Information & Brokerage Session

Horizon Europe Cloud-Edge-IoT Call 2022

2 February 2022 | 9:30 - 12:30 CET, online

Organised by



In collaboration with





Information & Brokerage Session Horizon Europe Cloud-Edge-IoT Call 2022

10:45 12:00

PITCH SESSION

Cognitive Cloud

10:50 - 10:55 Alberto P. Marti, OpenNebula

10:55 - 11:00 Cagatay Yilmaz, Green Continuum Information Technologies Ltd.

11:00 - 11:05 **Tuna Tugcu,** *Bogazici University*

11:05 - 11:10 **Tatiana Silva,** *Tree Technology*

11:10 - 11:15 Karthee Sivalingam, Huawei

Decentralised Intelligence

11:20 - 11:25 Santiago Macho Gonzalez, Tree Technology

11:25 - 11:30 **Andreas Harth,** Friedrich-Alexander-University Erlangen-Nuremberg and Fraunhofer IIS Nuremberg

Open Source

11:30 - 11:35 Ricardo Vitorino, Ubiwhere

Next Generation Edge Cloud Platform

Build & Manage an On-Demand Edge Cloud Using Resources from Bare-Metal Providers





This work has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement ONEedge 880412



OpenNebula Systems

The company developing OpenNebula and leading ONEedge.io



- European open source company.
- The only European open source laaS solution, born in 2008.
- A success story emerged from EU innovation programs.
- Members of EOSC and GAIA-X.
- Offices in the US (Massachusetts) and Czech Republic.





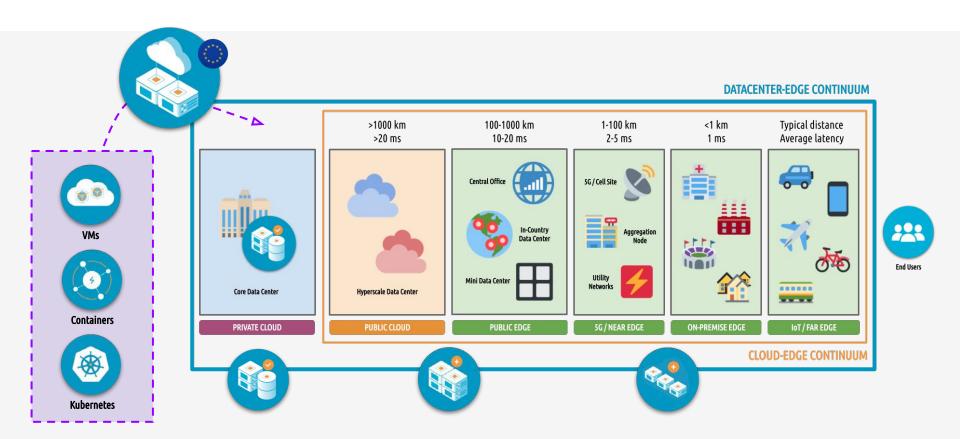






Edge Computing Innovation

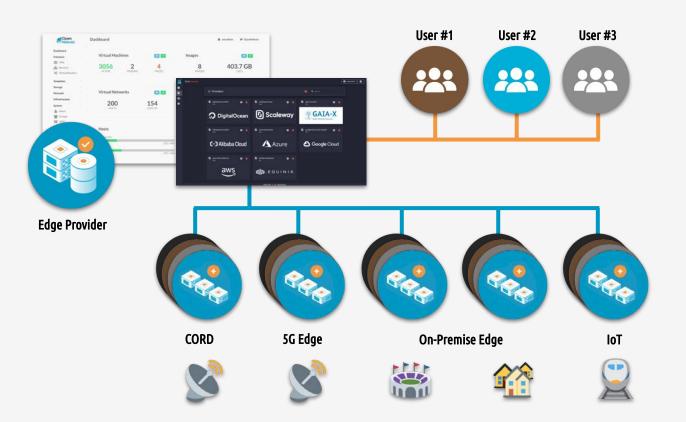
Towards a meta-orchestrator for the Datacenter-Cloud-Edge-IoT continuum





Our Vision on Edge Computing

Building an open source, vendor-neutral platform for Europe



- / Multi-Tenancy
- ✓ Self-Service Portal
- User Quotas
- ✓ Workload Isolation
- ✓ Multi-Tier Applications
- ✓ Elasticity Rules
- ✓ High Availability
- ✓ Federation
- ✓ Workload Portability
- ✓ Automatic Provisioning
- ✓ Public Marketplace of Virtual Appliances



