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Challenging vision tasks meeting depth sensing: an in-depth look

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Research pages: <http://ivs.ait.ac.at>

Short intro – who are we in 20 seconds

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Contents

- Motivate & stimulate
- Algorithms through applied examples

Optical flow driven motion analysis

2D

Left item detection

3D

Queue length and waiting time estimation

3D

Virtual shield

3D

Multi-modal sensing

3D+

Introduction

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A frequently asked question

Google

why is computer vision
why is computer vision **so difficult**

Press Enter to search.

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Motivation

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Why is Computer Vision difficult? (from a Bayesian perspective)

- Primary challenge in case of Vision Systems (incl. biological ones):

? uncertainty/ ambiguity ?

Example for robust vision

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Example: Crop detection

- Radial symmetry
- Near regular structure

Introduction

Motivation

- Challenges when developing Vision Systems:
 - Complexity ← Algorithmic, Systemic, Data
 - Non-linear search for a solution

IDEA APPLICATION PRODUCT

Depth sensing

- Emerging trends in sensing other modalities
 - Active (illumination) stereo → depth
 - Kinect (PrimeSense)
 - Time of Flight (TOF)
 - Passive stereo → depth
 - Light-field camera (Lytro, Raytrix) → depth
 - Thermal Infrared

Motivation

Visual Surveillance - Motivating example

Algorithmic units:

- Object detection and classification
- Tracking

Typical surveillance scenario:

Who : people, vehicle, objects, ...

Where is their location, movement? **Activity recognition**

What is the activity?

When does an action occur?

Motivation

Visual Surveillance - Motivating example

Algorithmic units:

- Object detection and classification
 - Counting, Queue length, Density, Overcrowding
 - Abandoned objects
 - Intruders
- Tracking
 - Single objects
 - Video search
 - Flow
- Activity recognition
 - Near-field (articulation)
 - Far-field (motion path)

Typical surveillance scenario:

Who : people, vehicle, objects, ...

Where is their location, movement? **Activity recognition**

What is the activity?

When does an action occur?

2D

Real-time optical flow based particle advection

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Optical flow driven advection

Advection: transport mechanism induced by a force field

A particle trajectory induced by the OF field

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Particle advection with FW-BW consistency

A simple but powerful test

Forward:

Backward:

Successful:

Failure:

Consistency check: $\Delta\epsilon < \beta \overline{\Delta x}$ $\overline{\Delta x}$: mean offset

Pedestrian Flow Analysis

Public dataset: Grand Central Station, NYC: 720x480 pixels, 2000 particles, runs at 35 fps

Wide-area Flow Analysis

Other examples: wide area surveillance (small objects, nuisance, clutter)

2D + 3D

Left-item detection using depth and intensity information

- Composite task:
 - Static object detection
 - Human detection and tracking

Detecting stationary objects

What is a static object?

- "non-human" foreground which keeps still over a certain period of time
- Two fundamentally different approaches:
 - Background modeling (foreground regions becoming static)
 - +: simple, pixel-based
 - : object removal, ghosts
 - Tracking detected foreground regions
 - +: many adequate tracking approaches (blob-based, correlation-based)
 - : crowd, occlusion → failure

Both techniques experience problems with illumination variations → motivation for depth-based sensing

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Detecting stationary objects

Method 1: single background models + sub-sampling of FG results

Liao H-H, Chang J-Y, Chen L-G. "A localized Approach to abandoned luggage detection with Foreground-Mask sampling". Proc. of AVSS 2008, pp. 132-139.

Detecting stationary objects

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Method 2: computing two background models at two different framerates

The diagram illustrates the process of detecting stationary objects. It starts with a sequence of 'Input frames' over 'time'. These frames are processed into two foreground models: 'Foreground Long' (represented by red blocks) and 'Foreground Short' (represented by blue blocks). These two models are then combined to produce a 'Motion Image' (represented by a yellow block).

Porikli, F.; Ivanov, Y.; Haga, T. "Robust Abandoned Object Detection Using Dual Foregrounds", Journal on Advances in Signal Processing, art. 30, 11 pp., 2008.

Detecting stationary objects

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Method 2: computing two background models at two different framerates

The flowchart shows the hypothesis testing process. An 'Image' is compared against two foreground models. If there is a 'Change' in the 'Foreground Long' model, it leads to a 'Change' hypothesis, which is identified as a 'Moving object'. If there is a 'Change' in the 'Foreground Short' model, it leads to a 'No Change' hypothesis, which is identified as a 'Candidate abandoned object'. If there is a 'No Change' in both models, it leads to a 'No Change' hypothesis, which is identified as 'Scene background'.

