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Training Aggregate Receiver- Transmitter Interference Classifiers

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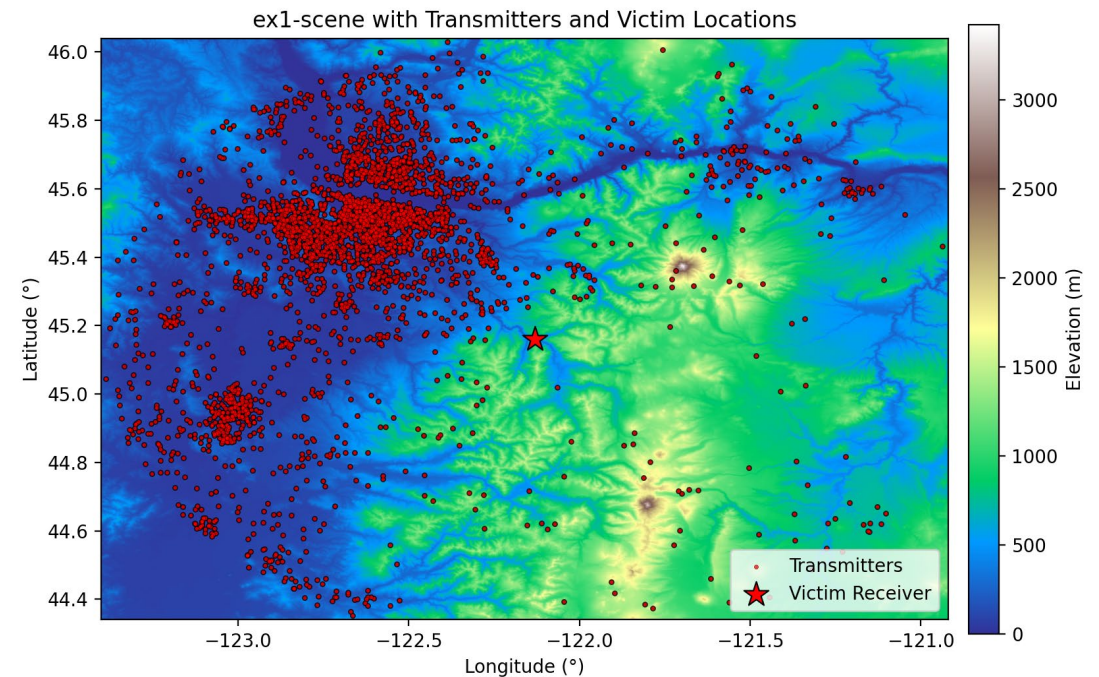
Agenda/Outline

