

# Vision and Image Processing: Camera Models

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# Plan for today

- Before we start: vectors and matrices.
- Motivation and a bit of history.
- A general Introduction to Camera Models, specifically the pinhole model.
- Notions of projection and projective geometry
- 2 Views and Epipolar Geometry.



# Outline

- ① Vectors and Matrices
- ② Introduction
- ③ The Pinhole Camera
- ④ A bit of History
- ⑤ Projection
- ⑥ More on Camera Models
- ⑦ Stereo



# Vectors

- A  $n$ -vector is a  $n$ -uple of real values:

$$v = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

- Addition

$$\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} = \begin{pmatrix} x_1 + y_1 \\ \vdots \\ x_n + y_n \end{pmatrix}$$

- Multiplication by a scalar

$$\lambda \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} \lambda x_1 \\ \vdots \\ \lambda x_n \end{pmatrix}$$



# Linear Mapping

- Mapping between vectors with only addition of coordinates, multiplications by scalar and no constant terms.
- Example

$$f \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x + 3y \\ z - 2x \end{pmatrix}$$

- Non linear example

$$g \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x^2 + 3yz \\ z - 2x^2 + 1 \end{pmatrix}$$

There are powers and constant terms.



# Linearity

- This means  $f(v + \lambda v') = f(v) + \lambda f(v')$

$$\begin{aligned}f\left(\begin{pmatrix}x \\ y \\ z\end{pmatrix} + \lambda \begin{pmatrix}x' \\ y' \\ z'\end{pmatrix}\right) &= f\begin{pmatrix}x + \lambda x' \\ y + \lambda y' \\ z + \lambda z'\end{pmatrix} \\&= \begin{pmatrix}x + \lambda x' + 3(y + \lambda y') \\ z + \lambda z' - 2(x + \lambda x')\end{pmatrix} \\&= \begin{pmatrix}x + 3y \\ z - 2x\end{pmatrix} + \lambda \begin{pmatrix}x' + 3y' \\ z' - 2x'\end{pmatrix} \\&= f\begin{pmatrix}x \\ y \\ z\end{pmatrix} + \lambda f\begin{pmatrix}x' \\ y' \\ z'\end{pmatrix}\end{aligned}$$

- $f$  is linear.



# Matrices

- A  $n \times m$  matrix is an array of numbers with  $n$  rows and  $m$  columns
- A  $2 \times 3$  matrix  $F$

$$F = \begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix}$$

- 2 matrices **of the same size** can be added together: just add the entries:

$$\begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix} + \begin{pmatrix} 4 & -2 & 1 \\ 7 & -3 & 0 \end{pmatrix} = ?$$



# Product of a Matrix and a Vector

- A matrix of size  $m \times n$  and a vector of length  $m$  can be multiplied to form a vector of length  $n$ .
- Formal rule:

$$A = \begin{pmatrix} a_{11} & \dots & a_{1m} \\ \vdots & \vdots & \vdots \\ a_{n1} & \dots & a_{nm} \end{pmatrix}, v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$$

$$Av = \begin{pmatrix} a_{11}v_1 + a_{12}v_2 + \dots + a_{1m}v_n \\ a_{21}v_1 + a_{22}v_2 + \dots + a_{2m}v_n \\ \vdots \\ a_{n1}v_1 + a_{n2}v_2 + \dots + a_{nm}v_n \end{pmatrix}$$

- Each line of  $A$  is multiplied in “inner product way” with  $v$ .



- Example: Compute the product of

$$A = \begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix} \text{ and } v = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$



- Example: Compute the product of

$$A = \begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix} \text{ and } v = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

- We find precisely the value of

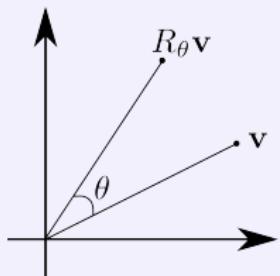
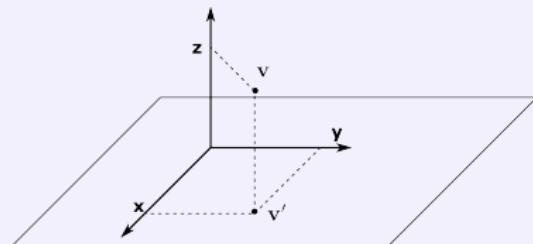
$$f \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x + 3y \\ -2x + z \end{pmatrix} = \begin{pmatrix} x + 3y \\ z - 2x \end{pmatrix}$$

- Each linear mapping can be written that way. Often use the same notation for the matrix and the linear mapping.

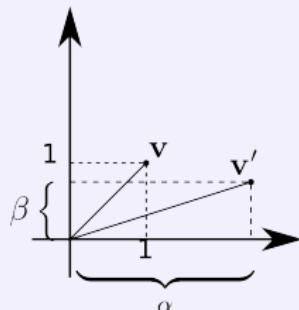


# Matrices /linear mappings as geometric transformations

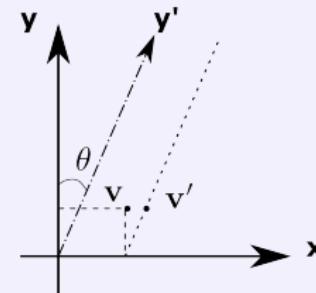
Projection on  $x - y$  plane



Rotation of angle  $\theta$



anisotropic scaling



shear



- projection  $\mathbb{R}^3 \rightarrow \mathbb{R}^2$

$$F \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix}, \quad F = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

- Rotation of angle  $\theta$  from  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$ :

$$R_\theta \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{pmatrix}, \quad R_\theta = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

- Scaling by a factor  $\alpha$  in  $x$  and  $\beta$  in  $y$ :

$$S \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \alpha x \\ \beta y \end{pmatrix}, \quad S = \begin{pmatrix} \alpha & 0 \\ 0 & \beta \end{pmatrix}$$

- Shear of the  $y$ -axis with angle  $\theta$ :

$$S \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x + \sin \theta y \\ y \end{pmatrix}, \quad S = \begin{pmatrix} 1 & \sin \theta \\ 0 & 1 \end{pmatrix}$$



# Product of Matrices

- A matrix of size  $m \times n$  and a matrix of size  $n \times p$  can be multiplied to produce a matrix of size  $m \times p$ .
- Algebraic rule:  $a_{ij}$  entry  $(i, j)$  of  $A$ ,  $b_{jk}$  entry  $(j, k)$  of  $B$

$$A = (a_{ij})_{\substack{i=1 \dots m \\ j=1 \dots n}} \quad B = (b_{jk})_{\substack{j=1 \dots n \\ k=1 \dots p}}$$

Denote entry  $(i, k)$  of product  $C = AB$  by  $c_{ik}$ :

$$c_{ik} = \sum_{j=1}^n a_{ij} b_{jk}$$

- Matrix vector multiplication is in fact a special case of it!
- Example

$$\begin{pmatrix} 2 & 2 \\ 1 & 3 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix} =$$



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$$\begin{pmatrix} 2 & 2 \\ 1 & 3 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix} = \begin{pmatrix} -2 & 6 & 2 \\ -5 & 3 & 3 \\ 3 & 3 & -1 \end{pmatrix}$$

- What does matrix multiplication means?



