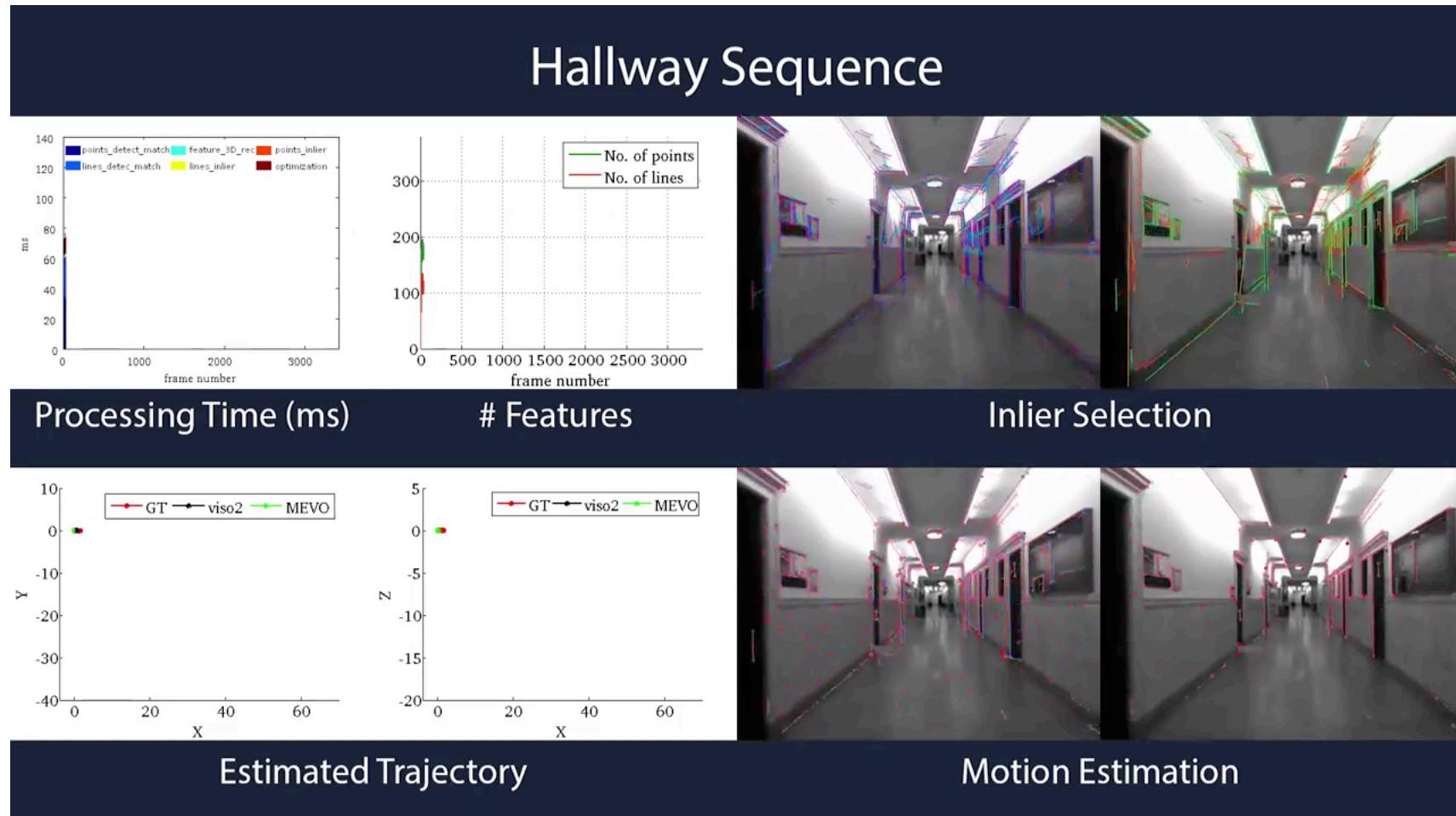


# 3D Motion from Two Views or Structure from Motion (SfM)

Kostas Daniilidis

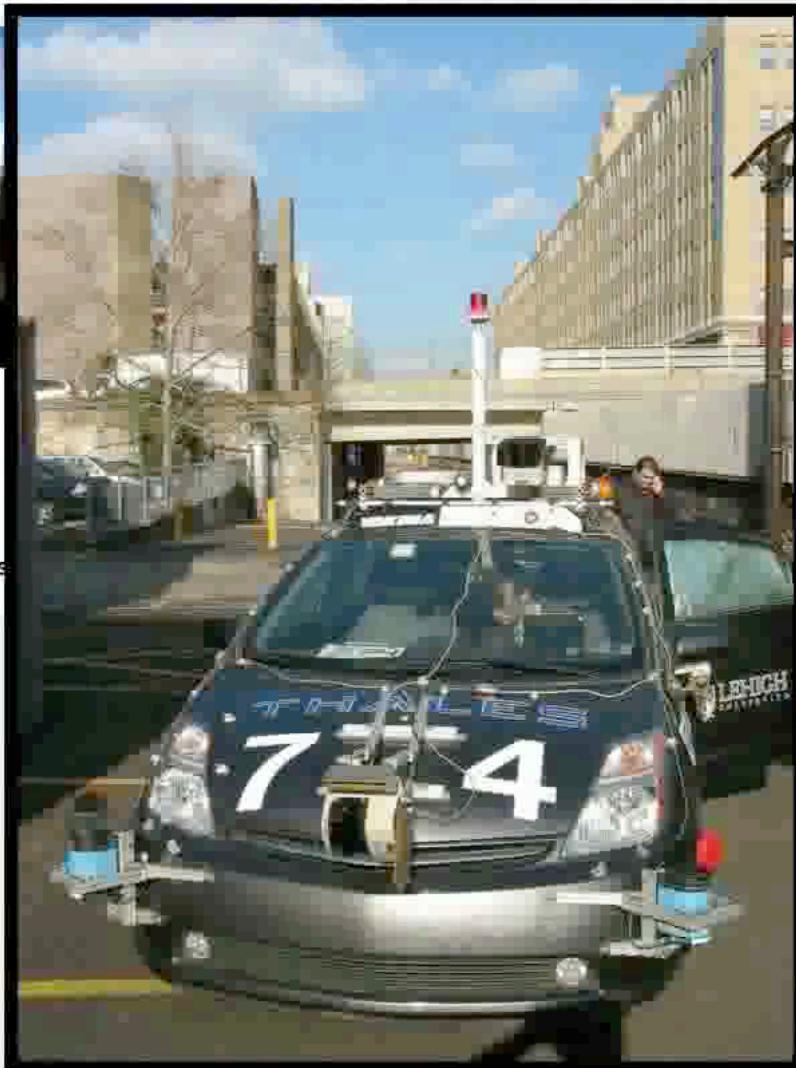
# Main ingredient of all visual odometry algorithms



