OIS -

AN OPTICAL INFORMATION SYSTEM FOR ROAD TRAFFIC MEASUREMENT AND MANAGEMENT

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ABSTRACT:

OIS is a new Optical Information System for road traffic observation and management. The complete system architecture from the sensor for automatic traffic detection up to the traffic light management for a wide area is designed under the requirements of an intelligent transportation system. Particular features of this system are the vision sensors with integrated computational and real-time capabilities, real-time algorithms for image processing and a new approach for dynamic traffic light management for a single intersection as well as for a wide area. The developed real-time algorithms for image processing extract traffic data even at night and under bad weather conditions. This approach opens the opportunity to identify and specify each traffic object, its location, its speed and other important object information. Furthermore the algorithms are able to identify accidents, and non-motorized traffic like pedestrians and bicyclists. Combining all these single information the system creates new derivate and consolidated information. This leads to a new and more complete view on the traffic management it is necessary to improve the modelling and forecasting of traffic flow. Therefore the information of the current Origin-Destination (OD) flow is essentially. Taking this into account OIS also includes an approach for anonymous vehicle recognition. This approach is based on single object characteristics, order of objects and forecast information, which will be obtained from intersection to intersection.

INTRODUCTION

Traffic observation, control and real-time management is one of the major components within future intelligent transportation systems. But the state-of-the-art traffic measurement equipment, e.g. induction loops, does not suffice anymore the growing demand of transport research and traffic control.

The project OIS uses optical and informational enabling technologies for an automatic traffic data generation with an image processing approach. Its main purpose is to acquire and evaluate autonomously traffic image sequences from roadside cameras. Traffic parameters will be obtained from extracted and characterised objects of this image sequences. To meet these requirements, numerous image processing algorithms have been developed since more then 20 years (e.g. special issue [Broggi and Dickmanns 2000]), with simple web-cameras and more complex systems (e.g. [Pedersini et al. 2001]).

Traffic scene information can be used to optimise traffic flow on intersections during busy periods, identify stalled vehicles and accidents, and is able to identify non-motorised traffic like pedestrians and bicyclists. Additional contributions can be obtained for the determination of the Origin-Destination (OD) matrix. The OD matrix contains the information where and when the traffic participants start and end their trip and which route they have chosen. OD matrix is one basic element for an optimised modelling and forecast of traffic flow. The estimated traffic flow is necessary for a dynamic wide area traffic control, management and travel guidance. Furthermore, the recent advances in computational hardware can provide high computational power with fast networking facilities at an affordable price. The availability of specific solutions in the low-cost general-purpose range allows special image processing and avoids some basic bottlenecks.

A couple of traffic data measurement systems already exist. Best known is the induction loop. Induction loops are embedded in the pavement. They are able to measure the present of a vehicle, its speed and rough classification. These are only local information but for a wide area traffic management a big coverage of the area is needed.

Another approach is based on the idea that moving vehicles transmit information about there position and velocity via mobile communication, e.g. GSM to a traffic management centre. These data are called Floating Car Data (FCD). To get an overview of the traffic situation of a complete city at any time a huge number of vehicles has to be equipped with hardware and mobile communication units. To get reliable data every one minute the current position and velocity of the vehicles is needed. This causes enormous costs for the mobile communication. The FCD approach provides spatial traffic information, but the spatial and timely resolution doesn't fit the requirement of a traffic signal control.

OIS as a new and innovative traffic observation system that opens the opportunity to deliver all necessary input for a local traffic signal control as well as for a dynamic wide area traffic management system.

Next challenge is the implementation of OIS in a wide area city.

SYSTEM REQUIREMENTS

A modern system for traffic control and real-time management has to meet the following requirements:

- · Reliability under all illumination and weather conditions
- Working period non-stop, 24hours 7 days a week
- complete overview over the intersection from at least 20m in front to 20m behind
- · working in real-time

• real-time (every half second a complete data set of the traffic situation) on an intersection

This requirements should be taken into account for the design of all part of the system, which covers all procedures and processes from image data acquisition, image processing and traffic data retrieval up to traffic control. For realizing an operating system for 24 hours and under different weather conditions infrared cameras should be used. Algorithms had to be developed for a special camera arrangement (with real-time demands) for vehicle detection and deriving relevant parameters for traffic description and control.

SYSTEM OVERVIEW

To get a complete overview over the intersection from at least 20m in front to 20m behind it is necessary to have more than one camera. The number of required cameras depends on the intersection geometry and installation possibilities for the cameras on house walls or lampposts.