# Meaning of the Product

- $M$  and  $N$  the linear mappings  $\begin{pmatrix} 2 & 2 \\ 1 & 3 \\ 1 & -1 \end{pmatrix}$  and  $\begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix}$
- Apply  $N$  to  $\mathbf{v} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$  and  $M$  to the result:

$$N\mathbf{v} = \begin{pmatrix} 1 & 3 & 0 \\ -2 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x + 3y \\ -2x + z \end{pmatrix}$$

- and

$$\begin{pmatrix} 2 & 2 \\ 1 & 3 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x + 3y \\ -2x + z \end{pmatrix} =$$



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- and
- $$\begin{pmatrix} 2 & 2 \\ 1 & 3 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x + 3y \\ -2x + z \end{pmatrix} = \begin{pmatrix} -2x + 6y + 2z \\ -5x + 3y + 3z \\ 3x + 3y - z \end{pmatrix} =$$



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- and

$$\begin{pmatrix} 2 & 2 \\ 1 & 3 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x + 3y \\ -2x + z \end{pmatrix} = \begin{pmatrix} -2x + 6y + 2z \\ -5x + 3y + 3z \\ 3x + 3y - z \end{pmatrix} = \begin{pmatrix} -2 & 6 & 2 \\ -5 & 3 & 3 \\ 3 & 3 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$



# Matrix Product as Chain Application of Linear Mappings

- We found that

$$M(N\mathbf{v}) = \underbrace{MN}_{\text{Matrix product}} \mathbf{v}$$

- Very Important Property: Matrix product corresponds to chain application (composition) of linear mappings!

Read the Linear Algebra Tutorial and Reference on Absalon!

This will also be useful for other courses!

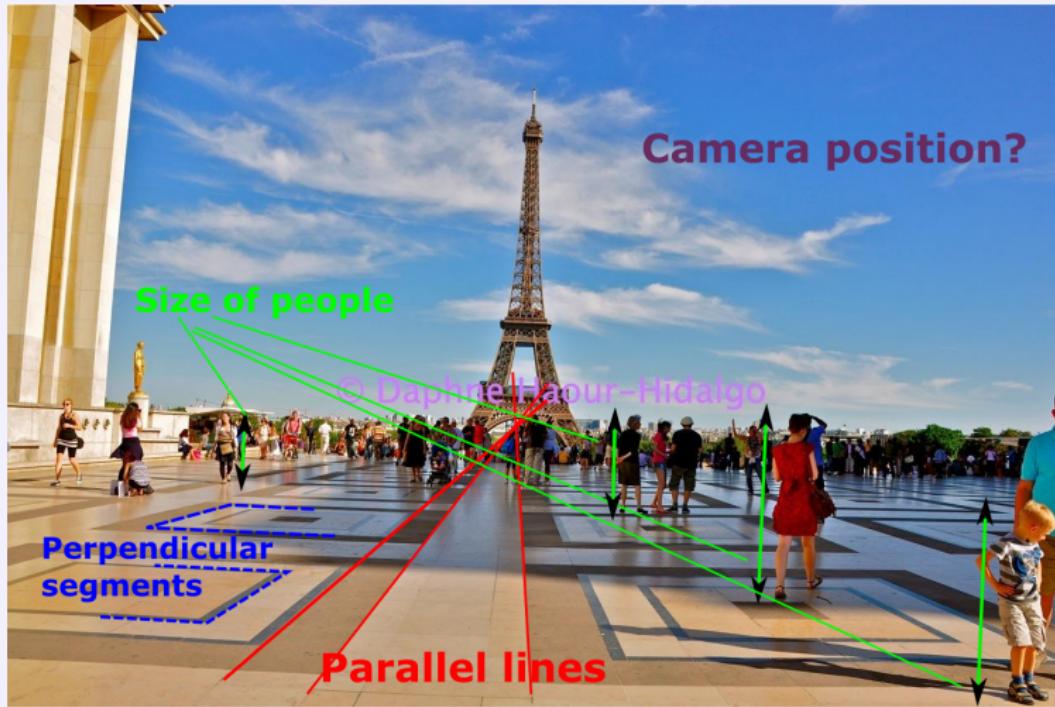


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# Motivation



# Questions

Previous picture raises some questions about:

- Lines?
- Parallelism?
- Angles / orthogonality?
- Sizes?
- Camera position / Horizon?

What happens when you take a picture (or Daphné Haour-Hidalgo in the previous case :-))

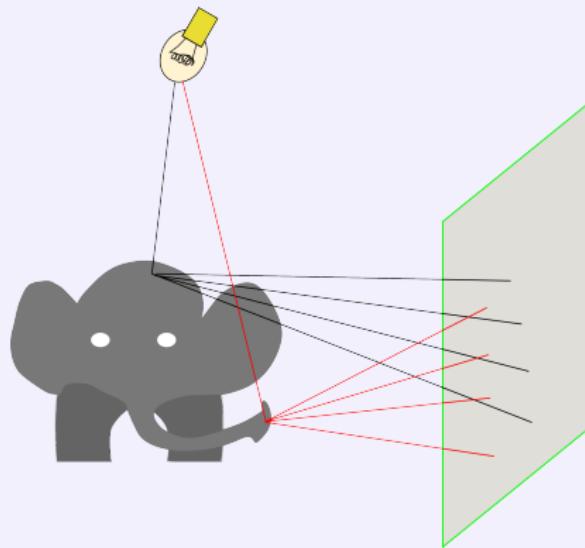


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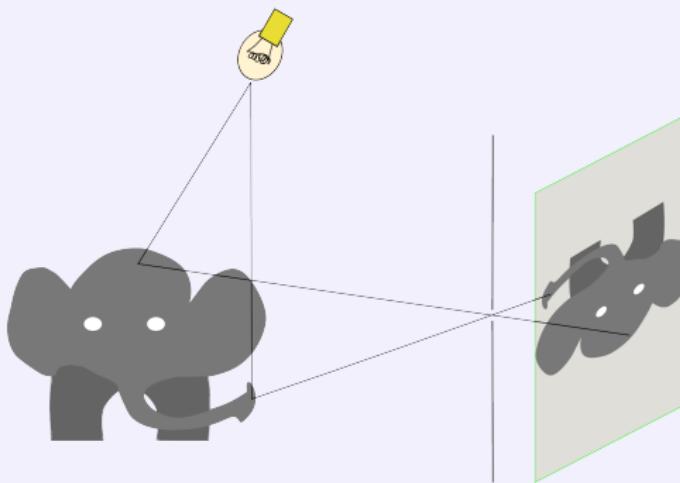
# Getting an Image – I



Many rays emanating from the same position touch the image sensitive array at many location: big blur!



# Getting an Image – II



Filtering the rays via a pinhole: get an (inverted) image.  
Principle of the **Camera Obscura** (dark room).

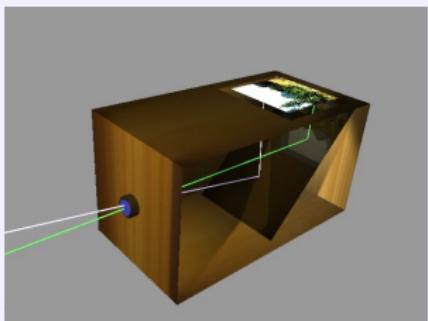


# Outline

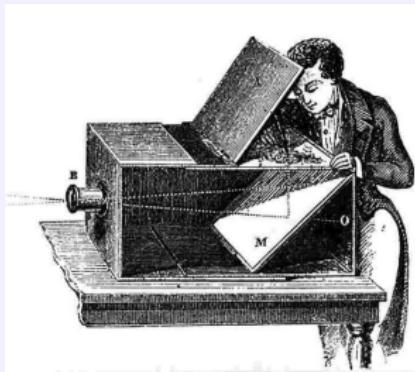
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# Camera Obscura



Principle of Camera Obscura



18th Century Camera Obscura

- Known from old Chinese writings
- Mentioned by Aristotle
- Plaque with photosensitive material: Photographic camera!



