

# Unfolding the Atmospheric Muon Flux with IceCube: Investigating Stopping Muons and High-Energy Prompt Contributions

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for the IceCube Collaboration

July 17, 2025

# Introduction

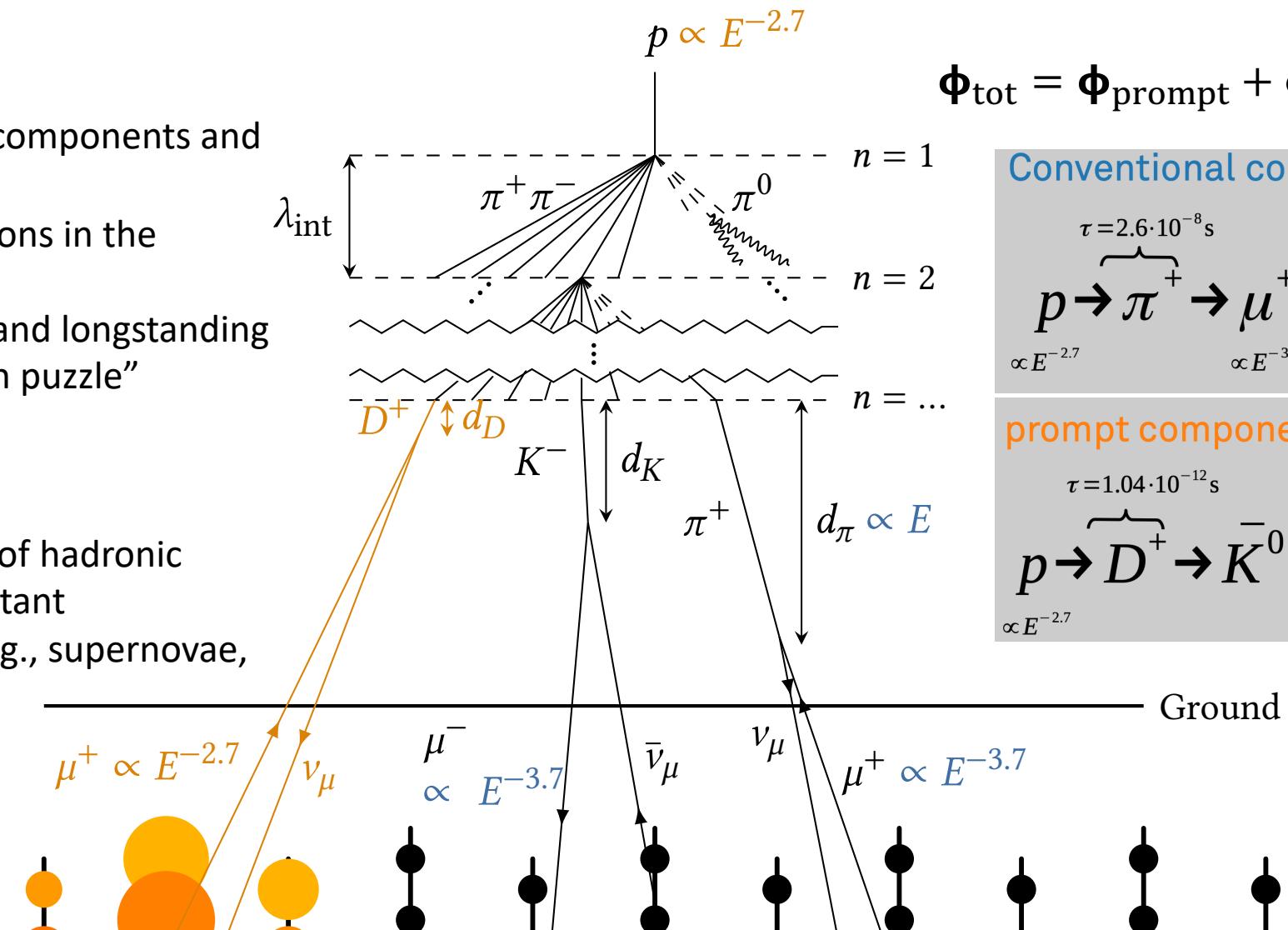
## Motivation

- Characterize muon flux components and depth intensity
- Probe hadronic interactions in the atmosphere
- Constrain uncertainties and longstanding questions like the “muon puzzle”

## Long term

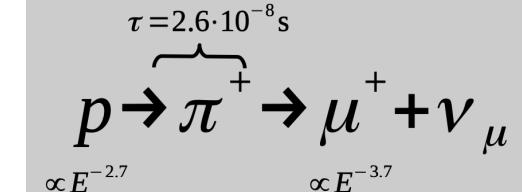
- Enhance understanding of hadronic processes relevant in distant astrophysical sources (e.g., supernovae, AGNs, ...)

**Conventional Muon:**  
Parent is pion or kaon

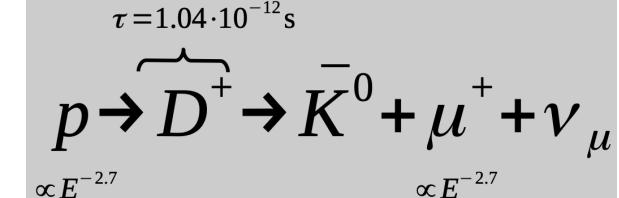


$$\Phi_{\text{tot}} = \Phi_{\text{prompt}} + \Phi_{\text{conventional}}$$

**Conventional component:**



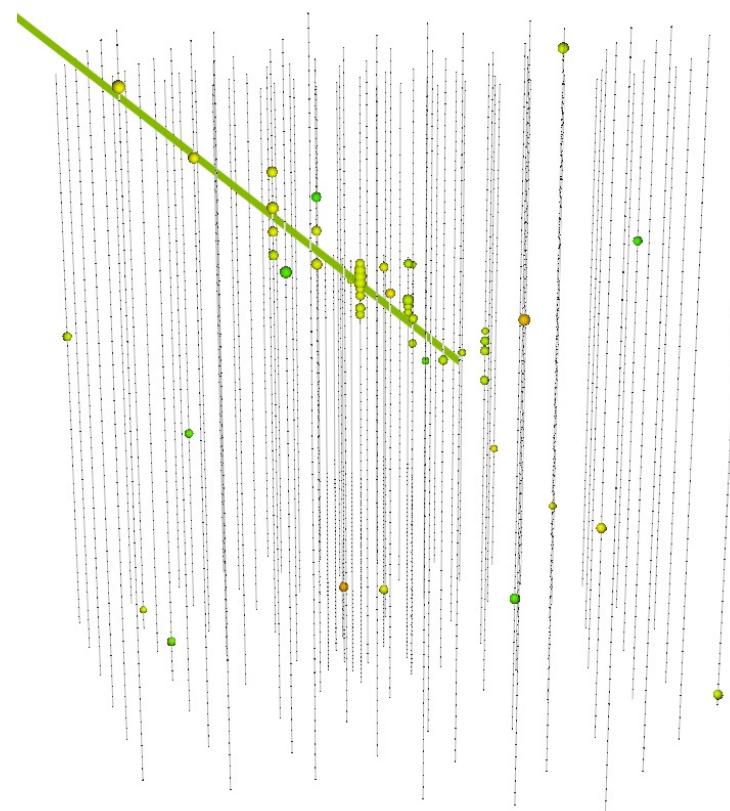
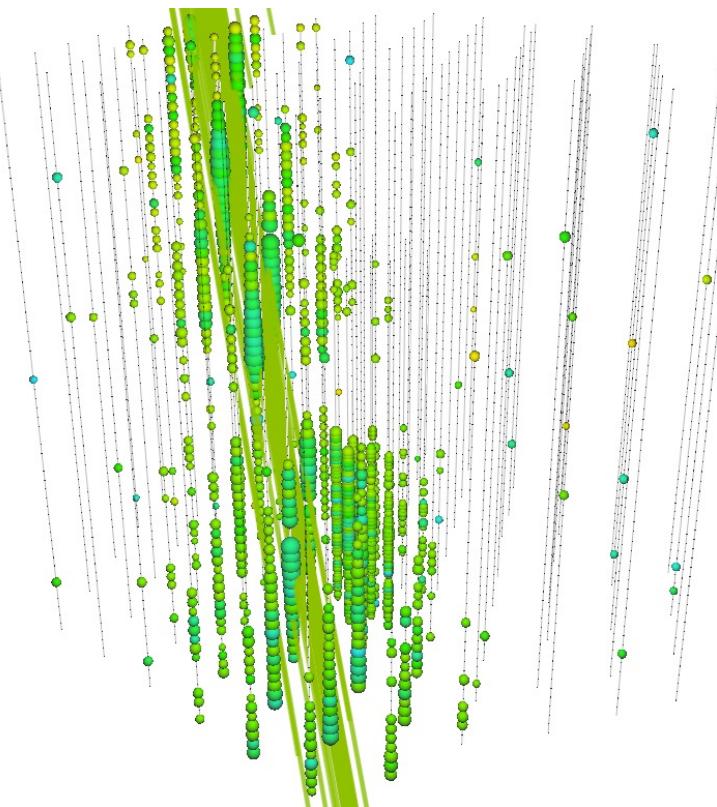
**prompt component:**



Credit: Ludwig Neste

# Stopping and Leading Muons

- Leading muons:
  - Most energetic muon in the bundle
  - High energies above 10 TeV
  - Focus on muons carrying > 40% of the entire bundle

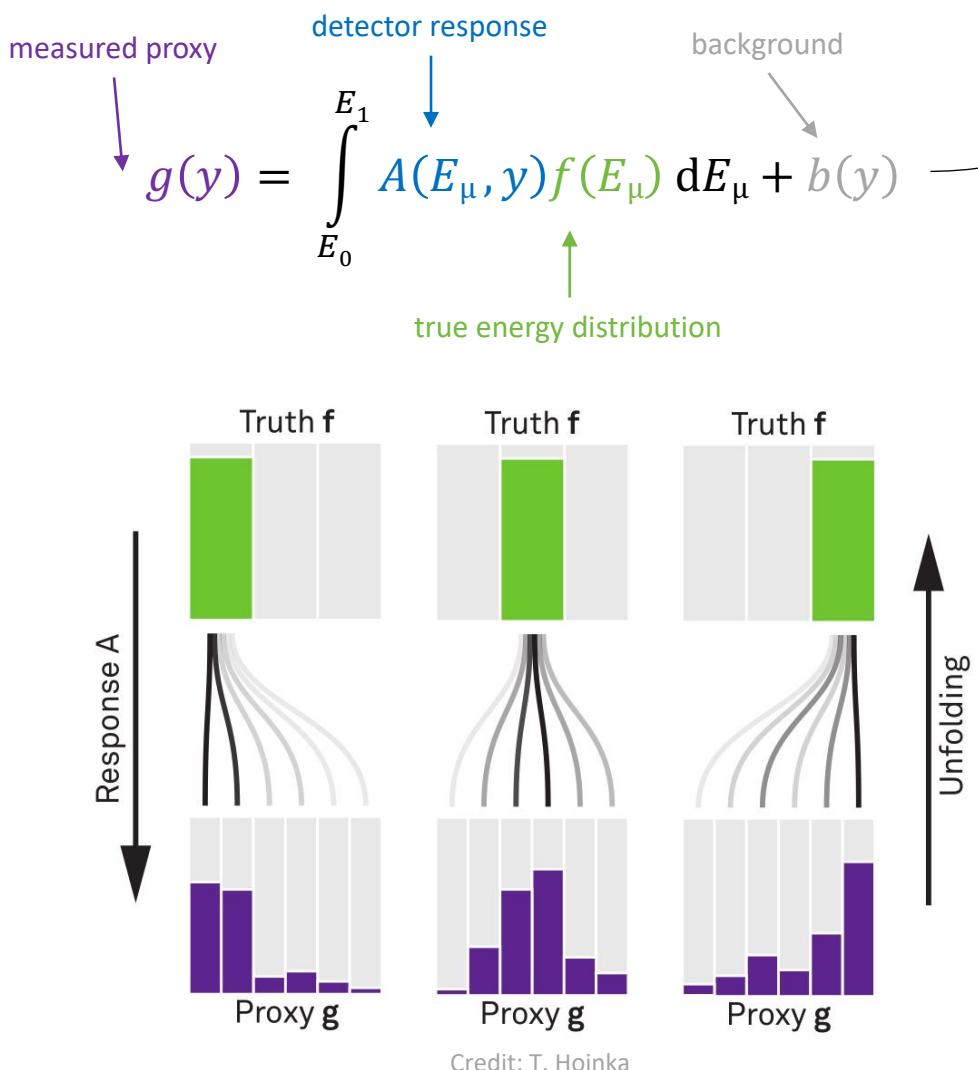


- Stopping muons:
  - Stop (decay) inside the in-ice array
  - Stopping point + direction → propagated length
  - Proxy to muon energy at the surface
  - Low-energy muons

