

COORDINATING EDGE AND CLOUD FOR BIG DATA ANALYTICS

When big data meets smart cities...



Chongquing, August 28th, 2019

Paolo Burgio, UNIMORE















Modena Automotive Smart Area



1 Km² urban area with connectivity enables IoT devices to exchange information

- Hundreds of smart cameras
- Traffic scanners
- Three highly-connected cars
 - UNIMORE's EU projects Class & Prystine
 - Vehicle-to-infrastructure (V2I), vehicle-to-cloud (V2C), vehicle-to-vehicle (V2V)
 - Cameras @4K, long-range and middle range radars and ultrasound sensors



V2V, V2I, V2C connectivity







Real value for the city: (Big) data analytics



Build a real-time dataset for Smart City Distributed Awareness

- Traffic/mobility analysis and prediction
- Critical scenarios identification
- Road user behaviour understanding and prediction
 - Improved algorithms for autonomous driving
- Data anonymization for GDPR compliance
 - Implemented at source/edge level



Distributed real-time urban awareness



Infrastructure sensors



In-vehicle sensors



Real-time detection





Low-latency V2X communication





Data analytics

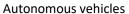


Public authorities



Traffic enforcement







CHONGQUING, Aug 28th, 2019

(Prototype) control center tools



- The presentation includes a video here
- For more information, please contact: class-project@bsc.es



HUGE credits to...



Dott. Roberto Cavicchioli

Prof. Nicola Capodieci

...and all of their team







Vulnerable Road User detection & avoidance

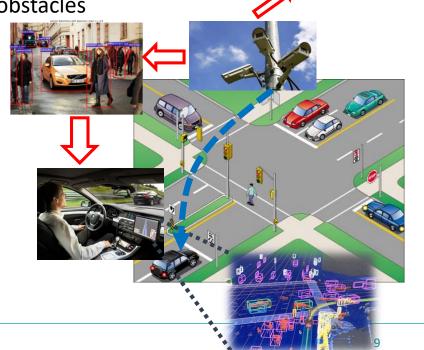
CLASS

(VRUs are: other vehicles, buses, trucks, pedestrians, bicycles, etc.)

To cloud infrastructure

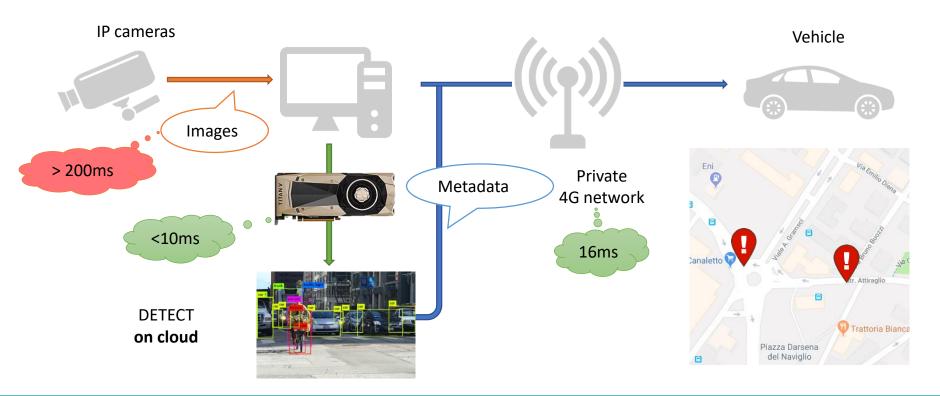
- Infrastructure cameras detect road users and obstacles
- Also, vehicle cameras detect objects!
- Sent to L3/L4 vehicles (V2I) in real-time for collision avoidance

The goal:
camera-to-car comm
to deliver high-criticality services
< 100ms!



