

# Epipolar geometry

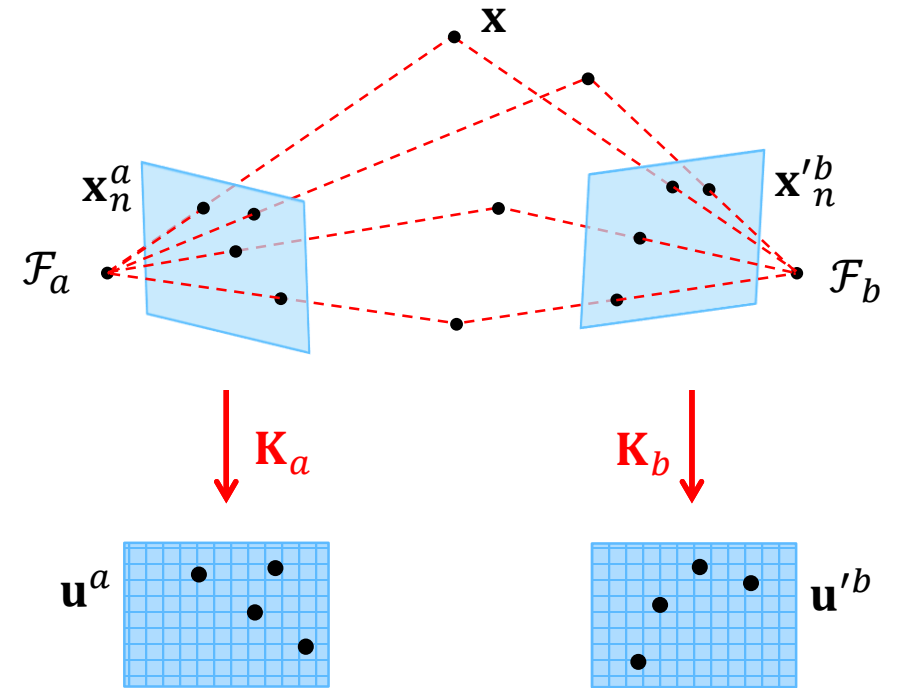
Thomas Opsahl

2023



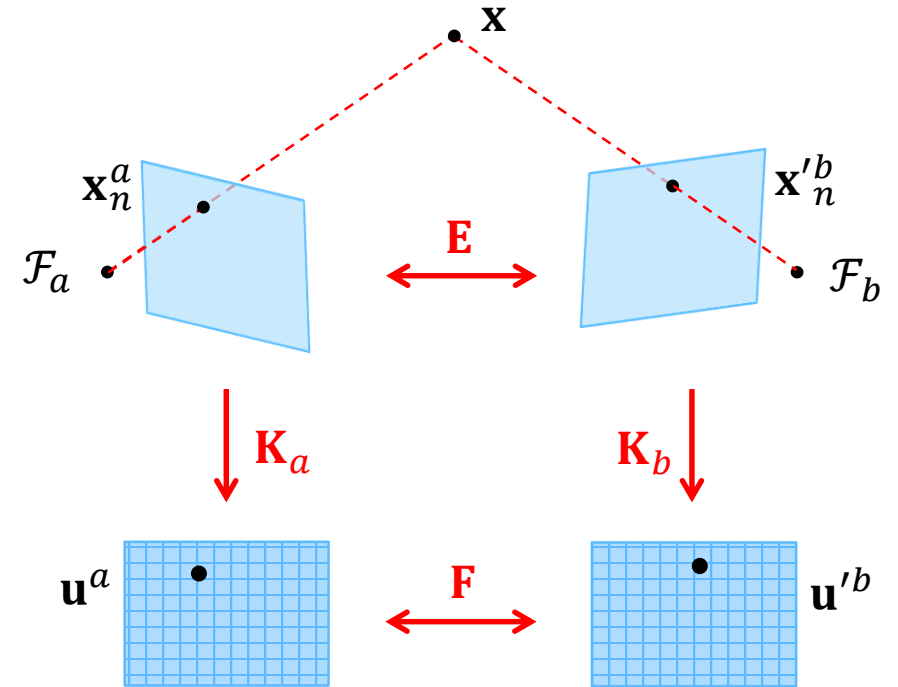
# Introduction

- Observing the same points with two cameras,  $\mathcal{F}_a$  and  $\mathcal{F}_b$ , puts a strong geometrical constraint on the point correspondences
- This epipolar constraint can be represented by two  $3 \times 3$  matrices

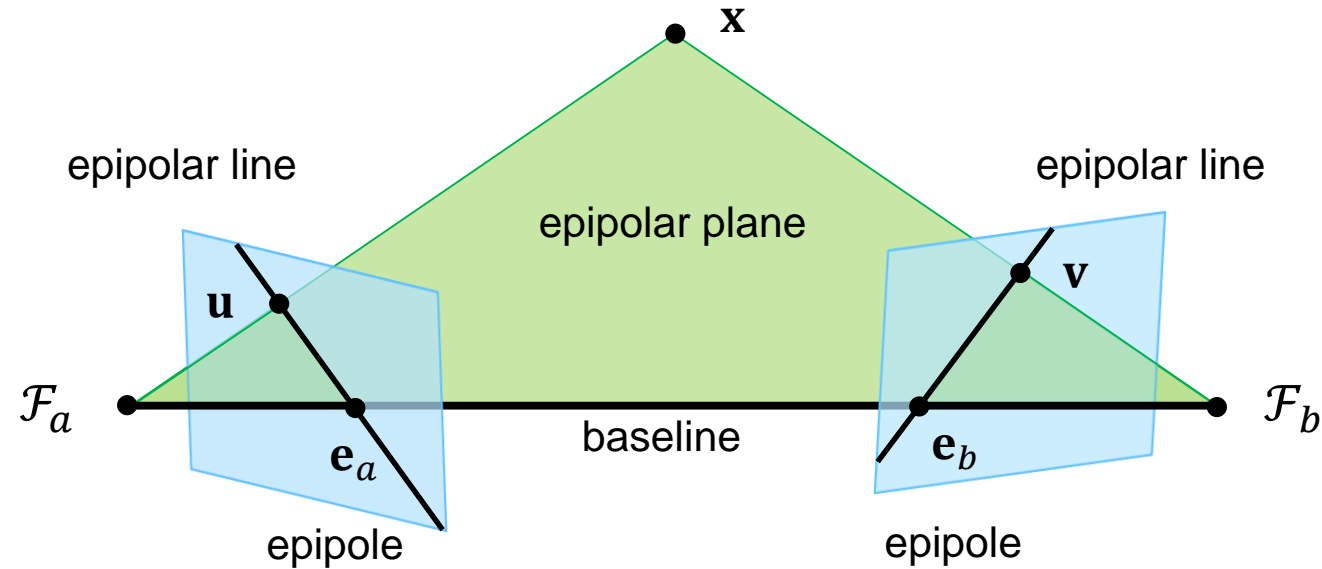


# Introduction

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- This epipolar constraint can be represented by two  $3 \times 3$  matrices
- The **essential matrix**  $\mathbf{E}$  represents the constraint for point correspondences  $\mathbf{x}_n^a \leftrightarrow \mathbf{x}'_n{}^b$
- The **fundamental matrix**  $\mathbf{F}$  represents the same constraint, but for point correspondences  $\mathbf{u}^a \leftrightarrow \mathbf{u}'^b$



# Introduction



- The **baseline** is the line joining the two camera centers
- The **epipolar plane** is the plane containing  $\mathbf{x}$  and the two camera centers  $\mathcal{F}_a$  and  $\mathcal{F}_b$
- The **epipolar lines** are where the epipolar plane intersects the image planes
- The **epipoles** are where the baseline intersects the two image planes
- Epipoles and epipolar lines can be represented in the normalized image plane as well as in the image

# Exploring the epipolar geometry

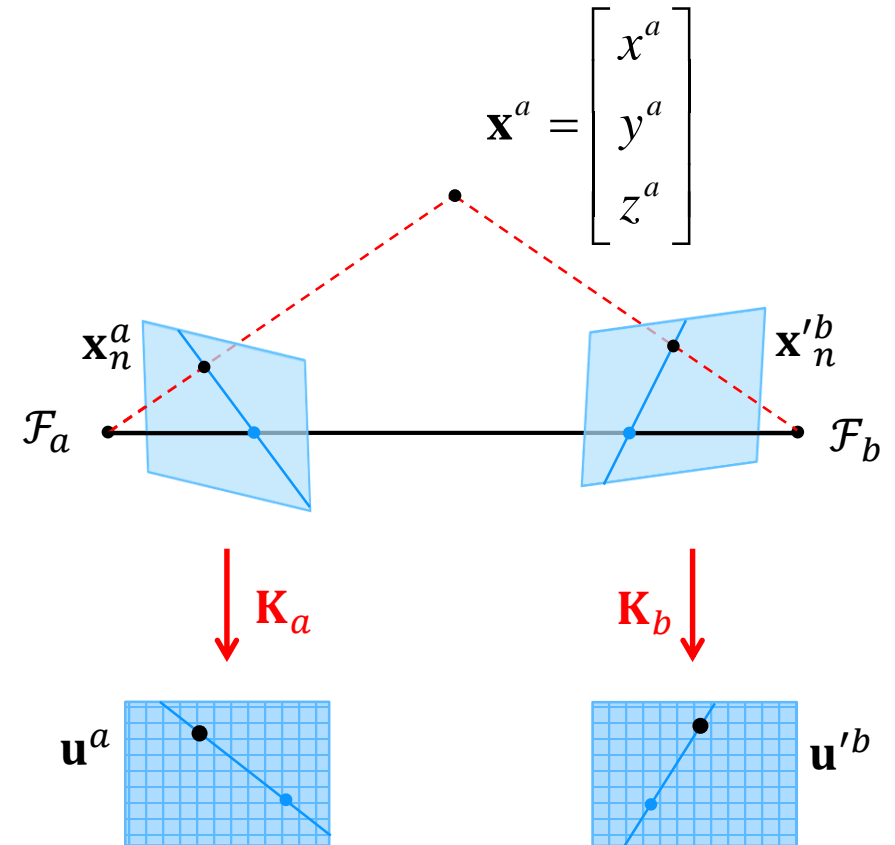
Let us consider two cameras,  $\mathcal{F}_a$  and  $\mathcal{F}_b$ , and let  $\mathcal{F}_a$  to be our “world frame”

Then we have the camera projection matrices

$$\mathbf{P}_a = \mathbf{K}_a [\mathbf{I} \quad \mathbf{0}]$$

$$\mathbf{P}_b = \mathbf{K}_b [\mathbf{R}_{ba} \quad \mathbf{t}_{ba}^b]$$

Assume that the two cameras project a 3D world point  $\mathbf{x}^a = [x^a \quad y^a \quad z^a]^T$  to  $\mathbf{u}^a$  and  $\mathbf{u}'^b$  correspondingly



# Exploring the epipolar geometry

Projecting  $\mathbf{x}$  into the first image yields

$$\tilde{\mathbf{u}}^a = \mathbf{K}_a [\mathbf{I} \quad \mathbf{0}] \tilde{\mathbf{x}}^a$$

$$\tilde{\mathbf{u}}^a = \mathbf{K}_a \mathbf{x}^a$$

$$\tilde{\mathbf{x}}^a = \begin{bmatrix} \mathbf{x}^a \\ 1 \end{bmatrix}$$

So up to scale we know that

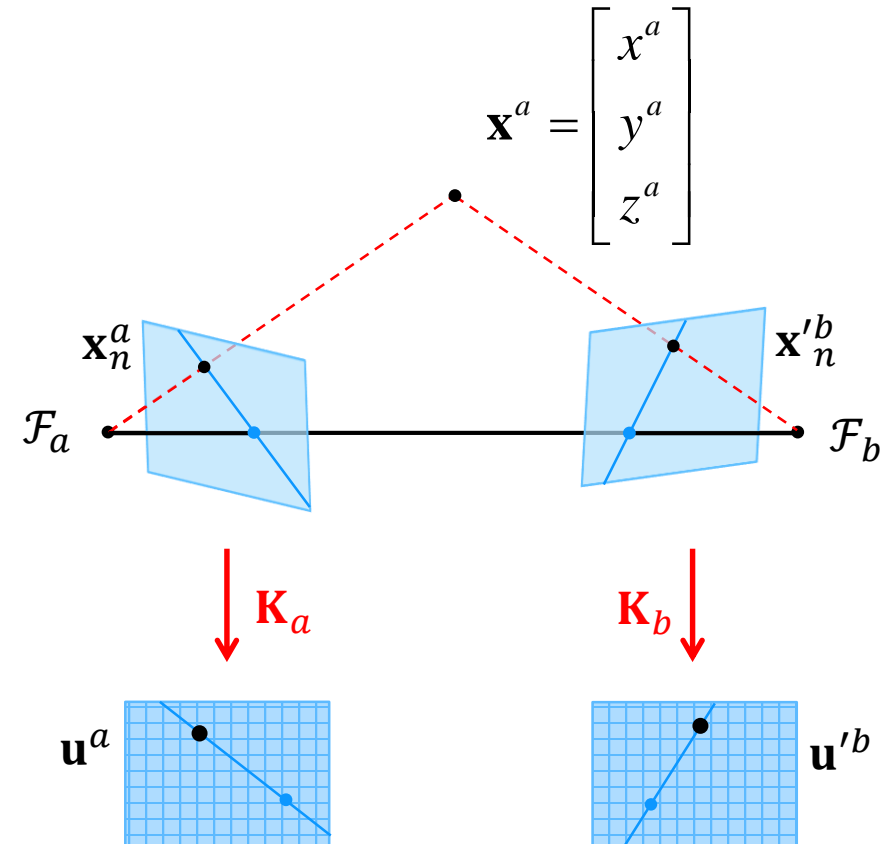
$$\mathbf{x}^a = \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a$$

