AI-RAN: Unveiling the Future of Telecom with **Breakthrough Innovations**

Feb 6, 2025 5:00 pm EST / Feb 7, 2025 7:00 am JST







Today's Speakers



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SoftBank Corp.



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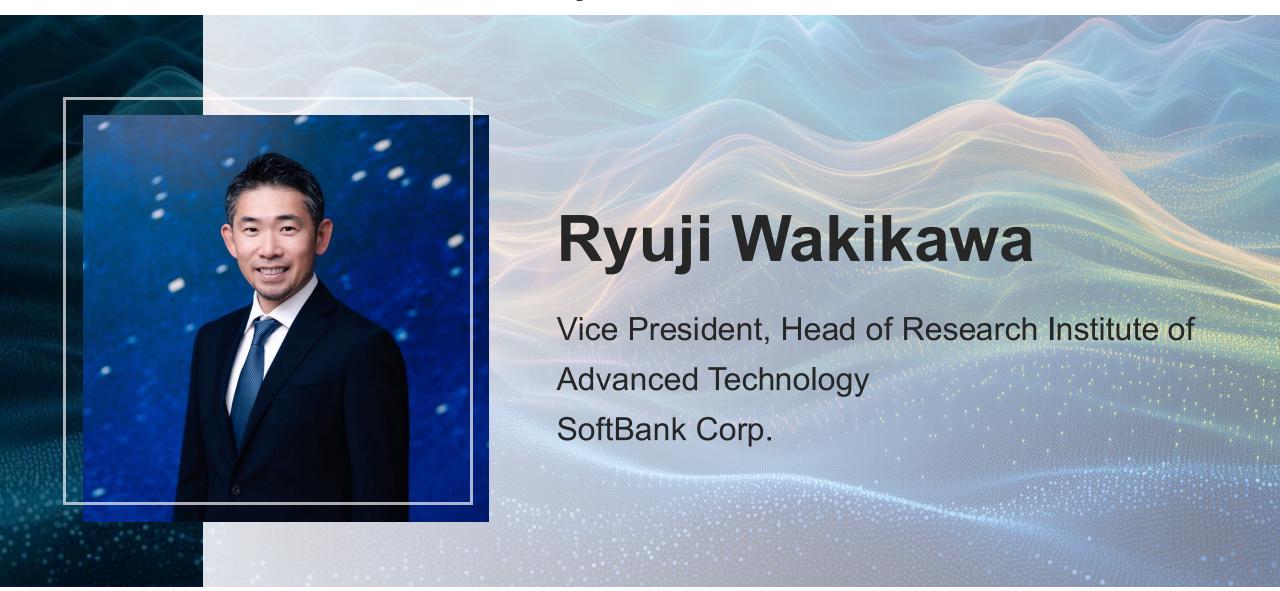
Development Department, Wireless

System Innovation Division

SoftBank Corp.

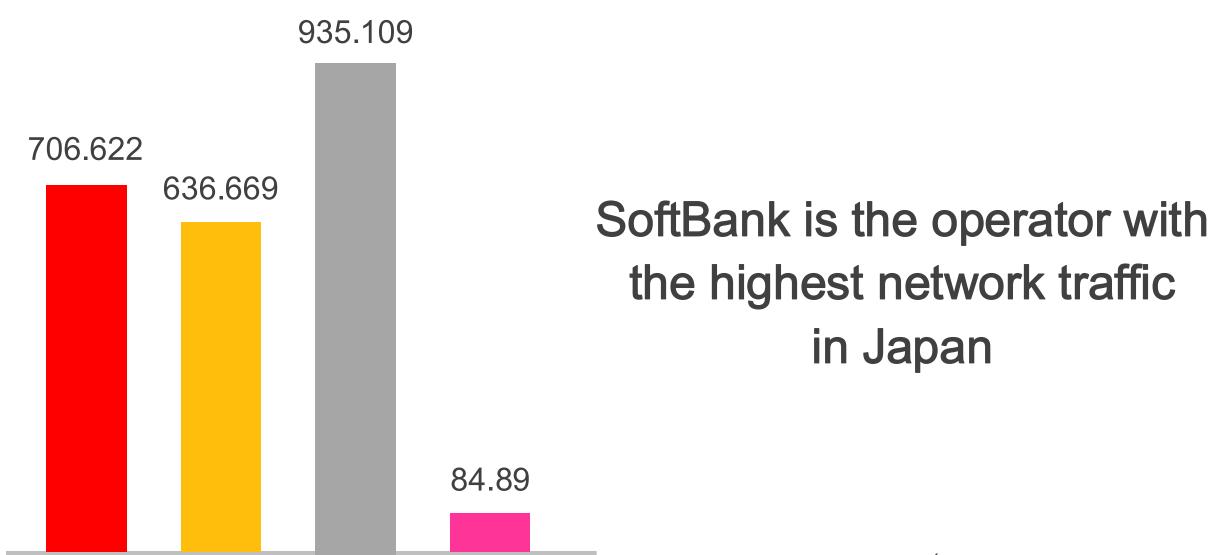


Speaker



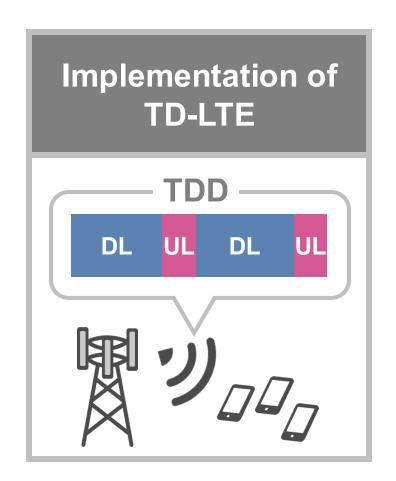
SoftBank's Al-RAN Initiative

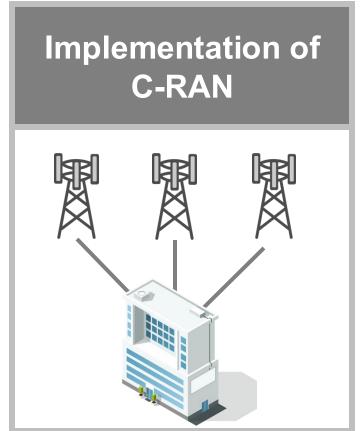
Monthly Aggregate Network Traffic of All Users (Unit : PB)





SoftBank's Measures on Network Capacity Expansion

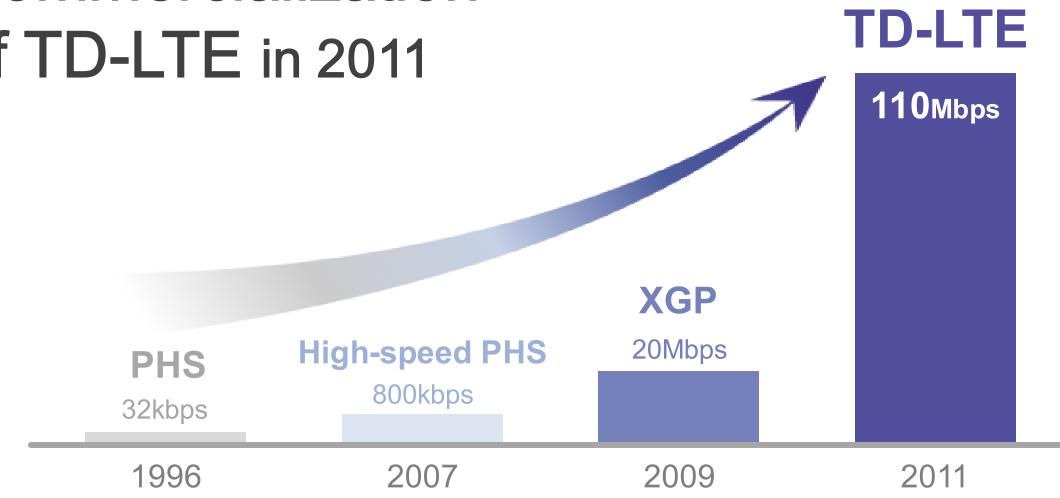








World's First Commercialization of TD-LTE in 2011





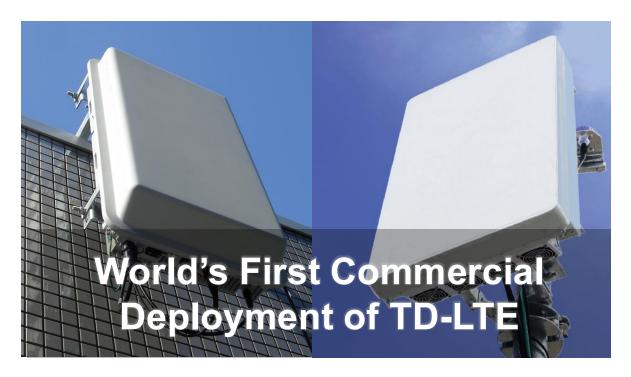
Difficulty of interference control in high-density areas

