

# Lecture 8.1 Multiple-View Geometry

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# Weekly overview





- Multiple-view geometry
  - Correspondences
- Structure from Motion (SfM)
  - Sparse 3D reconstruction
- Multiple-view stereo
  - Dense 3D reconstruction





- Epipolar geometry
  - The essential matrix  $E = [t]_{\times}R$  $\widetilde{\mathbf{x}}^{\prime T} E \widetilde{\mathbf{x}} = 0$
  - The fundamental matrix  $F = K'^{-T}EK^{-1}$  $\widetilde{\boldsymbol{u}}'^{T}F\widetilde{\boldsymbol{u}} = 0$
  - Estimating *F* from 7 or 8 correspondences  $u_i \leftrightarrow u_i'$
  - Estimating *E* from 5 correspondences  $x_i \leftrightarrow x_i'$





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- Pose from epipolar geometry
  - Decomposing *E* into *R* and *t* (up to scale)





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  - Triangulation based on known camera matrices
- Sequential visual odometry
  - $x_i^{(k)} \leftrightarrow x_i^{(k+1)} \longrightarrow E_{k,k+1} \longrightarrow {}^k \xi_{k+1}$
  - ${}^{0}\xi_{k+1} = {}^{0}\xi_{k} {}^{k}\xi_{k+1}$





#### **Correspondences (matching)**

- Correspondences must satisfy the epipolar constraint represented by the fundamental matrix  $\widetilde{\boldsymbol{u}}'^T F \widetilde{\boldsymbol{u}} = 0$
- Useful for reducing the number of mismatches

#### Scene geometry (structure)

- Sparse 3D from triangulating correspondences
- Dense 3D from stereo processing

### **Camera geometry (motion)**

- In the uncalibrated case, the camera matrices *P* and *P'* can be estimated from the fundamental matrix *F* up to a projective ambiguity
- In the calibrated case, the relative pose between cameras can be estimated up to scale by decomposing the essential matrix

$$(\mathbf{x}_j \leftrightarrow \mathbf{x}'_j) \xrightarrow{E = [\mathbf{t}]_{\times R}} (R, \lambda \mathbf{t})$$



### More-than-two-view geometry

#### **Correspondences (matching)**

- How does "more-than-two-view geometry" constrain our 2D matches?
- Algebraic description?

#### Scene geometry (structure)

- Effect of more views on determining the 3D structure of the scene?
- Next lecture

#### **Camera geometry (motion)**

- Effect of more views on determining camera poses?
- Next lecture





#### **Two views**

- Points  $u_1$  and  $u_2$  must satisfy the epipolar constraint
- The fundamental matrix *F* represents this constraint  $\widetilde{\boldsymbol{u}}_2^T F \widetilde{\boldsymbol{u}}_1 = 0$
- Epipoles img2 img1  $\boldsymbol{u}_2$ A point  $\boldsymbol{u}_2$  in img2 X correspond to a line in img1 img2 img1  $\boldsymbol{u}_1$ A point  $u_1$  in img1  $\boldsymbol{u}_2$  $\bigotimes$ correspond to a line in img2 img2 img1

• *F* also describes the correspondence between points and epipolar lines

$$\widetilde{\boldsymbol{l}}_2 = F\widetilde{\boldsymbol{u}}_1 \widetilde{\boldsymbol{l}}_1 = F^T\widetilde{\boldsymbol{u}}_2$$





















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 $\boldsymbol{u}_3$ 

 $\boldsymbol{u}_3$ 

img3

img3

img3

img3

img3



- This construction shows that the three points *u*<sub>1</sub>, *u*<sub>2</sub> and *u*<sub>3</sub> are connected by some geometric constraint
  - Any one of them can be computed from the two others
- But it is not clear if this three-view constraint governs more than the three epipolar constraints put together





- The difference between two-view geometry and three-view geometry becomes evident if we consider lines instead of points
- In two-view geometry no constraints are available for lines
- In three-view geometry, lines *l*<sub>1</sub> and *l*<sub>2</sub> in two views will in general generate a line *l*<sub>3</sub> in a third view





- The three view geometry has an algebraic representation known as the trifocal tensor *T* 
  - A  $3 \times 3 \times 3$  array with 18dof
- This tensor governs the relationship between points and lines in three views
  - Point-point-point
  - Point-point-line
  - Point-line-line
  - Point-line-point
  - Line-line-line
- It may be used to transfer a two-view point/line correspondence into a point/line in a third view





#### **Three views**

- As we just saw, point transfer can be done directly from the epipolar constraints  $\widetilde{u}_3 = (F_{31}\widetilde{u}_1) \times (F_{32}\widetilde{u}_2)$
- However, this fails for points in the plane defined by the three camera centers – the trifocal plane – since the epipolar lines then will coincide





• The trifocal tensor allows point transfer also for points in the trifocal plane

 $\widetilde{\boldsymbol{u}}_3 = (F_{31}\widetilde{\boldsymbol{u}}_1) \times (F_{32}\widetilde{\boldsymbol{u}}_2)$ 



### **Example** Point transfer based on epipolar constraints



Uncertainty in feature points transfer to uncertainty in the epipolar lines

Hence the reliability of the predicted point depends on the angle between the epipolar lines

A large angle is good!



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 $F_{12}$ 

 $F_{31}$ 

 $F_{32}$ 

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### More-than-two-view geometry

#### **Correspondences (matching)**

- More views enables us to reveal and remove more mismatches than we can do in the two-view case
- More views also enables us to predict correspondences that can be tested with or without the use of descriptors
- Uncertainties in these predictions will in general decrease with the number of views





# **Summary**



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- Multiple-view geometry
- Correspondences  $\bullet$ 
  - Two-view vs Three-view
  - Fundamental matrix vs Trifocal tensor