equal up to scale

But given that  $\mathbf{x}^a = [x^a \quad y^a \quad z^a]^T$ , we also know the scale

$$\mathbf{x}^a = z^a \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a$$

truly equal



# Exploring the epipolar geometry

Projecting  $\mathbf{x}$  into the second image yields

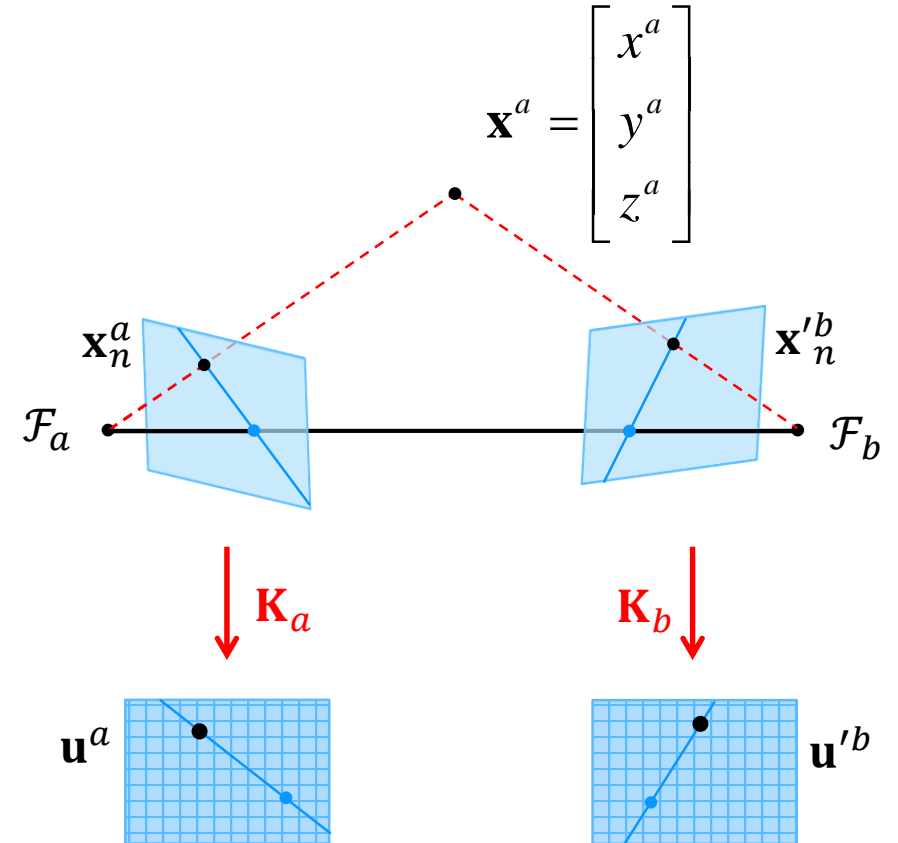
$$\tilde{\mathbf{u}}'^b = \mathbf{K}_b \begin{bmatrix} \mathbf{R}_{ba} & \mathbf{t}_{ba}^b \end{bmatrix} \tilde{\mathbf{x}}^a$$

$$\tilde{\mathbf{u}}'^b = \mathbf{K}_b \left( \mathbf{R}_{ba} \mathbf{x}^a + \mathbf{t}_{ba}^b \right)$$

$$\tilde{\mathbf{u}}'^b = \mathbf{K}_b \mathbf{R}_{ba} \mathbf{x}^a + \mathbf{K}_b \mathbf{t}_{ba}^b$$

$$\tilde{\mathbf{x}}^a = \begin{bmatrix} \mathbf{x}^a \\ 1 \end{bmatrix}$$

equal up to scale



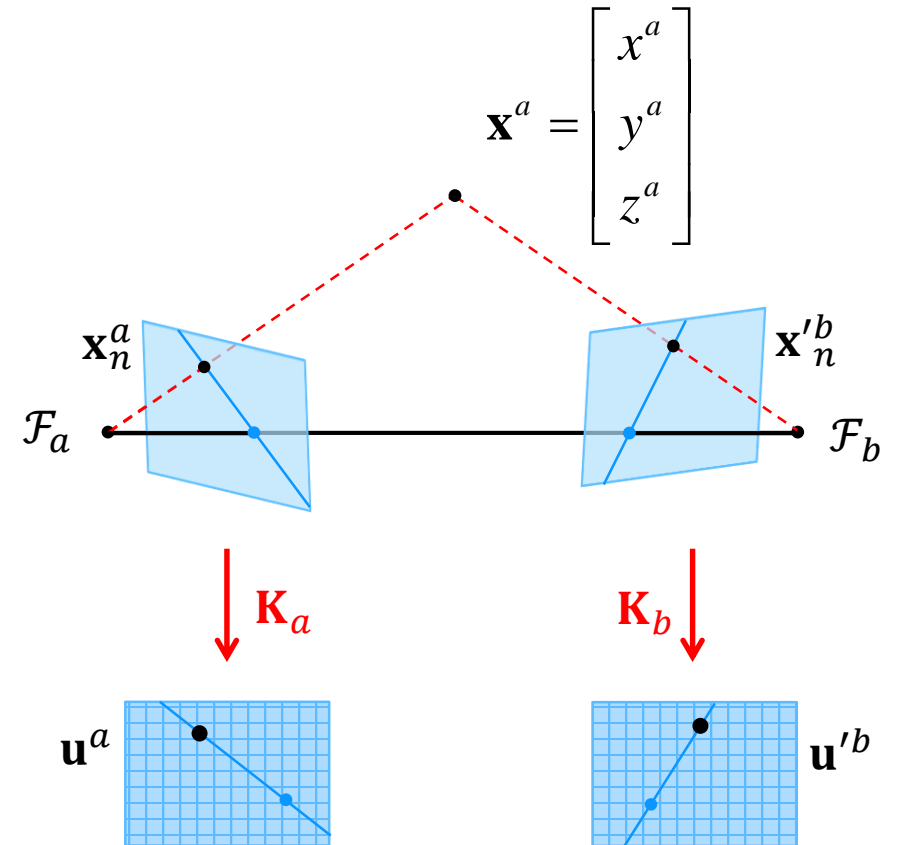
# Exploring the epipolar geometry

Combining these two results gives us

$$\tilde{\mathbf{u}}'^b = z^a \mathbf{K}_b \mathbf{R}_{ba} \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a + \mathbf{K}_b \mathbf{t}_{ba}^b$$

$$\tilde{\mathbf{u}}'^b = \mathbf{K}_b \mathbf{R}_{ba} \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a + \frac{1}{z^a} \mathbf{K}_b \mathbf{t}_{ba}^b$$

equal up to scale





# Exploring the epipolar geometry

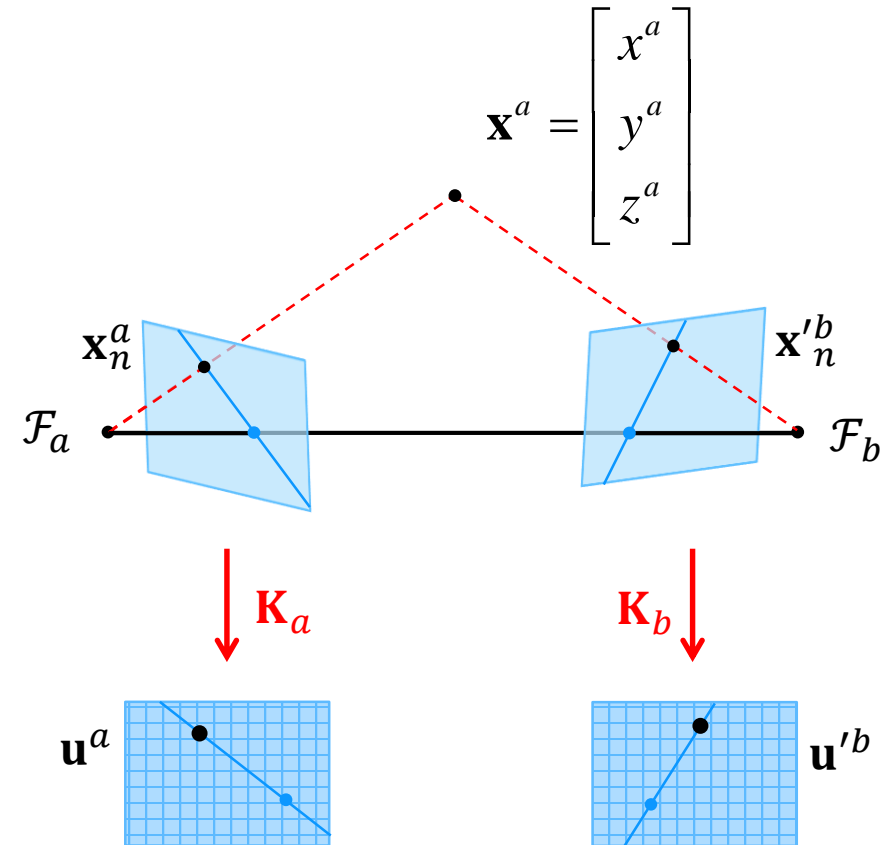
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equal up to scale

This describes how the position of  $\mathbf{u}'^b$  on the epipolar line varies with the depth  $z^a$  of the observed world point  $\mathbf{x}^a$



# Exploring the epipolar geometry

Combining these two results gives us

$$\tilde{\mathbf{u}}'^b = z^a \mathbf{K}_b \mathbf{R}_{ba} \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a + \mathbf{K}_b \mathbf{t}_{ba}^b$$

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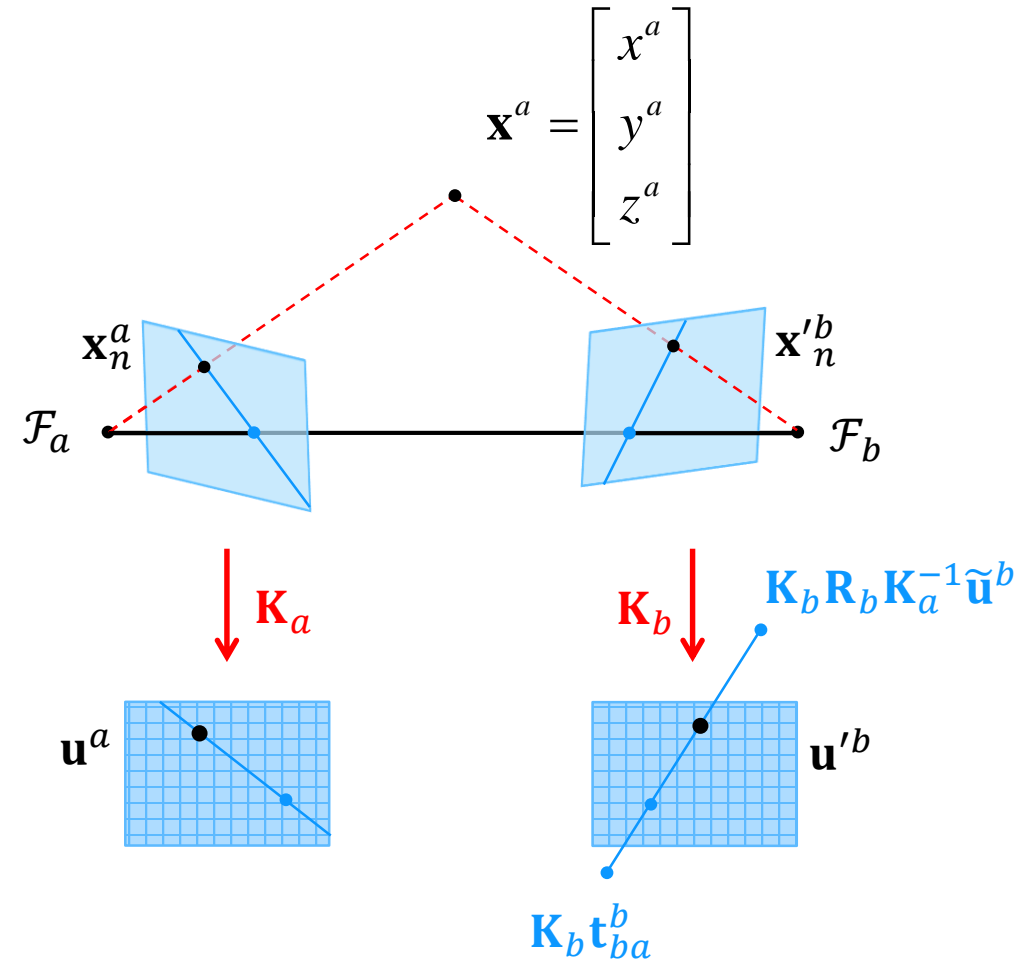
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This describes how the position of  $\mathbf{u}'^b$  on the epipolar line varies with the depth  $z^a$  of the observed world point  $\mathbf{x}^a$

It is clear that  $\tilde{\mathbf{u}}'^b$  is naturally restricted to an interval on the epipolar line

$$z^a = 0 \quad \Rightarrow \quad \tilde{\mathbf{u}}'^b = \mathbf{K}_b \mathbf{t}_{ba}^b$$

$$z^a = \infty \quad \Rightarrow \quad \tilde{\mathbf{u}}'^b = \mathbf{K}_b \mathbf{R}_{ba} \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a$$



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Combining these two results gives us

$$\tilde{\mathbf{u}}'^b = z^a \mathbf{K}_b \mathbf{R}_{ba} \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a + \mathbf{K}_b \mathbf{t}_{ba}^b$$

