

### **Intended Audience**

- The slides are intended for a graduate course of roughly 20 hours (14 lectures of 90 min each).
- Students are expected to possess basic knowledge in Computer Vision and Artificial Intelligence.

2

#### Website

The website for this course can be reached via

http://kogs-www.informatik.uni-hamburg.de/~neumann/HBD-SS-2008/

You will find PDF copies of the slides and possibly other useful information related to the course.

The website will be updated each week on Wednesday.

3

## Contents (1)

Lecture 1: Introduction

Contents overview, motivation, aims, problem areas

Lecture 2: Early work on scene interpretation

Badler, Tsotsos, Hogg, Nagel, Neumann

Lecture 3: Basic knowledge representation formalisms

Semantic Networks, Frames, Constraints, Relational

Structures

Lecture 4: Conceptual units for scene interpretation

Aggregates, situation trees, scenarios

Lecture 5: Interface to low-level vision

Primitive symbols, grounding

Lecture 6: Modelling spatial and temporal relations

Fuzzy predicates, Allen, RCC8, constraints

Lecture 7: Interpretation procedures

4

## Contents (2)

Lecture 8: Logical framework

Model construction, Decription Logics

Lecture 9: Scene interpretation as configuration

Stepwise construction, SCENIC

Lecture 10: Probabilistic Guidance

Hierarchical Bayesian Networks

**Lecture 11: Robot Localisation** 

Simultaneous localisation and Mapping

Lecture 12: Case study

Real-time scenario recognition (Orion/INRIA)

Lecture 13: Application development

Criminal act recognition (Orion/INRIA)

Lecture 14: Summary and outlook

5

### What is Computer Vision?

Computer Vision is the academic discipline dealing with task-oriented reconstruction and interpretation of a scene by means of images.

scene: section of the real world

stationary (3D) or moving (4D)

image: view of a scene

projection, density image (2D)

depth image (2 1/2D) image sequence (3D)

reconstruction computer-internal scene description and interpretation: quantitative + qualitative + symbolic

task-oriented: for a purpose, to fulfill a particular task

context-dependent, supporting actions of an agent

3

### What Is Scene Interpretation?

Scene Interpretation is the task of "understanding" or interpreting a scene <u>beyond single-object recognition</u>. Typical examples are traffic scene interpretation for driver assistance, inferring user intentions in smart-room scenarios, recognizing team behavior in robocup games, discovering criminal acts in monitoring tasks.

#### **Characteristics:**

- Interpretations involve several objects and occurrences.
- Interpretations depend on temporal and spatial relations between parts of a scene
- Interpretations describe the scene in qualitative terms, omitting geometric details.
- Interpretations include inferred facts, unobservable in the scene.
- Interpretations are based on conceptual knowledge and experience about the world.

"Scene interpretation" means roughly the same as "high-level vision".

## **Examples for Scene Interpretation (1)**



scene interpretation means understanding every-day occurrences

Garbage collection in Hamburg (1 frame of a sequence)

We want to recognize parts, activities, intentions, spatial & temporal relations

## **Examples for Scene Interpretation (2)**



Scene interpretation is silent movie understanding

Buster Keaton in "The Navigator"
We want to recognize episodes, the "story", emotions, funnyness

# Some Application Scenarios for Scene Interpretation

- Street traffic observations (long history)
- Cameras monitoring parking lots, railway platforms, supermarkets, nuclear power plants, ...
- Video archiving and retrieval
- Soccer game analysis
- Smart room cameras
- Autonomous robot applications
   (e.g. robot watchmen, playmate for children, assistance for elderly)

# **Technological Challenges of Scene Interpretation Tasks**

- Problem area combines Computer Vision (CV) and Artificial Intelligence (AI), not well attended by CV and AI research
- Interpretations may build on common sense knowledge, common-sense knowledge representation is an unsolved issue
- Application scenarios may be large and highly diverse, knowledge engineering is a challenge
- Visual learning and adaptation may be required
- Reliability and complexity management may become important issues
- Economical application development requires generic approach

