

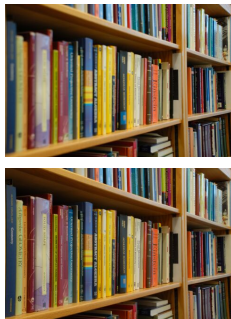
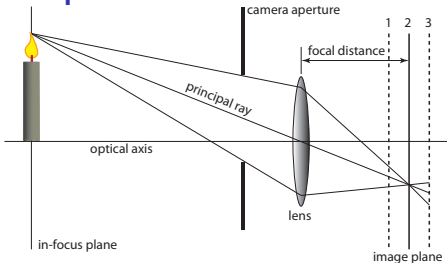
Real Cameras and their Calibration

COMPSCI 527 — Computer Vision

Outline

- 1 Real Cameras
 - Depth of Field
 - Distortion
- 2 Camera Calibration
 - A Camera Model
 - Parameter Optimization
 - Lab Setup and Imaging

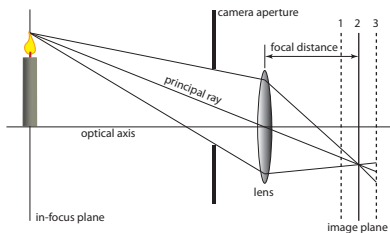
Depth of Field



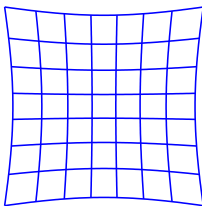
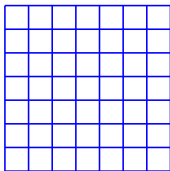
- Focal *length*: focal *distance* when an object at ∞ is in focus
- Focal length is a lens property
- Focal distance can be changed by rotating the focusing ring
- Nothing to do with *zoom*, which changes focal length
- Alas, f is often used for either focal distance or focal length

Changing Depth of Field

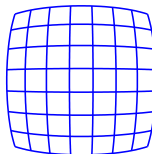
- *Aperture*: diameter of the hole in front of the lens
- Measured in *stops*, or *f-numbers* $n = \frac{f}{a}$
(a is aperture diameter, f is focal length)
- Area (light flux) is proportional to *square* of diameter
- Small aperture (big f -number) \Rightarrow great depth of field
- A shallow depth of field is sometimes desirable



Lens Distortion



pincushion



barrel

Camera Calibration

- Cameras have *intrinsic parameters*: focal distance, pixel size, principal point, lens distortion parameters
- ... and *extrinsic parameters*: Rotation, translation relative to some world reference system
- *Camera calibration* is a combination of lab measurements and algorithms aimed at determining both types of parameters
- We do not calibrate for finite depth of field (stop down the aperture, flood the scene with light)

Calibration as Learning

- Many variants, the general idea is the same
- Looks very much like machine learning:
 - 1 Make a parametric model of what a camera does: Inputs are world points \mathbf{W} in world coordinates, outputs are image points ξ in image pixel coordinates (“predictor architecture”)
 - 2 Collect a sufficiently large set T of input-output pairs (\mathbf{W}_n, ξ_n) (“training set”)
 - 3 “Loss function” measures discrepancy between $\hat{\xi}_n$ predicted by model and ξ_n measured in image
 - 4 Fit the parameters to T by numerical optimization (“training”)
- We even have generalization requirements: The parameters should be correct for pairs (\mathbf{W}, ξ) not in T
- We already know how to do 3, 4. Need to figure out 1, 2.

Camera Model

$$\mathbf{X} = R(\mathbf{W} - \mathbf{t})$$

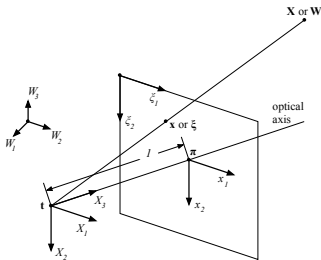
$$\mathbf{x} = p(\mathbf{X}) = \frac{1}{X_3} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix}$$