# The Very First Photography, 1826



J.N. Nièpce, View from the window at Le Gras, Saint Loup de Varennes, France – Now at University of Texas at Austin.



# The Pioneers



J. Nicéphore Nièpce



Louis Daguerre



Henri F. Talbot



# Now...

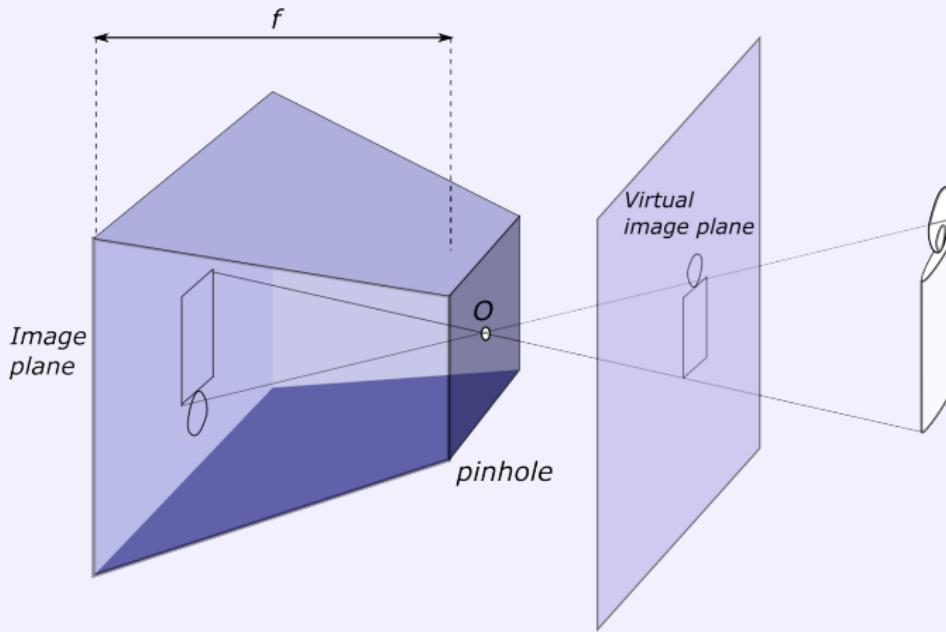


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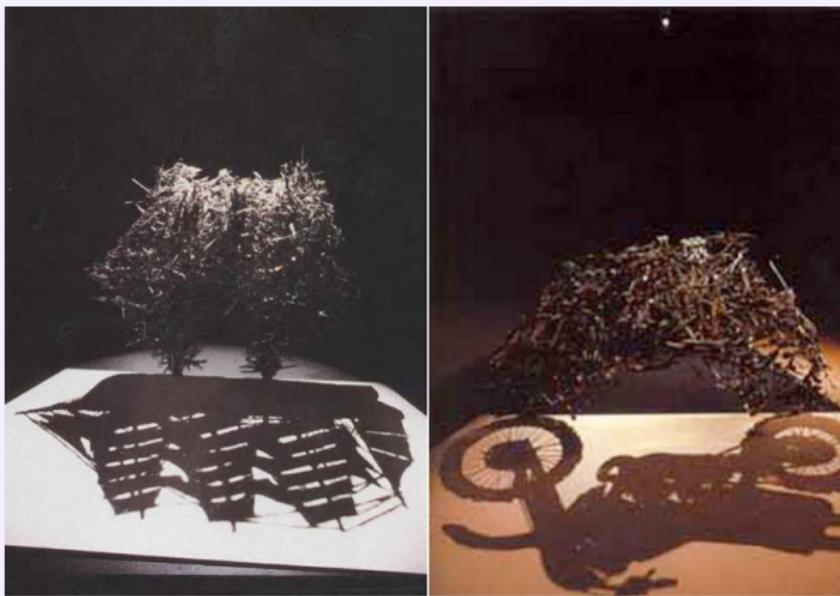
# The Pinhole Camera Model



- $f$  is the focal length,
- $O$  is the camera center.



# Projection is Tricky!

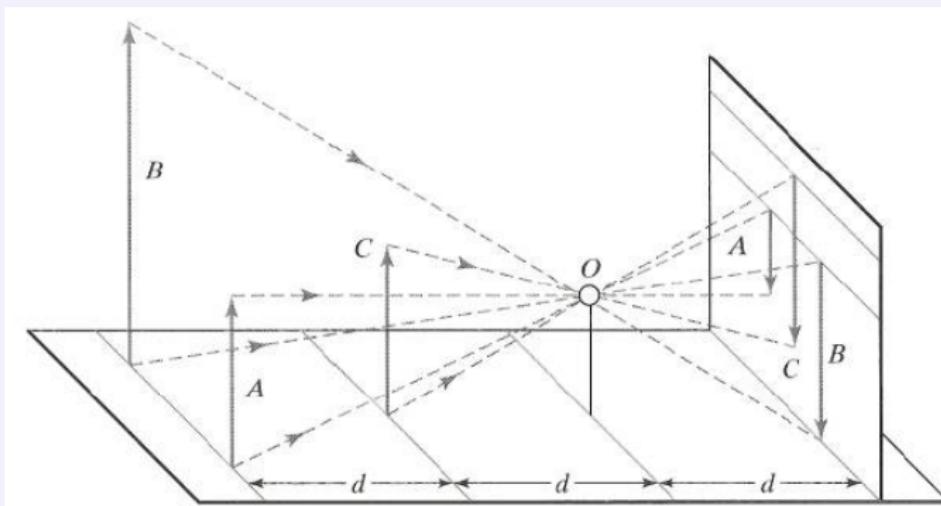


Some illusions from Shigeo Fukuda.



# Perspective Effects

Remember from Kim's first lecture

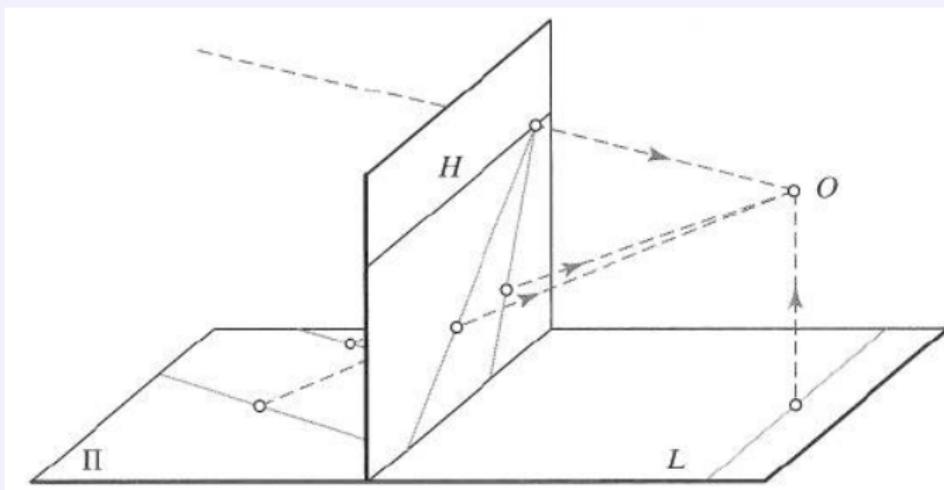


Far objects appear smaller than close ones.



