

**MultiDrone**



# **Image acquisition, camera geometry**

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## Image acquisition

- A still image visualizes a still object or scene, using a still picture camera.
- A video sequence (moving image) is the visualization of an object or scene illuminated by a light source, using a video camera.
- The captured object, the light source and the video camera can all be either moving or still.
- Thus, moving images are the projection of moving 3D objects on the camera image plane, as a function of time.
- Digital video corresponds to their spatiotemporal sampling.



## Light reflection

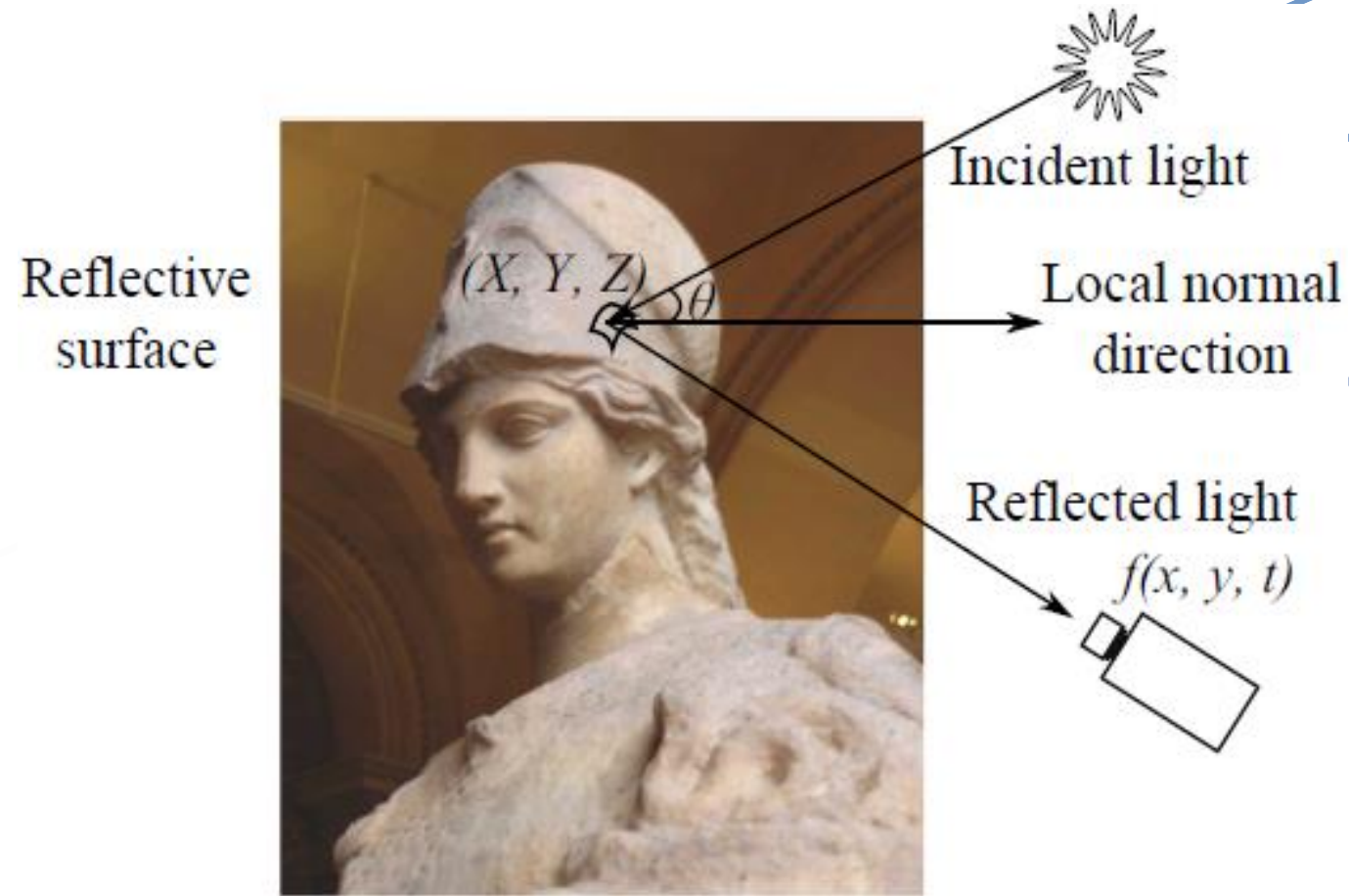
- Objects reflect or emit light.
- Reflection can be decomposed in two components:
  - *Diffuse reflection* (distributes light energy equally along any spatial direction, allows perceiving object color).
  - *Specular reflection* (strongest along the direction of the incident light, incident light color is perceived).
- *Lambertian surfaces* perform only diffuse reflection, thus being dull and matte (e.g., cement surface).



## Light reflection

- *Ambient illumination* sources emit the same light energy in all directions (e.g., a cloudy sky).
- *Point illumination* sources emit light energy isotropically or anisotropically (e.g., ordinary light bulbs) along various directions.

# Light reflection





# Light reflection

- Reflected irradiance when object surface produces diffuse reflectance and incident light source comes from:

- Ambient illumination:

$$f_r(X, Y, Z, t, \lambda) = r(X, Y, Z, t, \lambda) \cdot E_a(t, \lambda)$$

- Point light source:

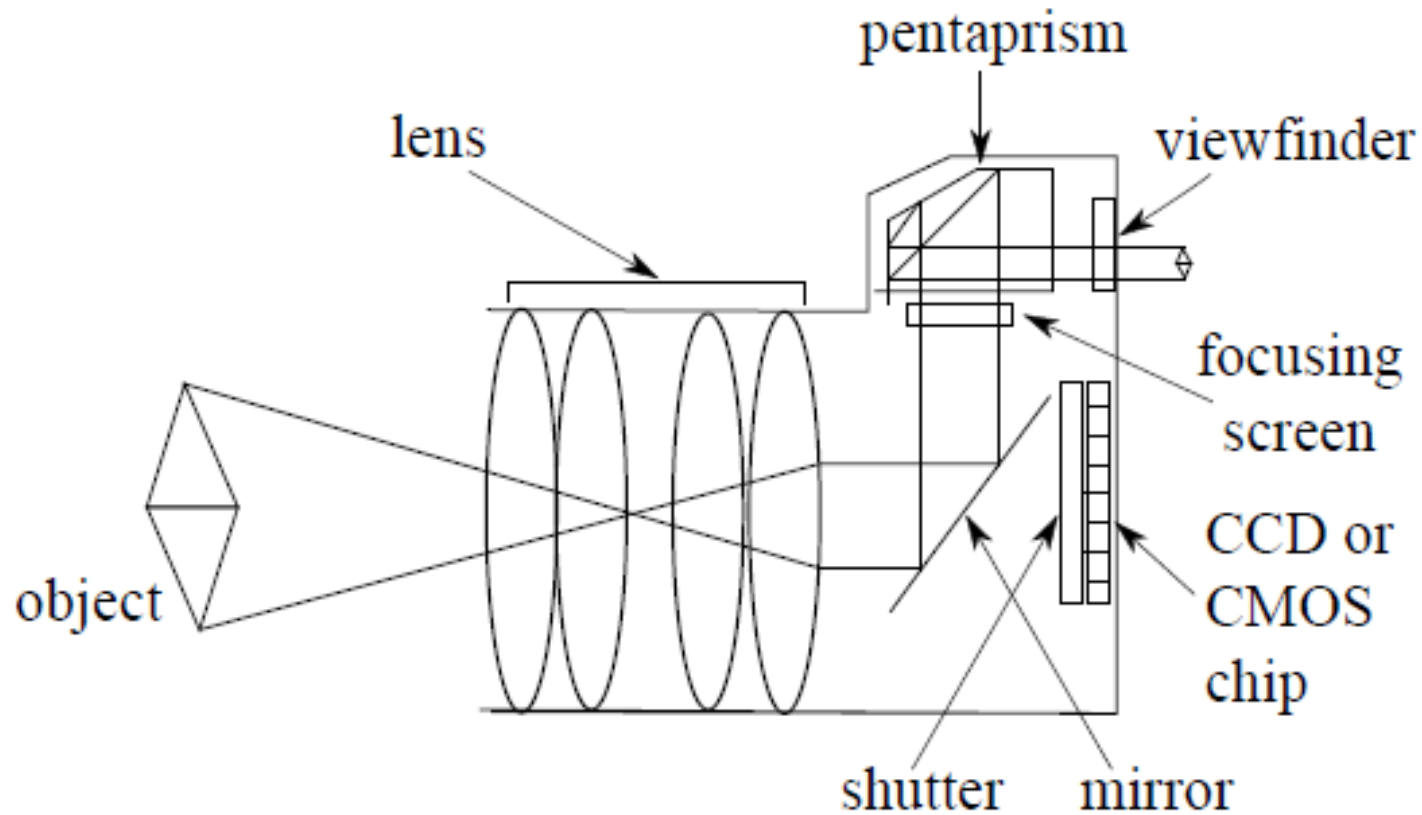
$$f_r(X, Y, Z, t, \lambda) = r(X, Y, Z, t, \lambda) \cdot E_p(t, \lambda) \cdot \cos \theta$$

- Distant point source and ambient illumination:

$$E(t, \lambda) = E_a(t, \lambda) + E_p(t, \lambda) \cdot \cos \theta$$

# Camera structure

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# Camera structure

- The *lens* is the most important part of the camera.
- Incident light rays pass through a lens (or a group of lenses) and get focused on the semiconductor chip.
- The distance between the lens center (*optical center*, **O**) and the point of convergence of the light rays inside the camera (*focal point*, **F**) is called *focal length*.
- Focal length characterizes the lens and determines the scene part to be captured as well as scene object sizes (*magnification*).





## Camera structure

- Two kinds of lenses:
  - Fixed (e.g., *prime*) and
  - Zoom (e.g., *telephoto*)
- Based on their focal length, lenses are categorized in wide-angle, normal and telephoto:
  - Wide-angle lenses have smaller focal length than normal, thus capturing wider parts of the scene and exaggerating differences in the relative distance and size between foreground and background objects.



# Camera structure

- The *shutter* opens and closes to control the time interval during which light rays can hit the CCD or CMOS chip.
- *Shutter speed* is the speed at which the shutter opens and closes and determines the amount of incoming light.
- Higher speed is required for capturing unblurred, fast moving scenes, while lower speed is used in night shooting, along with bigger *aperture* size.



