

COSC579: Scene Geometry

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Overview

- Linear Algebra Review
 - Homogeneous vs non-homogeneous representations
 - Projections and Transformations
- Scene Geometry
- The pinhole projection model
 - Qualitative properties
 - Perspective projection matrix
- Cameras with lenses
 - Depth of focus
 - Field of view
 - Lens aberrations



Scene Geometry, Projection, and Perspective



- Readings
 - Szeliski 2.1



Projection



- Readings
 - Szeliski 2.1



Dimensionality Reduction (3D to 2D)



3D world



2D image

Point of observation

What have we lost?

- Angles
- Distances (lengths)



Linear Algebra and Projections

- Before we dive into camera geometry, lets introduce some notation and review linear algebra.
- 2-D points will be used to represent pixel coordinates: points in a (image) plane. (non-homogeneous)

$$x=(x,y)\in \mathcal{R}^2,$$
 $x=\left[egin{array}{c}x\\y\end{array}
ight]$



Homogeneous and Non-homogeneous Vectors

- 2-D points can be represented by a vector and a weight.
 - Homogeneous
 - P² is the 2-d projective space

 $ilde{x} = (ilde{x}, ilde{y}, ilde{w}) \in \mathcal{P}^2$

 $\tilde{x} = (\tilde{x}, \tilde{y}, \tilde{w}) = \tilde{w}(x, y, 1) = \tilde{w}\bar{x}$

where $\bar{x} = (x, y, 1)$ is the *augmented vector*.



Why Homogeneous Coordinates?

Homogeneous representation permits linear matrix
 operations for transformations used in Computer Vision

How? add one more coordinate:

 $(x, y) \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \qquad (x, y, z) \Rightarrow \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$ homogeneous image coordinates $\begin{array}{c} x \\ y \\ z \\ 1 \end{bmatrix}$

Converting from homogeneous coordinates

$$\begin{bmatrix} x \\ y \\ w \end{bmatrix} \Rightarrow (x/w, y/w) \qquad \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} \Rightarrow (x/w, y/w, z/w)$$



Lines

- 2-D Lines can also be represented using homogeneous coordinates $\tilde{l} = (a, b, c)$
- Corresponding line equation

$$\bar{x} \cdot \tilde{l} = ax + by + c = 0.$$



• It is common to normalize the line equation vector

 $l = (\hat{n}_x, \hat{n}_y, d) = (\hat{n}, d)$ with $\|\hat{n}\| = 1$.

• Observe: in this formulation \hat{n} is the unit normal vector perpendicular to the line and d is its distance to the origin



Lines

 Computing the intersection of 2 lines using homogeneous coordinates

$$ilde{x} = ilde{l}_1 imes ilde{l}_2$$

Computing the line joining two points using homogeneous coordinates

$$\tilde{l} = \tilde{x}_1 imes \tilde{x}_2.$$



Similar representations exist to model 3-D points: points in a scene.

$$=(x,y,z)\in \mathcal{R}^3$$
 $ilde{x}=(ilde{x}, ilde{y}, ilde{z}, ilde{w})\in \mathcal{P}^3$

• 3-D planes specification using homogeneous coordinates

 $\tilde{m} = (a, b, c, d)$

• Equation of plane

 ${old x}$

$$\bar{x} \cdot \tilde{m} = ax + by + cz + d = 0.$$

• Normalized

 $m = (\hat{n}_x, \hat{n}_y, \hat{n}_z, d) = (\hat{n}, d)$ with $\|\hat{n}\| = 1$



Lines in 3-D

- Not as easy to represent
- One method: model line segment r as convex combination of two points on the line.
 - Assume p and q are points on the line.
 - $0 \leq \lambda \leq 1$ $ilde{r} = \mu ilde{p} + \lambda ilde{q}$
- Homogeneous coordinates

3D line equation, $r = (1 - \lambda)p + \lambda q$.



Lines in 3-D

 More generally, we can represent a line in 3-D using a point and a gradient

 $r = p + \lambda \, \hat{d}$ $\hat{d} = [\Delta x, \Delta y, \Delta z]$





2D Planar Transformations





Translations

• Here the augmented vector \bar{x} is used to account for translation t

Translation. 2D translations can be written as x' = x + t or

$$x' = \left[egin{array}{cc} I & t \end{array}
ight]ar{x}$$

where I is the (2×2) identity matrix or

• Simply a shift





Euclidean Transformations

Rotation + translation. This transformation is also known as 2D rigid body motion or the 2D Euclidean transformation (since Euclidean distances are preserved). It can be written as x' = Rx + t or

$$\boldsymbol{x}' = \begin{bmatrix} \boldsymbol{R} & \boldsymbol{t} \end{bmatrix} \bar{\boldsymbol{x}} \tag{2.16}$$

where

$$\boldsymbol{R} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$
(2.17)

is an orthonormal rotation matrix with $RR^T = I$ and |R| = 1.



