

# Camera Imaging of Various Geometrical Forms (including the Absolute Conic)

Reference: "Multiple View Geometry in Computer Vision" by Hartley and Zisserman

- In this lecture, we are interested in the images formed by a camera for various geometrical forms in the physical 3D space. We are specifically interested in the imaging of planes, lines, conics, and quadrics. We will also be interested in how a camera images the Absolute Conic because of the important role it plays in modern camera calibration algorithms.

## How a Plane in 3D is Imaged by a Camera

- This section **proves** that the relationship between a planar scene and its camera image **is always a homography** — that is, a linear relationship when we use homogeneous 3-vectors for representing the coordinates — regardless of the pose of the camera with respect to the scene.

- We assume that the planar scene is in the  $Z=0$  plane of the world coordinate frame, and that the camera is in an arbitrary pose with respect to this plane. Writing the  $3 \times 4$  camera projection matrix  $P$  as  $P = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_3 \ \vec{p}_4]$ , we have for the pixel coordinates

$$\vec{x} = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_3 \ \vec{p}_4] \vec{X} = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_3 \ \vec{p}_4] \begin{pmatrix} x \\ y \\ 0 \\ w \end{pmatrix} = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_3] \begin{pmatrix} x \\ y \\ w \end{pmatrix} = H \vec{x}_w$$

↳ pixel
↳ world point
↳ 3x3 homography

where  $\vec{x}_w$  is the homogeneous 3-vector representation of a world point in the  $Z=0$  plane of the world frame. If we prefer to write  $P$  as  $P = K[R|\vec{t}]$  (see page 18-3 of Lecture 18), it follows that the  $3 \times 3$  homography  $H$  shown above would become  $H = K[\vec{r}_1 \ \vec{r}_2 \ \vec{t}]$ , where  $\vec{r}_1$  and  $\vec{r}_2$  are the first two columns of the  $3 \times 3$  rotation matrix  $R$ , and  $\vec{t}$  the translation vector from the world frame origin to the camera frame origin.

## How a Line in 3D is Imaged by a Camera

- As you'd expect, it is trivial to show that the camera image of a 3D line is a line. Following Lecture 6, let's represent a 3D line as a vector span of two world points  $\vec{A}$  and  $\vec{B}$ . Any world point on this line can be expressed as  $\vec{X} = \lambda_1 \vec{A} + \lambda_2 \vec{B} \equiv \vec{A} + \lambda \vec{B}$  for arbitrary values of the coefficients involved. The image of such a world point is given by  $\vec{x}(\lambda) = P\vec{A} + \lambda P\vec{B} = \vec{a} + \lambda \vec{b}$  where  $\vec{a}$  and  $\vec{b}$  are the coordinates of the pixels for the images of  $\vec{A}$  and  $\vec{B}$ . Obviously, all the points  $\vec{x}(\lambda)$  in the image form a straight line.



- While we are on the subject of lines, let's talk about **backprojecting** image lines into the 3D space of the world coordinate frame. As you would expect, a line in the camera image backprojects into a plane in the world frame. If a 3-vector  $\vec{l}$  is the homogeneous representation of a line in a camera image, the homogeneous representation of the world plane that  $\vec{l}$  backprojects to is given by  $P^T \vec{l}$ . PROOF: The set of pixels  $\vec{x}$  on  $\vec{l}$  must obey  $\vec{x}^T \vec{l} = 0$ . Now let  $\vec{x}$  be the image of some world point  $\vec{X}$ . We have  $\vec{x} = P\vec{X}$ . Substituting  $\vec{x} = P\vec{X}$  in  $\vec{x}^T \vec{l} = 0$ , we get  $\vec{X}^T P^T \vec{l} = 0$ , which (from Lecture 6) implies that  $\vec{X}$  is on a world plane  $\Pi$  whose homogeneous representation is the 4-vector given by  $\vec{\Pi} = P^T \vec{l}$ .

