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Description Logic for Vision-Based Intersection Understanding

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- **Motivation:** Road recognition for toy worlds only?
- **DL Road Network Knowledge Base Development**
 - DL Tutorial
 - Hypothesis Space
 - Sensor Input
- **DL Inference for Semantic Road Recognition**
 - Deduction
 - Model Construction

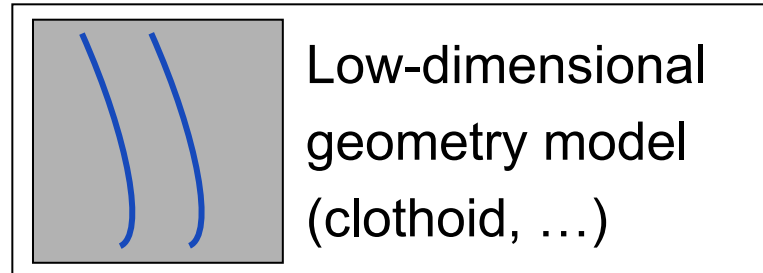
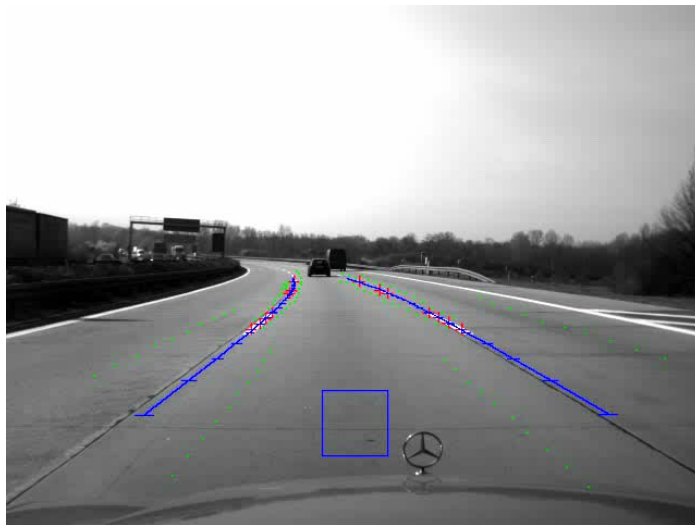
Road Recognition Status Quo

- Since mid 80ies
- Solved for highly restricted domains (highways)
- Few work for more complex domains, then only consideration of special cases
- Geometry only, no semantics

→ Toy worlds?

Common Approach to Road Recognition

1. Project



2. Compare with
image cues

3. Update

→ Cannot be generalized to more complex domains

1. **High-dimensional hypothesis space**
2. **Few features**
 - Narrow field of view
 - Massive occlusions
 - Omitted features
3. **Presence of noise**
 - Unmodelled objects
 - Decreased feature quality



→ Problem is ill-posed!

What is needed?

1. Geometrical Model of arbitrary roads and intersections
2. Massive reduction of hypothesis space size
 - using prior knowledge
 - using large set of complementary sensor data: video object detectors, digital map, positioning sensors, ...
3. Conceptual model of arbitrary roads and intersections
 - explicit representation (due to intensive HMI within a DAS)
4. Sound inference & retrieval services on the KB

→ Paradigm shift: Intersection recognition as

... scene understanding task

... ~~mid~~ high level vision task

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- **Decidable subset of 1st order logic**
- **Syntax:**

<i>Name</i>	<i>Description</i>	<i>Example</i>
Individuals	objects of the domain	<i>John</i>
Concepts	(\approx classes) sets of individuals	Human
Roles	binary relations on individuals	hasChild
Constructors	to build complex expressions $C, D \rightarrow C \cap D \mid C \cup D \mid$ $\neg C \mid \exists R.C \mid \forall R.C \mid$ $T \mid \perp \quad \dots$ $R, S \rightarrow R \mid R^{-} \mid R \circ S$	Man \cap $\exists \text{hasChild}.T$

- **Semantics:** Set-theoretic

- **Axioms** form sentences

Terminological Axioms	$C \equiv D$	$\text{Father} \equiv \text{Man} \cap \exists \text{hasChild.T}$
	$C \subseteq D$	$\text{Father} \subseteq \text{Person}$
	$R \subseteq S$	$\text{hasChild} \subseteq \text{hasDescendent}$
Assertional Axioms	$i : C$ $(i,j) : R$	$\text{John} : \text{Human}$ $(\text{John}, \text{Emily}) : \text{hasChild}$

- A **DL Knowledge Base** consists of
 - **Tbox**: Set of terminological axioms
 - state general domain knowledge here
 - **Abox**: Set of assertional axioms
 - state knowledge about particular situation here
 - (**Rulebox**)

- Classical DL inference services:

Satisfiability Check for TBox and ABox	$(\text{Mother} \cap \text{Male})$ is inconsistent
Classification of Tbox and Abox	$\mathcal{KB} \models \text{John: Father}$
Entailment	$\text{John: } \exists \text{hasChild.Female}$
Retrieval	Retrieve all individuals which are instance of: $\exists_{\geq 3} \text{hasChild} \cap \neg \text{Female}$