Traditional Smart City flow: high latency





Class & MASA: fog computing in Real-Time CLASS





Edge nodes

- Hundreds of smart cameras detecting VRUs in real-time
- Local high-performance embedded board
- Information sent in V2I to vehicles, and in ETH to Fog nodes

Fog nodes

- Four local servers
- High-end computing system
- Additional (complex) tasks not feasible on edge

Cloud

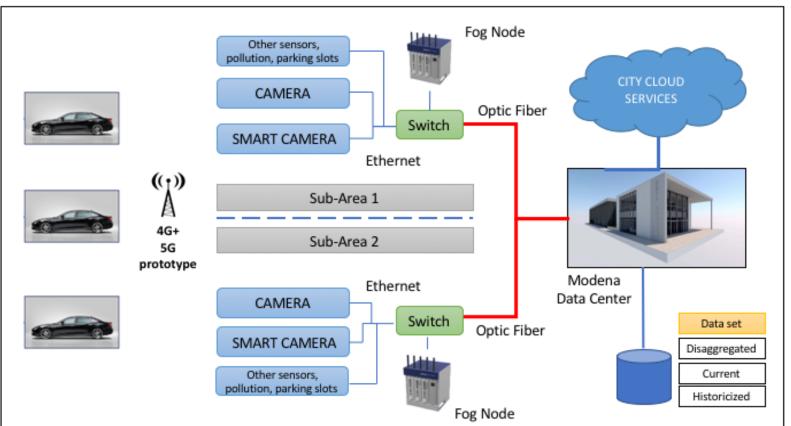
- Fog nodes are fiber connected to main control center
- Data storage, learning, low-criticality

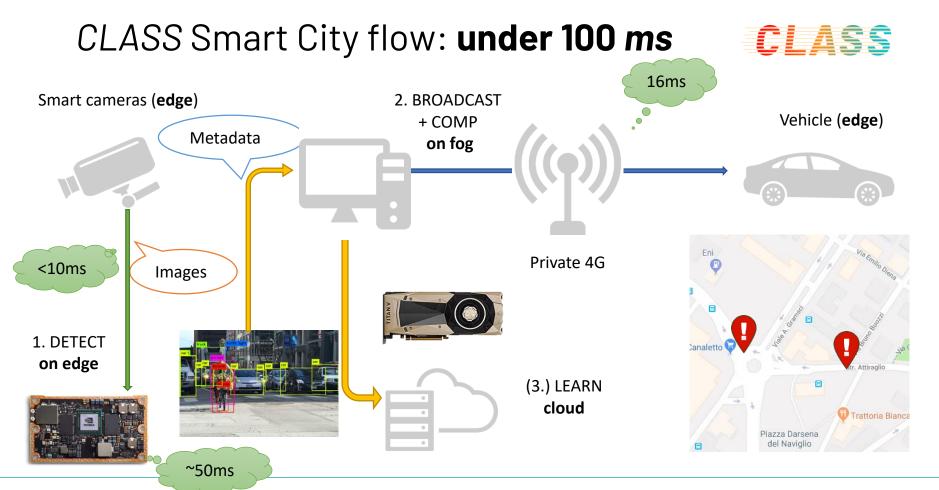




Edge + Fog + Cloud

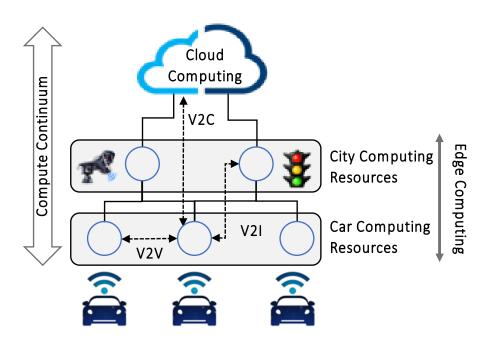






How much data?