Panoramic image (from 6 cameras)



Reconstruction (global view)



Recons

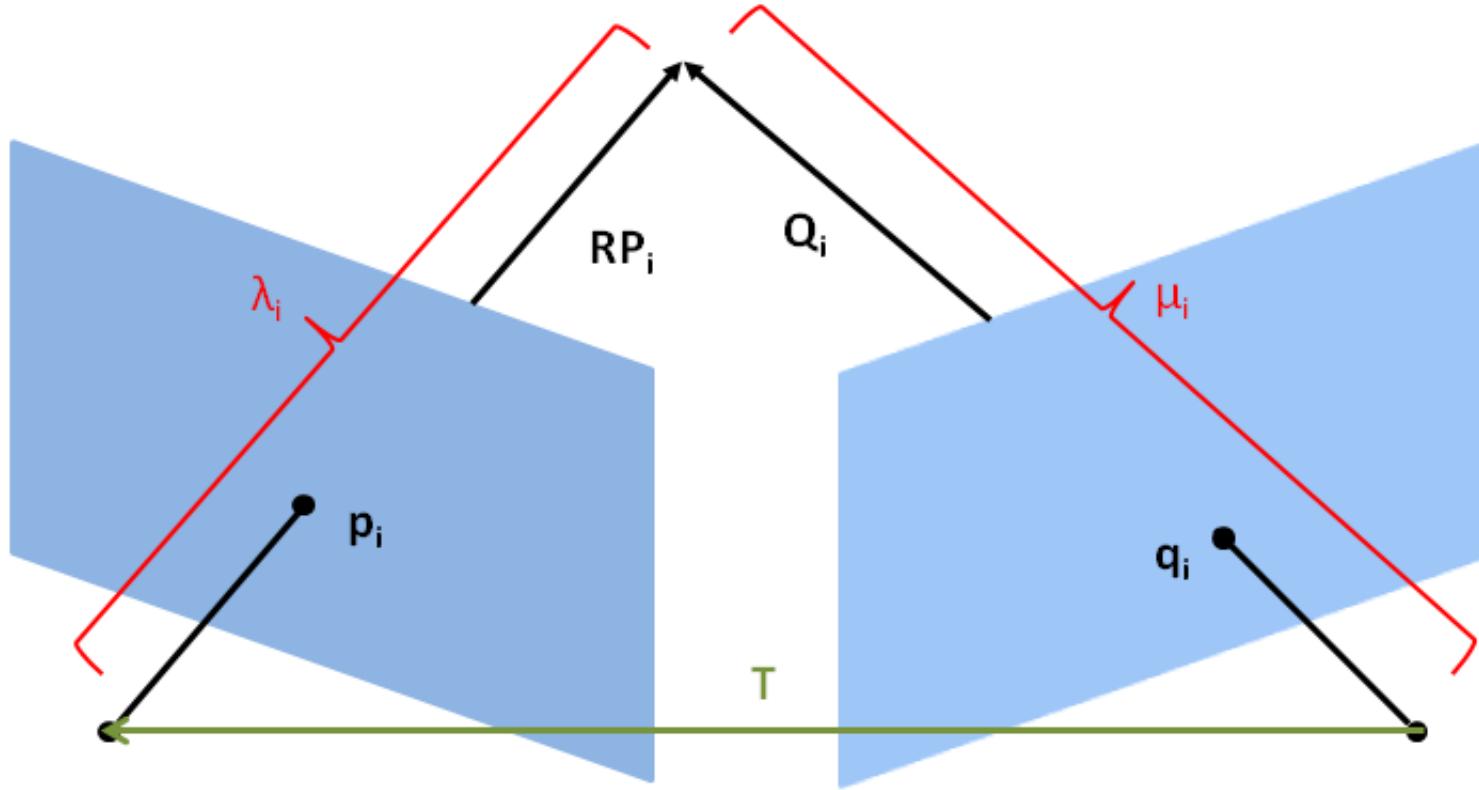


Map data ©2006 Tele Atlas -

# Two calibrated views of a scene



# Two calibrated views of a scene

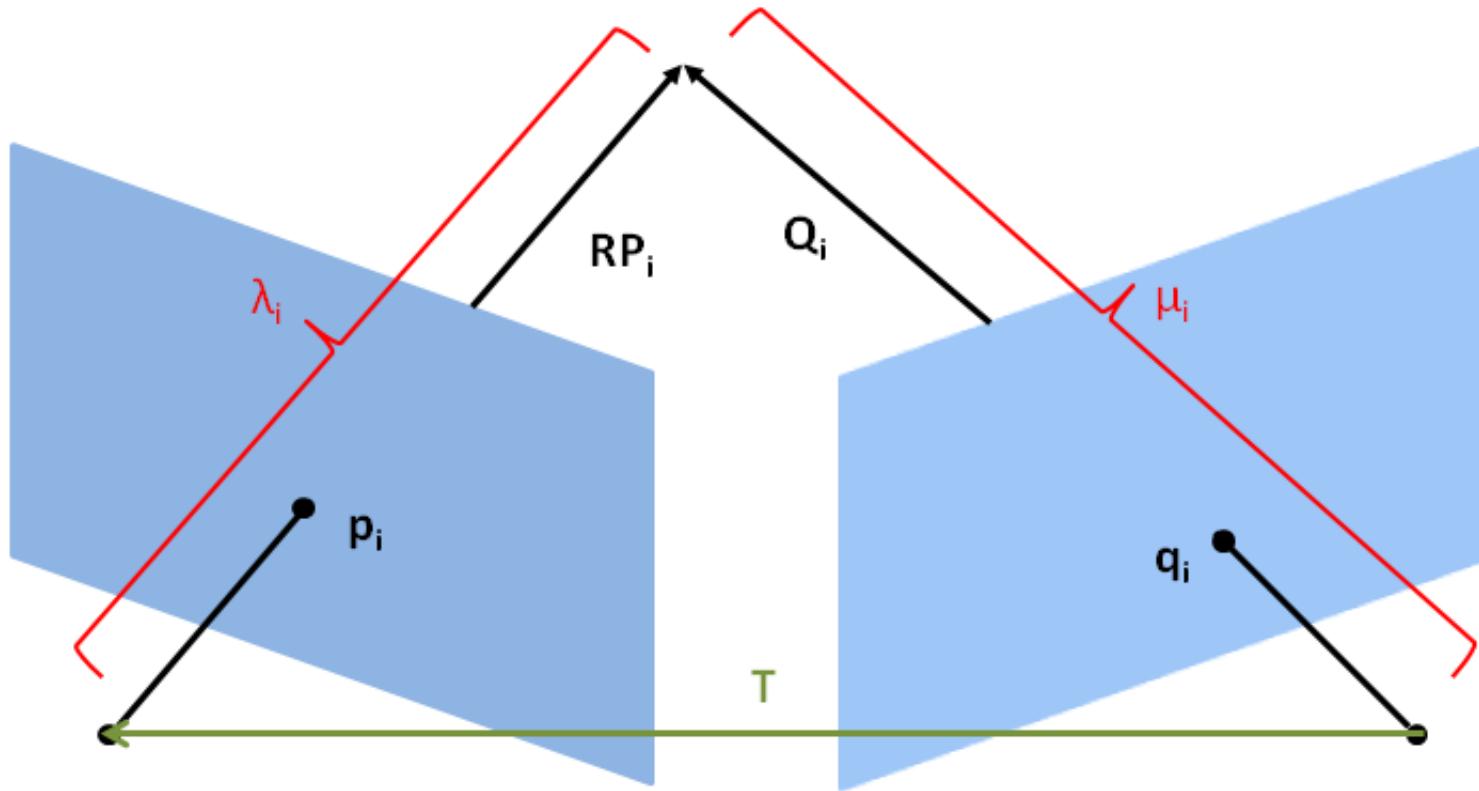