- Non-classical inference

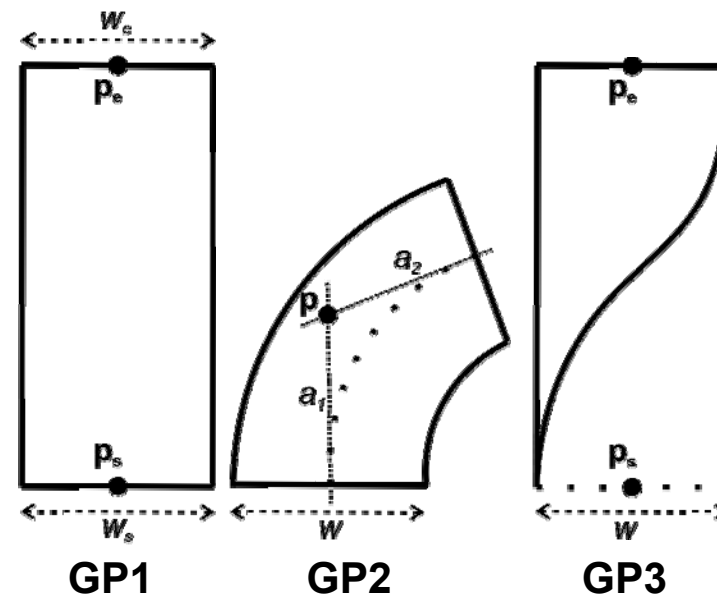
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Road Network Hypothesis Space

A Geometrical Modelling

B Conceptual Modelling

1. Qualitative spatial relations
2. Road Network taxonomy
3. Geometric Constraints
4. Road building / Semantic Constraints

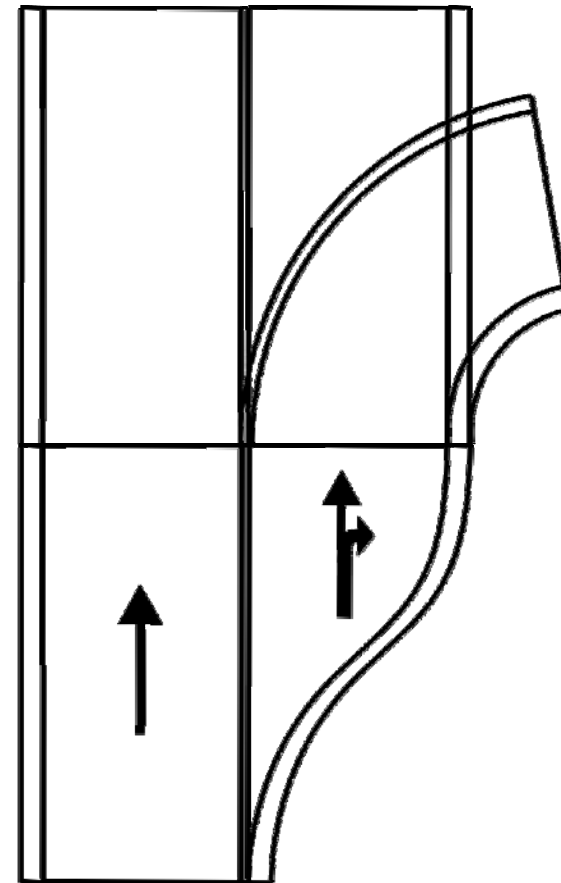


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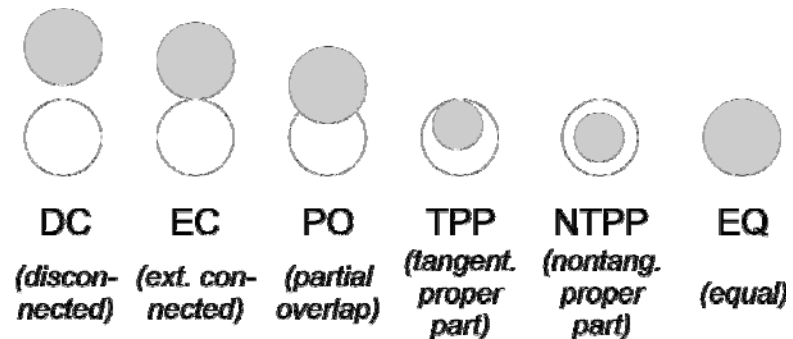


Road Network Hypothesis Space

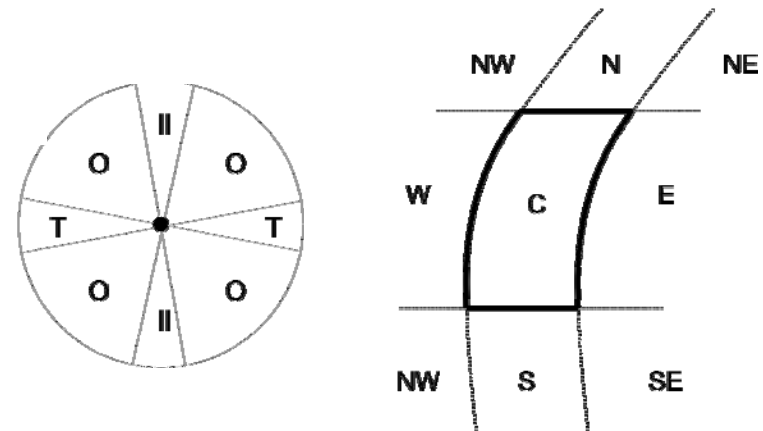
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B Conceptual Modelling

