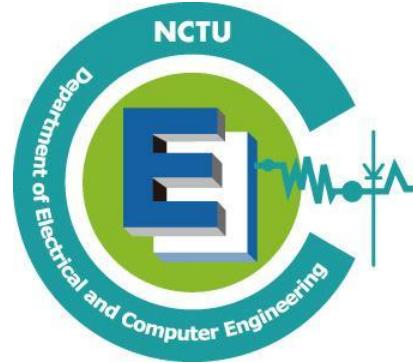




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Lecture 5 : Radio Maps for 6G

Class GitHub: https://github.com/NYCU-ICCLab/sionna_introduction

NVIDIA website: <https://developer.nvidia.com/sionna>

Sionna official tutorial: <https://nvlabs.github.io/sionna/index.html>

Sionna GitHub: <https://github.com/NVlabs/sionna>

Physical-Layer tooling, isn't Matlab enough?

- **Need for 6G-ready tools:** As the industry moves towards 6G, researchers require simulation frameworks capable of handling unprecedented modeling accuracy and scale.

- **Integration of AI/ML:** There's a growing need for link-level simulators with native machine learning integration and automatic gradient computation.

- **Environment-specific simulations:** Many 6G technologies demand simulations of specific environments with spatially consistent correspondence between physical location, wireless channel impulse response, and visual input.

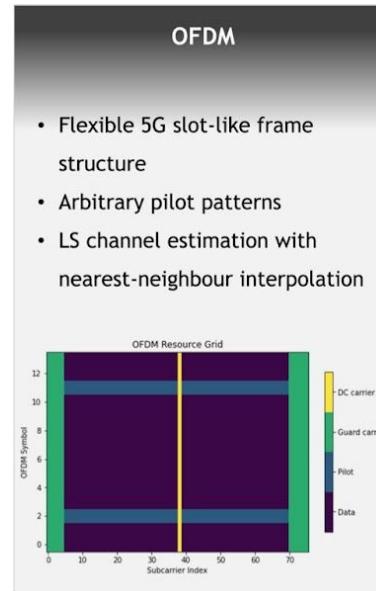
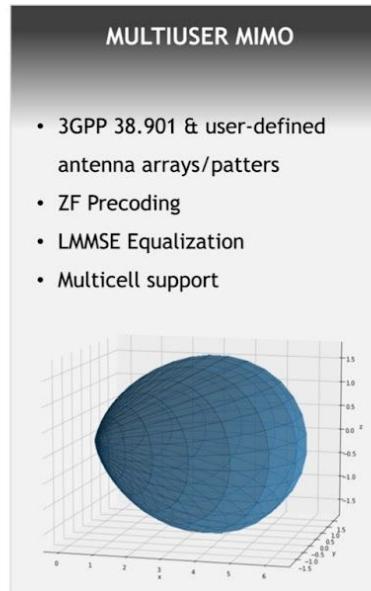
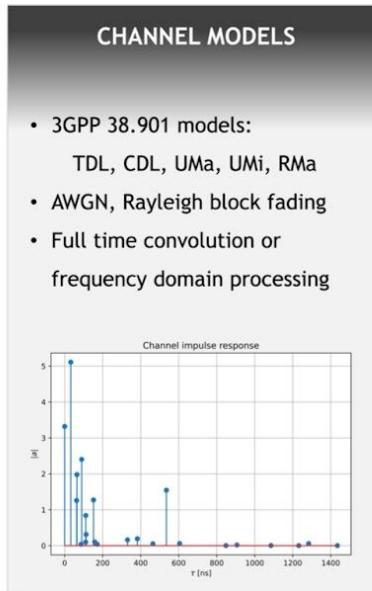
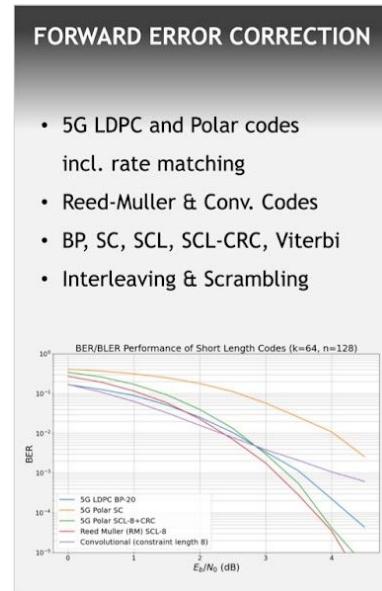


courtesy : NVIDIA Sionna

- **Computational efficiency:** The complexity of modern wireless systems necessitates GPU-accelerated simulations to enable interactive exploration and rapid prototyping

Importance of PHY Layer simulation

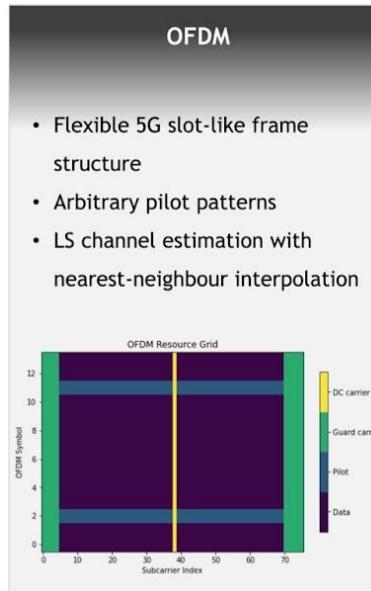
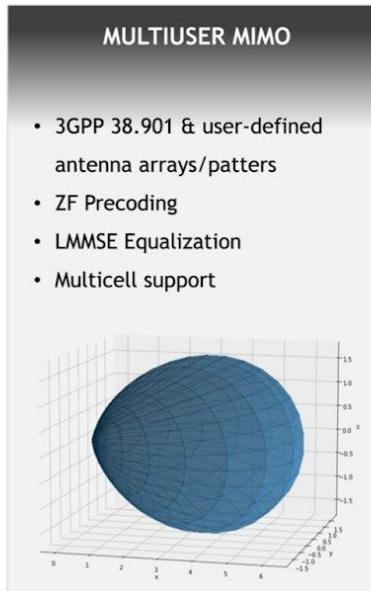
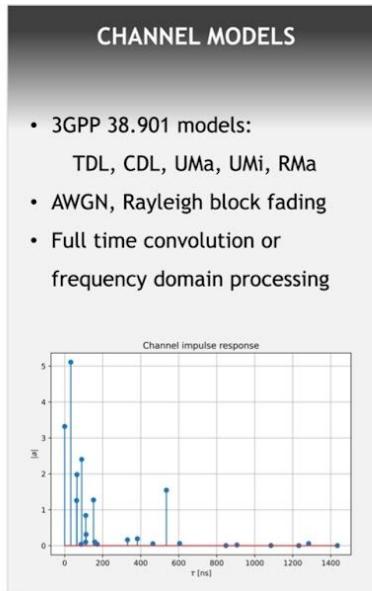
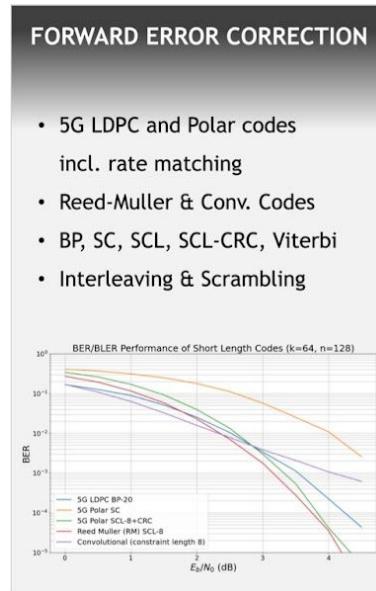
- **Open-source and accessible:** As researcher and student, the needs for an accessible tools becomes important → As an open-source library, Sionna democratizes 6G research, allowing researchers to focus on impactful and reproducible work.
- **Differentiability:** Sionna's differentiable architecture enables gradient-based optimization and seamless integration of neural networks into communication systems.



courtesy : Sionna Documentation

Importance of PHY Layer simulation

- **Modular design:** The framework's modular and extensible nature facilitates rapid prototyping of complex communication system architectures.
- **GPU acceleration:** Sionna leverages NVIDIA GPUs for orders-of-magnitude faster simulations, enabling interactive exploration of complex systems.
- **Integrated research platform:** It combines channel-, link-, and system-level simulation capabilities with machine learning support

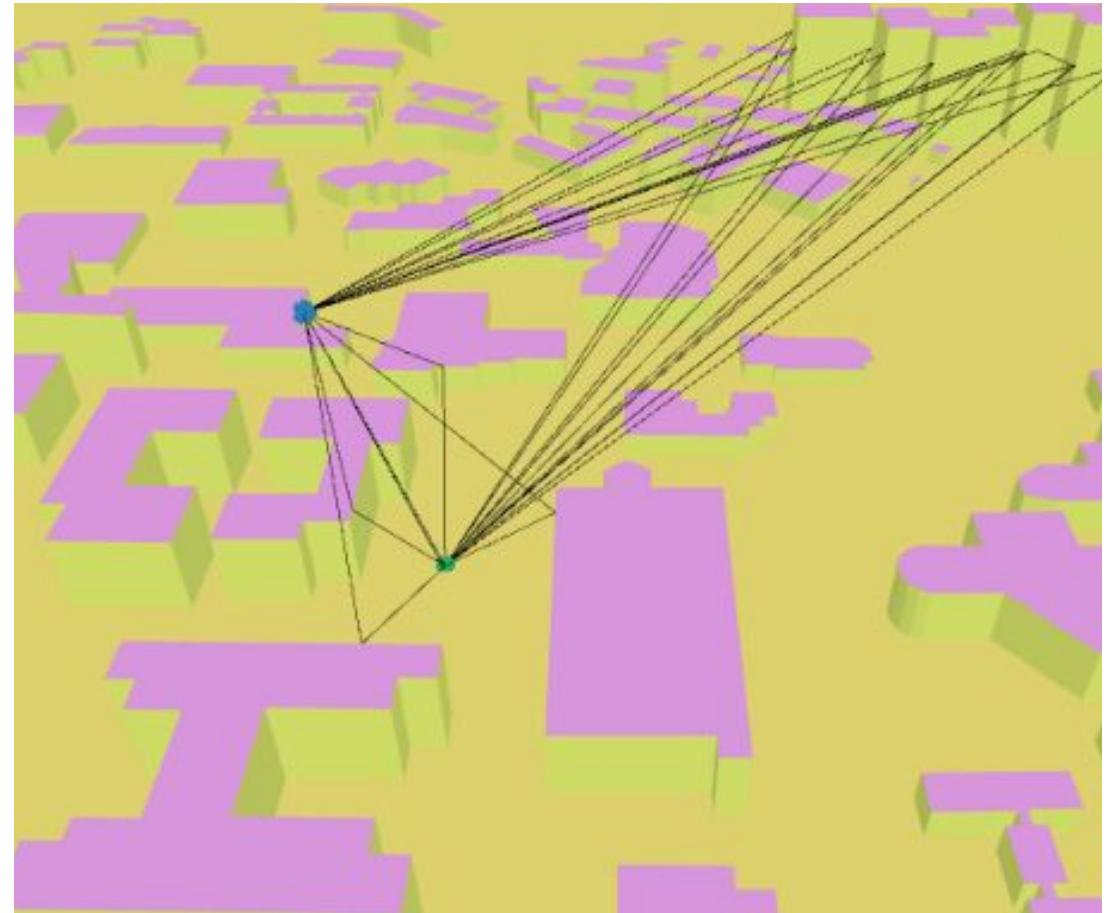


courtesy : Sionna Documentation

What is Sionna?

