

Camera Calibration

Readings in Vision, Graphics
and Interactive Systems

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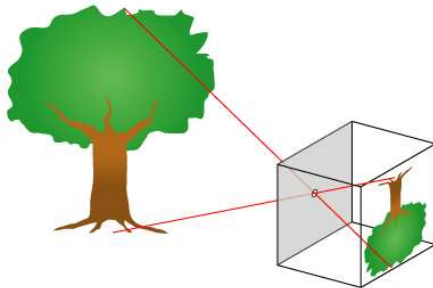
October 16, 2009

Introduction

Camera Models

Camera Calibration

A Picture Says More Than a Thousand Words



What is Camera Calibration?

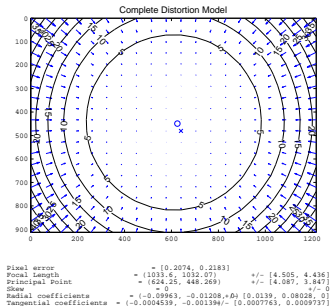
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What is Camera Calibration?

- Camera calibration “is the process of finding the true parameters of the camera that produced a given photograph or video,” *Wikipedia*
- Two sets of parameters are determined:
 - *Intrinsic* parameters
 - *Extrinsic* parameters

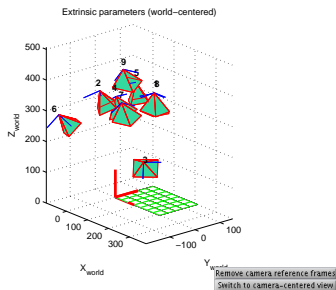
Camera Calibration Parameters

- **Intrinsic parameters**
- *how does the camera depict the scene?*



Camera Calibration Parameters

- **Intrinsic parameters**
- *how does the camera depict the scene?*
- **Extrinsic parameters**
- *where is the camera located?*



Motivation and Applications

- Calibration is necessary whenever we want to measure anything accurately from images.
- Accurate measurements are vital for many computer vision systems.

Motivation and Applications

- Calibration is necessary whenever we want to measure anything accurately from images.
- Accurate measurements are vital for many computer vision systems.
- Applications include:
 - Augmented reality
 - Robots picking up objects
 - Aerial photography

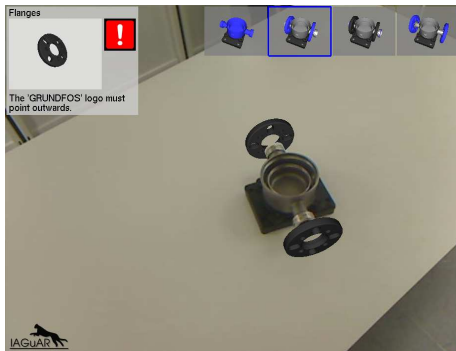
Augmented Reality

ARToolKit



Augmented Reality

Interactive Assembly Guide using Augmented Reality



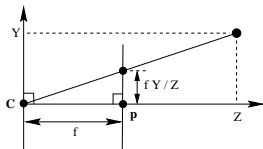
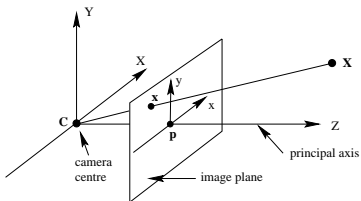
Robots Picking Up Objects

Bin Picking



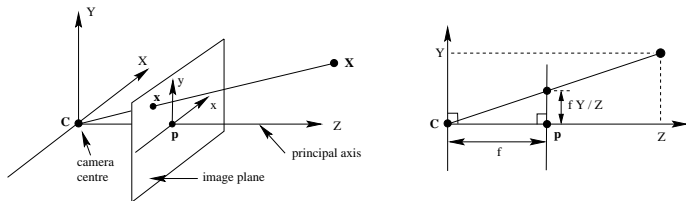
Camera Models

Basic Pinhole Model



- Central projection of points in space onto the image plane.
- Centre of projection is $\mathbf{C} = \mathbf{O}$, and image plane is $Z = f$.

