

Unfolding the Atmospheric Muon Flux with IceCube

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[wiki page](#)

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Collab: Anatoli

Tech: Karolin

Introduction

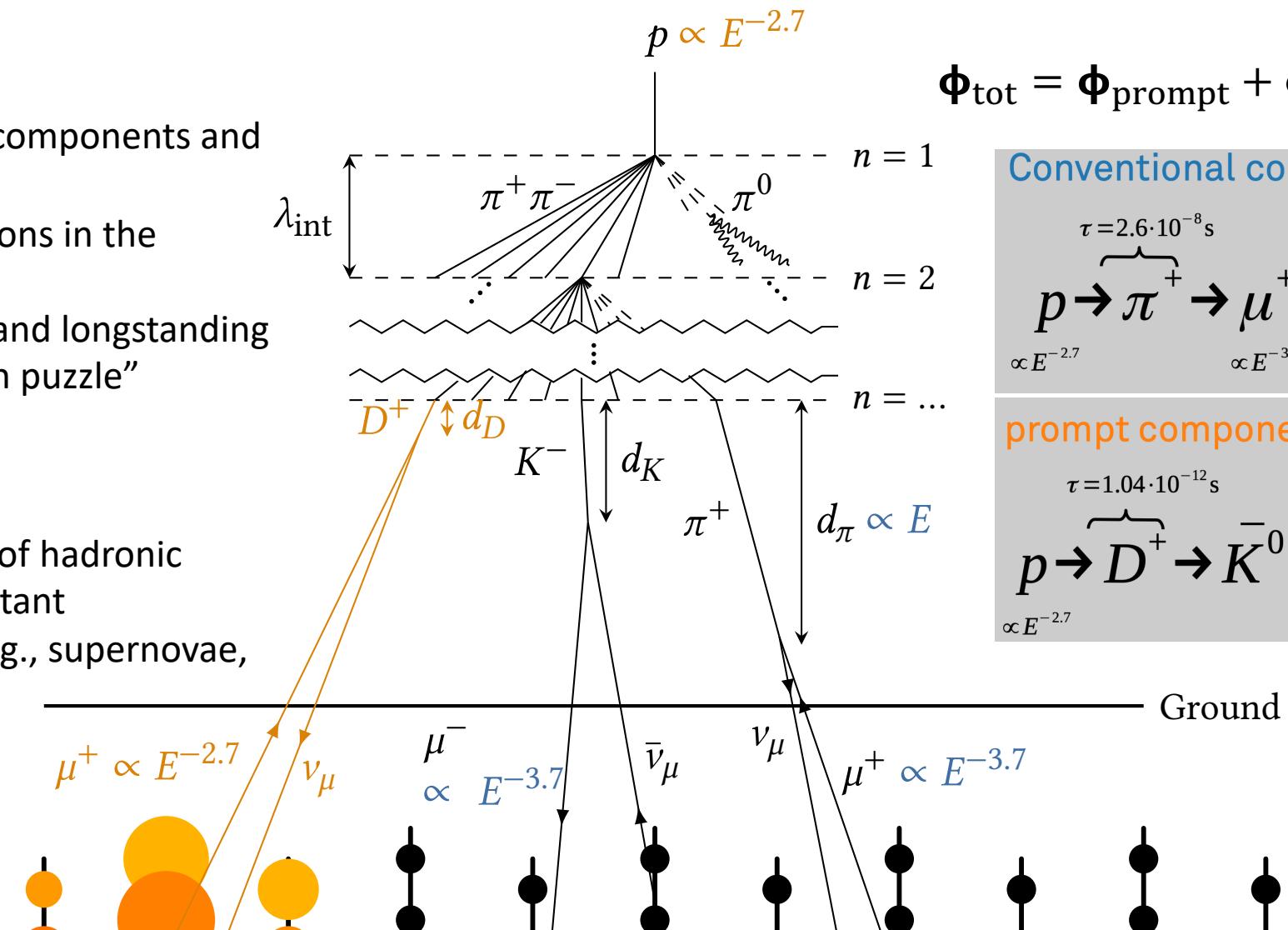
Motivation

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

Long term

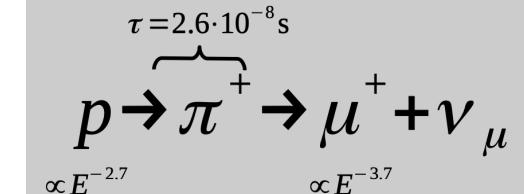
- Enhance understanding of hadronic processes relevant in distant astrophysical sources (e.g., supernovae, AGNs, ...)

Conventional Muon:
Parent is pion or kaon

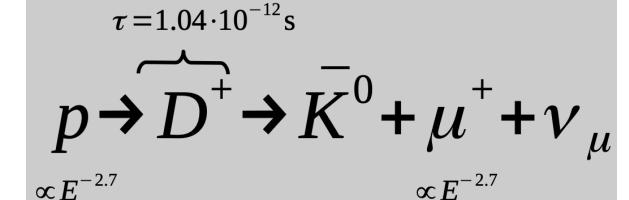


$$\Phi_{\text{tot}} = \Phi_{\text{prompt}} + \Phi_{\text{conventional}}$$

Conventional component:

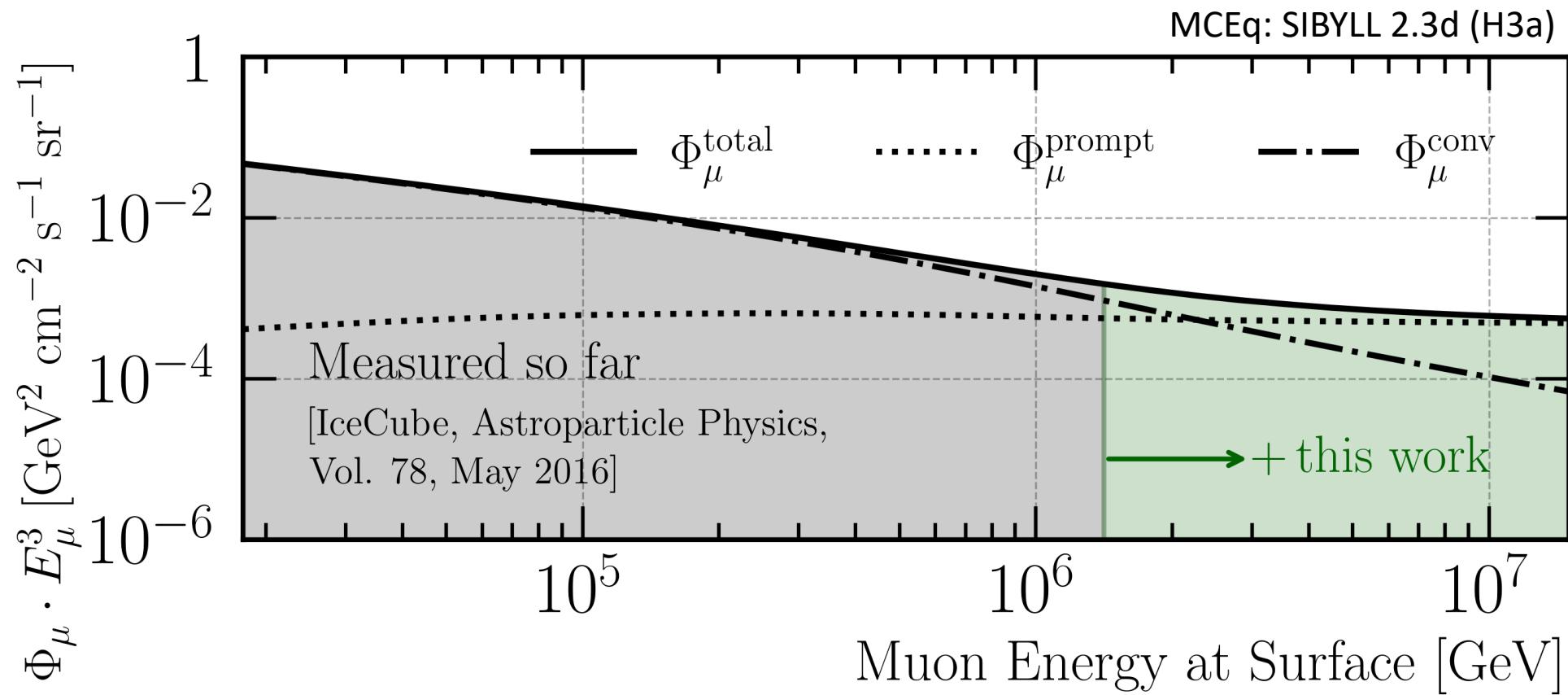


prompt component:

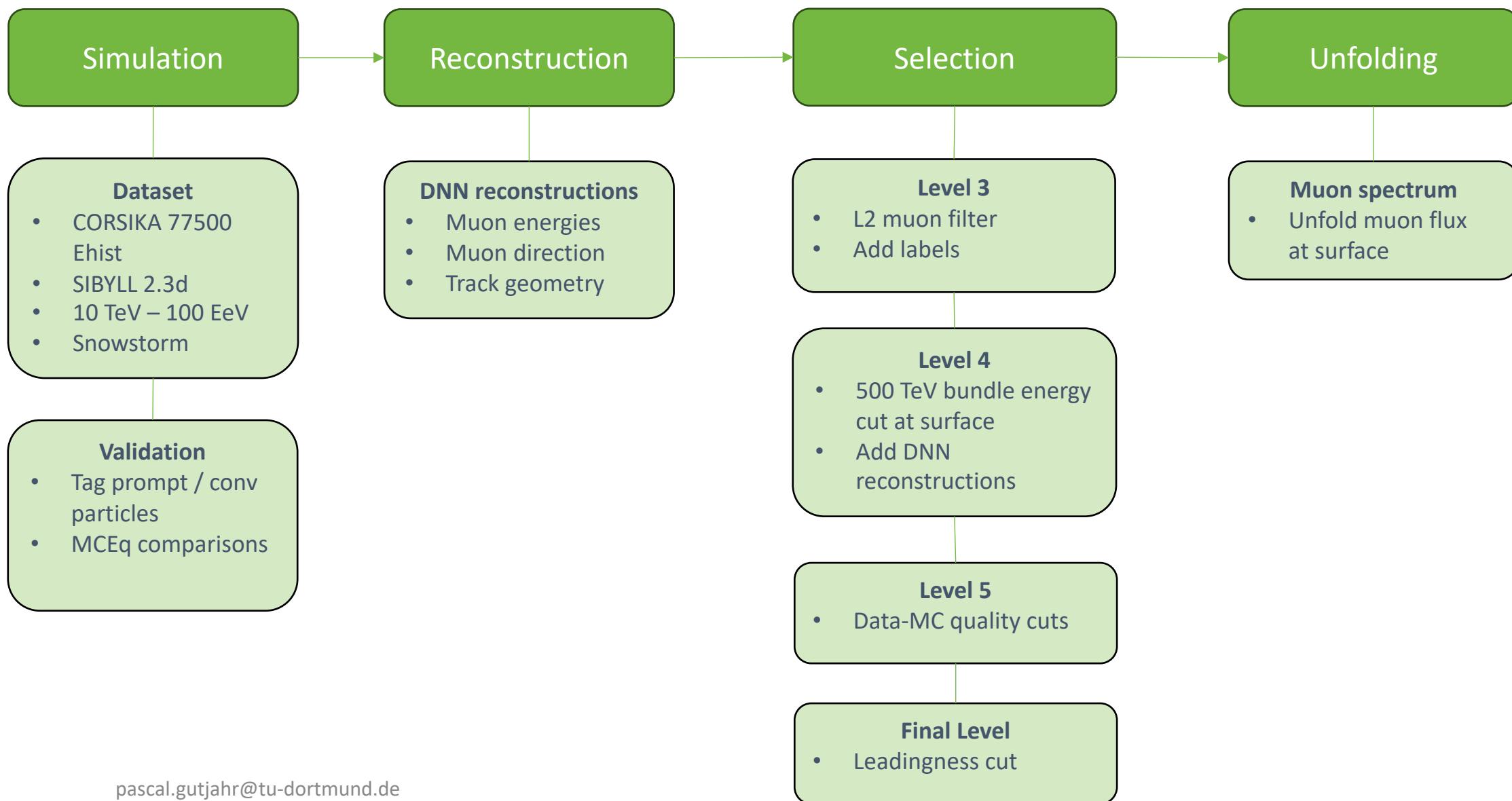


Credit: Ludwig Neste

Goal: Measure Muon Flux at Surface



Overview



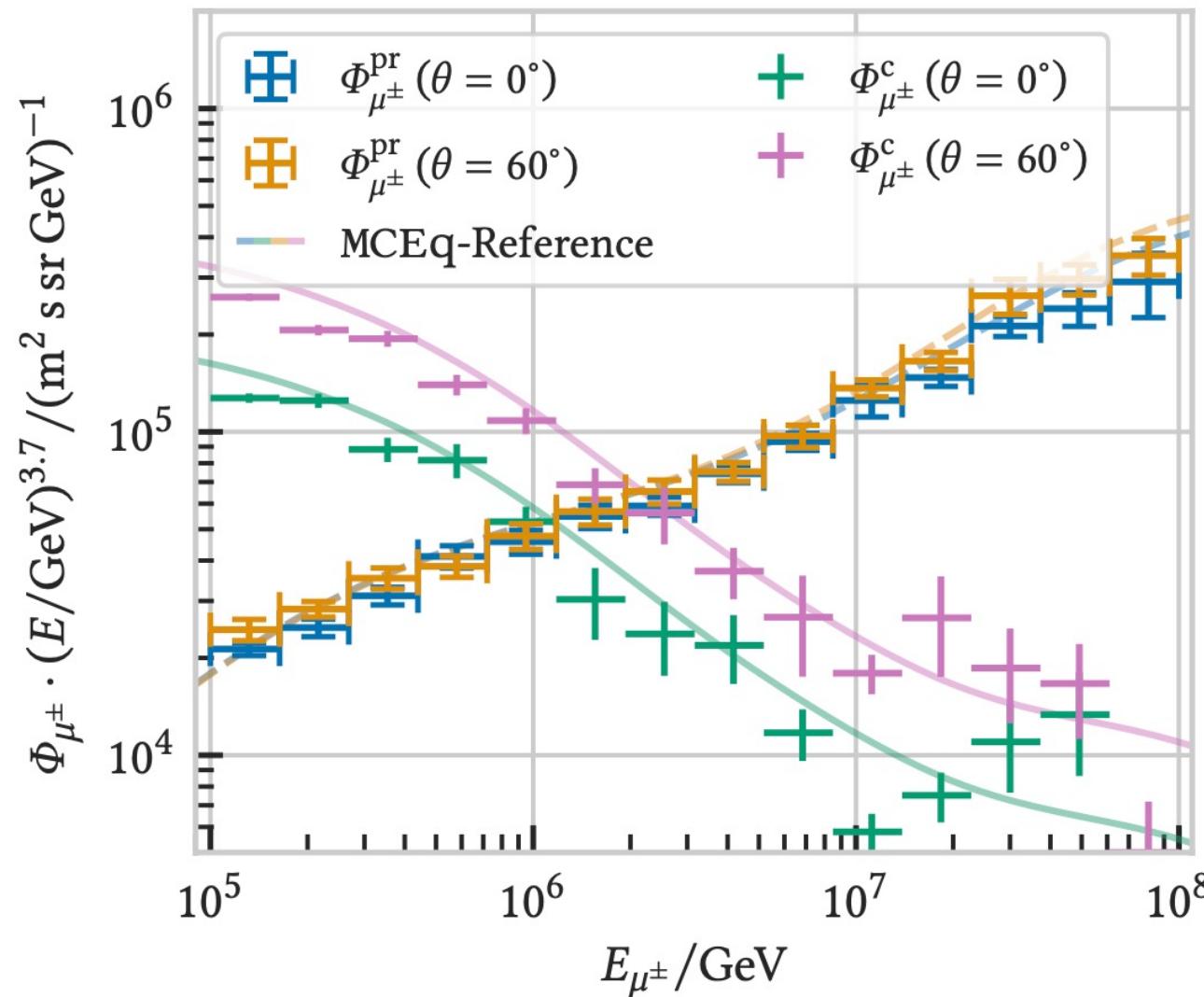
New CORSIKA simulation

with extended history option for information about the parent particles

6 Datasets: 22875, 22774–8

Primary Energy: 600 GeV – 1e11 GeV

CORSIKA 7 vs. MCEq

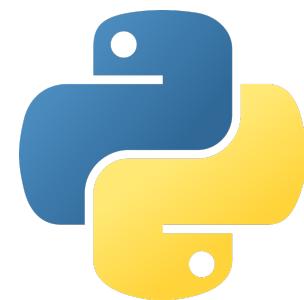


MCEq: tool to numerically solve the cascade equations that describes the evolution of particle densities as they propagate through a gaseous, dense medium
<https://github.com/mceq-project/MCEq>

➤ Good agreement for inclusive flux

Python package developed – PANAMA

- Execute CORSIKA 7 (multi core)
- Read DAT files → pandas DataFrames
- Parse EHIST option
- Calculate primary weightings



Reconstruction

Machine Learning (CNN)

