

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 3×4 camera projection matrix P as $P = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_3 \ \vec{p}_4]$, we have for the pixel coordinates

$$\underset{\substack{\leftarrow \text{pixel}}}{\vec{x}} = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_3 \ \vec{p}_4] \underset{\substack{\leftarrow \text{world point}}}{\vec{X}} = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_3 \ \vec{p}_4] \begin{pmatrix} x \\ y \\ 0 \\ w \end{pmatrix} = [\vec{p}_1 \ \vec{p}_2 \ \vec{p}_4] \begin{pmatrix} x \\ y \\ w \end{pmatrix} = \underset{\substack{\leftarrow 3 \times 3 \text{ homography}}}{H} \vec{x}_w$$

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 16-3 of Lecture 16), it follows that the 3×3 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 3×3 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 Π whose homogeneous representation is the 4-vector given by $\vec{\Pi} = P^T \vec{l}$.

Backprojecting Conics

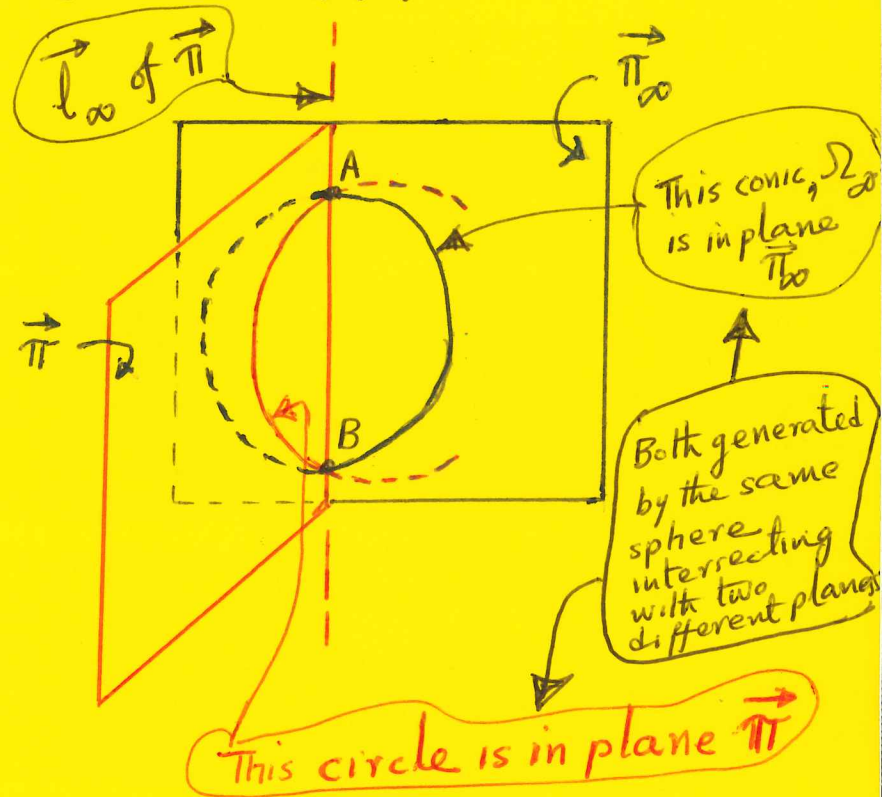
- A reader might ask: Shouldn't we first talk about the camera imaging of conics in world-3D before addressing the problem of backprojecting conics? **The reason we gloss over the problem of how a camera images a conic in world-3D is that there is nothing special about it.** A conic in the world coordinate 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 18-1 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 3×3 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 4×4 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 on the right 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

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 Ω_∞ 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}_∞ . Next, we need to show that \vec{l}_∞ intersects the conic Ω_∞ 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}_∞ line of plane $\vec{\pi}$ because that's where the plane $\vec{\pi}$ meets the plane $\vec{\pi}_\infty$. The blue conic Ω_∞ 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}_∞ 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 3×3 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 Ω_∞ 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 $H \cdot \vec{I}$ and $H \cdot \vec{J}$ as two points on the image conic ω defined at the bottom of the previous page. Since $H \cdot \vec{I} = \vec{h}_1 + i\vec{h}_2$ and $H \cdot \vec{J} = \vec{h}_1 - i\vec{h}_2$, both these points on ω must satisfy the $\vec{x}^T \omega \vec{x} = 0$ condition. So we get the two equations:

$$\begin{aligned} (\vec{h}_1 + i\vec{h}_2)^T \omega (\vec{h}_1 + i\vec{h}_2) &= 0 \\ (\vec{h}_1 - i\vec{h}_2)^T \omega (\vec{h}_1 - i\vec{h}_2) &= 0 \end{aligned} \Rightarrow \begin{aligned} \vec{h}_1^T \omega \vec{h}_1 - \vec{h}_2^T \omega \vec{h}_2 + i2\vec{h}_1^T \omega \vec{h}_2 &= 0 \\ \vec{h}_1^T \omega \vec{h}_1 - \vec{h}_2^T \omega \vec{h}_2 - i2\vec{h}_1^T \omega \vec{h}_2 &= 0 \end{aligned} \Rightarrow \begin{aligned} \vec{h}_1^T \omega \vec{h}_1 - \vec{h}_2^T \omega \vec{h}_2 &= 0 \\ \vec{h}_1^T \omega \vec{h}_2 &= 0 \end{aligned}$$
 where we get the middle pair of equations by ω 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 ω . We apply Cholesky decomposition to this ω to recover K^{-1} . We invert that to get K . Cholesky breaks ω into LL^T where L is lower triangular.