# The essential role of Al in the 5G future

How machine learning is accelerating wireless innovations in the new decade and beyond



# The essential role of AI in the 5G future



5G and AI are two synergistic, essential ingredients that are fueling future innovations



Applying AI to solve difficult wireless challenges and deliver new values



Al plays an expanding role in the evolution of 5G towards 6G



Unifying connectivity fabric that can efficiently connect virtually everything around us

# 5G

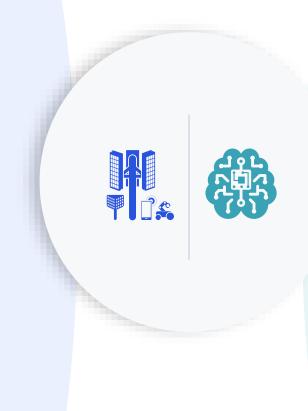
Extreme capacity

Multi-Gbps speeds

Ultra-high reliability and low latency

Robust end-to-end security and privacy

New and diverse services, spectrum, deployments



Learning platform that can make virtually everything around us intelligent



Contextual awareness
Personalization at scale
Intelligent, intuitive, and automated actions
Continual improvement through self-learning

Solving seemingly impossible-to-model problems

Two synergistic and essential ingredients fueling future innovations

#### Advancement in AI is making

# 5G better

Elevated level of performance

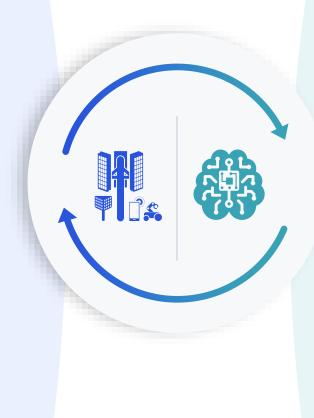
More efficient resource utilization

Energy reduction for longer battery life

Personalized security and privacy

Continuous enhancements over time

New and enhanced system capabilities



#### Proliferation of 5G is making

# Albetter

Responsive user experiences and services

Lifelong learning

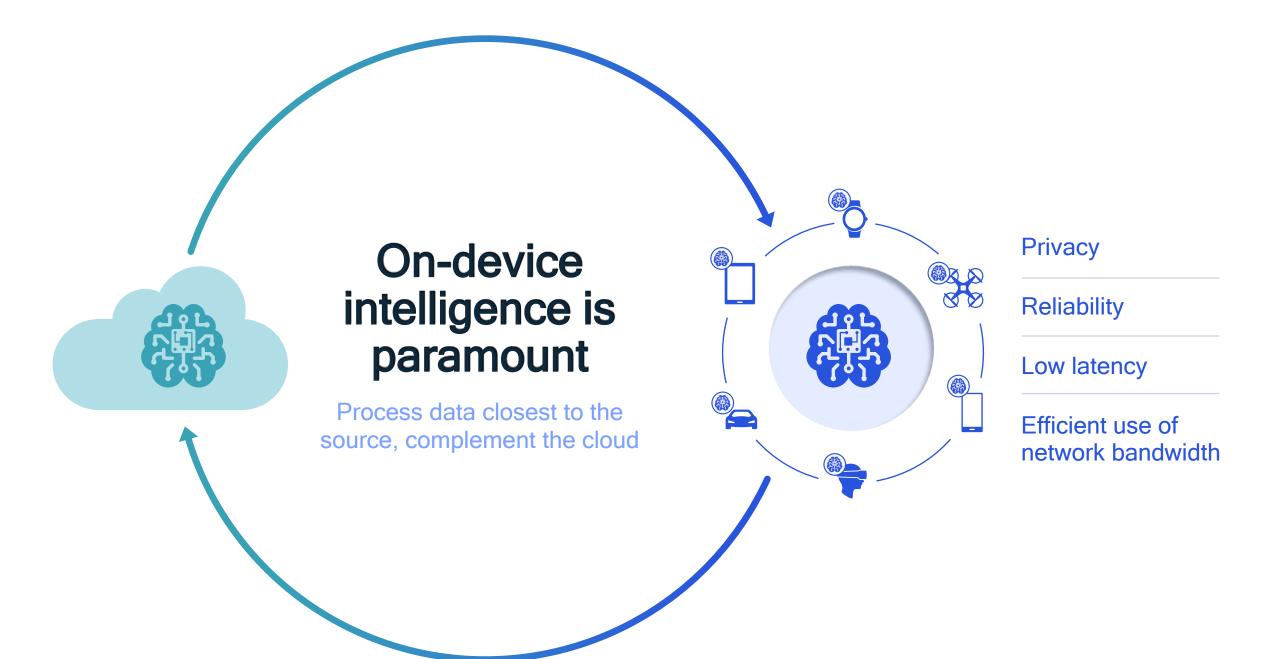
Flexibility for distributed functionality across devices

On-device intelligence assisted by cloud

Scale intelligence through distributed learning

Massive data aggregation for improved AI models

5G and AI are working together to accelerate innovations



# Transformation of the Connected Intelligent Edge has begun at scale

Processing data closer to devices at the edge derives new system values (e.g., lower latency, enhanced privacy)