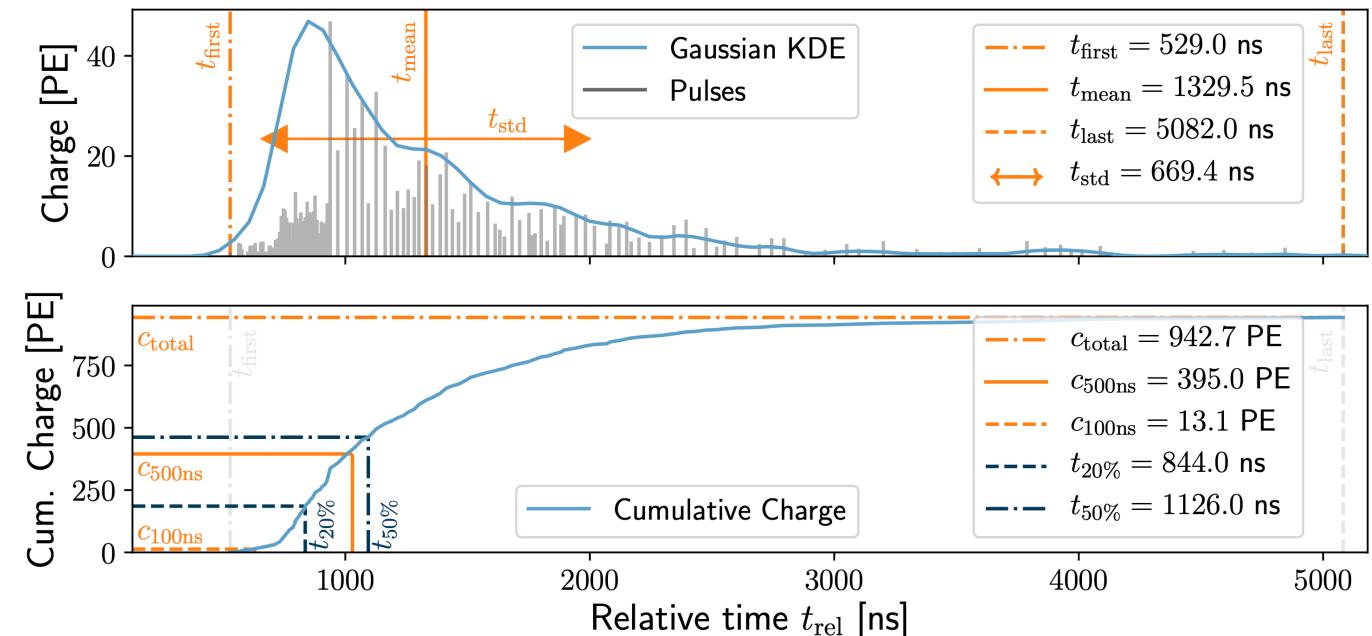
Input data per DOM

3 inputs

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

9 inputs

- t_{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_{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

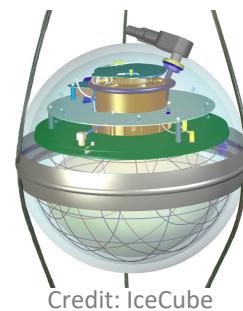


Input pulse cleaning

- 6000 ns

Training data

- Different simulations for robustness



Reconstructed properties

Energy

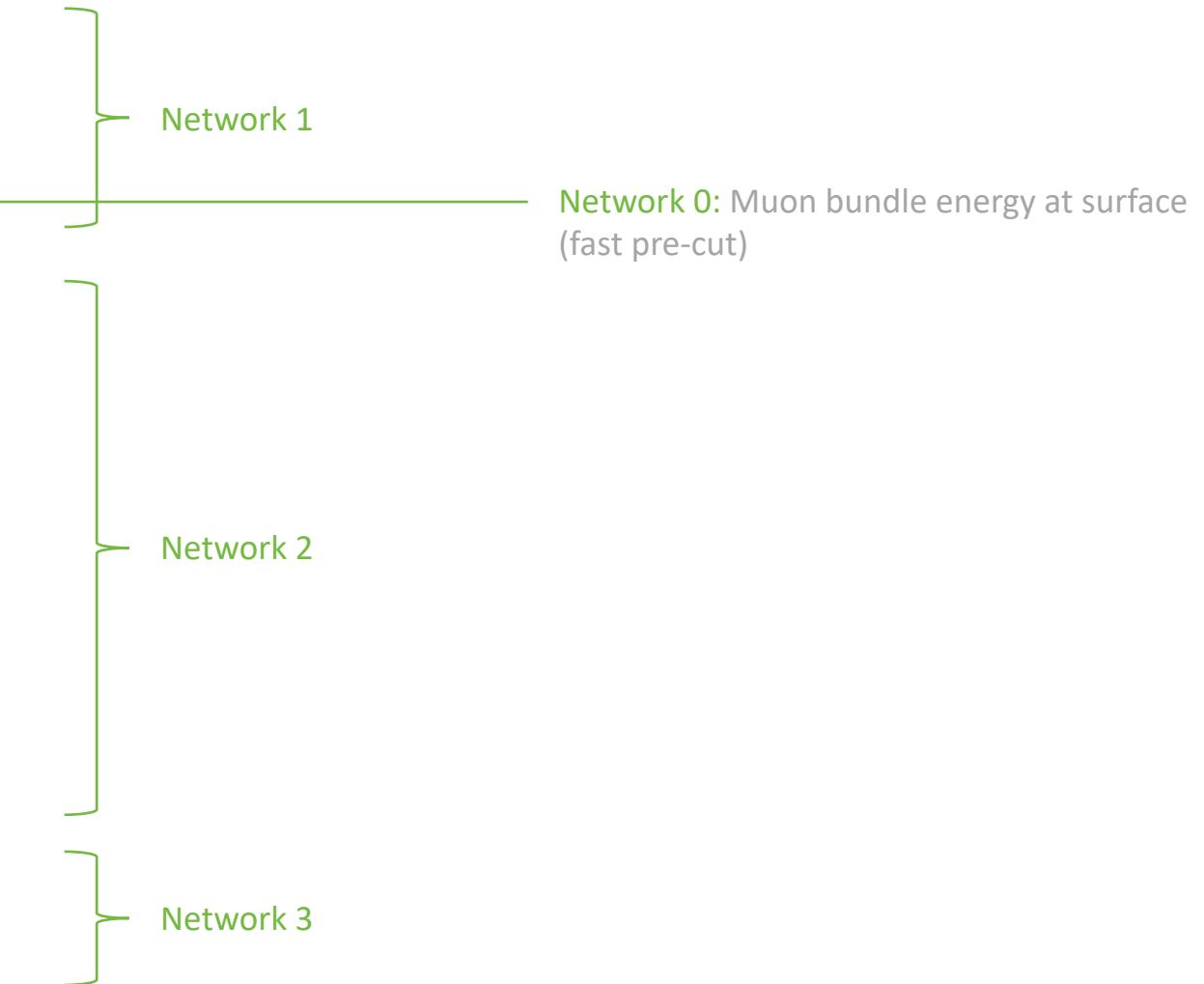
- `entry_energy`: Leading muon energy at the detector entry
- `bundle_energy_at_entry`: Muon bundle energy at the detector entry
- `muon_energy_first_mctree`: Leading muon energy at surface
- `bundle_energy_in_mctree`: Muon bundle energy at surface

Track geometry

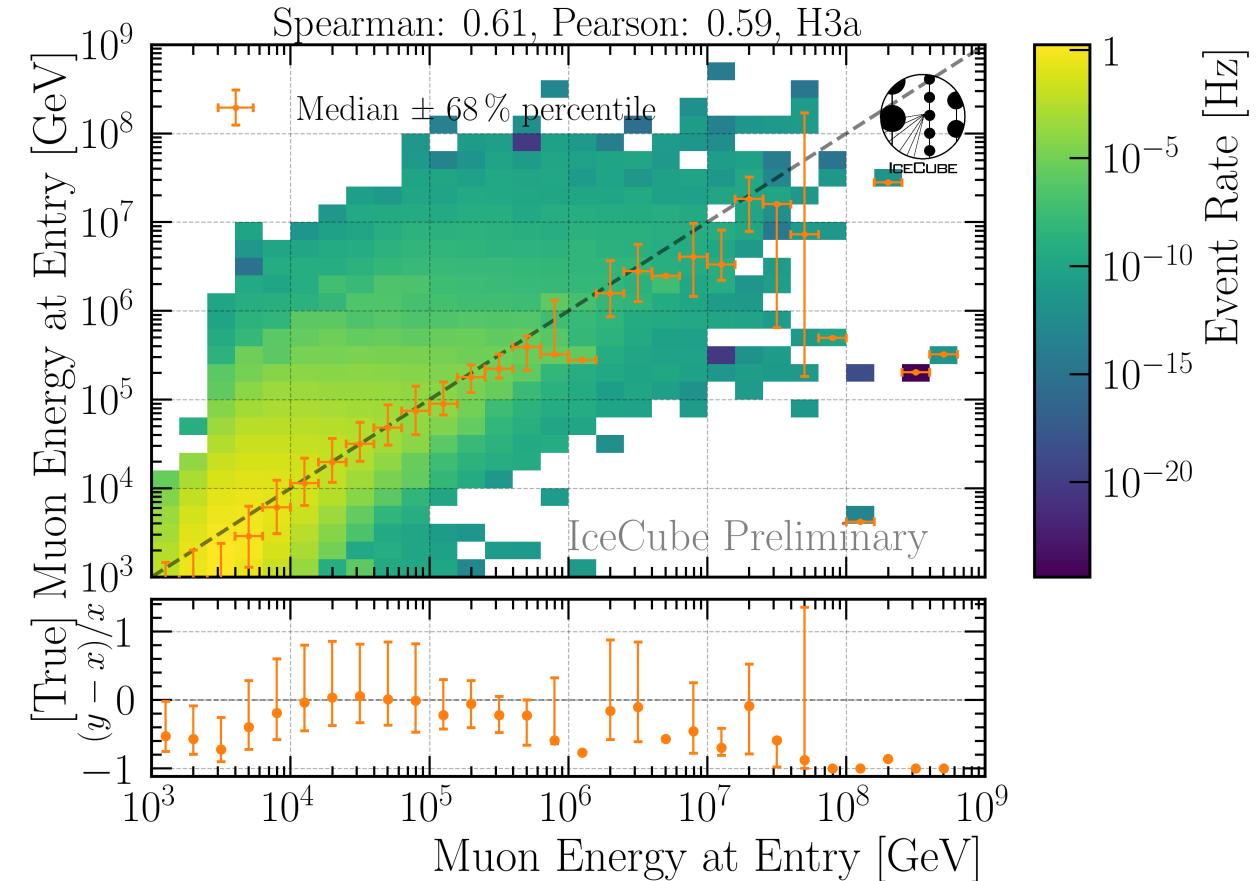
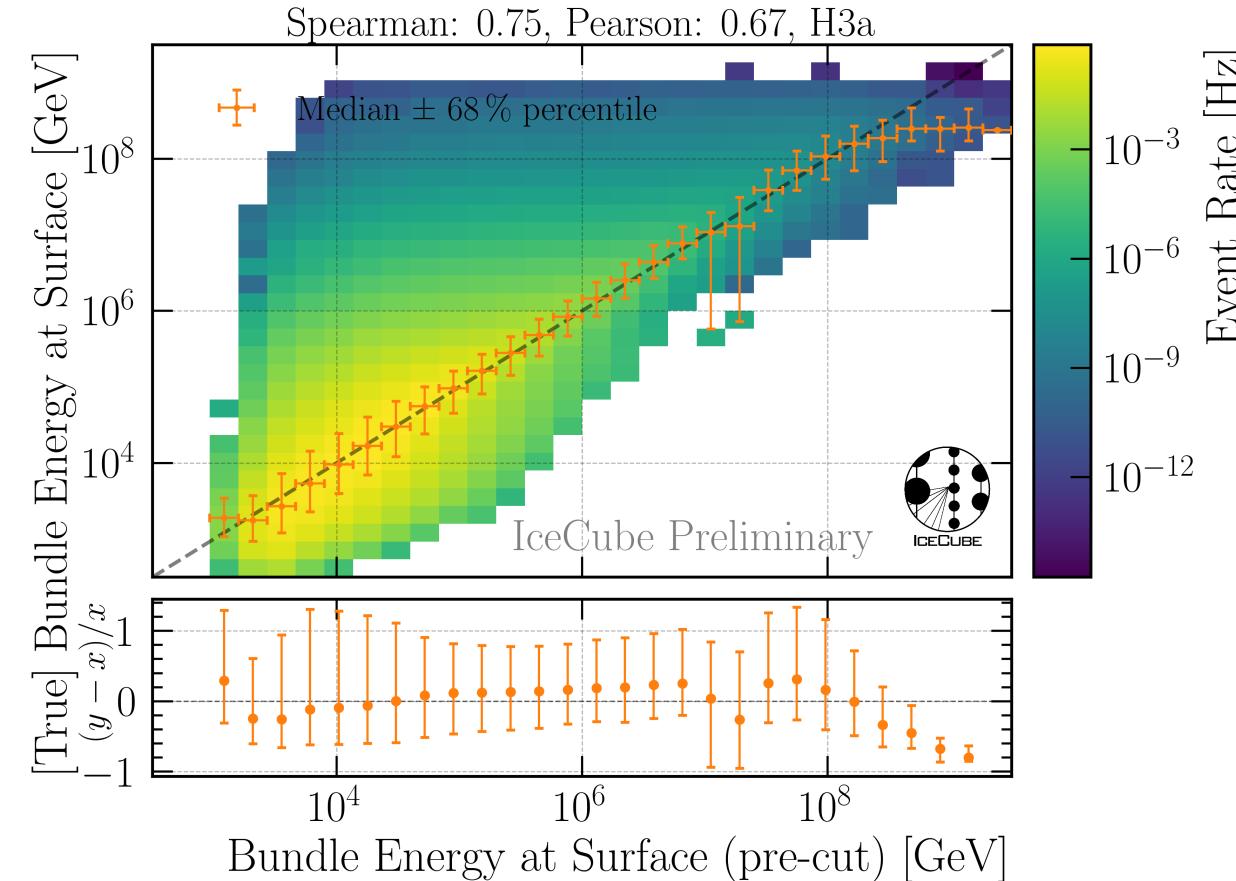
- `Length`: Propagation length of muon in the ice
- `LengthInDetector`: Propagation length of muon in the detector
- `center_pos_x`: Closest x position of muon to center of the detector
- `center_pos_y`: Closest y position of muon to center of the detector
- `center_pos_z`: Closest z position of muon 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



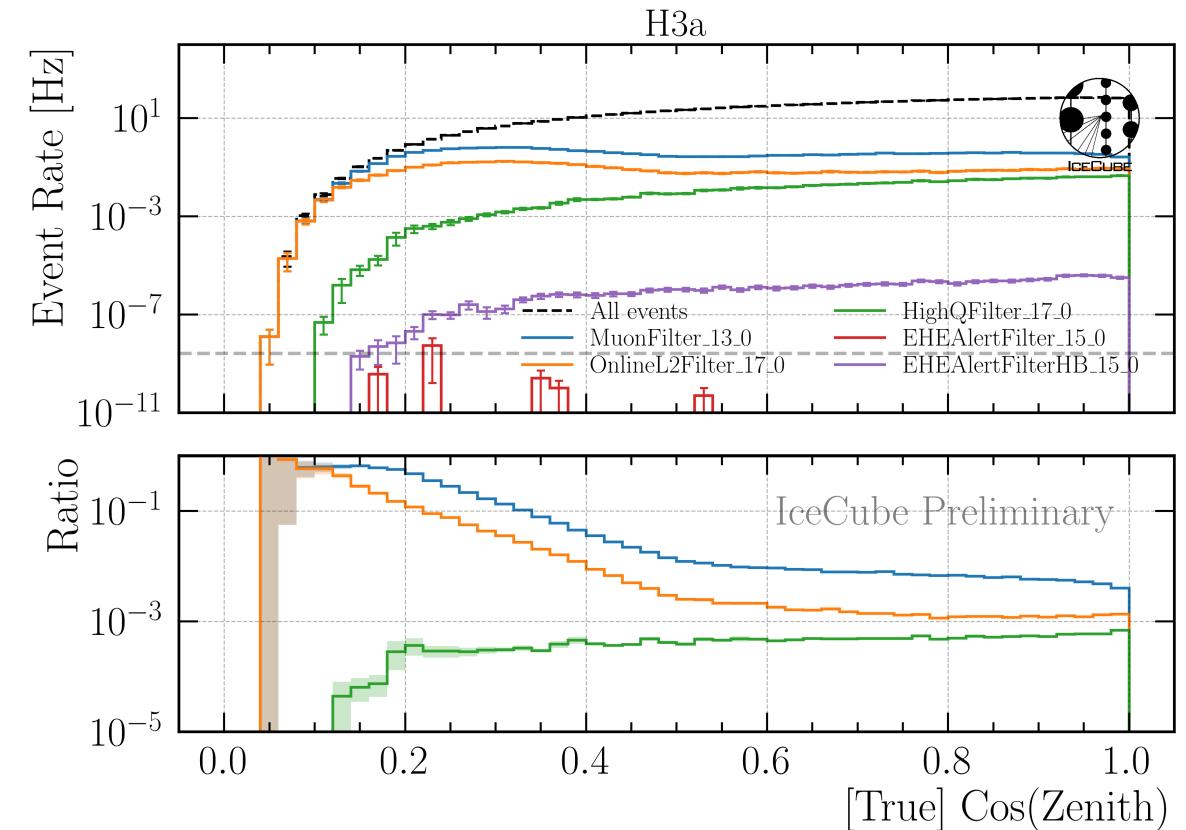
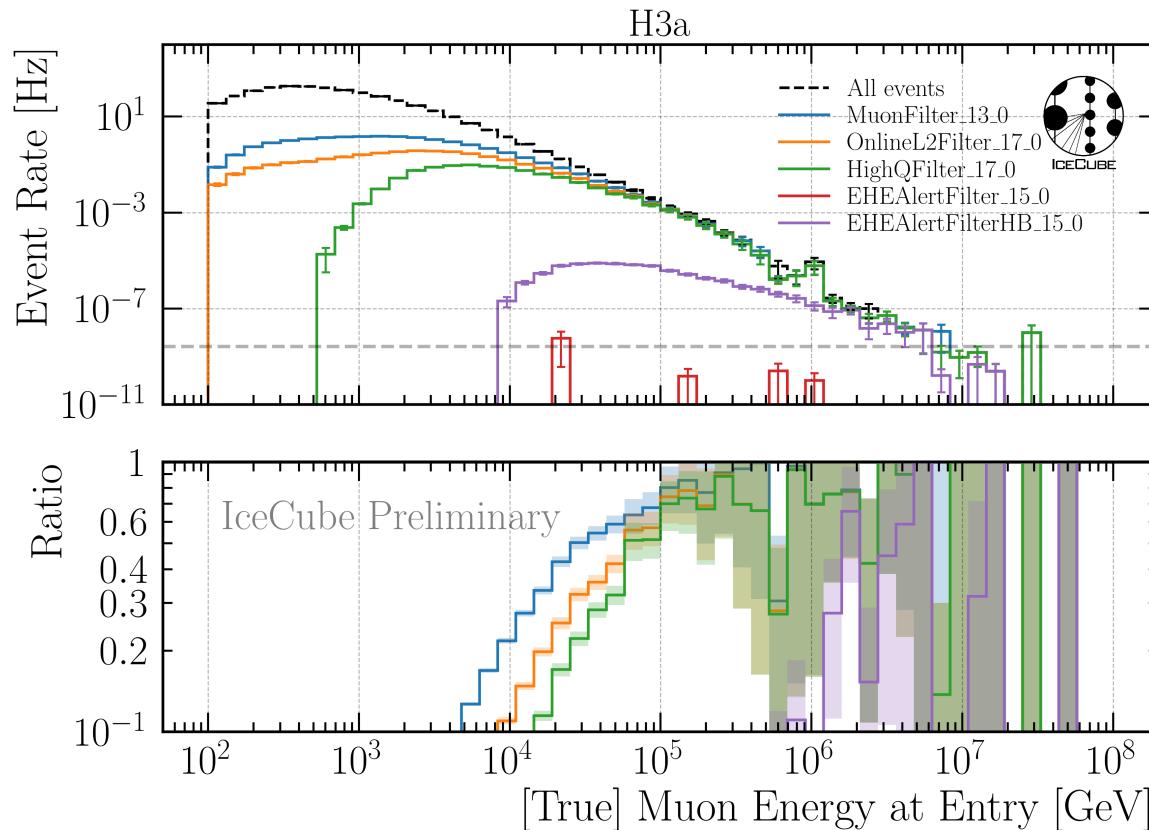
Energy Reconstructions