C-RAN implemented to enable better coordination between base stations



Implementation of Massive MIMO in LTE

Developed pricing strategies to address high traffic needs



Spectral efficiency:
Achieving breakthrough improvement in wireless capacity

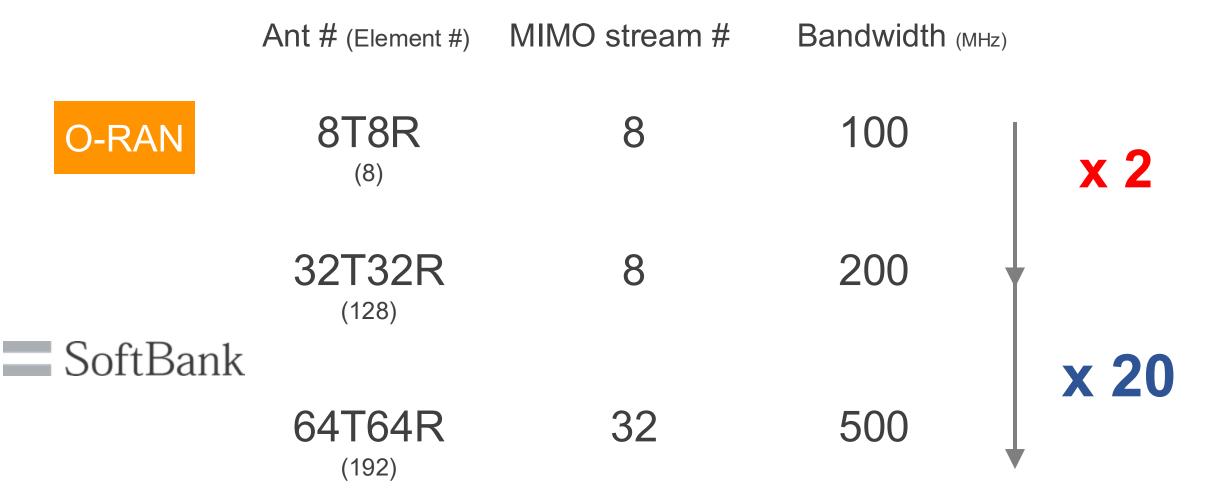
First launch in 2016



OpenRAN



Performance Limitations of oRAN (TDMA)



^{*} FD-LTE can be processed by software (bandwidth is ~20 MHz)



AI-RAN

oRAN + AI + Chip(GPU)

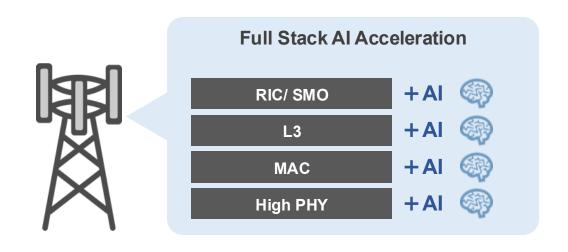
This is where Innovation occurs

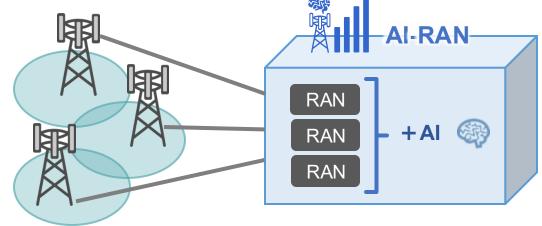


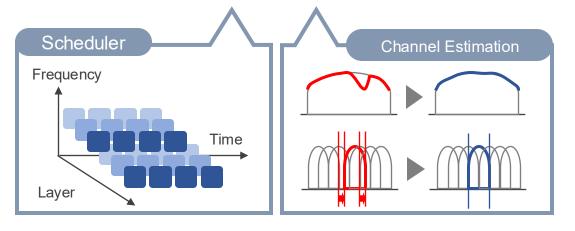
Radio Improvement (Al-for-RAN)

Individual Cell

Inter-Cell Coordination

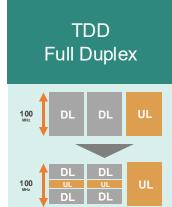












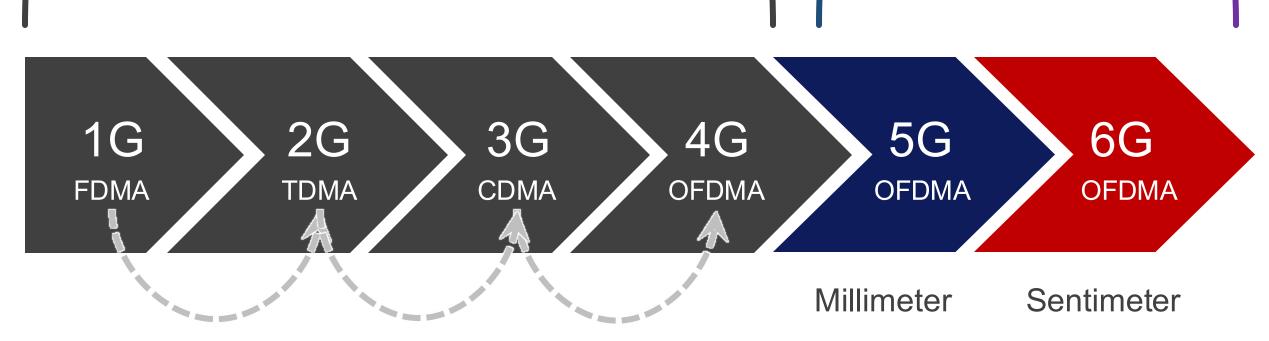
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Technological Advancements Limited with 5G

Technological Innovation

Frequency Expansion



10 years Cycle

15

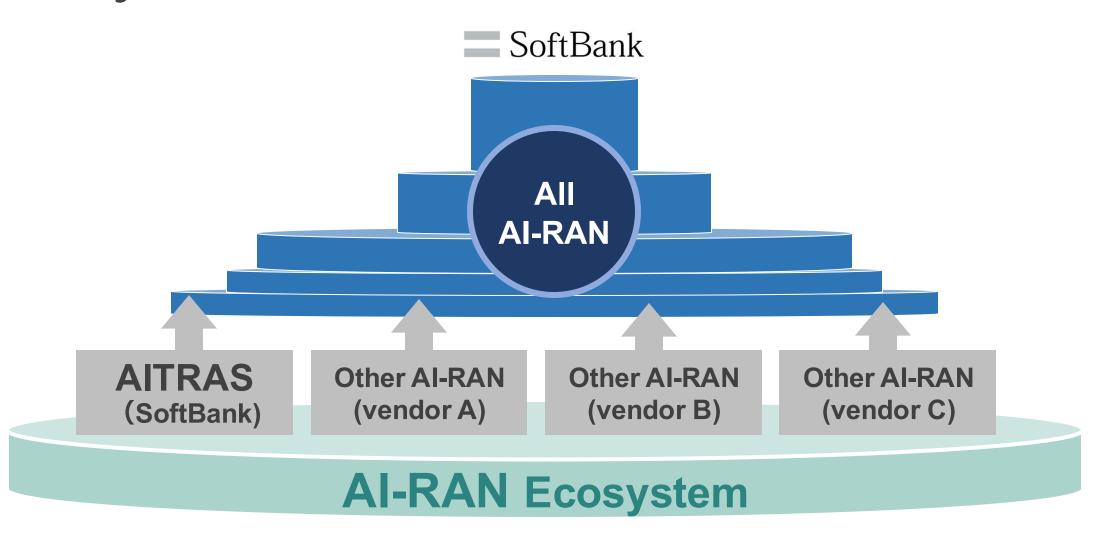


AI-RAN: Towards continuous evolution





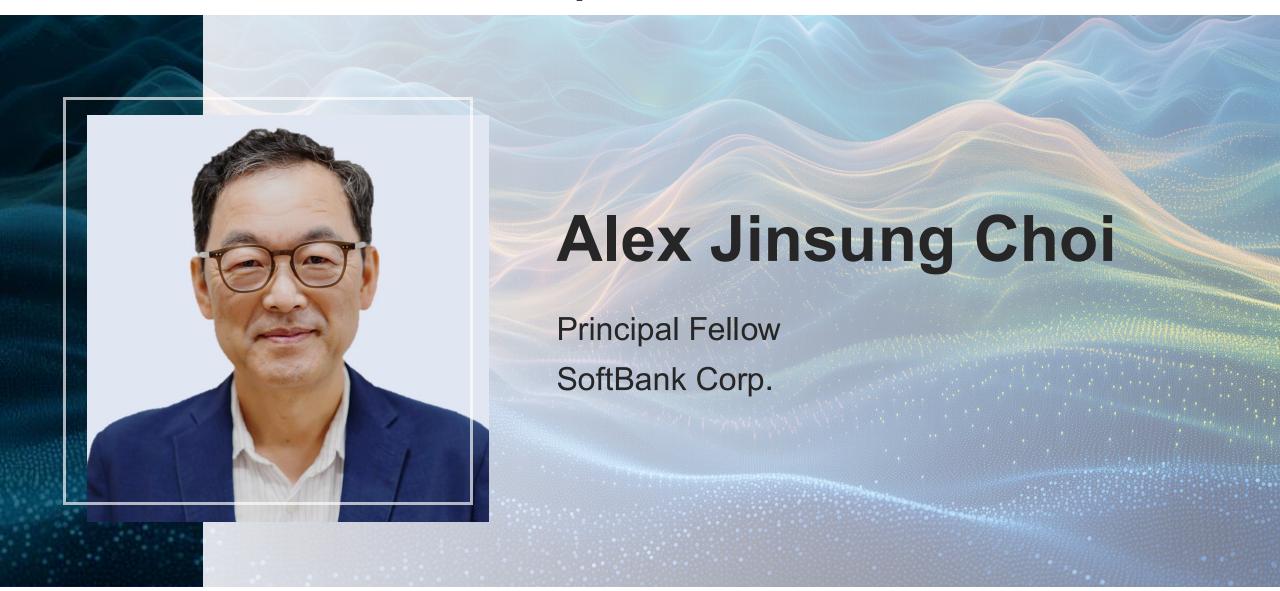
Beyond Full Al-RAN Transformation



SoftBank



Speaker



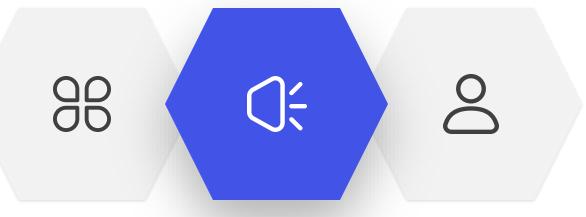




AI-RAN Definitions

Al-RAN (Artificial Intelligence Radio Access Network) refers to the integration of Al technologies into Radio Access Networks (RAN) to optimize network performance, automate resource management, and enable new Al-driven services. It transforms traditional network infrastructure into intelligent, adaptive, and revenue-generating platforms by leveraging Al-powered data analysis, prediction models, and automated control mechanisms.