Discover OpenNebula

Get in contact with us! <a>research@opennebula.io

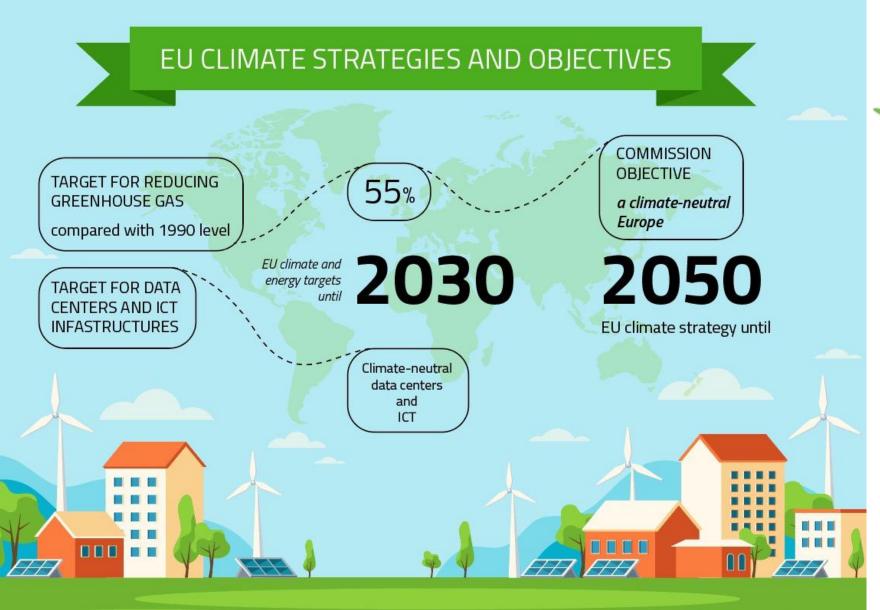








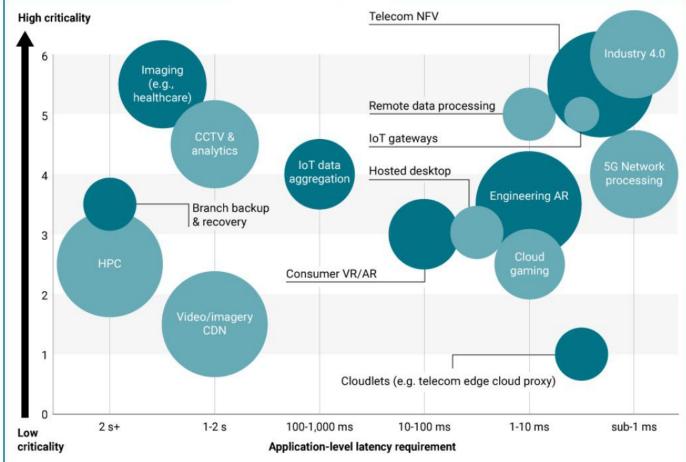






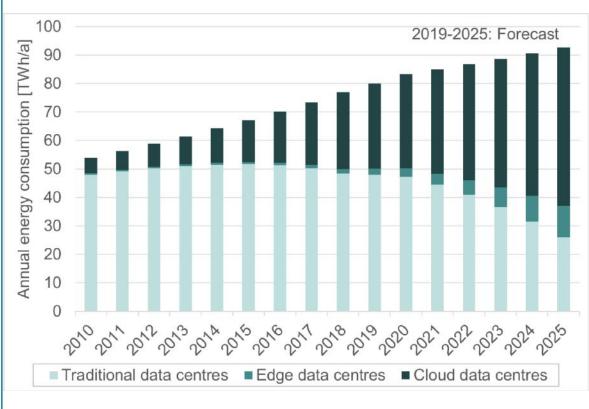


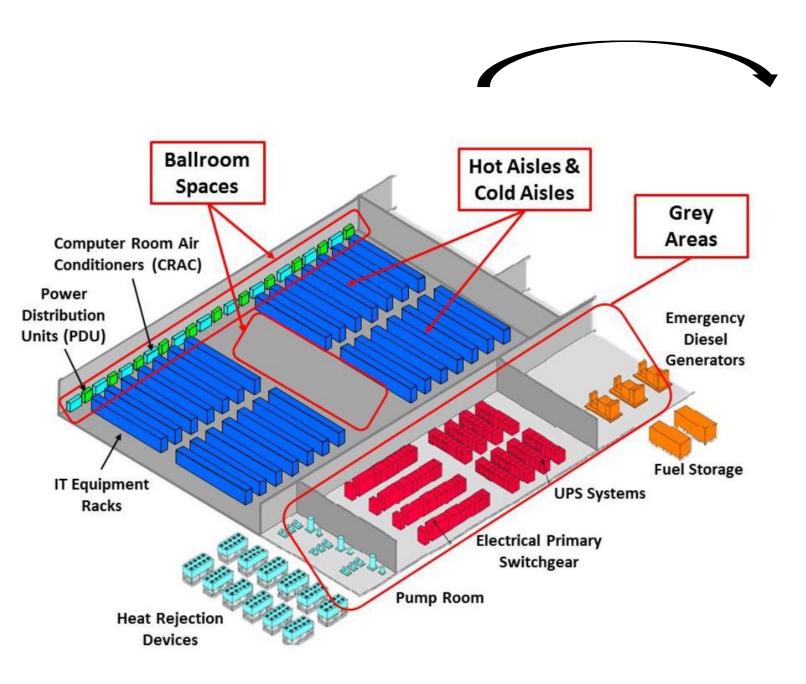
Uptime Institute



Score value	Volume / day / site	Criticality (availability of data or site)	Key for volume/day/site
6	100 TB+ (10 Gbps+)	0 downtime/extreme business risks	6
5	10-100 TB (1-10 Gbps)	few seconds/ high risks	5
4	1TB-10 TB (0.1-1Gbps)	few minutes / moderate risks	4
3	100-1,000 GB (10-100 Mbps)	few hours / low risks	3
2	10-100 GB (1-10 Mbps)	few days / little risk	1
1	10 GB (1 Mbps)	irrelevant / no risk	
time Institute Intelligence, 2021			Uptime Institute INTELLIGENCE