- Sufficient energy reconstructions
- Tight 68 % intervals, with outliers

Selection

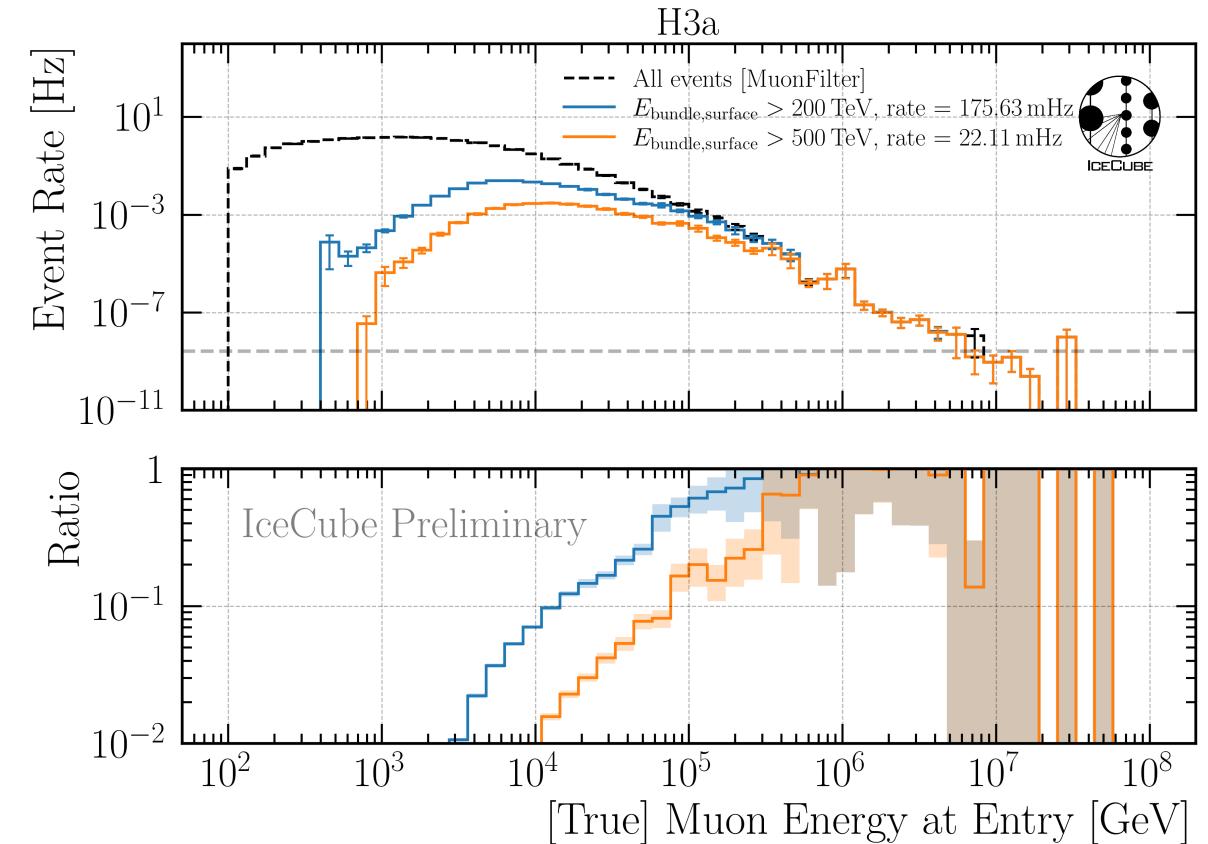
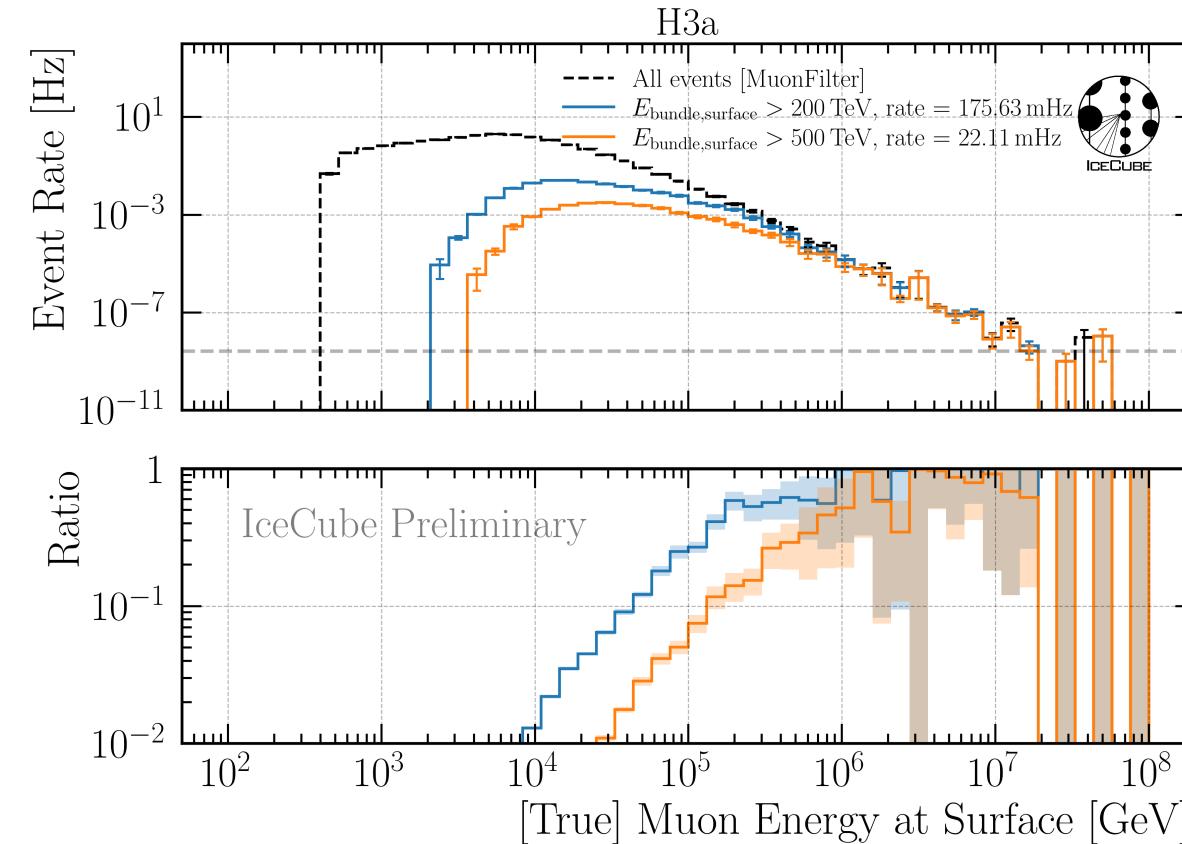
Level 3: Muon Filter



Muon filter: zenith-dependent charge and quality cut

➤ Choose muon filter to select as many high-energy muons as possible

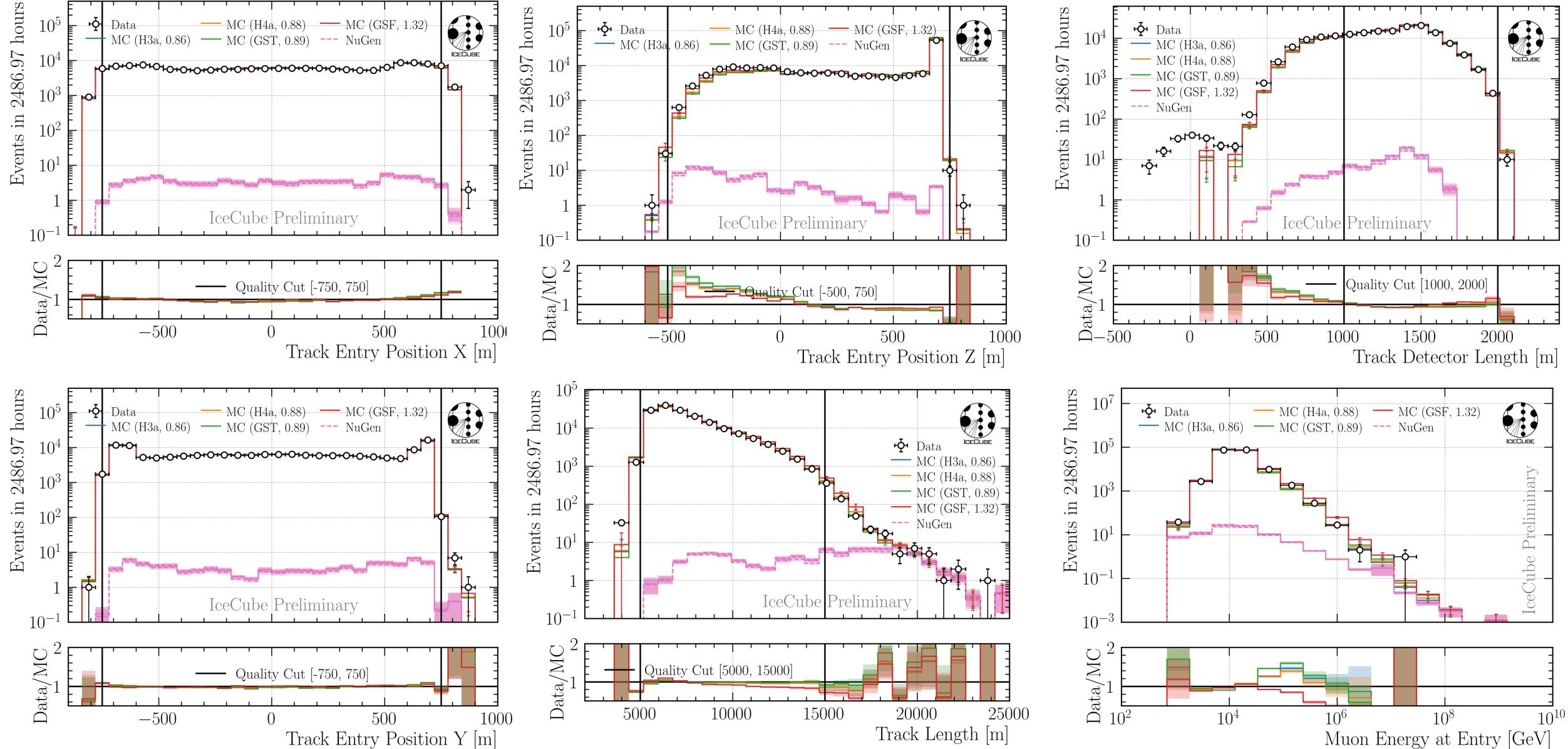
Level 4: Energy Cut



- 6 billion events expected in 10 years → computationally not feasible
- focus on high-energetic events
- Remove low-energy muons: bundle energy at surface > 500 TeV

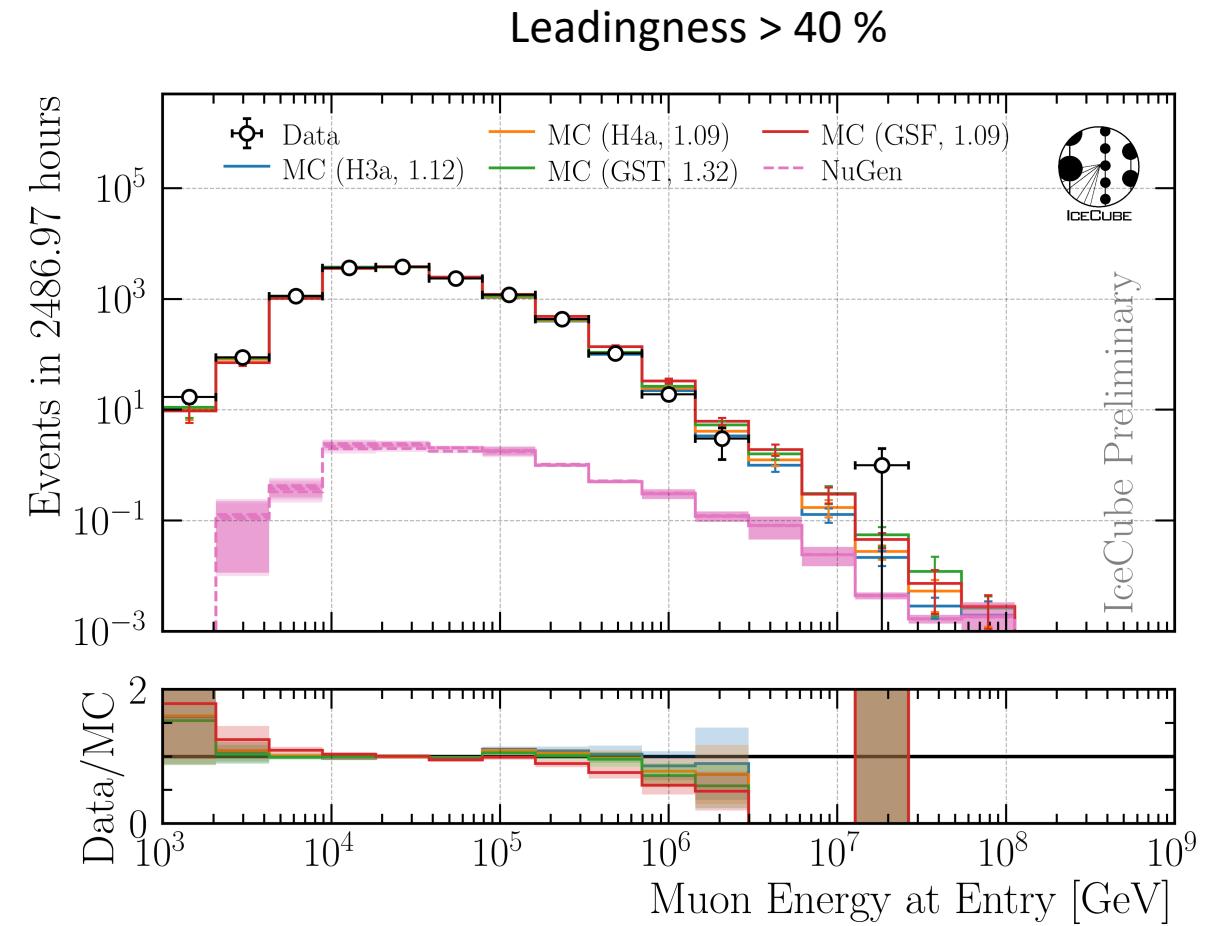
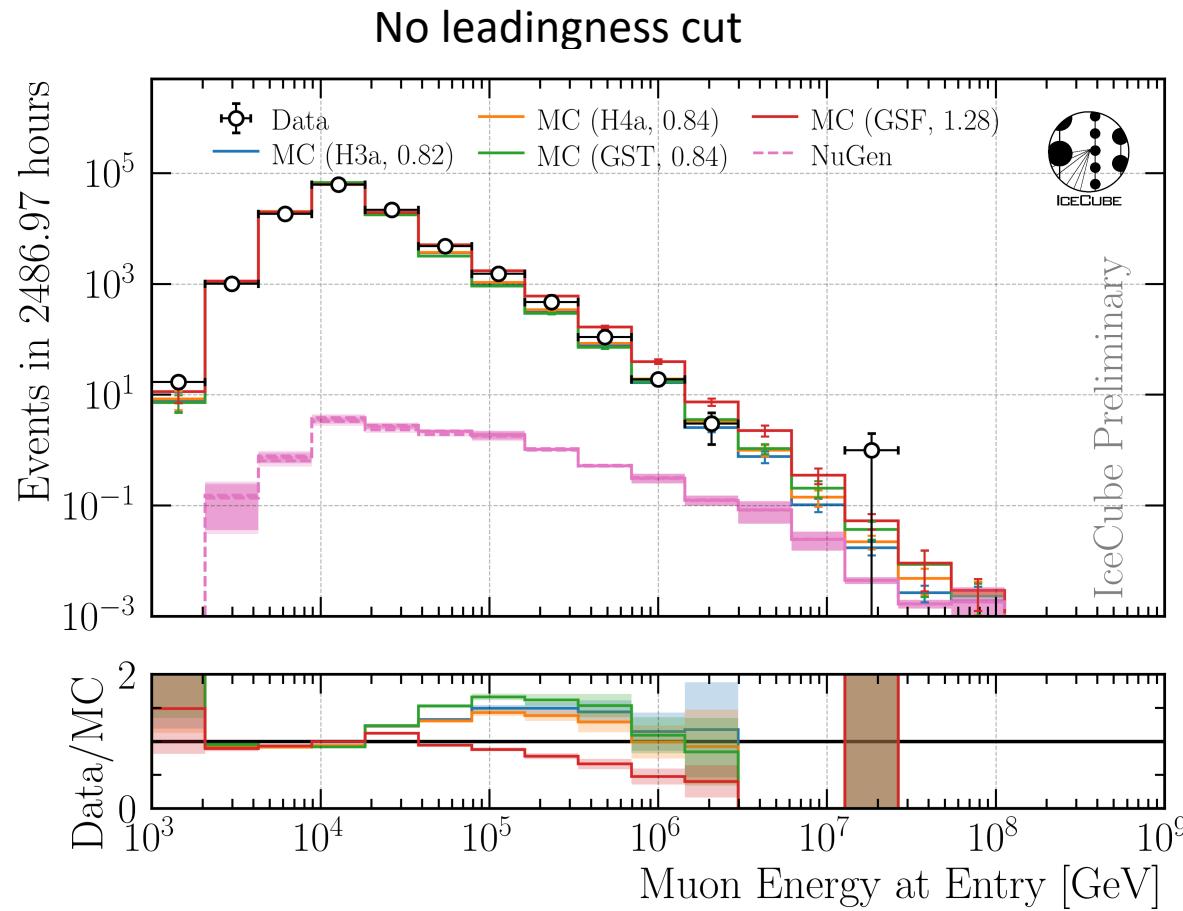
- Remove outliers
- Depth—dependent slope (no analysis relevance)
- CR—model impact on energy reconstruction
- 23 quality cuts in total