Foreground Long (F_L) and Foreground Short (F_S) computed for a sample scene:

The sequence shows the original scene (a), the foreground models (b, c), and the final result (e). The foreground models are shown as binary masks over the original scene. The final result shows the detected objects in the scene.

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Obtaining stereo depth information

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Passive stereo based depth measurement

- 3D stereo-camera system developed by AIT
 - Area-based, local-optimizing, correlation-based stereo matching algorithm
 - Specialized variant of the Census Transform
 - Resolution: typically ~1 Mpixel
 - Run-time: ~ 14 fps (Core-i7, multithreaded, SSE-optimized)
 - Excellent "depth-quality-vs.-computational-costs" ratio
 - USB 2 interface

The image shows a 3D stereo-camera system and a depth map of a scene. A red arrow indicates a distance of 12 m.

Advantage:

- Depth ordering of people
- Robustness against illumination, shadows,
- Enables scene analysis

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Stereo camera characteristics

Trinocular setup:

- 3 baselines possible
- 3 stereo computations with results fused into one disparity image

The image shows a trinocular camera setup with three cameras mounted on a tripod.

small medium

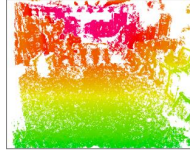
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Data characteristics

The image shows an 'Intensity image' and a 'Disparity image'. The disparity image is a color-coded map of the scene's depth. A graph below shows the relationship between disparity (d) and depth (y).

2.5D vs. 3D algorithmic approaches

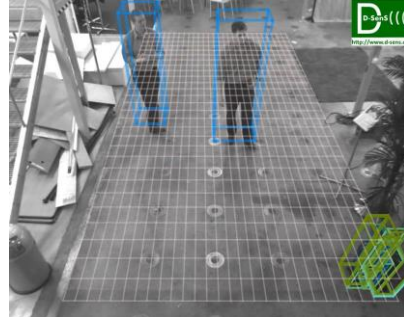
2.5D == using disparity as an intensity image



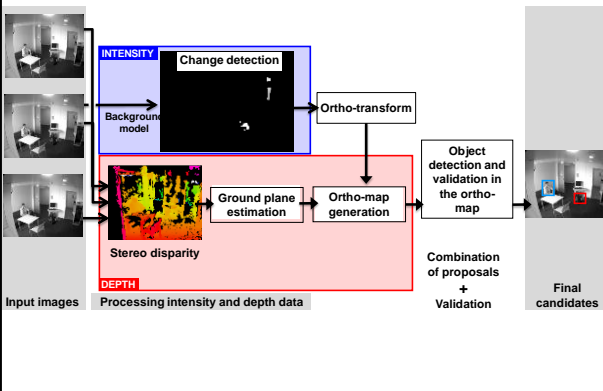
Left Item Detection

Additional knowledge (compared to existing video analytics solutions):

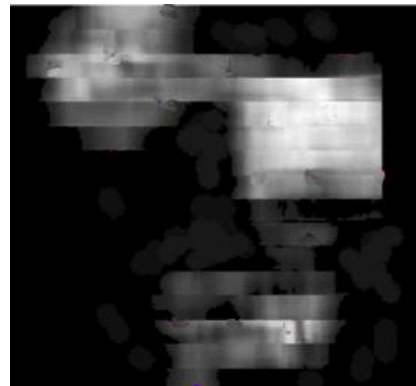
- Stationary object (Geometry introduced to a scene)
- Object geometric properties (Volume, Size)
- Spatial location (on the ground)



Methodology

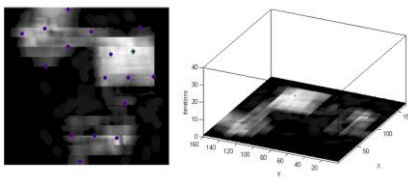


Human detection as clustering



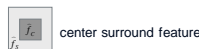
Scale-adaptive clustering using a stability criterion (1)

- Evolution of cluster centroid coordinates during CamShift



Algorithmic steps:

- (1) Computation of integral images: $ii_{00}(x, y)$ $ii_{02}(x, y)$
- (2) Locating the sample set: $\hat{f}_s / \hat{f}_c < T$



Scale-adaptive clustering using a stability criterion (2)

