



# Background Rejection in the ARA Experiment

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## Part 1: Background rejection in the ARA Testbed station

- Event analysis techniques

- GRB timing rejection

- Optimization of cuts

## Part 2: New algorithm for background rejection in stations with regular geometry

- Regular geometry advantages

- Efficiencies



# **BACKGROUND REJECTION IN THE ARA TESTBED GRB NEUTRINO SEARCH**

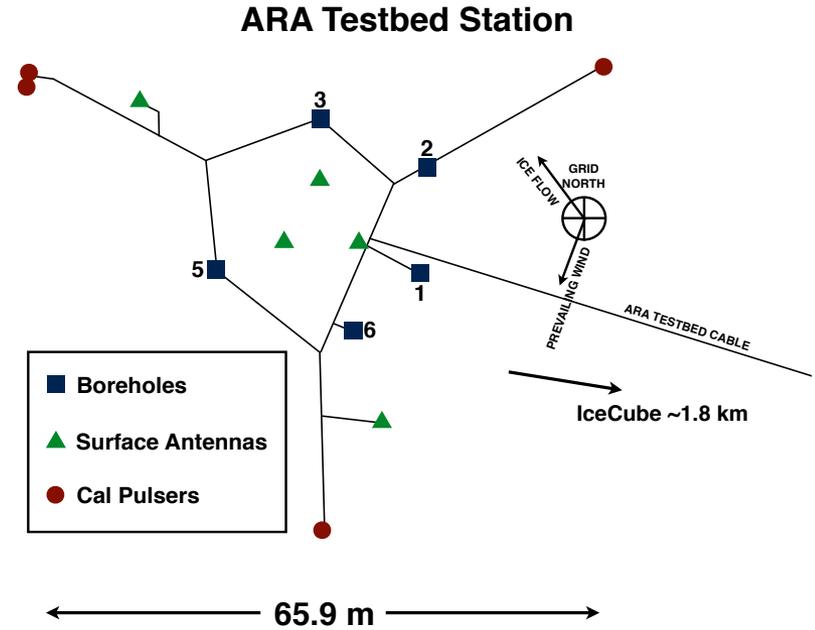


- 2 basic types of noise
  - CW
  - **Thermal**
    - **Characterized by (semi-)random fluctuations from surrounding environment**
- ARA trigger – based on tunnel diode output
  - Acts as a few-ns power integrator
  - Trigger rides a threshold determined by the thermal noise level
  - 100's of millions of events – almost all thermal noise
- How to reject these signals efficiently?
  - For analysis cuts
  - For filtering before transmission to the North

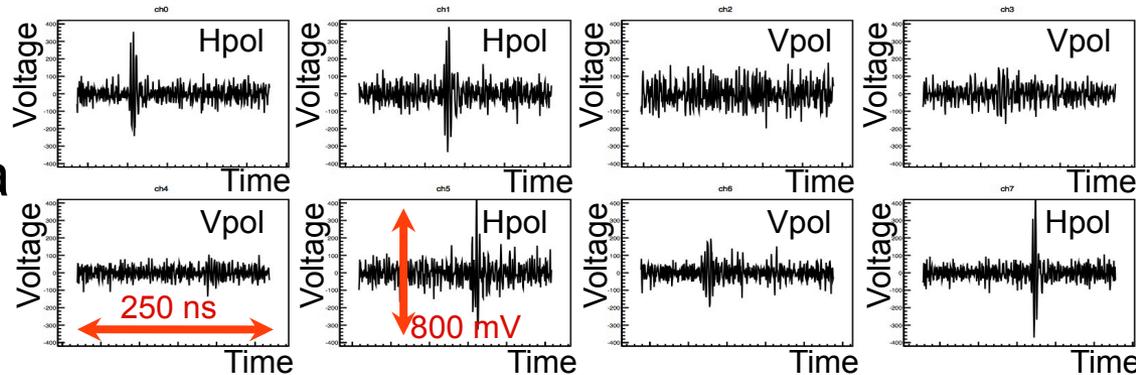


# Testbed Station

- Total 16 antennas, 8 borehole antennas at 150 MHz to 850 MHz
- Maximum depth of antennas ~ 30 m
- 3 sets (Vpol + Hpol) of calibration pulsers
- Deployed 2010-2011
- Ran for 2 years (2011 – 2012)
  - Not intended for long-term operation
- First ARA neutrino searches carried out with Testbed station data
  - Diffuse: [arxiv:1404.5285](https://arxiv.org/abs/1404.5285)
  - GRB: [arxiv:1507.00100](https://arxiv.org/abs/1507.00100)



Calibration pulser event waveform from 8 deep antennas in Testbed

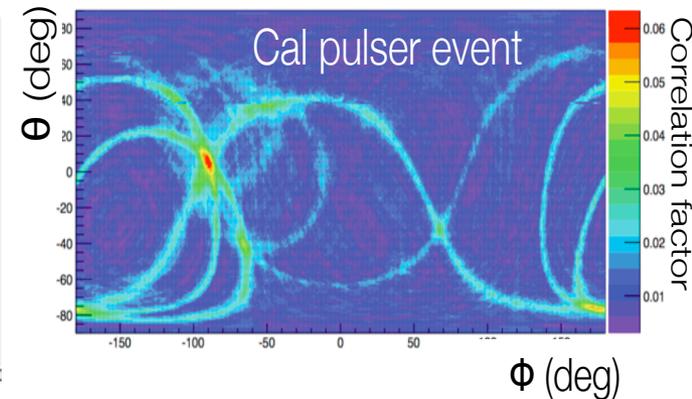
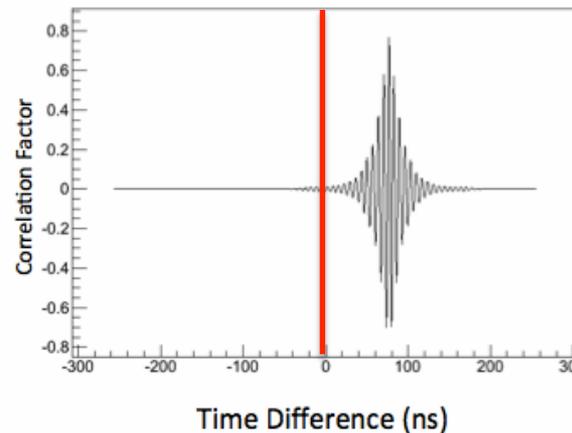
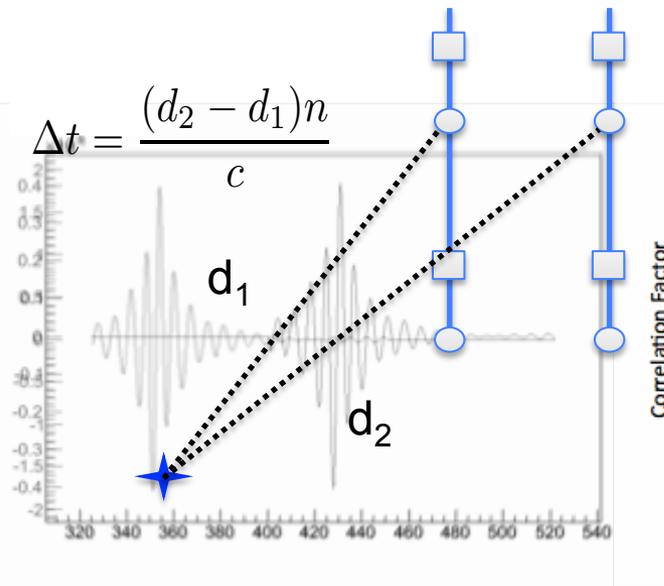


