

# *Edge AI 2.0* *for Future Computing*

**Aaron Ding**

Director of CPI, Senior Associate Professor (UHD1 Ius Promovendi)

# *Why + Lessons + Directions*

***Why*** + *Lessons* + *Directions*

# Challenges



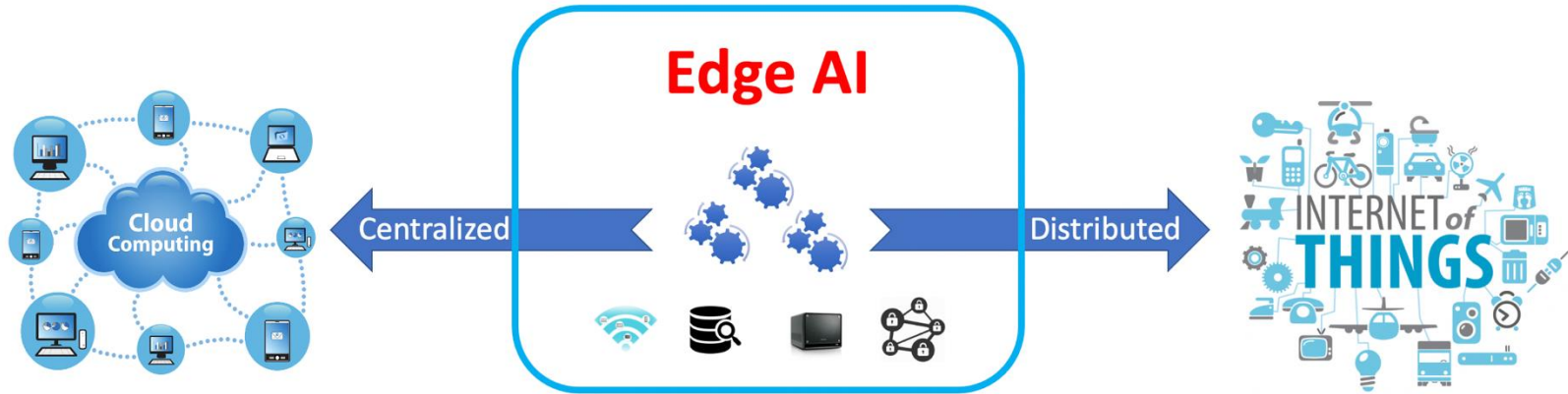
latency, reliability, security,  
data **privacy** and trust ...

You don't want your  
device to confirm with  
the cloud to take action



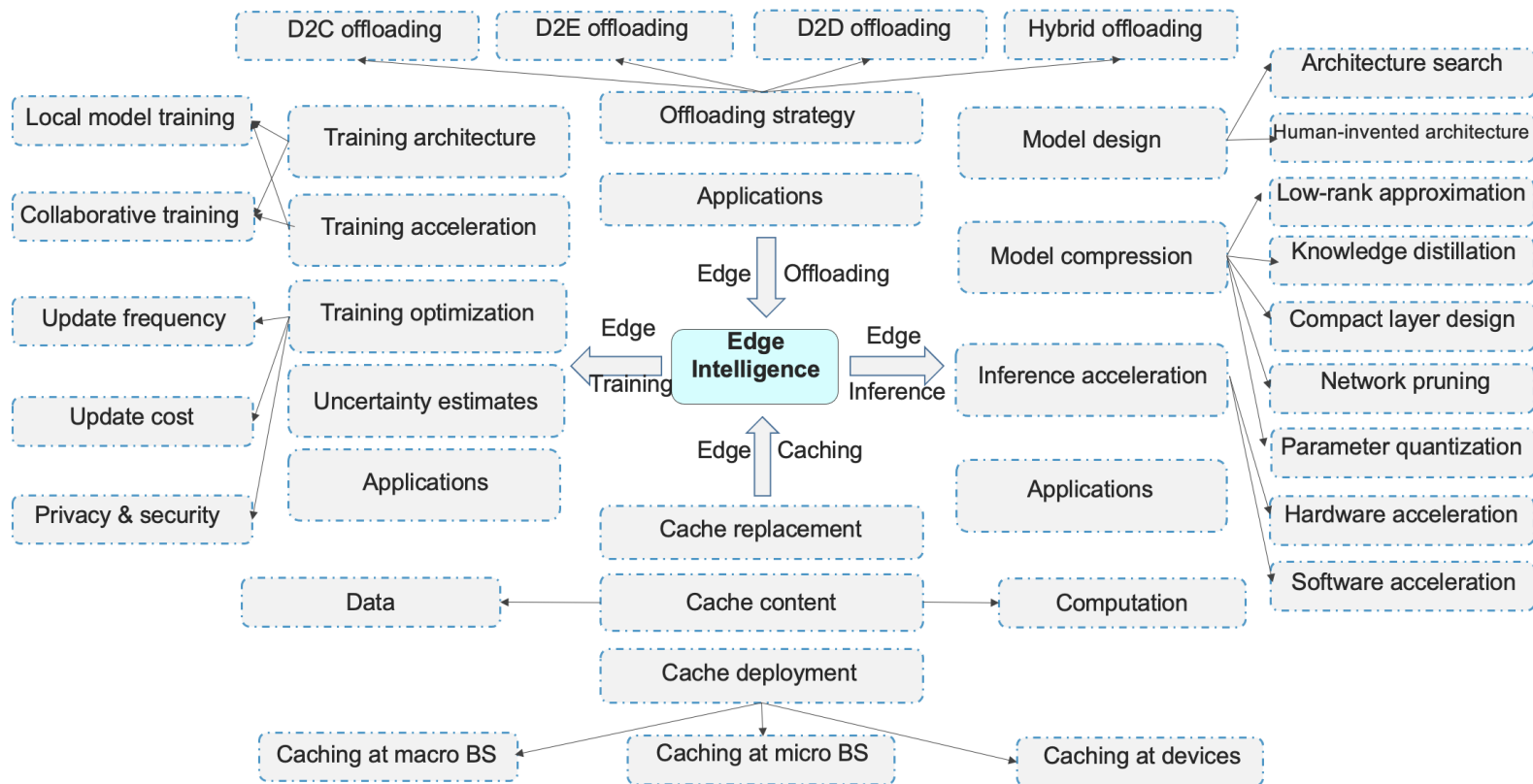
# Motivations

Bridge the Gap



**Consolidate Cloud & IoT**

# Enabling Techniques



*But **Where** Are They ?*

# Holistic View



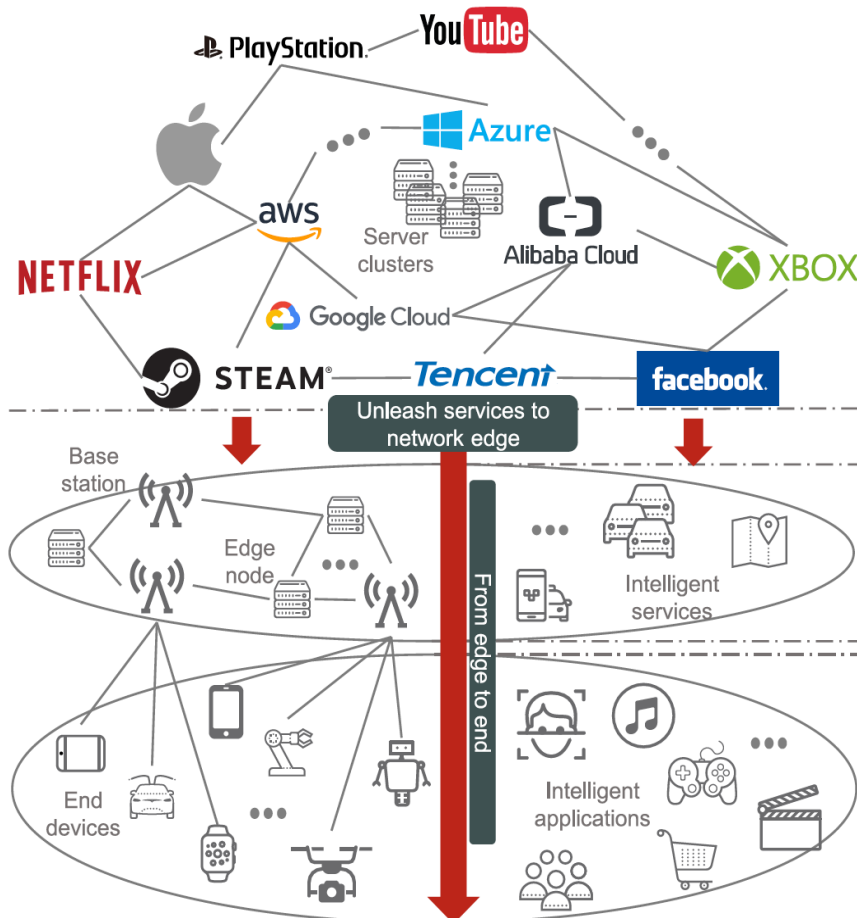


# Cloud View

Cloud Layer

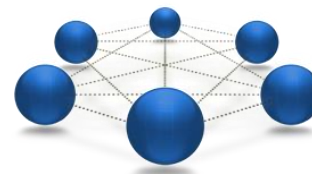
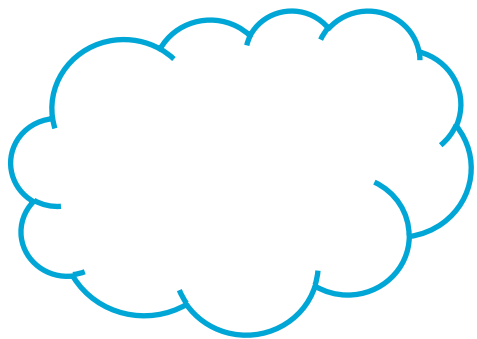
Edge Layer