AI-RAN integrates AI technology in RAN to improve mobile network efficiency.



Al-RAN Alliance include
Al-for-RAN,
Al-and-RAN,
Al-on-RAN.

- 1. Al-for-RAN: Enhancing network operations such as load balancing, interference management, and network optimization using Al/ML models.
- 2. Al-and-RAN: Sharing infrastructure to run both RAN and Al workloads, enabling parallel processing and creating new revenue streams.
- **3. Al-on-RAN:** Embedding Al capabilities directly into RAN components for real-time decision-making and operational automation.

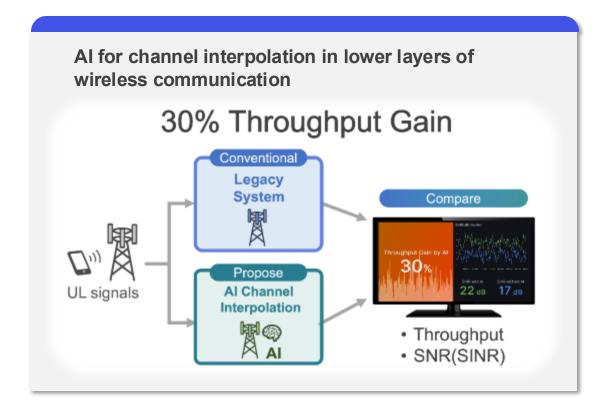
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Early Research and Al-RAN Development

Initial Efforts

SoftBank's early AI-RAN research focused on integrating AI-driven signal processing into RAN to enhance 5G performance. The research explored channel estimation, AI-based beamforming, and AI-powered traffic management to overcome network congestion and improve service reliability. The development of an AI-RAN data center enabled parallel operation of RAN and AI workloads, unlocking new revenue streams through AI-powered services.





Introduction of GPU based vRAN - gRAN



01

GPU for vRAN Evolution

GPUs handle highly parallel processing workloads efficiently.





02

Technological Leap with GPU-based vRAN

Enables modern AI services like LLM inferencing at the edge.

*NVIDIA GH200 Grace Hopper Superchip



Key Components of AITRAS

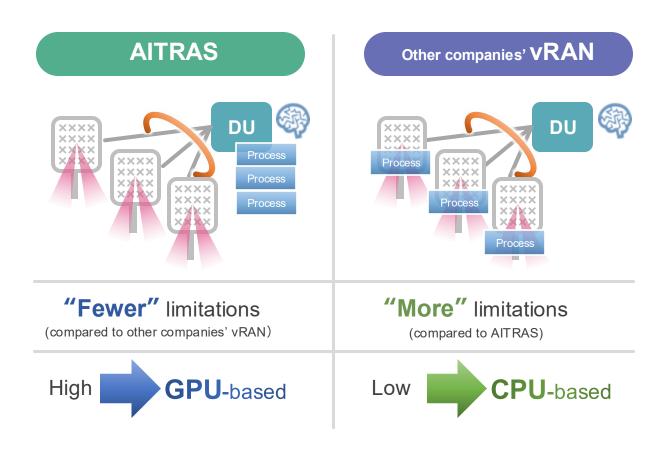
System Components and Architecture

01

Virtualization platform with RAN functions structured with L1, L2, and L3 layers.

02

Edge Al and orchestrator for dynamic resource allocation.





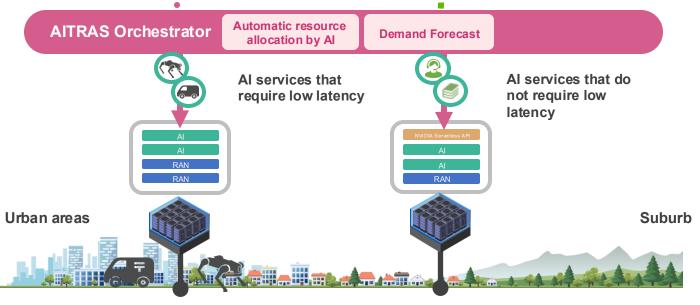
Al-Driven Orchestration

Al Orchestration Layer

Al-driven orchestration for managing and optimizing workloads.

Includes automatic resource allocation by Al and dynamic changes of server roles.

- Multi- Modal AI for remote autonomous vehicle support.
- Edge RAG for operational efficiency.
- Real- time robotic control with low- latency response.



Advanced

Unique Edge Al Services



Key Benefits of AITRAS

Core Benefits

Cost reduction through consolidated AI and RAN workloads.

Optimized resource utilization and new revenue opportunities.



Outdoor Testbed for AITRAS

Trial Results

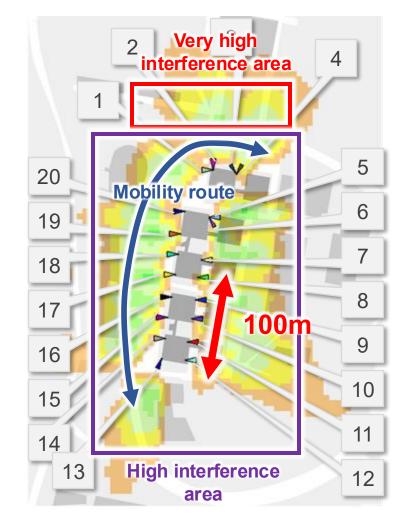




Carrier-grade stability in the AITRAS trial in Kanagawa.

Tested with 100 User Equipment streaming videos simultaneously.







AITRAS Performance Evaluation

Performance Metrics

Server's power consumption compared to current RANs.

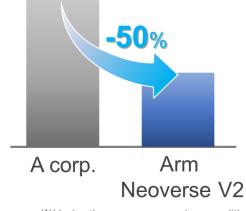
Evaluated UE accumulation and radio resource occupation.

Power rating 1,500W (CPU+GPU)





CPU power consumption*



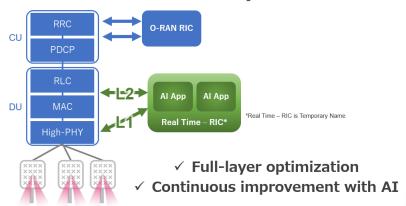
**Under the same processing conditions SoftBank lab measurement results



SoftBank's L1 Enhancements in AITRAS

RAN Performance Improvements

Enhancement by AI





GPU acceleration for higher processing speeds and throughput improvements.

Al- driven L1 optimizations for cell capacity and power consumption.

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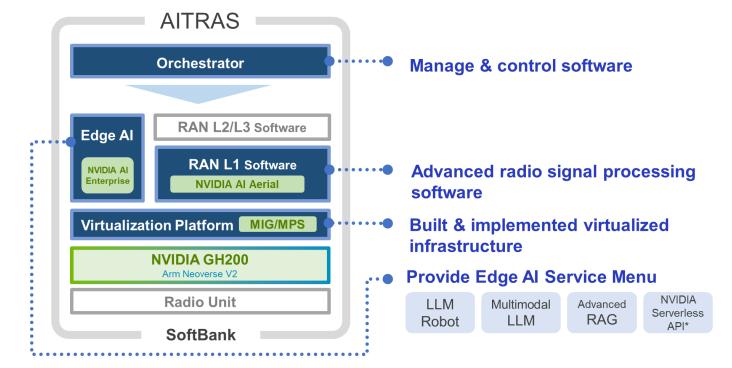


SoftBank Al-and-RAN Approach

Management Cluster and Workload Clusters for optimized placement.

Cluster level, node level, and core level resource management.

Virtualized Clusters



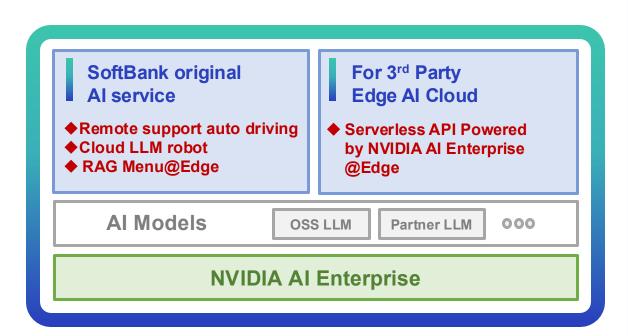


Use of CPU- GPU integrated servers for efficient handling of tasks.



Agentic AI – Serverless API powered by NVIDIA AI Enterprise

Serverless API Benefits





Supports flexible and scalable Al deployments.



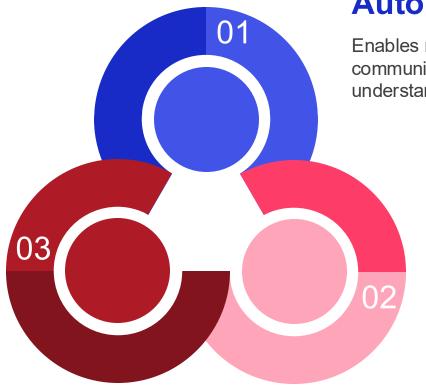
Interfacing with the Al-and-RAN infrastructure for optimized GPU usage.



Use Cases for the AITRAS Al-on-RAN

Retrieval-Augmented Generation (RAG) Chatbots

Use RAG chatbots for customer service and network support utilizing secure processing.



Autonomous Driving

Enables real-time V2X communication and multimodal traffic understanding AI.

