



# Mobile Visual Scene Understanding in Highly Dynamic Environments

**Bastian Leibe** 

Mobile Multimedia Processing

Computer Sciences 8 - Computergraphics & Multimedia

RWTH Aachen

MIRACLE Workshop, St. Augustin, 30.10.2009



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#### RWTH Computer Graphics & Multimedia Group

- Prof. Dr. Leif Kobbelt
  - Computer Graphics
  - Geometric Modelling



- Prof. Dr. Bastian Leibe
  - Computer Vision
  - Machine Learning







B. Leibe





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# Research Focus: Mobile Vision Applications

Three main scenarios



Mobile phones, Wearable computing



Mobile robotics, Personal mobility



Intelligent vehicles

**On-board computation** 

**Unrestricted environment** 



#### **Research Directions**

- Mobile Visual Search
  - Recognition from mobile phones
  - Automatic content creation
  - Towards mobile AR applications...



- Mobile Object Detection and Tracking
  - Object categorization
  - Scene geometry estimation
  - Multi-person tracking
  - Detailed body pose analysis





# Target Scenario: Pedestrian Navigation





#### Mobile visual search

- Simply point the camera to any object/building of interest.
- Images are transmitted to a central server for recognition.
- Object-specific information is sent back to be displayed on the mobile phone (mobile AR).

# **System Overview**

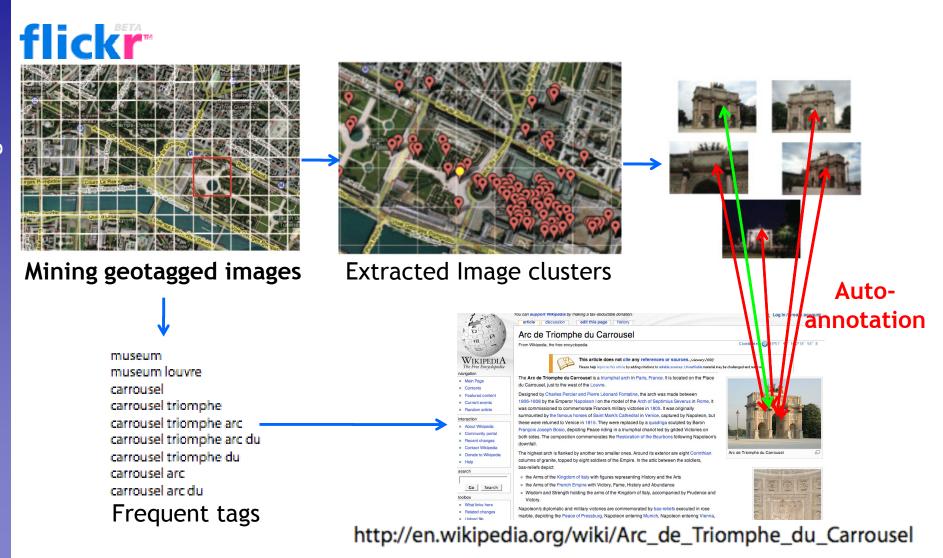
 How can we make this scalable for an entire city?





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#### **World-Scale Mining for Content Creation**



# **Example Results: Famous Tourist Sights**





http://en.wikipedia.org/wiki/Basilica\_of\_the\_Sacr%C3%A9\_C%C5%93ur 426 Elements, 233 users, 287 days. Precision: 100%





http://en.wikipedia.org/wiki/Tour\_Montparnasse 40 elements, 10 users, 11 days. Precision: 100%

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http://en.wikipedia.org/wiki/Notre\_Dame\_de\_Paris 588 elements, 287 users, 334 days. Precision: 100%

[Quack, Leibe, Van Gool, CIVR'08]

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# **Example Results: Matching Under Occlusion**





http://en.wikipedia.org/wiki/Old\_Town\_Square\_(Prague) 262 elements, 122 users, 195 days. Precision: 98%.





http://en.wikipedia.org/wiki/Colosseum
582 elements, 190 users, 252 days. Precision: 100%
B. Leibe | Quack, Leibe, Van Gool, CIVR'08]



