

## Stereo Geometry

COSC 470: Special Topic  
Computer Vision | 3D Reconstruction  
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## Stereo and Mosaicing

- We saw last time that in specific cases images are related by Homographies
- This is not the case in general
  - Two (perspective) images of the same scene are in general related by a *Fundamental Matrix*
  - For calibrated cameras this can be reduced to the *Essential Matrix*
  - These can be computed from feature matches

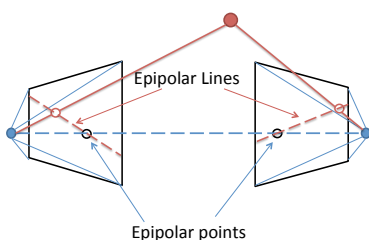
## The Fundamental Matrix

- Matching points,  $p$  and  $p'$ , in two views are related by the Fundamental Matrix
 
$$p'^T F p = 0$$
- $F$  has 9 elements, so looks like it has 9 degrees of freedom
  - Homogenous equation, so just 8 degrees
  - $F$  has rank 2, so just 7 degrees of freedom

## Epipolar Geometry

- $F$  defines the *epipolar geometry* for two views
  - The *epipolar points* are the projections of each camera centre into the other view
- Given the projection of a 3D point in one view
  - This corresponds to a ray from the camera centre
  - This projects to an *epipolar line* in the other view
  - This line passes through the image of the 3D point and the epipolar point in the second view

## Epipolar Geometry



## Epipolar Geometry and $F$

- The Fundamental matrix is closely related to the epipolar geometry, suppose  $p'^T F p = 0$ 
  - $l' = Fp$  is the epipolar line in the second view corresponding to point  $p$  in the first view
  - $l = F^T p'$  is the epipolar line in the first view corresponding to point  $p'$  in the second view
  - Here a line,  $l = [a, b, c]^T$  is interpreted as the line with formula  $ax + by + c = 0$

### The Essential Matrix

- Recall our camera projection model  

$$P = K[R|t]$$
- $K$  is the camera calibration matrix. If we know the calibration matrices we can form the *Essential matrix*  

$$E = K'^T F K$$
- This depends only on  $R$  and  $t$

### The Essential Matrix

- The Essential matrix has 9 elements but only 5 degrees of freedom
  - There are 3 degrees of freedom in  $R$
  - There are 3 degrees of freedom in  $t$
  - Less one for scale – we don't know the distance between the two cameras

### Computing $F$ and $E$

- There are several important algorithms based on different numbers of correspondences
  - The 8-point algorithm is a linear method for estimating  $F$ , but is over-constrained
  - The 7-point algorithm for  $F$  is minimal but can give 3 possible solutions from solving a cubic equation. These can be checked for consistency
  - The 5-point algorithm for  $E$  requires solving an order 10 polynomial and can give up to 10 solutions which need to be checked

### The 8-point algorithm

- In practice the 8-point algorithm is often used
  - Similar in essence to estimating a homography, but we need 8 points to get 8 equations
  - Similar issues of normalisation and outliers
- Each point correspondence gives us one homogeneous equation in the 9 elements of  $F$
- 8 of these let us solve for  $F$  (up to a scale) using the SVD as we did for finding  $H$

### Point Correspondence Equations

- Expanding out  $p'^T F p = 0$  gives

$$[x' \quad y' \quad 1] \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = 0$$

$$\begin{aligned} x'x f_{11} + y'x f_{21} + x f_{31} + \\ x'y f_{12} + y'y f_{22} + y f_{32} + \\ x'f_{13} + y'f_{23} + f_{33} = 0 \end{aligned}$$

### Basic 8-point Algorithm

- Given 8 (or more) correspondences  $p' \leftrightarrow p$ :
  - Form a linear equation for each correspondence
  - Stack these up to form  $Af = 0$
  - Find the Singular Value Decomposition  $A = USV^T$
  - $S$  should have one zero (or very small) value
  - The column of  $V$  corresponding to this value is  $f$
  - Reshape  $f$  into the  $3 \times 3$  matrix  $F$

### Robust 8-point Algorithm

- We need to normalise the data as before
  - Shift all the points in each view so that their mean value is zero
  - Scale them so that their average distance from the origin is  $\sqrt{2}$
- There will be outliers in the data
  - Use more than 8 points and RANSAC to find an inlier set, then use inliers to find least squares fit

### RANSAC and the 8-point Algorithm

- Recall that the number of iterations needed is

$$n = \frac{\log(p)}{\log(1 - \phi^s)}$$

- Here  $p$  is the probability of getting a bad result,  $\phi$  the inlier rate, and  $s$  the sample size
- Sample size here is 8, so really want  $\phi$  to be close to 1 – need reliable correspondences

### Refining $F$

- The solution we get from the 8-point algorithm usually isn't quite right
  - $F$  should have rank 2, which means that the each column is a linear combination of the other two
  - This is equivalent to having a zero singular value
  - So we can correct our estimate by taking the SVD,  $F = USV^T$ , and setting the smallest value in  $S$  to 0 to give  $S'$ , our corrected result is then  $F' = US'V^T$

### Estimating $E$

- If we know the calibration matrices we can compute  $E$  from  $F$
- This means we can use the 8-point algorithm and have a simple linear solution
- It is possible to compute  $E$  from just 5 points
  - This is much more complicated, and not as stable
  - However, it means that  $s = 5$  in RANSAC, so can be more robust to larger outlier rates

### Non-Linear Optimisation

- As with homographies, the solution we get from the linear equations doesn't quite minimise what we want
- We want to minimise the reprojection error for some unknown 3D point
- Next lecture we will see how to find 3D points from our 2D correspondences and  $E$