



Robustifying Predictive Intelligence for A New Spectrum Market

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AI Empowers Next-Gen Wireless

- AI is empowering our NextG spectrum and wireless systems



@Tmobile

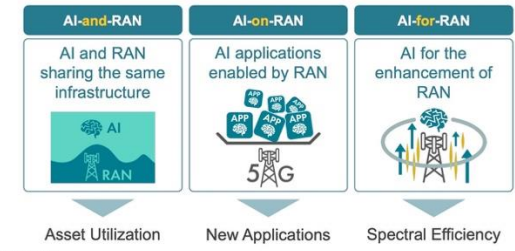
Future Technologies

LLM Application in Wireless Communication Knowledge Management

This paper details the use of RAG to design a Q&A solution for wireless communication knowledge bases and an accompanying evaluation solution.

Hua

Key Focus Areas



- But do we trust it yet?

DeepSeek-R1 LLM Fails Over Half of Jailbreak Attacks in Security Analysis

Technology

Deepfake makers can now evade an unusual detection method

Opinion - Chatbots, Robotics & AI

From "Gig" to "Glitch" Economy: Hospitality's New Battle Against AI Hallucinations

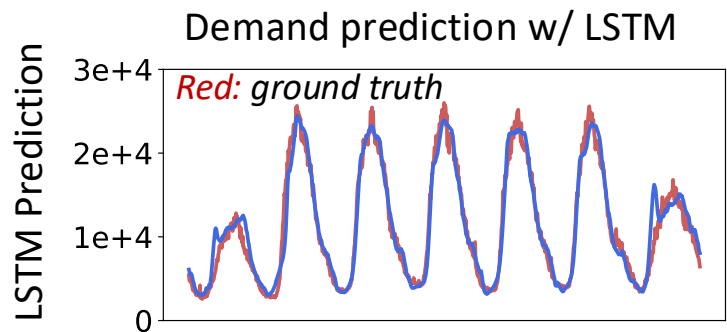
- Bottomline:** Unsafe and unrobust AI is not ready for wireless.

- ❖ **Risks:** infrastructure instability, constraint violation, exploit by malicious user, market manipulation, ...

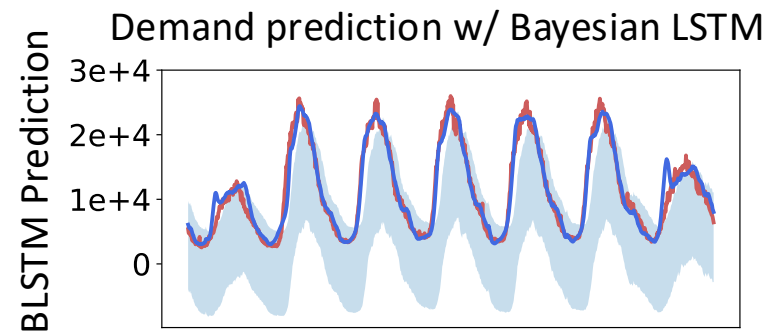
How This Project Differs from AI Safety Research?

❑ **Argument:** Robust AI does not necessarily make AI-empowered wireless sufficiently or more robust.

❑ **Example:** spectrum allocation based on demand prediction.



Maximum # violated cases: 49.51% - 53.23%



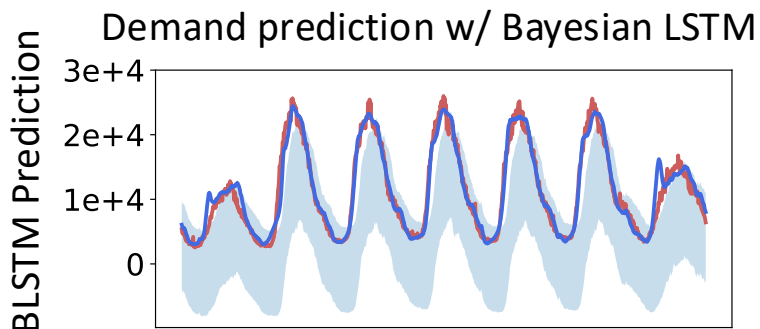
Maximum # violated cases: 5.67% - 22.26%

- ❖ Apply Bayesian uncertainty quantification and use 95% confidence bound.
 - Reduced violations (underprovisioning) => More robust than LSTM.
 - Violation still larger than estimated risk (95%) => Not robust enough.