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Joint work with: Andreas Ess
Stephan Gammeter

Luc Van Gool

(ETH Zurich)

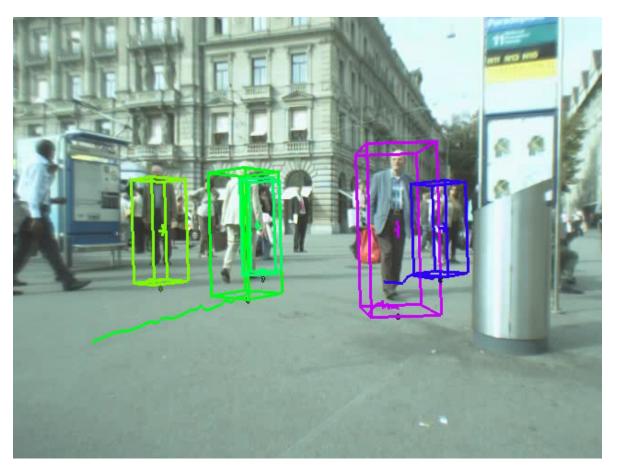
(TU Darmstadt)

**Konrad Schindler** 

B. Leibe



#### **Towards Visual Scene Understanding**



#### Objectives

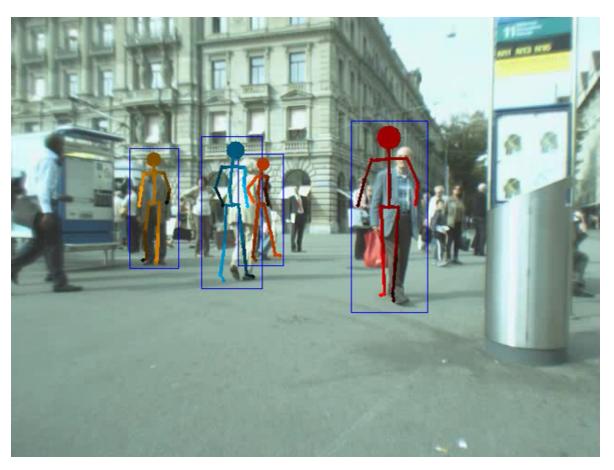
- Detect & track people in environment
- Interpret their motion
- Predict their future behavior

#### Challenges

- We are moving
- > Objects are moving
- > Significant occlusions



#### **Towards Visual Scene Understanding**



#### Objectives

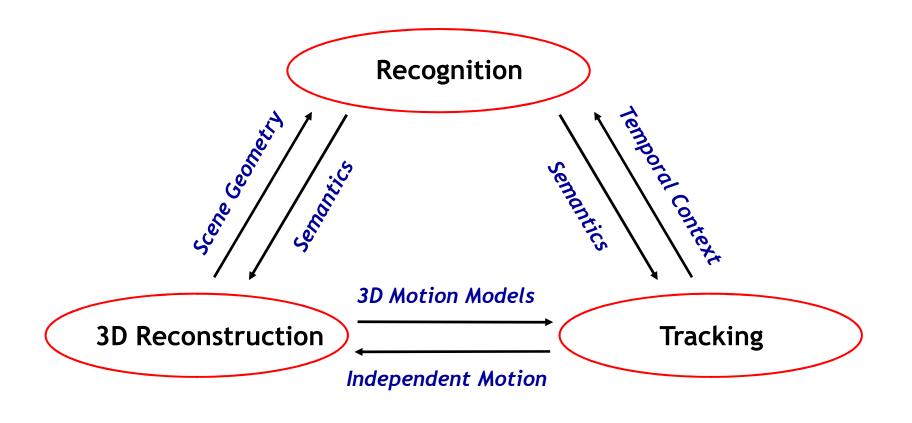
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# Integration Principle: Cognitive Loops