# Perspective Effects

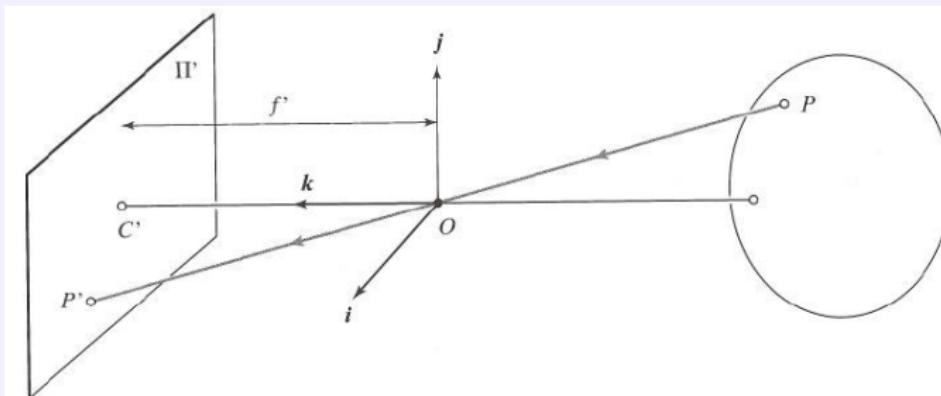
Remember from Kim's first lecture again



Images of parallel lines intersect at the horizon (virtual image plane).



# Projection Equations

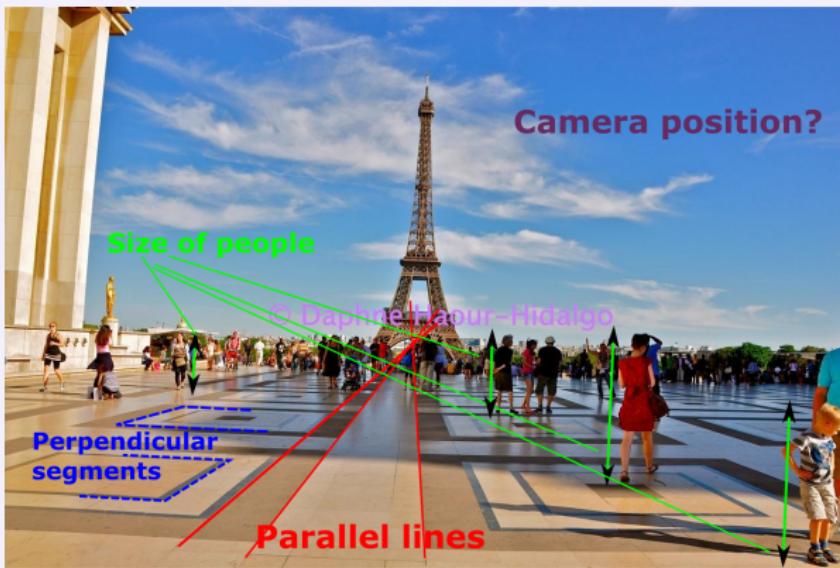


- $P(x, y, z), P'(x', y', z')$ .  $P'$  in the image plane  $\implies z' = f$
- Remember optical flow lecture: similar triangles:

$$x' = f \frac{x}{z}, \quad y' = f \frac{y}{z}$$



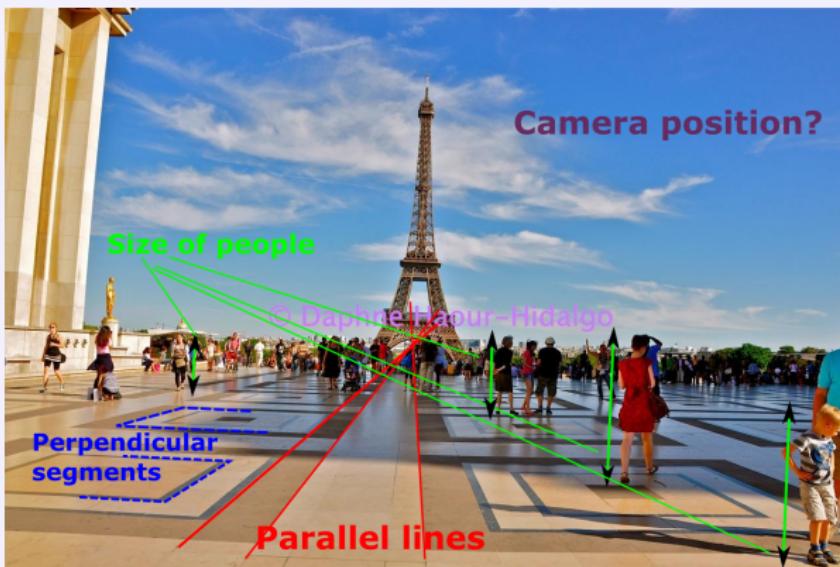
# Lost in projection



Angles, size, parallelism.



# Preserved by projection



Straight lines.



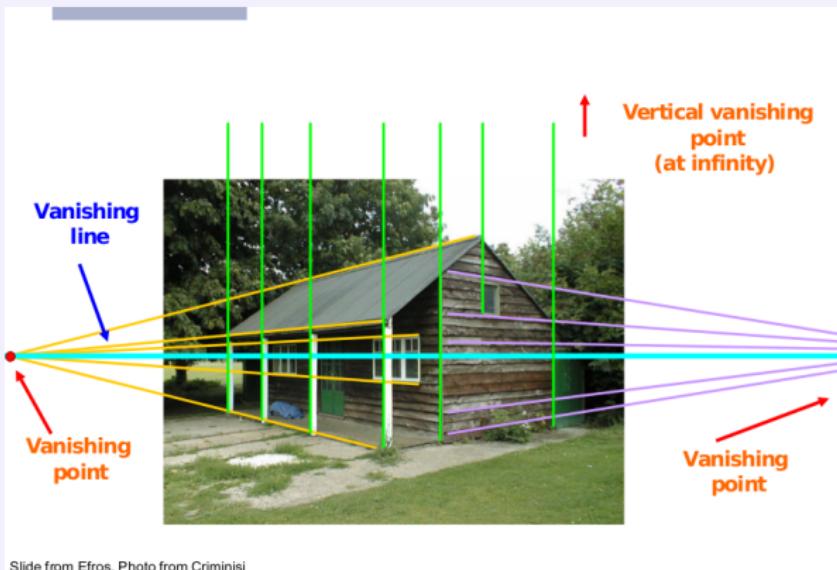
# Vanishing Points



Projections of parallel lines intersect at common points.



# Vanishing line



Vanishing points intersect at a common line. Intersection can be at infinity.



# Homogeneous coordinates 1D

- In 1D coordinate is just 1 number.