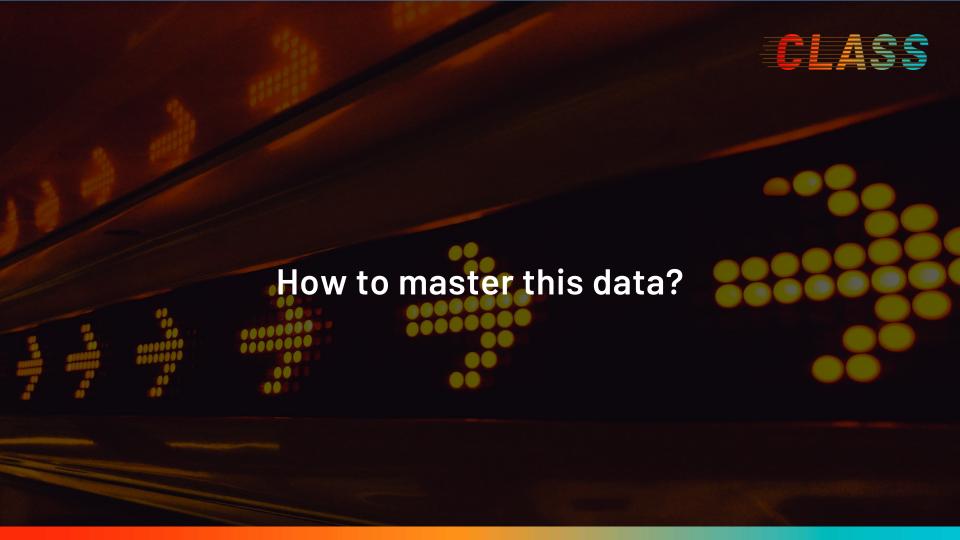


11.4 GB/s data-sets

(3 cars + 1 km2 sensing area)

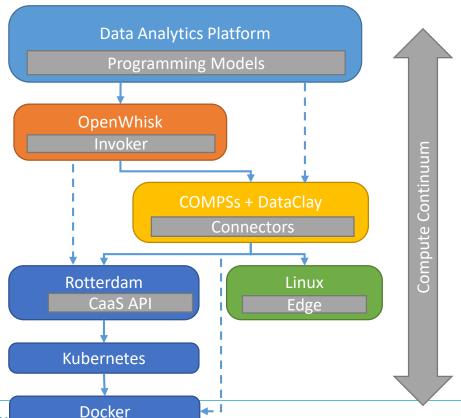
Goal: services in 100ms

- Acquisition
- Processing
- Transmission



The nerd corner: CLASS SW infrastructure





Edge computing nodes for smart cameras



NVIDIA Tegra X2 SoM

- Dev platform @~350€
- Production platform for autonomous driving @~5000€



- Complete Sw dev toolchain
 - GNU/Linux
- Support by NVIDIA

- Already two generation ahead!
 - Xavier SoM, Pegasus board

Multiple trackers: view from camera



- The presentation includes a video here
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Position of objects



- Detection on edge
 - Bounding boxes on the camera image
- GPS conversion
 - (x, y) → (GPS latitude, GPS longitude)
- Also GPS-to-meters
 - Tracking from a map point of view and not from the camera itself
 - Faster to fuse data from multiple cams



