

Atmospheric-Neutrino Flux-Integrated Differential Cross-Section Measurement in IceCube

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Overview

Overview

- Introduction
- The IceCube Experiment
- MC Sample
- Analysis
- Conclusion



Introduction

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Introduction

Standard Cross-section Measurements

- Modern neutrino experiments rely heavily on neutrino flux predictions, which have been historically difficult to define precisely

$$\text{rate} \sim \int \Phi \times \sigma \times \epsilon$$

- Hampers our ability to determine the cross-section from a measurement
- Common to model the neutrino flux, assuming a neutrino interaction cross-section model which leads to biases in the measurement

Many distributions

Divide by
integrated flux

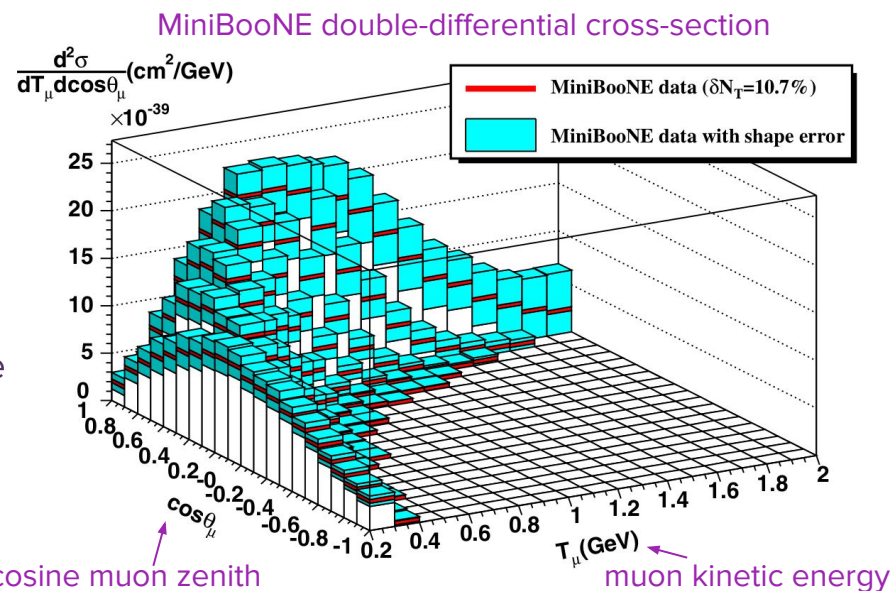
Flux-integrated
Differential
Cross-section

Introduction

No atmospheric based neutrino experiment has yet performed a differential cross-section measurement

Flux-integrated Differential Cross-section Measurement

- The cross-section has been normalised with an absolute predicted flux and has not been adjusted based on measured processes
- Multitude of cross-section distributions allow for the most stringent tests for a given physics mode
- Most model-independent measurement



PRD81(2010)092005
MiniBooNE collaboration

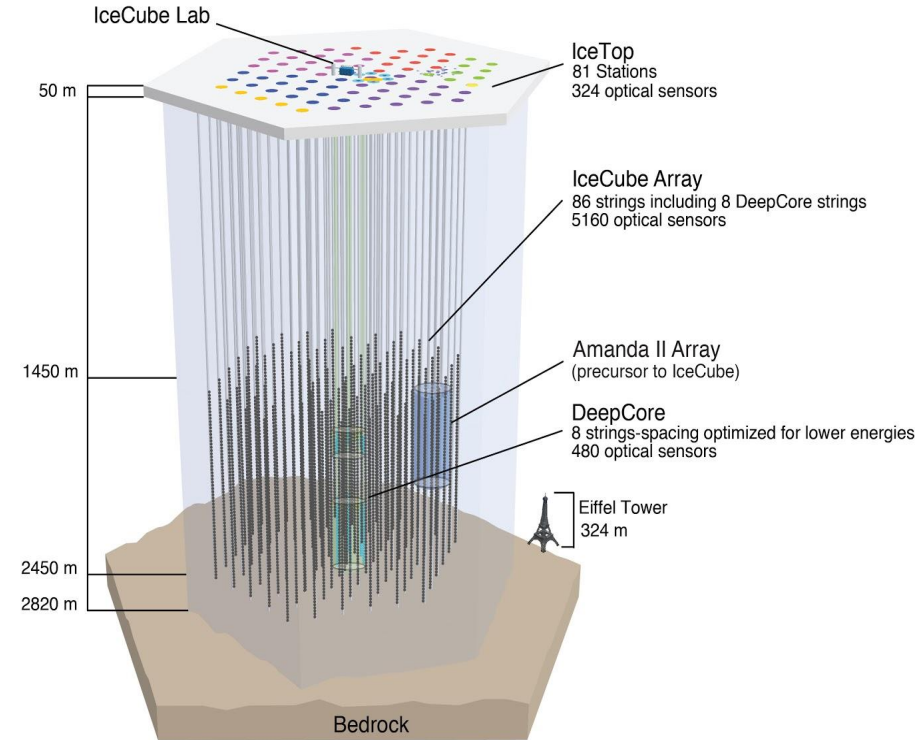


The IceCube Detector

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The IceCube Detector

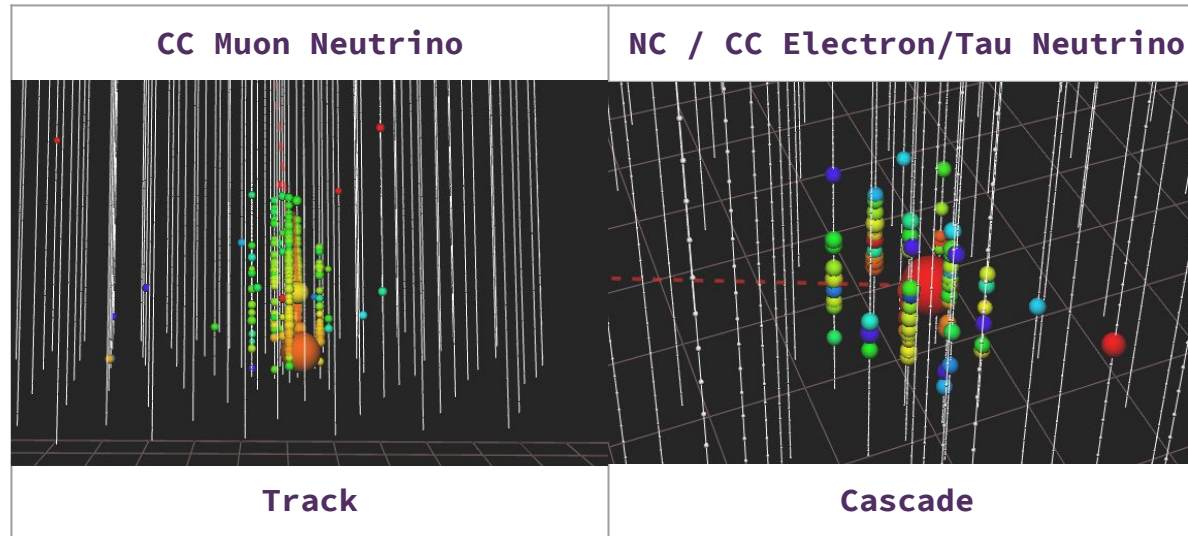
- Located at the South Pole
- 5160 Digital Optical Modules (DOMs)
embedded in a cubic km of clear Antarctic ice
- DOMs contain PMTs which detect Cherenkov light
- Main Physics Goals
 - ◆ Detection of astrophysical neutrinos
 - ◆ Measure atmospheric neutrino oscillations (DeepCore)