# Camera structure

- Aperture size is usually expressed in *f-numbers*. The bigger the f-number the smaller the aperture size.
- It controls the *depth of field* (DOF), the distance between the nearest and farthest focused objects in the image
- The smaller the aperture size is, the longer the depth of field, since less light rays are captured on the image for each visible 3D scene point.

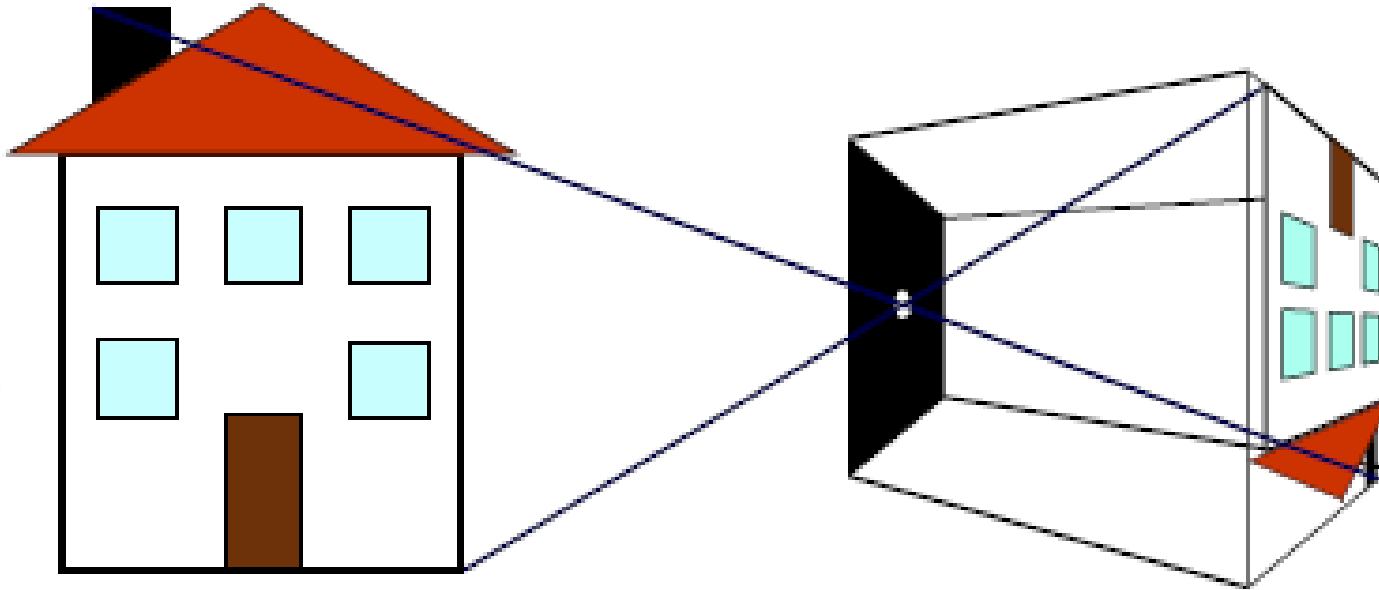
# Pinhole Camera and Perspective Projection



- The rather naïve *pinhole camera system* can be used to accurately model the geometric and optical aspects of most modern cameras through the *pinhole perspective projection model* or *central perspective projection model*.
  - A very small aperture size is considered
  - Camera pinhole coincides with the *optical center*, or *center of projection* or *camera center*.



# Pinhole Camera and Perspective Projection



# Pinhole Camera and Perspective Projection

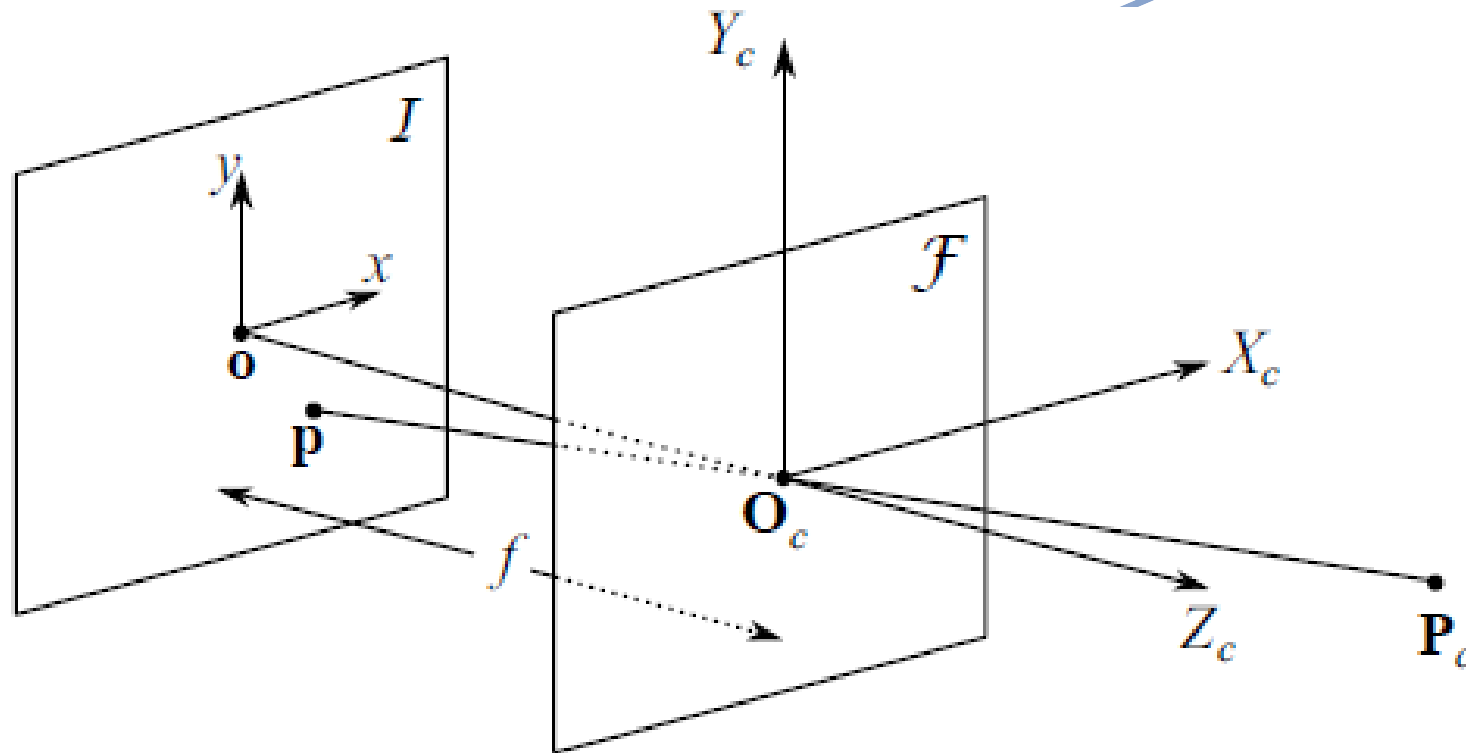


- Let us consider 2 coordinate systems:
  - the *camera (or standard) coordinate system* ( $\mathbf{O}_c, X_c, Y_c, Z_c$ ) and
  - the image coordinate system ( $\mathbf{o}, x, y$ ).
- $X_c$ , and  $Y_c$  define the plane  $\mathcal{F}$  that is parallel to the camera image plane  $\mathcal{I}$ , lying at a focal length  $f$  behind the optical center  $\mathbf{O}_c$  along the optical axis  $Z_c$ .



# Pinhole Camera and Perspective Projection

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# Pinhole Camera and Perspective Projection

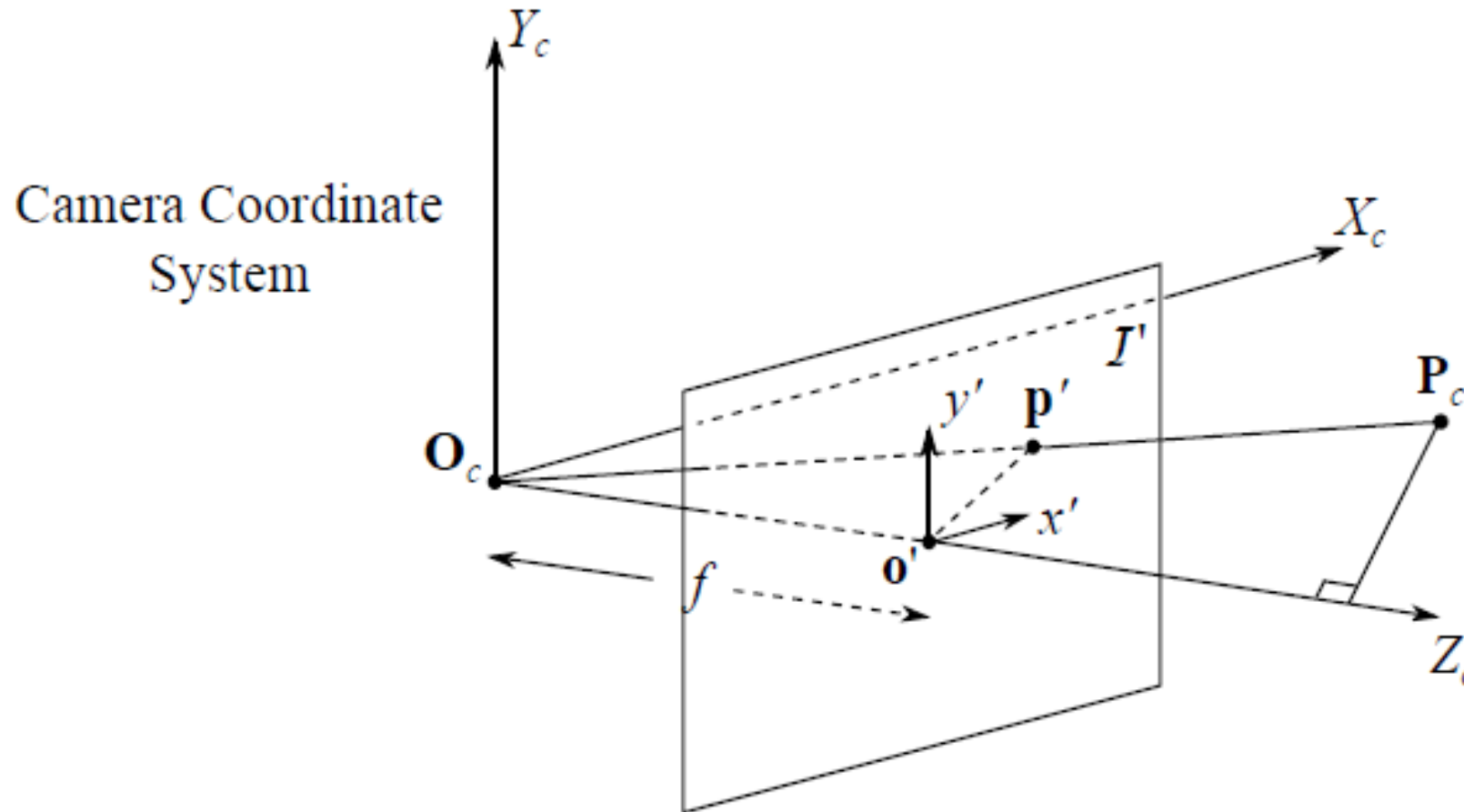


- Points projected on the image plane are assigned camera coordinates of opposite sign:
  - images are inverted.
- In order to facilitate the mathematical treatment, we can define a virtual image plane  $\mathcal{I}'$ , in front of  $\mathcal{F}$  at a positive distance  $f$ .