LLM Robots

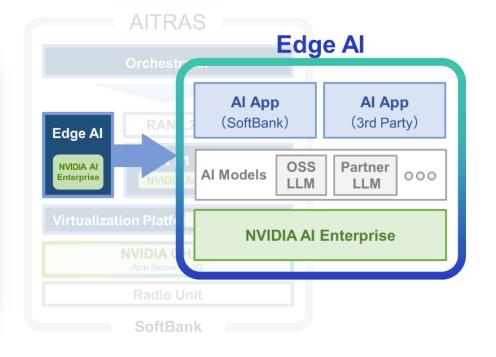
LLM robots perform with high- speed control LLM on AITRAS.



AITRAS Al-and-RAN for New Revenue Generation

Strategic Al Integration

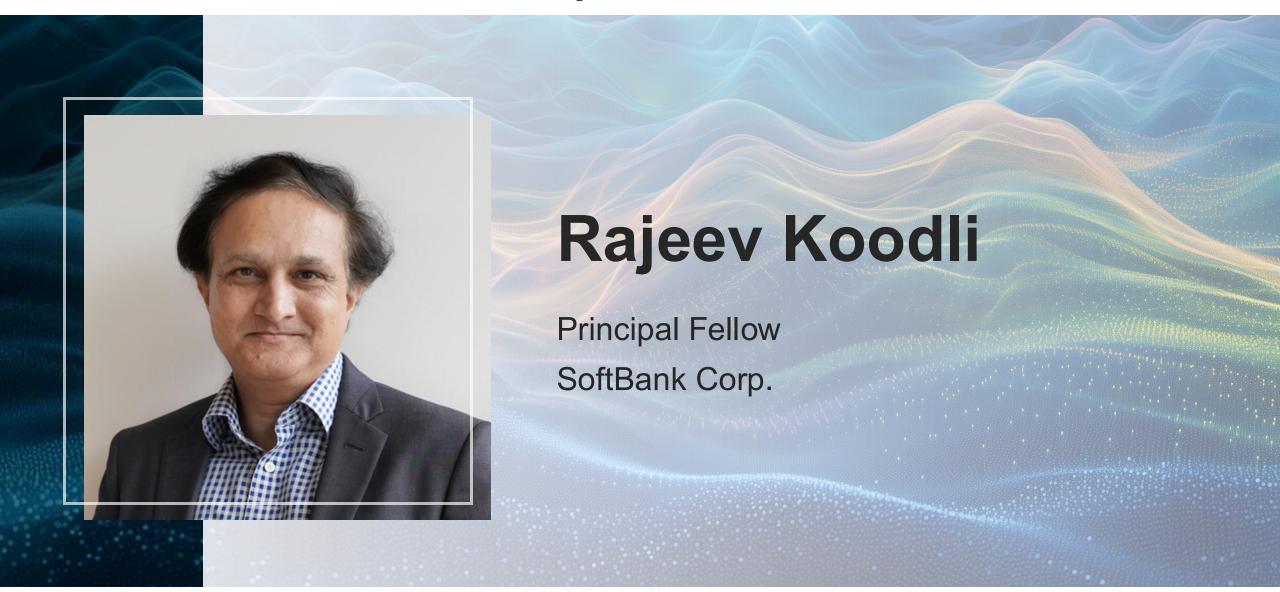
Edge AI inferencing and containerized infrastructure supporting digital transformation.



SoftBank



Speaker



Al for Telco Vision and Concept



Outline

- Introduction: Telco and Al Inflection point
- SoftBank Vision
- Use Cases: Human AI, Machine AI
- Al Pillars: Data, Infrastructure, Models
- Al Tool Kit, Operationalizing Al
- Next Steps & Conclusion



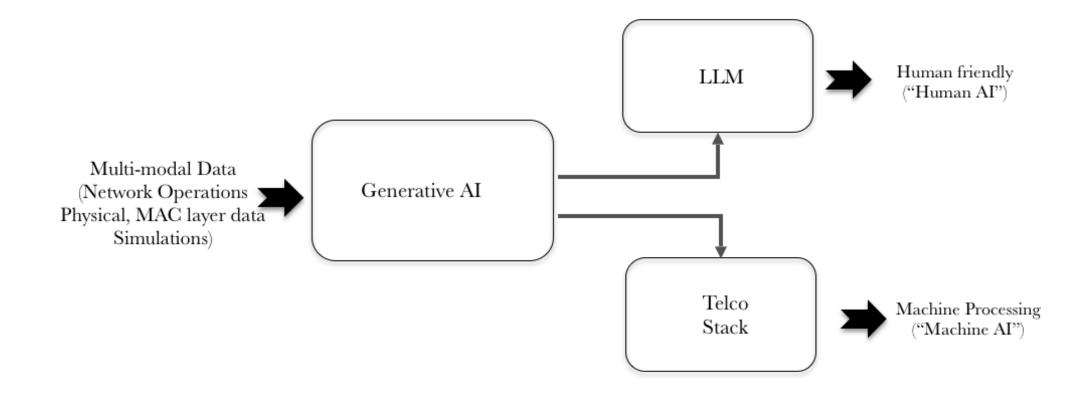
Telco and Al: Inflection Point?

	AI-RAN Alliance	O-RAN Alliance	3GPP & ITU	Global Telco AI Alliance
Key Initiatives	 AI-for-RAN: AI to enhance capability of RANs AI-and-RAN: Integration of AI and RAN processing infrastructure AI-on-RAN: Deploying AI services at the network edge 	 RAN Intelligent Controller (RIC): AI/ML-based solutions for traffic steering, energy savings, anomaly detection Open Software Development: Collaboration with Linux Foundation Global PlugFests: Testing and integration of O-RAN solutions 	 AI/ML for 5G and Beyond: Enhancements for beam management, positioning, energy savings Lifecycle Management: Frameworks for managing AI/ML models ITU AI for Good: Promoting AI to address global challenges. AI Standards: Framework for AI governance 	 Telco LLM Development: Alpowered customer service and network management Al Governance: Focus on ethical and sustainable Aldeployment Ecosystem Building: Integration of Al solutions into telco services and new business models
Outcomes	 Demonstrated feasibility of Alnative RANs in trials and pilot projects Blueprints and benchmarks 	 Standardization of AI/ML frameworks for RANs Acceleration of open RAN adoption 	Promoting AI standardization and governance in telecom networks	 Initial applications of Telco LLM in customer care and network optimization

Telcos and Industry Organizations are embracing Al

Developing and Deploying AI is necessary (prior to standardization)

SoftBank Vision

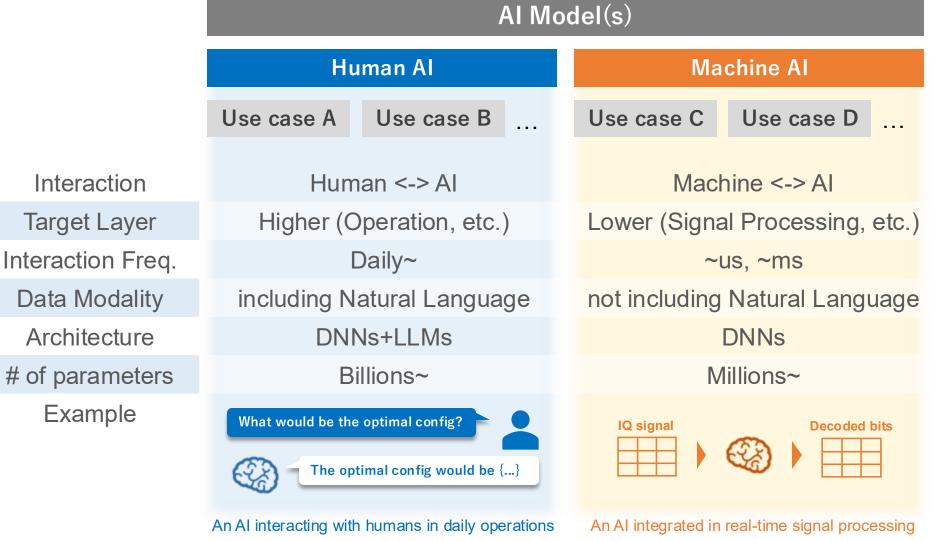


Generative AI for Telco:

- 1. Network Operations Efficiency (Opex)
- 2. Network Stack Processing (Capex)