Scaled Rotations

Scaled rotation. Also known as the *similarity transform*, this transformation can be expressed as x' = sRx + t where s is an arbitrary scale factor. It can also be written as

$$\boldsymbol{x}' = \begin{bmatrix} s\boldsymbol{R} & \boldsymbol{t} \end{bmatrix} \bar{\boldsymbol{x}} = \begin{bmatrix} a & -b & t_x \\ b & a & t_y \end{bmatrix} \bar{\boldsymbol{x}}, \quad (2.18)$$

where we no longer require that $a^2 + b^2 = 1$. The similarity transform preserves angles between lines.



Affine

Affine. The affine transformation is written as $x' = A\bar{x}$, where A is an arbitrary 2×3 matrix, i.e.,

$$x' = \begin{bmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \end{bmatrix} \bar{x}.$$
 (2.19)

Parallel lines remain parallel under affine transformations.



Projective Transformation (Homography)

Projective. This transformation, also known as a *perspective transform* or *homography*, operates on homogeneous coordinates,

$$\tilde{x}' = \tilde{H}\tilde{x},\tag{2.20}$$

where \tilde{H} is an arbitrary 3×3 matrix. Note that \tilde{H} is homogeneous, i.e., it is only defined up to a scale, and that two \tilde{H} matrices that differ only by scale are equivalent. The resulting homogeneous coordinate \tilde{x}' must be normalized in order to obtain an inhomogeneous result x, i.e.,

$$x' = \frac{h_{00}x + h_{01}y + h_{02}}{h_{20}x + h_{21}y + h_{22}} \text{ and } y' = \frac{h_{10}x + h_{11}y + h_{12}}{h_{20}x + h_{21}y + h_{22}}.$$
 (2.21)

Perspective transformations preserve straight lines (i.e., they remain straight after the transformation).



Properties of Transformations

Transformation	Matrix	# DoF	Preserves	Icon
translation	$\left[egin{array}{c c} I & t \end{array} ight]_{2 imes 3}$	2	orientation	
rigid (Euclidean)	$\left[egin{array}{c c} {m R} & t \end{array} ight]_{2 imes 3}$	3	lengths	\diamond
similarity	$\left[\begin{array}{c c} s oldsymbol{R} & t \end{array} \right]_{2 imes 3}$	4	angles	\diamond
affine	$\left[egin{array}{c} A \end{array} ight]_{2 imes 3}$	6	parallelism	\square
projective	$\left[egin{array}{c} ilde{H} \end{array} ight]_{3 imes 3}$	8	straight lines	\square



3D transformations

Translation. 3D translations can be written as x' = x + t or

$$x' = \left[egin{array}{cc} I & t \end{array}
ight] ar{x}$$



3D Euclidean Transformation

Rotation + translation. Also known as 3D *rigid body motion* or the 3D *Euclidean trans*formation, it can be written as x' = Rx + t or

$$\boldsymbol{x}' = \begin{bmatrix} \boldsymbol{R} & \boldsymbol{t} \end{bmatrix} \bar{\boldsymbol{x}} \tag{2.24}$$

where R is a 3 × 3 orthonormal rotation matrix with $RR^T = I$ and |R| = 1. Note that sometimes it is more convenient to describe a rigid motion using

$$x' = R(x - c) = Rx - Rc,$$
 (2.25)

where c is the center of rotation (often the camera center).



3D similarity transformation

Scaled rotation. The 3D *similarity transform* can be expressed as x' = sRx + t where s is an arbitrary scale factor. It can also be written as

$$x' = \begin{bmatrix} sR & t \end{bmatrix} \bar{x}. \tag{2.26}$$

This transformation preserves angles between lines and planes.