3. CamShift clustering iterations

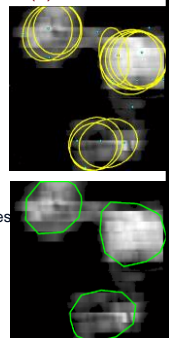
- Stopping criterion:

$$A_j^i = L_1 L_2$$

$$\gamma(i) = |A_j^{i-\delta} - A_j^{i+\delta}| / A_j^i$$

4. Spatial grouping of elliptic clusters

- Subsampling ellipse contours \rightarrow piecewise linear boundaries
- Pairwise check for overlap \rightarrow convex hull approximation (fast BFP algorithm)



Left Item Detection – Demos

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Left Item Detection – Interesting cases

Object form Transferred objects

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Quantitative evaluation

Detection results Ground truth Depth-based proposals Motion-based proposals

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2D + 3D

Queue length detection using depth and intensity information

Queue Length + Waiting Time estimation

What is waiting time in a queue?
Time measurement relating to last passenger in the queue

Why interesting?
Example: Announcement of waiting times (App) → customer satisfaction
Example: Infrastructure operator → load balancing

Queue analysis

- Challenging problem

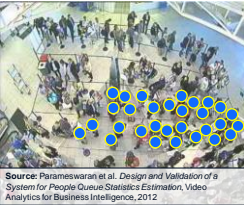
$$\text{Waiting time} = \frac{\text{Length}}{\text{Velocity}}$$

1. What is the shape and extent of the queue?
2. What is the velocity of the propagation?

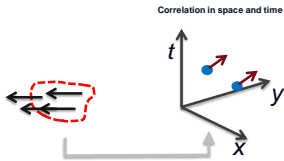
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Visual queue analysis - Overview

- How can we detect (weak) correlation?




Correlation in space and time

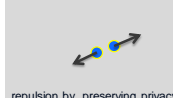

- Much data is necessary → Simulating crowding phenomena in Matlab
 - Social force model (Helbing 1998)



goal-driven kinematics – force field



repulsion by walls




repulsion by „preserving privacy“

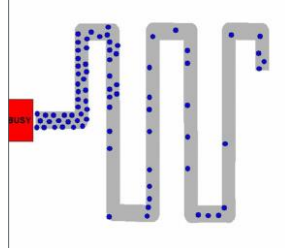
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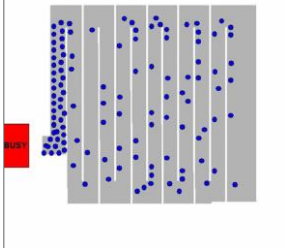
Queue analysis

Simulation tool → Creating infinite number of possible queueing zones



Two simulated examples (time-accelerated view):







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Queue analysis (length, dynamics)

Straight line



Meander style

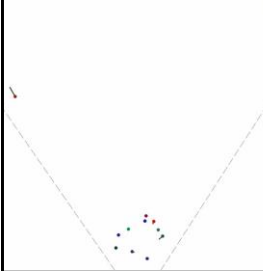


Staged scenarios, 1280x1024 pixels, computational speed: 6 fps


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Adaptive estimation of the spatial extent of the queueing zone

Estimated configuration
(top-view)



Detection results



Left part of the image is intentionally blurred for protecting the privacy of by-standers, who were not part of the experimental setup.