## Backprojecting Conics

- A reader might ask: Shouldn't we first talk about the camera imaging of conics in world-3D before addressing the problem of backprojecting conics? **The reason we gloss over the problem of how a camera images a conic in world-3D is that there is nothing special about it.** A conic in the world coordinate frame (the conic must reside in a plane in the world frame) is imaged by a camera using the same homography that you saw on page 20-1 when we talked about how a camera images a plane.
- So let's talk about backprojecting conics from the image plane into the world coordinate frame. Consider a conic whose homogeneous representation is a 3x3 matrix  $C$ .
- As you can imagine, the conic  $C$  will backproject into a cone-like object in the world frame. The apex of this cone will be at the camera center and its cross-sections in any plane parallel to the image plane a scaled version of  $C$ . **Such a cone in the world frame is a degenerate quadric** whose homogeneous representation is given by the 4x4 matrix  $Q_{co} = P^T C P$ . PROOF: All pixels  $\vec{x}$  on the conic  $C$  obey  $\vec{x}^T C \vec{x} = 0$ . Now let  $\vec{X}$  be a world point whose image is at the pixel  $\vec{x}$ . Obviously,  $\vec{x} = P\vec{X}$ . Substituting this in  $\vec{x}^T C \vec{x} = 0$ , we get  $\vec{X}^T P^T C P \vec{X} = 0$ , which completes the proof since all points  $\vec{X}$  on a quadric  $Q$  must obey  $\vec{X}^T Q \vec{X} = 0$ .
- Note that  $Q_{co} = P^T C P$  is NOT of full rank, because  $P$  is only of rank 3. It is trivial to show that the null vector of  $Q_{co}$  is the same as that of  $P$ . You can show that by multiplying <sup>on the right</sup> both sides of  $Q_{co} = P^T C P$  by the null vector of  $P$ . It is because  $Q_{co}$  is of reduced rank that it has a cone-like shape in 3D.

## How a Quadric is Imaged by a Camera

- The camera image of a point quadric  $Q$  — **the image is just a silhouette of the quadric** — is a point conic  $C$ . The  $Q$  and  $C$  are related by  $C^* = P Q^* P^T$  where  $Q^*$  is the dual of the point quadric  $Q$  and  $C^*$  the dual of the point conic  $C$ . PROOF: The lines  $\vec{l}$  that are tangent



to the point conic  $C$  must obey  $\vec{l}^T C^* \vec{l} = 0$ . Additionally, as you saw in the first bullet on the previous page, each such image line backprojects into the plane  $\vec{\pi} = P^T \vec{l}$  in the world frame. From the geometry of image formation, we know that such planes must be tangential to the point quadric  $Q$ . That is, we must have  $\vec{\pi}^T Q^* \vec{\pi} = 0$ . Substituting in this the relationship  $\vec{\pi} = P^T \vec{l}$ , we get  $\vec{l}^T P Q^* P^T \vec{l} = 0$ , which implies  $C^* = P Q^* P^T$ .

## How Does a Camera Image The Absolute Conic

- In Lecture 7, we defined the Absolute Conic  $\Omega_\infty$  as the intersection of  $\vec{\pi}_\infty$  and any arbitrary sphere in world 3D. The points on  $\vec{\pi}_\infty$  are all ideal and are defined by homogeneous vectors  $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ 0 \end{pmatrix}$ . On the plane  $\vec{\pi}_\infty$ ,  $\Omega_\infty$  is defined by  $(x_1 \ x_2 \ x_3) I \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ 0 \end{pmatrix} = 0$  where  $I$  is the 3x3 identity matrix.
- To find the camera image of  $\Omega_\infty$ , let's first focus on the camera image of a single point  $\vec{X} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ 0 \end{pmatrix}$  on the plane  $\vec{\pi}_\infty$ . The pixel coordinates of the image point for  $\vec{X}$  are:

pixel  $\rightarrow \vec{x} = P\vec{X} = KR [I_{3 \times 3} \mid -\vec{c}] \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ 0 \end{pmatrix} = KR \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = KR \vec{x}_d = H \vec{x}_d$

where  $\vec{x}_d = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$  means a "direction vector" to the point  $\vec{X} = \begin{pmatrix} \vec{x}_d \\ 0 \end{pmatrix}$  on the plane  $\vec{\pi}_\infty$ . To see that we can associate the meaning of **direction** with the first three elements of  $(x_1 \ x_2 \ x_3 \ x_4)^T$ , note that as we go from the point  $\vec{X} = (x_1, x_2, x_3, x_4)^T$  to the point  $(kx_1, kx_2, kx_3, x_4)^T$ , we will continue to travel along the ray that joins  $\vec{X}$  with the camera center.

