



Giant Radio Array for Neutrino Detection

Reconstruction of highly-inclined extensive air showers in GRAND

Oscar Macías
(SFSU)

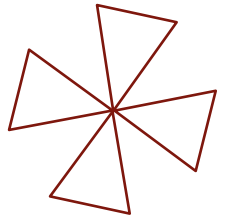
*On behalf of the GRAND
Collaboration*



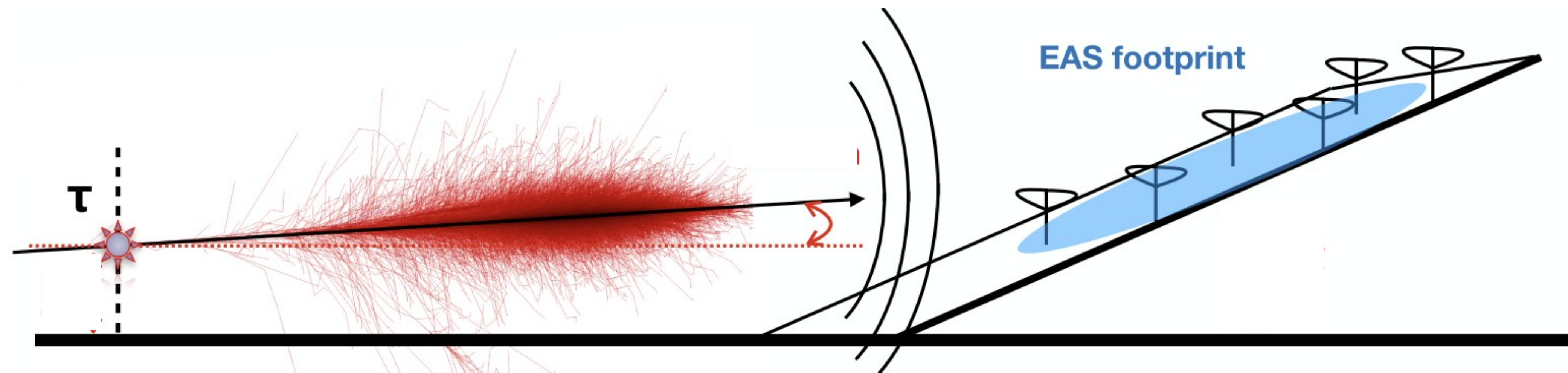
ARENA 2024
KICP, University of Chicago



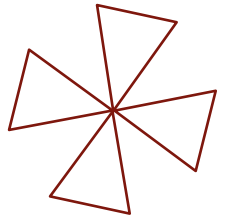
SAN FRANCISCO
STATE UNIVERSITY



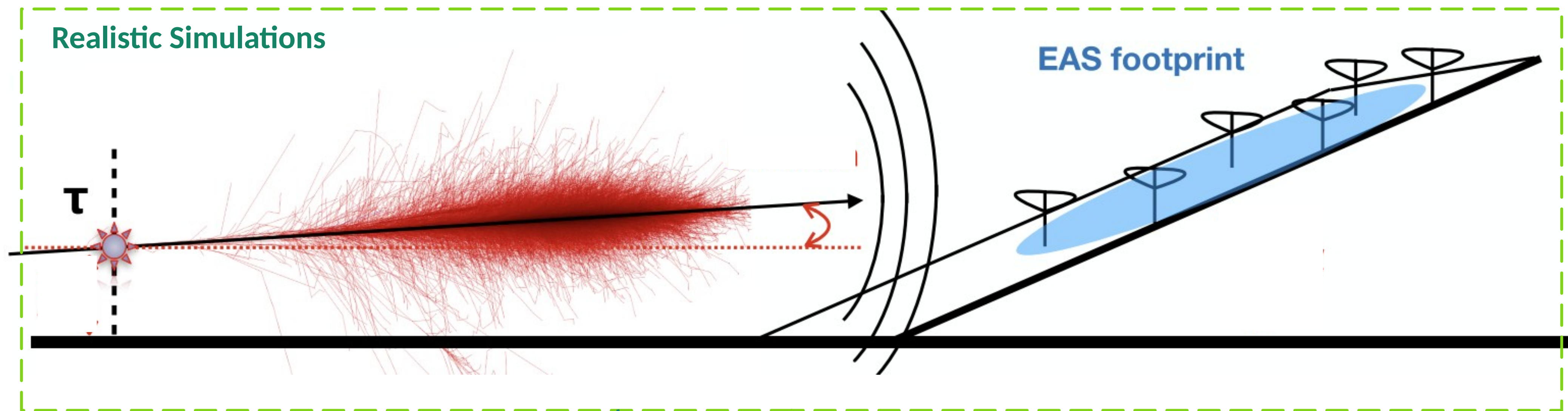
Reconstruction of highly-inclined Air Showers (conventional + ML methods)



Decoene (2021)



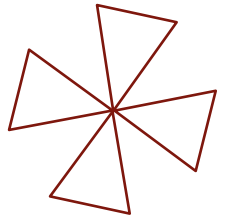
Reconstruction of highly-inclined Air Showers (conventional + ML methods)



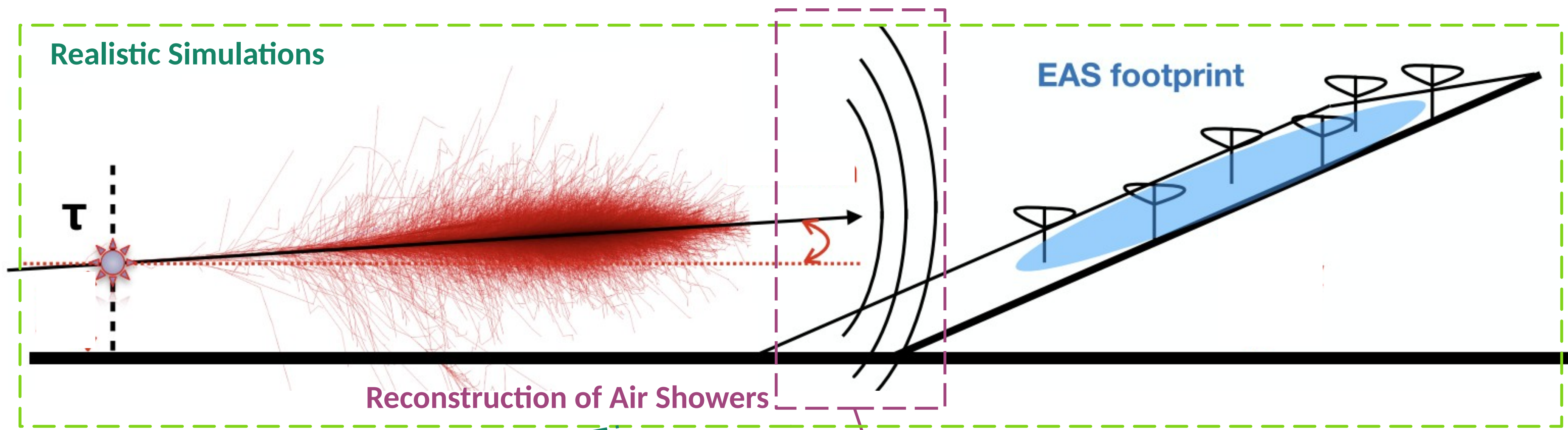
Realistic Data Simulation libraries

- ✓ Include Galactic noise
- ✓ Include antenna response + RF chain + GPS jitter
- ✓ More than 20,000 simulations





Reconstruction of highly-inclined Air Showers (conventional + ML methods)



Decoene (2021)

Realistic Data Simulation libraries

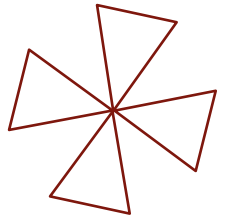
- ✓ Include Galactic noise
- ✓ Include antenna response + RF chain + GPS jitter
- ✓ More than 20,000 simulations

Reconstruction of Air Showers

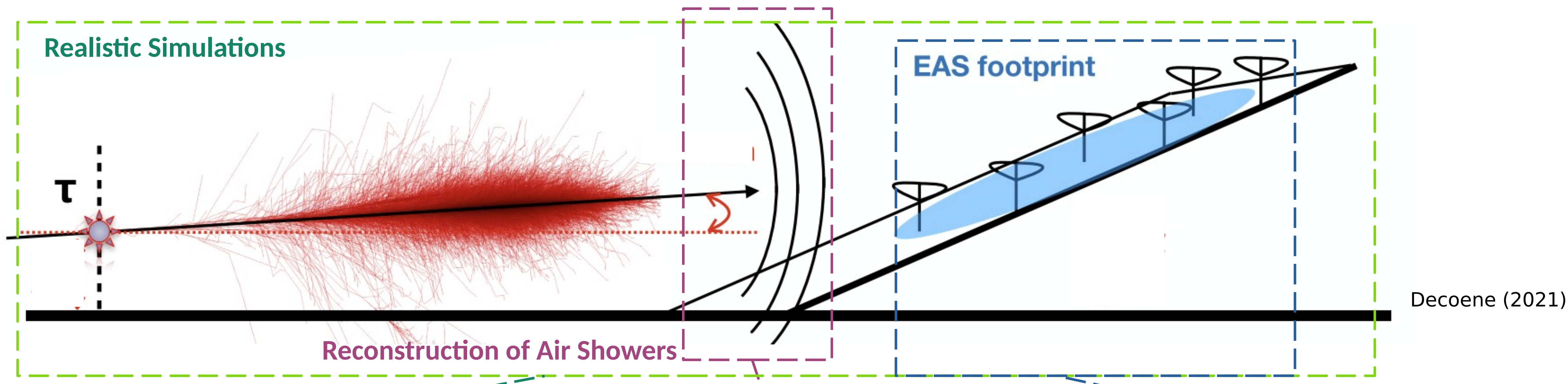
- ✓ Plane Wave Front (PWF): fast timing & direction reconstruction
- ✓ Fitting (empirical and Physics informed) of Angular Distribution Function: more precise shower parameter reconstruction
- ✓ Empirical fitting of lateral distribution function
- ✓ Graph Neural Networks for EAS studies



see talks by L. Gülzow & J. Köhler



Reconstruction of highly-inclined Air Showers (conventional + ML methods)



Realistic Data Simulation libraries

- ✓ Include Galactic noise
- ✓ Include antenna response + RF chain + GPS jitter
- ✓ More than 20,000 simulations

Reconstruction of Air Showers

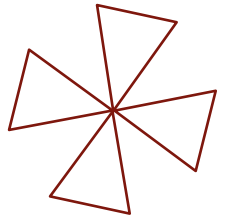
- ✓ Plane Wave Front (PWF): fast timing & direction reconstruction
- ✓ Fitting (empirical and Physics informed) of Angular Distribution Function: more precise shower parameter reconstruction
- ✓ Empirical fitting of lateral distribution function
- ✓ Graph Neural Networks for EAS studies