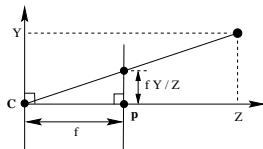
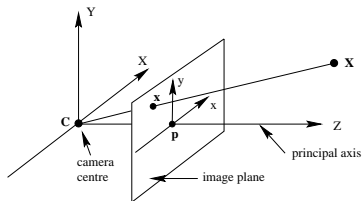
Basic Pinhole Model



- Central projection of points in space onto the image plane.
- Centre of projection is $\mathbf{C} = \mathbf{O}$, and image plane is $Z = f$.
- A point in space $\mathbf{X} = [X \ Y \ Z]^T$ is mapped to the image plane at the intersection with the line from \mathbf{C} to \mathbf{X} . By similar triangles

$$\mathbf{X} = [X \ Y \ Z]^T \mapsto [fX/Z \ fY/Z]^T = \mathbf{x}.$$

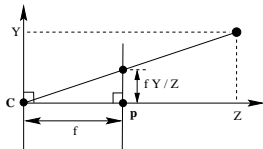
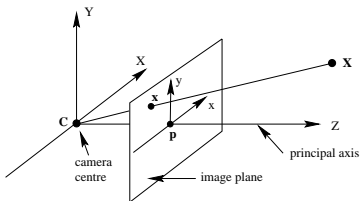
Central Projection using Homogeneous Coordinates



- Representing world and image coordinates by homogeneous vectors, this becomes a linear mapping:

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

Central Projection using Homogeneous Coordinates

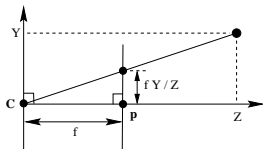
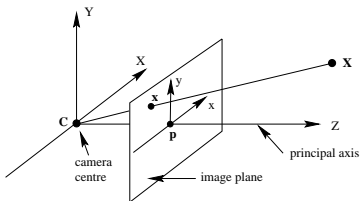


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- In homogeneous coordinates $[fX \ fY \ Z]^T \sim [fX/Z \ fY/Z \ 1]^T$.

Central Projection using Homogeneous Coordinates



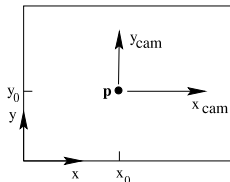
- With $\mathbf{X} = [X \ Y \ Z \ 1]^T$, and $\mathbf{x} = [fX \ fY \ Z]^T$, we have the *camera projection matrix*

$$P = \begin{bmatrix} f & & 0 \\ & f & 0 \\ & & 1 & 0 \end{bmatrix} = \text{diag}(f, f, 1)[I | \mathbf{0}].$$

- The mapping from world to image coordinates is now simply

$$\mathbf{x} = P\mathbf{X}.$$

Principal Point Offset



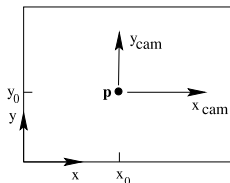
- In general the origin of coordinates in the image plane is not the principal point, $\mathbf{p} = [p_x \ p_y]^T$. Thus there is a mapping

$$[X \ Y \ Z]^T \mapsto [fX/Z + p_x \ fY/Z + p_y]^T.$$

- Expressed in homogeneous coordinates, this becomes:

$$\begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} \mapsto \begin{bmatrix} fX + Zp_x \\ fY + Zp_y \\ Z \end{bmatrix} = \begin{bmatrix} f & p_x & 0 \\ & f & p_y \\ & & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Principal Point Offset