- **Open-source** Wireless Communication Simulation Framework
- Built upon **TensorFlow**, designed for developing, analyzing, and validating wireless communication algorithms
- Particularly suited for research and development in modern wireless standards such as **5G/6G**

Video :
<https://youtu.be/cYUNE4i4Q4E>



Simulation Accuracy

1. **Digital Twin fidelity:** Research comparing Sionna to Wireless InSite (WI) and real-world data from the FLASH dataset shows that Sionna's accuracy generally improves as the fidelity of the digital twin increases.
2. **LOS vs NLOS scenarios:** Sionna's performance varies between Line-of-Sight (LOS) and Non-Line-of-Sight (NLOS) scenarios, with accuracy trends slightly outperforming WI in certain parameters.
3. **Beam selection accuracy:** When evaluating the top 10 beams, Sionna's accuracy in matching real-world data ranges from 54% to 82% depending on the simulation fidelity.
4. **Comparison metrics:** Various metrics such as Magnitude Difference (MD), Cosine Similarity (CS), and Normalized Discounted Cumulative Gain (NDCG) show high similarity between Sionna and other simulators, particularly for higher-fidelity simulations.
5. and many other ongoing development: As an actively developed tool, Sionna's accuracy is likely to improve over time, with future extensions such as integrated ray tracing promising even more precise simulations

Application Scenario Example

For first two : **Digital Twin fidelity** and **LOS vs NLOS scenarios** we could show some applications scenario



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3D Channel GPT :
Application of Large Language Model for
Signal Quality Estimation

NYCU Campus Case Study

Members
邱佳詮、陳正根、賴德強、廖秉豪

Advisor
王蒞君 國立陽明交通大學 電機學院

ICCL Intelligent Communication and Computing Lab



Background &

Motivation



UAV(Unmanned aerial vehicle) as aerial base station

→ 3D spatial **air-to-ground** communication

Current limitation :

- Traditional communication model is for simple environment
- **Mobility** and **obstacle** in UAV communication scenario

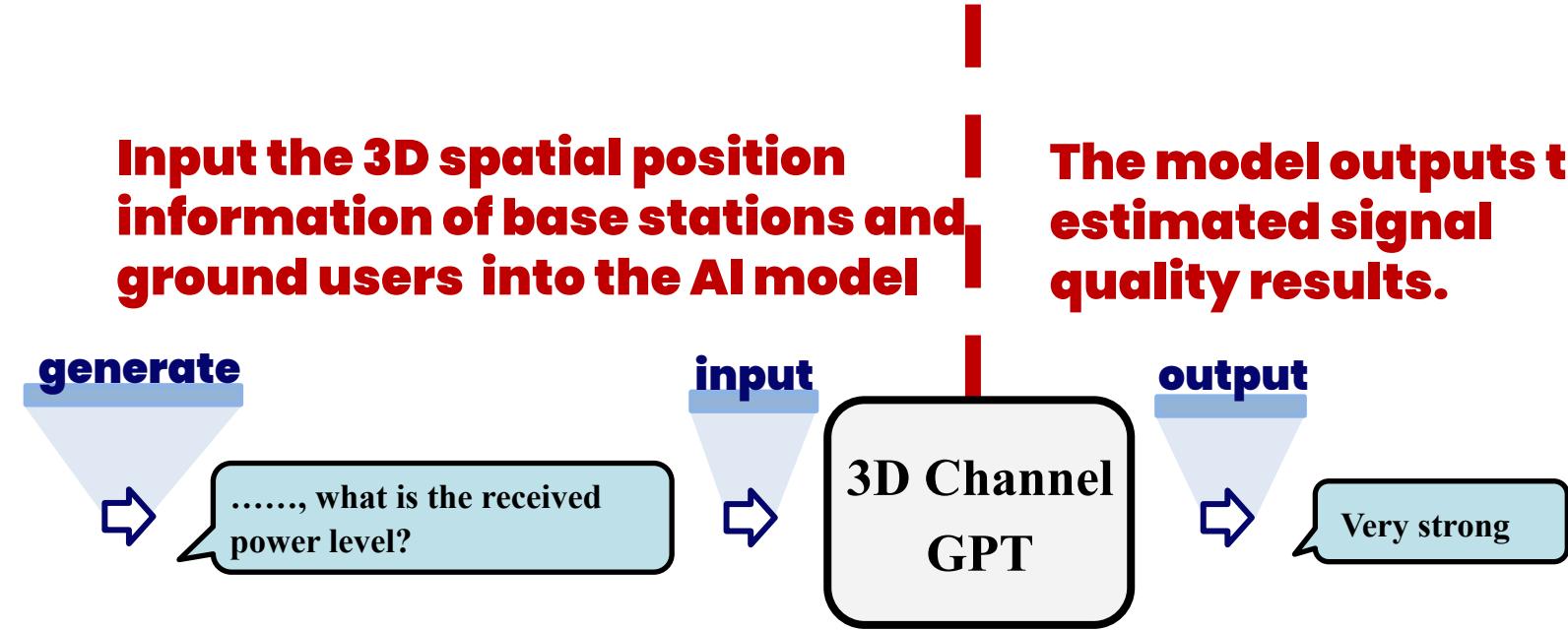
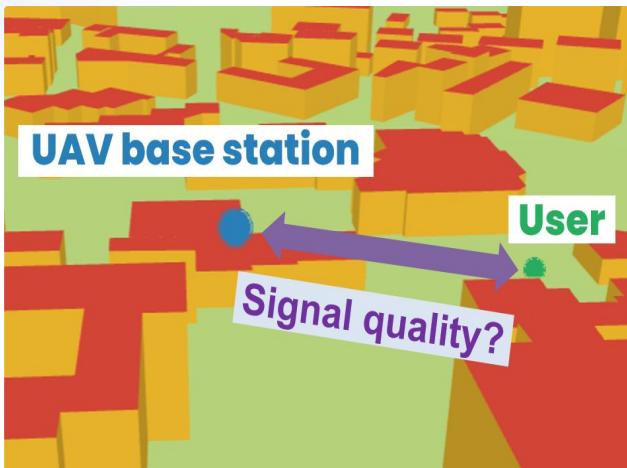
→ The estimation of signal quality becomes more **complex** and **challenging**

Objective : To understand the signal quality for users in different areas within an air-to-ground communication system

Signal Quality Estimation Framework Based on Large Language Model

- Fine-tuning the Large Language Model (LLM)
- Enabling users to understand the communication environment through text-based queries and answers
- Applying and validating this framework in NYCU campus

What we do? – 3D Channel GPT

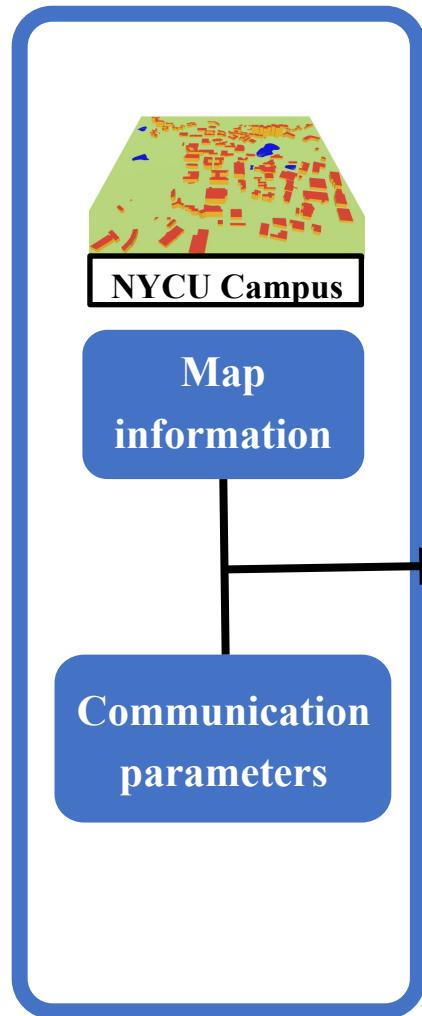


- We create **3D Channel GPT**, which provides **user-friendly Q&A dialogue service** to assess signal quality in **site-specific** 3D air-to-ground communication by extending DeepMentor solution.