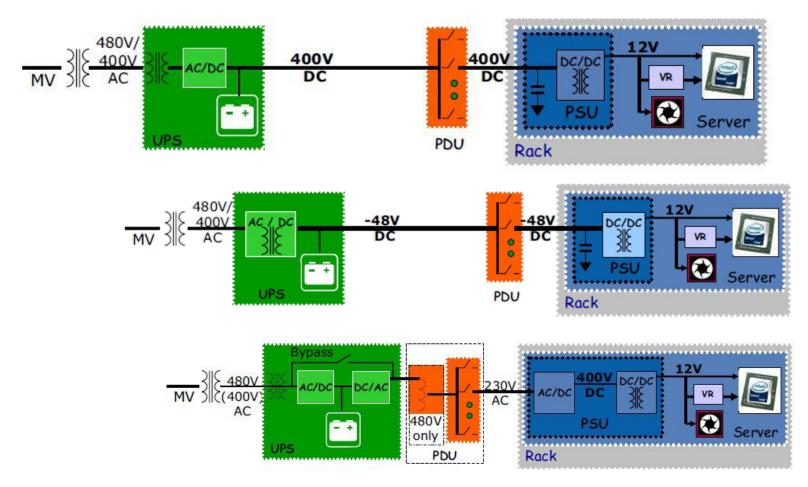
for volume/day/site



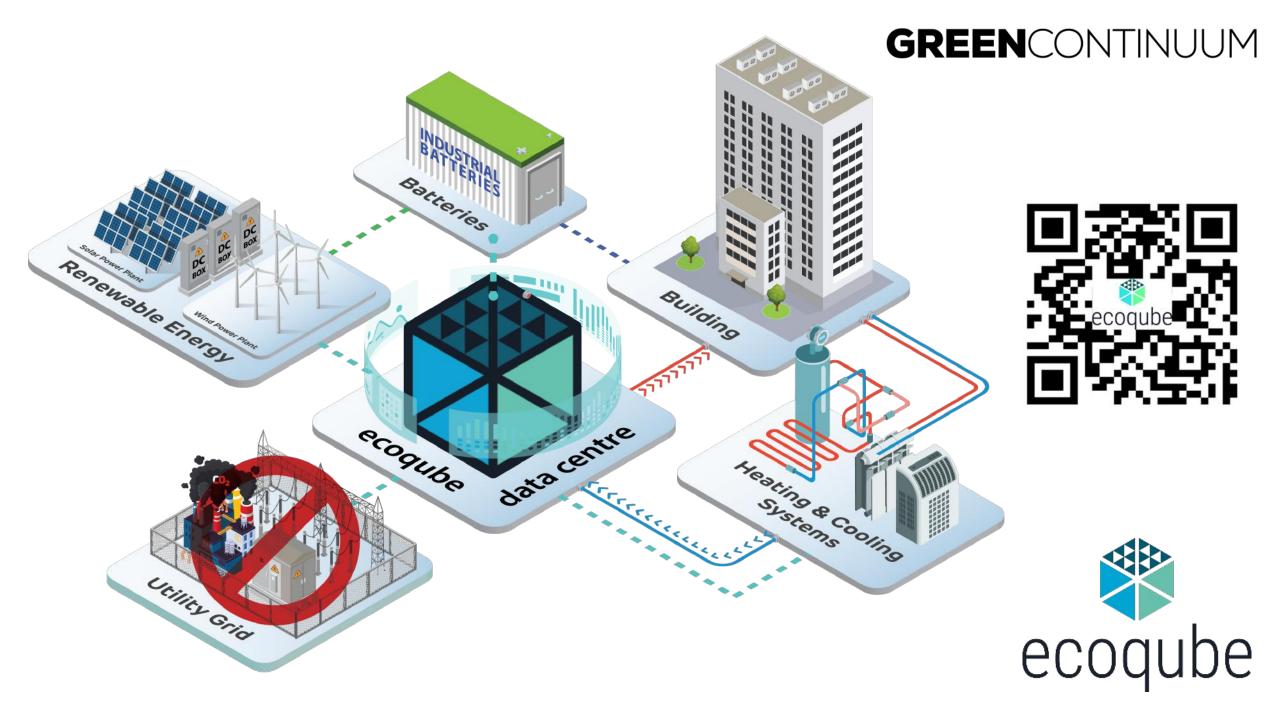


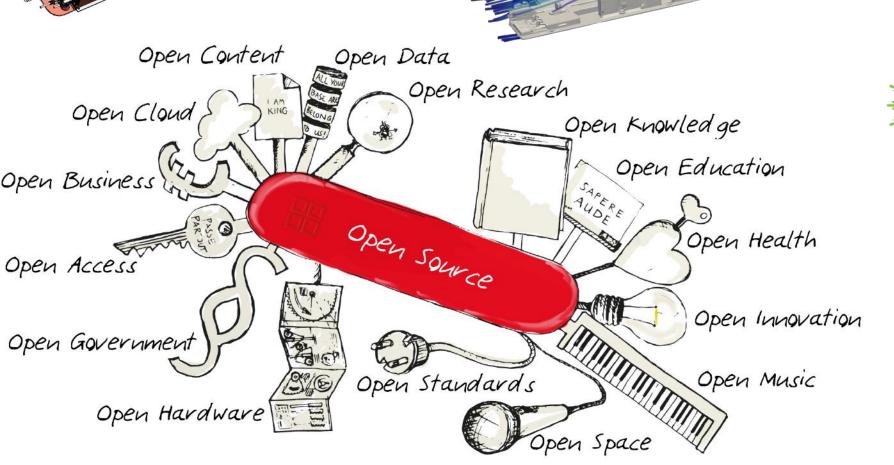
















Proposal by Bogazici University Team: Distributed PAN-European Cloud+Edge Platform

- Aim: Design an open interface cloud+edge framework to enable:
 - Flourishment of cloud & edge provider companies at different scales in all European countries
 - Competitive and efficient, yet constructive, business environment for cloud/edge/IoT/mobile services through data and service brokerage
 - Transnational service mobility across Europe
 - Europe-based data analytics and AI/ML services provided to businesses and services through standardized APIs

