CSCM77 Camera Model and Calibration

Prof. Xianghua Xie

x.xie@swansea.ac.uk

http://csvision.swan.ac.uk

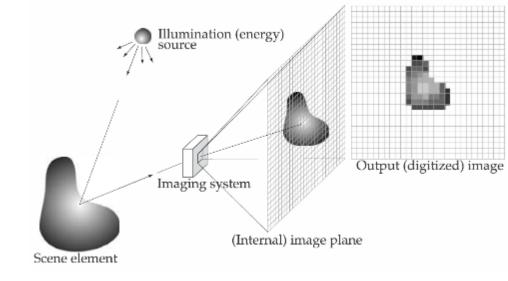
Content

- Camera models
 - Perspective or pin-hole
 - Weak perspective & orthographic projection
- Parameters
 - Extrinsic (camera position and orientation)
 - Intrinsic (relationship between pixel and image plane coordinates)
- Camera calibration
 - Calibration pattern and system
 - Direct calibration method

CSCM77: CVDI

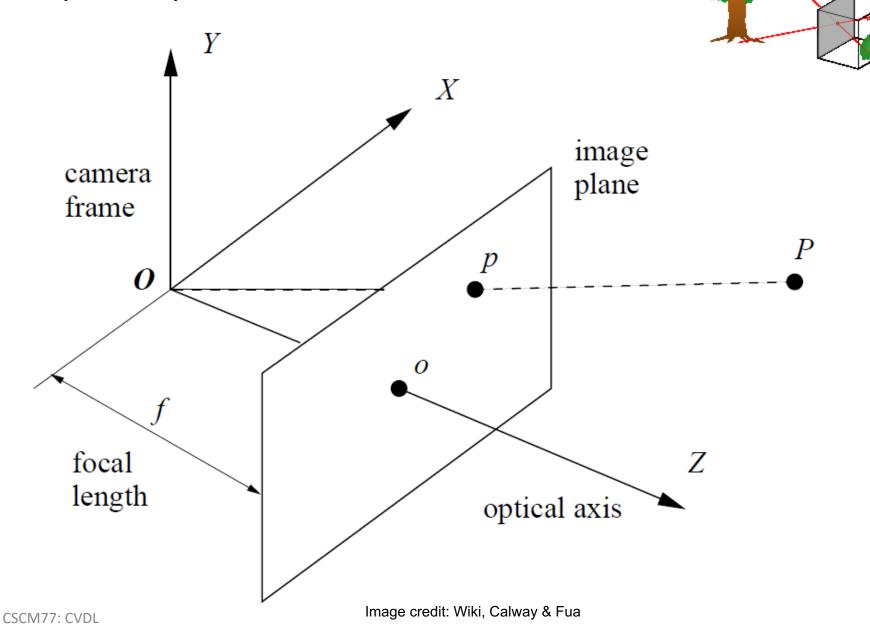
Digital image formation

- Light focused onto image plane using optical system
- Intensity variation across
 image plane captured by
 scene element
 array of photosensitive devices, e.g. CCD



- Intensity values captured at given instant in time to give array of pixel values
- Each pixel has a position in **pixel coordinates**, e.g. (i,j) that is (i-th row and j-th column) and an intensity value E(i,j)
- Intensity values usually in the range [0,255] for monochrome
- Spatial resolution determined by effective area of image plane occupied by each pixel, $s_x \times s_y$

Perspective pin-hole camera model



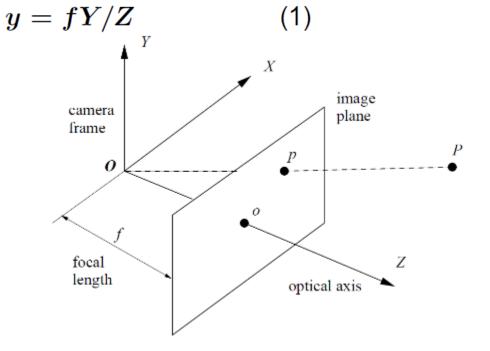
Perspective projection

- Assume all points defined with respect to camera frame
- Point $P = [X,Y,Z]^T$ in 3D space
- P projects to point $p = [x, y, f]^T$ in image plane
 - -f is **focal length:** the parameter of interest in perspective proj.
 - T denotes transpose

x = fX/Z

Relationship between P and p is defined as:

$$X$$
 X
 Z



CSCM77: CVDL

Weak perspective

- Let maximum difference in depth (distance to camera centre, focal point) between any two scene points = δZ
- Let average depth of all scene point = \bar{Z}
- If the variation in depth is negligible compared to average depth, then for each scene point:

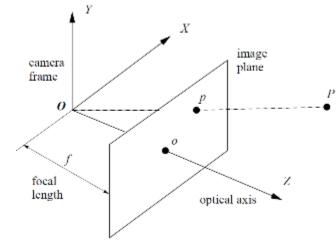
$$x \approx fX/\bar{Z}$$
 $y \approx fY/\bar{Z}$

- Since focal length and average depth are fixed constants, it becomes a linear camera model
- Viable approximation when variation in depth is less than 5%
- Known as "weak perspective": retains some sense of perspective
- If ignore the constant values, it becomes the orthographic projection (x=X, y=Y): no perspective

CSCM77: CVDL

Camera parameters

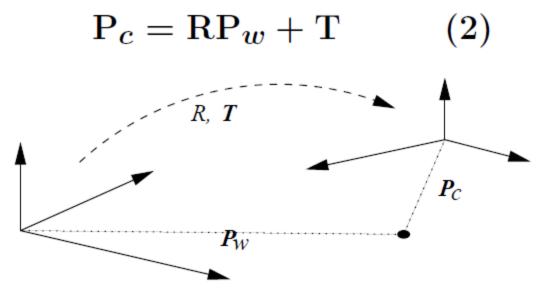
- Previously, points are defined with respect to camera reference frame, e.g. perspective camera model
- However, we require
 - Camera reference frame can be located with respect to some world reference
 - Establish the relationship between the coordinates of the image plane and pixel coordinates



- Working out the above two relationships require two sets of camera parameters
 - Extrinsic parameters
 - Intrinsic parameters

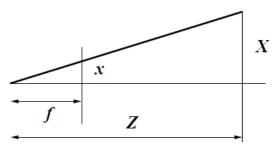
Extrinsic camera parameters

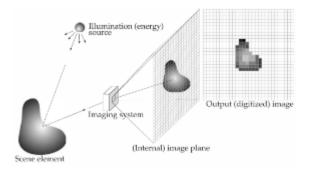
- Parameters that define location and orientation of camera reference frame with respect to known world reference frame
- Camera and world frame coordinates related by
 - 3 by 3 rotation matrix R, which defines rotations about X, Y and
 Z
 - 3D translation vector T, which defines translation between origins



Intrinsic camera parameters

- For pin hole camera, 3 sets of intrinsic parameters specifying
 - The perspective projection: focal length f
 - Transformation between image plane and pixel coordinates
 - Geometric distortion introduced by optics

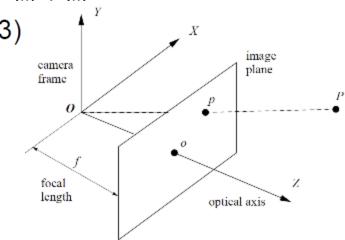




• Image plane (x,y) and pixel coordinates (x_{im}, y_{im}) :

$$x = (x_{im} - o_x)s_x$$
 $y = (y_{im} - o_y)s_y$ (3)

- $-o_x, o_y$ origin of image plane
- $-s_x$, s_y spatial resolution



Intrinsic camera parameters

 Geometric distortion can be modelled as displacements increasing with radial distance form image centre:

$$x = x_d(1 + k_1r^2 + k_2r^4)$$
 $y = y_d(1 + k_1r^2 + k_2r^4)$

 $-(x_d, y_d)$: coordinates of distorted points

$$-r^2 = x_d^2 + y_d^2$$





Linking pixel and world coordinates

- Given intrinsic and extrinsic parameters, we can link pixel and world coordinates together, i.e. mapping between pixel and world reference coordinates
- Intrinsic parameters allow us to map pixel coordinates to image plane/camera coordinates
- Perspective projection equation tells us how 3D points projected to image plane; those 3D points are defined with respect to camera coordinates
- Extrinsic parameters allow us to map the world reference and camera coordinates (rotation and translation)

CSCM77: CVDI

Linking pixel and world coordinates

- Link pixel and world coordinates together
- Recall (1-3):
 - (1): perspective projection
 - (2): rotation and translation between two coordinates (ext.)
 - (3): image plane and pixel coordinates (int.)

$$x = fX/Z$$
 $y = fY/Z$ (1)
 $P_c = RP_w + T$ (2)
 $x = (x_{im} - o_x)s_x$ $y = (y_{im} - o_y)s_y$ (3)

• We then have: $(x_{im}-o_x)s_x=frac{\mathrm{R}_1^T\mathrm{P}_w+T_x}{\mathrm{R}_3^T\mathrm{P}_w+T_z}$ (4) $(y_{im}-o_y)s_y=frac{\mathrm{R}_2^T\mathrm{P}_w+T_y}{\mathrm{R}_3^T\mathrm{P}_w+T_z}$

- R_i^T is *i*-th row of matrix R and $T=[T_x, T_y, T_z]^T$

Projection matrix

Equation (4) can be re-arranged:

$$\mathbf{x} = egin{bmatrix} x_1 \ x_2 \ x_3 \end{bmatrix} = \mathbf{M}_{int} \mathbf{M}_{ext} egin{bmatrix} \mathbf{P}_w \ 1 \end{bmatrix}$$