# Technical Basics

Unfolding and Machine Learning

# Unfolding in a nutshell



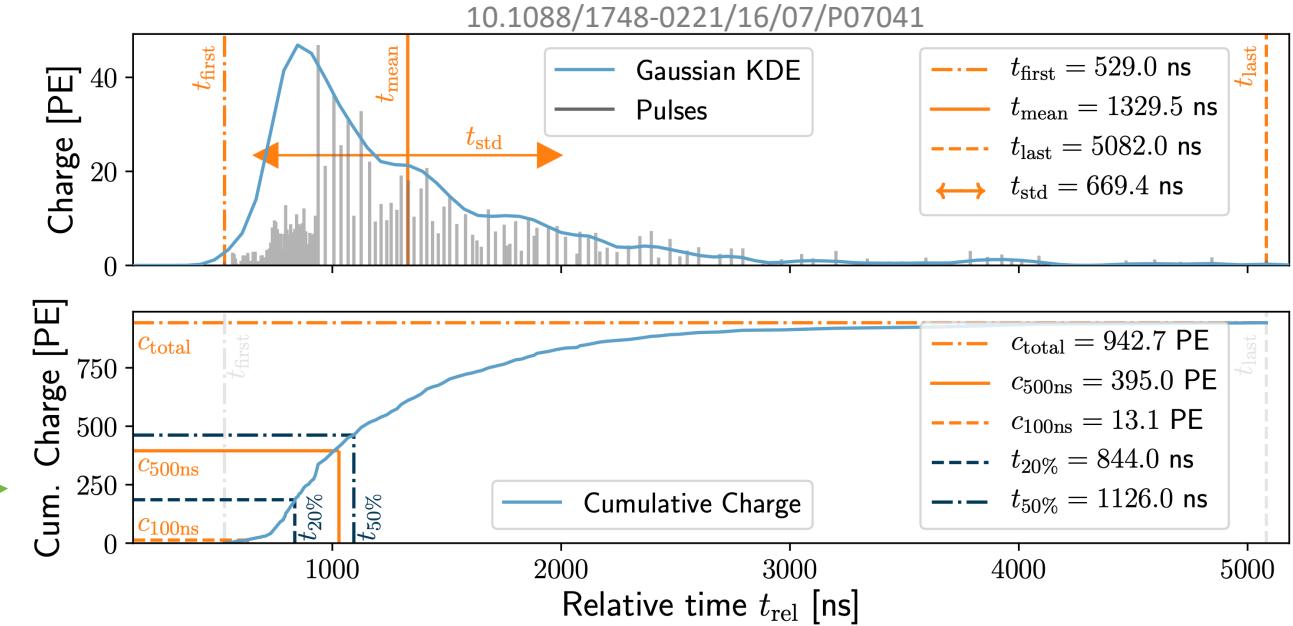
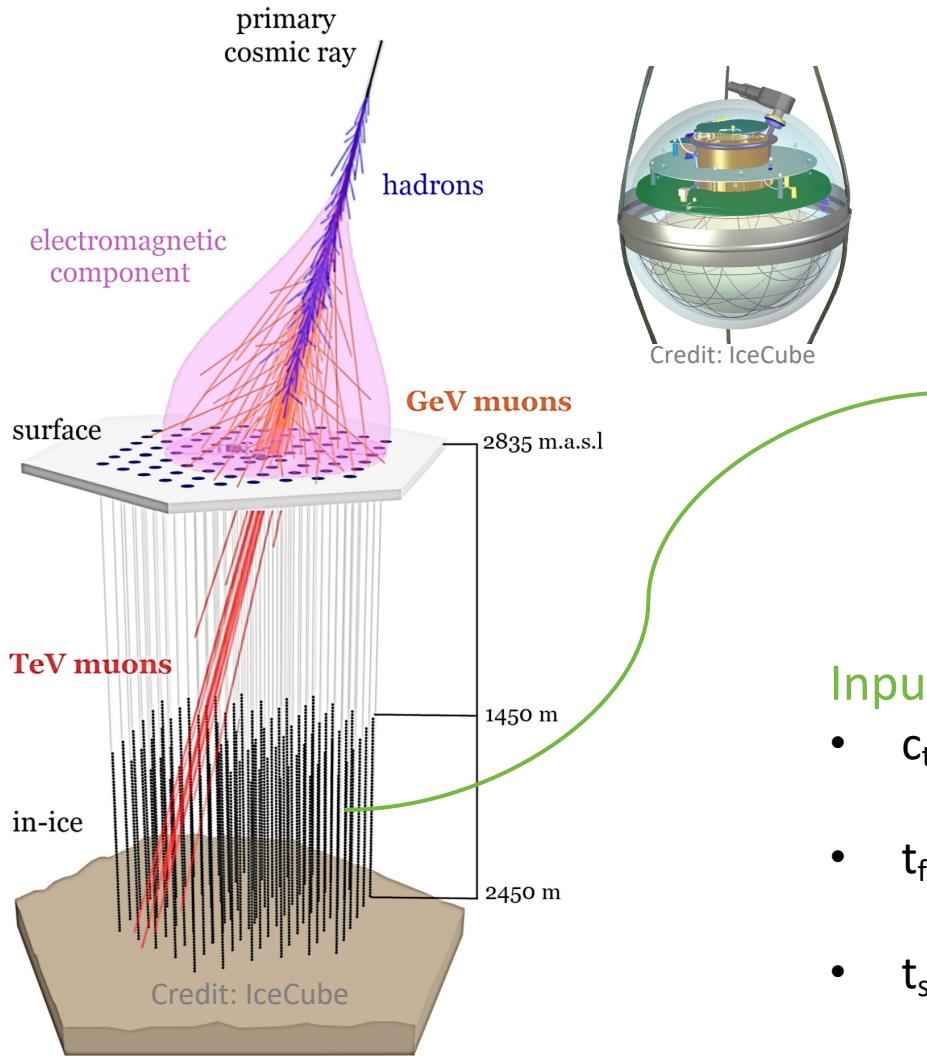
pascal.gutjahr@tu-dortmund.de

- 1. Discretized form:  $\vec{g} = A\vec{f} \leftrightarrow \vec{f} = A^{-1}\vec{g}$
  - 2. Maximum likelihood method:
  - 3. Tikhonov regularization:
  - 4. Maximize  $\log(\mathcal{L}(\vec{g}|\vec{f})) + t(\vec{f})$   
with respect to  $\vec{f}$  using  
Markov Chain Monte Carlo (MCMC)  
or Minuit
- folding                                      unfolding
- $$\mathcal{L}(\vec{g}|\vec{f}) = \prod_{j=1}^M \frac{\lambda_j^{g_j}}{g_j!} \exp(-\lambda_j)$$
- $$= \prod_{j=1}^M \frac{(A\vec{f})_j^{g_j}}{g_j!} \exp(-(A\vec{f})_j)$$
- $$t(\vec{f}) = -\frac{1}{2} (\vec{C}\vec{f})^T (\tau_1)^{-1} (\vec{C}\vec{f})$$



**funfolding**  
by M. Börner

# Machine Learning Reconstructions



## Convolutional Neural Network

### Inputs

- $c_{\text{total}}$ : Total charge
  - Sum of charge
- $t_{\text{first}}$ : Relative time of first pulse
  - Relative to total time offset
- $t_{\text{std}}$ : Standard deviation of first pulse
  - Charge weighted standard deviation of pulse times

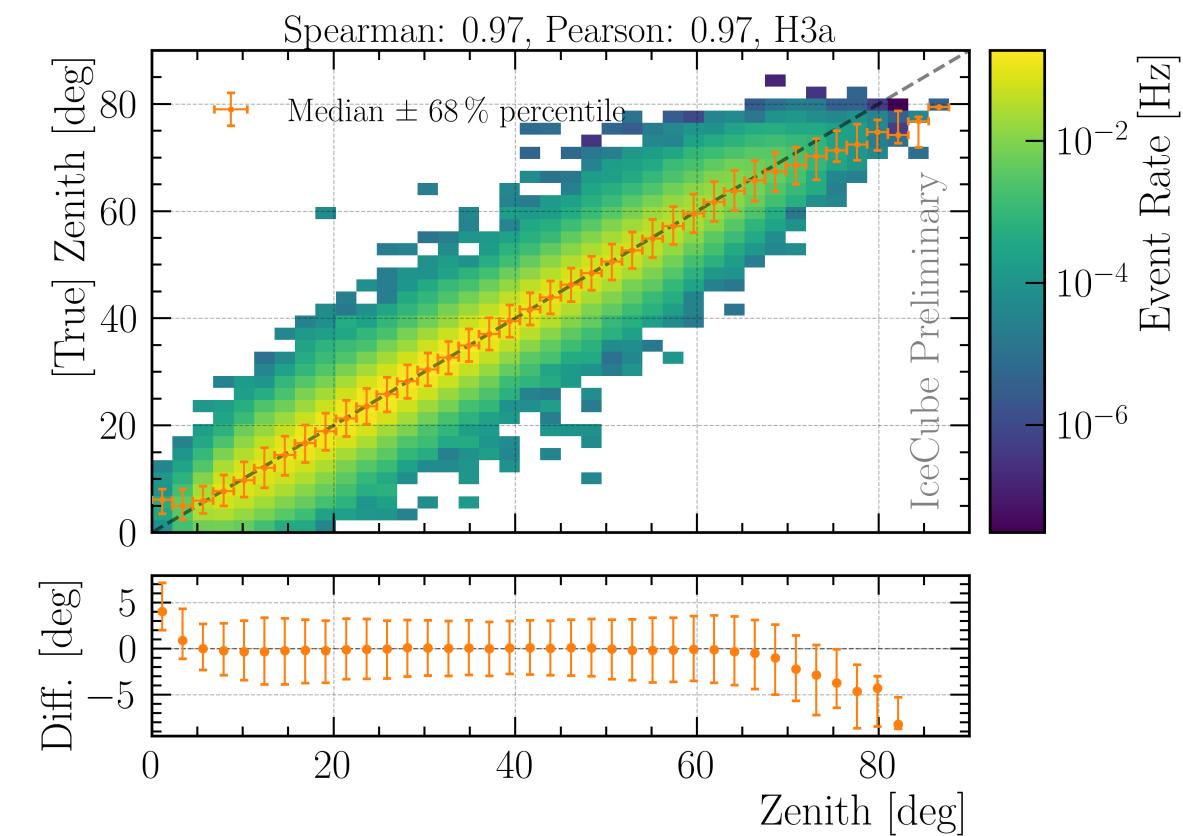
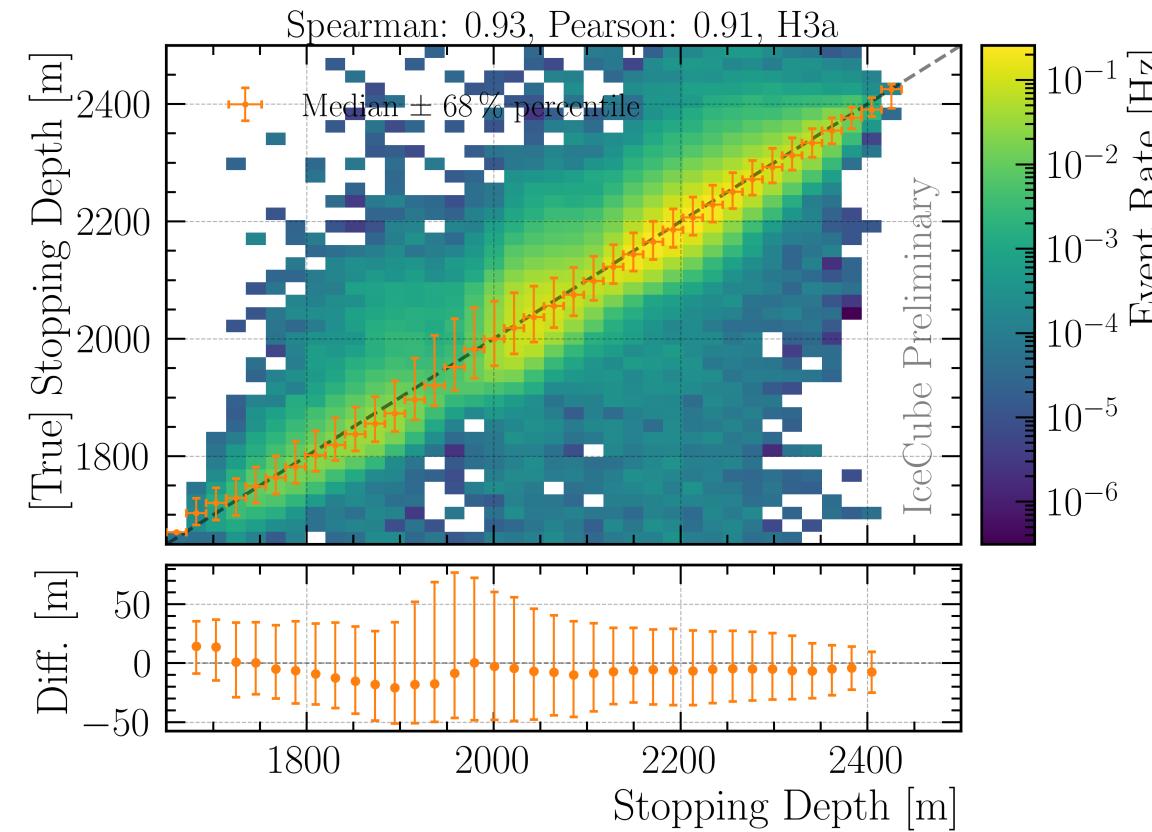
### Outputs

- Direction
- Stopping point
- Entry point
- Energy at entry/surface
- ...

# Depth Intensity Unfolding

# Reconstructions: Stopping Muons

- Good reconstruction of stopping depth and zenith angle
- Calculate propagation length → proxy variable for unfolding

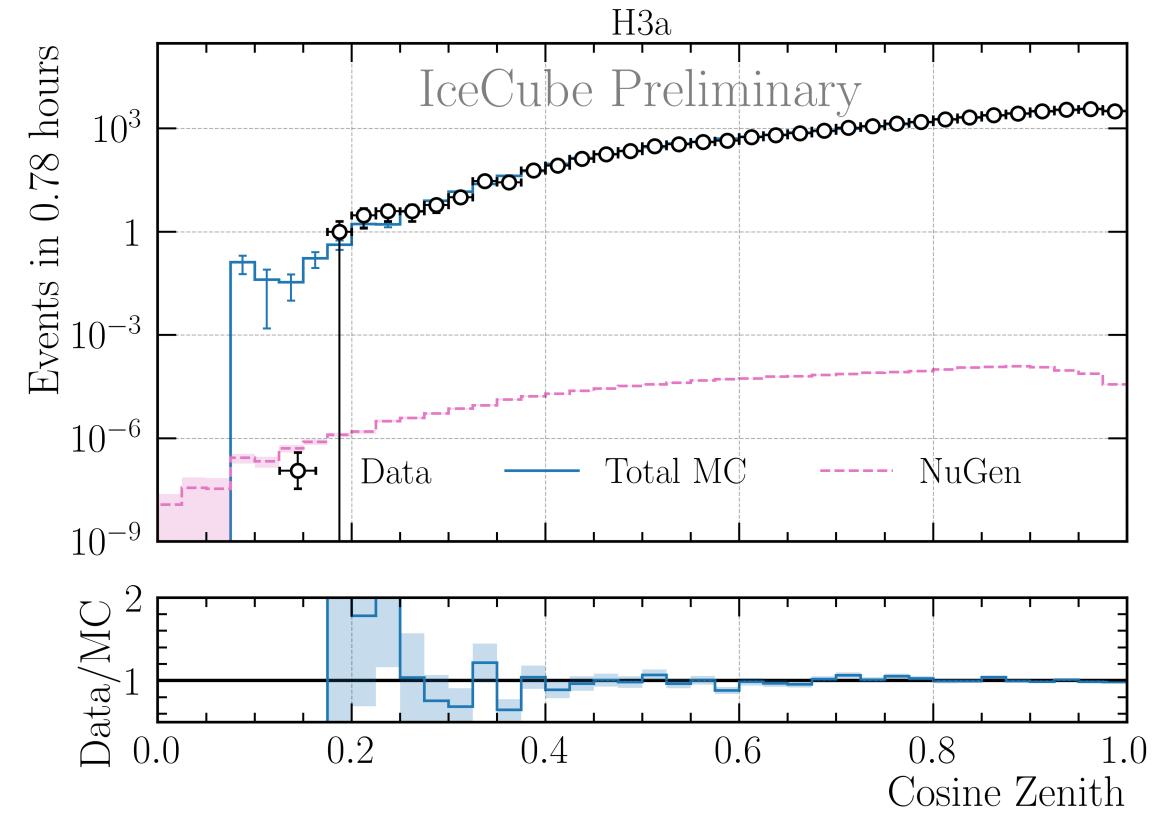
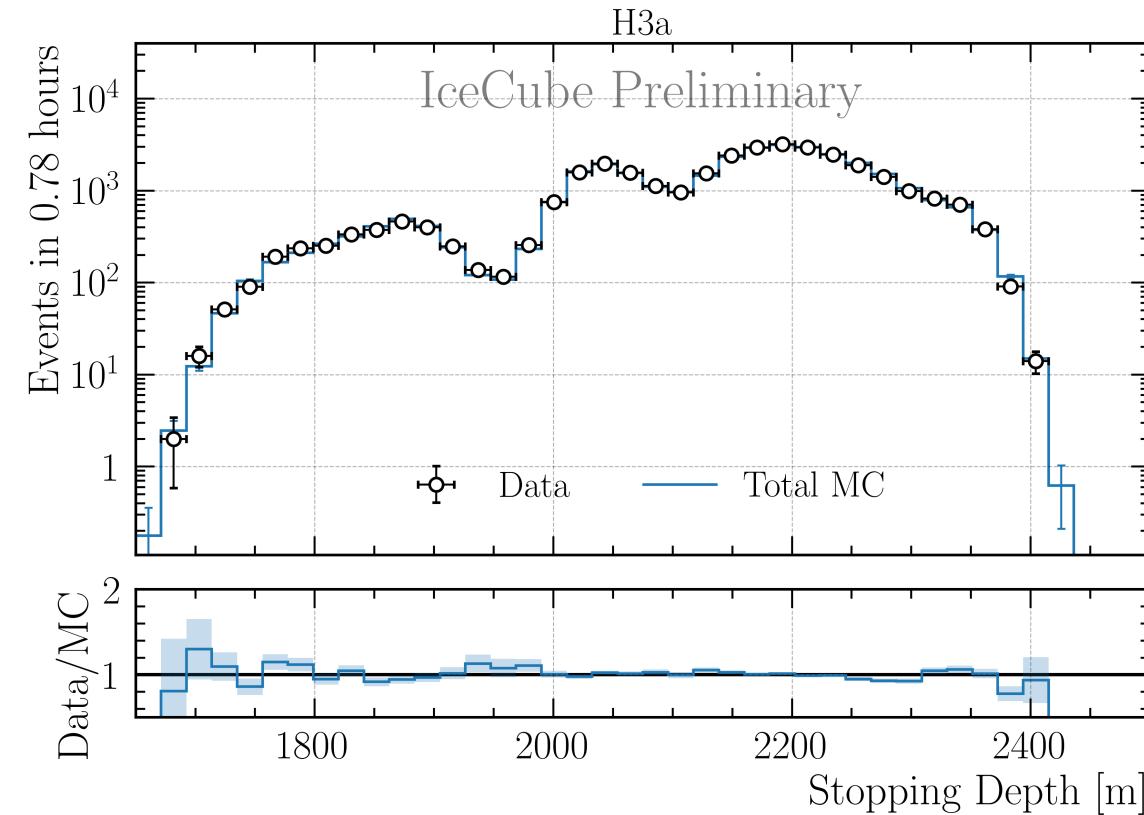