 Transparency and accountability to ensure user anonymity and data protection provided by data analytics companies

Sample Use Cases:

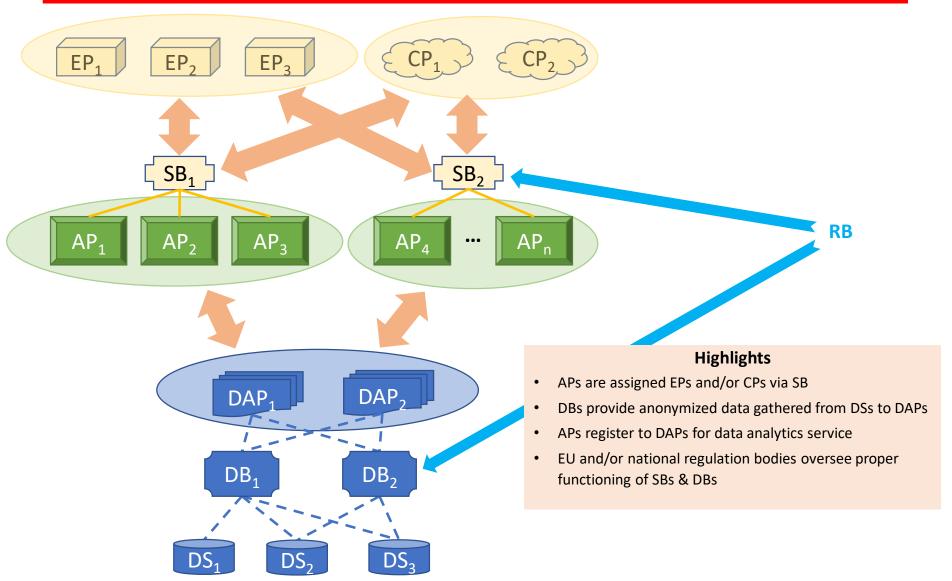
- 1) X-Company organizes a synchronized event in several e-sports arenas across Europe over edge servers of individual countries unified as a single e-sports challenge.
- 2) Y-Company wants to migrate to a different cloud server in another country for a better offer.
- 3) Z-Company offers courses and deploys exams over cloud servers in some countries and edge servers in others.



Stakeholders in the Proposed Architecture

- End User (EU): Subscriber of services (human, IoT device, etc.)
- Application Provider (AP): Company (at any scale) that provides applications/services to EUs
- Edge Provider (EP): Company (at any scale) that provides compute services to APs (via SBs) in a narrow zone (town or city)
- Cloud Provider (CP): Company (typically at a larger scale) that provides compute
 and storage services to APs (via SBs) in a region, country, or whole Europe
- Data Analytics Provider (DAP): Company that provides data analytics services to APs by pre-processing anonymized data from DBs (likely to be specific to a market such as mobile operators or online retailers)
- Service Broker (SB): Company or government institution that serves as a matchmaker for the requests of APs and available capacities of EPs/CPs
- Data Broker (DB): Company that collects data from various data sources, performs AI/ML-based algorithms, and provides anonymized analytics to DAPs
- Data Source (DS): Any company/institution that provides data to APs via DBs
- Regulation Body (RB): Body authorized by EU Commission or the local governments to oversee proper and lawful functioning of SBs & DBs

Conceptual System Architecture



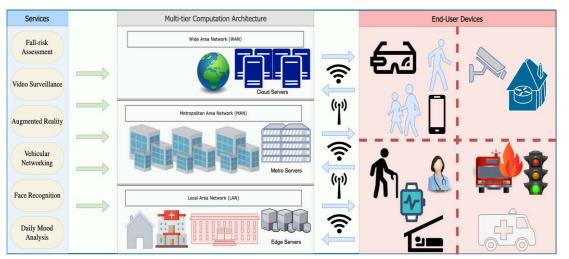
^{*} The links in the figure represent conceptual relations, not physical links

Related Previous Work by Bogazici Team

- Team members: Atay Ozgovde, Cem Ersoy, Birkan Yilmaz,
 Ali Emre Pusane, Tuna Tugcu
 - E-mails: {ozgovde, ersoy, birkan.yilmaz, ali.pusane, tugcu}@boun.edu.tr
 - Bogazici University, Computer Networks Research Laboratory

Multiaccess Multi-tier Edge-Cloud [1,2]:

- Optimal joint task offloading and resource planning
- Optimal multi-tier edge/RSU design for intelligent transportation
- Optimal slicing for service oriented multi-tier edge computing



EdgeCloudSimSimulator [3]:

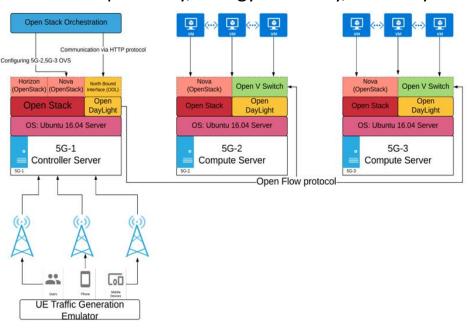
- Developer of popular EdgeCloudSim simulator for Edge Computing
- Incorporates 3G/4G/5G
- https://github.com/CagataySonmez/EdgeCloudSim

- [1] C. Baktir, B. Ahat, A. Ozgovde, N. Aras, C. Ersoy, "SLA-Aware Optimal Resource Allocation for Service-Oriented Networks" 2019
- [2] C. Sonmez, A. Ozgovde, C. Ersoy "Fuzzy Workload Orchestration for Edge Computing" 2019
- [3] C. Sonmez, A. Ozgovde, C. Ersoy "EdgeCloudSim: An Environment for Performance Evaluation of Edge Computing Systems" 2018

