

RNO-G Simulation Progress towards a Diffuse Neutrino Search



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Outline:

- What we have:
 - 5 Years of Physics data
- What we want:
 - Ultra High Energy Neutrinos (**UHEN**)
- What we need:
 - Simulation to match the detector

This Presentation will cover:

- RNO-G simulation update
- Introduction on diffuse search
- Analysis Simulation plan

Simulation structure: NuRadioMC

Event Generation

Generate neutrino
properties

Signal Generation

Generate
secondary
showers and
Radio Signal

Signal propagation

Calculate signal
path and
amplitude based
on Ice model

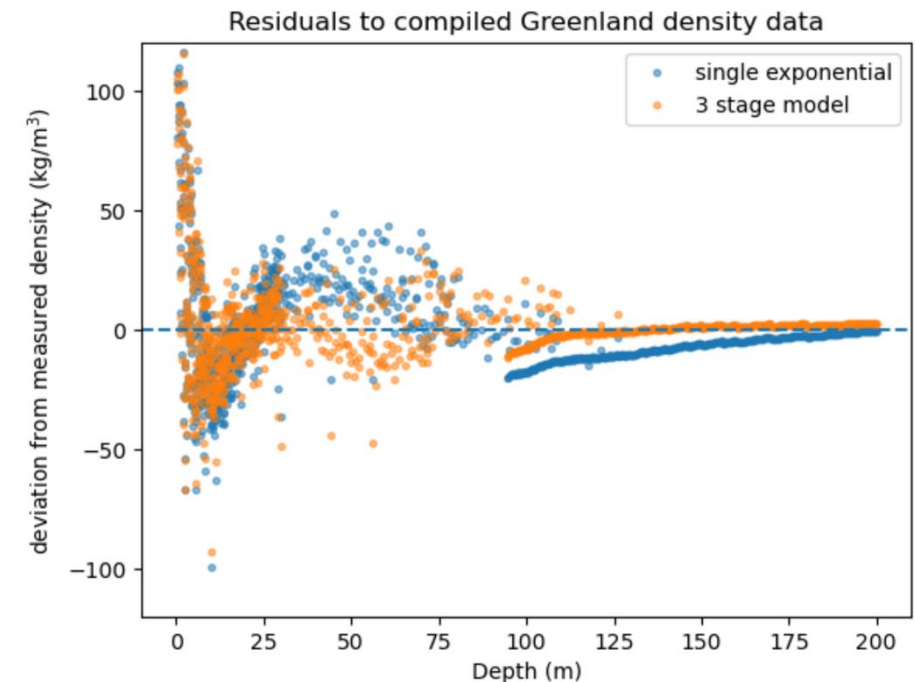
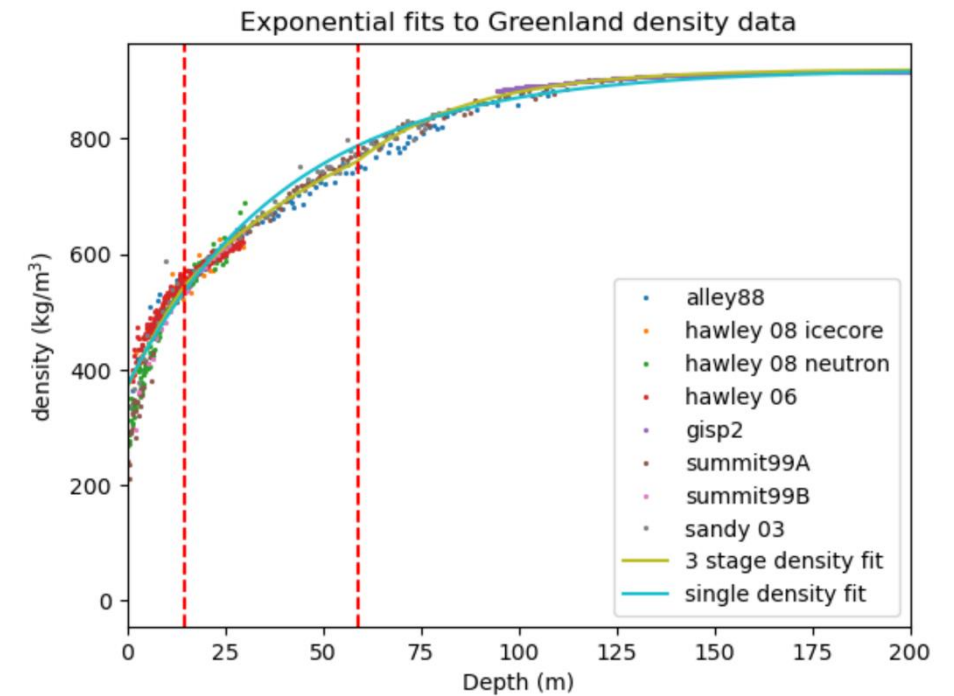
Detector simulation