IoT & End Devices



***What exactly is it ?***

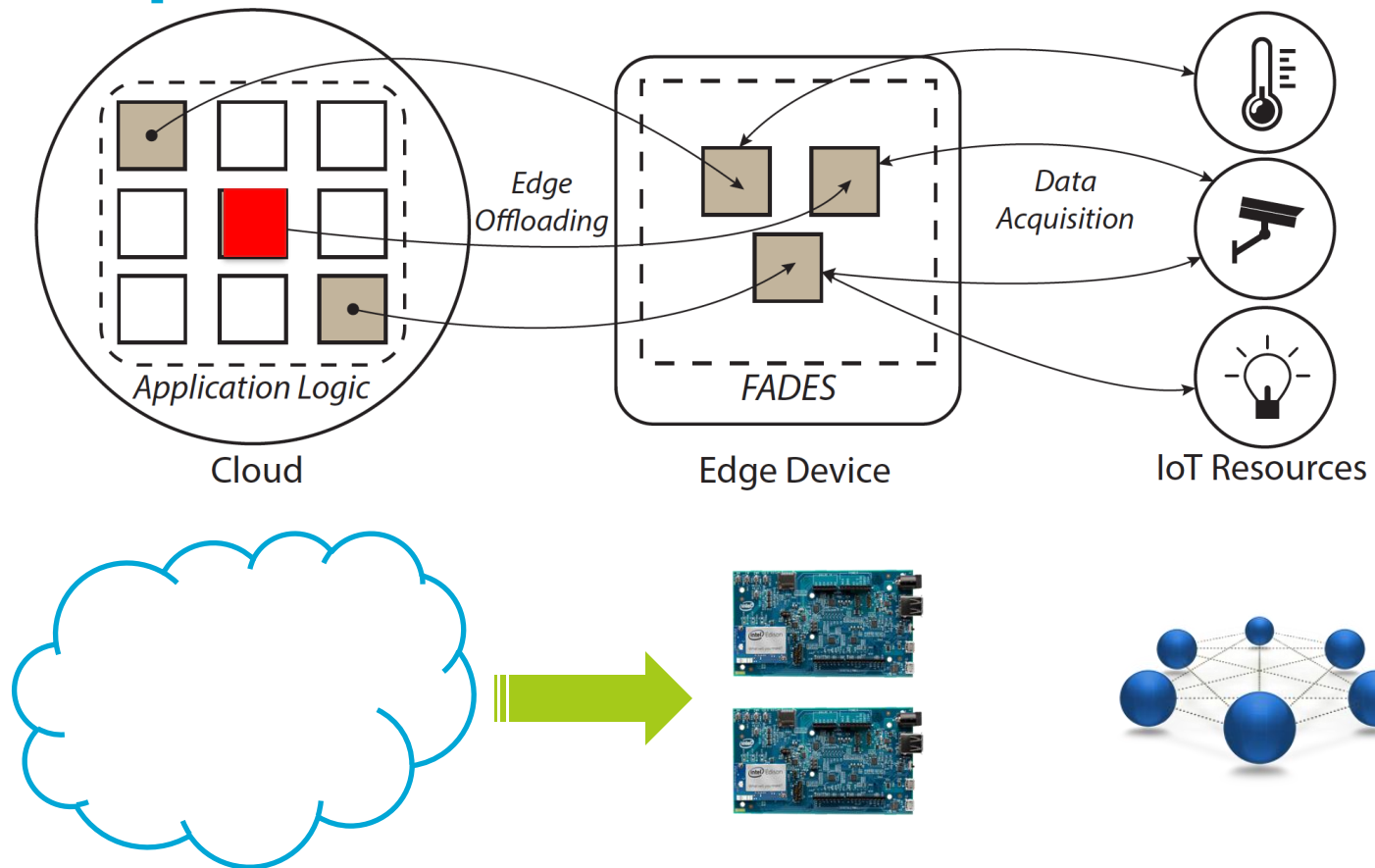
# Edge Paradigm



Edge

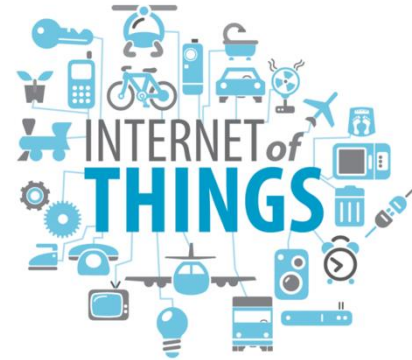
IoT

# Example



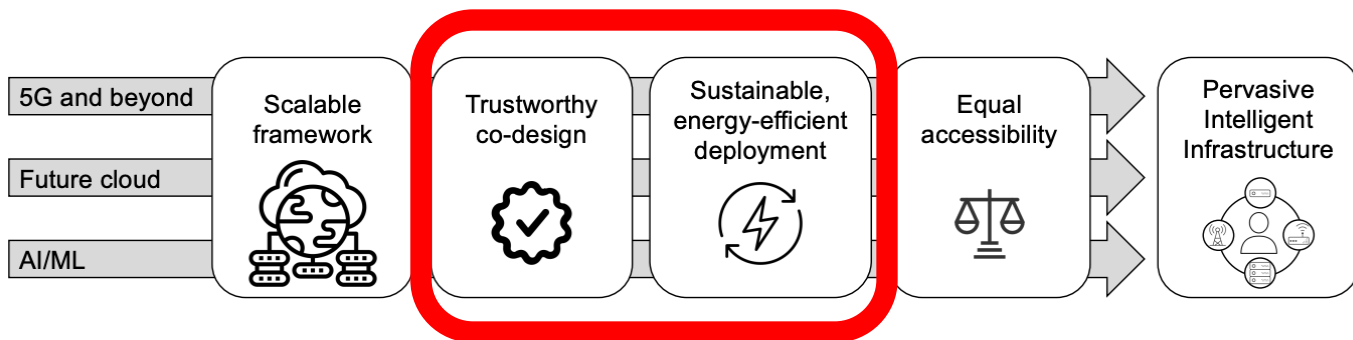
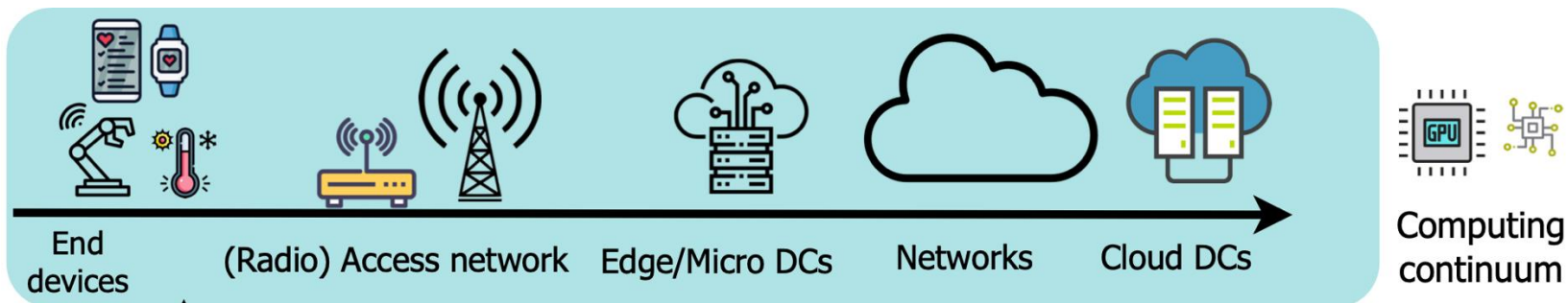
# Why Bringing Intelligence to Edge

**Problem Solved !?**



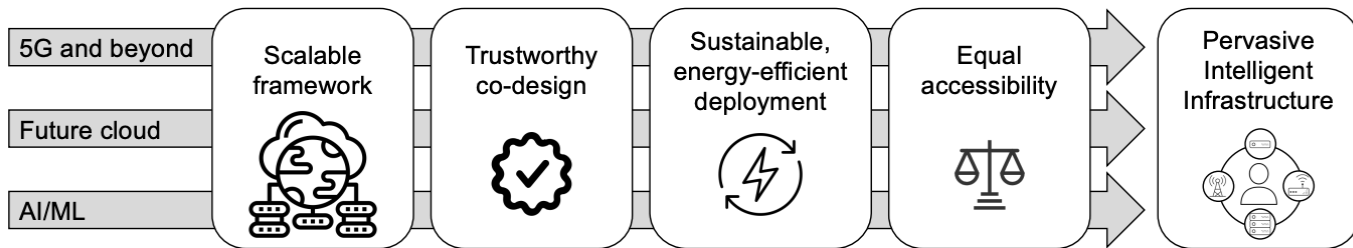
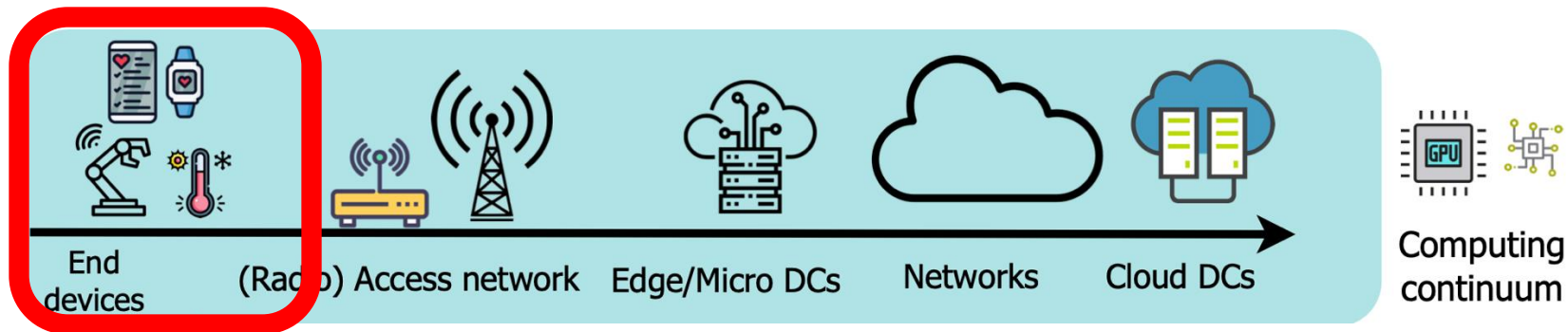
**Are You Sure ...?**