2016-06-09



## Adapted interferometric technique from diffuse search for GRB search

1. Impulsive waveform – ~1-10 ns time scale
2. Correlation factor - Convolution of the two waveforms including a timing offset
3. Calculate timing delays for all angles of approach
4. Sample correlation plot at these delays
5. Create a map for all pairs of antennas and the correlation

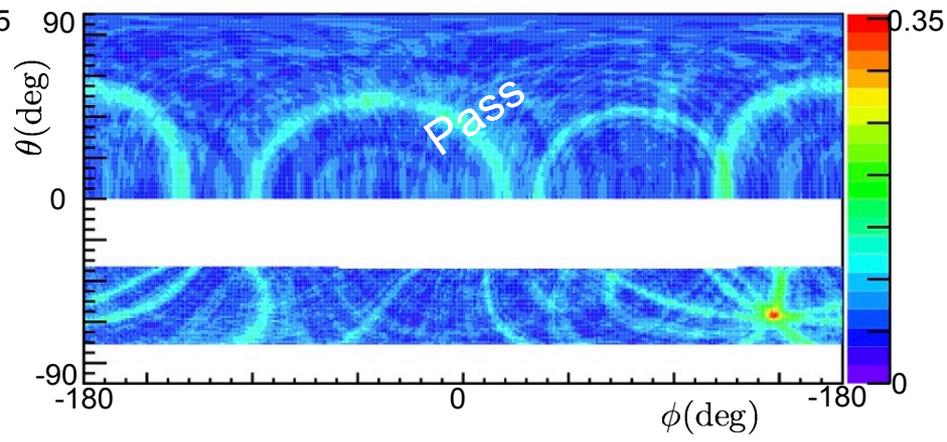
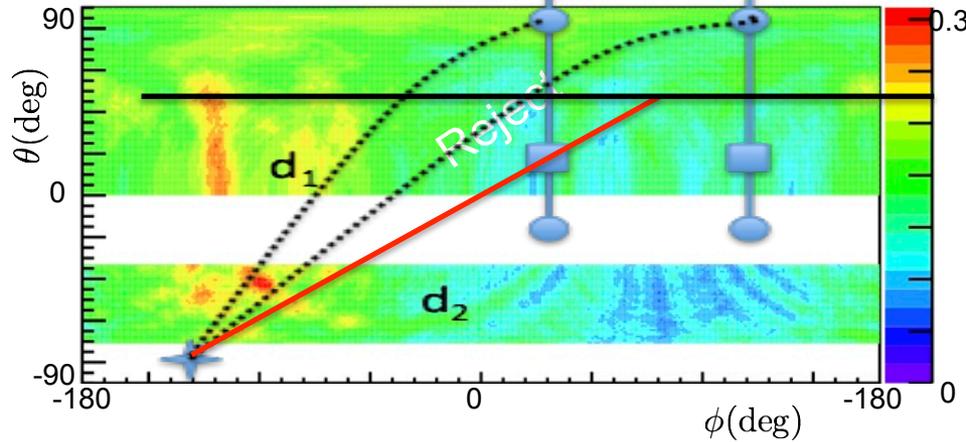




# Reconstruction Quality Cut

Known background event reconstruction map example

Simulated  $\nu$  event reconstruction map example



Rejected thermal noise by requiring strong reconstruction map peak that is unique

Reconstruction based on timing from ray-tracing

Use 30 m and 3 km maps in Hpol and Vpol

Requires at least one reconstruction map to be of good quality

$1 \text{ deg}^2 < \text{Area of 85\% contour surrounding the peak} < 70 \text{ deg}^2$

Total 85% contour peak area  $< 16.2 \times \text{Area of 85\% contour surrounding the peak}$

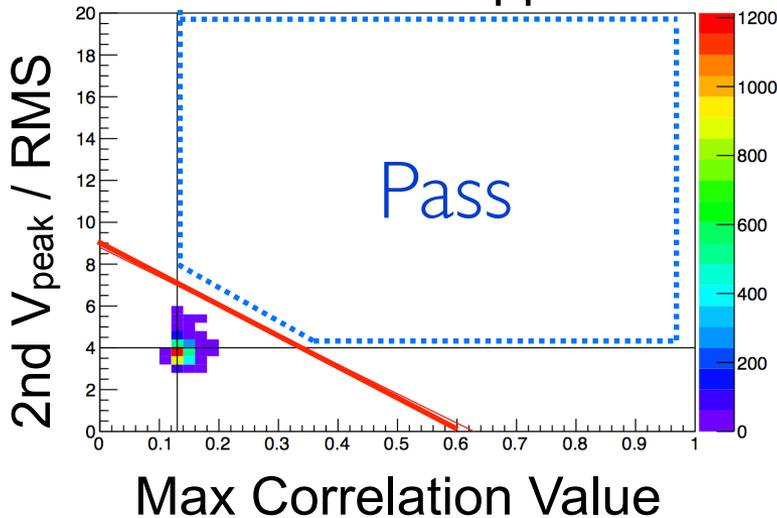
Depending on the polarizations which pass the cut, the event is separated into Vpol and/or Hpol channels