Readout signal
based on
Hardware
response

Signal Propagation

3 Part Ice model adoption

- updated ice model to better match density measurements
- computation time of new method is reasonable
- only minor changes to effective volume of detector
- better ice model = decreased uncertainties on reconstruction

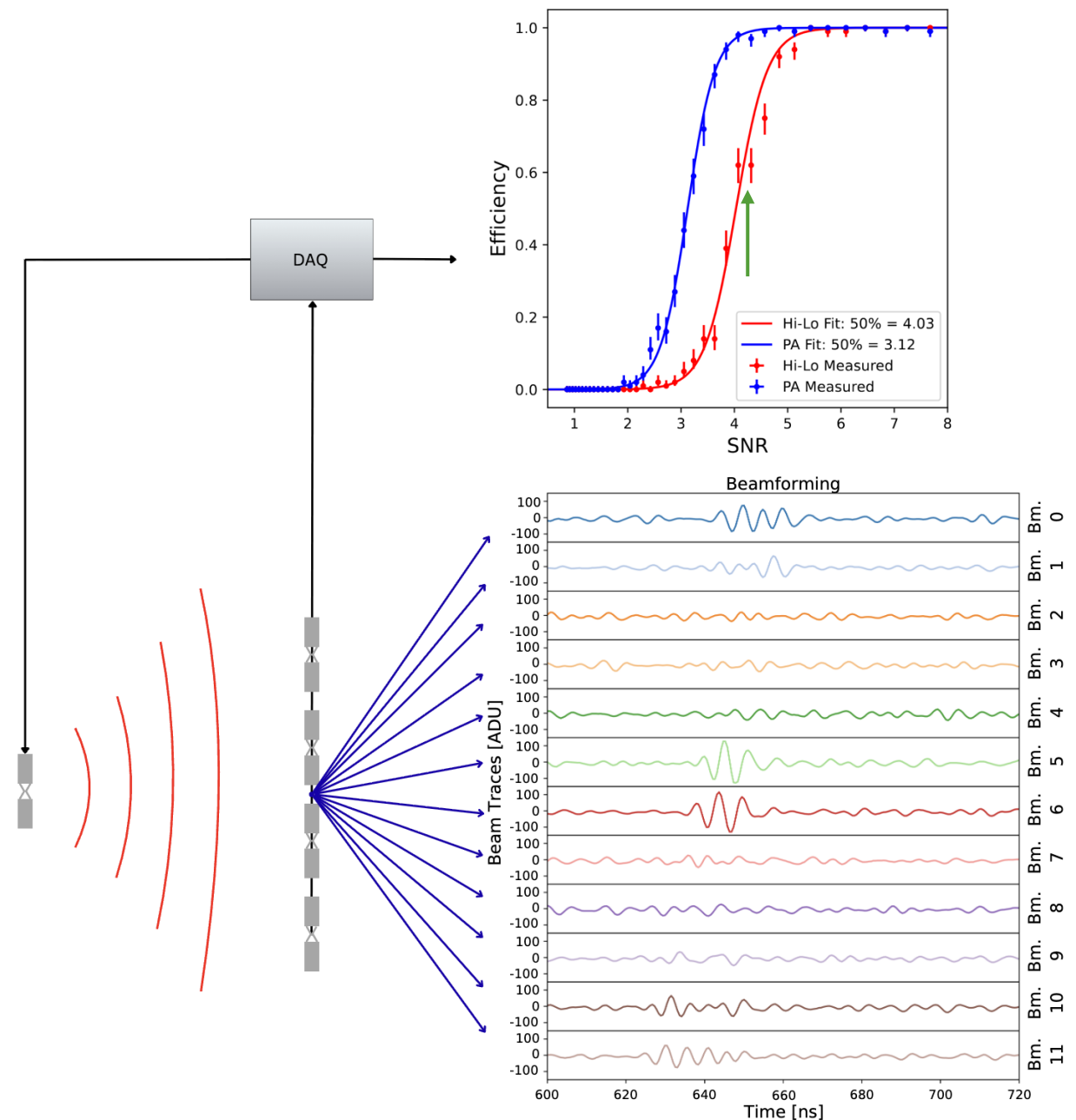


Trigger design: Phased array

Beamforming:

- Linearly adds signals in specific directions
- Boosts SNR by $\sqrt{N}$

This trigger was deployed in the
2024 off-season



Likelihood-based reconstruction

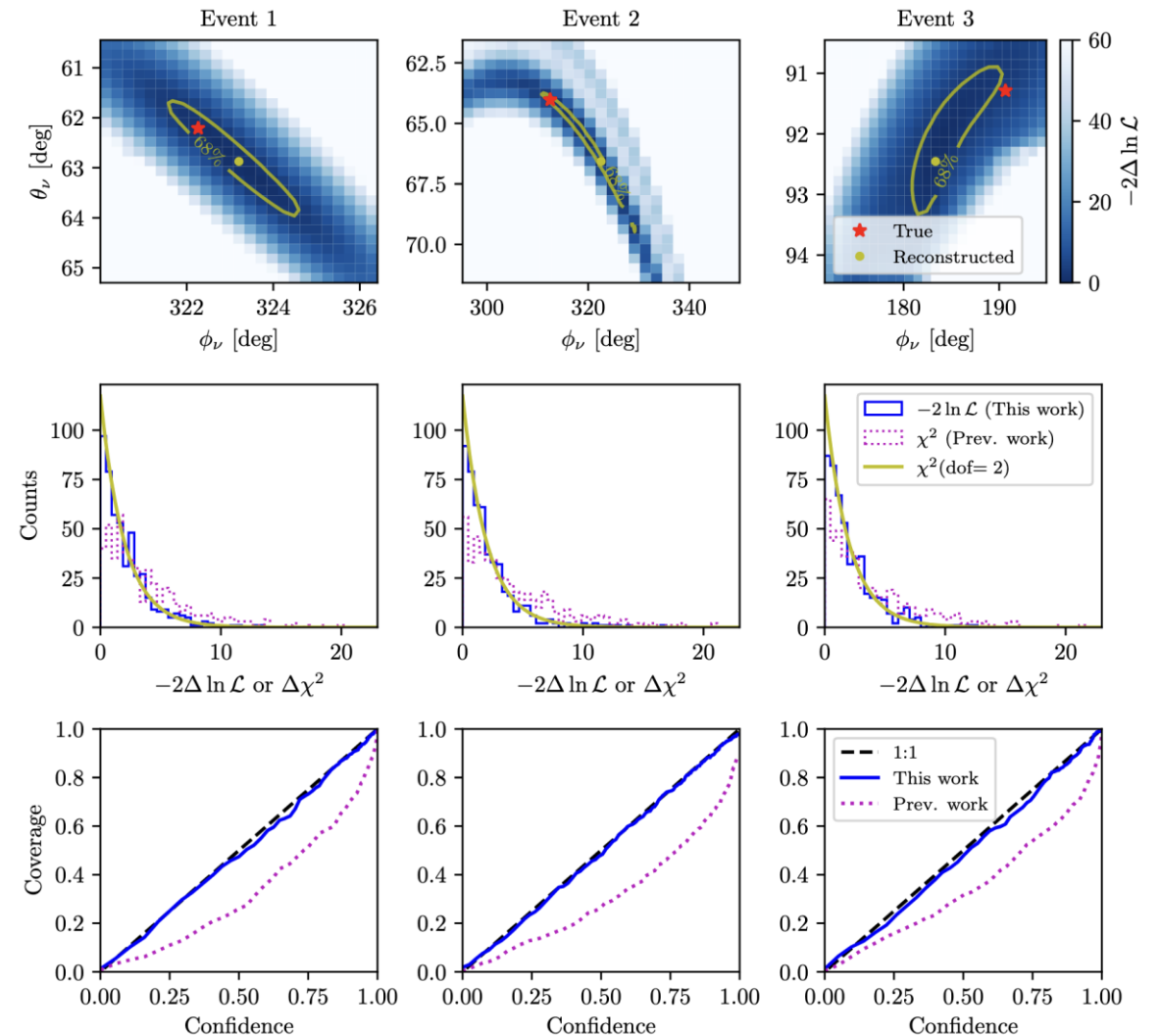
Find best fit by minimizing $-2 \ln \mathcal{L}$

$\mathcal{L}$: Likelihood

Neutrino reconstruction result based on new method

- Improve reconstruction precision
- Fast reliable uncertainty estimates
- Improve signal efficiency for future detectors