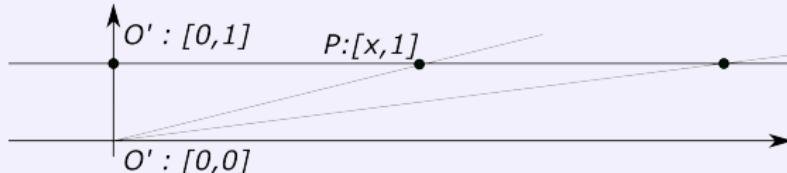
- 1D coordinate to 1D Homogeneous coordinates:

$$x \sim \begin{bmatrix} x \\ 1 \end{bmatrix}$$

- 1D homogeneous coordinate to 1 D coordinate

$$\begin{bmatrix} x \\ w \end{bmatrix} \sim x/w$$

- What can we do with that? we can “tame” infinity!
- A point with homogeneous coordinate  $[x, 0]^T$  has “normal” coordinate  $x/0 = \infty$  as if we took homogeneous coordinate  $[\infty, 1]^T$



# Homogeneous coordinates, 2D

“Natural Coordinates” for projective geometry

- From 2D point coordinate to 2D Homogeneous coordinate

$$(x, y) \Rightarrow \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

- From 2D homogeneous coordinates to 2D coordinates

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

- Exercise:  $P$  with coordinates  $(x, y, 1)$ .  $Q(x', y', w)$  point on the line through  $O(0, 0, 0)$  and  $P$ . Compute  $x'$  and  $y'$ . Considering  $[x', y', w]$  as homogeneous 2D coordinates, what are the corresponding 2D coordinates?



# Homogeneous coordinates, 3D

- From 3D point coordinate to 3D Homogeneous coordinate

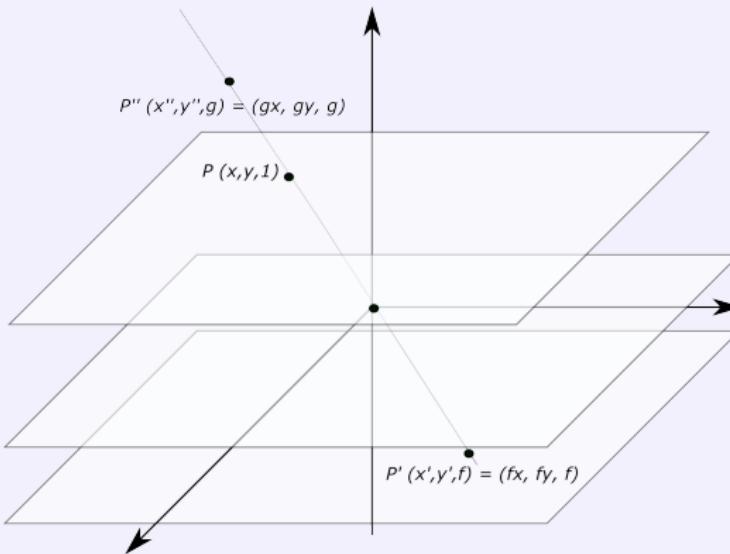
$$(x, y, z) \implies \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

- From 3D homogeneous coordinates to 3D coordinates

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



# Homogeneous coordinates



Homogeneous coordinates in 2D correspond to points in plane  $z = 1$  but also to lines through the origin and this point.



# Why Are They Useful

- Projection to image plane in standard coordinates:

$$P : (x, y, z) \mapsto P' : \left( f \frac{x}{z}, f \frac{y}{z} \right)$$

- In homogeneous coordinates:



# Why Are They Useful

- Projection to image plane in standard coordinates:

$$P : (x, y, z) \mapsto P' : (f \frac{x}{z}, f \frac{y}{z})$$

- In homogeneous coordinates:

$$P : \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \mapsto P' : \begin{bmatrix} fx \\ fy \\ z \end{bmatrix}$$

- Matrix notation

$$\begin{bmatrix} fx \\ fy \\ z \end{bmatrix} = \underbrace{\begin{pmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}}_M \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

- $M$  is the **Camera Matrix**



# World, Camera and Image Coordinates

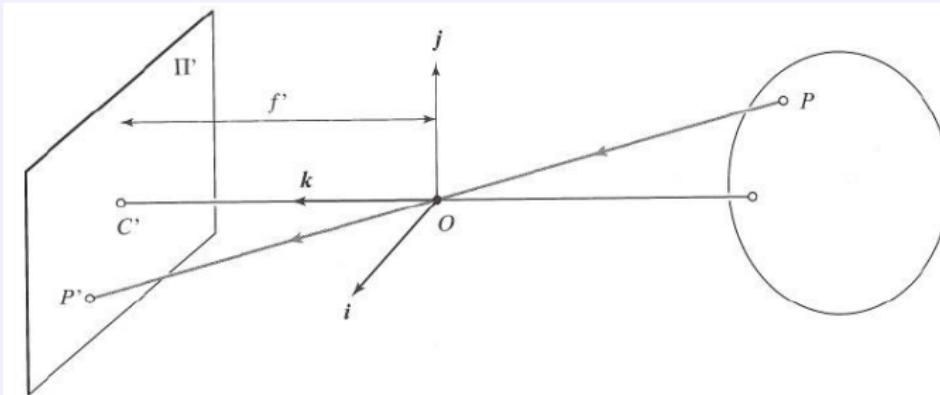
In the previous slides, Many coordinate systems are implicitly known:

- 3D World Coordinates: Coordinate system of the 3D world.
- Camera Coordinates: 3D coordinate system attached to the camera.
- Image Coordinates: 2D Coordinate system attached to the image plane.

Not that simple in practice!



# Assumptions Behind the Simple Model



- The camera center  $O$  is the same as the world coordinates origin,
- The axes  $i$ ,  $j$  and  $k$  are common for the camera and the world
- Pixels are squared and perfectly aligned with the camera coordinates.
- Image coordinates have their origin at  $C'$ , projection of the camera center  $O$ .



# Intrinsic vs Extrinsic Camera Parameters

Intrinsic parameters refer to internal parameters:

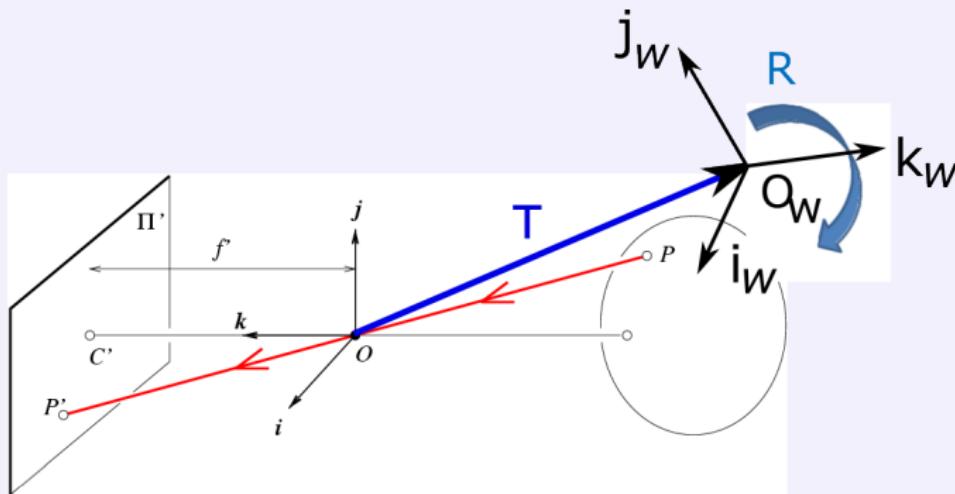
- Position of the image center wrt projection of the camera center: 2 parameters
- Scale factors for the pixels sizes in both x and y directions: 2 parameters,
- Skewness of pixels: 1 parameter.