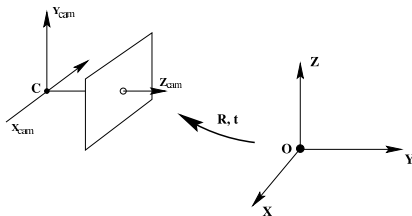
- Letting $P = K[I | \mathbf{0}]$, where K is the *camera calibration matrix*

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

- the mapping from camera space \mathbf{X}_{cam} to image coordinates \mathbf{x} becomes:

$$\mathbf{x} = P\mathbf{X}_{cam} = K[I | \mathbf{0}]\mathbf{X}_{cam}$$

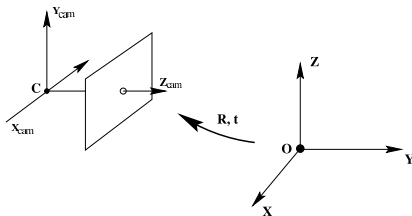
Camera Rotation and Translation



- Points are expressed in the *world coordinate frame*, which is related to the *camera coordinate frame* via a rotation R (3×3 matrix) and a translation \mathbf{t} .
- In homogeneous coordinates, this becomes:

$$\mathbf{X}_{\text{cam}} = \begin{bmatrix} R & \mathbf{t} \\ \mathbf{0} & 1 \end{bmatrix} \mathbf{x}$$

Camera Rotation and Translation



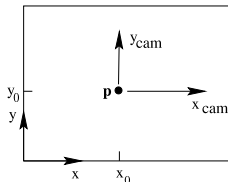
- This leads to a camera projection matrix of the form

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

- A direct mapping from world to image coordinates is now

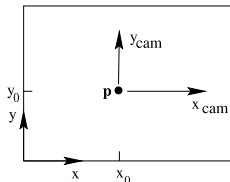
$$\mathbf{x} = P\mathbf{X} = K[R | \mathbf{t}]\mathbf{X}.$$

CCD Cameras



- In the previous slides, image coordinates were assumed to have equal scale on both axes.
- CCD cameras may have non-square pixels, and if image coordinates are measured in pixels, this changes the calibration matrix.

CCD Cameras



- Let m_x and m_y be the number of pixels per unit distance in the x and y directions, then

$$K = \begin{bmatrix} \alpha_x & & x_0 \\ & \alpha_y & y_0 \\ & & 1 \end{bmatrix},$$

where $\alpha_x = f m_x$, and $\alpha_y = f m_y$ represent the focal length in pixel dimensions. Similarly $x_0 = m_x p_x$, and $y_0 = m_y p_y$ represent the principal point in pixel dimensions.

Finite Projective Camera

- For added generality, a skew parameter s is included, and finally

$$K = \begin{bmatrix} \alpha_x & s & x_0 \\ & \alpha_y & y_0 \\ & & 1 \end{bmatrix} .$$

- If pixels in the CCD array are skewed, $s \neq 0$ meaning the x - and y -axes are not perpendicular.
- In general $s = 0$ for CCD cameras.

Lens Distortion

- A real camera may also have lens distortion, which can not be described by the camera projection matrix.
- This distortion is described by 5 distortion coefficients represented as a column vector \mathbf{d} .

Lens Distortion

- A real camera may also have lens distortion, which can not be described by the camera projection matrix.
- This distortion is described by 5 distortion coefficients represented as a column vector \mathbf{d} .
- Letting $\mathbf{X}_{\text{cam}} = [X_c \ Y_c \ Z_c]^T$, we have the normalized (pinhole) image projection:

$$\mathbf{x}_n = \begin{bmatrix} X_c/Z_c \\ Y_c/Z_c \end{bmatrix} = \begin{bmatrix} x_n \\ y_n \end{bmatrix}$$

Lens Distortion

- With $r^2 = x_n^2 + y_n^2$, the distorted normalized image coordinate becomes:

$$\mathbf{x}_d = \begin{bmatrix} x_d \\ y_d \end{bmatrix} = (1 + d_1 r^2 + d_2 r^4 + d_5 r^6) \mathbf{x}_n + \mathbf{dx},$$