New feature, Currently not used in diffuse search



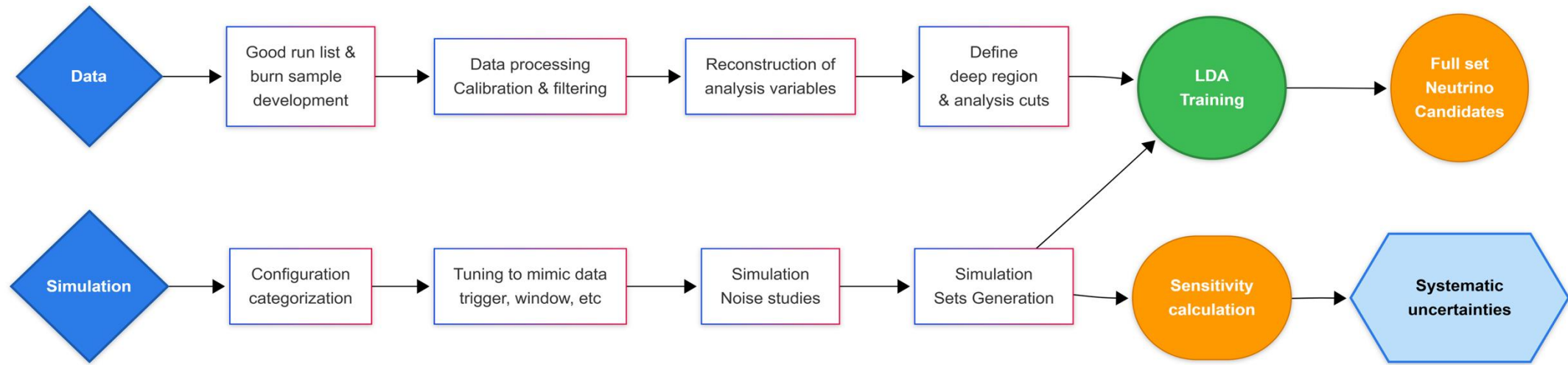
Neutrino diffuse search strategy

Definition of “diffuse”:

Search for neutrinos from any direction, staying agnostic to specific sources

Objective:

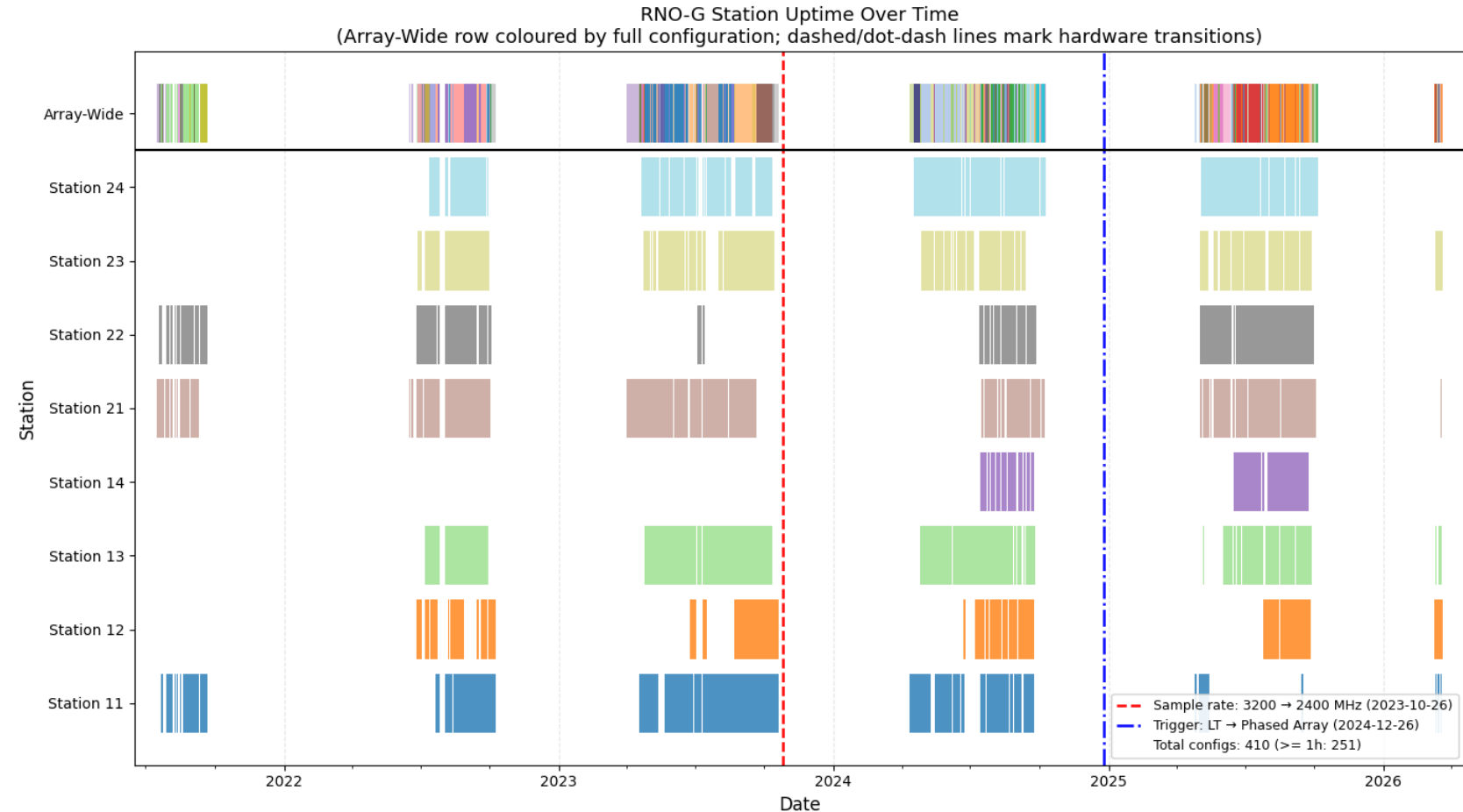
Generate possible neutrino candidates and provide a limit on the astrophysical neutrino flux