- Problem Statement
- Dataset
- Model
- Results
- Next Steps

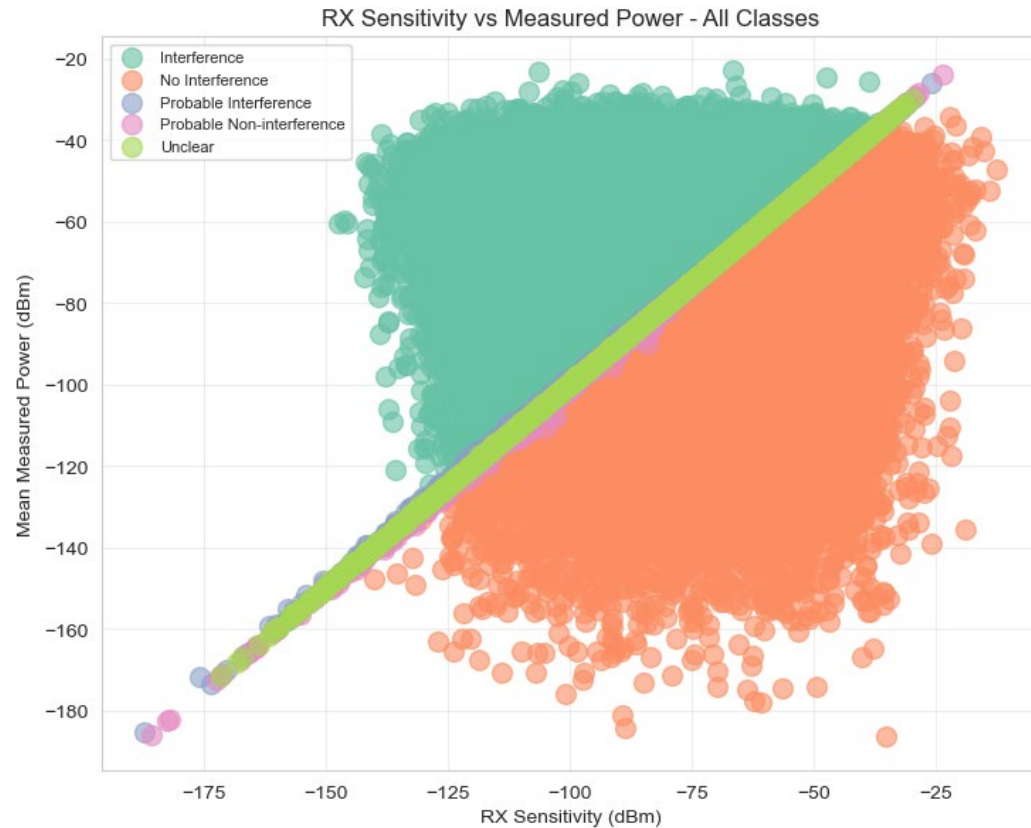


Problem Statement

- How do you know if a certain receiver will be interfered with?
- How much information do you need to decide?
- Can you explain the result?
With what confidence?



Problem Statement



- Knowing whether the model thinks this is an interfering or non-interfering scenario → now you can search for thresholds when the classification changes
- An explicit classification can be more useful than comparing a quantitative comparison against a reference value

Proposed Solution

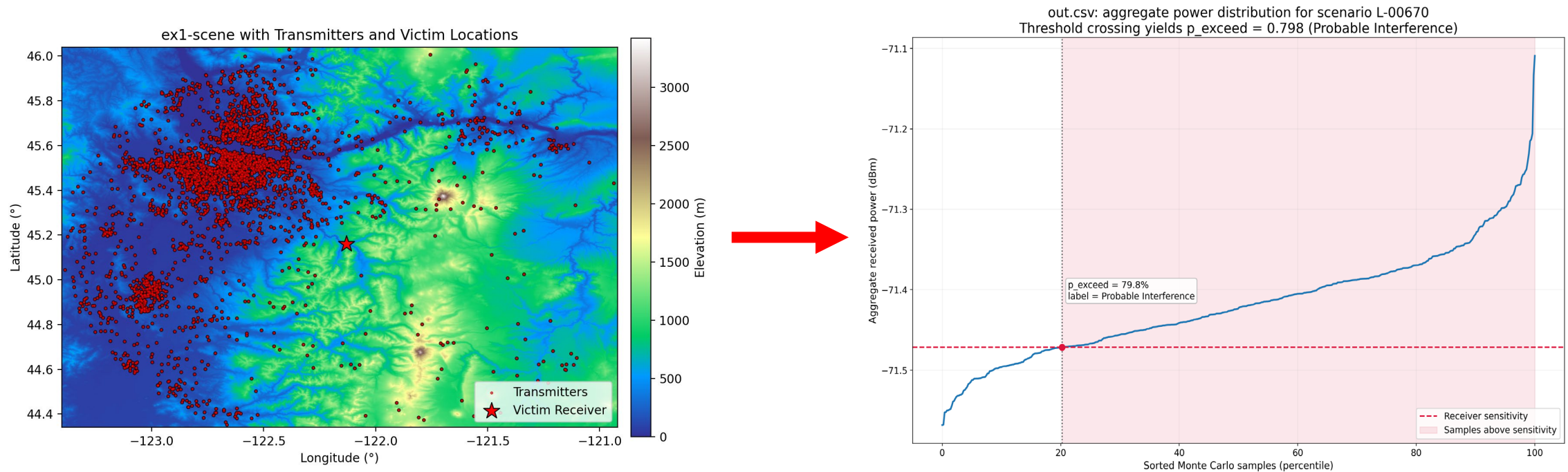
- Two key components:
 - A dataset that demonstrates this problem
 - A trained model reliably classifying when the victim is interfered with
- Intended outcome:
 - Develop a proof-of-concept against a gap in interference modeling
 - Learn strengths/weaknesses of the dataset, models, and approach