Atmospheric Neutrino Signal

IceCube triggers over 100,000 atmos. neutrino event per year - high statistics is ideal for a full 4π differential cross-section measurement

Event Signatures





MC Sample

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MC Sample

Using data sample optimised for a search for $\nu_{\mu} \rightarrow \nu_{\tau}$ appearance using the DeepCore inner detector

- Exposure = 3 years
- Multiple BDTs provide significant background rejection of muons
- Straight cuts focus on rejecting noise dominated triggers and muon backgrounds

Reconstruction used is based on the MultiNest algorithm

Atmospheric one-year-averaged neutrino flux at the South Pole (HAKKM, 2014)



Analysis

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Triple Differential Cross-section Formula

i, j, k = true index
 α, β, γ = reconstructed index

$$\left(\frac{d^3\sigma}{dE d\cos\theta d\phi} \right)_{ijk} = \frac{\sum_{\alpha\beta\gamma} U_{ijk\alpha\beta\gamma} (d_{\alpha\beta\gamma} - b_{\alpha\beta\gamma})}{\epsilon_{ijk} (\Phi T) \Delta E_i \Delta \cos\theta_j \Delta \phi_k}$$

Unsmearing Matrix

Data vector

Background vector

Efficiency

Flux integrated over time and $E, \cos\theta, \phi$

Target number

Bin widths

$i, j, k = \text{true index}$
 $\alpha, \beta, \gamma = \text{reconstructed index}$

$$\left(\frac{d^3\sigma}{dE d\cos\theta d\phi} \right)_{ijk} = \frac{\sum_{\alpha\beta\gamma} U_{ijk\alpha\beta\gamma} (d_{\alpha\beta\gamma} - b_{\alpha\beta\gamma})}{\epsilon_{ijk} (\Phi T) \Delta E_i \Delta \cos\theta_j \Delta \phi_k}$$

Labels in diagram:
 - **Unsmeared Matrix** (points to $U_{ijk\alpha\beta\gamma}$)
 - **Efficiency** (points to ϵ_{ijk})
 - **Flux integrated over time and E, cos θ , ϕ** (points to (ΦT))
 - **Target number** (points to $(d_{\alpha\beta\gamma} - b_{\alpha\beta\gamma})$)
 - **Bin widths** (points to $\Delta E_i \Delta \cos\theta_j \Delta \phi_k$)

Data vector

Background vector

- ➔ **Reconstructed signal**
- ➔ **Signal Definition**
 - ◆ Muon ν , $\bar{\nu}$ -H₂O CC induced muons
- ➔ **Predicted Background are cosmic rays**
 - ◆ Data consists of 20%–30% cosmic rays
 - ◆ For now just using MC sample, so only considering pure signal

Binning:

- Energy
 - 10 bins
 - log [1, 10^{2.5}] GeV
- Cos(zenith)
 - 10 bins
 - linear [-1, 1]
- Azimuth
 - 4 bins
 - linear [0, 360]°