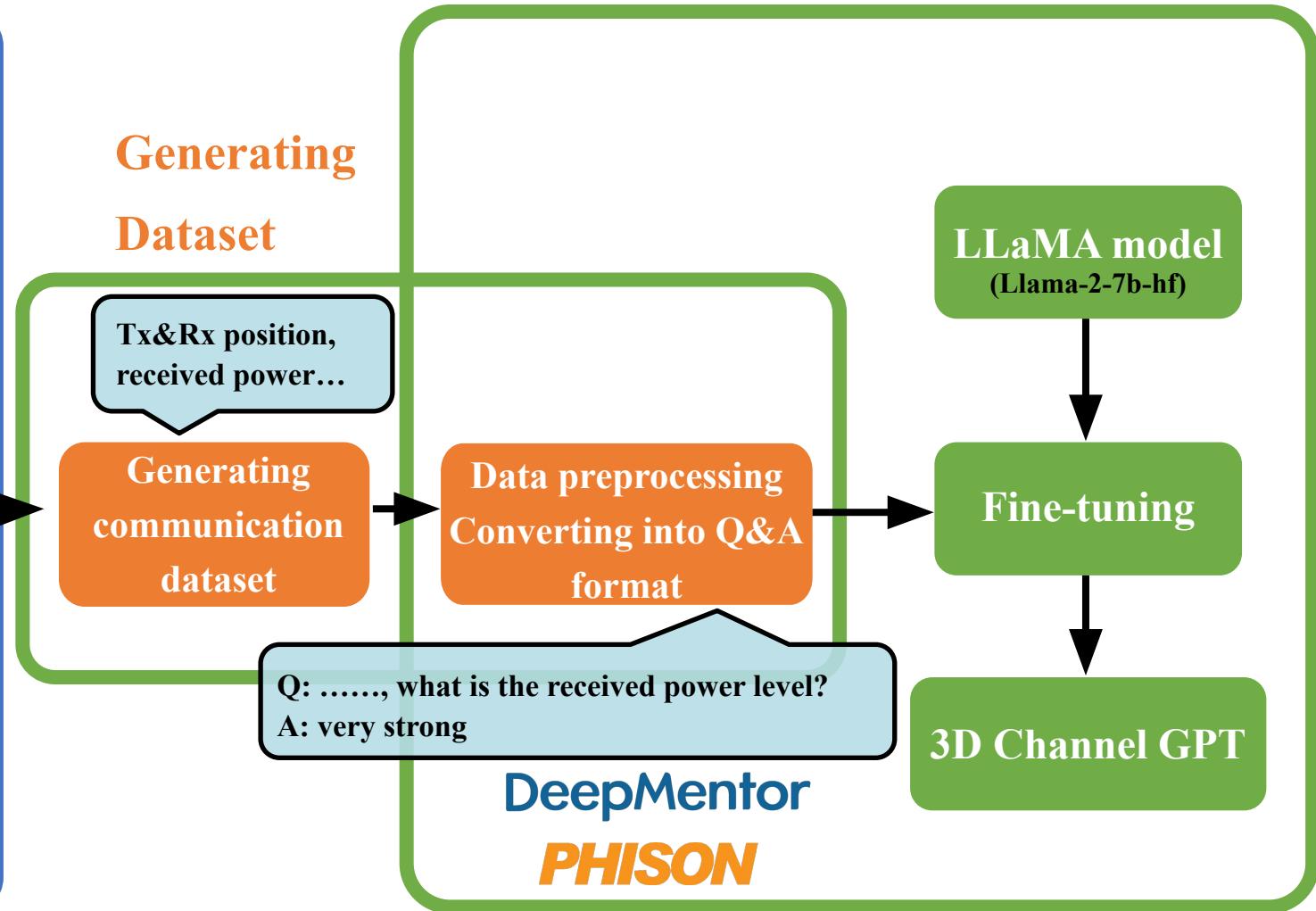


3D Channel GPT Framework

Communication Environment



Model training



3D Channel GPT: Workflows



Environment and dataset construction

- Radio Map Simulation
- Dataset Collection
- Dataset Preprocess



Model training

- LLaMA
- 3D Channel GPT



Model validation

- Case 1
- Case 2
- Conclusions

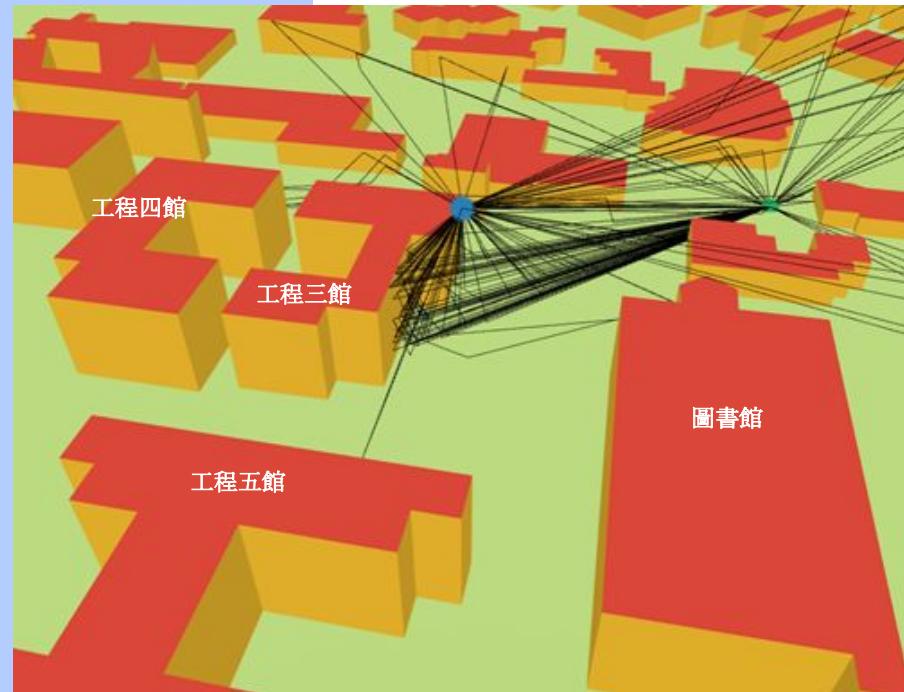


Environment and dataset construction

The First Open Data Generated Radio Map@ NYCU Campus



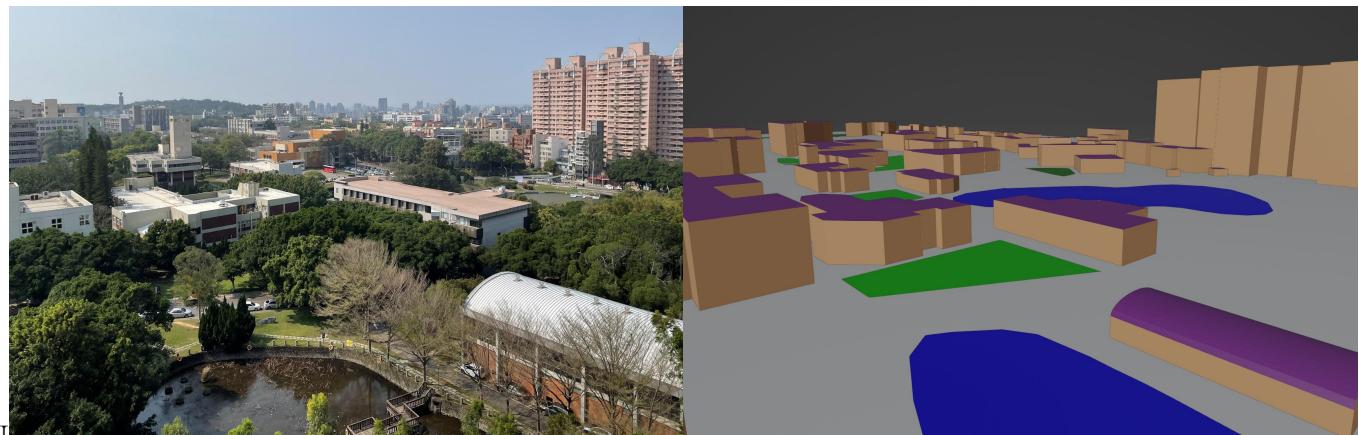
國立陽明



▲ Simulation of the Communication Environment of NYCU Campus using Ray Tracing

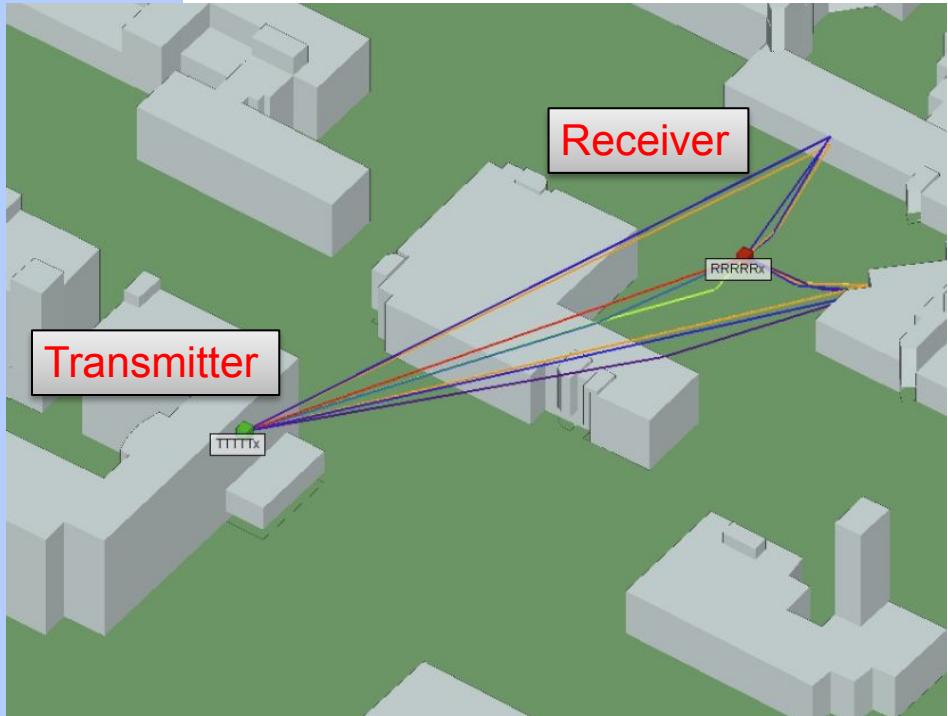
	Traditional Method	Our Method
Map information	Required On-Site Measurements	Open-Source Information
Ray tracing communication simulation	\$\$\$\$	Free (Open-Source)
AI support	Low	High

Our team has developed a simulated communication environment for NYCU campus by integrating open-source map data and employing Ray tracing technology.