The time synchronous image data acquisition from one observation point with different cameras will be done in a so-called camera node. The camera node is part of the OIS-philosophy and consists of different sensors as a combination of VIS and IR-cameras.

To fit the real-time processing requirement, time consuming image processing parts will be implemented in a special real time-unit.

This approach is in implementation at the moment for two intersections in Berlin, Germany equipped with camerasystems. The architecture of the complete system is shown in figure 1.



Figure 1: System design for the test bed

As shown in fig.1 the camera nodes are part of a hierarchy, which are linked via Internet or Wireless LAN with the computer systems on the intersection and the management centre. Due to the limited data transfer rate and possible failure of a system on an intersection the camera node works independently. Image processing is decentralized and will be done in the camera node, so that only objects and object features will be transmitted. Together with error information the object data will be collected on the next level in the hierarchy.

For synchronization purposes time signals can be incorporated over a network or from independently received signals (like GPS).

The camera node is part of a hierarchy starting with camera nodes, junctions, subregions, etc. The next so-called junction level unifies all cameras observing the same junction. Starting with the information from the camera level, relevant data sets are fused in order to determine the same objects in different images. After that, these objects are tracked as long as they are visible. So traffic flow parameters (e.g. velocity, traffic jam, car tracks) can be retrieved. Level number 3, the so-called region level, uses the extracted traffic flow parameters out of the lower level and feeds this data into traffic models in order to control the traffic (e.g. switching the traffic lights). Additional levels can be inserted.

For test applications the camera node will generate a compressed data stream. For typical working mode only object features in an image should be transmitted and collected in a computer of the hierarchy.

Generally, optical sensor systems in the visible and nearinfrared range of the electromagnetic spectrum have reached a very high quality standard, which meet the requirements even for high-level scientific and commercial tasks, above all concerning the radiometric and geometric resolution and data rate. Otherwise sensors working in the thermal infrared range (TIR) are still a research topic for traffic applications.

Technology development for the next few years will not be focused on higher resolutions or faster read-outs, because for most applications the performance of these sensors is sufficient. The emphasis will be put on smart, intelligent sensor systems with different measurement parameters (e.g. different resolutions, different spectral sensitivities) which are connected within a network similar to the internet and being able to convert the incoming physical signals not only to digital data but to process them to user needed information. Therefore, image fusion as well as fast and reliable algorithms are needed preferably near the sensor itself. Real-time processing and programmable circuits will play an important role.

HARDWARE CONCEPT

The hardware concept is oriented on the requirements, as described above. The system has to operate 24 hours and 7 days a week and to observe the whole junction and at least 20m of all related streets.

To improve the opportunities for image acquisition more than one camera system for one observation standpoint should be in use. Such a camera node fits the first requirement with a high (spatial) resolution camera and a low resolution thermal infrared camera. Also stereo and distance measurements are possible with two identical cameras. The camera node is able to acquire data from up to four cameras in a synchronised mode.

The camera node has real-time data processing capabilities and allows synchronous capture of GPS and INS (inertial navigation system) –data, which should be external mounted.

Due to limitations of camera observing positions and possible occlusions (e.g. from buildings, cars and other disturbing objects) for intersection observation mostly more than one standpoint for a camera node (camera-system) is necessary. Therefore communication or data transmission between camera nodes and the computer in the next higher hierarchy becomes critical. To reduce data volume in the network the real time data processing capability of the camera-node is used to speed up image processing and to transmit only object data. For the realtime unit a hardware implementation was chosen. Large -free programmable logic gate arrays (FPGA) are available now. A programming language is available (VHDA) and different image processing algorithms are implemented.

DATA PROCESSING

The processing of data in the sensor web follows successive steps. In the first step image data are generated and preprocessed. After that the objects are extracted out of the images. In the last step all the object features from different camera nodes will be collected, unified and processed into traffic information. The different data processing is optimised to the logical design of the system. There are operations for the system configuration, for the operational working for the data mining and visualisation. Results are new data and information. A complex data management system regulates the access, the transmission, handling and the analysis of the data.

Image Processing

The essential processing steps are the elimination of noise and systematic errors, compression/decompression, higher level image processing and spatial (or geo-)- and time-referencing. The last point is necessary to determine space and time coordinates of observed objects as an essential feature for data fusion.