- Where $x_{im} = x_1/x_3, y_{im} = x_2/x_3$

$$- \ \mathrm{M}_{int} = egin{bmatrix} f/s_x & 0 & o_x \ 0 & f/s_y & o_y \ 0 & 0 & 1 \end{bmatrix}$$

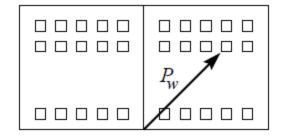
$$- M_{ext} = [R, T] \quad (3 \times 4)$$

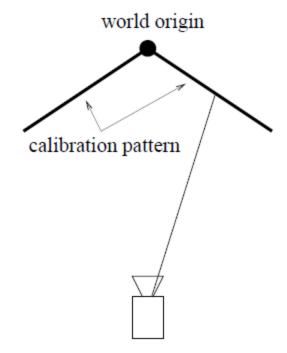
• The 3 by 4 matrix $M=M_{int}M_{ext}$ is known as the projection matrix

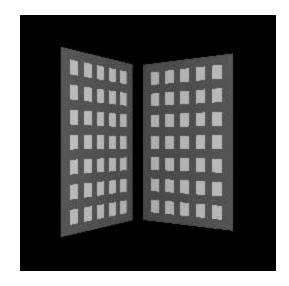
- Of course, camera intrinsic and extrinsic parameters are not known, i.e. the projection matrix is actually not determined
- However, we can use this projection matrix to estimate intrinsic and extrinsic parameters
- This is known as camera calibration
- Direct parameter calibration
 - Capture images containing calibration patterns with known 3D points
 - Determine pixel coordinates of projected points
 - Solve equations linking world and pixel coordinates using N such points and Least-Square optimisation

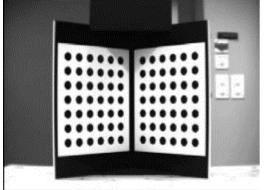
CSCM77: CVDI

- Calibration pattern is stationary
- Its origin is considered as the world reference origin









• Assume image centre (o_x, o_y) is known, so

$$x'_{im} = x_{im} - o_x \qquad y'_{im} = y_{im} - o_y$$

According to (4) we have

$$x'_{im} = f_x \frac{R_1^T P_w + T_x}{R_3^T P_w + T_z}$$
 $y'_{im} = f_y \frac{R_2^T P_w + T_y}{R_3^T P_w + T_z}$ (5)

- Where $f_x = f/s_x$ and $f_y = f/s_y$
- Hence from common denominator

$$x'_{im}(\mathbf{R}_2^T \mathbf{P}_w + T_y) - y'_{im}\alpha(\mathbf{R}_1^T \mathbf{P}_w + T_x) = 0$$

– Where aspect ratio $\alpha = f_x/f_y = s_y/s_x$

• Reminder:
$$(x_{im}-o_x)s_x=frac{\mathrm{R}_1^T\mathrm{P}_w+T_x}{\mathrm{R}_3^T\mathrm{P}_w+T_z}$$
 (4) $(y_{im}-o_y)s_y=frac{\mathrm{R}_2^T\mathrm{P}_w+T_y}{\mathrm{R}_3^T\mathrm{P}_w+T_z}$

CSCM77: CVDL

$$x'_{im}(\mathbf{R}_2^T \mathbf{P}_w + T_y) - y'_{im}\alpha(\mathbf{R}_1^T \mathbf{P}_w + T_x) = 0$$

 Thus, for N points above formulation will give us N linear equations with 8 unknowns:

$$[\mathbf{R}_2^T, T_y, \alpha \mathbf{R}_1^T, \alpha T_x]$$

 Solving this linear system (for example, using Singular Value Decomposition) and using constraints such as the length of rotation vectors should be one will give us

$$R_1, R_2, T_x, T_y, \alpha$$

- Rotation matrix has to be orthogonal: $R_3 = R_1 \times R_2$ (cross product)
- Finally, Tz and fx (and fy from aspect ratio) can be similarly worked out using equation (5) again, for example use leastsquare optimisation

$$x'_{im}\mathbf{R}_{3}^{T}\mathbf{P}_{w} + x'_{im}T_{z} = f_{x}(\mathbf{R}_{1}^{T}\mathbf{P}_{w} + T_{x})$$

Reminder:

$$x'_{im} = f_x \frac{R_1^T P_w + T_x}{R_3^T P_w + T_z}$$
 $y'_{im} = f_y \frac{R_2^T P_w + T_y}{R_3^T P_w + T_z}$ (5)

Application

 Work such as this one extensively involves camera parameter estimation (in a very large scale)