# Pinhole Camera and Perspective Projection

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# Pinhole Camera and Perspective Projection



- We want to derive the equations that connect a 3D point (3D vector)  $\mathbf{P}_c = [X_c, Y_c, Z_c]^T$  referenced in the camera coordinate system with its projection point (2D vector)  $\mathbf{p}' = [x', y']^T$  on the virtual image plane.
- By employing the similarity of triangles  $\mathbf{O}_c \mathbf{o}' \mathbf{p}'$  and  $\mathbf{O}_c \mathbf{Z}_c \mathbf{P}_c$ :

$$\frac{x'}{X_c} = \frac{y'}{Y_c} = \frac{f}{Z_c}, \quad x' = f \frac{X_c}{Z_c}, \quad y' = f \frac{Y_c}{Z_c}$$

- Coordinates on the real image plane are given by the same equations, differing only by a minus sign.



# The Weak-Perspective Camera Model

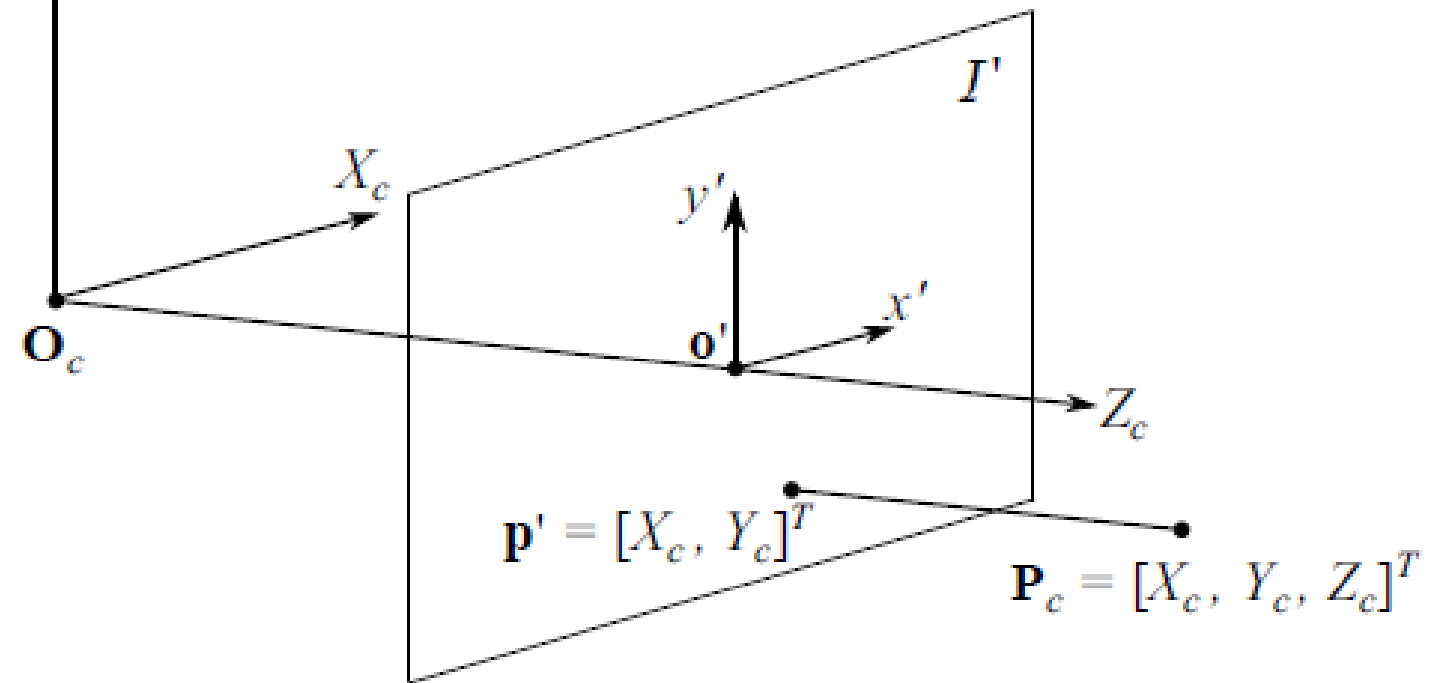


- Perspective projection equations are rational, rather than linear. Thus:
  - straight lines are mapped to straight lines but
  - distances between points and the angles between straight lines are not preserved after projection.
- We can linearise them applying two transformations:
  - *orthographic projection*  $x' = X_c$  and  $y' = Y_c$  and
  - *isotropic scaling*  $f / \bar{Z}$



# The Weak-Perspective Camera Model

Camera Coordinate System



Orthographic projection

# The Weak-Perspective Camera Model



- Isotropic scaling transformation leads to linearly approximate perspective projection equations, defining the so called *weak-perspective* camera model:

$$x' = f \frac{X_c}{Z_c} \approx \frac{f}{\bar{Z}} X_c, \quad y' = f \frac{Y_c}{Z_c} \approx \frac{f}{\bar{Z}} Y_c$$

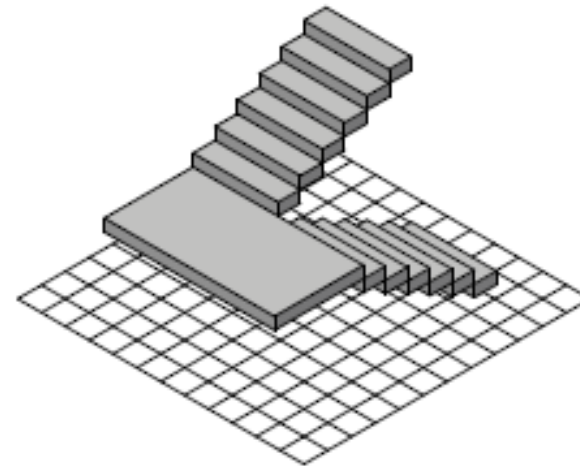
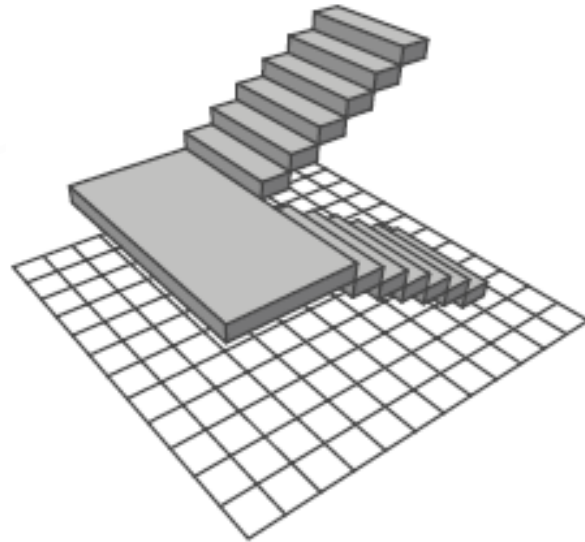
- The weak-perspective camera model is only an approximation of the pinhole camera imaging. It holds if the *relative distance*  $dZ_c$  for any pair of scene points along the optical axis is much smaller than  $\bar{Z}$ .



# The Weak-Perspective Camera Model



- While a weak-perspective camera preserves parallelism in the projected lines, as orthographic projection does (b), perspective projection (a) does not.





# Central Projection and Homogeneous Coordinates



- A way to linearize the perspective projection equations is by using the so-called *homogeneous coordinates*.
  - A 2D image point is mapped to a 3D point:  
 $\mathbf{p} = [x, y]^T \in \mathbb{R}^2 \rightarrow \mathbf{p}_H = [x, y, 1]^T \in \mathbb{P}^2.$
  - A 3D scene point is mapped to a 4D point:  
 $\mathbf{P} = [X, Y, Z]^T \in \mathbb{R}^3 \rightarrow \mathbf{P}_H = [X, Y, Z, 1]^T \in \mathbb{P}^3.$



# Central Projection and Homogeneous Coordinates



- The linear relationship that connects  $\mathbf{p}_H$  and  $\mathbf{P}_H$  is:

$$Z\mathbf{p}_H = \begin{bmatrix} Zx \\ Zy \\ Z \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} = \mathcal{P}\mathbf{P}_H$$

where  $\mathcal{P}$  is the so-called camera *perspective projection matrix*, a  $3 \times 4$  full row rank homogeneous matrix with 11 *degrees of freedom*.