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a) Degree of overlap (RCC-Calculus)



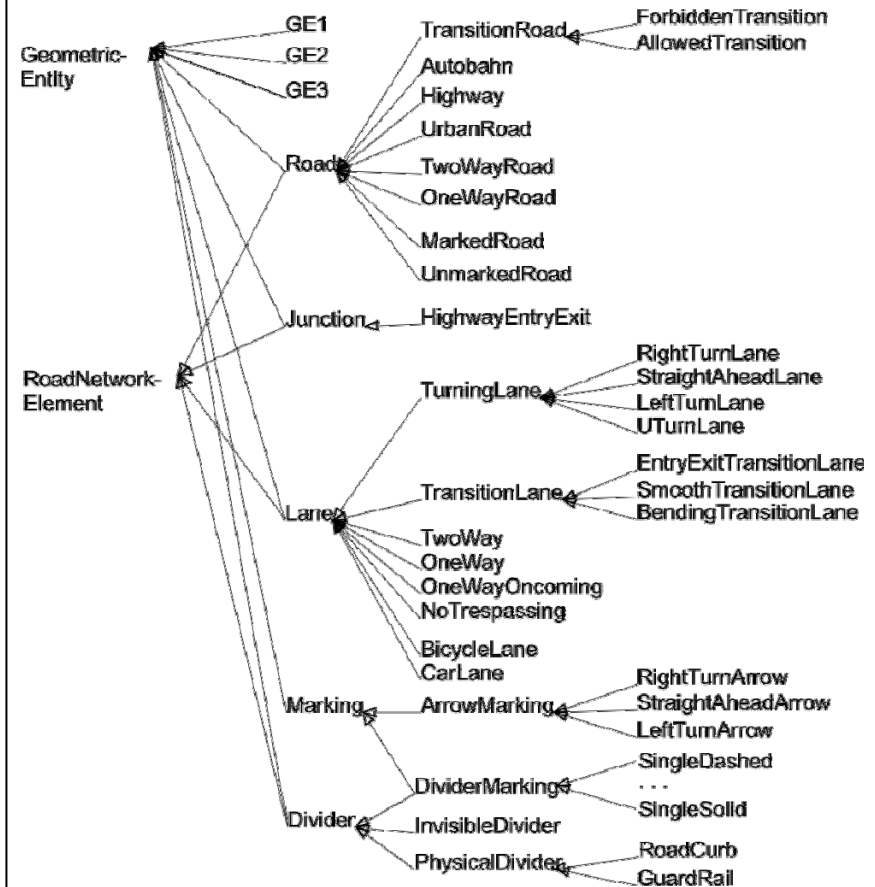
b) Rel. orientation c) Rel. position

Road Network Hypothesis Space

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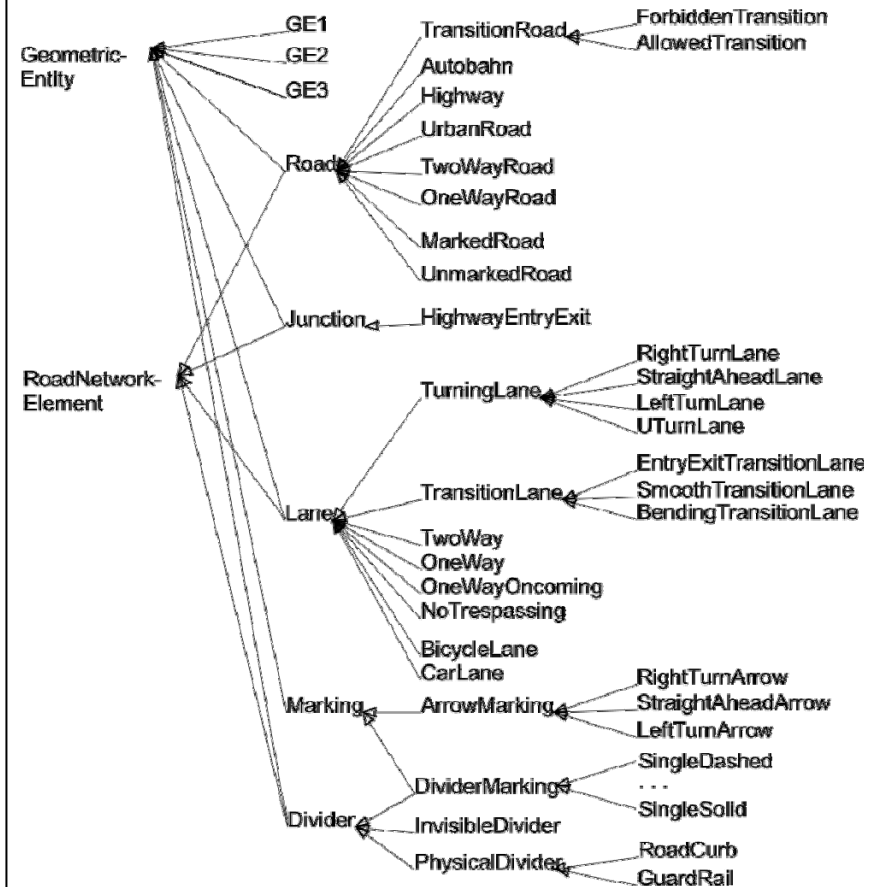


