

Introduction to Autonomous Systems

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Version 2.5

Autonomous Systems

- **Definitions**
- Applications
- Technologies
 - Mission Planning
 - Mission Control Perception
 - Intelligence
 - Embedded computing
 - Autonomous systems swarms
 - Communications

Autonomous System definitions

Autonomous system etymology:

- Greek words '**αυτο**' (auto, own) and '**νόμος**' (law, rule)
- A system that operates using own rules/laws
- A system that operates independently:
 - without reference to other entities
 - without reporting to any authority.

Autonomous System definitions

A fully autonomous system can:

- Gain information about the environment.
- Work for an ***extended period without human intervention.***
- Move either all or part of itself throughout its operating environment without human assistance.
- Avoid situations that are harmful to people, property, or itself unless those are part of its design specifications.

Autonomous System definitions

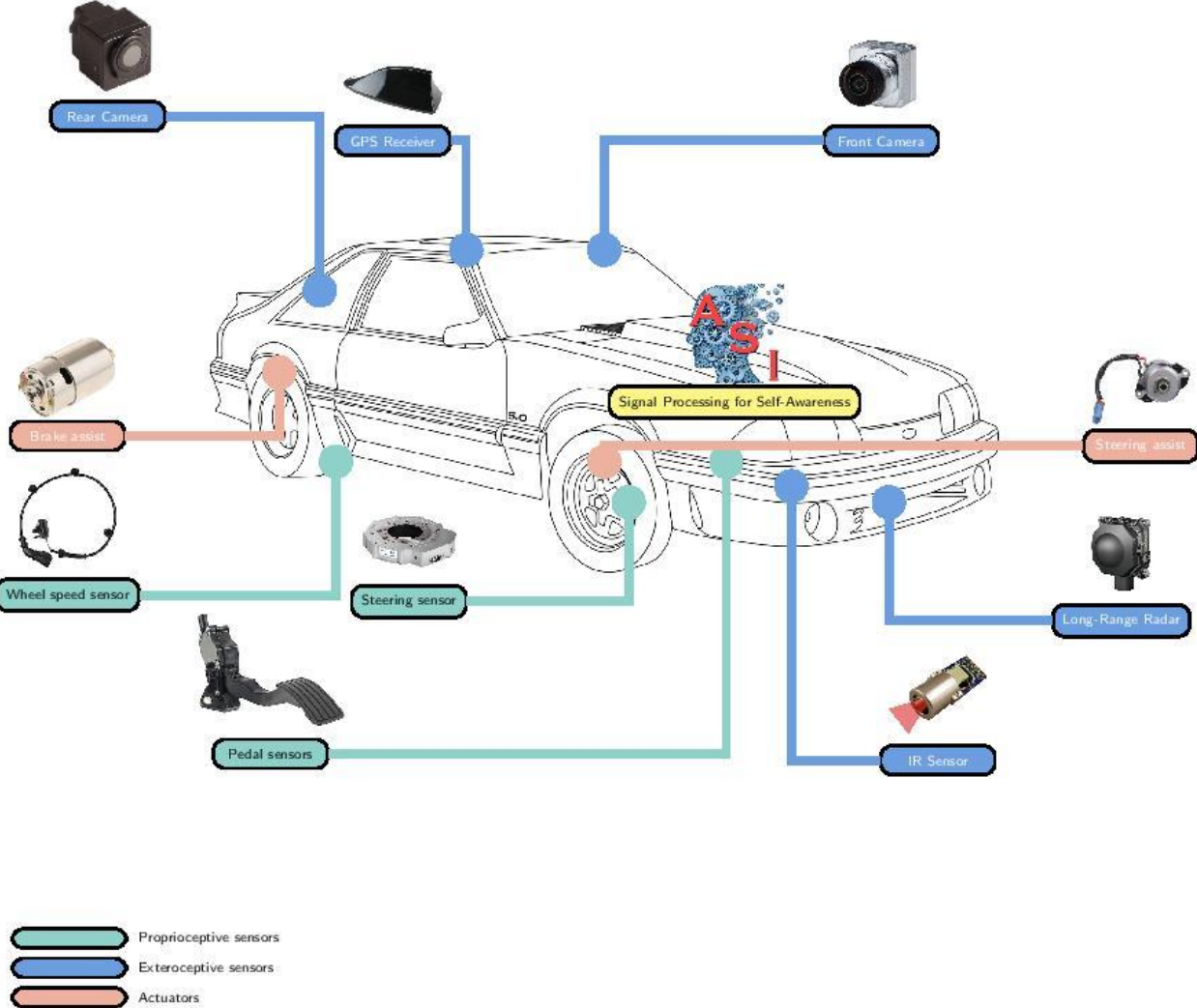
Sensorial signals (video, acoustic, tactile, radio signals) should be processed by an AS in real time to:

- ***interpret the external situation*** in which it operates;
- relate such a situation to its ***internal state***, by observing it with other proprioceptive sensors, so that it becomes self-aware;
- to use representations to help its own control blocks to ***drive its actuators***;
- ***to be able to explain*** at sub-symbolic and symbolic level the reasons of its own choices.

Autonomous System definitions



Autonomous car structure



Courtesy L. Marcenaro, C. Regazzoni

Autonomous Systems

- Definitions
- **Applications**
 - **Cars**
 - **Drones**
 - **Marine systems**
 - **Robots**
- Technologies