$$\lambda q = R\mu p + T$$

Given 2D correspondences  $(p, q)$

Find motion  $R, T$  and depths  $\lambda, \mu$ .

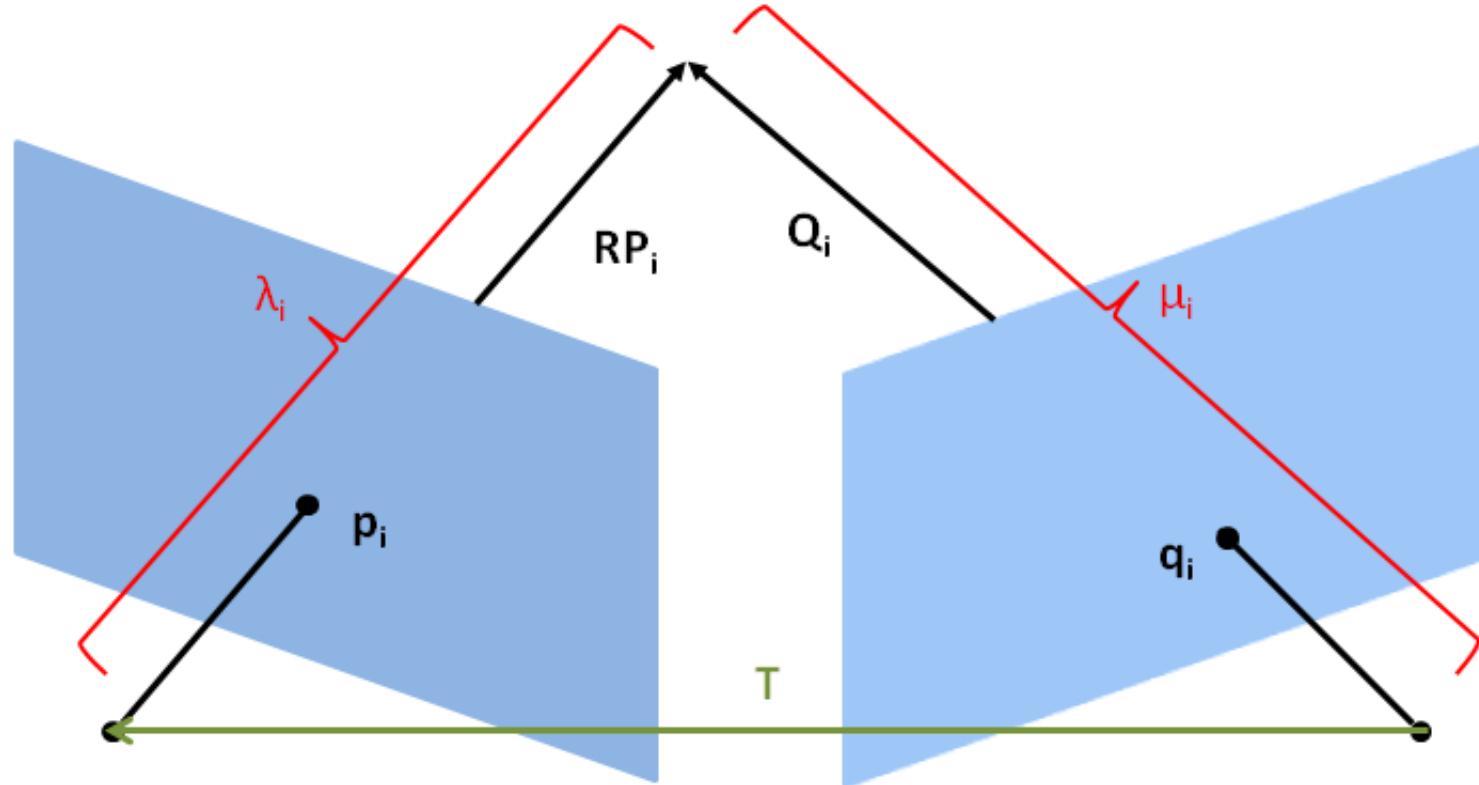
# Epipolar constraint



We can eliminate the depths from  $\lambda q = R\mu p + T$  and obtain the epipolar constraint:

$$q^T(T \times Rp) = 0$$

# Geometric meaning of the epipolar constraint

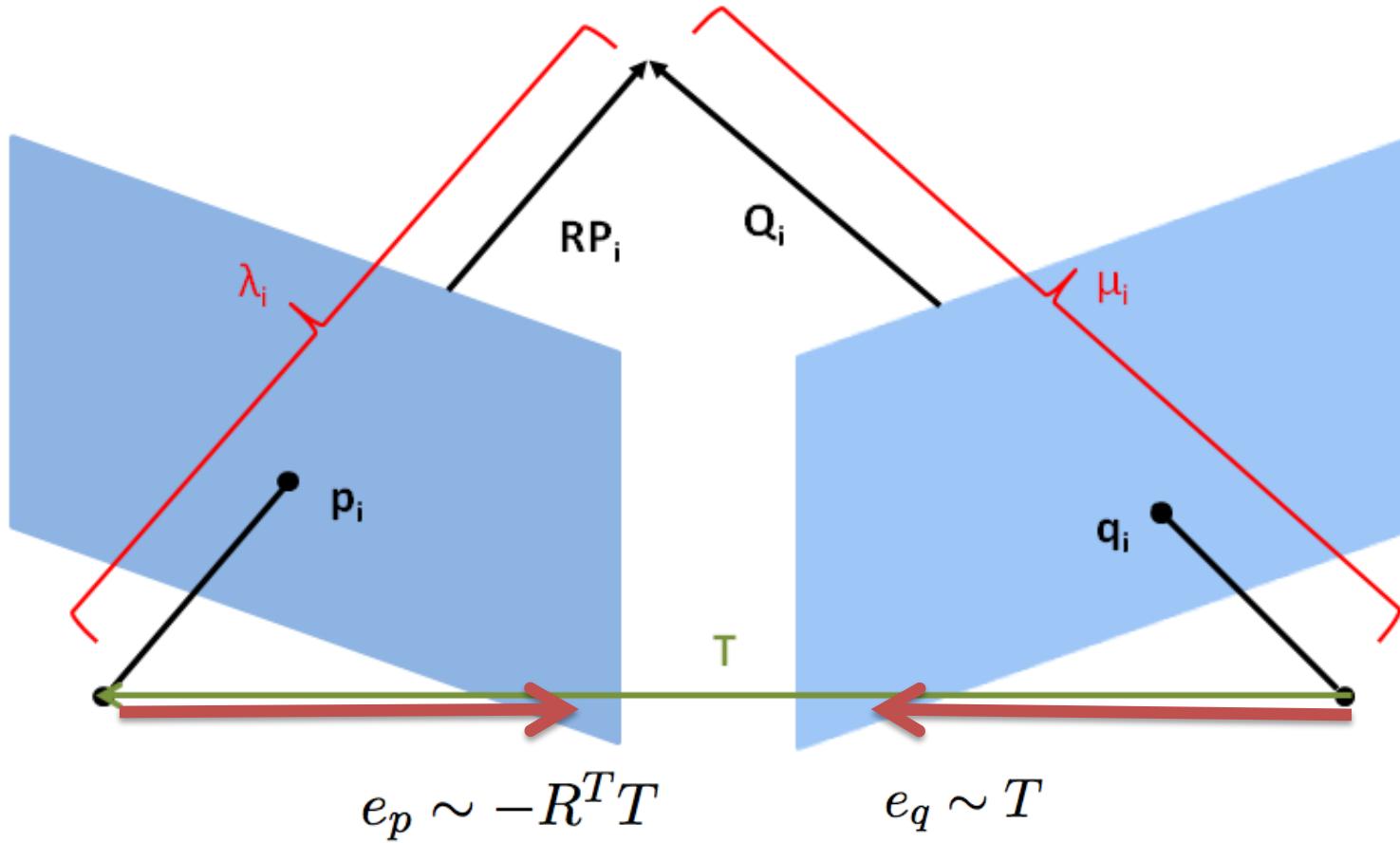


The two rays  $q$  and  $Rp$  intersect in space if and only if they are coplanar with the translation vector  $T$ . Three vectors are coplanar if their mixed product vanishes:

$$q^T(T \times Rp) = 0$$

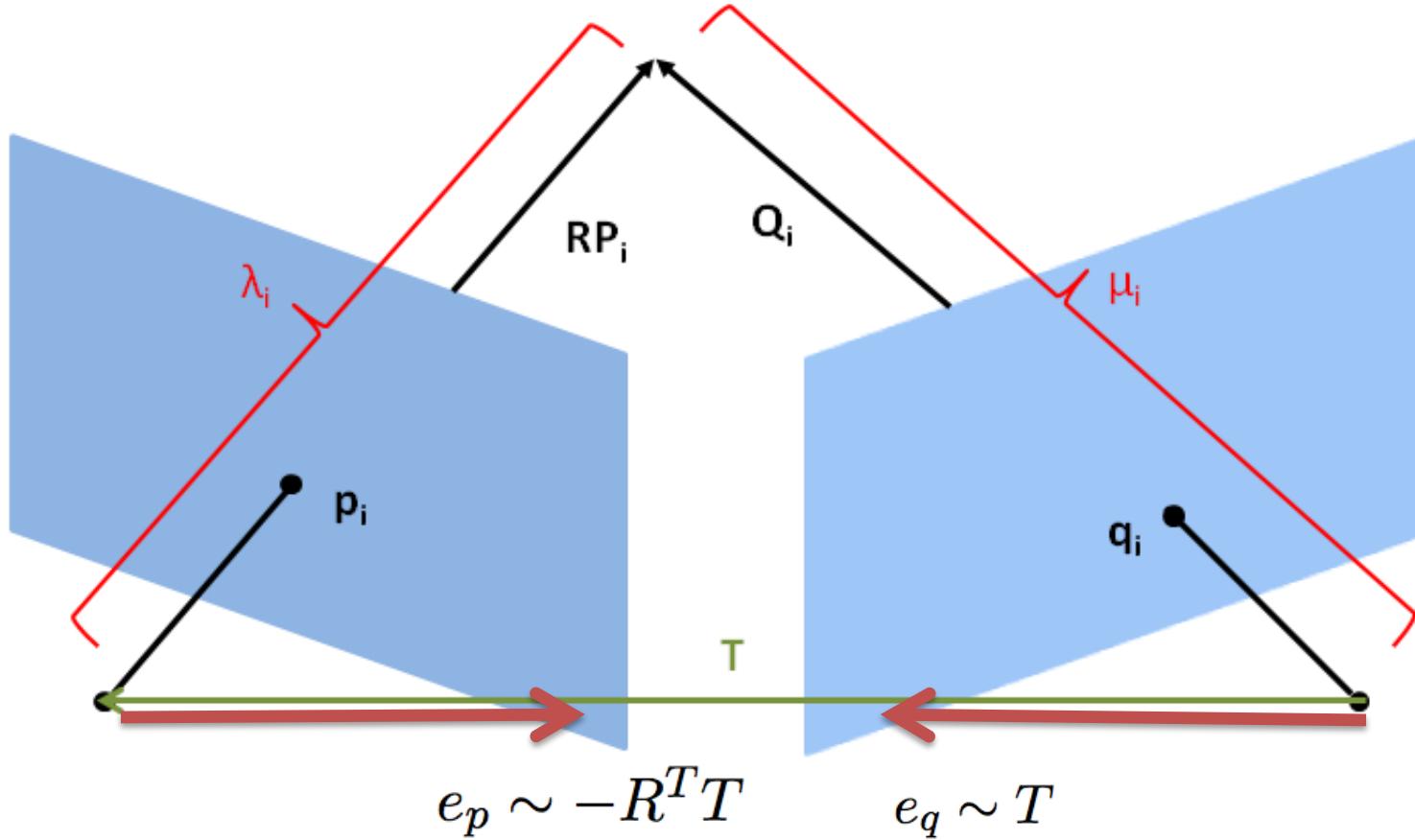
The plane spanned by the three vectors is called **epipolar** plane.

# Why is it called epipolar?



$e_p \sim -R^T T$  and  $e_q \sim T$  are the intersections of the baseline (translation) with the two image planes.

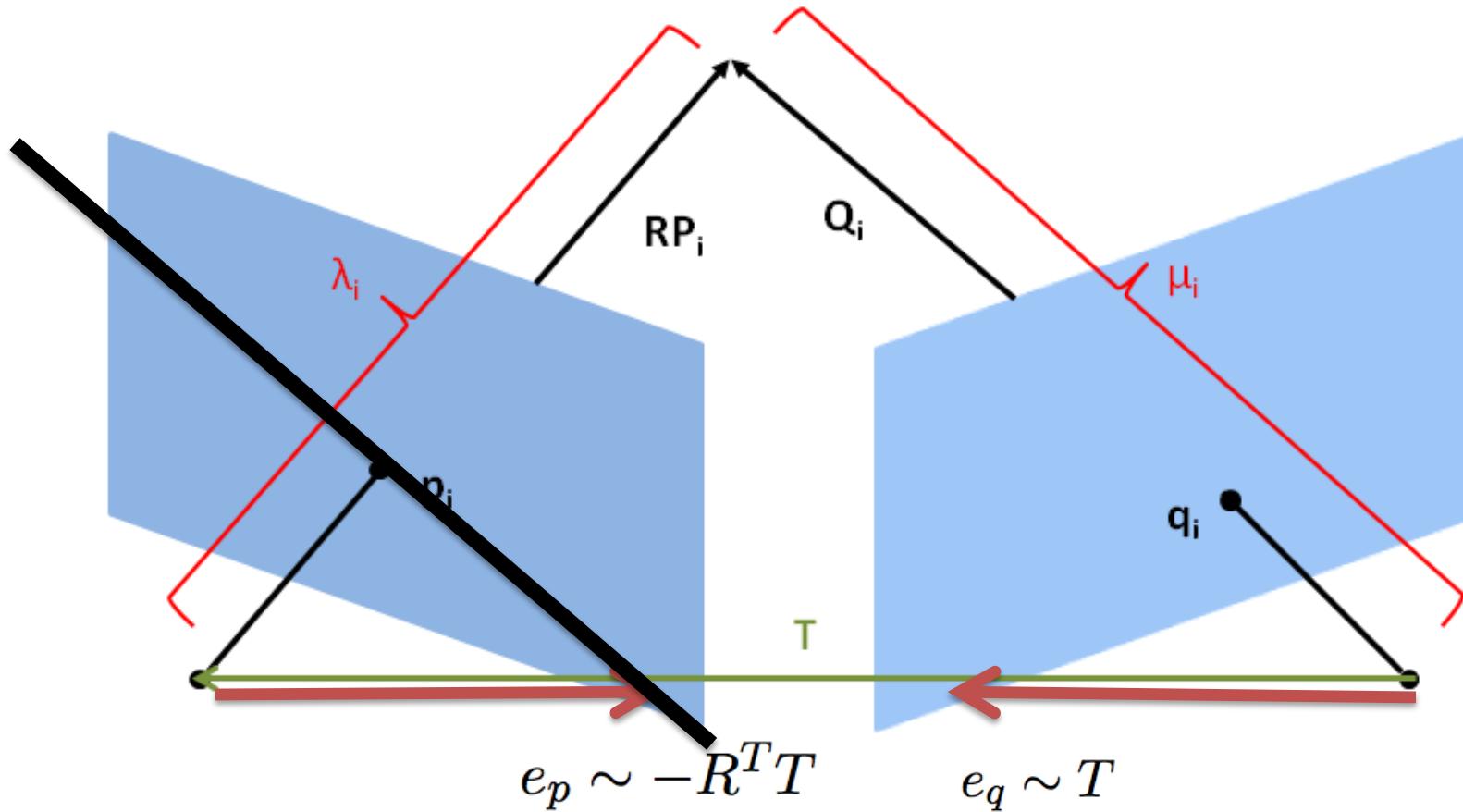
# Essential matrix



We can summarize the unknowns into a  $3 \times 3$  matrix  $E$  we will call the **essential** matrix:

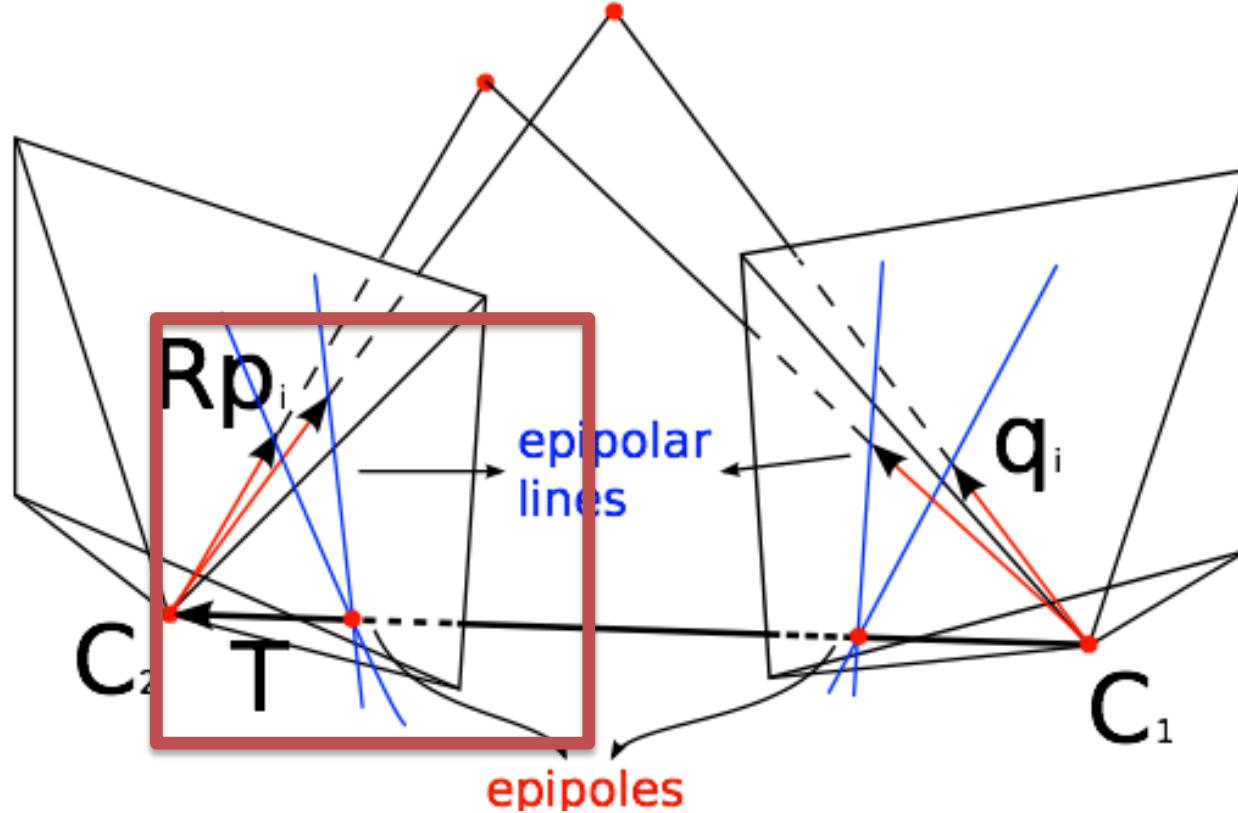
$$q^T E p = 0 \quad \text{where} \quad E = \hat{T} R$$

# Geometric properties



Equation  $q^T E p = 0$  is a line equation in the  $p$ -plane with line coefficients  $E^T q$ . It is called the epipolar line in  $p$ -plane.