Simulation Plan

- RNO-G has run in many configurations as stations have come online
- this will only continue as RNO-G grows to 35 total stations
- need to carefully consider a simulation strategy to reflect this

Current results in configuration categorization

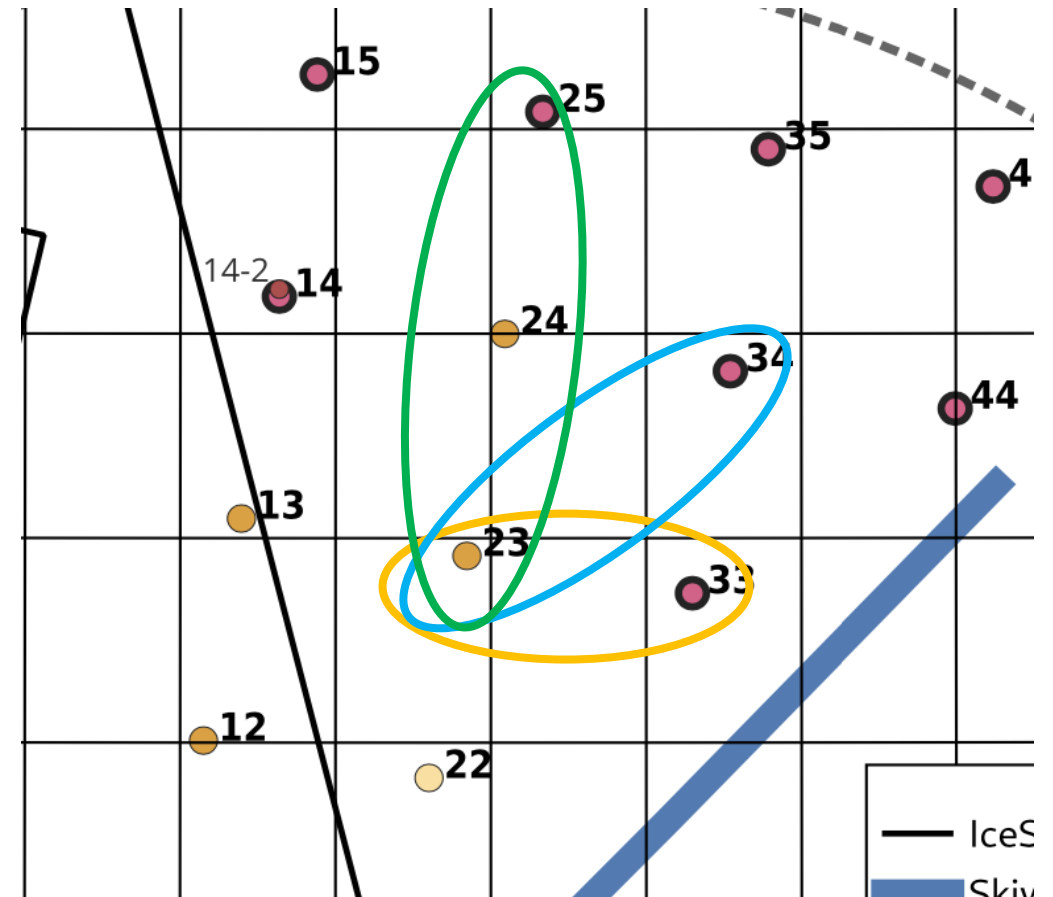


Simulation Strategy

- Major RNO-G station updates always applied off-season across all stations
- Reduce overcounting effect by computing neighboring pairs of single station sim by γ
- *Neighbor relationship: γ*
 - Orange: $\gamma = 1st$ Nearest
 - Blue: $\gamma = 2nd$ Nearest
 - Green: $\gamma = 3rd$ Nearest

Scalable! Reusable!