Extrinsic Camera parameters refer

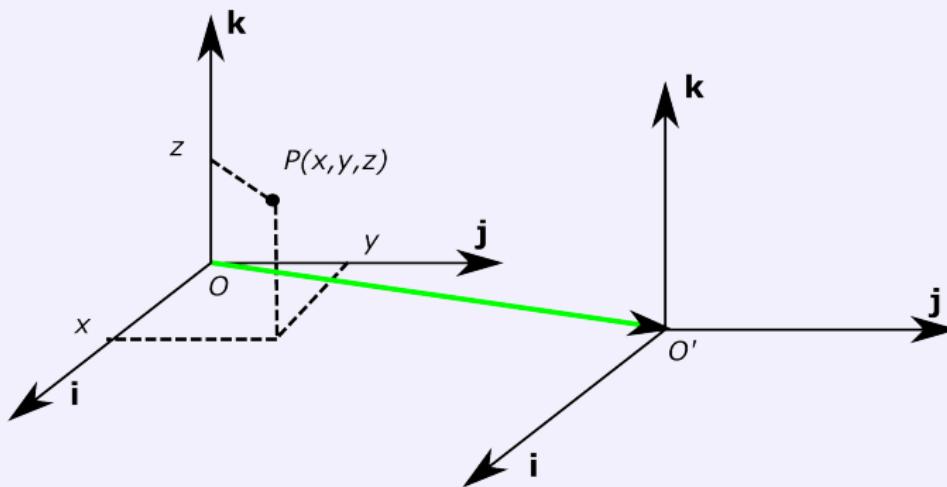
- position of the camera coordinate system vs the world coordinates system: translation,
- orientation of the camera coordinate system vs the world coordinates system: rotation.



# Oriented and Translated Camera



# Translating Coordinate System



Assume  $O'$  at coordinates  $(x_{O'}, y_{O'}, z_{O'})$  in coordinate system  $(O, \mathbf{i}, \mathbf{j}, \mathbf{k})$  and  $P$  has coordinates  $(x, y, z)$  also in  $(O, \mathbf{i}, \mathbf{j}, \mathbf{k})$ . What are its coordinates in  $(O', \mathbf{i}, \mathbf{j}, \mathbf{k})$ ?



- Let  $(x', y', z')$  its coordinates in  $(O, \mathbf{i}, \mathbf{j}, \mathbf{k})$ .
- Write  $\overrightarrow{O'P} = \overrightarrow{OP} - \overrightarrow{OO'}$
- Develop:

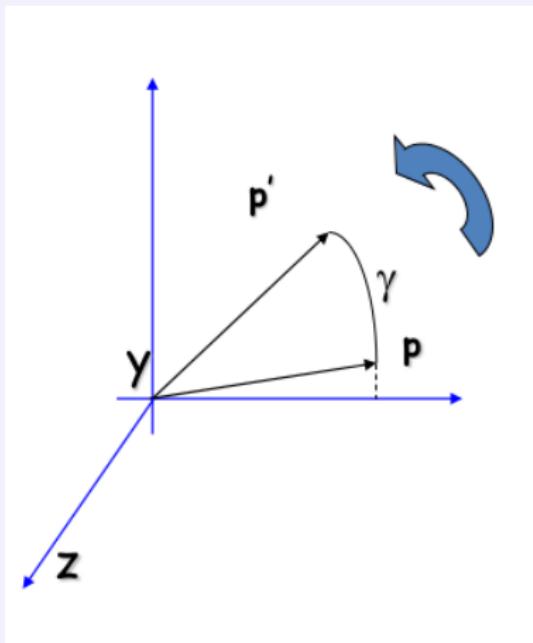
$$\begin{pmatrix} x' \\ y' \\ z' \end{pmatrix} = \begin{pmatrix} x \\ y \\ z \end{pmatrix} - \begin{pmatrix} x_{O'} \\ y_{O'} \\ z_{O'} \end{pmatrix} = \begin{pmatrix} x - x_{O'} \\ y - y_{O'} \\ z - z_{O'} \end{pmatrix}$$

- Transformation in homogeneous coordinates:

$$\begin{bmatrix} x' \\ y' \\ z' \\ 1 \end{bmatrix} = \begin{pmatrix} 1 & 0 & 0 & -x_{O'} \\ 0 & 1 & 0 & -y_{O'} \\ 0 & 0 & 1 & -z_{O'} \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$



# Rotating Coordinate System



Rotations along coordinate axes

$$R_{x\alpha} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha \end{pmatrix}$$

$$R_{y\beta} = \begin{pmatrix} \cos \beta & 0 & -\sin \beta \\ 0 & 1 & 0 \\ \sin \beta & 0 & \cos \beta \end{pmatrix}$$

$$R_{z\gamma} = \begin{pmatrix} \cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Can also be written in homogeneous coordinates



# Camera Matrix

- Combine world vs camera coordinates with
- Simple Camera matrix and
- Image plane transformation (axis scalings, shear, translation)

$$\mathbf{C} = \mathbf{K} [\mathbf{R} \; \mathbf{t}]$$

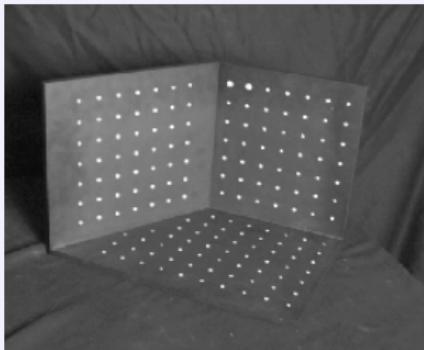
- $\mathbf{K}$   $3 \times 3$  matrix encoding the homogeneous transformations inside the camera: **calibration matrix**.
- $[\mathbf{R} \; \mathbf{t}]$  Concatenation of world coordinates rotation and origin translation to align camera and world coordinates.

$$w \begin{bmatrix} u \\ b \\ 1 \end{bmatrix} = \underbrace{\begin{pmatrix} \alpha & s & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{pmatrix}}_{\mathbf{K}} \underbrace{\begin{pmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \end{pmatrix}}_{[\mathbf{R} \; \mathbf{t}]} \begin{pmatrix} x \\ y \\ z \\ 1 \end{pmatrix}$$



# Geometric Calibration

- Computing the camera matrix is called geometric calibration.
- Extrinsic parameters: usually easy.
- Calibration focuses more on intrinsic parameters (calibration matrix  $\mathbf{K}$ ).
- Often difficult.



- Use an object with known geometry
- Use vanishing points / lines
- Use other cues...





Without knowledge of object geometry, calibration can be very problematic (Ames Room illusion).



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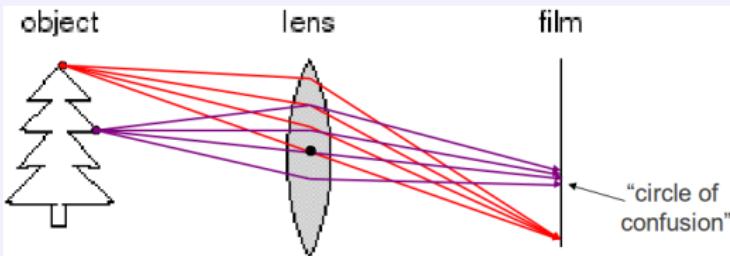
# Shrinking the aperture



Less light in, diffraction.