Rejects ~95% of noise-dominated events after initial quality cuts

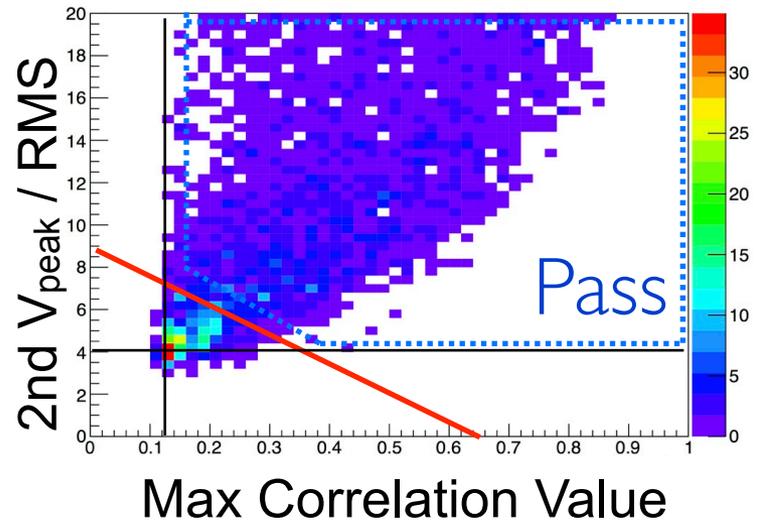


- Expect a correlation between signal strength from waveform and correlation value from reconstruction map for an impulsive event
- After removing known background events with other cuts, use this relation to get background estimation
  - Other cuts made: most reject specific anthropogenic signals

Testbed 10% data set  
With cuts applied



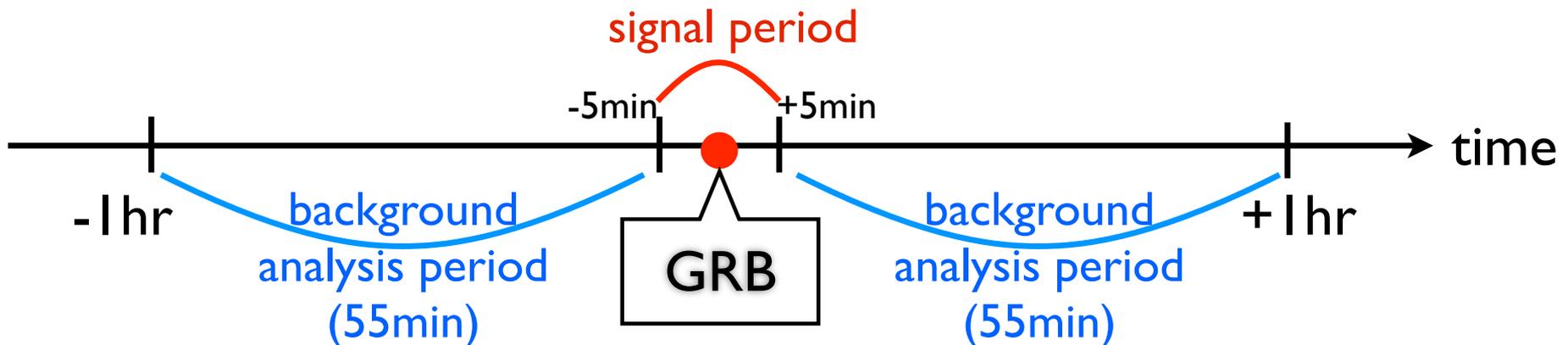
Simulated  $10^{18}$  eV  $\nu$  set  
with cuts applied





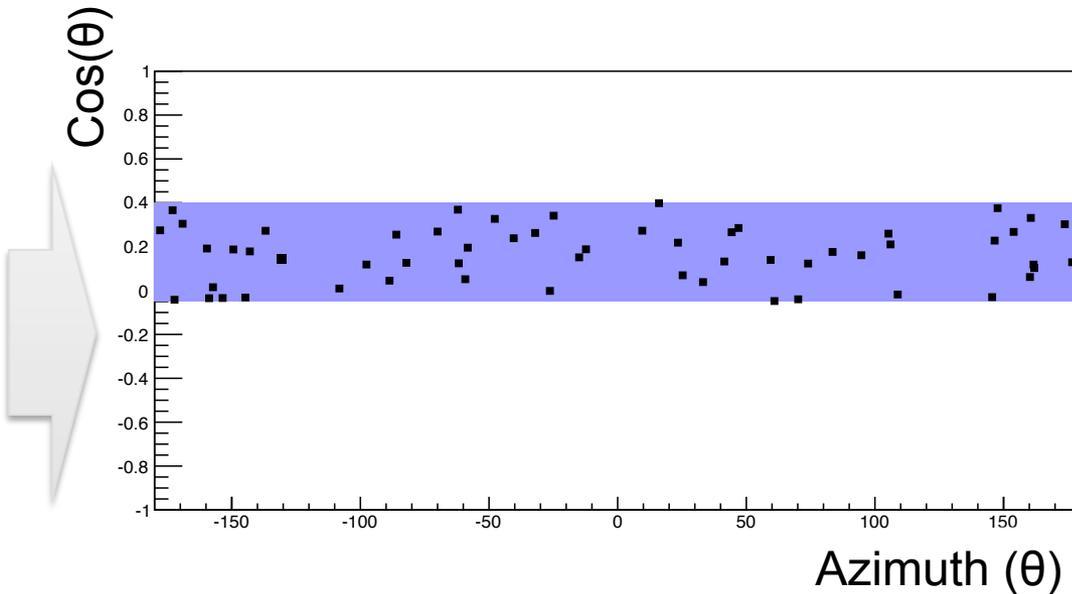
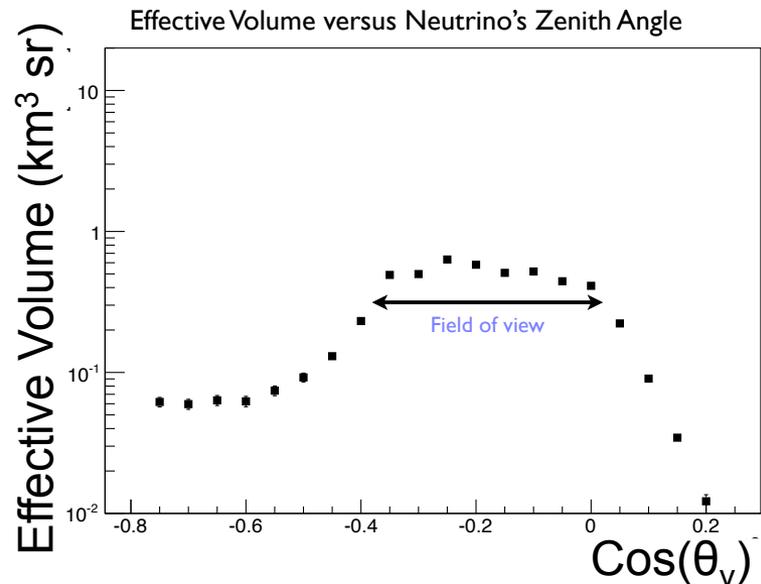
# Testbed GRB analysis