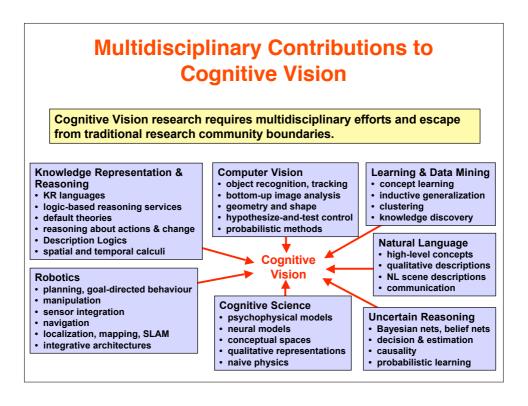
### **Cognitive Computer Vision**

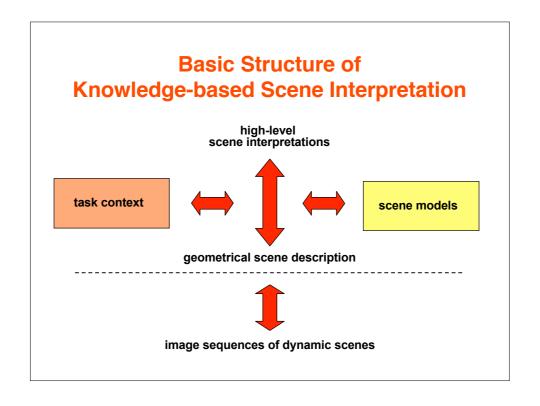
Scene interpretation is strongly related to "cognitive vision", a term created for vision comparable to human vision:

Cognitive computer vision is concerned with integration and control of vision systems using explicit but not necessarily symbolic models of context, situation and goal-directed behaviour. Cognitive vision implies functionalities for knowledge representation, learning, reasoning about events & structures, recognition and categorization, and goal specification, all of which are concerned with the semantics of the relationship between the visual agent and its environment.

#### **Topics of cognitive vision:**

- integration and control
- explicit models
- not necessarily symbolic
- context
- situation
- goal-directed behaviour
- knowledge representation
- learning
- reasoning
- recognition
- categorization
- goal specification
- visual agent





## Representation Levels for High-level Scene Interpretation

symbolic scene descriptions

symbolic domain

metric domain

pictorial scene representations

metric feature representations

map-type feature representations

signal domain

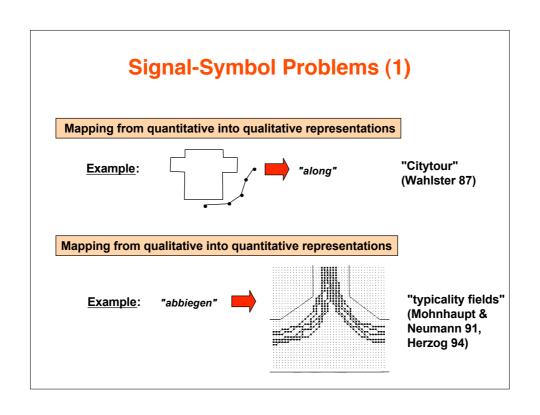
raw image sequence

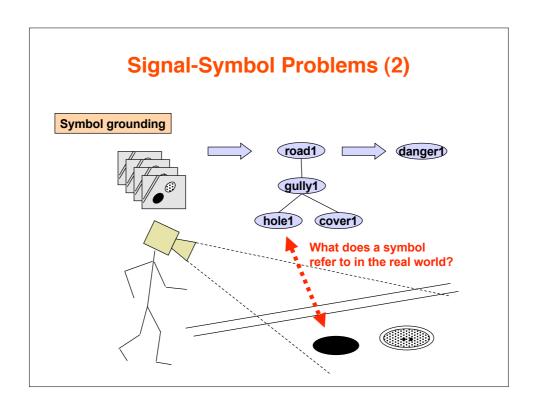
# **Context and Task Dependence**

Interpretations may depend on

- domain context
- spatial context
- temporal context
- intentional context
- task context
- communicative context
- focus of attention
- a priori probabilities

Constructing an interpretation is <u>not</u> a mapping from image data into interpretation space.





