

# SINGLE VIEW GEOMETRY

computer vision

# ABOUT THIS CLASS

- very basic elements of projective geometry :
  - the projective plane and homogeneous coordinates
  - points, lines in homogeneous coordinates
- the geometry of image acquisition
  - single view (finite) camera models
  - camera calibration

(A FEW SLIDES ON)  
PROJECTIVE GEOMETRY

computer vision

# POINTS

- we first consider points on a plane  $(x, y) \in \mathbb{R}^2$
- considering the plane as a vector space a point is identified by a vector
- Convention: points will be represented as column vectors  $\mathbf{x} = (x, y)^\top$

# LINES

$$ax + by + c = 0$$

- A line may also be represented as a vector  $(a, b, c)^\top$
- The line  $(ka)x + (kb)y + kc = 0$  is the same line, for each  $k \neq 0$
- then vectors  $(ka, kb, kc)^\top \quad \forall k \neq 0$  can be seen as equivalent (belong to the same **equivalence class**)
- such equivalence class is called a **homogeneous vector**

# PROJECTIVE PLANE

- the projective plane  $\mathcal{P}^2$  or 2D projective space is formed by any vector  $(a, b, c)^\top$  representative of an equivalence class:

$$\mathbb{R}^3 - (0, 0, 0)^\top$$

•

# LINES AND POINTS

- A point  $(x, y)^\top$  lies on the line  $\mathbf{l} = (a, b, c)^\top$  iff  $ax + by + c = 0$
- This may be written in terms of an inner product:

$$(x, y, 1)(a, b, c)^\top = \mathbf{x}^\top \mathbf{l} = 0$$

- A point on the Euclidean plane may be represented as a point in the projective space (a 3D vector with a final 1).
- Notice that here again points define equivalence classes

# LINES AND POINTS

- The point  $\mathbf{x}$  lies on the line  $\mathbf{l}$  iff  $\mathbf{x}^\top \mathbf{l} = 0$
- Intersection of lines:  $\mathbf{x} = \mathbf{l} \times \mathbf{l}'$  (since the intersection is a point lying on both lines, thus a point whose inner product with both lines is 0...)
- Lines joining points:  $\mathbf{l} = \mathbf{x} \times \mathbf{x}'$  (as above)



# IDEAL POINTS AND LINES AT INFINITY

- What is the intersection of parallel lines?

$$\begin{aligned} ax + by + c &= 0 & \mathbf{l} &= (a, b, c)^\top \\ ax + by + c' &= 0 & \mathbf{l}' &= (a, b, c')^\top \end{aligned}$$

$$\mathbf{x} = \mathbf{l} \times \mathbf{l}' = (b, -a, 0)^\top$$

# SUMMING UP

- Points  $\mathbf{x} = (x_1, x_2, x_3)^\top$  so that  $x_3 \neq 0$  correspond to finite points on the Euclidean plane  $(x_1/x_3, x_2/x_3)^\top$
- Augmenting the Euclidean plane by adding points with  $x_3 = 0$  produces the set of all homogeneous vectors
- Points with  $x_3 = 0$  are known as ideal points or points at infinity
- All points at infinity lie on a single line  $\mathbf{l}_\infty = (0, 0, 1)^\top$

# SUMMING UP

- a line  $(a, b, c)^\top$  intersects  $\mathbf{l}_\infty$  in the ideal point  $(b, -a, 0)^\top$
- the line  $(a, b, c')^\top$  intersects  $\mathbf{l}_\infty$  in the same point
- in inhomogeneous notation  $(b, -a)^\top$  is a vector tangent to the line representing the line's direction
- for these reasons the line at infinity can be thought of as the set of directions of lines in the plane

# PROJECTIVE GEOMETRY

- points at infinity simplify intersection properties
- in the projective plane
  - two distinct lines meet at a single point
  - two distinct points lie on the same line
- the geometry of the projective plane is known as **projective geometry**

# THE DUALITY PRINCIPLE

- to any theorem of 2D projective geometry there corresponds a dual theorem, which may be derived by interchanging the roles of lines and points in the original theorem

# PROJECTIVE TRANSFORMATIONS

(an anticipation...)

- A homography is a linear transformation of homogeneous vectors represented by a non-singular 3x3 matrix  $H$

$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

$$\mathbf{x}' = H\mathbf{x}$$

# A HIERARCHY OF TRANSFORMATIONS

all linear in homogeneous  
coordinates!