# Data—MC: Stopping Muons

- Good data—MC agreement
- Global offset → upscale MC by 10%

Neutrino weighting:  
SPL:  $n = 1.8, \gamma = 2.52$

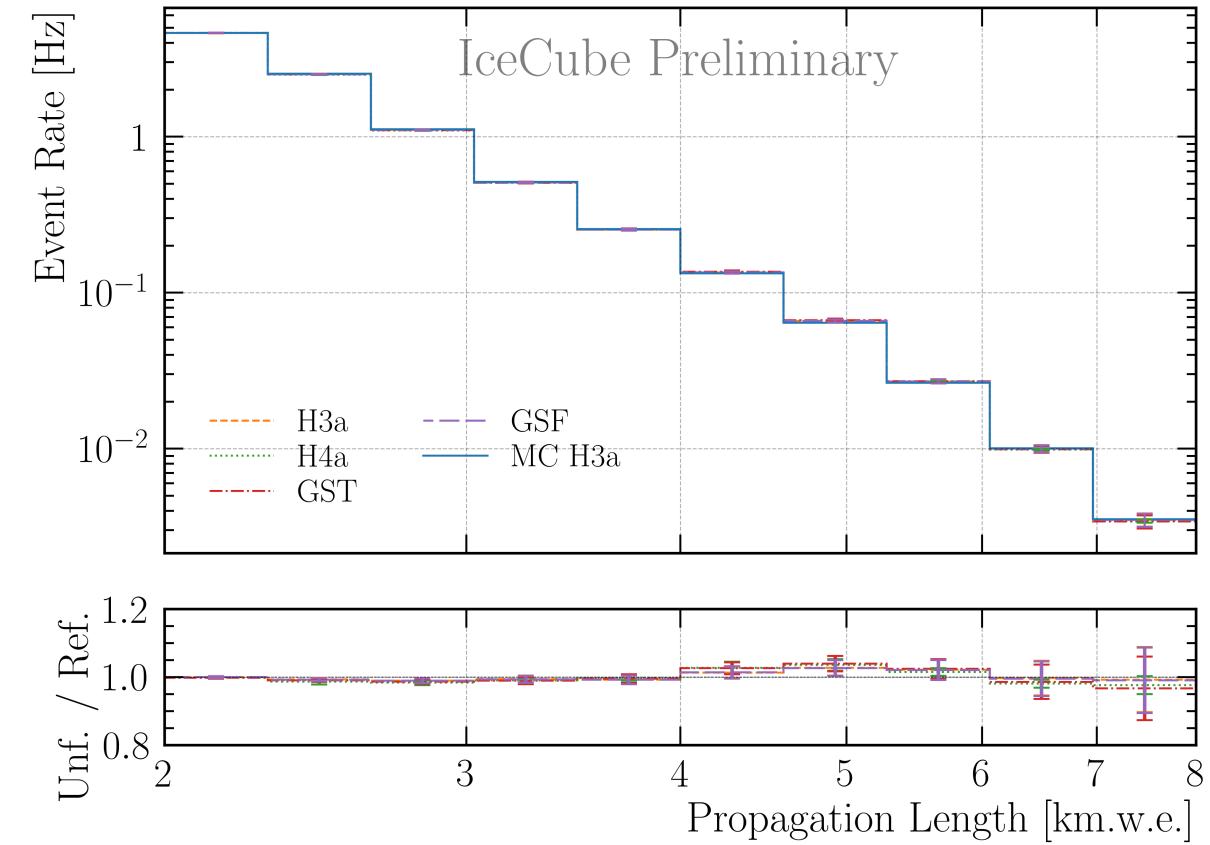
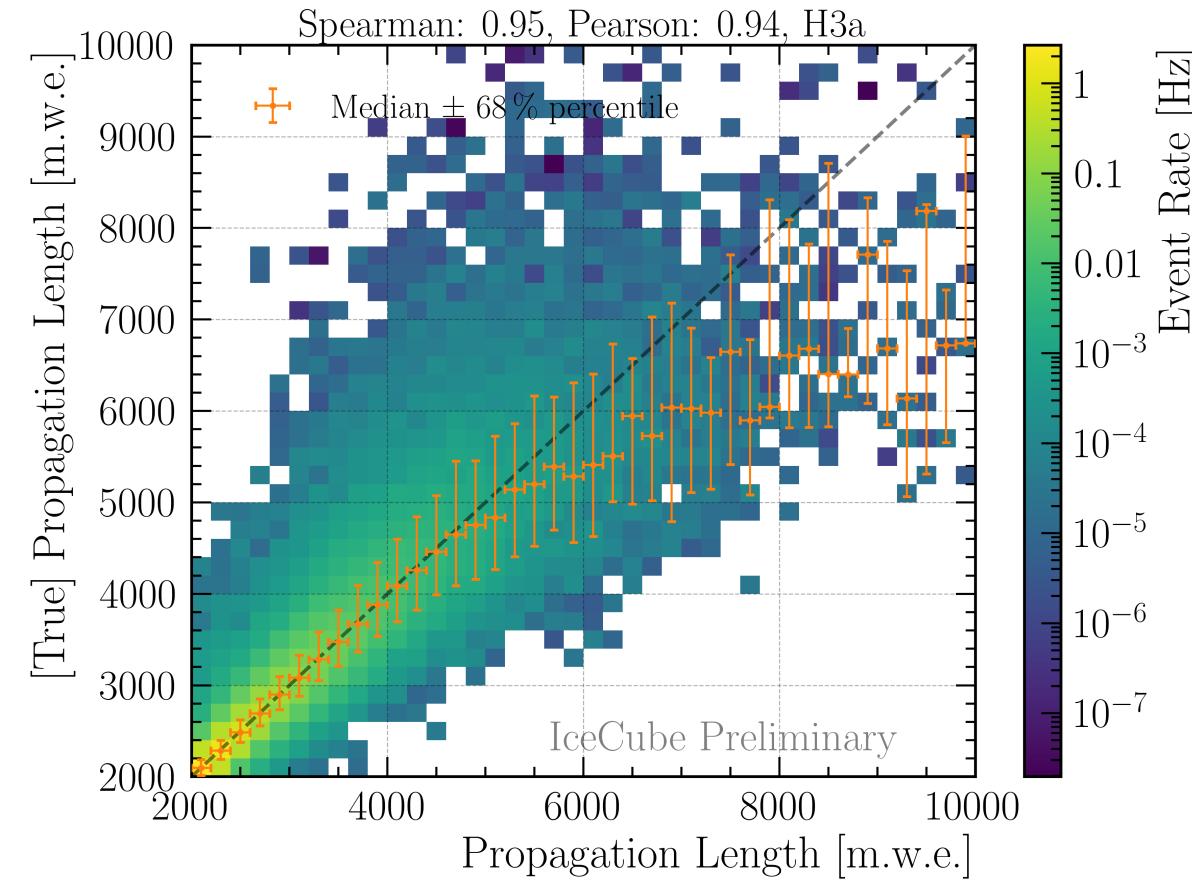
IceCube Collaboration PoS ICRC2023 1064



# Correlation and Robustness Test: Stopping Muons

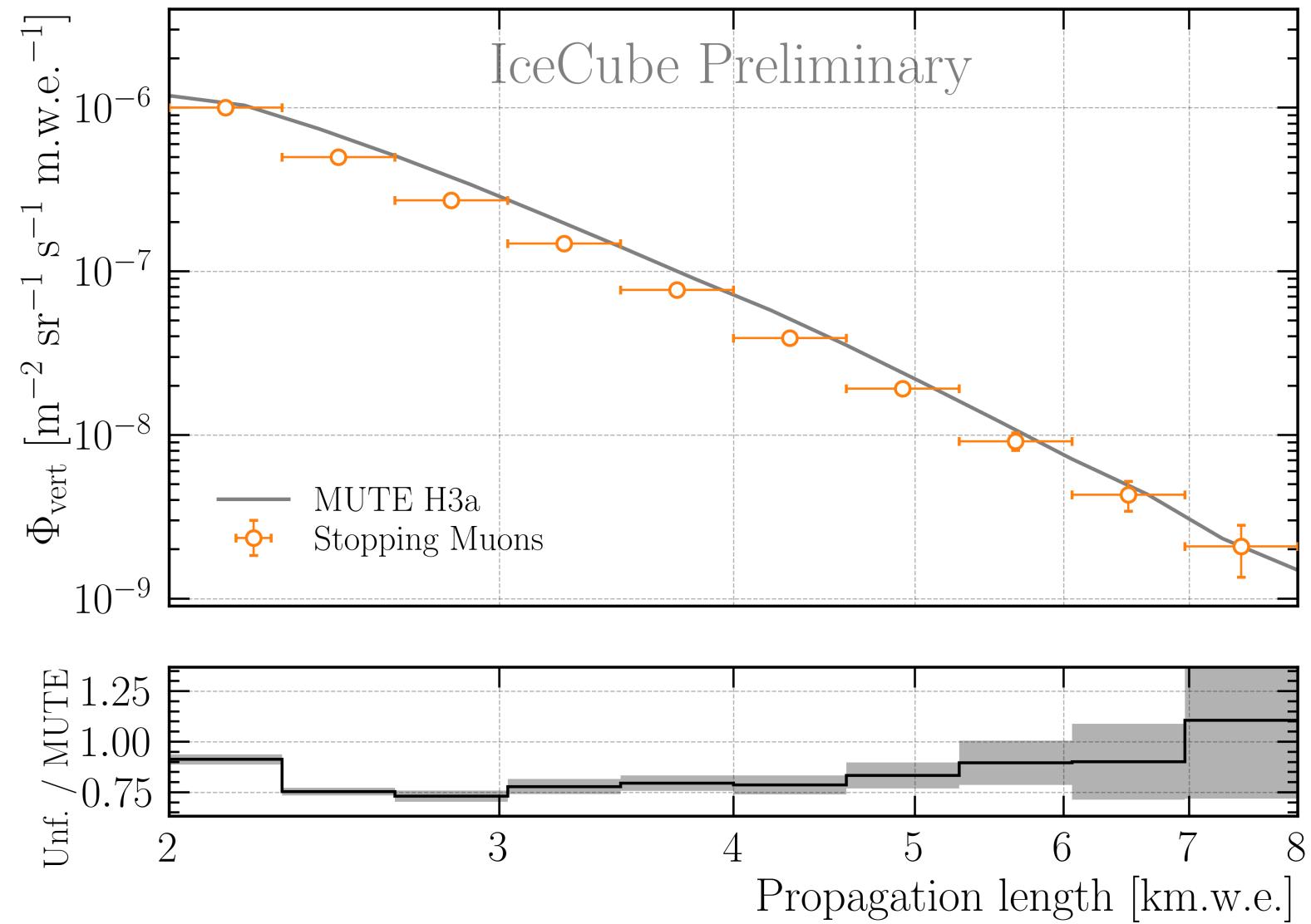
➤ Strong correlation between proxy and target

- Build unfolding matrix on H3a, H4a, GST & GSF
- Unfold H3a as test set
- Results are within uncertainties → robust



# Depth Intensity Unfolding

- 47 min IceCube data
- 32943 events
- Comparison to MUTE<sup>1</sup>
- Up to 25 % below prediction at 2.5 – 5.5 km.w.e.
- Mismodelling in energy losses?