Cloud

Edge cloud

Past
Cloud-centric Al

Al training and inference in the central cloud

Today

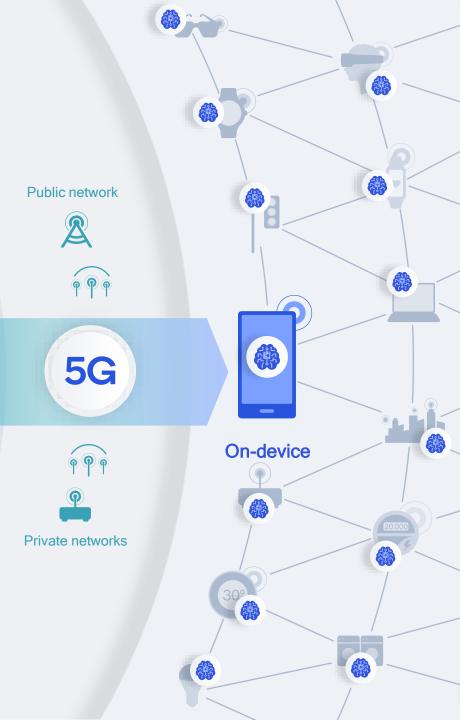
Partially-distributed Al

Power-efficient on-device Al inference

Future

Fully-distributed Al

With lifelong ondevice learning



Local network analytics

Low-latency interactive content

# Connected Intelligent Edge

brings new and enhanced services

Boundless XR

On-demand computing

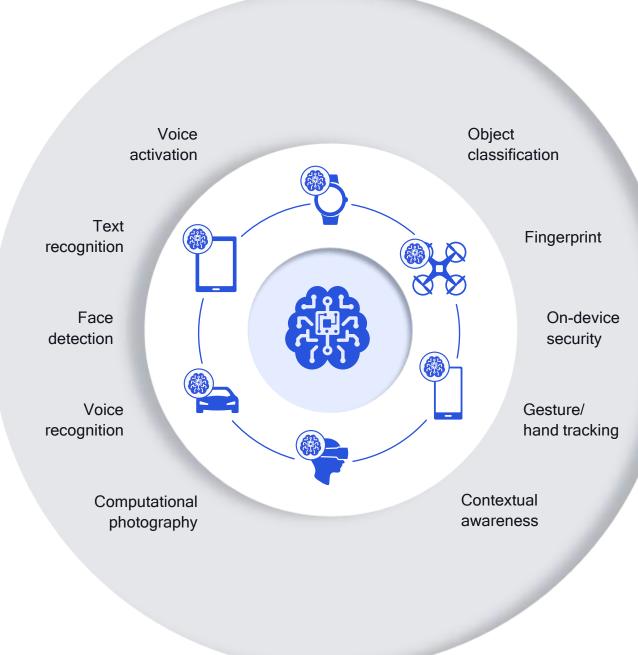
Edge Cloud Al



On-device Al

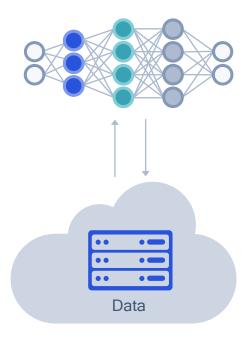
Industrial automation and control

Enterprise data

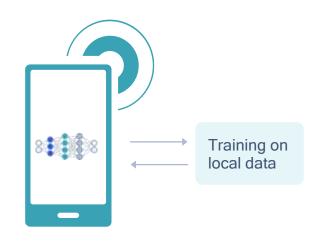


## Federated learning brings on-device learning to new level

#### Offline learning

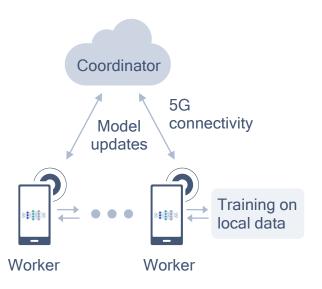


#### On-device learning



Locally adapt once to a few samples (e.g., few shot learning) or continuously (e.g., unsupervised learning)

#### Federated learning



Aggregate model updates across multiple users to globally improve model from more diverse data

Offline training prior to deployment

Local adaptation

Global adaptatior

# Our research focus to improve 5G with Al

With a cloud-to-network-to-device approach for data collection and learning

#### **Distributed Cloud**

Central and edge clouds

For end-to-end service optimization

#### **Disaggregated Network**

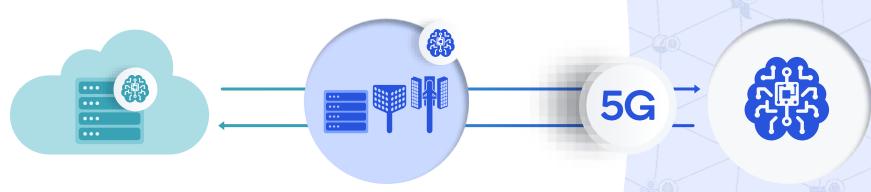
Virtualized and disaggregated RAN

For network and device optimization

#### **Edge Devices**

On-device intelligence

For local device optimization



Steps towards enabling AI/ML cloud and device platforms

#### **Shorter Term**

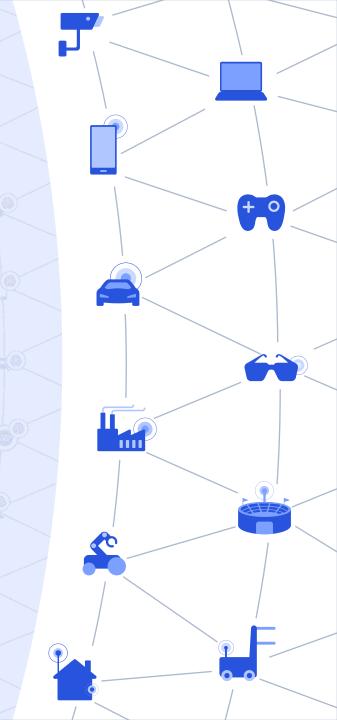
Continue to define data collection for new use cases hosted at the RAN

#### Medium Term

Enabling jointly optimized AI/ML use cases between RAN functions and device

#### Longer Term

Joint cloud, core, RAN, and device AI/ML functions



# Applying Al

to solve difficult wireless challenges and deliver new values

### Wireless challenges



Hard-to-model problems



Computational infeasibility of optimal solution



Efficient modem parameter optimization

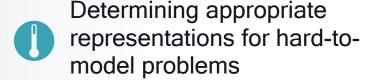


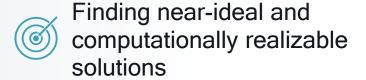
Dealing with non-linearity



Al-enhanced wireless communications

### Al strengths



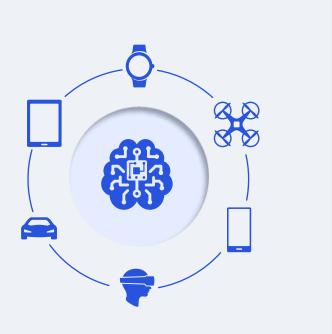




# Applying AI to solve difficult wireless challenges

Deep wireless domain knowledge is required to optimally use AI capabilities

## On-device AI improves the 5G end-to-end system



#### Radio awareness

Environmental and contextual sensing that reduces access overhead and latency



#### Enhanced device experience

More intelligent beamforming and power management improve throughput, robustness, and battery life



#### Improved system performance

On-device inference reduces network data traffic for more efficient mobility and spectrum utilization



#### Better radio security

Detecting and defending against malicious base station spoofing and jamming with fingerprinting



### Radio awareness

achieved by advanced on-device Al algorithms



#### **Spectrum sensing and access**

Predict activities of other devices for more efficient access and better scheduling to improve 5G system performance



#### **Contextual awareness**

Use device context (e.g., position, velocity, or in-car) derived from RF, sensors, traffic activities to improve device experience



