

Accident avoidance by active intervention for Intelligent Vehicles

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Multi Sensor Calibration

Tobias Hanning interactIVe Summer School 4-6 July, 2012

Agenda

- Multiple Sensors in Driver Assistance Systems
- Example: Camera calibration
 - Classic camera calibration
 - Camera calibration as multi-sensor calibration.
- Applications: (Optical) Multi-Sensor Systems
 - Calibration of Two Imaging Sensors (Stereo Camera Calibration)
 - Calibration of an Imaging Sensor to a 3D-Sensor
- Conclusion



Multi Sensor Calibration Overview

- Multiple Sensors In Automotive Applications:
 - Camera(s)
 - Lidar
 - Ultrasonic sensors
 - Accelerometer
 - and many more..

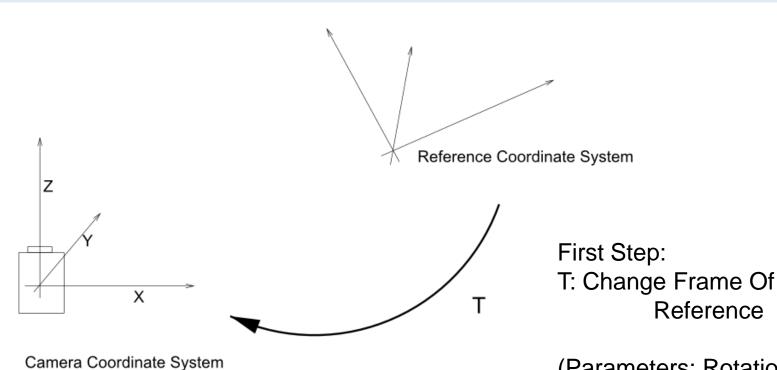




Multi Sensor Calibration Overview

- Objective: Find parameters of each sensor
- Objective in Talk: Find pose and orientation of each sensor
 - Example: Camera Calibration as Multi-Sensor Calibration (meaning each pixel is a sensor on its own)
 - Therefore: A (very) short introduction to camera calibration
 - A new view at camera calibration as multi-sensor calibration
 - Transfer of results
 - Stereo Camera Calibration
 - Calibration of a 3D-Sensor wrt Grayvalue Camera

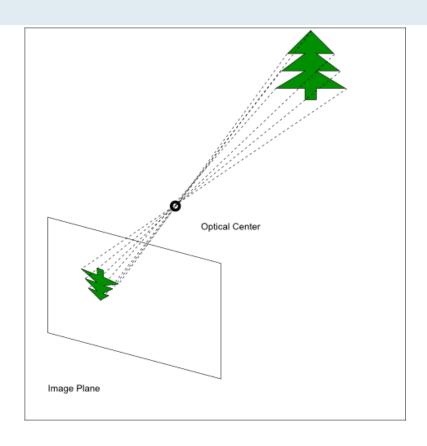




(Parameters: Rotation R and Translation t)

Reference

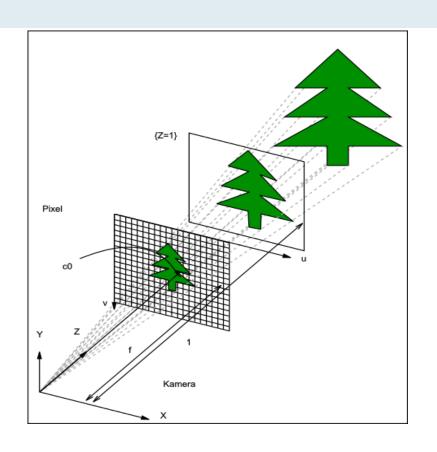




Second Step: Projection

- Modelled as pinhole projection
- Operation: Division by focal length
- But: Image is upside down