- Best case would be real-world data, but it doesn't exist at scale
- Therefore, we use a synthetic technique:



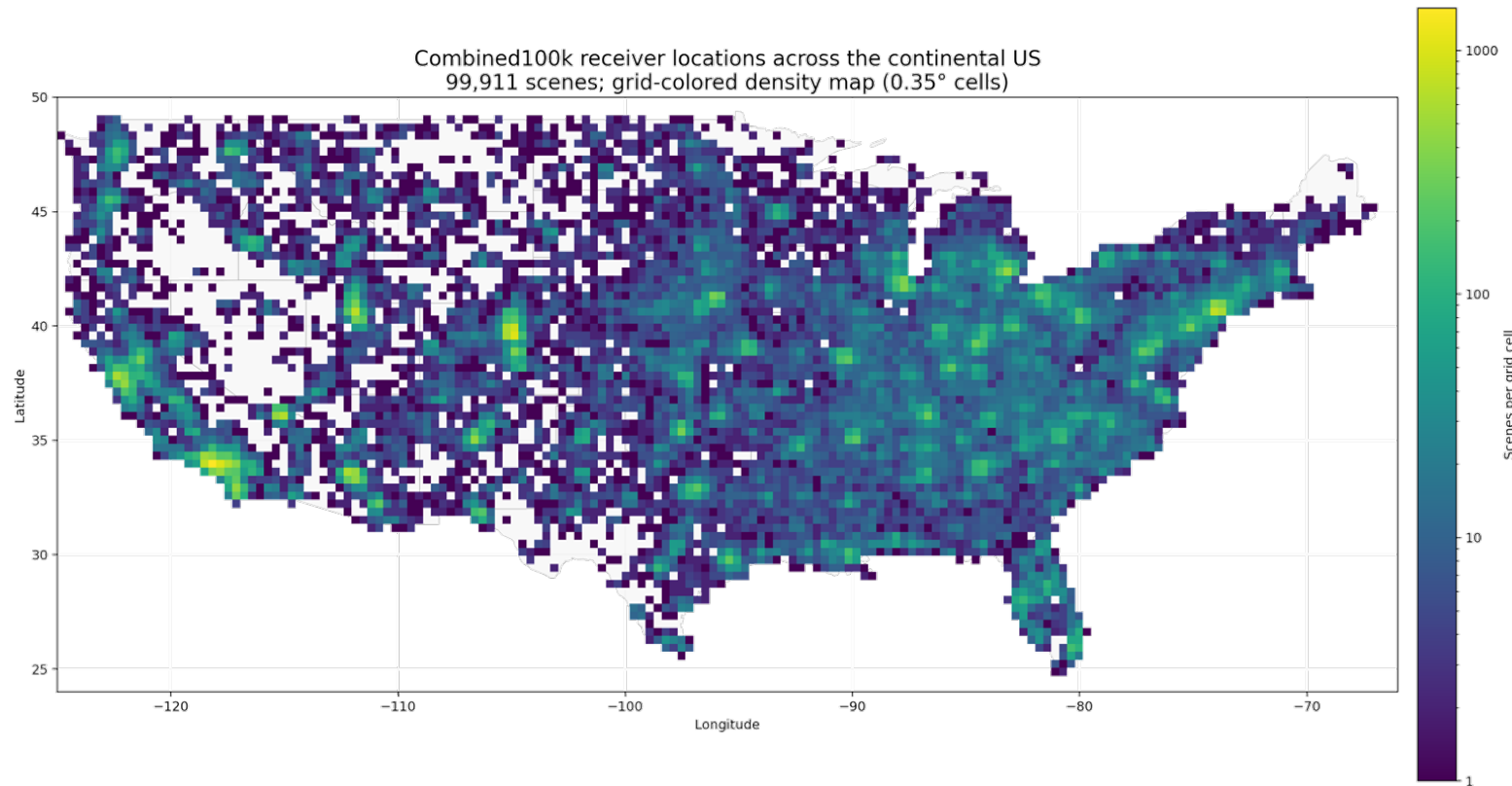
Dataset Concept

- Best case would be real-world data, but it doesn't exist at scale