Autonomous system applications

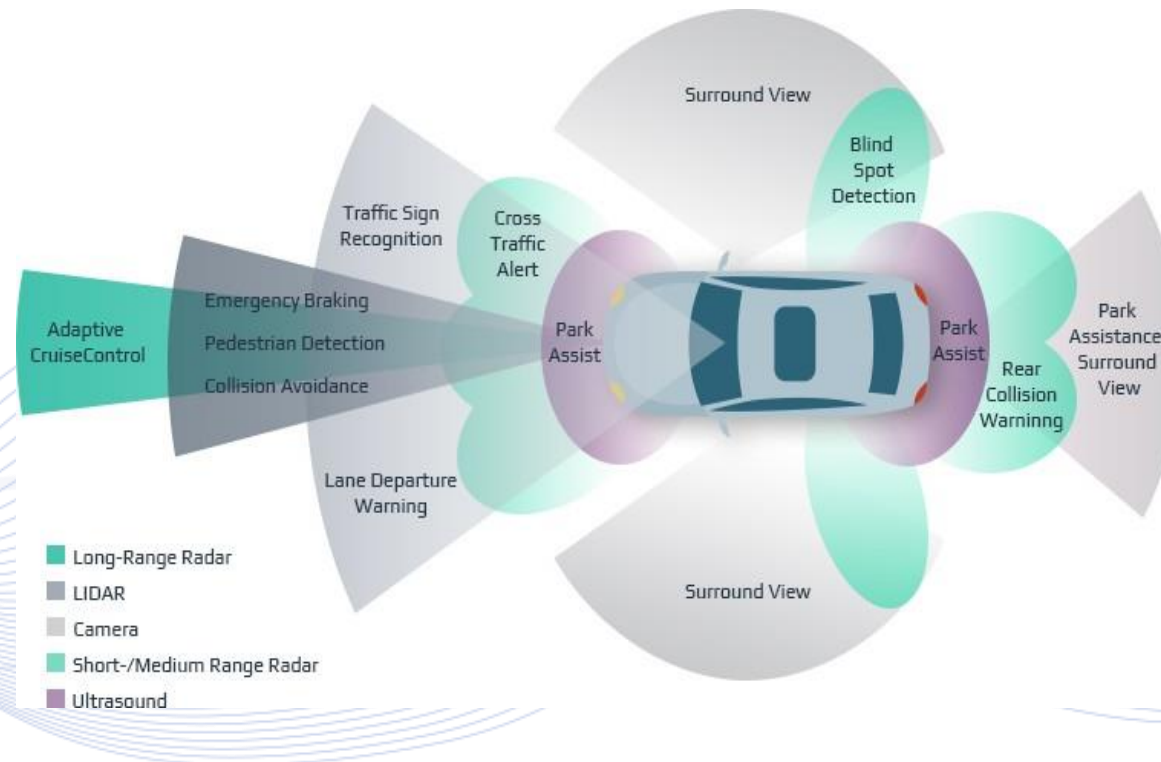
- **Autonomous cars**



Autonomous system applications

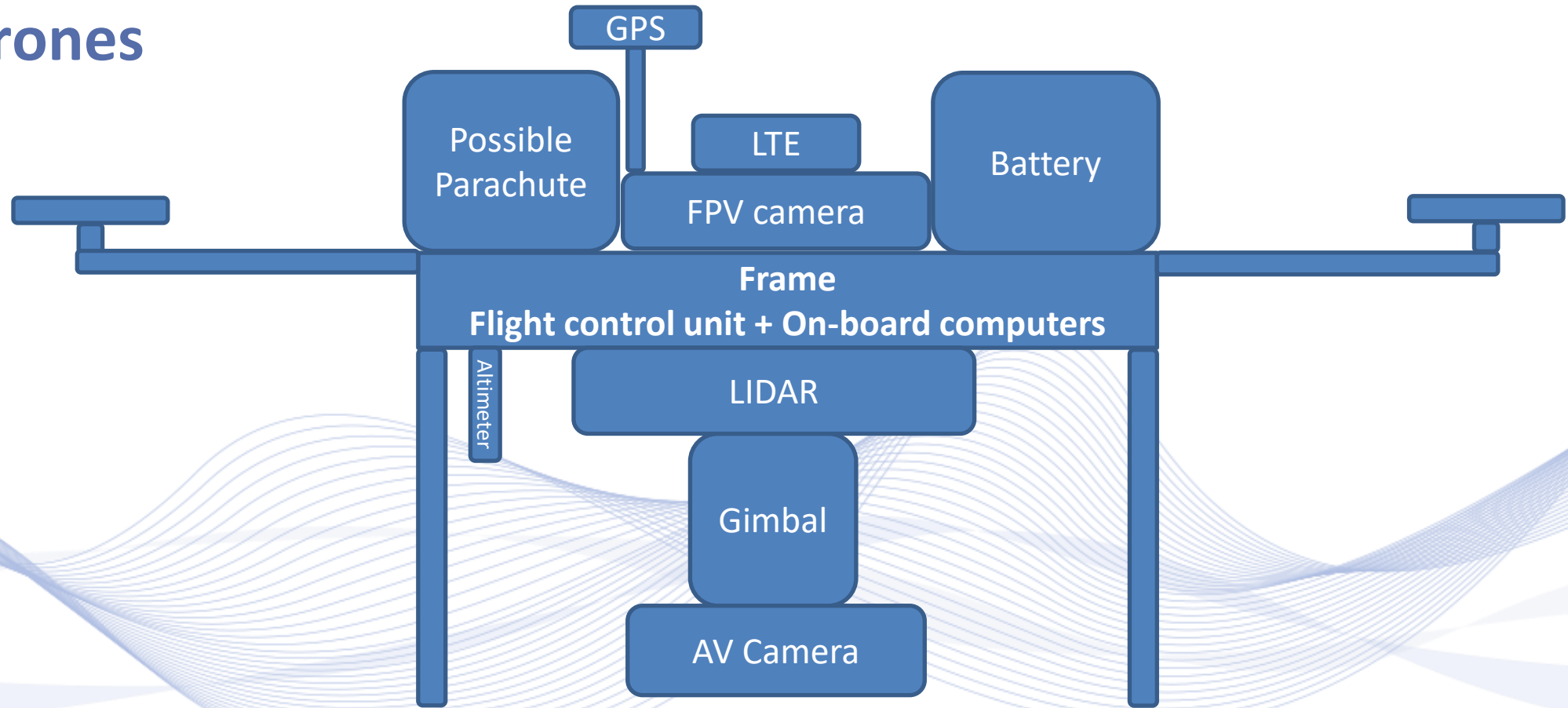


- **Autonomous car sensors and perception**



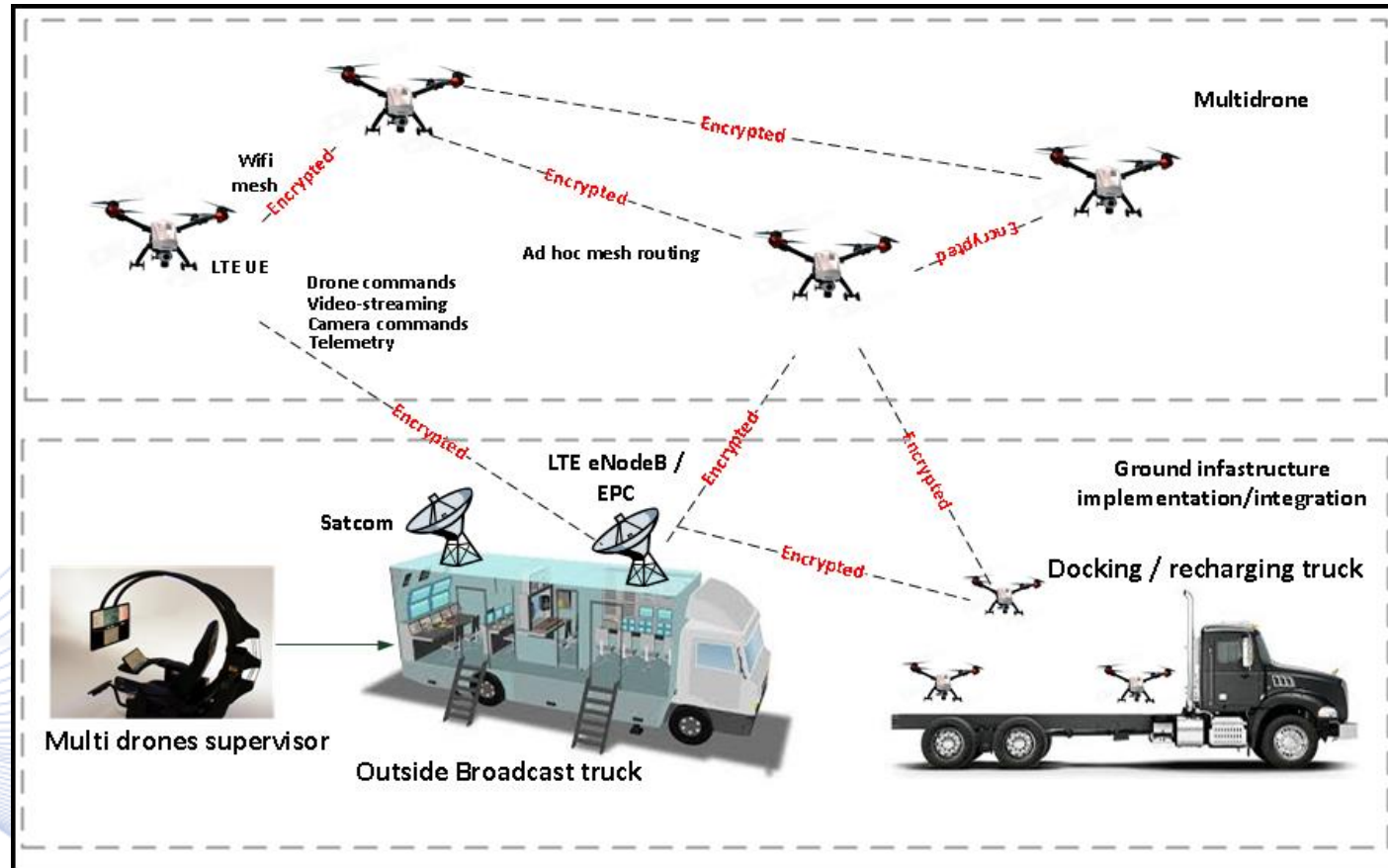
Autonomous system applications

Drones



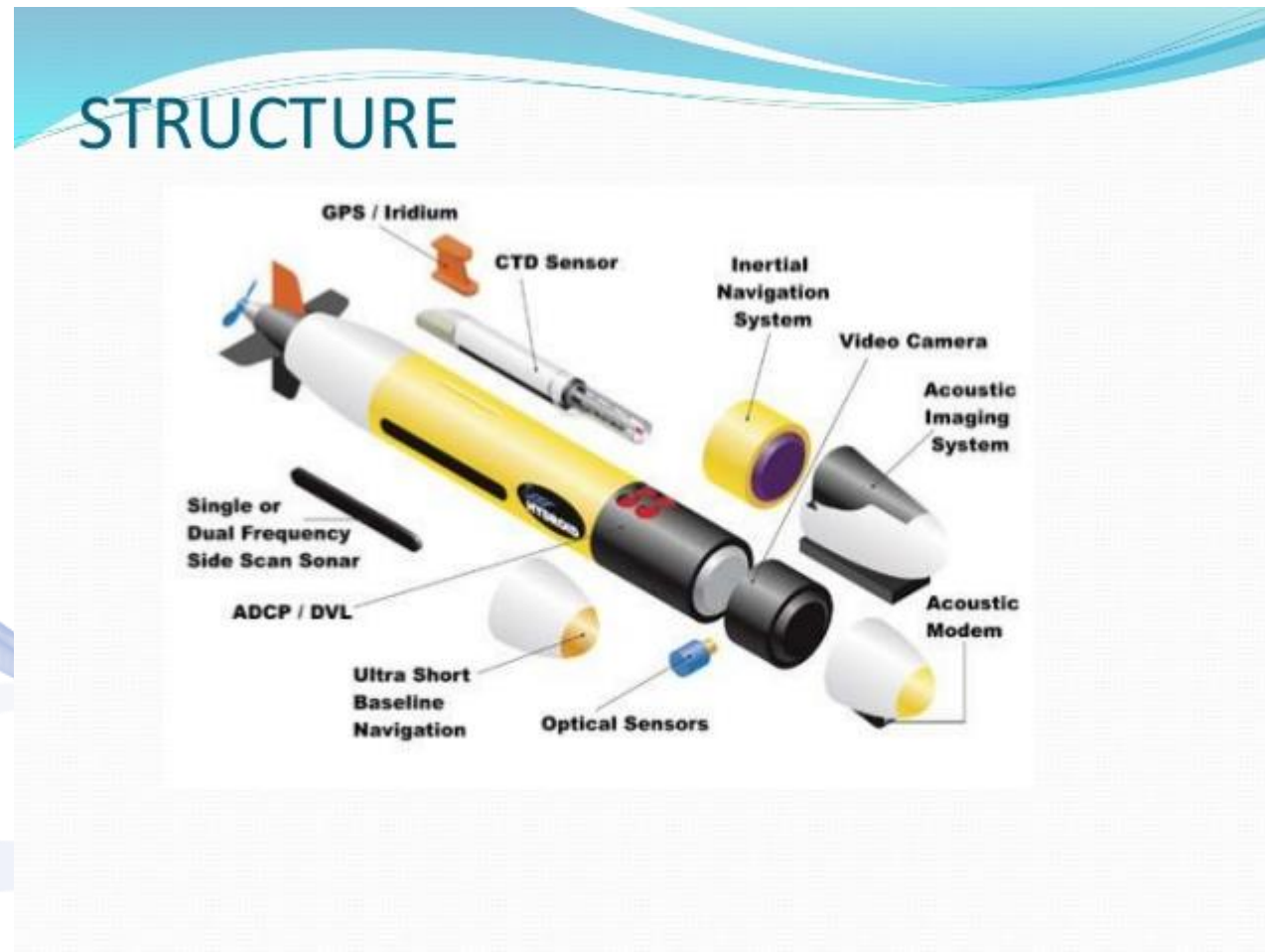
Autonomous system applications

Drone swarms



Autonomous system applications

Underwater vehicles



Autonomous system applications

Merchant ships

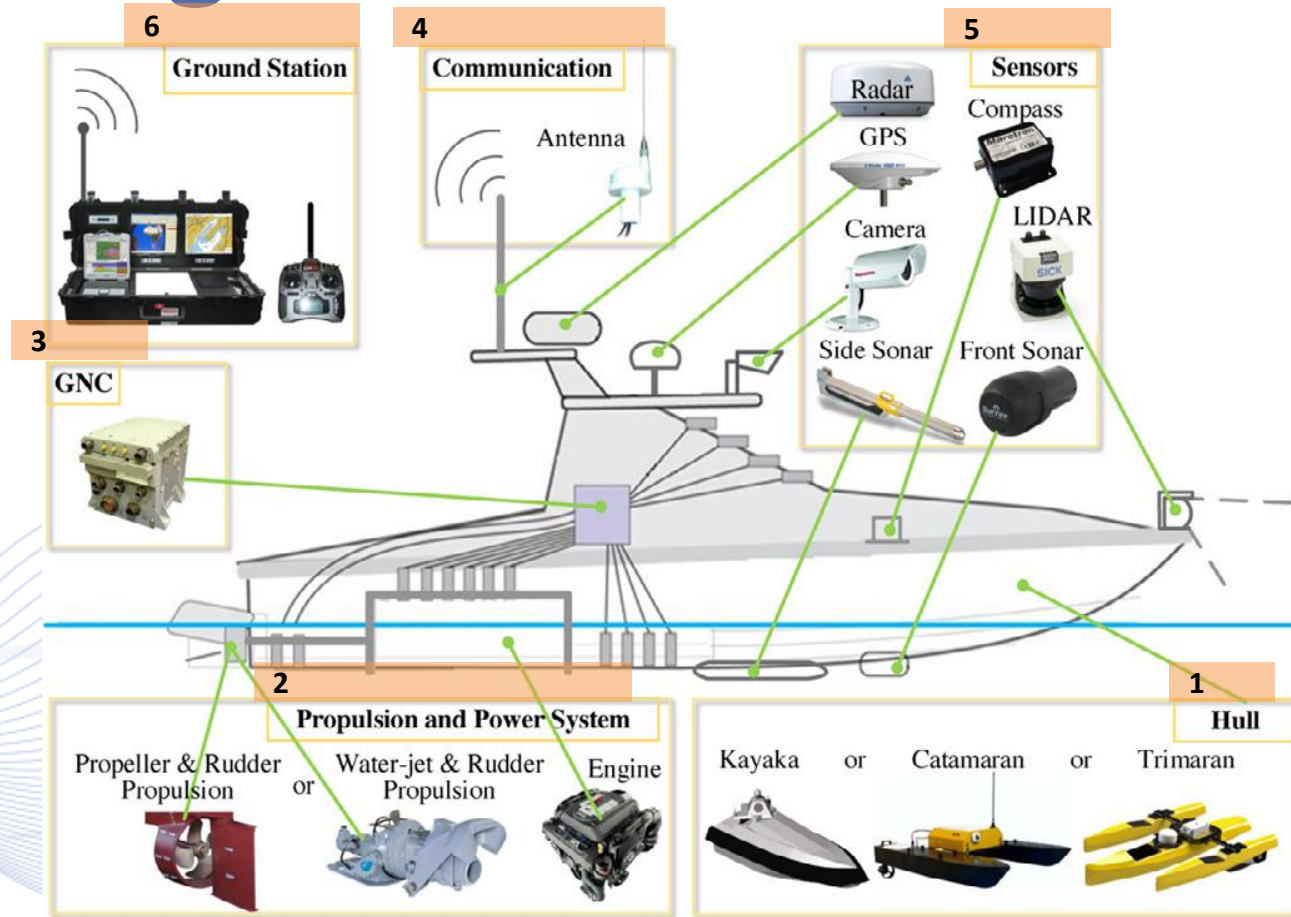


Autonomous system applications

Robots

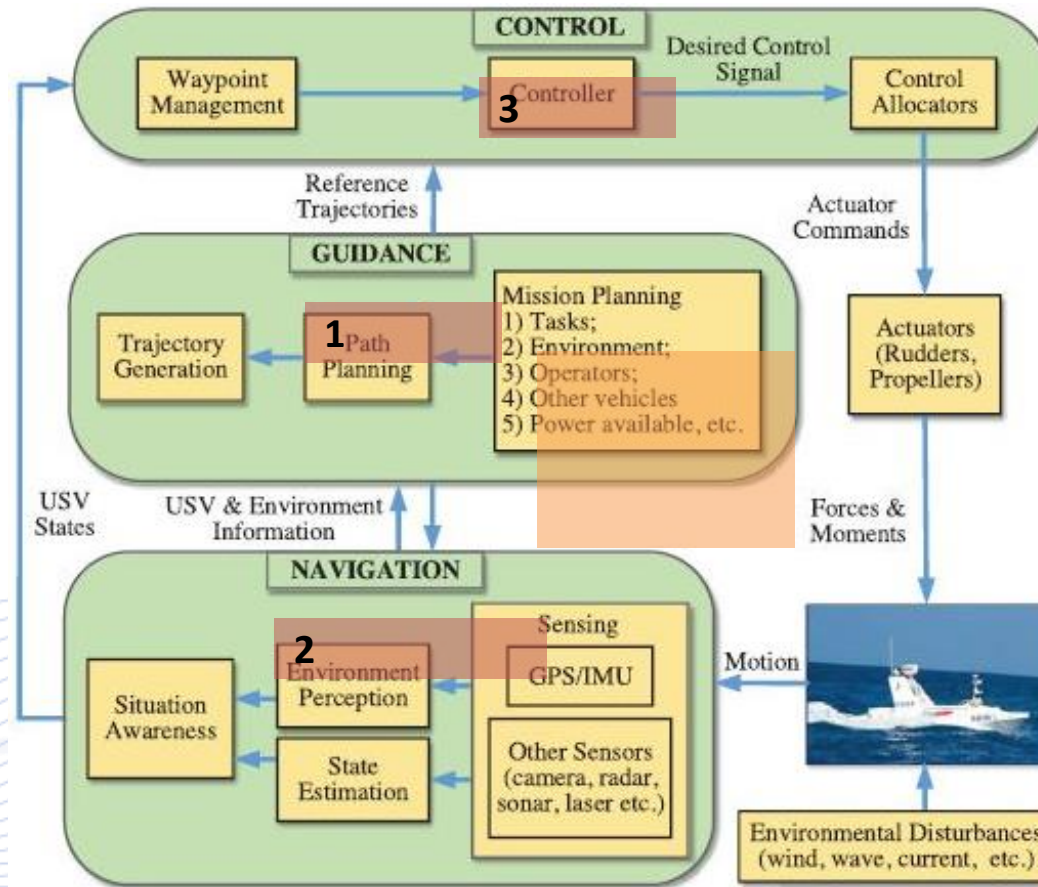


Autonomous System technologies



Basic elements of an Autonomous Surface Vessel [LIU2016].

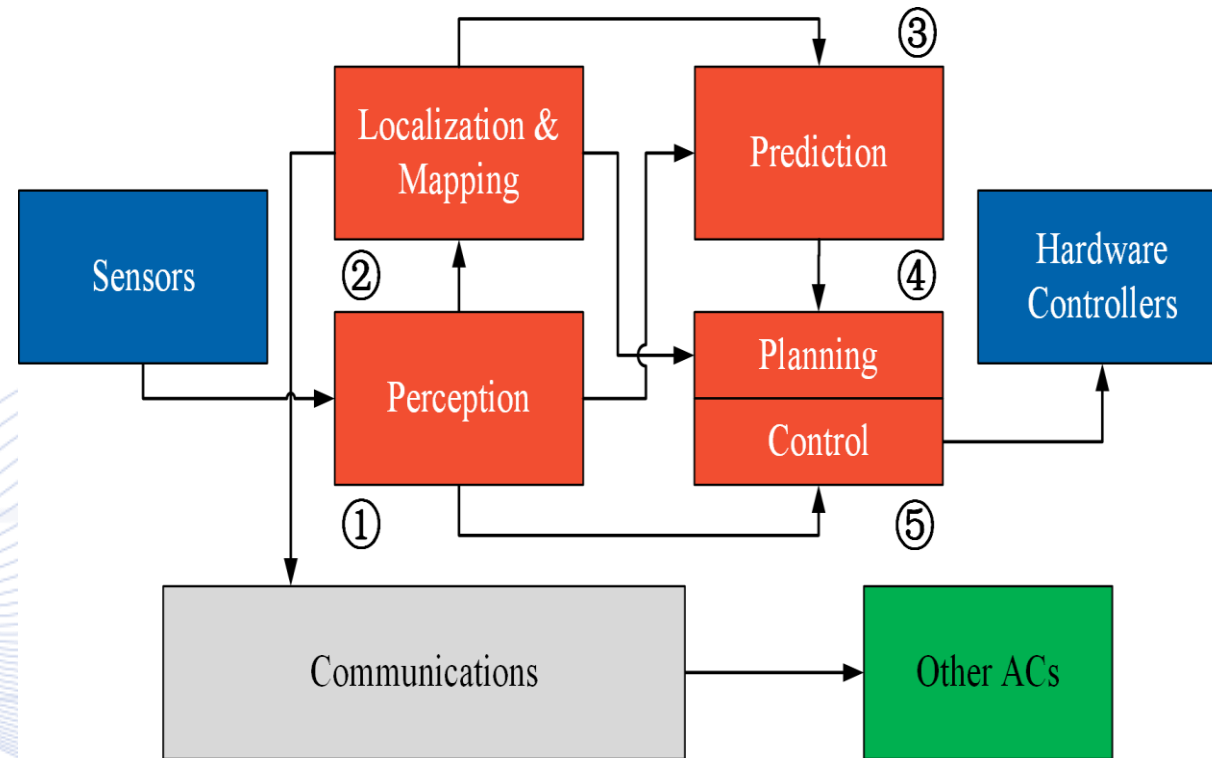
Autonomous System technologies



Autonomous Surface Vessel structure systems [LIU2016].

Autonomous System technologies

- Autonomous car structure



Autonomous System technologies

- **Mission Planning and Control**
- Perception and Intelligence
- Embedded computing
- Swarm systems
- Communications
- Societal technologies

Autonomous system mission

- **Autonomous car mission**
 - **List of navigation actions**
 - **Motion along a 2D trajectory (path)**
- **Autonomous drone AV Shooting Mission: list of actions**
 - **Shooting Actions:** drone + camera
e.g., Lateral Tracking, Fly-Over, Orbit, ...
 - **Navigation Actions:** drone action only, does not involve shooting
e.g., Take-off, Land, Go-to-waypoint, ...

Autonomous system mission planning

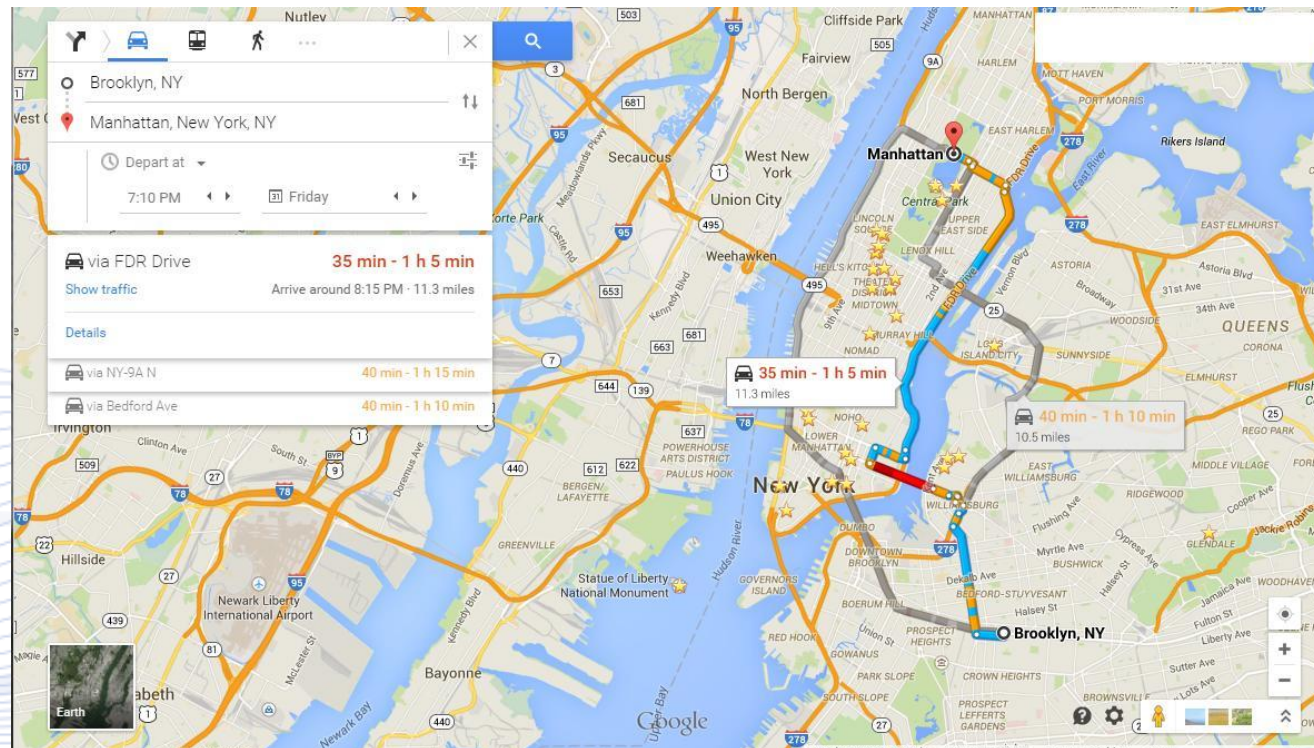


- **Autonomous car mission planning**
 - Find the best (2D) trajectory from start to destination
 - Planning constraints:
 - Road map (e.g., Google maps)
 - Regulatory restrictions (one way streets)
 - Traffic load
 - Use of semantic (2D) maps

Autonomous system mission planning

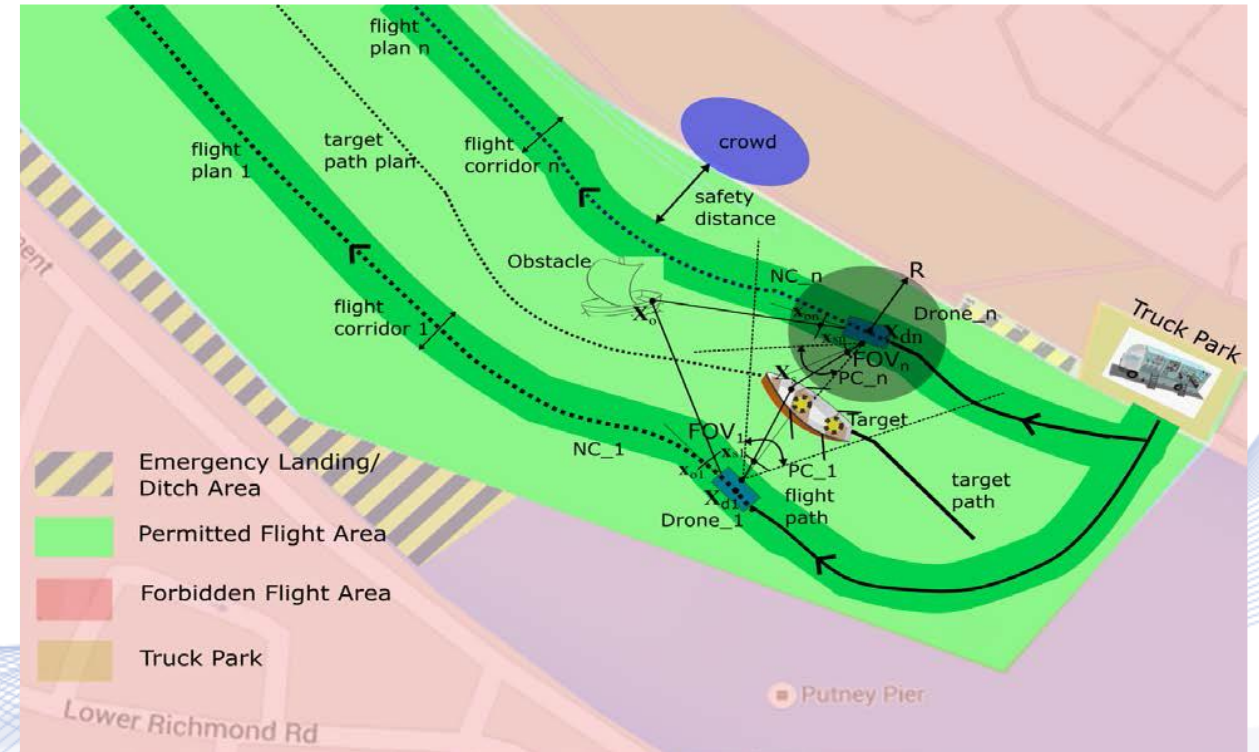


- Google maps path planning.

