# Visual odometry

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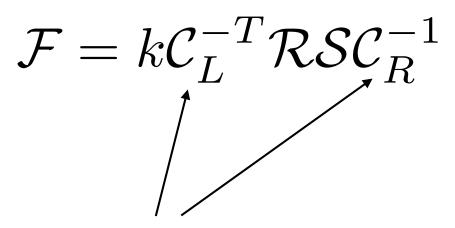
# Odometry from two camera geometry

## • Idea:

- use calibrated camera
- move; track some points
  - in reading slides
- compute essential matrix (calibrated fundamental matrix) to get
  - rotation
  - translation up to scale
- Options:
  - fix scale later
  - use (say) wheel measurements + Kalman filter to fix
  - use stereo

# RECALL: The Fundamental Matrix $\downarrow \mathbf{p_1}^T \mathcal{F} \mathbf{p_2} = 0$

- Can be fit a pair of images using feature correspondences
  - 8 point algorithm
  - robustness is an important issue
  - we'll do this



If we know these

we can recover info about R, T from F

# The Essential matrix

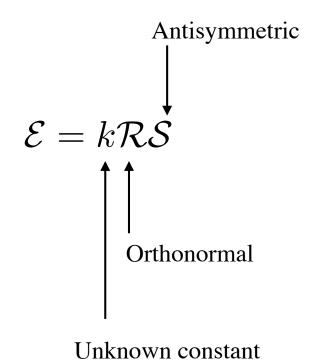
## • Assume camera calibration is known

• Cameras are normalized so that C=Id

$$\mathbf{p_1}^T \mathcal{F} \mathbf{p_2} = 0$$
 becomes  $\mathbf{p}_1^T \mathcal{E} \mathbf{p}_2 = 0$ 

$$\mathcal{F} = k \mathcal{C}_L^{-T} \mathcal{RSC}_R^{-1}$$
 becomes  $\mathcal{E} = k \mathcal{RS}$   
The essential matrix

# From fundamental matrix

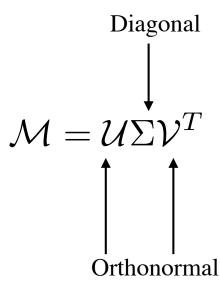


# Getting R, S from E

#### • Recall SVD:

- Notice that, for R a rotation,
  M and PM have the same singular value
  - M and RM have the same singular values
- So singularvalues(E)=singularvalues(S)
  - check:

$$\Sigma(\mathcal{S}) = \left(\begin{array}{rrrr} s & 0 & 0\\ 0 & s & 0\\ 0 & 0 & 0 \end{array}\right)$$



# Recovering R, S - I

#### • Write

 $\mathcal{W} = \begin{pmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$  $\mathcal{R} = \mathcal{U}\mathcal{W}^{-1}\mathcal{V}^{T}$  $\mathcal{S} = \mathcal{V}\mathcal{W}\Sigma\mathcal{V}^{T}$ 