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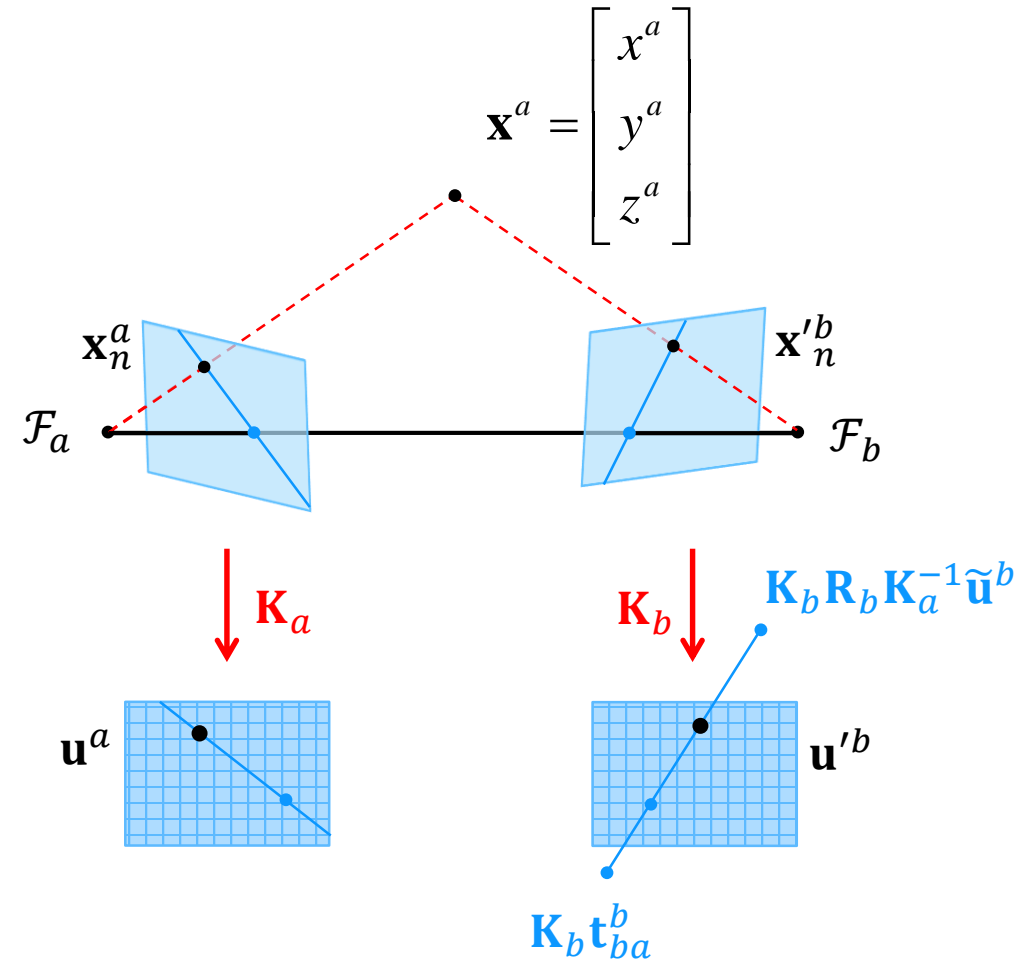
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Generalized disparity:  
 $d = \|\mathbf{u}^b - \mathbf{u}_\infty^b\|$  where  $\tilde{\mathbf{u}}_\infty^b = \mathbf{K}_b \mathbf{R}_{ba} \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a$

# Exploring the epipolar geometry

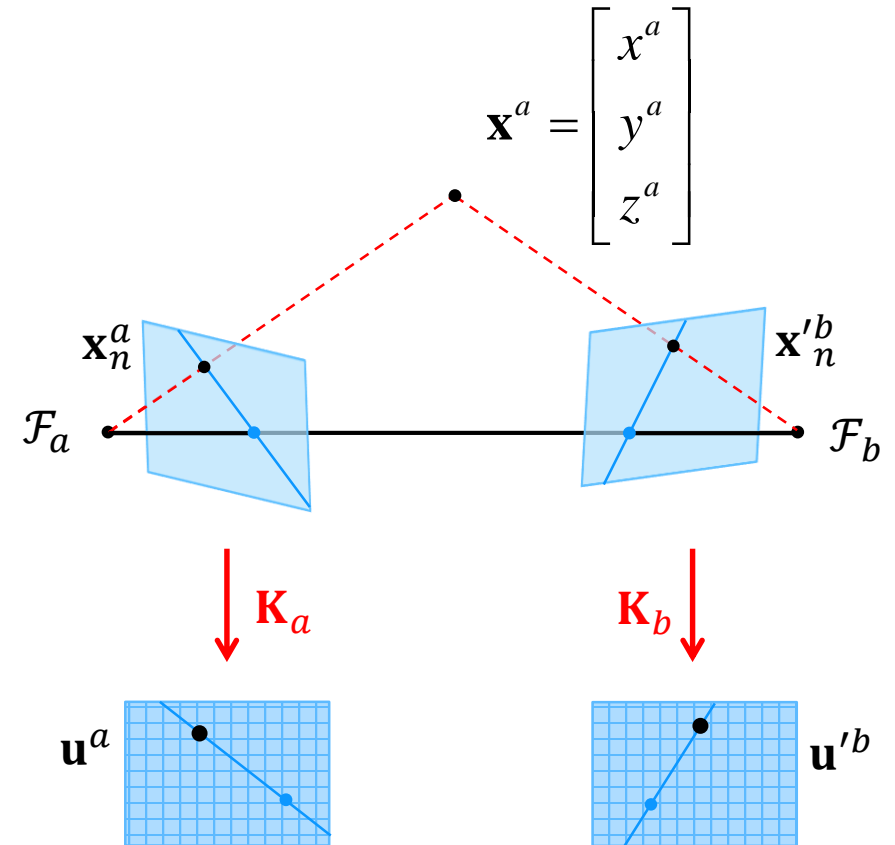
Since  $\tilde{\mathbf{u}}^a = \mathbf{K}_b \mathbf{R}_b \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^b$  for any correspondence  $\mathbf{u}^a \leftrightarrow \mathbf{u}'^b$  for a “far away” 3D point  $\mathbf{x}^a$ , it is clear that

Two overlapping perspective images of a “far away” scene is related by the homography

$$\mathbf{H}_{ba} = \mathbf{K}_b \mathbf{R}_{ba} \mathbf{K}_a^{-1}$$

The same is obviously true when  $\mathcal{F}_b$  is just a rotation of  $\mathcal{F}_a$ , i.e. when  $\mathbf{t}_{ba}^b = \mathbf{0}$

This explains why it is easy to coregister images of distant scenes even when the camera motion is not a pure rotation



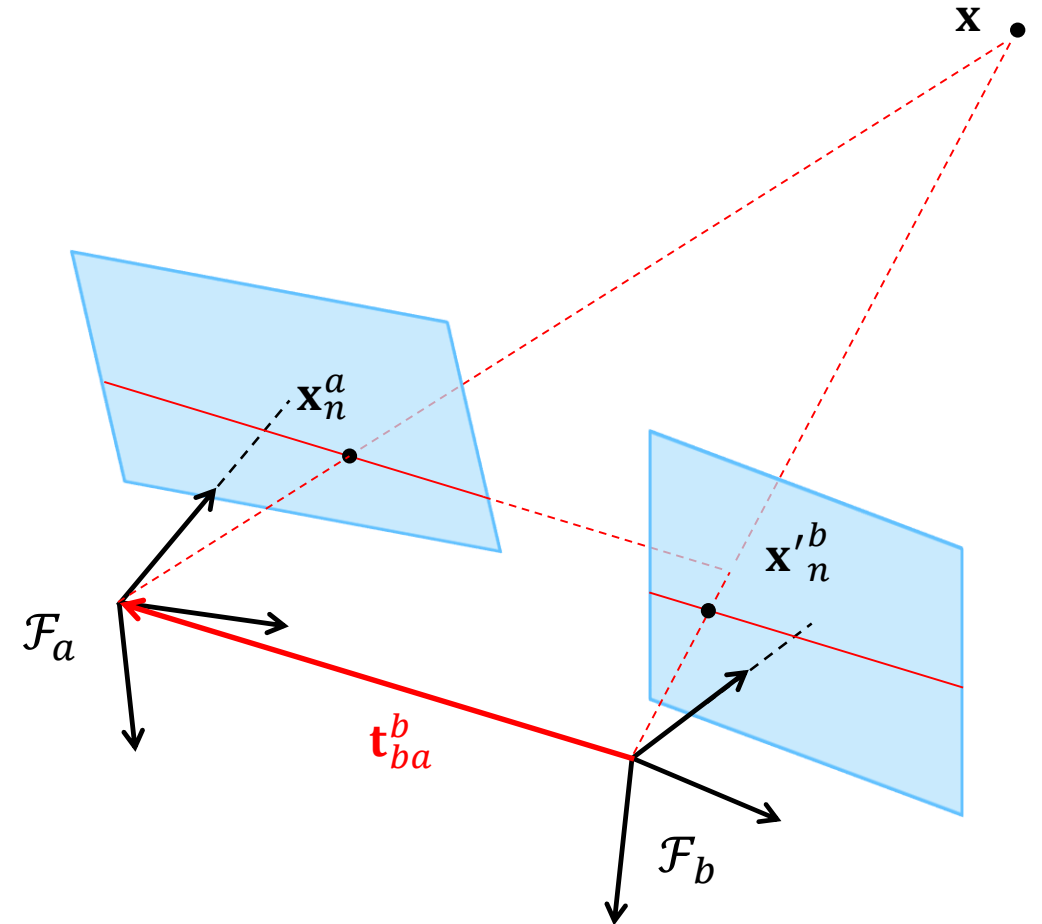
“far away”  $\Leftrightarrow z^a \gg \|\mathbf{K}_b \mathbf{t}_{ba}^b\|$

# Describing the epipolar geometry

Let  $\mathbf{x}$  project to  $\mathbf{x}_n^a$  in the normalized image plane of  $\mathcal{F}_a$   
and  $\mathbf{x}_n^b$  in that of  $\mathcal{F}_b$

Let the pose of  $\mathcal{F}_a$  relative to  $\mathcal{F}_b$  be

$$\mathbf{T}_{ba} = \begin{bmatrix} \mathbf{R}_{ba} & \mathbf{t}_{ba}^b \\ \mathbf{0} & 1 \end{bmatrix}$$



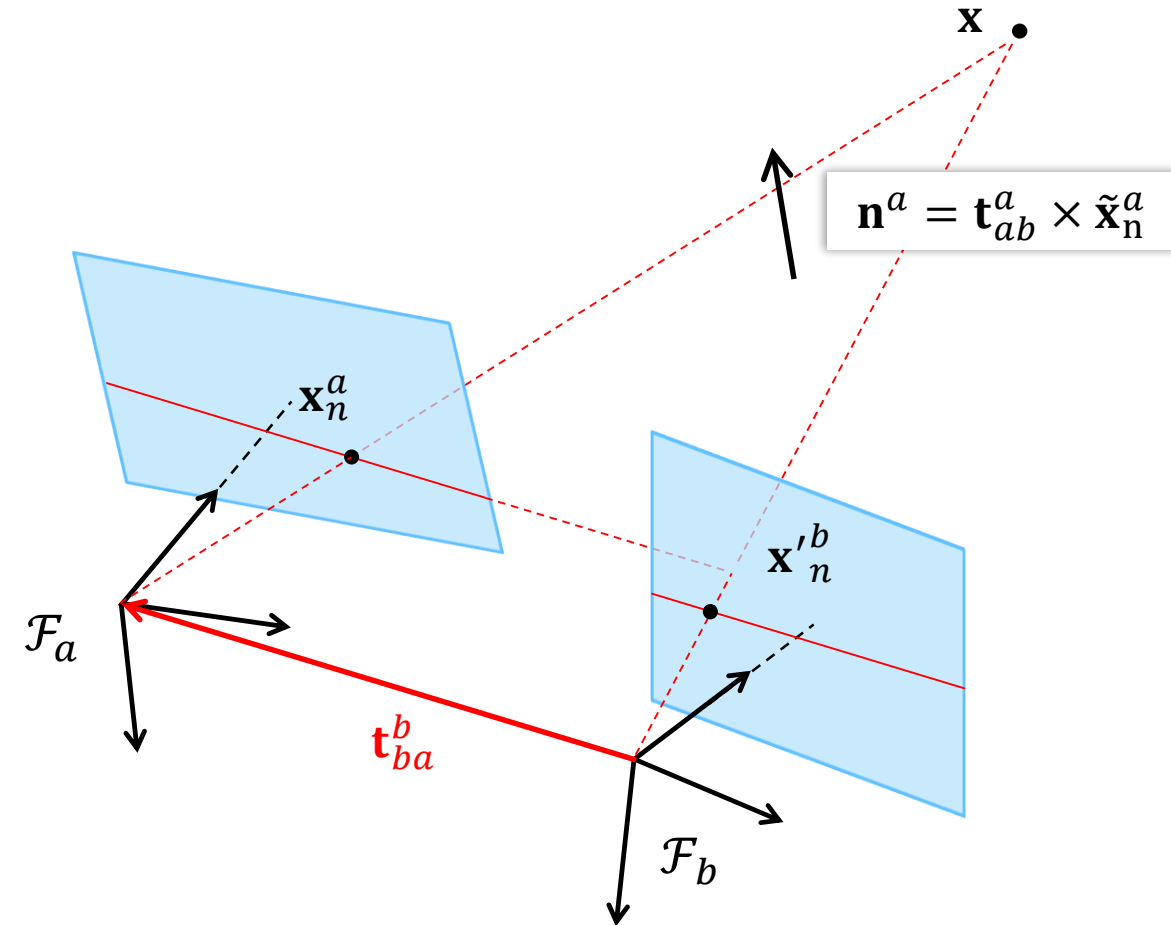
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It is clear from the illustration that  $\mathbf{n}^a = [\mathbf{t}_{ab}^a] \times \tilde{\mathbf{x}}_n^a$