# Camera Parameters and Projection Matrix

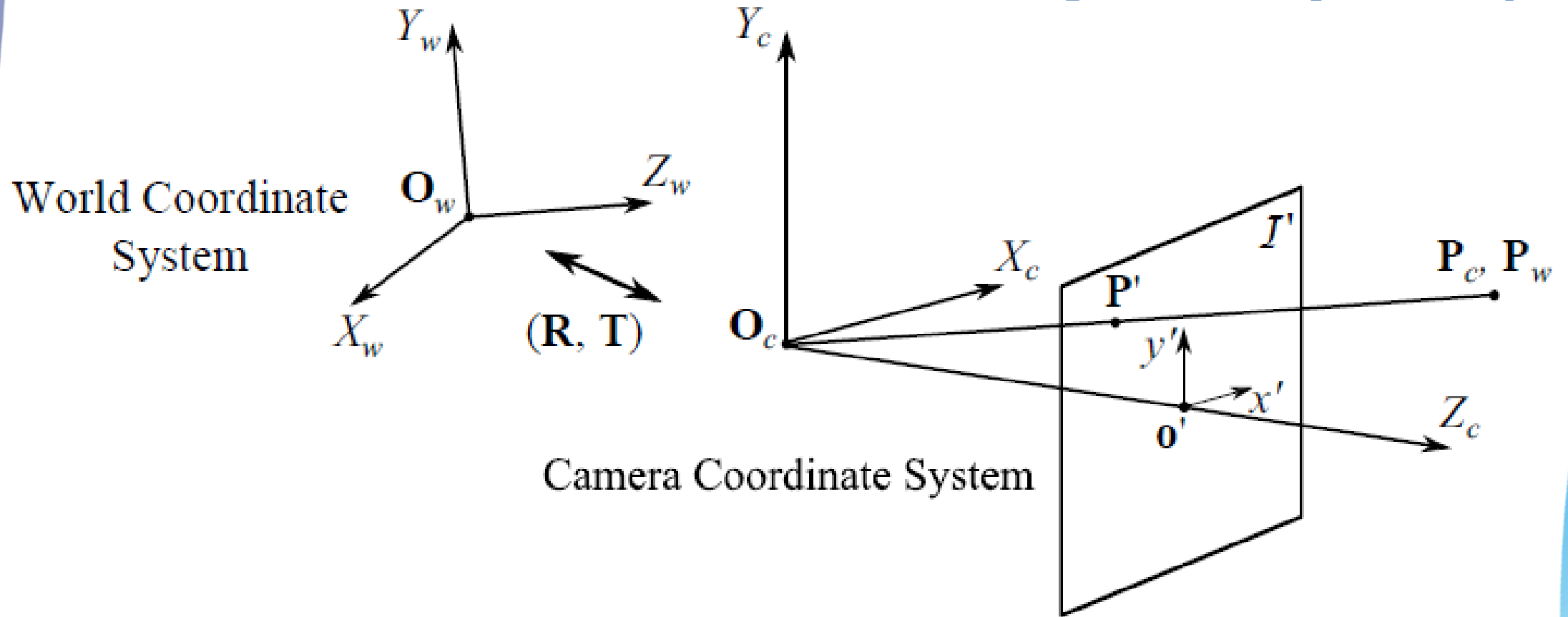


- Object coordinates  $\mathbf{P}$  in the camera coordinate system are, in most cases, unknown, whereas in the *world coordinate system* they may be known.
- The required transformation from the world to the camera coordinate system involves a translation followed by a rotation, based on the *extrinsic* camera parameters.
- Projection on the image plane requires the *intrinsic* camera parameters.



# Camera Parameters and Projection Matrix

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# Camera Parameters and Projection Matrix



- Extrinsic camera parameters:
  - *Translation vector*  $\mathbf{T} \in \mathbb{R}^3$  (3 degrees of freedom)
  - *Orthonormal rotation matrix*  $\mathbf{R} \in \mathbb{R}^{3 \times 3}$  (3 degrees of freedom: only 3 of the 9 rotation matrix entries are independent from each other).
- The relationship between a point  $\mathbf{P}_w \in \mathbb{R}^3$  in world coordinates and its camera coordinate counterpart  $\mathbf{P}_c \in \mathbb{R}^3$  is:

$$\mathbf{P}_c = \mathbf{R}(\mathbf{P}_w - \mathbf{T})$$



# Camera Parameters and Projection Matrix



- If the three rows of the rotation matrix and focal length  $f$  are known, the image coordinates  $x'$ ,  $y'$  on the virtual image plane are given by:

$$x' = f \frac{R_1^T (P_w - T)}{R_3^T (P_w - T)}$$

$$y' = f \frac{R_2^T (P_w - T)}{R_3^T (P_w - T)}$$



# Camera Parameters and Projection Matrix



- In reality, the virtual image plane  $\mathcal{I}'$  does not exist. The real 2D image plane (image sensor surface) is digitized.
- The transformation of an image point  $\mathbf{p} = [x, y]^T$  on the image plane coordinates to the corresponding discrete point  $\mathbf{p}_d = [x_d, y_d]^T$  in pixel coordinates, is given by:

$$x = -(x_d - o_x)s_x, \quad y = -(y_d - o_y)s_y$$



# Camera Parameters and Projection Matrix



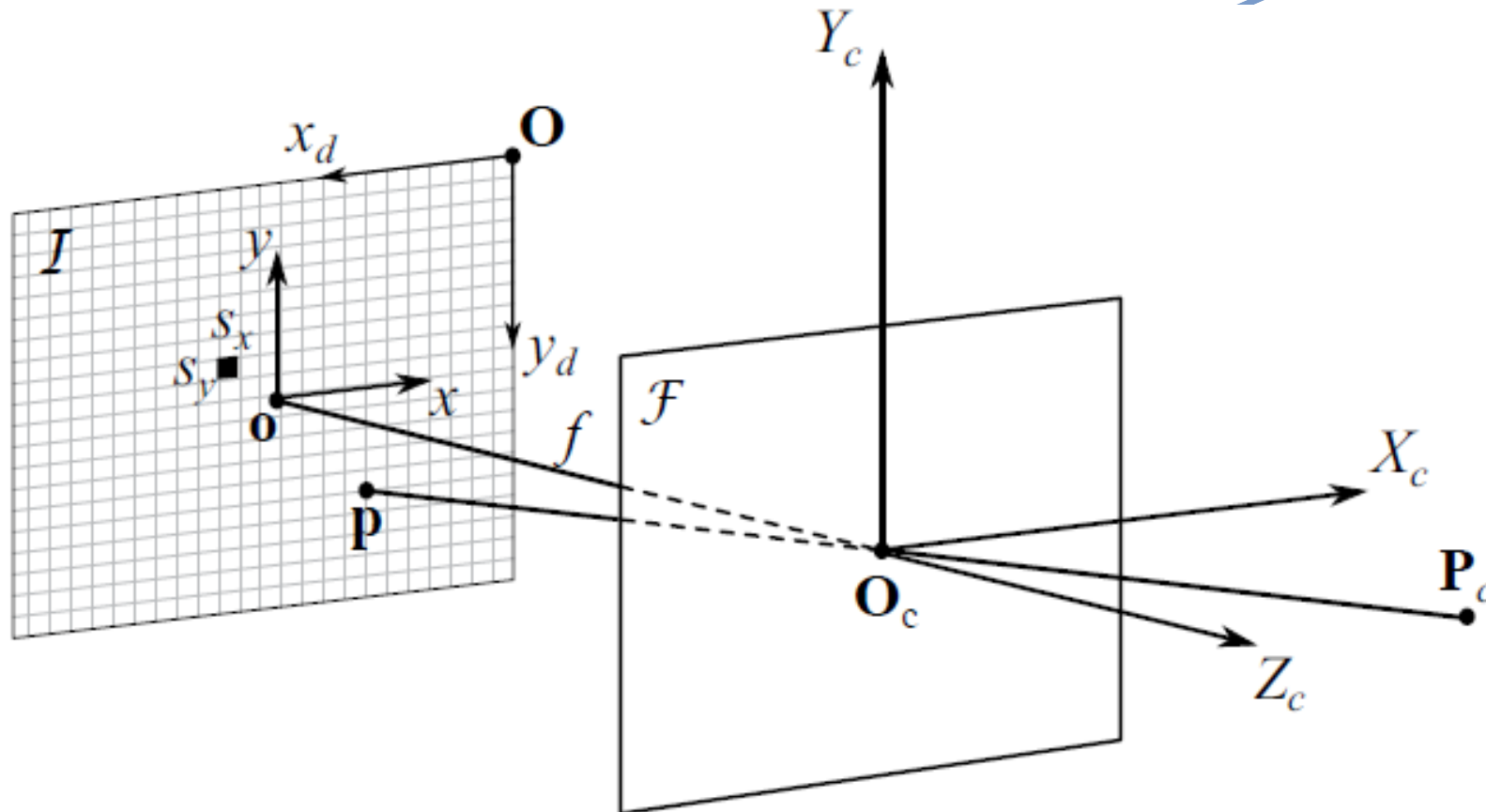
where:

- $[o_x, o_y]^T$ : the location of the principal camera point **o** in pixel coordinates.
- $s_x, s_y$ : the effective pixel sizes in millimeters
- Coordinate system origin: at the top left corner of the image, not at the image center.



# Camera Parameters and Projection Matrix

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# Camera Parameters and Projection Matrix



- The transformation relating the image pixel coordinates with the world coordinates is:

$$x_d = o_x - \frac{f}{s_x} \frac{R_1^T (P_w - T)}{R_3^T (P_w - T)} \quad y_d = o_y - \frac{f}{s_y} \frac{R_2^T (P_w - T)}{R_3^T (P_w - T)}$$

- It can be linearized in homogeneous coordinates, by decomposing the transformation into a sequence of two transformations:
  - Map a world coordinate point to camera coordinates.
  - Map the came coordinate point to homogeneous image pixel coordinates.

# Camera Parameters and Projection Matrix



- Definition of the  $3 \times 4$  matrix of extrinsic parameters  $\mathbf{P}_E$ :

$$\mathbf{P}_E = \begin{bmatrix} r_{11} & r_{12} & r_{13} & -\mathbf{R}_1^T \mathbf{T} \\ r_{21} & r_{22} & r_{23} & -\mathbf{R}_2^T \mathbf{T} \\ r_{31} & r_{32} & r_{33} & -\mathbf{R}_3^T \mathbf{T} \end{bmatrix}$$

- Definition of the  $3 \times 3$  matrix of intrinsic parameters  $\mathbf{P}_I$ :

$$\mathbf{P}_I = \begin{bmatrix} -\frac{f}{s_x} & 0 & o_x \\ 0 & -\frac{f}{s_y} & o_y \\ 0 & 0 & 1 \end{bmatrix}$$



# Camera Parameters and Projection Matrix



- The transformation of a point  $\mathbf{P} \in \mathbb{P}^3$  to  $\mathbf{p} \in \mathbb{P}^2$  is given by:

$$\begin{bmatrix} Zx_d \\ Zy_d \\ Z \end{bmatrix} = \mathbf{P}_I \mathbf{P}_E \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix} \quad \mathbf{p} = \mathbf{P}_I \mathbf{P}_E \mathbf{P} = \mathcal{P} \mathbf{P}$$

# Camera Parameters and Projection Matrix



- Where  $\mathcal{P} = \mathbf{P}_I \mathbf{P}_E$  is the  $3 \times 4$  camera projection matrix, also called camera calibration matrix

$$\mathcal{P} = \begin{bmatrix} -\frac{f}{s_x} & 0 & o_x \\ 0 & -\frac{f}{s_y} & o_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & -\mathbf{R}_1^T \mathbf{T} \\ r_{21} & r_{22} & r_{23} & -\mathbf{R}_2^T \mathbf{T} \\ r_{31} & r_{32} & r_{33} & -\mathbf{R}_3^T \mathbf{T} \end{bmatrix}$$

- $\mathbf{P}_E$  has the form  $\mathbf{P}_E = [\mathbf{R}_1 | \mathbf{R}_2 | \mathbf{R}_3 | -\mathbf{R}\mathbf{T}]$ , since we assume that the camera coordinate system is first translated and then rotated. Otherwise it would be  $\mathbf{P}_E = [\mathbf{R}_1 | \mathbf{R}_2 | \mathbf{R}_3 | \mathbf{T}]$ .