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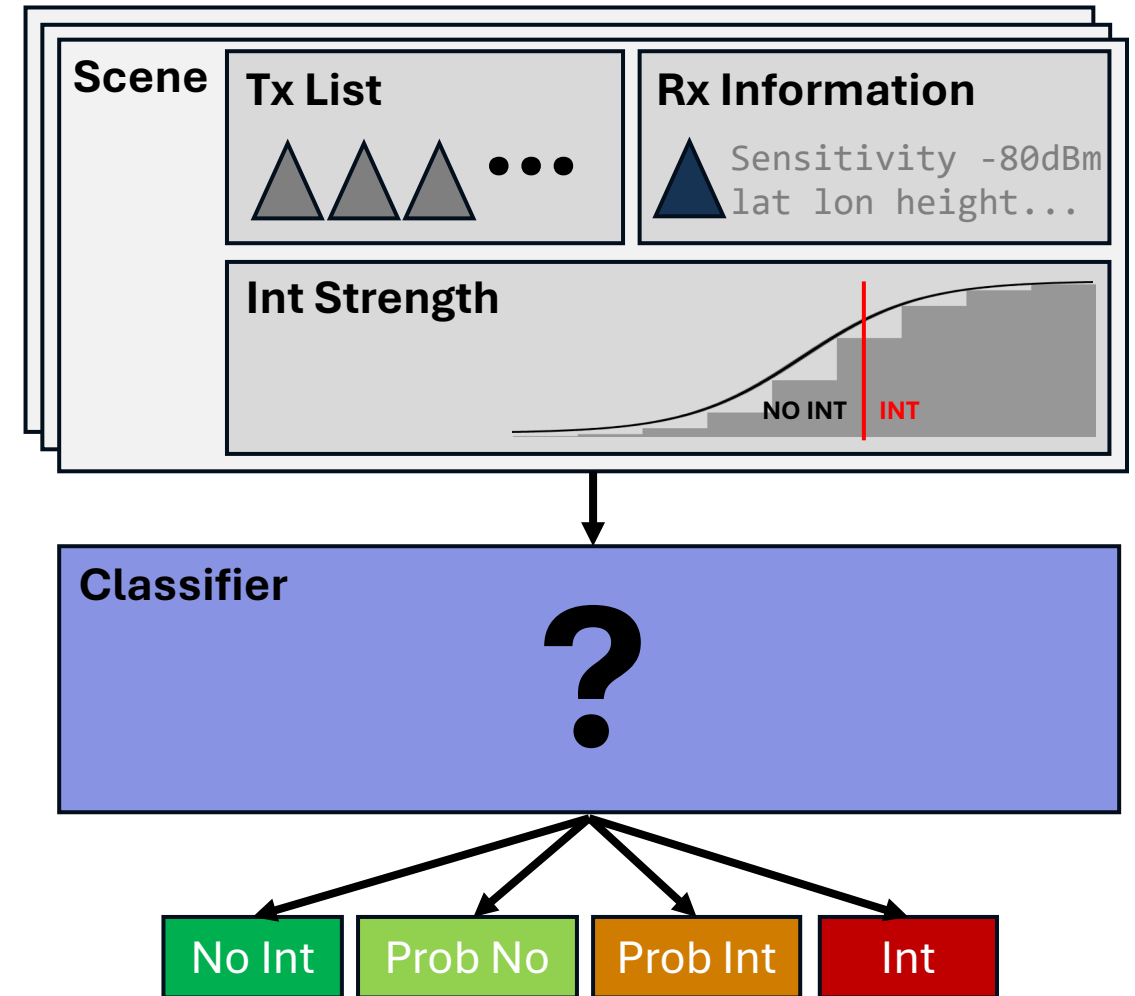
Dataset Result

- ~100k unique labeled scenes across the continental US
- 10s to 1000s of transmitters across rural and urban scenarios