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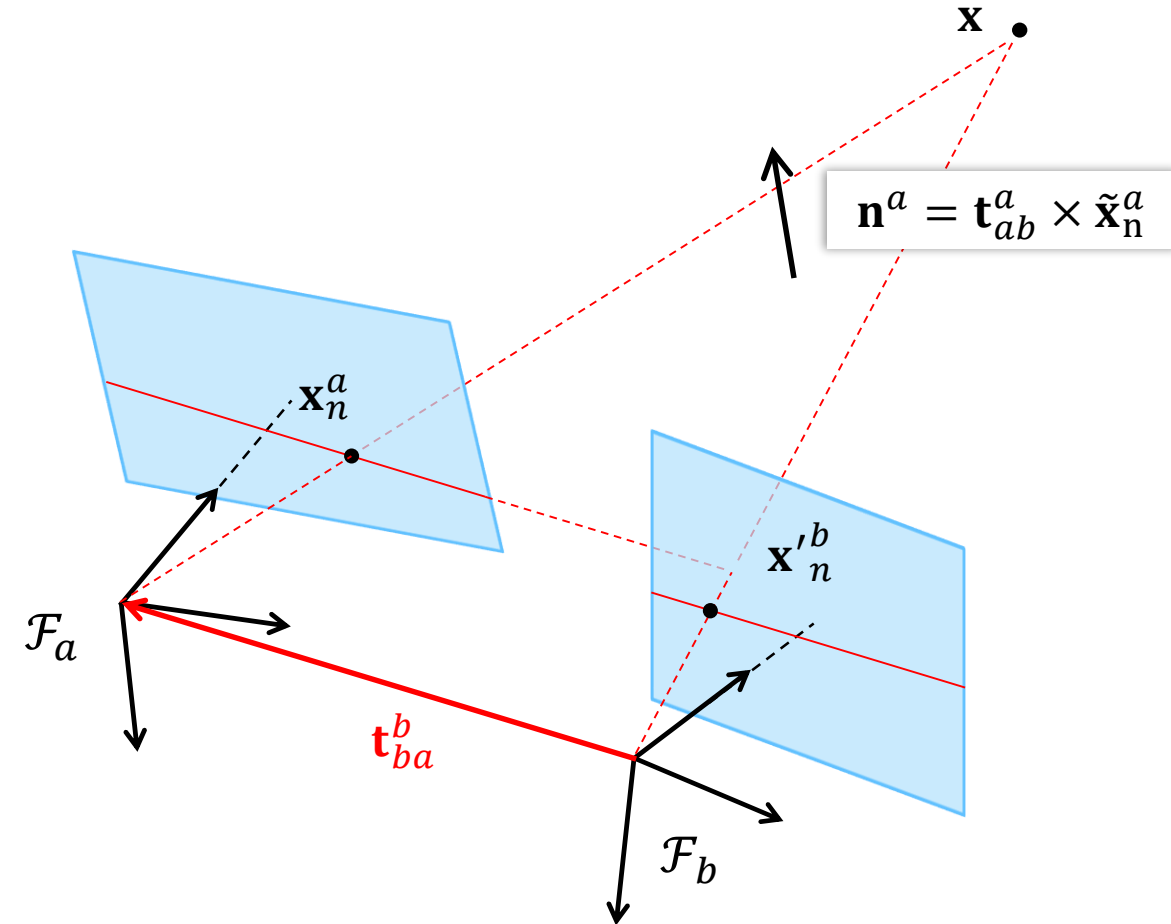
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It is clear from the illustration that  $\mathbf{n}^a = [\mathbf{t}_{ab}^a] \times \tilde{\mathbf{x}}_n^a$   
Transformed to  $\mathcal{F}_b$ , this becomes

$$\mathbf{n}^b = \mathbf{R}_{ba} [\mathbf{t}_{ab}^a] \times \tilde{\mathbf{x}}_n^a$$



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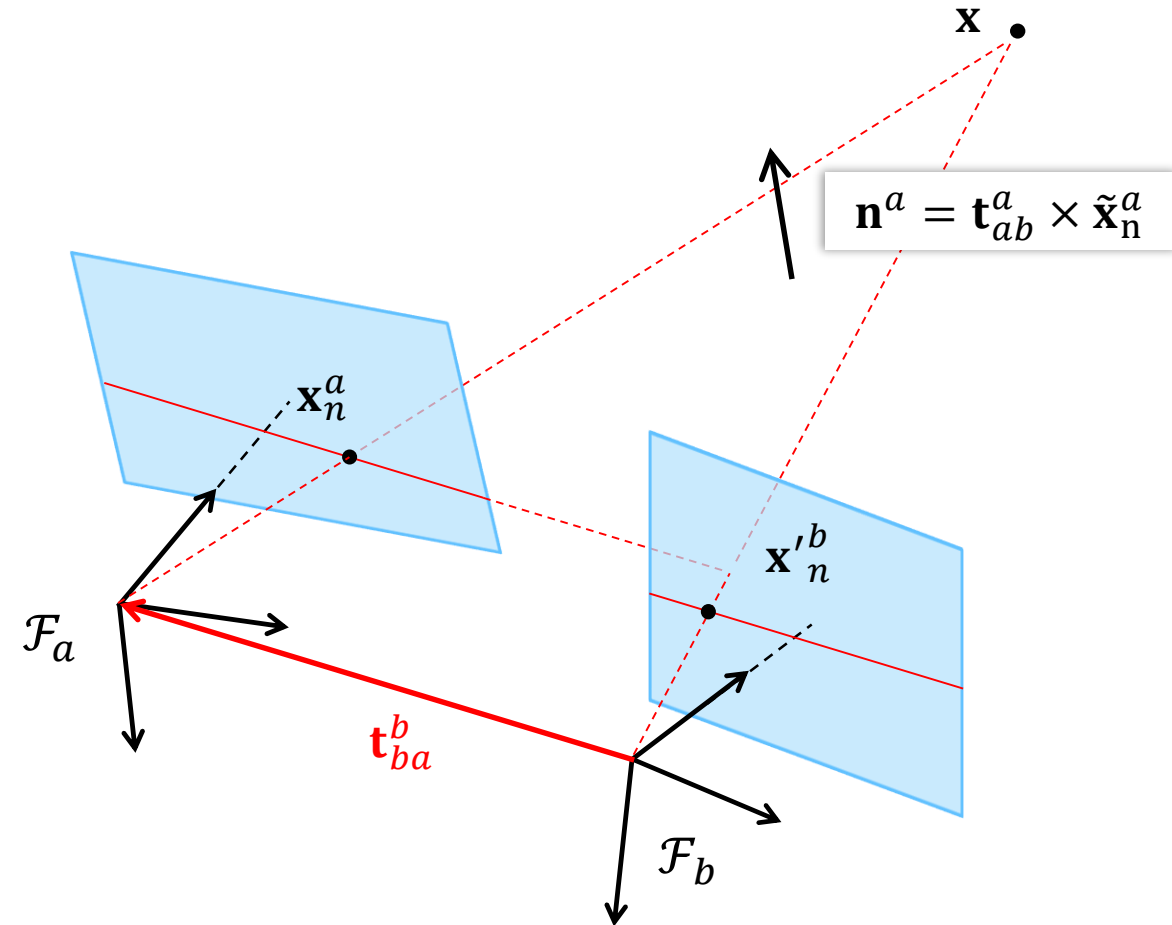
$$\mathbf{T}_{ba} = \begin{bmatrix} \mathbf{R}_{ba} & \mathbf{t}_{ba}^b \\ \mathbf{0} & 1 \end{bmatrix}$$

It is clear from the illustration that  $\mathbf{n}^a = [\mathbf{t}_{ab}^a]_{\times} \tilde{\mathbf{x}}_n^a$   
 Transformed to  $\mathcal{F}_b$ , this becomes

$$\mathbf{n}^b = \mathbf{R}_{ba} [\mathbf{t}_{ab}^a]_{\times} \tilde{\mathbf{x}}_n^a$$

The equation of the epipolar plane relative to  $\mathcal{F}_b$  is  $\mathbf{n}^{bT} \mathbf{x}^b = 0$ , so in particular we know that

$$\mathbf{n}^{bT} \tilde{\mathbf{x}}_n'^b = 0$$





# Describing the epipolar geometry

Combining these observations, we get a general constraint on the relationship between  $\mathbf{x}_n^a$  and  $\mathbf{x}_n^b$

$$(\mathbf{R}_{ba} [\mathbf{t}_{ab}^a]_{\times} \tilde{\mathbf{x}}_n^a)^T \tilde{\mathbf{x}}_n^b = 0$$

$$\tilde{\mathbf{x}}_n^{aT} [\mathbf{t}_{ab}^a]^T \mathbf{R}_{ba}^T \tilde{\mathbf{x}}_n^b = 0$$

$$\tilde{\mathbf{x}}_n^{aT} [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab} \tilde{\mathbf{x}}_n^b = 0$$

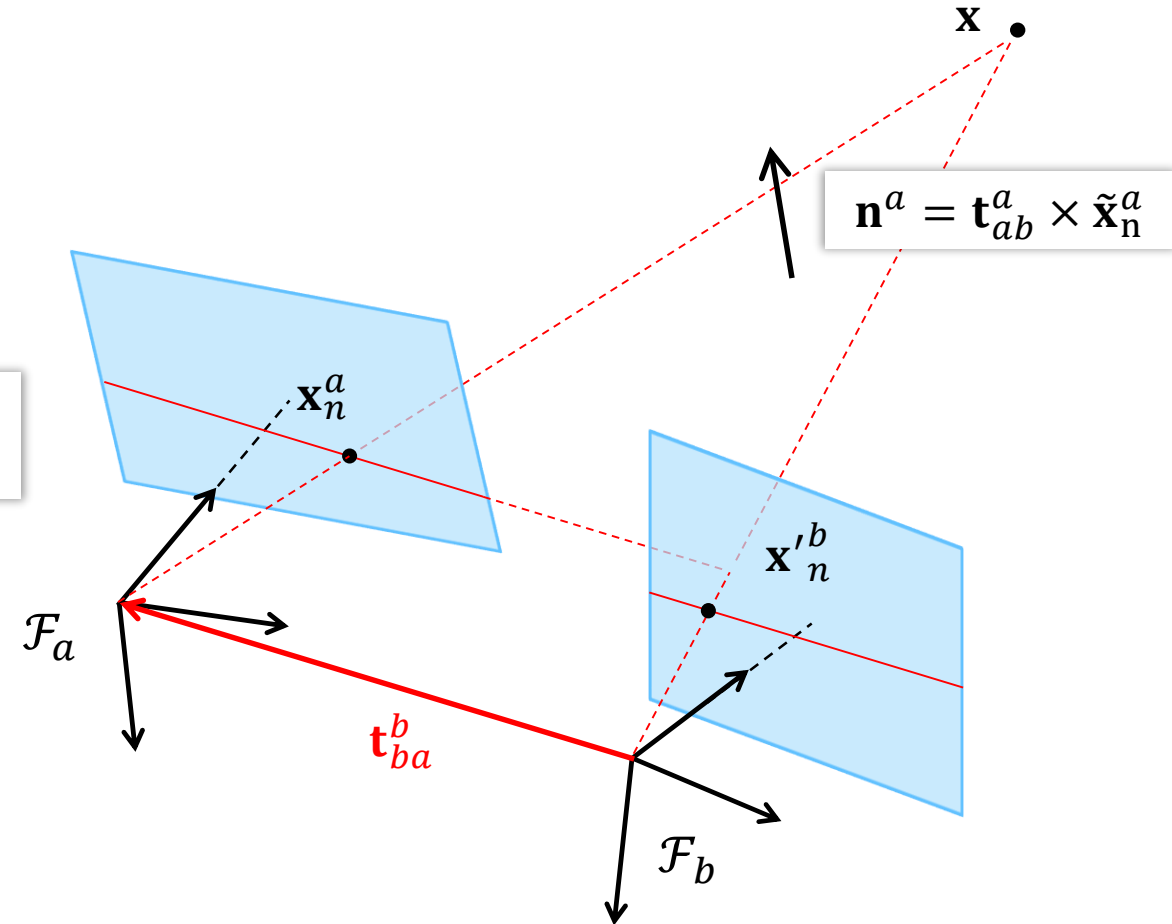
$$\begin{aligned} [\mathbf{t}_{ab}^a]^T &= -[\mathbf{t}_{ab}^a]_{\times} \\ \mathbf{R}_{ba}^T &= \mathbf{R}_{ab} \end{aligned}$$

This is known as the **epipolar constraint**

$$\tilde{\mathbf{x}}_n^{aT} \mathbf{E}_{ab} \tilde{\mathbf{x}}_n^b = 0$$

Where we define the **essential matrix** to be

$$\mathbf{E}_{ab} = [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab}$$



# Describing the epipolar geometry

Combining these observations, we get a general constraint on the relationship between  $\mathbf{x}_n^a$  and  $\mathbf{x}_n^b$

$$(\mathbf{R}_{ba} [\mathbf{t}_{ab}^a]_{\times} \tilde{\mathbf{x}}_n^a)^T \tilde{\mathbf{x}}_n^b = 0$$

$$\tilde{\mathbf{x}}_n^a T [\mathbf{t}_{ab}^a]^T \mathbf{R}_{ba}^T \tilde{\mathbf{x}}_n^b = 0$$

