

DriveSafe: an App for Alerting Inattentive Drivers and Scoring Driving Behaviors

Luis M. Bergasa, Daniel Almería, Javier Almazán, J. Javier Yebes, Roberto Arroyo

Department of Electronics, University of Alcalá, Madrid, Spain



Universidad de Alcalá

• **Contact:**
Luis Miguel Bergasa
Phone: (+34) 91 885 65 69 / 40
Fax: (+34) 91 885 65 91
Email: luism.bergasa@uah.es

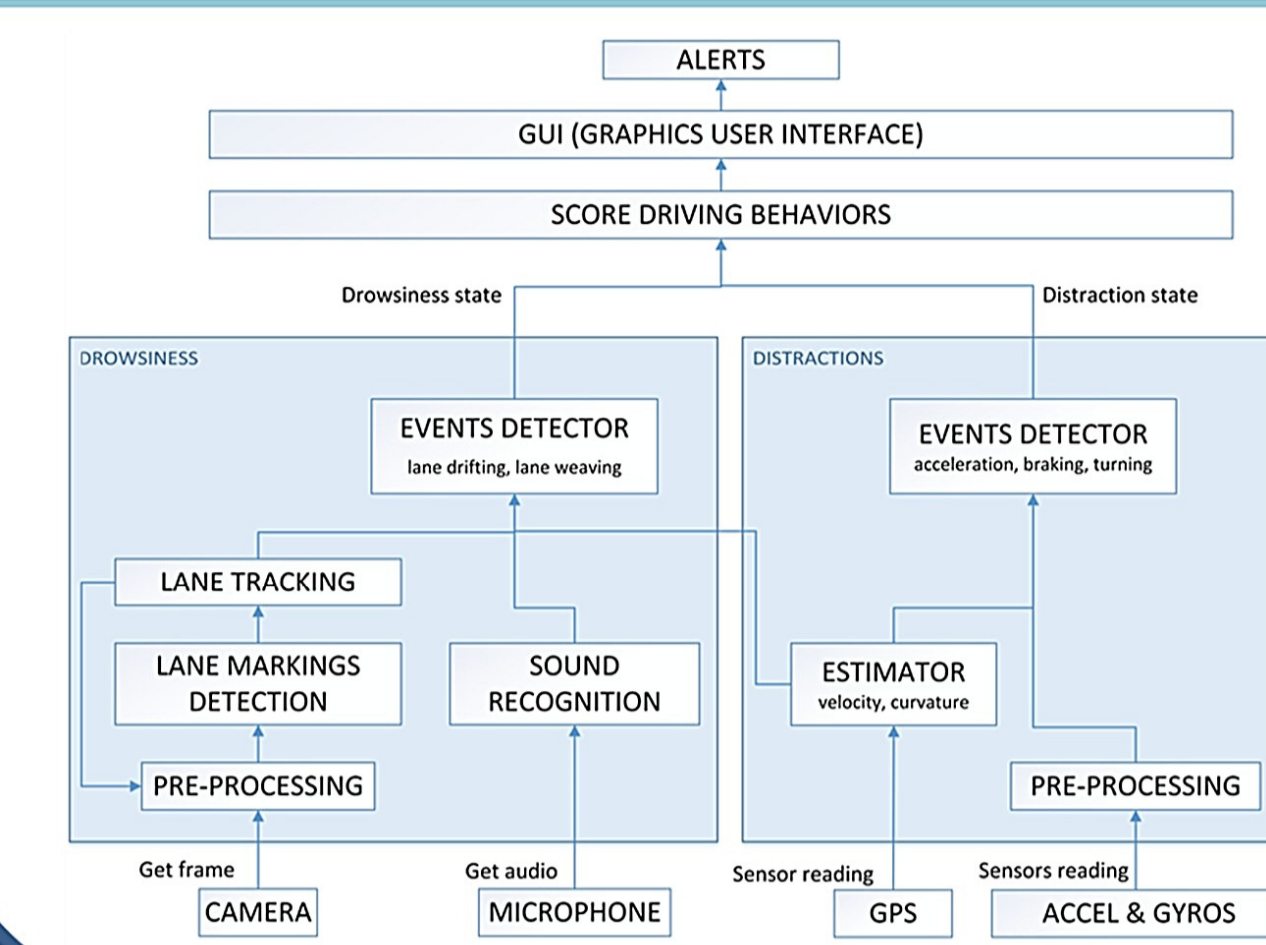
Problem Description

- Developing technologies to **detect inattentive drivers** is **essential** to **avoid vehicle accidents** and to **stimulate safe driving practices**
- ADAS** are commonly fitted only in **top-end vehicles**
- In the **motor insurance sector** consumers **reject the Pay-As-You-Drive** due to the required installation of intrusive **"black-boxes"** in vehicles and their additional cost
- DriveSafe** aims to **mimic some safety features** found in many **top-end vehicles** but using a commodity **iPhone**, and persuade insurers this app is an **interesting alternative** to **"black-boxes"**