stereo sensor ▼

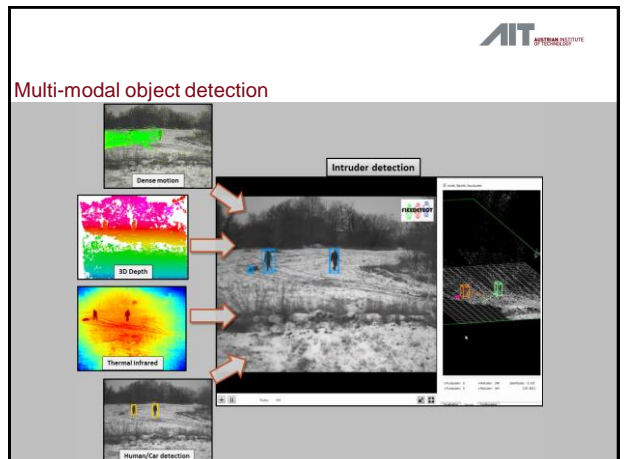
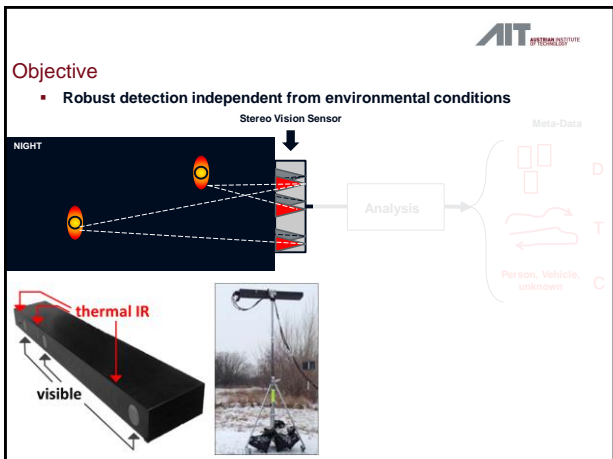
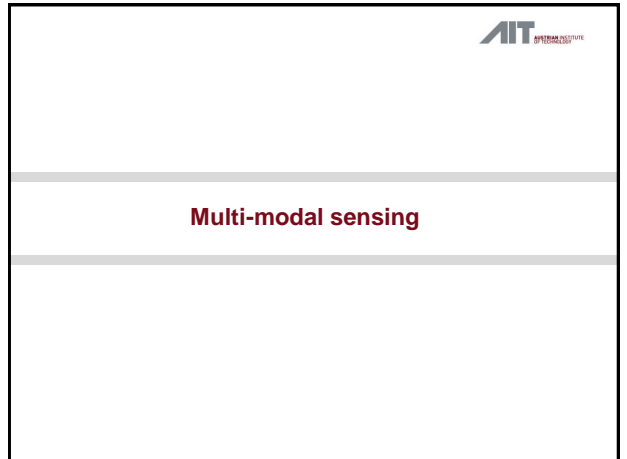
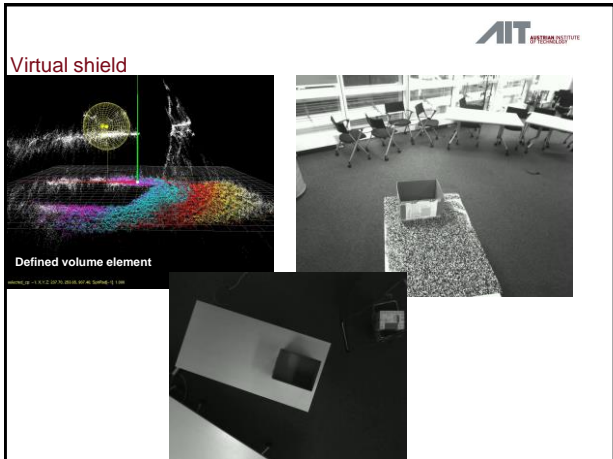
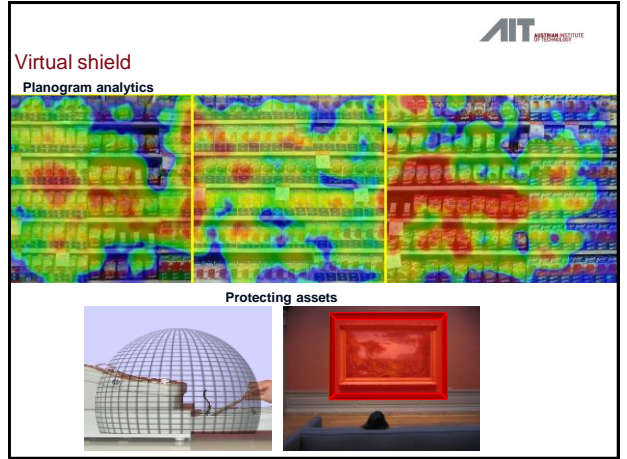
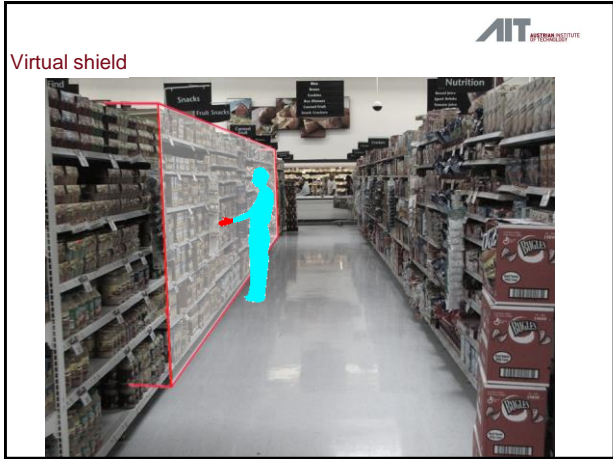
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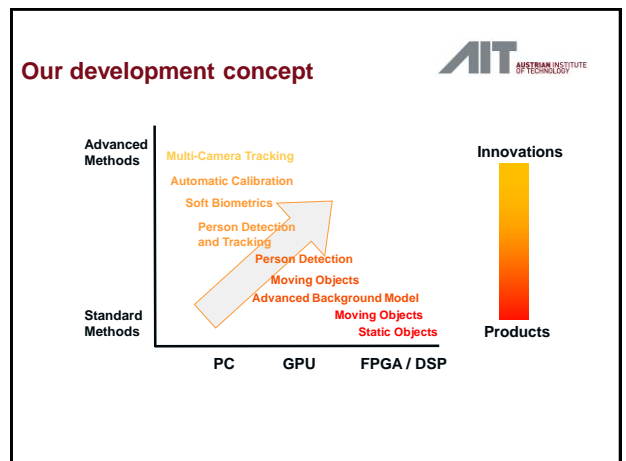
