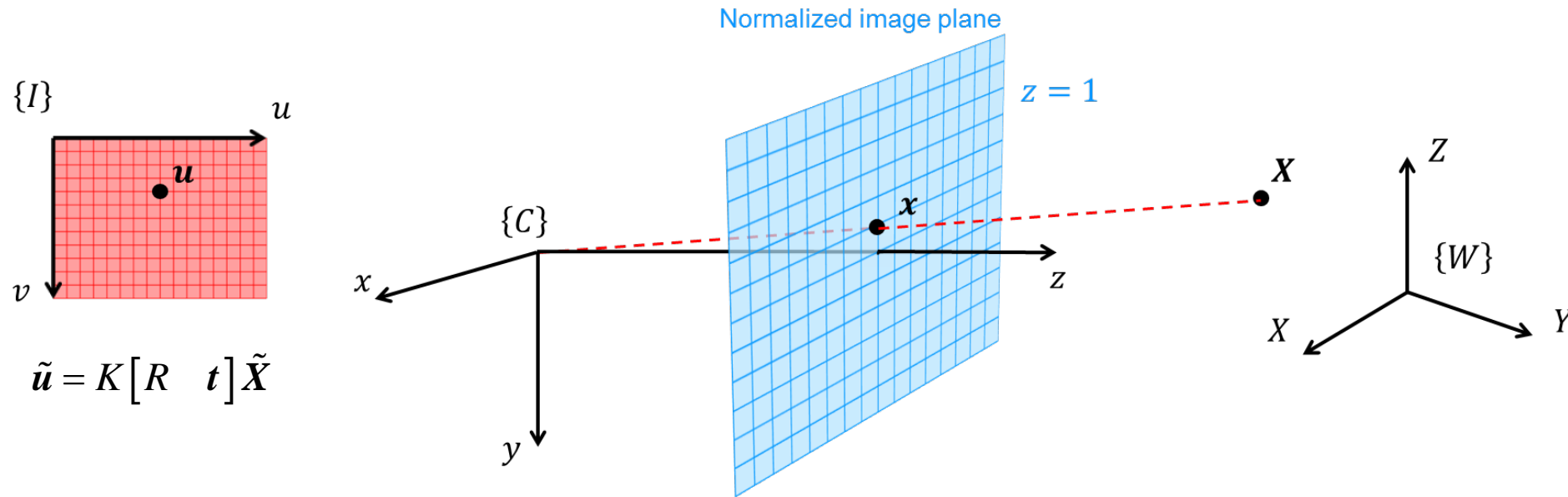


# **Lecture 5.1**

## **Camera calibration**

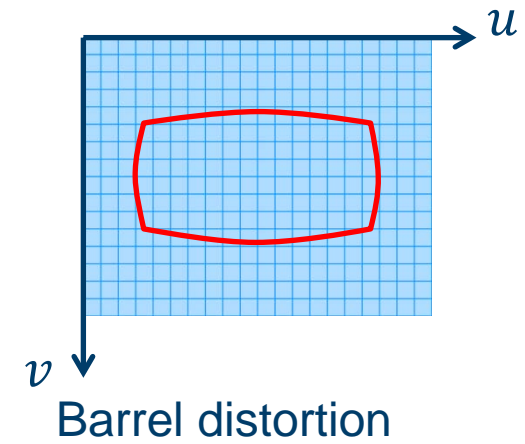