Integrate different vision modalities through Cognitive Feedback



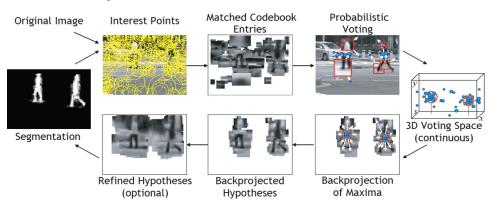
#### **Outline**

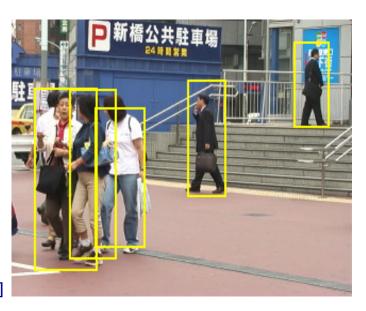
- Object Recognition
  - Implicit Shape Model (ISM) approach
- Integration with Scene Geometry
  - Coupled object detection & 3D estimation
- Temporal Integration
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- Visual Odometry
  - Feedback from detection and tracking
- Putting It All Together...
  - Mobile pedestrian tracking
  - > Articulated tracking under egomotion



# **Object Categorization & Detection**

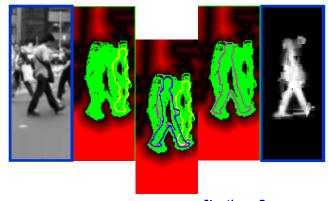
ISM Object Detection





[Leibe, Leonardis, Schiele, IJCV'07]

Pedestrian Detection in Crowds





[Leibe, Seemann, Schiele, CVPR'05]



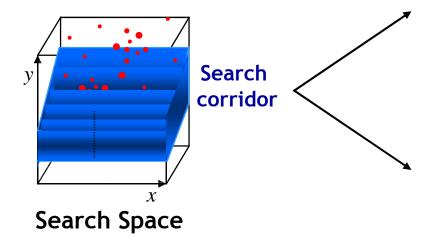
#### **Outline**

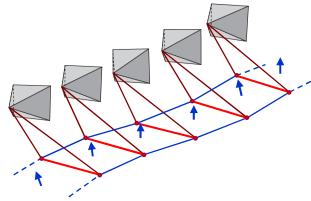
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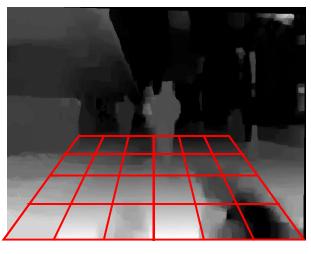
### **Scene Geometry Estimation**

- Goal: Find the ground plane
  - Restrict object location
  - Assume Gaussian size prior
  - ⇒ Significantly reduced search space





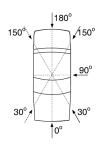
Structure-from-Motion



Dense stereo

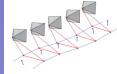
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# **Detections Using Ground Plane Constraints**







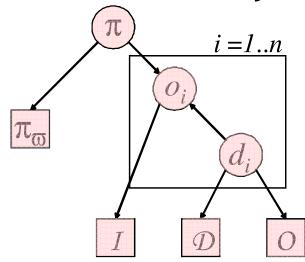


left camera 1175 frames



# Object & Ground-plane Reasoning

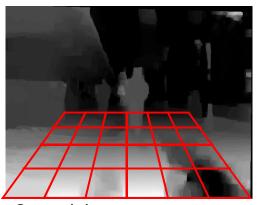
Probabilistic combination in Bayesian network



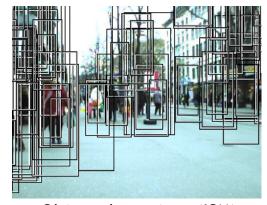
Groundplane  $\pi$ 

Object detections  $O_i$ 

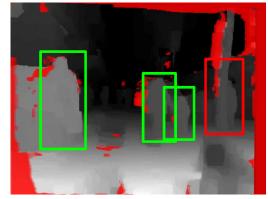
Depth measurements  $d_i$ 



Groundplane measurements



Object detections (ISM)



Depth verification



# Object & Ground-plane Reasoning





- Effect:
  - Reliable detections from scene context
  - Accurate 3D positioning from depth map