3D Affine Transformation

Affine. The affine transform is written as $x' = A\bar{x}$, where A is an arbitrary 3×4 matrix, i.e.,

$$\boldsymbol{x}' = \begin{bmatrix} a_{00} & a_{01} & a_{02} & a_{03} \\ a_{10} & a_{11} & a_{12} & a_{13} \\ a_{20} & a_{21} & a_{22} & a_{23} \end{bmatrix} \bar{\boldsymbol{x}}.$$
 (2.27)

Parallel lines and planes remain parallel under affine transformations.



Projective Transformation

Projective. This transformation, variously known as a 3D perspective transform, homography, or collineation, operates on homogeneous coordinates,

$$\tilde{x}' = \tilde{H}\tilde{x},$$
 (2.28)

where \tilde{H} is an arbitrary 4 × 4 homogeneous matrix. As in 2D, the resulting homogeneous coordinate \tilde{x}' must be normalized in order to obtain an inhomogeneous result x. Perspective transformations preserve straight lines (i.e., they remain straight after the transformation).



3D Rotations

- Sequential Rotations Approach (Euler Angle).
 - Perform 3 sequential rotations around 3 cardinal axes.
 - Cons
 - Result depends on order of rotations.
 - Not always practical
- Axis / Angle Approach
 - Rotation can be determined by 1 rotation axis and 1 angle



Rotation around an axis \hat{n} by an angle θ .



3D Axis/Angle Rotation

• Constructing the rotation matrix R, given \hat{n} and θ

$$\boldsymbol{R}(\hat{\boldsymbol{n}},\theta) = \boldsymbol{I} + \sin\theta[\hat{\boldsymbol{n}}]_{\times} + (1-\cos\theta)[\hat{\boldsymbol{n}}]_{\times}^2$$

$$\hat{\boldsymbol{n}} = (\hat{n}_x, \hat{n}_y, \hat{n}_z)$$

$$[\hat{n}]_{ imes} = \left[egin{array}{ccc} 0 & -\hat{n}_z & \hat{n}_y \ \hat{n}_z & 0 & -\hat{n}_x \ -\hat{n}_y & \hat{n}_x & 0 \end{array}
ight]$$



For Rotation around an axis \hat{n} by an angle θ .



Projecting from 3D to 2D

- Projections
 - Orthographic and scaled orthographic (weak perspective) projection
 - Para-perspective projection
 - Perspective projection
- These projects are used often during image formation.



Modeling projection



- The coordinate system
 - We will use the pin-hole model as an approximation
 - Put the optical center (Center Of Projection) at the origin
 - Put the image plane (Projection Plane) in front of the COP
 - Why?
 - The camera looks down the *negative* z axis
 - · we need this if we want right-handed-coordinates



Orthographic projection

- Special case of perspective projection
 - Distance from the COP to the PP is infinite (ignore z-axis)



- Good approximation for telephoto optics
- Also called "parallel projection": $(x, y, z) \rightarrow (x, y)$
- What's the projection matrix?

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \Rightarrow (x, y)$$



Variants of orthographic projection

- Scaled orthographic
 - Also called "weak perspective"

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1/d \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 1/d \end{bmatrix} \Rightarrow (dx, dy)$$

- Affine projection
 - Also called "para-perspective"

$$\left[\begin{array}{cccc}a&b&c&d\\e&f&g&h\\0&0&0&1\end{array}\right]\left[\begin{array}{ccc}x\\y\\z\\1\end{array}\right]$$



Modeling projection



- Projection equations
 - Compute intersection with PP of ray from (x,y,z) to COP
 - Derived using similar triangles (Try this at Home!)

$$(x, y, z) \rightarrow (-d\frac{x}{z}, -d\frac{y}{z}, -d)$$

• We get the projection by throwing out the last coordinate:

$$(x, y, z) \rightarrow (-d\frac{x}{z}, -d\frac{y}{z})$$



Perspective Projection

• Projection is a matrix multiply using homogeneous coordinates:

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1/d & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ -z/d \end{bmatrix} \Rightarrow (-d\frac{x}{z}, -d\frac{y}{z})$$

divide by third coordinate

This is known as perspective projection

- The matrix is the **projection matrix**
- Can also formulate as a 4x4

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & -1/d & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ z \\ -z/d \end{bmatrix} \Rightarrow (-d\frac{x}{z}, -d\frac{y}{z})$$
divide by fourth coordinate GEORGETOR OF CONTINUERS (The second sec

Projection Review

- Reviewed
 - Linear algebra and notation
 - Projections
- Projections are used to model image formation and analysis
- To see where projections fit in, lets investigate camera models





Camera Models

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Image formation



- Let's design a camera
 - Idea 1: put a piece of film in front of an object
 - Do we get a reasonable image?