- where \mathbf{dx} is the tangential distortion vector:

$$\mathbf{dx} = \begin{bmatrix} 2d_3 x_n y_n + d_4 (r^2 + 2x_n^2) \\ d_3 (r^2 + 2y_n^2) + 2d_4 x_n y_n \end{bmatrix}$$

Lens Distortion

- Finally, the pixel coordinate $\mathbf{x}_p = [x_p \ y_p]^T$ of the projection of \mathbf{X}_{cam} when distortion is applied is:

$$\begin{bmatrix} x_p \\ y_p \\ 1 \end{bmatrix} = \mathbf{K} \begin{bmatrix} x_d \\ y_d \\ 1 \end{bmatrix}$$

- In other words,

$$\mathbf{x}_p = \begin{bmatrix} \alpha_x x_d + s y_d + x_0 \\ \alpha_y y_d + y_0 \end{bmatrix}.$$

Summary of Camera Models

- For *finite projective cameras* points in space \mathbf{X} are mapped onto image coordinates \mathbf{x} through the linear transformation

$$\mathbf{x} = \mathbf{P}\mathbf{X},$$

where \mathbf{P} is the 3×4 *camera projection matrix*.

- Defining the *camera calibration matrix*

$$\mathbf{K} = \begin{bmatrix} \alpha_x & s & x_0 \\ & \alpha_y & y_0 \\ & & 1 \end{bmatrix},$$

we can express \mathbf{P} as:

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

where \mathbf{R} and \mathbf{t} are the rotation and translation transforming points from world to camera space.

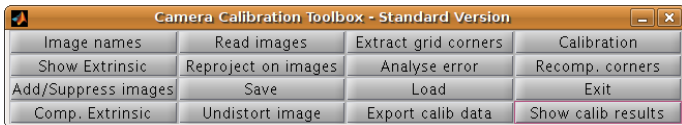
Summary of Camera Models

- The *intrinsic* parameters of the camera are:
 - the calibration matrix K , and
 - the lens distortion parameters $d_i, i = 1, 2, \dots, 5$.
- The *extrinsic* parameters are:
 - the rotation matrix R , and
 - the translation vector \mathbf{t} .

Camera Calibration

Camera Calibration Toolbox for Matlab

- Recall that we want to find the true parameters of the camera, i.e. the intrinsic and extrinsic parameters.
- This process is greatly simplified using the Camera Calibration Toolbox for Matlab.
- The following is a minimal introduction on how to use the toolbox. Really good tutorials are available on its homepage.



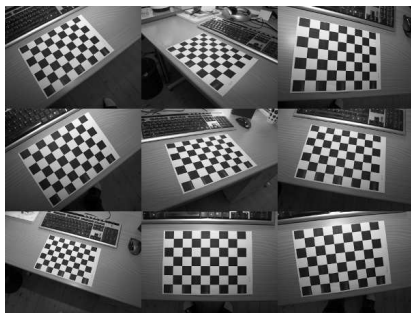
Intrinsic Calibration

Steps involved:

1. Load input images with calibration pattern.
2. Extract grid corners on all images.
3. Compute calibration parameters.
4. Inspect the result.

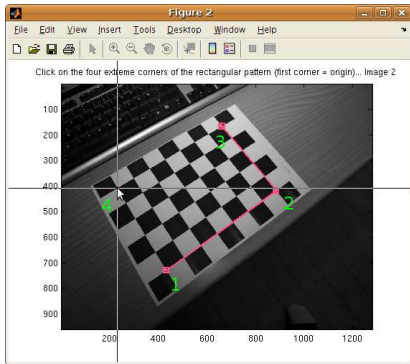
Loading Calibration Images

- Best results are achieved with more images, e.g. 20, of the calibration pattern from different angles.
- Images are loaded using the *Image names* button in the GUI of the calibration toolbox.
- In this example 9 images are loaded.