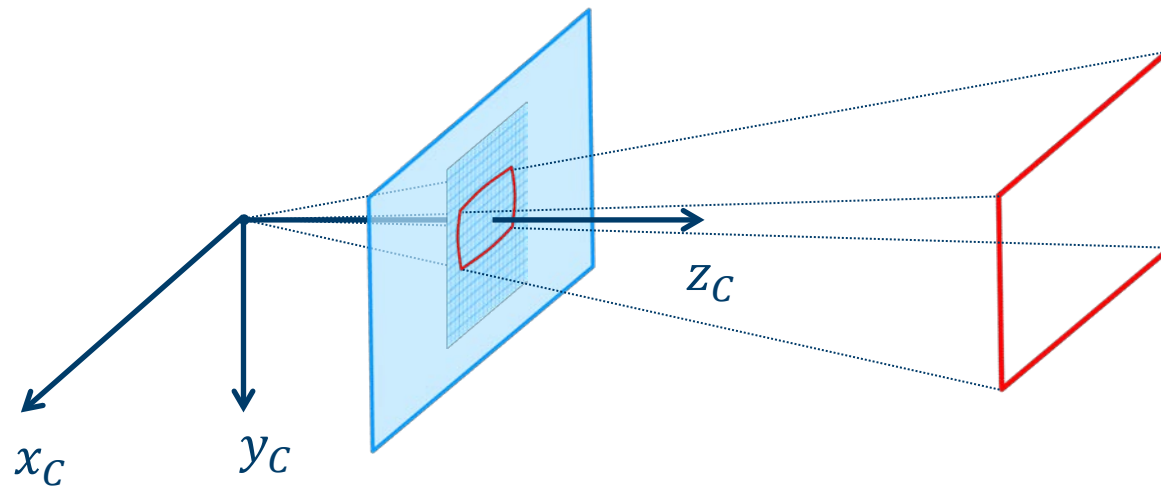
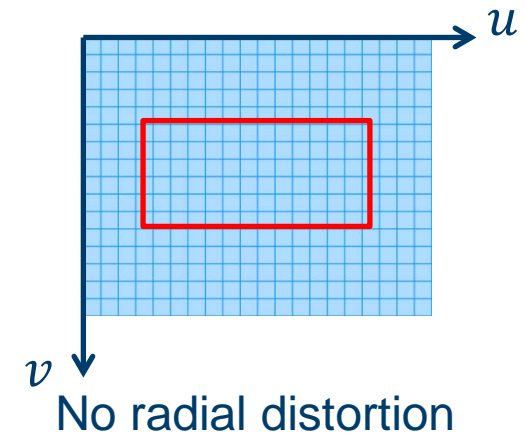
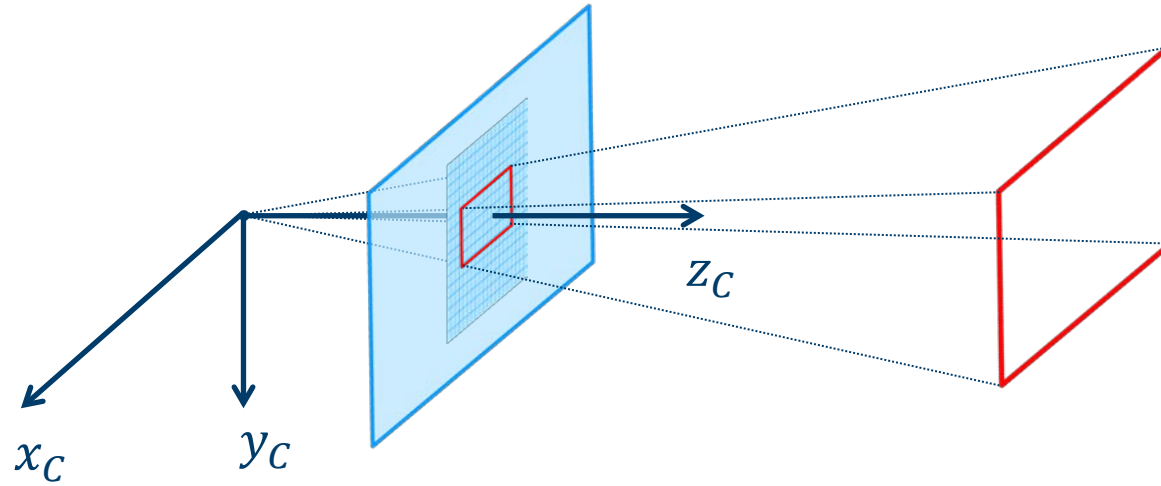
Thomas Opsahl