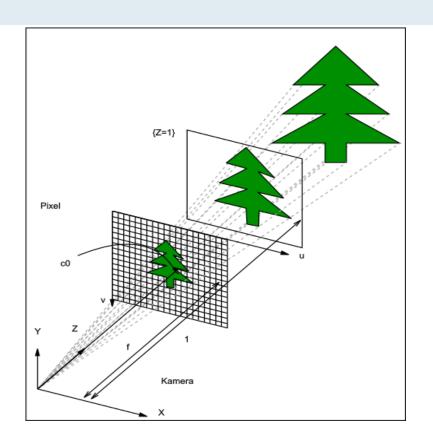
Second Step: Projection

- Project onto (virtual) plane {z=1}
- Conversion to Image Coordinates:
 - Shift
 - Scale

$$\Pi: \begin{pmatrix} X \\ Y \\ Z \end{pmatrix} \rightarrow \begin{pmatrix} X/Z \\ Y/Z \end{pmatrix}$$

$$I: \begin{pmatrix} u \\ v \end{pmatrix} \rightarrow \begin{pmatrix} \alpha & 0 \\ 0 & \beta \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} + \begin{pmatrix} u_0 \\ v_0 \end{pmatrix}$$





Overall Camera Mapping

$$C = I \circ \Pi \circ T$$

Camera Calibration:

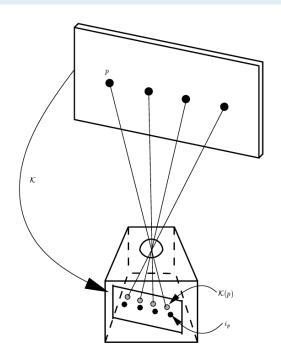
Input:

- 1. Set P of points in 3D
- 2. Observation i_p for every p in P

Task:

Find $(R, t, \alpha, \beta, u_0, v_0)$ minimizing

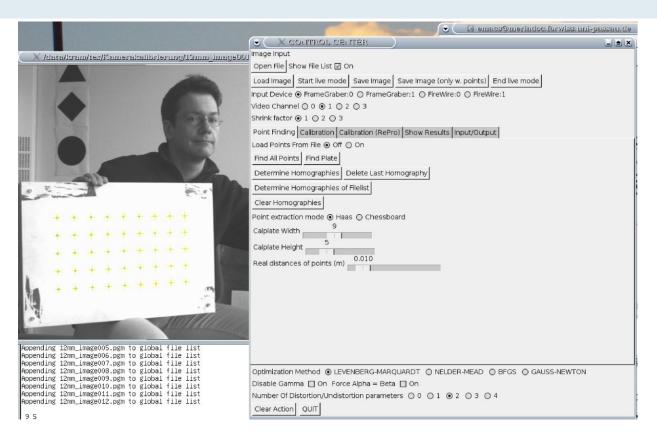




$$(R, t, \alpha, \beta, u_0, v_0) \mapsto \sum_{p \in P} \left\| i_p - I \circ \Pi \circ T(p) \right\|^2$$

"Minimize the distance between model projection and observation"





Sucessful Calibration: Prototype projected in input image



Summary: Classic Camera Calibration

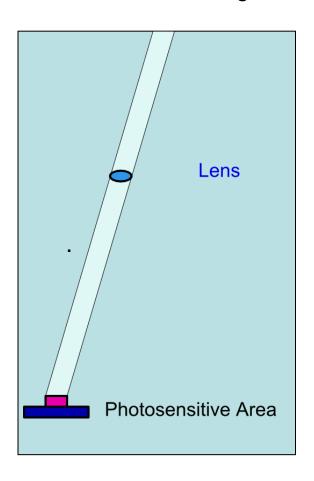
Camera is modelled as pinhole camera
Calibration input on point to point base
Sensor (= Imager) not mentioned in model

Let's take a different point of view:

Each pixel is a sensor on its own!



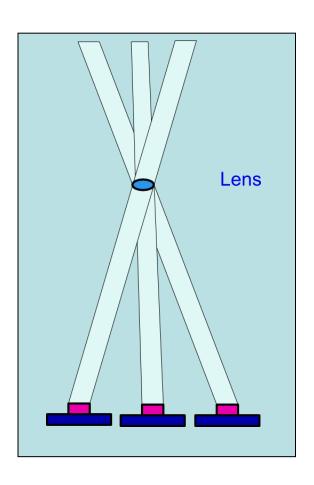
Camera Pixel As Single Sensor



- Sensor measures intensity of light
- Together with lens:
 - Measuring intensity of a ray of light