Level 5: Data/MC Quality Cuts



Final Level: Leadingness > 40 %

$$L = \frac{\text{Leading Energy at Entry}}{\text{Bundle Energy at Entry}}$$



➤ Improve Data/MC by leadingness cut

Unfolding

Unfolding in a nutshell

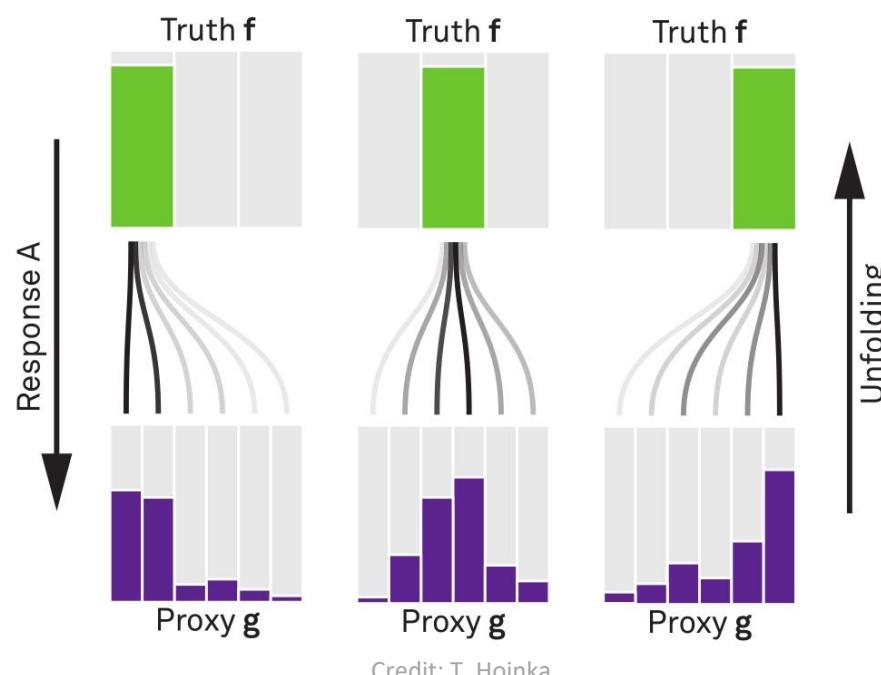
measured proxy

$$g(y) = \int_{E_0}^{E_1} A(E_\mu, y) f(E_\mu) dE_\mu + b(y)$$

detector response

background

true energy distribution



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- 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
- $$\mathcal{L}(\vec{g}|\vec{f}) = \prod_{j=1}^M \frac{\lambda_j^{g_j}}{g_j!} \exp(-\lambda_j)$$
- $$= \prod_{j=1}^M \frac{(A\vec{f})_j^{g_j}}{g_j!} \exp(-(A\vec{f})_j)$$
- $$t(\vec{f}) = -\frac{1}{2} (\vec{C}\vec{f})^T (\tau_1)^{-1} (\vec{C}\vec{f})$$

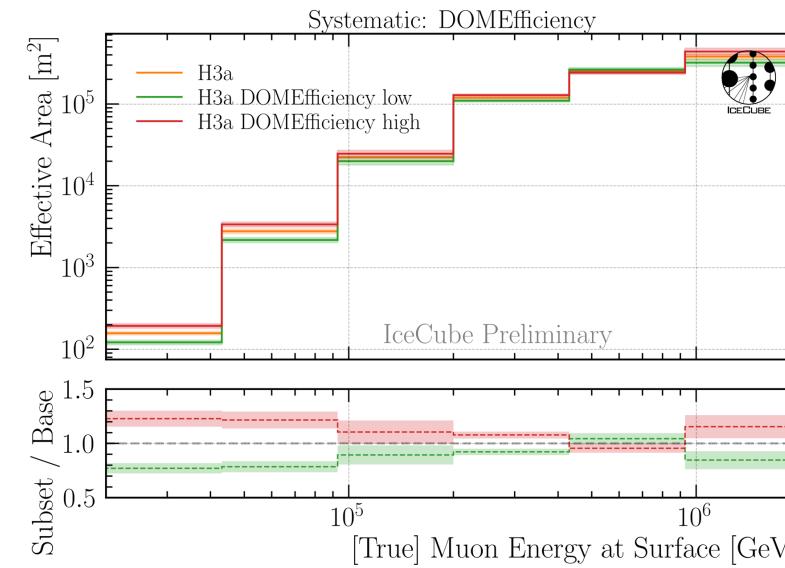
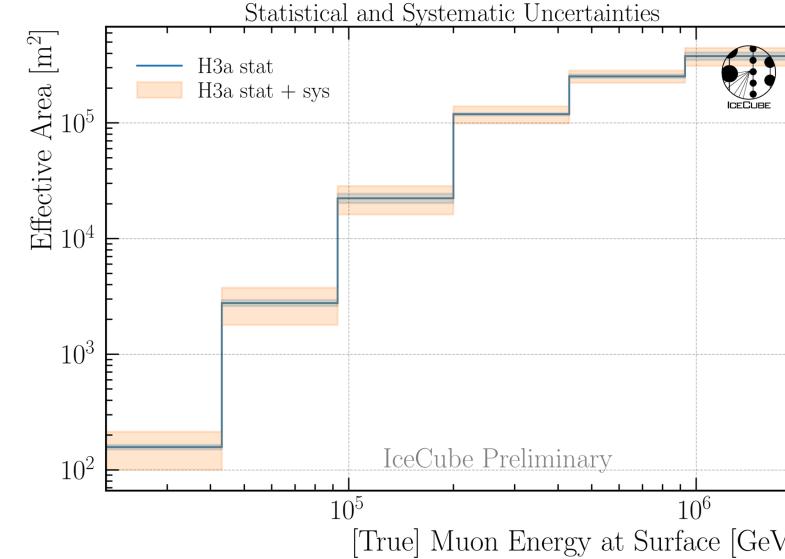


funfolding
by M. Börner

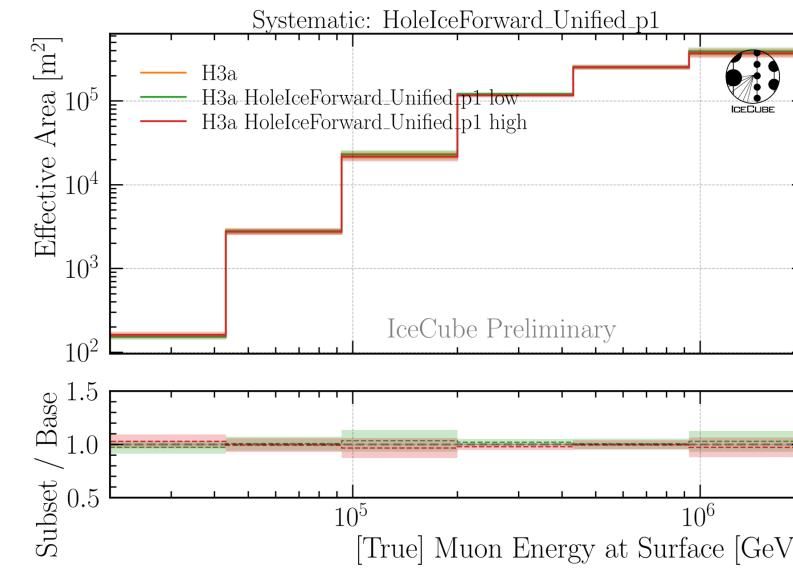
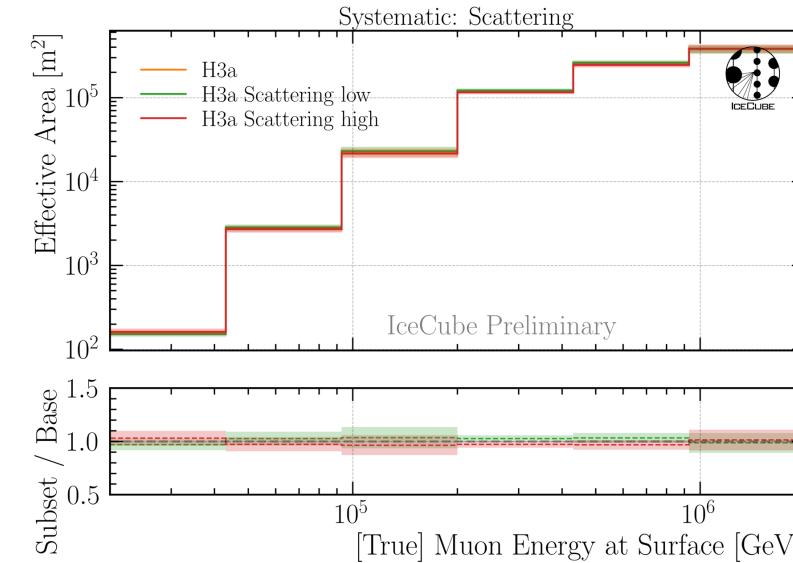
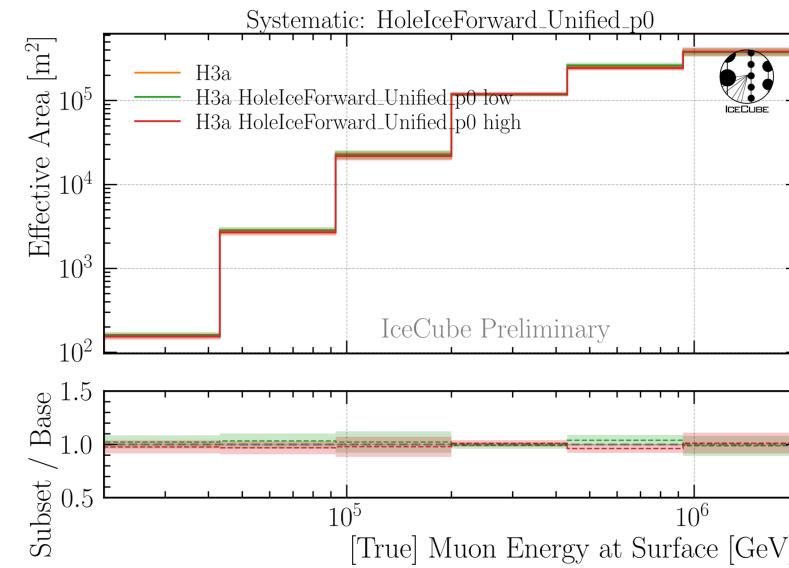
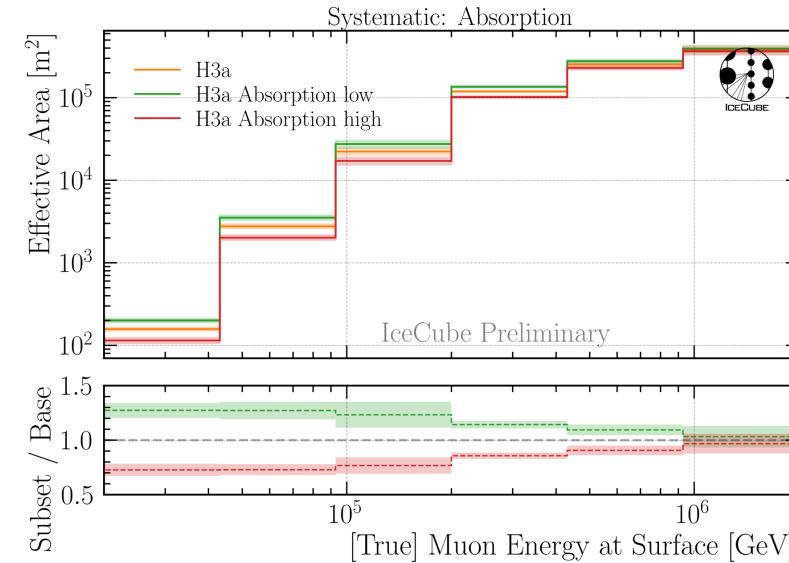
Muon Flux Unfolding

- ❑ Effective area
- ❑ Systematics
- ❑ Proxy that correlates with target
- ❑ Regularization
 - Unfolding

Ice & Detector Systematics



$$\sigma_{\text{sys},i} = \max(A_{\text{low}}, A_{\text{high}}) - A_{\text{base}}$$



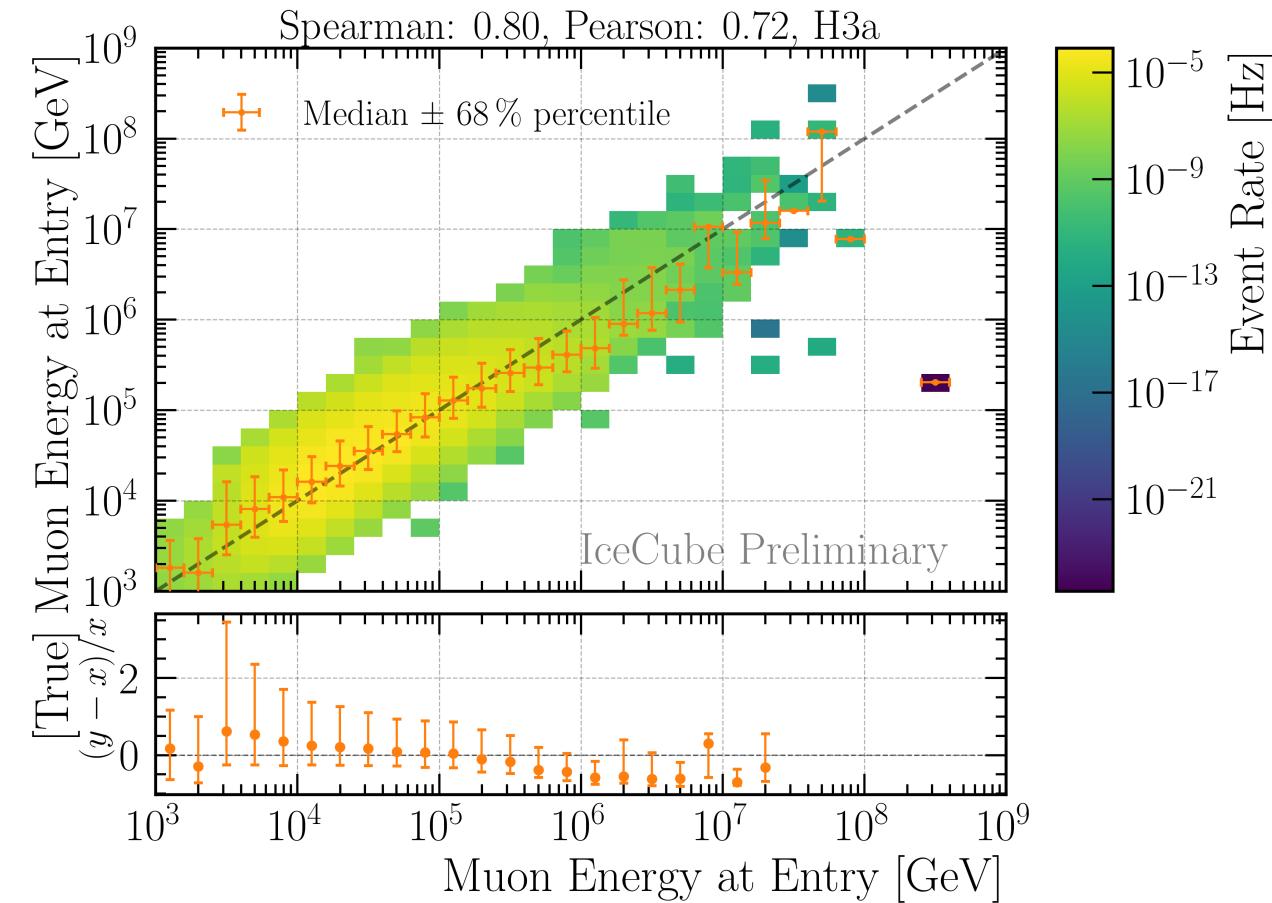
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- Baseline: entire set
- Subset: above/below center