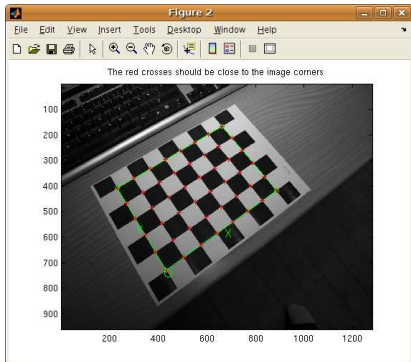
Extracting Grid Corners

- Press *Extract grid corners* in the GUI.
- Click the four extreme grid corners (first click defines the origin).



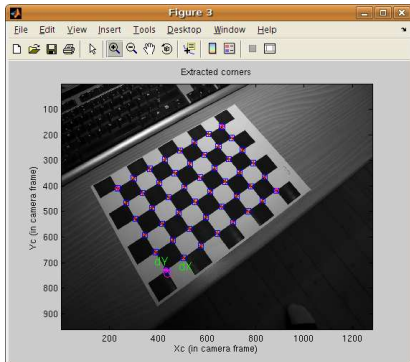
Extracting Grid Corners

- Press *Extract grid corners* in the GUI.
- Click the four extreme grid corners (first click defines the origin).
- Gussed corner locations are marked.



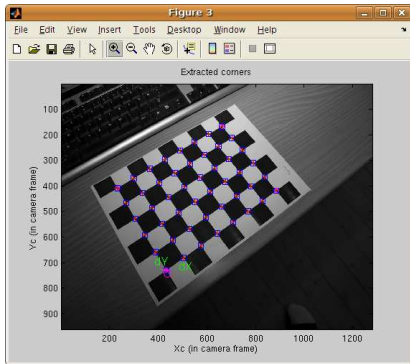
Extracting Grid Corners

- Press *Extract grid corners* in the GUI.
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- Gussed corner locations are marked.
- Optimized corner locations are found.



Extracting Grid Corners

- Press *Extract grid corners* in the GUI.
- Click the four extreme grid corners (first click defines the origin).
- Gussed corner locations are marked.
- Optimized corner locations are found.
- Repeat for all images.



Computing Calibration Parameters

- Press the *Calibration* button in the GUI to start calibration.

```

Aspect ratio optimized (est_aspect_ratio = 1) -> both components of fc are estimated (fc = [ 1007.79386 1007.79386 ])
Principal point optimized (center_optim=1) - (DEFAULT). To reject principal point, set center_optim=0.
Skew not optimized (est_alpha=0) - (DEFAULT)
Distortion not fully estimated (defined by the variable est_dist):
  Sixth order distortion not estimated (est_dist(5)=0) - (DEFAULT) .
Initialization of the principal point at the center of the image.
Initialization of the intrinsic parameters using the vanishing points of planar patterns.

```

Initialization of the intrinsic parameters - Number of images: 9

Calibration parameters after initialization:

```

Focal Length:      fc = [ 1007.79386  1007.79386 ]
Principal point:   cc = [ 639.50000  479.50000 ]
Skew:             alpha_c = [ 0.00000 ] => angle of pixel axes = 90.00000 degrees
Distortion:       kc = [ 0.00000  0.00000  0.00000  0.00000  0.00000 ]

```

Main calibration optimization procedure - Number of images: 9

```

Gradient descent iterations: 1...2...3...4...5...6...7...8...9...10...11...12...13...14...15...16...17...18...19...20...
Estimation of uncertainties...done

```

Calibration results after optimization (with uncertainties):

```

Focal Length:      fc = [ 1033.59588  1032.06788 ] ± [ 4.50528  4.43636 ]
Principal point:   cc = [ 624.24952  448.26889 ] ± [ 4.08749  3.84683 ]
Skew:             alpha_c = [ 0.00000 ] ± [ 0.00000 ] => angle of pixel axes = 90.00000 degrees
Distortion:       kc = [ -0.09963  -0.01208  -0.00045  -0.00139  0.00000 ] ± [ 0.00000  0.00000  0.00000  0.00000  0.00000 ]
Pixel error:      err = [ 0.20743  0.21826 ]

```

Note: The numerical errors are approximately three times the standard deviations (for the 95% confidence interval).