Electric field reconstruction

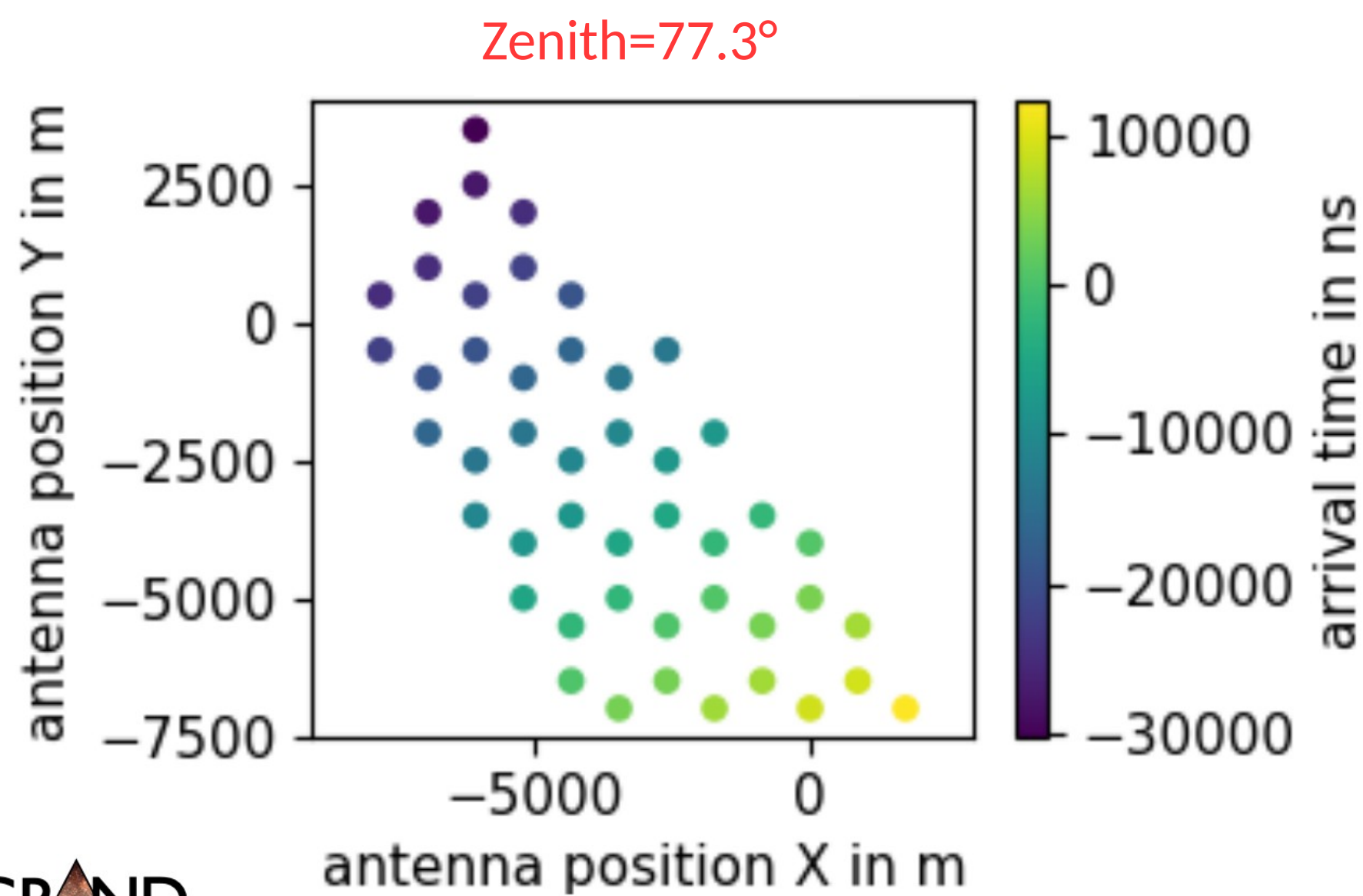
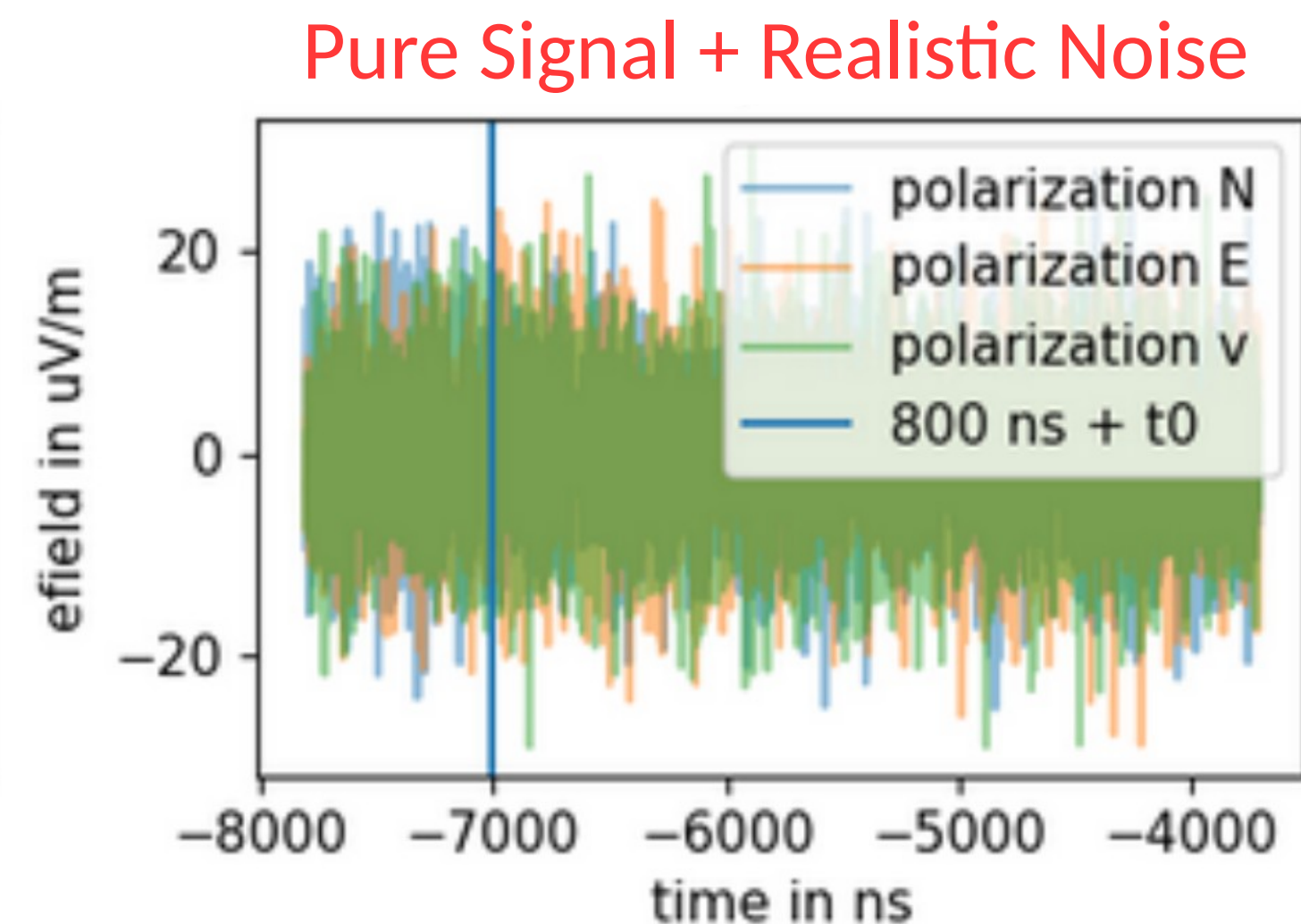
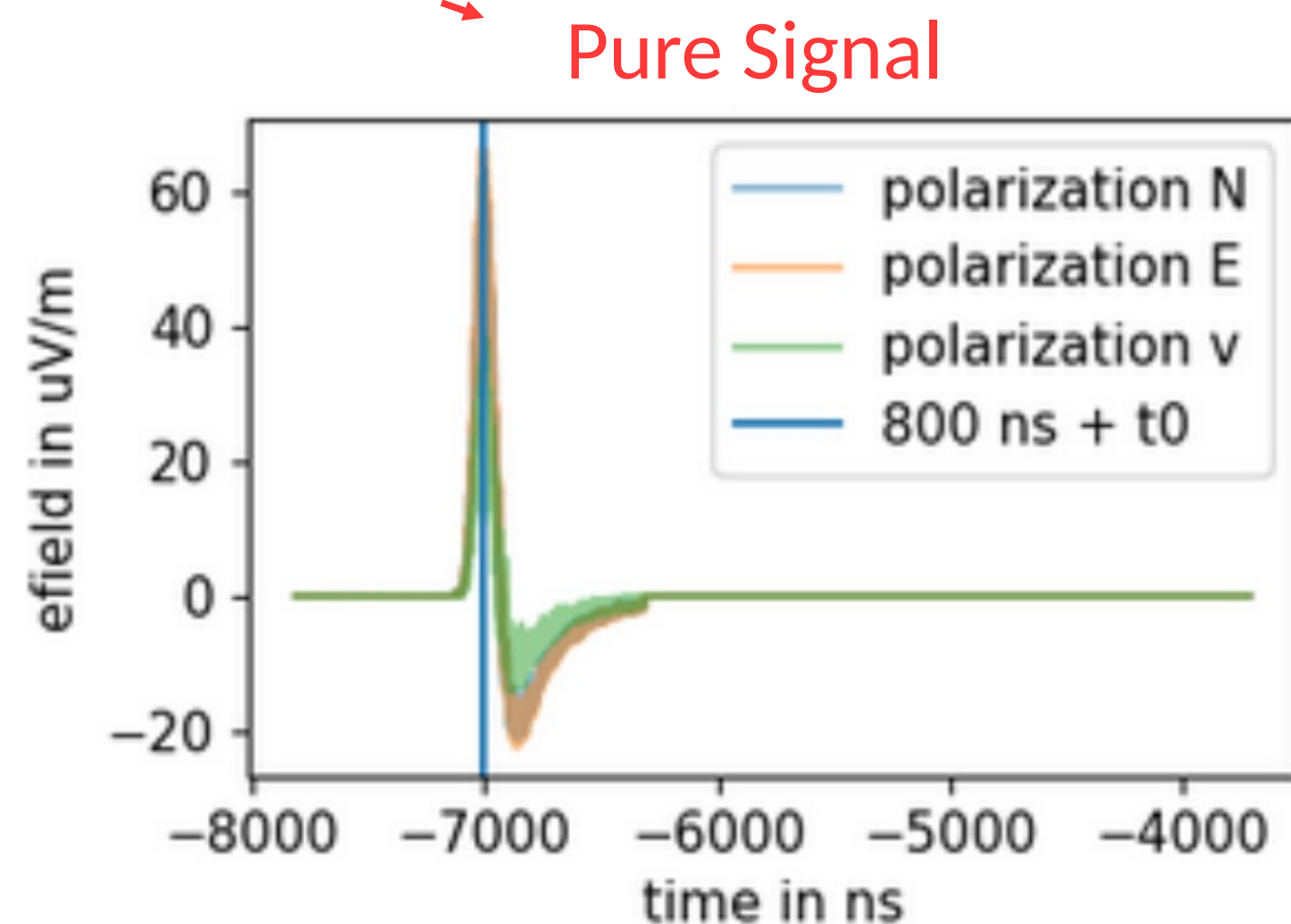
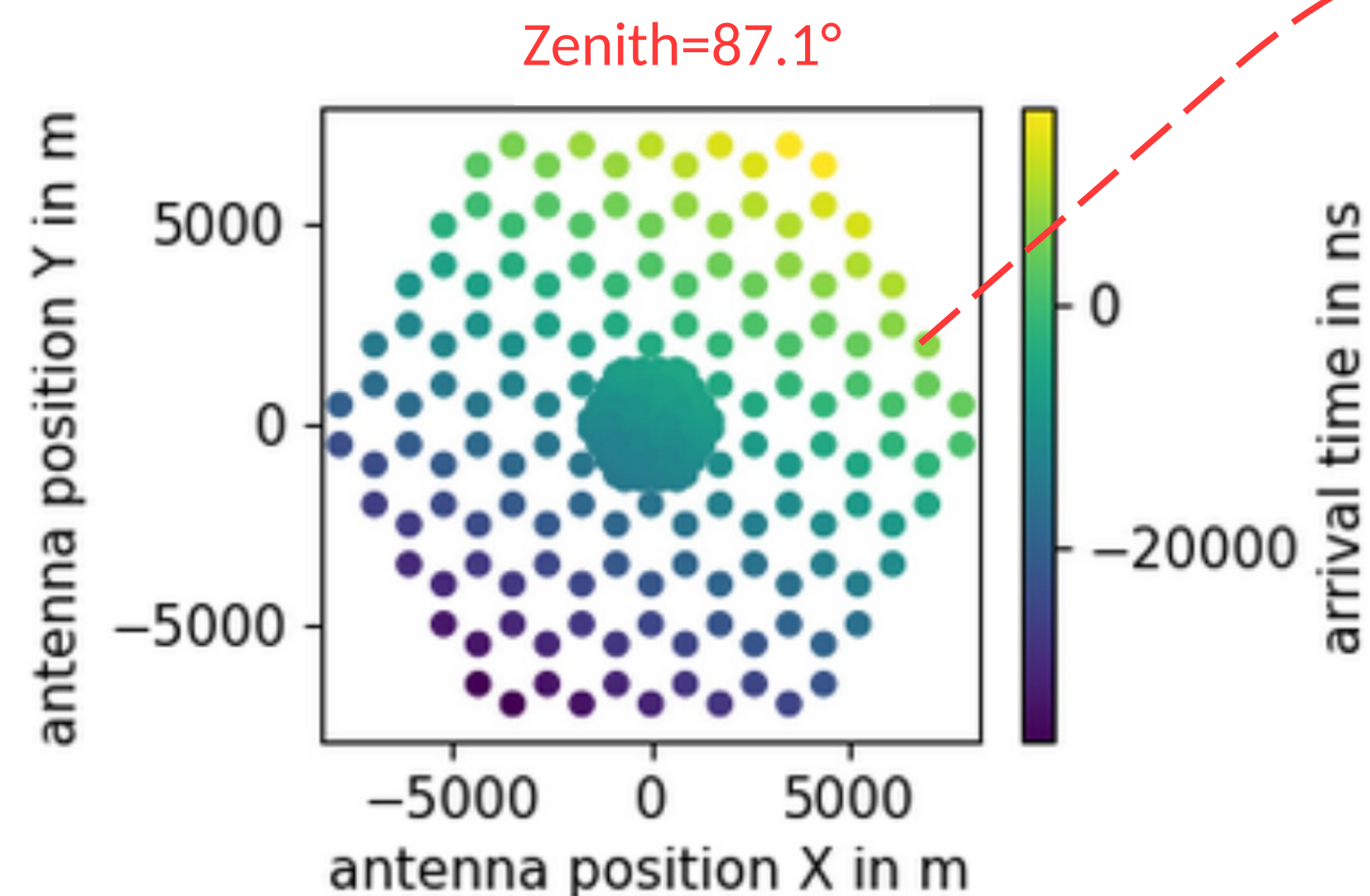
- ✓ E-field reconstruction with CNN
- ✓ Direction reconstruction based on polarization
- ✓ Denoising of E-field/ADC using ML



see talks by L. Gülzow & J. Köhler



GRAND Data Challenge 2: a complete realistic simulation library



DC2 cosmic ray simulation set:

- ZHAireS and CoREAS Simulations
- Pure Signals: without noise for ML applications
- Pure Signals + Hardware-like jitter
(this is going to be the "Data Challenge 2" blind data)

Included in simulations with jitter:

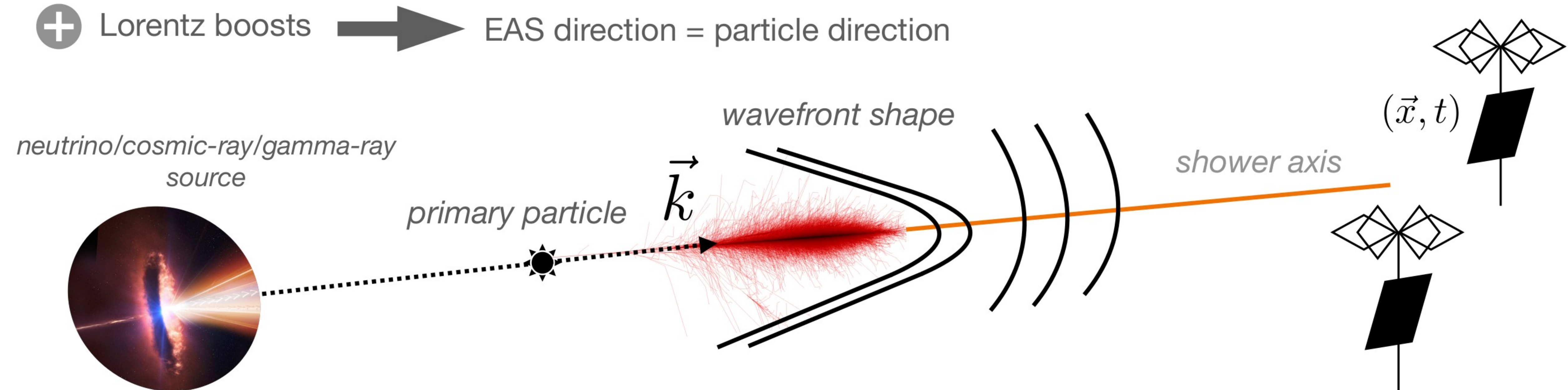
- ◆ 5 ns Gaussian smeared "trigger" time (GPS jitter)
- ◆ "Amplitude Calibration" Gaussian smeared 7.5%
- ◆ Galactic Noise

Note: Only illuminated antennas are simulated

Study of the wavefront shape

The radio wavefront allows to reconstruct the EAS direction

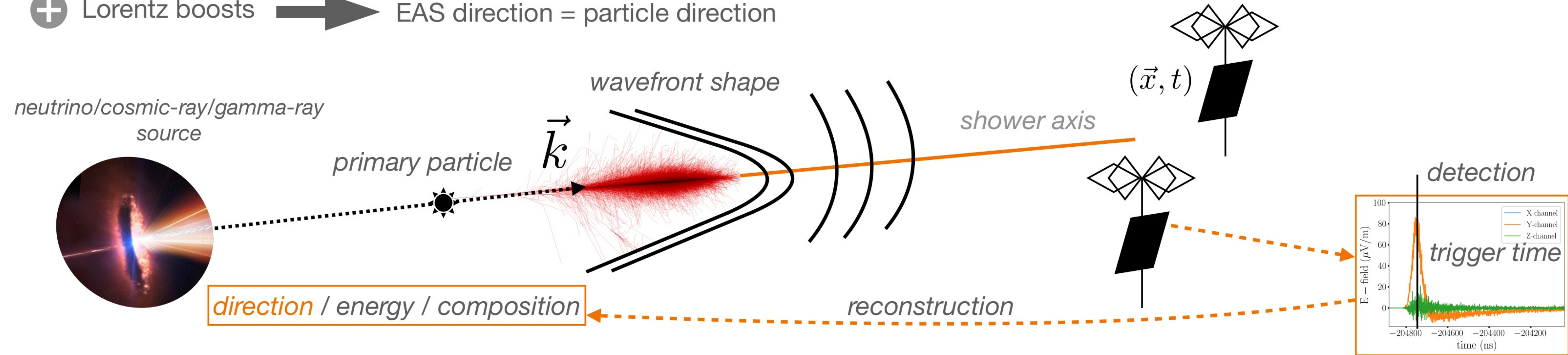
 Lorentz boosts  EAS direction = particle direction



Study of the wavefront shape

The radio wavefront allows to reconstruct the EAS direction

+ Lorentz boosts → EAS direction = particle direction



Method: adjust the wavefront model to the trigger times → **direction accuracy = wavefront shape correctness**

see talk by Kumiko Kotera

EAS Reconstruction Procedure

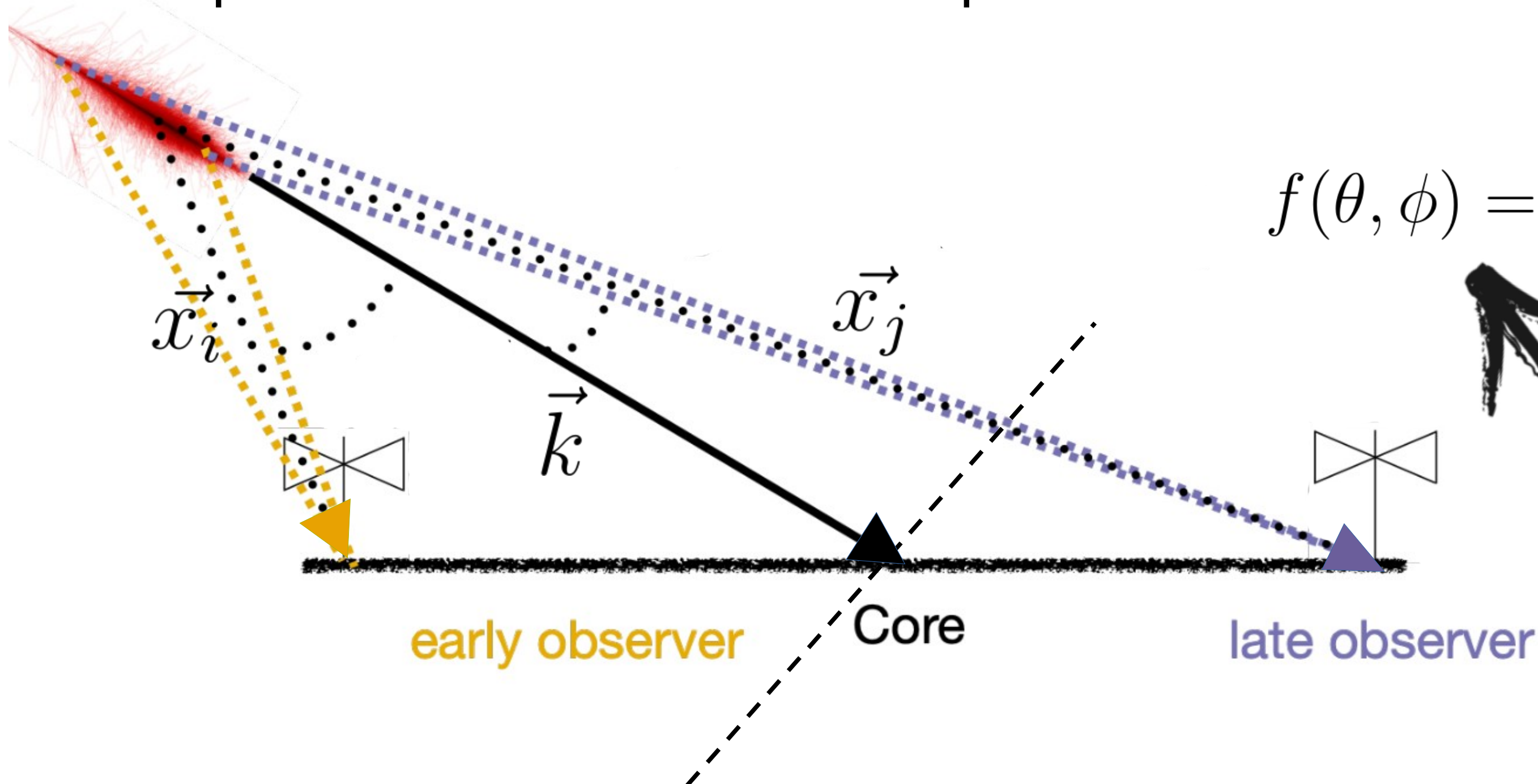


EAS Reconstruction Procedure

1) The plane wave reconstruction

The procedure relies on the comparison of the **relative trigger times** from one antenna to another

Decoene+(2021)



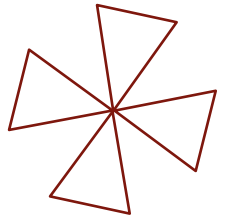
$$f(\theta, \phi) = \sum_{i,j}^{N_{\text{antenna}}} \left[\frac{c}{n} (t_i - t_j) - \vec{k}(\theta, \phi) \cdot (\vec{x}_i - \vec{x}_j) \right]^2$$