- I. isometries
- II. similarities
- III. affinities
- IV. projectivities

# PROJECTIVE SPACE

- A point  $X$  in 3D space is represented in homogeneous coordinates as a 4-vector
- A projective transformation of the 3D space is a  $4 \times 4$  non singular matrix, with 15 d.o.f.



# PROJECTIVE TRANSFORMATIONS

- since homogeneous points define equivalence classes,  
*projective transformations are unique up to a scale factor*
- $x = Hx'$
- $x = 3Hx'$
- ...
- $x = aHx'$  for each non zero constant  $a$

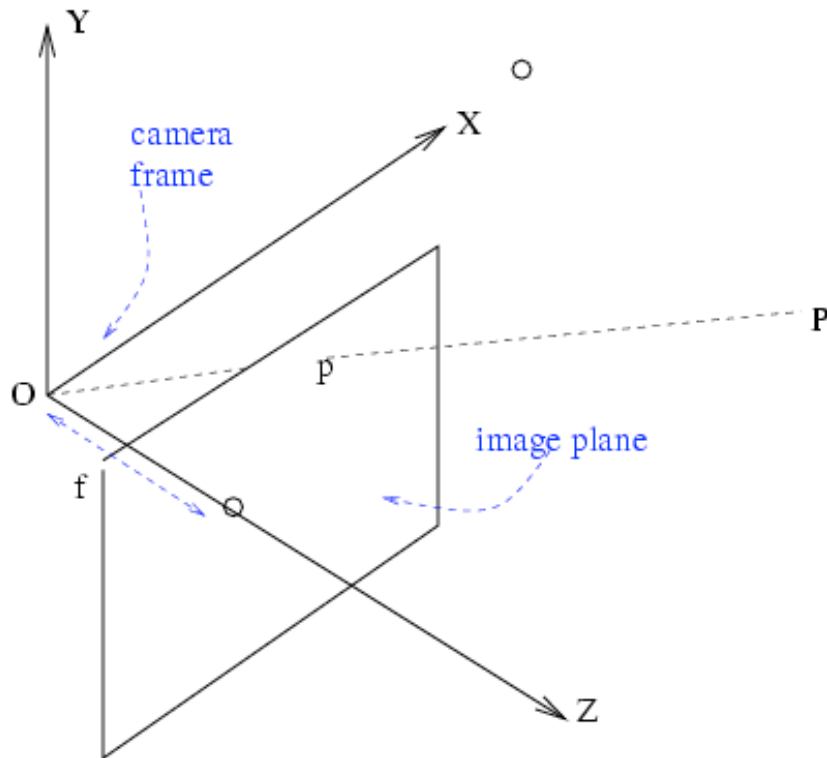
# FROM SPACE TO PLANE

- a perspective projection can be represented as a projective transformation  $x=PX$
- not true the vice-versa...

# CAMERA MODELS AND CAMERA PARAMETERS

# BASIC PIN HOLE CAMERA

- a camera is a mapping between the 3D world and a 2D image



$$x = f \frac{X}{Z}$$
$$y = f \frac{Y}{Z}$$

in matrix form (homogeneous coordinates)

$$\begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = \begin{pmatrix} fX \\ fY \\ Z \end{pmatrix} = \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix}$$

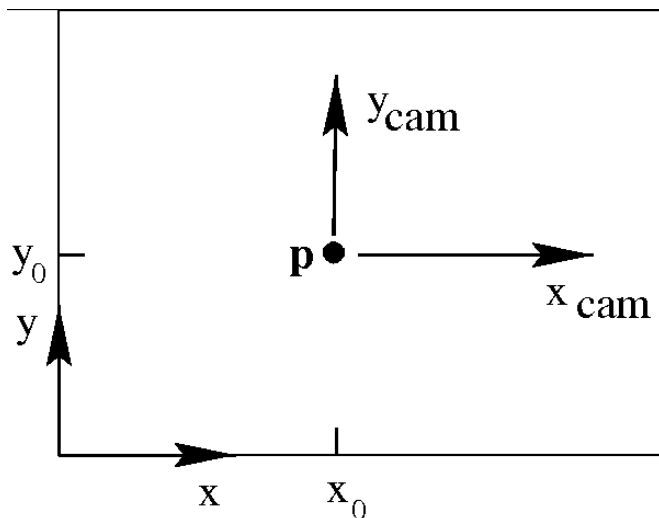
more compactly:  $\mathbf{x} = \mathbf{P} \mathbf{X}$       $\mathbf{P} = \text{diag}(f, f, 1) [\mathbf{I} ; \mathbf{0}]$