Reconstructed muon energy spectrum for $\cos\theta - [0.00 - 0.20]$

$i, j, k = \text{true index}$
 $\alpha, \beta, \gamma = \text{reconstructed}$

$$\left(\frac{dE}{d\Omega dt} \right)$$

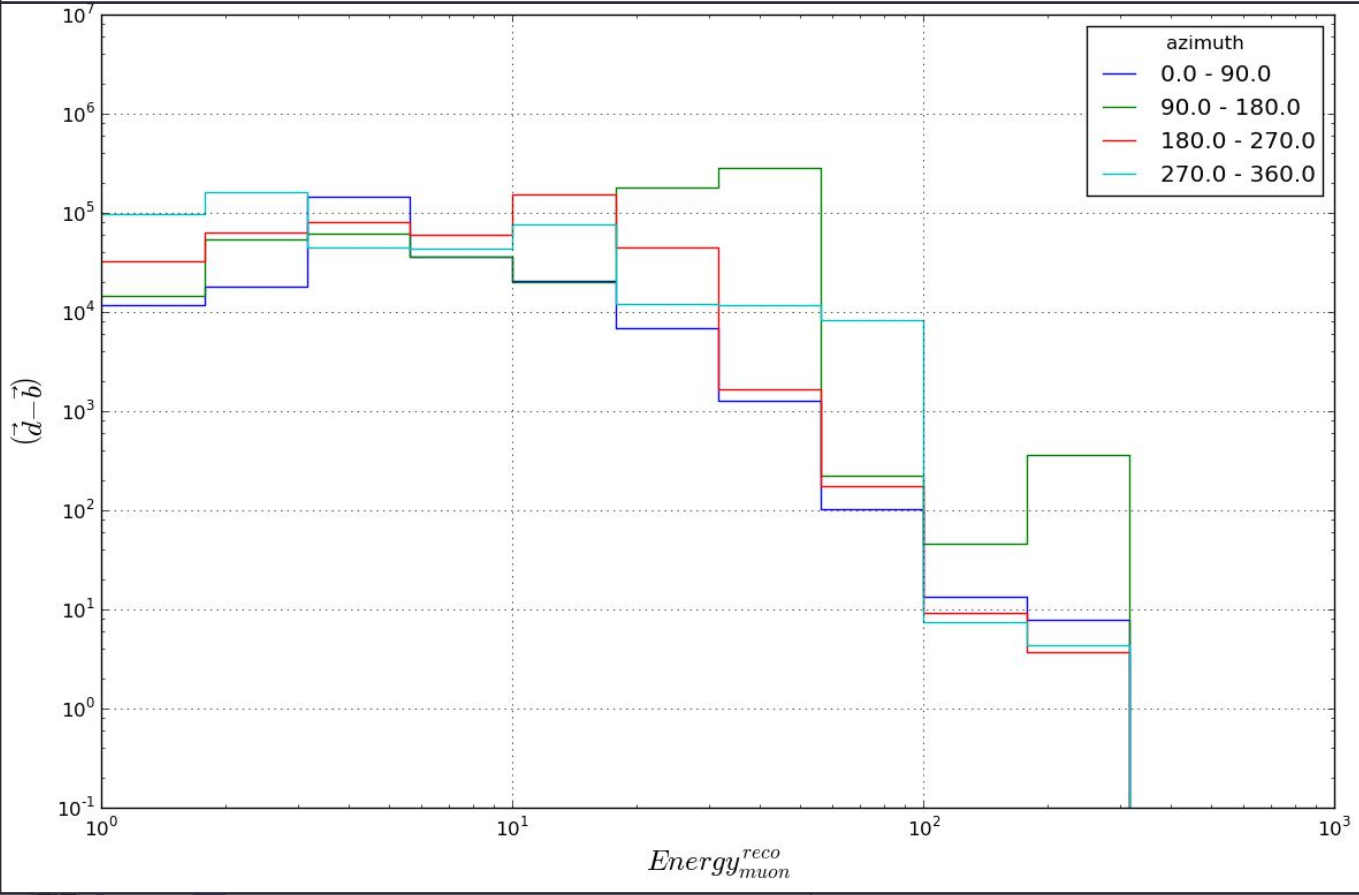
Data

→ Reco

→ Signal



→ Prediction



GeV

o

$i, j, k = \text{true index}$
 $\alpha, \beta, \gamma = \text{reconstructed index}$

$$\left(\frac{d^3\sigma}{dE d\cos\theta d\phi} \right)_{ijk} = \frac{\sum_{\alpha\beta\gamma} U_{ijk\alpha\beta\gamma} (d_{\alpha\beta\gamma} - b_{\alpha\beta\gamma})}{\epsilon_{ijk} (\Phi T) \Delta E_i \Delta \cos\theta_j \Delta \phi_k}$$

Unfolding

Labels in diagram:
Data vector
Background vector
Unsmearing matrix
Efficiency
Bin widths
Target number
Flux integrated over time and E, cos theta, phi

→ Unsmearing matrix

- ◆ Corrects for the ‘smearing’ of the data due to detection related effects and biases from event selection
- ◆ Quantifies the probability for an event in truth bin ijk to be reconstructed to bin α, β, γ