# Camera Parameters and Projection Matrix



- For reasons of simplicity, it is common to assume that:
  - the origins of both the pixel coordinate system and the image plane coordinate system coincide with the principal point,  $o_x = o_y = 0$  and
  - pixels are square having unit edge length  $s_x = s_y = 1$ .
- The projection matrix, can thus be rewritten as:

$$\mathcal{P} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fr_{13} & f\mathbf{R}_1^T \mathbf{T} \\ -fr_{21} & -fr_{22} & -fr_{23} & f\mathbf{R}_2^T \mathbf{T} \\ r_{31} & r_{32} & r_{33} & -\mathbf{R}_3^T \mathbf{T} \end{bmatrix}$$





# Camera Parameters and Projection Matrix



- If the two axes of the coordinate system  $(x_d, y_d)$  are not exactly perpendicular (non-rectangular pixels), the projection matrix takes the form:

$$\mathcal{P} = \begin{bmatrix} -\frac{f}{s_x} & s_\theta & o_x \\ 0 & -\frac{f}{s_y} & o_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{R} & -\mathbf{RT} \\ \mathbf{0} & 1 \end{bmatrix}$$

- $s_\theta$ : *skew factor*, proportional to  $\frac{1}{\tan \theta}$ ,
- $\theta$ : the angle between the pixel coordinate system axes.
- Typically  $\theta = 90^\circ$ , and hence  $s_\theta = 0$ .

# Camera Parameters and Projection Matrix



- Assuming  $s_\theta = 0$  and treating the ratios  $a_x = -\frac{f}{s_x}$  and  $a_y = -\frac{f}{s_y}$  as single quantities, by expressing the focal length in terms of pixel dimensions along the horizontal and vertical dimension,  $\mathbf{P}_I$  can be rewritten as:

$$\mathbf{P}_I = \begin{bmatrix} a_x & 0 & o_x \\ 0 & a_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$



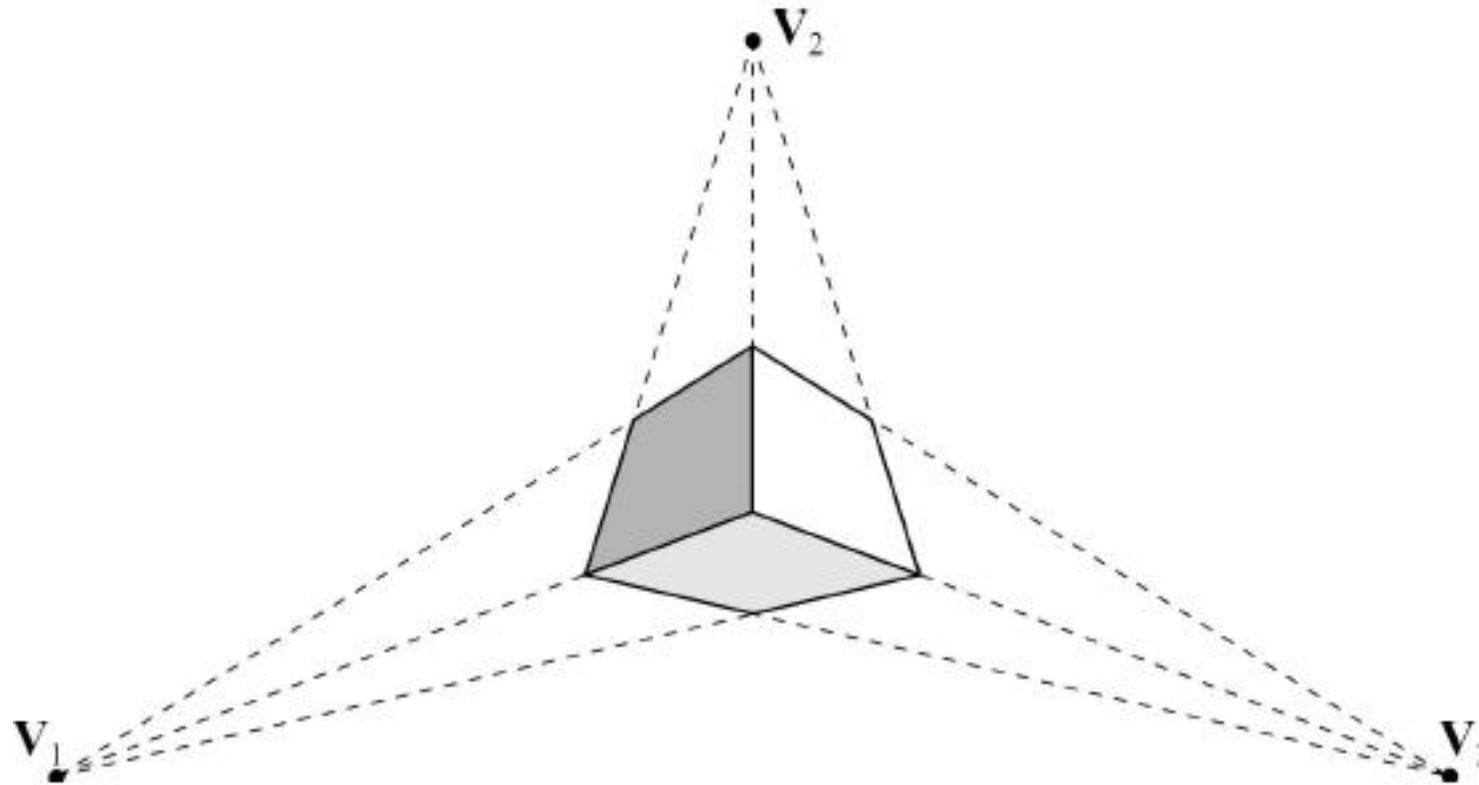
# Properties of the Projective Transformation



- The projective transformation does not change the cross-ratio  $C_r$  of four collinear points.
- *Line at infinity*: the set of all points at infinity in  $\mathbb{P}^2$ .
- *Plane at infinity*: the set of all points at infinity in  $\mathbb{P}^3$ , formed by an infinite number of lines at infinity corresponding to different plane directions.
- *Vanishing points*: the points of intersection of the projected lines formed by parallel lines in the 3D Euclidean space.

# Properties of the Projective Transformation

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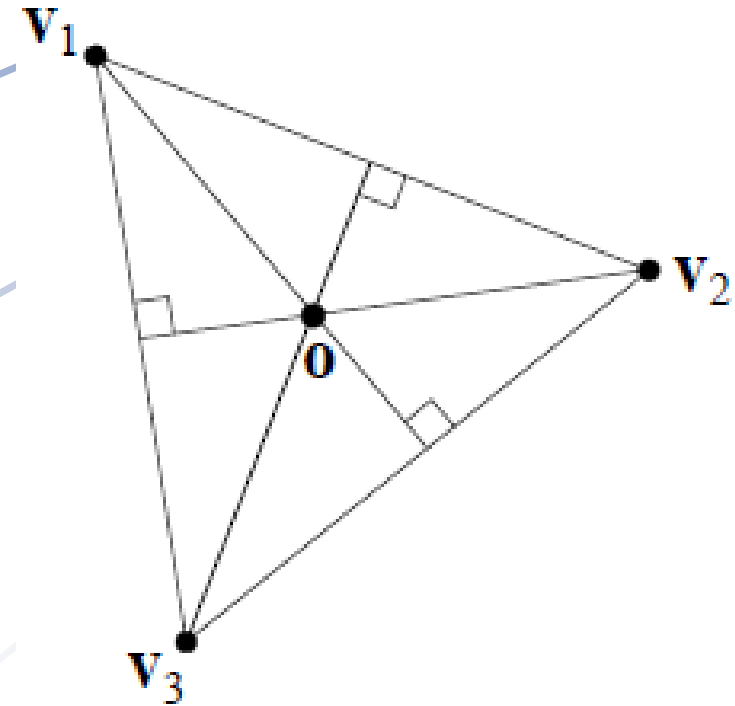
Vanishing points



# Properties of the Projective Transformation



- The orthocenter of the triangle formed by the vanishing points  $v_1, v_2, v_3$ , corresponding to three perpendicular parallel line directions in the world reference system, is the principal point  $o$  on the image plane.



# Properties of the Projective Transformation



- *Cross-ratio (or anharmonic ratio)  $C_r$* : ratio of ratios of distances between collinear points.
  - It is a geometric property invariant under a projective transformation.
- For four points  $\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3, \mathbf{p}_4$  in  $\mathbb{P}^2$ :

$$C_r(\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3, \mathbf{p}_4) = \frac{\Delta_{13}\Delta_{24}}{\Delta_{14}\Delta_{23}}$$

$\Delta_{ij}$ : the Euclidean distance between points  $\mathbf{p}_i$  and  $\mathbf{p}_j$ .