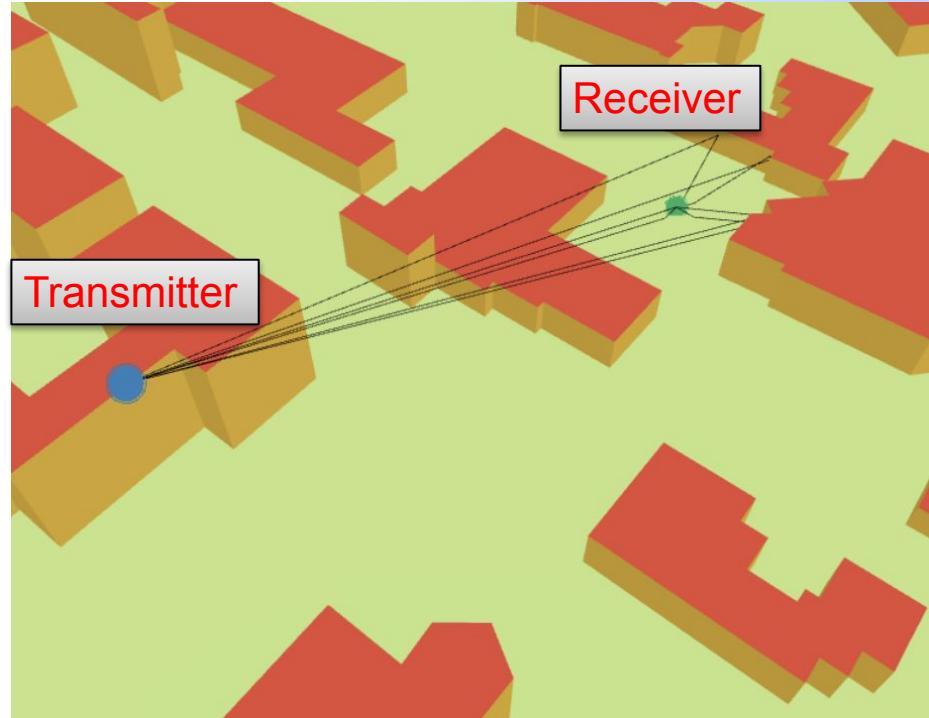


▲ Comparison between NYCU Real World and Rendered Scenes

Comparison : Traditional Radio Map vs. Our Approach



▲ Simulation of Traditional Method

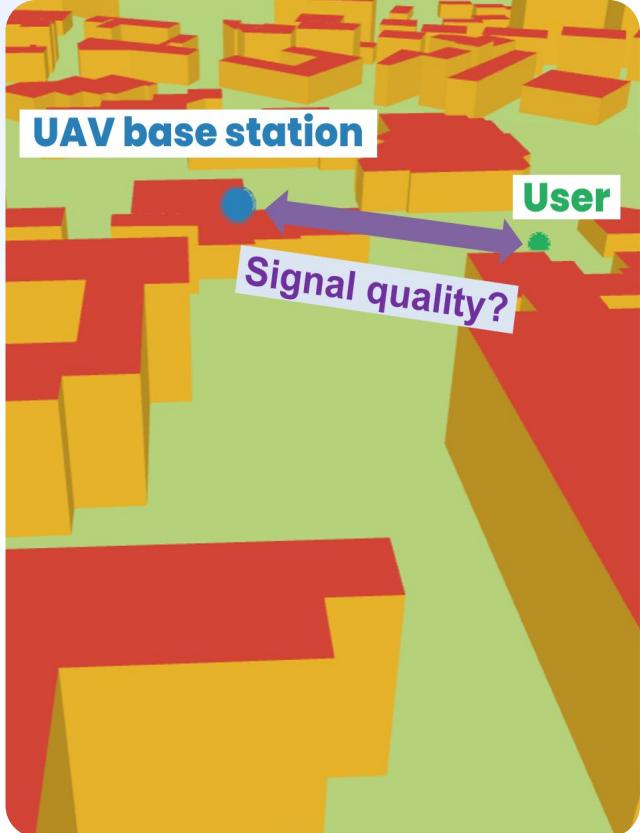


▲ Simulation from our Method

	Traditional Method	Our Method
Path loss	84.346 dB	84.377 dB
Computing time	50.506 s	0.268 s

Similar Results
Faster Computing Time

Environment-Aware Radio Signal Map by Ray Tracing



▲ Wireless Communications' Simulation Parameters
based on UAV and User Locations

Wireless communication parameter setting

Parameters	Description
UAV Position	(0, 0, 100)
Carrier Frequency	2.65 GHz (4GLTE band 7)
User Position	<u>Random</u>



Randomly Sample
10,000 Data Points



Record
User Location
Distance (UAV-User)
LoS / NLoS
User's Received Power

Classify Radio Signal Strength to Signal Quality



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Opinion Original Data

Base_station_ID	Tx_x	Tx_y	Tx_z	Rx_x	Rx_y	Rx_z	Distance	Frequency	LOS_Flag	Transmit power (dBm)	Received power (dBm)
Base_Station_UAV	0	0	100	-183.333	7.265218	1.5	208.2448	2.65E+09	1	0	-87.2452



Signal Quality Classification	Path Loss	Signal quality opinion
	$PL \leq 88$	Very strong
	$89 \leq PL \leq 92$	Strong
	$93 \leq PL \leq 96$	Fair
	$97 \leq PL \leq 100$	Weak
	$101 \leq PL$	Very weak
	No signal Received	No signal Received

Transfer Ray Tracing Signal to Q&A Syntax



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Original Data

Base_station_ID	Tx_x	Tx_y	Tx_z	Rx_x	Rx_y	Rx_z	Distance	Frequency	LOS_Flag	Transmit power (dBm)	Received power (dBm)
Base_Station_UAV	0	0	100	-183.333	7.265218	1.5	208.2448	2.65E+09	1	0	-87.2452



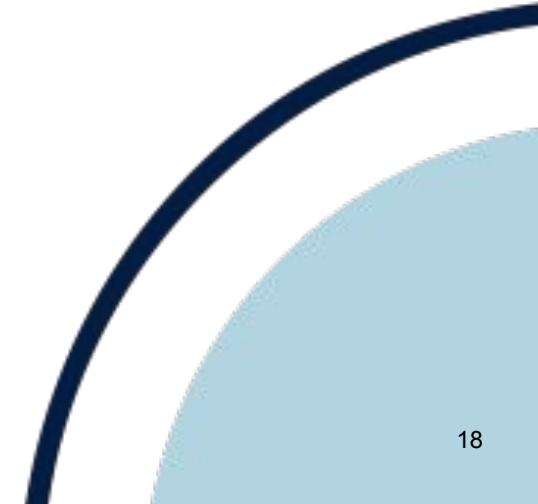
Classify Radio Signal Strength to Signal Quality Opinion

Question (Input)	Answer (Output)
<p>In wireless communication environment, the base station is positioned at $(X,Y,Z)=(0,0,100)$, and the receiver is at $(X,Y,Z)=(-183,7,1)$. <u>What is the received power level?</u></p>	<p>very strong</p>





Model Training



LLaMA2 (Large Language Model Meta AI)



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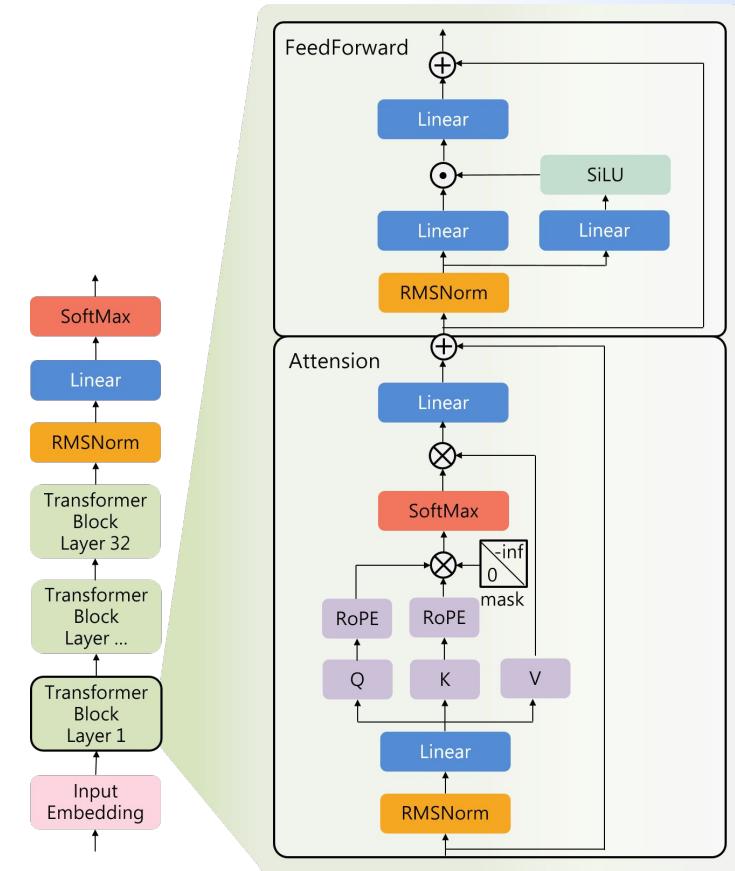


- Meta's Open-Source Large Language Model (LLM)
- Foundation Language Model

	Training Data	Params	Context Length	GQA	Tokens	LR
LLAMA 1	See Touvron et al. (2023)	7B	2k	✗	1.0T	3.0×10^{-4}
		13B	2k	✗	1.0T	3.0×10^{-4}
		33B	2k	✗	1.4T	1.5×10^{-4}
		65B	2k	✗	1.4T	1.5×10^{-4}
LLAMA 2	A new mix of publicly available online data	7B	4k	✗	2.0T	3.0×10^{-4}
		13B	4k	✗	2.0T	3.0×10^{-4}
		34B	4k	✓	2.0T	1.5×10^{-4}
		70B	4k	✓	2.0T	1.5×10^{-4}

▲ Comparison table outlining the differences between LLAMA1 and LLAMA2 models [1]

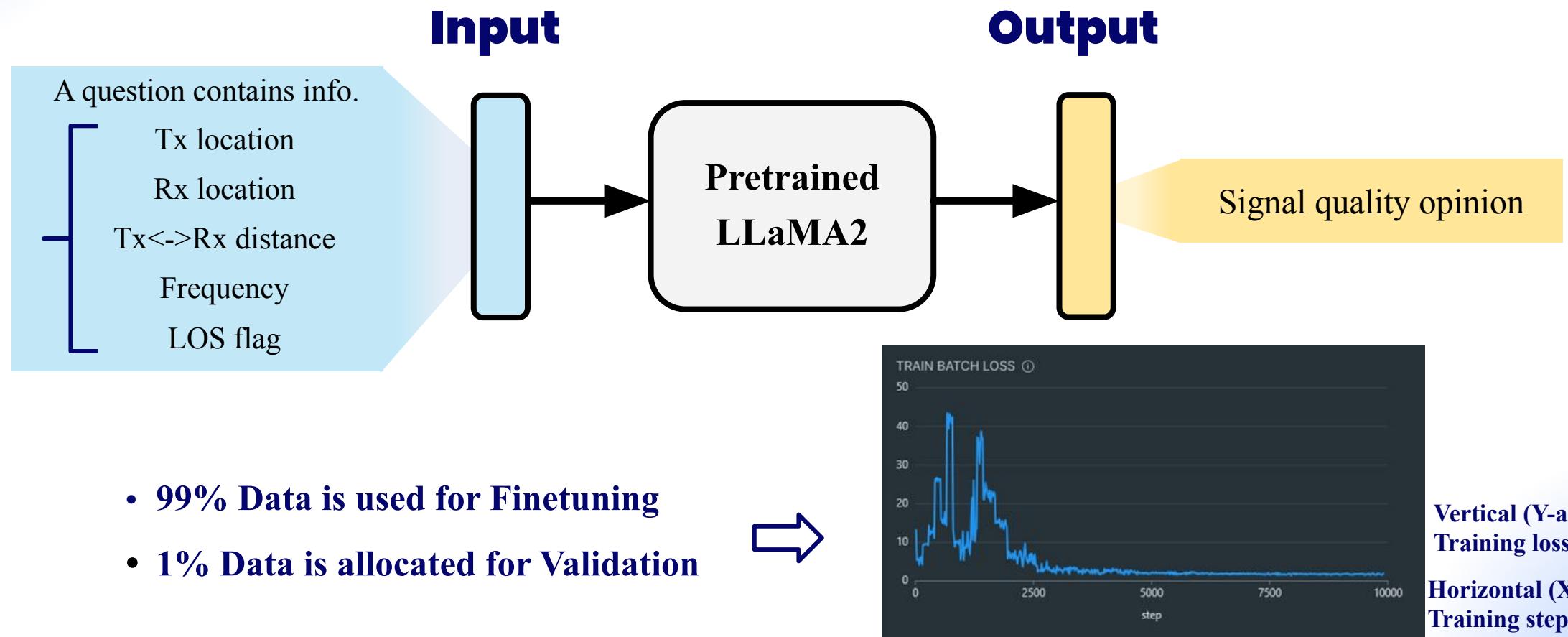
- Increased Training Data by 40%
- Utilized Grouped-query attention (GQA) to accelerate inference
- Implemented RLHF (reinforcement learning from human feedback) in Llama2-chat models to enhance model robustness.