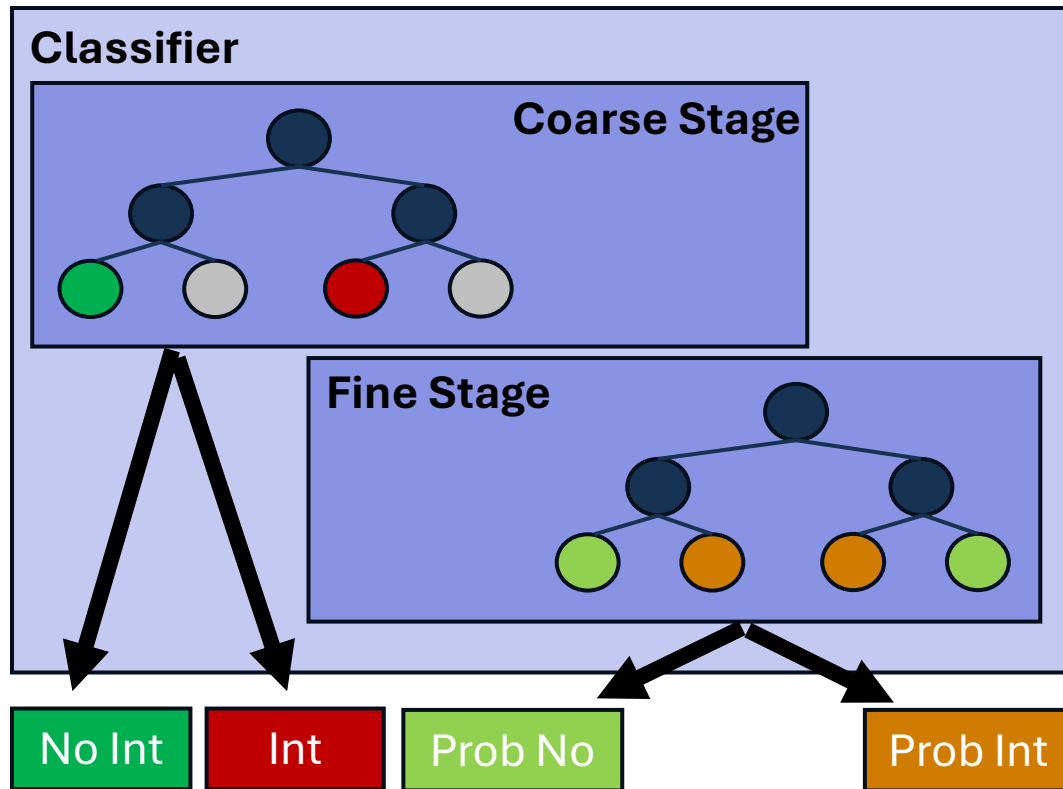


Model Concept

- Aggregate Receiver-Transmitter Interference Classifier (ARTIC)
- Train against the corpus of scenes and determine how well it predicts an interference class
- No Interference $\leftrightarrow$ Interference



Classifier Architecture



- Many architectures considered
- Strongest builds separate endpoint cases from close calls into multiple stages
- Main frontrunning technologies
 - gradient-boosted machines (GBMs)
 - graph neural networks (GNNs)