Pinhole camera



- Add a barrier to block off most of the rays
 - This reduces blurring
 - The opening known as the aperture
 - How does this transform the image?



Camera Obscura: Pinhole model



- The first camera
 - Known to Aristotle
 - How does the aperture size affect the image?
- Pinhole model:
 - Captures **pencil of rays** all rays through a single point
 - The point is called **Center of Projection (focal point)**
 - The image is formed on the Image Plane



Shrinking the aperture



- Why not make the aperture as small as possible?
 - Less light gets through
 - Diffraction effects...



Shrinking the aperture



2 mm

1 mm







0.15 mm



- The human eye is a camera
 - Iris colored annulus with radial muscles
 - **Pupil** the hole (aperture) whose size is controlled by the iris
 - What's the "film"?
 - photoreceptor cells (rods and cones) in the retina



Adding a lens



- A lens focuses light onto the film
 - Rays passing through the center are not deviated



Adding a lens



- A lens focuses light onto the film
 - There is a <u>specific distance</u> at which objects are "<u>in</u> <u>focus</u>"
 - other points project to a "circle of confusion" in the image
 - Changing the shape of the lens changes this distance EORGETOWN

Lenses



- A lens focuses parallel rays onto a single focal point
 - focal point at a distance *f* beyond the plane of the lens
 - f is a function of the shape and index of refraction of the lens
 - Aperture of diameter D restricts the range of rays
 - aperture may be on either side of the lens
 - Lenses are typically spherical (easier to produce)



Thin lenses



• Thin lens equation:
$$\frac{1}{d_o} + \frac{1}{d_i} = \frac{1}{f}$$

- Any object point satisfying this equation is in focus
- What is the shape of the focus region?
- How can we change the focus region?
- Thin lens applet: http://www.phy.ntnu.edu.tw/java/Lens/lens_e.html (by Fu-Kwun Hwang)



Depth of field





f/5.6





- Changing the aperture size affects depth of field
 - A smaller aperture increases the range in which the object is approximately in focus



Projection within the context of camera model

- Given our camera model, lets revisit transformations
- Image plane vs. Virtual image plane
- Image Formation
 - Translation projection: center the image coordinates
 - Affine Projection: accounts for camera position and orientation
 - Perspective Projection: image formation, 3d to 2d



Pinhole camera



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Pinhole camera terminology



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Deriving a Model for a Pinhole Camera

- We will first derive the pinhole camera model algebraically.
- Then we will revisit from a geometric perspective.



Camera parameters

• How can we model the geometry of a camera?





"The World"



Camera parameters

- To project a point (x, y, z) in world coordinates into a camera
- First transform (*x*,*y*,*z*) into *camera* coordinates
- Need to know
 - Camera position (in world coordinates)
 - Camera orientation (in world coordinates)
- The project into the image plane
 - Need to know camera *intrinsics*



Perspective Projection Matrix

• Projection is a matrix multiplication using homogeneous coordinates:

Observe: We can solve for image coordinates in terms of real-world coordinates and transformation parameters

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/f' & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ z/f' \end{bmatrix} \Rightarrow (f'\frac{x}{z}, f'\frac{y}{z})$$

divide by the third coordinate

In practice: lots of coordinate transformations...



Perspective projection (Intrinsics)



 Ω : aspect ratio (1 unless pixels are not square)

S : skew (0 unless pixels are shaped like rhombi/parallelograms)

 (c_x, c_y) : **principal point** ((0,0) unless optical axis doesn't intersect projection plane at origin)



Projection matrix (Extrinsics)





- How do we get the camera to "canonical form"?
 - (Center of projection at the origin, x-axis points right, yaxis points up, z-axis points backwards)



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- How do we get the camera to "canonical form"?
 - (Center of projection at the origin, x-axis points right, yaxis points up, z-axis points backwards)



Camera parameters

A camera is described by several parameters

- Translation t of the optical center from the origin of world coords
- Rotation R of the image plane
- focal length f, principle point (x'_c, y'_c),
- blue parameters are called "extrinsics," red are "intrinsics"