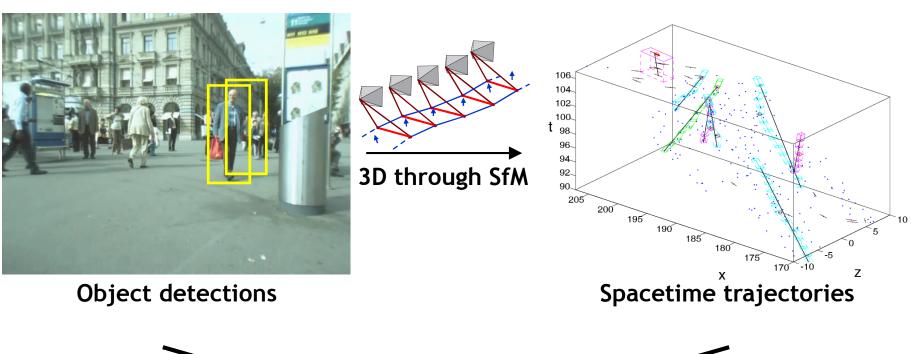


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## **Coupled Detection and Tracking**

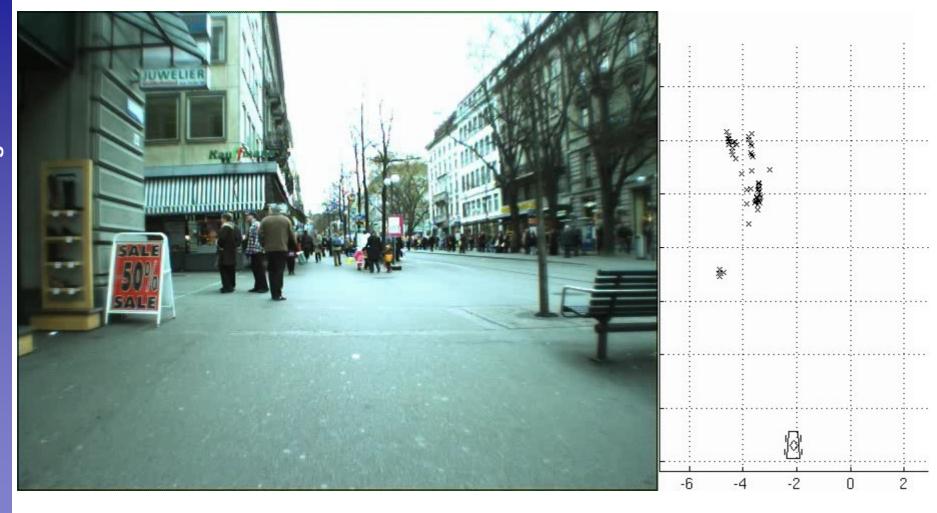




(Quadratic Boolean Optimization)

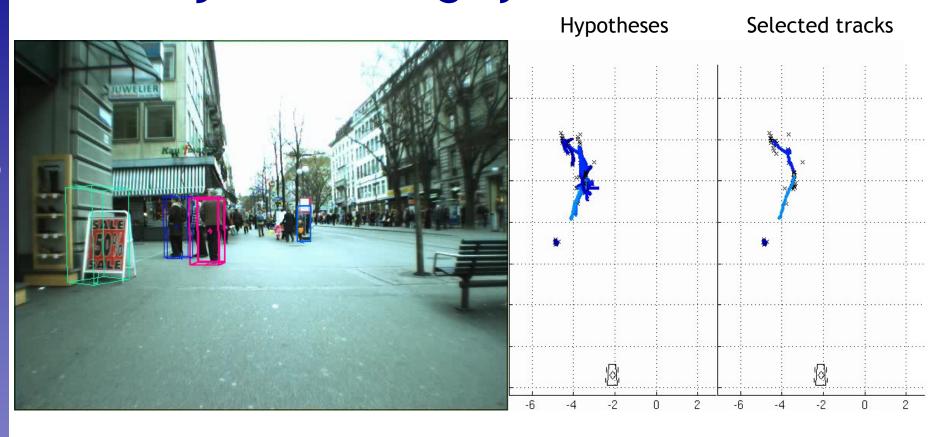


# **Multi-Object Tracking by Detection**





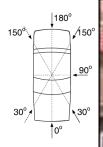
#### Multi-Object Tracking by Detection



- Multi-hypothesis tracking with model selection in each frame
- Ability to recover temporarily lost tracks