#### Environment (RF) sensing

Detect gestures, movements, and objects by monitoring signal reflection patterns to enable new use cases

# on-device Al enhances 5G device experience

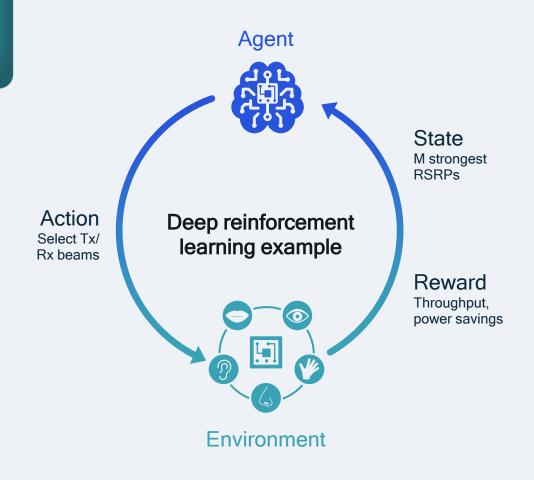


#### Better beam management

Incorporate location, velocity, other aspects of environmental and application awareness to improve robustness and throughput

#### More power saving

Optimize performance/power consumption tradeoffs by taking advantage of better contextual awareness on device

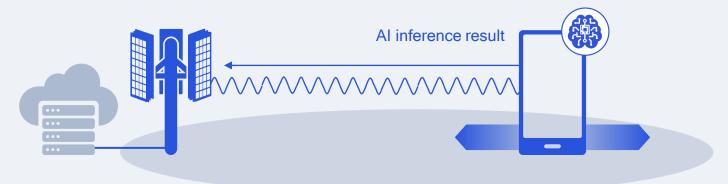


# On-device Al improves 5G system performance



#### **Better link adaptation**

Position-aware interference prediction can improve overall system throughput and spectral efficiency





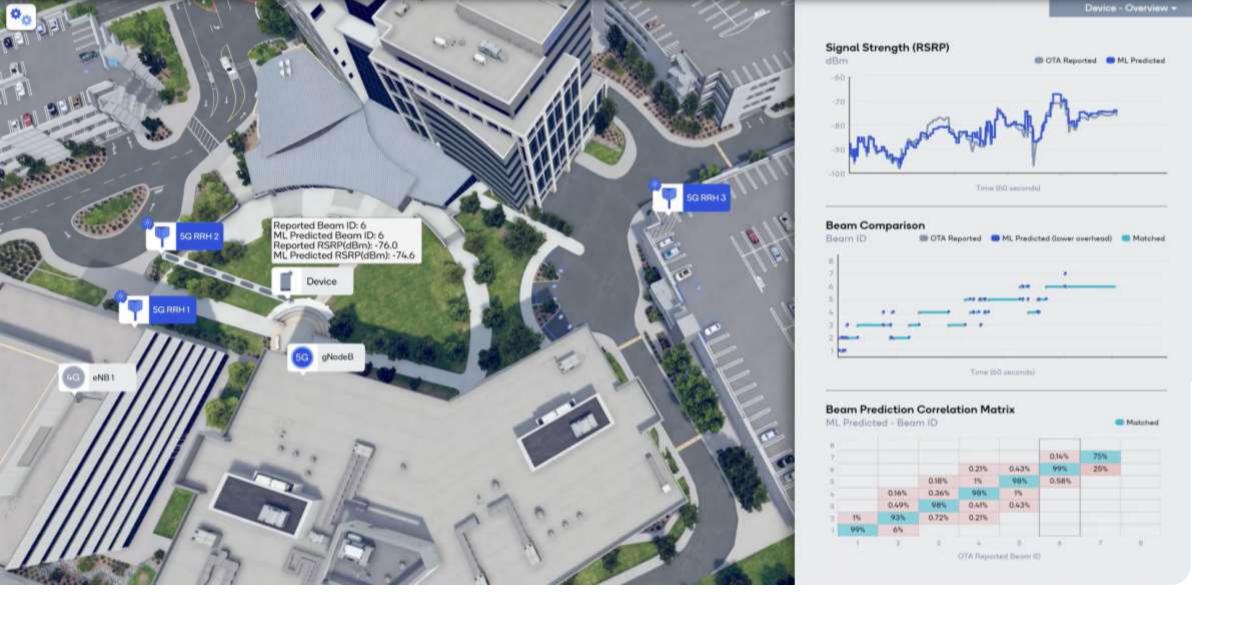
#### Reduced network loading

On-device Al inference reduces the amount of raw data needed be sent across the network



#### More seamless mobility

Device-centric mobility utilizes on-device Al and sensors to predict handovers



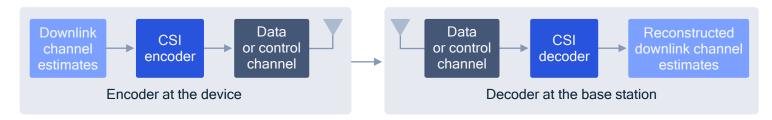
Demo: Mobile 5G mmWave beam prediction

# Applying Al for enhanced 5G air interface efficiency

Example: for uplink transmissions

#### Improving system spectral efficiency

Implementing a neural network framework for CSI<sup>1</sup> on non-linear temporal encoding and decoding



#### Improving device power efficiency

Optimizing transmit waveform to reduce peak-to-average power ratio (PAPR)

Allowing receiver to recover signal from a device operating in the non-linear PA region

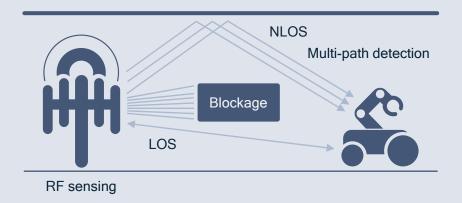


1 Channel State Information 17



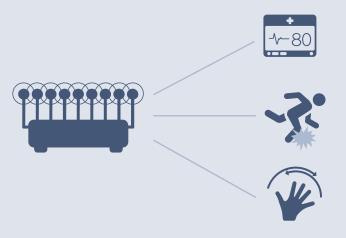
Demo: Enhanced 5G massive MIMO channel state feedback

#### More accurate device positioning



Learning device position over time without prior knowledge with RF sensing – complementing existing positioning methodologies<sup>1</sup>