Most of the work on vehicle detection or recognition was done on ground images, mainly as pre-processing before tracking for surveillance or traffic applications ([Betke et al. 2000], [Coifmana et al. 1998], [Malik et al. 1997]). Different approaches, e.g. finding edges [Canny 1986], [Fathy and Siyal 1995], [Ikeda et al. 1994] or deformable model for vehicle [Ferryman et al. 1995] can be found. Generally, there are problems in image processing with car occlusion [Koller et al. 1993], [Ikeda et al. 1996] and shadows [Fung et al. 2002]. Beside stationary image acquisition from ground also from moving platforms and arial images [Chellappa 1996] are used. Image processing approaches will also be used for detecting lanes and obstacles by fusing information [Beauvais and Lakshmanan 2000].

Image processing for OIS was described in detail in [Reulke, et. al., 2002]. The main processing is a classical object detection and identification task. Following major problems have to be solved

- Object discrimination from a spatial and time variable background (cross, street, buildings, etc.)
- Removing disturbing structures between object and camera, as well as shadow regions around the object
- Identification of cars in a raw, which are occluded by other

The first problem can be solved at least in two different ways: 1) Working with the image sequences, which are subtracted from the image before or

2) following the time changing background.

For the first approach background can be eliminated very simple, but stalled vehicles are invisible. The other approach is an on-going update of the background. This needs much more expense, but allows a detection of moving and also stalled cars. As a result of this procedure, background can be subtracted from the actual image and objects can be derived as shown in figure 2. Additional morphological operation removes clutter and close objects structures. The right image shows the grey coded objects after labelling.



Figure 2: Object detection in a traffic scene (left: original image, right processed image)

The other major problem occurs after detecting the object. To determine size and shape, disturbing effects like shadows have to be removed. A straightforward way is the analysis of grey and colour values, as well as texture within the found object boundaries.

Data and object fusion

To ensure an operation system 24 hours a day even under bad weather conditions, e.g. rain, fog, and at night the fusion of a visible and an IR sensor is a promising approach.

The fusion is possible on different stages of information processing:

- data level (e.g. image data from different cameras)
- object level (objects extracted from the image data)
- information level.

Image matching and registration as one part of the data fusion is a procedure that determines the best spatial fit between two or more images acquired at the same time and depicting the same scene, by identical or different sensors. To fuse the different images and/or object data a synchronisation of time is necessary. 'Time synchronization can be realized by internal clocks (e.g. computer) or external time information (e.g. GPS). For the application of OIS we merge images on camera node level and fuse the position information and object features on junction or region level. The object information are fed in from different camera nodes. Both procedures should be explained more detailed.

Fusion on data level: To fit the requirement for 24-hour observations, typical CCD-cameras fail because of limited illumination of the objects. Car headlights and rear a lamp seems not be sufficient. To overcome this problem the car self-radiation, which has a maximum in the thermal spectral region (TIR), can be use. There are a bunch of different detectors sensitive in this spectral range. Most of the sensors are expensive and needs additional cooler. Recent developments shows, that bolometer arrays are a candidate for cheaper and uncooled detector arrays. Therefore a camera development was started, which gives full access on the sensor, control, data correction and dataflow.

First experiments were done with a commercial system. An example of the data and the fusion of both on data level is shown in Fig. 3.



Figure 3: Sensor fusion of VIS an IR images (I)

Observation was done on late afternoon. The left image is a typical CCD-image. Contrast becomes smaller, only reflections from sun glitters on car roofs are visible. The middle image was taken from a bolometer sensor. The whole intersection and the street are visible. The right image is the fusion of both. For visualization grey level image was put in the green channel and the TIR image in the red. Merging the visible image and the IR-image a affine transformation was used.

To fuse data from different sources is obvious, but need spatial and time synchronisation, because of different imaging system and observation conditions. For this example the synchronisation task is based on manual procedures, like finding equivalent points in both IR and VIS images and the calculation of the necessary transformation. Especially the automatically spatial synchronisation is a research topic. All these operations are done in each camera node.

Another advantage of TIR images shows the following image pair, which was taken from the same observation point, but at daytime.



Figure 4: Sensor fusion of VIS an IR images (II)

Figure 4 shows an example of fusing RGB and TIR images at daytime. After coregistration the IR-image to the RGB-image and applying affine transform, a direct comparison is possible. In difference to figure 1 the infrared image was fit directly into the grey level image (a colour separation of the RGB-image).

The RGB image has a much more higher resolution than the TIR-image (720x576 pixel). Spatial and true colour object data can be derived from this image. The TIR-image has a smaller resolution (320x240 pixel). The combination of both images gives a more complex result, because of the thermal features, which can be observed on the engine bonnet and as reflected radiation from under the car. These are also new features for object detection and description within the image-processing task.