### **Common-Sense Problems**

#### Common-sense reasoning

Deductions from symbolic knowledge about a scene should not only be correct w.r.t. to domain-related definitions but also w.r.t. to common sense.

**Examples:** (implies (and house (some near lake)) mosquito-house)

(instance house1 house) (instance lake1 lake) (related house1 lake1 near)

(instance house1 (not (mosquito-house)))

=> inconsistent by domain-related definitions

(instance house1 house)
(instance cup1 cup)
(related house1 cup1 inside)
=> inconsistent by common sense

## **Uncertainty Problems (1)**

#### **Fuzzyness of concepts**

Many high-level concepts have unsharp boundaries.

"behind" "overtake" "meet"

- => mapping into logical propositions may be problematic
- Fuzzy set theory offers "degree of applicability"
- Probability theory offers statistical measures for language use



Fuzzy definition of behind

## **Uncertainty Problems (2)**

#### Uncertainty of data

**Example: Object boundaries** 





Strict bottom-up image interpretation is fundamentally ill-defined

# **Uncertainty Problems (3)**

#### **Exploring multiple hypotheses**

Answers from several disciplines:

- · graph matching
- heuristic search
- optimization theory
- · logic theories
- probability & utility theory
- · case-based reasoning
- · neural networks
- particle physics (and others)

Mixed bottom-up and top-down interpretation strategies have been rarely explored

### **Uncertainty Problems (4)**

#### Cultural clash betweeen logical and probabilistic reasoning

Probabilistic methods are nor yet seamlessly integrated with logical calculi

#### **Interesting recent developments:**

- First-order probabilistic inference (Poole 03)
- Probabilistic relational models (http://dags.stanford.edu/PRMs/)

**Example for reasoning in image interpretation:** 

(from Kanade's invited lecture at IJCAI-03: "Computer Vision: Al or Non-Al Problem?")

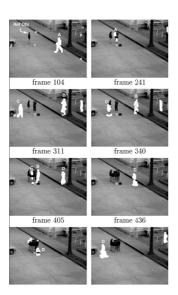
car on left side of street

(uncertain orientation of car)

japanese signs => left-hand traffic

orientation of car resolved

# State-of-the-art Example of Scene Interpretation



S. Hongeng, R. Nevatia and F. Bremond.
Video-Based Event Recognition: Activity
Representation and Probabilistic
Recognition Methods.

Computer Vision and Image Understanding, Vol. 96 (2004), 129 - 162.

#### Recognising "Stealing by Blocking":

"A" approaches a reference object (a person standing in the middle with his belongings on the ground). "B" and "C" then approach and block the view of "A" and the reference person from their belongings. In the mean time, "D" comes and takes the belongings.

# **Learning and Recognising Structures in Buildings (1)**

EU-funded project eTRIMS\* at the Cognitive Systems Laboratory of Hamburg University

Rectangular objects recognised by low-level vision

Window-arrays recognised by highlevel vision using a learnt model



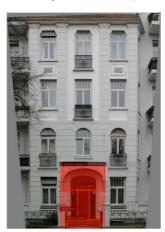
\*) E-Training for the Interpretation of Man-made Scenes

# **Learning and Recognising Structures in Buildings (2)**

Interpretation of a facade, entrance is not recognised



Entrance is recognised after learning from the example



# **Monitoring Airport Activities** in the EU-Project Co-Friend

**EU-funded project Co-Friend\* at the Cognitive Systems Laboratory of Hamburg University** 



#### **Application scenario**

- Aircraft servicing operations at Toulouse-Blagnac Airport are observed by eight cameras
- Moving objects are tracked by a low-level vision system
- Activities such as refueling or baggage unloading are recognised by a high-level vision system

#### **Project goals**

- · Reliable on-line interpretation of extended multi-camera video sequences
- · Learning new activities from examples
- Robust recognition performance based on a rich domain ontology

<sup>\*)</sup> Cognitive & Flexible learning system operating Robust Interpretation of Extended real scenes by multi-sensors Datafusion