Related Previous Work by Bogazici Team

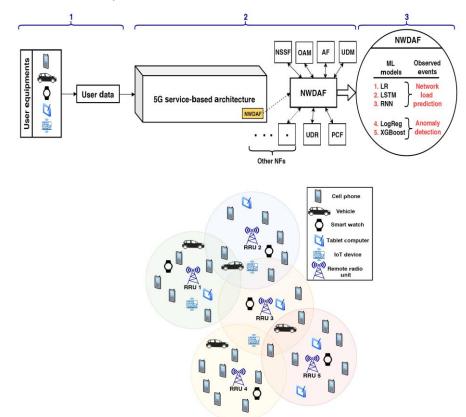
C-RAN (CloudRAN) Testbed [4]:

- Virtual BBUs as VMs
- Live migration
- Low latency, scalability, low OPEX/CAPEX, interoperability, energy efficiency, flexibility



NWDAF w/ Machine Learning [5]:

- NWDAF gathers network data and provides data analytics for 5G NFs
- Designed according to 3GPP specs
- Several machine learning techniques implemented

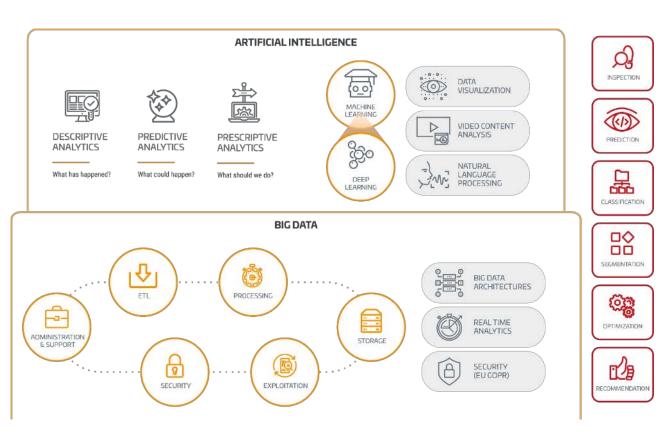


TREE TECHNOLOGY

https://www.treetk.com/en



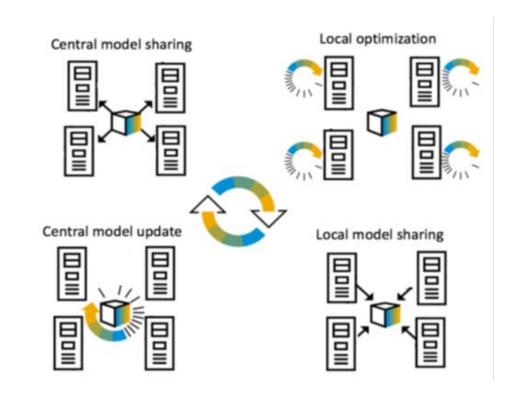
- Expertise in Big Data and Artificial Intelligence
- Spanish SME
- Large experience in EU projects (17 ongoing H2020 projects)
- Role:
 - Coordinator?
 - WP leader
- Contact:
 - Tatiana Silva (<u>tatiana.silva@treetk.com</u>)



Federated Learning

Federated Learning

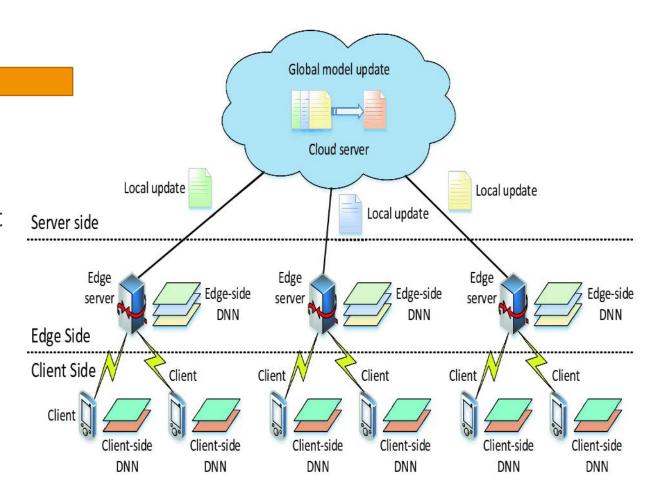
- Different devices collaborative learn a shared predictive model.
- Data never leave the devices
- Decouples machine learning from the need to store the data in a single place.
- Brings the training to the devices



Federated Learning and Edge Computing

Problems in edge computing

- A hierarchical approach can be used.
- In every use case, the cloud, edge servers and IoT devices have completely different computational power, memory or communication capabilities.
- Many decisions may produce bottlenecks in a federated architecture and incurs in computational cost not affordable by the architecture.



Federated Learning and Edge Computing

Improvements

- Having three different levels (devices, edge servers and cloud servers) we propose a framework to adapt intelligently where data must be allocated, the predictive model must be trained and the global model updated.
- This framework will respond and adapt intelligently to changes in application behavior such as DNN size or data workload by applying AI techniques to automatically adapt connectivity, computing and storage in cloud and edge to optimize where the ML training should be performed.
- Trustworthiness will be provided by including defenses to avoid data poisoning attacks.



Federated Learning and Edge Computing

Looking for

- Potential coordinator
- Experts in edge computing
- Case studies to test and validate the approach



Huawei, a leading global provider of ICT infrastructure and smart devices



Bring digital to every person, home and organization for a fully connected, intelligent world



197,000

Employees



49 on Fortune Global 500



105,000+

R&D employees



2TOP EPO
Applicant



170+
Countries and regions



Rank 5th

In Global R&D Investment

Huawei's research in Europe at a glance

370+) Research projects

230+) Academic institutions are partners

Participation in H2020 projects.