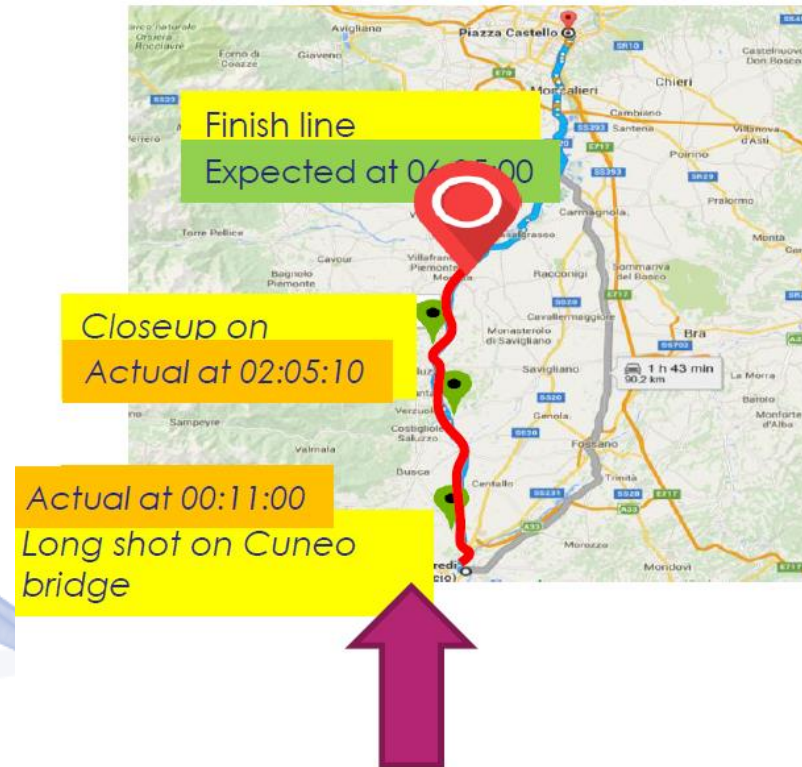


Autonomous system mission planning

- Drone mission planning.
- Planning of:
 - Drone flight
 - Payload (e.g., camera) actions
- Use of semantic 3D maps



Mission example: Giro d'Italia



<<Accident Detected>>



Other view 1

Other view 2

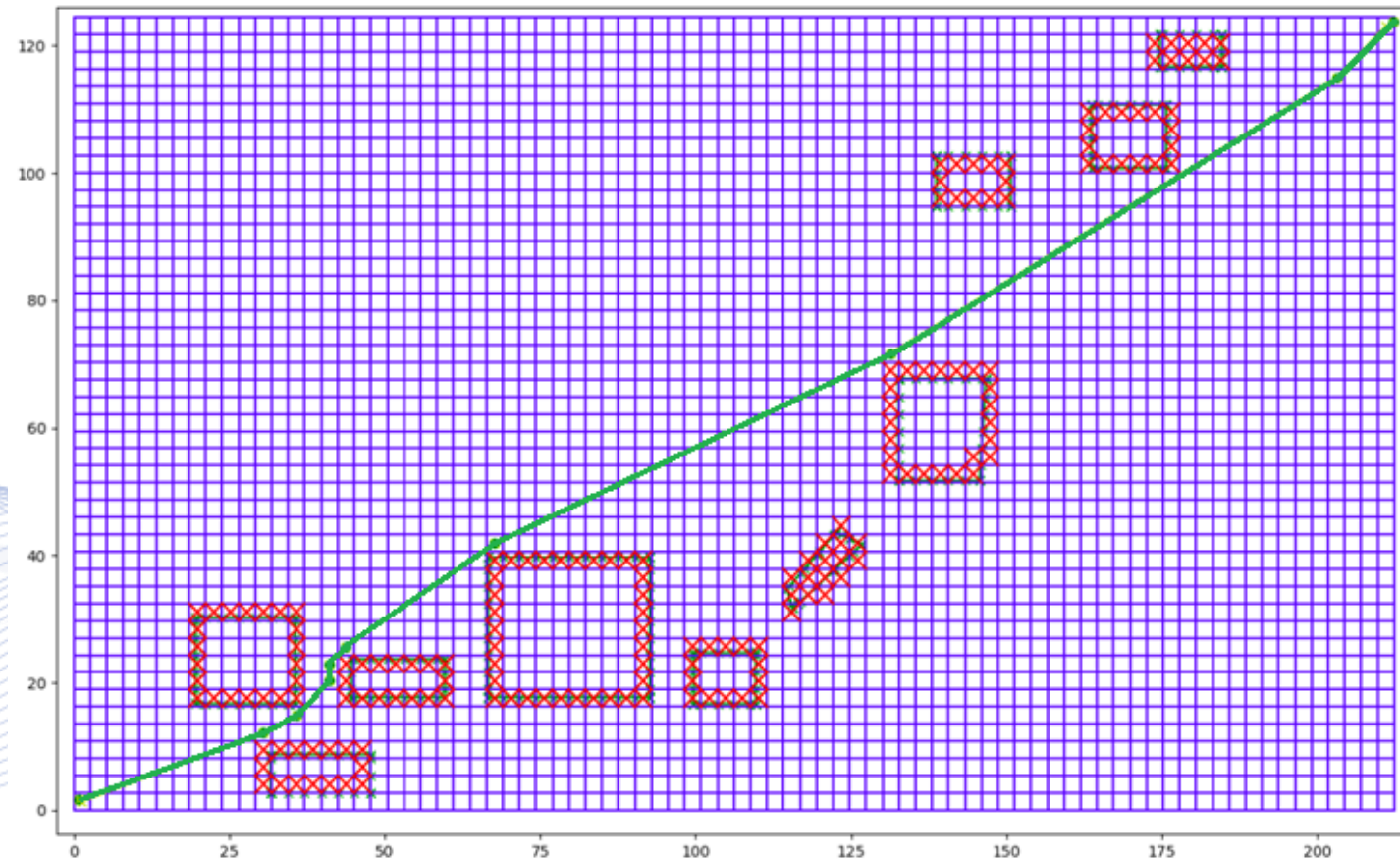
Panoramic (stitched)

Path Planner

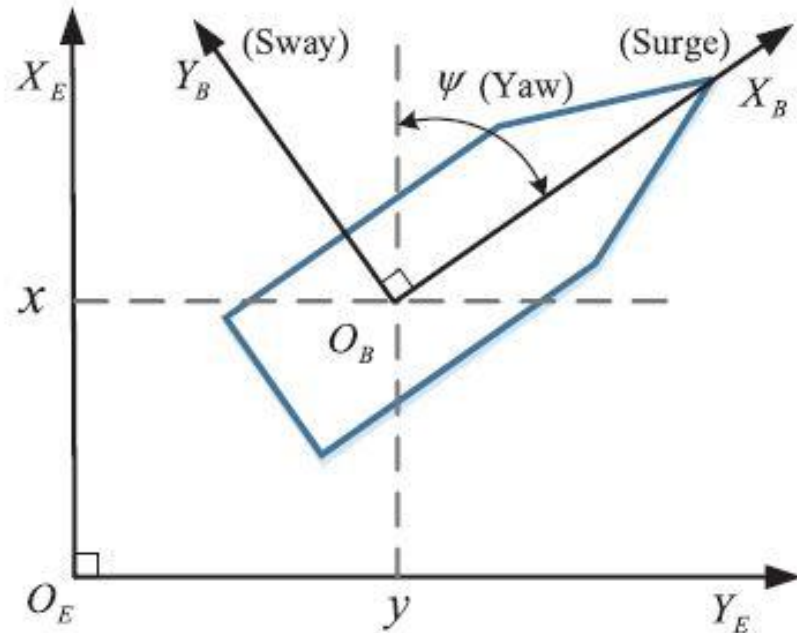
- This submodule is used by:
 - High-level Planner to estimate drone paths and flying times.
 - Onboard Scheduler to compute a path to a landing position in case of emergency.
- Navigation map implemented as a grid. Obtained from Semantic Map.
 - Semantic annotations are indicated as KML features.
 - Geodesic coordinates translated into Cartesian.
 - No-fly polygons become occupied cells in grid.
- Safe path computed using A* search algorithm. Fast for simple solution spaces.

Path Planner Example

- Path from one corner to the other. Buildings labeled as no-fly zones (obstacles represented as red crosses in the grid).
- Solved in 66 ms.



Autonomous vehicle control



ASV planar motion [LIU2016].

Autonomous vehicle control is based on:

- Kinematic vehicle model
- dynamic vehicle model.

They are drastically vehicle-dependent.

Autonomous Surface Vessel (ASV):

- kinematic (motion) model:

$$\dot{\eta} = \mathbf{R}(\psi)\mathbf{v}.$$

- dynamic model [FOS1994]:

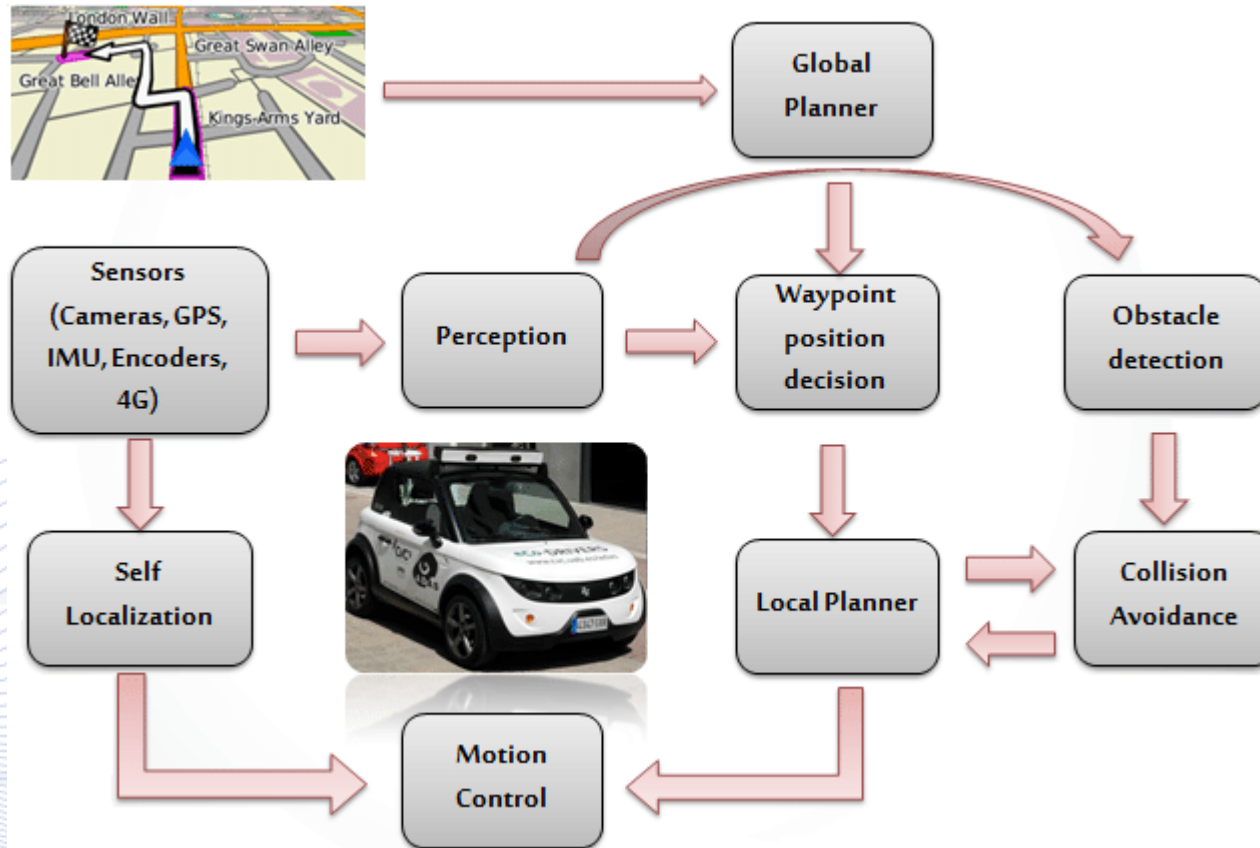
$$\mathbf{M}\dot{\mathbf{v}} + \mathbf{C}(\mathbf{v})\mathbf{v} + \mathbf{D}(\mathbf{v})\mathbf{v} + \mathbf{g}(\eta) = \boldsymbol{\tau} + \boldsymbol{\tau}_E.$$

Autonomous car control

- Car dynamic modelling
- Interfacing car perception to car control
- Levels of car control automation

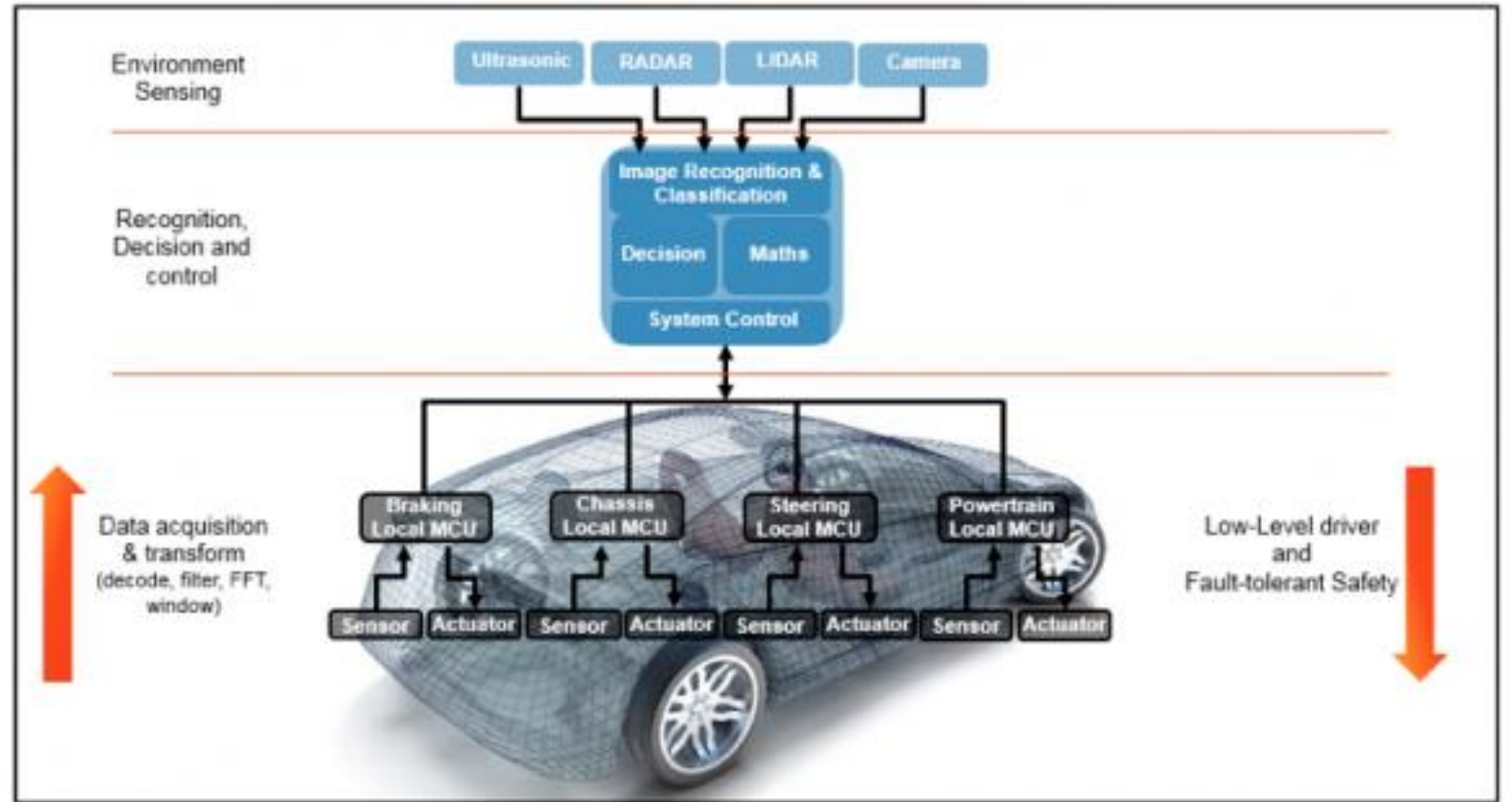
Level	Name	Driver	DEM2	DDTF3
0	No automation	HD4	HD	HD
1	Driver assistance	HD & system	HD	HD
2	Partial automation	System	HD	HD
3	Conditional automation	System	System	HD
4	High automation	System	System	System
5	Full automation	System	System	System