# Future Computing – Edge AI 2.0



*Why + **Lessons** + Directions*

# Tiny but not Wasted: Circular Economy?





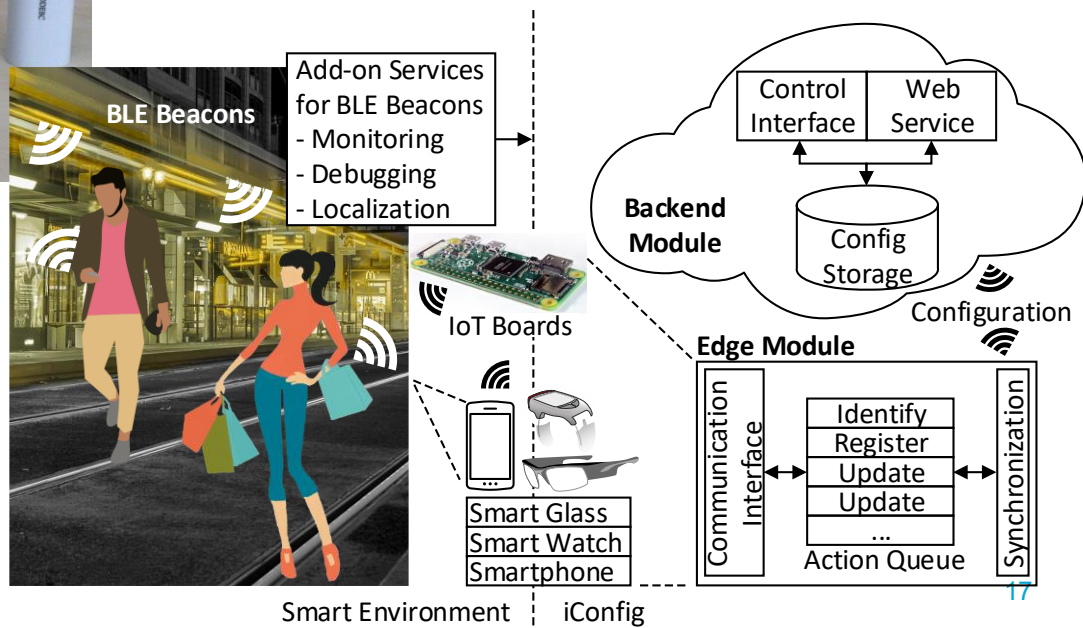
# Case: Tiny IoT



what I see is what i **Configure**

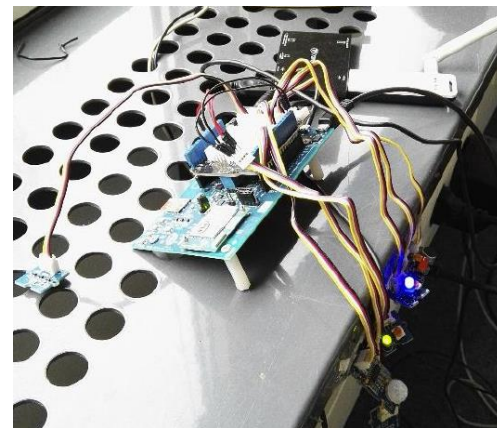
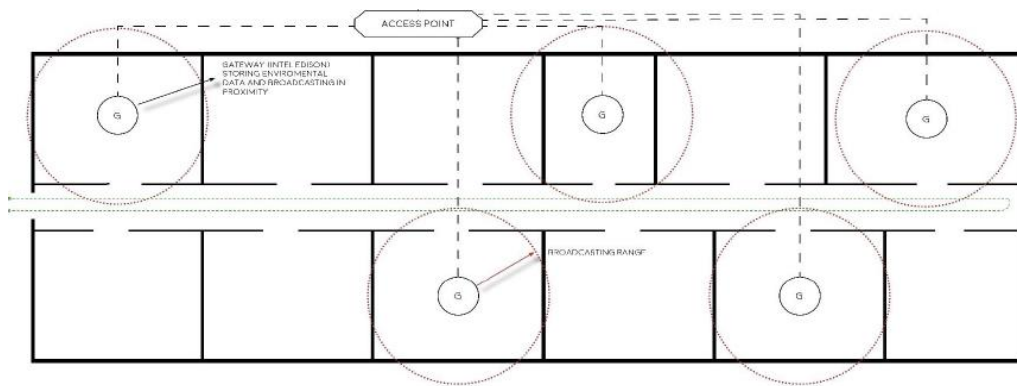
*“ I don’t get what’s this really useful for ... ”*

*“ Why those BLEs are ever needed ? Why configure them ? ”*



# Starting on Campus

- Live testbed on campus
  - Office room context monitoring (IoT boards)
  - Proximity reasoning (BLE beacons)



# How to Configure ?

- Complications started
  - No backend, no fixed connectivity
  - Poor UI
  - Purely manual
  - Error prone
  - Time consuming
- Scalable ?
  - 10-20, perhaps doable
  - 100+, will drive people mad

The screenshot shows a mobile application interface for configuring a beacon. At the top, there's a status bar with icons for play, home, and back, and a clock showing 15:41. Below the status bar, the title "Beacon #6" is displayed on the left, and a "CONNECT" button is on the right. The main content area is divided into two sections. The first section, titled "Beacon General Information", contains fields for NAME (Beacon #6), TYPE (sBeacon, Eddystone TLM, Eddystone URL, Eddystone UID), RSSI (-89 dBm), STATUS (Disconnected), and PASSWORD. The second section, titled "SBeacon", contains fields for ID (17224680162821758480/EF0A5B590AB55E10), TEMPERATURE (26.0 °C), BATTERY VOLTAGE (3.63 V), and FIRMWARE REVISION. The bottom of the screen shows a standard Android navigation bar with back, home, and recent apps buttons.

Beacon #6	
CONNECT	
<b>Beacon General Information</b>	
NAME	Beacon #6
TYPE	sBeacon, Eddystone TLM, Eddystone URL, Eddystone UID
RSSI	-89 dBm
STATUS	Disconnected
PASSWORD	
<b>SBeacon</b>	
ID	17224680162821758480/EF0A5B590AB55E10
TEMPERATURE	26.0 °C
BATTERY VOLTAGE	3.63 V
FIRMWARE REVISION	

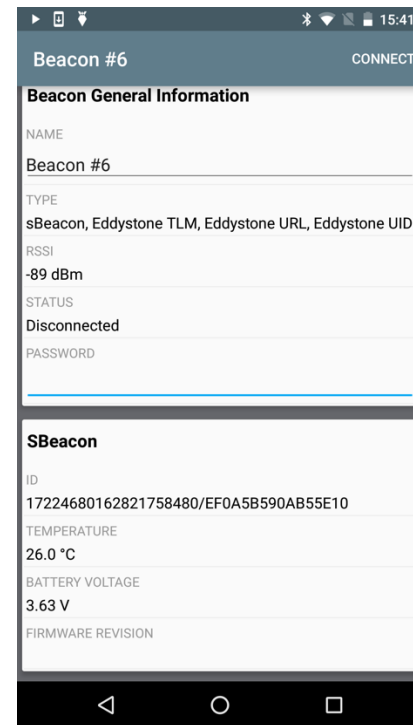
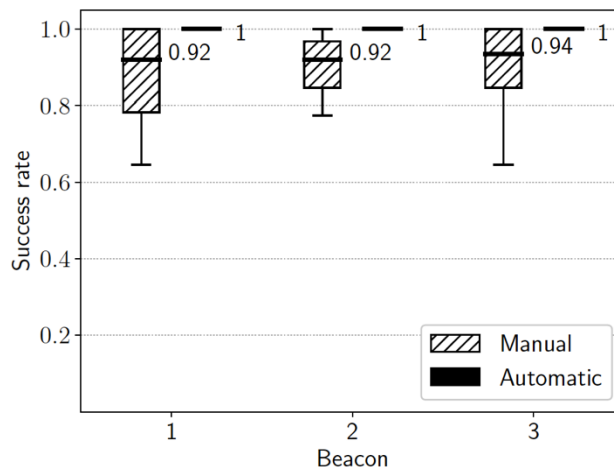
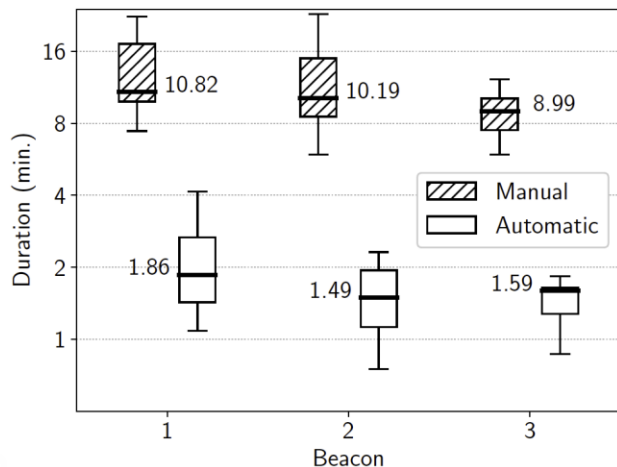
# Where to track ?