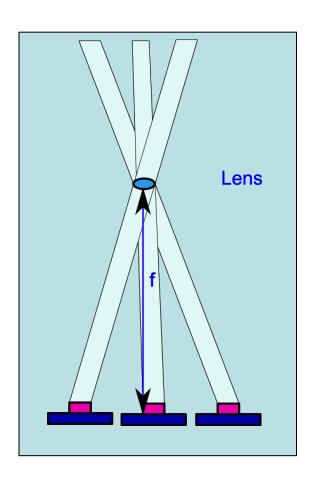
Camera Modelled As Multi-Sensor



- Camera = Grid of Sensors
- Parameters:
 - Position and Orientation wrt lens
- Constraints:
 - All rays share one point (optical sensor)
 - Sensors are arranged in fixed grid (perpendicular to the optical axis)
- Camera parameters



Camera Modelled As Multi-Sensor

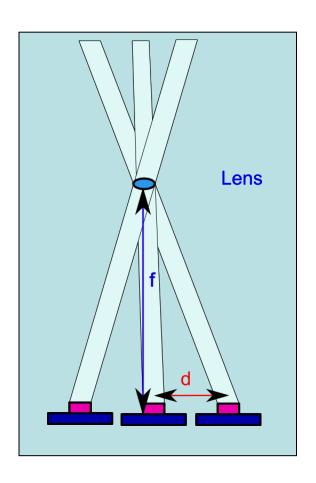


Parameters:

Distance to Lens f



Camera Modelled As Multi-Sensor

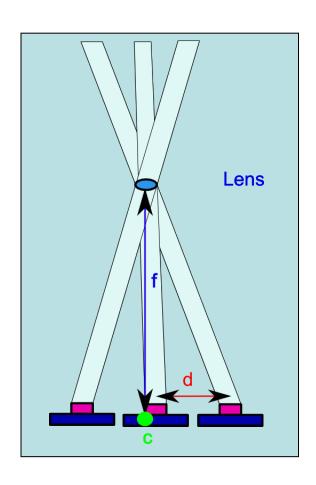


Parameters:

- Distance to Lens f
- Distance between Sensors d



Camera Modelled As Multi-Sensor

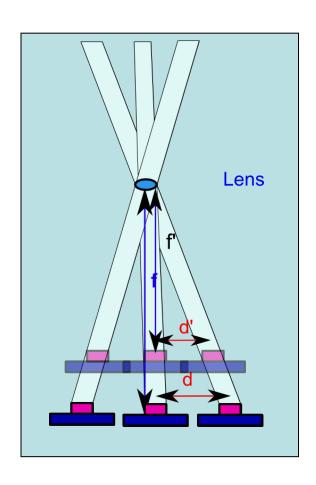


Parameters:

- Distance to Lens f
- Distance between Sensors d
- Point of orthogonal Projection of Lens c



Camera Modelled As Multi-Sensor



Parameters:

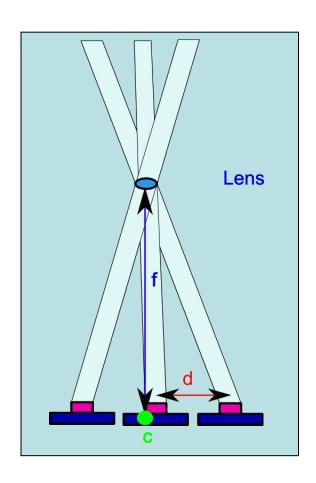
- Distance to Lens f
- Distance between Sensors d
- Point of orthogonal Projection of Lens c

But: Any f'/d' with f'/d' = f/d will deliver the same sensor!

Thus: Not f,d are not free parameters, but f/d is!



Camera Modelled As Multi-Sensor



Link to classic camera parametrization:

$$\frac{f}{d_{u}} \cong \alpha$$

$$\frac{f}{d_{v}} \cong \beta$$

$$c \cong \begin{pmatrix} u_{0} \\ v_{0} \end{pmatrix}$$

(R,t)



Scene Interpretation





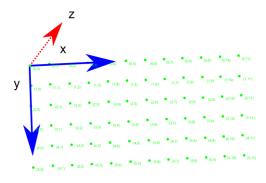
Scene Interpretation





Scene Interpretation:

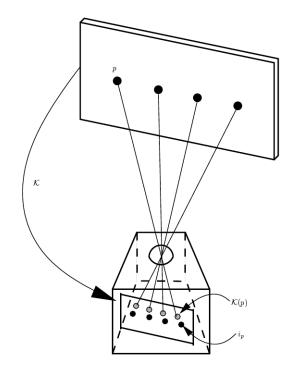
- Each sensor shares an subset of the same interpretation of the scene
- Sensor may not observe the whole scene
- Scene introduces new parameters (e.g. position of the calibration target)