Minimize residual differences

Reduces the parameter space from all the directions down to a cone of a few square degrees

$$\theta_{\text{true}} \in [\theta_{\text{plan}} - 2^\circ, \theta_{\text{plan}} + 2^\circ]$$

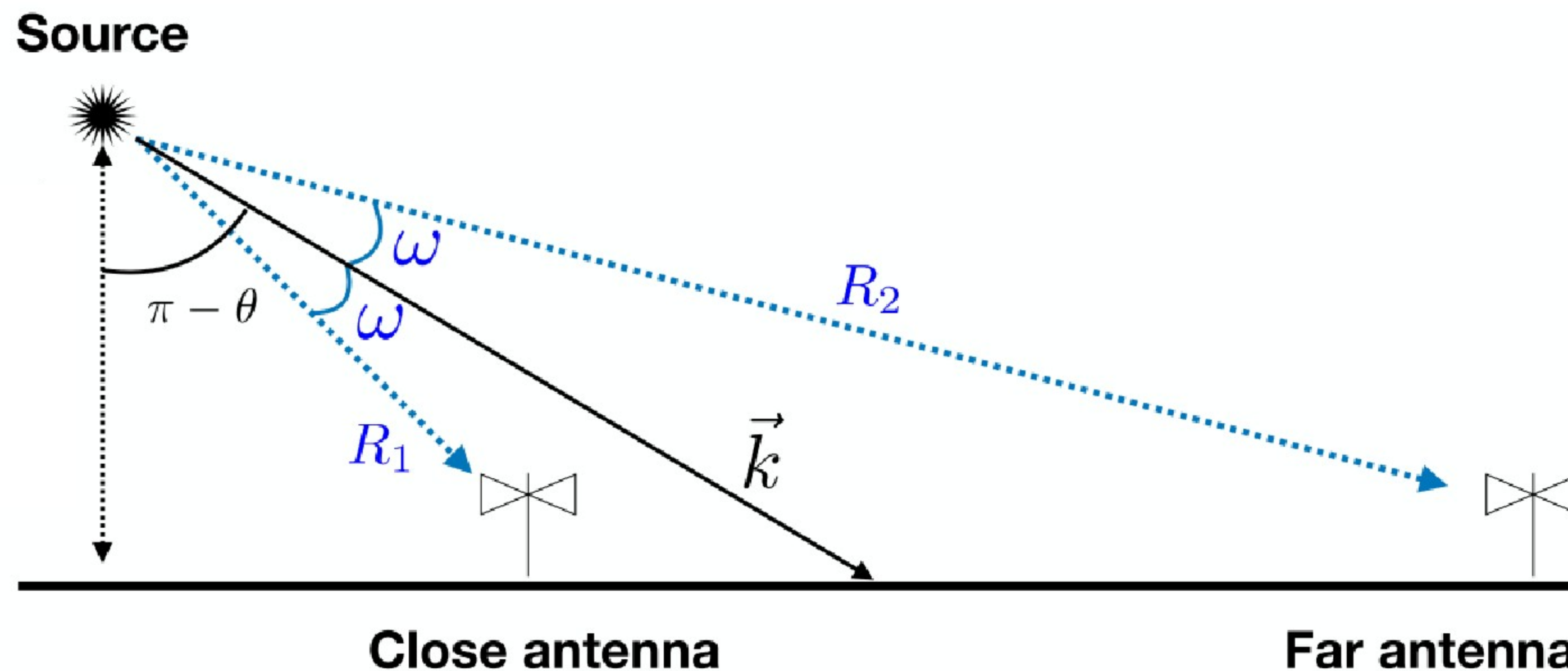
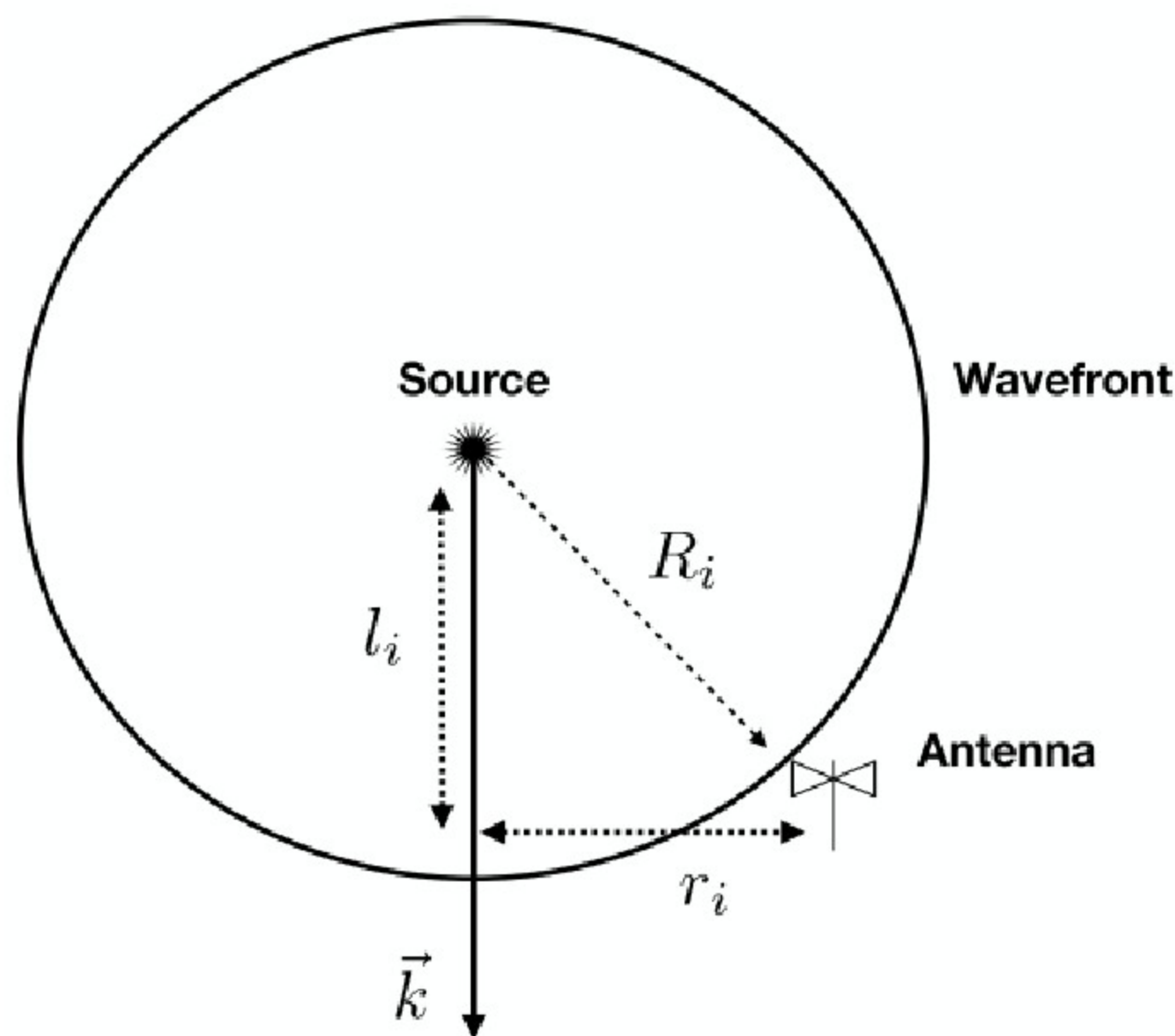
$$\phi_{\text{true}} \in [\phi_{\text{plan}} - 1^\circ, \phi_{\text{plan}} + 1^\circ]$$



EAS Reconstruction Procedure

2) Spherical wave reconstruction

Determines the best position of the point-source through the minimization of $f(\theta, \phi, \rho, t_{\text{source}})$



Decoene (2021)

$$f(\theta, \phi, \rho, t_{\text{source}}) = \sum_i^{N_{\text{antenna}}} \left(\frac{c}{n_i} (t_i - t_{\text{source}}) - \sqrt{(x_i - x_{\text{source}})^2 + (y_i - y_{\text{source}})^2 + (z_i - z_{\text{source}})^2} \right)^2$$

$$\begin{aligned} x_{\text{source}} &= \rho \cos(\phi) \sin(\theta), \\ y_{\text{source}} &= \rho \sin(\phi) \sin(\theta), \\ z_{\text{source}} &= \rho \cos(\theta), \end{aligned}$$



EAS Reconstruction Procedure

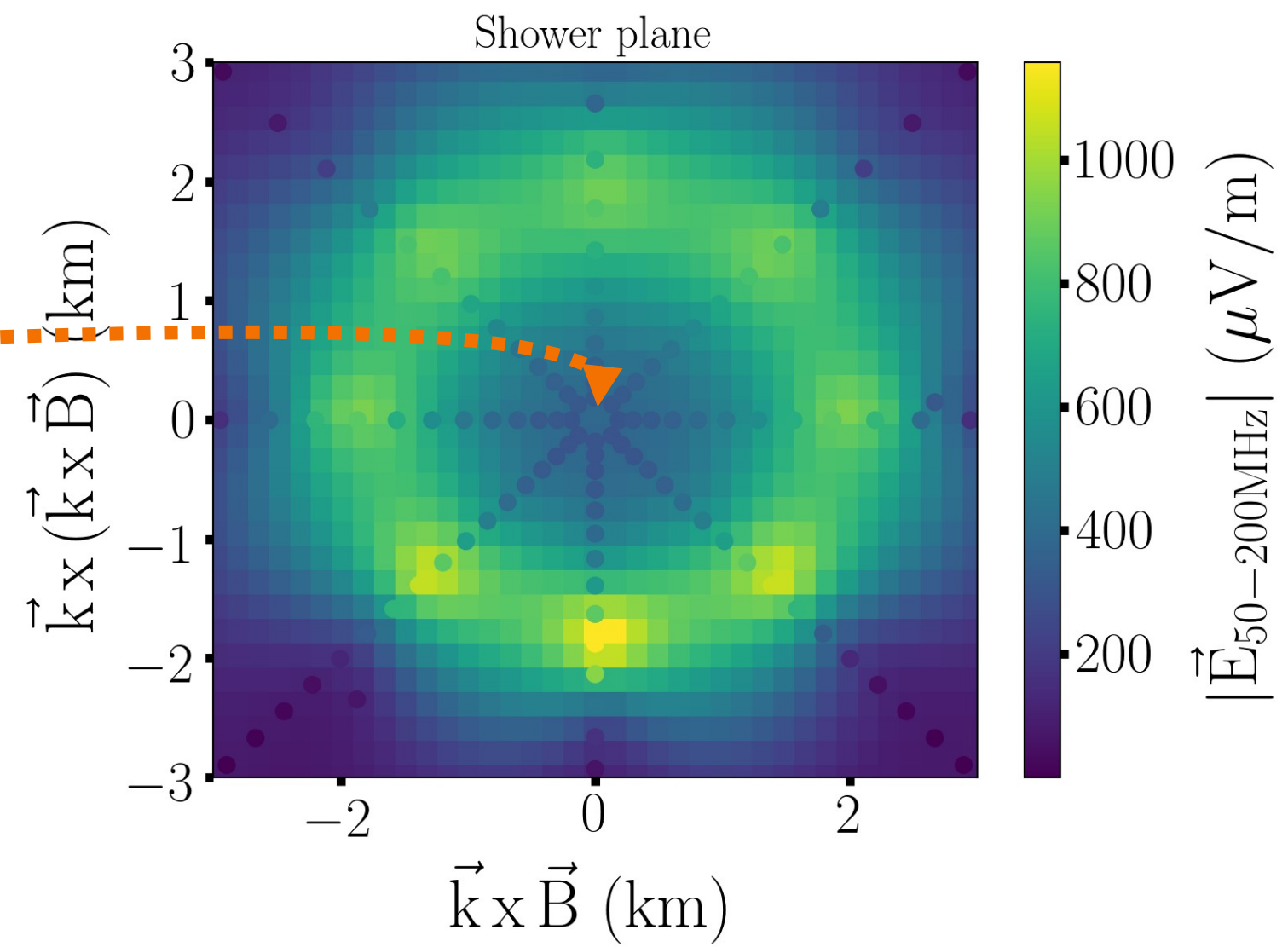
3) Angular Distribution of the Signal

straightforward handle on the core position (hence direction!)

→ beaming effect + Cerenkov effect + asymmetry features (Geomagnetic/Askaryan emissions)

$$f^{\text{ADF}}(\omega, \eta, \alpha, l; \delta\omega, \mathcal{A}) = \frac{\mathcal{A}}{l} f^{\text{GeoM}}(\alpha, \eta, \mathcal{B}) f^{\text{Cerenkov}}(\omega, \delta\omega)$$

empirical model!



- Geomagnetic asymmetry $f^{\text{GeoM}}(\alpha, \eta, \mathcal{B}) = 1 + \mathcal{B} \sin(\alpha)^2 \cos(\eta)$

α magnetic field inclination \mathcal{B} geomagnetic strength η polarisation angle

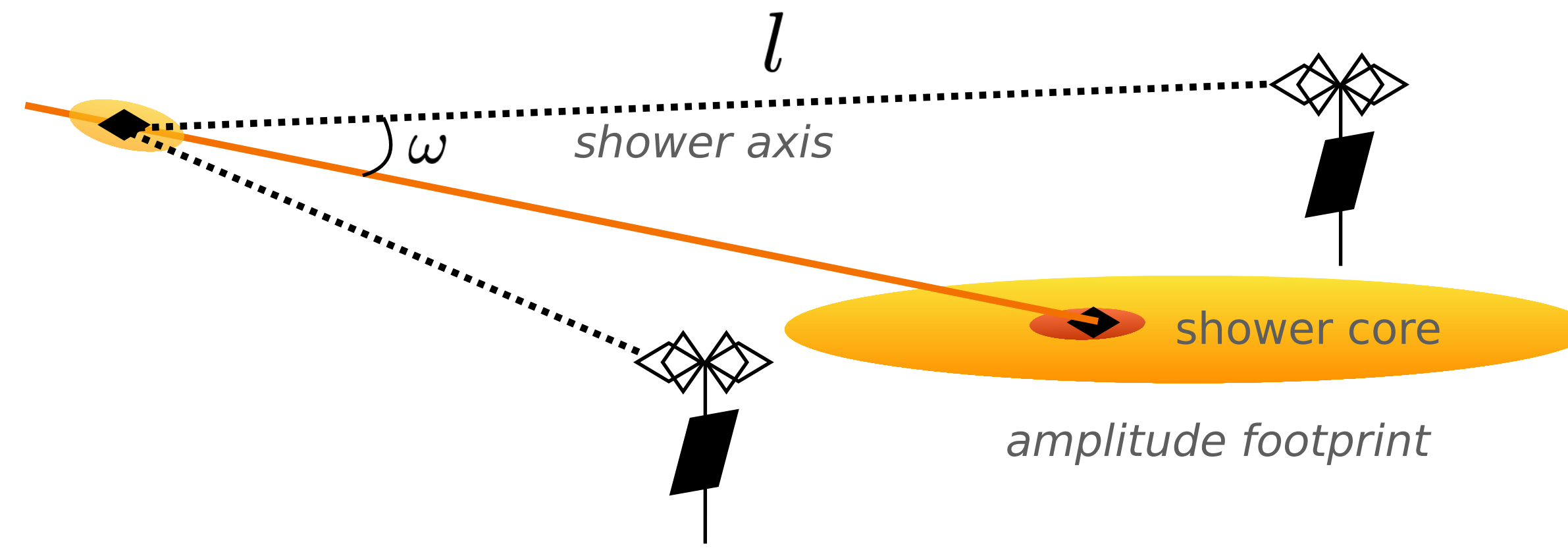
interplay between emission mechanisms
→ signal excess along the Lorentz force direction

- Early-late asymmetry $\frac{\mathcal{A}}{l}$ → energy dilution

- Cerenkov cone $f^{\text{Cerenkov}}(\omega, \delta\omega) = \frac{1}{1 + 4 \left[\frac{(\tan(\omega)/\tan(\omega_C))^2 - 1}{\delta\omega} \right]^2}$

→ geometrical Cerenkov effect description

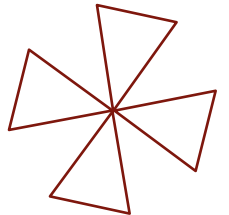
→ ω_C



4 fitting parameters only: $\{\theta, \phi, \mathcal{A}, \delta\omega\}$

Results I: Plane Wave Reconstruction On Data Challenge 2 Simulations





EAS Direction Reconstruction on "Data Challenge 2" Simulations (Analytical solution)

DC2 simulation set:

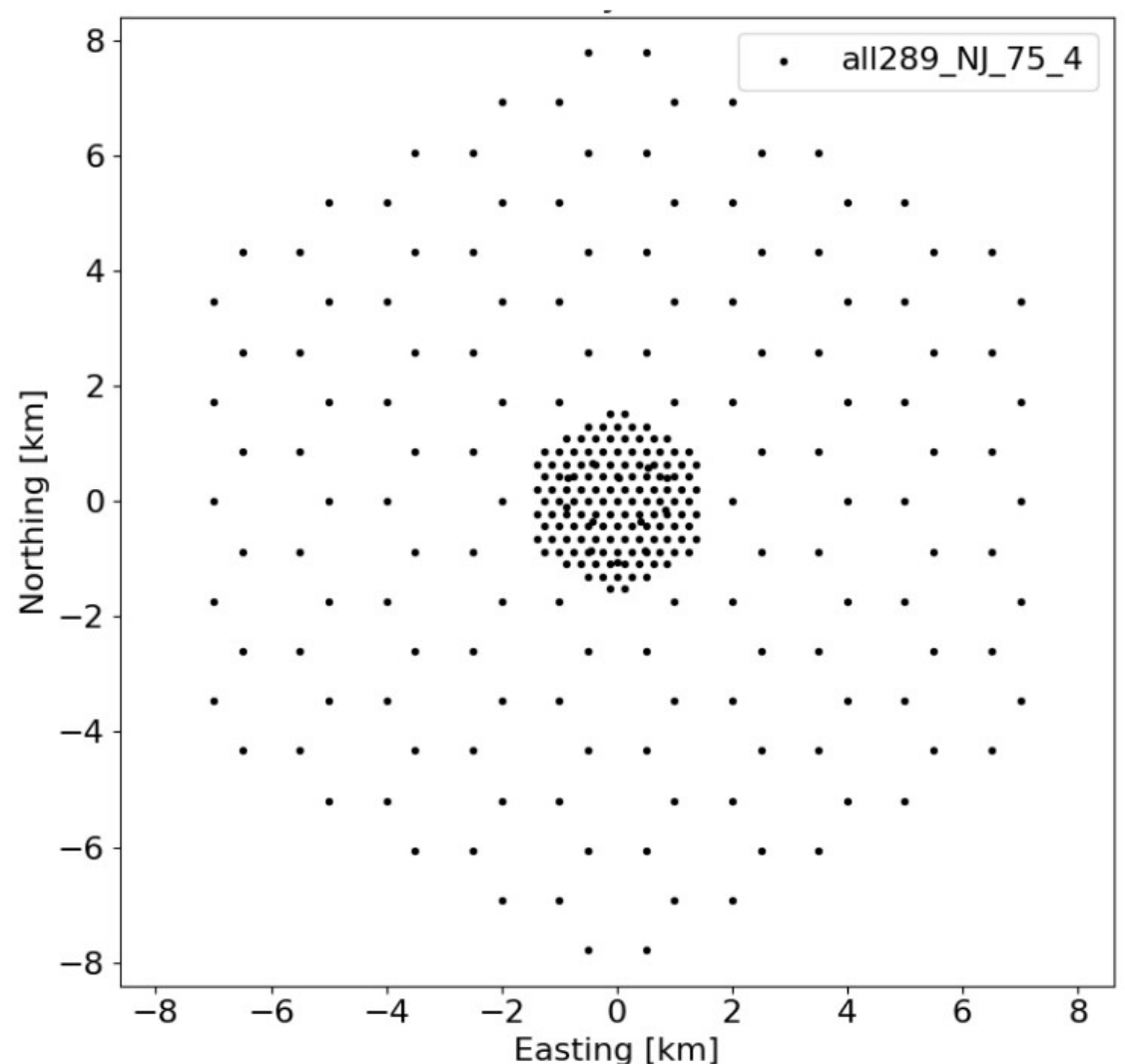
- Simulated data including realistic noise

Processing:

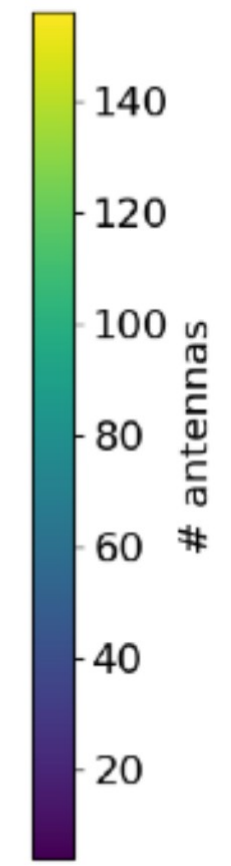
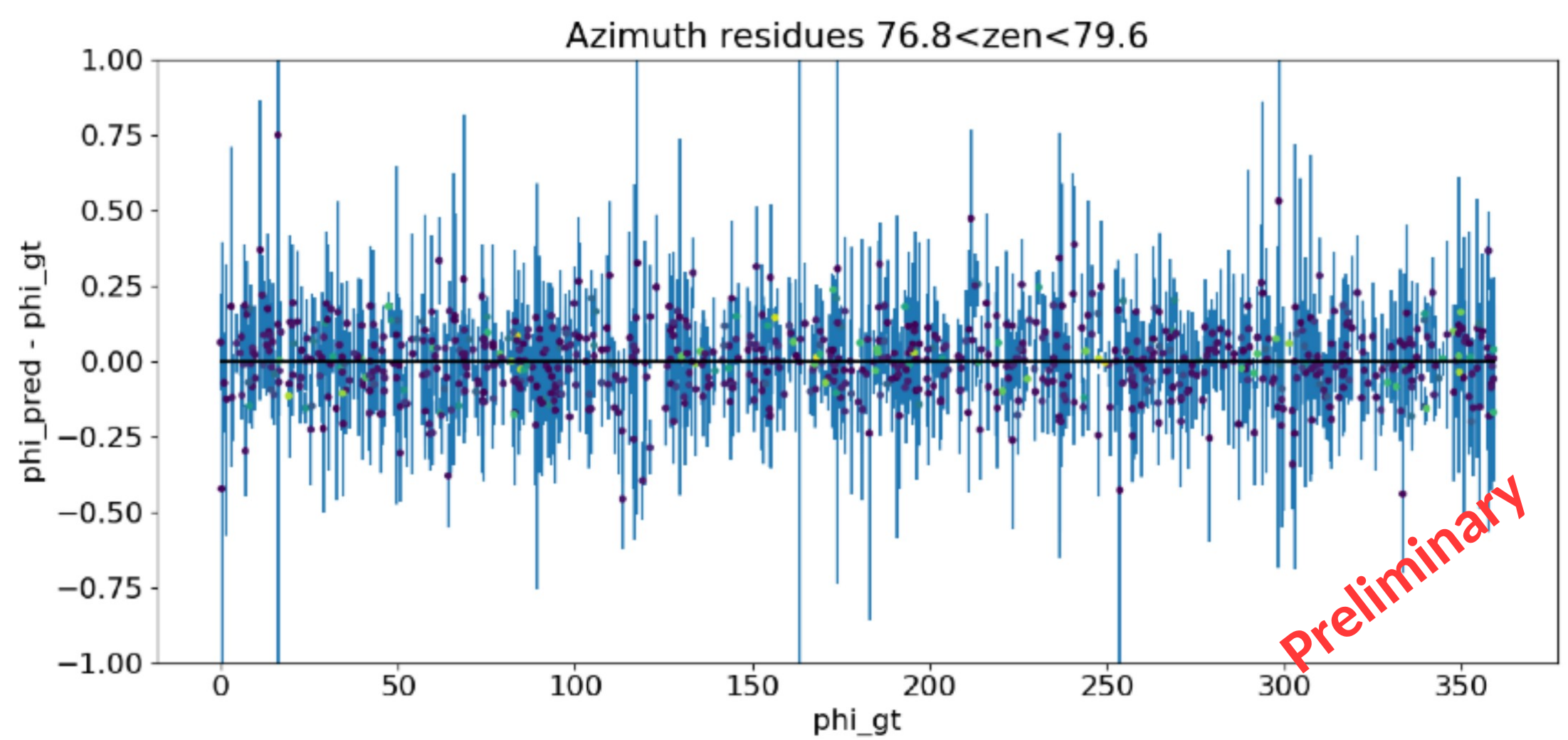
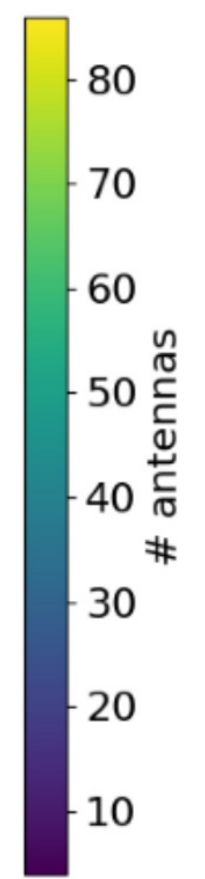
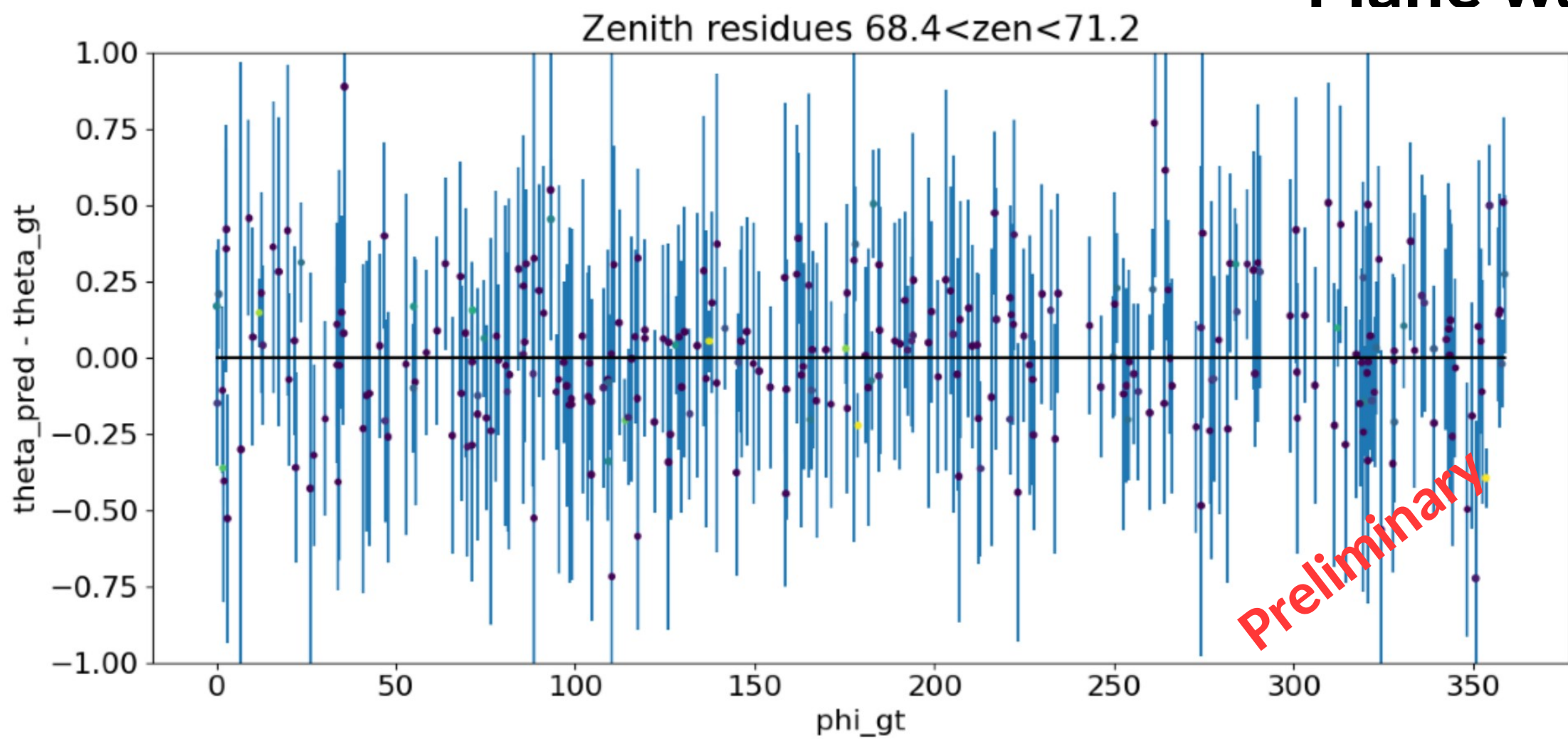
- ◆ Filter in the [50, 200] MHz frequency range
- ◆ Amplitude: Hilbert peak amplitude/Trigger time read from root files
- ◆ Quality cuts:
 - Amplitude threshold = 110
 - Antenna threshold: 5 antennas



All 289 antennas

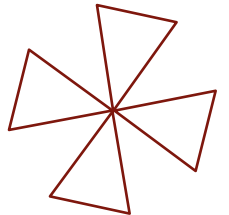


Plane wave reconstruction



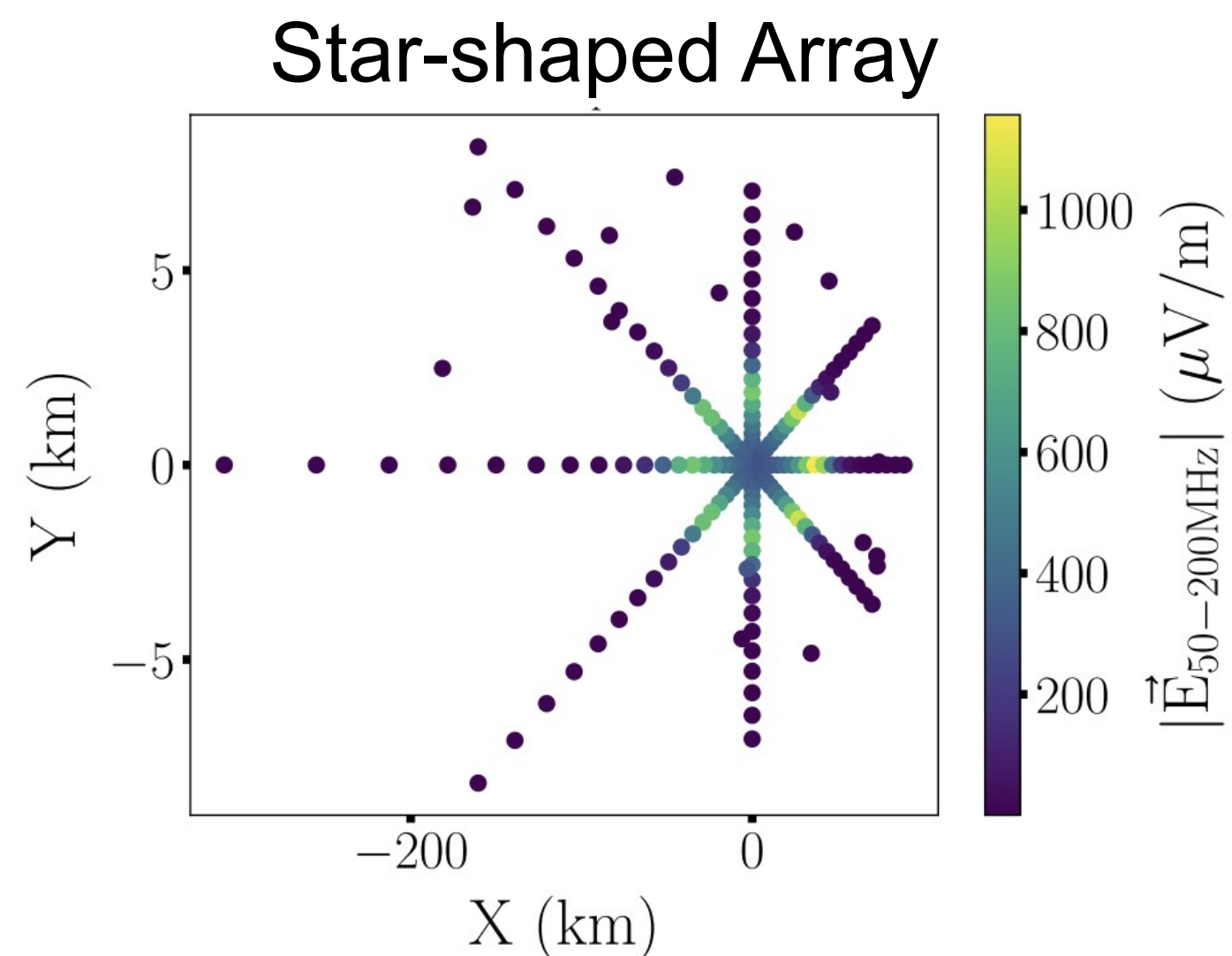
Results II: Reconstruction on star-shaped simulations