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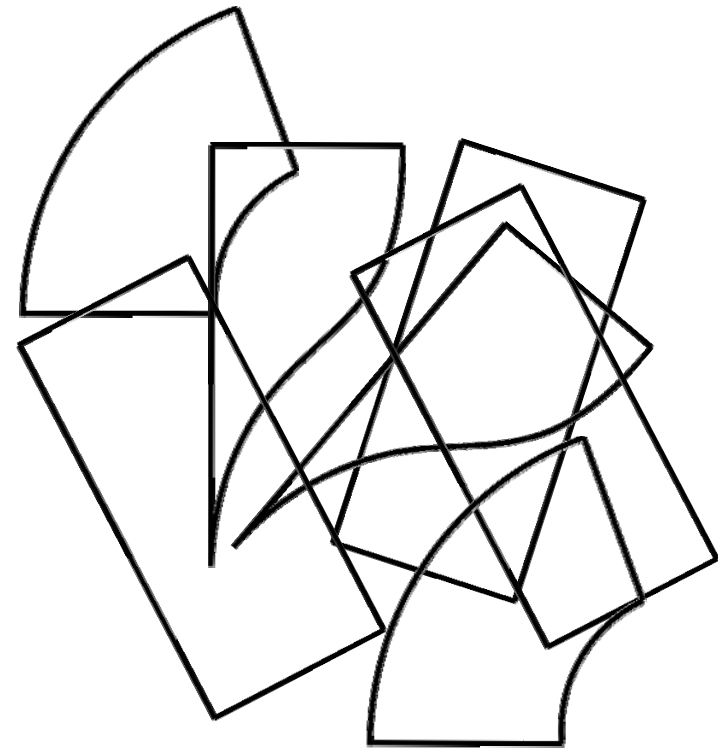


Road Network Hypothesis Space

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Arbitrary Sample

Road Network Hypothesis Space

A Geometrical Modelling

B Conceptual Modelling

1. Qualitative spatial relations
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// Lanes are externally connected

// only to lanes or dividers.

$\text{Lane} \sqsubseteq \forall EC. (\text{Lane} \sqcup \text{Divider})$

// Lanes only have arrows

// as proper parts.

$\text{Lane} \sqsubseteq \forall PP. \text{Arrow}$

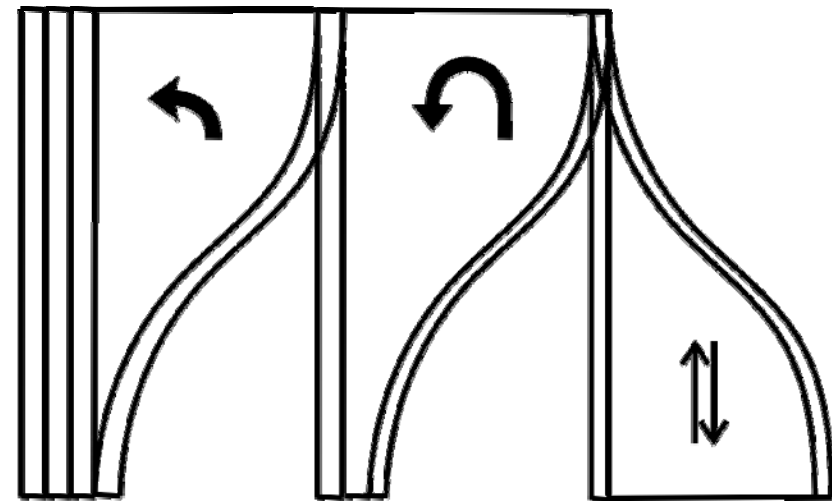
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Road Network Hypothesis Space

A Geometrical Modelling

B Conceptual Modelling

1. Qualitative spatial relations
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3. Geometric Constraints
4. Road building / Semantic Constraints



Arbitrary Road Sample

Road Network Hypothesis Space

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B Conceptual Modelling

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2. Road Network taxonomy
3. Geometric Constraints
4. Road building / Semantic Constraints

// Only right turn lanes can be right of right turn lanes

$\text{RightTurnLane} \sqsubseteq \forall \text{hasEastNeighbor. RightTurnLane}$

// A one way road does not have a uturn lane.

$\text{OneWayRoad} \sqsubseteq \forall \text{NTPP. } \neg \text{UTurnLane}$

// All autobahns and highways are one way roads.

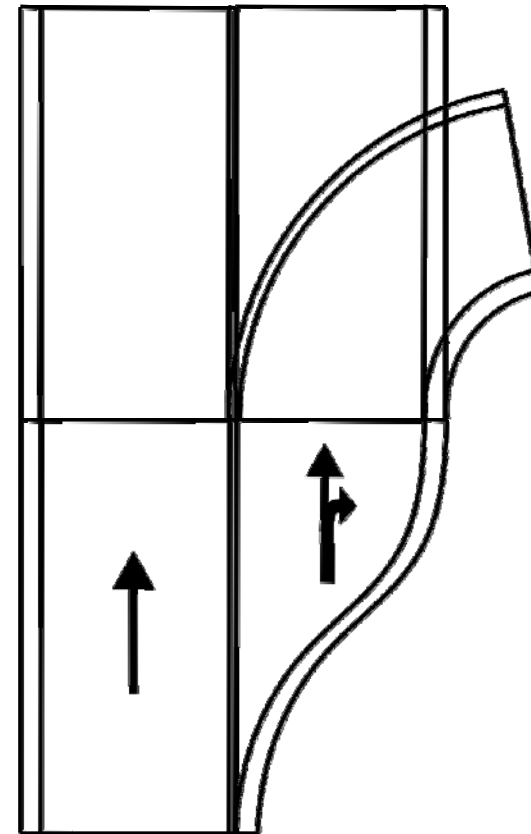
$\text{Autobahn} \sqcup \text{Highway} \sqsubseteq \text{OneWayRoad}$