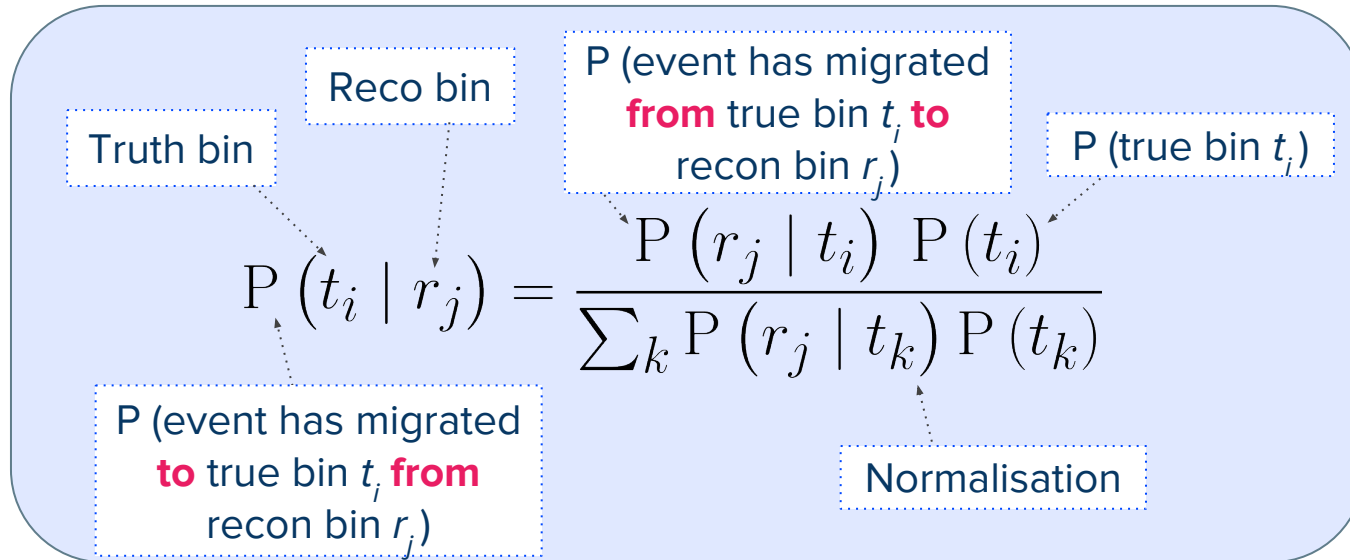
→ Efficiency

- ◆ Accounts for events lost due to selection cuts

Unsmearing Matrix

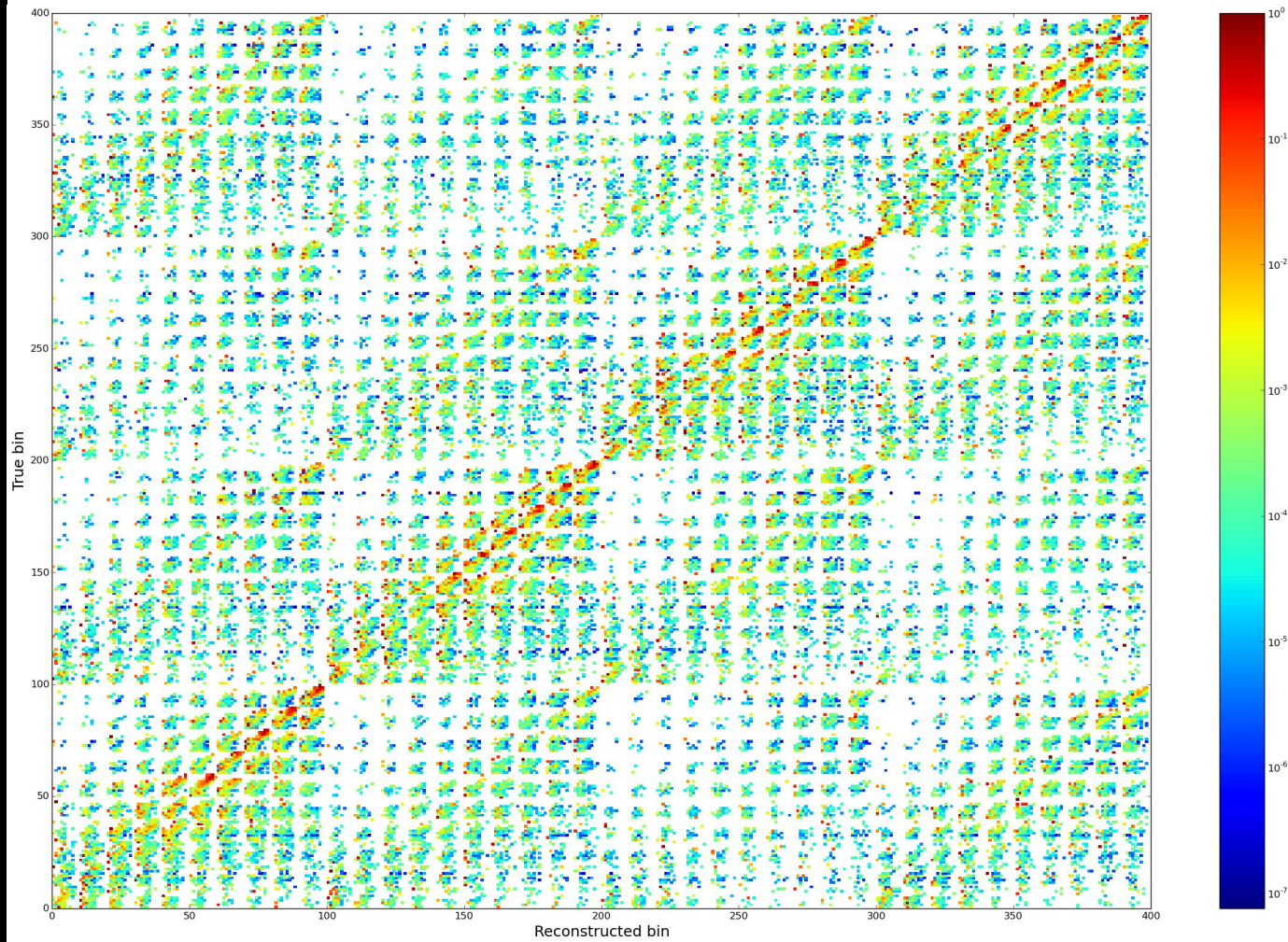
400 bins

Bayesian Unsmearing allows for unfolding in all three dimensions with full correlations