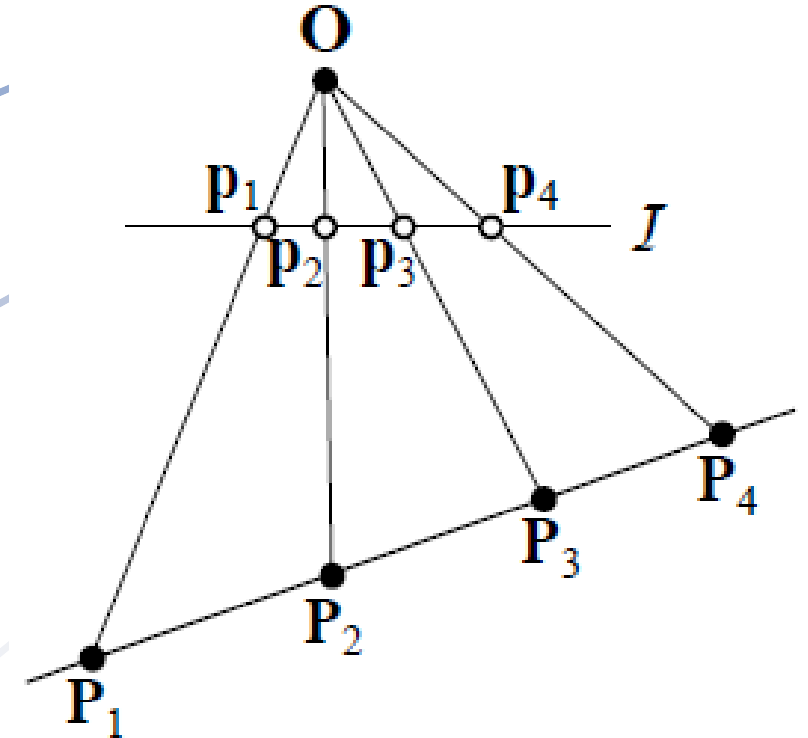
# Properties of the Projective Transformation

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- The cross-ratio of four collinear points remains invariant under a projective transformation:

$$C_r(\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3, \mathbf{p}_4) = C_r(\mathbf{P}_1, \mathbf{P}_2, \mathbf{P}_3, \mathbf{P}_4).$$





# Properties of the Projective Transformation



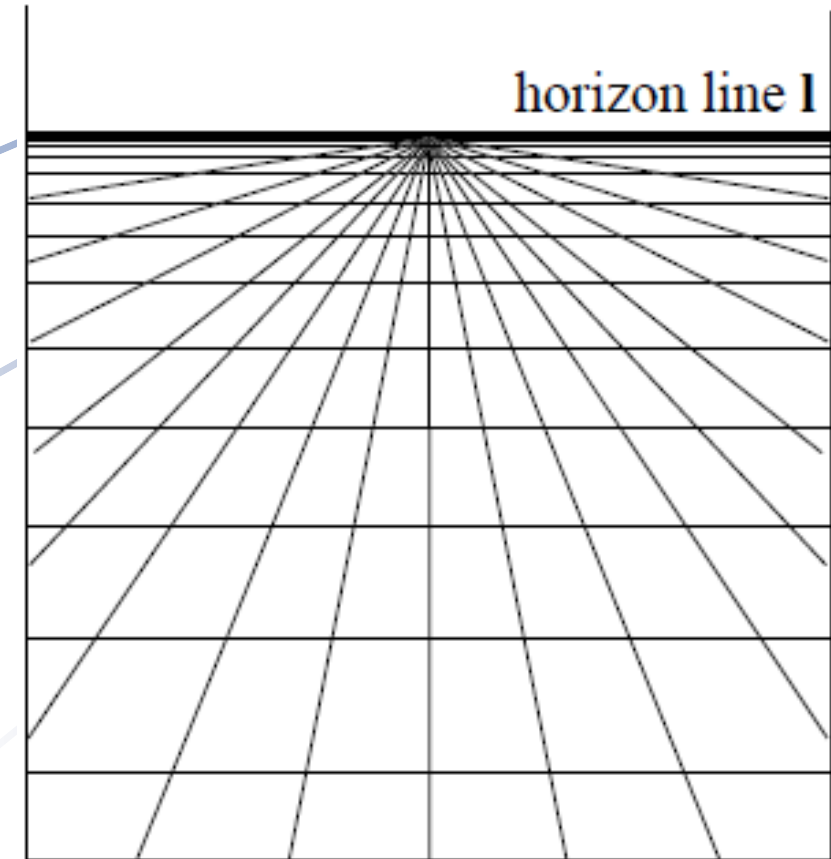
- *Vanishing lines*: the projections on the image of lines at infinity in  $\mathbb{P}^3$ , where parallel planes in the 3D Euclidean space intersect.
- *Horizon line*: the vanishing line of the ground plane and its parallel planes.
  - The projections of parallel lines lying on a plane that forms an angle  $\theta$  with the ground, intersect either above or below the horizon line, depending on the sign of  $\cos \theta$ .



# Properties of the Projective Transformation



- *Chirp effect*: the increase in local image spatial frequency proportionally to the distance of the projected scene area from the camera.
- It is evident in 2D image regions where distant and close-up scene parts are projected.





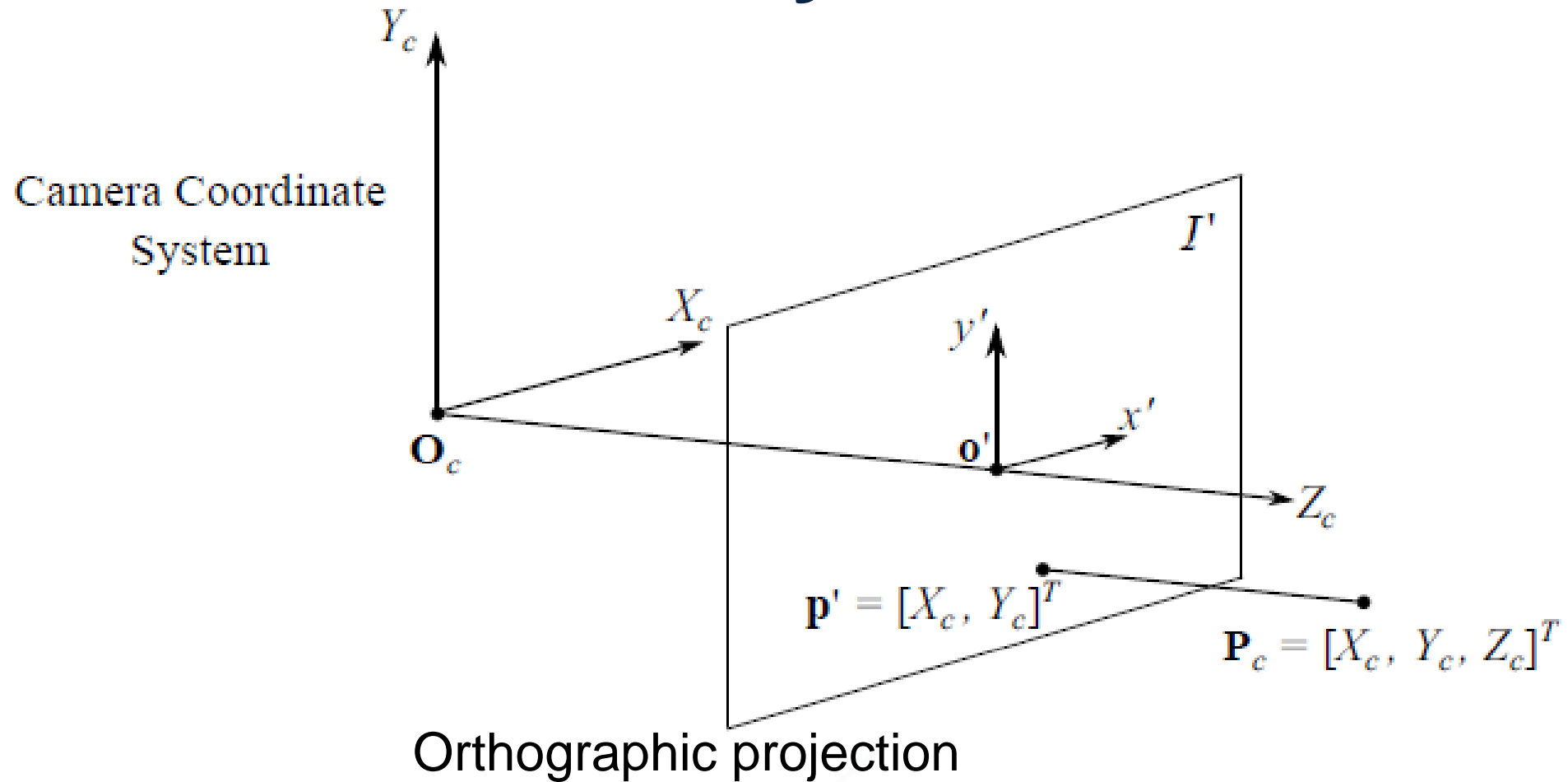
# Cameras at Infinity

- When the center of projection lies at infinity, instead of perspective projection, the orthographic camera and the weak-perspective camera model are used.
- Weak-perspective camera model projection matrix  $\mathcal{P}_{wp}$  :

$$\mathcal{P}_{wp} = \begin{bmatrix} -fr_{11} & -fr_{12} & -fr_{13} & f\mathbf{R}_1^T \mathbf{T} \\ -fr_{21} & -fr_{22} & -fr_{23} & f\mathbf{R}_2^T \mathbf{T} \\ 0 & 0 & 0 & \mathbf{R}_3^T (\bar{\mathbf{P}} - \mathbf{T}) \end{bmatrix}$$

$\bar{\mathbf{P}}$ : centroid of the viewed object.

# Cameras at infinity





# Cameras at Infinity

- The weak-perspective projection is valid, if the depth variations amongst the points of a viewed object are small, in comparison with its average distance from the camera (object depth), represented by  $\mathbf{R}_3^T (\bar{\mathbf{P}} - \mathbf{T})$ .
- Another camera-at-infinity model is the *affine camera model*, with projection matrix:

$$\mathcal{P}_{af} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ 0 & 0 & 0 & a_{34} \end{bmatrix}$$



# Cameras at Infinity

- The left  $3 \times 3$  sub-matrix of  $\mathcal{P}_{af}$  is singular.
- $\mathcal{P}_{af}$  has 9 independent entries.
- Main difference between the affine camera model and the projective framework:
  - In the projective framework, no distinction exists between points at infinity and the usual finite points of the affine space.



## Cameras at Infinity

- The affine camera model does **not** really describe any physical camera, but is used due to its simplicity.
- Main difference between the affine camera model and the weak-perspective model:
  - It preserves straight line parallelism, but **not** angles, since it may entail anisotropic scaling.





# Camera Calibration

- Camera calibration deals with determining the extrinsic and intrinsic camera parameters.
- To this end, a *calibration pattern*, also called *calibration grid* is employed, so that the projection matrix  $\mathcal{P}$  of the camera can be computed from a single view by mapping known points  $\mathbf{P}_w = [X_w, Y_w, Z_w]^T$  in world coordinates to their projections  $\mathbf{p} = [x, y]^T$  on the image plane.