- Distributed across buildings
  - Paper, manual logs got lost...
  - Lack of tools, process
- Management **nightmare**

ACTION	UPDATE STATUS	MAC ADDRESS	BEACON ID	NEAREST ROOM	MAINTENANCE		
<a href="#">Edit</a>	up-to-date	C9:ED:20:D2:6A:5F	8EC968106F2	<a href="#">01.05.038</a>	PKT COUNT	VOLTAGE	UPTIME
					35155830	3.6000001	459853408
<a href="#">Edit</a>	up-to-date	EB:26:C9:E4:DC:5B	C28C3BC4AE8	<a href="#">01.05.038</a>	PKT COUNT	VOLTAGE	UPTIME
					24238715	3.6000001	459856770
<a href="#">Edit</a>	up-to-date	D6:82:27:05:E4:0B	2F0446B54A6	<a href="#">01.05.038</a>	PKT COUNT	VOLTAGE	UPTIME
					31240056	3.6000001	459853016

# Usability Study

- Basic smartphone tool
  - Six times longer, manual vs. automatic
  - Lowest rate **58%** + only **1/3** got entirely right



# Challenges behind

- The issues of being tiny
  - Battery only
  - No wire connectivity



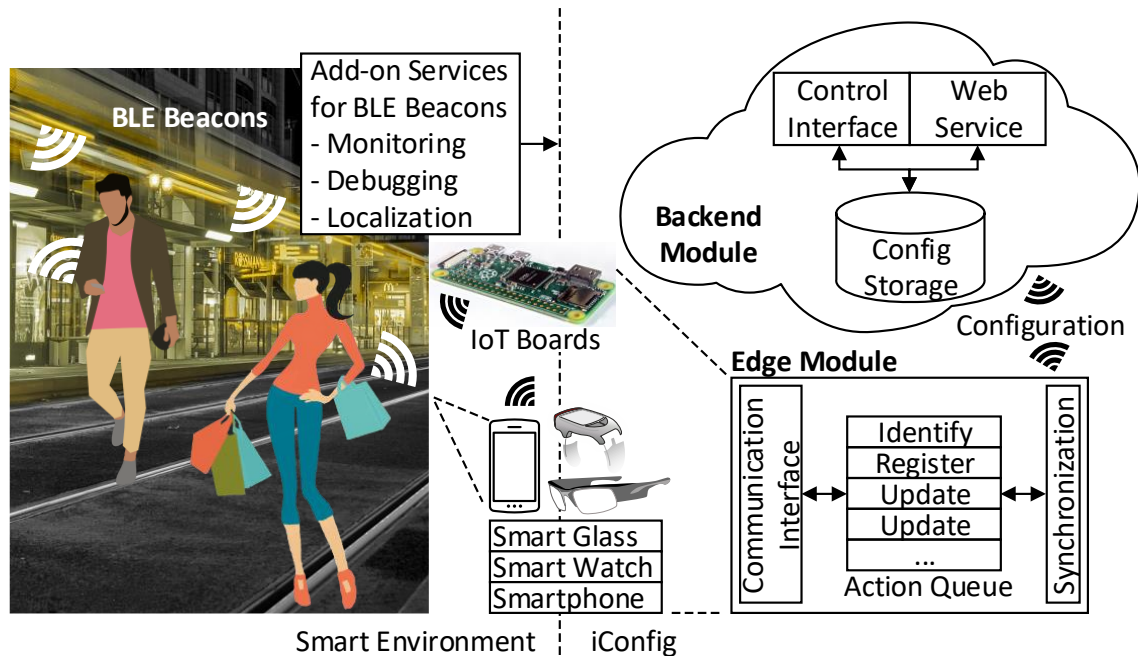
# What's missing ?

- Device life cycles
  - Registration: location, tags
  - Configuration
  - Update
  - Monitoring
  - Debugging
- More automation



# iConfig

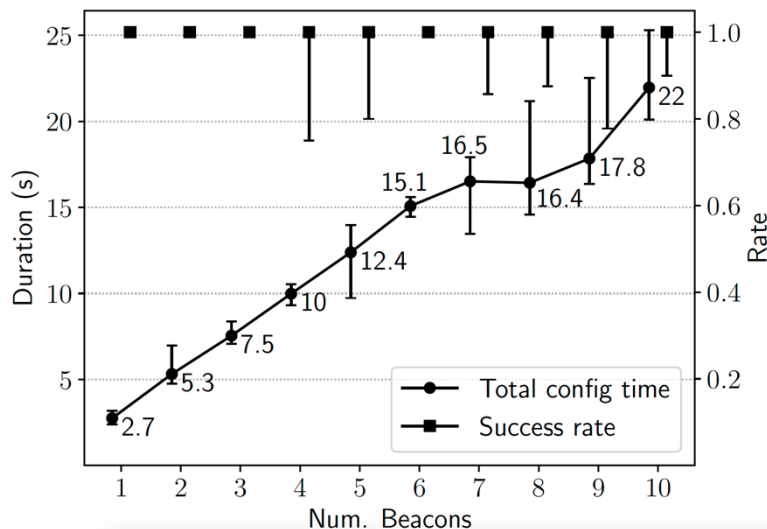
- For configuration automation
  - Portable frontend + central backend





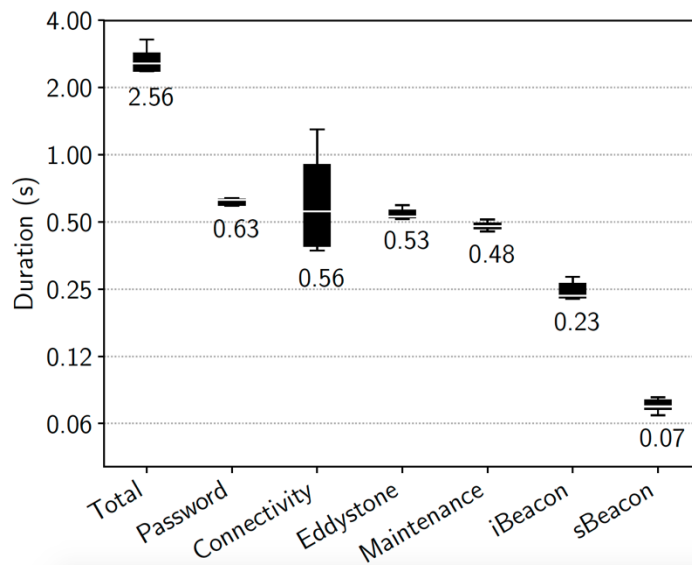
# Does it work ?

- Testbed experiments
  - ~2.2s per beacon
  - Password 25%



But not enough:

- Novelty ?
- Another App ?



# Wait ...

- Something's missing
  - Conventional design
  - Web UI, cloud backend
  - Old-fashion interaction
  - Poor usability



ACTION	UPDATE STATUS	MAC ADDRESS	BEACON ID	NEAREST ROOM	MAINTENANCE		
					PKT COUNT	VOLTAGE	UPTIME
<a href="#">Edit</a>	up-to-date	C9:ED:20:D2:6A:5F	8EC968106F2	<a href="#">01.05.038</a>	35155830	3.6000001	459853408
<a href="#">Edit</a>	up-to-date	EB:26:C9:E4:DC:5B	C28C3BC4AE8	<a href="#">01.05.038</a>	24238715	3.6000001	459856770
<a href="#">Edit</a>	up-to-date	D6:82:27:05:E4:0B	2F0446B54A6	<a href="#">01.05.038</a>	31240056	3.6000001	459853016

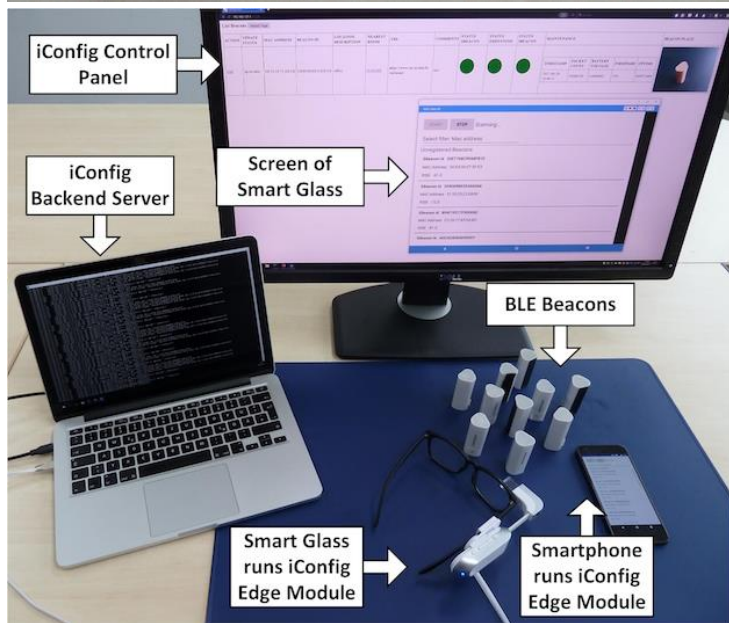


# Next shot

- Voice!
  - No hands

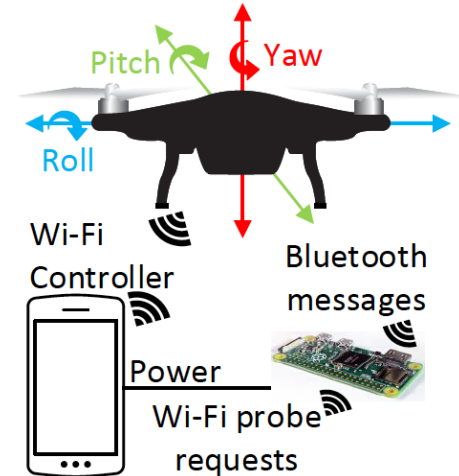
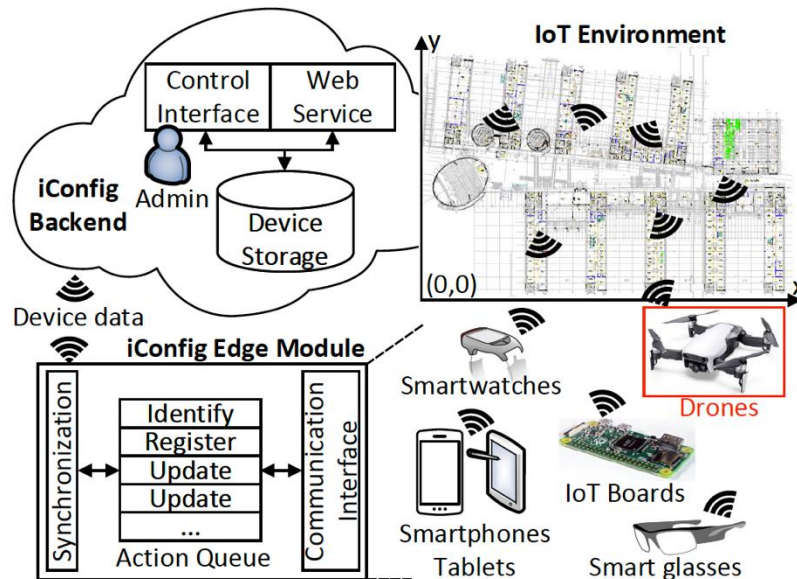
what I see is what i Configure