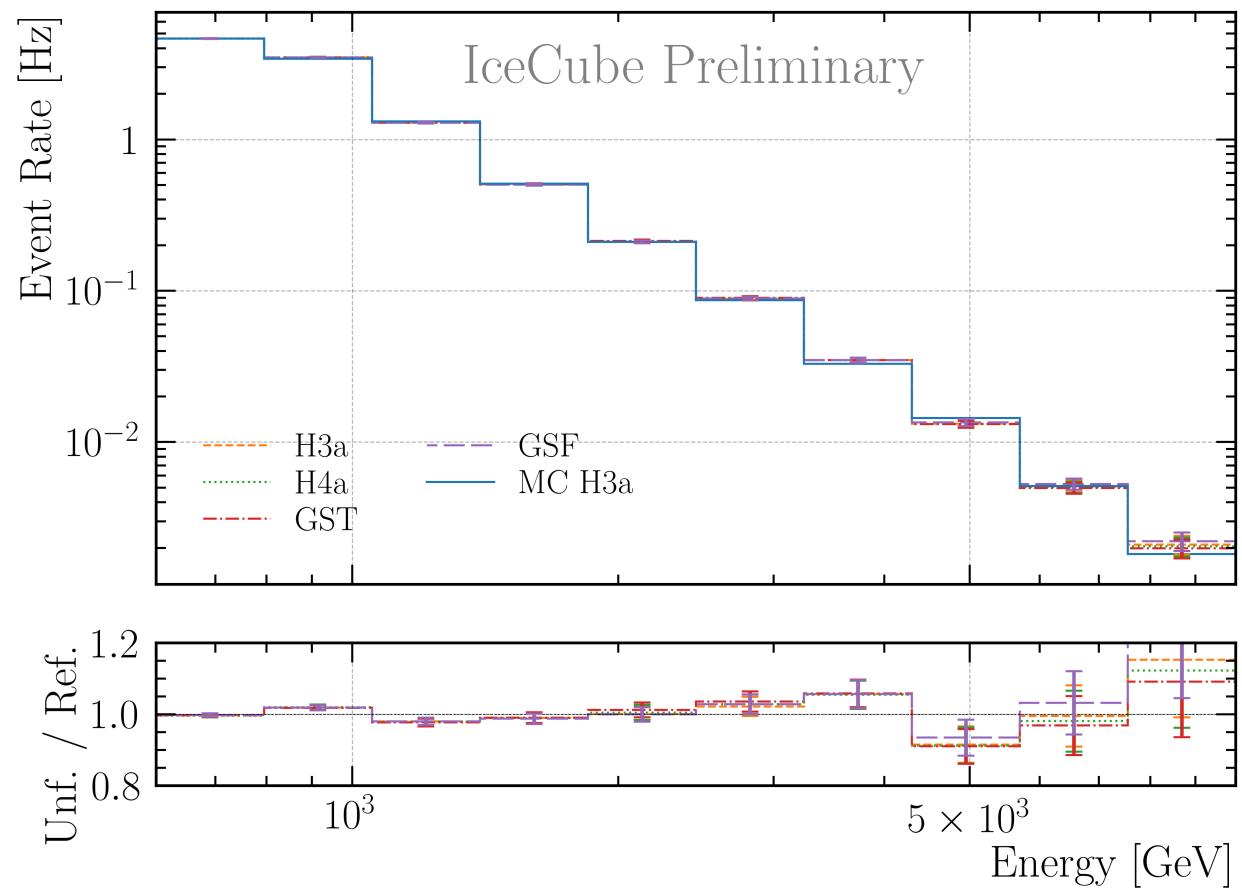
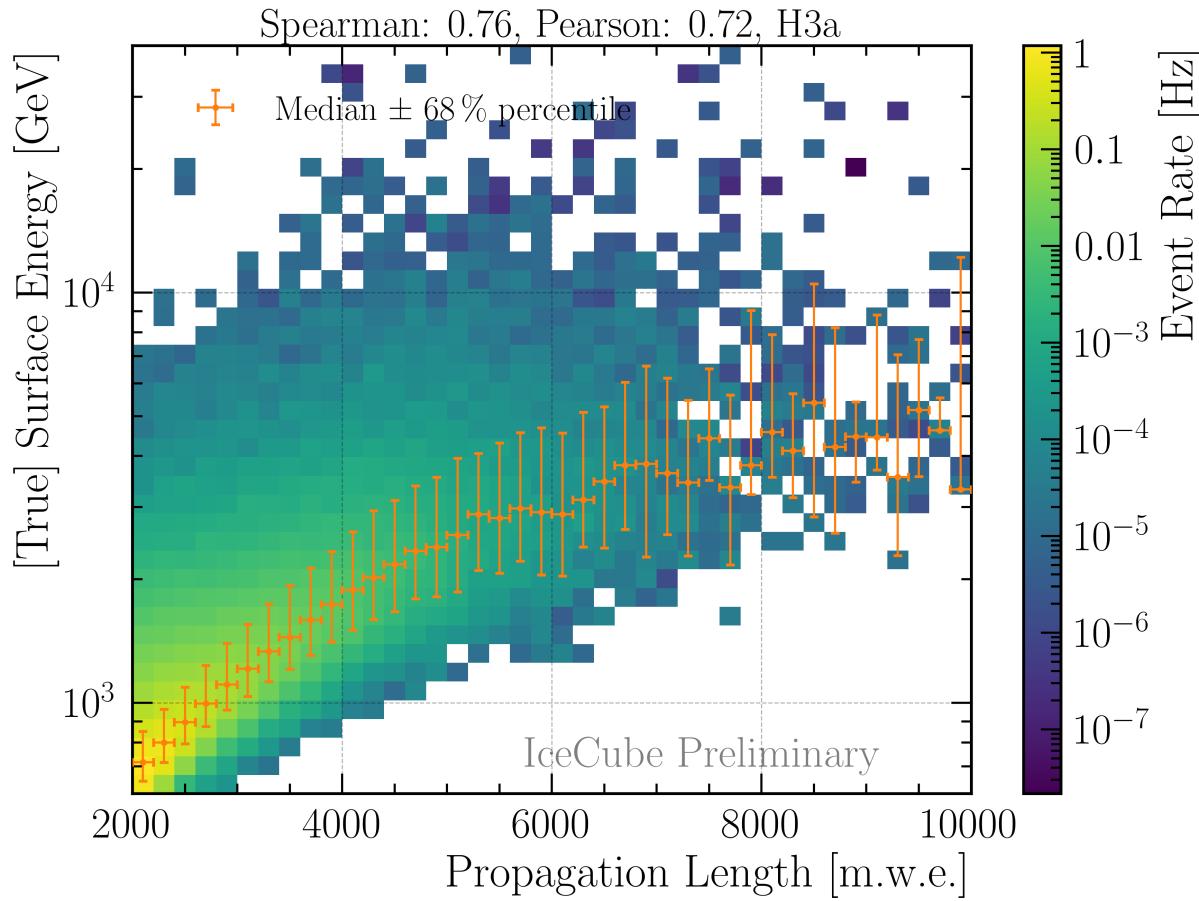


# Muon Flux Unfolding

# Correlation and Robustness Test: Stopping Muons

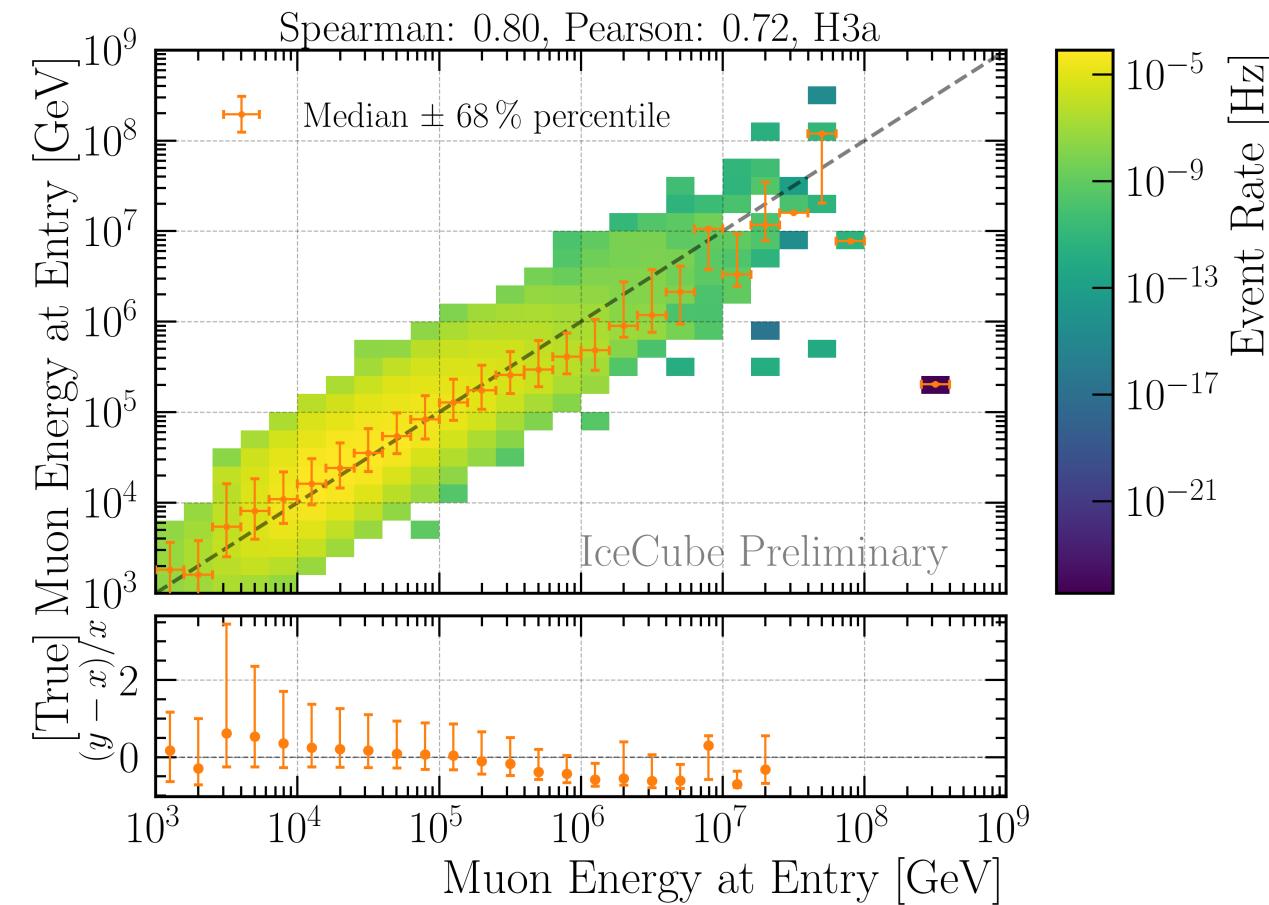
- Correlation between proxy and target

- Build unfolding matrix on H3a, H4a, GST & GSF
- Unfold H3a as test set
- Results are within uncertainties → robust