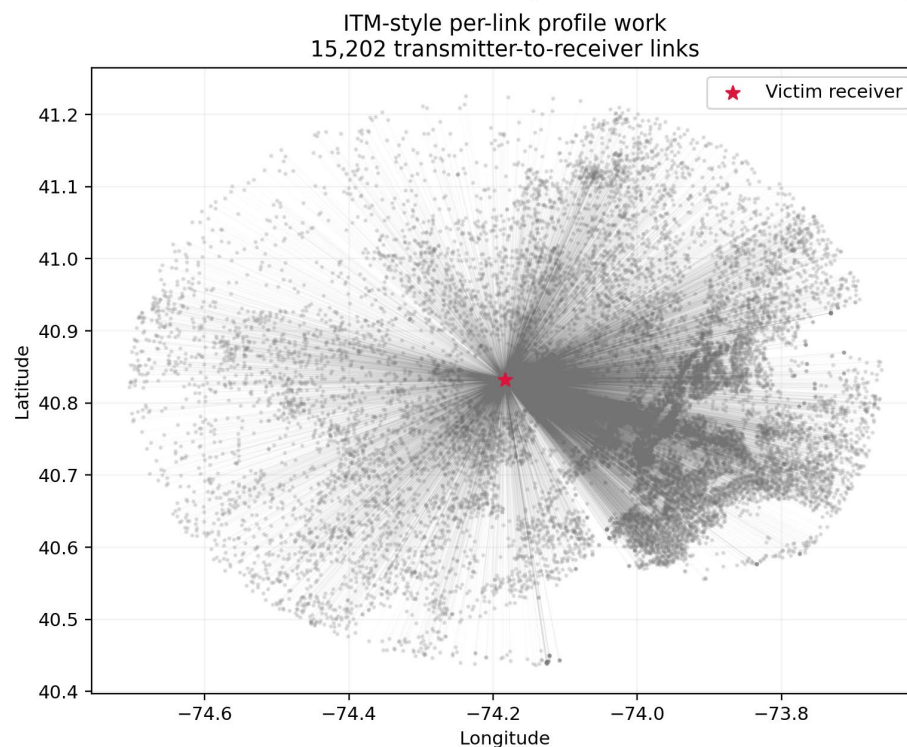
Feature Engineering

- This system operates at a scene level, not a link level
- Therefore, features can come from scene context or from meta-model behavior

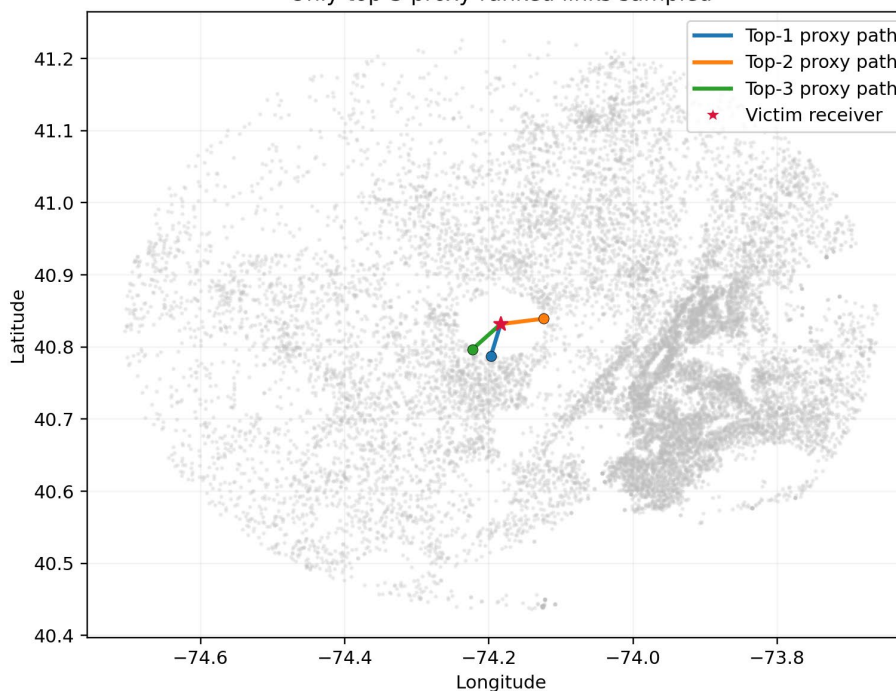
Feature Engineering

- Ex - only evaluate line-of-sight from top N links based on a free-space-like power function