$$\mathbf{y} = d(\mathbf{x}) \quad (\text{lens distortion})$$

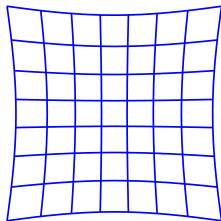
$$\xi = S\mathbf{y} + \pi$$

$$S = f \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix}$$

Can only determine
products $f s_x$ and $f s_y$,
not f , s_x , s_y individually



Lens Distortion Model



- Distortion is radial around the principal point:
 $\mathbf{y} = d(\mathbf{x}) = \delta(r) \mathbf{x}$ where $r = \|\mathbf{x}\|$
- *Radial distortion function* $\delta(\cdot)$ is nonlinear
- Must be analytical everywhere (Maxwell). Implication:
 - Restrict to x axis: $\delta(r(\mathbf{x})) = \delta(|x|)$
 - Odd powers of $|x|$ have a cusp at the origin
 - Therefore, $\delta(r) = 1 + k_1 r^2 + k_2 r^4 + \dots$
 ($k_0 \neq 1$ can be folded into f)
- Large powers of r only affect peripheral areas and cannot be determined well
- Typically, $\delta(r) = 1 + k_1 r^2 + k_2 r^4$

Camera Parameters

$$\mathbf{X} = R(\mathbf{W} - \mathbf{t})$$

$$\mathbf{x} = \rho(\mathbf{X}) = \frac{1}{X_3} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix}$$

$$\mathbf{y} = \mathbf{x} (1 + k_1 \|\mathbf{x}\|^2 + k_2 \|\mathbf{x}\|^4)$$

$$\xi = \mathbf{S}\mathbf{y} + \pi$$

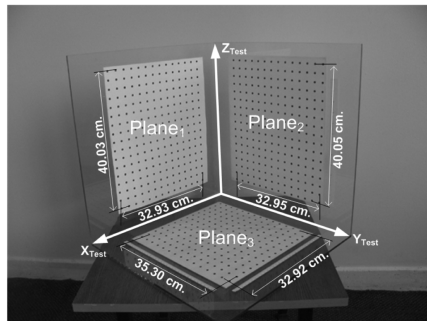
- Extrinsic parameters: R, \mathbf{t} (6 degrees of freedom)
- Intrinsic parameters: $\pi, f s_x, f s_y, k_1, k_2$ (6 numbers)

$$\xi = \mathbf{c}(\mathbf{W}; \mathbf{p}) \text{ where } \mathbf{p} \in \mathbb{R}^{12}$$

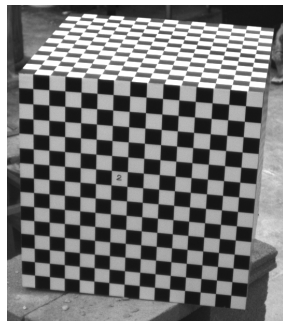
Data Fitting

- Collect input-output pairs (\mathbf{W}_n, ξ_n) for $n = 1, \dots, N$
 $\hat{\xi} = \mathbf{c}(\mathbf{W}; \mathbf{p})$ where $\mathbf{p} \in \mathbb{R}^{12}$
 $\mathbf{p}^* = \arg \min_{\mathbf{p}} e(\mathbf{p})$ where $e(\mathbf{p}) = \frac{1}{N} \sum_{n=1}^N \|\xi_n - \mathbf{c}(\mathbf{W}_n; \mathbf{p})\|^2$
- e is nonlinear
- To initialize: clamp $k_1 = k_2 = 0$, solve a linear system
- Approximate because of clamping *and* because the residual is different from $e(\mathbf{p})$
- Use any optimization algorithm to refine

Calibration Target

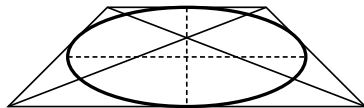
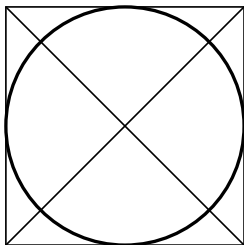


<http://www.mdpi.com/1424-8220/9/6/4572/htm>



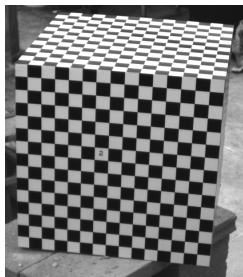
Duke Computer Vision Lab

Circles are Problematic



Calibration Protocol Summary

- Place calibration target in front of camera (fill the image)
- Measure image coordinates (with software help?)
- Make a file with (\mathbf{W}_n, ξ_n) pairs
- Fit parameters by numerical optimization
- Redo if you touch the lens or the camera!



An Example for Distortion Only