# Introduction

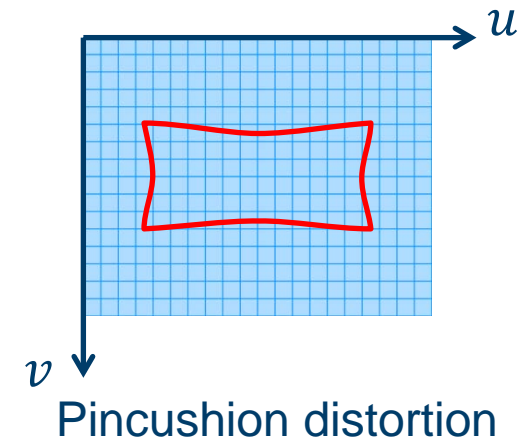
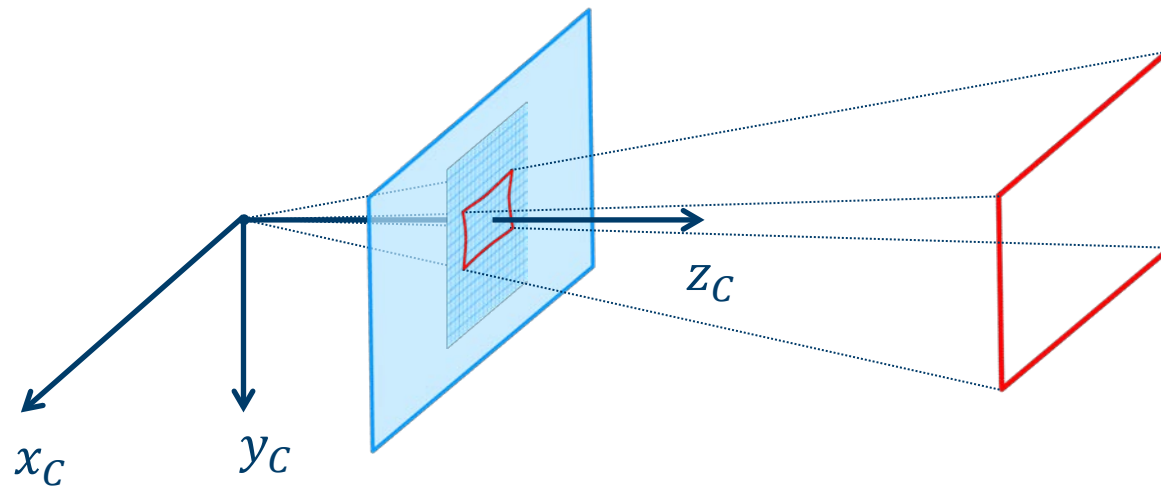
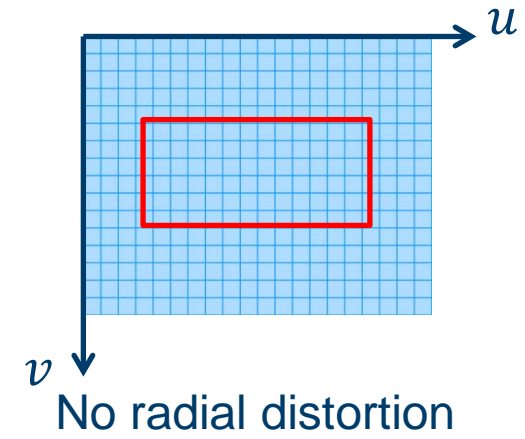
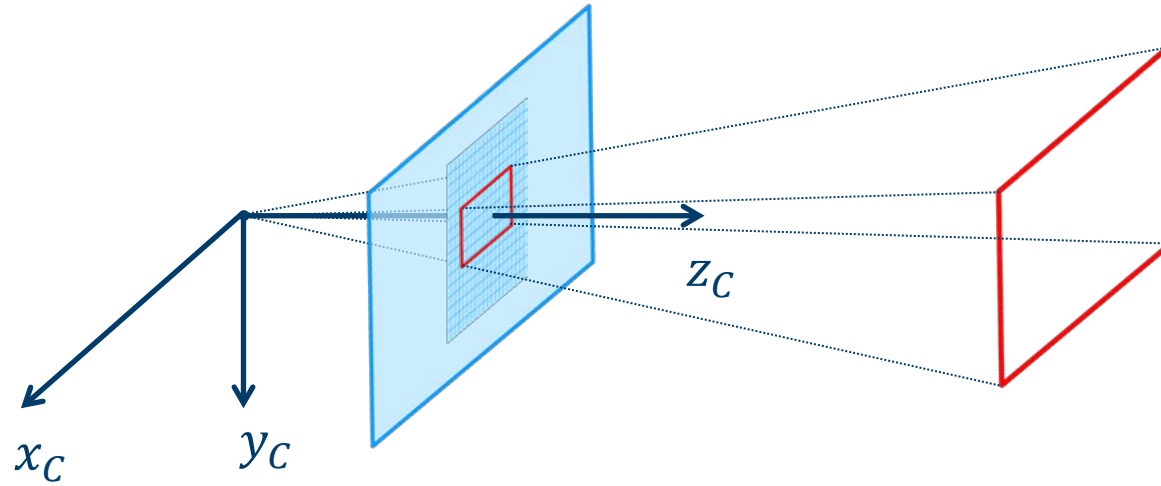


