1. check if input is valid:

- (1) correspondences $\geq = 8$
- (2) points 0.size() = points 1.size()

2. normalize process - to get W(Wf = 0):

• construct transform matrix:

```
tx = sum(x coordinates) / N
```

ty = sum(y coordinates) / N

(tx, ty) - use the image center as new origin of coordinate system avg distance = sum(distance from p to (tx, ty) - NOT squared distance):

```
// scale factor
double sum_dist = 0;
for (const auto& p : points)
    sum_dist += sqrt((p.x() - tx) * (p.x() - tx) + (p.y() - ty) * (p.y() - ty));
double avg_dc = sum_dist / N;
where 'points' is the input 2d points set
```

scale = sqrt(2) / avg distance

normalize points:

p_normalized = transform_matrix * p
points_0 and points_1 both are normalized

- 3. get Fundamental matrix
 - use SVD to solve W and get initial_F construct W - use <u>normalized</u> points

$$\begin{bmatrix}
p_i = (u_i, v_i, 1) \\
p'_i = (u'_i, v'_i, 1)
\end{bmatrix} \quad + \quad p'^T F p = 0$$

$$\begin{bmatrix}
u_i u'_i & v_i u'_i & u'_i & u_i v'_i & v_i v'_i & v_i & v_i & 1
\end{bmatrix} \begin{bmatrix}
F_{11} \\
F_{12} \\
F_{13} \\
F_{21} \\
F_{22} \\
F_{23} \\
F_{31} \\
F_{32} \\
F_{32} \\
F_{33}
\end{bmatrix} = 0$$

$$\begin{bmatrix} u_1u_1' & v_1u_1' & u_1' & u_1v_1' & v_1v_1' & v_1' & u_1 & v_1 & 1 \\ u_2u_2' & v_2u_2' & u_2' & v_2v_2' & v_2' & u_2 & v_2 & 1 \\ u_3u_3' & v_3u_3' & u_3' & u_3v_3' & v_3v_3' & v_3' & u_3 & v_3 & 1 \\ u_4u_4' & v_4u_4' & u_4' & u_4v_4' & v_4' & u_4 & v_4 & 1 \\ u_5u_5' & v_5u_5' & u_5' & u_5v_5' & v_5' & v_5 & v_5 & 1 \\ u_6u_6' & v_6u_6' & u_6' & u_6v_6' & v_6' & v_6 & u_6 & v_6 & 1 \\ u_7u_7' & v_7u_7' & u_7' & u_7v_7' & v_7v_7' & v_7 & v_7 & 1 \\ u_8u_8' & v_8u_8' & u_8' & u_8v_8' & v_8v_8' & v_8' & u_8 & v_8 & 1 \end{bmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix}$$

solve W:

- (1) Matrix W(m, n); // W is a m by n matrix, m(=rows), n(=cols)
- (2) svd decompose(W, U, S, V); // W = U * S * V.transpose()

```
// solve W using SVD
Matrix U(m, m, 0.0);  // initialized with 0s
Matrix S(m, n, 0.0);  // initialized with 0s
Matrix V(n, n, 0.0);  // initialized with 0s
svd_decompose(W, U, S, V);

// Form F using the last column of V
Vector vlc = V.get_column(V.cols() - 1);  // get the last column of V
F.set_row(0, { vlc[0], vlc[1], vlc[2] });
F.set_row(1, { vlc[3], vlc[4], vlc[5] });
F.set_row(2, { vlc[6], vlc[7], vlc[8] });
```

- use SVD to decompose initial_F
- rank-2 approximation make S(2, 2) = 0
- denormalize F

$$F = T'^T F_q T$$

• scale F such that F(2, 2) = 1.0

```
// decompose F using SVD
int m = initial F.rows();
int n = initial_F.cols();
Matrix U(m, m, 0.0); // initialized with 0s
Matrix S(m, n, 0.0); // initialized with 0s
Matrix V(n, n, 0.0); // initialized with 0s
svd_decompose(initial_F, U, S, V);
// rank-2 approximation - make S(2, 2) = 0
S(2, 2) = 0;
// update F
F = U * S * V.transpose();
// denormalize F: F = T'_transpose() * F * T scale F such that F(2, 2) = 1.0
F = T .transpose() * F * T;
// scale F such that F(2, 2) = 1.0 (F is up to scale)
const double scale = 1.0 / F(2, 2);
F = F * scale; // scale F
```