Autonomous car control



Autonomous car control

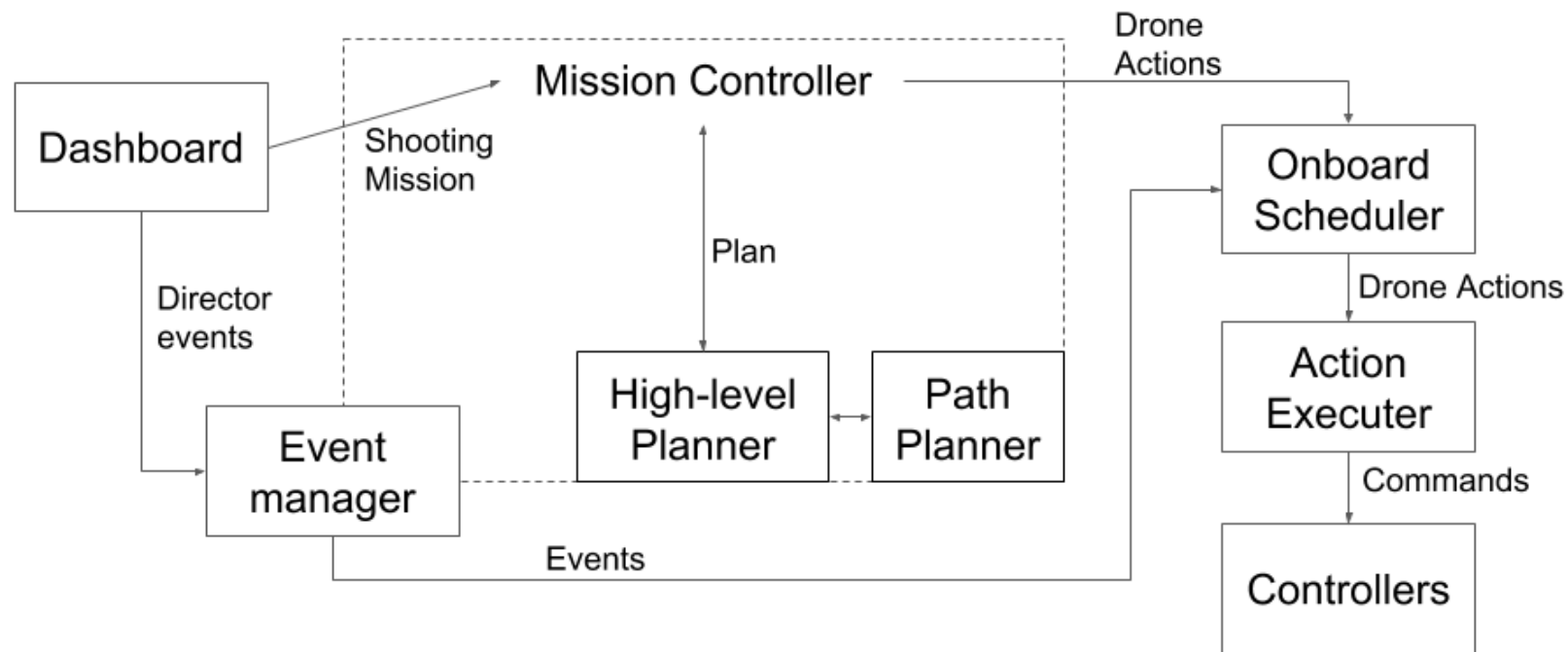
- Steering control
- Braking control
- Power control



Drone Mission Planning and Control Architecture

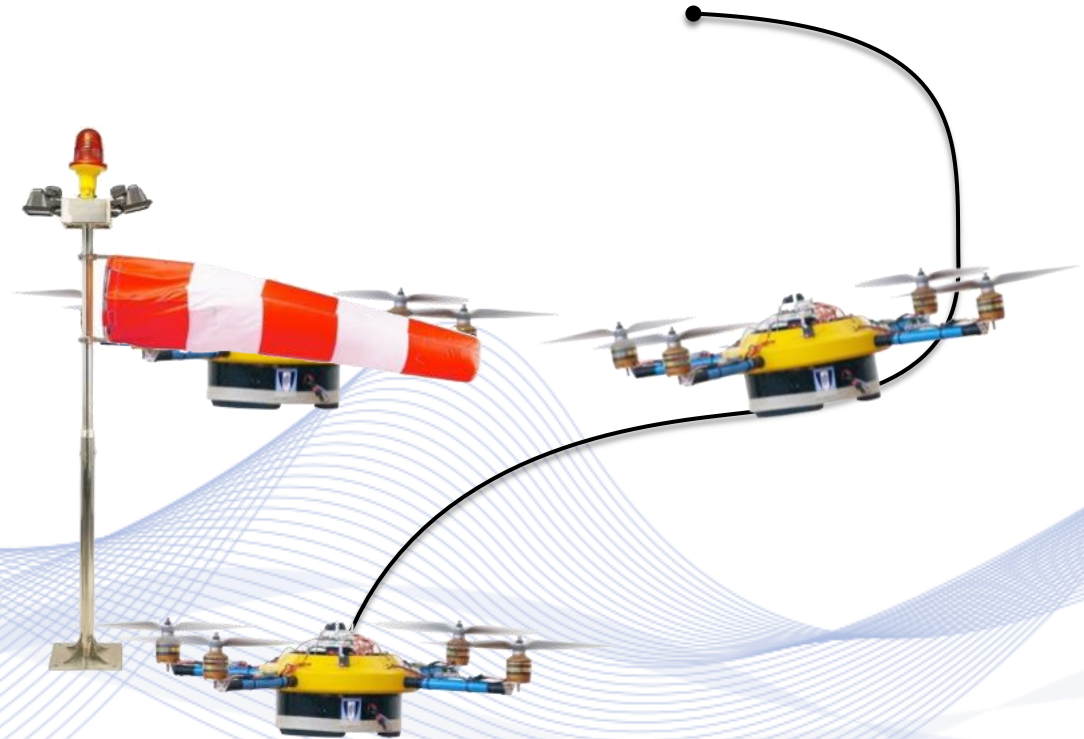


MULTIDRONE Planning

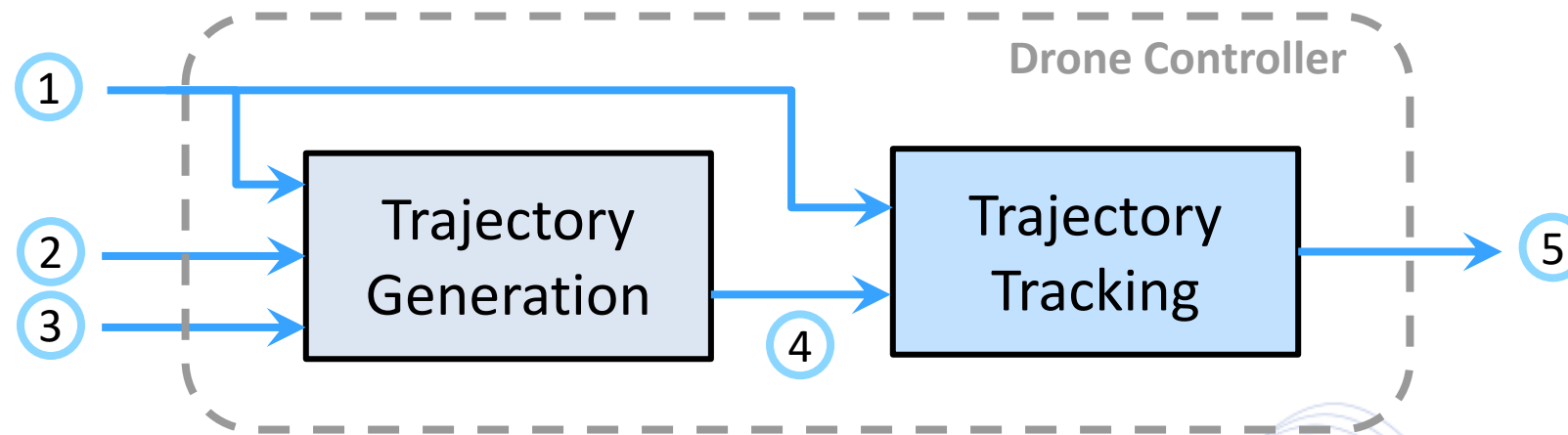


Drone Control Objectives – Trajectory Tracking

- Track a trajectory.
- Realistic model.
- Robustness to disturbances.
- Bounded actuation.
- Large basin of attraction.

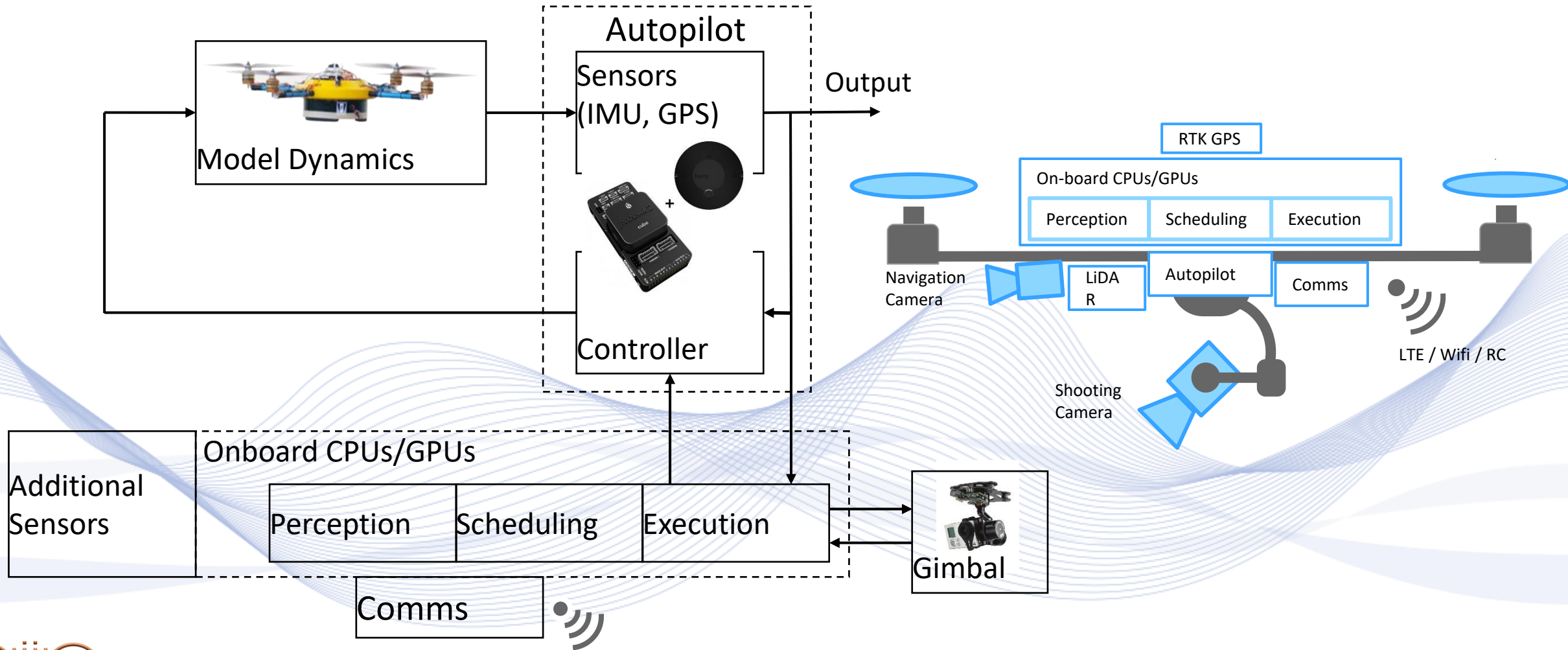


Drone Controller



- 1 Drone Status
- 2 Target Status
- 3 Shooting Action parameters
- 4 Reference
- 5 Drone Velocity Command

Onboard Drone control Architecture



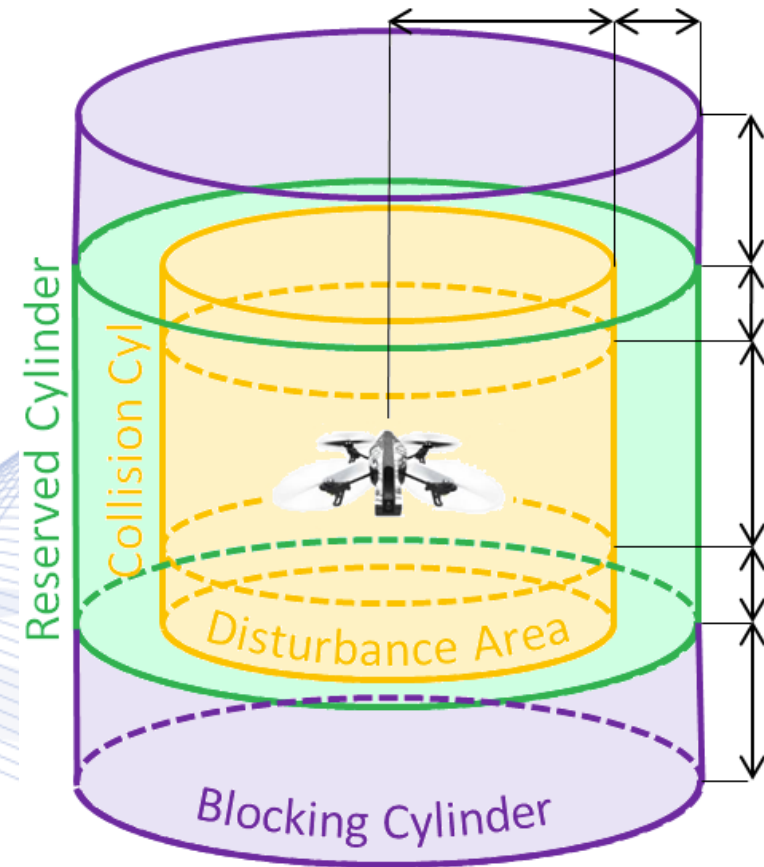
Car collision avoidance

- Sensors for:
- Vehicle detection/localization
- Pedestrian detection
- Real-time car trajectory replanning for collision avoidance.