#### Motion and gesture detection



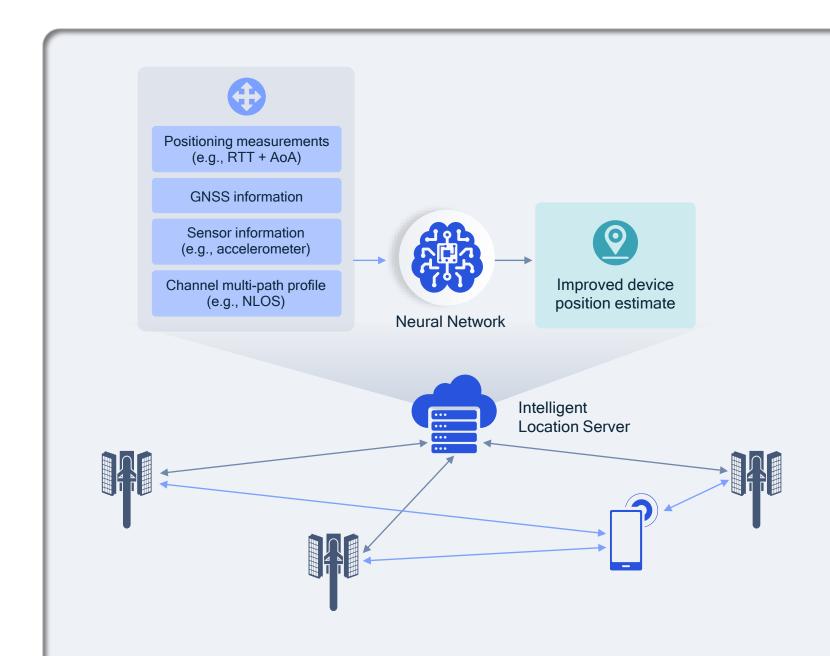
Sensing changes in environment to infer location and type of motion for a wide range of use cases (e.g., vital sign tracking, fall detection)

1. For example, Observed Time Difference of Arrival (OTDOA), Multiple Round Trip Time (Multi-RTT), Angle of Arrival (AoA)

# Applying AI for contextual awareness and environmental sensing

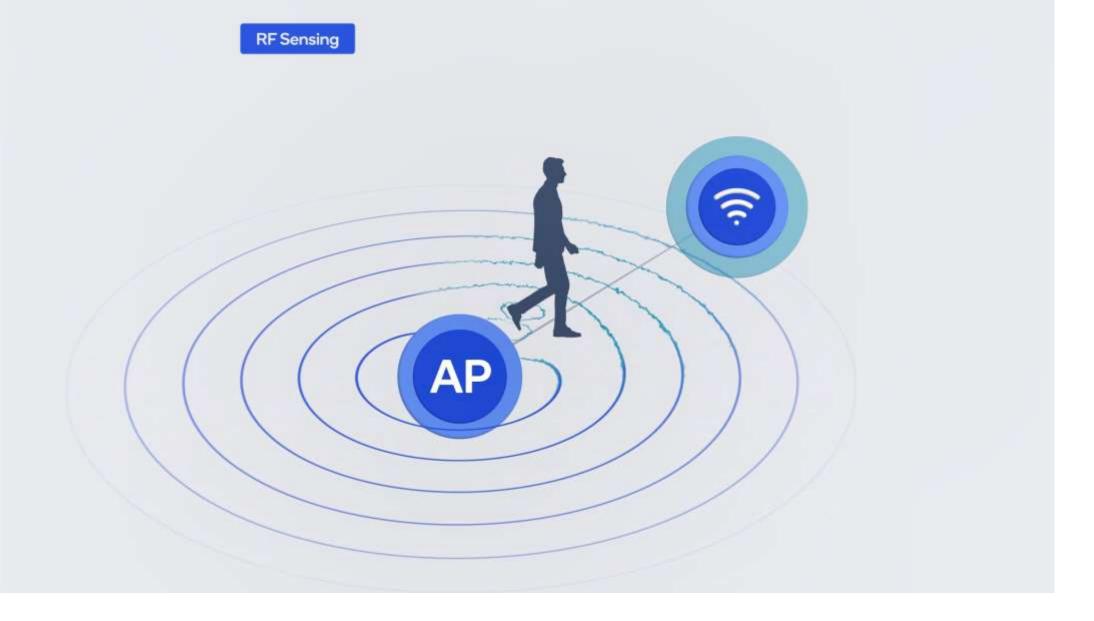
# Al/ML for enhanced 5G positioning performance

5G positioning is supplemented by various assisting information, such as GNSS, multi-path profiles, and other sensors





Demo: Precise 5G positioning with machine learning fusion



# Demo: Wireless Al-assisted indoor positioning

Leveraging unsupervised/weakly supervised learning – also applicable to 5G RF sensing (e.g., for positioning, motion and gesture detection)

## Al enables intelligent 5G network management

#### Enhanced service quality

Better mobility management, user localization, and user behavior and demand prediction



#### Simplified deployment

More capable Self Organizing Networks (SON) for e.g., mmWave network densification

#### Higher network efficiency

More efficient scheduling, radio resource utilization, congestion control and routing

#### Improved network security

More effective detection and defense against malicious attacks by analyzing a massive quantity of data



# A more intelligent way to deploy 5G mmWave



Create a digital twin of targeted deployment based on readily available sources/databases (e.g., Google Street/Aerial view, GIS, ...)



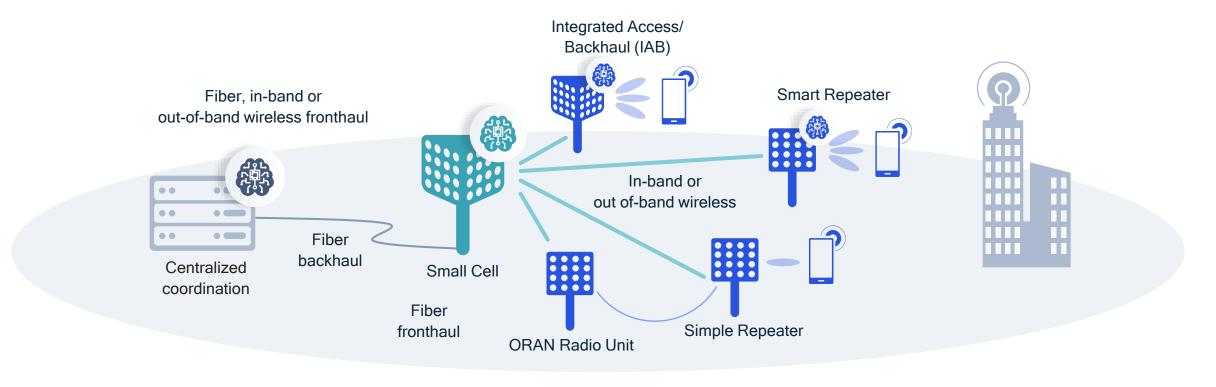
Reduce model complexity and balance tradeoffs through ML-based object recognition, clustering, and pruning



Optimize mmWave network deployment to provide focused capacity using diverse existing and new infrastructure (e.g., repeater, IAB,...)