- Use cases
  - Configuration Automation
  - Debug + Monitor
  - Energy-aware management



# Sky is the limit 😊

- Drone based IoT management
  - ACM SIGCOMM MAGESys



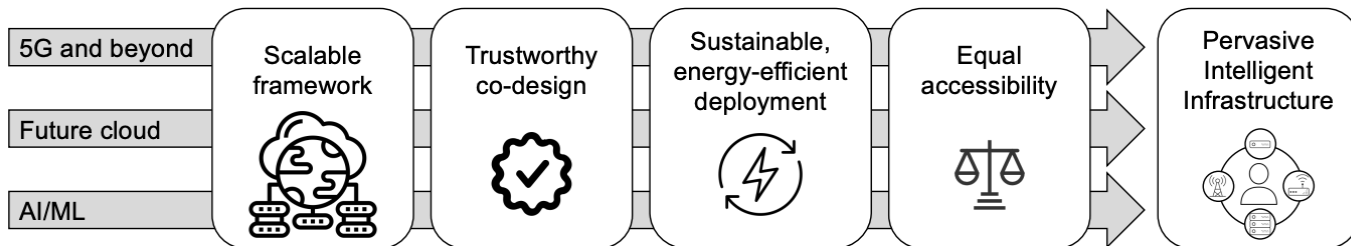
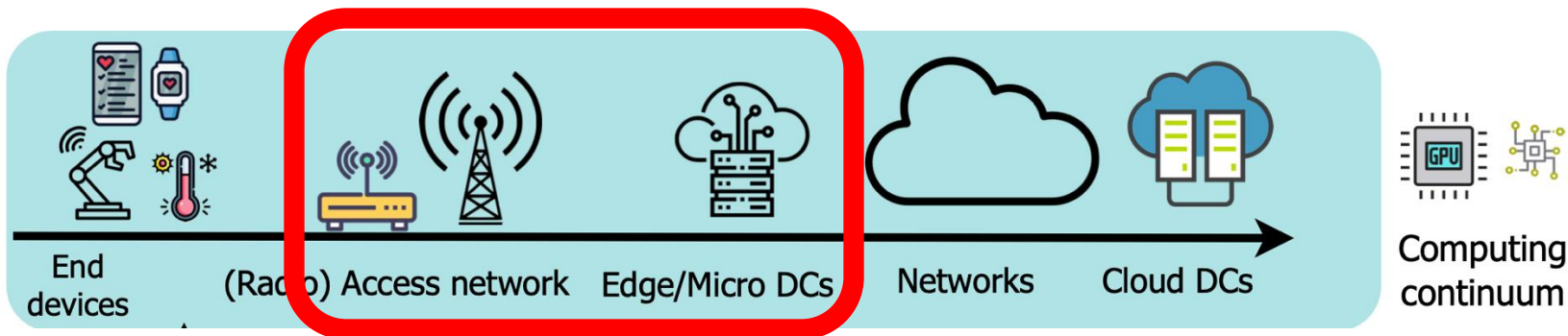
# Lesson #1

**Tiny things hard to **manage****

**Recycle is a MUST for **Sustainability****

**Tools (Edge) are **Missing****

# Bigger but not Static: Programmable ?



# Lesson #2

“ What’s Killer App on the Edge? ”

# Case: Crowd Intelligence on Edge

- **Societal impact** of past years
  - Responding and coping with emergency/pandemics
  - **Urban activity/mobility sensing** on the edge





# Motivation

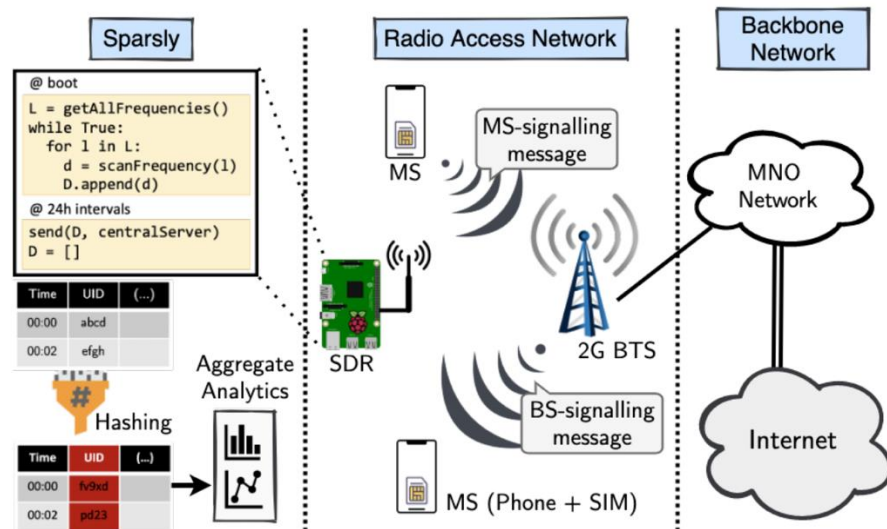
- Low cost and scalable
  - User equipment
  - Deployment and coverage
- Passive (non-intrusive)
  - No need to force user interactions nor mandatory engagement
- Privacy-aware/friendly
  - Balance fidelity and data (local) regulations



**Gap:** high fidelity entails high cost, infrastructure dependency, privacy intervention

# Design

- SDR: Software Defined Radio
- Cellular signal probing
- Paging requests (PR, BS-signaling) and location update requests (LUR, MS-signaling)
- International Mobile Subscriber Identity (IMSI) for sensing purpose
- R-Pi with Noolec NESDR dongle, spectrum (900MHz)



# Unexpected

- Project ends...
  - Regulatory and legal considerations
  - Privacy in local context



# Reusable is Key

- Reflections
  - Privacy on Edge? regulations and legal
  - Difference across countries
  - Programmability has saved us



## Where is my Tag? Unveiling Alternative Uses of Apple FindMy Service

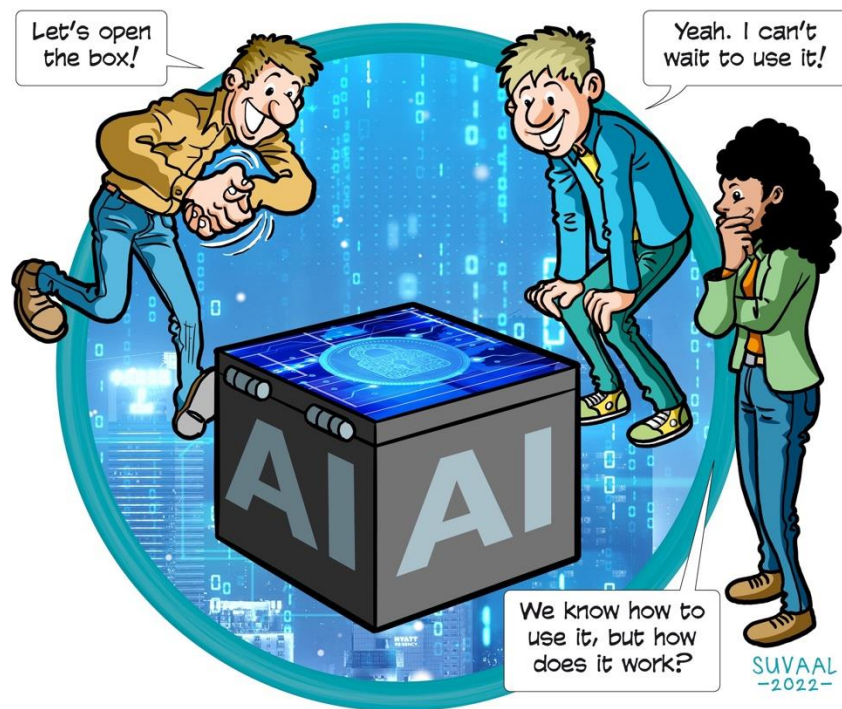
"Learn from the mistakes of others. You can't live long enough to make them all yourself." - Eleanor Roosevelt