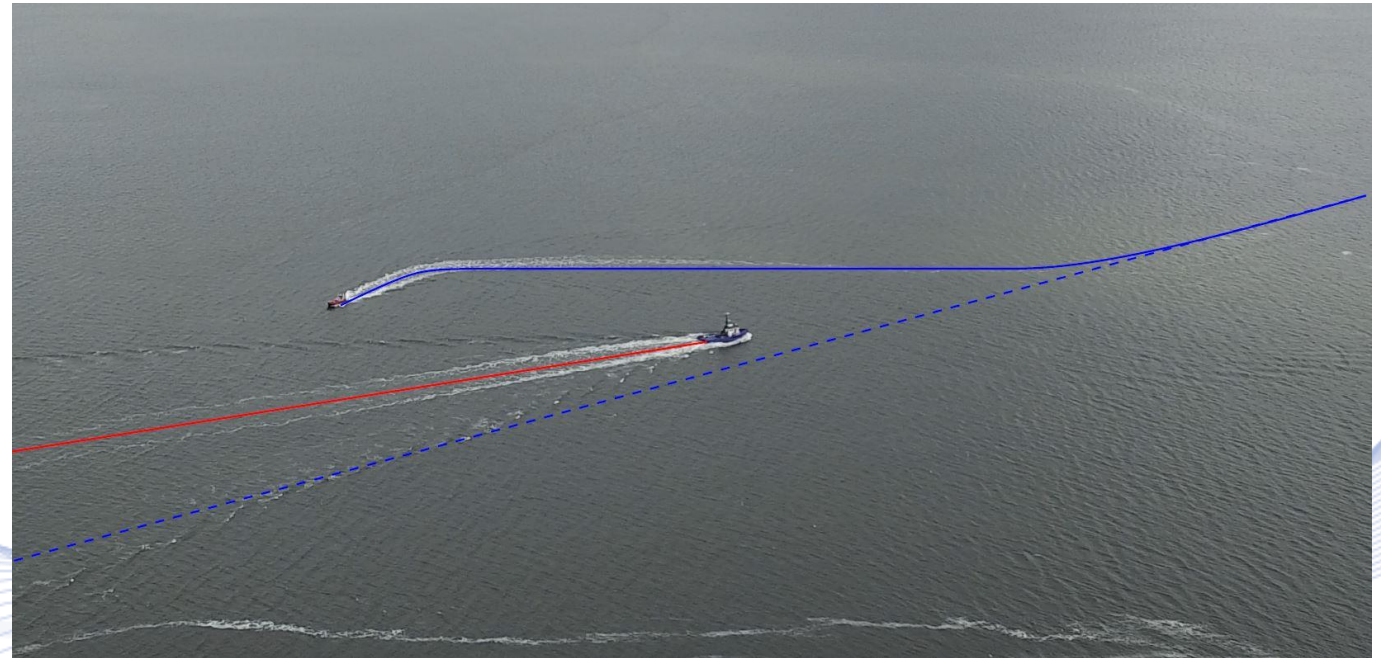
Drone collision avoidance

- Collision hull defined as a cylinder (yellow).
- Horizontal conflict when reserved cylinder (green) overlaps with others.
- Vertical conflict when blocking cylinder overlaps with others.
- Cylinders allow drones to brake on time and maneuver to avoid collision.



AVS collision avoidance

Autonomous surface
vessel Collision avoidance:
Autosea project



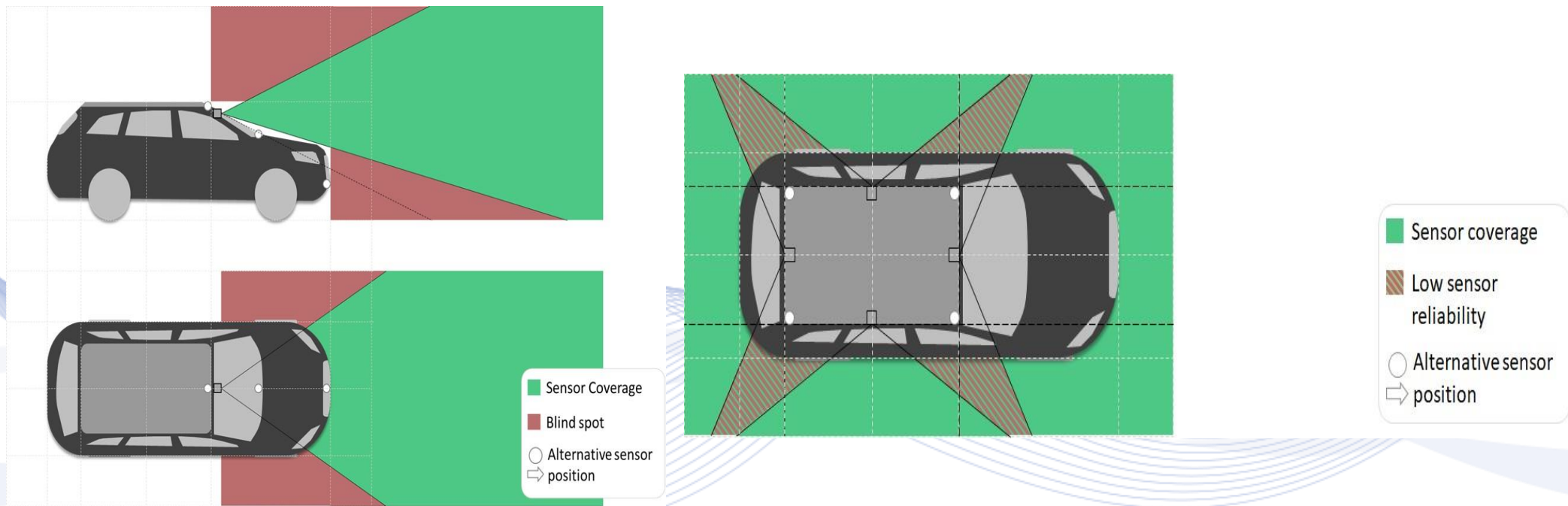
Example of AVS collision avoidance [AUT].

Autonomous System technologies

- Mission Planning and Control
- **Perception and Intelligence**
- Embedded computing
- Swarm systems
- Communications
- Societal technologies

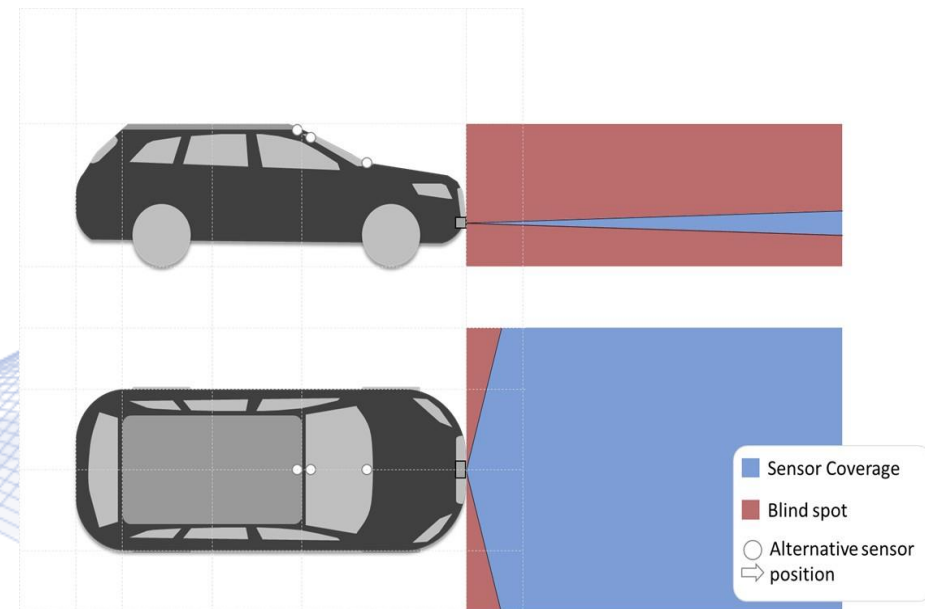
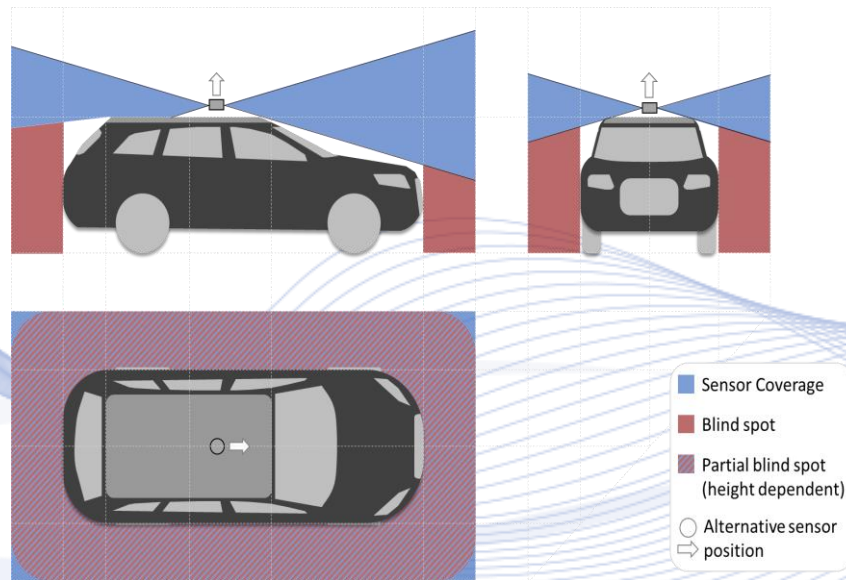
Autonomous car sensors

- Front/roof cameras



Autonomous car sensors

- Front/roof Lidars

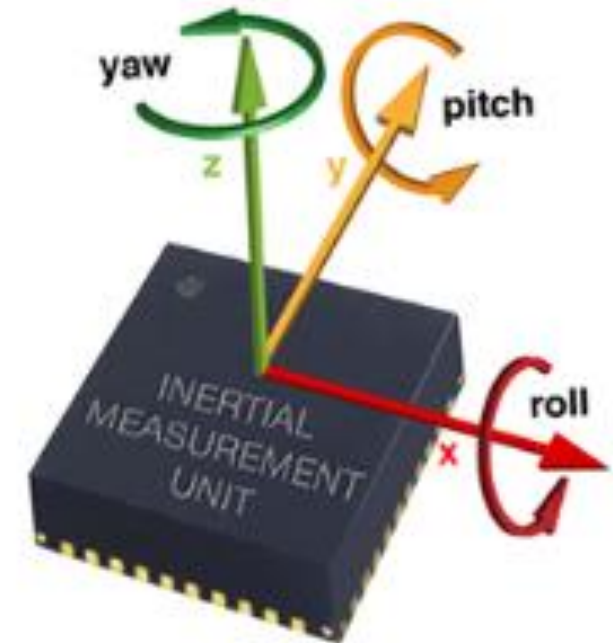


3D localization sensors: GPS

- Other Satellite systems: GLONASS (Russia), BeiDou (China), Galileo (EU).
- **RTK-GPS** uses measurements of the phase of the signal carrier wave, in addition to the information content of the signal and relies on a single reference ground station (or interpolated virtual station) to provide real-time corrections, providing up to cm-level accuracy.

Drone Sensors: IMU

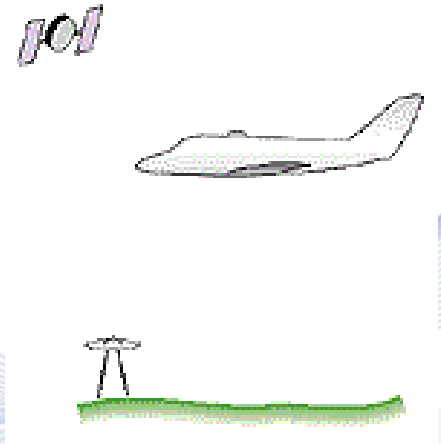
- ***Inertial Measurement Unit (IMU):***
 - It measures and reports a body's specific force, angular motion rate and, sometimes, the magnetic field surrounding the body.
 - It uses a combination of **accelerometers**, **gyroscopes** and, sometimes, also magnetometers.



Drone Sensors: Laser altimeter

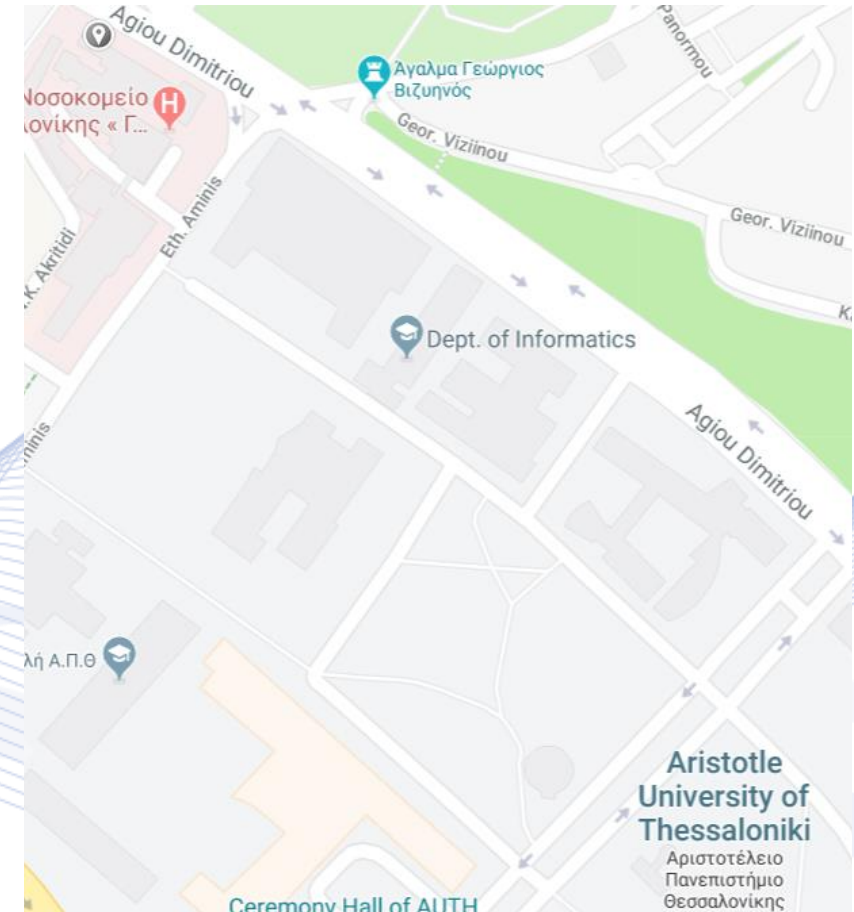
Laser altimeter:

- It measures the altitude (height) above a fixed ground level.
- It emits laser pulses which travel to the ground surface, where they are reflected.
- Part of the reflected radiation returns to the laser altimeter, is detected, and stops a time counter started when the pulse was sent out.
- The distance is then easily calculated by taking the speed of light into consideration.



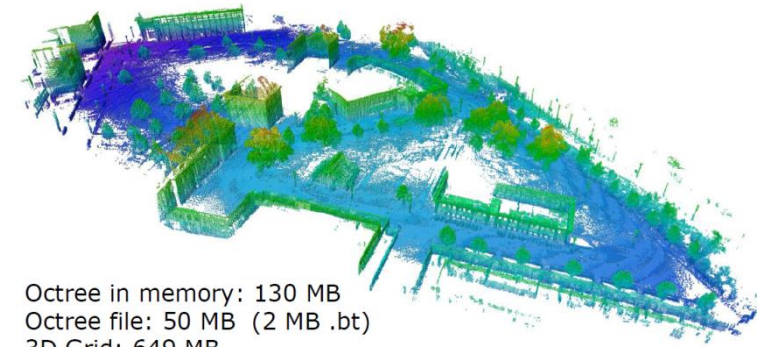
2D maps

- **Google maps.**
- **OpenStreetMaps.**
- Semantic annotated information:
 - (roads, POIs, landing sites) in KML format in Google Maps.
 - roads in OSM (XML) in case of OpenStreetMaps.
- **Google Maps JavaScript API.**
- **OpenStreetMaps API.**



3D maps

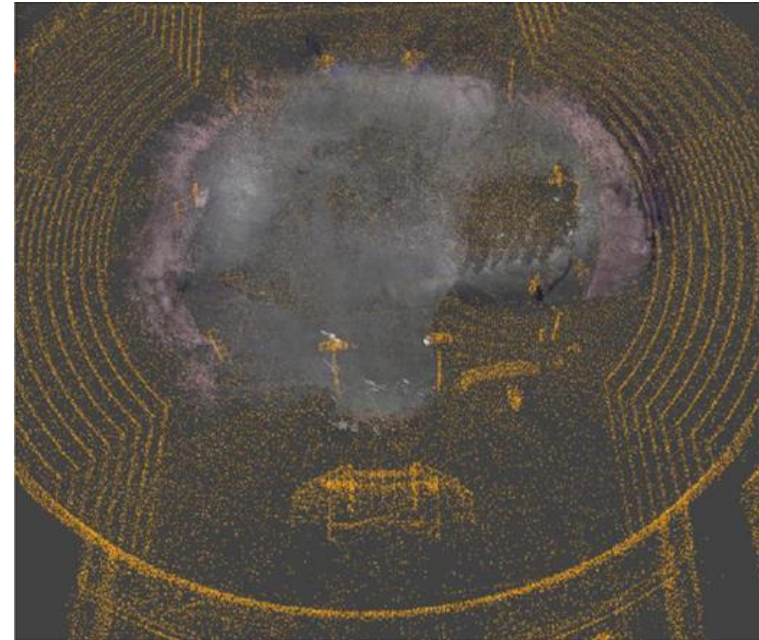
- Formats:
 - 3D triangle mesh.
 - 3D Octomap.
- **Octomap:**
 - The octomap is a fully 3D model representing the 3D environment, where the UAV navigates.
 - It provides a volumetric representation of space, namely of the occupied, free and unknown areas.
 - It is based on octrees and using probabilistic occupancy estimation.