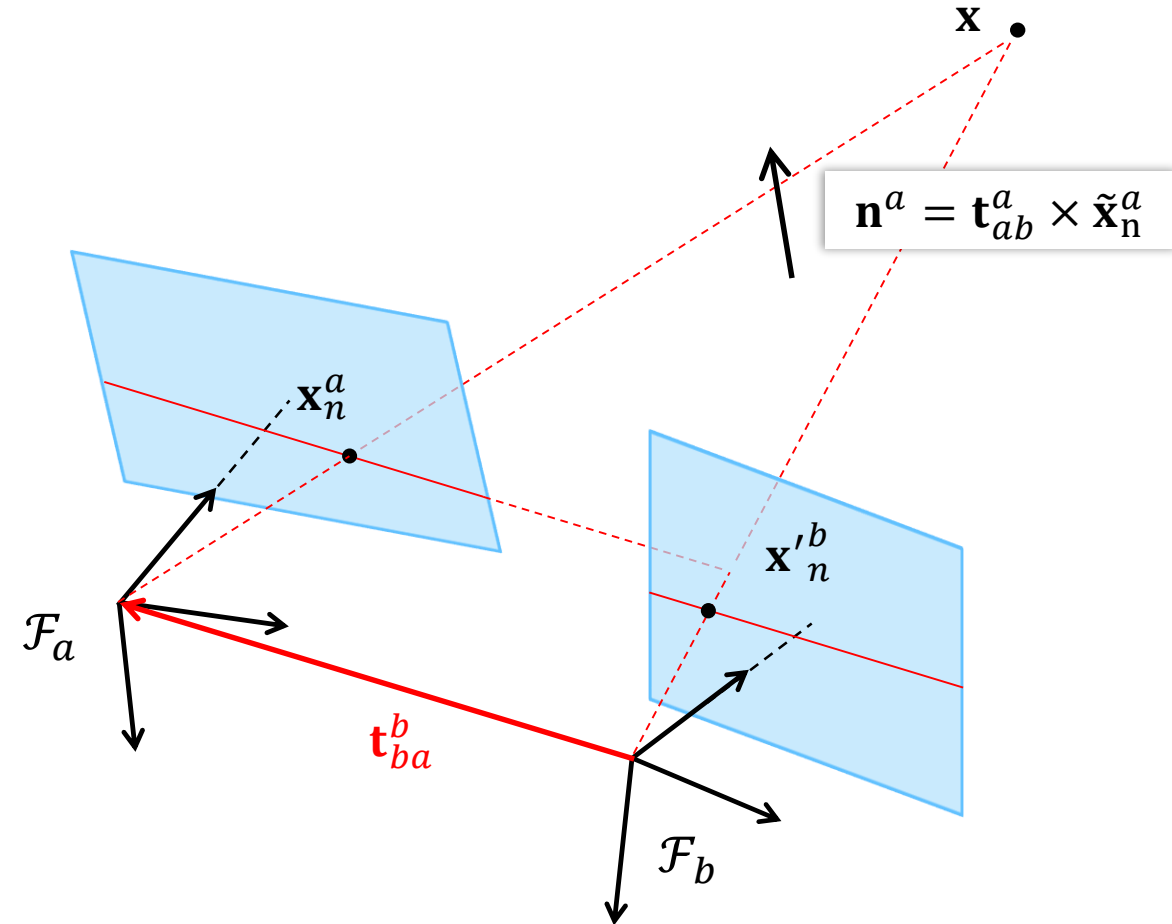
$$\tilde{\mathbf{x}}_n^a T [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab} \tilde{\mathbf{x}}_n^b = 0$$

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Where we define the **essential matrix** to be

$$\mathbf{E}_{ab} = [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab}$$



Notice that this derivation is independent of  $\|\mathbf{t}_{ba}^b\|$ !  
Hence,  $\mathbf{E}_{ab}$  is homogeneous by nature

# The essential matrix E

For a correspondence  $\mathbf{x}_n^a \leftrightarrow \mathbf{x}'_n{}^b$  to be geometrically viable, it must satisfy the equations

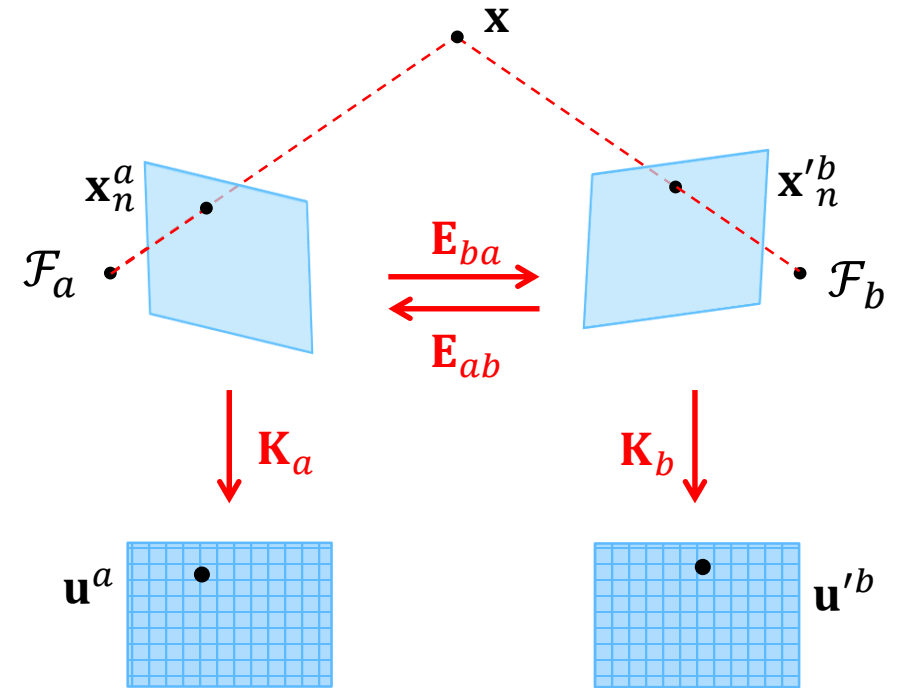
$$\tilde{\mathbf{x}}_n^{aT} \mathbf{E}_{ab} \tilde{\mathbf{x}}_n'^b = 0$$

$$\tilde{\mathbf{x}}_n'^b{}^T \mathbf{E}_{ba} \tilde{\mathbf{x}}_n^a = 0$$

where the essential matrices  $\mathbf{E}_{ab}$  and  $\mathbf{E}_{ba}$  are homogeneous and given (up to scale) by

$$\mathbf{E}_{ab} = [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab}$$

$$\mathbf{E}_{ba} = [\mathbf{t}_{ba}^b]_{\times} \mathbf{R}_{ba}$$



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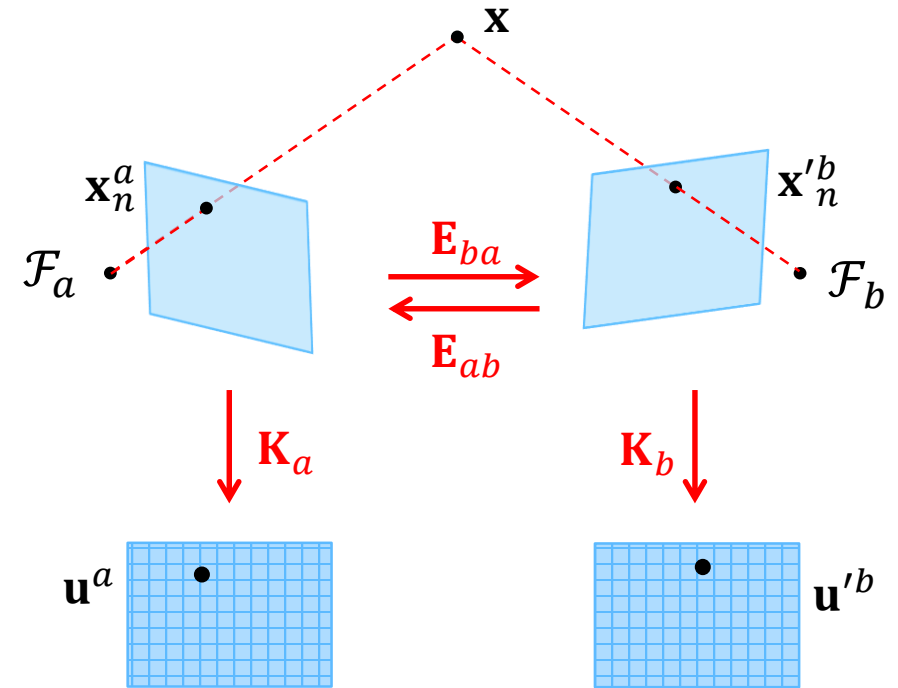
$$\tilde{\mathbf{x}}_n^{aT} \mathbf{E}_{ab} \tilde{\mathbf{x}}_n'^b = 0$$

$$\tilde{\mathbf{x}}_n'^{bT} \mathbf{E}_{ba} \tilde{\mathbf{x}}_n^a = 0$$

where the essential matrices  $\mathbf{E}_{ab}$  and  $\mathbf{E}_{ba}$  are homogeneous and given (up to scale) by

$$\mathbf{E}_{ab} = [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab}$$

$$\mathbf{E}_{ba} = [\mathbf{t}_{ba}^b]_{\times} \mathbf{R}_{ba}$$



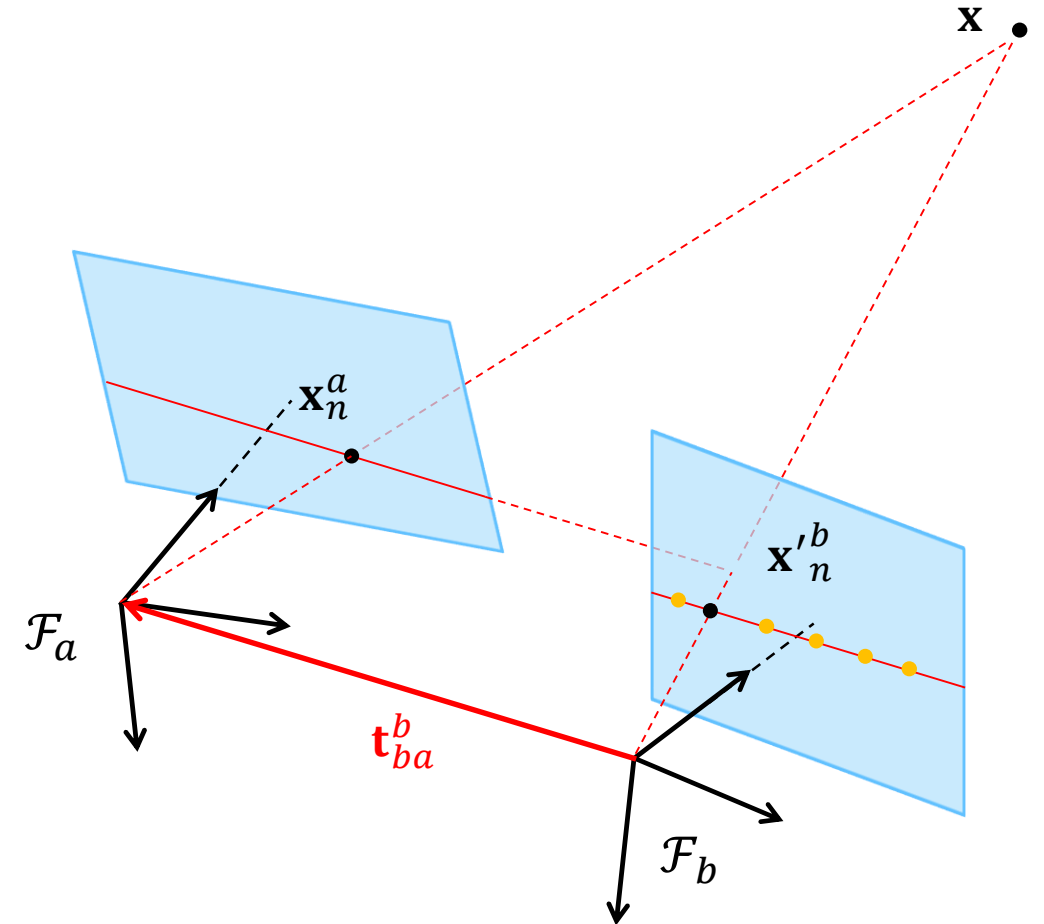
From the equations it is clear that  $\mathbf{E}_{ba} = \mathbf{E}_{ab}^T$ , so that the two equations are equivalent representations of the same constraint

# The essential matrix E

## Note

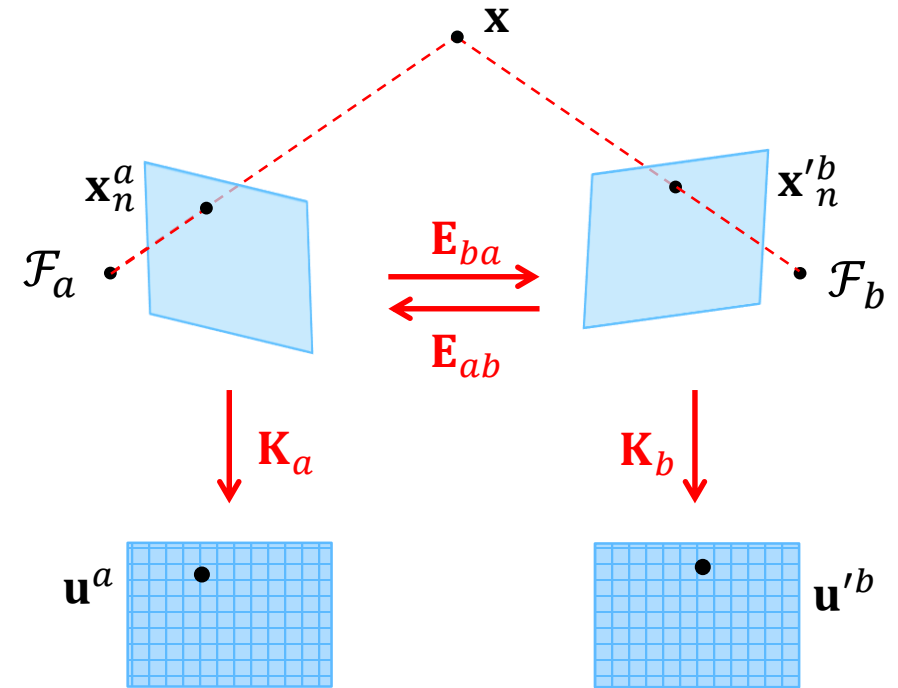
Although  $\tilde{\mathbf{x}}_n'^b{}^T \mathbf{E}_{ba} \tilde{\mathbf{x}}_n^a = 0$  is a *necessary* requirement for the correspondence  $\mathbf{x}_n^a \leftrightarrow \mathbf{x}_n'^b$  to be correct, it is *not sufficient* to guarantee its correctness