▲ Architecture diagram of the LLaMA2 model



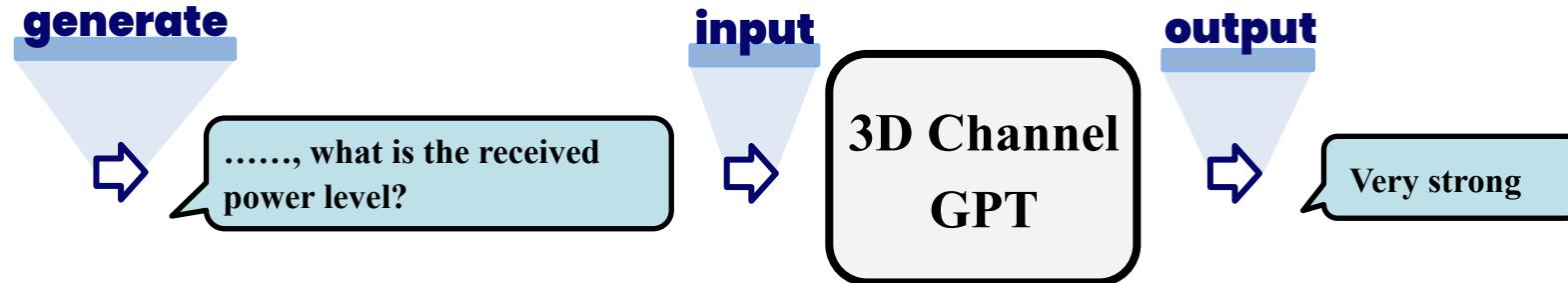
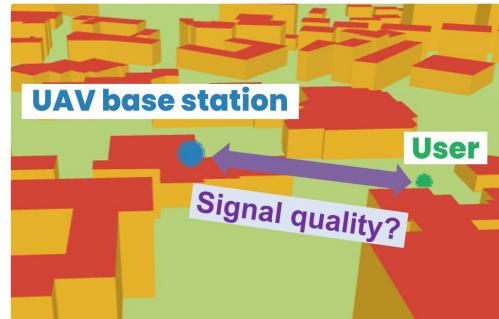
Fine-tuning : LLaMA2 to 3D Channel GPT



▲ Training Loss Convergence Curve for the Model

Model Validation

Case 1: Single Base Station Scenario



- Randomly selecting 100 data points yields an accuracy of 84%

⇒ Evaluating the effectiveness of LLM model in wireless communication scenarios.

Input Text (tokenization max length setting may tr... Question: In wireless communication environment, the Answer:	Target Text	Predicted Text
	strong	fair
Question: In wireless communication environment, the Answer:	strong	strong
Question: In wireless communication environment, the Answer:	fair	fair
Question: In wireless communication environment, the Answer:	no signal received	no signal received
Question: In wireless communication environment, the Answer:	very strong	very strong
Question: In wireless communication environment, the Answer:	fair	fair

▲ Comparison of the actual results and model output

Case 1: Single Base Station Scenario



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User Location

(-226,365,1)

(232,-146,1)

(-594,-215,1)

(-213,-351,1)

(-17,67,1)

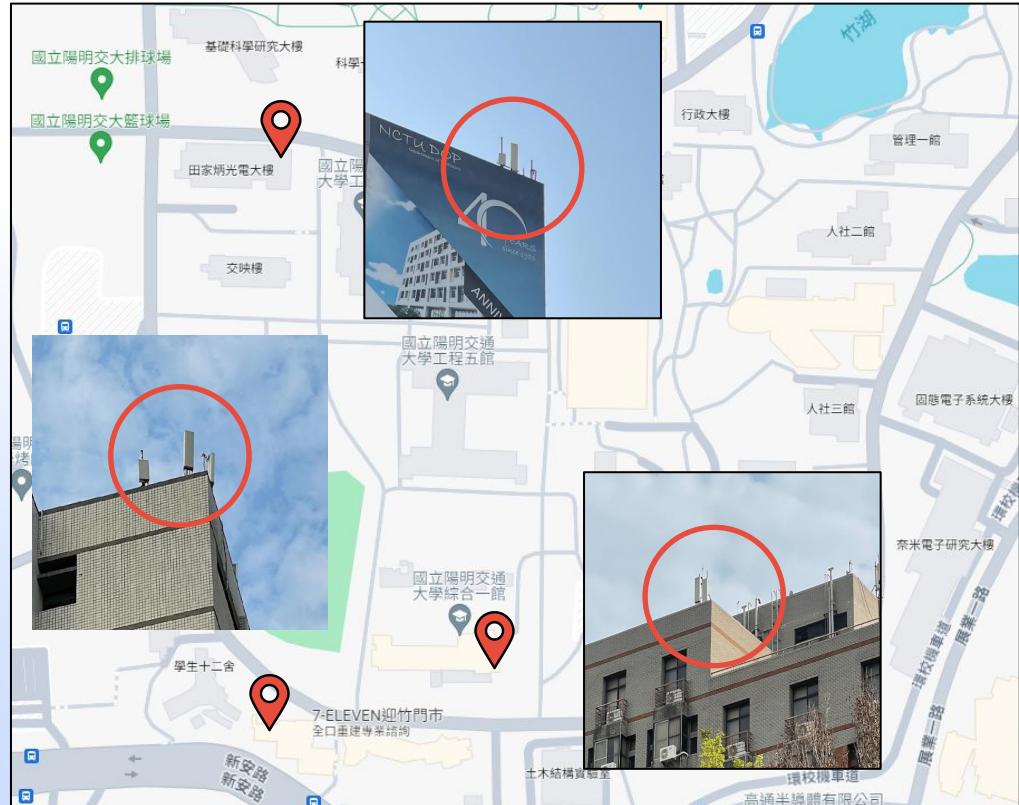
(-559,-317,1)

Result Comparison : Model Prediction vs. Actual Results

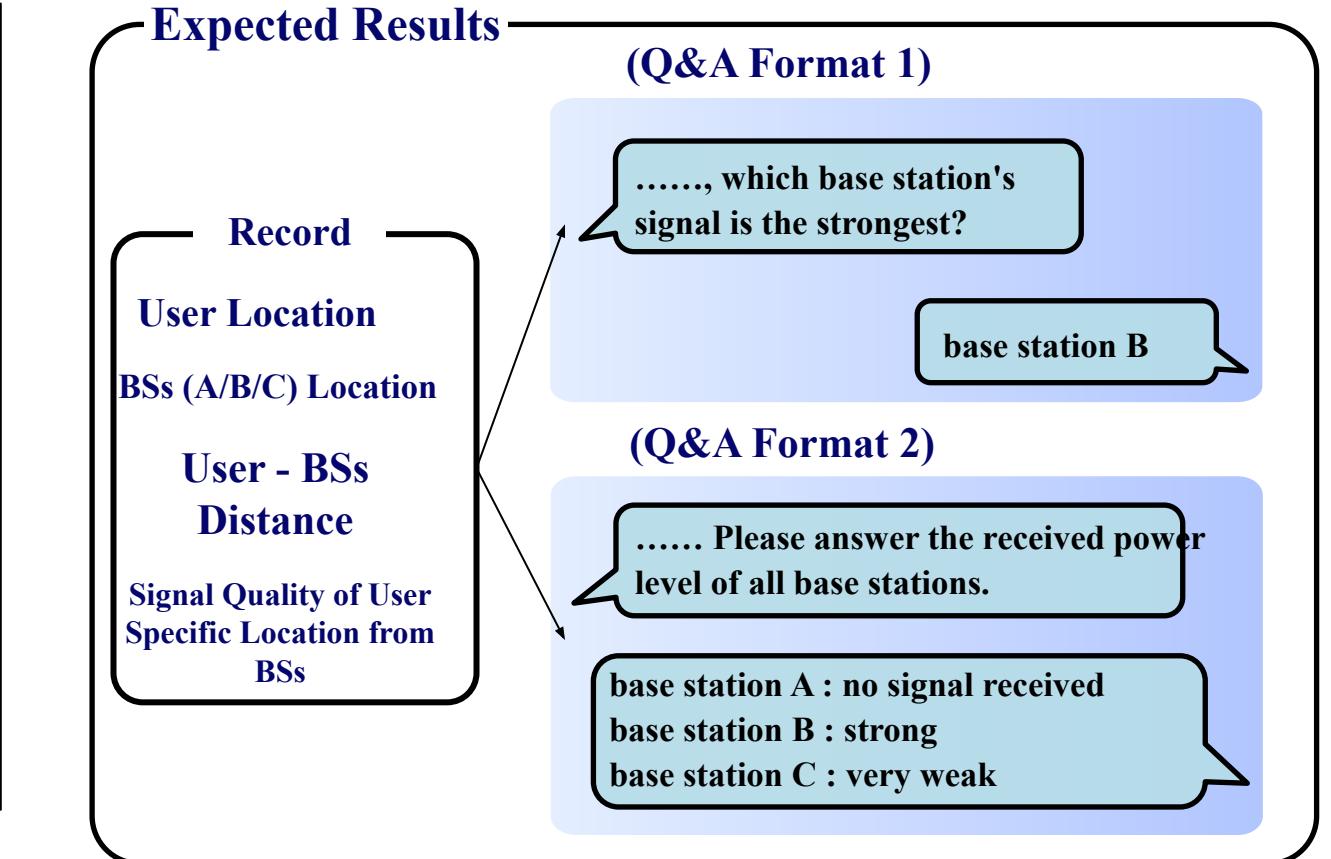
User Location	Input Text (tokenization max length setting may affect the result)	Target Text	Predicted Text	
(-226,365,1)	Question: In wireless communication environment, the signal strength is strong.	strong	fair	Fail
(232,-146,1)	Question: In wireless communication environment, the signal strength is strong.	strong	strong	Correct
(-594,-215,1)	Question: In wireless communication environment, the signal strength is fair.	fair	fair	Correct
(-213,-351,1)	Question: In wireless communication environment, the signal strength is no signal received.	no signal received	no signal received	Correct
(-17,67,1)	Question: In wireless communication environment, the signal strength is very strong.	very strong	very strong	Correct
(-559,-317,1)	Question: In wireless communication environment, the signal strength is fair.	fair	fair	Correct