Octree in memory: 130 MB
Octree file: 50 MB (2 MB .bt)
3D Grid: 649 MB

Geometrical mapping

Lidar mapping



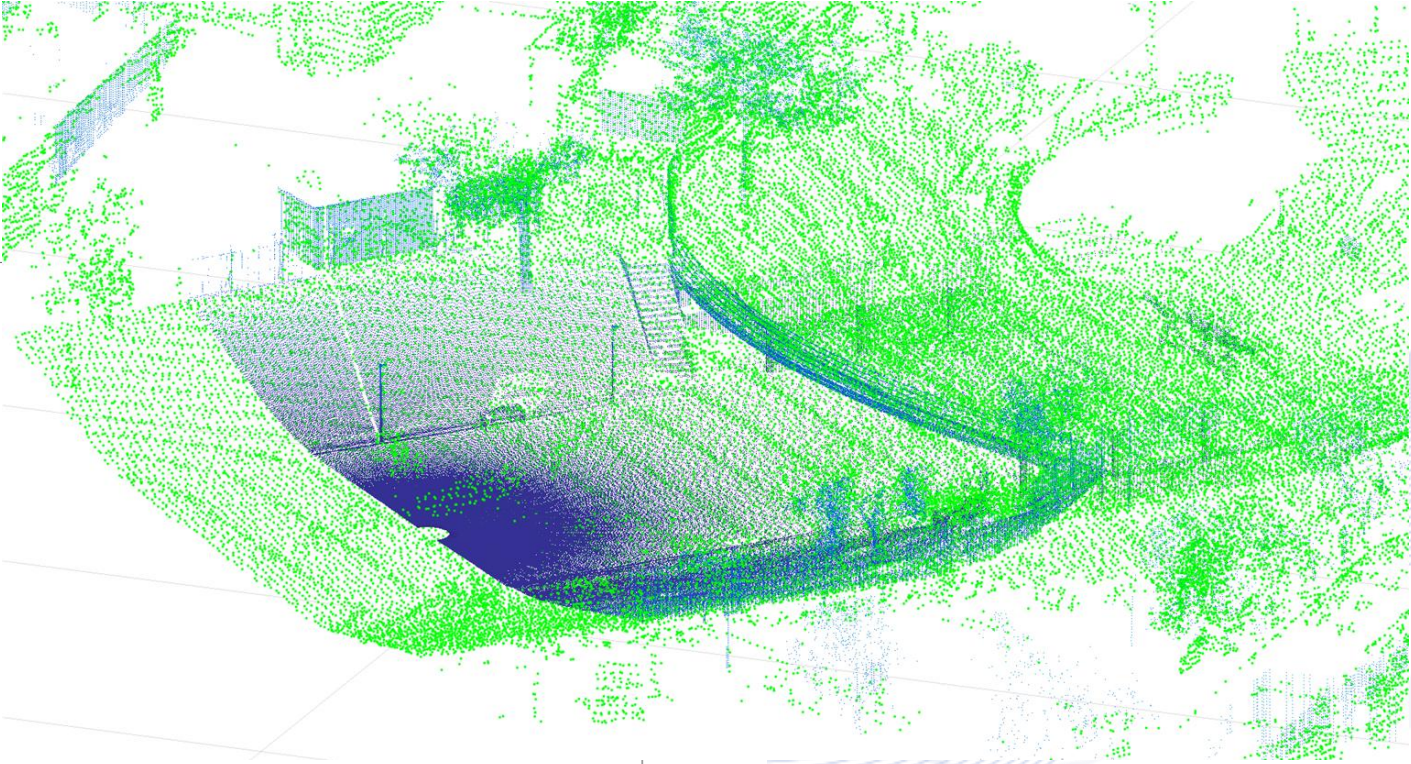
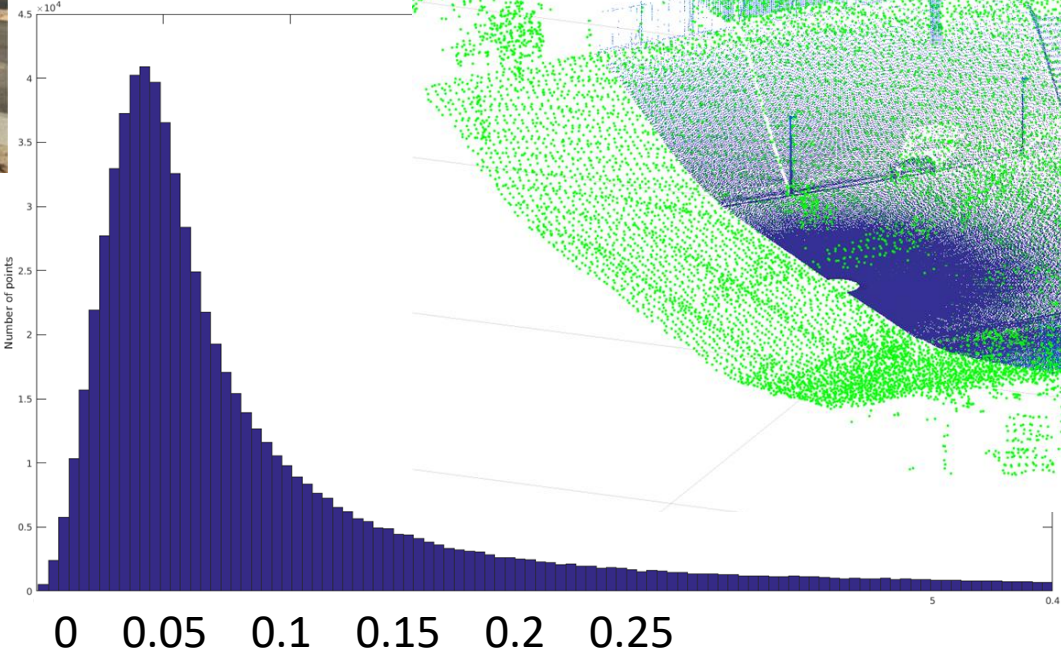
Repeatability

Dataset	Mean Error (m)	Median Error (m)	Min Error (m)
1	0,1377	0,1073	0,00098
2	0,1053	0,0769	0,00045
3	0,0847	0,0578	0,00083
4	0,1074	0,0792	0,00078
5	0,1722	0,1560	0,00130

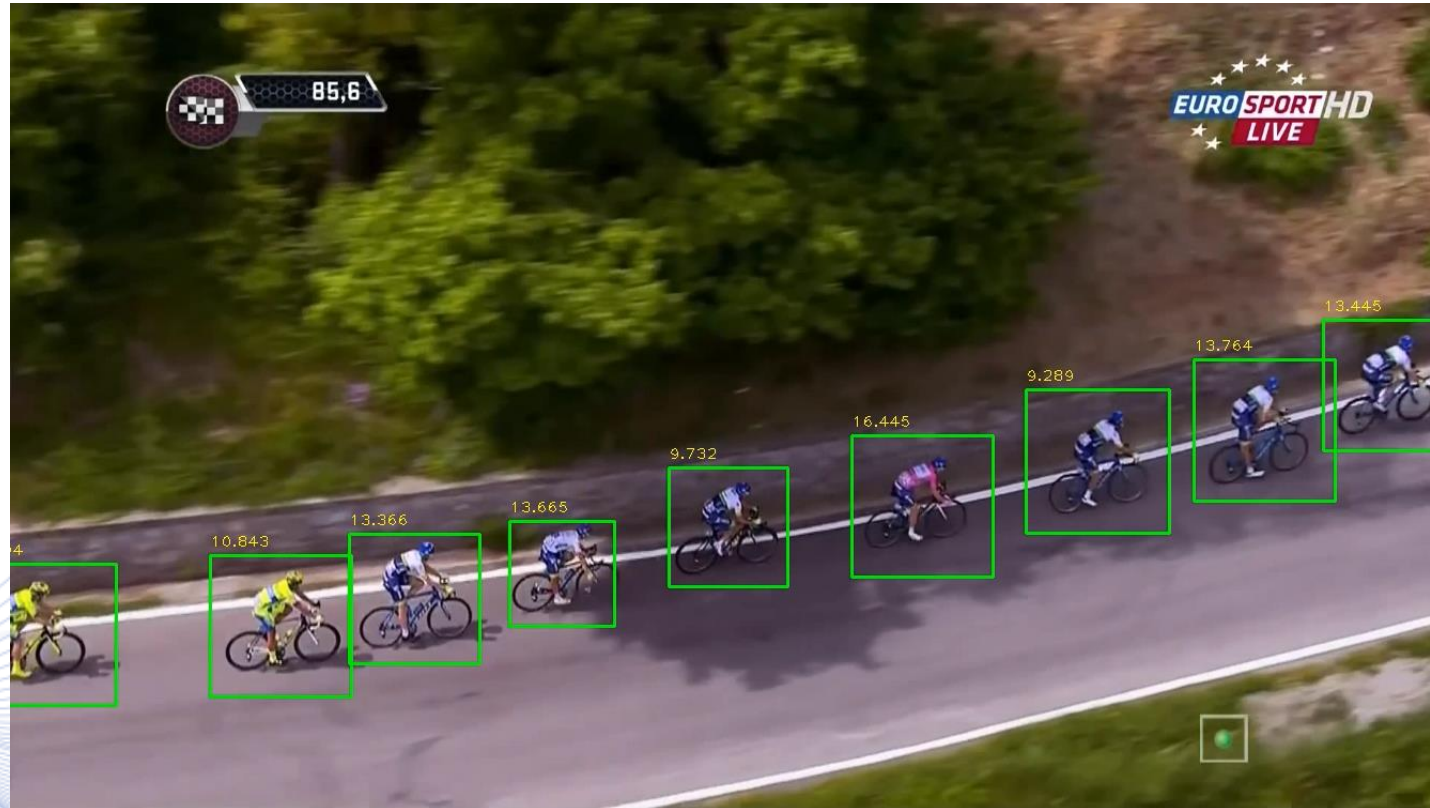
Geometrical mapping



Validation with a TOTAL STATION



UAV Object detection & tracking



Advanced autonomous car Intelligence

- Self-awareness
- Driver status modelling/recognition
- Affective computing
- Attention
- Human (e.g., pedestrian) intention prediction

Autonomous System technologies

- Mission Planning and Control
- Perception and Intelligence
- **Embedded computing**
- Swarm systems
- Communications
- Societal technologies

GPU and multicore CPU architectures. Algorithm mapping

- **NVIDIA embedded processing boards**
 - NVIDIA Jetson TX2
 - **NVIDIA Jetson Xavier**
- GPU and multicore CPU architectures
 - » Multicore CPUs
 - GPUs
- Algorithm mapping:
Convolutions

GPU and multicore CPU architectures. Algorithm mapping

- NVIDIA embedded processing boards
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- **GPU and multicore CPU architectures**
 - **Multicore CPUs**
 - GPUs
- Algorithm mapping:
Convolutions

Processing Units

- **Multicore (CPU):**
 - MIMD.
 - Focused on latency.
 - Best single thread performance.
- **Manycore (GPU):**
 - SIMD.
 - Focused on throughput.
 - Best for embarrassingly parallel tasks.

NVIDIA Jetson Xavier

- AI Computer for autonomous machines
- Designed for robots, drones and other
- Multiple operating modes (10/15/30 W)
- Comparison to TX2:

Greater than 10x the energy efficiency.

More than 20x the performance



CUDA

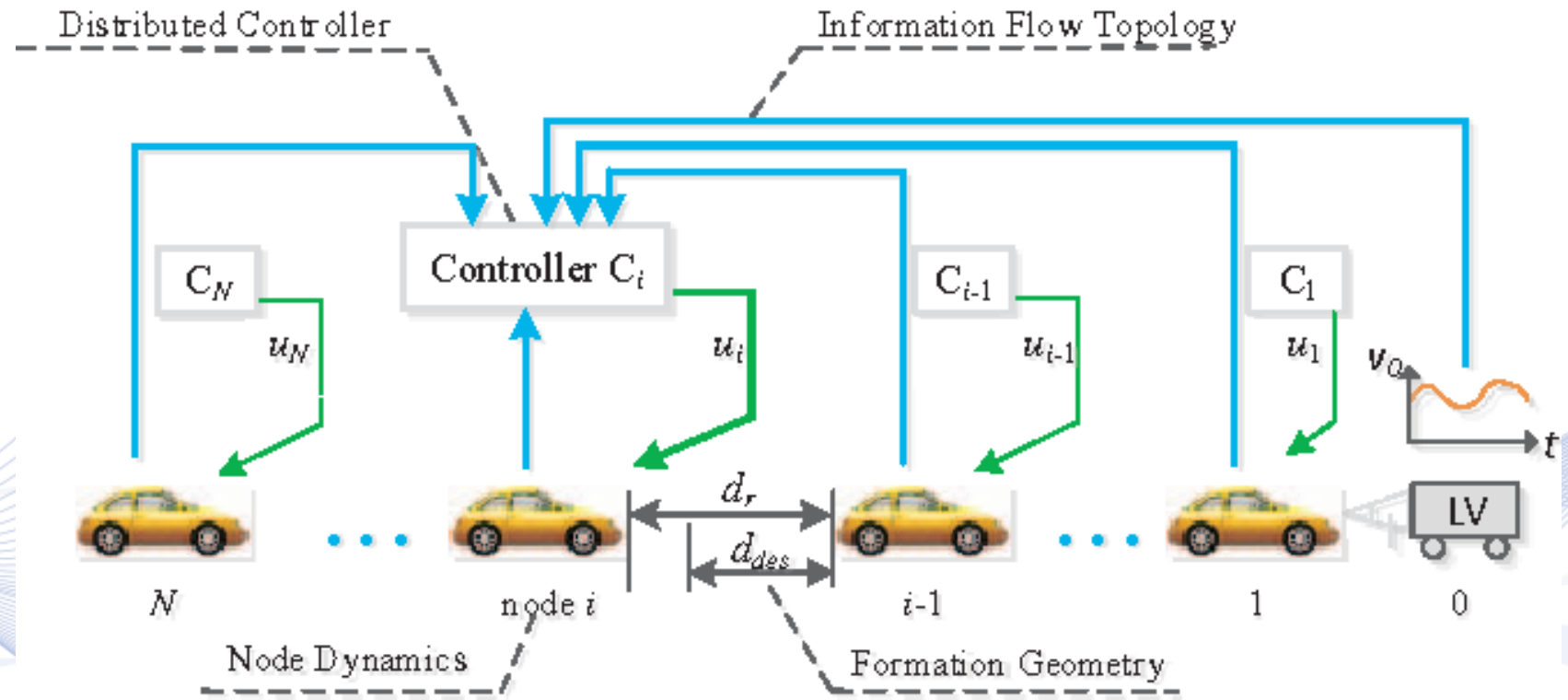
- Compute Unified Device Architecture (CUDA) is a parallel programming framework.
- Developed by Nvidia.
- Started as an attempt to give C/C++ programs access to GPU resources.
- Microarchitectures are named after famous physicists (Kepler, Maxwell, Pascal, Turing, Volta).

Autonomous System technologies

- Mission Planning and Control
- Perception and Intelligence
- Embedded computing
- **Swarm systems**
- Communications
- Societal technologies

Autonomous Systems swarms

- Car platoon control



Autonomous Systems swarms

- Lane-less highways
- Collision avoidance
- Fluid dynamics principles

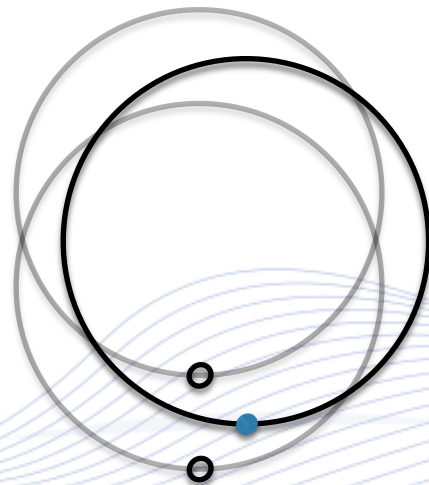


Drone swarms

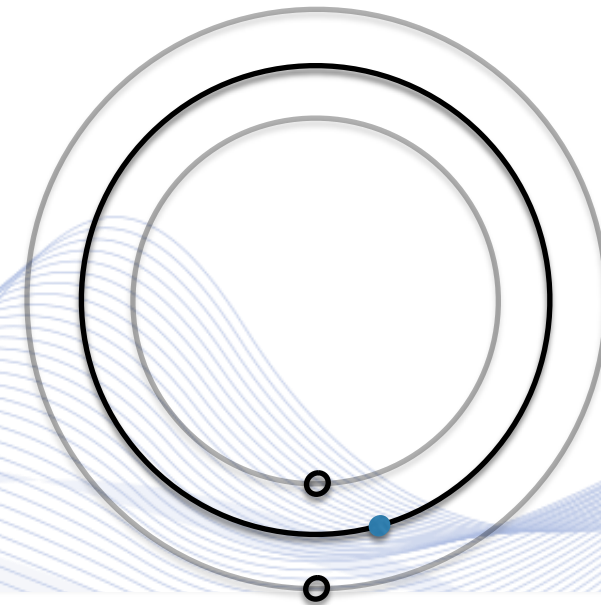


Leader-following for drone formation control

- Main idea:
Trailer-like behavior for the followers.