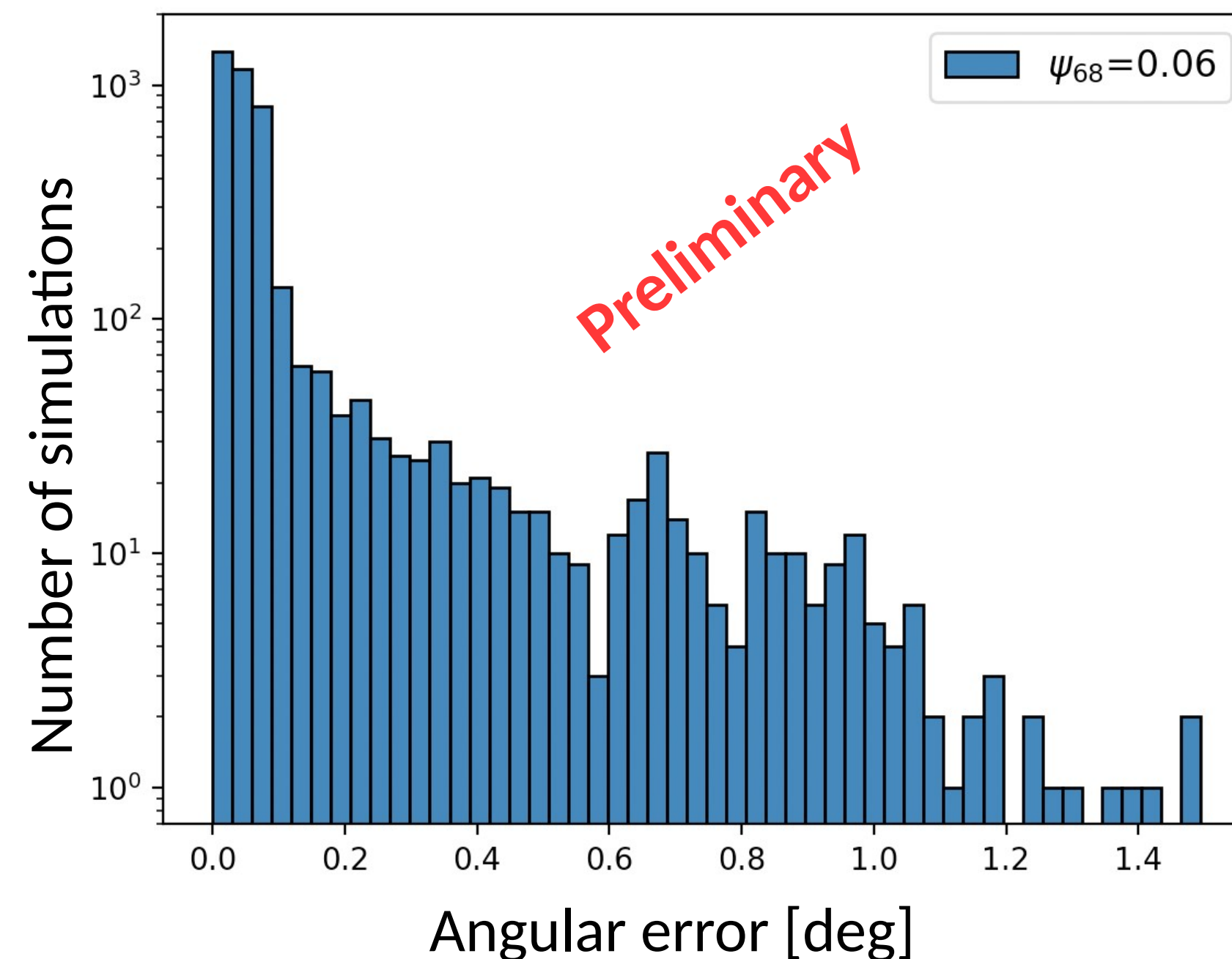
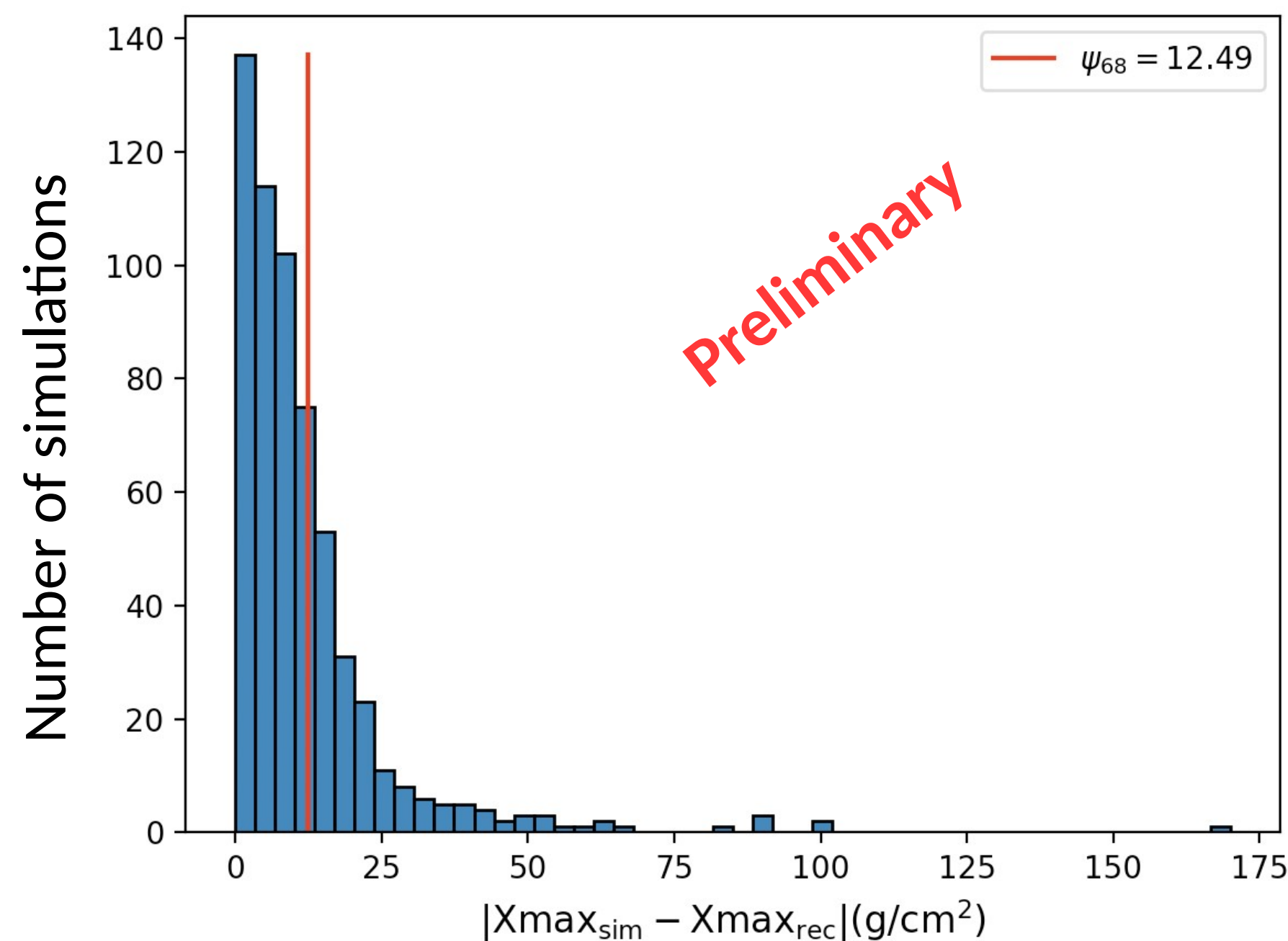
Performance on Star-shaped antenna layout

Decoene (2021)



EAS simulations without realistic noise.

Zenith=84.2°



Reconstruction of Xmax and direction of EAS on Star-shaped simulations:

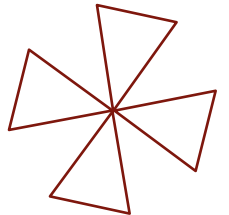
- Xmax reconstruction resolution around 10g/cm² for very inclined showers (and 5g/cm² for vertical showers)
- Angular reconstruction has a resolution of about ~0.1°



Follows the LOFAR methodology introduced in Buitink et al, 2014

Results III: Angular Distribution Function On toy GP300 simulations





Description of the toy GP300 EAS simulations

ZHAireS Simulations:

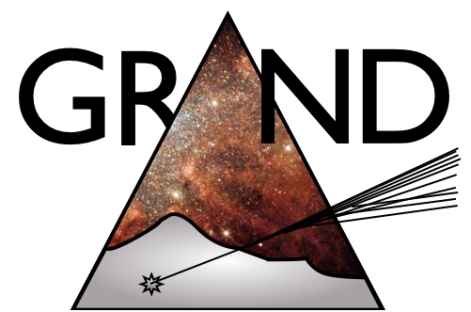
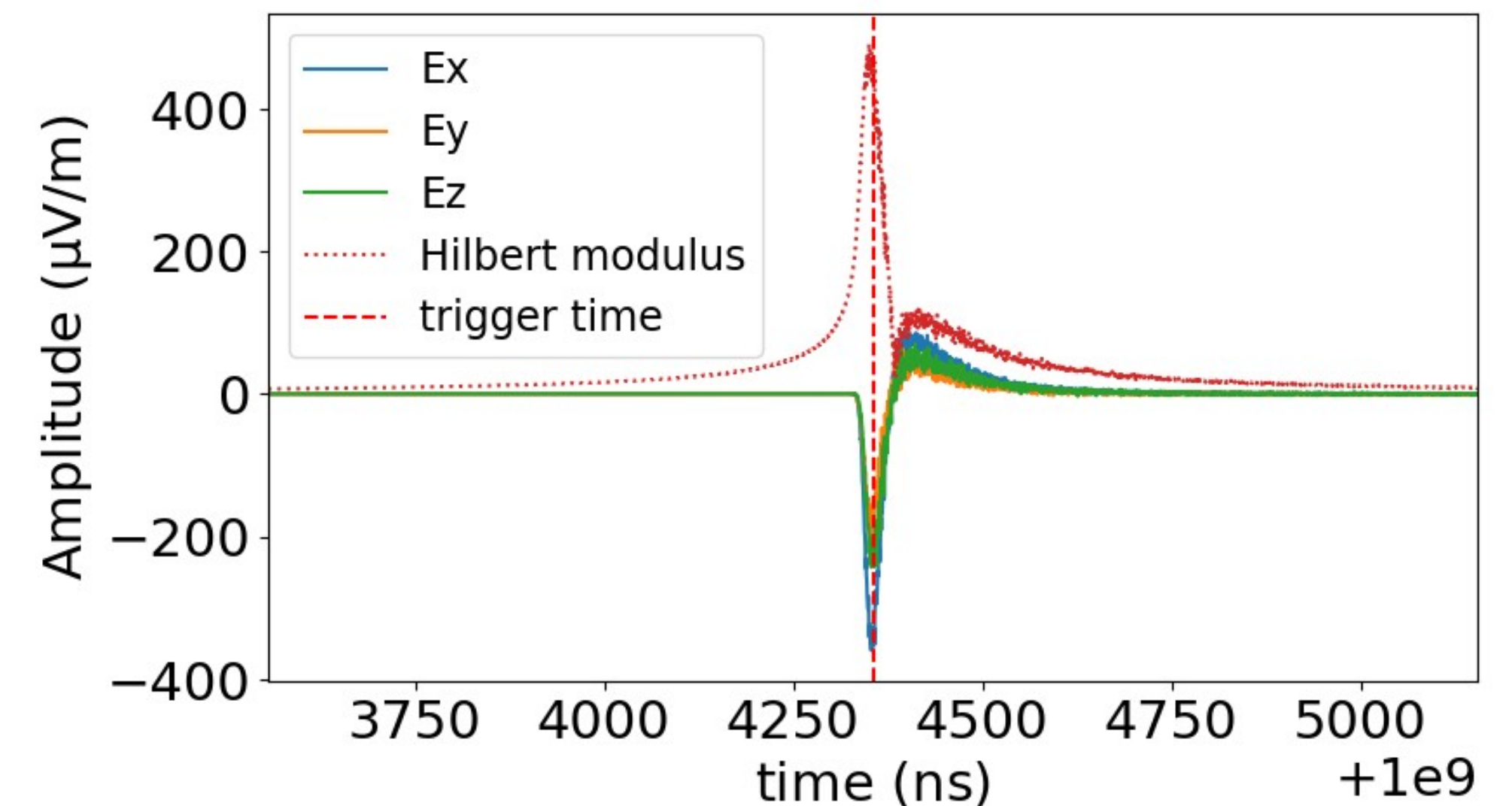
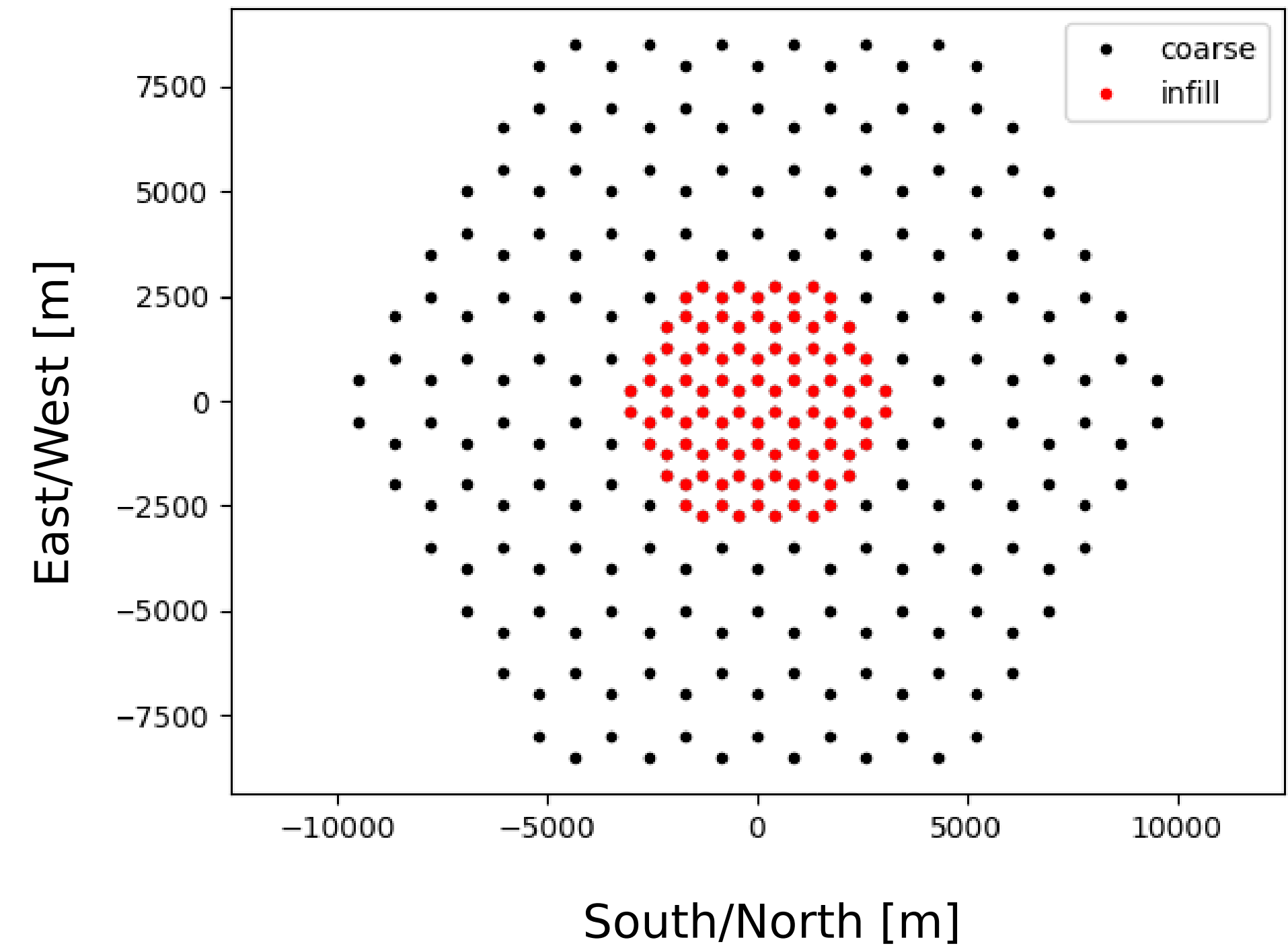
- Primaries: Proton, Iron, Gamma
- Energy: 0.251, 0.631, 1.58, 3.98 EeV
- Zenith: [63°, 87°]

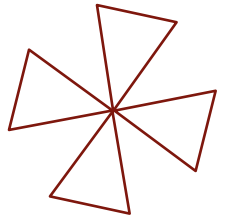
Toy GP300 layout with infill

- ◆ Shower core always contained in the layout
- ◆ Raw electric field data without galactic noise
- ◆ Random gaussian error = 5 ns on trigger times (GPS)
- ◆ Random gaussian error of = 10% on signal amplitudes (calibration)

Processing:

- ◆ Filter in the [50, 200] MHz frequency range
- ◆ Amplitude: Hilbert peak amplitude/Trigger time
- ◆ Quality cuts:
 - Amplitude threshold = **110**
 - Antenna threshold: **5 antennas**