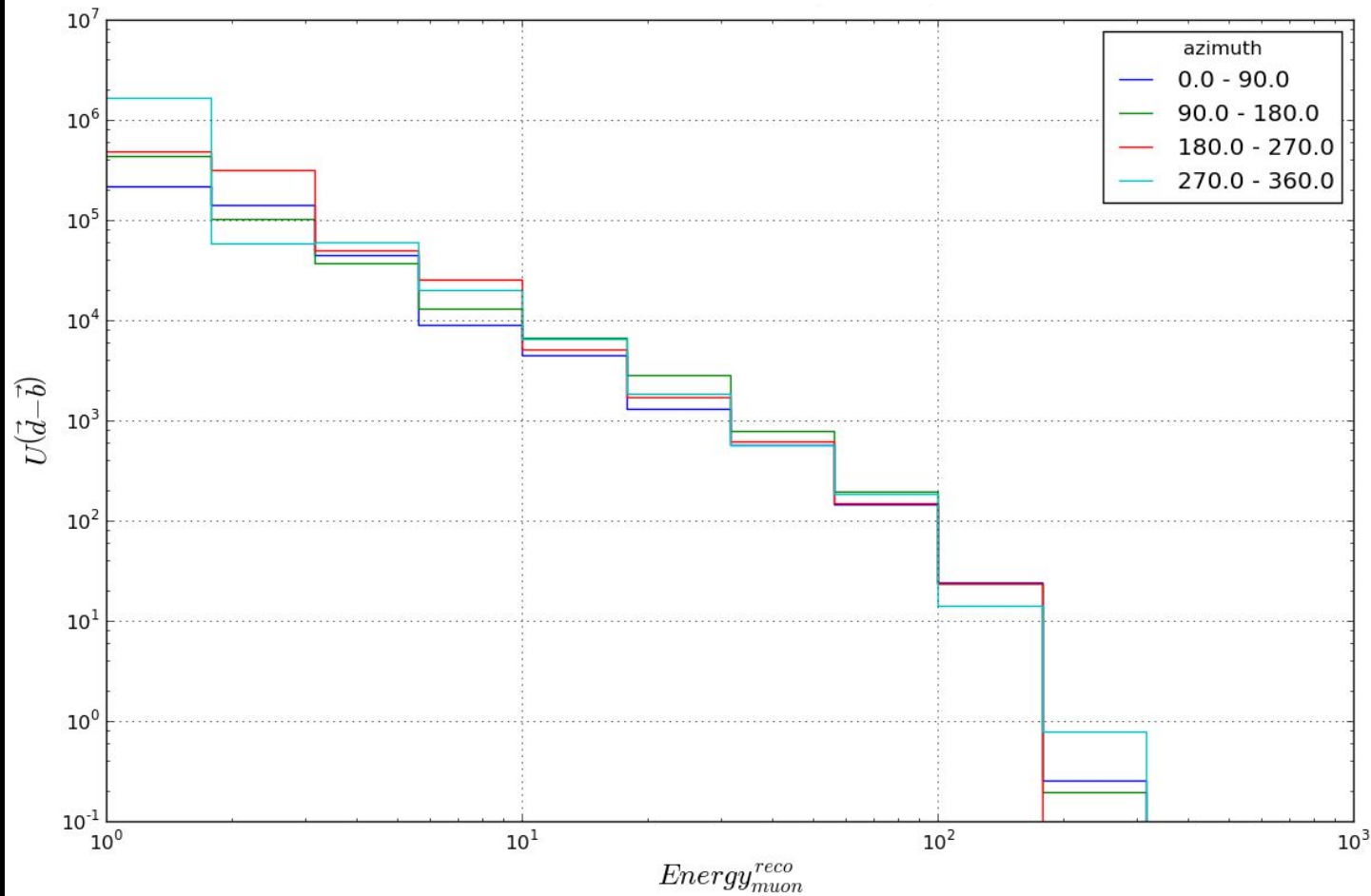


D'Agostini
Nucl.Instrum.Meth. A362 (1995) 487-498

Unsmearing Matrix



Unsmearred reconstructed muon energy spectrum for $\cos\theta - [0.00 - 0.20]$



Efficiency

$$\epsilon_{ijk} = \frac{N_{ijk}^{\text{after}}}{N_{ijk}^{\text{before}}}$$

- Measure by comparing to generated GENIE MC events with correct normalisation
 - ◆ Normalisation found by comparing GENIE number of events with Honda Flux times the cross section of the spline which was input into GENIE to generate the events
- Normalised using
 - ◆ Target Number = 4.183×10^{37} water molecules
 - ◆ Exposure = 3 years

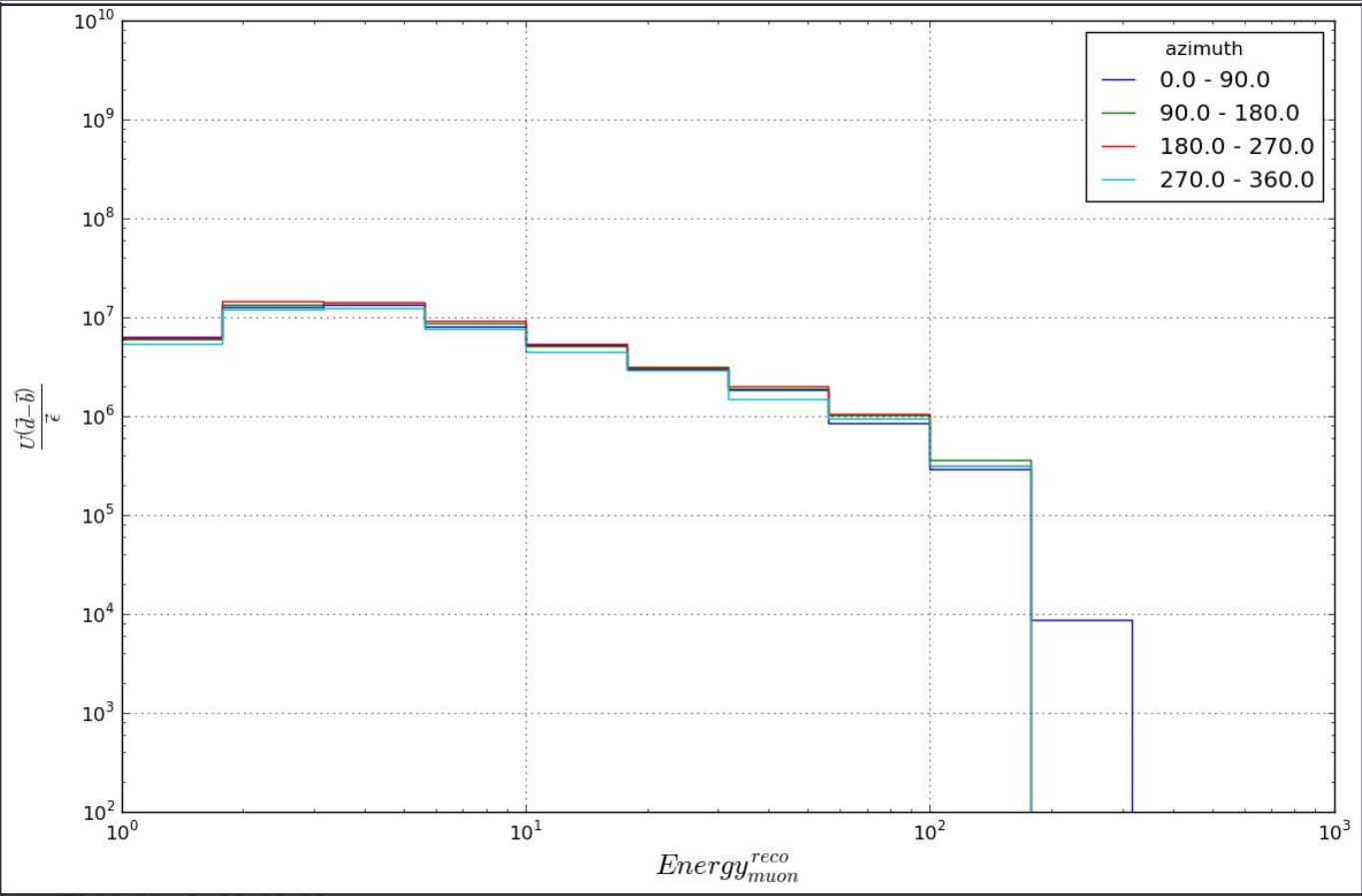
Unsmeared reconstructed muon energy spectrum with **efficiency correction** for $\cos\theta - [0.00 - 0.20]$

Efficiency

→ Meas



→ Norm



on
flux times
events

$i, j, k = \text{true index}$
 $\alpha, \beta, \gamma = \text{reconstructed index}$

Unsmearing Matrix Data vector Background vector

$$\left(\frac{d^3\sigma}{dE \, d\cos\theta \, d\phi} \right)_{ijk} = \frac{\sum_{\alpha\beta\gamma} U_{ijk\alpha\beta\gamma} (d_{\alpha\beta\gamma} - b_{\alpha\beta\gamma})}{\epsilon_{ijk} (\Phi T) \Delta E_i \Delta \cos\theta_j \Delta \phi_k}$$

Efficiency

Flux + Factors

Bin Widths

Target Number

Flux integrated over time and $E, \cos\theta, \phi$

- Bin Width
- Flux integrated over time and energy, $\cos\theta, \phi$
 - ◆ Exposure = 3 years
- Target Number = 4.183×10^{37} water molecules
 - ◆ Cylinder with $R = 600 \text{ m}$, $L = 1200\text{m}$
 - ◆ Ice density = 0.922 M g m^{-3}
 - ◆ Molar mass of water = $18.01528 \text{ g mol}^{-1}$

Flux-Integrated Triple-Differential Cross-section for $\cos\theta - [0.00 - 0.20]$