# GENERAL MODEL

- a general model for a camera should take into account:
  - rigid transformation between the camera and the 3D world reference frames
  - the transformation from metric coordinates to pixels coordinates

# INTRINSIC PARAMETERS

- in the basic model we assumed the origin of the image plane to be the principal point
- in practice this will not be
  - we may add a translation of the points in the image plane



$$\mathbf{x} = P \mathbf{X} \quad P = K [I ; \mathbf{0}]$$

$$K = \begin{bmatrix} f & 0 & p_x \\ 0 & f & p_y \\ 0 & 0 & 1 \end{bmatrix}$$

# INTRINSIC PARAMETERS

- in images units are expressed in pixels
- then the matrix  $K$  of intrinsic parameters becomes

$$K = \begin{bmatrix} \alpha_x & 0 & x_0 \\ 0 & \alpha_y & y_0 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{all entries in pixel coordinates}$$

- with

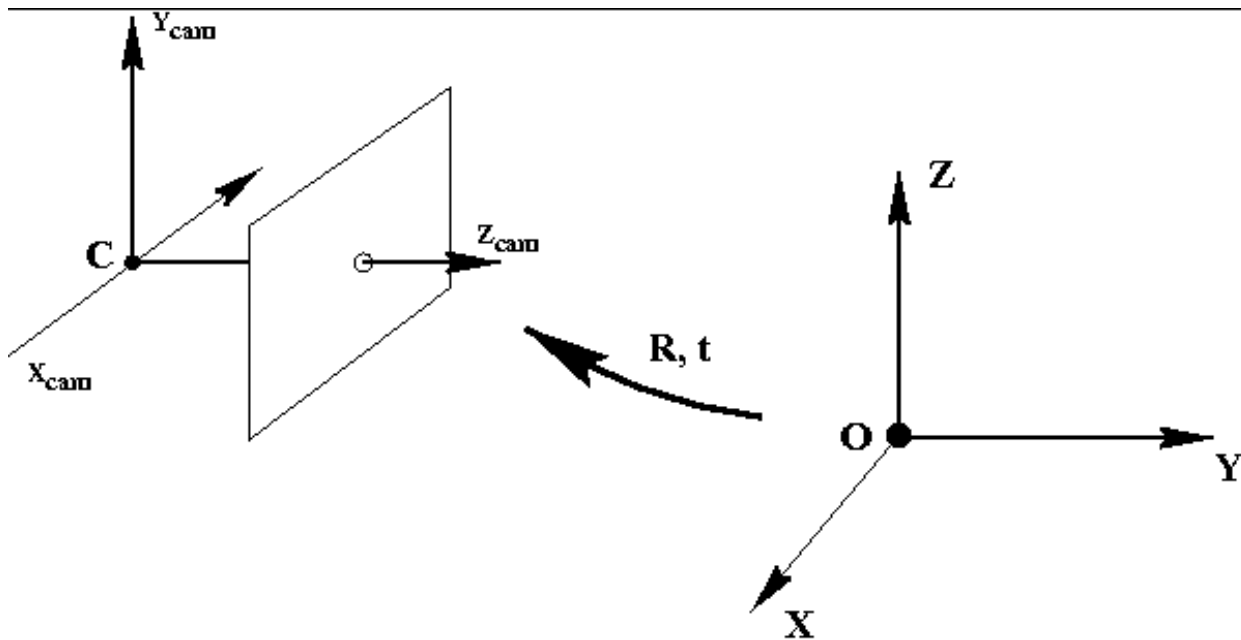
$$\begin{aligned} \alpha_x &= f m_x \\ \alpha_y &= f m_y \end{aligned} \quad \begin{array}{l} \swarrow \searrow \\ \text{number of pixels per unit distance} \end{array}$$

$$x_0 = m_x p_x$$

$$y_0 = m_y p_y$$

# EXTRINSIC PARAMETERS

- world reference system in general does not coincide with the camera reference frame