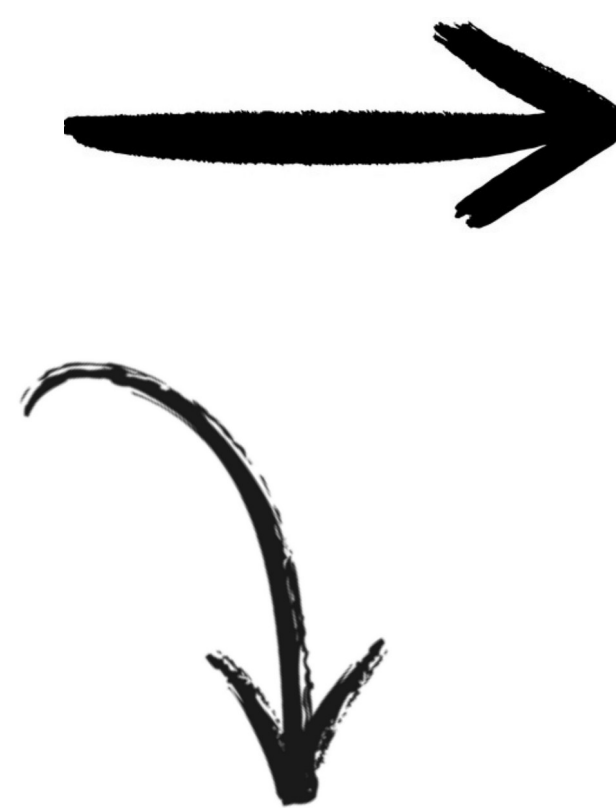




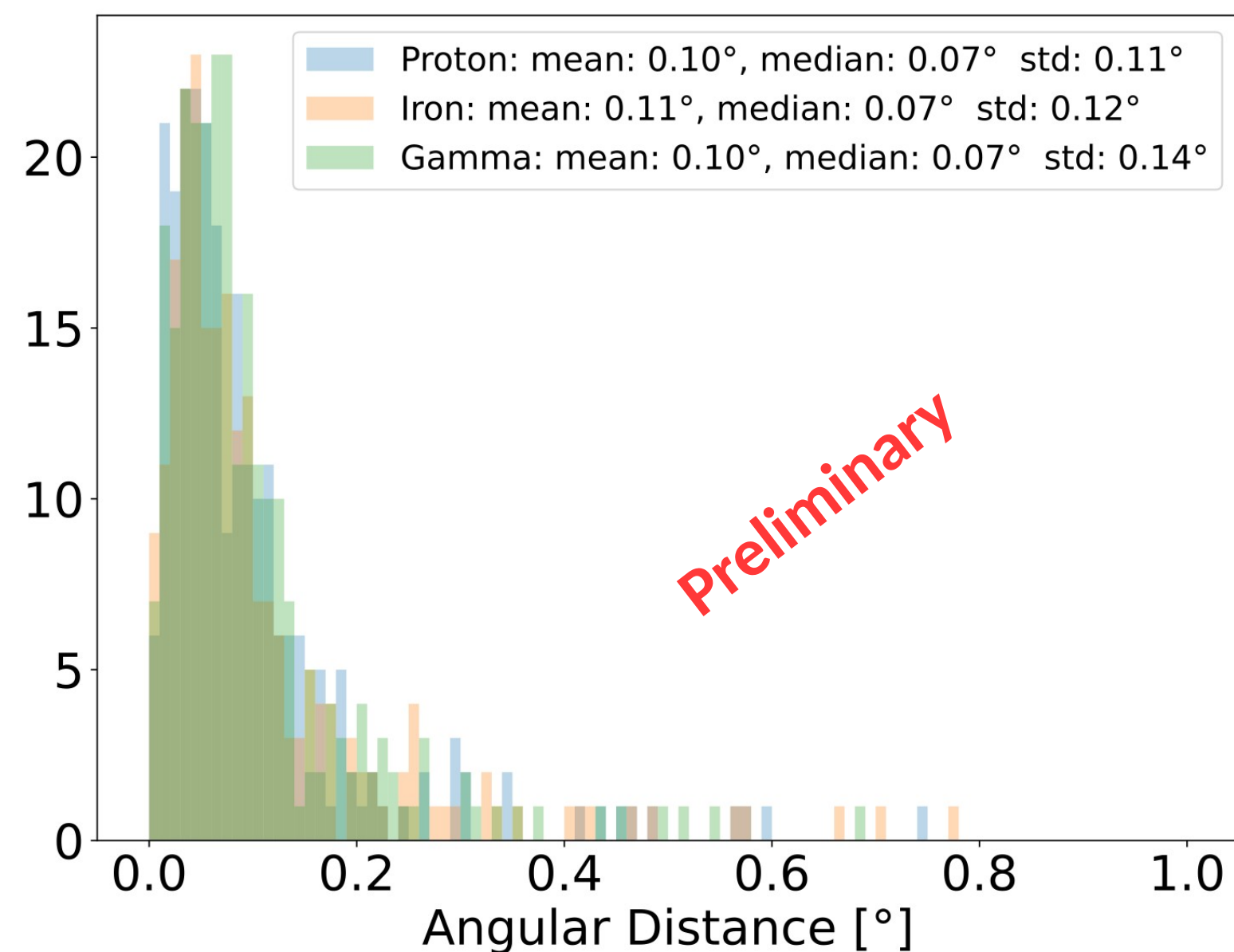
EAS Direction Reconstruction on toy GP300 Simulations

Main Results:

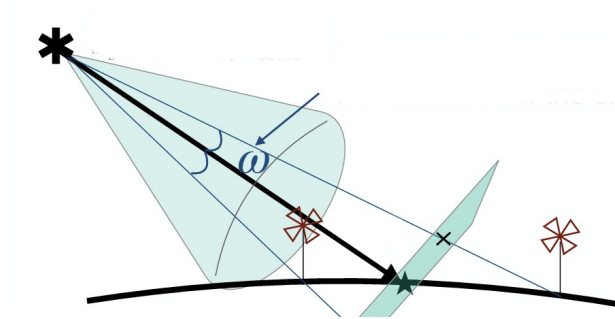
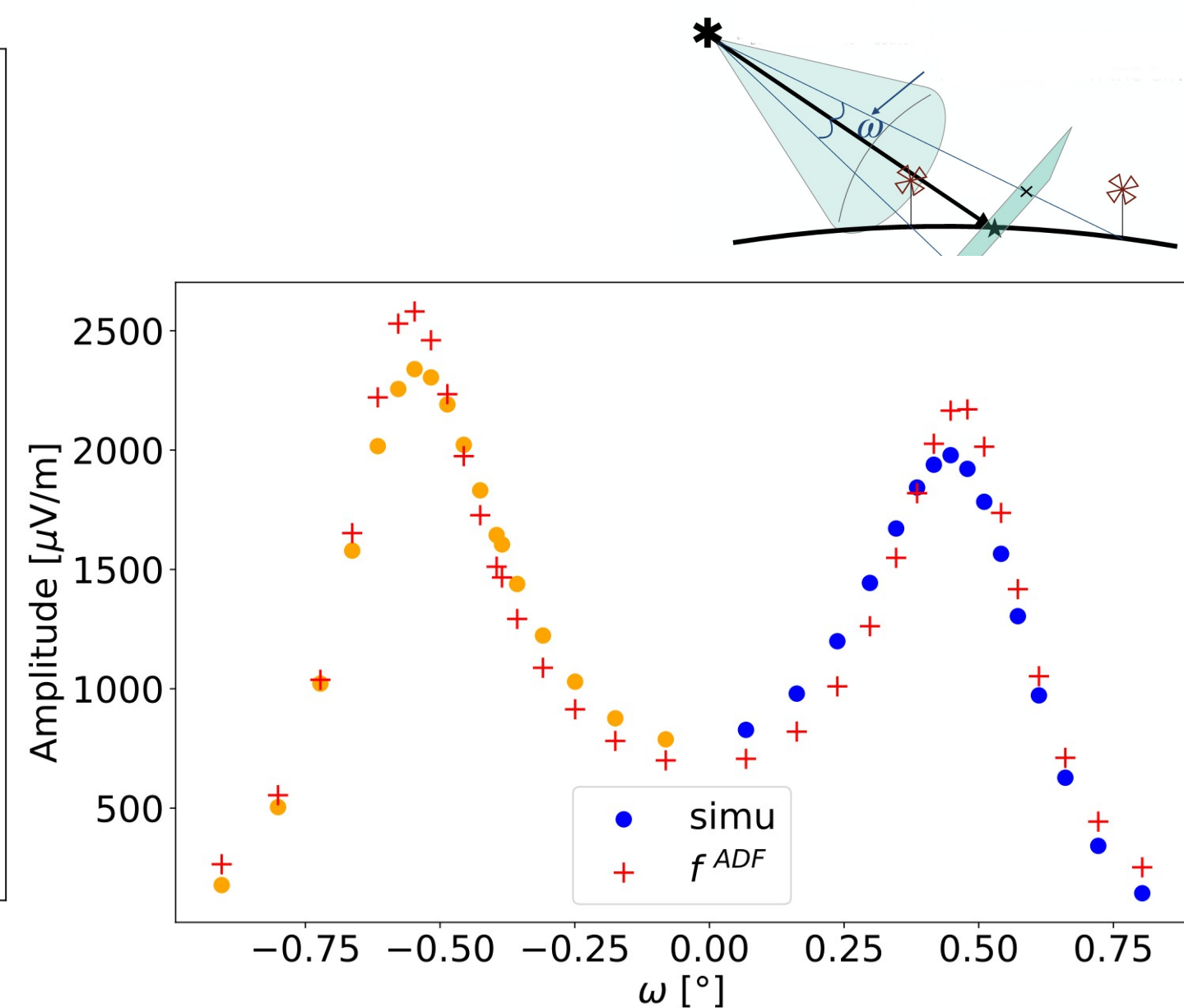
- Reconstruction on GP300 ZHAireS simulations with experimental uncertainties
- Excellent angular reconstruction: approx. $\sim 0.1^\circ$
- ADF approx. matches angle and amplitude peak for zenith angles greater than $\sim 70^\circ$



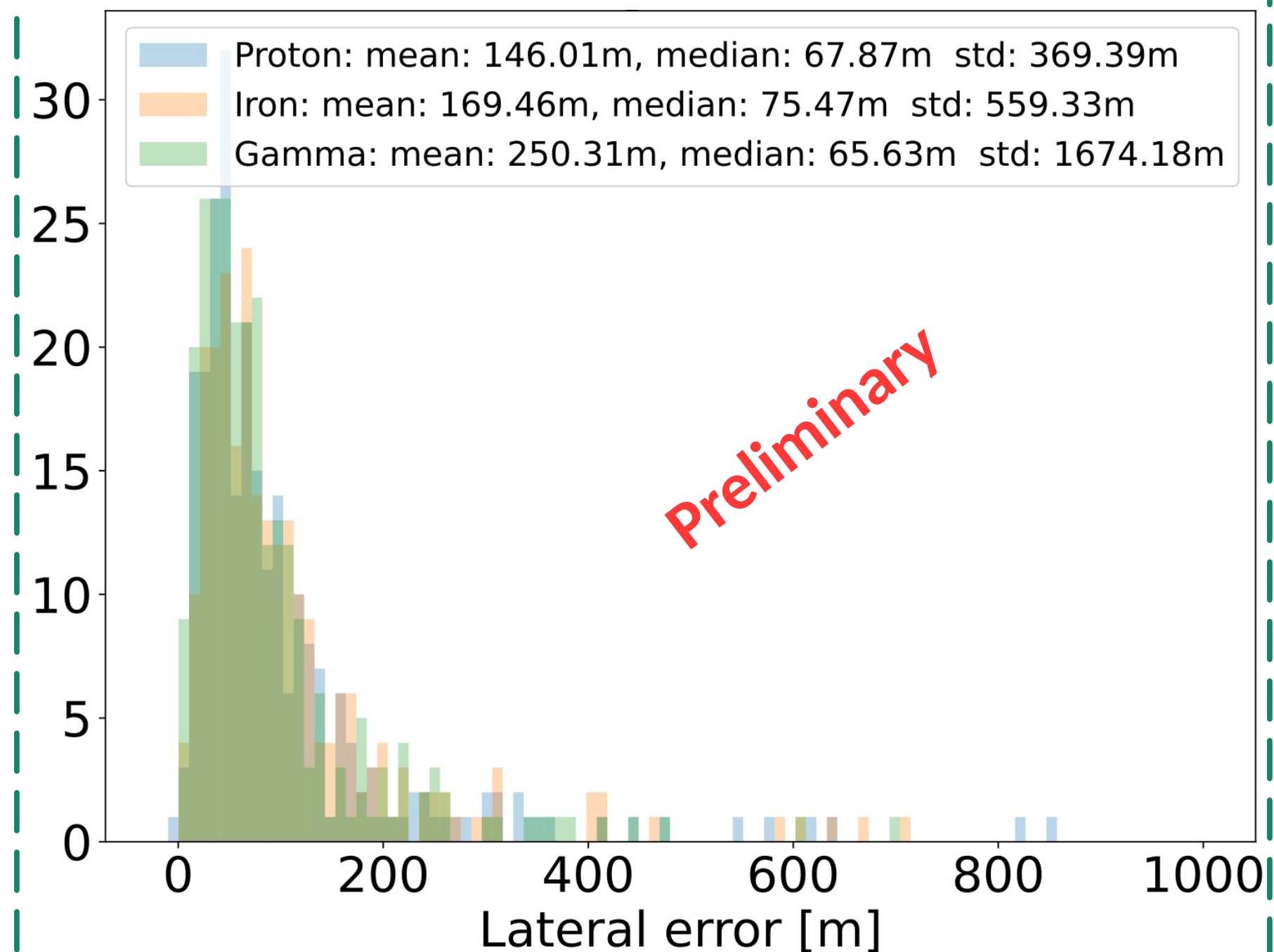
ADF reconstruction



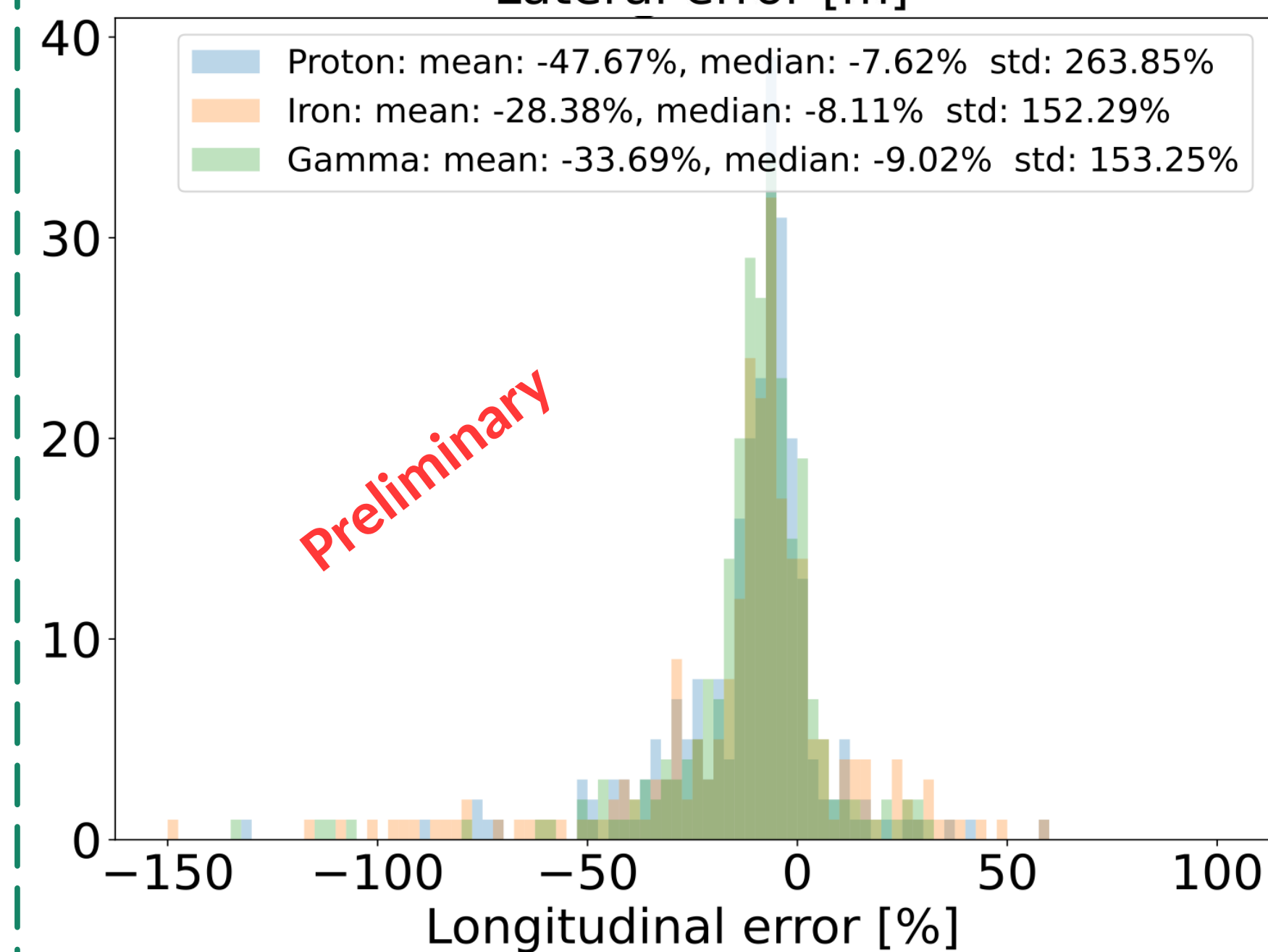
Preliminary



Spherical wave reconstruction



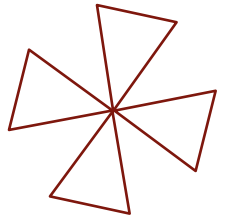
Preliminary



Preliminary

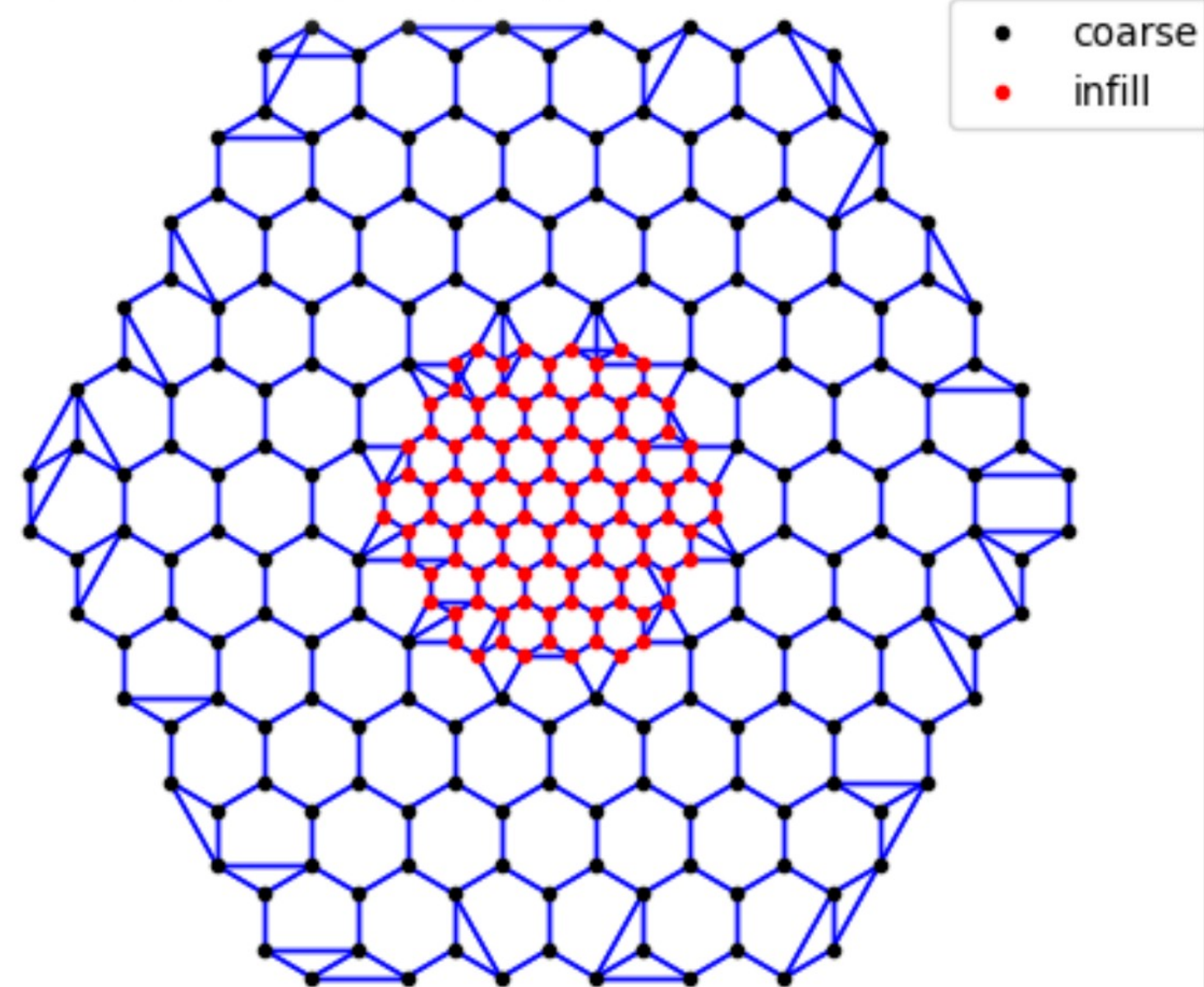
Results IV: EAS Reconstruction using Machine Learning methods