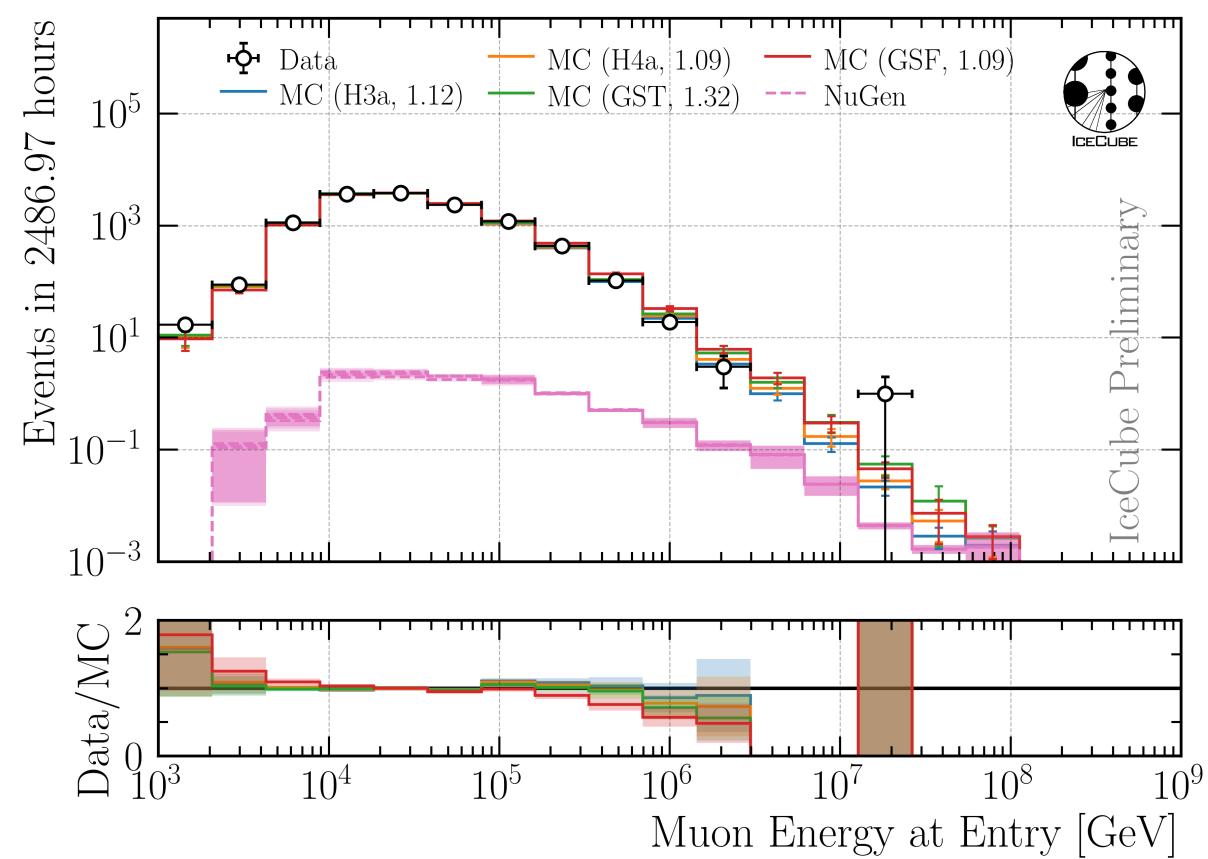
$$\sigma_{\text{tot}} = \sqrt{\sigma_{\text{stat}}^2 + \sum_i \sigma_i^2}, i: \text{scat, abs, DOME, holeice p0, p1}$$

Reconstruction and Data—MC: Leading Muons

- Good reconstruction of leading muon energy → proxy

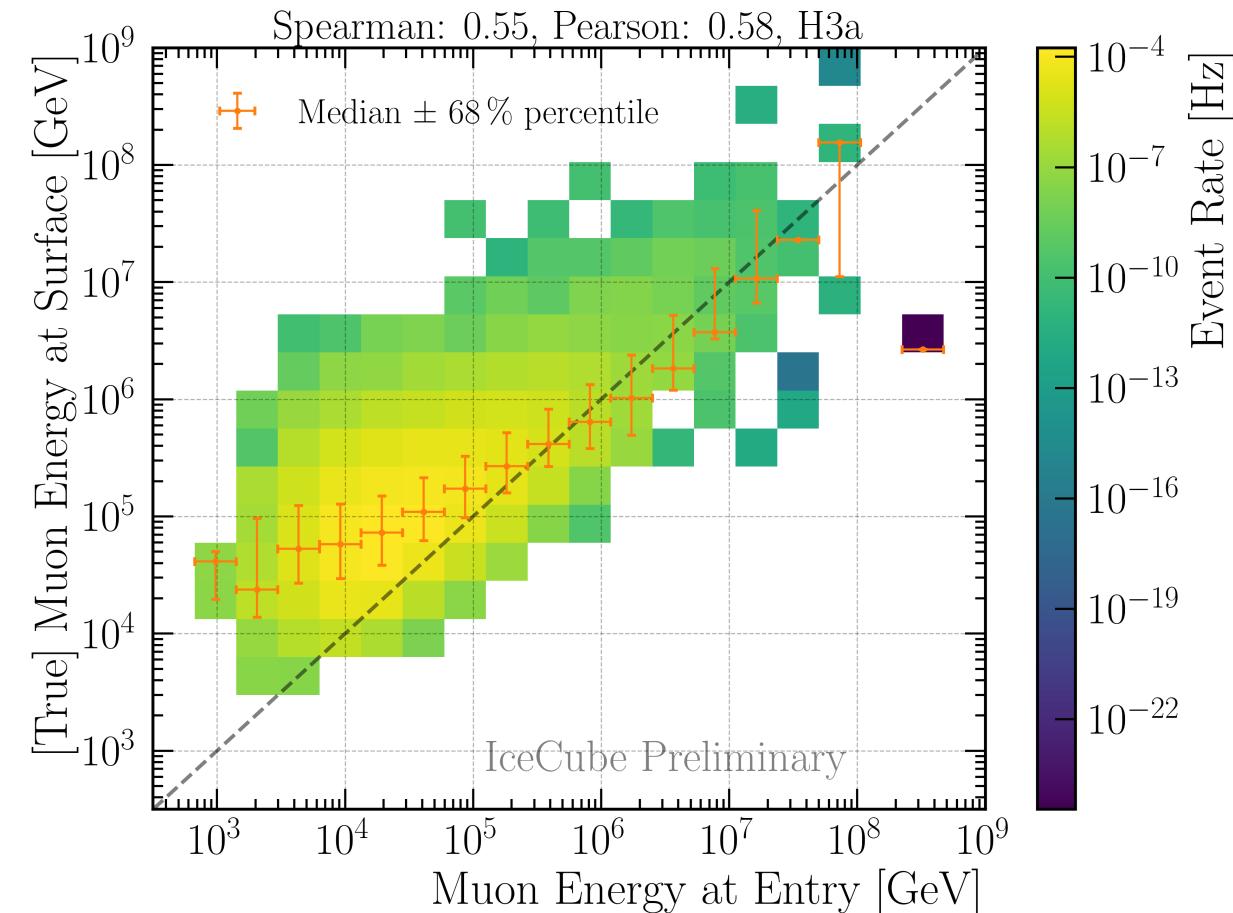


- Good data—MC agreement
- Global offset → upscale MC by 12%

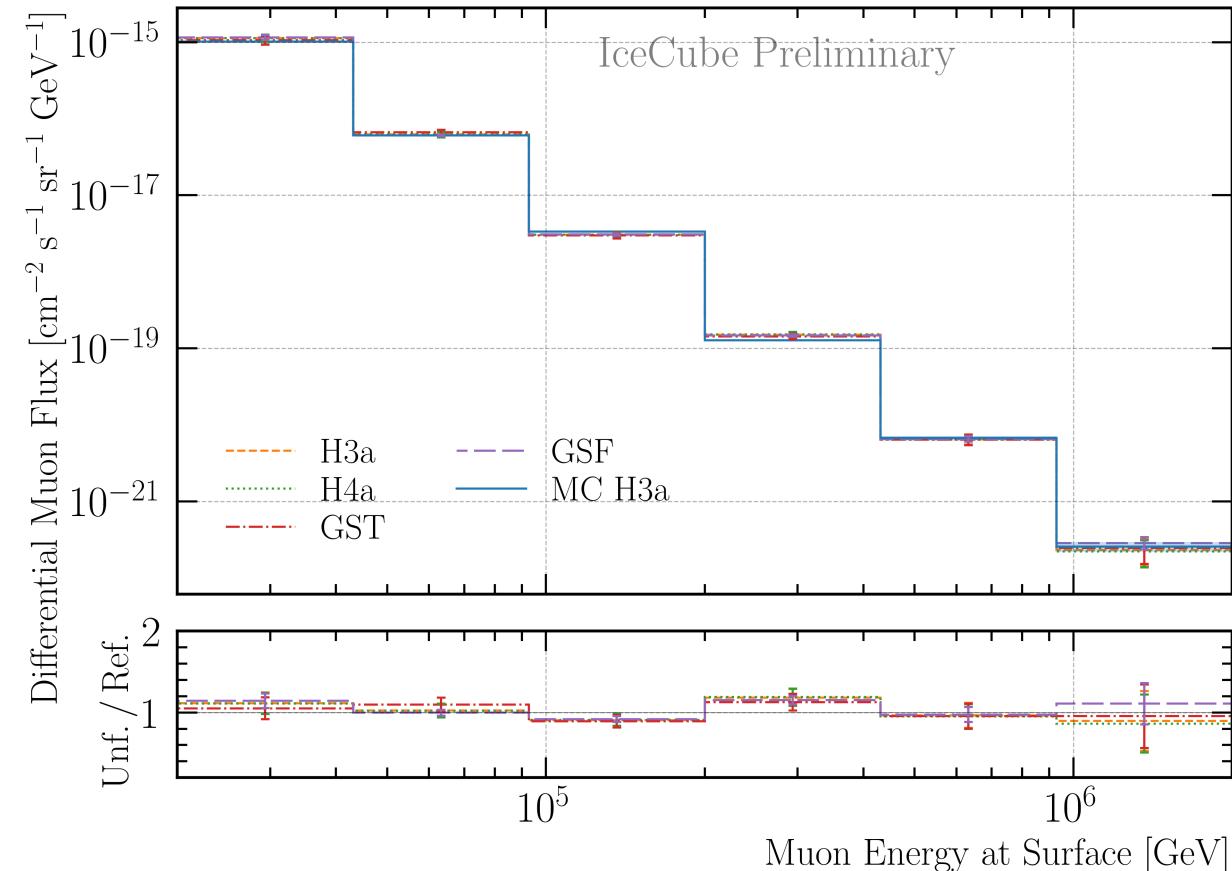


Correlation and Robustness Test: Leading Muons

- Correlation between proxy and target



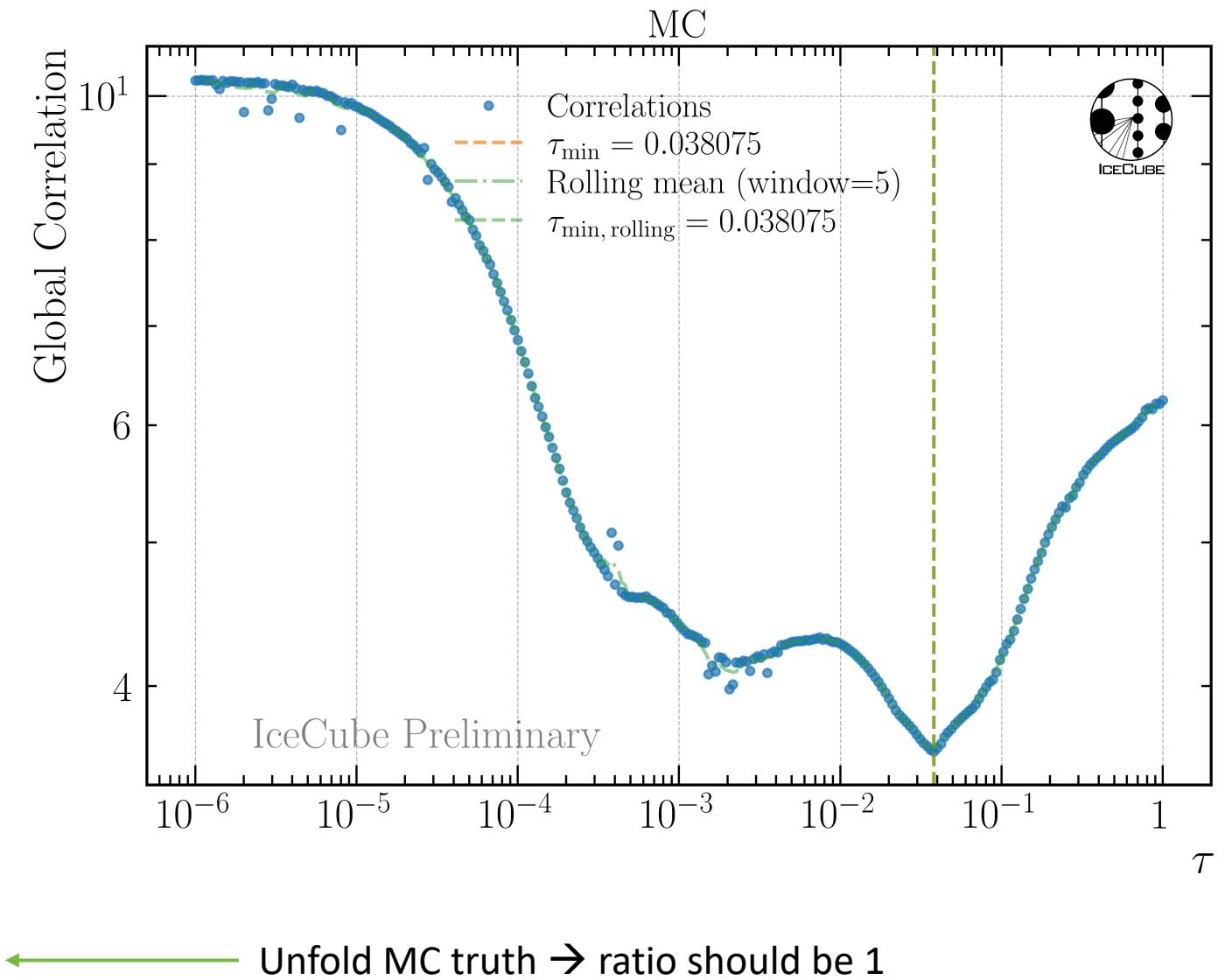
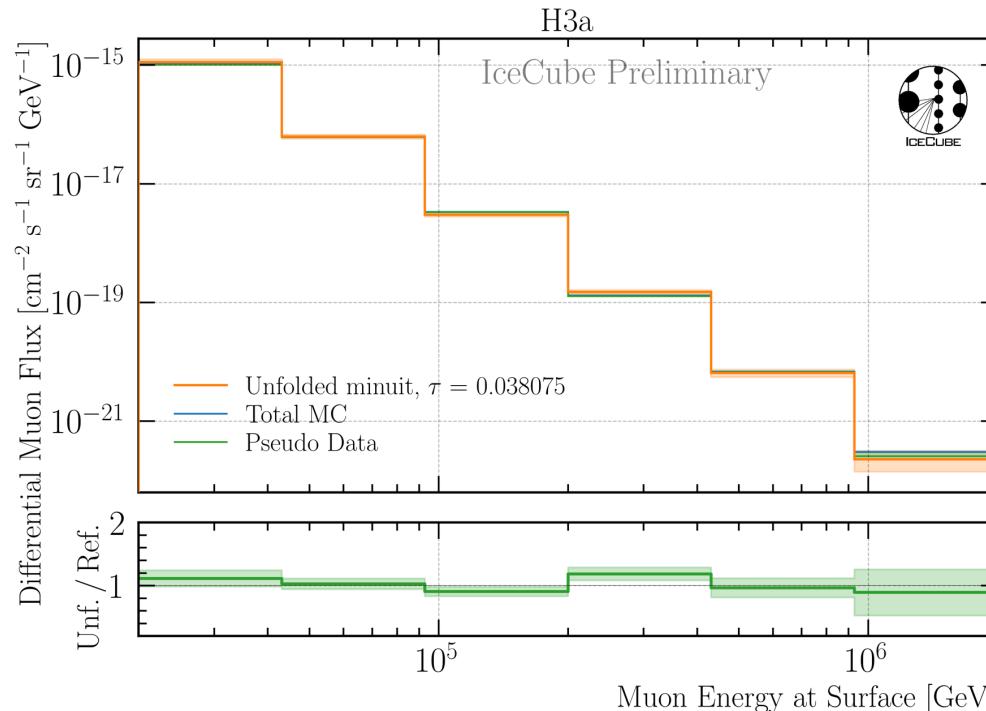
- Build unfolding matrix on H3a, H4a, GST & GSF
- Unfold H3a as test set
- Results are within uncertainties → robust