$$G = \begin{bmatrix} R & \mathbf{t} \\ \mathbf{0}^\top & 1 \end{bmatrix}$$

$$P = K[I; \mathbf{0}]G$$

equivalent  
formulation

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



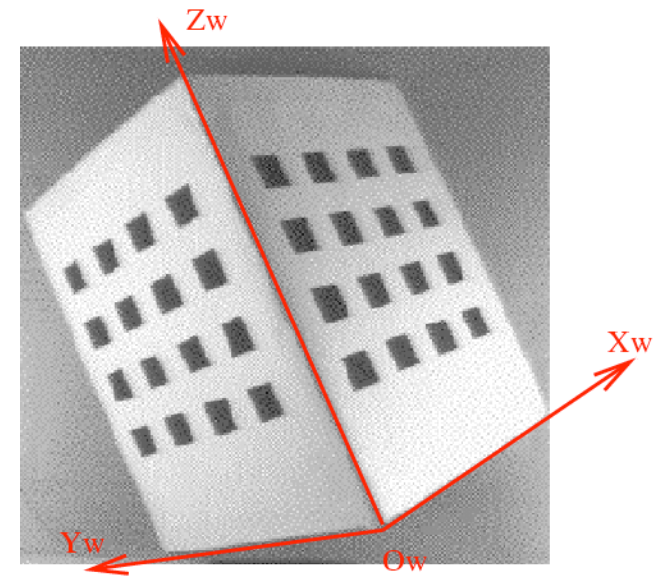
# CAMERA CALIBRATION (WORLD & IMAGE PLANE)

- we start off from 2D - 3D correspondences:

$$\{(\mathbf{x}_i, \mathbf{X}_i)\}_{i=1}^N$$

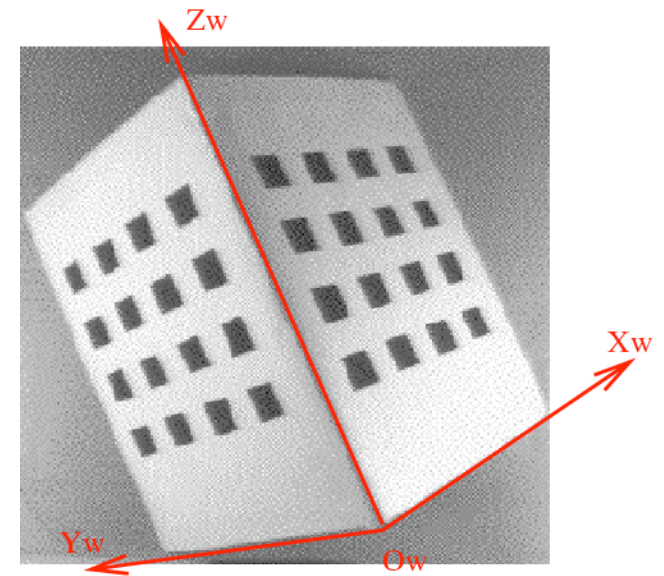
P has 11 d.o.f. then we need  $N \geq 6$

1. We may first estimate the entries of P
2. and then derive the internal and external parameters



# CAMERA CALIBRATION (WORLD & IMAGE PLANE)

- we start off from 2D - 3D correspondences:  $\{(\mathbf{x}_i, \mathbf{X}_i)\}_{i=1}^N$
- **$P$  has 11 d.o.f. then  $N \geq 6$**
- the position of 3D points is known in advance (measured on the pattern)
- the position of 2D points may be estimated for instance by:
  1. canny edge detection
  2. straight line fitting
  3. intersecting the lines to obtain the imaged corners



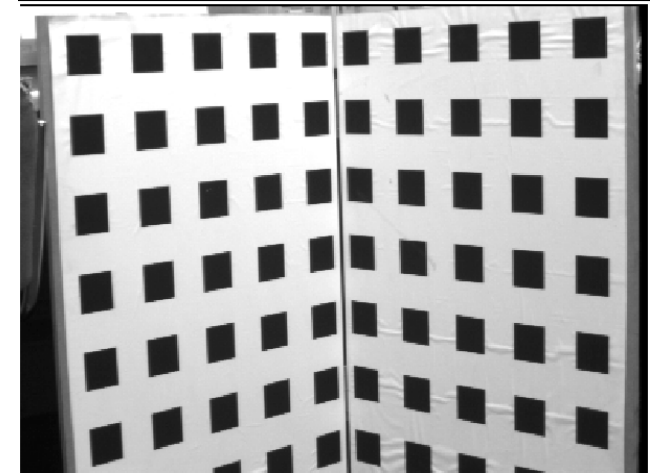
with a careful implementation we may reach very good sub-pixel precisions

# CAMERA CALIBRATION (WORLD & IMAGE PLANE)

- The main steps of a standard procedure (details in the additional material - next slides)
  1. Estimate the projection matrix  $P$  from world-image points correspondences
  2. Extract parameters from the estimated  $P$