# Camera Calibration

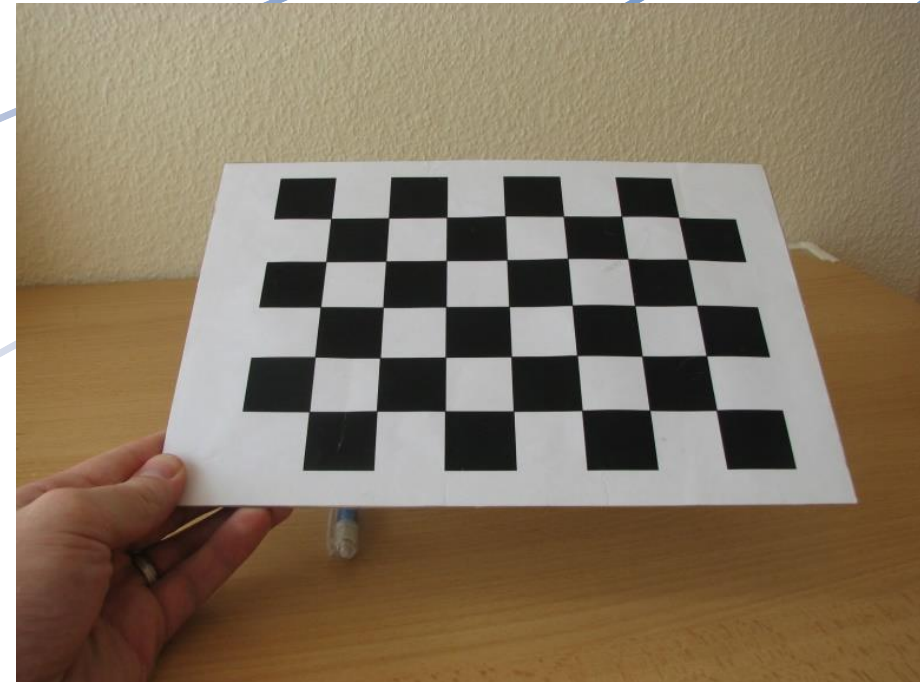
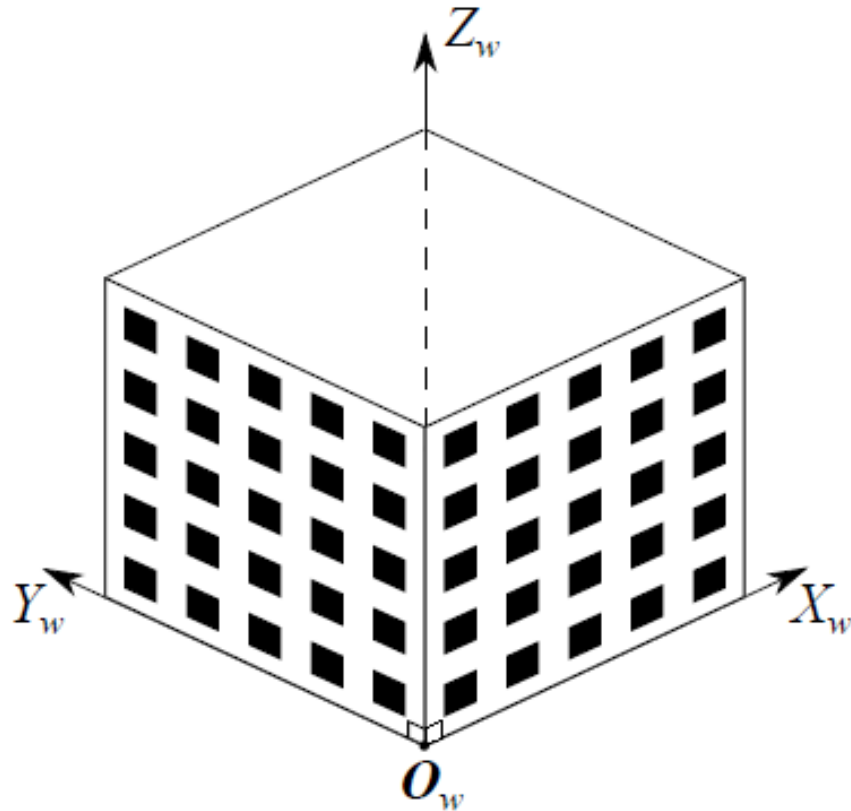


- Taking as many such mappings as needed, a system of equations is formed.
- The solution of the system leads to the determination of the unknown extrinsic and intrinsic camera parameters.



# Camera Calibration

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Calibration patterns.

# Camera Calibration



- The calibration pattern is a 3D object of common, known dimensions and positioning, with a checkerboard pattern clearly visible on each side.
- Pattern dimensions must be known, in an accuracy much greater than the desired calibration accuracy.



# Camera Calibration

- The most popular calibration techniques utilizing solely a planar calibration pattern are:
  - *Direct camera parameter estimation* and
  - *Zhang's calibration method*.
- Other calibration methods do not require a calibration object and are jointly referenced by the term *self-calibration* or *autocalibration*.

# Direct camera parameter estimation



- $\mathbf{P}_w = [X_w, Y_w, Z_w]^T$ : a known point in world coordinates.
- $\mathbf{P}_c = [X_c, Y_c, Z_c]^T$ : the same point in camera coordinates.
- $\mathbf{p}_d = [x_d, y_d]^T$ : its image point in pixel coordinates.
- The transformation between the world and camera coordinate systems involves an orthonormal  $3 \times 3$  rotation matrix  $\mathbf{R}$  and a  $3 \times 1$  translation vector  $\mathbf{T}$  (equivalent to determining the extrinsic camera parameters).



# Direct camera parameter estimation



$$\begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix} = \mathbf{R} \begin{bmatrix} X_w \\ Y_w \\ Z_w \end{bmatrix} + \mathbf{T} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} X_w \\ Y_w \\ Z_w \end{bmatrix} + \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix}$$

- It can be decomposed into:

$$X_c = r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x$$

$$Y_c = r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y$$

$$Z_c = r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z$$



# Direct camera parameter estimation



- Assuming  $o_x = o_y = 0$ , the point  $\mathbf{P}_c$  in the camera coordinate system is related to the pixel coordinates of point  $\mathbf{p}_d$  by:

$$x_d = -\frac{f}{s_x} \frac{X_c}{Z_c} \quad y_d = -\frac{f}{s_y} \frac{Y_c}{Z_c}$$

and finally:

$$x_d = -\frac{f}{s_x} \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$
$$y_d = -\frac{f}{s_y} \frac{r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

# Direct camera parameter estimation



- Using a sufficient number of world coordinate and pixel coordinate point pairs, equations can be formulated and solved for the unknown camera parameters:
  - $r_{11}, r_{12}, \dots, r_{33}, T_x, T_y, T_z, s_x, s_y, f$ .
- It should be noted that knowledge of the ratios  $f/s_x, f/s_y$ , rather than of all internal camera parameters suffices for camera calibration.



# Direct camera parameter estimation



- Internal camera parameters:
  - $f$ : focal length in pixel length.
  - $s_x, s_y$ : pixel size.
  - $o_x, o_y$ : camera center coordinates.



# Direct camera parameter estimation



$$x_d = -\frac{f}{s_x} \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$
$$y_d = -\frac{f}{s_y} \frac{r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

- Since the two equations have the same denominator, for each pair of 3D points  $\mathbf{P}_{wi} = [X_{wi}, Y_{wi}, Z_{wi}]^T$  and their image points  $\mathbf{p}_{di} = [x_{di}, y_{di}]^T$ ,  $i = 1, \dots, N$ , we have:

$$x_{di} \frac{f}{s_y} (r_{21}X_{wi} + r_{22}Y_{wi} + r_{23}Z_{wi} + T_y) = y_{di} \frac{f}{s_x} (r_{11}X_{wi} + r_{12}Y_{wi} + r_{13}Z_{wi} + T_x)$$

# Direct camera parameter estimation



- By using the pixel aspect ration  $a = s_x/s_y$ , putting all the equation terms on the left side and employing  $N$  point pairs, we get:

$$\begin{array}{ccccccc}
 x_{d1}r_{21}X_{w1} & + \cdots & + x_{d1}T_y & - y_{d1}\alpha r_{11}X_{w1} & - \cdots & - y_{d1}\alpha T_x & = 0 \\
 x_{d2}r_{21}X_{w2} & + \cdots & + x_{d2}T_y & - y_{d2}\alpha r_{11}X_{w2} & - \cdots & - y_{d2}\alpha T_x & = 0 \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \\
 x_{dN}r_{21}X_{wN} & + \cdots & + x_{dN}T_y & - y_{dN}\alpha r_{11}X_{wN} & - \cdots & - y_{dN}\alpha T_N & = 0.
 \end{array}$$

# Direct camera parameter estimation



- Each of the linear homogeneous equations has 8 unknown parameters:
  - $\mathbf{u} = [ar_{11}, ar_{12}, ar_{13}, r_{21}, r_{22}, r_{23}, T_x, T_y]^T \triangleq [u_1, u_2, \dots, u_8]^T$
- Expressing the homogeneous system of equation as a product of the matrix  $\mathbf{X}$

$$\mathbf{X} \triangleq \begin{bmatrix} x_{d1}X_{w1} & x_{d1}Y_{w1} & x_{d1}Z_{w1} & x_{d1} & -y_{d1}X_{w1} & -y_{d1}Y_{w1} & -y_{d1}Z_{w1} & -y_{d1} \\ x_{d2}X_{w2} & x_{d2}Y_{w2} & x_{d2}Z_{w2} & x_{d2} & -y_{d2}X_{w2} & -y_{d2}Y_{w2} & -y_{d2}Z_{w2} & -y_{d2} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x_{dN}X_{wN} & x_{dN}Y_{wN} & x_{dN}Z_{wN} & x_{dN} & -y_{dN}X_{wN} & -y_{dN}Y_{wN} & -y_{dN}Z_{wN} & -y_{dN} \end{bmatrix}$$

$$\mathbf{X}\mathbf{u} = \mathbf{0}$$

and the vector  $\mathbf{u}$ :

# Direct camera parameter estimation



- Thus, the desired solution is the *null space* of the matrix  $\mathbf{X}$ .
- Provided that  $N \geq 7$  pairs are not coplanar:
  - Matrix  $\mathbf{X}$  will have rank 7.
  - The system of equations will have one non-trivial solution  $\mathbf{u} \neq \mathbf{0}$ , obtained via the *singular value decomposition* (SVD) of matrix  $\mathbf{X}$ :

$$\mathbf{X} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$$

- $\mathbf{\Sigma}$  is a diagonal matrix containing the singular values. The solution  $\mathbf{u}$  is the column of matrix  $\mathbf{V}$  corresponding to the zero singular value of  $\mathbf{\Sigma}$ .