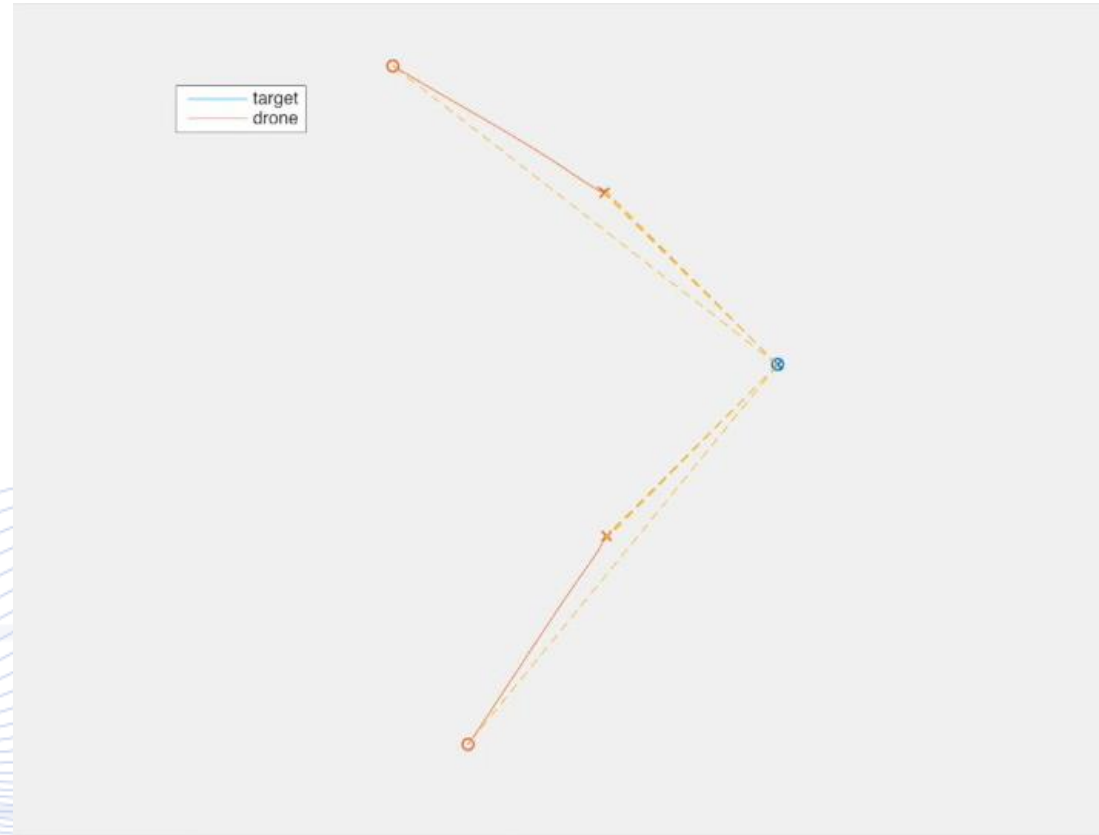


In inertial frame:
Translated identical paths



In trailer frame:
Different paths, no superposition

SA1 - Constant relative positions



Autonomous System technologies

- Mission Planning and Control
- Perception and Intelligence
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- Swarm systems
- **Communications**
- Societal technologies

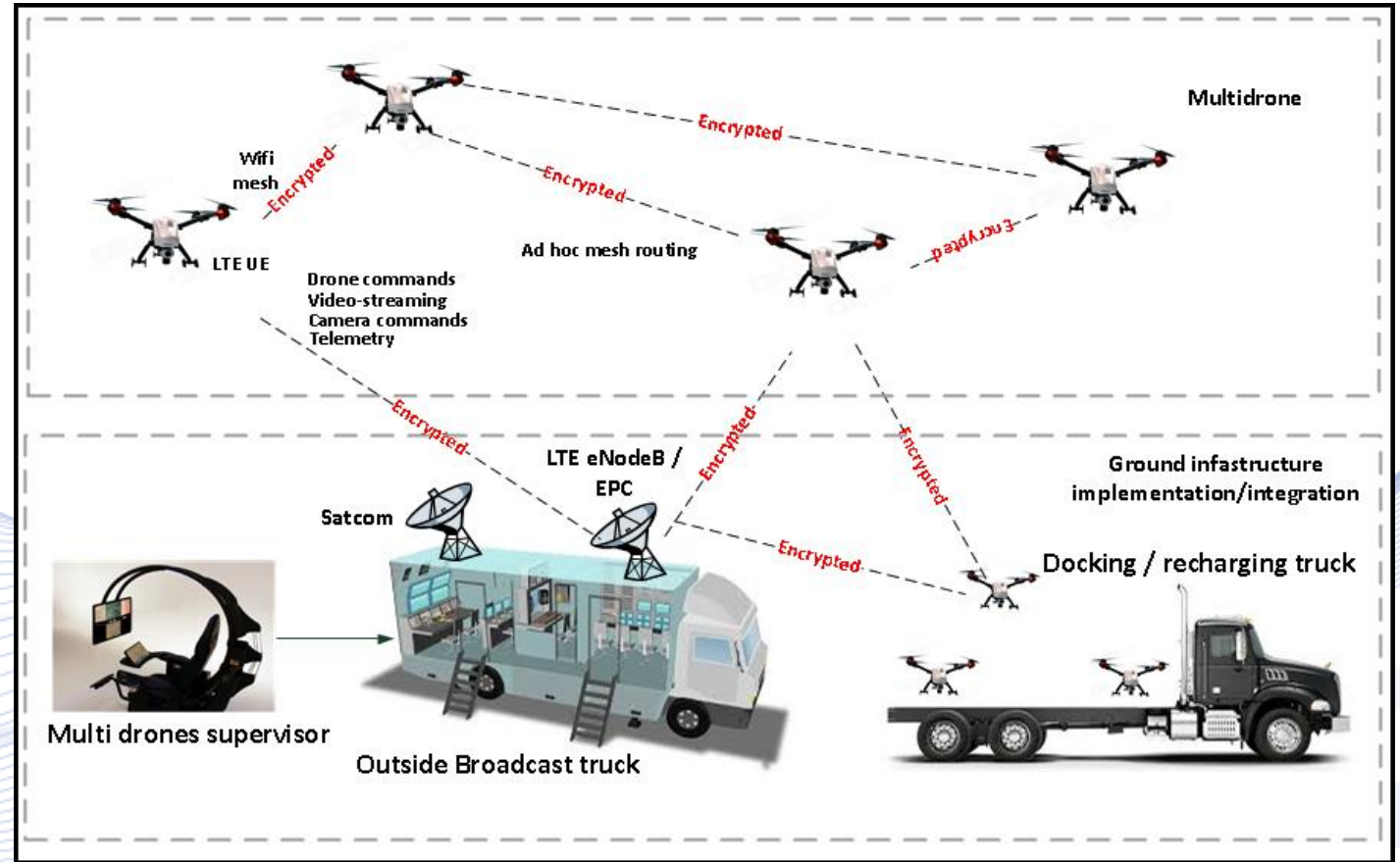
Autonomous System Communications

- **Communication infrastructure**
- Video streaming

Drone Swarm Communication infrastructure

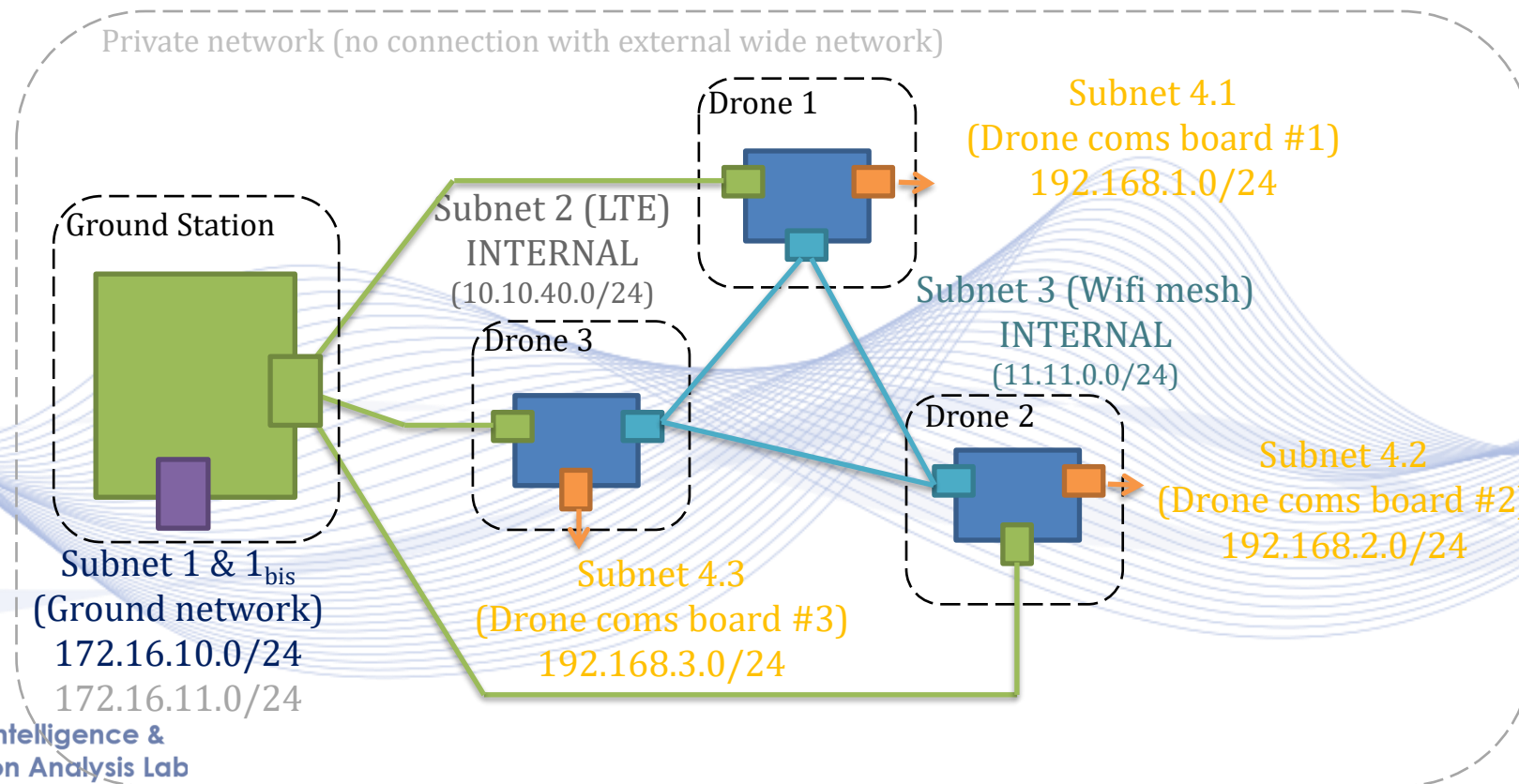


- Drone2Drone Communication.
- Drone2Ground communication.
- Live broadcasting.



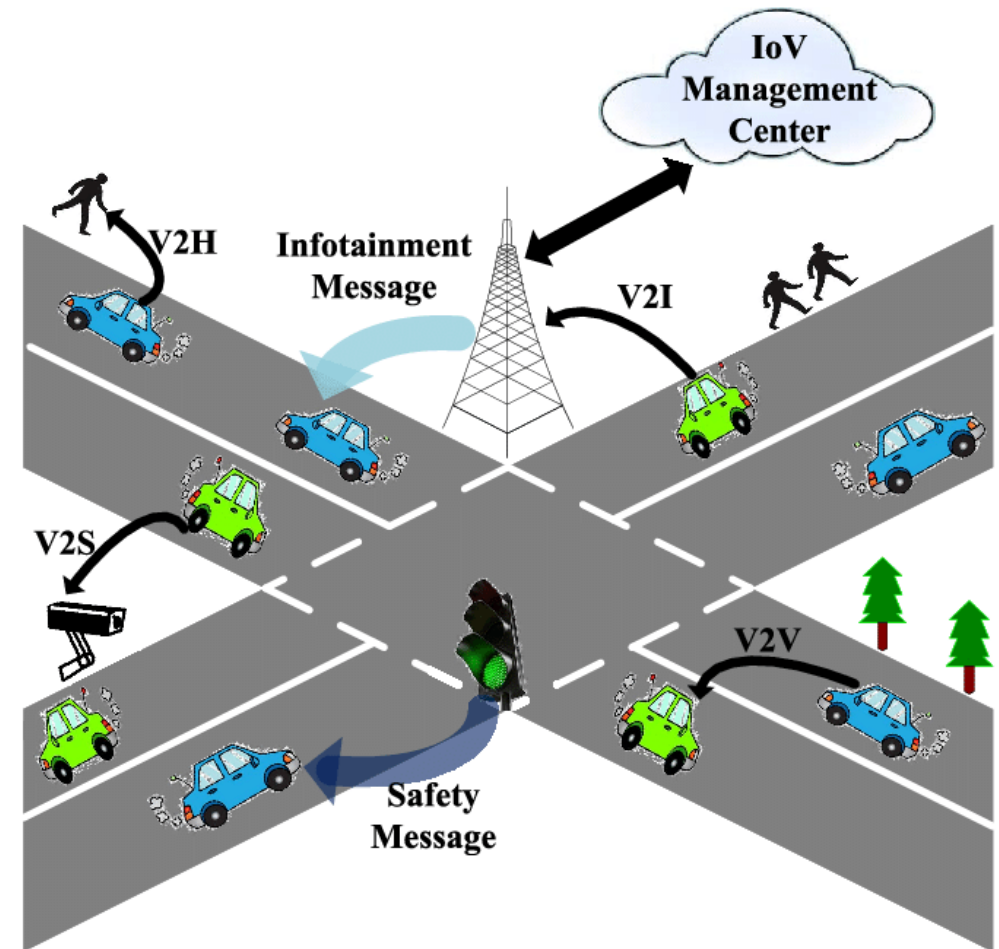
Drone Communications Infrastructure

Objective: Secured and resilient transparent IP access to drones / ground station (LTE and WiFi).



5G Communications Infrastructure

- Internet of Vehicles
- Massive deployment, throughput
- Ultra low latency networks
- Robustness
- Edge/cloud computing



Autonomous System Communications

- Communication infrastructure
- **Video streaming**

Drone Digital Video Streaming

Problem 1: Network



Considerations:

Wireless communication with receiver - weak & subject to failure (distance, obstacles, other wireless networks etc).

Good quality video is **massive** in terms of **Mbps** required to transfer it

1 second of 720p (1280x720) **8-bit video** requires 65.92MBytes – prohibitive.

Video **compression** must be used prior to streaming:

H264 & H265 coding are great candidates...

... but they inevitably introduce **delays** (compression + decompression)

Lossy: must find **trade-off between latency & quality**.

Which **network protocol** should be used?

Real-time Transport Protocol (**RTP**) with User Datagram Protocol (**UDP**)

TCP is also standardized for use with RTP, but favors reliability instead of timeliness.

Drone Digital Video Streaming

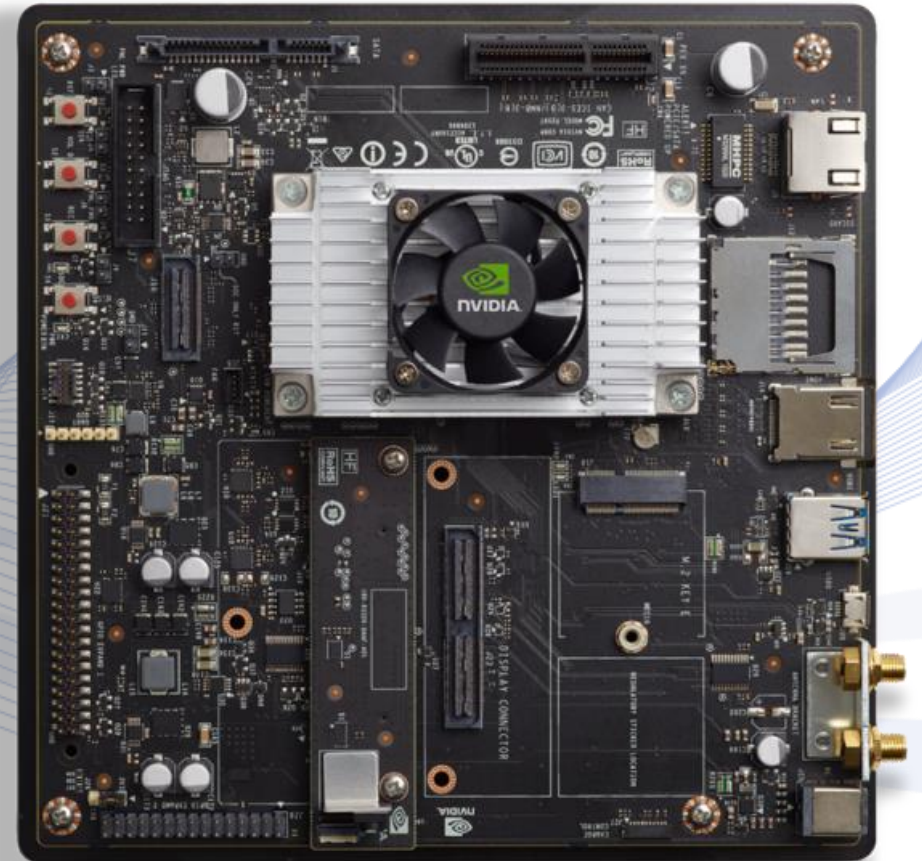
Problem 1: Network



Compression takes place on-board the drone
NVIDIA's Jetson TX2 module offers.

hardware accelerated image/video
compression.