Case 2: CHT BSs Signal Quality Opinion by 3D Channel GPT

Objective: Estimating signal quality with multiple base stations in the campus environment of NYCU Campus



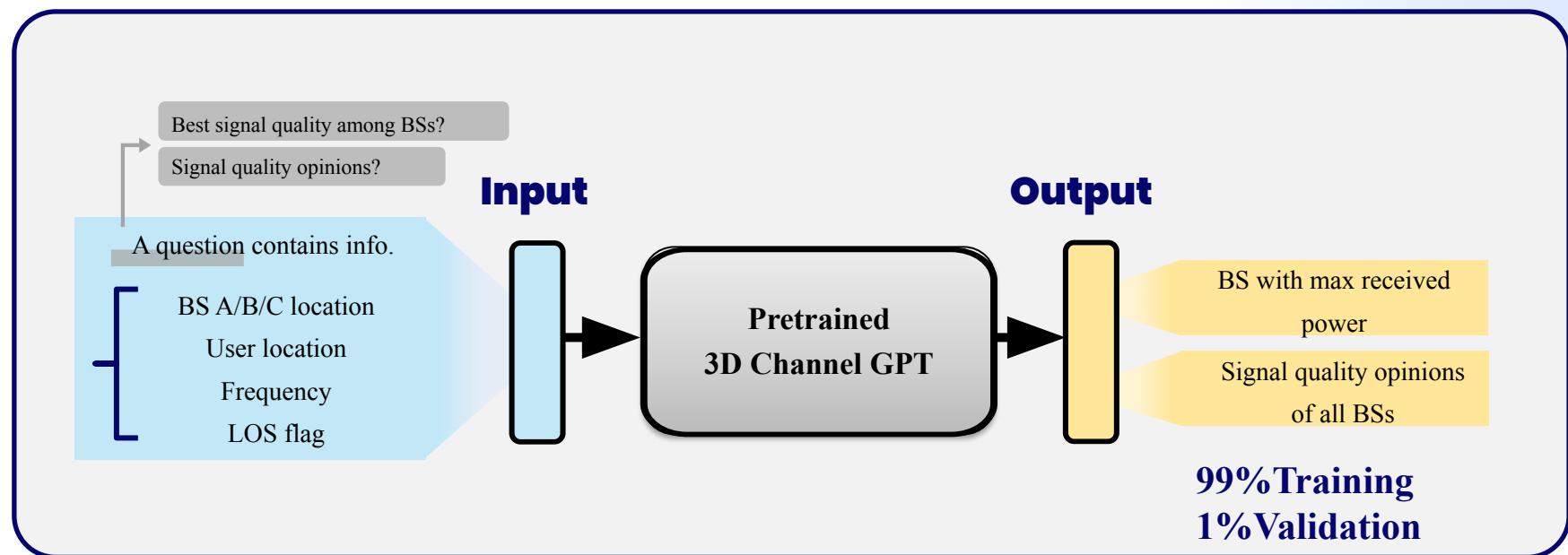
▲ Actual deployed locations of Chunghwa Telecom base stations in the Guangfu Campus of Yang Ming Chiao Tung University



Case 2: CHT BSs Signal Quality Opinion by 3D Channel GPT

Wireless communication parameter setting

Parameters	Description
BS A Position	(-150, -330, 40)
BS B Position	(0, -275, 40)
BS C Position	(-150, 50, 40)
Carrier Frequency	2.65 GHz (4GLTE band 7)
User Position	<u>Random</u>



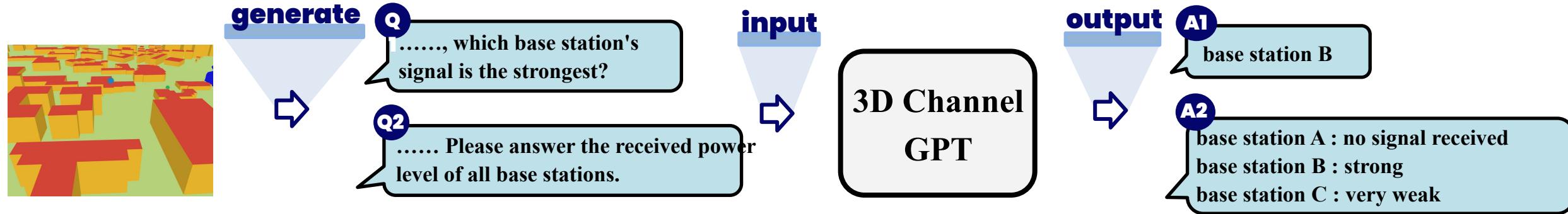
Randomly sample
10,000 data



Conversion to 20,000
Q&A format

(Format 1 10,000 data)
(Format 2 10,000 data)

Case 2: CHT BSs Signal Quality Opinion by 3D Channel GPT



- (Q&A Format 1) Randomly sample 100 data, resulting in an accuracy of **67%**
- (Q&A Format 2) Randomly sample 100 data, resulting in an accuracy of **62%**

(Due to the increased complexity of the scenario, the accuracy has decreased compared to Case 1.)

⇒ Furtherly extend the capabilities of 3D Channel-GPT model through users and Multiple base stations through simple Q&A interactions

Conclusions

- We(NYCU ICCLab) creates **3D Channel GPT** (Generative Pretrained Transformer), achieving signal quality opinion score up to 84% accuracy with the help of DeepMentor solution.
- Develop a methodology to generate the first **open data radio signal map @** NYCU campus, which can be extended further for 3D+Radio dataset in other cities.
- 3D Channel GPT provides **user-friendly Q&A dialogue** service for assessing signal quality in **site-specific** 3D air-to-ground communication.

However underlying Challenges exists....

1. **Stochastic vs Deterministic Models:**

While supporting 3GPP 38.901 channels, Sionna's accuracy in NLOS scenarios still trails specialized tools like Wireless InSite by 15-20% in beam selection tests

2. **Quantization Effects:**

Hardware implementation mismatches emerge from ideal simulations vs actual RF impairments

3. **Mobility Handling:**

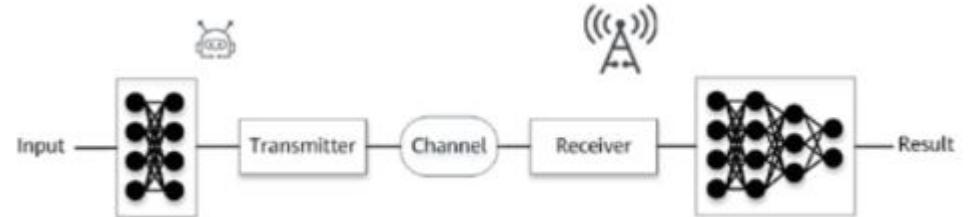
Simulating high-velocity scenarios (e.g., >500 km/h) strains current ray tracing capabilities

Learning Curve

1. Cross-Disciplinary Expertise

Requires simultaneous mastery of:

- TensorFlow/Keras for neural network integration
- Communication theory (LDPC, MIMO, OFDM)
- Ray optics for propagation modeling



source : beyond traditional communications, Huawei Research



Learning Curve

2. API Complexity

The framework's flexibility introduces challenges in:

- **Debugging gradient flows through hybrid AI/DSP pipelines**
- **Customizing lower PHY components while maintaining differentiability**

```
batch_size = 1024
n = 1000 # codeword length
k = 500 # information bits per codeword
m = 4 # bits per symbol
snr = 10 # signal-to-noise ratio

c = Constellation("qam",m,trainable=True)
b = BinarySource(([batch_size, k])
u = LDPC5GEncoder (k,n)(b)
x = Mapper (constellation=c)(u)
y = AWGN(([x,1/snr])
llr = Demapper("app", constellation=c)([y,1/snr])
b_hat = LDPC5GDecoder(LDPC5GEncoder (k, n))(llr)
```



```
c = Constellation("qam",m,trainable=True)
b = BinarySource(([batch_size, k])
u = LDPC5GEncoder (k,n)(b)
x = Mapper (constellation=c)(u)
y = AWGN(([x,1/snr])
llr = NeuralDemapper()([y,1/snr])
b_hat = LDPC5GDecoder(LDPC5GEncoder (k, n))(llr)
```

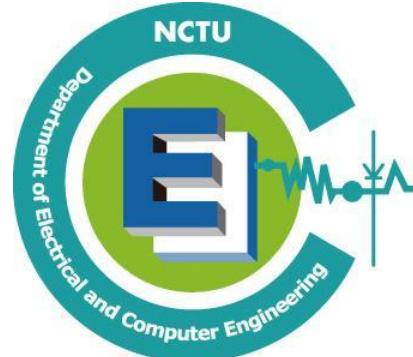
Customization :
replacing Demapper with NeuralDemapper

Operational Constraints

1. **Version Compatibility**: Strict requirements for Python 3.8-3.12 and TensorFlow 2.14-2.19 create dependency conflicts
2. **Multi-GPU Scaling**: Distributed training needs careful resource management for large batch sizes
3. **Reproducibility**: Differences in GPU architectures (Ampere vs Hopper) affect benchmark consistency



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6G Physical-Layer Research Tools

NVIDIA Sionna: An Open-Source Library

Presenter: Chia-Chuan Chiu

Advisor: Li-Chun Wang

Class GitHub: https://github.com/NYCU-ICCLab/sionna_introduction

NVIDIA website: <https://developer.nvidia.com/sionna>

Sionna official tutorial: <https://nvlabs.github.io/sionna/index.html>

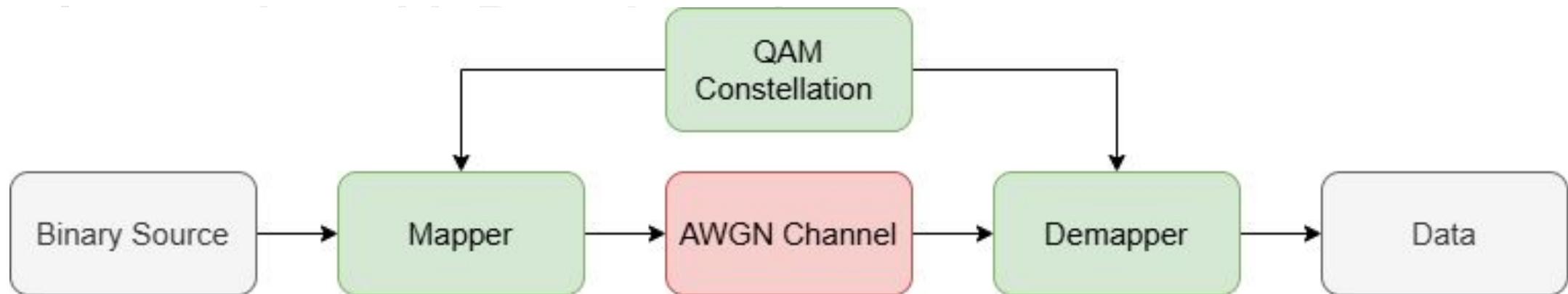
Sionna GitHub: <https://github.com/NVlabs/sionna>