## Distributed topology enables more efficient deployments

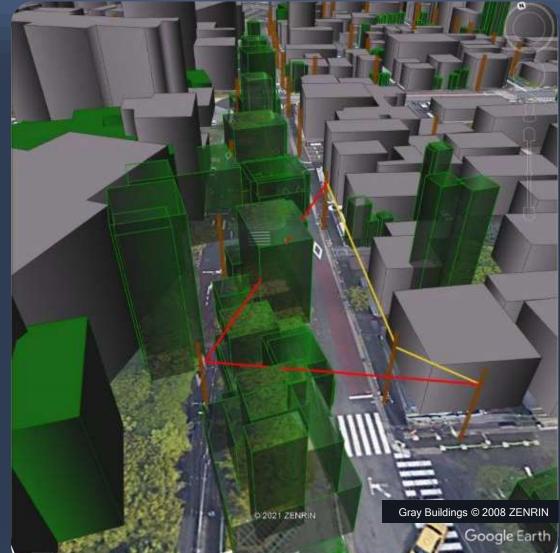
Standardization in e.g., 3GPP, O-RAN Alliance



# Creating a digital twin of targeted deployment using readily available databases

Example in Tokyo, Japan



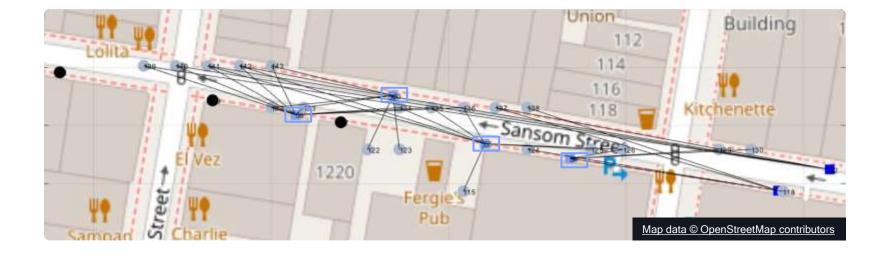


## 5G mmWave topology optimization example

Philadelphia design using small cells only

Two existing small cells serving a portion of device traffic

Four additional small cells to support DL/UL traffic of remaining devices



Unused

small cells

small cells

Device

hotspots

In-band

access

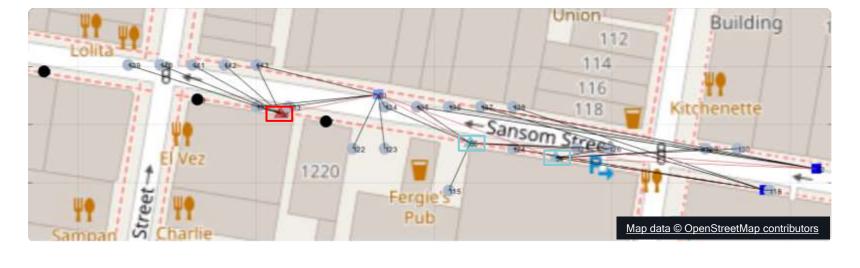
# 5G mmWave topology optimization example

Design with integrated access/backhaul, repeaters, small cells with in-band and out-of-band backhaul

Three new small cells are replaced with one IAB and two repeaters

One repeater is placed to bend signal around the corner to provide coverage

IAB now provides coverage to devices on the left, as new small cell cannot meet demand of those devices with its in-band bandwidth





Unuse

poles

Device

hotspots

Repeaters

Integrated access/

backhaul

In-band backhaul Out-of-band backhaul

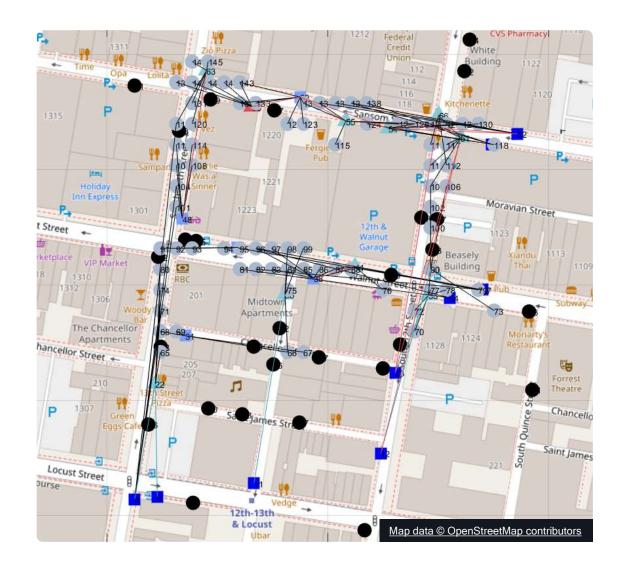
In-band access

### **Expanding design** to larger deployment area with diverse mmWave topologies

Refreshed design based on new traffic requirements of the larger area

Additional small cells, repeaters, IABs to address increased traffic demand

Utilizing both in-band and out-of-band backhaul links for even more capacity





Newly added small cells



poles



hotspots





backhaul

In-band

backhaul

Out-of-band

In-band backhaul access

# Machine learning

plays an expanding role in the evolution of 5G towards 6G



## Standardizing AI/ML in cellular communication systems

Broad range of work across standards and industry organizations



Developing AI use cases Architectural framework for ML Framework for evaluating intelligence level Framework for data handling to enable ML Al for autonomous and assisted driving



Developing an Al Mobile Device Requirement Spec (TS.47)

Focusing on Al mobile phone and tablet (may extend to IoT/wearable in future releases)



Defining reference architecture for Radio Intelligence Controller (RIC) and interfaces

Developing technical report: "AI/ML Workflow Description and Requirement"



Network automation and autonomy based on Al

Defining requirements and platform recommendations for reference implementation and interface standards



Data collection for network performance enhancements (RAN2/3 – Rel-16/17)