Calibration Results

- After calibration, the intrinsic parameters of the camera are available.
- The variable `KK` contains the camera calibration matrix:

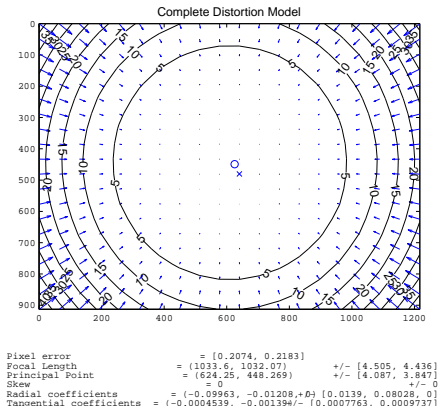
$$\mathbf{KK} = \begin{bmatrix} 1033.6 & 0 & 624.25 \\ 0 & 1032.1 & 448.27 \\ 0 & 0 & 1 \end{bmatrix}$$

- The distortion coefficients are available in `kc`:

$$\mathbf{kc} = [-0.0996 \quad -0.0121 \quad -0.0005 \quad -0.0014 \quad 0]^T$$

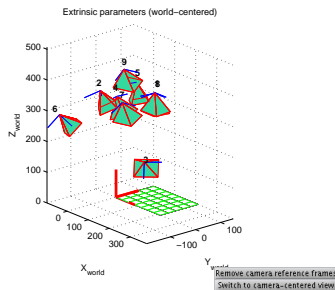
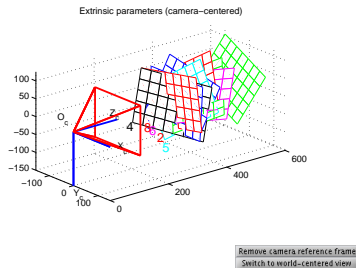
Calibration Results

- Visualizing the effect of distortions on the pixel image (run the script *visualize_distortions* from the Matlab prompt):



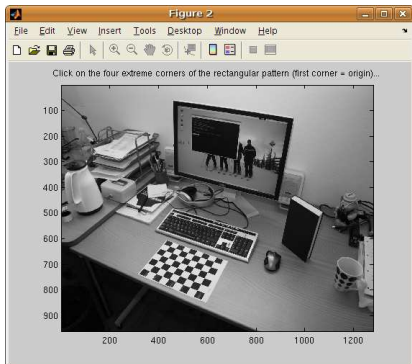
Calibration Results

- Visualization of extrinsic parameters for calibration images (press *Show extrinsic* in the GUI):



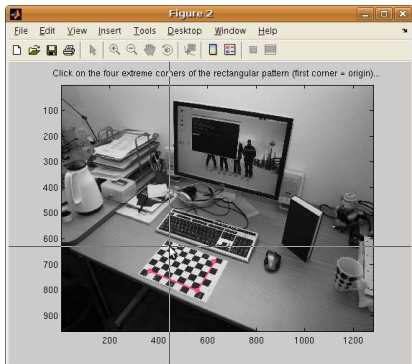
Extrinsic Calibration

- Once the intrinsic parameters of the camera have been found, extrinsic parameters can be computed for other images from the same camera.
- Press the *Comp. extrinsic* button in the GUI.