Multi-Sensor Calibration:

- Determine parameters
 - of the scene
 - of the Multi-Sensor-Array
- such that distance of the
 - interpretation of the observation
 - and the interpretation of the scene
- becomes minimal





Further Topics of Classic Camera Calibration:

(Not covered in this talk:)

- (Radial) Distortion ("fish-eye")
- Sensor Correlations (e.g. blurring introduced by lens)
- Error Functions
- Initial Solution (Projective Geometry)
- Optimization Techniques



Applications: Multi-sensor systems

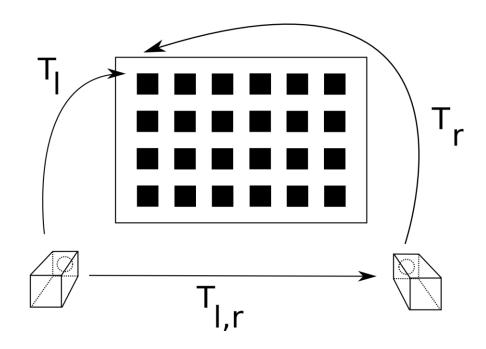
Examples:

- Homogeneous Multi-Sensor-System: Stereo Camera
- Heterogeneous Multi-Sensor-System: 3D Camera to Mono-Camera

In both examples the sensors are related by rotation and translation



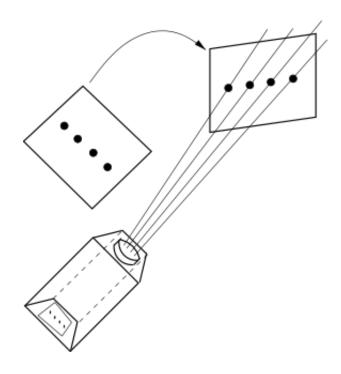
Stereo-Camera-Calibration



Left Camera and right camera observe the same scene



Stereo-Camera-Calibration

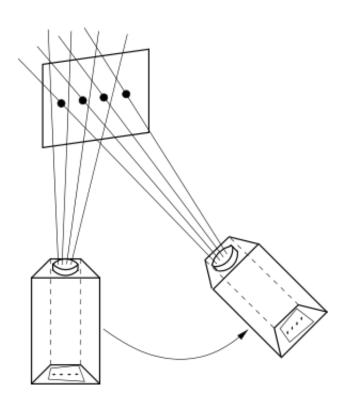


Left Camera:

Determine position of observed calibration pattern wrt left camera



Stereo-Camera-Calibration

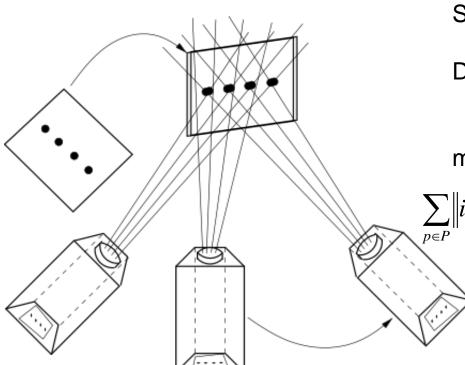


Right Camera:

Determine position of right camera wrt observed calibration pattern



Stereo-Camera-Calibration



Stereo Camera Calibration:

Determine parameters

$$(R_l, t_l, \alpha_l, \beta_l, u_{0,l}, v_{0,l}, R_r, t_r, \alpha_r, \beta_r, u_{0,r}, v_{0,r})$$

minimizing

$$\sum_{p \in P} ||i_{p,l} - I_l \circ \Pi \circ T_l(p)||^2 + ||i_{p,r} - I_r \circ \Pi \circ T_r(p)||^2$$

 $i_{p,l}$ observation of p in left image $i_{p,r}$ observation of p in right image

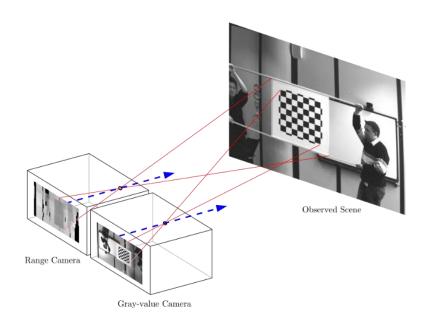


Stereo-Camera-Calibration

- Scene interpretation by two sensors of same kind
- Calibration: Determine parameters of both sensors such that the observation matches the interpretation of the model