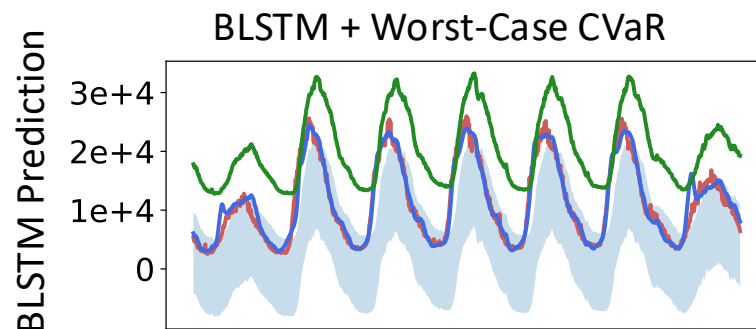
Project Goals

❑ Thrust I: Robust AI Algorithms for New Spectrum AI

- ❖ **Goal:** Certifiable & distributional robustness for dynamic spectrum access
- ❖ **Tools:** (1) Bayesian uncertainty quantification, (2) distributionally robust risk, (3) conformal calibration (certifiable statistical guarantee)
- ❖ **Preliminary Results:**



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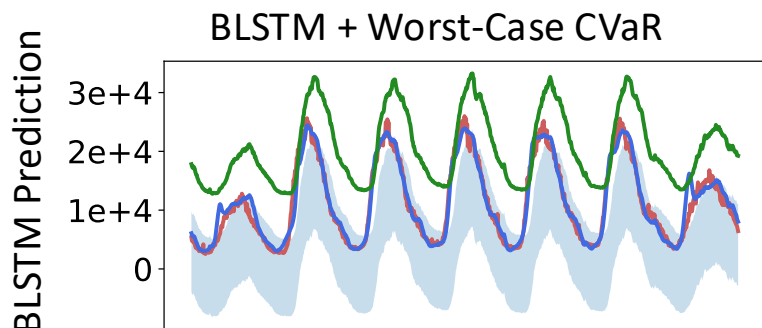
Maximum # violated cases: 0.02% - 9.68%

- ❖ Apply distributionally robust (CVaR) upper bound on Bayesian posterior.
 - Conservative statistical bound w.r.t. distributional uncertainty.
 - Inefficient, and still not *certifiably* robust.

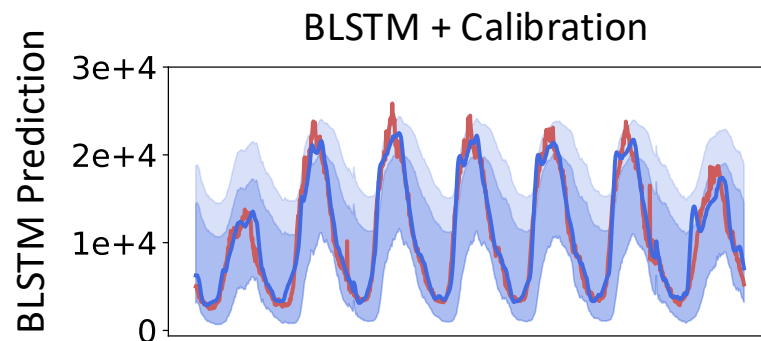
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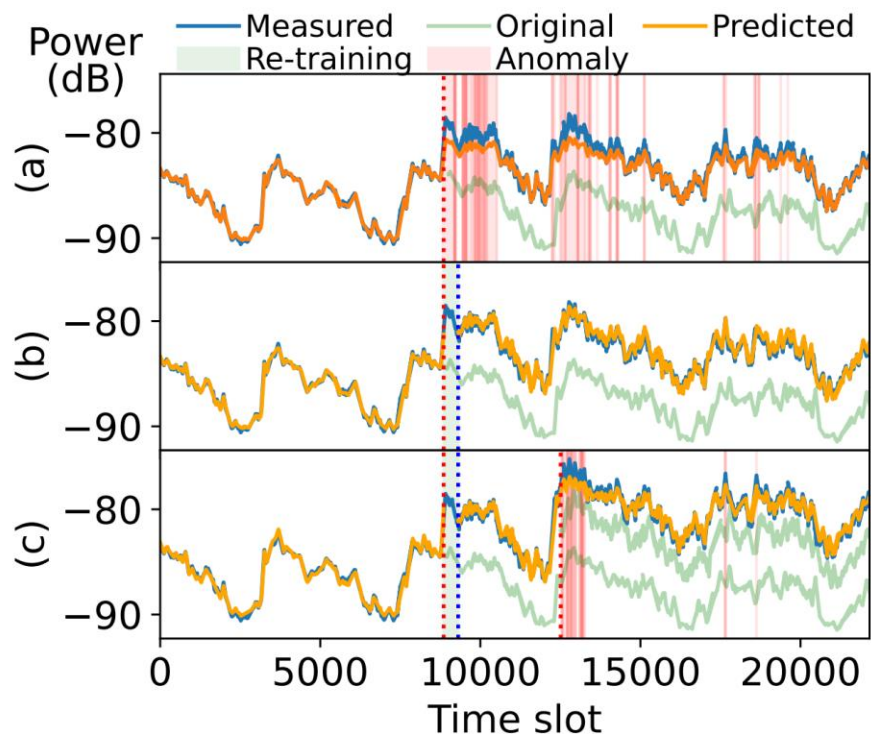