It only guarantees that the two points lie in the same epipolar plane



# Properties of $\mathbf{E}$

- $\mathbf{E}_{ab} = [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab}$
- $\mathbf{E}_{ab}$  is homogeneous
- $\text{rank}(\mathbf{E}_{ab}) = 2$
- $\det(\mathbf{E}_{ab}) = 0$
- $\mathbf{E}_{ab}$  has five degrees of freedom
  - $\mathbf{R} \Rightarrow 3$ ,  $\mathbf{t} \Rightarrow 3$ , homogeneous  $\Rightarrow -1$
  - It can be estimated from as little as five point correspondences  $\mathbf{x}'_n \leftrightarrow \mathbf{x}_n^a$



$$\tilde{\mathbf{x}}_n^a{}^T \mathbf{E}_{ab} \tilde{\mathbf{x}}_n'^b = 0$$

$$\tilde{\mathbf{x}}_n'^b{}^T \mathbf{E}_{ba} \tilde{\mathbf{x}}_n^a = 0$$



# Properties of E

- $\tilde{\mathbf{l}}'^a = \mathbf{E}_{ab} \tilde{\mathbf{x}}'^b_n$  is the homogeneous representation of the epipolar line in the normalized image plane of  $\mathcal{F}_a$  corresponding to the point  $\mathbf{x}'^b_n$

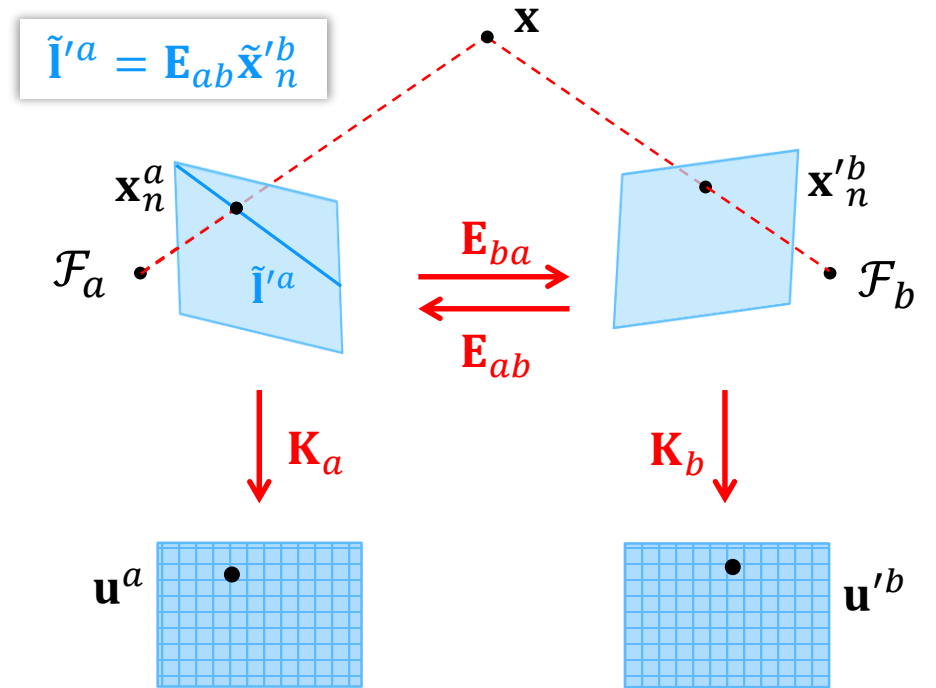
Line in  $\mathbb{R}^2$ :

$$ax + by + c = 0$$

Line in  $\mathbb{P}^2$ :

$$\tilde{\mathbf{x}}^T \tilde{\mathbf{l}} = 0$$

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix}^T \begin{bmatrix} a \\ b \\ c \end{bmatrix} = 0$$



$$\tilde{\mathbf{l}}'^a = \mathbf{E}_{ab} \tilde{\mathbf{x}}'^b_n$$

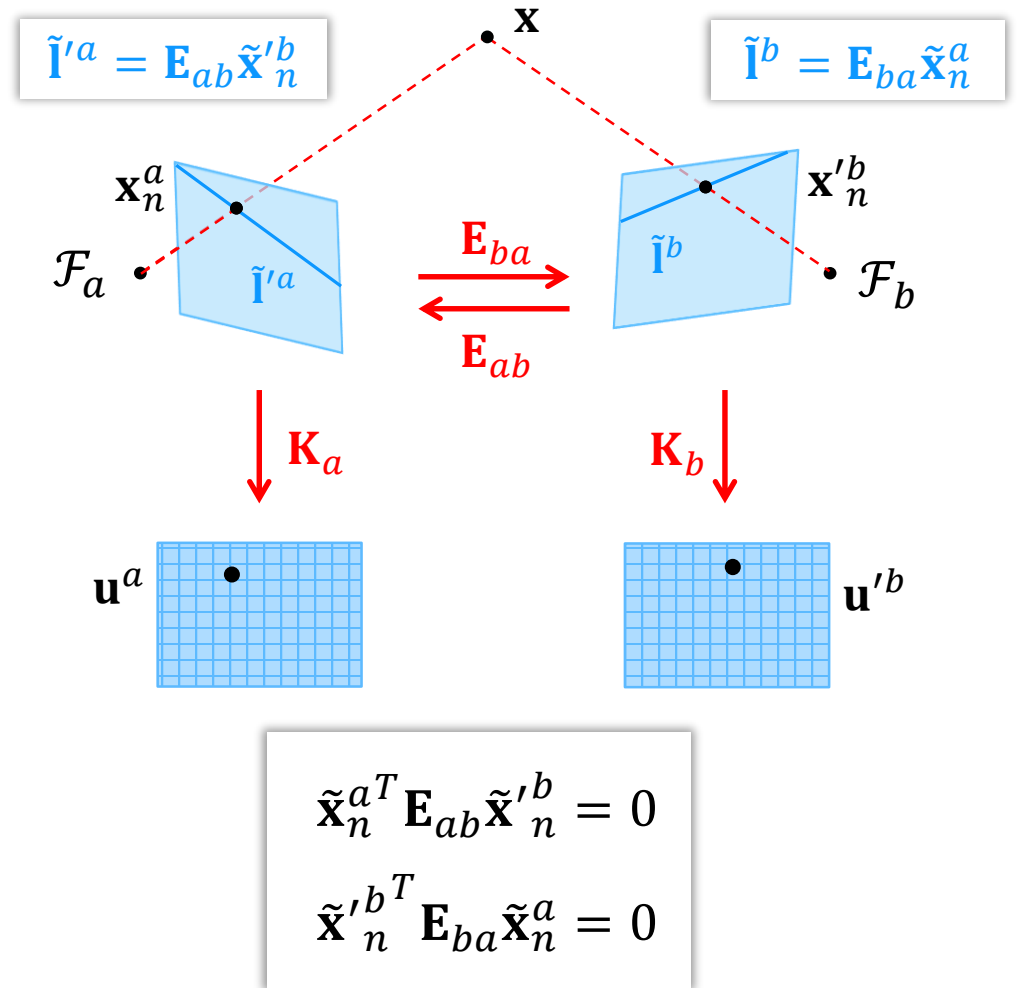
$$\tilde{\mathbf{x}}_n^{aT} \mathbf{E}_{ab} \tilde{\mathbf{x}}'^b_n = 0$$

$$\tilde{\mathbf{x}}'^b_n{}^T \mathbf{E}_{ba} \tilde{\mathbf{x}}_n^a = 0$$



# Properties of E

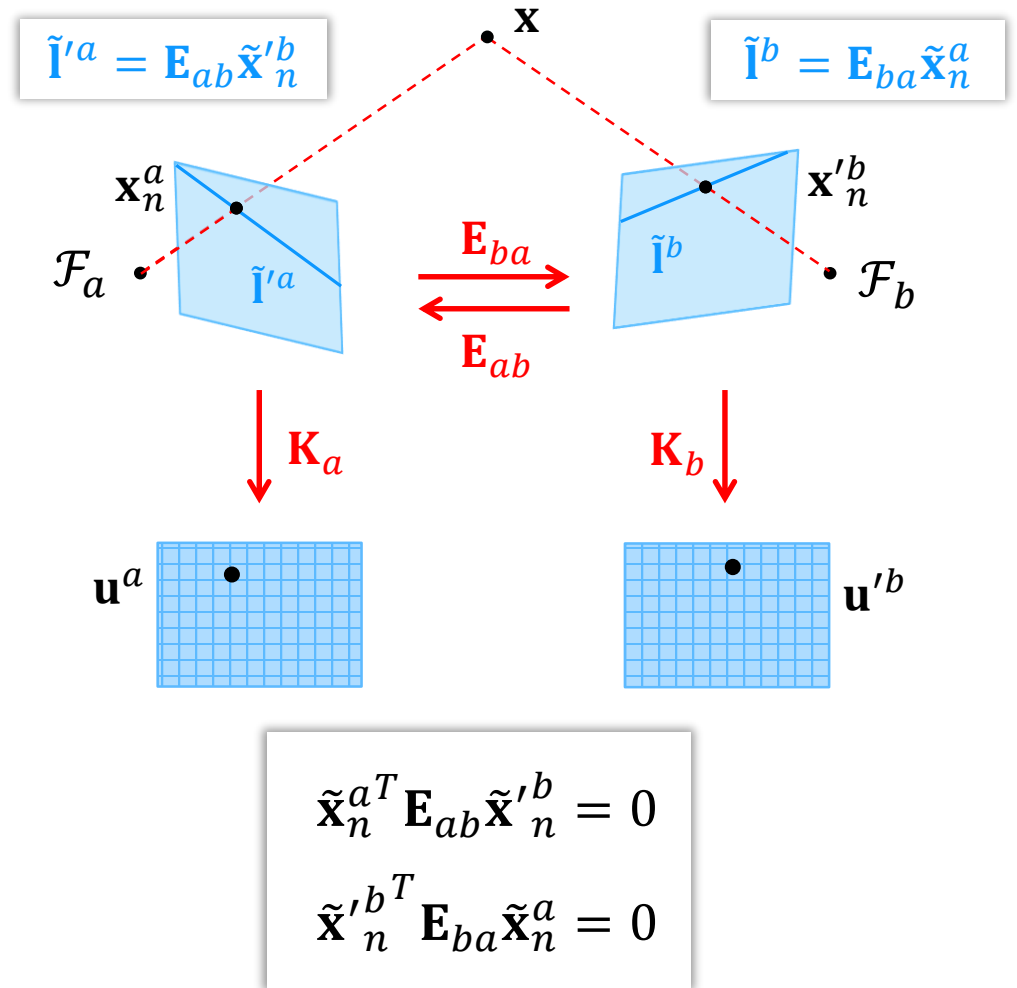
- $\tilde{\mathbf{l}}'^a = \mathbf{E}_{ab} \tilde{\mathbf{x}}'^b_n$  is the homogeneous representation of the epipolar line in the normalized image plane of  $\mathcal{F}_a$  corresponding to the point  $\mathbf{x}'^b_n$
- $\tilde{\mathbf{l}}^b = \mathbf{E}_{ba} \tilde{\mathbf{x}}^a_n$  is the epipolar line in the normalized image plane of  $\mathcal{F}_b$  corresponding to the point  $\mathbf{x}^a_n$



# Properties of E

- $\tilde{\mathbf{l}}^a = \mathbf{E}_{ab} \tilde{\mathbf{x}}_n^{\prime b}$  is the homogeneous representation of the epipolar line in the normalized image plane of  $\mathcal{F}_a$  corresponding to the point  $\mathbf{x}_n^{\prime b}$
- $\tilde{\mathbf{l}}^b = \mathbf{E}_{ba} \tilde{\mathbf{x}}_n^a$  is the epipolar line in the normalized image plane of  $\mathcal{F}_b$  corresponding to the point  $\mathbf{x}_n^a$
- It is possible to determine  $\mathbf{R}_{ab}$  and  $\mathbf{t}_{ab}^a$  (up to scale) by decomposing  $\mathbf{E}_{ab}$

$$\mathbf{E}_{ab} = [\mathbf{t}_{ab}^a]_{\times} \mathbf{R}_{ab}$$



# The fundamental matrix F

The epipolar constraint extends naturally to point correspondences  $\mathbf{u}^a \leftrightarrow \mathbf{u}'^b$  via the camera calibration matrices  $\mathbf{K}_a$  and  $\mathbf{K}_b$