- The perspective camera model describes a 3D to 2D transformation that is consistent with the pinhole geometry
- No cameras fit this model perfectly – They all suffer from some kind of distortion
- If we want to use images for geometrical computations we need to take this distortion into account

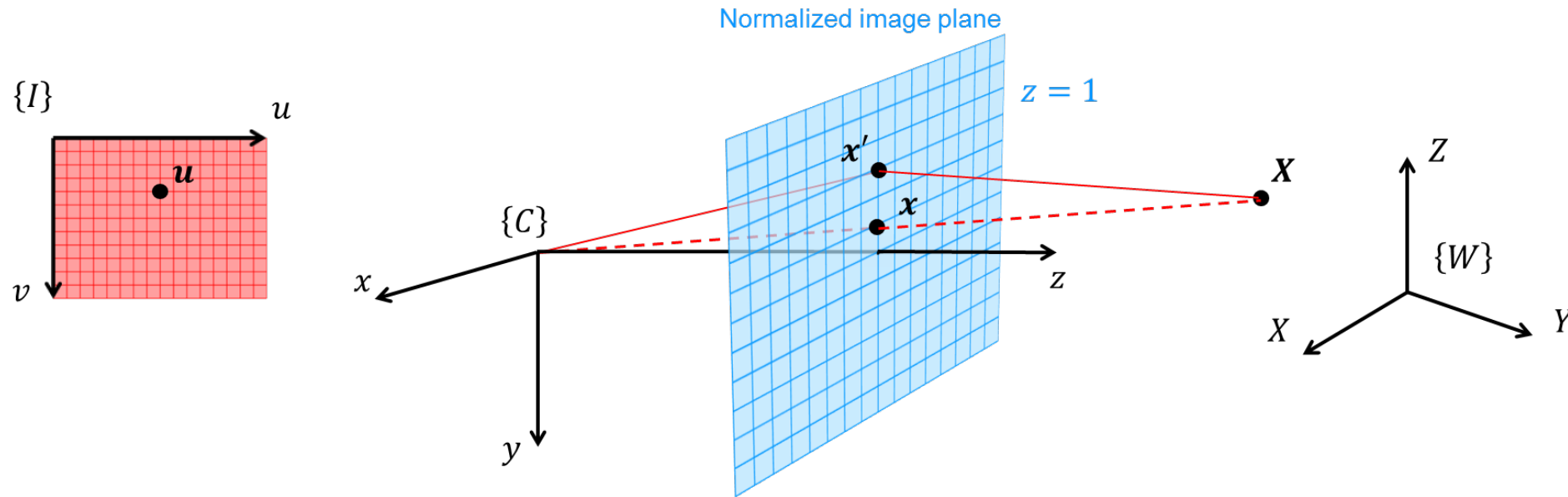
# Introduction



# Introduction



# Undistortion



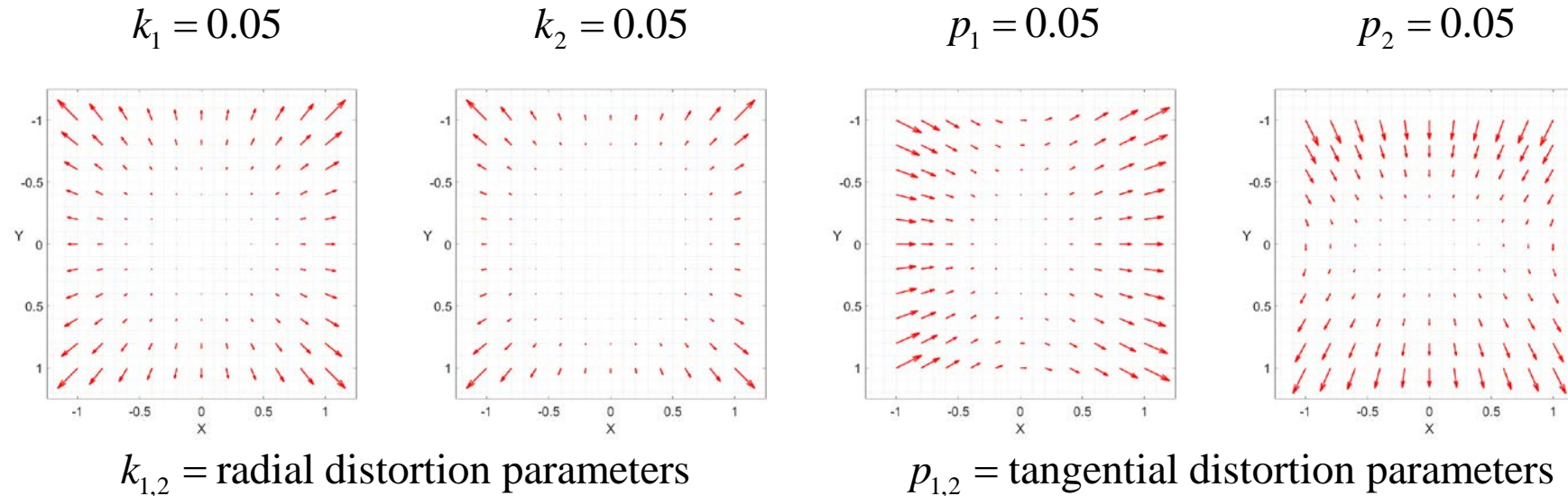
- A distortion model describes the relationship between undistorted coordinates  $x$  and distorted coordinates  $x'$  of the normalized image plane
- This example model describes both radial distortion and tangential distortion