Therefore, what remains invariant as we travel along the same ray all the way to  $\vec{\pi}_\infty$  are the ratios of the three elements  $x_1, x_2$ , and  $x_3$ .

- Therefore, if the 3-vector  $\vec{x}_d$  in the derivation shown above is interpreted as a direction vector, it is homogeneous in the same sense as the pixel coordinate vector  $\vec{x}$  on the left. The derivation shown above says that the direction vectors to points on  $\vec{\pi}_\infty$  are related to their corresponding pixels by the homography  $H = KR$ .

- The Absolute Conic  $\Omega_\infty$  is defined by the direction vectors  $\vec{x}_d$  that obey  $\vec{x}_d^T I_{3 \times 3} \vec{x}_d = 0$ . You know from Lecture 3 that under a homography  $H$ , a conic  $C$  transforms as  $C' = H^T C H^{-1}$ . Since the image formation from the direction vectors  $\vec{x}_d$  to the pixels  $\vec{x}$  is the homography  $H = KR$ , the image of the Absolute Conic is given by

$\omega = H^{-T} I_{3 \times 3} H^{-1} = (KR)^{-T} (KR)^{-1} = ((KR)^T)^{-1} (KR)^{-1}$   
 $= (R^T K^T)^{-1} (KR)^{-1} = K^{-T} R^{-T} R^{-1} K^{-1} = K^{-T} (R R^T)^{-1} K^{-1} = K^{-T} K^{-1}$

The actual pixels on the image conic  $\omega$  would be  $\vec{x}^T \omega \vec{x} = 0$ . As you would expect, all these pixels are imaginary because  $K^{-T} K^{-1}$  is positive definite.

Inverse of upper triangular matrix is always upper triangular  
 " " " " " lower " " " " "

When you multiply a pos-def matrix by its transpose, you get a pos-def matrix  
 If  $x^T A x > 0$  for all  $x$