No multi-station configuration required!



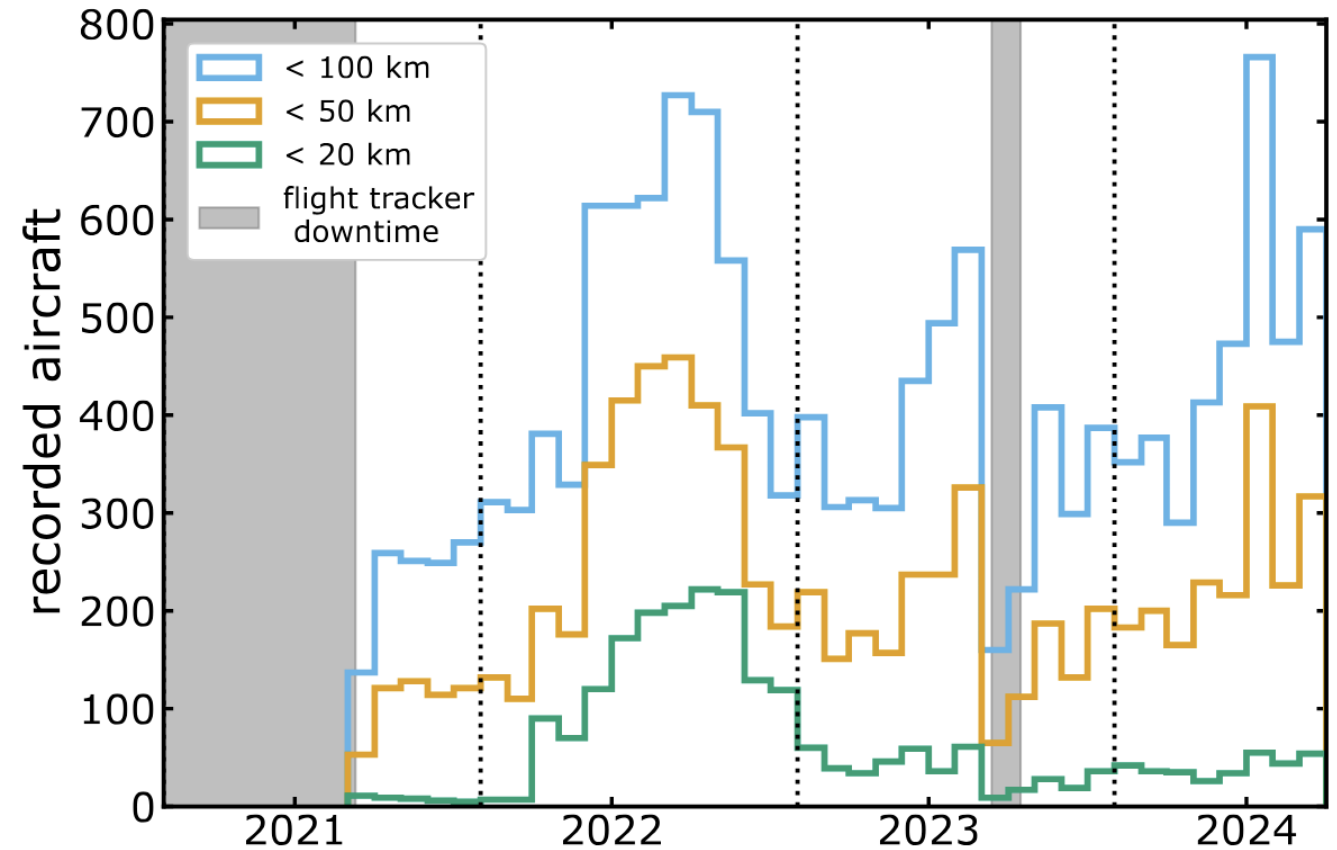
Data Cleaning

Known background sources based Cuts:

- Airplane Cut
- Spatiotemporal Cut
- Wind Cut
- Direction Cut
- Intra-run Rate Cut
- Glitches event cut

Example:

Airplane flying near the station
picked up by station plane tracker



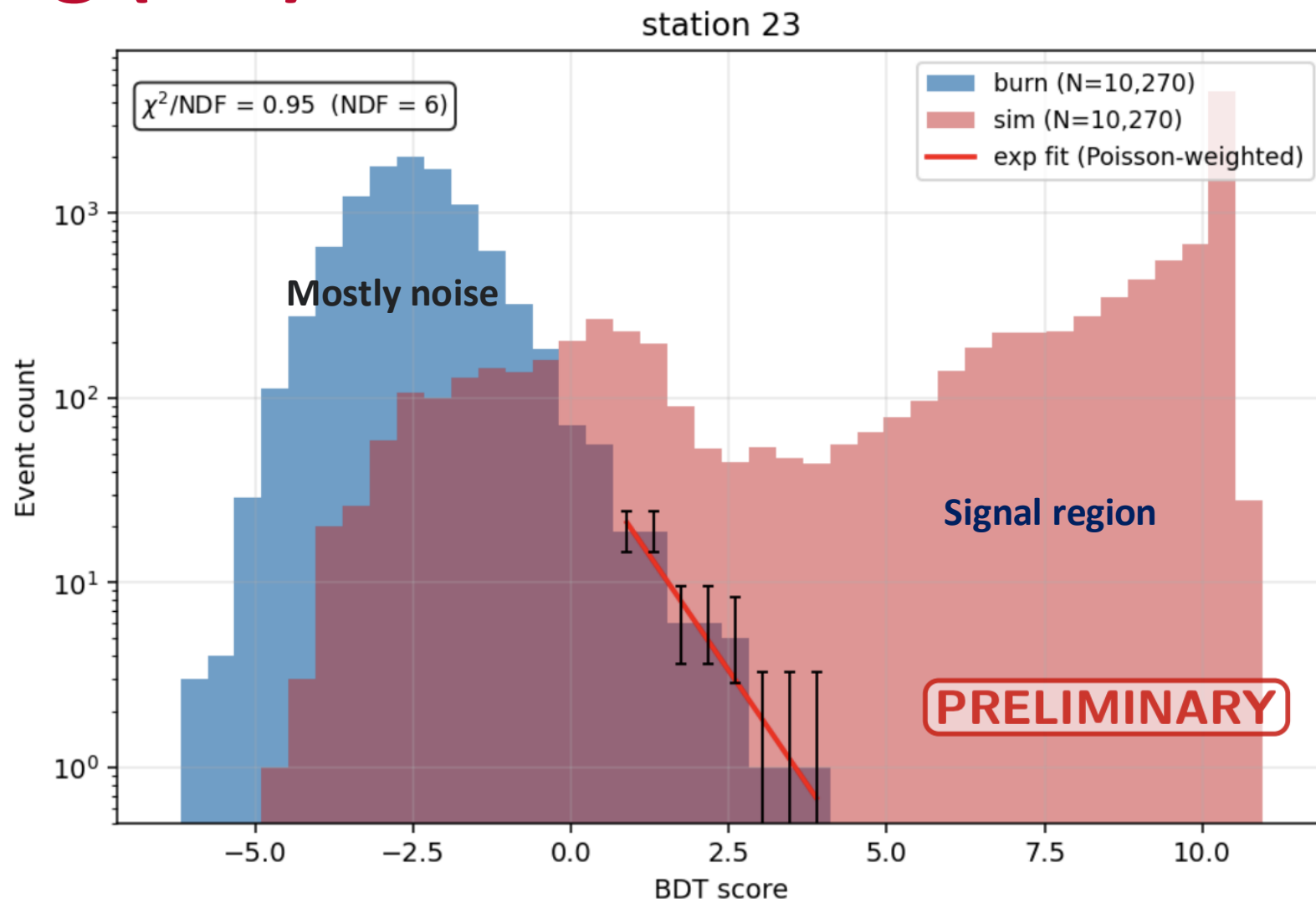
Machine learning (ML)

various ML techniques are informing our cut strategy

(Deep CR as an example)

here, a boosted decision tree is distinguishing detector noise from simulated impulses

next step is to integrate this method into full neutrino search pipeline



Future plan of RNO-G

- RNO-G is currently drilling for more stations in Greenland right now
- Deep CR Analysis and diffuse neutrino search is on-going

Thanks for Every RNO-G member's contribution!