## • Check that

- RS=E
- R is orthonormal
- S is antisymmetric

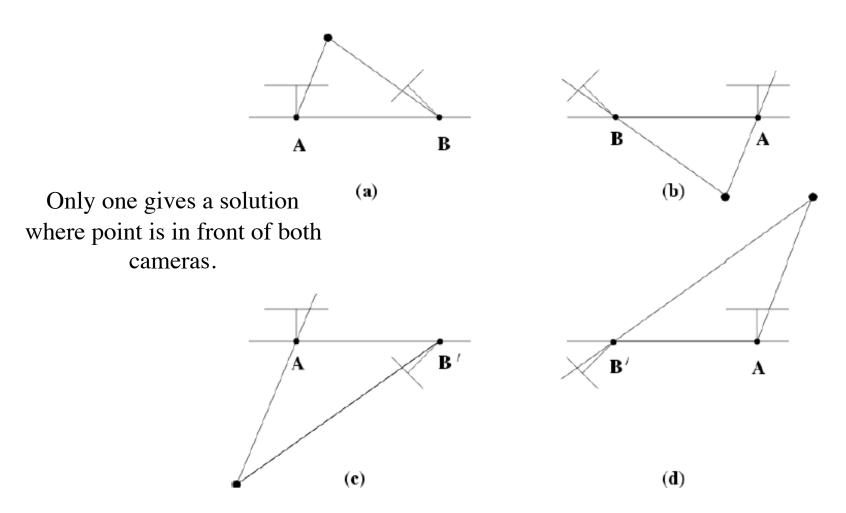
# BUT

- There are ambiguities
  - check that for any Q of the form
    - square root of identity
  - R', S' as given also work
    - R' is orthonormal
    - S' is antisymmetric
- Four of these don't matter
  - cause det(R')=-1
  - implies camera was reflected as well as rotated
    - and that doesn't happen

$$\mathcal{Q} = \operatorname{diag}(\pm 1, \pm 1, \pm 1)$$

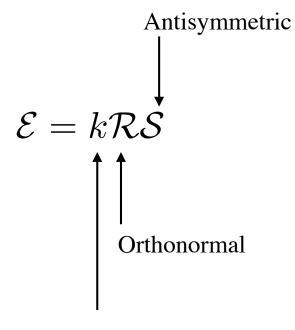
$$\mathcal{S}' = \mathcal{QS}$$
  
 $\mathcal{R}' = \mathcal{RQ}$ 

## The other four....



S. Weiss' notes on visual odometry from CVPR 14 tutorial

# But the unknown constant is unknown...



Unknown constant

Different values of k will lead to different scales of S - equivalently, different scales of translation between cameras - you need extra information to sort this out.

# What we have

### • Can determine

- the rotation between two cameras
- the translation \*up to scale\*
- From this, we can recover 3D points
  - up to scale

## What we have

• 3D points: 
$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$
 And  $\begin{pmatrix} x_1^t \\ x_2^t \\ x_3^t \end{pmatrix} = \mathcal{R} \begin{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} - \mathbf{t}$ 

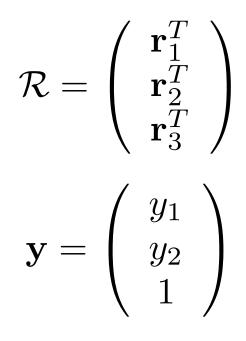
Original point in camera two's coordinate system

• normalized image points:

$$\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} x_1/x_3 \\ x_2/x_3 \end{pmatrix} \qquad \begin{pmatrix} y_1^t \\ y_2^t \end{pmatrix} = \begin{pmatrix} x_1^t/x_3^t \\ x_2^t/x_3^t \end{pmatrix}$$

# Recovering the point in 3D

• Write





$$x_3 = \frac{(\mathbf{r}_1 - y_1^t \mathbf{r}_3)^T \mathbf{t}}{(\mathbf{r}_1 - y_2^t \mathbf{r}_3)^T \mathbf{y}}$$

And we have everything in 3D!

# The effect of scale

$$x_3 = \frac{(\mathbf{r}_1 - y_1^t \mathbf{r}_3)^T \mathbf{t}}{(\mathbf{r}_1 - y_2^t \mathbf{r}_3)^T \mathbf{y}}$$

- Notice that if k changes, t gets bigger or smaller
  - point coordinates scale
    - x\_1=y\_1 x\_3, x\_2=y\_2 x\_3
- So if we can match points across more than two cameras
  - there is only one scale ambiguity in the whole sequence
  - This could be quite easy to sort out
    - eg you know the size of high bay
    - eg you know some reference scale
    - etc

# Alternatives

- Filter the scale using estimates from wheels
  - etc
- Stereo odometry
  - If I have two cameras then there is no issue with scale

# Pragmatics

- Need
  - good fast feature computation and tracking
    - fast features and good robust methods seem to beat good features
  - reliable camera calibration
    - and robust FM/EM estimation
    - Ransac remains reliable
  - OR good stereo
- See slides+notes