- Adapt the above techniques from the Testbed diffuse neutrino search (arxiv:1404.5285) to search for events coincident with known Gamma Ray Bursts
  - Stricter requirements in time → relaxation of cut values
- 2 unblinding stages
  - Tune cuts on 10% of data in the **background estimation window**
  - 1: Check remaining 90% in **background estimation window**
  - 2: **Signal search** – 100% of data +/- 5 minutes around GRB event
  - Timing technique adapted from ANITA (arxiv: 1102.3206)

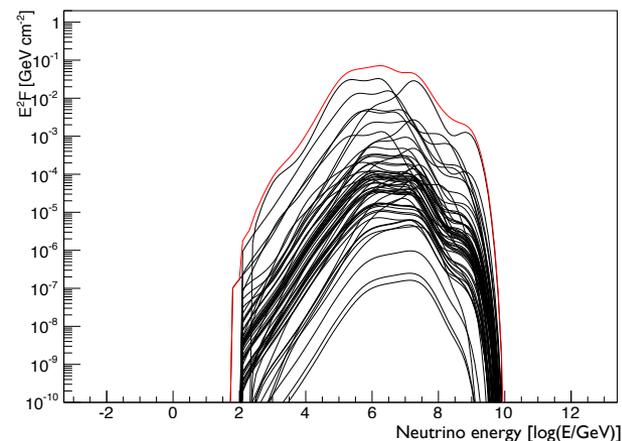




# GRB Selection

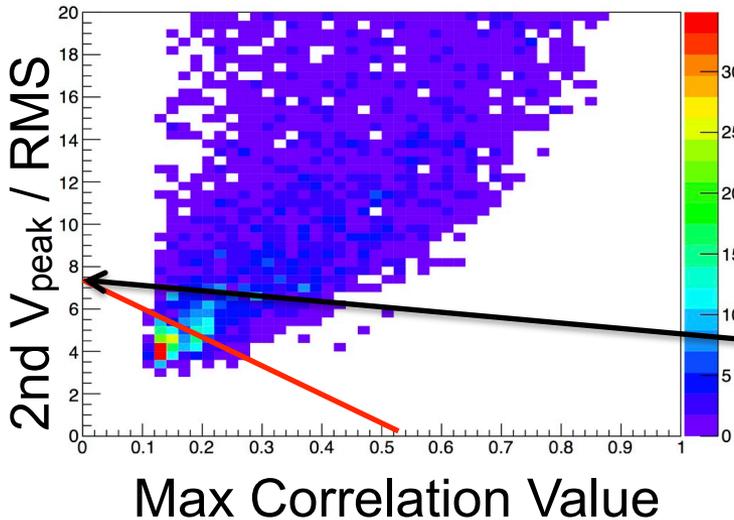


- Selected 57 GRBs based on livetime and geometric acceptance
- Get fluences for each GRB from NeuCosmA simulation and then total
- Tune cuts based on modeled neutrino fluence

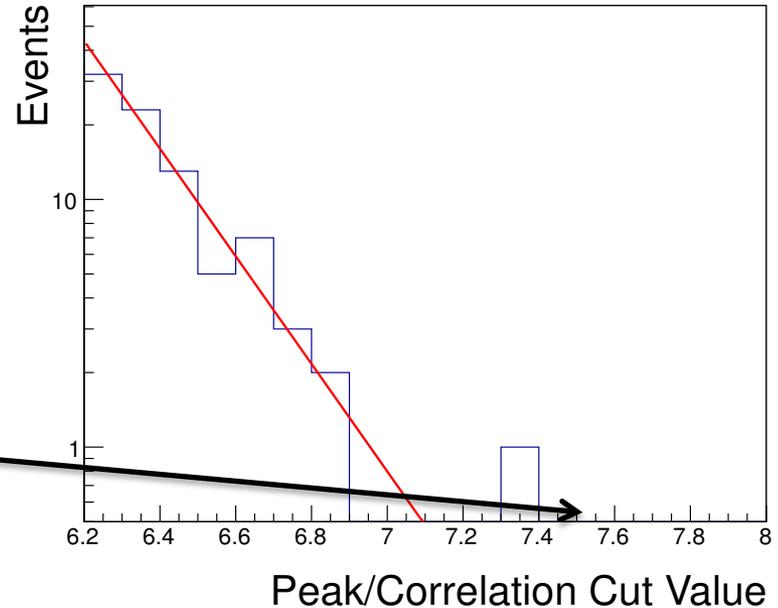




Simulated  $10^{18}$  eV neutrino set  
with cuts applied



Testbed Data



Optimize the cut parameters:

- Fit the background distribution with an exponential
- Integrate extrapolation to get expected background
- $S_{upper}$  is the 90% confidence limit on the signal for an expected background
- $N_{passed,sim}$  is the weighted number of passed simulated neutrinos from an expected flux
- Maximize R to optimize for best limit

$$R = \frac{N_{passed,sim}}{S_{upper}}$$



- All optimized cut parameters relaxed for GRB neutrino search when compared with diffuse neutrino search
- Factor of 2.4 improvement in efficiency against a simulated GRB flux
- Another cut for rejecting CW was removed

Cut	Reconstruction Quality Cut		Peak/Correlation Cut
	Parameter	$A_{\text{peak}}$	$A_{\text{peak}}/A_{\text{total}}$
Diffuse Neutrino Search	50 deg <sup>2</sup>	1.5	8.8
GRB Neutrino Search	70 deg <sup>2</sup>	16.2	7.5



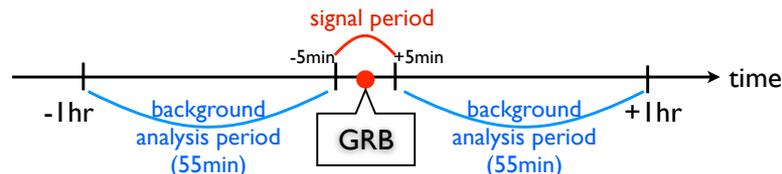
# Preliminary Results

Stage 1 (90% background period unblinding):