Representative scene A-02800: pruning before terrain/LOS analysis

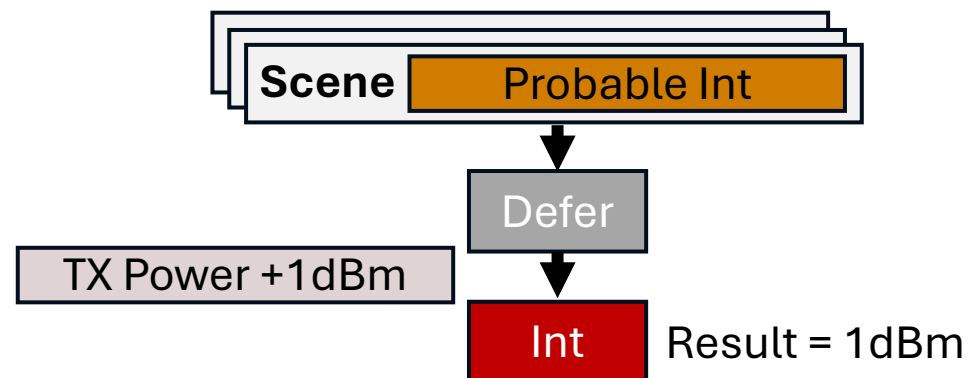
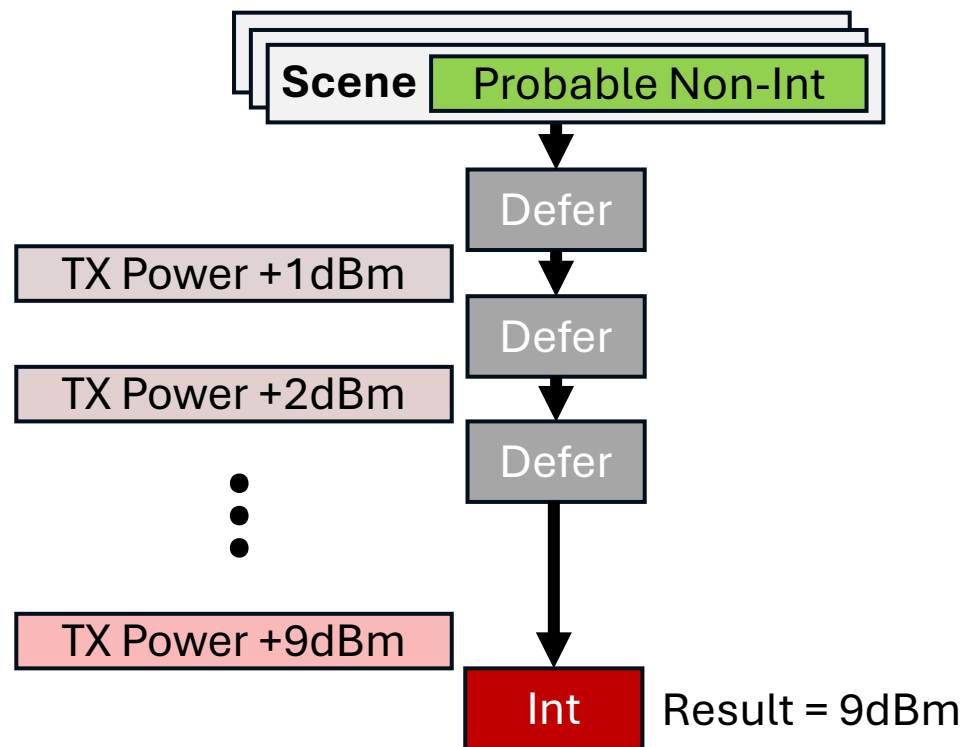