# Epipolar pencil of lines

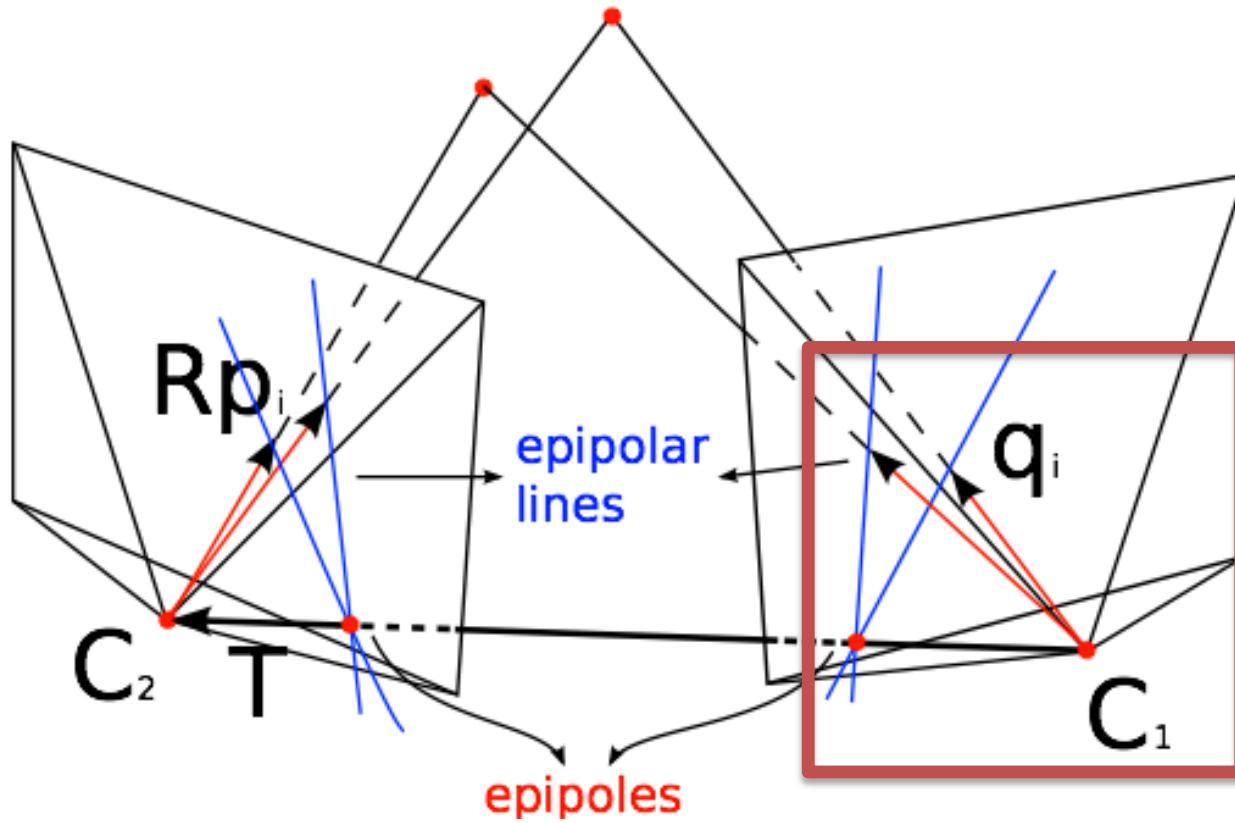


All epipolar lines  $q^T E p = 0$  in  $p$ -plane go through the epipole  $e_p$  because

$$E e_p = \hat{T} R (-R^T T) = \hat{T} T = T \times T = 0$$

They build a pencil of epipolar lines with one line per point correspondence.

# Epipolar pencil of lines

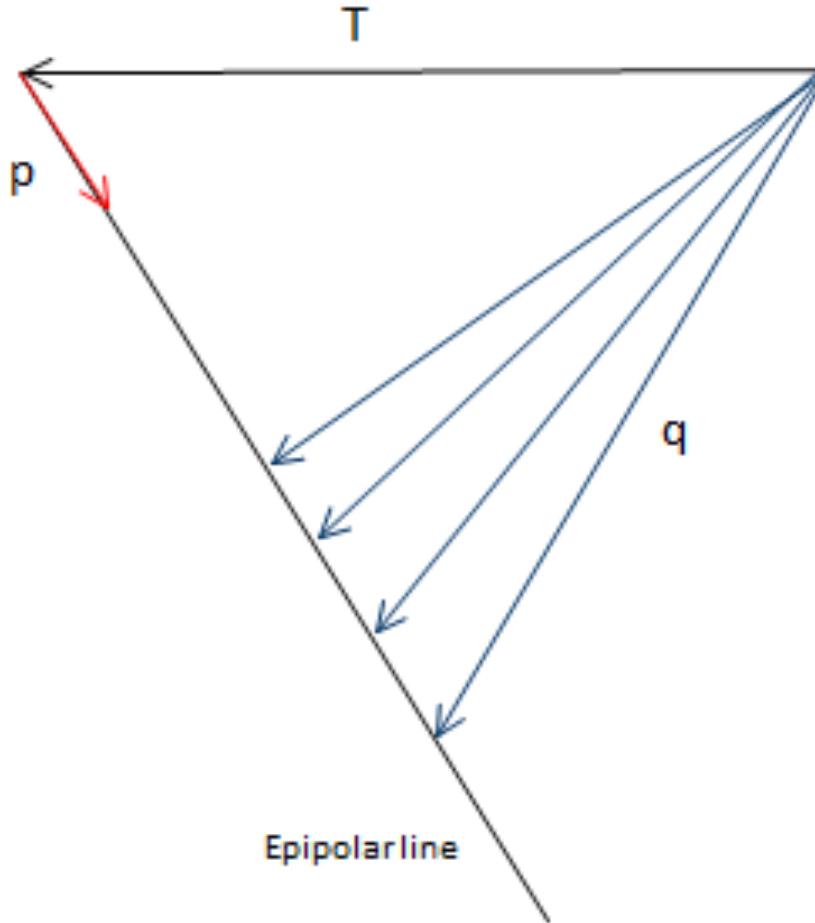


All epipolar lines  $p^T E^T q = 0$  in the  $q$ -plane go through the epipole  $e_q$  because

$$E^T e_p = R^T \hat{T}^T T = R(\hat{T}T) = R(T \times T) = 0$$

They also build a pencil of epipolar lines with one line per point correspondence.

# Simpler correspondence



Knowledge of the  $E$ -matrix allows us to search for points  $q$  corresponding to points  $p$  along the epipolar line, reducing correspondence to 1D-search.

Position of the corresponding point  $q$  along epipolar line varies with depth of the 3D points which is still constrained to lie on the ray through  $p$ .

# How can we compute the E-matrix?

If

$$E = (e_1 \ e_2 \ e_3)$$

then epipolar constraint can be rewritten as

$$\begin{aligned} q^T (e_1 \ e_2 \ e_3) \begin{pmatrix} p_x \\ p_y \\ p_z \end{pmatrix} &= q^T (p_x e_1 + p_y e_2 + p_z e_3) \\ &= (p_x q^T \ p_y q^T \ p_z q^T) \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = 0 \end{aligned}$$

This equation is linear and homogeneous in  $E' = \begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix}$ .

# The 8-point algorithm

Let  $\vec{a} = (p_x q^T \quad p_y q^T \quad p_z q^T)$

$$\begin{pmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_n^T \end{pmatrix} E' = 0$$

where  $a_i$  is the known  $1 \times 9$  vector of image points and  $E'$  is the essential matrix re-organized into a  $9 \times 1$  column vector.

$E'$  has to be in the null-space of  $\begin{pmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_n^T \end{pmatrix}$ .