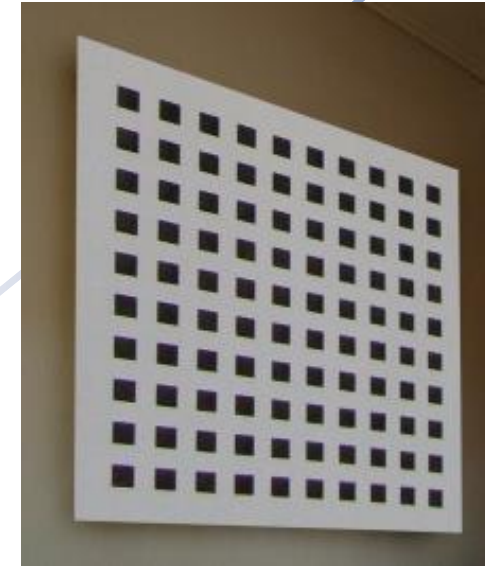






# Zhang's calibration method

- It uses a planar calibration pattern, posing in  $N$  different orientations ( $N \geq 2$ ), by moving either the pattern or the camera.
  - Precise knowledge of this motion is not required.
  - The calibration pattern can simply be printed on a paper and attached to any planar surface.
  - It has reduced complexity.





# Zhang's calibration method

- Let  $\mathbf{P}_I$  be the intrinsic parameters matrix:

$$\mathbf{P}_I = \begin{bmatrix} \alpha_x & s_\theta & o_x \\ 0 & \alpha_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$

where  $\begin{bmatrix} o_x & o_y \end{bmatrix}^T$  the principal point coordinates,  $s_\theta$  the skew factor and  $a_x = -\frac{f}{s_x}$ ,  $a_y = -\frac{f}{s_y}$ .



# Zhang's calibration method

- Assuming the calibration pattern lies on the scene plane  $Z_w = 0$  (in world coordinates):

$$s \begin{bmatrix} x_d \\ y_d \\ 1 \end{bmatrix} = \mathbf{P}_I[\mathbf{R}_1|\mathbf{R}_2|\mathbf{R}_3|\mathbf{T}] \begin{bmatrix} X_w \\ Y_w \\ 0 \\ 1 \end{bmatrix} = \mathbf{P}_I[\mathbf{R}_1|\mathbf{R}_2|\mathbf{T}] \begin{bmatrix} X_w \\ Y_w \\ 1 \end{bmatrix}$$

where  $\mathbf{P}_w = [X_w, Y_w, 0]^T$  a scene point on the calibration pattern,  $\mathbf{p} = [x_d, y_d]^T$  its two-dimensional image and  $s$  is just a scale factor.



# Zhang's calibration method

- A  $3 \times 3$  homography matrix  $\mathbf{H}$  can be defined up to a scale factor, relating  $\mathbf{P}$  and  $\mathbf{p}$ :

$$s\mathbf{p} = \mathbf{H}\mathbf{P} = \mathbf{P}_I[\mathbf{R}_1|\mathbf{R}_2|\mathbf{T}]\mathbf{P}$$

Therefore:

$$\mathbf{H} = \lambda \mathbf{P}_I[\mathbf{R}_1|\mathbf{R}_2|\mathbf{T}]$$

where  $\lambda$  is a scale factor and  $\mathbf{H} \triangleq [\mathbf{h}_1\mathbf{h}_2\mathbf{h}_3]$



# Zhang's calibration method

- From known 3D and 2D correspondences, such a homography can be estimated iteratively and by utilizing

$$\mathbf{sp} = \mathbf{HP} = \mathbf{P}_I[\mathbf{R}_1|\mathbf{R}_2|\mathbf{T}]\mathbf{P}$$

to form an objective function for optimization with the aid of a non-linear optimization algorithm, like Levenberg-Marquardt:

$$\min_{\mathbf{H}_i} E(\mathbf{H}_i) = \sum_{j=1}^M \|\mathbf{p}_{ij} - \hat{\mathbf{H}}_i \mathbf{P}_j\|^2, \quad i = 1, \dots, N.$$

# Zhang's calibration method



- Given a known  $\mathbf{H}$ , let  $\omega$  be a symmetric matrix:

$$\omega \triangleq \begin{bmatrix} \omega_{11} & \omega_{12} & \omega_{13} \\ \omega_{21} & \omega_{22} & \omega_{23} \\ \omega_{31} & \omega_{32} & \omega_{33} \end{bmatrix} = \mathbf{P}_I^{-T} \mathbf{P}_I^{-1} =$$

$$= \begin{bmatrix} \frac{1}{\alpha_x^2} & \frac{-s_\theta}{\alpha_x^2 \alpha_y} & \frac{o_y s_\theta - o_x \alpha_y}{\alpha_x^2 \alpha_y} \\ \frac{-s_\theta}{\alpha_x^2 \alpha_y} & \frac{s_\theta^2}{\alpha_x^2 \alpha_y^2} + \frac{1}{\alpha_y^2} & \frac{-s_\theta (o_y s_\theta - o_x \alpha_y)}{\alpha_x^2 \alpha_y^2} - \frac{o_y}{\alpha_y^2} \\ \frac{o_y s_\theta - o_x \alpha_y}{\alpha_x^2 \alpha_y} & \frac{-s_\theta (o_y s_\theta - o_x \alpha_y)}{\alpha_x^2 \alpha_y^2} - \frac{o_y}{\alpha_y^2} & \frac{(o_y s_\theta - o_x \alpha_y)^2}{\alpha_x^2 \alpha_y^2} + \frac{o_y^2}{\alpha_y^2} + 1 \end{bmatrix}$$

which can be represented by  $\mathbf{b} = [\omega_{11}, \omega_{12}, \omega_{22}, \omega_{13}, \omega_{23}, \omega_{33}]^T$



# Zhang's calibration method

- Since  $\mathbf{R}_1$  and  $\mathbf{R}_2$  are orthogonal,  $s\mathbf{p} = \mathbf{H}\mathbf{P} = \mathbf{P}_I[\mathbf{R}_1|\mathbf{R}_2|\mathbf{T}]\mathbf{P}$  and the definition of  $\omega$  entail that:

$$\mathbf{h}_1^T \omega \mathbf{h}_2 = 0$$

$$\mathbf{h}_1^T \omega \mathbf{h}_1 = \mathbf{h}_2^T \omega \mathbf{h}_2.$$

$$\mathbf{h}_i^T \omega \mathbf{h}_j = \mathbf{v}_{ij}^T \mathbf{b}$$

where:

$$\mathbf{v}_{ij}^T \triangleq [h_{i1}h_{j1}, h_{i1}h_{j2}+h_{i2}h_{j1}, h_{i2}h_{j2}, h_{i3}h_{j1}+h_{i1}h_{j3}, h_{i3}h_{j2}+h_{i2}h_{j3}, h_{i3}h_{j3}]$$

- Based on the way  $\mathbf{v}$  is defined,

$$\begin{bmatrix} \mathbf{v}_{12}^T \\ (\mathbf{v}_{11} - \mathbf{v}_{22})^T \end{bmatrix} \mathbf{b} = \mathbf{0}$$





# Zhang's calibration method

- By taking simultaneously into account  $N$  different images of the calibration pattern, and using  $N$  different homography matrices  $\mathbf{H}_i, i = 1, \dots, N$ , the system of  $N$  corresponding equations can be compactly restated as:

$$\mathbf{A}\mathbf{b} = 0,$$

where  $\mathbf{A}$  is a  $2N \times 6$  matrix.

- This system can be solved for  $\mathbf{b}$  applying SVD to  $\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$ .



# Zhang's calibration method

- The intrinsic camera parameters can be estimated from  $\omega$ :

$$o_y = \frac{(\omega_{12}\omega_{13} - \omega_{11}\omega_{23})}{\omega_{11}\omega_{22} - \omega_{12}^2}$$

$$\alpha_y = \sqrt{\frac{\lambda\omega_{11}}{\omega_{11}\omega_{22} - \omega_{12}^2}}$$

$$\lambda = \omega_{33} - \frac{\omega_{13}^2 + o_y(\omega_{12}\omega_{13} - \omega_{11}\omega_{23})}{\omega_{11}}$$

$$s_\theta = -\frac{\omega_{12}\alpha_x^2\alpha_y}{\lambda}$$

$$\alpha_x = \sqrt{\frac{\lambda}{\omega_{11}}}$$

$$o_x = \frac{s_\theta o_y}{\alpha_y} - \frac{\omega_{13}\alpha_x^2}{\lambda}$$



# Zhang's calibration method

- Having computed the intrinsic camera parameters, the matrix  $\mathbf{P}_I$  can also be estimated.
- Thus, by substituting  $\mathbf{P}_I$  in the equations:

$$\mathbf{p}_{ij} = \lambda \mathbf{P}_I [\mathbf{R}_{1i} | \mathbf{R}_{2i} | \mathbf{T}_i] \mathbf{P}_j, \quad i = 1, \dots, N, \quad j = 1, \dots, M$$

we can solve for the columns of the rotation matrix and the translation vector, in order to obtain the extrinsic camera parameters (rotation matrix  $\mathbf{R}_i$ , translation vector  $\mathbf{T}_i$ ).

# Zhang's calibration method

MultiDrone



$$\mathbf{R}_{1i} = \lambda \mathbf{P}_I^{-1} \mathbf{h}_{1i}$$

$$\mathbf{R}_{2i} = \lambda \mathbf{P}_I^{-1} \mathbf{h}_{2i}$$

$$\mathbf{R}_{3i} = \mathbf{R}_{1i} \times \mathbf{R}_{2i}$$

$$\mathbf{T}_i = \lambda \mathbf{P}_I^{-1} \mathbf{h}_{3i}$$

$$\lambda = \frac{1}{\|\mathbf{P}_I^{-1} \mathbf{h}_{1i}\|} = \frac{1}{\|\mathbf{P}_I^{-1} \mathbf{h}_{2i}\|}.$$

- The above estimated results can be used as initializations for some repetitive optimization algorithm, so that refined results ones can be derived.



## Self-calibration

- These approaches:
  - do not require a calibration object and recover the camera parameters from image information alone;
  - are flexible but not very robust;
  - exploit properties of the absolute conic of the projective geometry;
  - typically require information equivalent to a partial 3D reconstruction of the scene.
- Self-calibration is strongly related to the geometry of multiple cameras.

# Q & A



**Thank you very much for your attention!**

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