4. get Essential matirx

$$E = K^T F K$$

• where K is the intrinsic matrix

If TWO different cameras are used:

$$E = [t_{\times}]R = K'^T F K$$

K - camera 1 (usually camera 1 CRS is also the World CRS)

K' - camera 2

5. get possible R, t combinations from E (4 combinations in total)

- · Essential matrix from fundamental matrix
- Relative pose from essential matrix
 - SVD of E
 - determinant(R) > 0

$$W = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$E = U\Sigma V^T$$

• Two potential values

T up to a sign

$$R = (\det UWV^T)UWV^T$$
 or $(\det UW^TV^T)UW^TV^T$

• Two potential values

• Last column of U

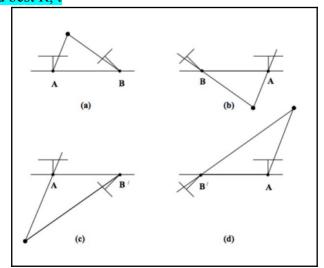
Corresponds to smallest singular value

$$t = \pm U \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \pm u_3$$

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In practice, det(UWV.transpose()) and det(UW.transpose()V.transpose()) are both > 0 and (should be) equal(or close) to 1

6. find best R, t



Four different settings of R, t

in order to find best R,t, we need to use correspondences from two pictures to recover the 3d points, and for these 3d points, they need to:

(1) z coordinates in World CRS \geq 0 (camera-1 World CRS)

(2) z coordinates in Relative CRS >0 (camera-2 Relative CRS)

transform point3d from camera 1 to camera 2:

$$Q = R * P + t$$

recover 3d points from correspondences

Two image points

$$p = MP = (x, y, 1)$$

 $p' = M'P = (x', y', 1)$

By the definition of the cross product

$$p\times (MP)=0$$
 Similar constraints can also be formulated for p' and M'.
$$A=\begin{bmatrix}xM_3-M_1\\yM_3-M_2\\x'M_3'-M_1'\\y'M_3'-M_2'\end{bmatrix}$$

$$y(M_3P)-(M_2P)=0$$

$$x(M_2P)-y(M_1P)=0$$

$$AP=0$$

 $M = K[I \ 0] - M$ is 3 by 4 matrix $M' = K[R \ t] - M'$ is 3 by 4 matrix M1, M2, M3 - each row of M the same with M'

A is a 4 by 4 matrix;

use SVD to decompose A, the last column of V has 4 elements, and its cartesian() coordinates is the coordinates of 3d points in World CRS

```
// construct matrix A:
 // xM3 - M1
// yM3 - M2
 // x'M3' - M1'
 // y'M3' - M2'
double x = points_0[i].x(), y = points_0[i].y();
double x_ = points_1[i].x(), y_ = points_1[i].y();
Matrix44 A;
A.set_row(0, x * M3 - M1);
A.set_row(1, y * M3 - M2);
A.set_row(2, x_ * M3_ - M1_);
A.set_row(3, y_* M3_ - M2_);
// solve A using SVD
int m = A.rows();
int n = A.cols();
Matrix U(m, m, 0.0); // initialized with 0s
Matrix S(m, n, 0.0); // initialized with 0s
Matrix V(n, n, 0.0); // initialized with 0s
svd_decompose(A, U, S, V);
Vector4D vlc = V.get_column(V.cols() - 1); // get the last column of V
points_3d.emplace_back();
points_3d.back() = vlc.cartesian(); // add the 3d point to the result vector
transform these recovered 3d points to camera-2 CRS:
Q = R * P + t where P is the 3d point in World CRS
```

Then for best R, t:

```
3D points must be in front of both cameras
```

- First camera
 - -P.z > 0?
- Second camera
 - P in 2nd camera's coordinate system: Q = R * P + t
 - -Q.z > 0?