Key Features of Sionna

- **Modular & Flexible Design**
 - Modular architecture that can be easily customized or extended.
 - Allows quick prototyping of complex communication systems.
- **Integration with Deep Learning**
 - Parameters are constructed using TensorFlow
- **High Computational Efficiency**
 - Accelerated by GPU computing

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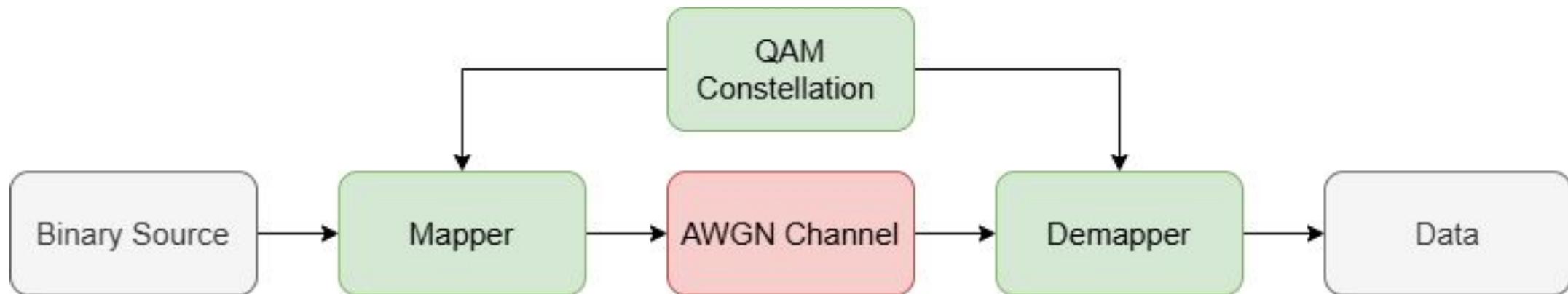
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- Parameters are constructed using **TensorFlow**
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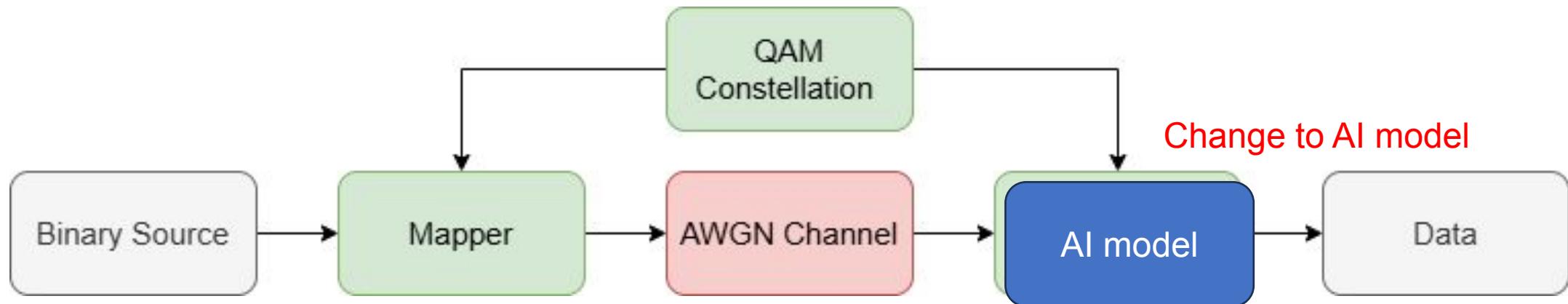
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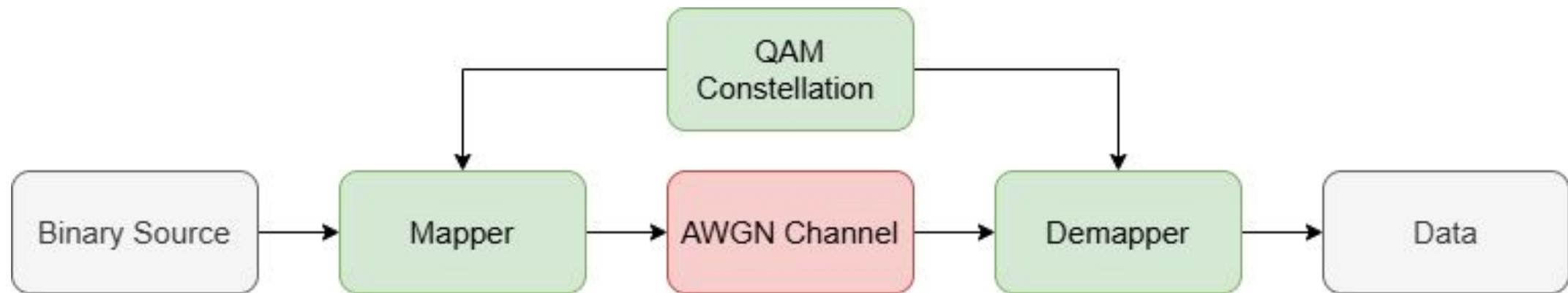
- **Integration with Machine Learning**

- Parameters are constructed using TensorFlow
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- **High Computational Efficiency**

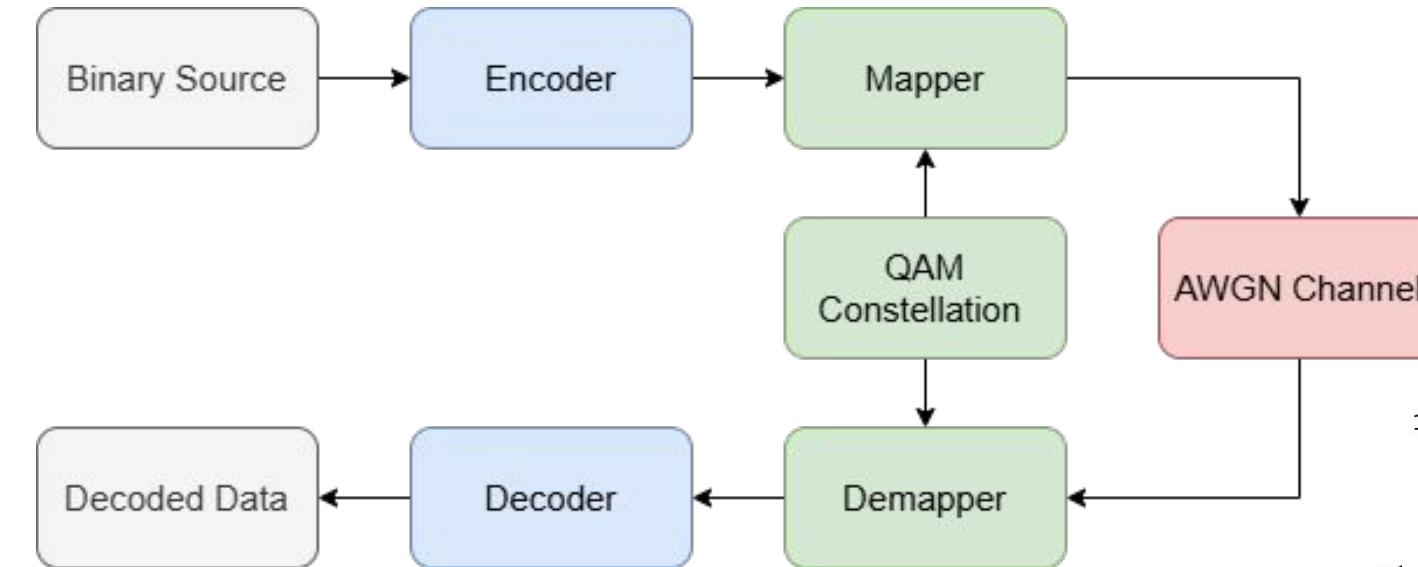
- Accelerated by GPU computing

Basic Usage of Sionna

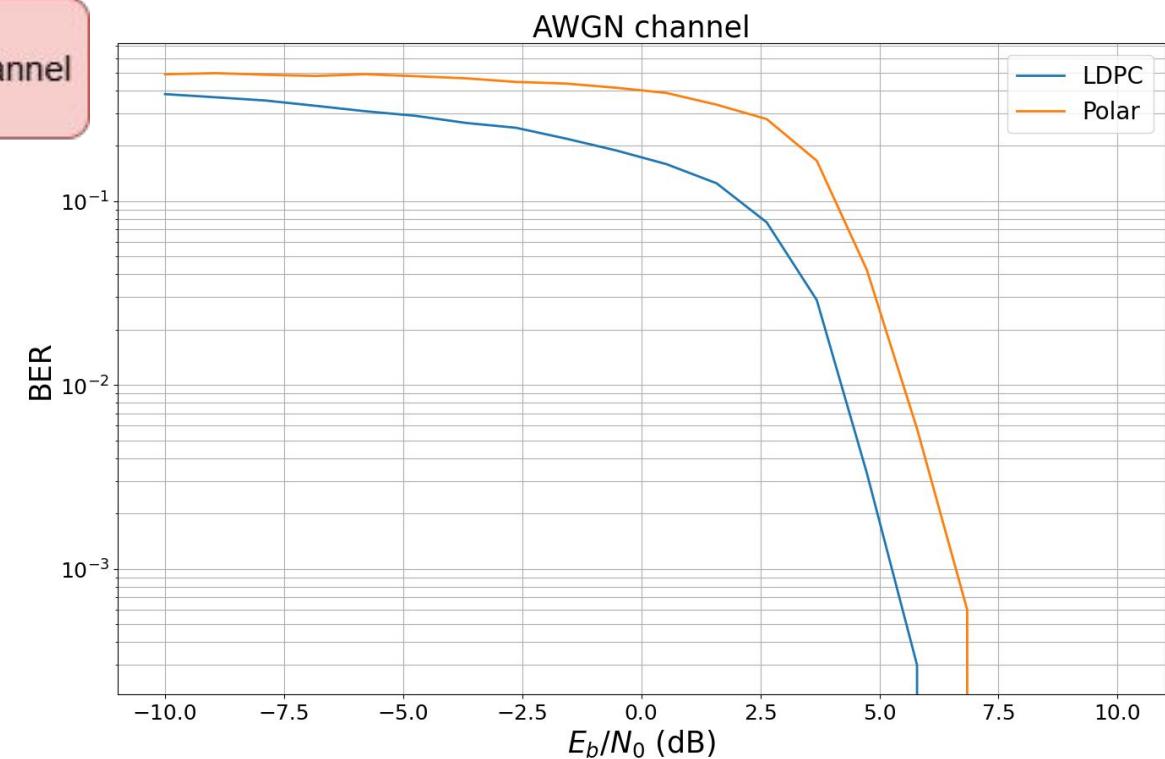


https://github.com/NYCU-ICCLab/sionna_introduction/blob/main/sample_code/Demo_sionna_basic.ipynb