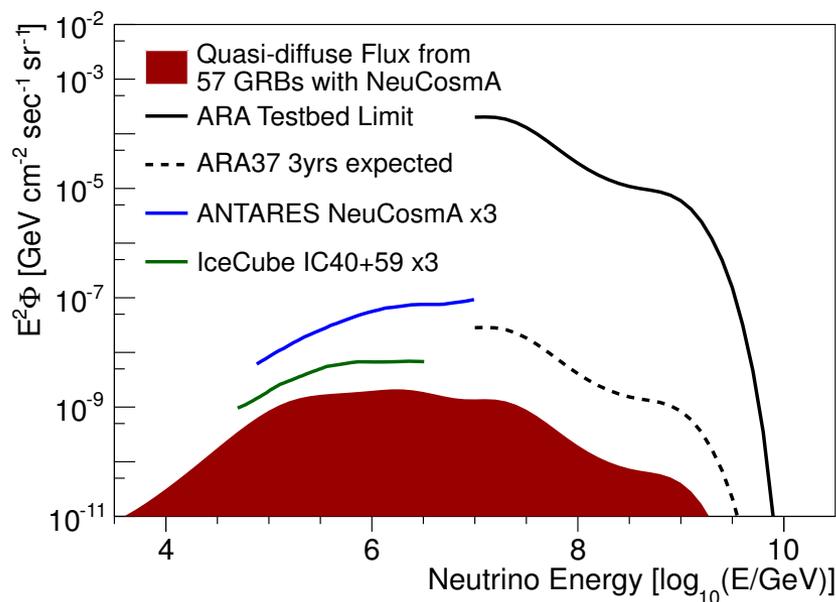
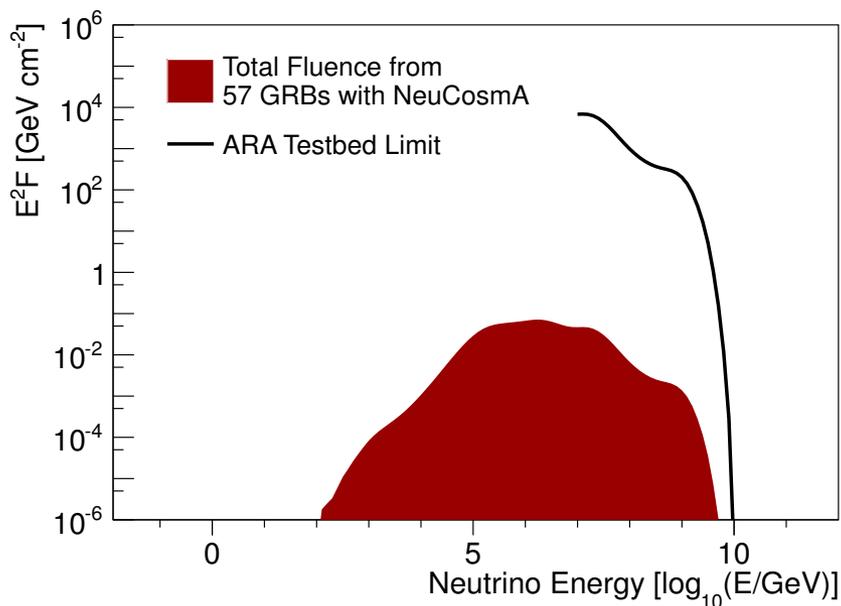
Expected background events: 1.2  
2 events survived

Stage 2 (signal period unblinding):

Expected background: 0.12, Expected neutrinos:  $1.7 \times 10^{-5}$   
0 events survived



First quasi-diffuse flux limit above  $10^{16}$  eV (arxiv:1507.00100)





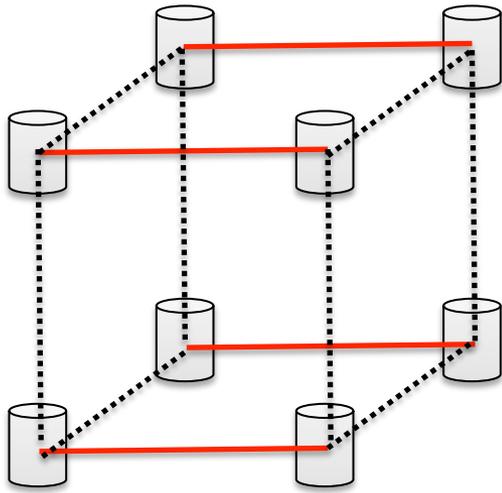
# BACKGROUND REJECTION FOR A REGULAR ARRAY



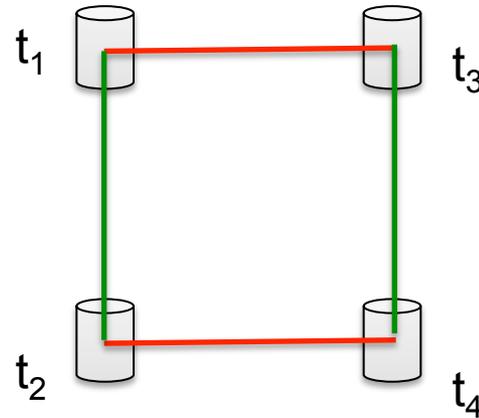
- 100's of millions of events – too many to efficiently use complex reconstruction methods
  - Need  $< 0.1\%$  thermal acceptance to be efficient
- Can we create an adaptable, efficient filter-level algorithm
- Goals:
  - Computationally simple
  - Easily differentiates between signal and noise
  - Decrease volume of data to then use more computationally intensive techniques (ray-tracing, etc)
  - Single understandable output
  - Easily optimizable
- Ultimate goal is a deep station analysis of current data
  - Perhaps use algorithm as a trigger or filter to the North?