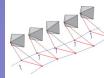


# **Dynamic Scene Analysis**









[Leibe, Cornelis, Cornelis, Van Gool, CVPR'07]

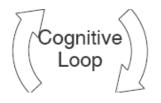
# Application: Augmented 3D City Model

Enhancing your driving experience...





Original



**3D Reconstruction** 



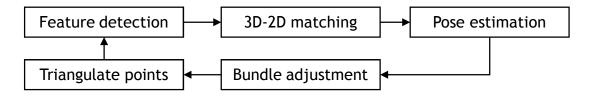
#### **Outline**

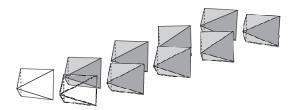
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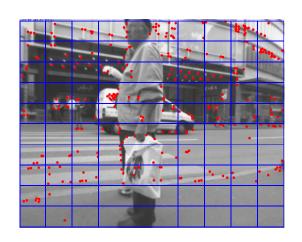


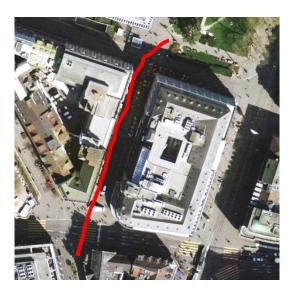
# **Visual Odometry**

- Defines common coordinate frame
  - Basis for tracking-by-detection in 3D
- Stereo-based Structure-from-Motion
  - Similar to Nister et al., CVPR'04