Case 1 : System Performance Evaluation

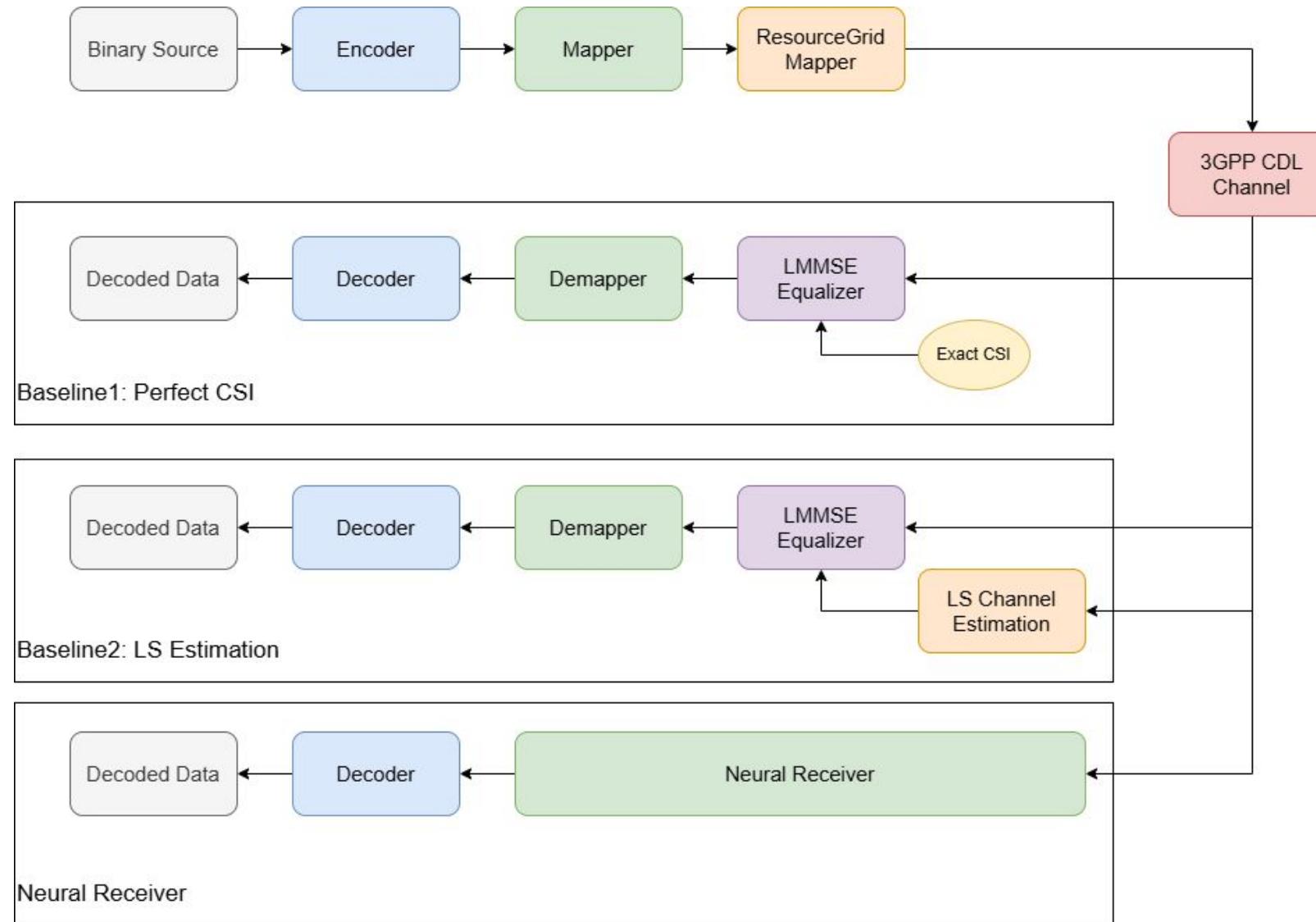


Channel coding comparison: LDPC vs Polar



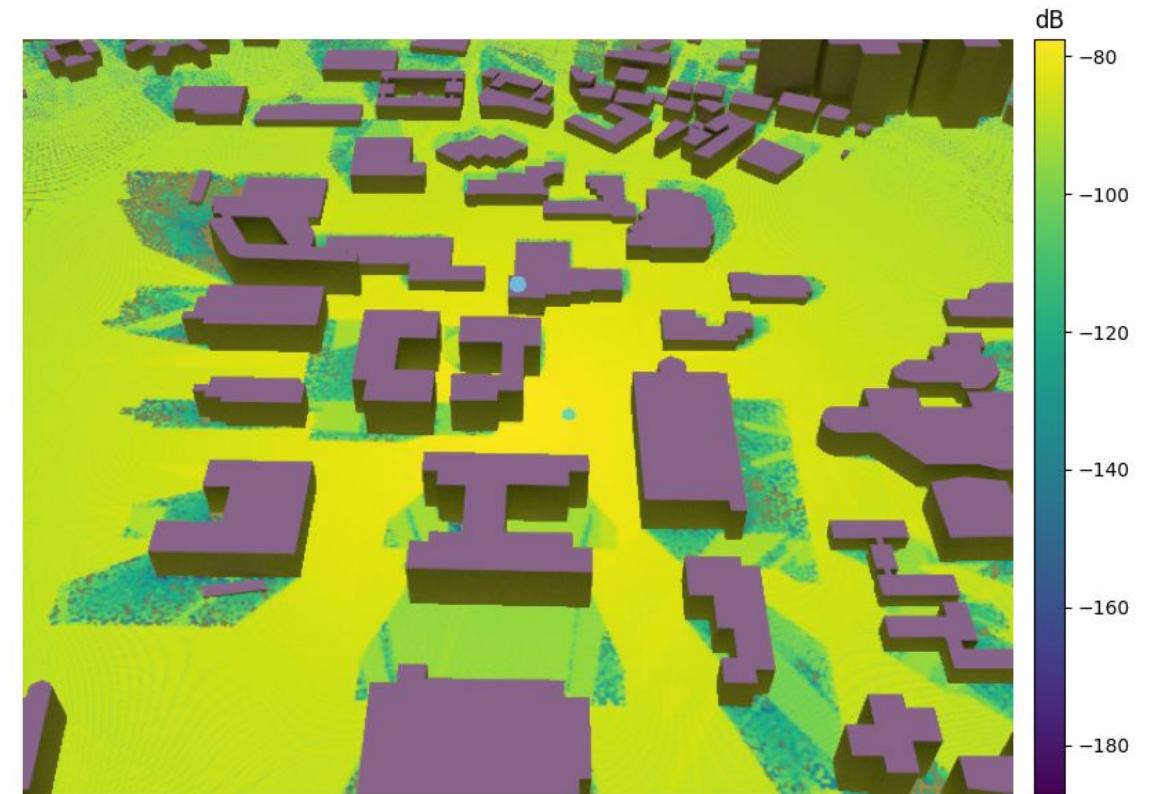
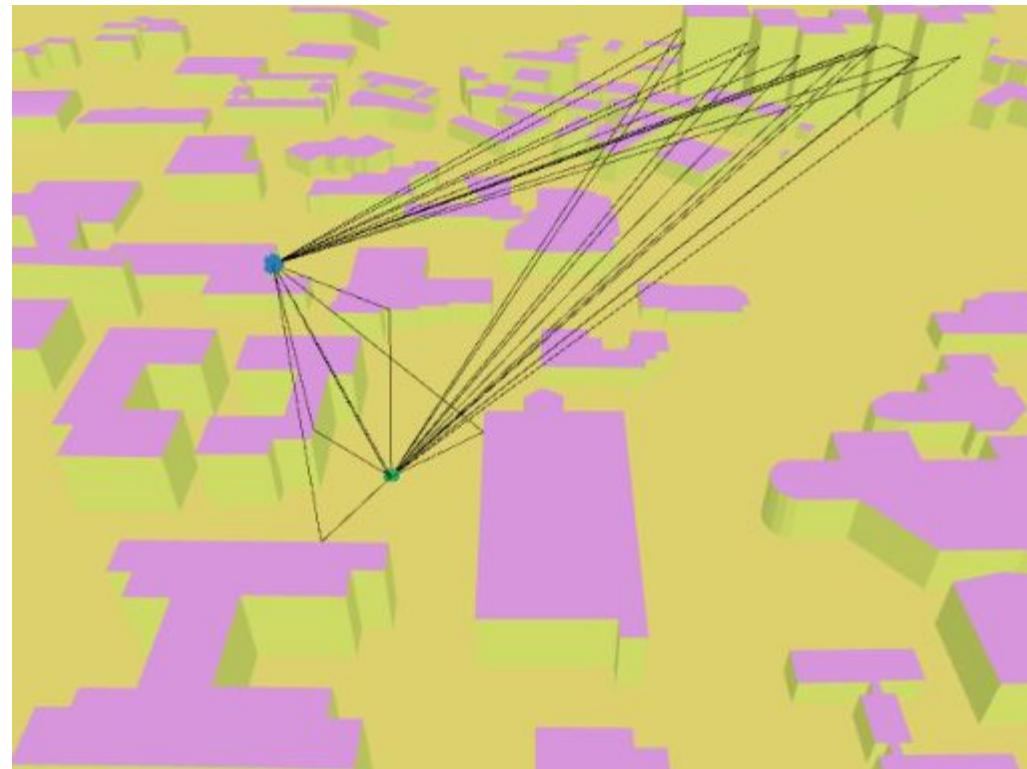


Case 2 : Neural Receiver



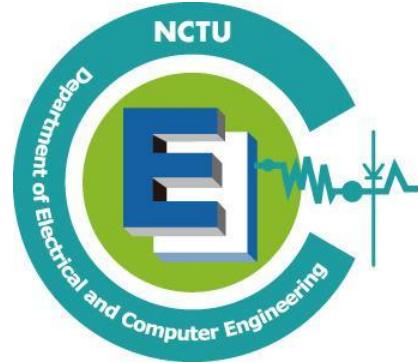
Case 3 : Ray Tracing

Ray tracing simulation in NYCU campus





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Thank You

Class GitHub: https://github.com/s87315eve/sionna_introduction

NVIDIA website: <https://developer.nvidia.com/sionna>

Sionna official tutorial: <https://nvlabs.github.io/sionna/index.html>

Sionna GitHub: <https://github.com/NVlabs/sionna>