Research in Europe

- 1. 2400 employees, 84% recruited locally
- 2. 27 R&D centres in EU, Russia, Ukraine, Belarus and Israel, with ERI HQ based in Leuven/BE
- 3. Munich Research Center is the largest European Huawei Research Center, established in 2008.



Research Activities (a few examples)

5G & Wireless Optical Communications Data Communications Photonics Cyber security & Privacy Quantum Technologies Semiconductors Thermal Technologies AI/ML/CV **RFIC** NFV/SDN **Cloud/Big Data Aesthetics** User experience

Focus on innovation, continuous dedication and backward compatibility

Atlas 300

Al Accelerator Card

- 64 TOPS of INT8 @ 67 W
- 32 GB memory
- 64-channel HD video real-time analytics
- Standard half-height half-length PCIe card form factor. applicable to general-purpose servers





reddot award 2019

Al Accelerator Module 16 TOPS of INT8

Atlas 200

52 mm x 38 mm x 10 mm

- 16-channel HD video real-time analytics, JPEG decoding
- 4 GB/8 GB memory, PCIe 3.0 x4 interface
- Operating temperature: -25°C to +80°C

Ascend 310

Al SoC with ultimate efficiency

- Half precision (FP16): 8 TFLOPS
- Integer precision (INT8): 16 TOPS
- 16-channel full-HD video decoder: H.264/265
- 1-channel full-HD video encoder: H.264/265
- Max. power consumption: 8 W
- 12nm

Atlas 200 DK

Quickly build development environments in 30 minutes

- 16 TOPS of INT8 @ 24 W
- 1 USB type-C, 2 camera interfaces, 1 GE port, 1 SD card slot
- 4 GB/8 GB memory

Storage-intensive

Atlas 500

Al Edge Stations

- 16 TOPS of INT8
- 25-40 W
- Wi-Fi & LTE
- 16-channel HD video real-time analytics
- Fanless design, -40°C to +70°C environments

Ascend 910





reddot award 2019

Atlas 800

Deep Learning System



- Plug-and-play installation
- Ultimate Performance
- Integrated Management



Atlas 900 Al Cluster

The pinnacle of computing power

- Thousands of Ascend 910 AI processors
- High-speed interconnection
- Delivers up to 256 to 1024 PetaFLOPS at FP16
- · Can complete model training based on ResNet-50 within 59.8 seconds
- 15% faster than the second-ranking product
- Faster AI model training with images and speech

Kunpeng 920

The industry's highest-performance ARM-based server CPU

Highest compute density on a single chip

Half precision (FP16): 256 TFLOPS

Integer precision (INT8): 512 TOPS

Max. power consumption: 350 W

128-channel full-HD video decoder: H.264/265

ARM v8.2-architecture

7nm

- up to 64 cores, 2.6 GHz
- 8 DDR4 memory channels
- PCIe 4.0 and CCIX
- Integrated 100GE LOM and encryption and compression engines
- Supports 2- or 4-socket interconnects

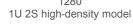
Computing-intensive



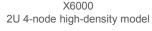


5280 4U 40-drive storage model























Huawei pilot cluster

Hosted at Goethe University of Frankfurt (FIAS)
Supported by Open Edge and HPC Initiative (OEHI)

Cluster Configuration

28	Standard Compute Node - TaiShan 200 (Model 2280)		
1	Development Compute Node - TaiShan 200 (Model 2280)		
2	IO Node - TaiShan 200 (Model 5280)		
1	Al Training Node - Atlas 800 (Model 9000)		
1	Al Interference Node - Atlas 800 (Model 3000)		

Connected by 6 Mellanox Infiniband – Each switch provides 36 x 100 Gb/s InfiniBand EDR ports.

Future addition

- Edge intelligence
- OceanStor Pacific storage
- Cloud capabilities





Taishan 200 server

- support of two Kunpeng 920 processors as well as DDR4-DIMM-Slots
- 2 x Kunpeng 920 processor (ARMv8 64bit):
 - o 64 core; 2.60 GHz base frequency; 180 Watt TDP;
 - o 2933 MHz max. Memory speed incl. heatsinks
- 128 GB main memory

IO nodes

2 x 960 GB SSD local storage + 36 x 1.2 TB HDD data storage

Al Training Node

8 x Huawei Ascend 910 with 32 Al cores 1024 GB main memory + 2 x 960 GB SSD + 4 x 3,2 TB SSD NVMe