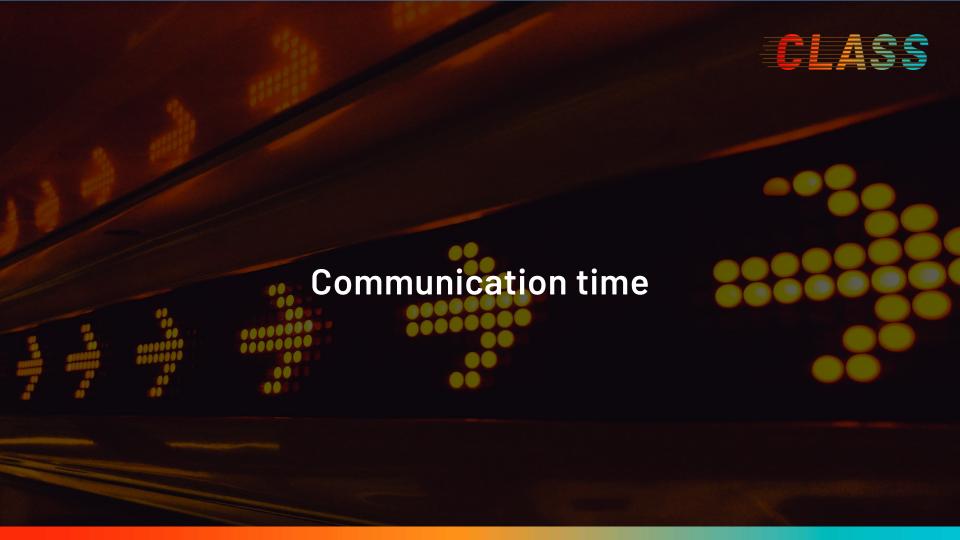
Detection performance (80 objects)



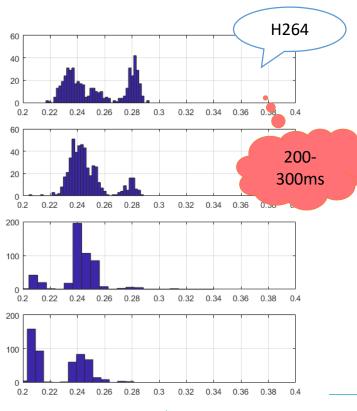
- NVIDIA's high-end Titan V (fog node): ~ 3 ms
- NVIDIAS's Tx2 (edge node): ~ 10 ms

Comparable!!

(and, still under 100ms)