A conic  $C$  must NOT be pos-def since  $x^T C x = 0$

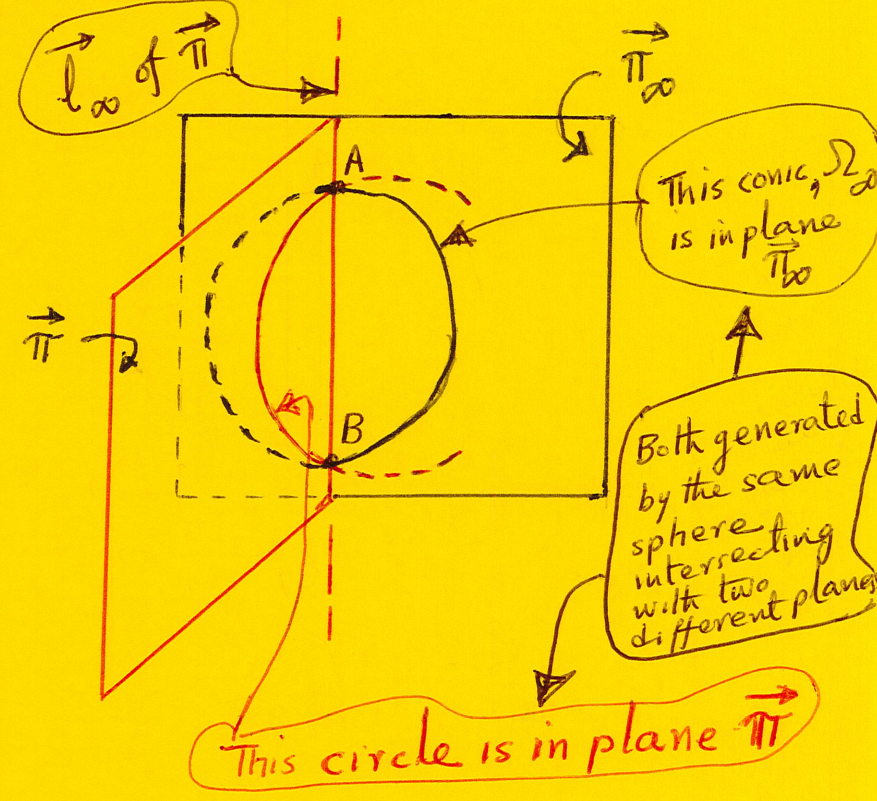
Image of the Absolute Conic



# A Most Important Property of the Absolute Conic

- Any arbitrary plane  $\vec{\pi}$  in world 3D samples the Absolute Conic at exactly two points — the two circular points  $\vec{I}$  and  $\vec{J}$  of  $\vec{\pi}$ . (See Lecture 4 for the definition of the Circular Points.)
- In order to establish this property, the first thing to note is that the Absolute Conic  $\Omega_\infty$  resides in the plane  $\vec{\pi}_\infty$  and any given plane  $\vec{\pi}$  intersects the plane  $\vec{\pi}_\infty$  in the former's line  $\vec{l}_\infty$ . Next, we need to show that  $\vec{l}_\infty$  intersects ~~the~~ the conic  $\Omega_\infty$  at exactly two points — the circular points of  $\vec{\pi}$ . This is best done with the help of the figure shown below:

The two points A and B shown are on the  $\vec{l}_\infty$  line of plane  $\vec{\pi}$  because that's where the plane  $\vec{\pi}$  meets the plane  $\vec{\pi}_\infty$ . The blue conic  $\Omega_\infty$  is formed by the intersection of an arbitrary sphere with  $\vec{\pi}_\infty$ . And the red circle is formed by the intersection of the same sphere with the plane  $\vec{\pi}$ . The two points A and B where the red circle on plane  $\vec{\pi}$  meets the plane  $\vec{\pi}_\infty$  must be on the line  $\vec{l}_\infty$  of plane  $\vec{\pi}$ . That implies that A and B are the two Circular Points of the plane  $\vec{\pi}$ .



# Calibrating a Camera's Intrinsic Parameters by Waving a 2D Pattern in Front of It

- The property of the Absolute Conic described leads to a novel algorithm for calculating the intrinsic parameters (as represented by the elements of the 3x3 matrix K) of a camera. **Estimating K is a very important part of camera calibration.**
- The idea is to let the camera record at least three different images of a 2D pattern as you "wave" it in front of the camera. The orientation of the pattern must be different for each image that is recorded. The translation does not matter.
- The plane corresponding to each pose of the pattern will sample  $\Omega_\infty$  at the two Circular Points,  $\vec{I} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$  and  $\vec{J} = \begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix}$ , in that plane.
- Let's assume the pattern is visually rich enough to allow us to compute the homography  $H = [\vec{h}_1 \ \vec{h}_2 \ \vec{h}_3]$  from the plane of the pattern to the camera image plane. Applying H to  $\vec{I}$  and  $\vec{J}$ , we get  $H\vec{I}$  and  $H\vec{J}$  as two points on the image conic  $\omega$  defined at the bottom of the previous page. Since  $H\vec{I} = \vec{h}_1 + i\vec{h}_2$  and  $H\vec{J} = \vec{h}_1 - i\vec{h}_2$ , both these points on  $\omega$  must satisfy the  $\vec{x}^T \omega \vec{x} = 0$  condition. So we get the two equations:
 
$$\begin{cases} (\vec{h}_1 + i\vec{h}_2)^T \omega (\vec{h}_1 + i\vec{h}_2) = 0 \\ (\vec{h}_1 - i\vec{h}_2)^T \omega (\vec{h}_1 - i\vec{h}_2) = 0 \end{cases} \Rightarrow \begin{cases} \vec{h}_1^T \omega \vec{h}_1 - \vec{h}_2^T \omega \vec{h}_2 + i2\vec{h}_1^T \omega \vec{h}_2 = 0 \\ \vec{h}_1^T \omega \vec{h}_1 - \vec{h}_2^T \omega \vec{h}_2 - i2\vec{h}_1^T \omega \vec{h}_2 = 0 \end{cases} \Rightarrow \begin{cases} \vec{h}_1^T \omega \vec{h}_1 - \vec{h}_2^T \omega \vec{h}_2 = 0 \\ \vec{h}_1^T \omega \vec{h}_2 = 0 \end{cases}$$
 where we get the middle pair of equations by  $\omega$  being symmetric, and the final pair by setting to zero separately the real and the imaginary parts. only for pos-def
- Each image gives us 2 equations for the 5 unknowns of  $\omega$ . We apply Cholesky decomposition to this  $\omega$  to recover  $K^{-1}$ . We invert that to get K. Cholesky breaks  $\omega$  into  $LL^T$  where L is lower triangular.