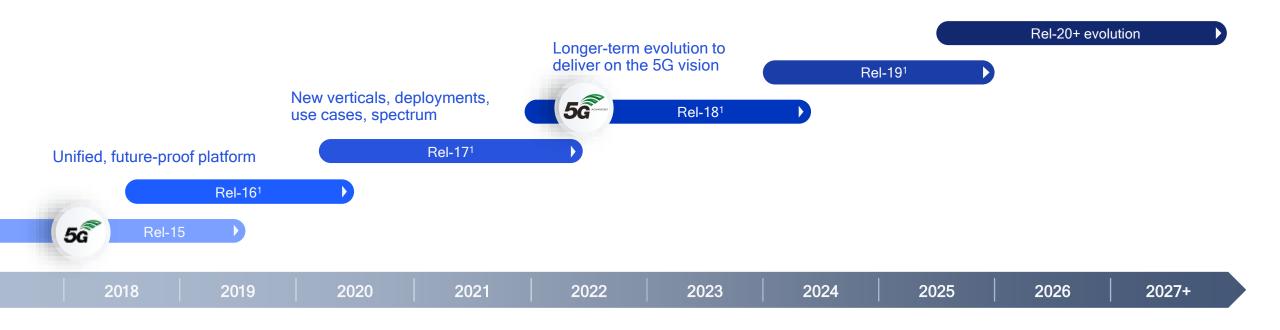
Study on AI/ML functional frameworks and use cases (RAN3 – Rel-17)

Network data analytic function for core AI/ML use cases (SA2 - Rel-16/17)

Management data analytic service, autonomous network (SA5 – R17)

Study on AI/ML model transfer performance requirements over 5G (SA1 – Rel-17)

## Driving the 5G technology evolution in the new decade



#### Rel-15 eMBB focus

- 5G NR foundation
- Smartphones, FWA, PC
- Expanding to venues, enterprises

#### Rel-16 industry expansion

- eURLLC and TSN for IIoT
- NR in unlicensed
- 5G V2X sidelink multicast
- In-band eMTC/NB-IoT
- Positioning

#### Rel-17 continued expansion

- Lower complexity NR-Light
- Higher precision positioning
- Improved IIoT, V2X, IAB, and more...

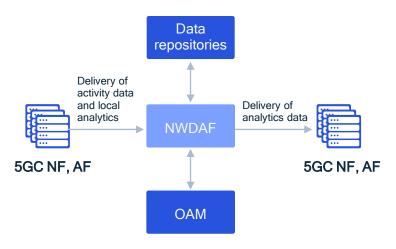
#### Rel-18+ 5G-Advanced

- Next set of 5G releases (i.e., 18, 19, 20, ...)
- Potential projects in discussions
- Rel-18 expected to start in 2022

## Data collection for network performance enhancements

Part of 3GPP Release 16

#### **Enhanced Network Automation (eNA)**



#### Minimization of Drive Testing (MDT)



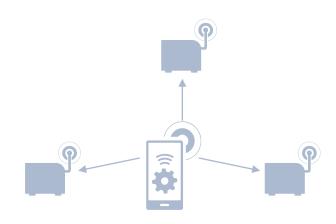
New enhanced core network function for data collection and exposure

Expanding NWDAF<sup>1</sup> from providing network slice analysis in Rel-15 to data collection and exposure from/to 5G core NF, AF, OAM<sup>2</sup>, data repositories

# Logged and immediate MDT, mobility history information, accessibility and L2 measurements<sup>3</sup>

Specifying features for identified use cases, including coverage, optimization, QoS verification, location information reporting, sensor data collection

#### Self Organizing Network (SON)



Mobility robust optimization (MRO), mobility load balancing (MLB), and RACH optimization

Specifying device reporting needed to enhance network configurations and inter-node information exchange (e.g., enhancements to interfaces like N2, Xn)

# Expanding 5G system support for wireless machine learning

Part of 3GPP Release 17

#### Enhancements for 5G network interfaces

Facilitating machine learning procedures such as model training and inference, as well as actions to enforce model inference output

#### Augmented network and device data collection

Supporting targeted applications (e.g., energy saving, load balancing, mobility management), operations enhancements, expanded use case<sup>1</sup>

#### Support for over-the-top AI/ML services

Introducing new QoS (Quality of Service) definitions that are tailored for machine learning model delivery over 5G













## Network architecture enhancements

Allowing for machine learning to run over different HW/SW and future RAN function split to improve flexibility and efficiency



## AI/ML procedure enhancements

Optimizing system for model management, training (e.g., federated and reinforcement learning), and inference



## Data management enhancements

Standardizing ML data storage/ access, data registration/ discovery, and data request/ subscription



## New and expanded use cases

Supporting traffic/mobility prediction, coverage/capacity optimization, massive MIMO, SON, CSI feedback, beam management, and other PHY/MAC and upper layer improvements



5G Advanced (Rel-18+) targets to expand wireless machine learning to the end-to-end system across RAN, device, and air interface

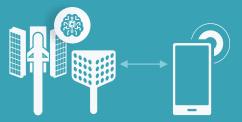
# Machine learning can bring continuous wireless enhancements

Al-native air interface design can enable continual system improvements in between major 3GPP releases through self-learning

Approximately 1.5-year cycle

Release X Release X+1

No standardized improvement during nominal Work/Study Item phase towards subsequent release



# Data-driven communication and network design

Data-driven system configuration provides end-to-end optimizations

Dynamic parameter adaptation based on fast machine learning algorithms

Neural network system design can customize to given wireless environment

# Machine learning is a key technology vector on the path to 6G



#### Wireless AI/ML

Data-driven communication and network design, with joint training, model sharing and distributed interference across networks and devices



#### Scalable network architecture

Disaggregation and virtualization from cloud-to-edge, and use of advanced relay/mesh topologies to address growing demand



#### New radio designs

Waveform/coding for MHz to THz, intelligent surfaces, joint comms. and sensing, large-scale MIMO, advanced duplexing, energy-efficient RF



#### Communications resiliency

End-to-end configurable security, post quantum security, robust networks tolerant to failures and attacks



#### Coordinated spectrum sharing

New paradigms for more efficient use of spectrum, leveraging location /environmental awareness for dynamic/adaptive coordination



#### Merging of worlds

Physical, digital, virtual, e.g., ubiquitous, low-power sensing/monitoring, immersive interactions taking human augmentation to next level





Link parameter prediction

Multi-cell interference learning

Mobility parameter prediction

Fully autonomous networks
Interference coordination/scheduling
Mobility handoff decisions



Data-driven propagation models

Data-driven optimization of signaling, measurements, and feedback



Predictive beam management
Channel state measurements
Device positioning

Joint sensing-communications
Dynamic ML model adaptation
Personalized lifelong learning