# I. ESTIMATION OF THE PROJECTION MATRIX

- we start off from 2D - 3D correspondences:  $\{(\mathbf{x}_i, \mathbf{X}_i)\}_{i=1}^N$
- our goal is to estimate the entries of matrix  $P$  ( $3 \times 4$ ):  $\mathbf{x}_i = P\mathbf{X}_i$



$$\mathbf{x}_i \times P\mathbf{X}_i = \mathbf{0}$$

$$\mathbf{x}_i = (x_i, y_i, w_i)^\top$$

$$P = \begin{bmatrix} \text{p}^1 \\ \text{p}^2 \\ \text{p}^3 \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{0}^\top & -w_i\mathbf{X}_i^\top & y_i\mathbf{X}_i^\top \\ w_i\mathbf{X}_i^\top & \mathbf{0}^\top & -x_i\mathbf{X}_i^\top \\ -y_i\mathbf{X}_i^\top & x_i\mathbf{X}_i^\top & \mathbf{0}^\top \end{bmatrix} \begin{pmatrix} \mathbf{P}^1 \\ \mathbf{P}^2 \\ \mathbf{P}^3 \end{pmatrix} = \mathbf{0}$$

← 2th dim vector of unknowns

# I. ESTIMATION OF THE PROJECTION MATRIX DIRECT LINEAR TRANSFORM (DLT) ALGORITHM

we build a homogeneous system of equations  $\mathbf{A}\mathbf{p} = \mathbf{0}$

$$\mathbf{A} = \begin{bmatrix} \mathbf{0}^\top & -w_1 \mathbf{X}_1^\top & y_1 \mathbf{X}_1^\top \\ w_1 \mathbf{X}_1^\top & \mathbf{0}^\top & -x_1 \mathbf{X}_1^\top \\ \dots & \dots & \dots \\ \mathbf{0}^\top & -w_N \mathbf{X}_N^\top & y_N \mathbf{X}_N^\top \\ w_N \mathbf{X}_N^\top & \mathbf{0}^\top & -x_N \mathbf{X}_N^\top \end{bmatrix}$$
$$\mathbf{p} = \begin{bmatrix} P_{11} \\ P_{12} \\ \vdots \\ P_{33} \\ P_{34} \end{bmatrix} = \begin{bmatrix} \mathbf{P}^1 \\ \mathbf{P}^2 \\ \mathbf{P}^3 \end{bmatrix} \quad \mathbf{P} = \begin{bmatrix} \text{p}^1 \\ \text{p}^2 \\ \text{p}^3 \end{bmatrix}$$

# I. ESTIMATION OF THE PROJECTION MATRIX DIRECT LINEAR TRANSFORM (DLT) ALGORITHM

- P has 11 degrees of freedom (homogeneous transformation)
- with at least 6 correspondences we may obtain a solution
  - with 11 equations or in noise free cases  $\text{rank}(A) = 11$
  - the non trivial solution may be found in the null space of A
- if  $\text{rank}(A)$  is full we can look for a least squares solution

$$\begin{array}{ll} \text{minimize} & ||A\mathbf{x}|| \\ \text{subj to} & ||\mathbf{x}|| = 1 \end{array} \quad \text{it avoids the trivial solution}$$

# DLT SOLUTION OF A [additional material] GENERAL PROBLEM

$$\begin{array}{ll} \text{minimize} & ||A\mathbf{x}|| \\ \text{subj to} & ||\mathbf{x}|| = 1 \end{array}$$

- Let  $A = UDV^T$
- Equivalent problem  $||DV^T\mathbf{x}||$  subj to  $||V^T\mathbf{x}|| = 1$
- Change of variable  $||D\mathbf{y}||$  subj to  $||\mathbf{y}|| = 1$   $\mathbf{y} = V^T\mathbf{x}$
- D is diagonal with elements in descending order, thus a solution that minimizes the norm is  $\mathbf{y} = (0, 0, \dots, 1)^T$
- thus the solution  $\mathbf{x}$  is the last column of V  $\mathbf{x} = V\mathbf{y}$

## 2. MATRIX DECOMPOSITION

- the  $P$  we estimate is a  $3 \times 4$  which we may write as  $P = [M | \mathbf{p}_4]$
- we now want to extract the intrinsic and extrinsic parameters
- we know that  $P = K[R | -R\tilde{C}]$
- $R, K$  can be derived from a QR decomposition of  $M$
- the center can be computed as the null space of  $P$