Overview

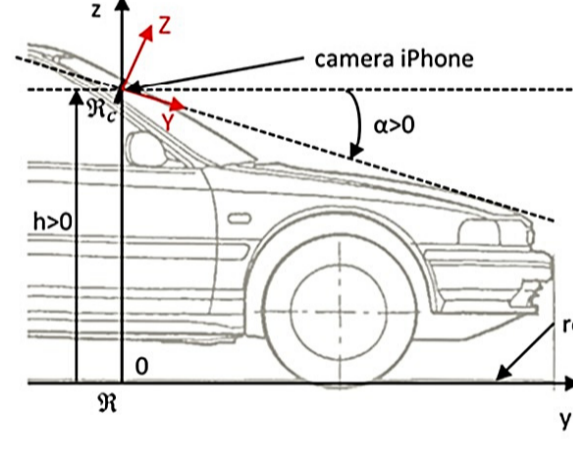


System Architecture



Drowsiness Detection

- Based on vehicle lane position and lane changes** obtained by computer vision
 - Resize images to 320x240
 - ROIs in the gray scale images
 - Switching among calibration, initialization and detection
 - Adaptive Canny edges
 - Representative line per ROI
 - Clothoidal ROI model
 - Lane markings detection
- 3D road and ego-vehicle model** from a modification of the Dickmans road model-base method



$$\theta = a \tan\left(\frac{h}{L}\right)$$

$$v = f_v \cdot \tan(\theta - \alpha)$$

$$L_{cv} = L + d$$

$$u = \frac{f_u}{L} \cdot \left(C_0 \cdot \frac{L_{cv}^2}{2} + C_1 \cdot \frac{L_{cv}^3}{6} - x_0 \pm 0.5 \cdot w - L_{cv} \cdot \psi \right)$$
- Lane tracking** using Kalman filtering

$$x = [C_0 \quad C_1 \quad x_0 \quad \psi \quad w]^T$$
- Event detector**
 - Lane drifting:** evaluate the driver's tendency to exit the lane. Fraction of a given time interval (60 s) spent outside a virtual driving lane of around the center 1.2 m width
 - Lane weaving:** evaluates involuntary lane changes. Based on the presence or absence of the directional indicator, the event detector module concludes whether a lane change is intentional or not

Distraction Detection and Score Driving Behaviors

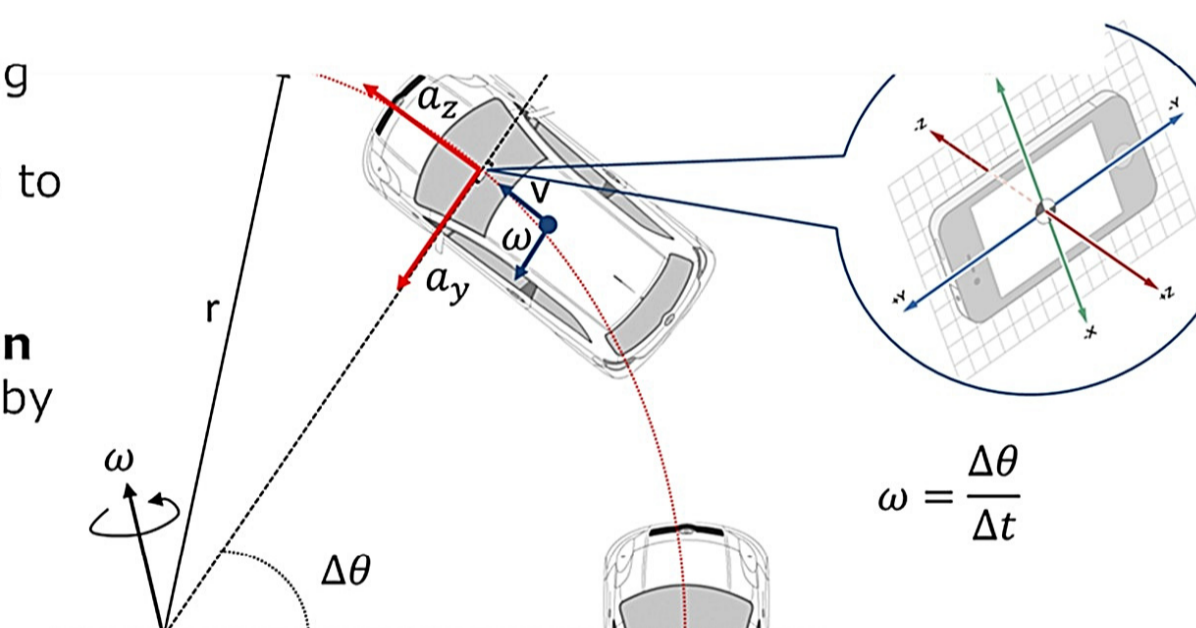
- Based on the frequency of critical driving events** considered as violations of certain thresholds imposed on vehicle acceleration measured by iPhone IMU
 - Acceleration:** Abrupt positive peaks along the z-axis indicate aggressive increases of velocity
 - Braking:** Abrupt negative peaks along the z-axis is indicative of harsh braking, and therefore indirectly of not retaining a minimum distance to the vehicle ahead
 - Turning:** High lateral acceleration along the y-axis, points towards excessive velocity in left or right turns, and may result in the vehicle losing traction