Reconstruction of EAS with Graph Neural Networks (GNN)

Graph generation

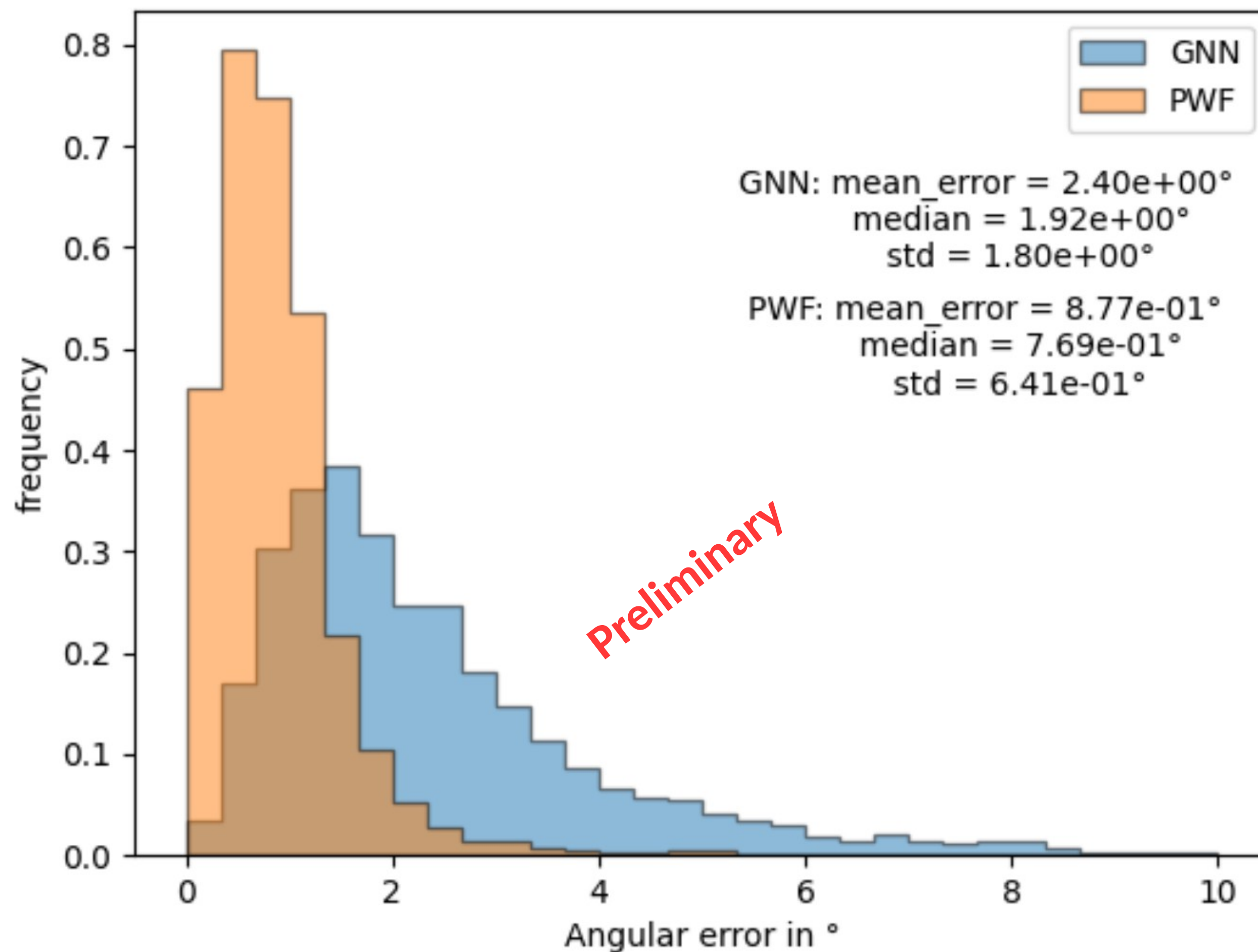


- Each antenna is linked to at least its three neighbors.
- Some antennas have more neighbors due to being neighbors of neighbors.

Training data:

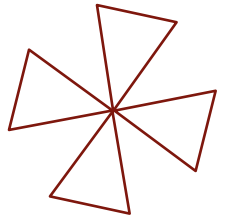
- ◆ Use only antennas with $E\text{-field} > 30 \mu\text{V/m}$
- ◆ Keep only events with >7 triggered antennas
- ◆ Features:
 - Peak time and amplitude
 - Antenna position

Prediction error

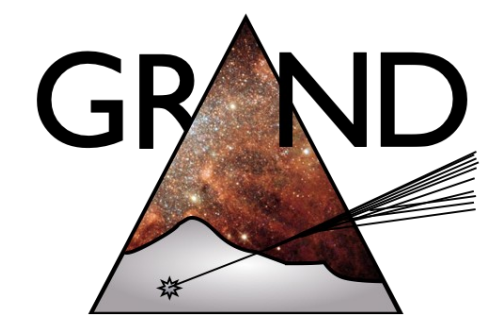
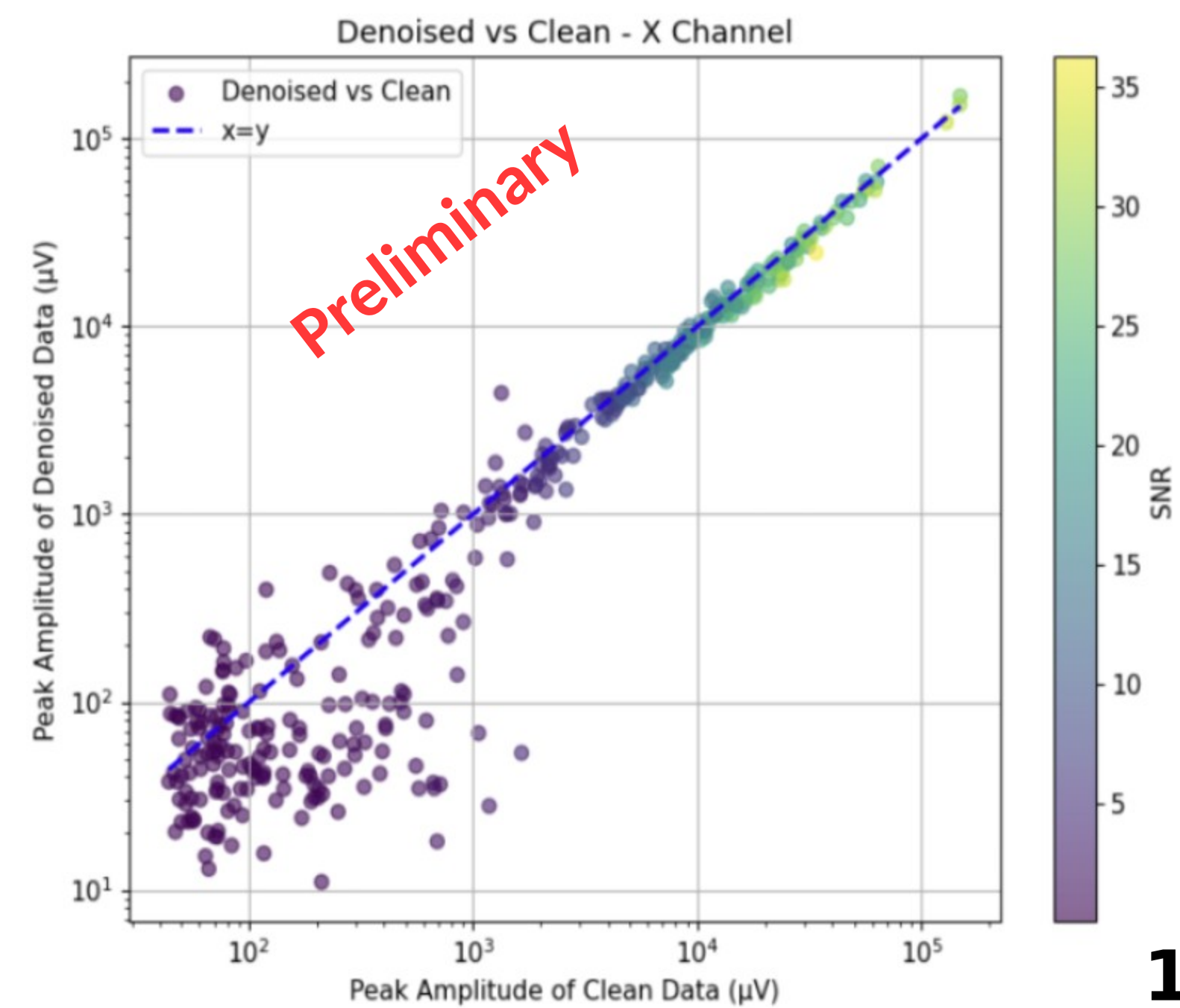
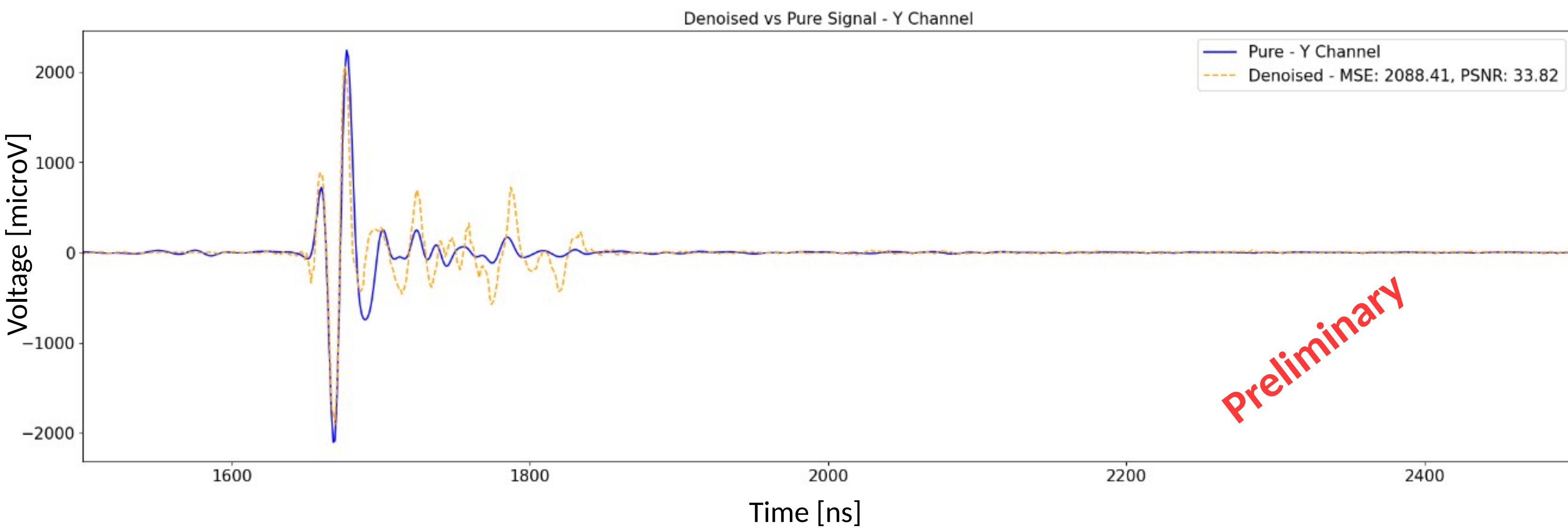
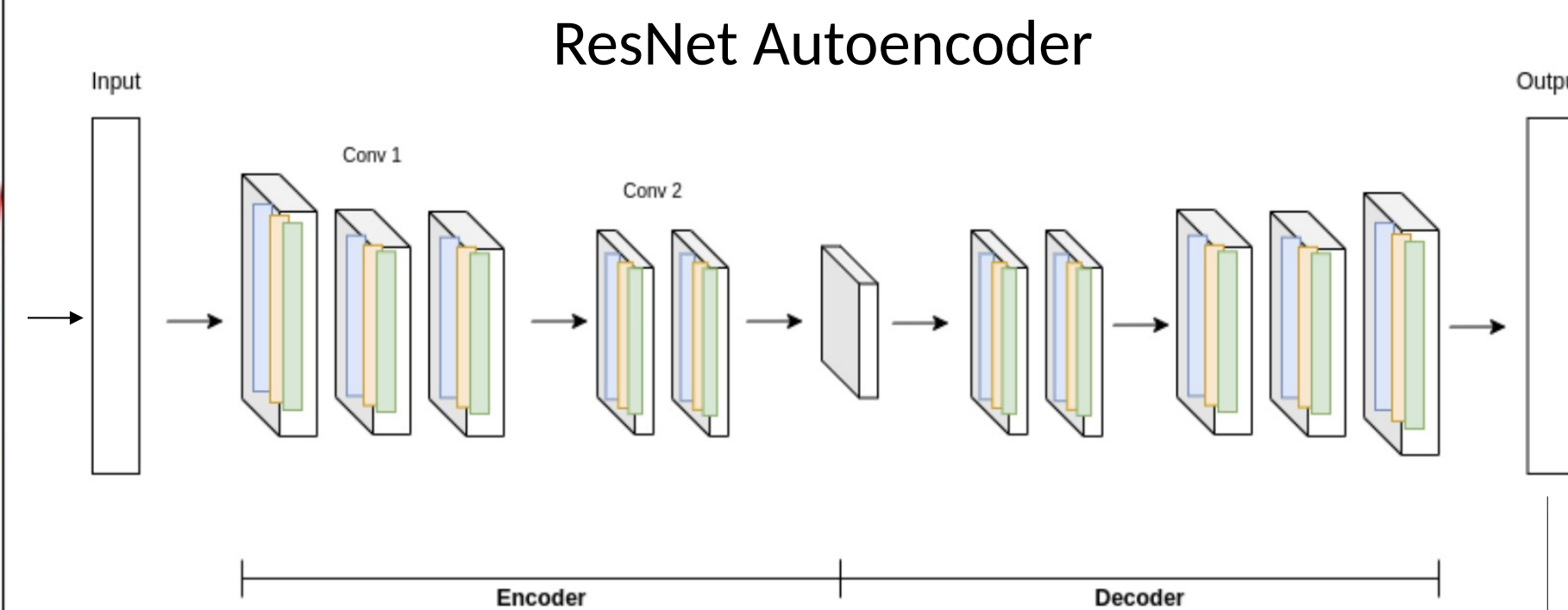
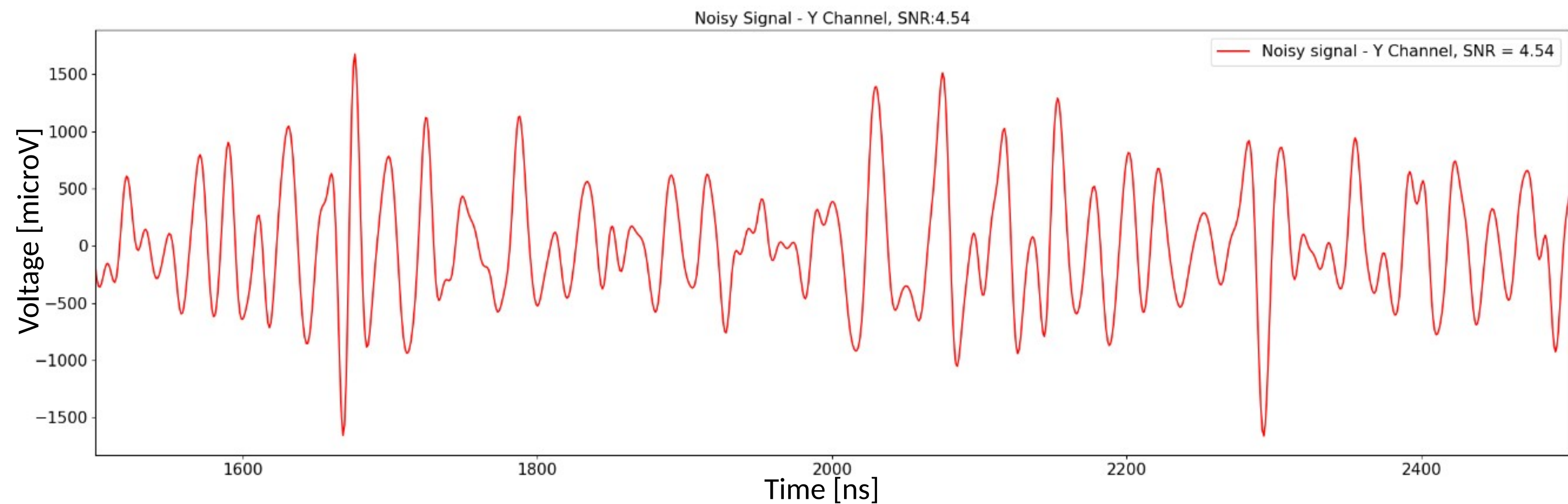


Results V: Signal Denoising using an Autoencoder





Voltage Denoising with a ResNet Autoendoer



Conclusions

- ✓ Developed and tested direction reconstruction on the new *Data Challenge 2* simulations with and without realistic noise. Preliminary results are promising.
- ✓ Fitted an empirical Angular Distribution Function (ADF) on various sets of simulations. Early results indicate that the ADF method has the potential to increase the reconstruction sensitivity.
- ✓ Preliminary results using machine learning methods achieve a sensitivity close to standard methods. Further studies are ongoing to enhance its capabilities.

