigma U^T)$$
 (28)

$$= \underset{R}{\operatorname{arg\,max}} tr(\Sigma U^T R V) \tag{29}$$

Define:

$$M = U^T R V (30)$$

Since  $R, U^T, V$  are orthogonal matrixs: M is a special orthogonal matrix:

$$\forall i, j, \quad |m_{ij}| < 1, \quad m \in M \tag{31}$$

$$\Sigma = \begin{pmatrix} \sigma_1 & & \\ & \sigma_2 & \\ & & \sigma_3 \end{pmatrix}, \quad \sigma_1, \sigma_2, \sigma_3 < 1 \tag{32}$$

$$tr(\Sigma M) = \begin{pmatrix} \sigma_1 & \\ & \sigma_2 & \\ & & \sigma_3 \end{pmatrix} \begin{pmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{pmatrix}$$

$$= \sum_{i,j} \sigma_i m_i j$$
(33)

$$=\sum_{i,j}\sigma_i m_i j \tag{34}$$

$$\leq \sum_{i,j} \sigma_i \tag{35}$$

Take the equal sign (maximum value) at  $m_{ij} = 1$ , since M is a orthogonal matrix:

$$M = \begin{pmatrix} 1 & & \\ & 1 & \\ & & 1 \end{pmatrix} = I \tag{36}$$

Thus:

$$M = U^T R V = I (37)$$

$$R = UV^T (38)$$

$$t = p - Rp' = p - UV^T p' \tag{39}$$