# Lesson to Action

Industry + Academia

EU Horizon Project

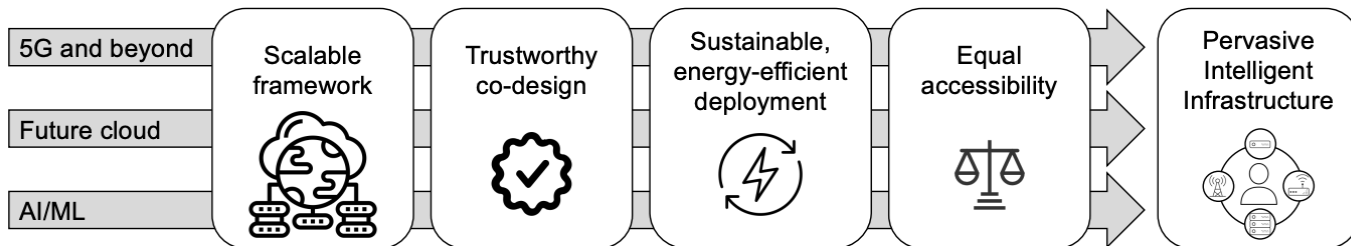
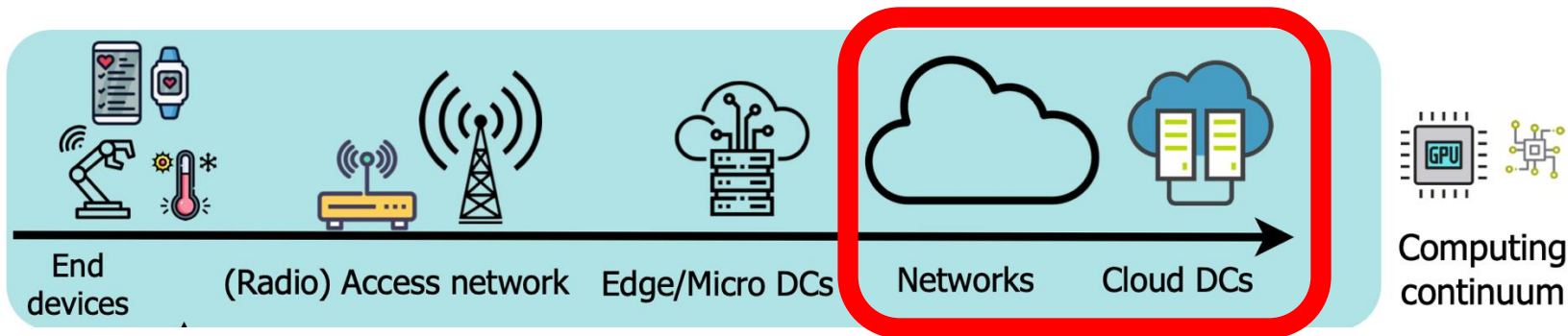
Score: **98/100** | Rate: **8%**



Funded by  
the European Union

This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement N° 101021808.

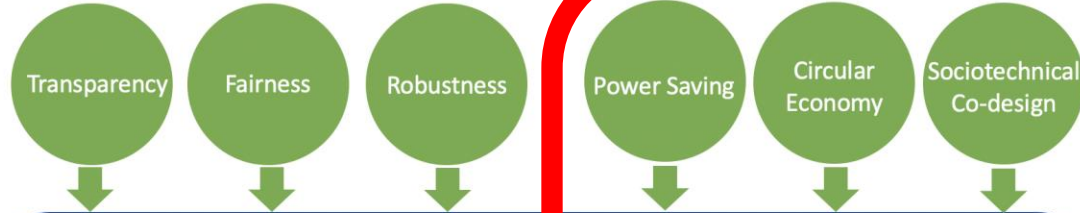
# Huge but not Monster: Energy ?



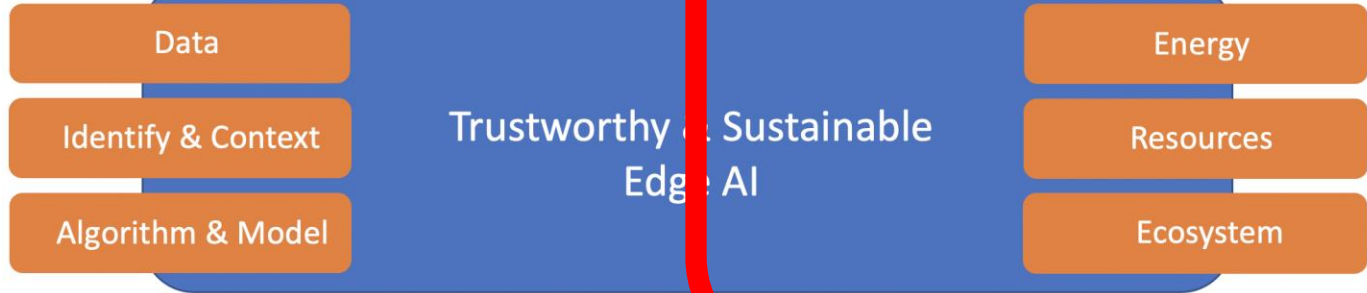




## Enablers



## Scoping



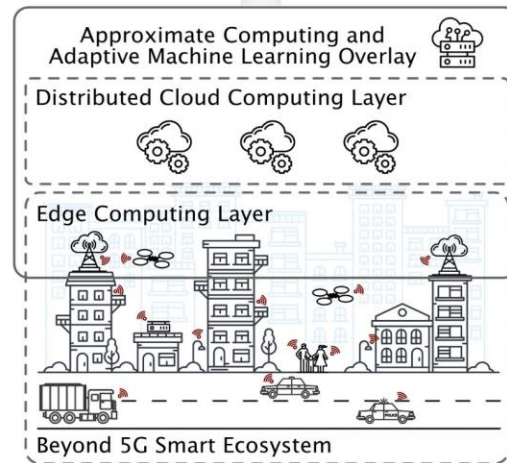
## Meta Schemes



# Sustainable Edge AI

- EU Marie Curie grant of €4M
- 20+ industrial academic partners

Approximate Computing for Power  
and Energy Optimisation



**APROPOS Project**  
**Sustainable AI**

**Score: 14.5/15 Rate: 3%**



# Vehicular Data



?



- Data increases
- Electric cars: **battery life** matters!

750MB per second, as Google's driverless car prototype reported

Autonomous Car - Sensors & Data		
Sensors	# Number	Data Volume
Camera	(8-12)	500 - 3500 Mbit/s
LiDAR	(2-4)	20-100 Mbit/s
Radar	(4-6)	0.1-15 Mbit/s
GPS		50 Kb/s
Ultrasonic	(8-16)	500-3500 Mb/s
20 TB Car/Day		

Source: Elektrobit

# Lesson #4



There is **no free lunch**

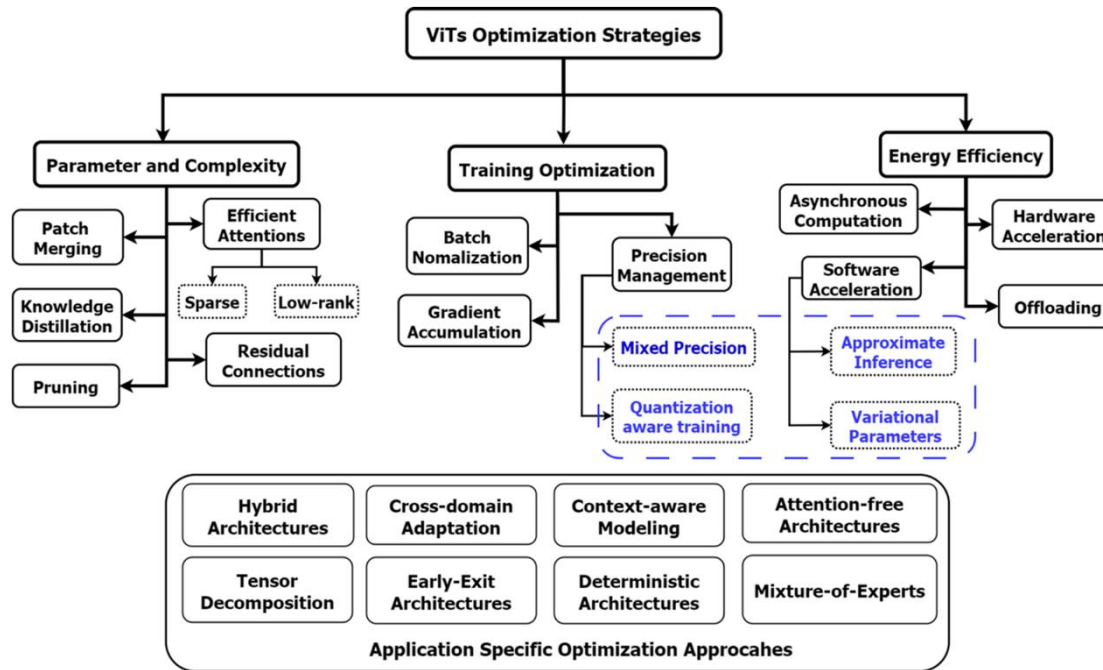


**Safety Critical**  
VS  
**Energy Saving**

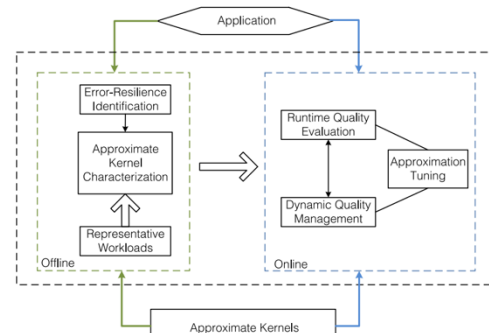
Can we strike both ?