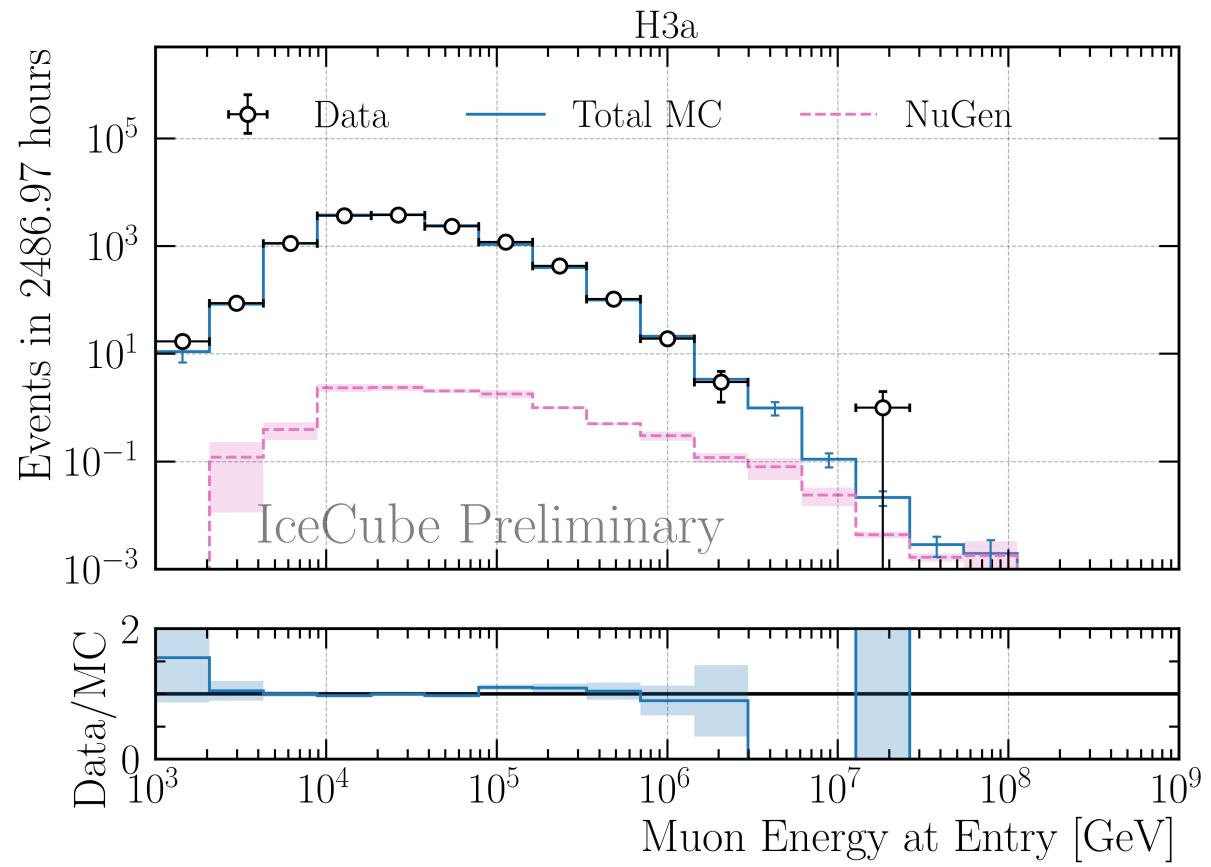


# Reconstruction and Data—MC: Leading Muons

- Good reconstruction of leading muon energy → proxy

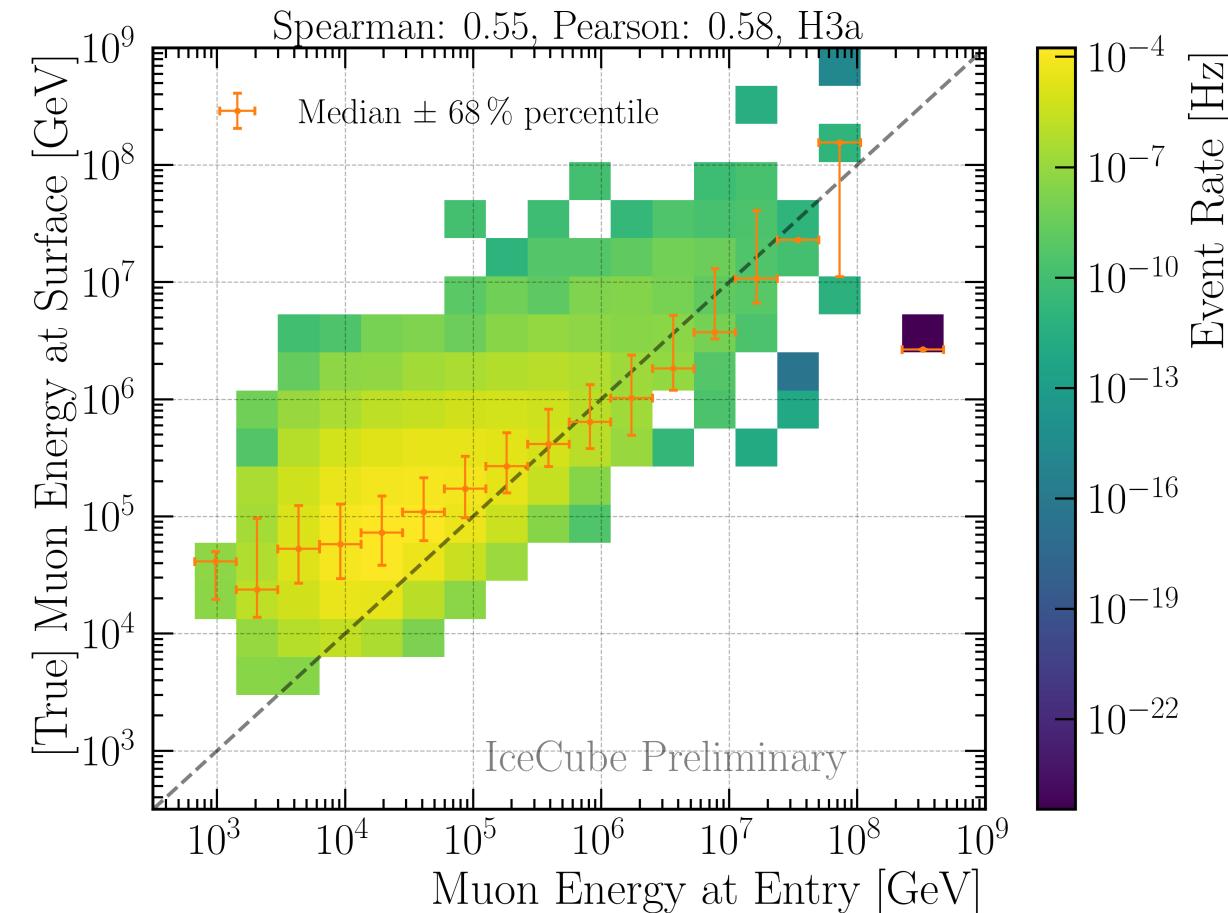


- Good data—MC agreement
- Global offset → upscale MC by 12%

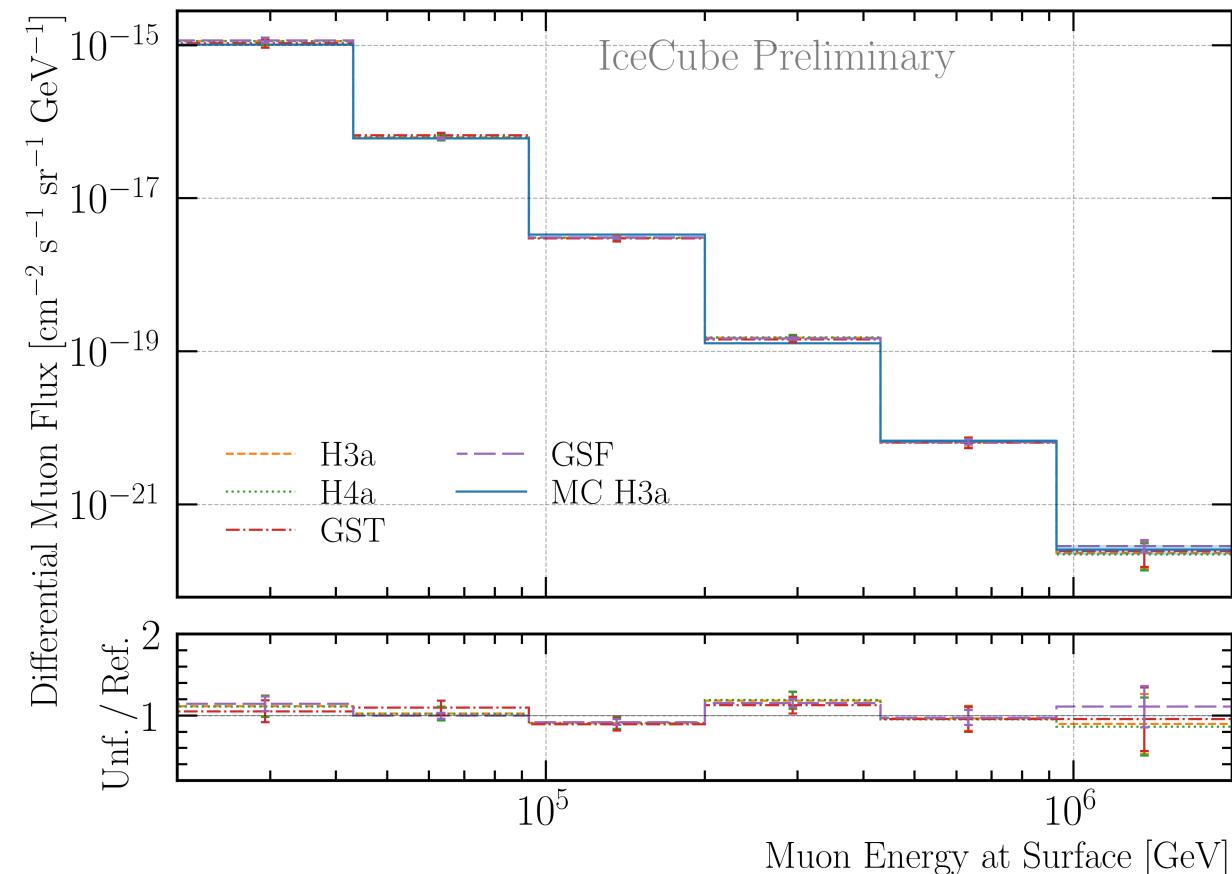


# Correlation and Robustness Test: Leading Muons

- Correlation between proxy and target



- Build unfolding matrix on H3a, H4a, GST & GSF
- Unfold H3a as test set
- Results are within uncertainties → robust



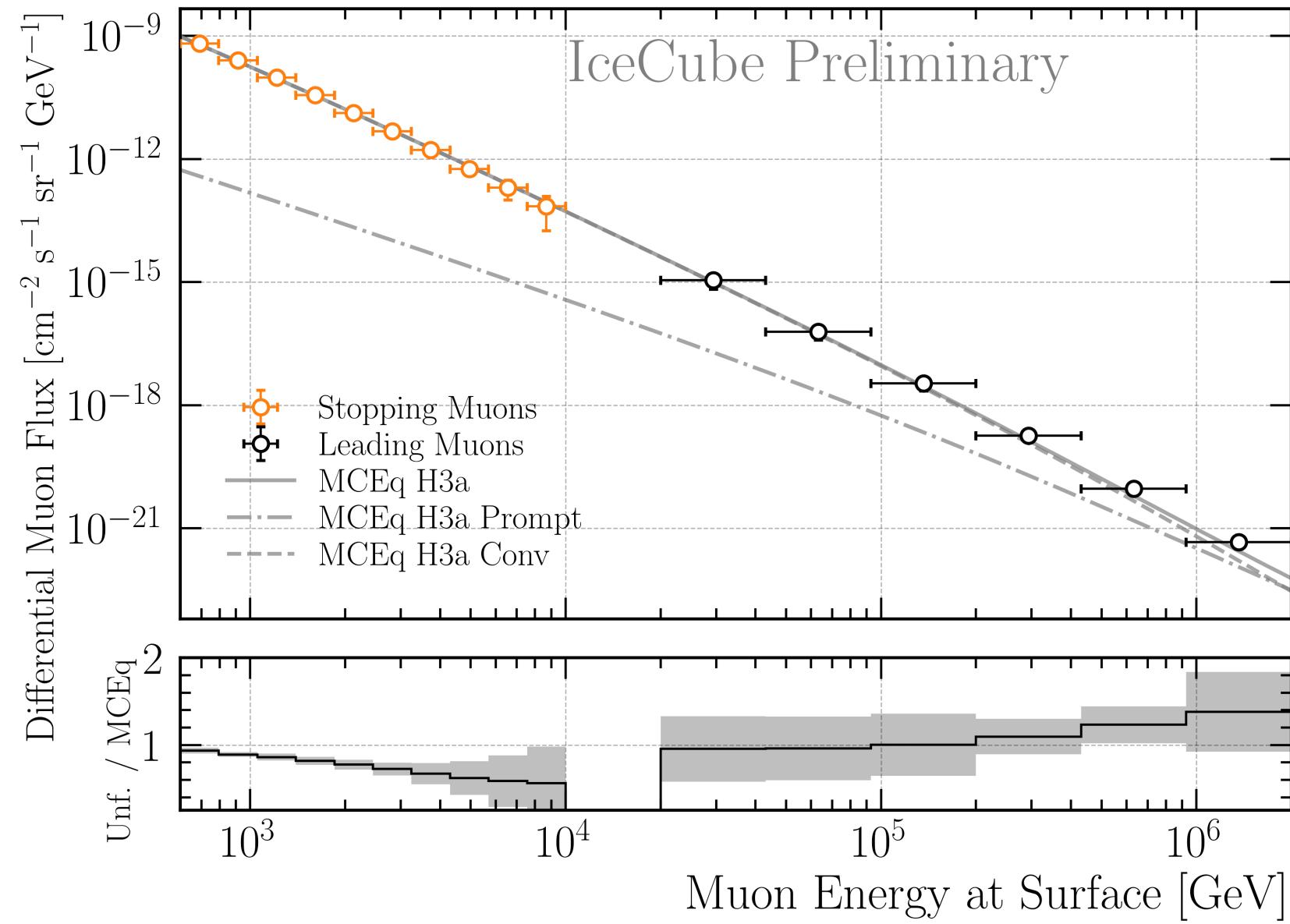
# Muon Flux Unfolding

## Stopping muons

- 47 min IceCube data
- 32943 events
- Below MCEq prediction

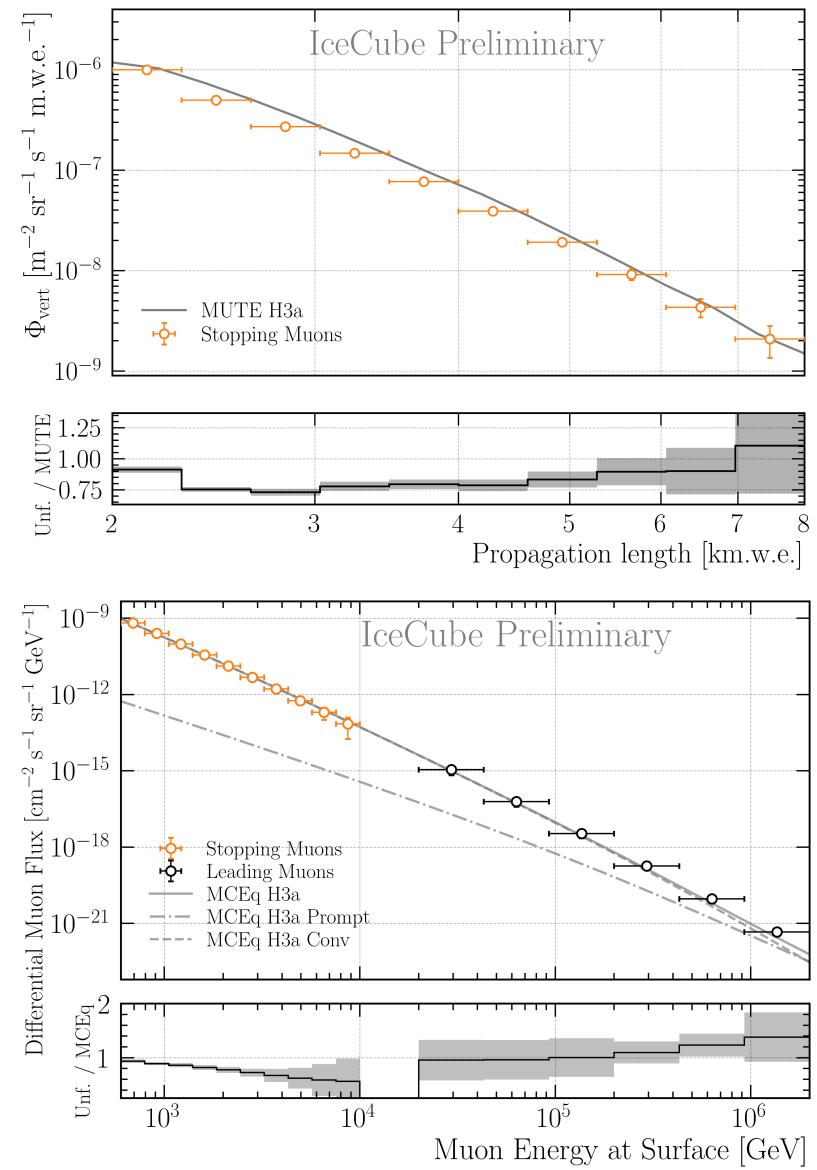
## Leading muons

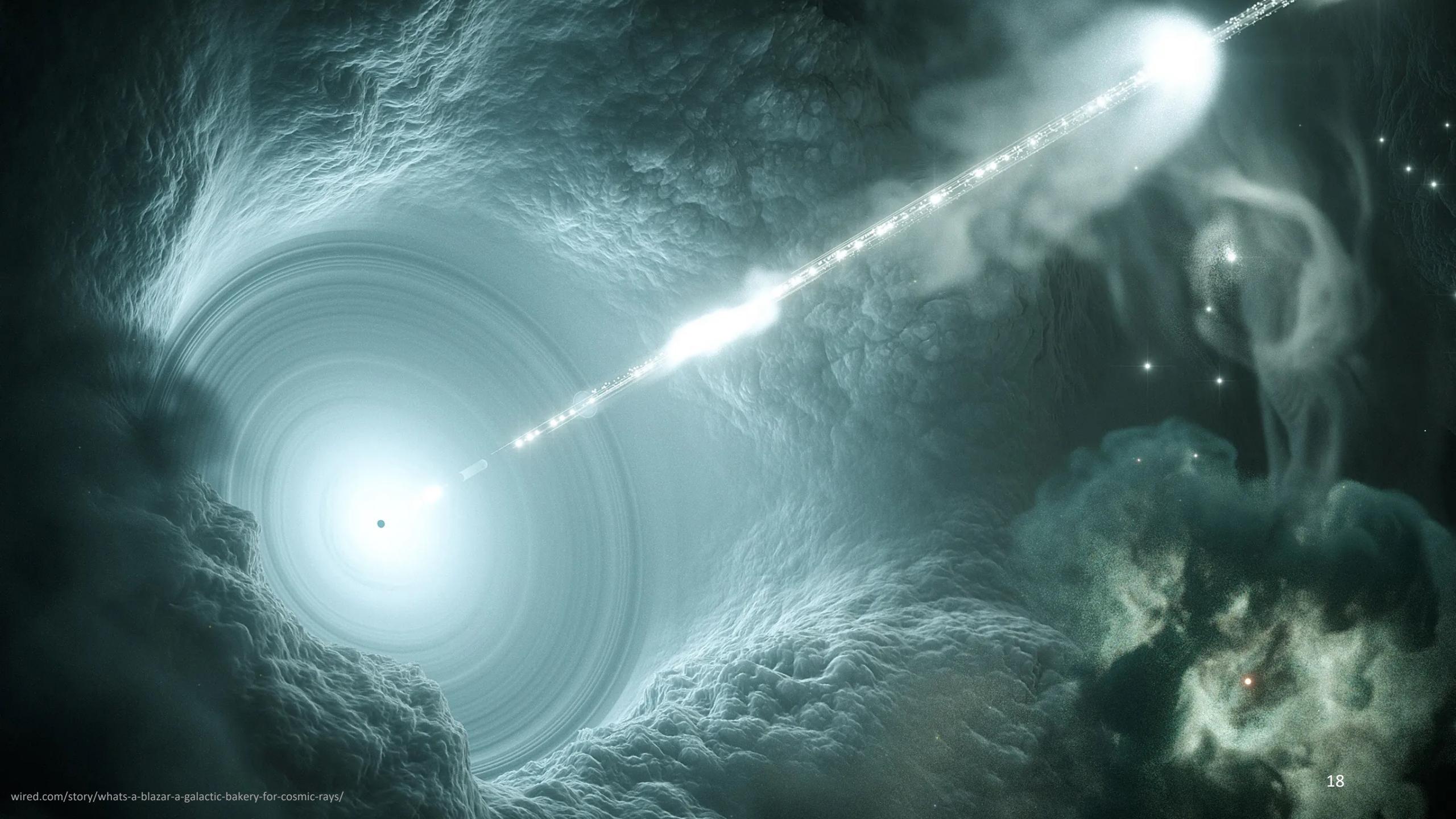
- 2487 h IceCube data
- 12754 events
- Agrees with MCEq



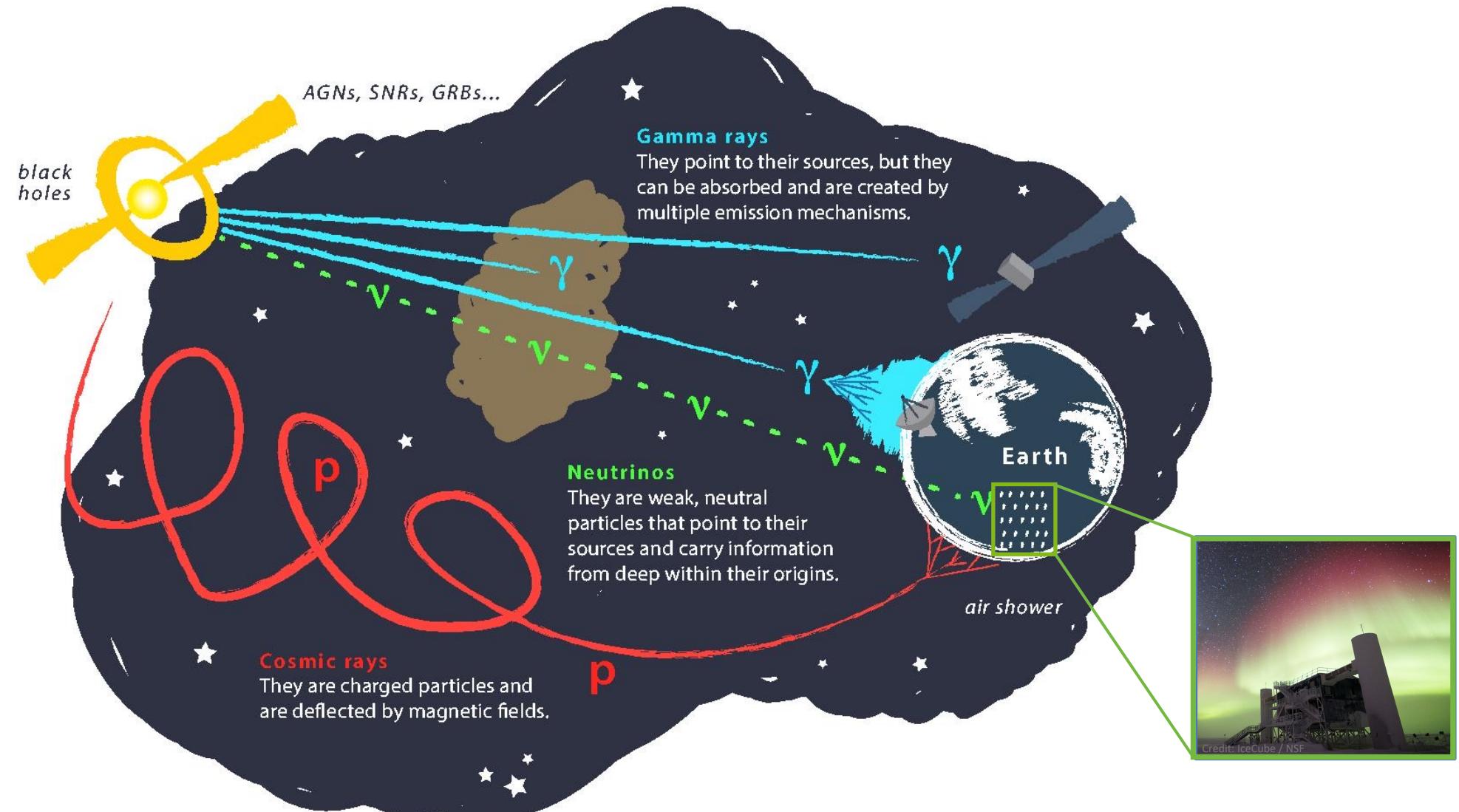
# Conclusion & Outlook

- DNN-based reconstructions performed
  - Stopping muon selection
  - Leading muon selection
  - Good Data—MC agreement
- Depth intensity unfolded using 47 min of IceCube data
  - Undershoots MUTE prediction up to 5.5 km.w.e.
- Muon flux at surface unfolded using stopping muons and 2487 h of leading muon IceCube data
  - Flux undershoots up to 10 TeV (MCEq prediction) – stopping muons
  - Flux agrees from 20 TeV to 2 PeV with MCEq – leading muons
    - Slope increases towards higher energies
    - Prompt and conventional flux contributes similar to the flux at  $\sim$ 2 PeV
- Finalize both analyses → unfold with more data
  - Extend energy range up to  $\sim$ 10 PeV