Event Type	Threshold sensitivity		
	Low	Medium	High
Acceleration	$0.1g < a_z < 0.2g$	$0.2g < a_z < 0.4g$	$a_z > 0.4g$
Braking	$-0.1g > a_z > -0.2g$	$-0.2g > a_z > -0.4g$	$a_z < -0.4g$
Turning	$0.1g < a_y < 0.2g$	$0.2g < a_y < 0.4g$	$ a_y > 0.4g$

Score driving behaviors

- Raw measurements filtering** by Kalman filtering
- Hysteresis control** to avoid near activations due to the same event
- Proposal for **decoupling the lateral acceleration due to the road curvature** from the one caused by wrong driver movements

$$a_y^c = \frac{v^2}{r} = r \cdot \omega^2 = \omega r v$$



Event detector

- Drowsiness takes into account the mean and the standard deviation of Lane Drifting and Lane Weaving

$$Score_{drows} = 1 - \left(\frac{m_{LD} + \sigma_{LD}}{2} + \left(\frac{m_{LW} + \sigma_{LW}}{2} \right) \right)$$

- We score the indicators taking into account the number and intensity of the events detected per Km

$$Event_km_i = [(k_1 \cdot Low_i + k_2 \cdot Medium_i + k_3 \cdot High_i) / Km]$$

$$Score_{drow} = 1 - CDF_e(Event_km_i)$$

Experimental results

Test-bed

- 12 participants** (9 males and 3 females) of our Lab
- The **test vehicle** was a **Renault Laguna** with manual shift
- Each participant** carried out the **two tests (aggressive and normal)**, **20 min long** each one, in **different days** and varied daytime (4 at morning, 4 at afternoon and 4 at night) and with **different weather conditions**: 20 mainly bright and sunny, 2 raining and 2 foggy
- Aggressive driving** under controlled vehicle maneuvers: 2 x lane drifting, 6 x lane weaving, 4 x sudden acceleration, 6 x sudden brake and 2 x sudden turn
- Normal driving**, which contains data collected during everyday driving routines

TABLE II. EVENTS DETECTION PERFORMANCE. TRUE POSITIVES (TP), FALSE POSITIVES (FP), GROUND-TRUTH (GT), PRECISION (PR), RECALL (RC)

Event	TP	FP	GT	PR	RC
Lane Drifting (LD)	25	3	25	0.89	1.00
Lane Weaving (LW)	75	6	78	0.93	0.96
Acceleration (AC)	51	19	58	0.74	0.88
Braking (BR)	80	20	91	0.80	0.88
Turning (TN)	28	10	30	0.74	0.93
Overall	260	57	282	0.82	0.92

TABLE III. COMPARATIVE DRIVING BEHAVIOR EVALUATION.

Event Type	Normal driving				Whole subjective score (%)	
	Detections	Score (Mean)	Detections	Score (Mean)	DS	AXA
Acceleration	12	15	8.85	6.37	62.50	37.50
Braking	13	11	9.92	7.27	87.50	12.50
Turning	8	28	8.10	6.18	80.00	20.00
Event Type	Aggressive driving					
	Detections	Score (Mean)	Detections	Score (Mean)	DS	AXA
Acceleration	58	56	4.54	5.00		
Braking	87	80	4.73	3.70		
Turning	30	58	3.60	2.20		

Conclusions and Future Works

- DriveSafe**, a new driver safety app for iPhones has been presented
- First app** that **detects inattentive driving behaviors** and **scores driving style** at the same time
- Quantitative evaluation** based on a **controlled test-bed** in **real scenarios** has been carried out from 12 drivers
- Detection performance** of some **inattentive driving events** reached an **overall precision of 82% at 92% of recall**
- Driving scores comparison **DriveSafe vs AXA Drive app** obtained a **better valuation for DriveSafe**
- Future guidelines:** upload Drivesafe version 2.0 to the Apple store after deeper tests with more vehicles, roads and users

Acknowledgments

This work has been funded by the Spanish Ministry of Economy and Competitiveness through the project Smart Driving Applications (TEC2012-37104) and by Comunidad de Madrid through the project RoboCity2030-II under Grant CAM-S2009/DPI-1559

Thanks also to Sergio Gámez for his support in the experimental tests



IEEE

INTELLIGENT VEHICLES SYMPOSIUM
 8 - 11 June 2014, Dearborn, Michigan, USA