E2E approach with machine learning to improve 5G system performance and efficiency

## Transition to ML data-driven air interface design and operation

Neural network air interface design for coding, waveform, and multiple access

Joint training, model sharing, and distributed inference across network & devices

Dynamic air interface operation and adaptation



# Signal intelligence, baseband and medium access

ML-based channel feedback
Channel estimation & pilot optimization
MIMO detection
Link prediction & adaptation
Beam management and optimization
Spectrum sensing and sharing
Radio resource scheduling



# Network intelligence and system optimization

Coverage and capacity optimization
Traffic and mobility prediction
Energy saving
Cooperative edge caching
Content-aware X-layer optimization
Enhanced personalized security
TCP optimization



# Device intelligence and optimization

Digital front-end optimization
Antenna and RF optimization
Full duplex
Battery saving
Reflective intelligent surface



# Vertical intelligence and other capabilities

High-precision positioning Environmental sensing Contextual awareness Sensor fusion Vehicular communication

Our Al research areas to advance wireless communication

5G+AI

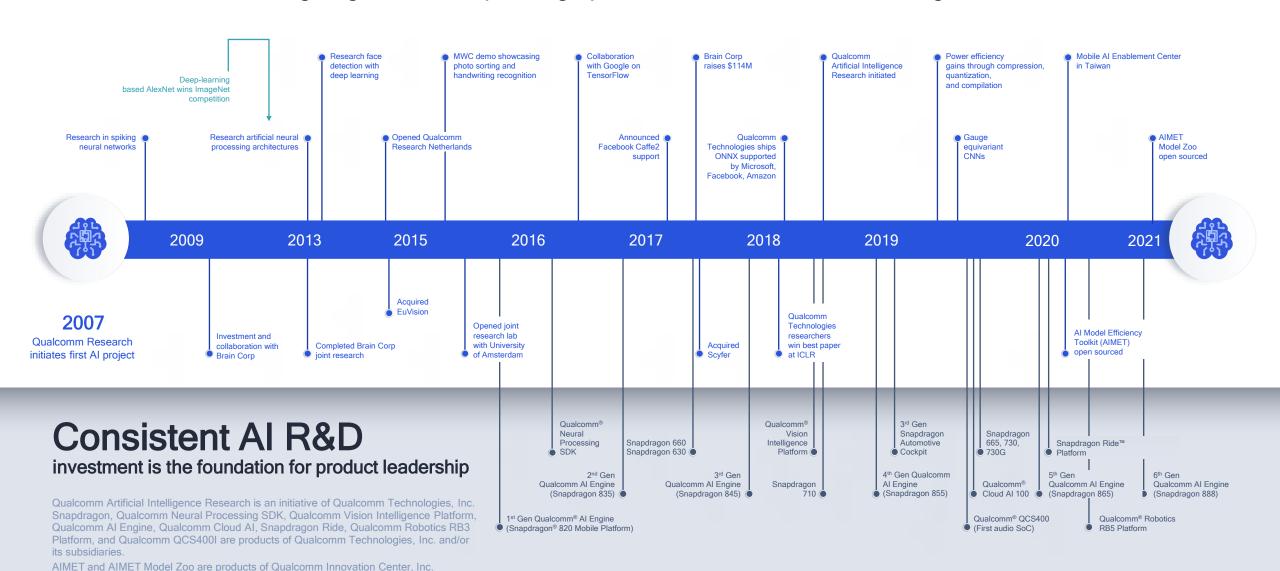


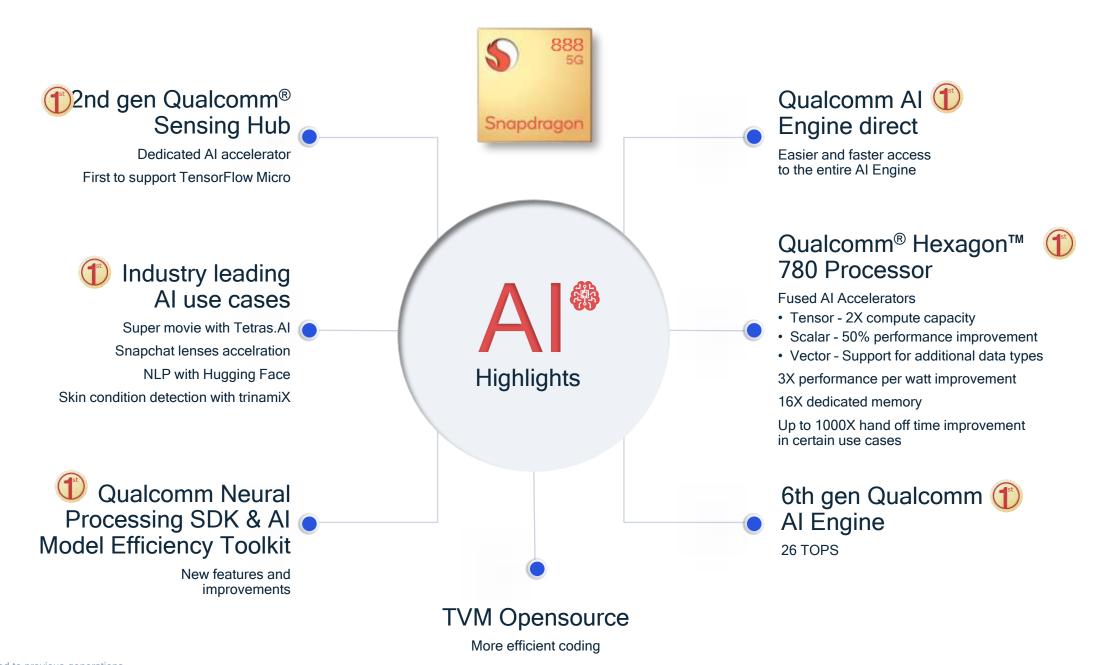
Qualcomm

Showcasing advanced wireless machine learning at MWC21 5G OTA prototypes and system simulations