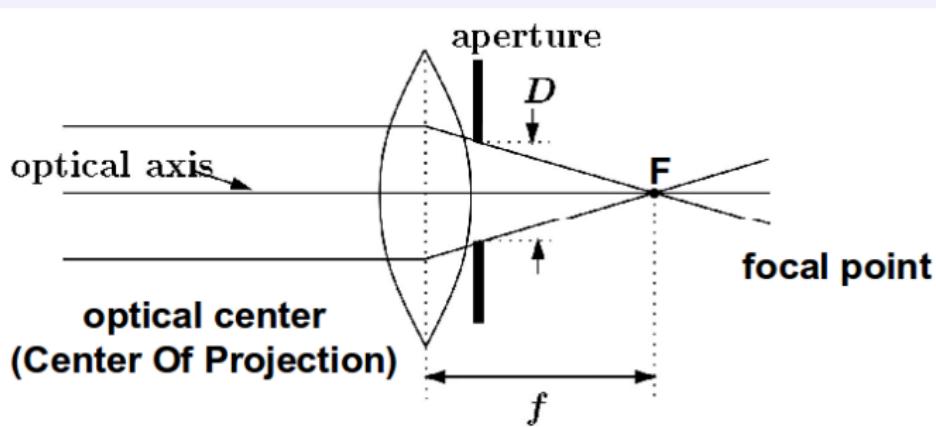
# Adding Lens



- Specific distance for which objects are in focus
- Changing shape of lens changes the focus distance.



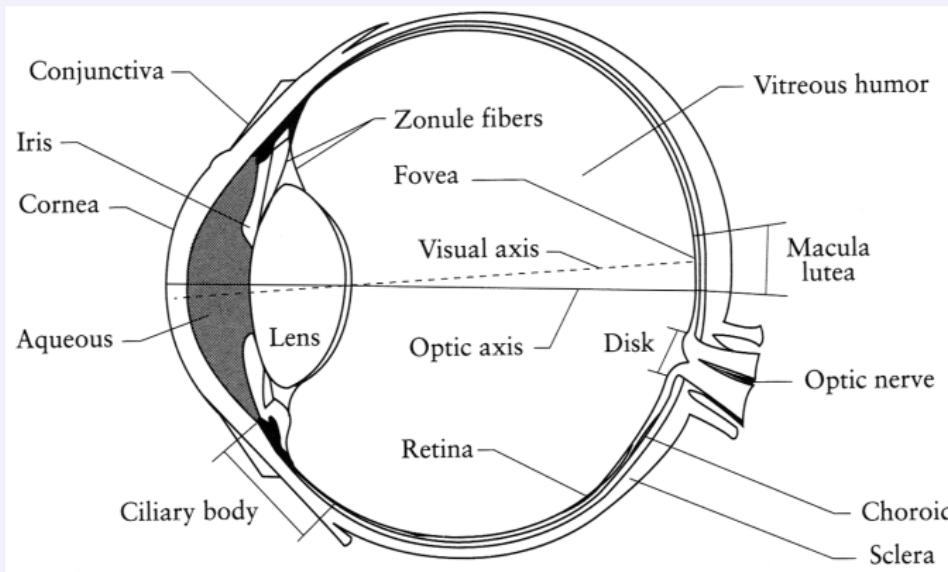
# Focal Length, Aperture, Depth of Field



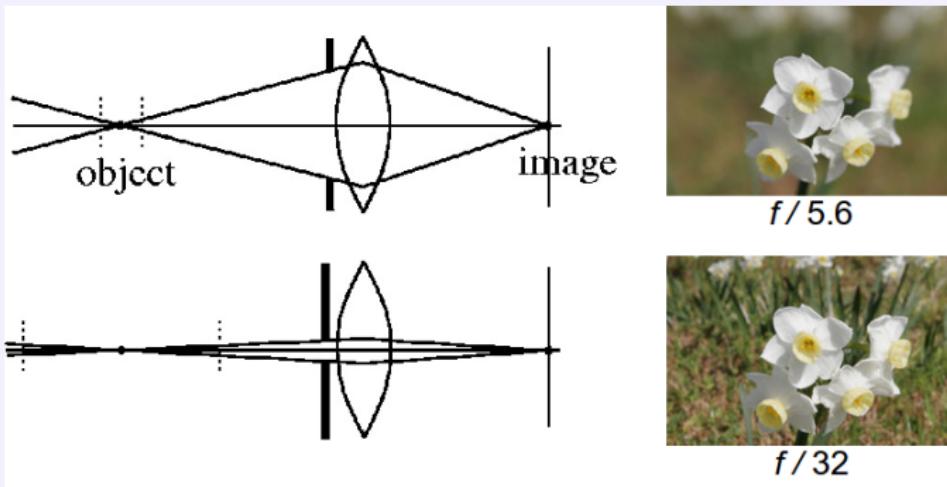
- Lens focuses parallel rays into a single point.
- Aperture restricts range of rays.



# The Eye is a Camera with Lens



# Depth of Field



Controlled by aperture size and focal length.

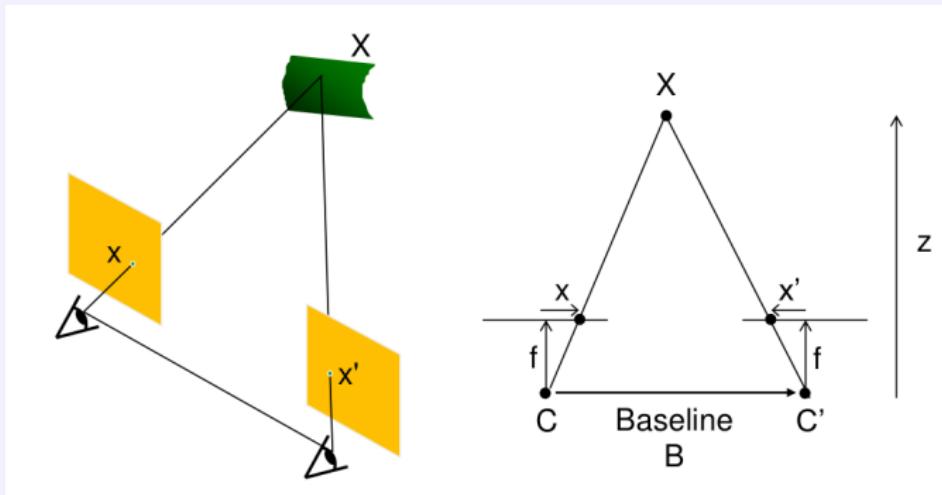


# Outline

- ① Vectors and Matrices
- ② Introduction
- ③ The Pinhole Camera
- ④ A bit of History
- ⑤ Projection
- ⑥ More on Camera Models
- ⑦ Stereo



# Multiple View Correspondences



If we can recover  $x'$  from  $x$  we can recover depth.