# Energy Awareness

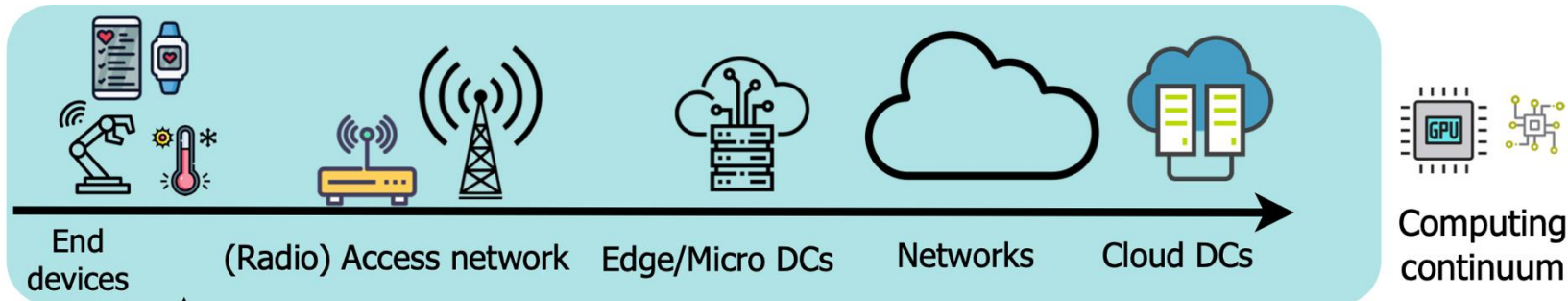
- Adaptive Approximation



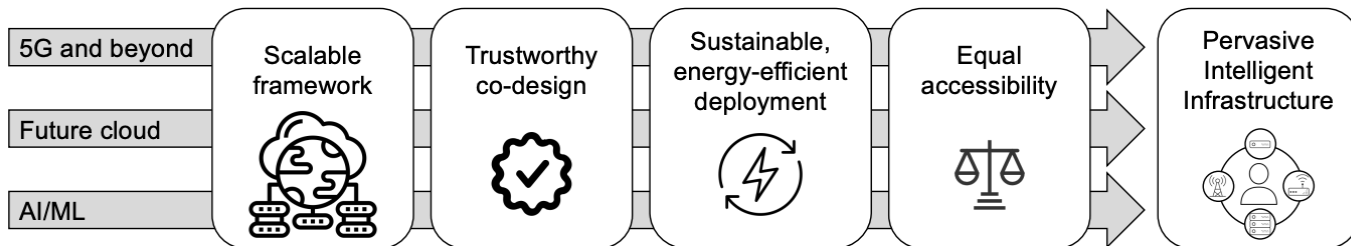
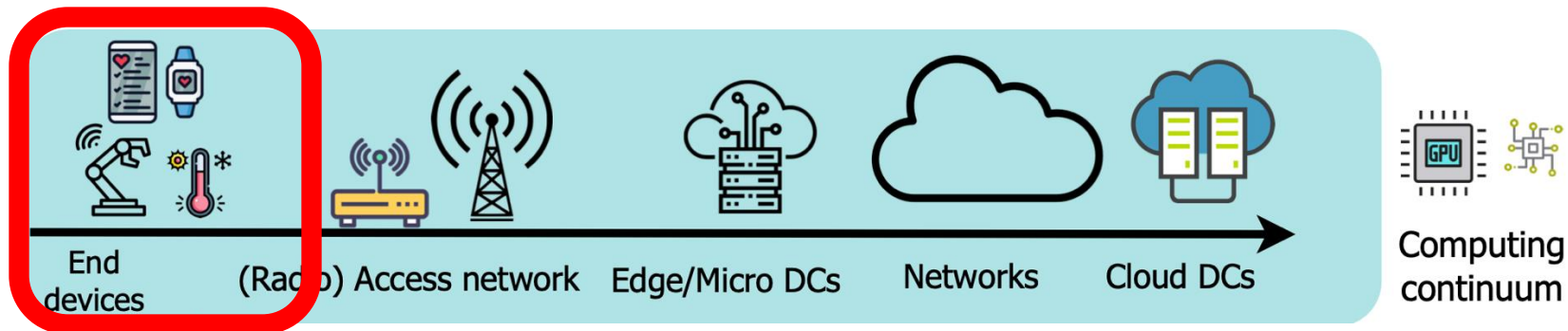
Year	Dataset	Camera	LiDAR	Radar	Radar	Sensors Included	GPUS	GPU	HD MAP	URL
2012 - 2022	KITTI [205]	Y	Y	N	N	Y	Y	N	Y	KITTI
2013 - 2019	KITTI Dataset [31]	Y	Y	N	Y	Y	Y	N	Y	KITTI
2016	HDK [143]	Y	Y	N	N	Y	Y	N	Y	HDK
2016	CVC-14 [136]	Y	N	N	N	N	N	N	N	CVC-14
2016	BrainCam [126]	Y	N	Y	Y	Y	Y	N	N	BrainCam
2016	JAAD [146]	Y	N	N	N	N	N	N	N	JAAD
2016	Cityscapes [53]	Y	N	N	N	Y	Y	N	N	CITYSCAPES
2016	Ucity	Y	N	N	N	Y	Y	N	N	Ucity
2016 - 2019	commal driving dataset [264]	Y	N	Y	Y	Y	Y	N	N	Commal dataset
2017	TRAM [102]	Y	N	Y	Y	Y	Y	N	N	TRAM
2017	Raincovert [109]	Y	N	N	N	N	N	N	N	Raincovert
2017	VPCNet [159]	Y	N	N	Y	N	N	N	N	VPCNet
2017	Teddlers	Y	N	N	N	N	N	N	N	Teddlers
2017	TorontoCity [156]	Y	Y	N	N	N	N	N	N	TorontoCity
2017	CityPersons	Y	N	N	N	N	N	N	N	CityPersons
2017	Mapillary Vista [221]	Y	N	N	N	N	N	N	N	Mapillary Vista
2017	Multi-spectral (Univ of Tokyo) [92]	Y	N	Y	N	N	N	N	N	Multi-spectral
2018	CULane [233]	Y	N	N	Y	N	N	N	N	CULane
2018	DBNet [47]	Y	Y	Y	Y	Y	Y	N	N	DBNet
2018	HD [151]	Y	N	N	N	N	N	N	N	HD
2018	MVSEC (U Penn) [403]	Y	Y	N	N	N	N	N	N	MVSEC
2018	NightOwls [222]	Y	N	N	N	N	N	N	N	NightOwls
2018	Road Damage [199]	Y	N	N	N	N	N	N	N	Road Damage
2018	WildData [377]	Y	N	N	N	N	N	N	N	wildData
2018 - 2020	BDD-100K [370]	Y	Y	N	Y	Y	Y	N	N	Berkeley
2018 - 2020	Apolliscape [113]	Y	Y	N	Y	Y	Y	N	N	Apollis
2018 - 2020	Honda Driving [239]	Y	Y	N	Y	Y	Y	N	N	HDD
2019	Argoverse [41]	Y	Y	N	Y	Y	Y	N	N	Argo
2019	Astra Hike [207]	Y	Y	N	N	N	N	N	N	Astra
2019	BLVD [152]	Y	Y	N	N	N	N	N	N	BLVD
2019	Bosch Driving [34]	Y	N	N	N	N	N	N	N	BOSCH
2019	EuroCity [154]	Y	N	N	N	N	N	N	N	Eurocity Persons
2019	EU Long-term Dataset [354]	Y	Y	Y	Y	Y	Y	N	N	EU Dataset
2019	IoVInterSec [240]	Y	Y	N	N	N	N	N	N	IoVInterSec
2019	StreetLearn [209]	Y	N	N	N	N	N	N	N	Street Learn
2019	Fundata	Y	Y	N	Y	Y	Y	N	N	Fundata
2019	WoodScape [168]	Y	Y	N	Y	Y	Y	N	N	WoodScape
2019	Unsupervised Llama - Bosch [25]	Y	Y	N	Y	Y	Y	N	N	Bosch
2020	L-Seasons [186]	Y	N	N	N	N	N	N	N	L-Seasons
2020	A*3D [242]	Y	Y	N	N	N	N	N	N	ASTAR-3D
2020	subscans [17]	Y	Y	N	N	N	N	N	N	subscans
2020	POSS [234]	Y	Y	N	N	N	N	N	N	POSS
2020	DDO20 [108]	Y	N	N	N	Y	Y	N	N	DDO20
2020	Highway Driving [139]	Y	N	N	N	N	N	N	N	Kate
2020	Lyft Level 3 [106]	Y	Y	N	N	N	Y	Y	N	lyft
2020	Beta Urban Adventure [107]	Y	Y	Y	Y	Y	Y	Y	N	BREND
2020	Ford Motor AV [14]	Y	Y	N	Y	Y	Y	Y	N	Ford Seasonal
2020	A2D1 [80]	Y	Y	N	N	N	N	N	N	Audi
2020	LiBBE [36]	Y	Y	Y	N	N	N	N	N	LiBBE
2020	Toronto-3D	Y	Y	N	Y	Y	Y	N	N	Toronto-3D
2021	NEALIX [124]	Y	Y	Y	Y	Y	Y	N	N	Nealia
2021	CADIC [244]	Y	N	N	Y	Y	Y	N	N	CADIC
2021	RadarScenes [266]	Y	N	Y	Y	Y	Y	N	N	RadarScenes
2021	CARLADA [246]	Y	N	N	N	N	N	N	N	CARLADA
2021	Waymo [290]	Y	N	N	N	N	N	N	N	Waymo Open
2021	SODA10M [93]	Y	N	N	N	N	N	N	N	SODA10M
2021	PredictLadderTech [61]	Y	Y	Y	Y	Y	Y	N	N	PredictLadderTech
2021	ONCE [202]	Y	Y	N	N	N	N	N	N	ONCE
2021	Deep Route All	Y	Y	Y	Y	Y	Y	N	N	Deep Route
2021	DeepAR [69]	Y	Y	Y	Y	Y	Y	N	N	DeepAR
2022	MUAD75	Y	N	N	N	N	N	N	N	MUAD
2022	SHIP	Y	N	N	N	N	N	N	N	SHIP
2022	Rope3D [64]	Y	Y	N	N	Y	Y	N	N	Rope3D
2022	CUDA [68]	Y	Y	N	N	Y	Y	N	N	CUDA
2022	View-of-Iso [132]	Y	Y	N	Y	Y	Y	N	N	View-of-Iso
2023	Libra-CS [76]	N	Y	N	N	N	N	N	N	Libra-CS
2023	Zad [11]	N	Y	N	Y	Y	Y	N	N	Zad
2023	Race-Car [151]	N	Y	Y	Y	Y	Y	N	N	Race Car



# Why + Lessons + *Directions* ?



# Tiny but not Waste: Safe, Circular



# Tiny but not Waste: Safe, Circular

“IoT-KEEPER: Detecting Malicious IoT Network Activity using Online Traffic Analysis at the Edge” *IEEE TNSM 2020, Q1 Journal IF 4.7*