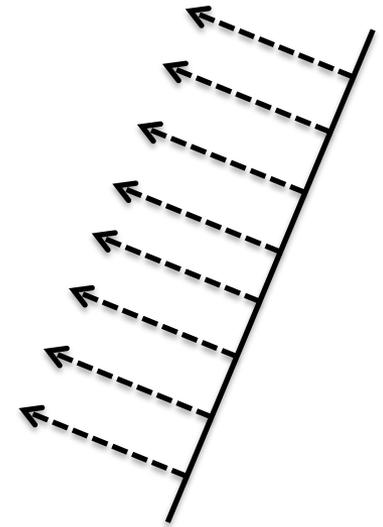
# Planar Signal Wavefront



— = similar pairs



— = A-type pairs  
 — = B-type pairs



$$\Delta t_{A,i} = t_3 - t_1$$

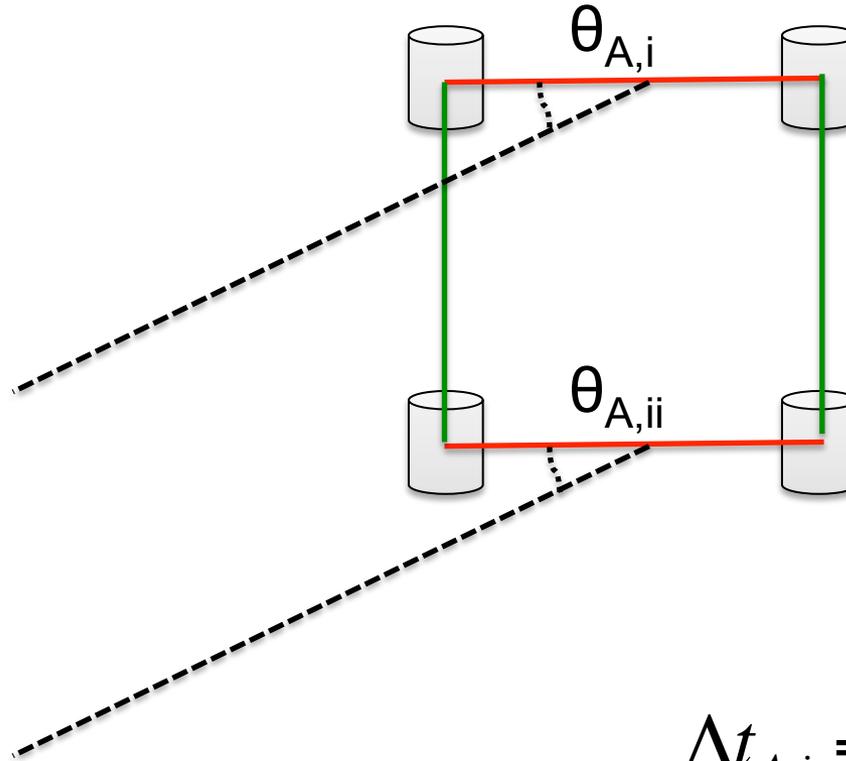
$$\Delta t_{A,ii} = t_4 - t_2$$

$$\Delta t_{A,i} \approx \Delta t_{A,ii}$$

- Divide array into faces
- Difficult to directly compare timing from different sets of pair-types – what to do?



# Angle of Incidence



- Use the angle from the baseline
- Comparable between different pair types

$$\theta_{A,i} \approx \theta_{A,ii} \quad \cos(\theta_{A,i}) \approx \cos(\theta_{A,ii})$$

$$\Delta t_{A,i} = \frac{n}{c} \cos(\theta_{A,i}) \Delta d_{A,i}$$

$$\cos(\theta_{A,i}) = \frac{c \Delta t_{A,i}}{n \Delta d_{A,i}}$$



- Similar time differences  $\rightarrow$  small variation
  - Find the “RMS” around their average

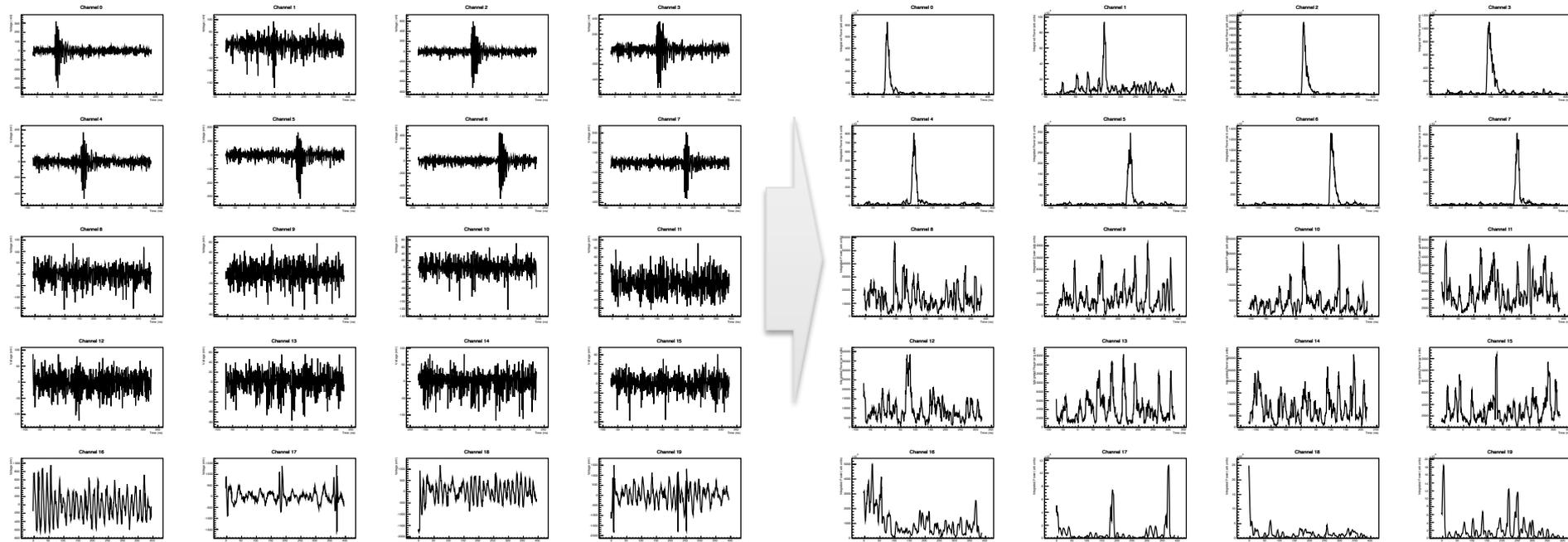
$$\overline{\cos(\theta_A)} = \frac{\cos(\theta_{A,i}) + \cos(\theta_{A,ii})}{2}$$

$$\text{RMS}(\cos(\theta_A)) = \sqrt{\frac{(\cos(\theta_{A,i}) - \overline{\cos(\theta_A)})^2 + (\cos(\theta_{A,ii}) - \overline{\cos(\theta_A)})^2}{2}}$$

- $\text{RMS}(\cos(\theta)) < 0.1$  if the arrival directions agree
- Also corrects for differences in baseline lengths



