Laser-Based Feature Extraction and Pattern Recognition in Intersection Management Systems

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Pattern Recognition, 2014



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Context Problem Statement Aims and Conditions

Context

Master's Research Project: Multisensor Architecture for a Vehicular Intersection Management System

Transportation Systems

Issues in traditional transportation systems

Congestion

Traffic rules violation

Vehicle interaction

Transportation Systems

Issues in traditional transportation systems

Congestion

Traffic rules violation

Vehicle interaction

Intersections are critical places in transportation systems

Intelligent Transportation Systems

Objectives of ITS

Increase safety
Increase efficiency

Reduce costs

Intersection Management Systems

Tasks

Traffic Monitoring
Traffic Management
Warning Advertisement

Intersection Scenario

Pedestrians, Vehicles (Cars, Two-wheeled vehicles, Big vehicles)

Recognition, Classification, Tracking

Incident detection, Intersection Management

Context Problem Statement Aims and Conditions

Main Objective

To develop a feature extraction and pattern recognition laser-based module for an intersection management system

- Review of laser-based feature extraction and pattern recognition in ITS and IMS

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- Evaluate pros and cons of the reviewed methods

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- Evaluate pros and cons of the reviewed methods
- Implement at least one method
- Evaluate implemented module and compare it with similar developments

Context Problem Statement Aims and Conditions

Conditions

- The information source will be a dataset.

Conditions

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- [New!] Just one laser.

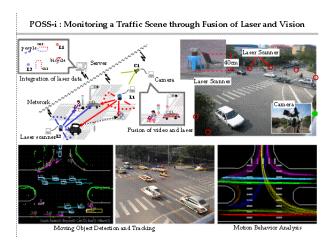
Research Groups

- PKU Omni Smart Sensing (POSS) Research group at Peking University (POSS-i project)
- Institute of Measurement, Control and Microtechnology at Ulm University (Ko-PER program)

PKU Omni Smart Sensing (POSS)

- POSS is leaded by Prof. Huijing Zhao, Ph.D.
- Focus on perception technologies using an intelligent vehicle, a network sensing system or a collaboration of them

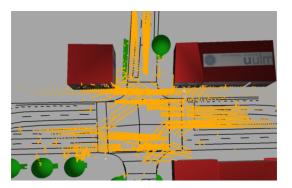
POSS-i



Ko-PER

- Ko-PER from Cooperative Perception
- Included in Forschungsinitiative Ko-FAS from Bundesministerium für wirtschaft und Technologie (Germany)
- Cooperative and collaborative sensors system for perception and preventive road safety.
- Daniel Meissen from Ulm University as leader researcher.

Projects



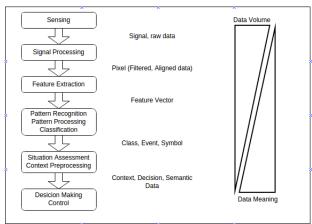
3D-recreated intersection scene with laser beams depicted [Meissner12, 13a, 13b, 13c, 14][Striegel13]

Applications, Methods and Techniques

Project	POSSi	Ko-PER				
Applications	Recognition, Classification and Tracking of Vehicles and Pedestrians					
Methods and Techniques	ClusteringMarkov ChainsKalman FilteringAdaBoost	 DBSCAN Multi-object Bayes Filter Sequential Monte Carlo Methods Dempster-Shafer Theory Multiple-Model Probability Hypothesis Density Filter (in Gaussian Mixture representation) 				

POSSi and PKU projects comparison

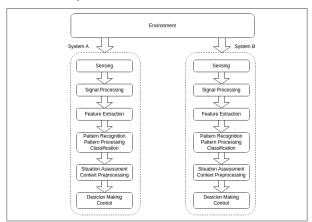
Typical System for one source of data



Single source system block diagram

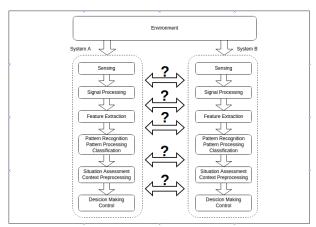


Multisensor Data System



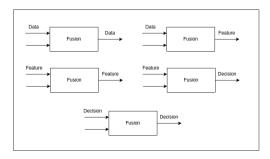
Multisensor system block diagram

How to fuse information?