Determine Regularization

- Find regularization τ with minimal bin-to-bin correlation
- LLH minimization (unfolding) provides full covariance matrix V

➤ Minimize global correlation $\rho = \sum_{i > i} V_{ij}$

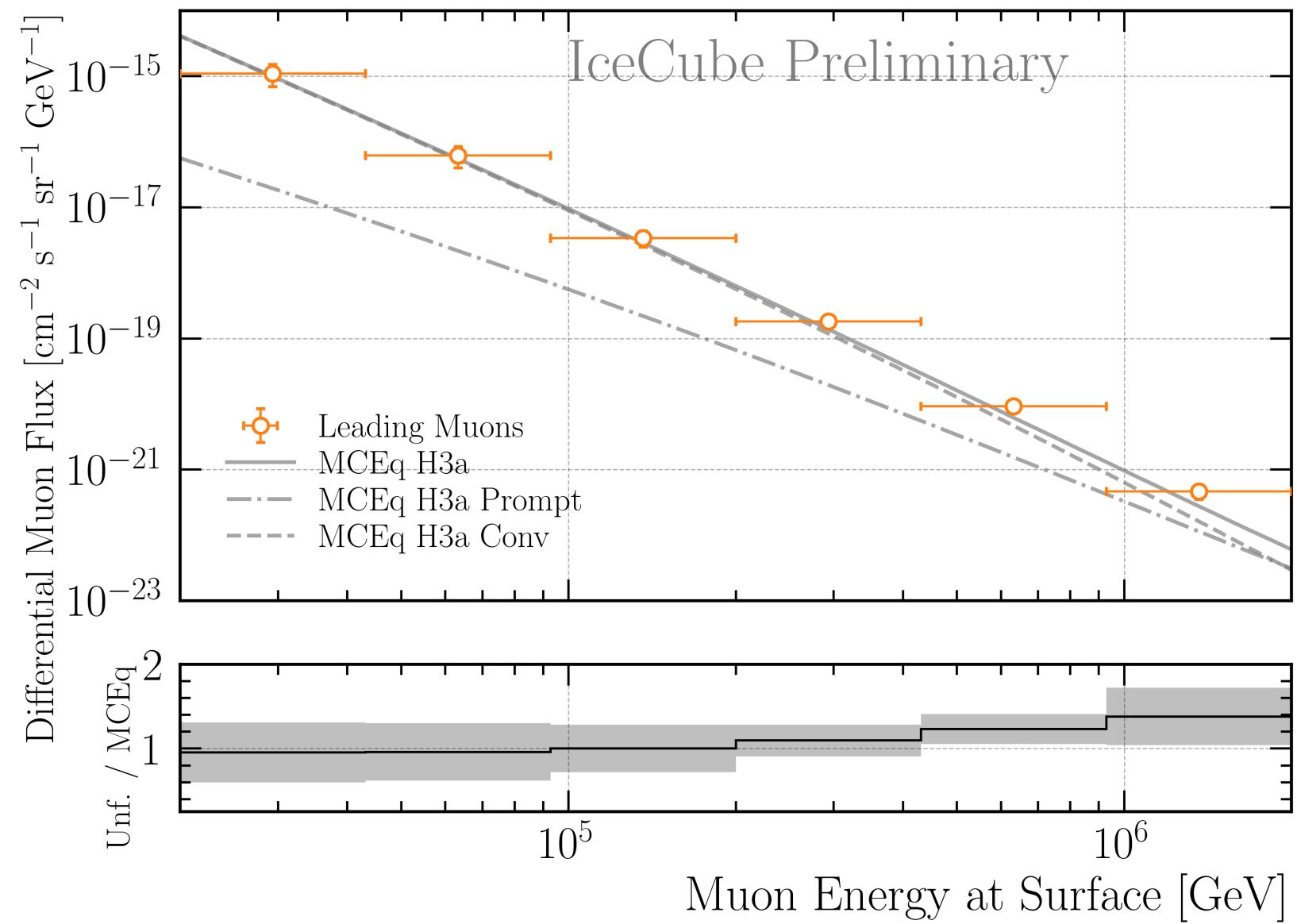


Burnsample

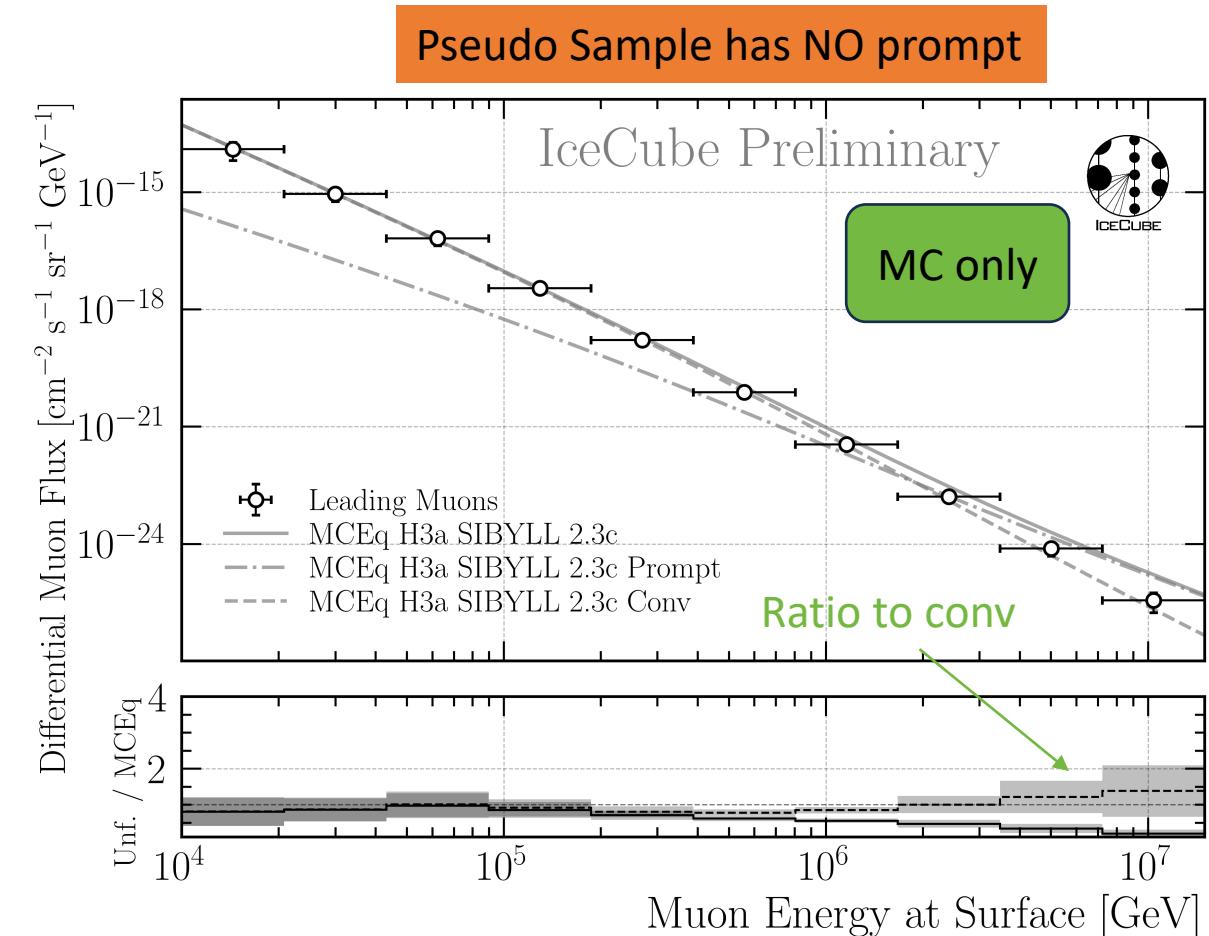
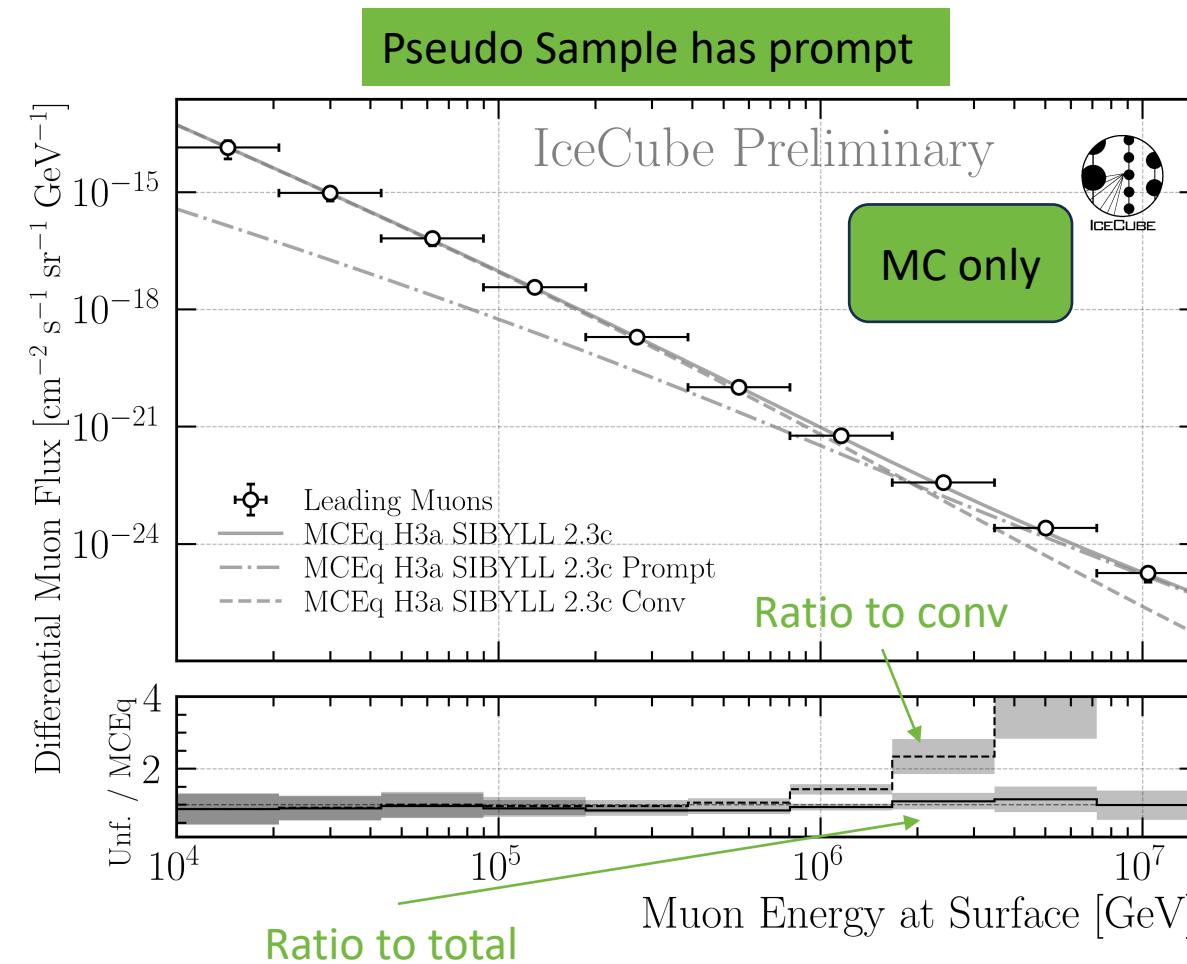
Burnsample Unfolding

Leading muons

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



Prediction: 12 Years of IceCube Data Unfolding



➤ Unfolding is sensitive to the prompt component

Conclusion & Outlook

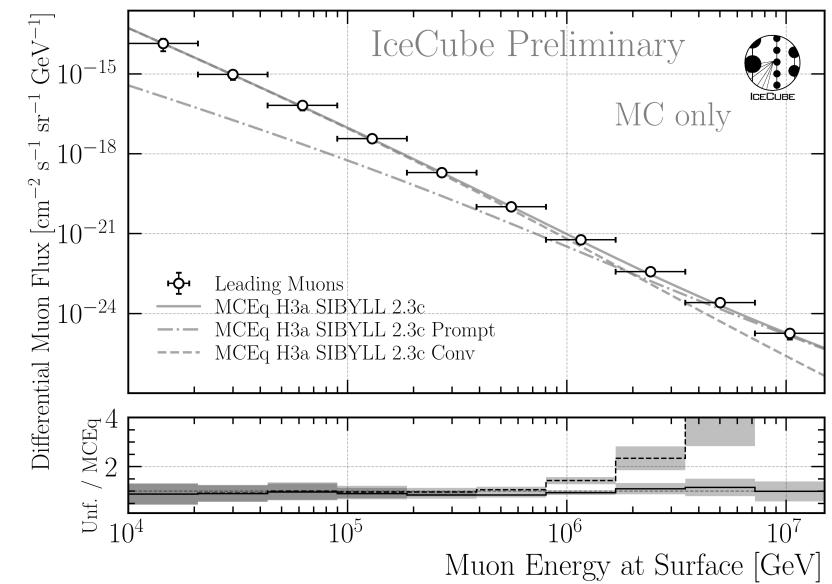
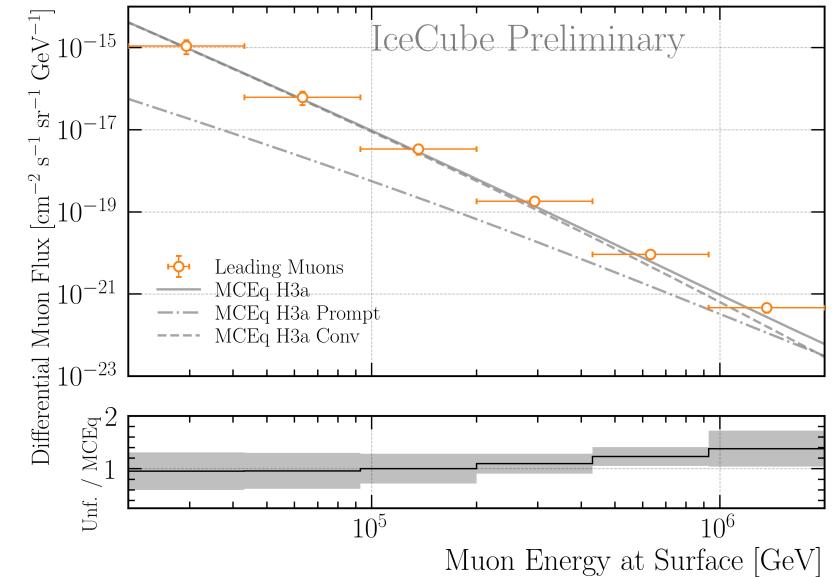
- New CORSIKA EHIST simulations
 - Prompt tagging agrees with MCEq
- DNN-based reconstructions
- Leading muon selection
- Burnsample unfolding agrees with MCEq prediction