IP camera transmission latency

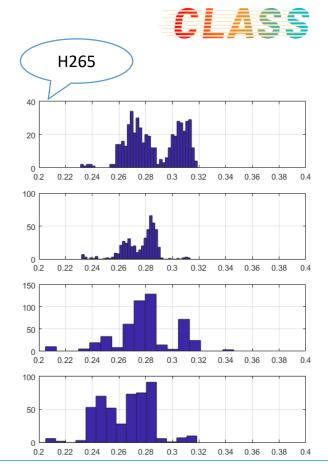


1920 x 1080

1280 x 720

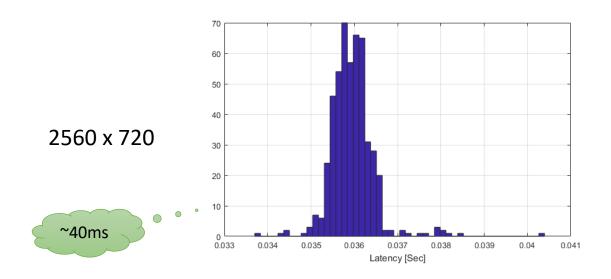
768 x 432

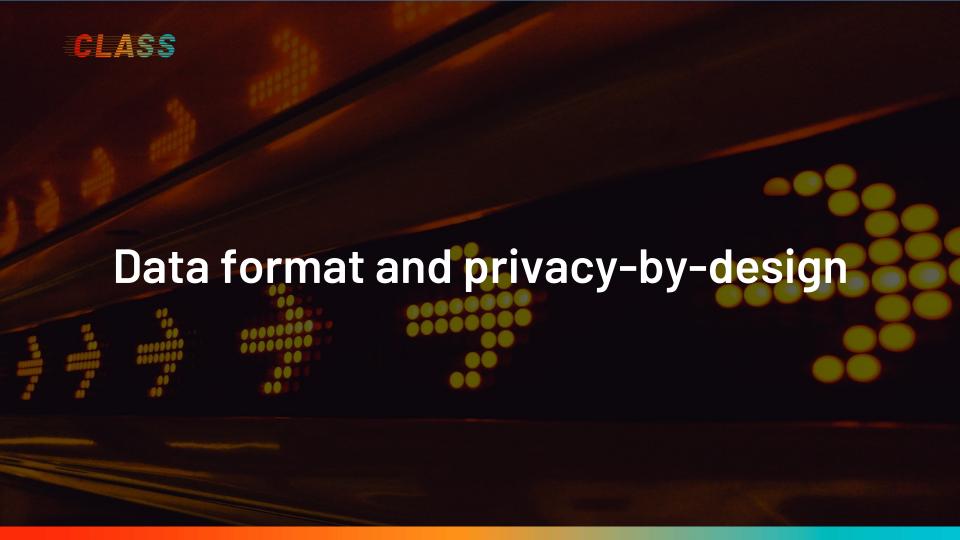
640 x 480



USB3 camera latency (Zed Stereo Camera)







Network Connection capabilities



Edge-to-fog (cameras)

- Ethernet LAN
 - Up to **1Gbit/s**

Fog-to-cloud

- Optical Fiber
 - Up to 100Gbit/s

Edge-to-all (V2X)

- Private 4G/LTE
 - 16 Mbit/s
 - <20 ms for CLASS metadata transmission
- **5G** to come soon (?)
 - Negligible overhead for CLASS metadata transmission (antenna in low-power mode!)
 - Not for single-device BW, but for multple-edge scalability!

Compact data format (aka: nerd time #2) CLASS



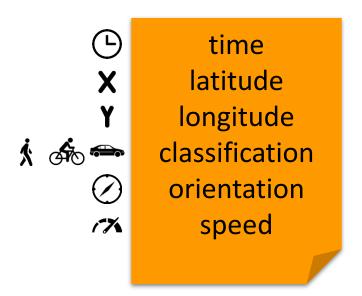
Standard UDP (will enhance with cryptography)

- Sender Identifier: 1B
- Timestamp: 1B
 - 1/200 sec resolution
- # objects that will be sent: 1B
- For each object:
 - Latitude + longitude: 8B
 - **Speed**: 4B (0 to 128 KM/h with 0,5 KM/h resolution)
 - Orientation: 1B (0 to 360° with ~2 degrees resol.)
 - Category identifier: 1B (people, dog, vehicle, bike...256 categories)

Around half KB for 50 objects sent



Data Anonymization



| LAT/LON PRECISION | MEANING |
|---|---|
| 28°N, 80°W | YOU'RE PROBABLY DOING SOMETHING SPACE-RELATED |
| 28.5°N, 80.6°W | YOU'RE POINTING OUT A SPECIFIC CITY |
| 28.52°N, 80.68°W | YOU'RE POINTING OUT A NEIGHBORHOOD |
| 28.523°N, 80.683°W | YOU'RE POINTING OUT A SPECIFIC SUBURBAN CUL-DE-SAC |
| 28.5234°N, 80.6830°W | YOU'RE POINTING TO A PARTICULAR CORNER OF A HOUSE |
| 28.52345°N, 80.68309°W | YOU'RE POINTING TO A SPECIFIC PERSON IN A ROOM, BUT SINCE YOU DIDN'T INCLUDE DATUM INFORMATION, WE CAN'T TELL WHO |
| 28.5234571°N, 80.6830941°W | YOU'RE POINTING TO WALDO ON A PAGE |
| 28.523457182°N 80.683094159°W | "HEY, CHECK OUT THIS SPECIFIC SAND GRAIN!" |
| 28.523457182918284°N, 80.683094159265358°W | EITHER YOU'RE HANDING OUT RAW FLOATING POINT VARIABLES, OR YOU'VE BULT A DATABASE TO TRACK INDIVIDUAL ATOMS. IN EITHER CASE, PLEASE STOP |

https://git.hipert.unimore.it/rcavicchioli/masa_protocol

Thanks for your attention.



Stay tuned!

www.class-project.eu

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