Road Network Hypothesis Space

A Geometrical Modelling

B Conceptual Modelling

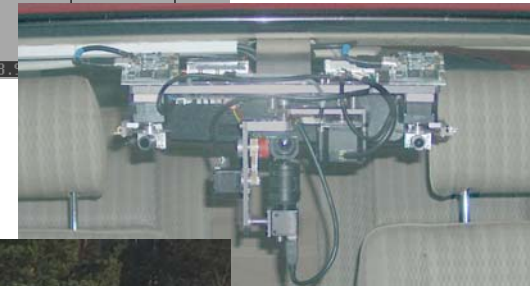
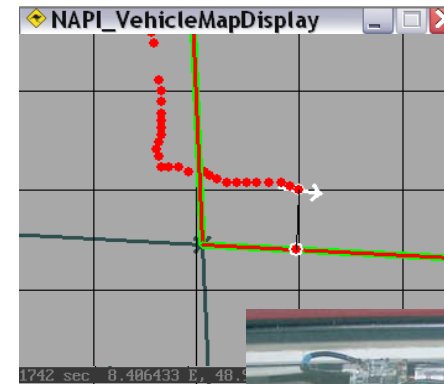
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Arbitrary Sample

Sensor Data Input

1. Digital Map
2. GPS & Map Matching
3. Video/Laser-based object detectors

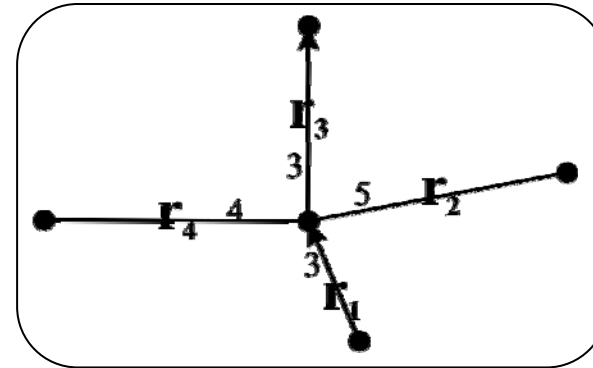


Sensor Data Input

1. Digital Map

2. GPS & Map Matching

3. Video/Laser-based object detectors



Symbol Grounding

[...] // *r4: not(OneWayRoad)*

[...] // *r4: has 4 Lanes*

[...] // *r4: leads to junction with 4 branches*

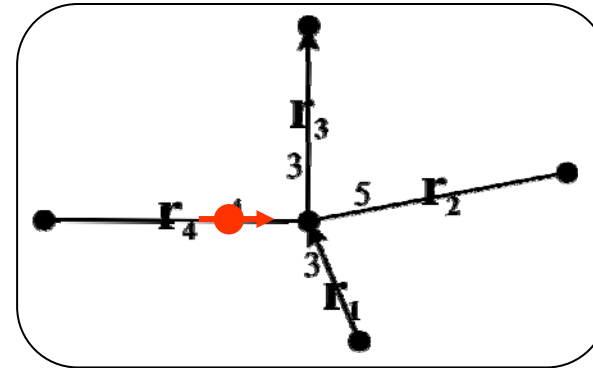
[...] // *Closed World Assumption*

Sensor Data Input

1. Digital Map

2. GPS & Map Matching

3. Video/Laser-based object detectors

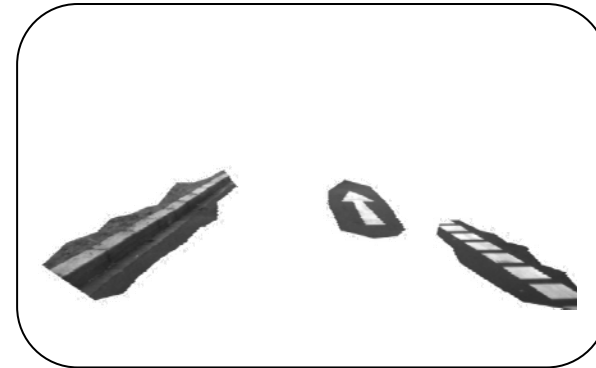


Symbol Grounding

```
egovehicle : Vehicle // create new vehicle  
(egovehicle, r4) : NTPP- // is inverse proper part of road r4  
(egovehicle, r4) : || // is parallel to road r4
```

Sensor Data Input

1. Digital Map
2. GPS & Map Matching
3. Video/Laser-based object detectors



Symbol Grounding

[...] // *i1: RoadCurb*
[...] // *i2: StraightAheadArrow*
[...] // *i3: DividerMarking50-20*
[...] // *state spatial relations*
[...] // *Open World Assumption*