Outlook

- Estimate unfolding uncertainty
- Unfold 12 years of data

Future

- Forward folding to fit prompt and conv normalization



Backup

Unblinding Proposal

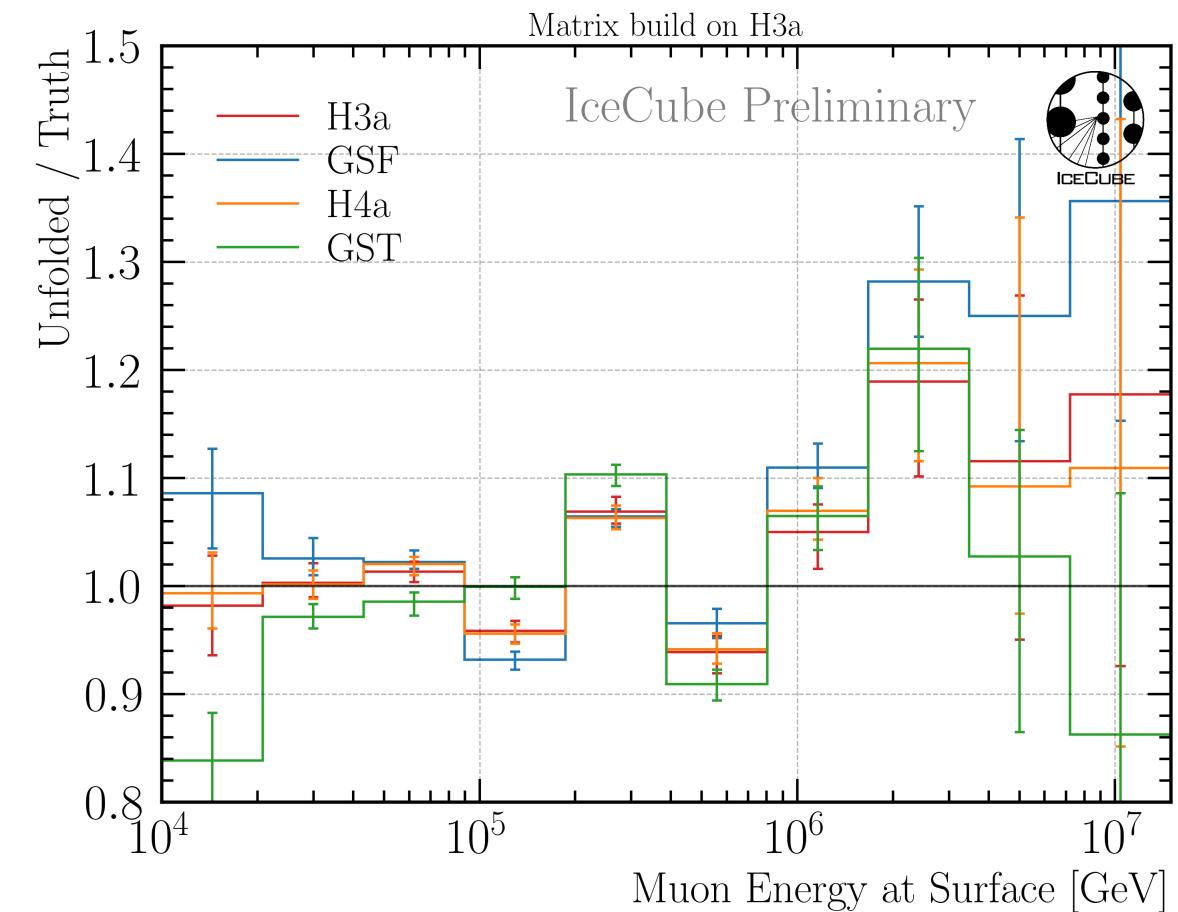
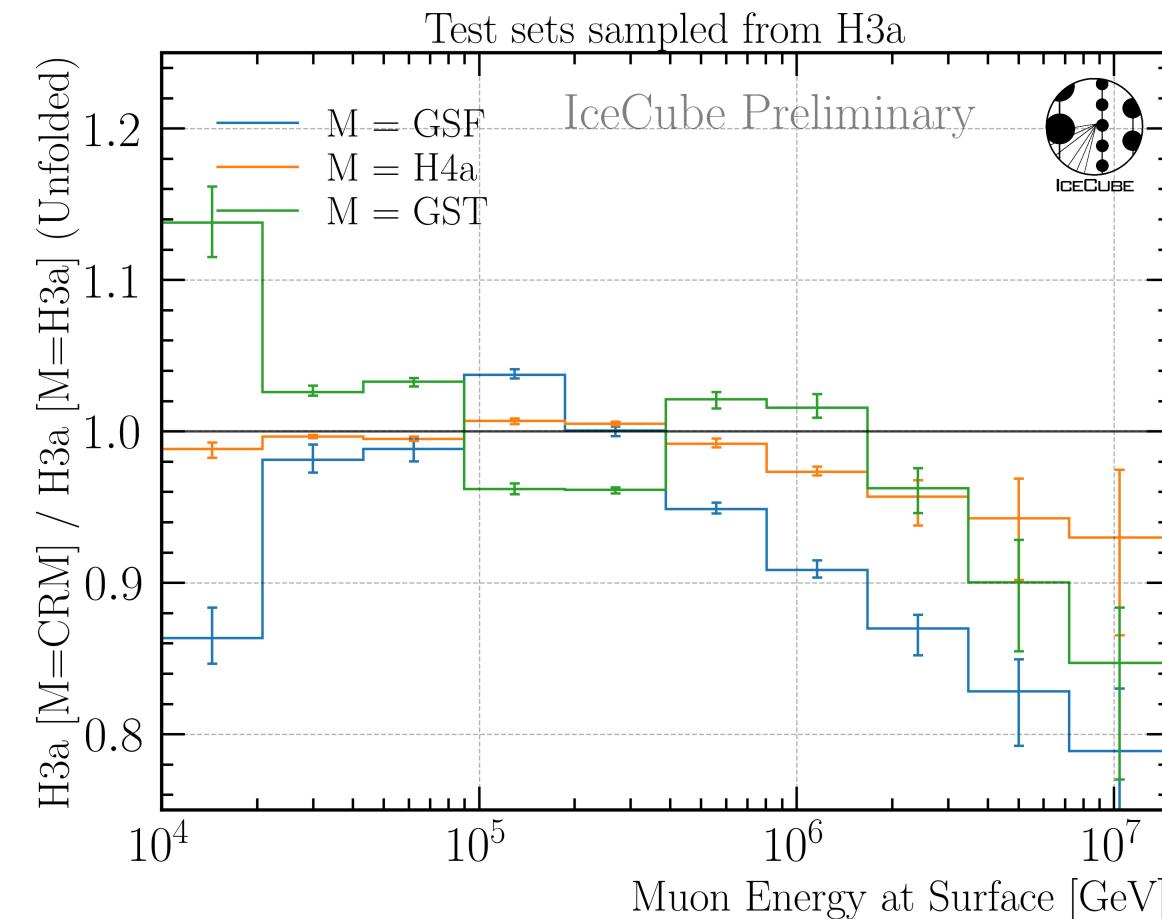
Unblinding plans

- Unfold 12 years of data for IC86 from 2011 to 2022
- Determine regularization strength on data
- Perform unfolding on data

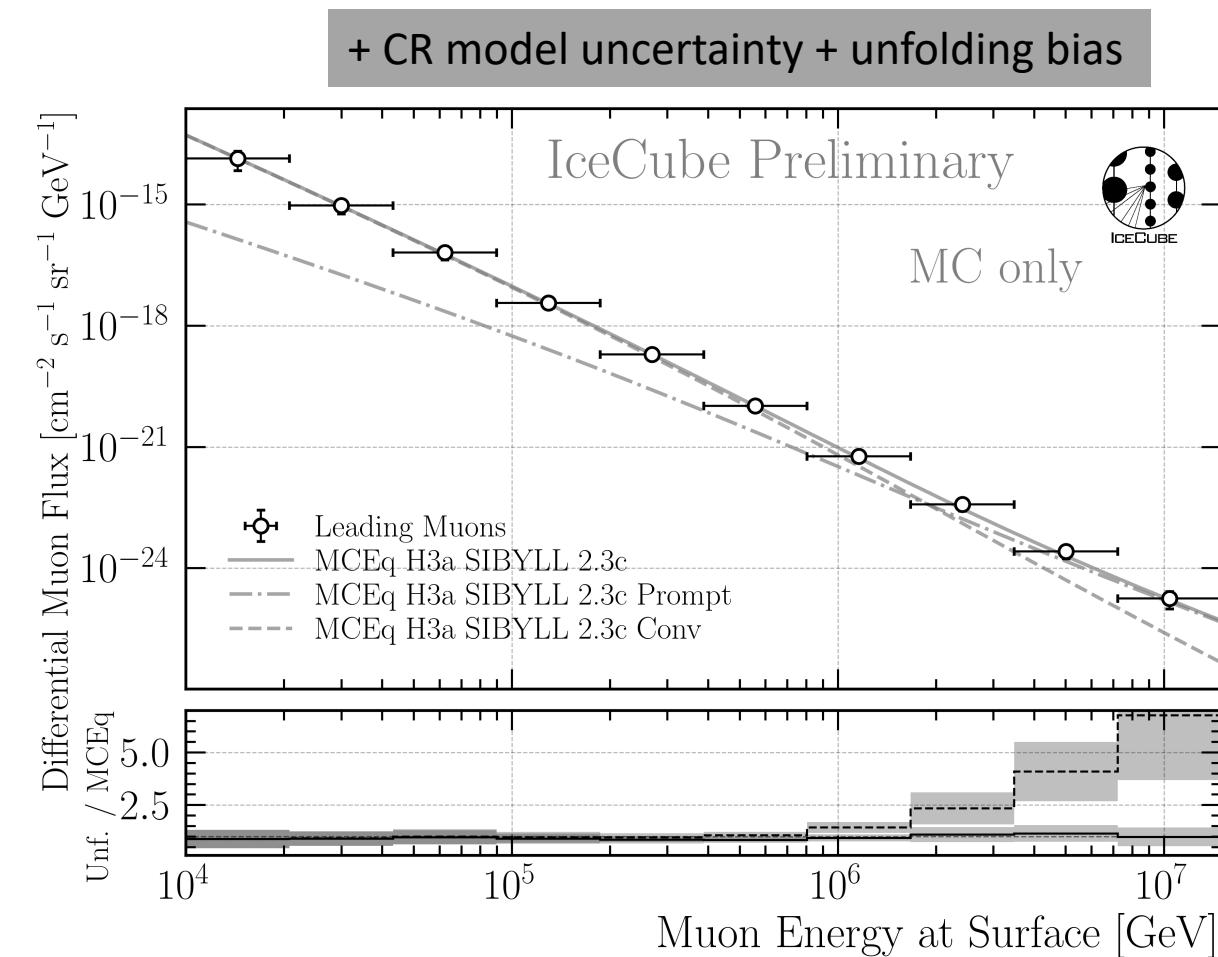
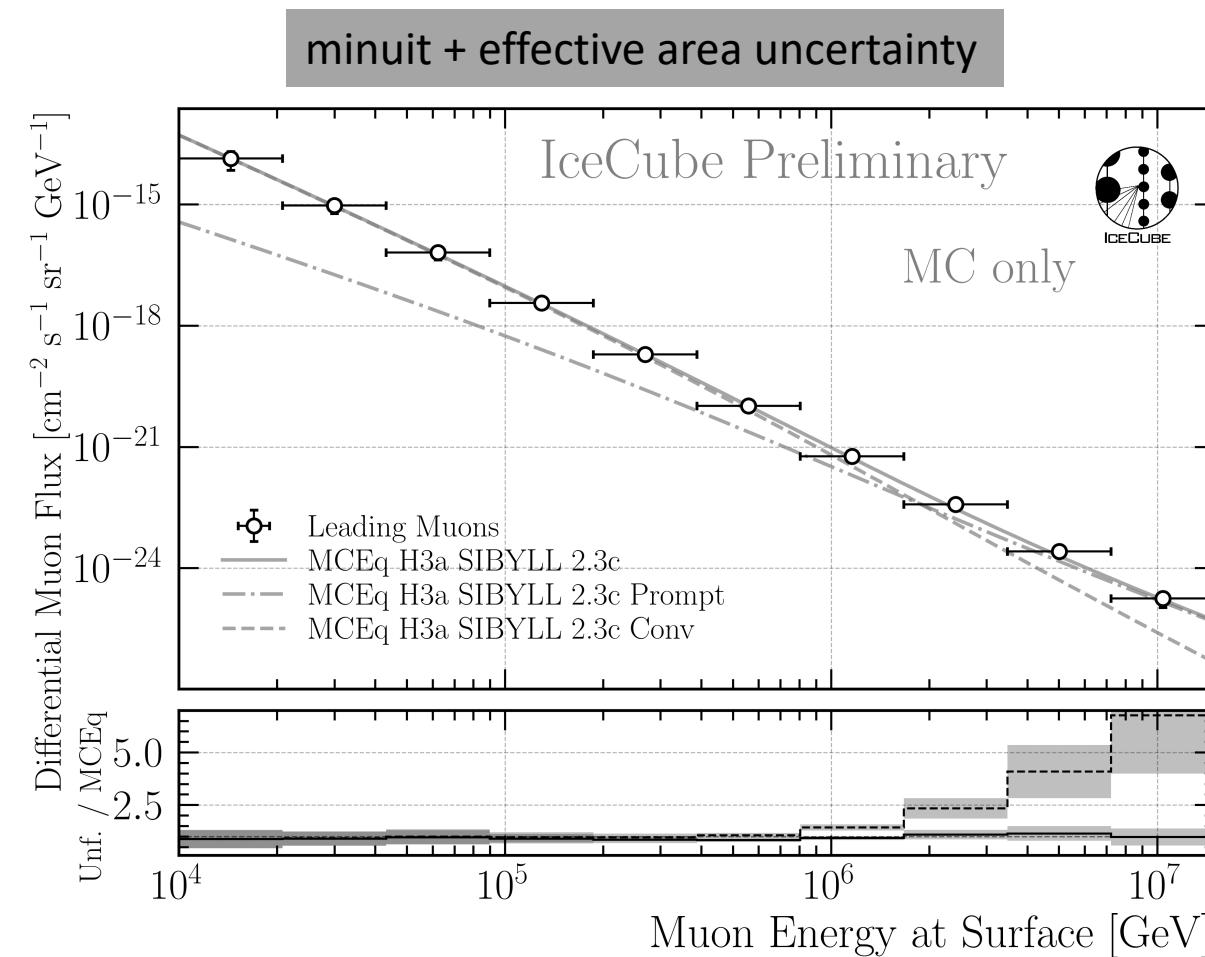
Post-unblinding checks

- Check the data-MC agreement for proxy variable (leading energy)
 - pre-fit (on the entire Snowstorm dataset)
 - post-fit (re-weight to fitted systematics)
- Divide total dataset in 3 subsets á 4 years to analyze systematic impacts over the years
- Unfold spectrum with all 4 CR model matrices

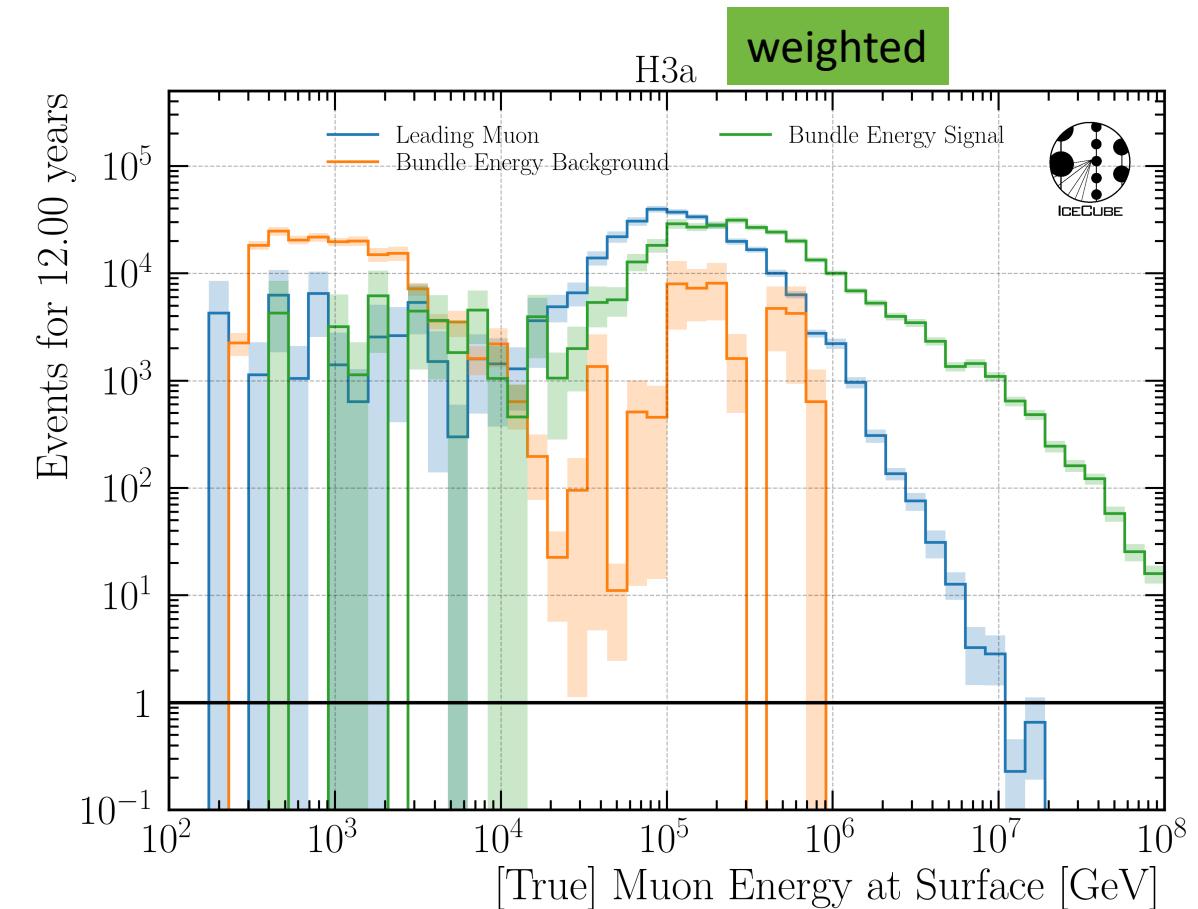
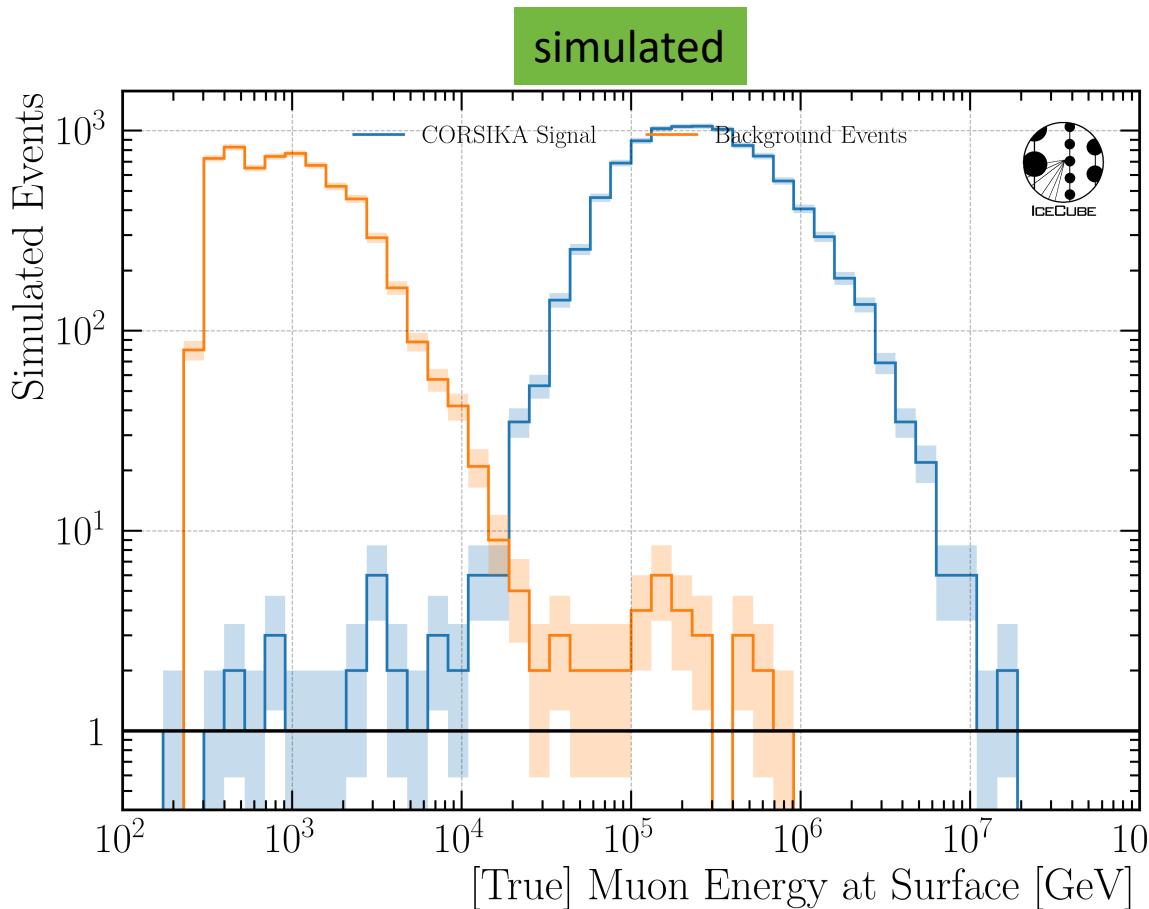
CR Model Impact and Unfolding Bias