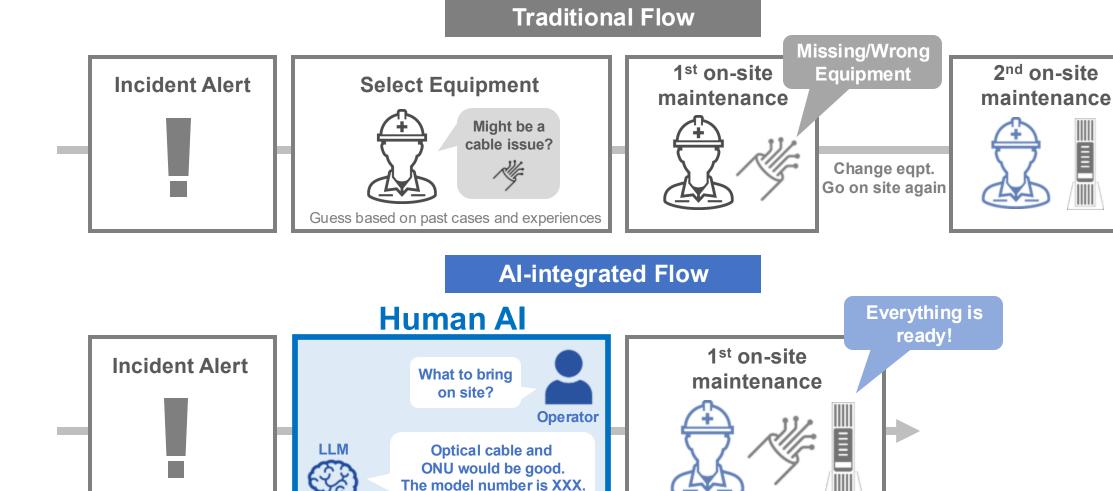


Human Al vs Machine Al





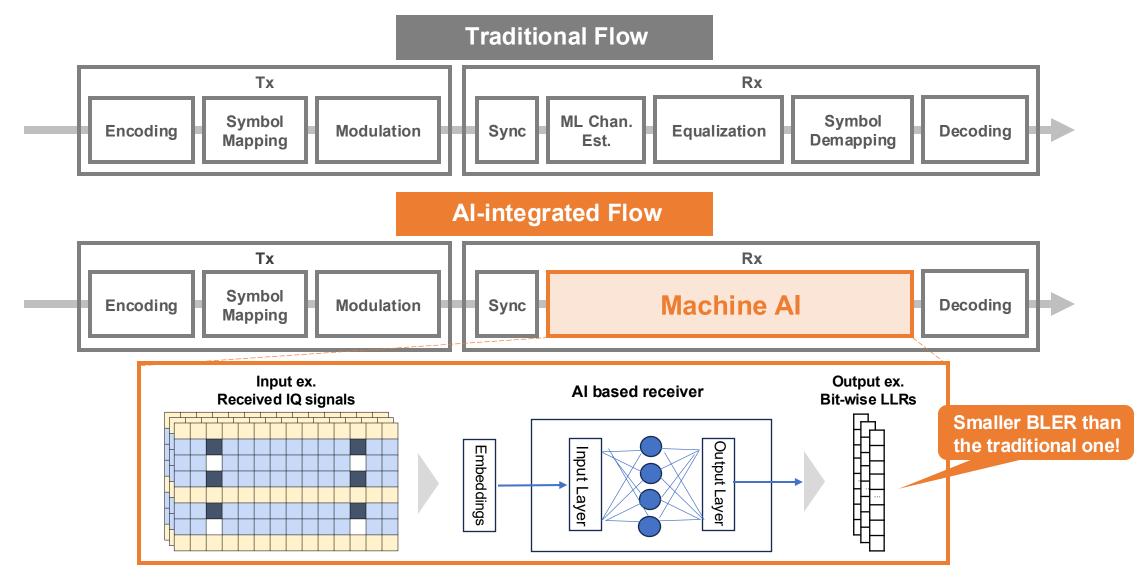
Human Al Use Case



Inference based on the hardware alarm, transmission-side behavior, actual data logs from multiple nodes, and past cases.



Machine Al Use Case





Al Pillars



Al Model End-End Pipeline

Data collection

Data processing

Training & Evaluation

Containerization

Serving as an API



- Collect necessary data without loss
- Design for different data granularity
- Confidentiality, integrity, availability
- Smooth export to machine learning cluster



- Import into databases
- SQL queries, ETL
- Data-grounded synthetic data generation with/without LLM
- Distributed processing on a data lake



- Supervised learning/reinforcement learning
- Distributed learning
- Job scheduling
- Hardware/job monitoring
- Evaluation using public/private datasets



- Model format conversion (ONNX, etc.)
- Integration with inference server
- Continuous updates of environment
- · Security scanning



- Deployment on cloud infrastructure
- Auto-scaling
- Observability
- · Performance optimization

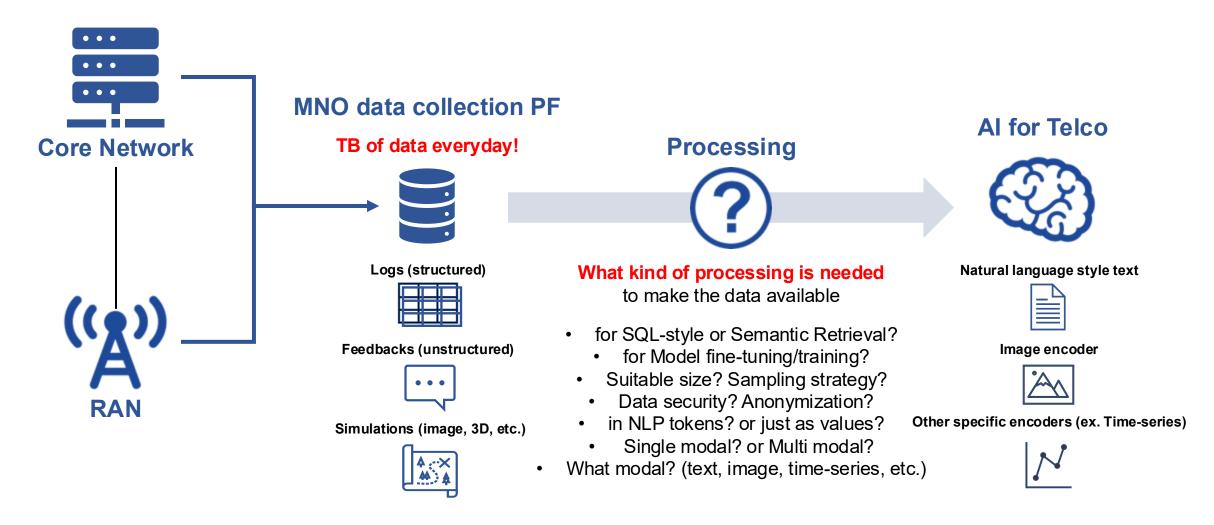
Loop

- Continuous updates with latest data
- Continuous performance improvement



Challenges with data

For all Al generation, we need to tailor the data (Data Processing)





Example Data Size

Total of ~4TB of Network Data

	15 minutes	hourly	daily
RAN	2024/05/07-2024/08/12 (1TB~)	2023/08/01-2024/08/12 (1TB~)	-
Core	2024/05/05-2024/08/12 (~50GB)	2023/08/01-2024/08/12 (~50GB)	-
Config	-	-	2024/07/16-2024/08/12 (500GB~ *zipped)

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Challenges with infrastructure

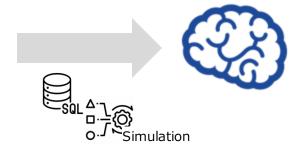
We need infrastructure to process the data and train the models



Huge raw data

Al ready data





Significant compute capacity needed to process the raw data

Training

Fine tuning



Full training



Computation resource required (Estimated, Unit: GPU Hour = 1 Hour spent on 1 DGX A100)
Based on LoRA Parameter Efficient Fine Tuning (PEFT)

Data size	1k cells	5k cells	10k cells	200k cells
7 days of data	42 GPU Hours	9x24 GPU Hours	18x24 GPU Hours	24x365 GPU Hours
30 days	8x24 GPU Hours	38x24 GPU Hours	75x24 GPU Hours	4x24x365 GPU Hours
365 days	91x24 GPU Hours	24x365 GPU Hours	3x24x365 GPU Hours	50x24x365 GPU Hours

GPU infrastructure is assumed for any kind of AI training

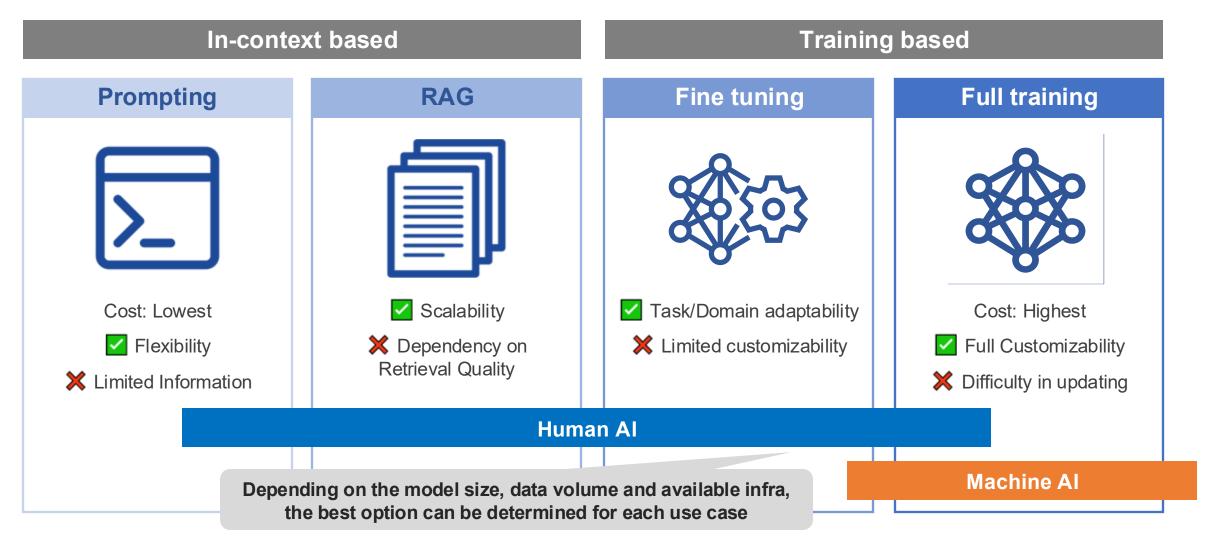
In LLM full training,
Data Center level GPU
infrastructure is needed



Al Tool Kit, Operationalizing Al



Al for Telco: Tool Kits



In-context approaches augment the base or finetuned models (without updating model weights)

Training Paradigm can start with Finetuning and move onto Continual Pretraining (and possibly Full Training)

