

Embedded Perception & Decision-making for Safe Navigation in Uncertain, Dynamic and Human-populated Environments

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Embedded Perception & Decision-making for Safe Navigation in Uncertain, Dynamic and Human-populated Environments

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Invited Talk – Coboteam Workshop on "Robots Navigation" France, On-line event, May 28th 2020







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Technology status & Ongoing challenges for AVs



- Strong involvement of Car Industry & GAFA + Large media coverage + Increasing Governments supports
- An expected market of 515 B€ at horizon 2035 (~17% world automobile market, Consulting agency AT Kearney, Dec 2017)
- But Validation & Certification issues are still unclear ... idem for Legal & Regulation issues

Informatics mathematics

=> Numerous experiments in real traffic conditions since 2010 (Disengagement reports $\{Masghts on Stem maturity\}$ => Rut still insufficient Realistic Simulation & Formal methods are also under the powert (No FU Enable-S3)



Fatal accidents involving AVs – Perception failure



- □ Tesla driver killed in a crash with Autopilot "level 2" active (ADAS mode) May 2016
 - The Autopilot <u>failed to detect</u> a white moving truck, with a
 brightly lit sky (Camera Mobileye + Radar)
 - The human driver was not vigilant & didn't took over

□ Self-driving Uber L3 vehicle killed a woman

=> First fatal crash involving a pedestrian Temple, Arizona, March 2018

✓ Despite the presence of multiple sensors (lidars, cameras ...), the perception system failed to detect the pedestrian & didn't disengaged
 ✓ The Section Detect of the left of the lef

✓ The Safety Driver reacted too lately (1s before the crash)









AVs have to face two main challenges

Challenge 1: The need for Robust, Self-diagnosing & Explainable Embedded Perception

2019.5.15 - TRY 3



Video Scenario:

- The Tesla perception system failed to detect the barriers blocking the left side route
- *The driver has to take over and steer the vehicle away from the blocked route (for avoiding the collision)*

Video source: AutoPilot Review @ youtube.com

Challenge 2: The need for Understandable Driving Decisions (share the road with human drivers)

Human drivers actions are determined by a complex set of interdependent factors difficult to model (e.g. intentions, perception, emotions ...)
⇒Predicting human driver behaviors is inherently <u>uncertain</u>
⇒AV have to reason about <u>uncertain intentions of</u> the surrounding vehicles

DELEC.



Video scenario:

Scene observed by the dash cam of a **bus** moving behind the Waymo AV

- Waymo AV is blocked by an obstacle and it decides to execute a left lane change
- The bus driver misunderstood the Tesla's intention and didn't yield
- The two vehicles collided

Video source: The Telegraph

Embedded Bayesian Perception & Decision-making

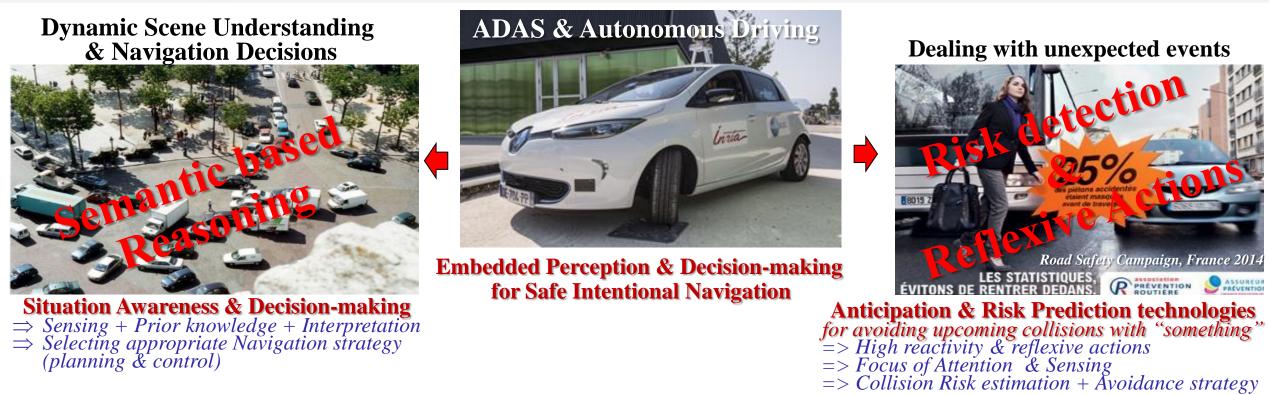


- Main Challenges & Required Technological Breakthrough
 - => Robustness, Efficiency, Dynamic Human Environments (Safety is still not guaranteed)
 - => Real-time integration of Perception & Motion planning & Control
 - => Integration into Embedded Hardware & Software (future products)
 - => Validation & Certification



ADAS & Autonomous Driving

Intra Perception & Decision-making requirements for AVs



Main features

- ✓ Dynamic & Open Environments => *Real-time processing & Reactivity (several reasoning levels are required)*
- Incompleteness & Uncertainty => Appropriate Model & Algorithms (probabilistic approaches)
- ✓ Sensors limitations (no sensor is perfect) => *Multi-Sensors Fusion*
- ✓ Hardware / Software integration => *Satisfying Embedded constraints*
- Human in the loop (mixed traffic) => Human Aware Decision-making process (AI based technologies)
 Taking into account Interactions + Behaviors + Social rules (including traffic rules)

1st Paradigm : Embedded Bayesian Perception

Sensors Fusion

=> Mapping & Detection

Dynamic scene interpretation

Using Context & Semantics



informatics mathematics

Embedded Multi-Sensors Perception ⇒ Continuous monitoring of the dynamic environment

□ Main challenges

- ✓ Noisy data, Incompleteness, Dynamicity, Discrete measurements
- ✓ Strong Embedded & Real time constraints

Our Approach: Embedded Bayesian Perception

- ✓ Reasoning about Uncertainty & Time window (Past & Future events)
- ✓ Improving robustness using Bayesian Sensors Fusion
- ✓ Interpreting the dynamic scene using Contextual & Semantic information
- ✓ Software & Hardware integration using GPU, Multicore, Microcontrollers...