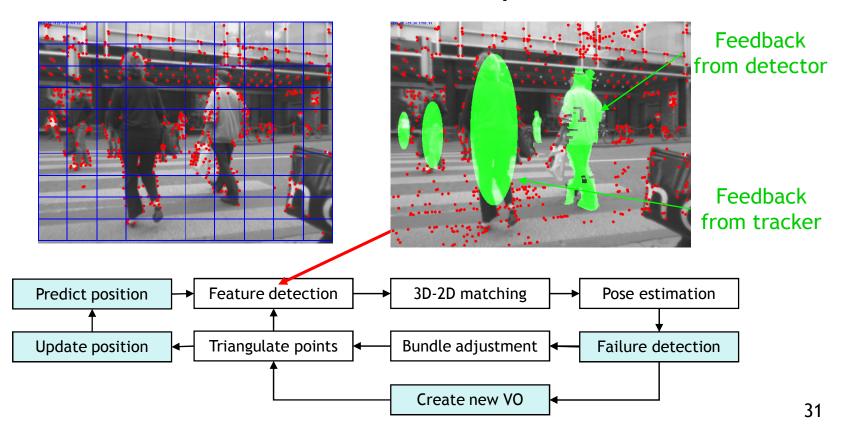






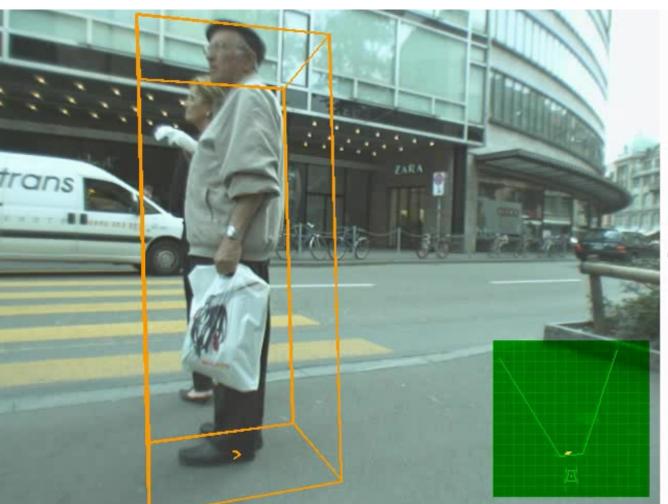
### Feedback to Visual Odometry

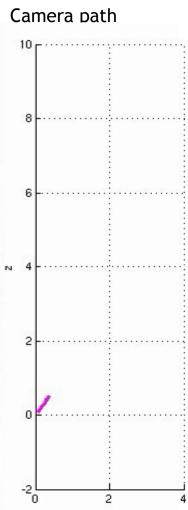
- Not all parts of scene are static
  - Detector / Tracker give semantic information
  - Mask out moving parts
    - ⇒ Restrict localization efforts to static parts





# Feedback to Visual Odometry



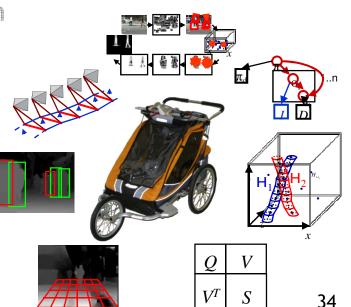


Standard VO
Failure detection, no masking
Failure detection + masking
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# **Mobile Tracking Through Crowds**





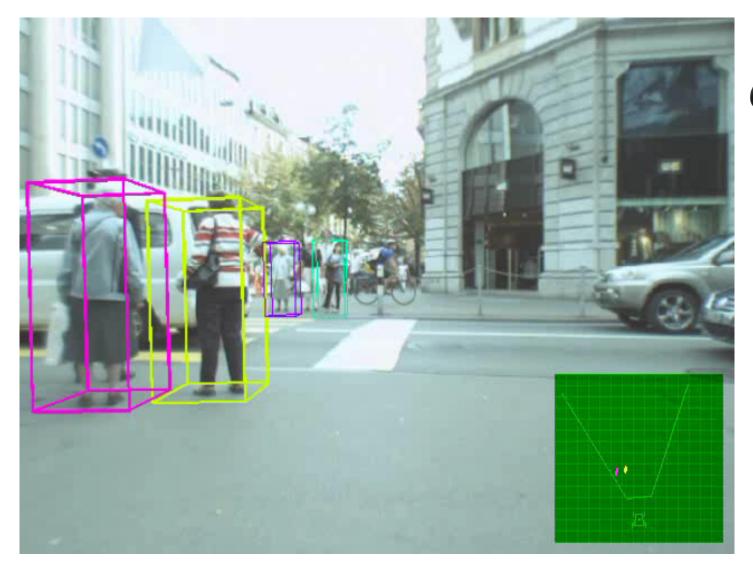




#### An Extreme Case...







# **Predicting Behavior of Dynamic Obstacles**











# **Application in Cars**





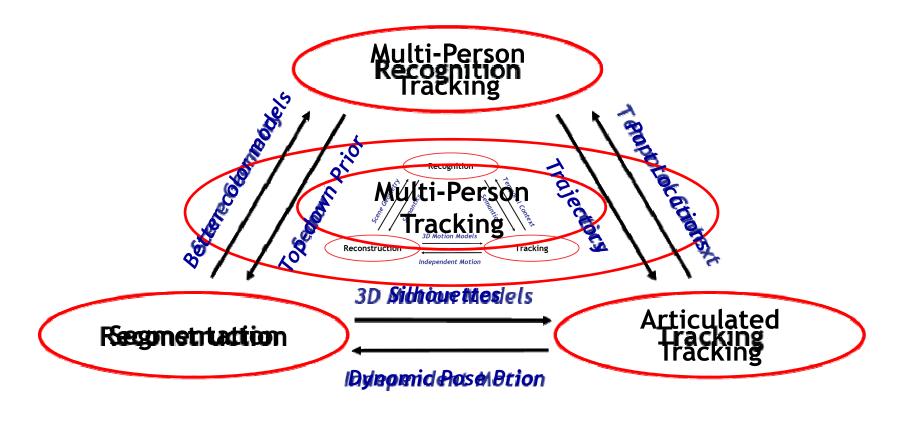




(Cooperation with Toyota Motor Corporation)