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Operationalizing Al Models

Data collection

Data processing

Training & Evaluation

Containerization

Serving as an API



- Collect necessary data without loss
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- Import into databases
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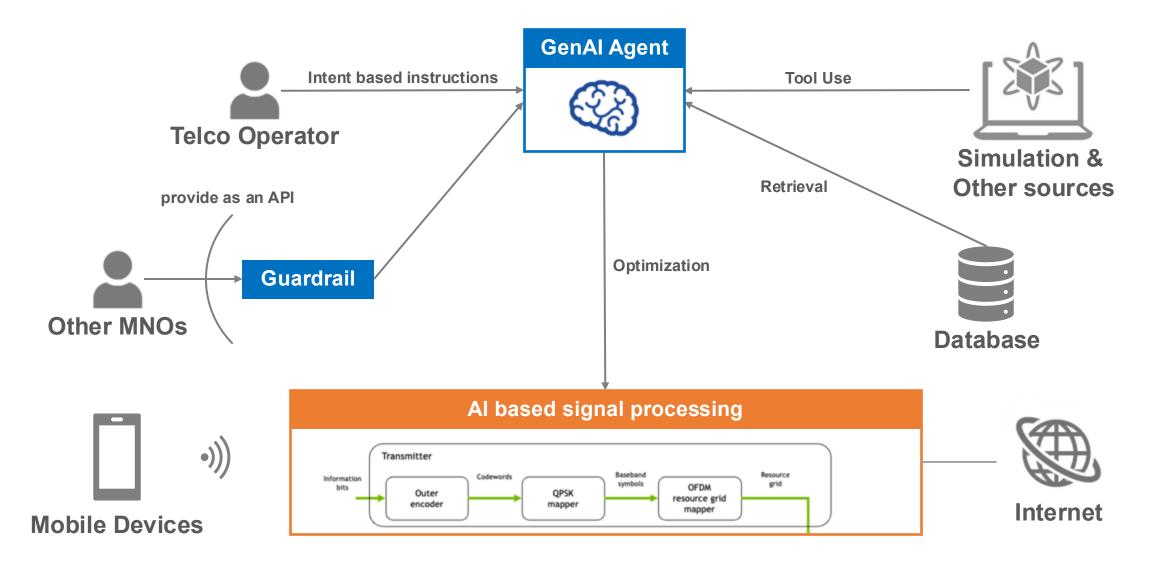
- Deployment on cloud infrastructure
- Auto-scaling
- Observability
- Performance optimization

Loop

- Continuous updates with latest data
- Continuous performance improvement



Towards Productionization



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Conclusions & Next Steps

- Several bodies actively pursuing Telco Al initiatives. 6G is seen as a pivotal moment for Al-native air interfaces (around 2030). Al-RAN Alliance is spearheading the industry activities.
- Human Al and Machine Al have significant potential in improving the operational expenses (opex) and fundamentally reshaping the capital expenses (capex) for Telcos.
- Telcos have enormous amounts of invaluable multi-modal data. However, creating programmatic and flexible data sets out of the voluminous data is daunting yet crucial.
- Creating Telco AI models with general expertise or finetuned for specific use cases requires significant amount of GPU computing infrastructure.
- Operationalizing Telco AI models follows the same trajectory as other AI models but the scale and new skill set needed for Telcos can be vastly different compared to the current practices.

SoftBank has embraced this journey of adopting Al for both its operational and infrastructure redesign purposes. We plan to further share our experiences in the near future.

SoftBank



Speaker



Al-for-RAN R&D Update



- Agenda -

What's Al-for-RAN?

- The Significance and Expectations of AI-for-RAN
- Target Goal
- SoftBank's Approach

Under Researching Themes of Al-for-RAN

- Reason why We Choose These Themes:
- Details of Each Theme status:
 - MU-MIMO Advanced scheduler
 - SRS Prediction
 - UL CH Interpolation

SoftBank Developed System Level Simulator

Development and Purpose of SoftBank-SLS

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What's Al-for-RAN?

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Al-RAN Alliance Key Focus Areas

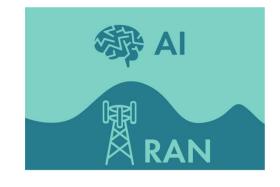
Al-for-RAN

Al for the enhancement of RAN



Al-and-RAN

Al and RAN sharing the same infrastructure



Al-on-RAN

Al applications enabled by RAN





Improving TCO and User Experience

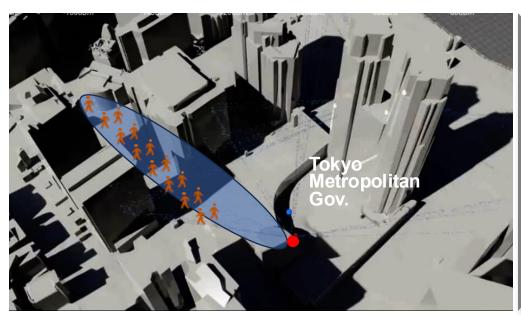
Biz Model Transformation

New Service Creation

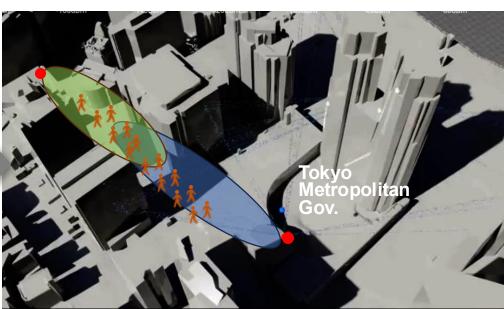


Improving TCO and User Experience

Usually, the improvement of user throughput has been achieved primarily through the expansion of deployed cell sites. This increase in the number of cell sites has led to an optimized expansion of coverage along user behavior and an increase in cell density, thereby enhancing the average user throughput.



Legacy Approach

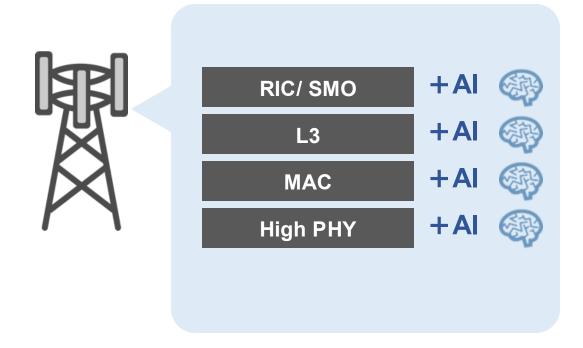


If SINR will be up (=Throughput up) without increasing number of Cell by Al-for-RAN, it is equal improving TCO.



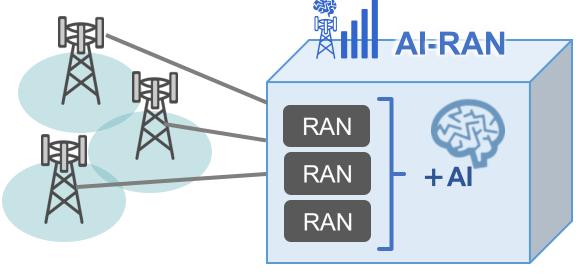
Target of Al-for-RAN

Individual Cell



Full Stack Al Acceleration

Inter-Cell Coordination



Enhanced MIMO



User Centric

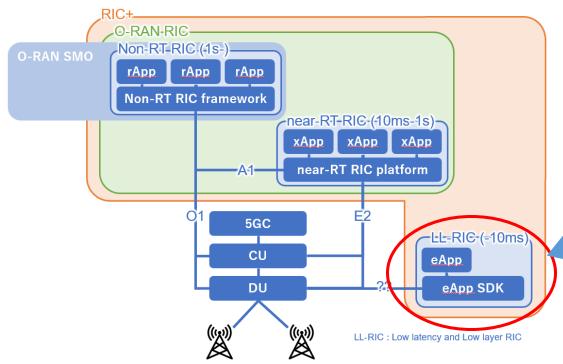


Multi Cell Drive



Approach of Al-for-RAN

- As a part of our initiative, we are starting with AI implementation at the lower layers.
- While the O-RAN Alliance is advancing the standardization of AI implementation at the upper layers by using Non-real time RIC and Near-real time RIC, there remains extensive research needed for the lower layers.
- However, this approach is extremely challenging and has a high failure rate.
- Through AI-RAN Alliance, we intend to explore diverse use cases and continuously expand them, promoting active utilization among our members.



This low layer part of RAN Intelligence Controller is ongoing research activity now. It has an opportunity to realize and extend new technologies of AI.

**LL-RIC and eAPP are defined by SoftBank, it might be different naming discussed within O-RAN Alliance.



Under Researching Themes of Al-for-RAN

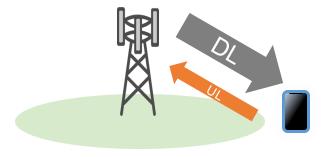
- Reason why We Choose These Themes:
- Details of Each Theme status:
 - MU-MIMO Advanced scheduler
 - SRS Prediction
 - UL CH Interpolation
- Evaluation results: MWC'25



Reason why We Choose These Themes

UL Data included ACK/NACK transmission opportunity has limited

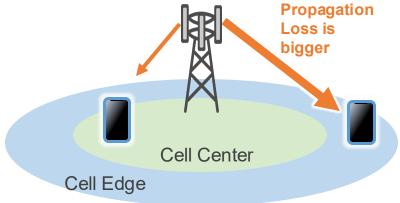
mMIMO performance equal NW Capacity!