Multisensor system block diagram

Fusion Levels



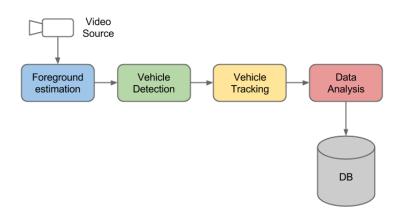
Fusion Levels [Luo11]

Fusion Algorithms

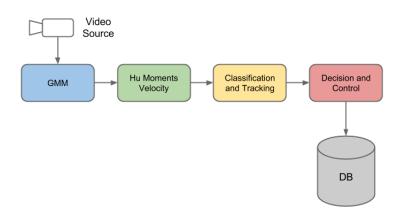
Low level fusion		Medium level fusion	High level fusion	
Estimation methods		Classification methods	Inference methods	
Recursive: • Kalman filter • Extended Kalman filter • Extended Kalman filter Non-recursive: • Covariance intersection • Covariance union		Parametric templates Cluster analysis K-means clustering Learning vector quantization Kohonen feature map Artificial neural network Support vector machines	Bayesian inference Particle filters Dempster-Shafer theory Expert system Fuzzy logic	

Classification of Fusion Algorithms [Luo11]

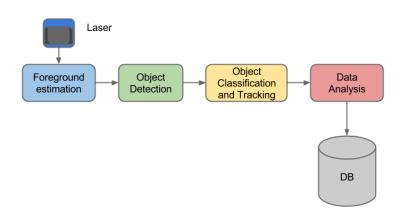
Video-Based System Block Diagram



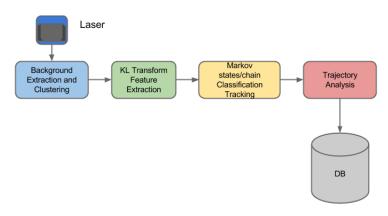
Video-Based System Block Diagram



Laser-Based System Block Diagram



Laser-Based System Block Diagram



Based on [Zhao06]



Background Extraction Object Classification Feature Extraction Implementation Next Steps

Background Extraction

- Histogram-based background extraction
- Done for each angle
- When a pick value is detected, tells that an object is detected

Classification

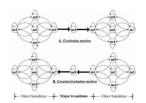
- Classes are proposed based on distribution of points in clusters
- Karhunen-Loeve Transform to detect number of axis

Objects in	Example of laser data					Class
cross road t ₁	t_1	t ₂	t ₃	<i>t</i> ₄	ts	definition
car				i	:	2-axis
bicycle						1-axis
pedestrian	٠.	٠.	٠.	٠.	٠.	0-axis

Markov States



There are 8 patterns that can happen

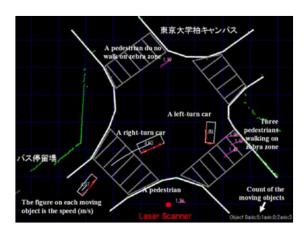


Possible transitions

Features

- Normal Vectors
- Number of axis
- Axis lengths
- Directional vector, Motion speed
- Markov States

Dataset



Capture of application [Zhao06]

Next Steps

- Implement Dataset handler
- Implement Clustering and KL Transform to classify in 0, 1 or 2 axis object
- Get features from objects and obtain trajectory

Background Extraction Object Classification Feature Extraction Implementation Next Steps

Clustering



Background Extraction Object Classification Feature Extraction Implementation Next Steps

Clustering



Background Extraction Object Classification Feature Extraction Implementation Next Steps

Clustering



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