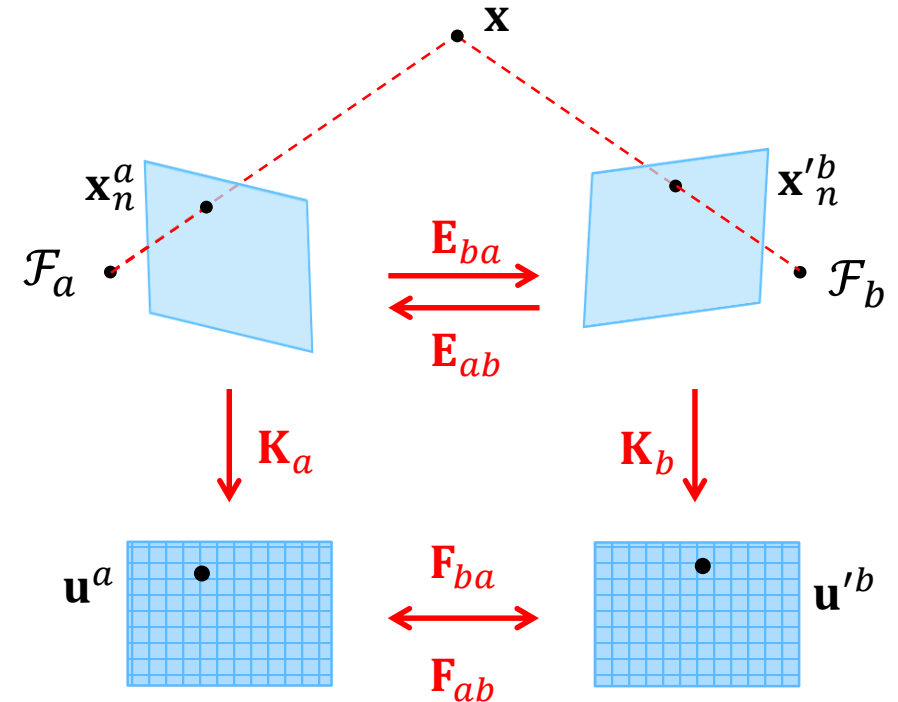
$$\tilde{\mathbf{u}}^{aT} \mathbf{F}_{ab} \tilde{\mathbf{u}}'^b = 0$$

$$\tilde{\mathbf{u}}'^{bT} \mathbf{F}_{ba} \tilde{\mathbf{u}}^a = 0$$

where the fundamental matrices  $\mathbf{F}_{ab}$  and  $\mathbf{F}_{ba}$  are given by

$$\mathbf{F}_{ab} = \mathbf{K}_a^{-T} \mathbf{E}_{ab} \mathbf{K}_b^{-1}$$

$$\mathbf{F}_{ba} = \mathbf{K}_b^{-T} \mathbf{E}_{ba} \mathbf{K}_a^{-1}$$



# The fundamental matrix $\mathbf{F}$

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$$\tilde{\mathbf{u}}^{aT} \mathbf{F}_{ab} \tilde{\mathbf{u}}'^b = 0$$

$$\tilde{\mathbf{u}}'^bT \mathbf{F}_{ba} \tilde{\mathbf{u}}^a = 0$$

where the fundamental matrices  $\mathbf{F}_{ab}$  and  $\mathbf{F}_{ba}$  are given by

$$\mathbf{F}_{ab} = \mathbf{K}_a^{-T} \mathbf{E}_{ab} \mathbf{K}_b^{-1}$$

$$\mathbf{F}_{ba} = \mathbf{K}_b^{-T} \mathbf{E}_{ba} \mathbf{K}_a^{-1}$$

$$\left(\tilde{\mathbf{x}}_n^a\right)^T \mathbf{E}_{ab} \tilde{\mathbf{x}}_n'^b = 0$$

$$\Downarrow \begin{cases} \tilde{\mathbf{x}}_n^a = \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a \\ \tilde{\mathbf{x}}_n^b = \mathbf{K}_b^{-1} \tilde{\mathbf{u}}^b \end{cases}$$

$$\left(\mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a\right)^T \mathbf{E}_{ab} \mathbf{K}_b^{-1} \tilde{\mathbf{u}}^b = 0$$

$$\left(\tilde{\mathbf{u}}^a\right)^T \underbrace{\mathbf{K}_a^{-T} \mathbf{E}_{ab} \mathbf{K}_b^{-1}}_{=\mathbf{F}_{ab}} \tilde{\mathbf{u}}^b = 0$$

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$$\tilde{\mathbf{u}}'^bT \mathbf{F}_{ba} \tilde{\mathbf{u}}^a = 0$$

where the fundamental matrices  $\mathbf{F}_{ab}$  and  $\mathbf{F}_{ba}$  are given by

$$\mathbf{F}_{ab} = \mathbf{K}_a^{-T} \mathbf{E}_{ab} \mathbf{K}_b^{-1}$$

$$\mathbf{F}_{ba} = \mathbf{K}_b^{-T} \mathbf{E}_{ba} \mathbf{K}_a^{-1}$$

From the equations it is clear that  $\mathbf{F}_{ba} = \mathbf{F}_{ab}^T$ , so that the two equations are equivalent representations of the same constraint

$$\left(\tilde{\mathbf{x}}_n^a\right)^T \mathbf{E}_{ab} \tilde{\mathbf{x}}_n'^b = 0$$

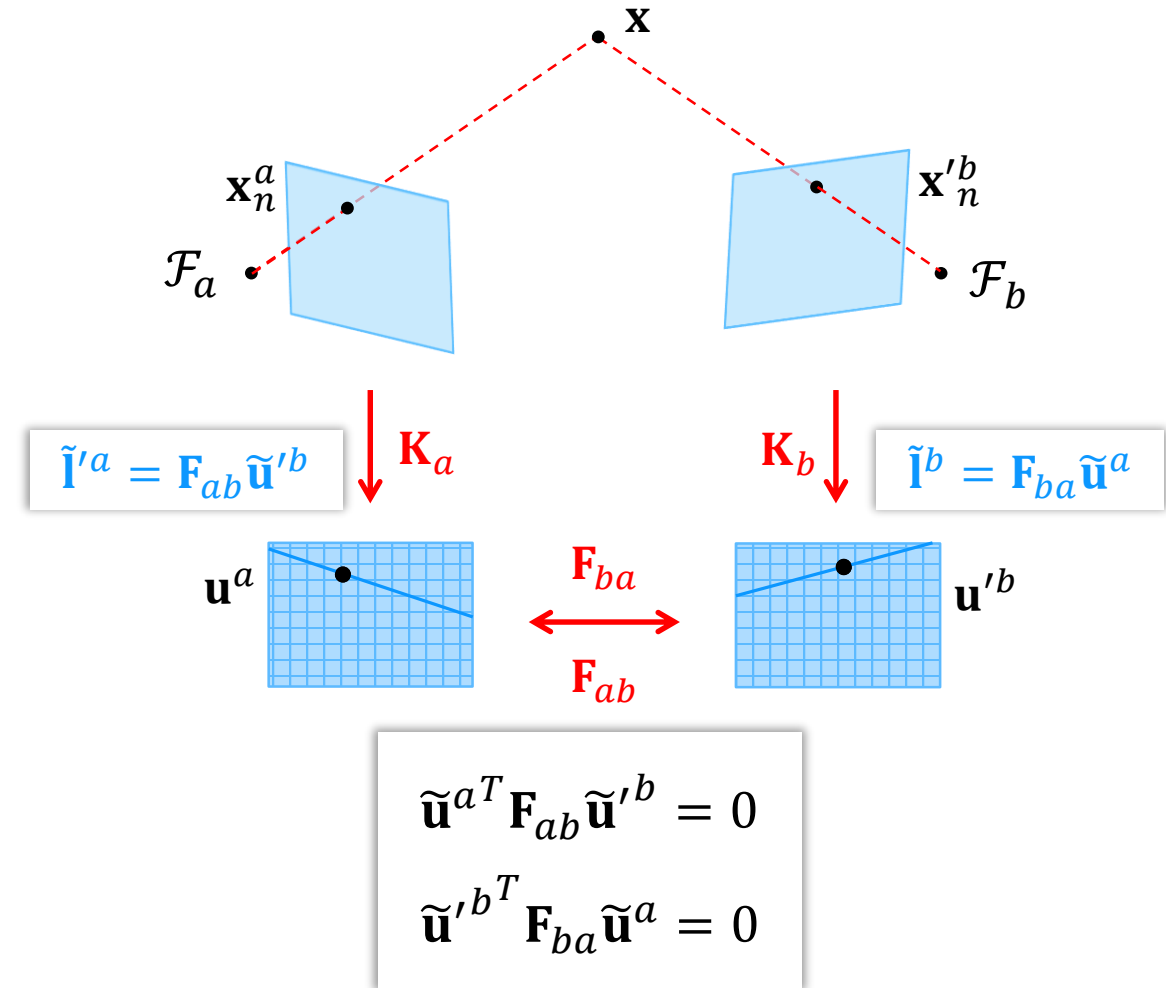
$$\Downarrow \begin{cases} \tilde{\mathbf{x}}_n^a = \mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a \\ \tilde{\mathbf{x}}_n^b = \mathbf{K}_b^{-1} \tilde{\mathbf{u}}^b \end{cases}$$

$$\left(\mathbf{K}_a^{-1} \tilde{\mathbf{u}}^a\right)^T \mathbf{E}_{ab} \mathbf{K}_b^{-1} \tilde{\mathbf{u}}^b = 0$$

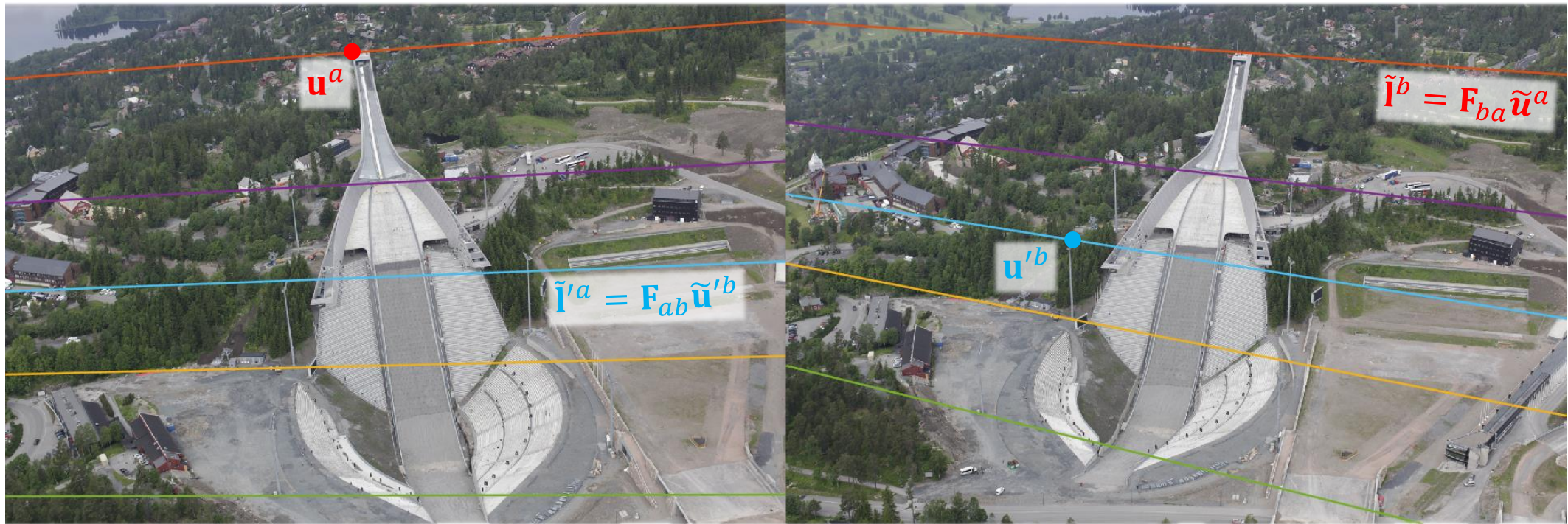
$$\left(\tilde{\mathbf{u}}^a\right)^T \underbrace{\mathbf{K}_a^{-T} \mathbf{E}_{ab} \mathbf{K}_b^{-1}}_{=\mathbf{F}_{ab}} \tilde{\mathbf{u}}^b = 0$$

# Properties of F

- $\mathbf{F}_{ab}$  is homogeneous
- $\text{rank}(\mathbf{F}_{ab}) = 2$
- $\det(\mathbf{F}_{ab}) = 0$
- $\mathbf{F}_{ab}$  has seven degrees of freedom
  - It can be estimated from seven or more point correspondences  $\mathbf{u}^a \leftrightarrow \mathbf{u}'^b$
- Epipolar line corresponding to  $\mathbf{u}'^b$  is
 
$$\tilde{\mathbf{l}}'^a = \mathbf{F}_{ab} \tilde{\mathbf{u}}'^b$$
- Epipolar line corresponding to  $\tilde{\mathbf{u}}^a$  is
 
$$\tilde{\mathbf{l}}^b = \mathbf{F}_{ba} \tilde{\mathbf{u}}^a$$



# Example



$img_a$

$img_b$

# Estimating F

Several algorithms can be used

- Linear: 7-pt, 8-pt
- Non-linear: Minimize total epipolar distance

Due to potential erroneous correspondences, it is natural to begin with a RANSAC estimation

The 8-pt algorithm is very similar to the homography estimation we have already seen

To simplify notations let us consider the correspondence  $\mathbf{u} \leftrightarrow \mathbf{u}'$  and the fundamental matrix

$$\tilde{\mathbf{u}}'^T \mathbf{F} \tilde{\mathbf{u}} = 0$$



# Estimating F

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$$\tilde{\mathbf{u}}'^T \mathbf{F} \tilde{\mathbf{u}} = 0$$

For each point correspondence  $\mathbf{u} \leftrightarrow \mathbf{u}'$  we have that

$$\tilde{\mathbf{u}}'^T \mathbf{F} \tilde{\mathbf{u}} = 0$$
$$\begin{bmatrix} u' & v' & 1 \end{bmatrix} \begin{bmatrix} f_1 & f_2 & f_3 \\ f_4 & f_5 & f_6 \\ f_7 & f_8 & f_9 \end{bmatrix} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = 0$$
$$\begin{bmatrix} uu' & vu' & u' & uv' & vv' & v' & u & v & 1 \end{bmatrix} \mathbf{f} = 0$$

# Estimating F

Several algorithms can be used

- Linear: 7-pt, 8-pt
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$$\begin{bmatrix} uu' & vu' & u' & uv' & vv' & v' & u & v & 1 \end{bmatrix} \mathbf{f} = 0$$