Al Interference Node

5 x Atlas 300 Al Inference Card; 32 GB; PCle3.0 x16

Huawei cluster pilot to streamline HPC software stack

European **HPC-AI** applications



Cluster Management / **Scheduler**













Middleware / Libraries















Filesystem & OS















Storage

HW Infrastructure





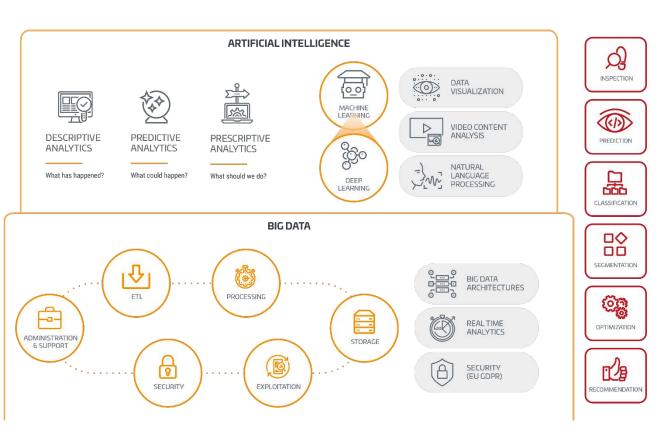




TREE TECHNOLOGY



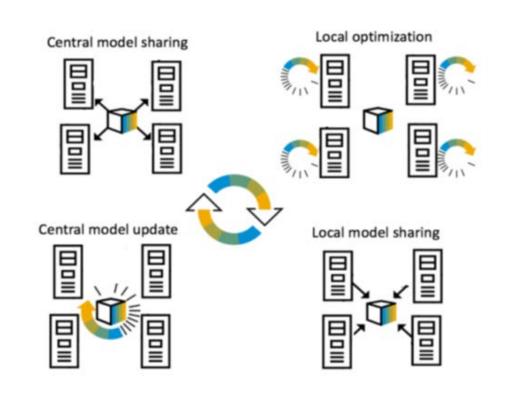
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- Spanish SME
- Large experience in EU projects (17 ongoing H2020 projects)
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Capabilities: Federated Learning

Federated Learning

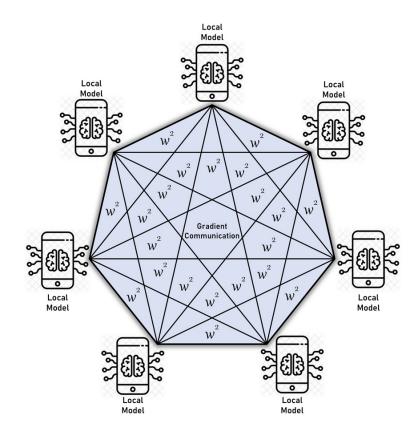
- Different devices collaborative learn a shared predictive model.
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- Brings the training to the devices



Challenges: Decentralized Federated Learning

Challenges

- In decentralized federated learning, the nodes should be able to coordinate themselves to obtain the global model.
- The specific network and device topology highly affect the performance of the learning process.
- The design and development of this kind of training architectures for real environments is extremely hard.
- In addition, the security threads increase due the lack of a central node to control the training.
- It is one of the open problems in the federated learning field.



Challenges: Decentralized Federated Learning

IDEA

 Creation of a new integrated environment for developing decentralized federated learning solutions



Looking for

- Potential Coordinator
- A consortium to join
- Experts in programming environments
- Case studies to test and validate the approach



From Internet of Things Platforms to Web of Things User Agents

Prof. Dr. Andreas Harth

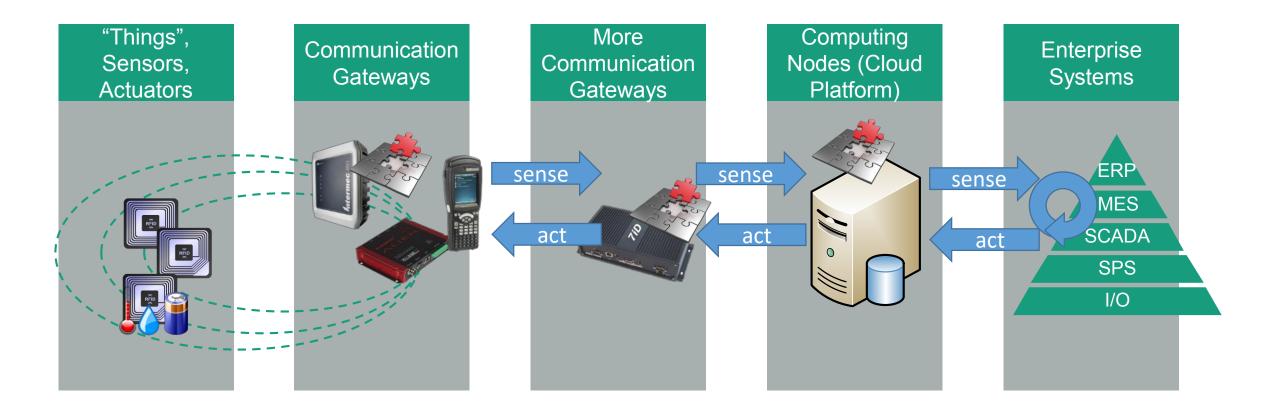
Department Data Spaces and IoT Solutions, Fraunhofer IIS Nuremberg

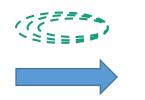
Chair of Technical Information Systems, Friedrich-Alexander University Erlangen-Nuremberg

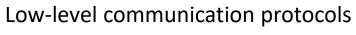




Traditional System Architecture







Sense-act cycle for control

Communication via TCP/IP ("narrow waist")

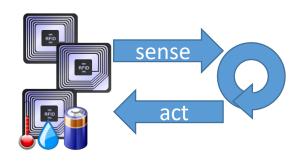
Selected Research Challenges

- Interoperability: different protocols to access to "things"
 - Standardise data representation and communication protocols
 - Use formal languages for data to enable data portability, data integration and querying
 - Use formal languages to describe network-accessible operations
- Complexity: writing programs that access "things" is hard
 - Provide higher-level programming abstractions
 - Use formal languages to support programming, enable automated composition of processes and program synthesis
- Communication: latency for network connectivity to the cloud
 - Bring computation closer to the devices ("edge computing")
 - Use formal languages for programs to enable flexible execution at different nodes
- Safety: systems with actuators might harm equipment or people
 - Use formal languages to reason about programs and verify their properties
- Resilience: intermittent communication links, reliance on business decisions and risk of censorship
 - Implement decentralised architectures that reduce or remove centralised components
 - Use formal languages to be able to split up monolithic systems into pieces under diverse ownership



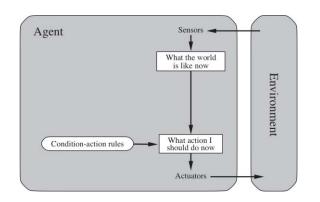
From Representing Knowledge to Representing Behaviour

- We assume a uniform interface to read data from sensors and manipulate actuators
- Access data from sensors
 - Linked Data: GET (read)
 - WoT: read Property, receive Events
- Effect change in actuators
 - Linked Data: PUT (write)
 - WoT: write Property, invoke Actions
- How to represent behaviour?