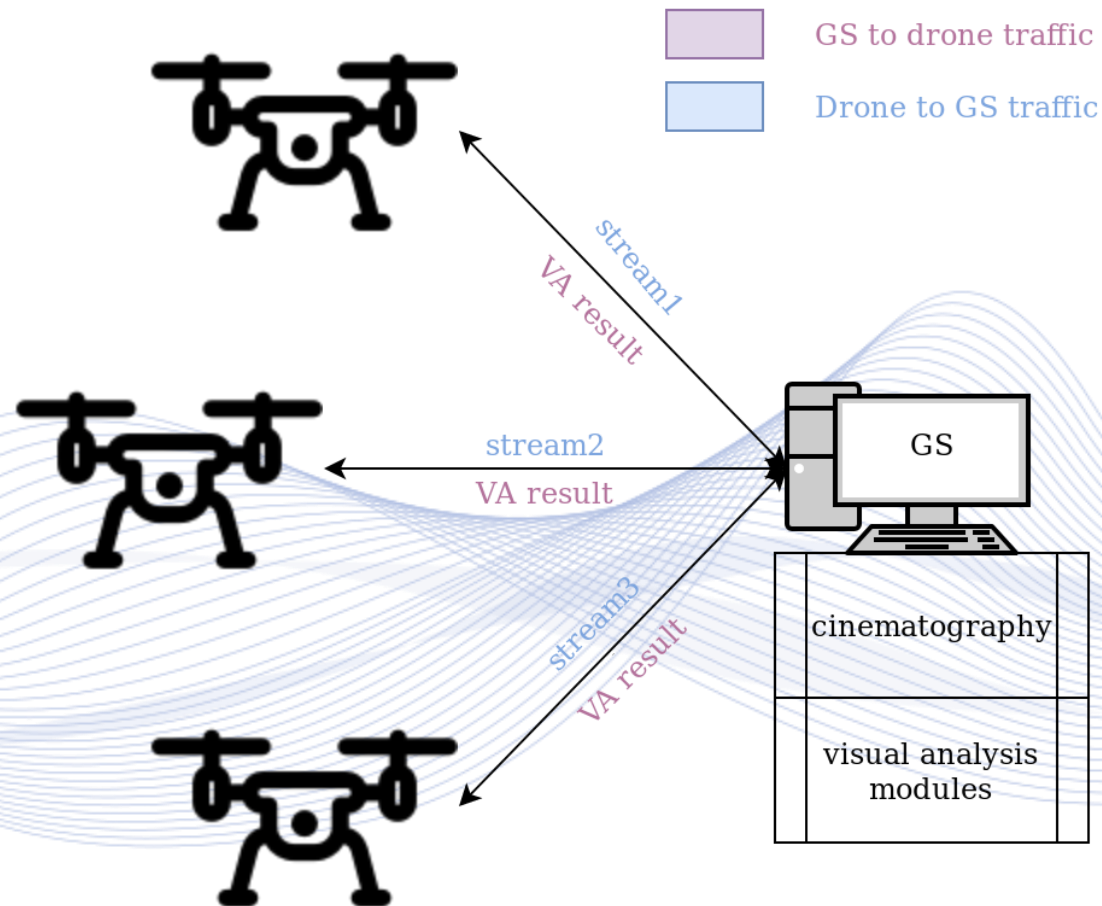
Also a 256-core Pascal @ 1300MHz GPU
with capability comparable to an
Intel Xeon E5-2960 v4 CPU in **Deep
Learning** tasks.



Drone Digital Video Streaming

Problem 2: Synchronization

Scenario 2: multiple drones - one ground station.



Solutions & Tools: Gstreamer

Gstreamer is written in C, but offers bindings in multiple languages:

<https://gstreamer.freedesktop.org/bindings/>

Recommended: original C or C++ or Python.

Sample streamer + receiver are provided in Python:

They show how to access pipeline elements & modify them, intercept buffers etc

<https://lazka.github.io/pgi-docs/#Gst-1.0> python bindings

Gstreamer official documentation:

<https://gstreamer.freedesktop.org/documentation/>

Useful elements for custom streams: appsrc and appsink.

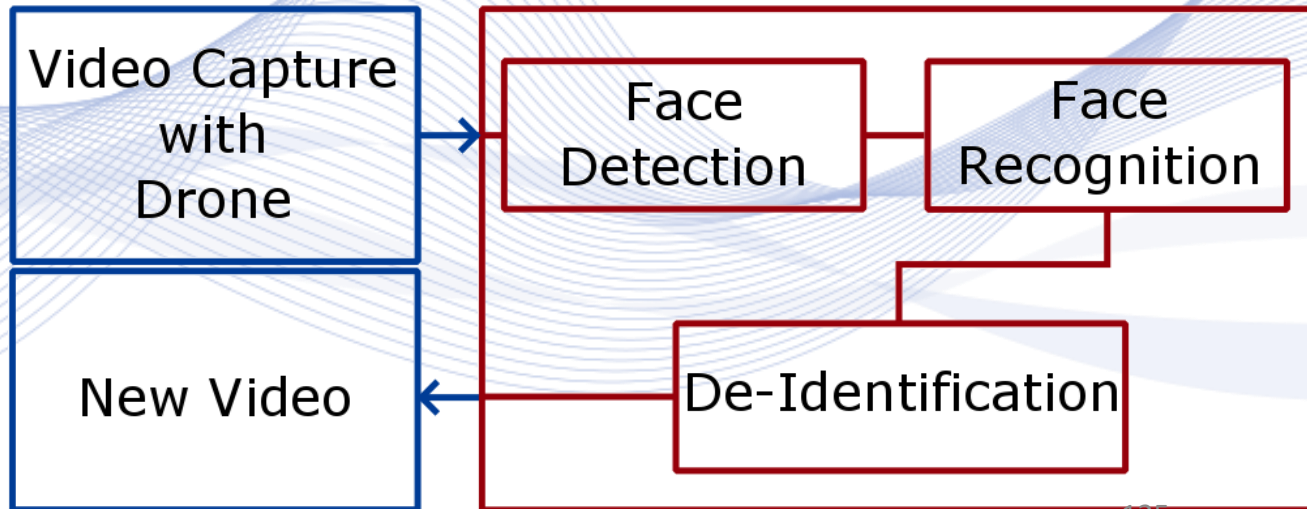
Autonomous System technologies

- Mission Planning and Control
- Perception and Intelligence
- Embedded computing
- Swarm systems
- Communications
- **Societal technologies:**
 - Security
 - Safety
 - Privacy protection

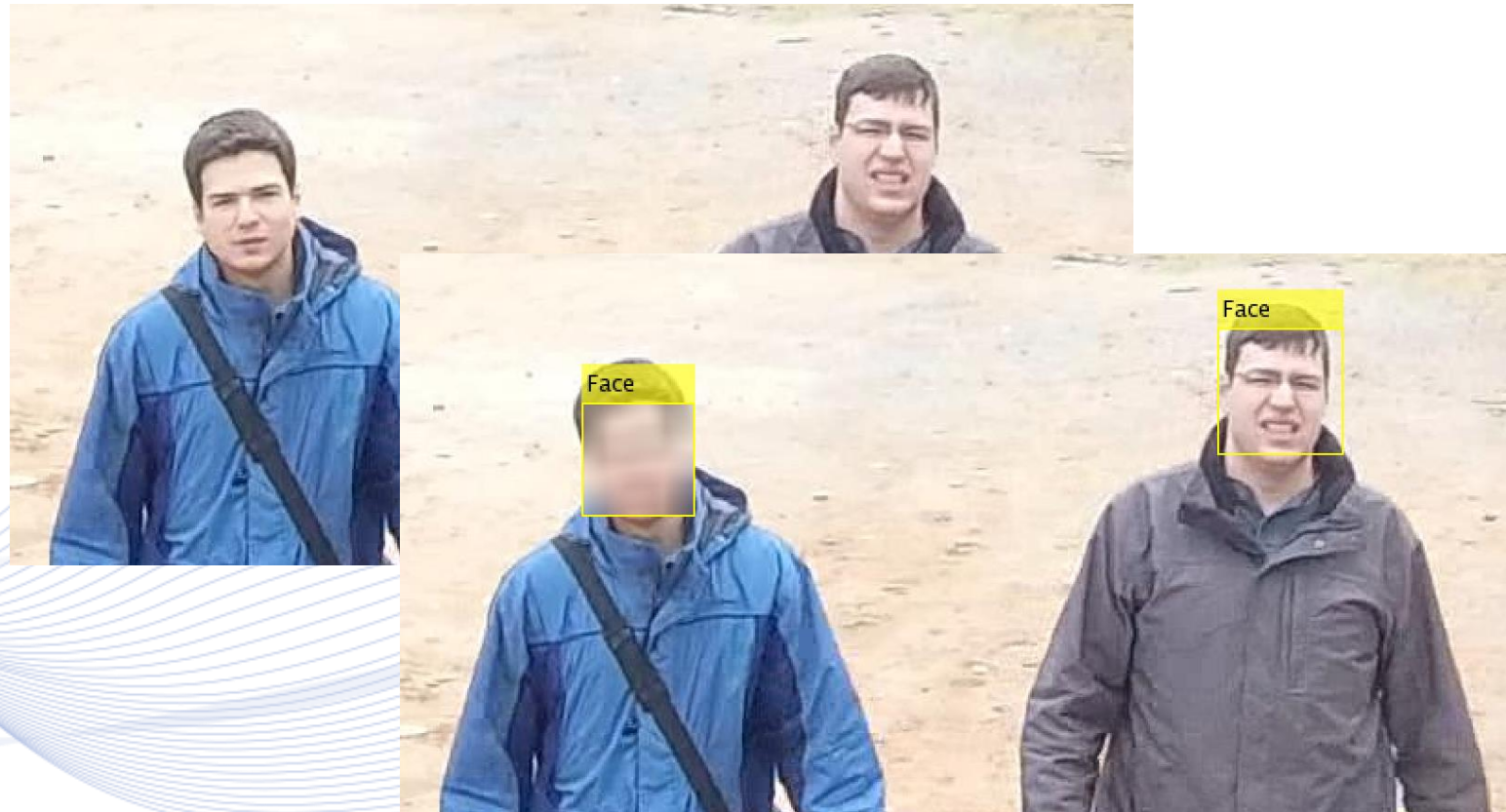
Privacy Protection

- An issue of ethics and security
- Post-production stage
- Approaches
 - Face de-detection (Face detector obfuscation)
 - Naïve approach
 - SVD-DID
 - Face de-identification (face recognizer obfuscation)
 - Gaussian blur
 - Hypersphere projection

Application on drone videos



Face recognition/de-identification/privacy protection



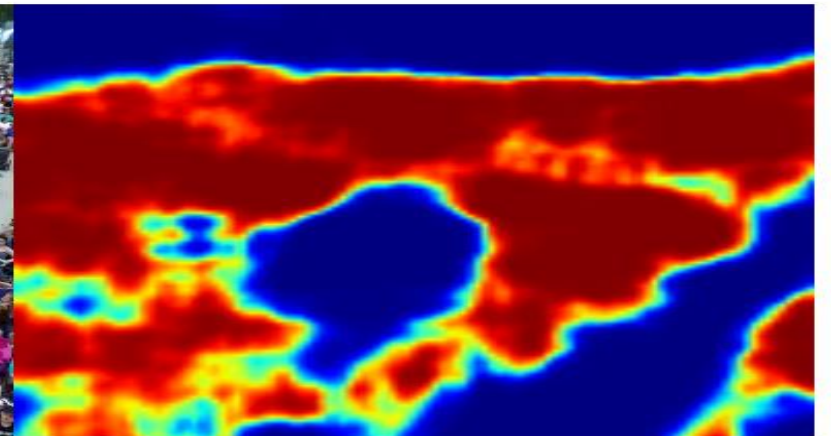
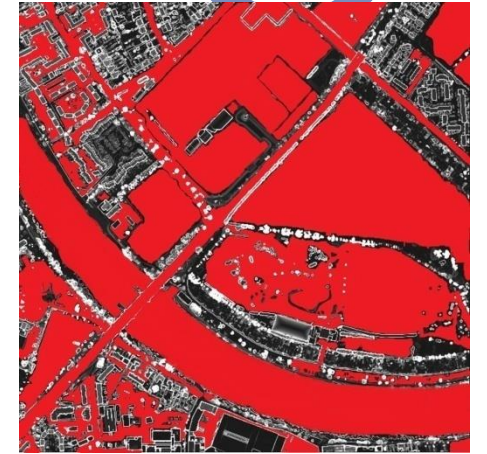
UAV flight regulations in EU



- UAVs < 2kg are allowed within a 50m flight radius without professional pilot license.
- Pilot license and drone insurance are required for all professional applications.
- UAVs > 2kg of weight may be required to carry emergency parachutes (France).
- UAVs exceeding 15kg of weight might require special license or even be prohibited (Germany).

Other UAV safety issues

- Potential Landing Site Detection
- Crowd detection and avoidance



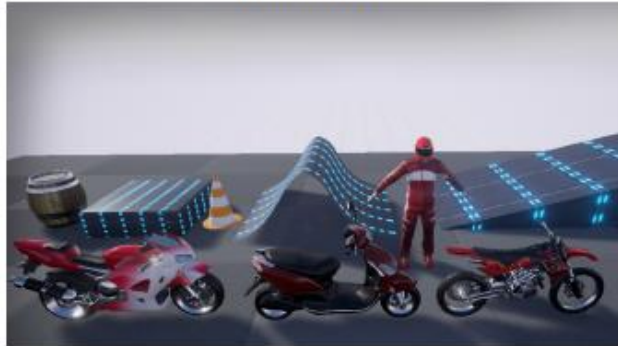
Mission simulations

- Simulations in Gazebo
- Simulations in Unreal Engine and AirSim
- Simulations for training data generation

Pilot Study - Test Content

Object Models

O1 Motorcycles



O2 Sports Car



Background Environment

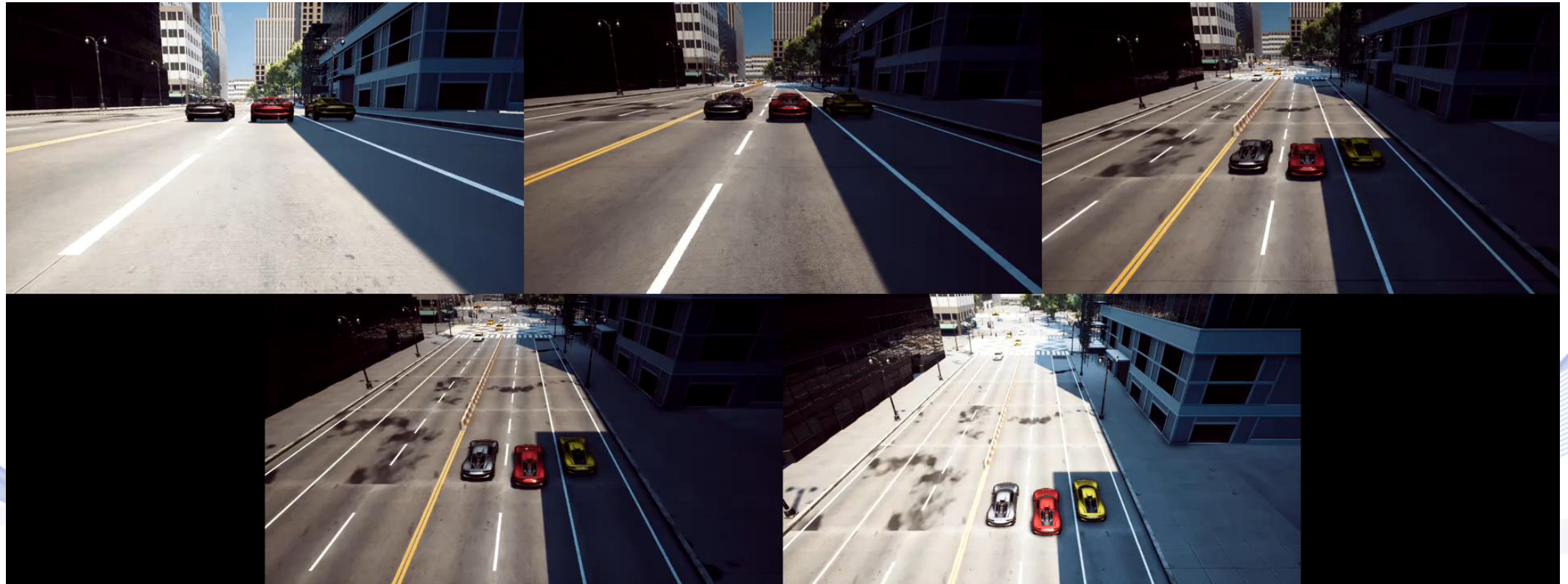
B1 Down Town



B2 Industrial City



Test Sequence Example II: S2



VIDEO: Scenario 2 with drone height of 1, 2, 6, 10 and 14m.

Bibliography

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Q & A

Thank you very much for your attention!

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