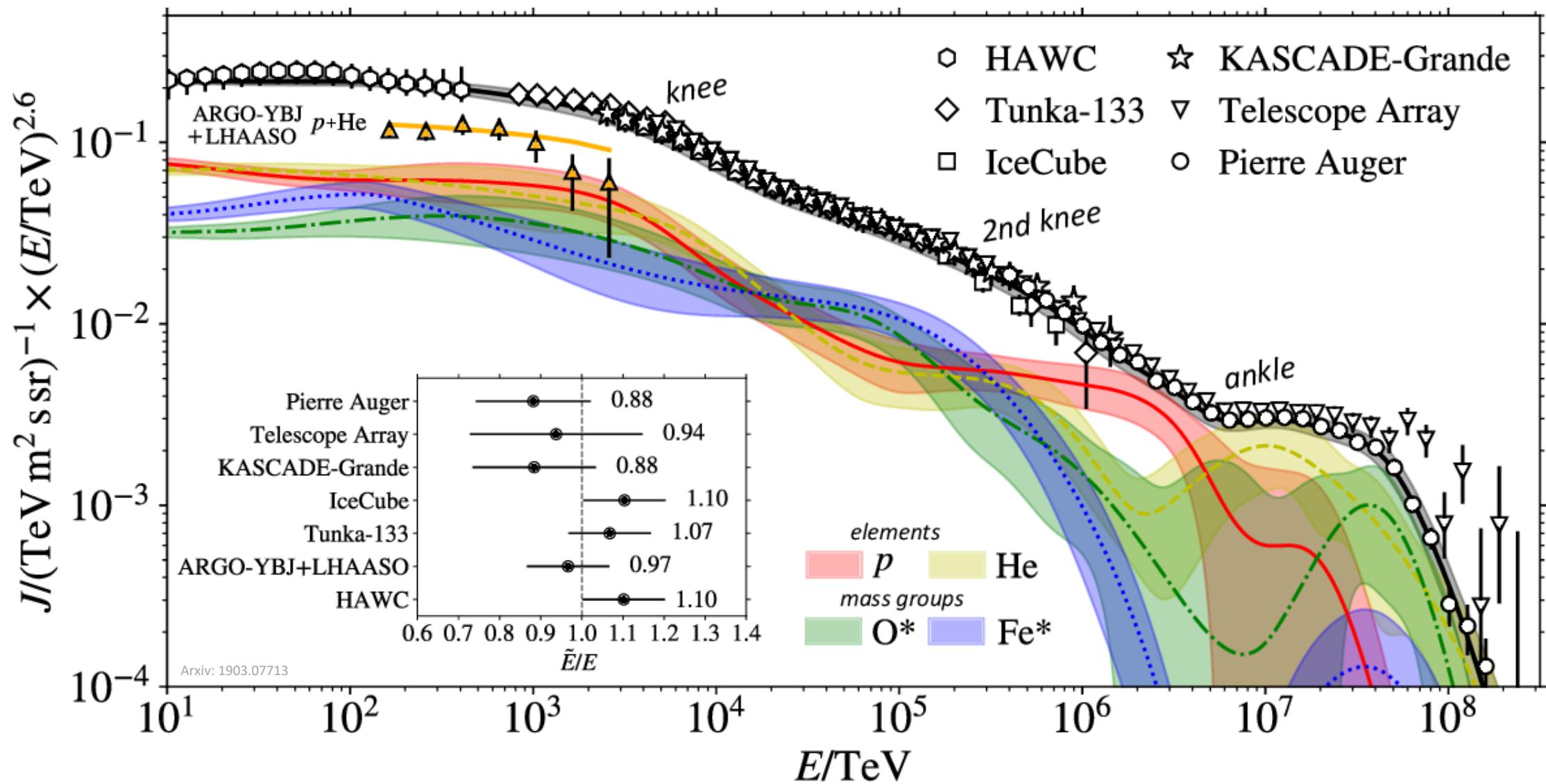




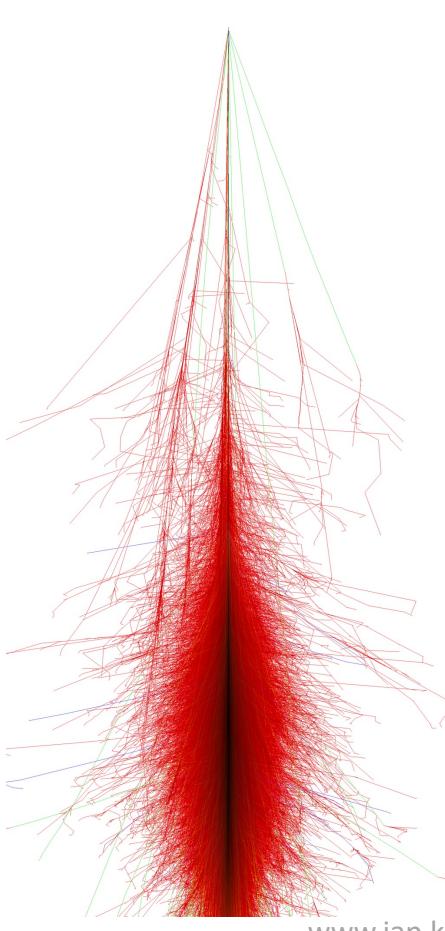
# Astroparticle physics



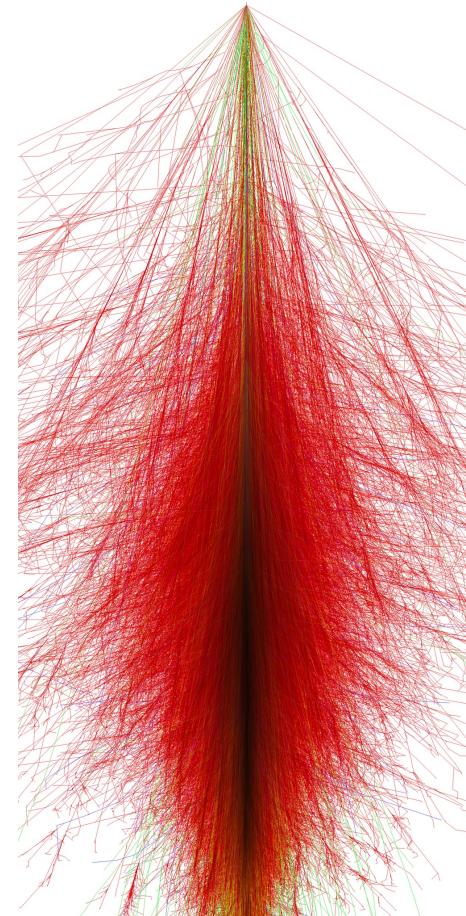
## Cosmic ray flux



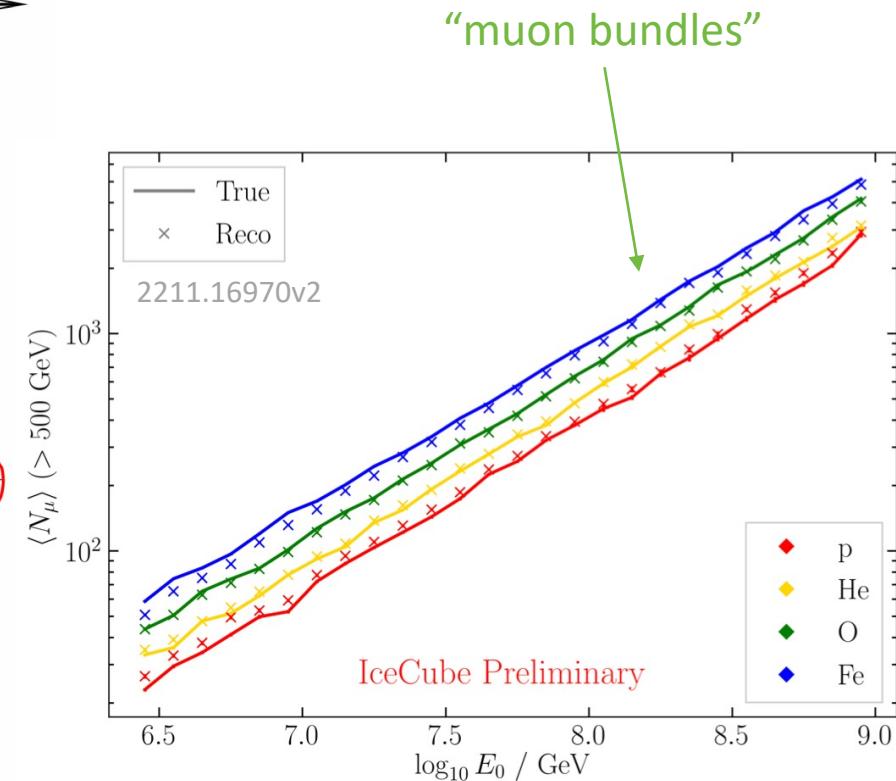
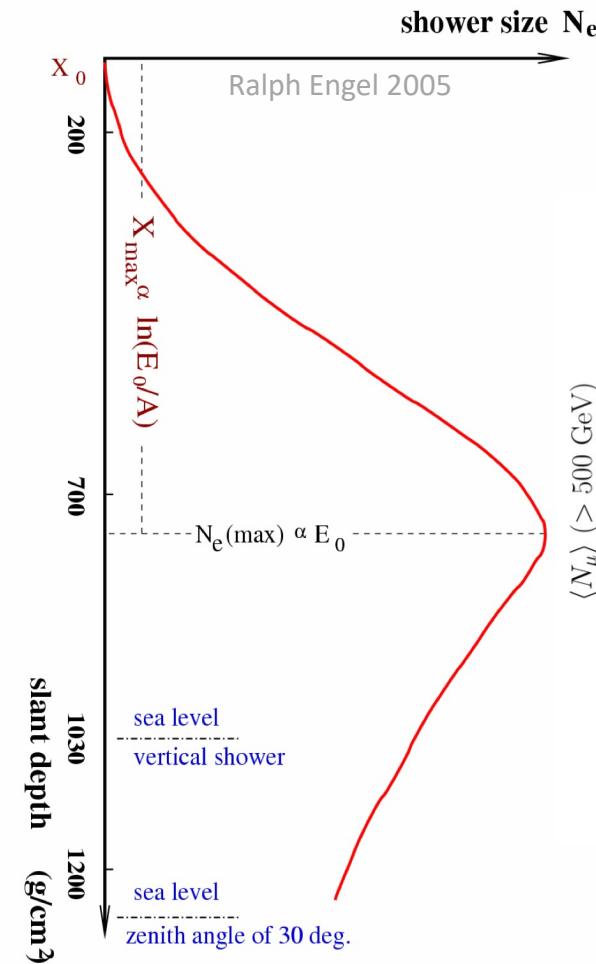
## Air shower – 10 TeV



Proton

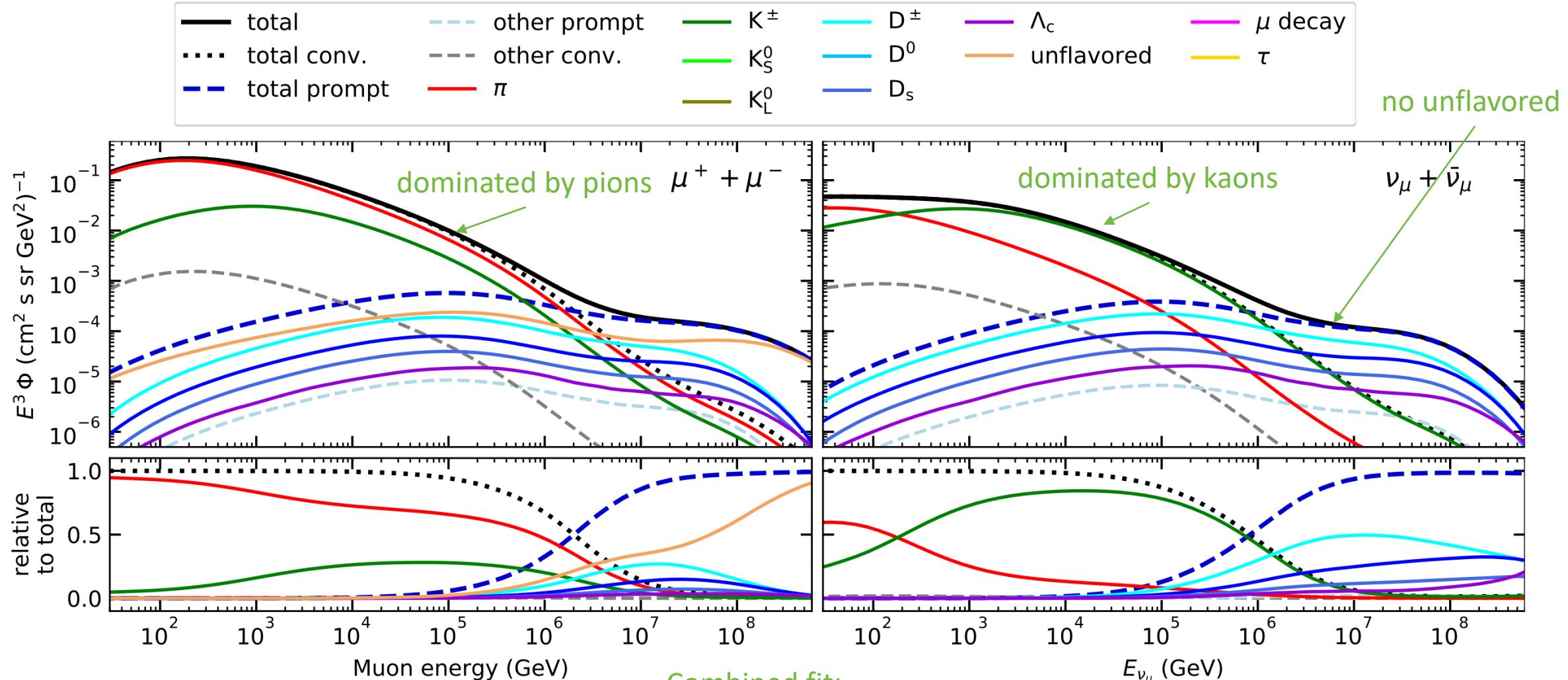


Iron



# Prompt atmospheric muons and neutrinos

10.1103/PhysRevD.100.103018



Combined fit:

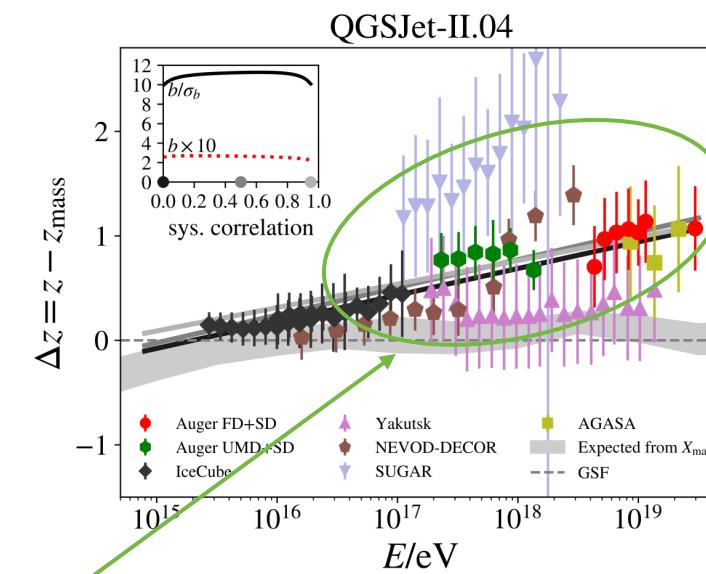
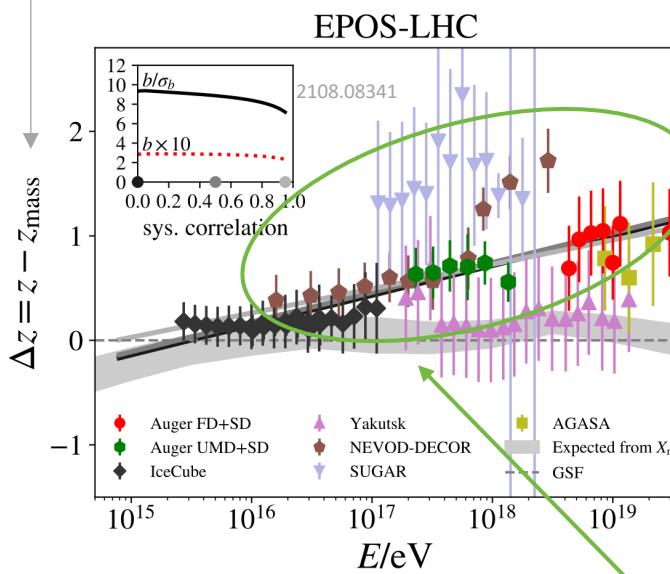
- handle on pion/kaon ratio
- handle on charmed mesons

# Muon puzzle and model uncertainties

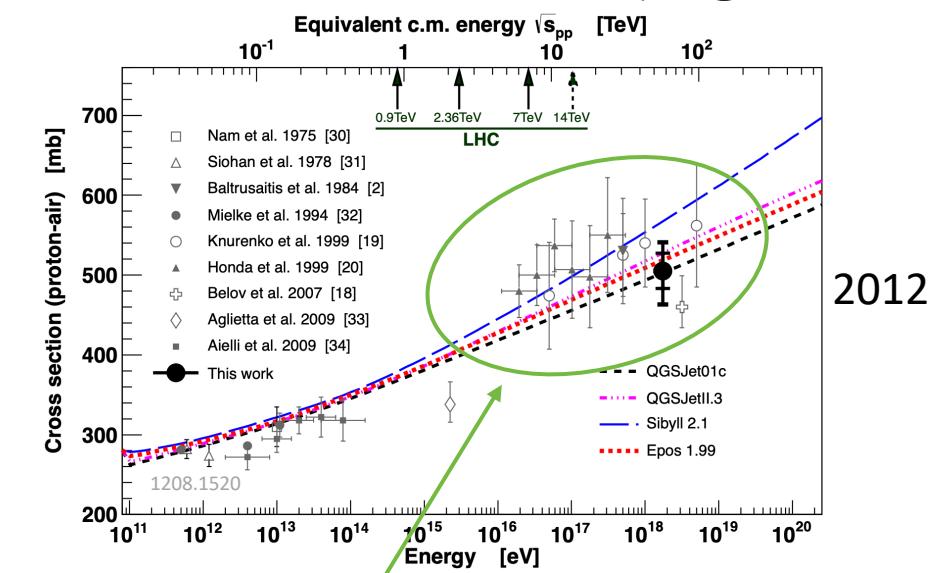
"muon number"

$$z = \frac{\ln\langle N_\mu \rangle - \ln\langle N_\mu \rangle_p}{\ln\langle N_\mu \rangle_{\text{Fe}} - \ln\langle N_\mu \rangle_p}$$

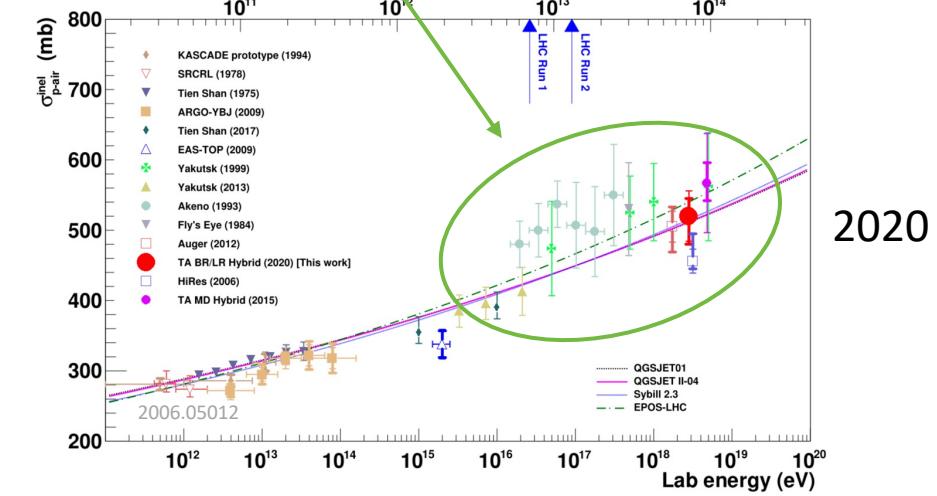
Expected  $z$   
("muon number")

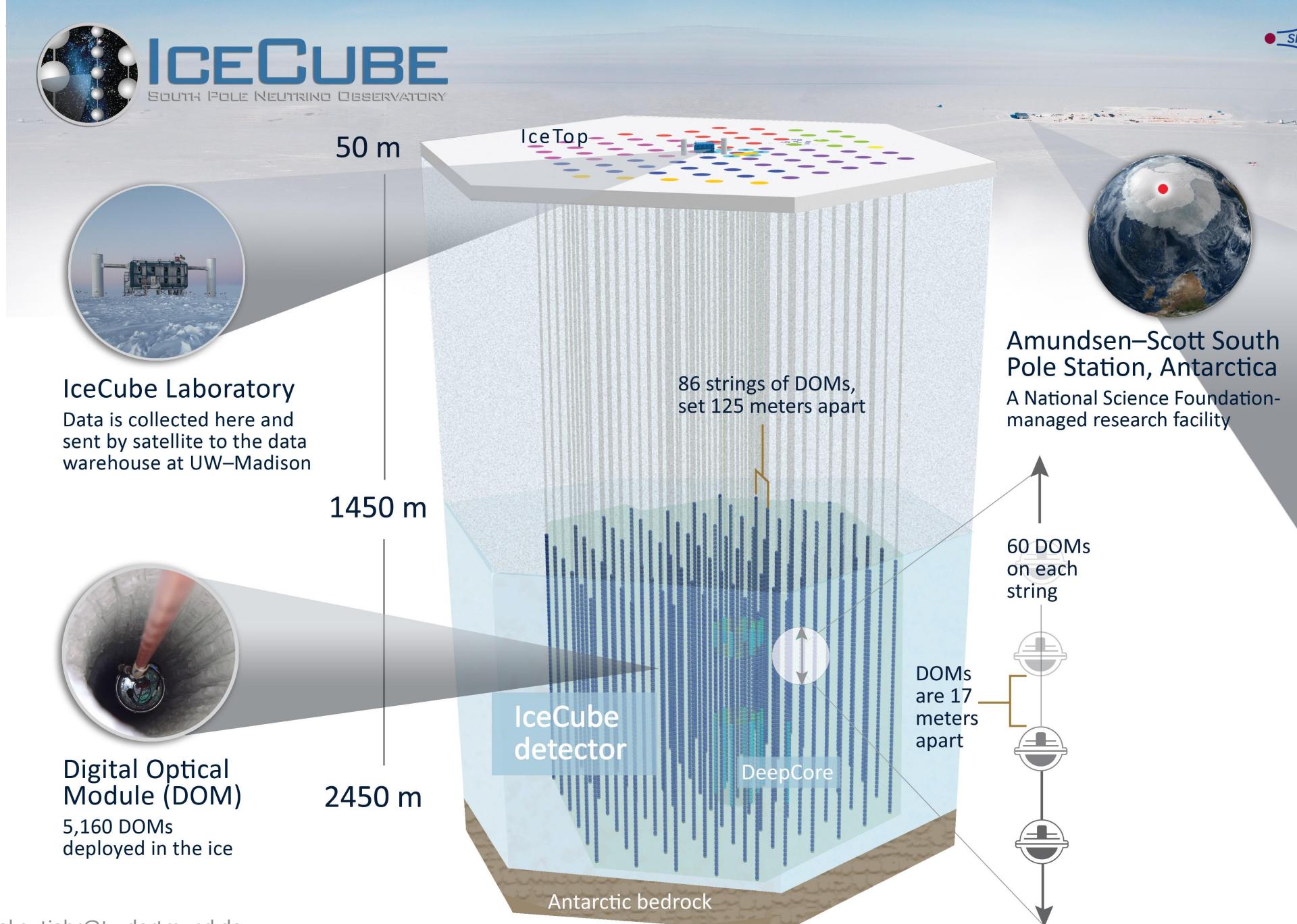


➤ More muons measured than simulated for  $E > 40 \text{ PeV} \sim \text{ cms } 8 \text{ TeV}$



➤ Uncertainties at  $E > 10 \text{ PeV}$

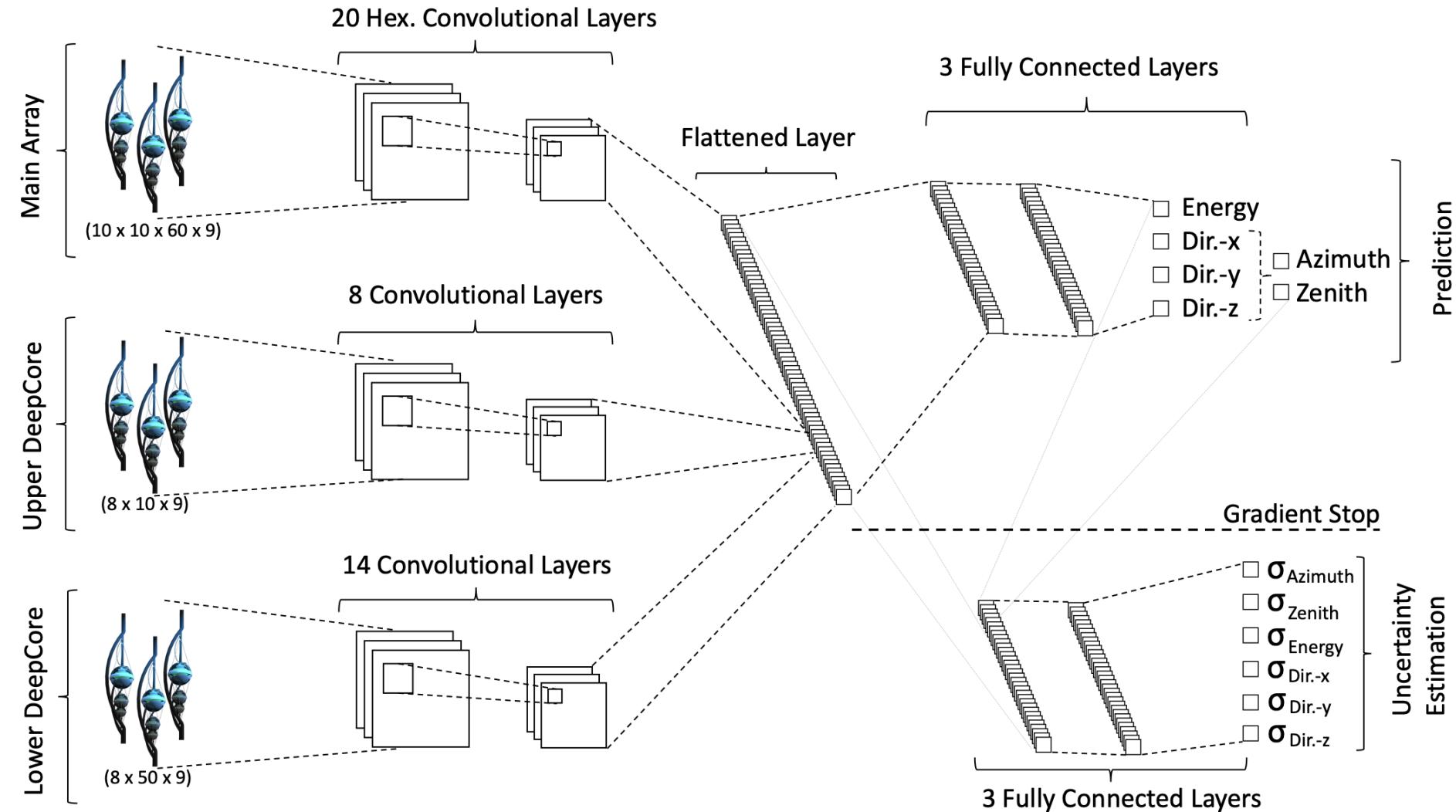




# Convolutional Neural Network (CNN)

Machine learning approach:

- fast
- identifying spatial patterns
- better event reconstruction and classification