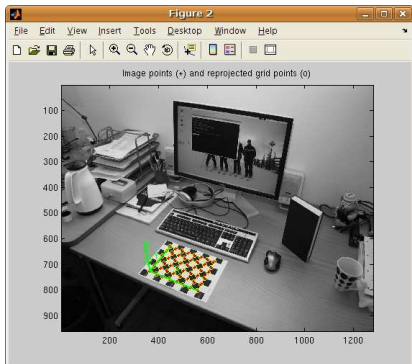
Extrinsic Calibration

- Once the intrinsic parameters of the camera have been found, extrinsic parameters can be computed for other images from the same camera.
- Press the *Comp. extrinsic* button in the GUI.
- As before, click the four extreme corners of the grid.



Extrinsic Calibration

- Once the intrinsic parameters of the camera have been found, extrinsic parameters can be computed for other images from the same camera.
- Press the *Comp. extrinsic* button in the GUI.
- As before, click the four extreme corners of the grid.
- The result is shown as an overlay on the image.



Extrinsic Calibration

- Recall that the extrinsic parameters are a rotation matrix R , and a translation vector \mathbf{t} .
- The variable `Rc_ext` now contains the rotation matrix:

$$Rc_ext = \begin{bmatrix} 0.9145 & 0.4046 & 0.0093 \\ 0.2263 & -0.4921 & -0.8406 \\ -0.3355 & 0.7708 & -0.5415 \end{bmatrix}$$

- The translation vector is available in `Tc_ext`:

$$Tc_ext = [-250.2 \ 267.5 \ 980.4]^T$$

Complete Calibration of a Single Camera

- Having computed both the intrinsic and extrinsic parameters of the camera, the mapping between world and image coordinates is known.
- Using the finite projective camera model, the camera projection matrix of the calibrated camera becomes:

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

- Disregarding lens distortion, the world coordinate \mathbf{X} is mapped to the image coordinate \mathbf{x} by the linear transformation

$$\mathbf{x} = P\mathbf{X}.$$

Image Rectification

- Using the toolbox, images can be rectified (undistorted). Here is an example:

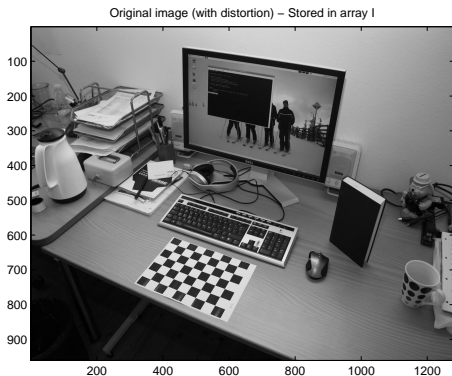
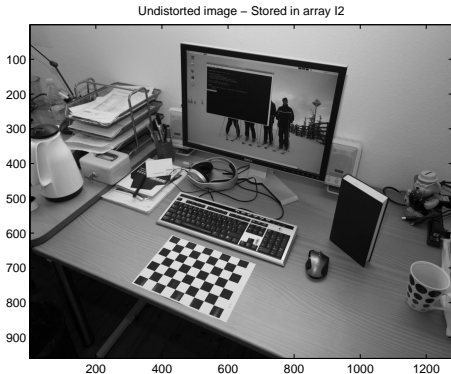


Image Rectification

- Using the toolbox, images can be rectified (undistorted). Here is an example:



Summary of Camera Calibration

- Estimation of the true parameters of a camera.
- Intrinsic parameters are computed based on a set of calibration images.
 - Camera calibration matrix is stored in **KK**.
 - Distortion coefficients are stored in **kc**.
- Afterwards extrinsic parameters can be calculated for other images from the same camera.
 - Rotation matrix is stored in **Rc_ext**.
 - Translation is stored in **Tc_ext**.

Any questions?

References

- Multiple View Geometry in Computer Vision (2nd Edition), *Richard Hartley and Andrew Zisserman*, chapters 6 and 9.
- Camera Calibration Toolbox for Matlab,
http://www.vision.caltech.edu/bouguetj/calib_doc/
- Camera Resectioning (Calibration), *Wikipedia*,
http://en.wikipedia.org/wiki/Camera_calibration