NANOELEC.

Characterization of the local

Safe Navigable Space & Collision Risk

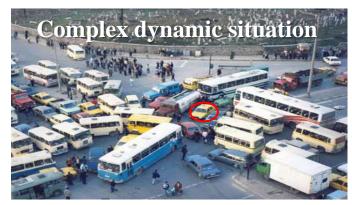
Dynamic Probabilistic Grid & Bayesian Filtering – Main Features Exploiting the <u>dynamic information</u> for a better understanding of the scene informatics mathematics NANOELEC. **Occupancy Grid** Motion fields (Static part (Dynamic part) Sensors data fusion **Bayesian Filtering Extracted Motion Fields** 3 pedestrians⁴ 1st Embedded & Optimized version **Free space** (HSBOF, patent 2014) Moving car Static obstacles 2 pedestrians Front camera view of the ego vehicle (urban scene) Patented Improvements & Implementations (2015, 2017) ery-high Obstacles Non-Ground Ground elevation o

Grid & Pseudo-objects

Classification (using Deep Learning)

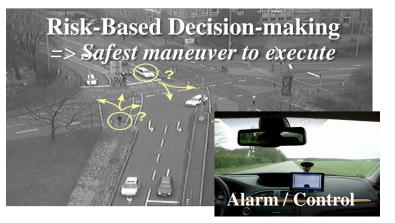
Detection & Tracking + Moving Objects Classification => CMCDOT 2015 (including a "Dense Occupancy Tracker") Ground Estimation & Point Cloud Classification (patent 2017)

2nd Paradigm: Collision Risk Assessment & Decision-making for avoiding Pending & Future Collisions



Main challenges

Human-Aware Situation Assessment



Uncertainty, Partial Knowledge, World changes, Real time Human in the loop + Unexpected events + Navigation Decision based on <u>Perception & Prior Knowledge</u>

□ Approach: Prediction + Risk Assessment + Bayesian Decision-making

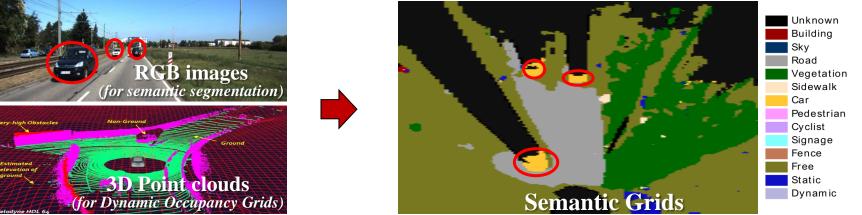
- ✓ Reason about Uncertainty & Contextual Knowledge (using History & Prediction)
- ✓ Estimate <u>Probabilistic Collision Risk</u> at a given time horizon $t+\delta$ ($\delta = a$ few seconds)
- ✓ Make <u>Driving Decisions</u> by taking into account the Predicted behavior of <u>all the observed surrounding traffic</u> participants (cars, cycles, pedestrians ...) & Social / Traffic rules

Decision-making: *Two types of "collision risk" have to be considered*

- ✓ *Short-term collision risk* => *Imminent collisions with "something" (unclassified), time horizon <3s, conservative hypotheses*
- ✓ Long-term collision risk => Future potential collisions, horizon >3s, Context + Semantics, Behavior models

3rd Paradigm: Models improvements using Machine Learning

□ **Perception level:** Construct "Semantic Grids" using Bayesian Perception & DL



Decision-making level: *Learn driving skills for Autonomous Driving*

1st Step: Modeling Driver Behavior using Inverse Reinforcement Learning (IRL)
 2nd Step: Predict motions of surrounding vehicles & Make Driving Decisions for Ego Vehicle





4th Paradigm: Combining Motion Planning & Safe navigation

Informatics mathematics

Global planner – Hybrid-State E*

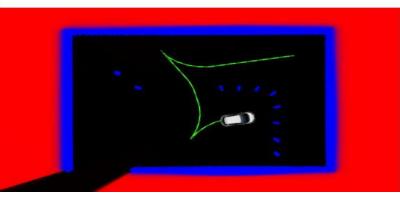
Pros & cons

- + Considers occupancy uncertainty
- + Continuous path even at low resolution
- + Allows complex maneuvers
- + Favors simple maneuvers
- Discrete space representation
- No obstacle motion

Future work

- · Use exact vehicle shape
- Online replanning

Example of a computed path



Local planner – DWA

Accuracy

- Accurate trajectory prediction
- Accurate ego vehicle shape

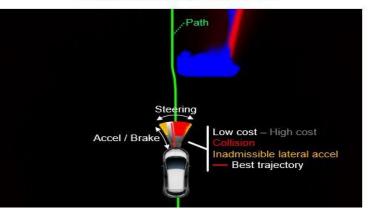
Computing efficiency

 Massively parallel computations over ego vehicle positions and trajectories

Simplicity

- Only simple trajectories
- Short term prediction (5-10s)

Command sampling illustration



NANOELEC.



Illustration: Video demos IROS 2018 (Madrid)

HADRID2018



Autonomous Driving Events

uc3m Carlos III & CSIC & Universidad Universidad