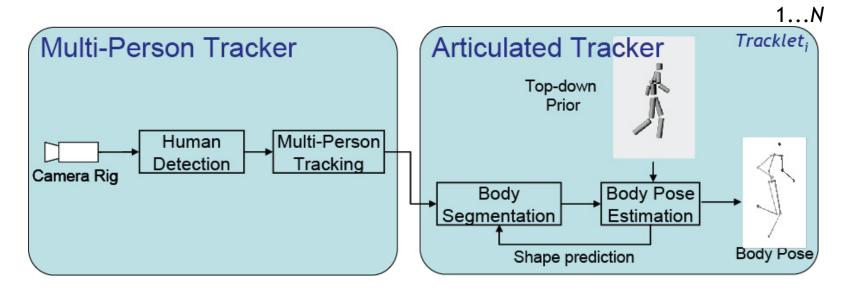
# Multi-Person Tracking as new Basic Unit



Many interesting ways to go on from here...



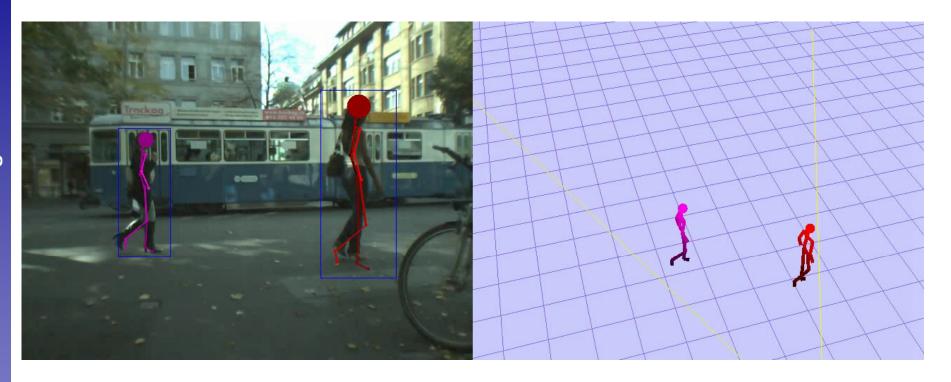
## **Recovering Articulations**



- Idea: Only perform articulated tracking where it's easy!
- Multi-person tracking
  - Solves hard data association problem
- Articulated tracking
  - Only on individual "tracklets" between occlusions



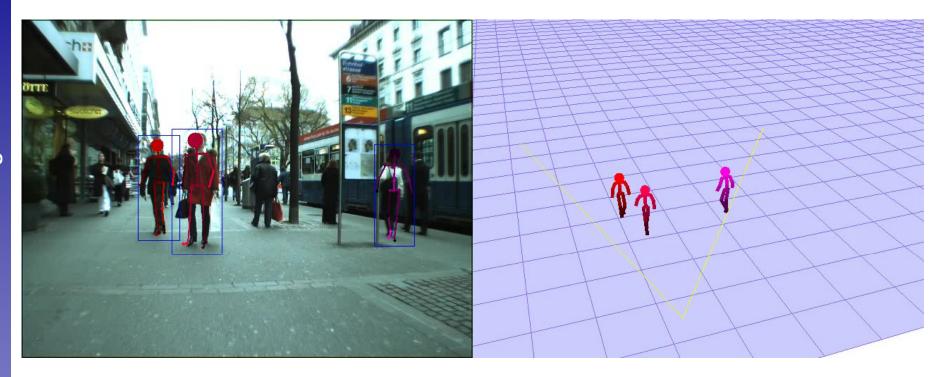
## **Articulated Multi-Person Tracking**



- Multi-Person tracking
  - Recovers trajectories and solves data association
- Articulated Tracking
  - Estimates detailed body pose for each tracked person



## **Articulated Tracking under Egomotion**



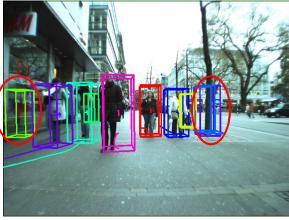
- Guided segmentation for each frame
  - No reliance on background modeling
  - Approach applicable to scenarios with moving camera
  - Feedback from body pose estimate to improve segmentation