## Calibration pulser event



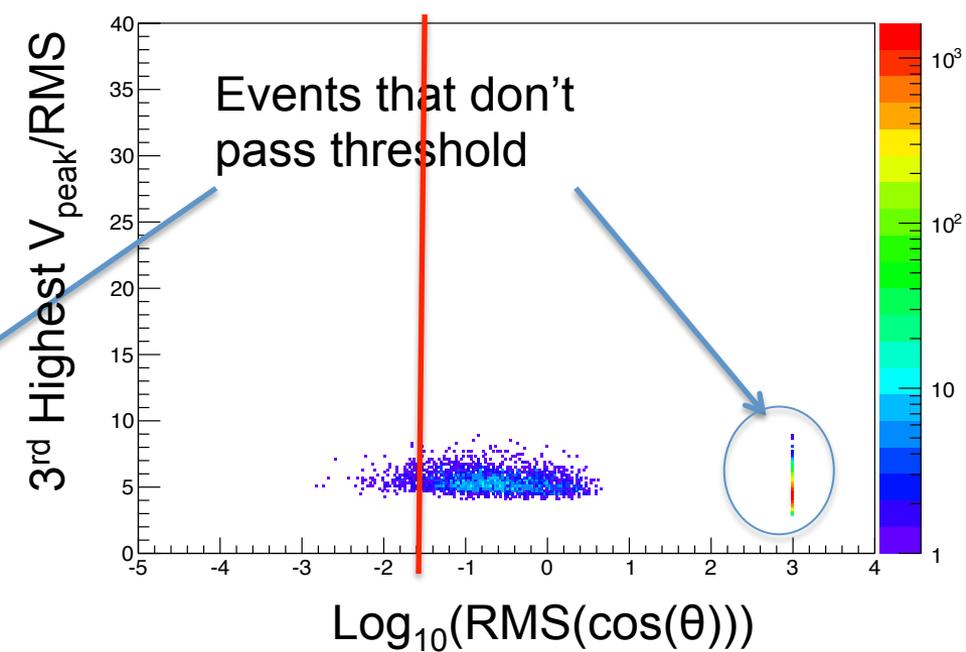
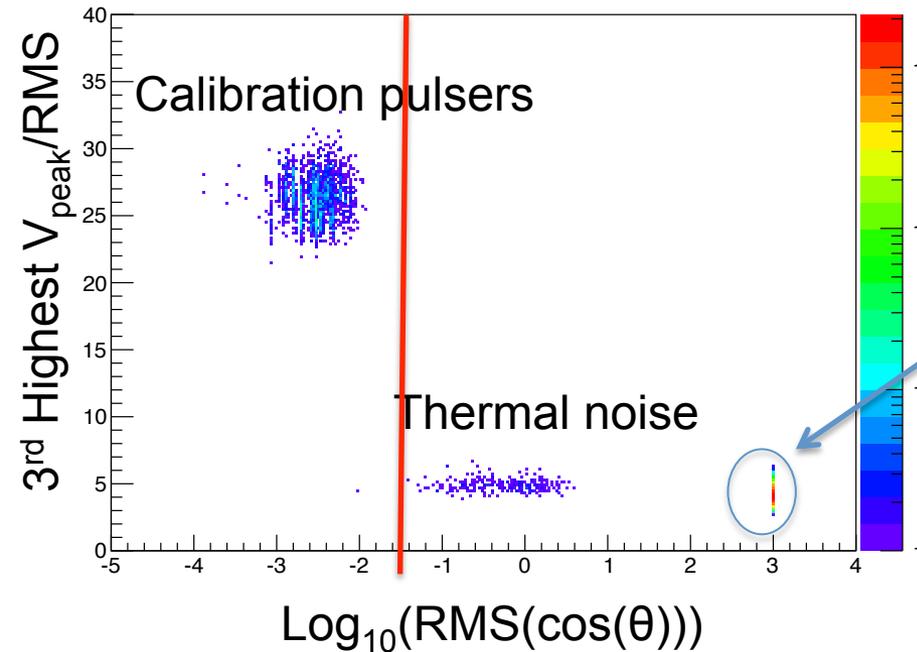
- To decrease noise fluctuations, scan an integrated power window of 5 ns
- Find the two highest peaks, use these as “hit times” for that channel
- Apply a threshold:
$$\frac{\text{RMS}(5 \text{ ns around the peak})}{\text{RMS}(\text{waveform})} > \text{Threshold}$$
- Find the face with the timing that agrees best with incoming signal (lowest face RMS)



Station A2, Run 1798  
Threshold = 2.5

Vpol

Hpol

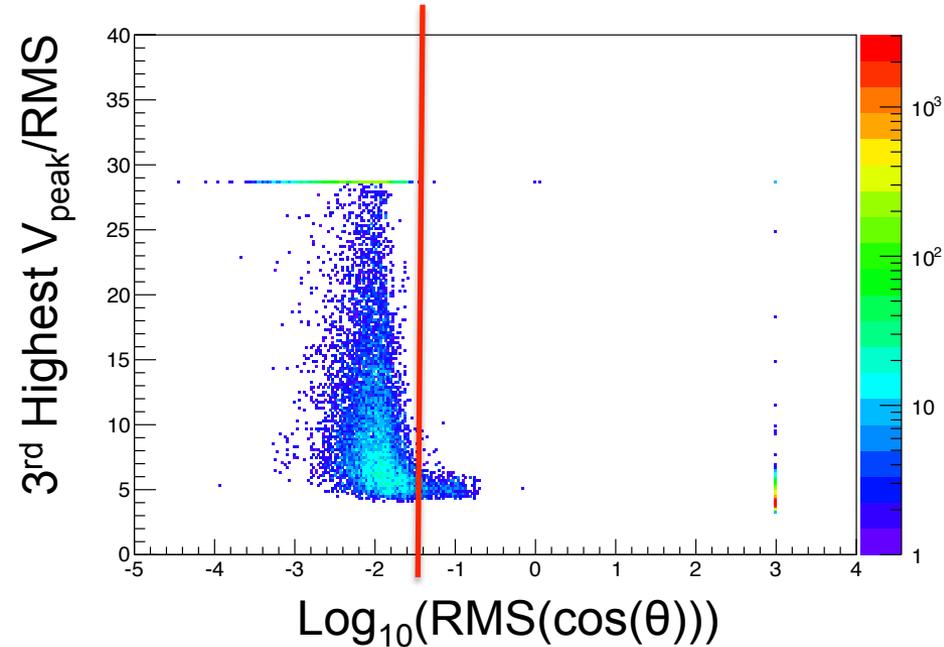
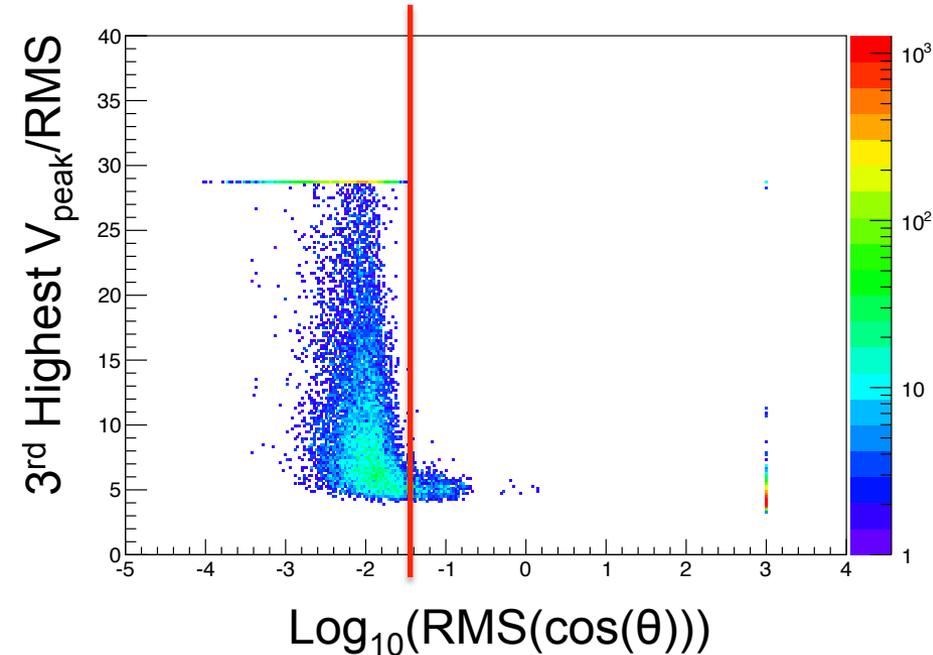