Backup Slides



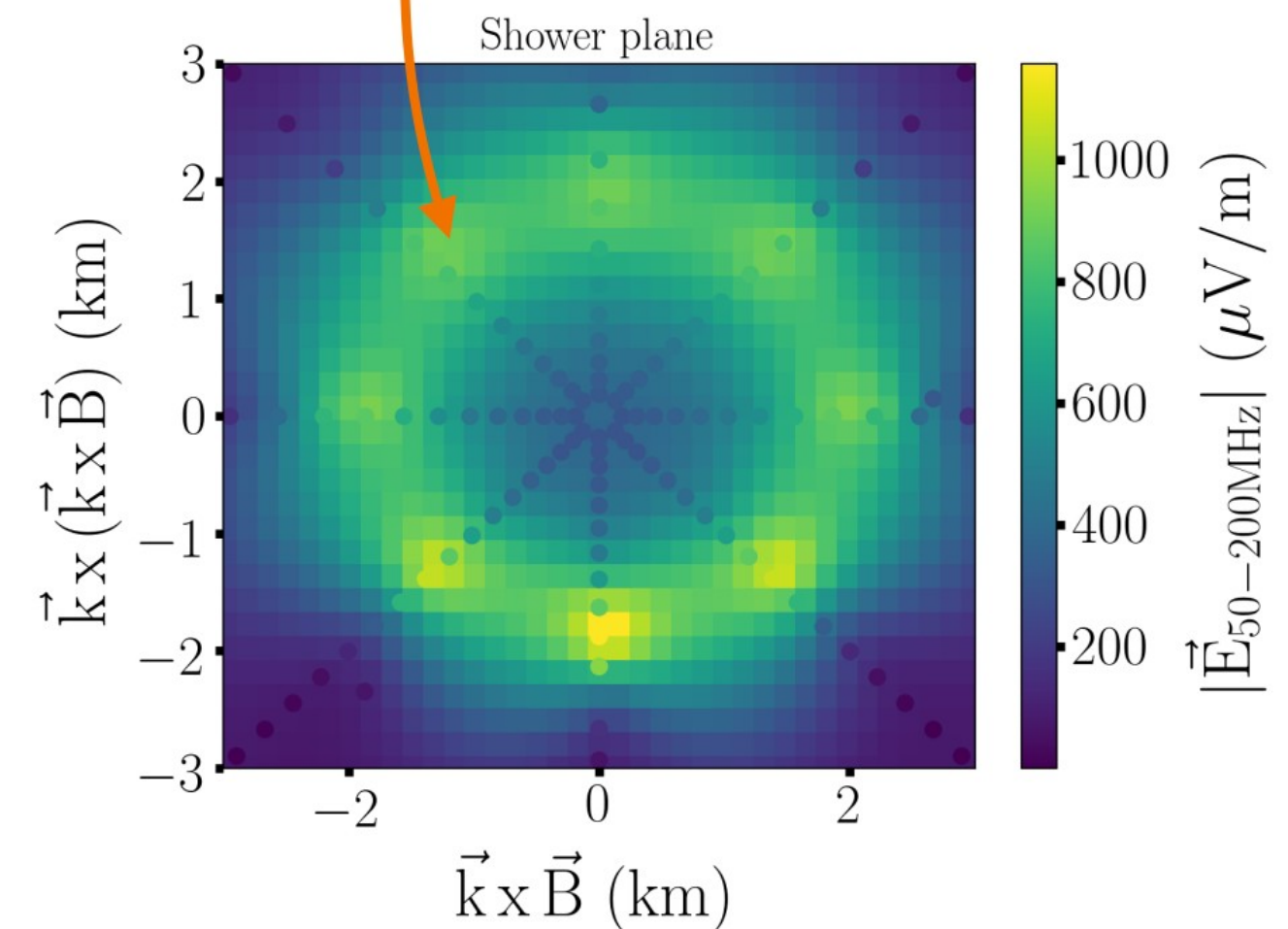
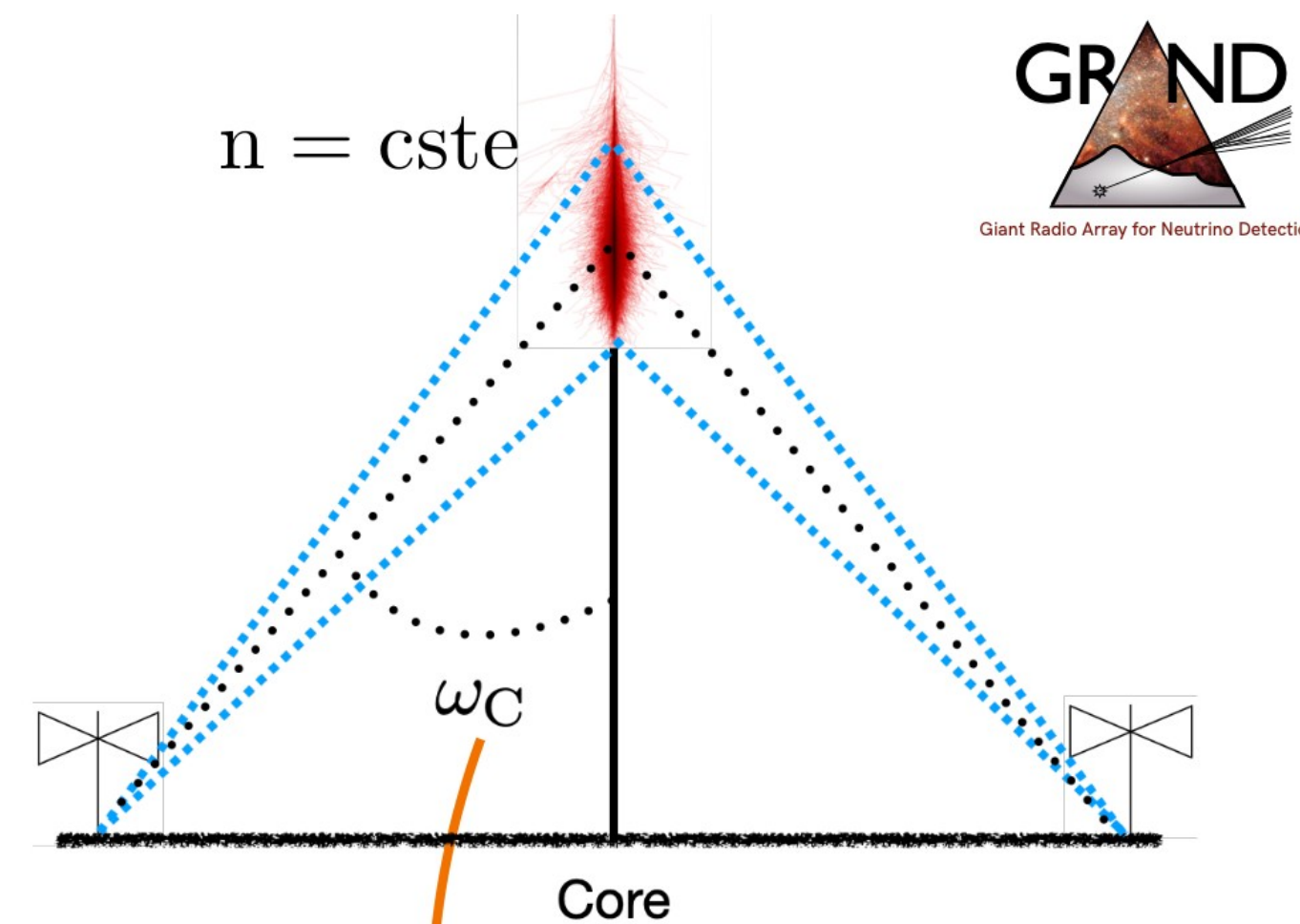
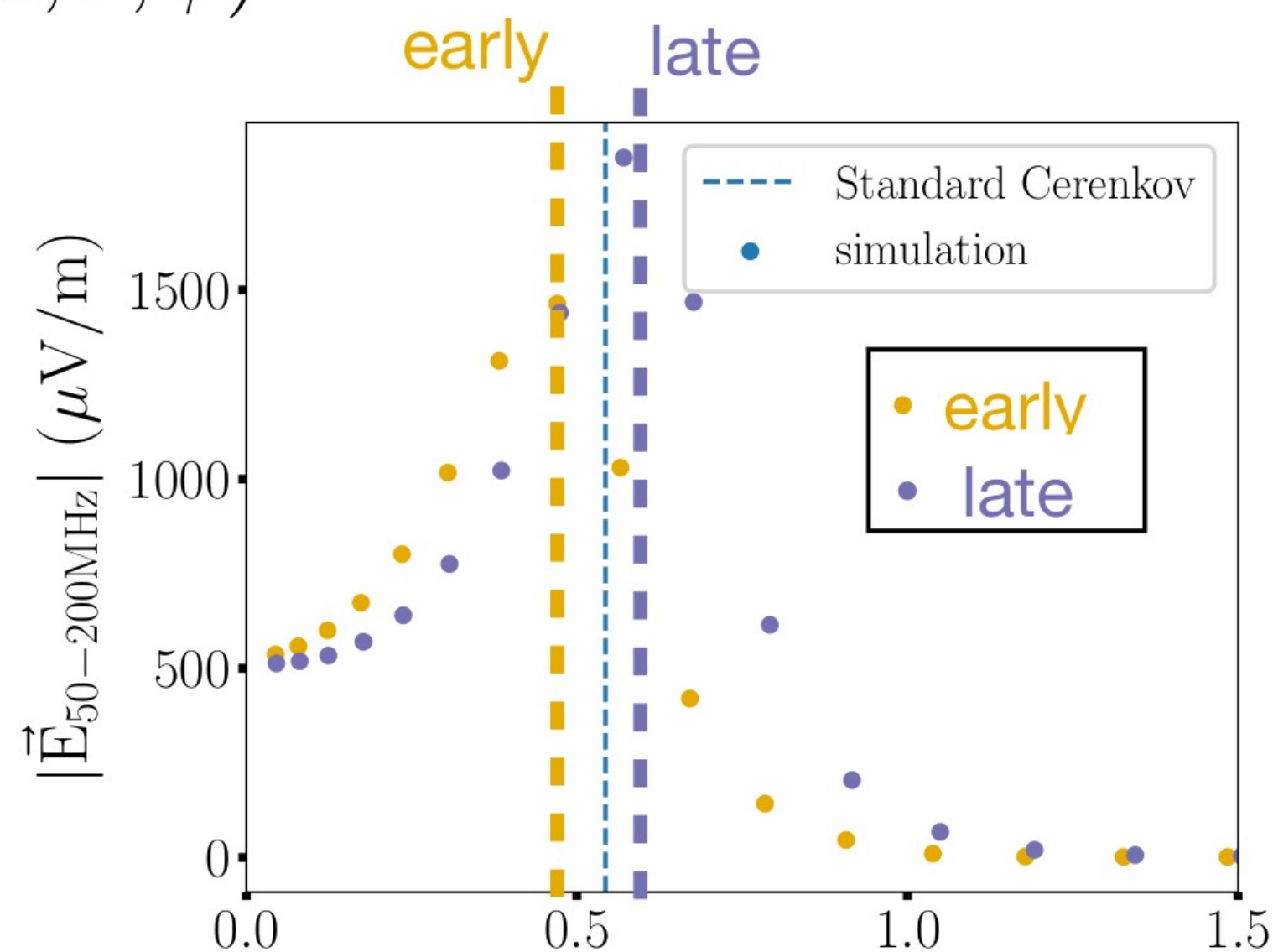
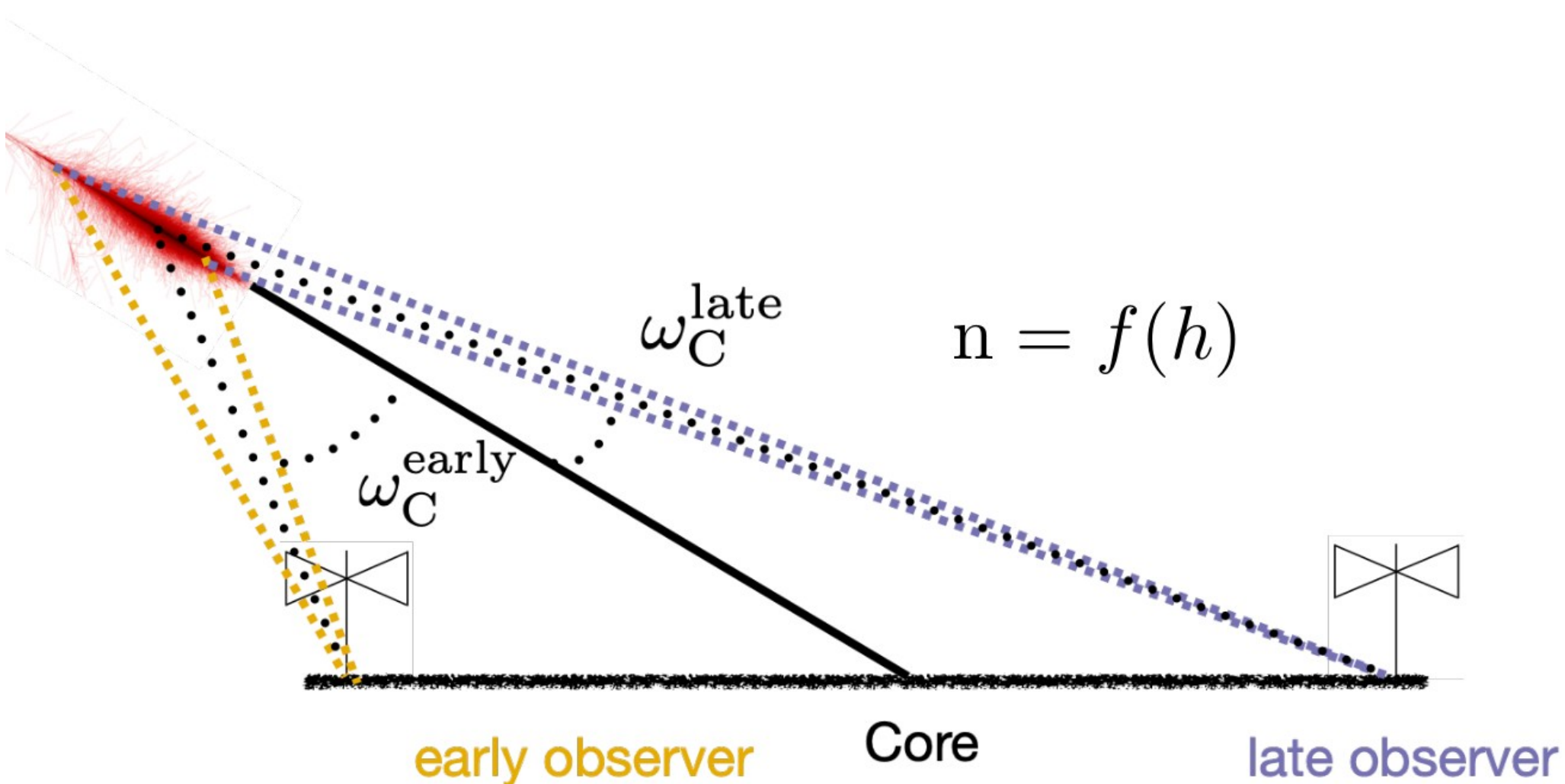
The Angular Distribution Function (ADF)

Cerenkov Asymmetry

Cerenkov cone:

- geometrical effect → angle where all emissions arrive at same time
- signal compression → high amplitudes
- standard computation: $\omega_C = \arccos(1/n)$ (equal optical paths = constant n)

But if optical paths are different (varying n) $\omega_C = f(\vec{x}, \theta, \phi)$



analytical description of $\omega_C = f(\vec{x}, \theta, \phi)$

used into the amplitude model: each antenna “sees” a different Cerenkov cone

The analytical description of the Cerenkov asymmetry matches the simulated data