$i, j, k = \text{true index}$
 $\alpha, \beta, \gamma = \text{reconstruct}$

$$\left(\frac{d^3\sigma}{dE d\cos\theta d\phi} \right)$$

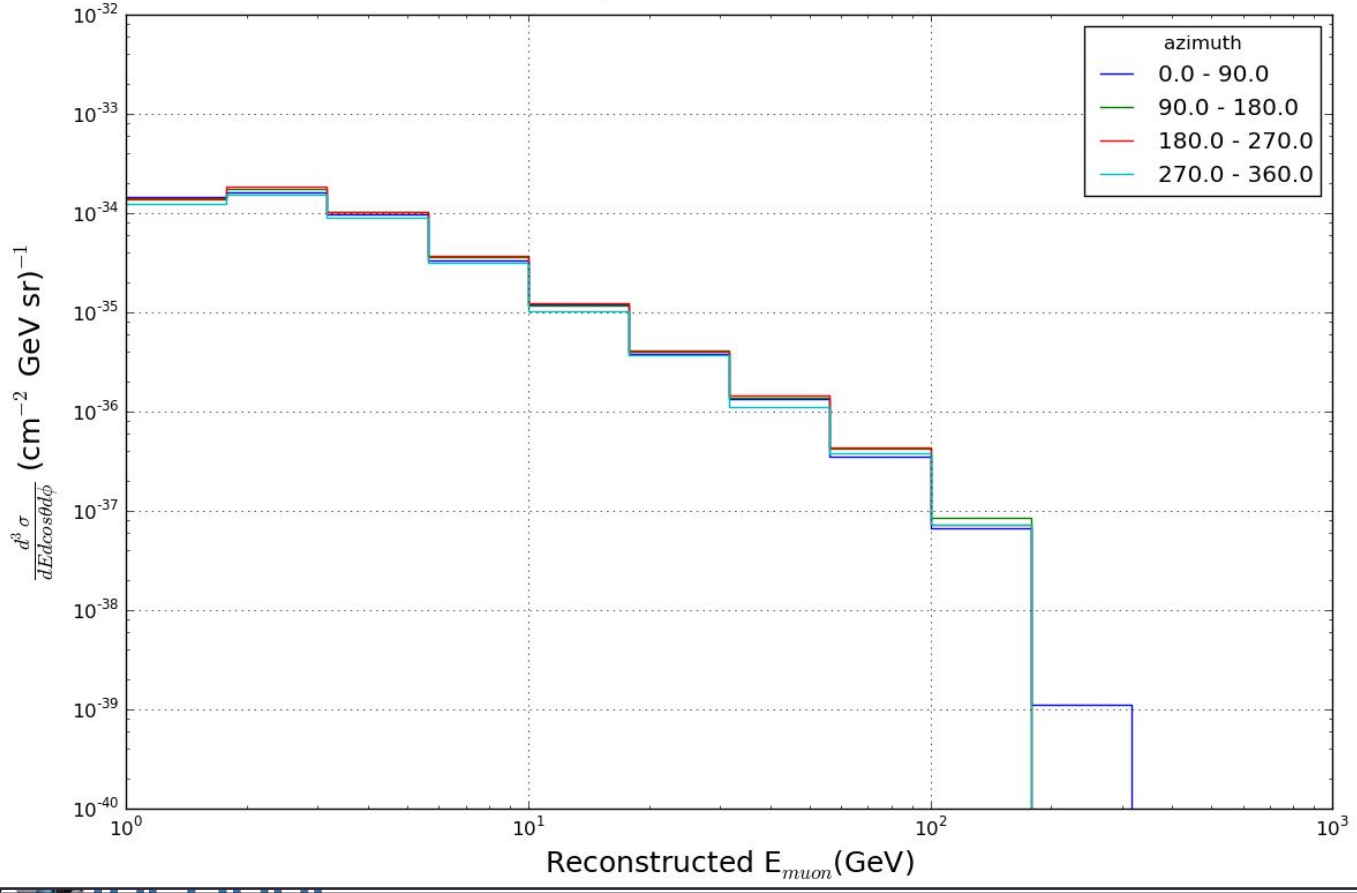
Flux +

→ Bin Width

→ Flux

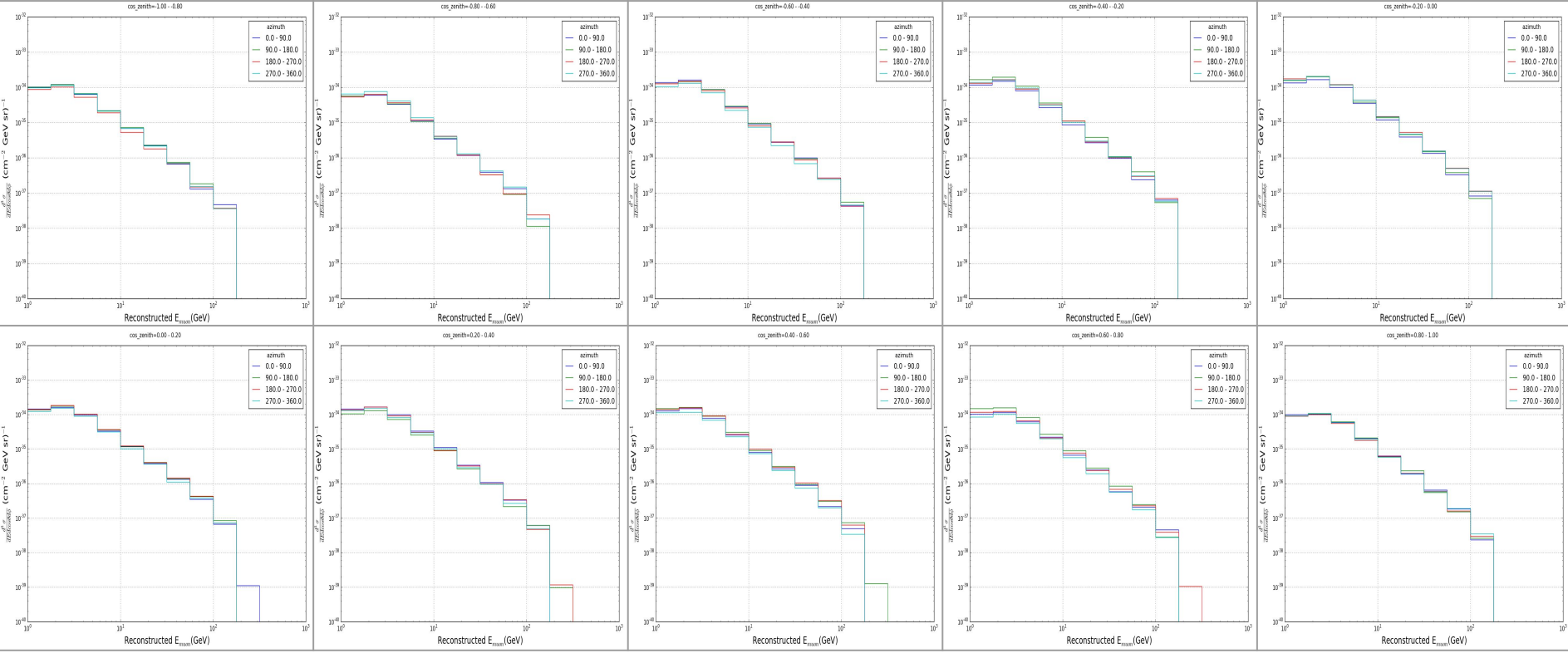