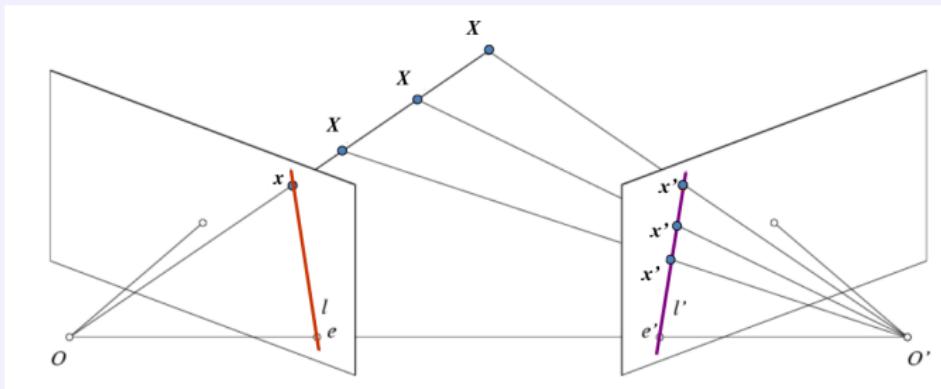
# Image Correspondence



How do we match points from image 1 to image 2: you should have seen some of it with Kim, but this is not the end of the story!



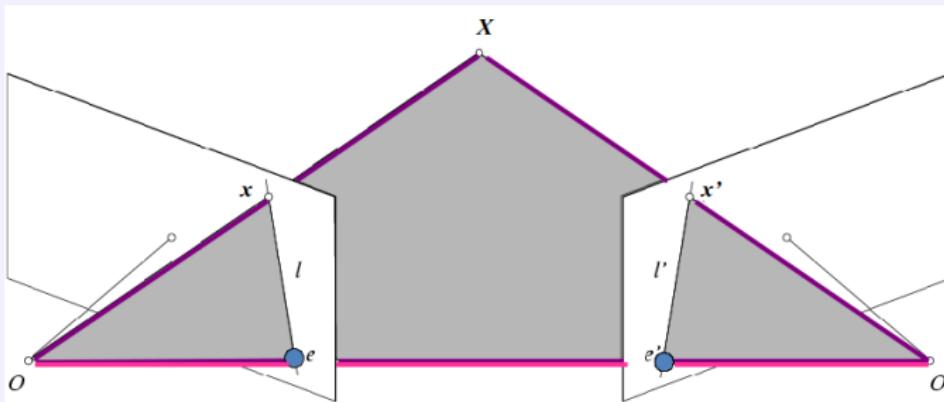
# Epipolar Constraints



- Potential match for  $x$  must lie in corresponding line  $l'$
- Potential match for  $x'$  must lie in corresponding line  $l$



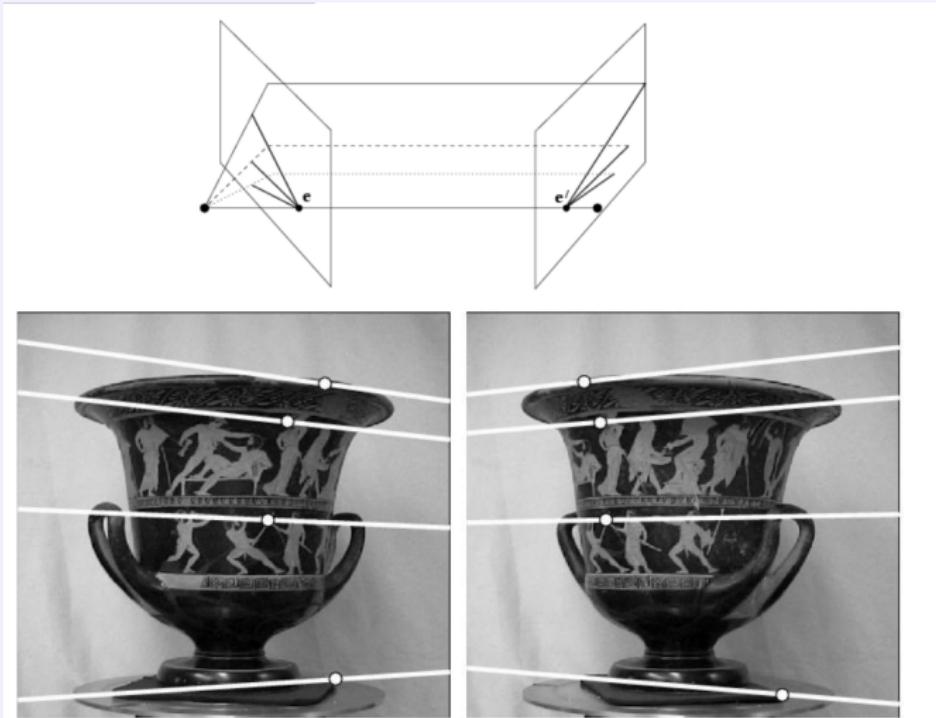
# Epipolar Constraints



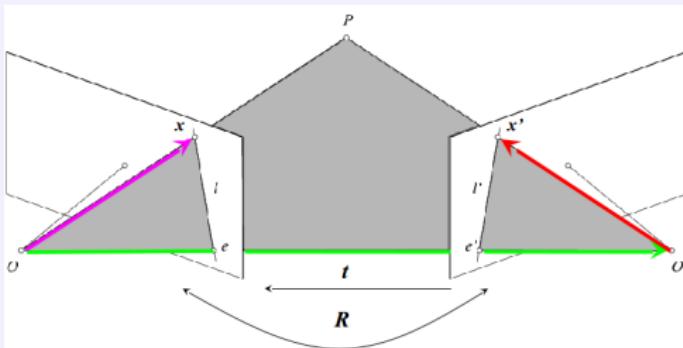
- Line connecting  $O$  and  $O'$ : **baseline**
- Plane through baseline  $x$  and  $x'$ : **Epipolar Plane**
- Epipoles: intersection of baseline and image planes: projection of the other camera center.
- Epipolar Lines - intersections of epipolar plane with image planes (always come in corresponding pairs)



# Example: Converging cameras



# Calibrated Case



Camera parameters known for the two cameras: calibration matrices  $K$  and  $K'$



- $x$  and  $x'$  (in 3D coords but not the same) are related by rotation and translation.

$$x' = R(x - \mathbf{t})$$

- Their homogeneous coordinates  $y$  and  $y'$  are related by a simple matrix  $E$  built from  $R$  and  $t$

$$y^T E y = 0.$$

- $E$  is called the **essential matrix** (Longuet-Higgins 1981).
- $E$  can be estimated from images.
- The position and orientation of camera 1 vs camera 2 (i.e.,  $R$  and  $\mathbf{t}$ ) can be recovered from  $E$ .



# Note on the geometric transformation between image 1 and image2

- The transformation between  $y$  and  $y'$  is not linear: it is an **homography** between the 2 image planes:
- An homography conserves straight lines, but not parallelism.
- Two parallel lines intersect at infinity: after an homography they may intersect at finite distance.



# Uncalibrated Case

- No way to directly compare  $x$  and  $x'$ : they relate through the unknown transformations  $K$  and  $K'$ .
- A relation however still exists (Faugeras, Luong, 1992)

$$y^T \underbrace{K^{-\top} E K'^{-1}}_F y = 0$$

- $F$  is called the **fundamental matrix**.
- Though  $K$  and  $K'$  are unknown  $F$  can be **estimated** (complicated) thus used for the correspondence problem!
- $K^\top$  is the transposed of  $K$ : inverse row and line indices. For a vector in column, write it in line.
- $K^{-1}$  is the inverse of  $K$ : provides the inverse change of coordinates from unnormalized to normalized image coordinate.
- $K^{-\top}$  means inverse of the transposed matrix: The same as transposed of the inverse.