Model terrain/LOS path work
Only top-3 proxy-ranked links sampled



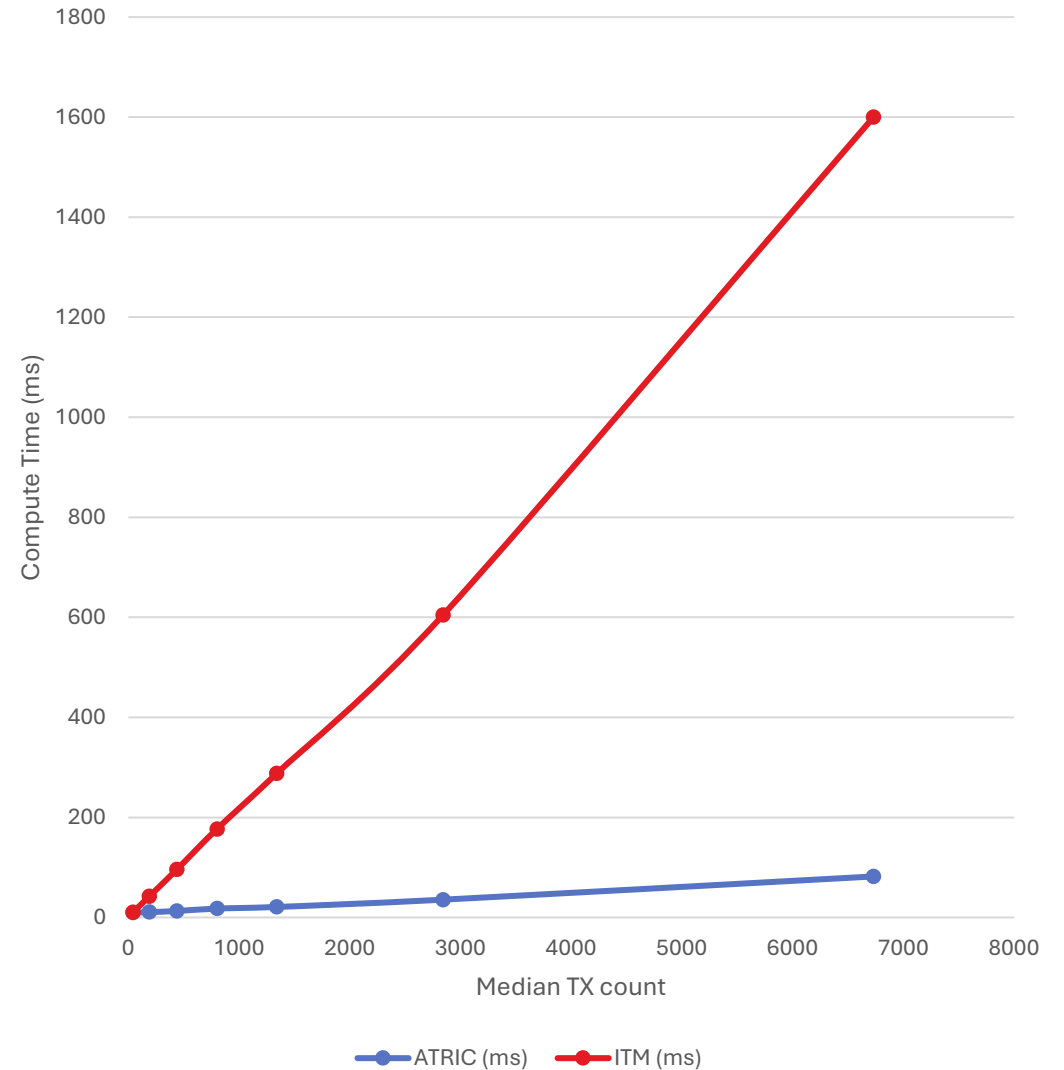
Feature Engineering

- Ex - informing fine sieve by varying coarse sieve TX power to see when it changes its prediction



Results - Speed

- Tested vs running ITM once against every link in a scene, both with cold cache
- ITM linearly adds about $\sim 240\mu\text{s}$ each TX
- ARTIC optimizes as TX count increases
 - $\sim 240\mu\text{s}/\text{TX}$ like ITM early on, $\sim 10\mu\text{s}/\text{TX}$ at $>5\text{k}$ TXs



Results - Accuracy

- ~81% accuracy vs ITM predictions using a 5,000-scene validation set
- ~86% accuracy allowing for near-correct predictions (e.g., Int $\leftrightarrow$ Prob Int)
- ~2% catastrophic misprediction (Int $\leftrightarrow$ No Int)
 - ~0.3% when limited to only high-confidence scenarios

n=5000	No Int	Prob No	Prob Int	Int (Pred)
No Int	810	139	29	19
Prob No	39	791	54	45
Prob Int	31	199	258	27
Int (Real)	77	247	63	2172
Precision	85%	57%	64%	96%
Recall	81%	85%	50%	85%

Strengths/Weaknesses

Core strengths

- **Functional** – proof-of-concept typically accurate
- **Fast** – soundly beats ITM when there's lots of TXs
- **Flexible** – can be retrained to different scenarios, and architectures can be mix and matched

Core weaknesses

- **Limited data** – synthetic, terrestrial-only, purely isotropic antennae, no near-band, purely proof-of-concept
- **Risky** – 2% catastrophic misprediction is too high for a real-world deployment

Next Steps

- Soon: data + software + paper release
- Future potential directions:
 - Feature analysis → what does this mean for ITM?
 - Feature performance/explainability
 - Incorporating features used in band studies/spectrum assignment
 - Temporal scenes (e.g., a transmitter on a schedule)
 - Validation against limited real-world datasets
 - ...

Q&A



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Thank you

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