$$x = x' \left( 1 + k_1 r^2 + k_2 r^4 \right) + 2p_1 x' y' + p_2 \left( r^2 + 2x'^2 \right)$$

$$y = y' \left( 1 + k_1 r^2 + k_2 r^4 \right) + p_1 \left( r^2 + 2y'^2 \right) + 2p_2 x' y'$$

where  $r^2 = x'^2 + y'^2$

# Undistortion

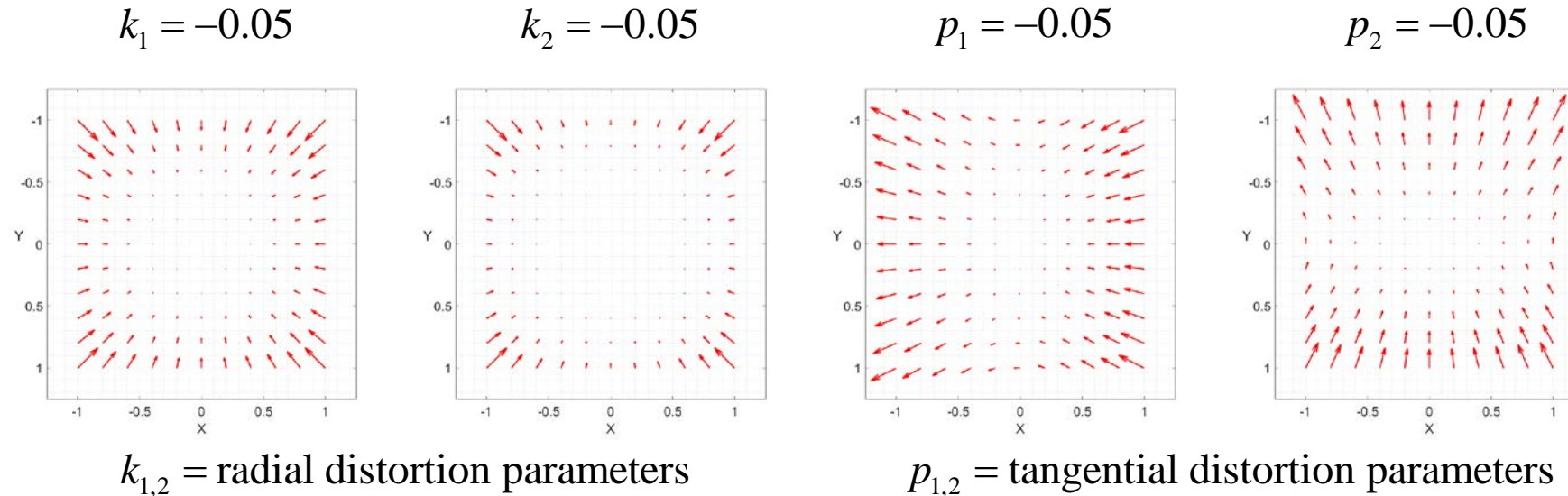


- A distortion model describes the relationship between undistorted coordinates  $x$  and distorted coordinates  $x'$  of the normalized image plane
- This example model describes both radial distortion and tangential distortion