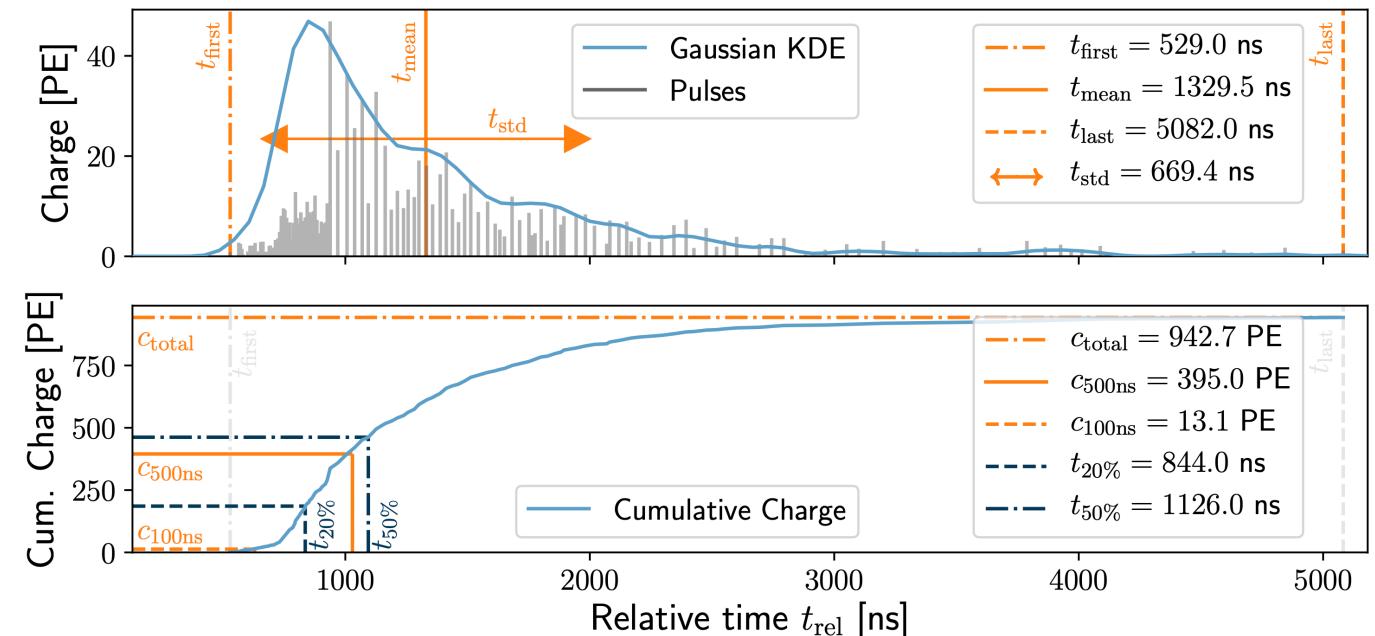
# Input data per DOM

## 3 inputs

- $c_{\text{total}}$ : Total charge
  - Sum of charge
- $t_{\text{first}}$ : Relative time of first pulse
  - Relative to total time offset, calculated as the charge weighted mean time of all pulses
- $t_{\text{std}}$ : Standard deviation of first pulse
  - Charge weighted standard deviation of pulse times relative to total time offset

## 9 inputs

- $t_{\text{last}}$ : Relative time of last pulse
  - Relative to total time offset, calculated as the charge weighted mean time of all pulses
- $t_{20\%}$ : Relative time of 20% charge
  - Relative to total time offset, calculated as the charge weighted mean time of all pulses
- $t_{50\%}$ : Relative time of 50% charge
  - Relative to total time offset, calculated as the charge weighted mean time of all pulses
- $t_{\text{mean}}$ : Mean time
  - Charge weighted mean time of all pulses relative to total time offset
- $c_{500\text{ns}}$ : Charge at 500ns
  - Sum of charge after 500ns
- $c_{100\text{ns}}$ : Charge at 100ns
  - Sum of charge after 100ns



## Input pulses

- SplitInIceDSTPulses
- SplitInIceDSTPulsesTWCleaning6000ns
- (DNN framework performs an internal cleaning)

## Training datasets

- 20904
- 21962
- 22020
- 22187

# Reconstructed properties

## Energy

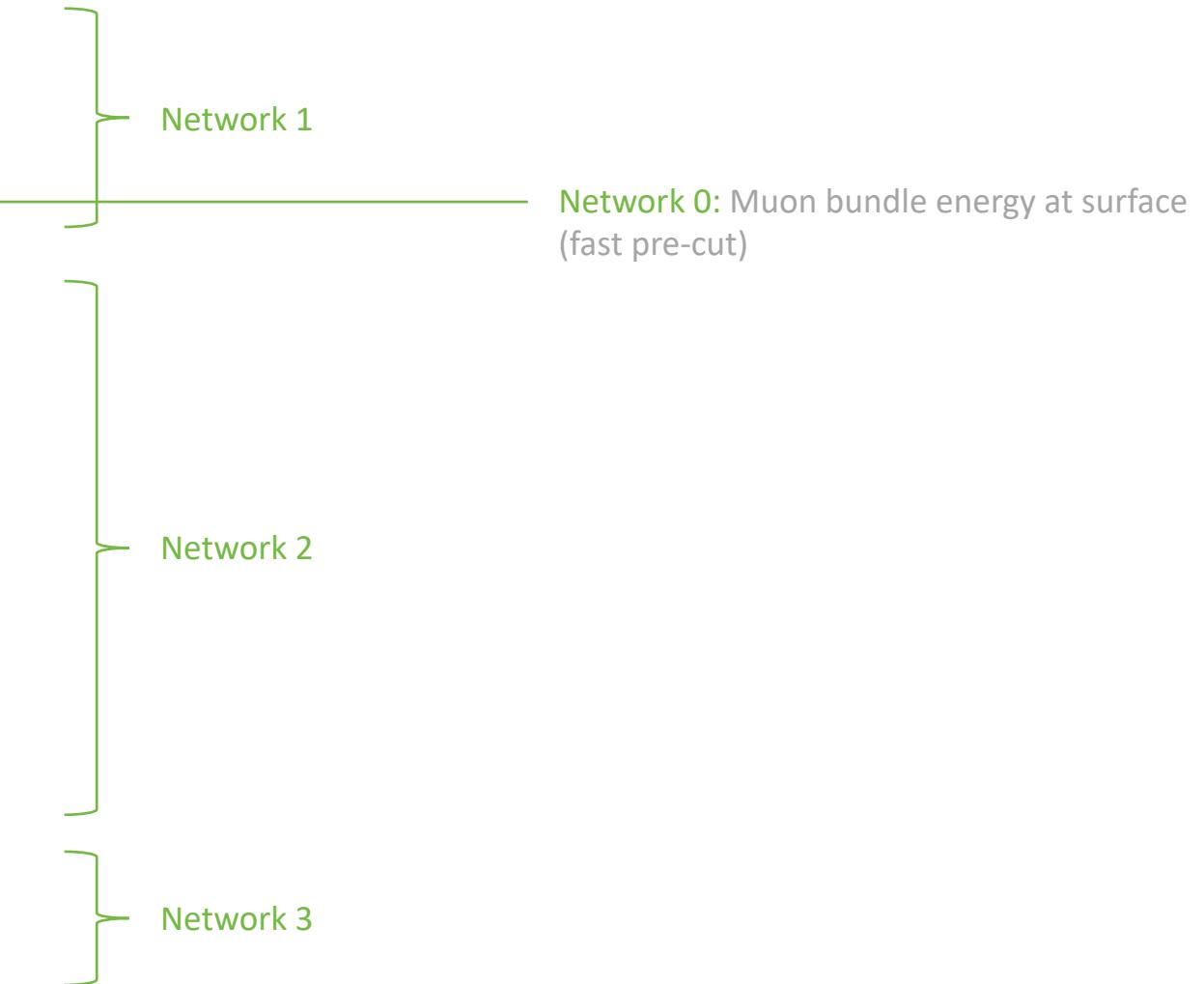
- `entry_energy`: Leading muon energy at the detector entry
- `bundle_energy_at_entry`: Muon bundle energy at the detector entry
- `muon_energy_first_mctree`: Leading muon energy at surface
- `bundle_energy_in_mctree`: Muon bundle energy at surface

## Track geometry

- `Length`: Propagation length of muon in the ice
- `LengthInDetector`: Propagation length of muon in the detector
- `center_pos_x`: Closest x position of muon to center of the detector
- `center_pos_y`: Closest y position of muon to center of the detector
- `center_pos_z`: Closest z position of muon to center of the detector
- `center_pos_t`: Time of closest approach to the center of the detector
- `entry_pos_x`: x position of muon at the detector entry
- `entry_pos_y`: y position of muon at the detector entry
- `entry_pos_z`: z position of muon at the detector entry
- `entry_pos_t`: Time of muon at the detector entry

## Direction

- `zenith`: Zenith angle of muon
- `azimuth`: Azimuth angle of muon



# Level5: quality cuts

containment cuts	>	<
length in detector	1000 m	2000 m
entry pos x, y	-750 m	750 m
entry pos z	-500 m	750 m
center pos x, y	-550 m	550 m
center pos z	-650 m	650 m

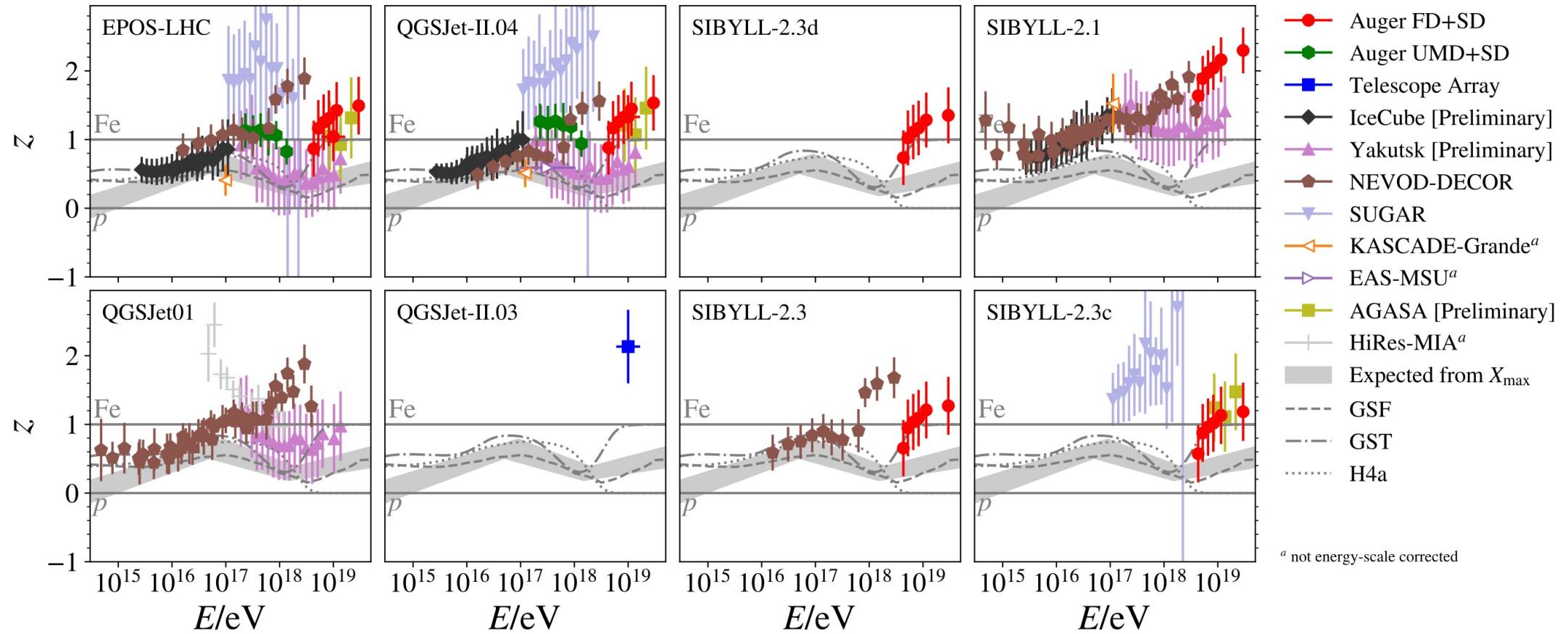
neutrino cuts	>	<
$\cos(\text{zenith})$	0.2	
length	5000 m	15000 m

uncertainty cuts	<
bundle energy at entry	$0.9 \log_{10}(\text{GeV})$
bundle energy at surface	$2.0 \log_{10}(\text{GeV})$
zenith	0.1 rad
azimuth	0.2 rad
entry pos x, y, z	42 m
center pos x, y, z	50 m
entry pos time	200 ns
center pos time	600 ns
length in detector	160 m
length	2000 m

# The Muon Puzzle

"muon number"

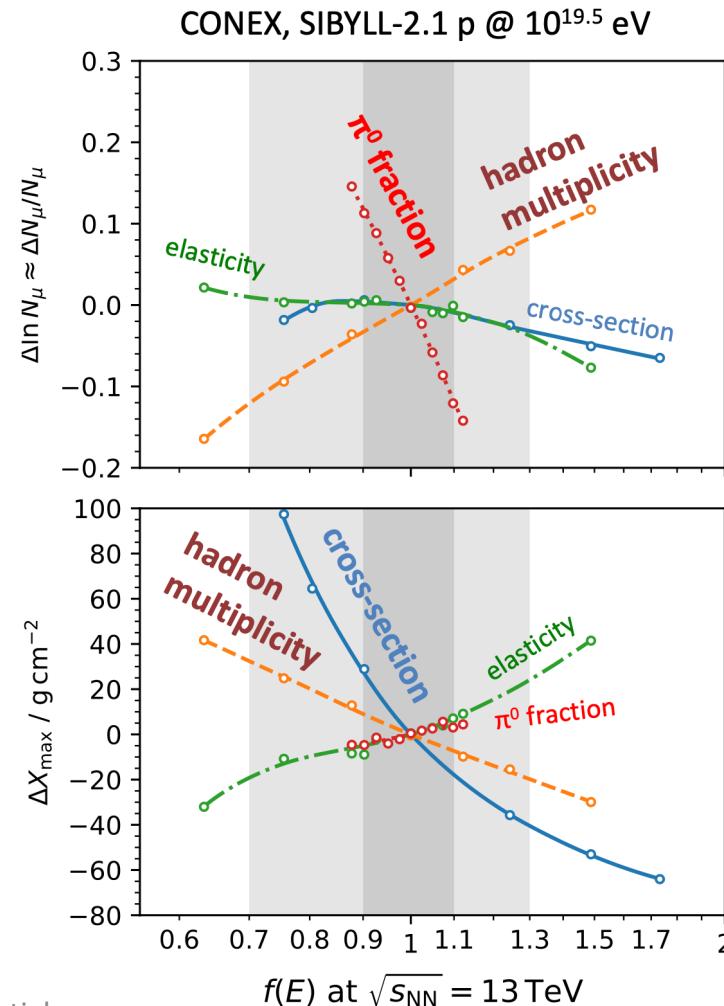
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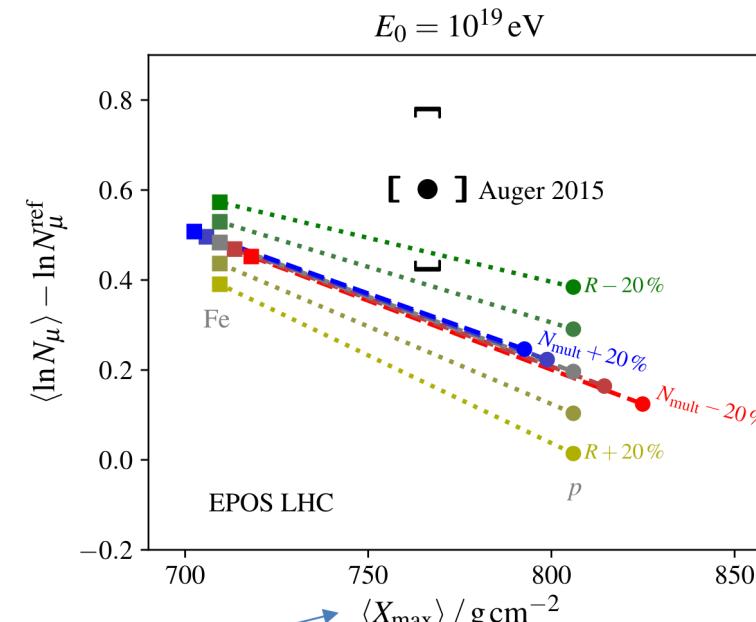
Arxiv: 2108.08341

# Possible Solutions

R. Ulrich, R. Engel, M. Unger, PRD 83 (2011) 054026



S. Baur, HD, M. Perlin, T. Pierog, R. Ulrich, K. Werner,  
arXiv:1902.09265



$$R = \frac{E_{\pi^0}}{E_{\text{other hadrons}}}$$

- Only changes to  $R$  can solve muon puzzle
- Small changes have large effect,  
 $R$  needs to be known to about 5 %