Object Data Fusion: Occluded regions can only be analysed with additional views on these objects. Because camera nodes always transmit object information, a data fusion process on object level is necessary.

The object list from different camera nodes has to be analysed and unified. Object features like position, size, and shape vary from each view angle. Therefore the different object features from different perspectives have to be compared. The result of this operation is one traffic object with an exact position, size and shape, etc. The assumptions for this operation are time and spatial synchronised image data. The sequential processing of this list allows the derivation of more detailed information, e.g. track of the traffic participants, removing erroneous objects, etc.

Data Acquisition and Georeferencing

Transformation from image to world coordinates is an essential need in order to calculate metric based traffic data. Standard photogrammetric procedures for the transformation of coordinates within monocular images are used.

Basic assumptions are well distributed and accurate measured ground control points (GCP) in object and image space as well as an exact camera calibration. The GCP's are calculated via DGPS within WGS84 and UTM-projection.

The calculated camera calibration parameters (interior and exterior orientation) and image coordinates are input for the transformation equations.

Due to monocular image accusation the object or vehicle positions have to be projected on a XY-plane in object space. The vehicle positions projected on that plane depend on the camera distance, the camera declination and the position of the point within the vehicle representing its position.

Once the camera calibration is set the vehicle positions can be transformed to world coordinates within image sequences until the camera position changes. Additionally intersection geometry for example lanes etc... can be transformed from image into object space.

AN ADAPTIVE AND HIGH DYNAMIC NETWORK CONTROL

The in this project developed video based traffic sensor creates possibilities for new concepts of traffic control for intersections and wide area networks. This approach also includes the development of a new dynamic and adaptive traffic control models for traffic lights.

State-of-the-art in traffic observation is the induction loop. Induction loops are embedded in the pavement and register about 1*2 m. This kind of sensor is able to measure the present of a vehicle, its speed and rough classification. In a next processing step it is possible to calculate time intervals between vehicles. This data is needed to control traffic light signals on intersections.

For a real dynamic and demand based traffic signal light control, you would need the data of numerous induction loops on one single intersection. This is neither efficient nor realisable.

OIS sensor web offers a new kind of traffic data/information. It is based on so called traffic-actuated signals. This means, the system is detecting information about the real-time traffic situation on an intersection automatically. For example the length of the queue for all different lanes, the traffic flow at the intersection, the density of traffic and the current velocity of each vehicle.

The OIS sensor web automatically processes data-position, speed vectors of each vehicle, queue length as well as other relevant features. This leads to a complete and real-time "overview" at least 20 m before and behind an intersection. Based on this new quality of data, new approaches are developed to control traffic light signals on dynamic demand.

At the moment most of traffic lights are controlled by two different ways:

- 1. control by "fix-time signals"
- 2. control by so called "actuated signals"

"Fix-time signals" means: green and red time is fixed over the time and independent of the actual situation on the intersection. "Actuated signals" means: a number of fix-time signals are used

for different demands and situations.

At this time there are nearly no sensor webs or measurement systems available, that are able to measure each object on a intersection and that provides all necessary spatial information for a real dynamic traffic control system.

OIS is a system that is able to acquire features like size, shape

and other object features to classify and identify traffic objects. Summarising this, OIS gives a complete data information (overview) over a whole intersection, not just a single point.

Therewith the system creates OD-flows at each intersection of a street network. As mentioned above these flows are basic elements for a new dynamic traffic control model. Additionally the sensor system collects a lot of different features of each object, like shape and colour. So the vehicles can be recognized on the surrounding intersections. Based on this feature a dynamic OD-matrix not only for one intersection but also for the whole street network will be developed. Combining this information with the time stamp and the current velocity of each object it is furthermore possible to calculate the real dynamic or middle travel time between to intersections.

The dynamic and adaptive traffic control model for the single intersection is based on the current number of vehicles, length of all queues, the velocities and also on the classification of the vehicles. Between two adjacent intersections an information transfer is possible. Therewith the control model enables a short time prediction of the next queue lengths.

The Institute of Transport Research develops an open source traffic simulation named **SUMO** tool, (http://ivf.dlr.de/berlin/projekte/sumo/index.html). SUMO is a microscopic traffic simulation tool for large scale networks based on the traffic flow model from Stefan Krauss. This simulation will be integrated in the 'OIS-traffic control model'. The outcome of the combination of this with the current and dynamic OD-matrix will be a new and high dynamic traffic management model for large scale street networks traffic control by traffic lights. In these projects the models will be tested in a simulation shell based on traffic data of the equipped Berlin intersections.

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