→ Target

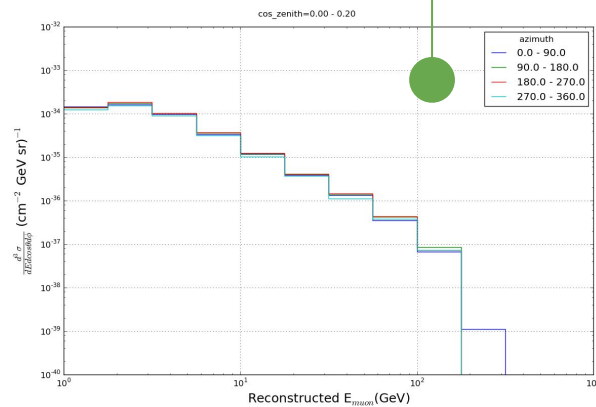
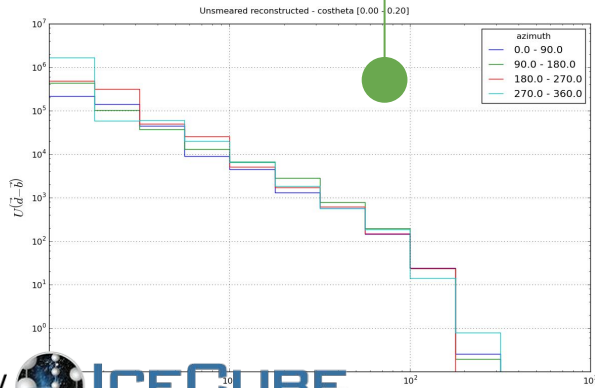
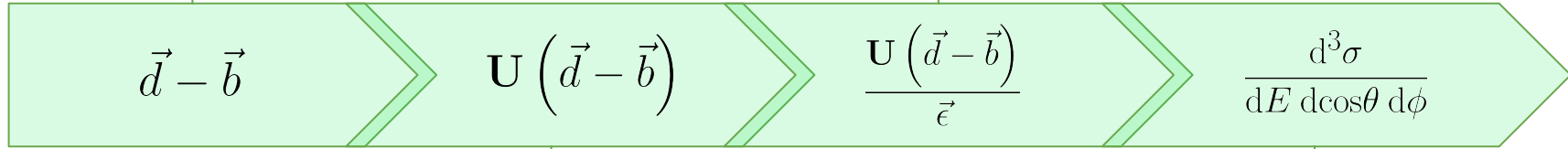
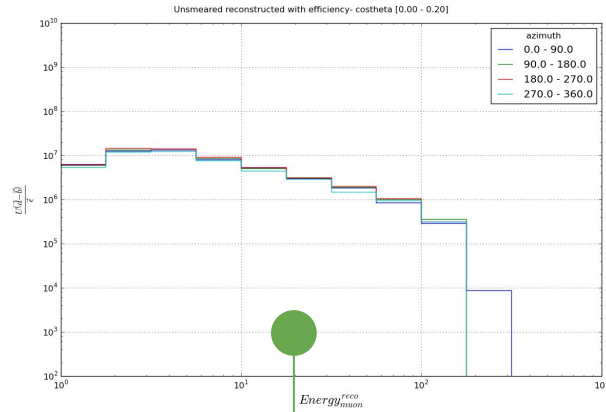
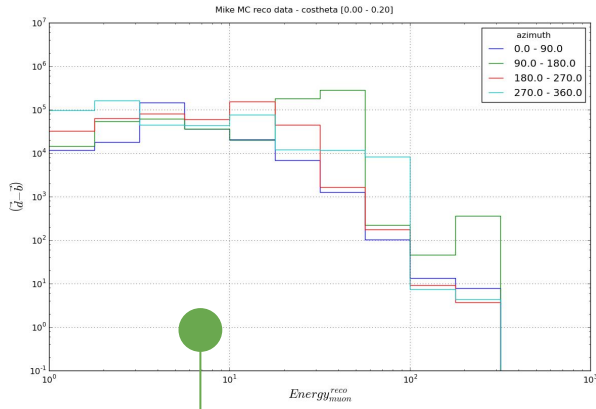