### Our Al leadership

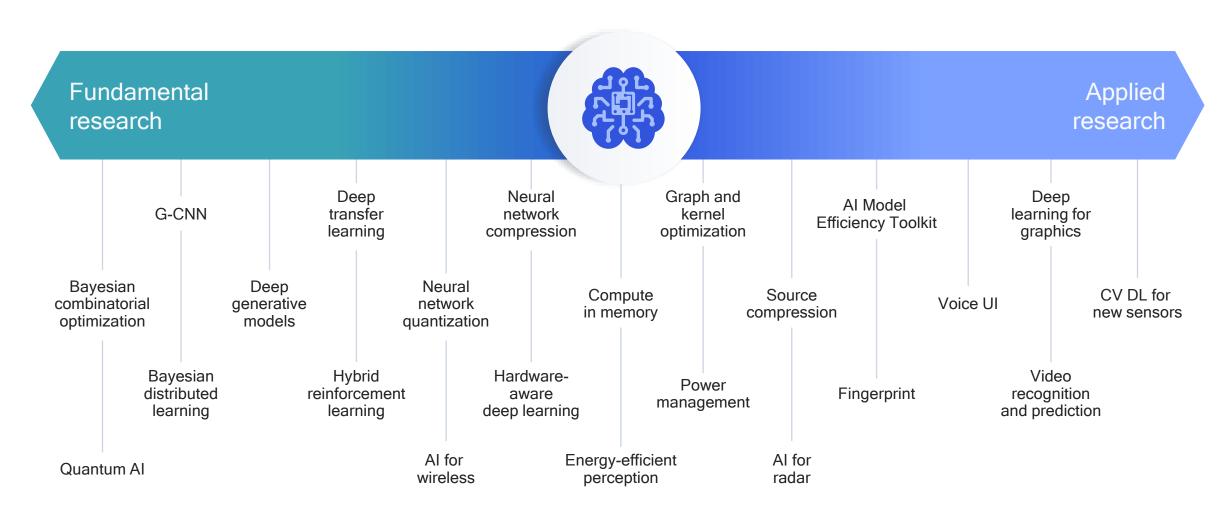
Over a decade of cutting-edge AI R&D, speeding up commercialization and enabling scale





### Leading research and development

Across the entire spectrum of Al



### Advancing Al research to make efficient Al ubiquitous

#### Power efficiency

Model design, compression, quantization, algorithms, efficient hardware, software tool

#### Personalization

Continuous learning, contextual, always-on, privacy-preserved, distributed learning

#### Efficient learning

Robust learning through minimal data, unsupervised learning, on-device learning

A platform to scale Al across the industry



#### Perception

Object detection, speech recognition, contextual fusion









#### Reasoning

Reinforcement learning for decision making

Action

Scene understanding, language understanding, behavior prediction

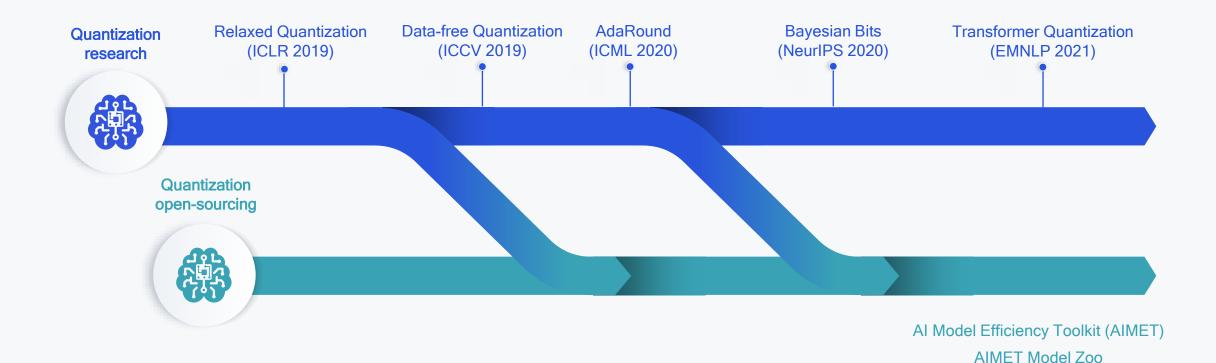


Cloud





Automotive



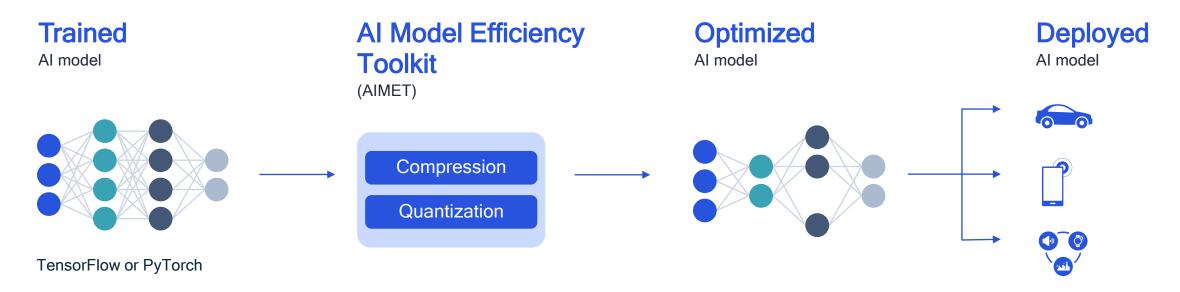
AIMET and AIMET Model Zoo are products of Qualcomm Innovation Center, Inc.

### Leading AI research and fast commercialization

Driving the industry towards integer inference and power-efficient Al

#### AIMET makes AI models small

Open-sourced GitHub project that includes state-of-the-art quantization and compression techniques from Qualcomm Al Research



If interested, please join the AIMET GitHub project: <a href="https://github.com/quic/aimet">https://github.com/quic/aimet</a>

Features:

State-of-the-art network compression tools

State-of-the-art quantization tools

Support for both TensorFlow and PyTorch

Benchmarks and tests for many models Developed by professional software developers

# AIMET

Providing advanced model efficiency features and benefits

#### Benefits



Lower



Lower memory bandwidth



Maintains model accuracy



Lower storage



Higher performance



Simple ease of use

#### **Features**

#### Quantization

State-of-the-art INT8 and INT4 performance

Post-training quantization methods, including Data-Free Quantization and Adaptive Rounding (AdaRound) – coming soon

Quantization-aware training
Quantization simulation

#### Compression

Efficient tensor decomposition and removal of redundant channels in convolution layers

Spatial singular value decomposition (SVD)
Channel pruning

#### Visualization

Analysis tools for drawing insights for quantization and compression

Weight ranges
Per-layer compression sensitivity

# AIMET Model Zoo

Accurate pre-trained 8-bit quantized models









Object detection



Pose estimation



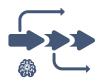
Speech recognition



5G and AI are two synergistic, essential ingredients that are fueling future innovations



Applying AI techniques to solve difficult wireless challenges and deliver new values



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#### The essential role of AI in the 5G future

How machine learning is accelerating wireless innovations in the new decade and beyond

#### Qualcomm

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