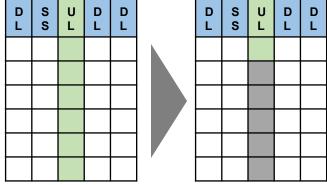


TDD frame configuration in Japan DL:UL=7:2

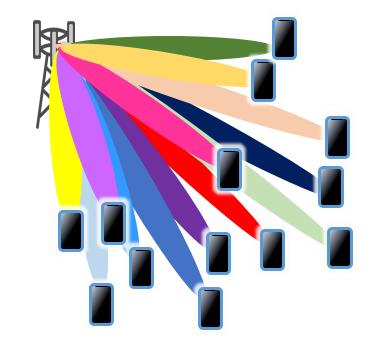
For Improving TCO and

User Experience



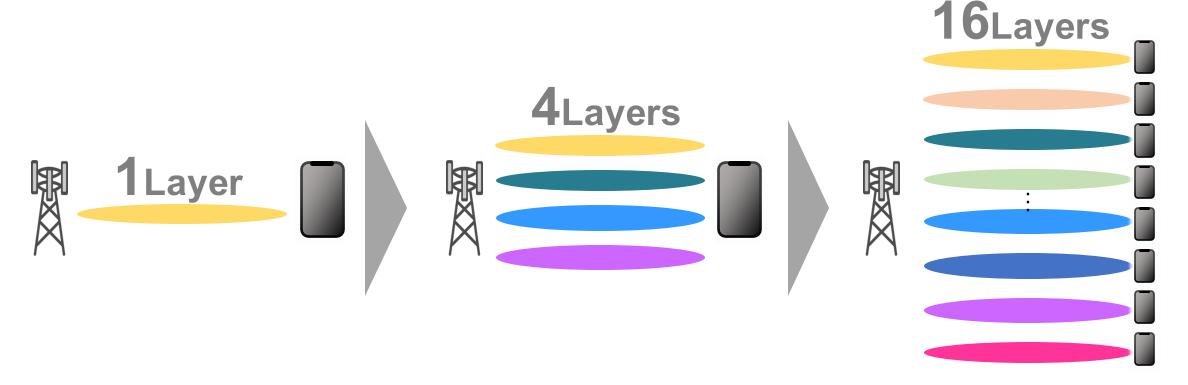


Reduced PRB # for power concentration



MU-MIMO is more important feature for higher density population area

Theme1:MU-MIMO Advanced Scheduler Necessity

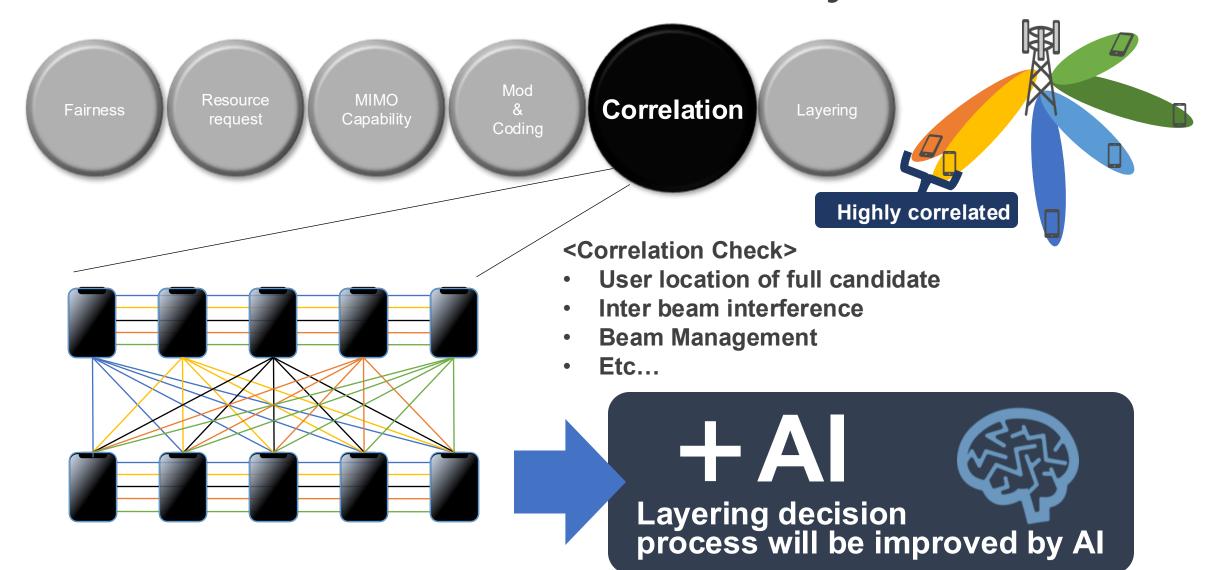


without MIMO case, resources will be shared.1data stream only

- Single User-MIMO is effective for
 The Peak User Throughput
 4data stream at DL by smartphone
- Multi Users-MIMO can achieve
 Peak Cell Throughput
 16data stream at DL
 NW Capacity improved!



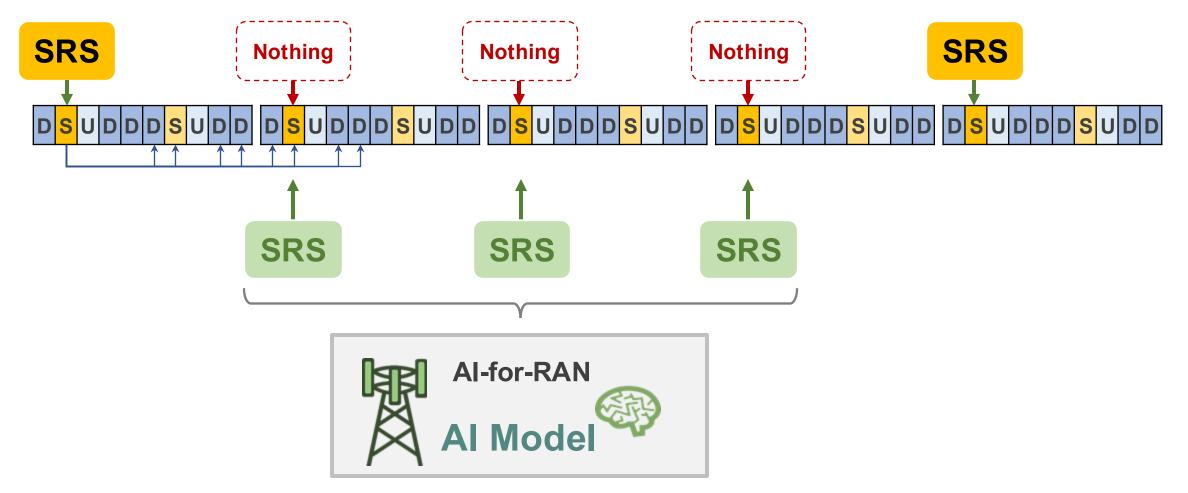
MU-MIMO Difficulty





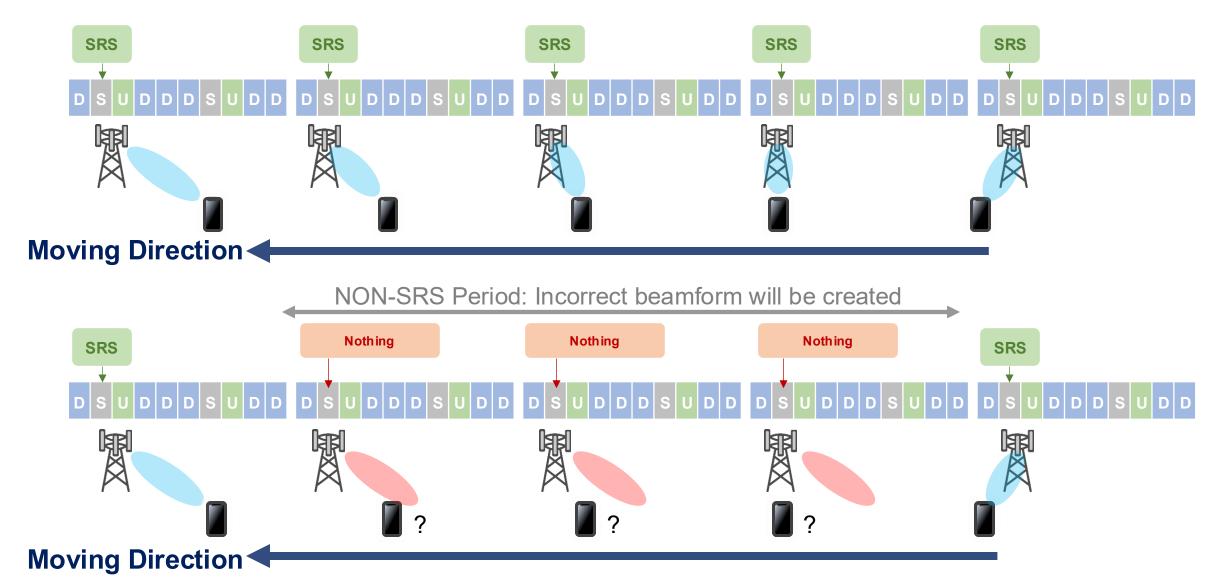
Theme2: SRS Prediction Necessity

Non-SRS Period: Beam Forming Accuracy will decrease





Beamforming Track Accuracy



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Theme3: UL Performance Improvement Necessity