Work in Progress

Work in Progress



Results





Conclusion

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Conclusion

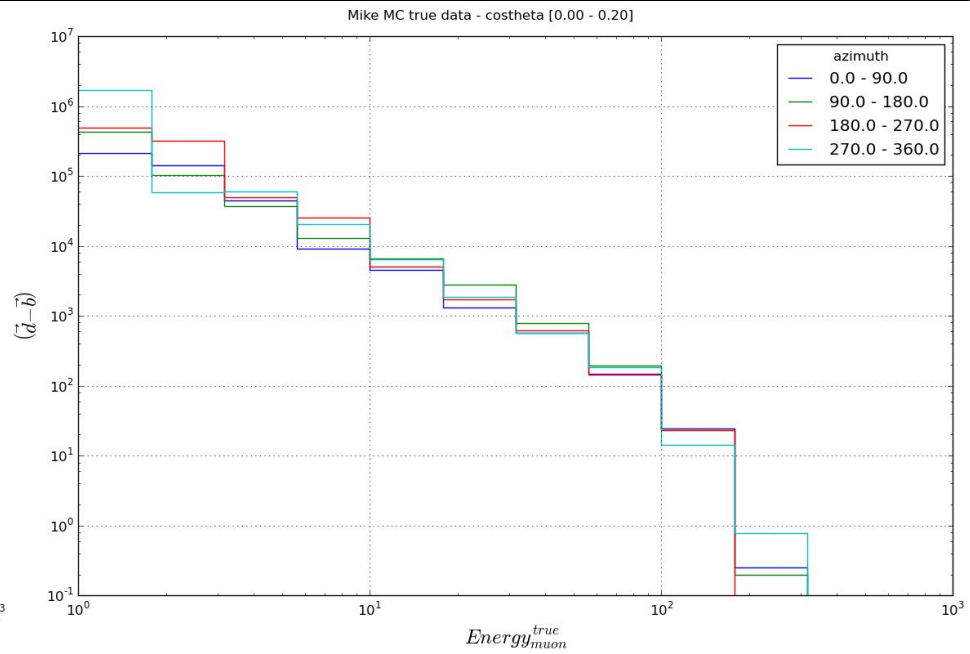
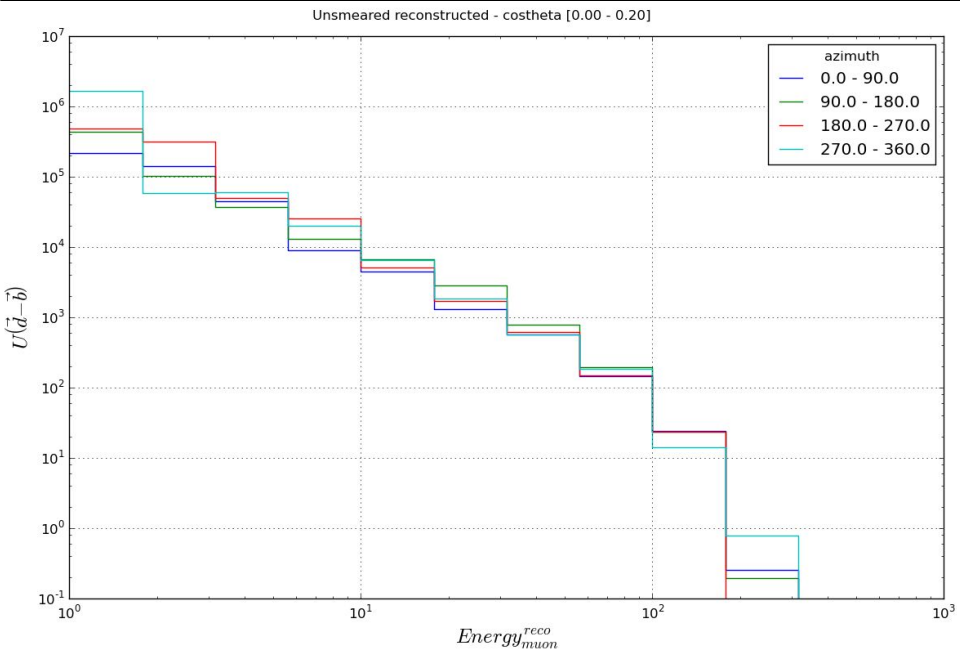
- Next steps are to use fake data
- ◆ Background subtraction
 - ◆ Statistical + systematic uncertainties

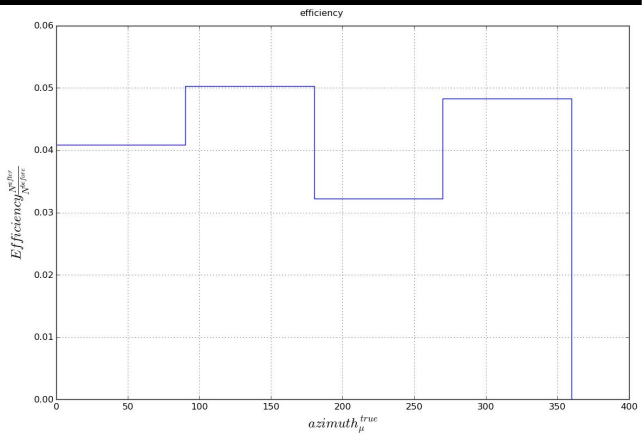
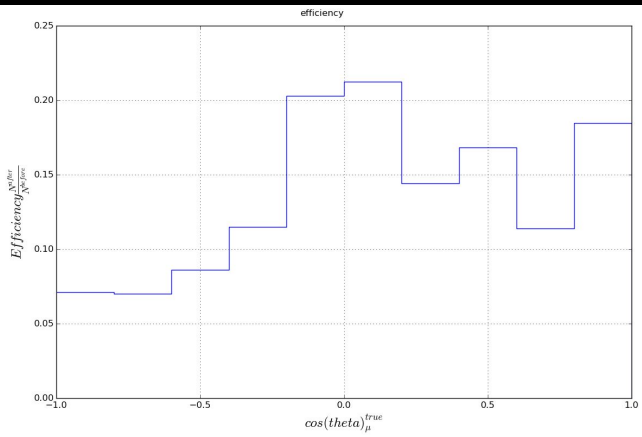
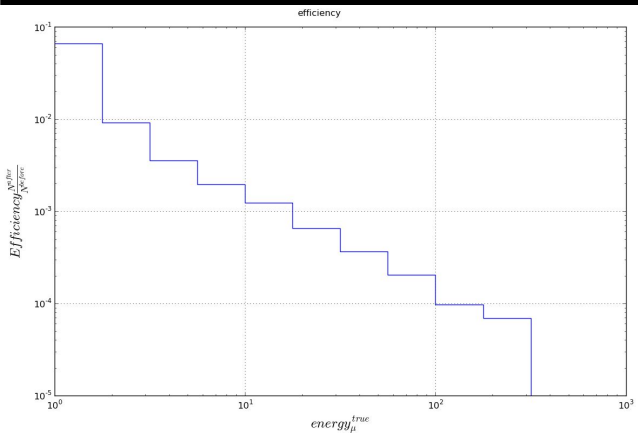
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**Flux-integrated Triple-Differential
Cross-section is the least model-
dependent data!**

Backup

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MC Sample

- Volume = cylinder with $R=600\text{m}$ and $H=1200\text{m}$
- Data
 - Two BDTs provide significant background rejection of muons and straight cuts focus on rejecting noise-dominated triggers. At final level, the sample includes both upgoing and downgoing events with roughly 26% of the rate consisting of atmospheric muons and 1% noise triggers.
- Level 4
 - BDT trained to reduce atmospheric background
- Level 5
 - A second BDT is trained at L5 to further reduce the atmospheric muon background by a factor of 10x.
- Level 6
 - Dedicated to muon and noise trigger removal