$$\begin{aligned}x &= x' \left( 1 + k_1 r^2 + k_2 r^4 \right) + 2 p_1 x' y' + p_2 \left( r^2 + 2 x'^2 \right) \\y &= y' \left( 1 + k_1 r^2 + k_2 r^4 \right) + p_1 \left( r^2 + 2 y'^2 \right) + 2 p_2 x' y'\end{aligned}$$

where  $r^2 = x'^2 + y'^2$

# Undistortion



- A distortion model describes the relationship between undistorted coordinates  $x$  and distorted coordinates  $x'$  of the normalized image plane
- This example model describes both radial distortion and tangential distortion

$$\begin{aligned} x &= x' \left( 1 + k_1 r^2 + k_2 r^4 \right) + 2p_1 x'y' + p_2 \left( r^2 + 2x'^2 \right) \\ y &= y' \left( 1 + k_1 r^2 + k_2 r^4 \right) + p_1 \left( r^2 + 2y'^2 \right) + 2p_2 x'y' \end{aligned} \quad \text{where } r^2 = x'^2 + y'^2$$

# Undistortion

Original image



$$\tilde{u} \neq K \begin{bmatrix} R & t \end{bmatrix} \tilde{X}$$

undistortion



Undistorted image



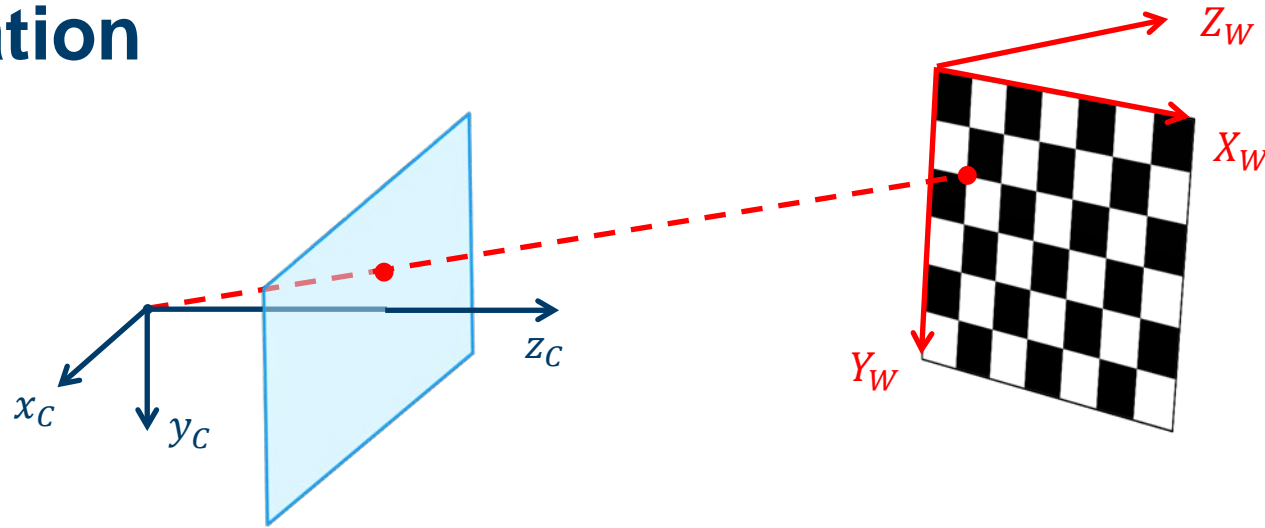
$$\tilde{u} = K \begin{bmatrix} R & t \end{bmatrix} \tilde{X}$$

- We can use the distortion model to warp the original image into the so called undistorted image
- The undistorted image satisfy the perspective camera model and are thus well suited for geometrical computations
- Since the distortion model depends on  $K$  for the undistorted camera, it is custom to estimate both in a common calibration process