3D-Camera and 2D-Camera



- 2D Camera and
- 3D Camera observe
- the same scene

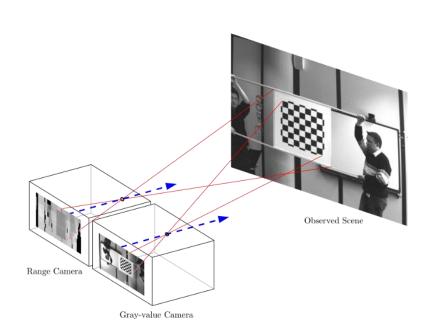
Sensors are related by Rotation and Translation

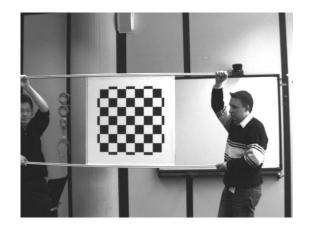


- Output: Image-like distance information (64 × 8-pixel matrix, 42° horizontal, 16° vertical angular resolution)
- Distance measurements in outdoor environments up to 25m
- Automotive and safety-applications capable



3D-Camera and 2D-Camera





Scene observed by grayvalue camera



Scene observed by 3D-Camera



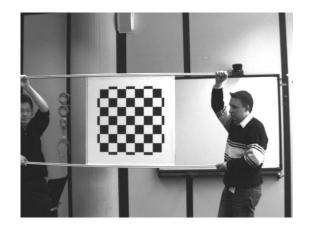
3D-Camera and 2D-Camera

Canonical scene interpretation of both sensors differ:

- Light intensity per pixel
- Depth information per pixel

Task:

Find an interpretation of the scene that applies for both sensors!



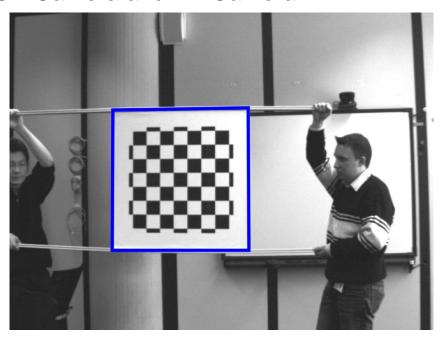
Scene observed by grayvalue camera



Scene observed by 3D-Camera



3D-Camera and 2D-Camera



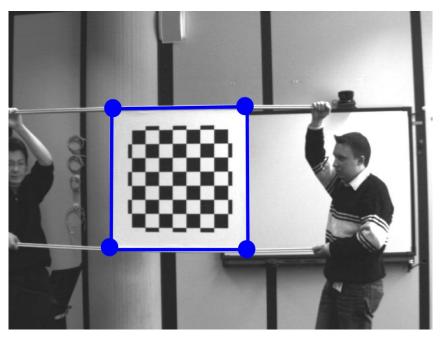
Shared Scene Interpretation:

Plane (with defined borders) at fixed position





3D-Camera and 2D-Camera



Shared Scene Interpretation:

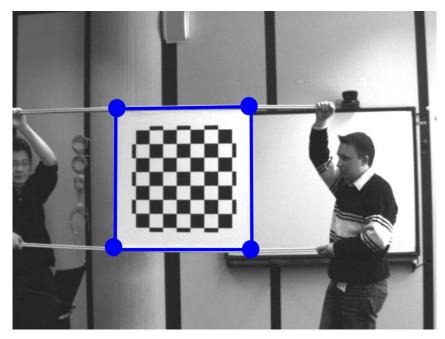
Points determining borders of plane

Calibration: Determine Rotation R and translation t that maps 3D points on 3D points





3D-Camera and 2D-Camera



Monocular Prototype Reconstruction:

Input:

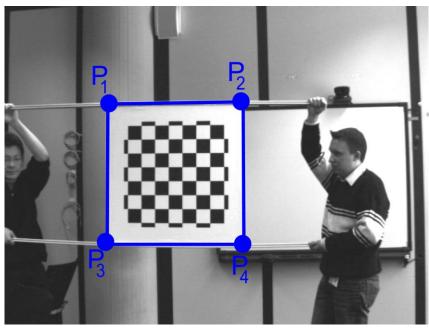
- Set P of four points in 3D (planar)
- 2. Observation i_p for every p in
- 3. Calibrated Camera C Task:

Determine (R₀,t₀) minimizing

$$\sum_{p \in P} \left\| C(Rp + t) - i_p \right\|^2$$



3D-Camera and 2D-Camera



 $(R_{0,t_0}) = \underset{(R,t)}{\operatorname{arg \, min}} \sum_{p \in P} ||C(Rp+t) - i_p||^2$

Monocular Prototype Reconstruction:

Result:

$$P_1 = R_0 p'_1 + t_0$$

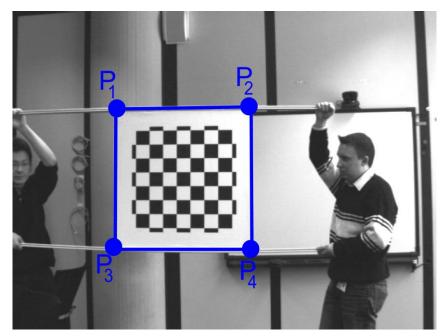
 $P_2 = R_0 p'_2 + t_0$

$$P_3 = R_0 p'_3 + t_0$$

$$P_4 = R_0 p'_4 + t_0$$



3D-Camera and 2D-Camera



Result of Object Segmentation in 3D Camera:

Q

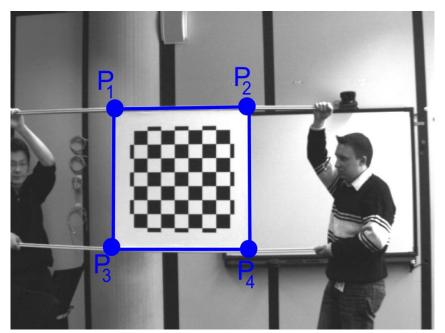
 Q_2

 Q_3

 Q_4



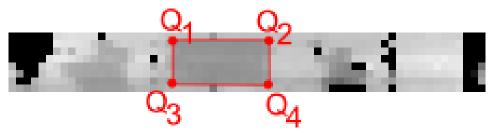
3D-Camera and 2D-Camera



Relative Position of 2D-Camera to 3D-Camera:

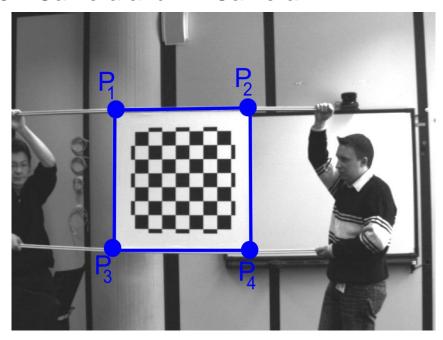
Determine

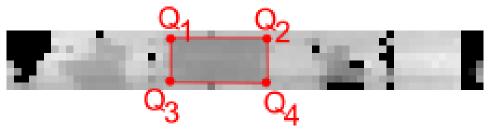
$$\underset{(R,t)}{\arg\min} \sum_{1}^{4} ||RQ_{i} + t - P_{i}||^{2}$$





3D-Camera and 2D-Camera





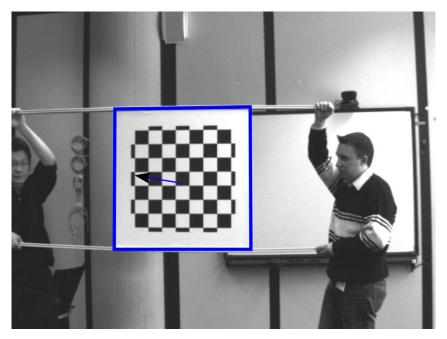
But:

Acquisition of 3D point information by 3D-Camera is not reliable!

- Low Resolution
- Noise
- Difficult Feature Extraction in 3D
- → Weak Correspondencies!