- More event pass threshold in Hpol antennas
  - use separate thresholds for Vpol and Hpol

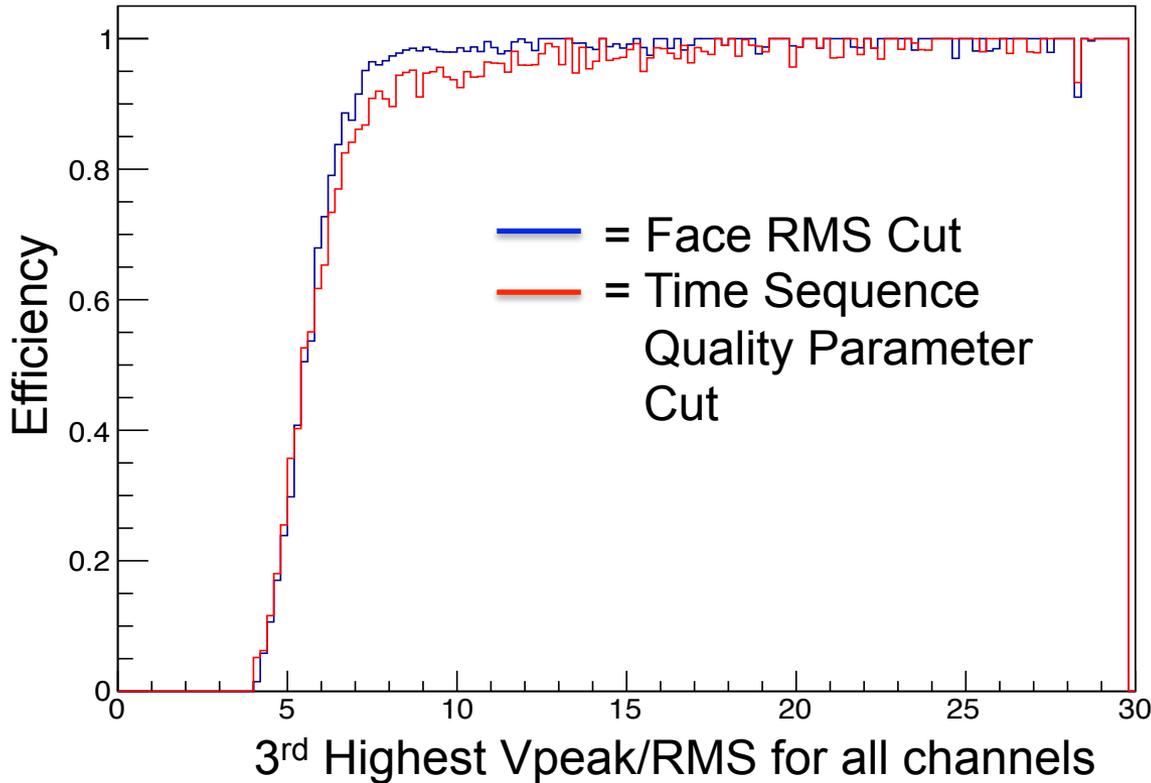


Vpol

Hpol



- Simulated  $10^{19}$  eV neutrino events generated with AraSim simulation package
- Good separation at high signal strength
- Reasonable separation at lower signal strength
- Noise starts to dominate over low SNR signals – difficult to reconstruct anyway



Vpol thresh = 2.5  
Hpol thresh = 2.9  
 $\text{Log}_{10}(\text{RMS}(\cos(\theta))) < -1.5$

Time Sequence Quality  
Parameter > 0.6  
(value used in 2013  
A2/3 analysis,  
described by Kael in  
earlier talk)

- Data RF events - Face RMS efficiency = 0.08 %, TSQP = 0.08 %
- Simulation - Face RMS efficiency = 83.1%, TSQP efficiency = 81.6%
- Currently filter algorithms comparable
- Face RMS not optimized, may improve even more



- Testbed GRB neutrino search
  - Optimized search cuts
  - Limiting background search window → cut relaxation
  - New quasi-diffuse flux limit above  $10^{16}$  eV
  - Projected limit for ARA37
- New filter-level cut
  - Efficient in rejecting thermal noise – 0.08% acceptance
  - Efficient in retaining simulated neutrinos
    - > 95% at high SNR
  - Flexible
    - Can characterize individual faces separately
    - Can treat hpol and vpol separately
  - Can improve event selection at the analysis level and maybe even the trigger level
  - Will optimize cut in full analysis (later this year!)



## ***Computing in High-Energy Astro-Particle Research***

Topics: Genetic programming, analytics, data analysis, feature selection, high-performance computing

Activities: tutorials, lectures, example code packages

Who: Members of ANITA, ARA, LIGO, SKA, others

Experts in genetic programming from industry and academia



***When: August 24th – 26th, 2016***

***Where: Center for Cosmology and AstroParticle***  
Contact: Carl Pfendner [pfendner.1@osu.edu](mailto:pfendner.1@osu.edu) or

***Physics (CCAPP), The Ohio State University***  
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# Questions?

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