Images: <http://www.robots.ox.ac.uk/~vgg/hzbook/>

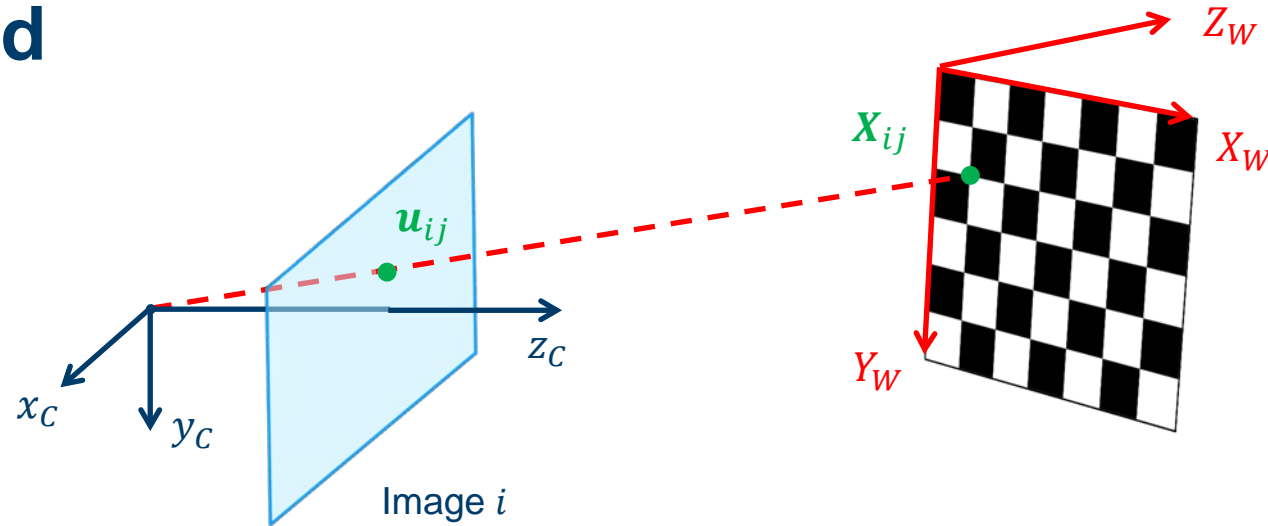


# Camera calibration



- Camera calibration is a process where we estimate the intrinsic parameters  $f_u$ ,  $f_v$ ,  $s$ ,  $c_u$ ,  $c_v$  and distortion parameters for a camera
- One of the most commonly used calibration algorithms was described by Zhengyou Zhang in the paper "*A Flexible New Technique for Camera Calibration*" in 2000
- Zhang's method is based on using a planar calibration object, e.g. a chessboard where we know the size of the tiles

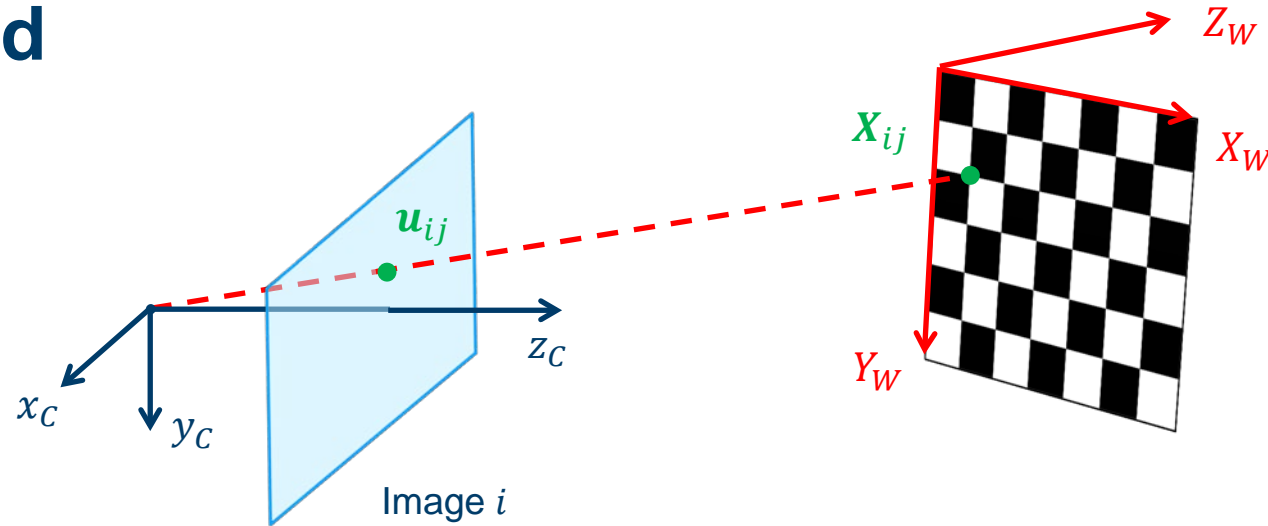
# Zhang's method



1. Capture multiple images (at least 3) of the planar calibration object
2. For each image, estimate the homography  $H$  between the 2D surface of the calibration object and the image
3. Based on these homographies, estimate  $K$  (DLT) and from this all  $R_i$ 's and  $t_i$ 's
4. Use the estimated parameters as the starting point of an iterative non-linear optimization of the full set of intrinsic parameters (including distortion parameters  $\kappa$ )

$$\min_{K, \kappa, R_i, t_i} \sum_i \sum_j \left\| u_{ij} - \hat{u}(K, \kappa, R_i, t_i; X_{ij}) \right\|^2$$