3D-Camera and 2D-Camera



Shared Scene Interpretation:

More reliable: Plane information

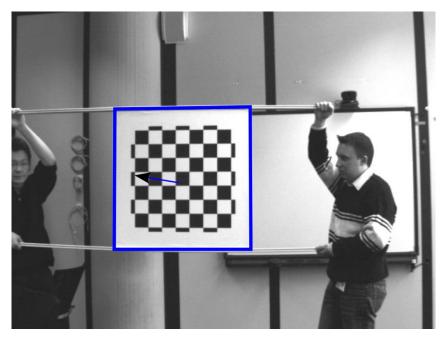
Idea: Represent plane by point and vector

Calibration: Determine Rotation R and translation t that maps 3D plane to 3D plane





3D-Camera and 2D-Camera



Plane matching:

Input:

- Normal n_{3d}, distance d_{3d} of plane obtained by 3D-Camera
- Normal n_{2d}, distance d_{2d} of plane obtained by 3D-Camera

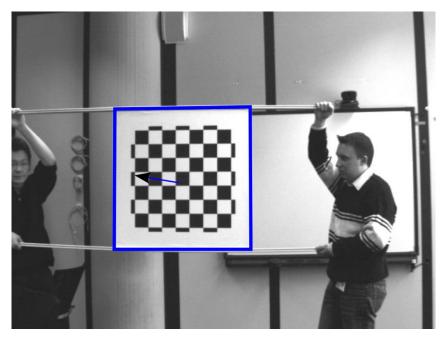
Output: (R₀,t₀) minimizing

$$\sum_{planes} w_n \left\langle Rn_{3d}, n_{2d} \right\rangle^2 + w_d \left| d_{3d} + t - d_{2d} \right|^2$$





3D-Camera and 2D-Camera



Plane matching:

Attention:

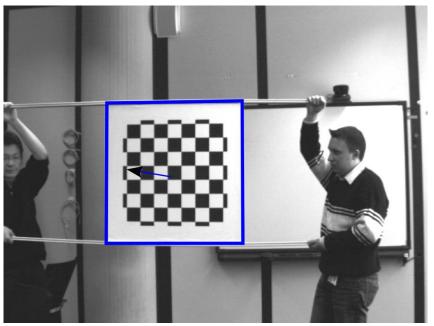
If one of the weights w_n or w_d is zero, the problem looses uniqueness. In that case we obtain rather a constraint than an optimization problem!

$$\sum_{planes} w_n \langle Rn_{3d}, n_{2d} \rangle^2 + w_d |d_{3d} + t - d_{2d}|^2$$





3D-Camera and 2D-Camera



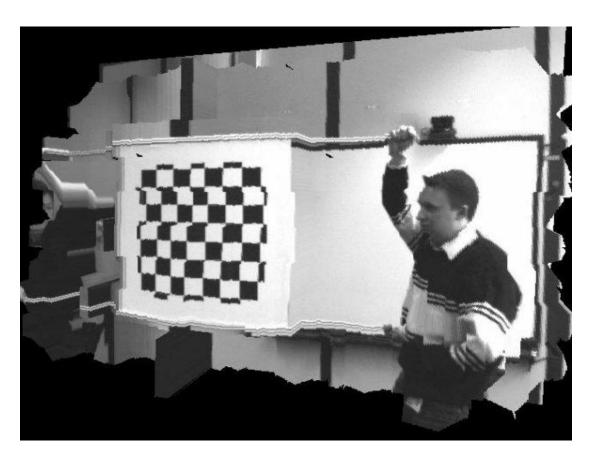
Algorithm:

- 1. Determine initial solution by point matching (closed form!)
- 2. Refine solution by plane matching (non-linear)

$$\underset{(R,t)}{\arg\min} \sum_{planes} w_n \langle Rn_{3d}, n_{2d} \rangle^2 + w_d |d_{3d} + t - d_{2d}|^2$$



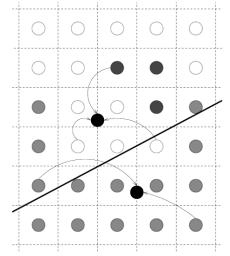
3D-Camera and 2D-Camera: Result



Application:

Integrated Depth Image

Depth information is supplied by segmentation of 2D image





Conclusion

- Multi-Sensor Calibration means sharing the same interpretation of the observed scene
- Shared Interpretation may mean applying a function resulting in an uncommon interpretation by the sensor
- The shared interpretation may only supply a necessary condition (constraint)
- Applications: Multi-sensor systems
 - Calibration of Two Imaging Sensors (Stereo Camera Calibration) modelled as multi-sensor arrays
 - Calibration of an Imaging Sensor to a 3D-Sensor



interactive (interactive)

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Thank you.

Tobias Hanning
University of Passau
Innstr. 43
D-94033 Passau

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SEVENTH FRAMEWORK PROGRAMME