Maximum # violated cases: 2% - ~5%

- ❖ Conformally calibrate Bayesian posterior upper 95%-CI.
 - Nice conformal certifiable robustness.
 - Does not handle well distribution shifts.

Project Goals

❑ Thrust 2: Robust Spectrum Data for New Spectrum AI

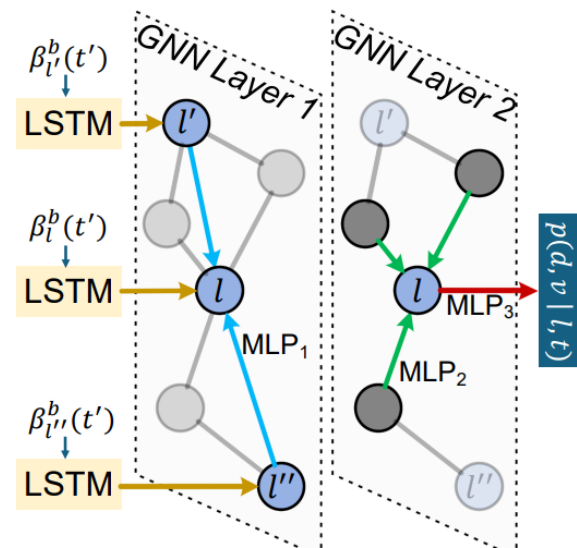
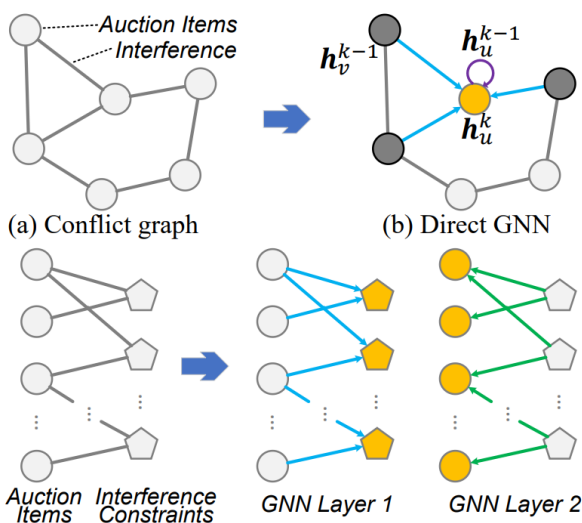


- ❖ **Goal:** Obtain trustworthy spectrum data during **spectrum handovers**
- ❖ **Tools:** (1) Cryptographic commitments and random sampling, (2) spatio-temporal-spectral learning & federated calibration

Project Goals

❑ Thrust 3: Robust Market Mechanism with New Spectrum AI

- ❖ **Goal:** Design efficient & robust market mechanism in face of AI players
- ❖ **Tools:** learning-based market mechanism design, with provable properties such as truthfulness, individual rationality and budget balance



Thank you very much!

Q&A?

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