# THE CAMERA CENTRE

[additional material]

- $P$  has a **1**-dim right null space [ $\text{rank}(P)=3$  but  $P$  has 4 columns]
- therefore there exists a (non trivial) vector  $C$ , so that  $PC=0$
- consider a line containing  $C$  and any other point  $A$  in 3-space  
In parametric coordinates:

$$\mathbf{X}(\lambda) = \lambda A + (1 - \lambda)C$$

$$\mathbf{x} = P\mathbf{X}(\lambda) = \lambda PA + (1 - \lambda)PC = \lambda PA \quad \text{line projection}$$

- all points are mapped to the same image point  $PA$  which means that the line is a ray through the camera centre
- since this holds for every  $A$  it follows that  $C$  must be the camera centre in homogeneous coordinates

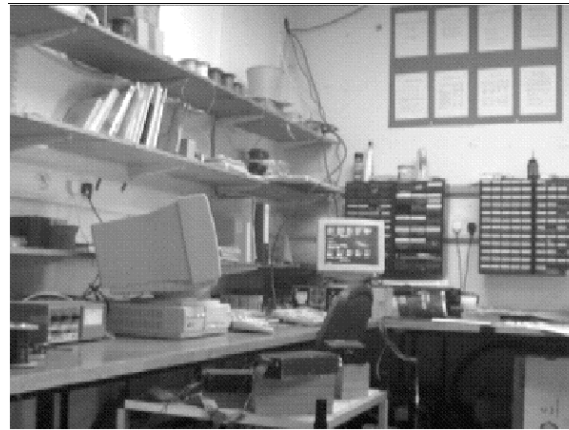
## RESTRICTED CAMERA ESTIMATION

- the decomposition method we describe does not take into account some “reasonable” assumptions on the internal parameters which may be
  - skew is 0
  - pixels are squared
  - principal point is known
  - $K$  is all known
- *there exist algorithms that allow us to impose constraints*

# RADIAL DISTORTION

[additional material]

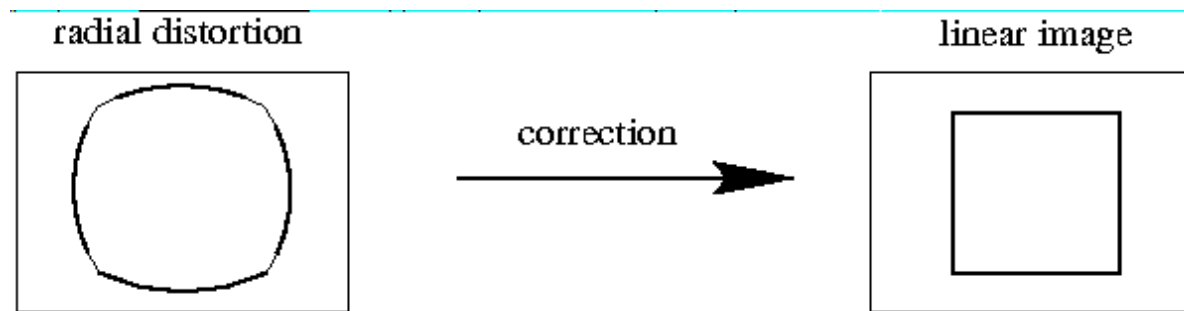
- the linear camera model discussed so far is not perfect for the real world
- short focal length and cheap cameras show the non linearity quite clearly



long focal length



short focal length



radial distortion correction allows us to obtain a corrected image “as if”  
generated by a linear model

# RADIAL DISTORTION

[additional material]

$$\begin{aligned}\hat{x} &= x_c + L(r)(x - x_c) \\ \hat{y} &= y_c + L(r)(y - y_c)\end{aligned}\quad r^2 = (x - x_c)^2 + (y - y_c)^2$$

distortion centre measured corrected

distortion function

- *L may be written as a polynomial approximation*

$$L(r) = 1 + k_1 r + k_2 r^2 + k_3 r^3 + \dots$$

- *the coefficients may be included in the calibration process (including the distortion centre which is not necessarily the principal point)*

# WRAPPING UP

- Camera calibration is the process of estimating camera parameters from “ad hoc” images
- It allows us to (geometrically) relate the images acquired from the camera with the real 3D world
- Many available algorithms...
  - today we discussed a classical procedure (estimate  $P$  and then extract the parameters)