Hope RNO-G can bring us more physics on UHEN in the near future



Detector response

Gain Calibration

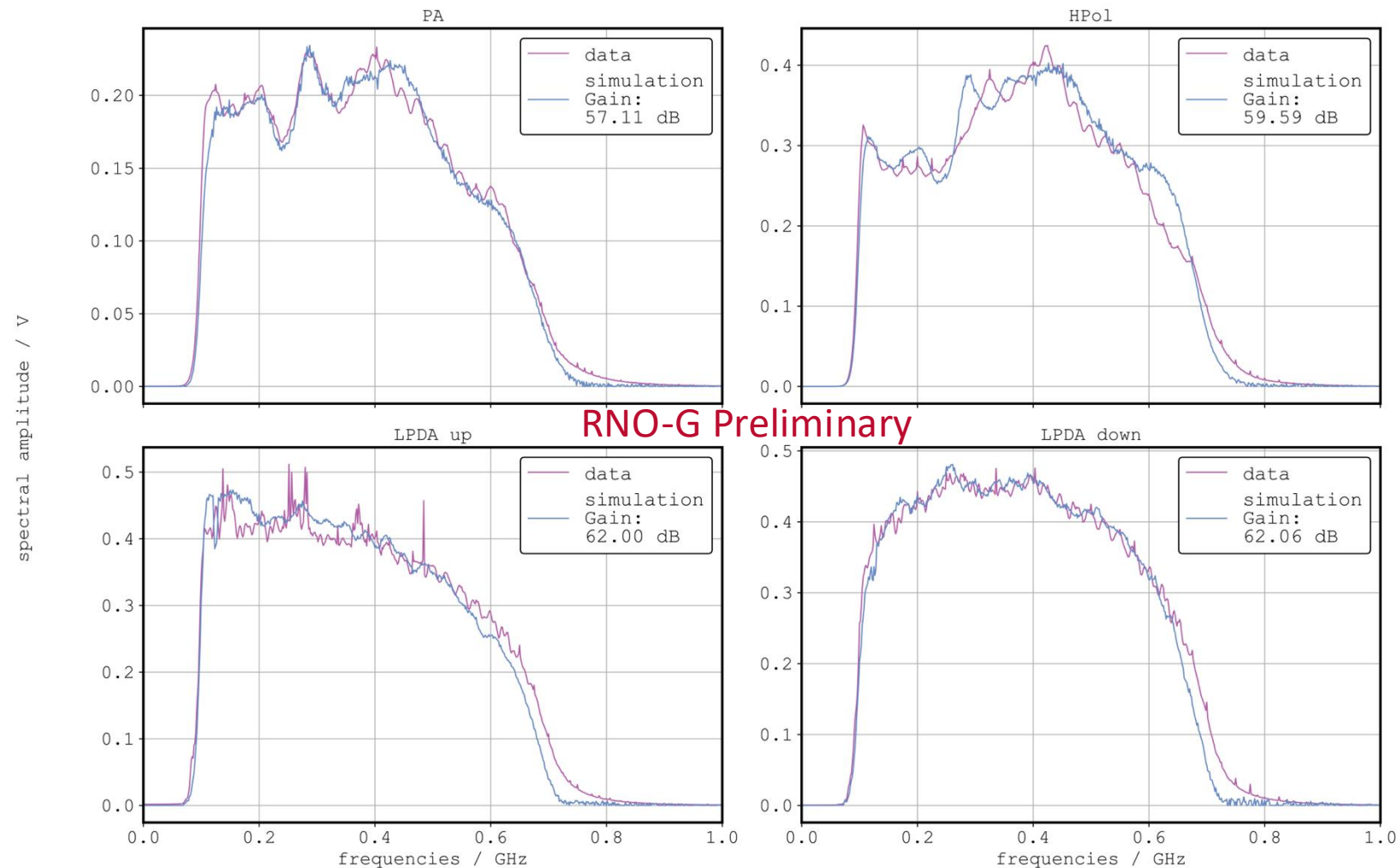
Gain Calibration is conducted

- per channel
- per operation season

Currently calibrated:

Season: 2022 - 2024

Credit: Ruben Camphyn, Felix Schlüter, Simona Toscano



Likelihood-based reconstruction

Find best fit by minimizing $-2 \ln \mathcal{L}$

$$-2 \ln \mathcal{L}(\boldsymbol{\mu}(\boldsymbol{\theta}); \mathbf{x}, \boldsymbol{\Sigma}) = \\ n_t \ln(2\pi) + n_t \ln(|\boldsymbol{\Sigma}|) + (\mathbf{x} - \boldsymbol{\mu}(\boldsymbol{\theta}))^T \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu}(\boldsymbol{\theta})).$$

$\mathcal{L}(\boldsymbol{\mu}(\boldsymbol{\theta}); \mathbf{x}, \boldsymbol{\Sigma})$: likelihood

$\mathbf{x}$: trace

$\boldsymbol{\mu}$: signal

$\boldsymbol{\Sigma}$: covariance matrix

$\boldsymbol{\theta} = (E_{\text{shower}}, \theta_v, \phi_v, r_{\text{vertex}}, \theta_{\text{vertex}}, \phi_{\text{vertex}}, t_0)$