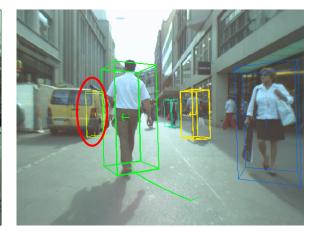


## **Typical Failure Cases**

- Too big pedestrians
  - Not completely visible
  - Separate detector necessary
- False positives on reflections, trees, trashcans, ...
  - Multi-class detector?
  - > OK for path planning: in most cases, still an obstacle



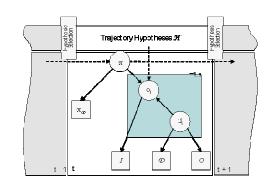


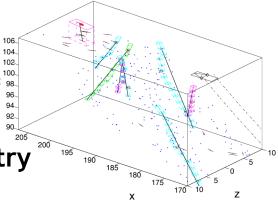


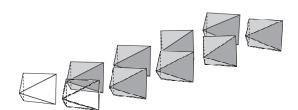


## **Keys for Success**

- Bayesian network for detection
  - Allow modification of bounding boxes
  - Use of confident depth information
  - No hard decision about ground plane
  - Spatial prior from tracker
- Multi-hypothesis tracking
  - Per-frame model selection
  - Depth information for localization
  - World-coordinate frame from visual odometry
- Visual Odometry
  - Feedback from Tracking/Detection
  - > Failure detection









## **Conclusion**

- Visual scene understanding
  - Vision is becoming feasible in the real world.
  - Many individual components are getting sufficiently mature.
  - Robust performance possible through combination.
- Perspective for Augmented/Mixed Reality
  - Currently still restricted to high-power hardware...
  - But real-time reachable within next 2 years.
  - Novel capabilities for AR applications?
    - Object categorization
    - Reaction to people
    - Augmenting categorical objects
- What use can we make of this for AR/MR applications?



## Thank you very much!

### **Collaborators**

#### **Local Features**

- K. Mikolajczyk
- N. Cornelis
- T. Quack

#### Obj. Detection

- E. Seemann
- M. Fritz
- A. Thomas
- A. Lehmann
- K. Mikolajczyk
- A. Leonardis
- B. Schiele

#### **Tracking**

- K. Schindler
- A. Ess



#### **Body Pose Est.**

- T. Jaeggli
- S. Gammeter

#### SfM & Stereo

- N. Cornelis
- K. Cornelis
- A. Ess
- T. Weise

#### System Integr.

- A. Ess
- K. Schindler
- L. Van Gool

http://mmp.rwth-aachen.de



## **Timing**

- C/C++ implementation
  - Without detector currently 3-4 fps
- Many calculations can be cached
- Detector current bottleneck
  - Faster detectors can be plugged in (e.g. fastHOG on GPU)
  - > Parallelization possible

Component	CPU	GPU	Time
Detector	×		2 x 15s
Depth map	×	×	15s / 0.020s
Bayesian Network	×		0.200s
Visual odometry	×	×	0.020s
Tracking	×		0.100s

(per frame)



## Large Datasets Available

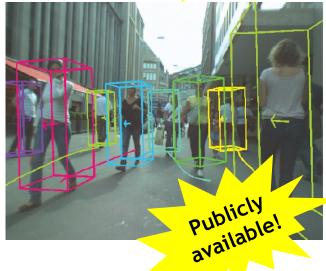
#### ICCV'07 Data

- 4 Sequences
- ~2200 frame pairs total
- ~10,900 Pedestrian annotations
- Cameras + groundplane from SfM
- Various baseline performances

#### CVPR'08 Data

- > 3 New sequences
- ~2750 additional frame pairs
- Pedestrian annotations (every 4<sup>th</sup> frm)
- Camera + groundplane from SfM
- Various baseline performances









# Mobile Multi-Person Tracking in Highly Dynamic Environments

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