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Inference

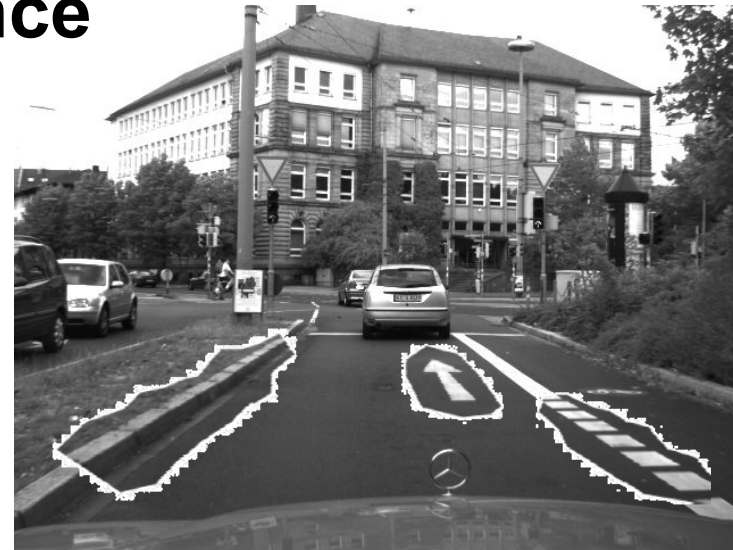
1. Draw all conclusions, given the KB (i.e. prior knowledge & sensor data)

→ Deduction

2. Generate all intersection hypotheses

→ Model konstruktion

→ Video-based hypothesis test / Deformable Model matching



$\mathcal{KB} \models l_{4.1} : \text{OneWayN} \sqcap \text{BicycleLane} \sqcap$
 StraightAheadLane

$l_{4.2} : \text{OneWayN} \sqcap \text{CarLane} \sqcap$
 StraightAheadLane

$l_{4.3} : \text{OneWayS}$

$l_{4.4} : \text{OneWayS}$

$\mathcal{KB} \models (\text{egovehicle}, l_{4.2}) : \text{NTPP}^-$

Inference

1. Draw all conclusions, given the KB (i.e. prior knowledge & sensor data)

→ **Deduction**

2. Generate all intersection hypotheses

→ **Model construction**

→ **Video-based hypothesis test / Deformable Model matching**

$$\mathcal{H} = \left(\begin{array}{c} \text{Abox } \mathcal{A}_1 \\ \vdots \\ \text{Abox } \mathcal{A}_n \end{array} \right)$$

Inference

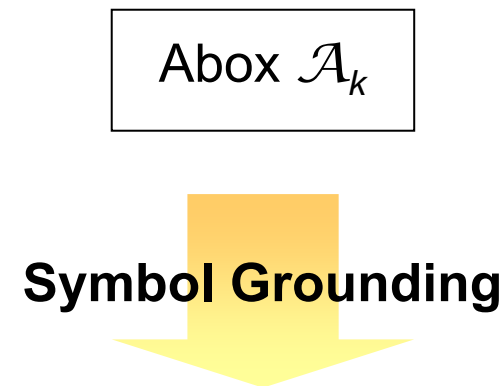
1. Draw all conclusions, given the KB (i.e. prior knowledge & sensor data)

→ **Deduction**

2. Generate all intersection hypotheses

→ **Model konstruktion**

→ **Video-based hypothesis test / Deformable Model matching**



Demo: Video


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Demo: Video

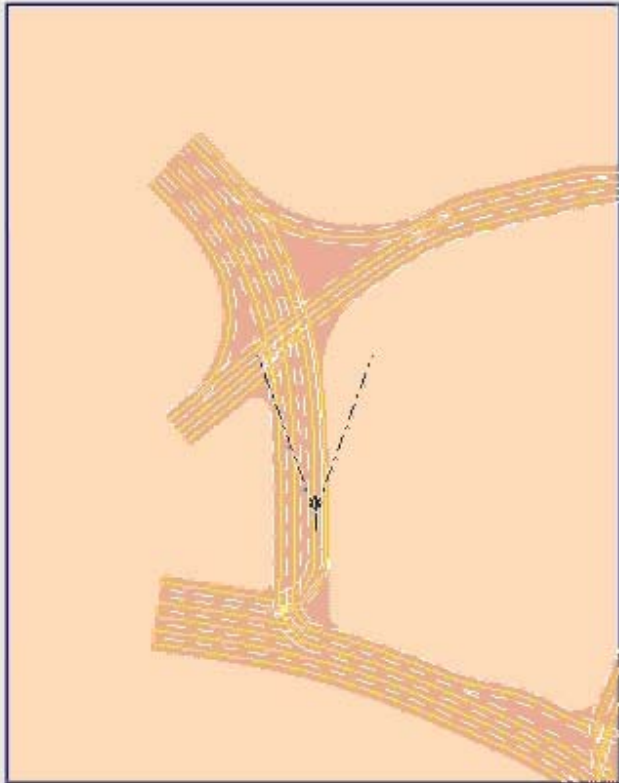
Sequence Nr:

Sensor I frame 1080 date 09-02-2005 time 15:39:12.6998



Latitude 49.0096 Longitude 8.4181

Orientation	
Abs.	Rel.
94.4256	0



Navigation controls: << >> [] || >

Buttons: Create Avifile Disp overlay Disp Lane Disp Border

Motivation:

- Road recognition works in toy worlds only
- A massive augmentation of prior knowledge is needed

Knowledge Representation:

- Introduction of DL as a formal, explicit, and readable knowledge representation formalism for real-world high-level scene interpretation
 - Development of a DL RoadNetwork KB
 - High-level conceptual constraints
 - Map, Positioning, Video/Laser sensors
- } constrain hypothesis space

Inference:

- Deductive reasoning for querying the KB
- Model Construction for returning the set of intersection hypotheses
- Natural handling of partial observability, differing abstraction layers and incremental additions

- **Hypothesize & Test**
- **Migration to probabilistic logic (MLN, SLP)**
- **Rule Learning from Training Data**

Thanks...

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Results 1-4: $\mathcal{KB} \models l_{4.1} : \text{OneWayN} \sqcap \text{BicycleLane} \sqcap$
 StraightAheadLane
 $l_{4.2} : \text{OneWayN} \sqcap \text{CarLane} \sqcap$
 StraightAheadLane
 $l_{4.3} : \text{OneWayS}$
 $l_{4.4} : \text{OneWayS}$

Result5: *// The vehicle is on lane $l_{4.2}$.*
 $\mathcal{KB} \models (\text{egovehicle}, l_{4.i}) : \text{NTPP}^- , \text{ iff } i = 2 .$

Proof sketch: The TBox of the \mathcal{KB} contains the following statements:

// Marking50-20 is a divider for bicycle lanes.

$\text{Marking50-20} \sqsubseteq \exists \text{isLaterallyConnectedTo.BicycleLane}$

// This type of arrow only occurs on car lanes.

$\text{Arrow} \sqsubseteq \forall \text{NTPP}^- . \text{CarLane}$

// Bicycle lanes are not next to each other.

$\text{BicycleLane} \sqsubseteq \forall \text{hasNeighbor.CarLane}$

These deductively lead to:

Result 5a:

// The driver's lane is a car lane (fortunately :)),

// and right of it, there is a bicycle lane.

$\mathcal{KB} \models \text{egovehicle} : \exists \text{NTPP}^-. (\text{CarLane} \sqcap$
 $\exists \text{hasEastNeighbor. BicycleLane})$

Result 5b:

// Right of the bicycle lane there can only be a car lane.

$\mathcal{KB} \models \text{egovehicle} : \exists \text{NTPP}^-. (\text{CarLane} \sqcap$

$\exists \text{hasEastNeighbor}. (\text{BicycleLane} \sqcap$

$\forall \text{hasEastNeighbor}. \text{CarLane}))$

Example: Proof of Result 5

// A lane with a straight ahead arrow is a straight ahead lane

[...]

// Right neighbors of straight ahead lanes are straight ahead or right turn lanes only.

[...]

// Bicycle lanes do not occur between lanes of the same turning lane type

[...]

Result 5c:

// Right of the bicycle lane can only be a right turn lane.

$\mathcal{KB} \models egovehicle : \exists NTPP^-(CarLane \sqcap$

$\exists hasEastNeighbor.(BicycleLane \sqcap$

$\forall hasEastNeighbor.RightTurnLane))$

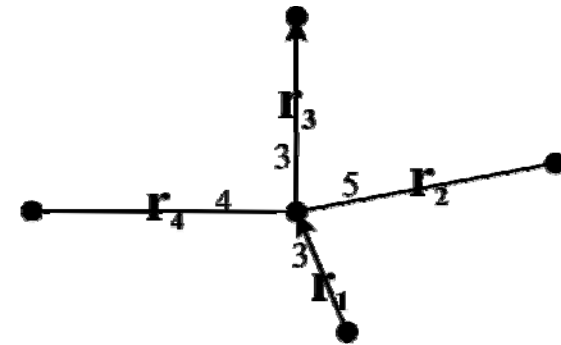
Example: Proof of Result 5

- Intermediate result:
As r1 is a OneWayRoad towards the junction, r4 cannot have a right turn lane.

Result 5d:

// There is no more lane right of the bicycle lane.

$\mathcal{KB} \models egovehicle : \exists NTPP^-(CarLane \sqcap$
 $\exists hasEastNeighbor.(BicycleLane \sqcap$
 $\neg \exists hasEastNeighbor))$



Example: Proof of Result 5

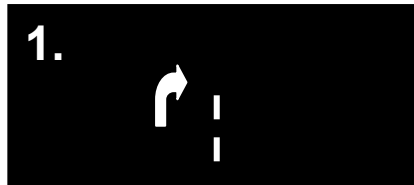
- *Closed World Assumption* for map data

==>

// The vehicle is on lane $l_{4.2}$.

$KB \models (egovehicle, l_{4.i}) : NTPP^-$, iff $i = 2$. \square

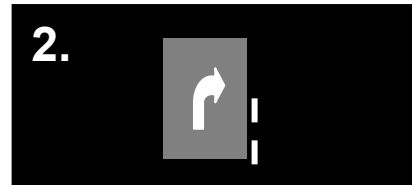




1.