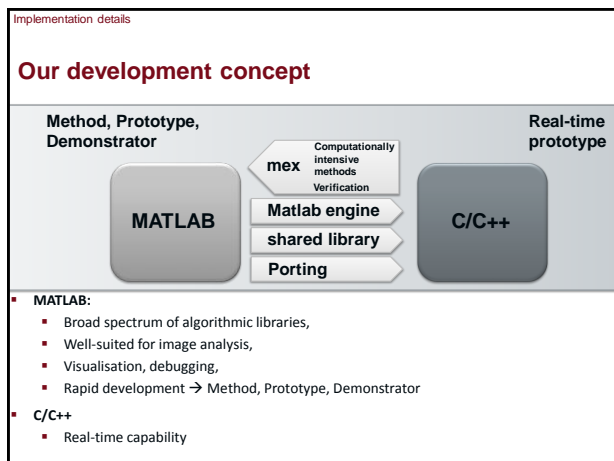
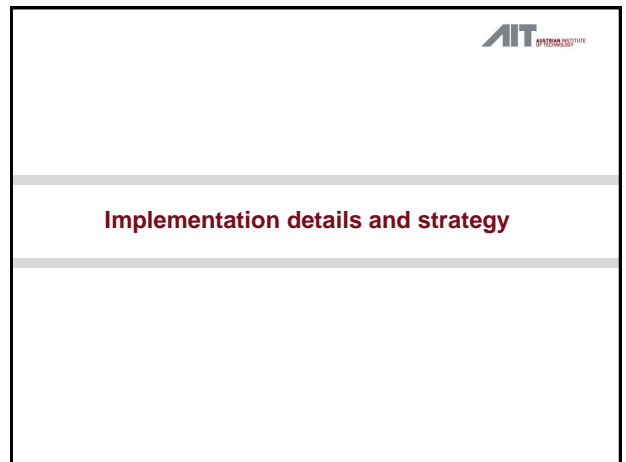
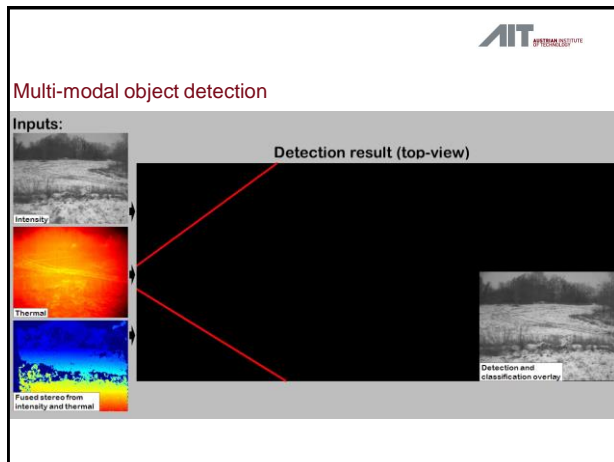
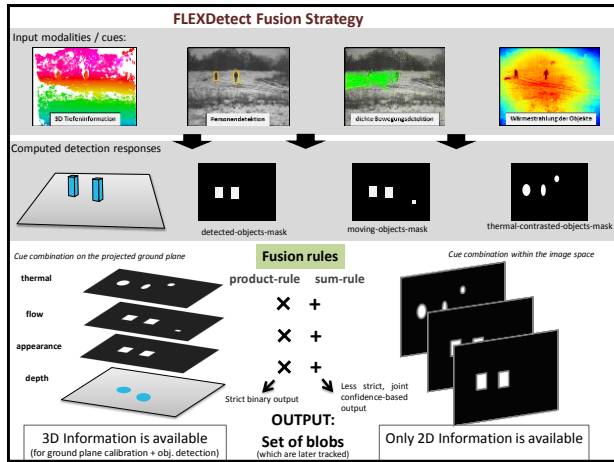
Scene-aware heatmap



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Virtual Shield





Thank you for your attention!



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