Options for Representing and Executing Behaviour

- Imperative programming languages (e.g., JavaScript)
- Declarative programming approaches
 - (Event-)Condition-Action rules (c.f. Simple Reflex Agents)
 - Process languages, workflow languages
- Declarative programming approaches allow for
 - programming at a higher level of abstraction
 - automated composition of processes and program synthesis
 - flexible execution at different computing nodes
 - reasoning about programs and verifying their properties
 - splitting up monolithic systems under central control into pieces under diverse ownership







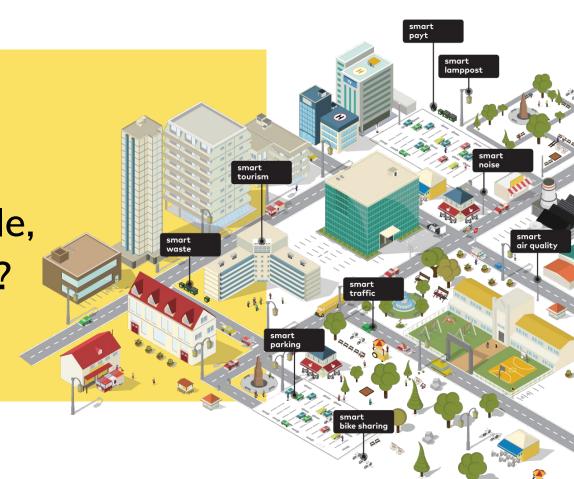
From
Portugal
to the world
with love.





2 —

Can you imagine your city as a single, integrated system?



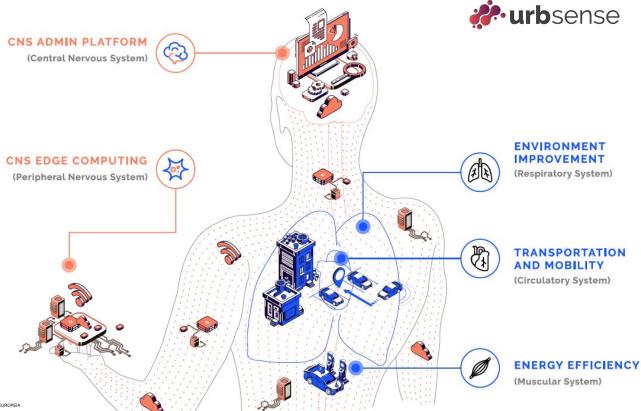
The City Nervous System

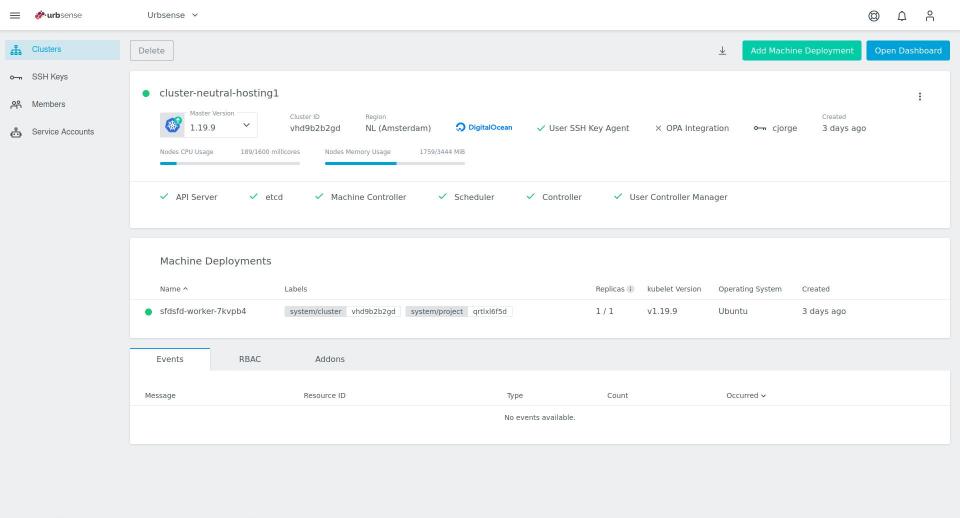
An organic approach to Smart City Management

→ urbsense.com









Suiting the future of cities and regions.

- + blueprint and reference architecture for smart cities
- + real-time data collection and analysis and digital tools for the community
- + trusted execution environment for restrained devices
- + life-cycle assessment, verification and validation

We're here for the long run.



Information & Brokerage Session Horizon Europe Cloud-Edge-IoT Call 2022

12:0012:30

WRAP-UP AND CLOSING SESSION



Rolf Riemenschneider Head of Sector IoT European Commission



Maria Tsakali
Programme Officer, Cloud and
Software Unit, European Commission



Luis Busquets PérezProgramme Officer, Cloud and
Software Unit, European
Commission





Monique Calisti
CEO
Martel Innovate



Jan Komarek
Topic Coordinator, IoT Unit,
European Commission