# Properties of the Essential matrix

$$E^T = R^T \hat{T} T = 0$$

E is a singular matrix,  $\det(E) = 0$ .

$$\begin{aligned} EE^T &= \hat{T} \hat{T}^T \\ &= TT^T - T^T T I \\ &= \begin{bmatrix} t_x^2 & t_x t_y & t_x t_z \\ t_x t_y & t_y^2 & t_y t_z \\ t_x t_z & t_y t_z & t_z^2 \end{bmatrix} - \|T\|^2 I \end{aligned}$$

# Properties of the Essential matrix

$$\begin{aligned} EE^T &= \hat{T}\hat{T}^T \\ &= TT^T - T^TTI \\ &= \begin{bmatrix} t_x^2 & t_xt_y & t_xt_z \\ t_xt_y & t_y^2 & t_yt_z \\ t_xt_z & t_yt_z & t_z^2 \end{bmatrix} - \|T\|^2 I \end{aligned}$$

If we solve the characteristic polynomial  $\det(EE^T - \lambda I) = 0$  we will find two eigenvalues both equal to  $\|T\|^2$ .

# Properties of the Essential matrix

Recall that the singular values of  $E$  are the square-roots of the eigenvalues of  $EE^T$  if  $E$  is a square matrix.

Hence, we have proved that **if a matrix is essential, namely, can be decomposed as the product of an antisymmetric  $\hat{T}$  and a special orthogonal  $R$  then its singular values are  $\sigma_1 = \sigma_2 > 0$  and  $\sigma_3 = 0$ .**

This is helpful in checking of a matrix is essential but is not constructive on how to decompose it.

# Properties of the Essential matrix

We have to prove the sufficient condition:

**If the singular values of a matrix are are  $\sigma_1 = \sigma_2 > 0$  and  $\sigma_3 = 0$  then the matrix can be decomposed into the product of an antisymmetric  $\hat{T}$  and a special orthogonal  $R$ .**

We need a lemma !

If  $Q$  is orthogonal ( $Q^T Q = I$ ), then

$$\widehat{Qa} = Q\widehat{a}Q^T$$

**Proof:**  $\widehat{Qab} = Qa \times b = Q(a \times Q^T b) = Q\widehat{a}Q^T b.$

and the following simple fact

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = T_z^T R_{z,\pi/2}$$

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$$\begin{aligned}
E &= U \begin{bmatrix} \sigma & 0 & 0 \\ 0 & \sigma & 0 \\ 0 & 0 & 0 \end{bmatrix} V^T \\
&= \sigma U \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} V^T \\
&= \sigma \widehat{U T_z}^T R_z V^T \\
&= \sigma \underbrace{\widehat{U T_z}}_{\text{antisymmetric}} \underbrace{U R V^T}_{\text{orthogonal}}
\end{aligned}$$

Observe  $U T_z = U \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ , which is the last column of  $U$

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**Necessary and sufficient condition:**  $E$  is essential iff  
 $\sigma_1(E) = \sigma_2(E) \neq 0$  and  $\sigma_3(E) = 0$ .

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We just showed that there is at least one such decomposition  $\hat{T}R$ , but is it unique?

We showed the following decomposition:

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But we could similarly write  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \hat{T}_z R_{z,-\pi/2}$ .

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If  $E = U\Sigma V^T = U \begin{bmatrix} \sigma & 0 & 0 \\ 0 & \sigma & 0 \\ 0 & 0 & 0 \end{bmatrix} V^T$ , there are four solutions for the pair  $(\hat{T}, R)$ :

$$(\hat{T}_1, R_1) = (UR_{z,+π/2}\Sigma U^T, UR_{z,+π/2}^T V^T)$$

$$(\hat{T}_2, R_2) = (UR_{z,-π/2}\Sigma U^T, UR_{z,-π/2}^T V^T)$$

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Please remember that we have to make  $R$  have  $\det(R) = 0$ , see our Procrustes lecture.

**Mirror ambiguity:** If  $T$  is a solution, then  $-T$  is a solution, too. There is no way to disambiguate from the epipolar constraint:  $q^T(-T \times Rp) = 0$ .

**Twisted pair ambiguity:** If  $R$  is a solution, then also  $R_{T,\pi}R$  is a solution. The first image is “twisted” around the baseline 180 degrees.

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# The full two-view algorithm

- ① Build the homogeneous linear system by stacking epipolar constraints  $q_i^T(T \times Rp_i) = 0, i = 1, \dots, 8$ :

$$\begin{bmatrix} \vdots \\ (q_i \otimes p_i)^T \\ \vdots \end{bmatrix} \begin{bmatrix} e'_1 \\ e'_2 \\ e'_3 \end{bmatrix}$$

$A \ (8 \times 9)$

- ② Let  $\begin{bmatrix} e'_1 \\ e'_2 \\ e'_3 \end{bmatrix}$  be the nullspace of  $A$  (if  $\sigma_8 \approx 0$  give up)

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# The full two-view algorithm

- ③  $[ e'_1 \ e'_2 \ e'_3 ] = U \text{diag} (\sigma'_1, s'_2, \sigma'_3) V^T$ . Then use the following estimate of the essential matrix:

$$E = U \text{diag} \left( \frac{\sigma'_1 + \sigma'_2}{2}, \frac{\sigma'_1 + \sigma'_2}{2}, 0 \right) V^T$$

- ④  $T = \pm \hat{u}_3 \quad R = UR_{Z,\pi/2}V^T$  or  $R = R_{T,\pi}R$

- ⑤ Try all four pairs  $(T, R)$  to check if reconstructed points are **in front** of the cameras  $\lambda q = \mu Rp + T$  give  $\lambda, \mu > 0$ .

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Triangulation is possible if we have computed  $R$  and  $T$  but again up to a scale factor. Set  $\|T\| = 1$ :

$$\underbrace{(q_i - Rp_i)}_{3 \times 2} \underbrace{\begin{pmatrix} \mu_i \\ \lambda_i \end{pmatrix}}_{2 \times 1} = \underbrace{T}_{3 \times 1}$$

There are then 3 equations with 2 unknowns  $\lambda_i$  and  $\mu_i$  for each point.

Solve with pseudo-inverse.

# Structure and motion from two views