From several correspondences, we get a system of linear equations that we can solve by SVD

$$\begin{bmatrix} u_1 u'_1 & v_1 u'_1 & u'_1 & u_1 v'_1 & v_1 v'_1 & v'_1 & u_1 & v_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ u_k u'_k & v_k u'_k & u'_k & u_k v'_k & v_k v'_k & v'_k & u_k & v_k & 1 \end{bmatrix} \mathbf{f} = 0$$

$$\mathbf{A} \mathbf{f} = 0$$

# Estimating F – The 8-point algorithm

Given eight (or more) correspondences  $\mathbf{u}_i \leftrightarrow \mathbf{u}'_i$

1. Normalize point sets  $\{\mathbf{u}_i\}$  and  $\{\mathbf{u}'_i\}$  using similarity transforms  $\mathbf{T}$  and  $\mathbf{T}'$
2. Build matrix  $\mathbf{A}$  from correspondences  $\hat{\mathbf{u}}_i \leftrightarrow \hat{\mathbf{u}}'_i$  and compute its SVD
3. Extract the estimate  $\hat{\mathbf{F}}$  from the right singular vector corresponding to the smallest singular value

4. Perform SVD on  $\hat{\mathbf{F}}$ :

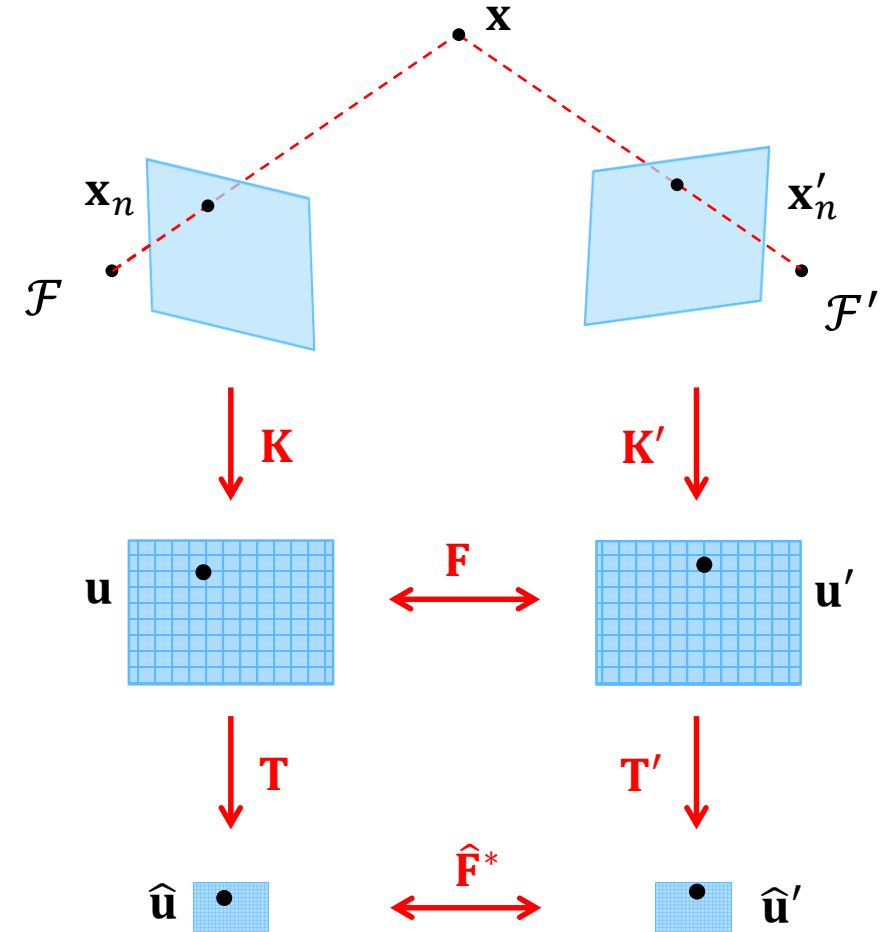
$$\hat{\mathbf{F}} = \mathbf{U}\mathbf{S}\mathbf{V}^T$$

5. Enforce zero determinant by setting the smallest singular value ( $s_{33}$  in  $\mathbf{S}$ ) to zero and compute a proper fundamental matrix

$$\hat{\mathbf{F}}^* = \mathbf{U}\mathbf{S}\mathbf{V}^T$$

6. Denormalize

$$\mathbf{F} = \mathbf{T}'^T \hat{\mathbf{F}}^* \mathbf{T}$$



# Estimating F – The 7-point algorithm

Given seven correspondences  $\mathbf{u}_i \leftrightarrow \mathbf{u}_i'$

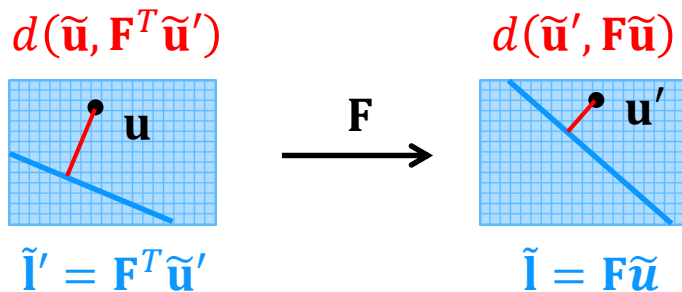
- The matrix  $\mathbf{A}$  is a  $7 \times 9$  matrix, so in general  $\text{rank}(\mathbf{A}) = 7$  and the null space of  $\mathbf{A}$  is 2-dimensional
  - Then the fundamental matrix must be a linear combination of the two right singular vectors of  $\mathbf{A}$  which correspond to the two singular values that are zero
- $$\mathbf{F}(\alpha) = \alpha \mathbf{F}_1 + (1 - \alpha) \mathbf{F}_2$$
- The additional constraint  $\det(\mathbf{A}) = 0$  leads to a cubic polynomial equation in  $\alpha$  which has one or three solutions
  - Hence the 7-pt algorithm returns one or three possible fundamental matrices
  - In a RANSAC scheme, the 7-pt algorithm is better than the 8-pt algorithm
    - It is minimal, since we only need to sample seven random correspondences per iteration
    - Each sampled set of correspondences can return three fundamental matrices for testing

# Estimating F – Beyond linear estimation

Improved estimates of  $\mathbf{F}$  can be obtained using iterative methods

One possibility is to determine the matrix  $\mathbf{F}$  that minimizes the total squared **epipolar distance**

$$\varepsilon = \sum_i d(\tilde{\mathbf{u}}'_i, \mathbf{F}\tilde{\mathbf{u}}_i)^2 + d(\tilde{\mathbf{u}}_i, \mathbf{F}^T\tilde{\mathbf{u}}'_i)^2$$



The distance between a homogeneous point  $\tilde{\mathbf{u}}$  and a homogeneous line  $\tilde{\mathbf{l}} = [\tilde{l}_1 \ \tilde{l}_2 \ \tilde{l}_3]^T$  is

$$d(\tilde{\mathbf{u}}, \tilde{\mathbf{l}}) = \frac{\tilde{\mathbf{u}}^T \tilde{\mathbf{l}}}{\sqrt{\tilde{l}_1^2 + \tilde{l}_2^2}}$$

Iterative methods typically achieve a noticeably better estimate than the linear methods

But linear methods typically provide quite good estimates

# Estimating E

For calibrated cameras we can first estimate  $\mathbf{F}$  and then compute  $\mathbf{E}$  by

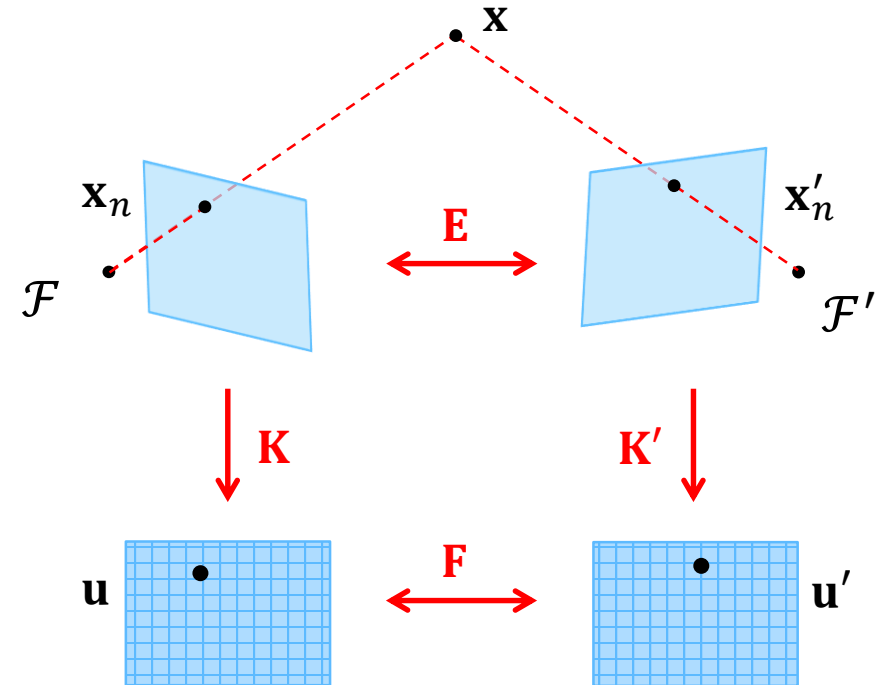
$$\mathbf{E} = \mathbf{K}'^T \mathbf{F} \mathbf{K}$$

One can also estimate  $\mathbf{E}$  directly from five normalized point correspondences  $\mathbf{x}_n \leftrightarrow \mathbf{x}'_n$  using an algorithm called the **5-pt algorithm**

- Involves finding the roots of a 10<sup>th</sup> degree polynomial

In a RANSAC scheme, the 5-pt algorithm is the preferred alternative

- To achieve 99% confidence with 50% outliers, requires 145 tests with using the 5-pt algorithm versus 1177 tests using the 8-pt algorithm



# Summary

- The essential matrix  $\mathbf{E}$  and the fundamental matrix  $\mathbf{F}$  represent the epipolar constraint

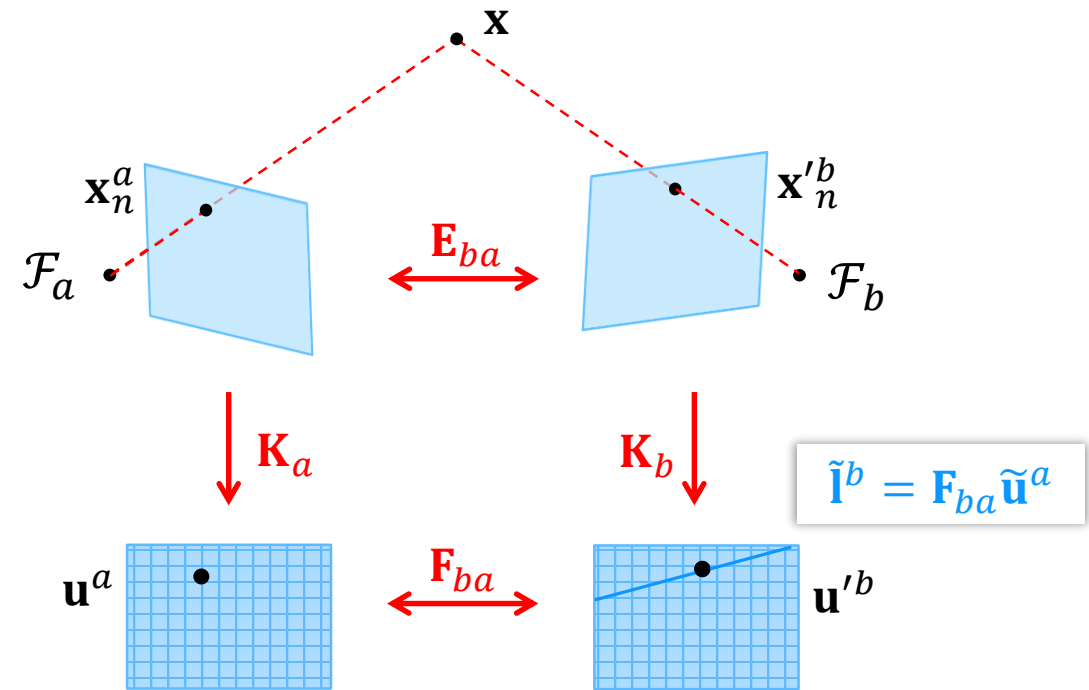
$$\left(\tilde{\mathbf{x}}_n^{\prime b}\right)^T \mathbf{E}_{ba} \tilde{\mathbf{x}}_n^a = 0$$

$$\left(\tilde{\mathbf{u}}^{\prime b}\right)^T \mathbf{F}_{ba} \tilde{\mathbf{u}}^a = 0$$

- $\mathbf{E}$  and  $\mathbf{F}$  can be estimated from correspondences
  - $\mathbf{F} \leftarrow$  RANSAC, 7-pt or 8-pt
  - $\mathbf{E} \leftarrow$  RANSAC, 5-pt

- $\mathbf{E}$  and  $\mathbf{F}$  maps points to their corresponding epipolar lines

$$\tilde{\mathbf{l}}^b = \mathbf{F}_{ba} \tilde{\mathbf{u}}^a$$



$$\mathbf{E}_{ba} = [\mathbf{t}_{ba}^b]_{\times} \mathbf{R}_{ba}$$

$$\mathbf{F}_{ba} = \mathbf{K}_b^{-T} \mathbf{E}_{ba} \mathbf{K}_a^{-1}$$

# Supplementary material

## Recommended

- *Richard Szeliski: Computer Vision: Algorithms and Applications 2<sup>nd</sup> ed*
  - Chapter 11 “Structure from motion and SLAM”, in particular section 11.3 “Two-frame structure from motion”
- *T. V. Haavardsholm: A Handbook In Visual SLAM*
  - Chapter 3 “Camera geometry”, in particular section 3.2 “Epipolar geometry”

## Other

- David Nistér, *An Efficient Solution to the Five-Point Relative Pose Problem*, 2004
- Richard I. Hartley, *In Defense of the Eight-Point Algorithm*, 1997