Projection equation

- The projection matrix models the cumulative effect of all parameters
- Useful to decompose into a series of operations

$$\mathbf{P}_{3x4} = \mathbf{K}[\mathbf{R} | \mathbf{t}] = \begin{bmatrix} -f & s & x'_c \\ 0 & -f & y'_c \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{R}_{3x3} | \mathbf{t}_{3x1} \end{bmatrix}$$
$$\widetilde{\mathbf{P}}_{4x4} = \widetilde{\mathbf{K}}\mathbf{E} = \begin{bmatrix} \mathbf{K} & \mathbf{0}^T \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} \mathbf{I}_{3x3} & \mathbf{0} \\ \mathbf{0}_{1x3} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{R}_{3x3} & \mathbf{t}_{3x1} \\ \mathbf{0}_{1x3} & 1 \end{bmatrix}$$
intrinsics rotation translation

Note: The definitions of these parameters are **not** completely standardized

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Pinhole Camera Model

- Derive from a geometric perspective.
- Using Princes Notation.



Geometric interpretation of homogeneous coordinates



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Normalized Camera



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Focal length parameters



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Focal length parameters

Can model both

- the effect of the distance to the focal plane
- the density of the receptors

with a single **focal length parameter f = d = \phi**

$$x = \frac{\phi u}{w} \quad y = \frac{\phi v}{w}$$

<u>To be overly cumbersome note:</u> In practice, the receptors may not be square:

$$x = \frac{\phi_x u}{w} \quad y = \frac{\phi_y v}{w}$$

So use different focal length parameter for x and y dims

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Offset parameters

- Current model assumes that pixel (0,0) is where the principal ray strikes the image plane (i.e. the center)
- Model offset to center

$$x = \frac{\phi_x u}{w} + \delta_x$$
$$y = \frac{\phi_y v}{w} + \delta_y$$

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Skew parameter

- Finally, add skew parameter
- Accounts for image plane being not exactly perpendicular to the principal ray

$$x = \frac{\phi_x u + \gamma v}{w} + \delta_x$$
$$y = \frac{\phi_y v}{w} + \delta_y$$

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Pinhole camera in homogeneous coordinates



In homogeneous coordinates:

$$\lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} \phi_x & \gamma & \delta_x & 0 \\ 0 & \phi_y & \delta_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix}$$
(linear!)

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Pinhole camera in homogeneous coordinates

$$\lambda \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} \phi_x & \gamma & \delta_x & 0 \\ 0 & \phi_y & \delta_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix}$$

Writing out these three equations

$$\lambda x = \phi_x u + \gamma v + \delta_x w$$
$$\lambda y = \phi_y v + \delta_y w$$
$$\lambda = w.$$

Eliminate λ to retrieve original equations

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Position and orientation of camera

- Position w=(u,v,w)^T of point in the world is generally not expressed in the frame of reference of the camera.
- Transform using 3D transformation

$$\begin{bmatrix} u'\\v'\\w' \end{bmatrix} = \begin{bmatrix} \omega_{11} & \omega_{12} & \omega_{13}\\ \omega_{21} & \omega_{22} & \omega_{23}\\ \omega_{31} & \omega_{32} & \omega_{33} \end{bmatrix} \begin{bmatrix} u\\v\\w \end{bmatrix} + \begin{bmatrix} \tau_x\\\tau_y\\\tau_z \end{bmatrix}$$

or





Complete pinhole camera model

$$x = \frac{\phi_x(\omega_{11}u + \omega_{12}v + \omega_{13}w + \tau_x) + \gamma(\omega_{21}u + \omega_{22}v + \omega_{23}w + \tau_y)}{\omega_{31}u + \omega_{32}v + \omega_{33}w + \tau_z} + \delta_x$$

$$y = \frac{\phi_y(\omega_{21}u + \omega_{22}v + \omega_{23}w + \tau_y)}{\omega_{31}u + \omega_{32}v + \omega_{33}w + \tau_z} + \delta_y.$$

 Intrinsic parameters matrix)

 $\{\phi_x, \phi_y, \gamma, \delta_x, \delta_y\}$

(stored as intrinsic

$$\mathbf{\Lambda} = \begin{bmatrix} \phi_x & \gamma & \delta_x \\ 0 & \phi_y & \delta_y \\ 0 & 0 & 1 \end{bmatrix}$$

• Extrinsic parameters

 $\{ oldsymbol{\Omega}, oldsymbol{ au} \}$

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At the end of the day...

- Scene geometry consists of internal and external parameters
- Observed light is focused onto an image plane
- The image is then captured (Image Acquisition)
- But first we will review radiometry: physics of LIGHT and COLOR