Evidence (ABOX):

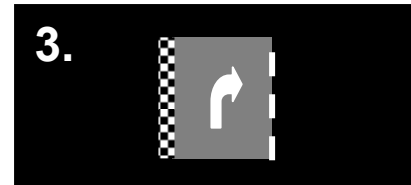
```
RightTurnArrow(
    arrow1);
SingleLongDashed
Divider(divi-
der1);
```



2.

Hypothesization:

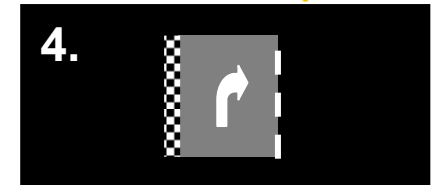
```
Lane(lane1);
hasMarking(
    lane1,
    arrow1);
```



3.

Hypothesization:

```
isConnectedWith-
RightDivider(
    divider1);
isConnectedWith-
LeftDivider(
    divider2);
```

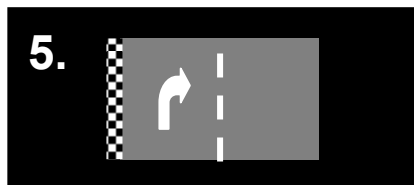


4.

Hypothesization:

```
Road(road1);
isPartOf(lane1,
    road1);
```

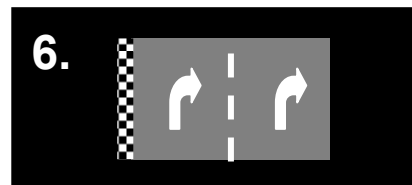
**non-trivial
Hypothesization:**



5.

Hypothesization:

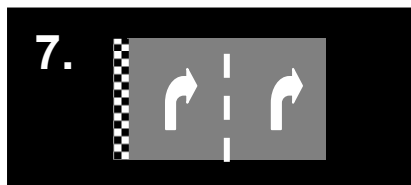
```
Lane(lane2);
hasRightNeighbor
(lane1, lane2);
```



6.

Deduction:

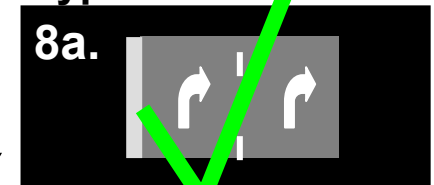
```
isPartOf(lane2,
    road1);
RightTurnLane(
    lane2);
```



7.

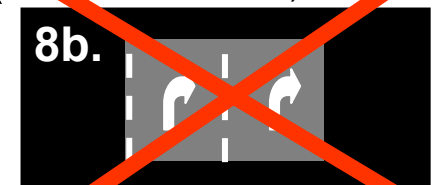
Deduction:

```
MarkedRoad(
    road1);
```



8a.

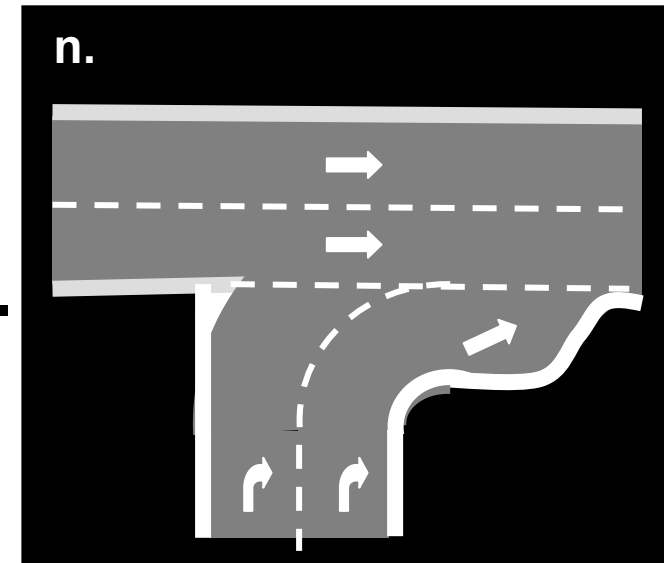
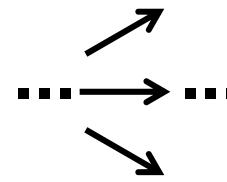
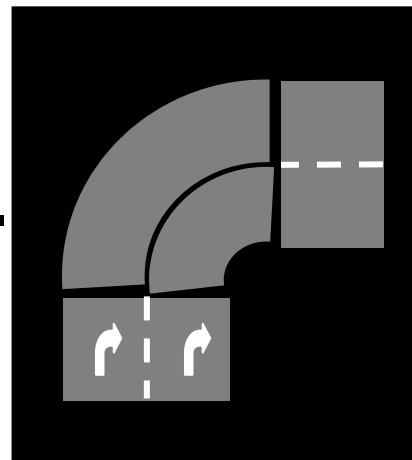
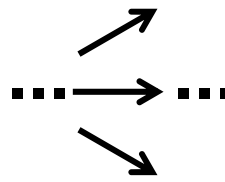
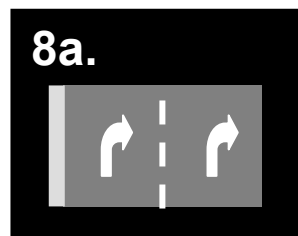
```
PhysicalDivider(
    divider2);
```



8b.

```
DividerMarking(
    divider2);
```

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model complete