“Context-dependent Trade-offs around Platform-to-platform Openness: The Case of IoT” *Elsevier Technovation 2021, Q1 Journal IF 11.1*

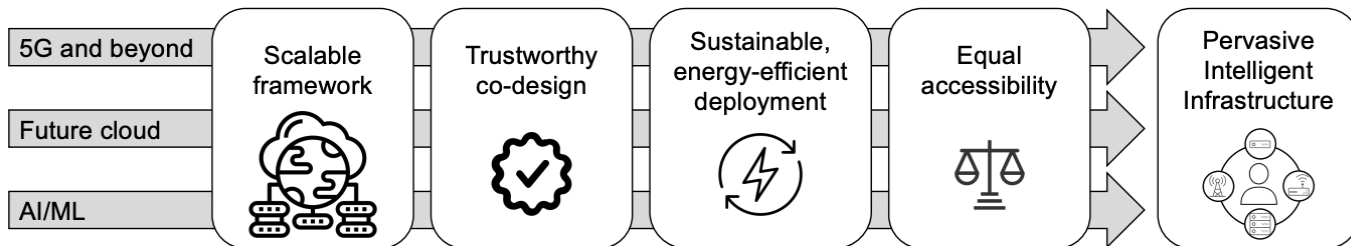
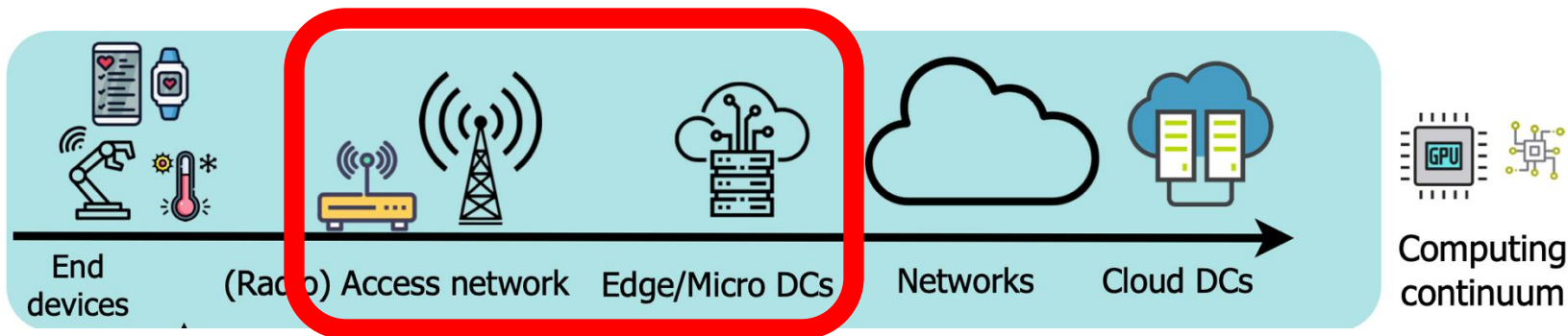
“Where Is My Tag? Unveiling Alternative Uses of the Apple FindMy Service” *IEEE WoWMoM 2022, Core A*

“XAI for Accountable MI Detection in IoT Emergency Communication Systems” *ACM IoT 2023, Best Paper Runner-up*



“ARASEC: Adaptive Resource Allocation and Model Training for Serverless Edge Computing” *IEEE Internet Computing 2024, Q1 Journal IF 3.7*

# Bigger but not Static: Trust + Reusable



# Bigger but not Static: Trust + Reusable

“Characterising the Role of Pre-Processing Parameters in Audio-based Embedded Machine Learning” *ACM SenSys 2021, Core A\**

“Bias Detection and Generalization in AI Algorithms on Edge for Autonomous Driving” *ACM/IEEE SEC 2022*

“Bias in Automated Speaker Recognition” *ACM FAccT 2023 Best PhD Paper*

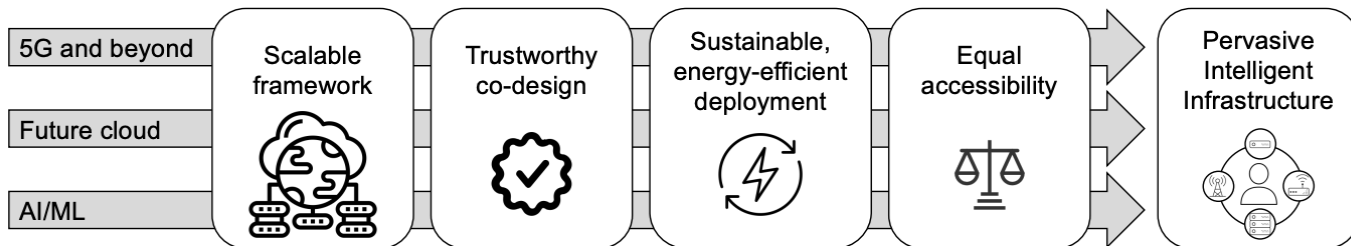
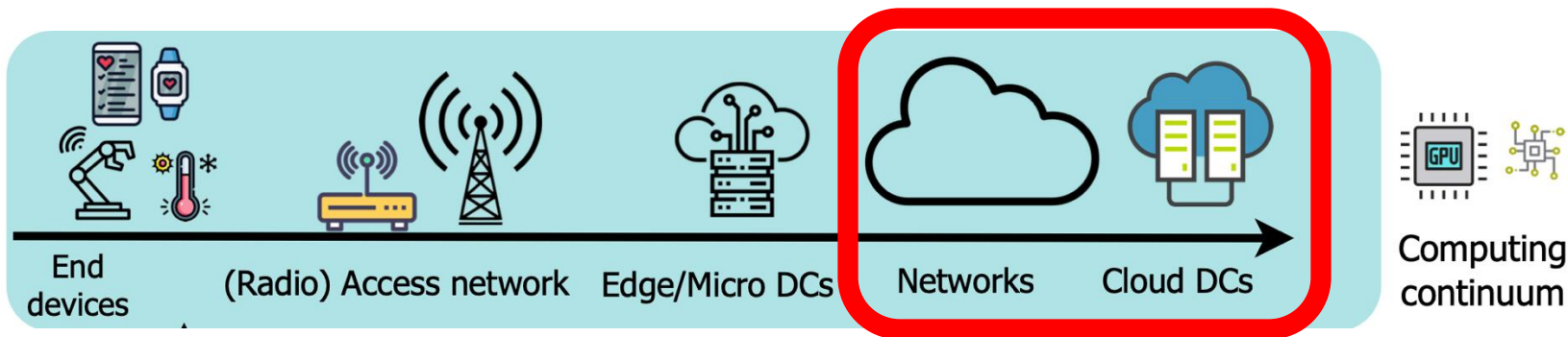


“Bias Propagation in On-device Machine Learning Workflows” *ACM TOSEM 2023, Q1 Journal IF 6.6*

“SPATIAL Architecture: Gauging and Monitoring the AI Inference Capabilities of Modern Applications” *IEEE ICDCS 2024, Core A\**



# Huge but not Monster: Approximate AI



# Huge but not Monster: Approximate AI

“Transfer Learning-Based Outdoor Position Recovery with Cellular Data”

*IEEE TMC 2021, Q1 Journal IF 7.7*

“Approximate Edge AI for Energy Efficient Autonomous Driving Services”

*IEEE COMST 2023, Q1 Journal IF 35,6*

“Nimbus: Towards Latency-Energy Efficient Task Offloading for AR Services”

*IEEE TCC 2023, Q1 Journal IF 5,3*

“Test-time Specialization of Dynamic Neural Networks”

*IEEE CVPR MAT 2024, Best Paper Award*



“Approximating Vision Transformers for Edge: Variational Inference and Mixed-precision for Multi-modal” *Spring Nature Computing 2025, Q1*

# Takeaway

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DEPARTMENT: INTERNET OF THINGS, PEOPLE, AND PROCESSES

## Revisiting Edge AI: Opportunities and Challenges

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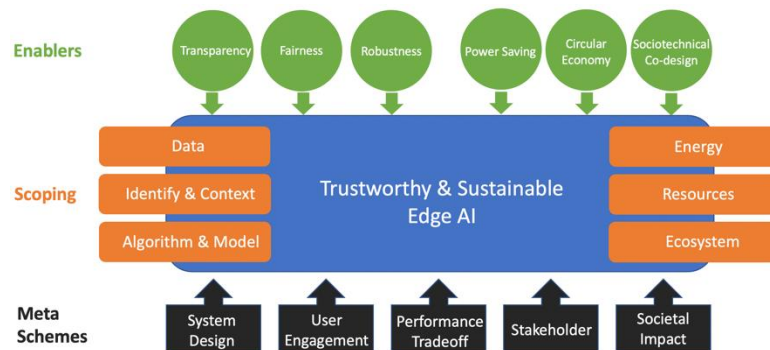
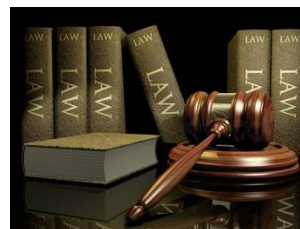
Edge artificial intelligence (AI) is an innovative computing paradigm that aims to shift the training and inference of machine learning models to the edge of the network. This paradigm offers the opportunity to significantly impact our everyday lives with new services such as autonomous driving and ubiquitous personalized health care. Nevertheless, bringing intelligence to the edge involves several major challenges, which include the need to constrain model architecture designs, the secure distribution and execution of the trained models, and the substantial network load required to distribute the models and data collected for training. In this article, we highlight key aspects in the development of edge AI in the past and connect them to current challenges. This article aims to identify research opportunities for edge AI, relevant to bring together the research in the fields of artificial intelligence and edge computing.

## Roadmap for Edge AI: A Dagstuhl Perspective

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"Trustworthy and Sustainable Edge AI: A Research Agenda"

# Outlook



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