# Zhang's method

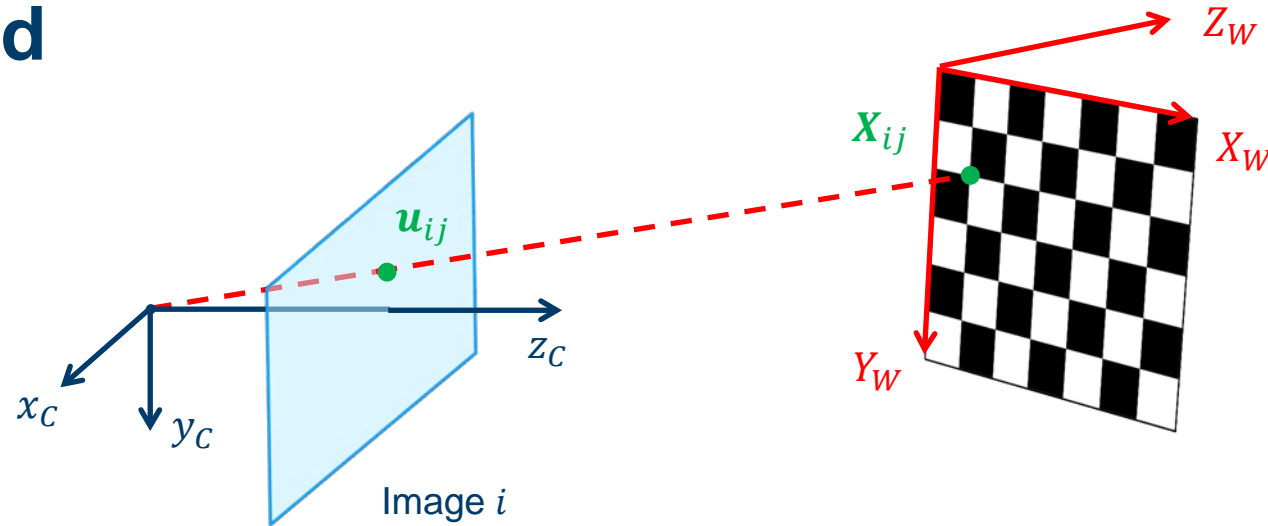


- Given that the calibration object is planar, the 3D-2D relationship between points on the calibration object and points in the image is described by a homography

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \underbrace{\begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{r}_3 & \mathbf{t} \end{bmatrix}}_{[R \quad t]} \begin{bmatrix} X \\ Y \\ 0 \\ 1 \end{bmatrix} \Rightarrow \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix} \Rightarrow \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = H \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

- We can estimate the homography  $H$  from a minimum of 4 point correspondences between the image and the calibration object

# Zhang's method



- Since  $\mathbf{r}_1$  and  $\mathbf{r}_2$  are columns of a rotation matrix, they impose 2 constraints on the elements of  $H$

$$\left. \begin{array}{l} \mathbf{r}_1^T \mathbf{r}_2 = 0 \\ \mathbf{r}_1^T \mathbf{r}_1 = \mathbf{r}_2^T \mathbf{r}_2 \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} \mathbf{h}_1^T K^{-T} K^{-1} \mathbf{h}_2 = 0 \\ \mathbf{h}_1^T K^{-T} K^{-1} \mathbf{h}_1 = \mathbf{h}_2^T K^{-T} K^{-1} \mathbf{h}_2 \end{array} \right.$$

- From  $N$  images we get  $N$  homographies, giving rise to  $2N$  constraints allowing us to solve for the different elements of  $K$  using linear methods (DLT)

# Summary



- Undistortion
  - For geometrical computations we work on undistorted images/feature points
- Calibration
  - Estimate  $K$  + distortion model parameters from images
  - Zhang's method
  - OpenCV, Matlab, Kalibr (<https://github.com/ethz-asl/kalibr>)
- Additional reading
  - Szeliski: 6.3
- Optional reading
  - *A flexible new technique for camera calibration*, by Z. Zhang