Desired Signal

Interference+ AWGN

SINR

Doppler frequency shift occurs

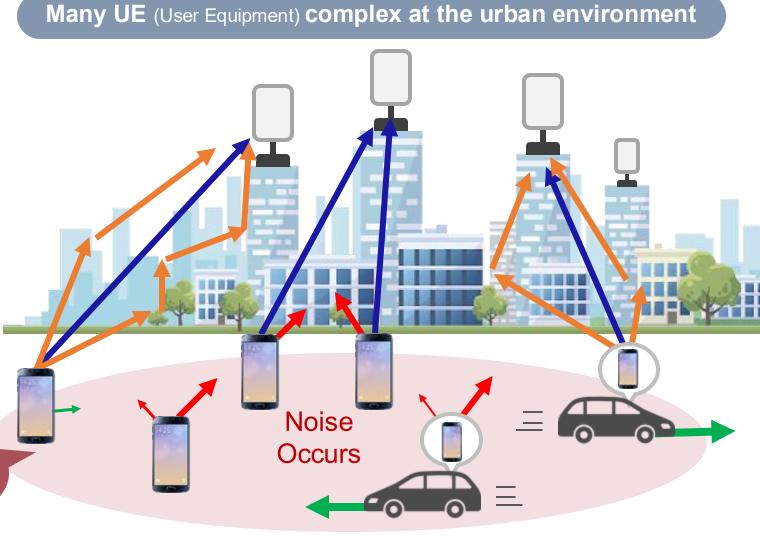
Delayed signal due to multipath

Desired Signal

Other Signal

Moving direction

Delayed signal



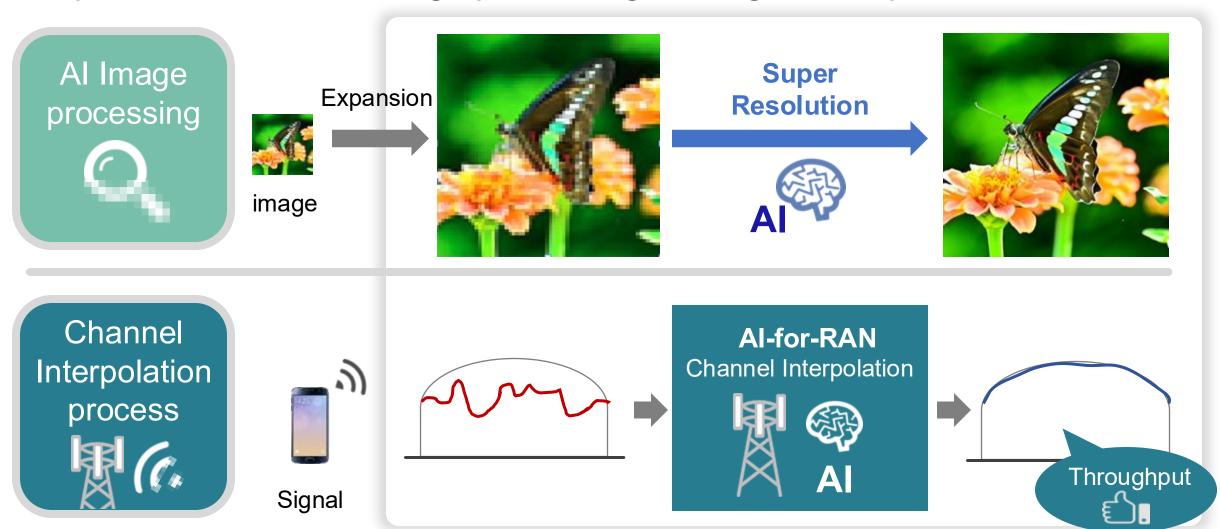
68

SINR =



Applied CNN Algorism

Super resolution in Al image processing and signal interpolation are similar

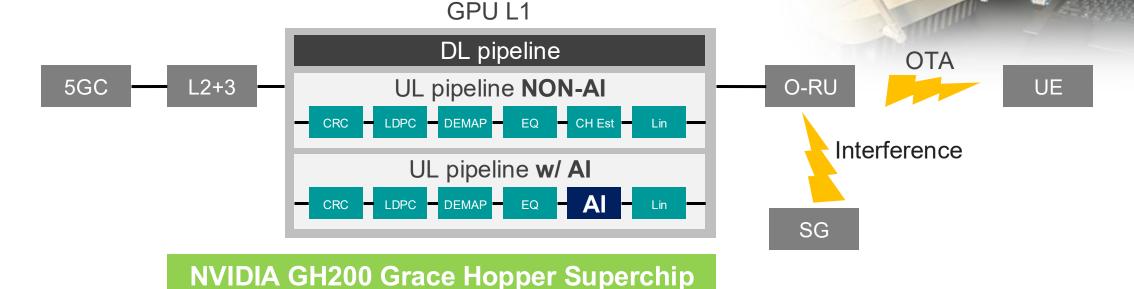


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Lab Test Environment

- Built a lab environment by implementing CPU Layer 2+3 and GPU Layer 1 on NVIDIA GH200.
- Modified the GPU Layer1 stack to embedded AI capabilities.
- Evaluated Uplink Channel Interpolation with AI on this circumstance.





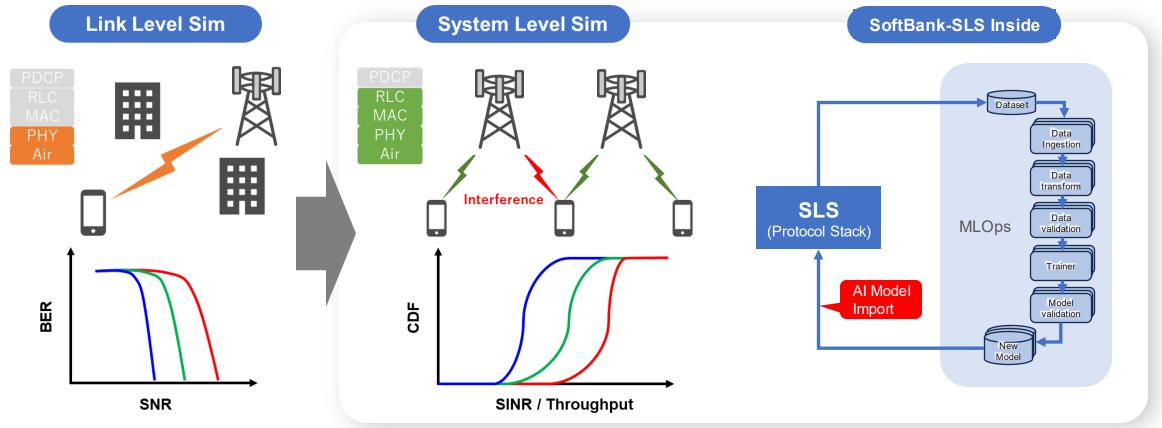
SoftBank developed System Level Simulator

- Development and Purpose of SoftBank-SLS
- Features and Implementation Details: MWC'25



System Level Simulator Necessity

- Why need to develop the SLS (System-Level Simulation)?
 - Link Level Simulation is involved 1 on 1 connection by Layer 1 only. It can NOT estimate neighbor cell interference, Multiple UE conditions and etc with PHY, MAC, RLC, and application layers protocol stack.
 - And supported AI model porting with MLops inside of SoftBank-SLS.



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MWC'25

Please keep an eye out and look forward to more exciting announcements on Al-for-RAN at MWC!

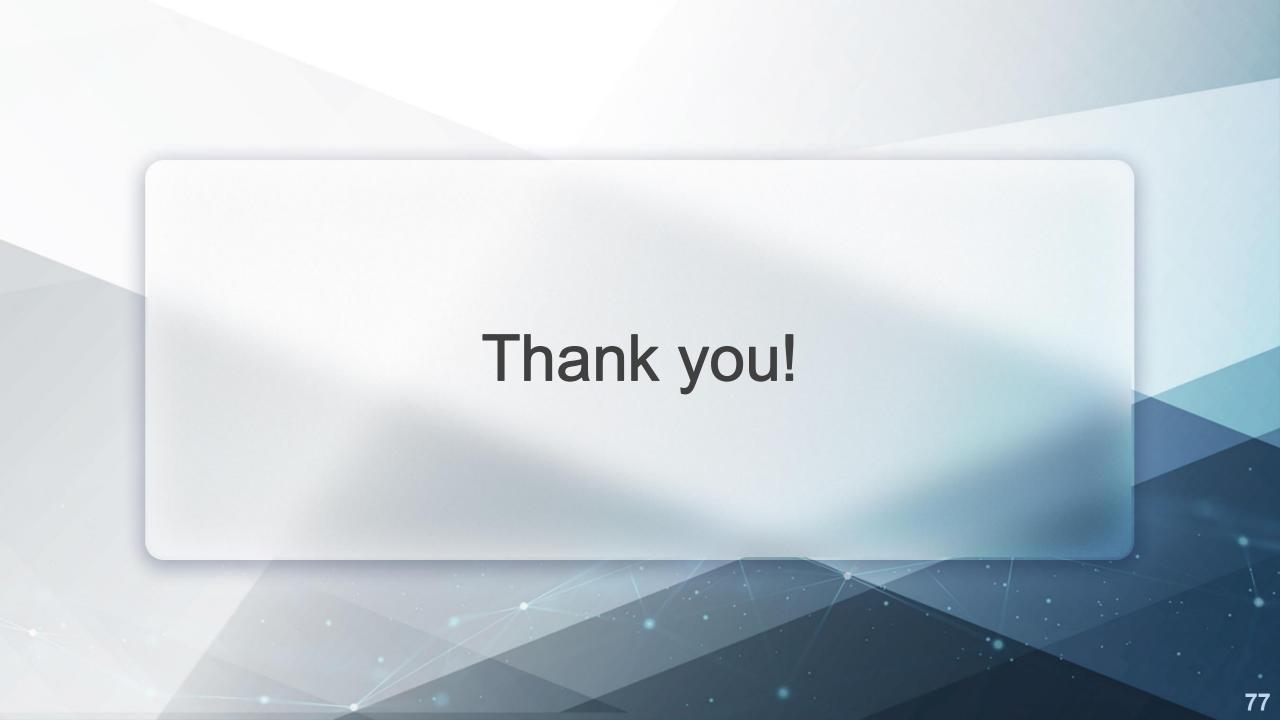
SoftBank





Visit our website for the latest updates on SoftBank's Al-RAN initiatives!

https://x.gd/SoftBank_RIAT



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