12 Years Prediction: All Uncertainties

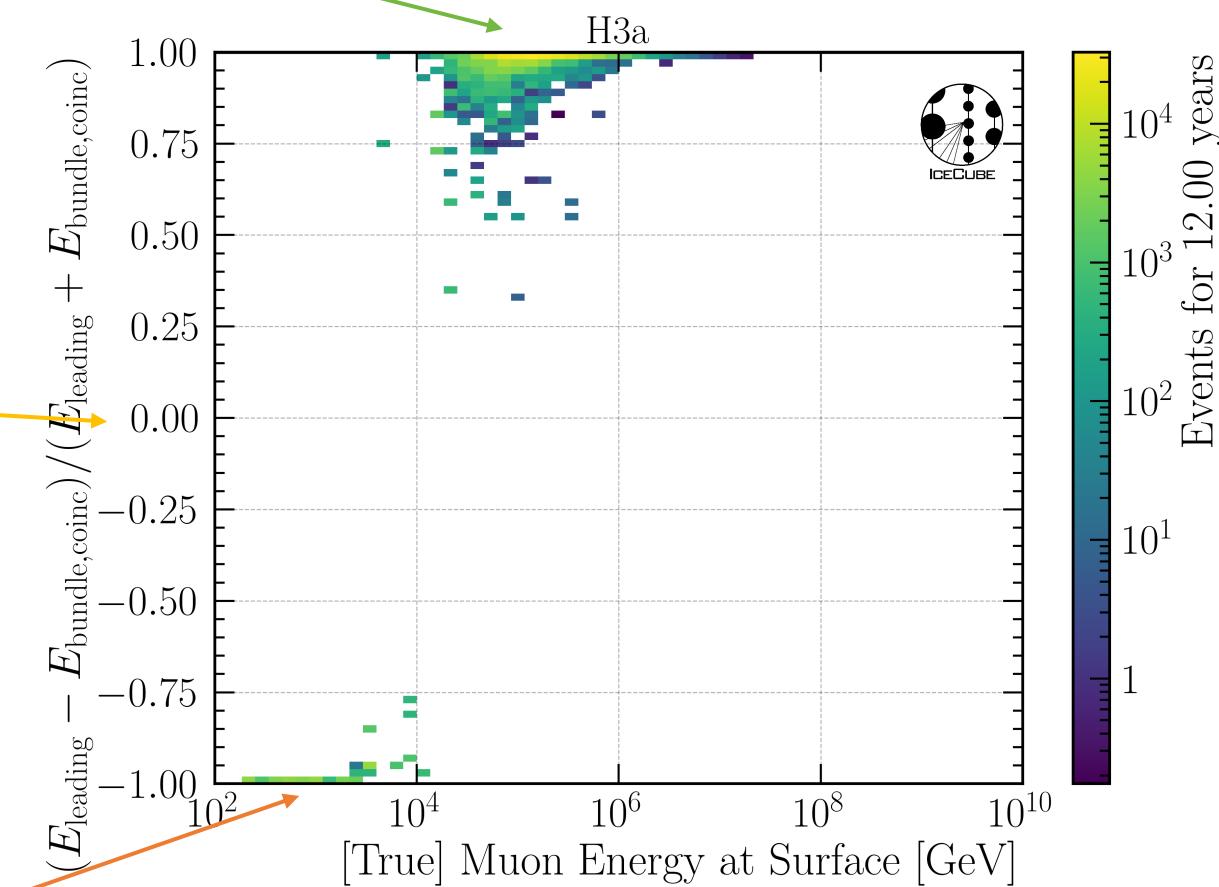
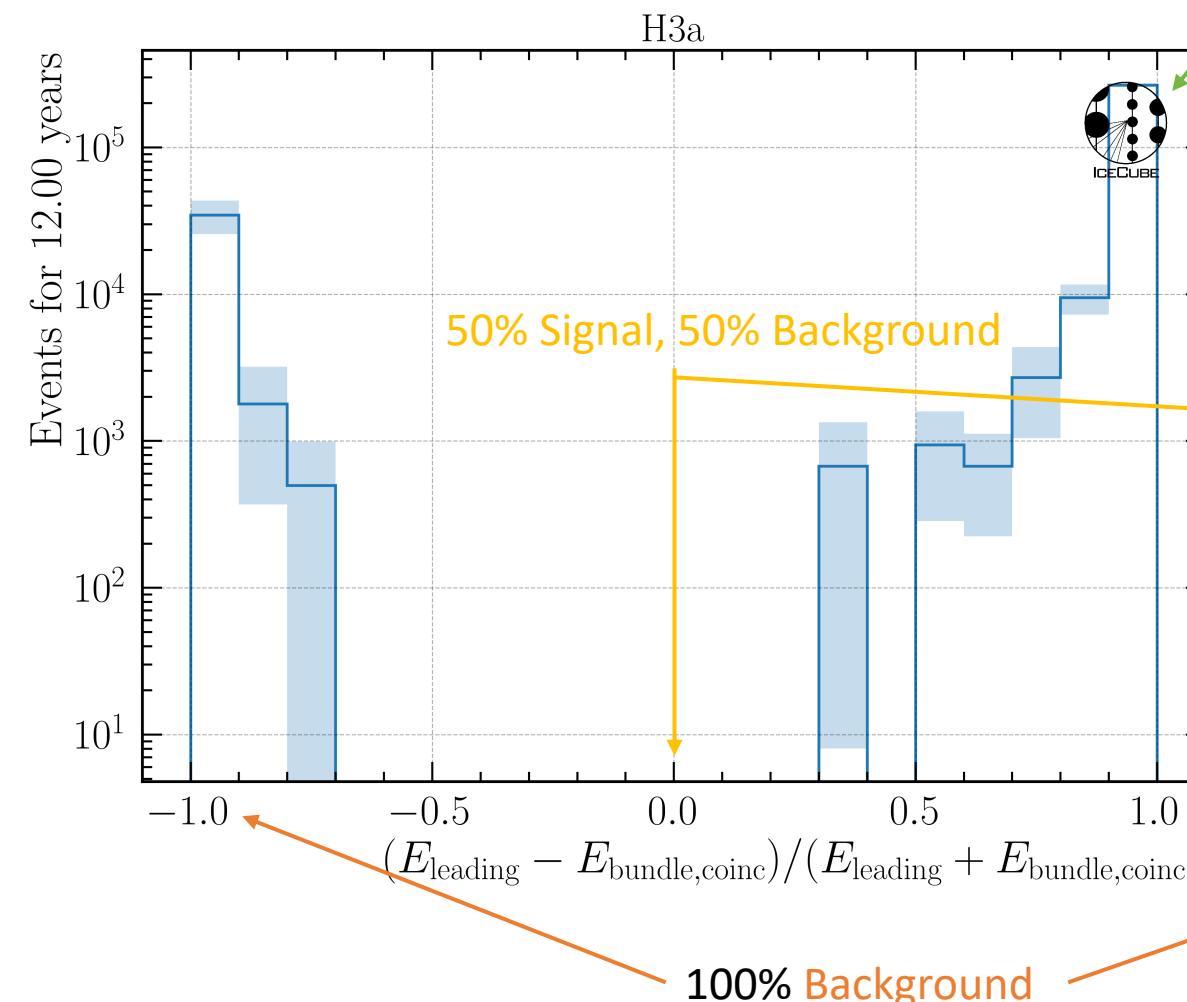


Muon Energy at Surface



- Leading Muon: most energetic muon in tree
- Bundle Energy Signal: sum of all muons of "signal" primary
- Bundle Energy Background: sum of all muons of coincident primaries

Coincidence: Muons on Event Level



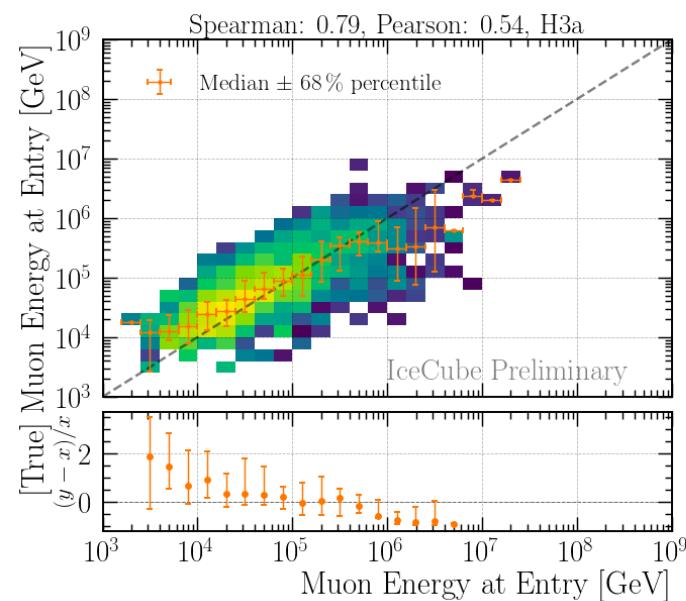
➤ above 10 TeV, no background dominated events

Estimate Rates with H3a

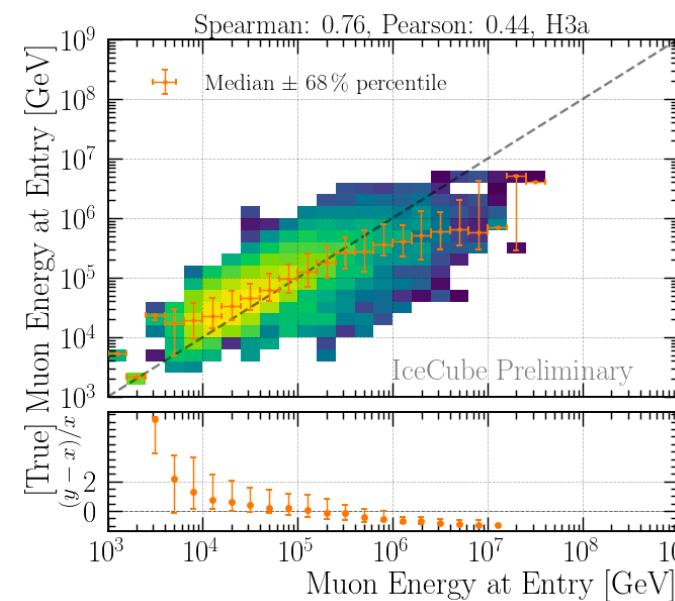
- Signal rate: 0.84 mHz (total event rate)
 - Background rate: 0.68 mHz (event rate with at least 1 coincident primary)
 - Signal rate: 0.74 mHz (leading muon energy at surface > 10 TeV)
 - Background rate: 0.58 mHz (event rate with at least 1 coincident primary & lead. muon E. > 10 TeV)
 - Signal rate: 0.74 mHz (leading muon energy at surface > 10 TeV)
 - Background rate: 0.02 mHz (event rate with at least 1 coincident primary & lead. muon E. > 10 TeV & coincident muon bundle energy at surface has at least 10% of leading muon energy)
- The light of the background events overlaps with the signal → no chance to separate
- A 10% bundle energy contribution would shift the measured light up by roughly 10% → within the uncertainties of the energy prediction
- Networks have been trained on MC with coincident events → they are able to subtract a little light in case they assume there is a background event, however, this is not quantified

Reconstructions: Coincident Events

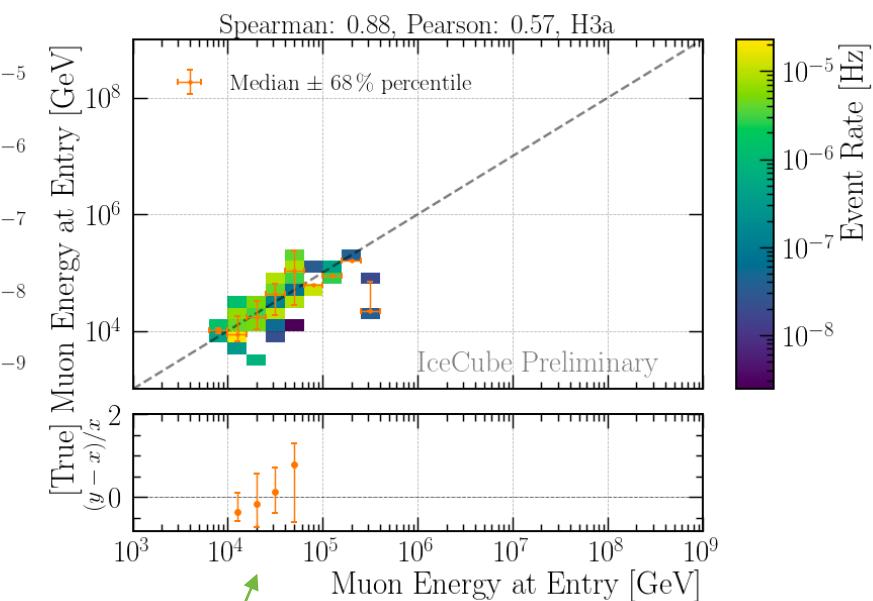
no background event



$\frac{\text{bundle energy at surface}}{\text{leading energy at surface}} < 0.1$

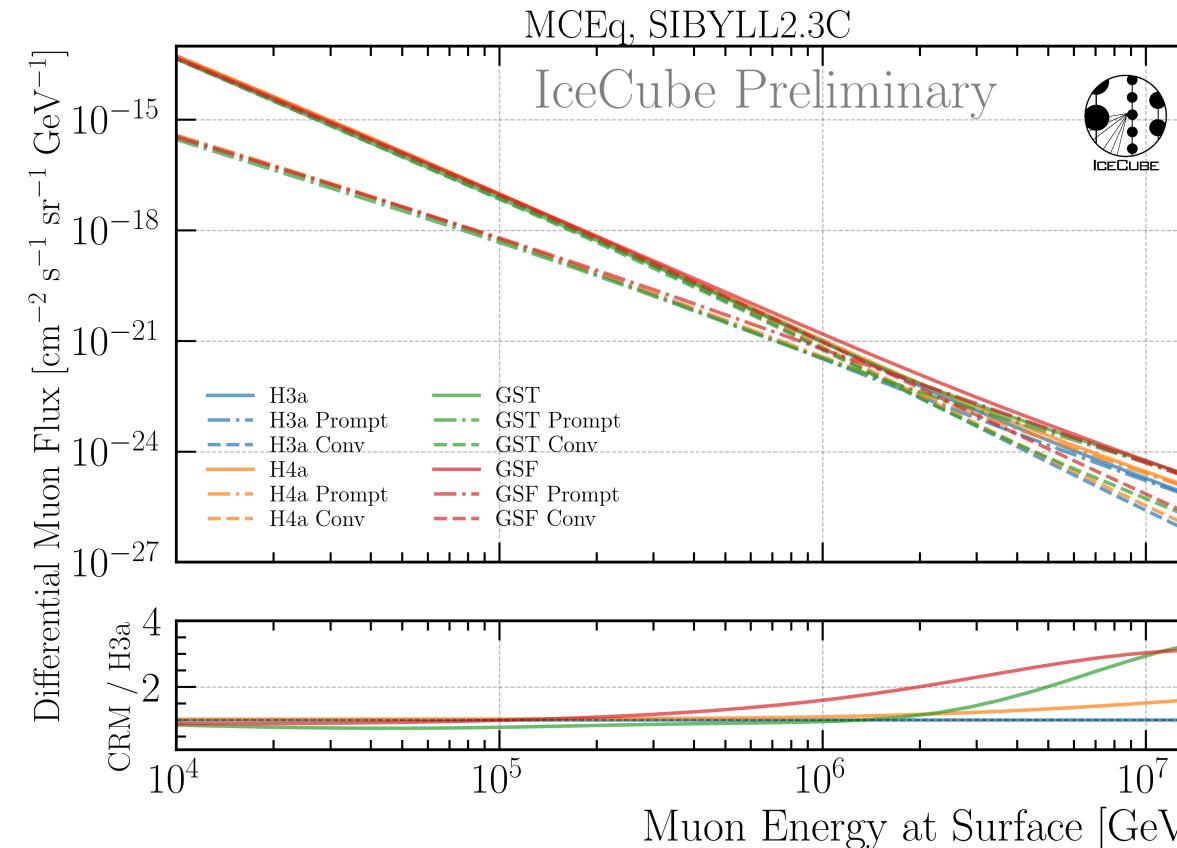
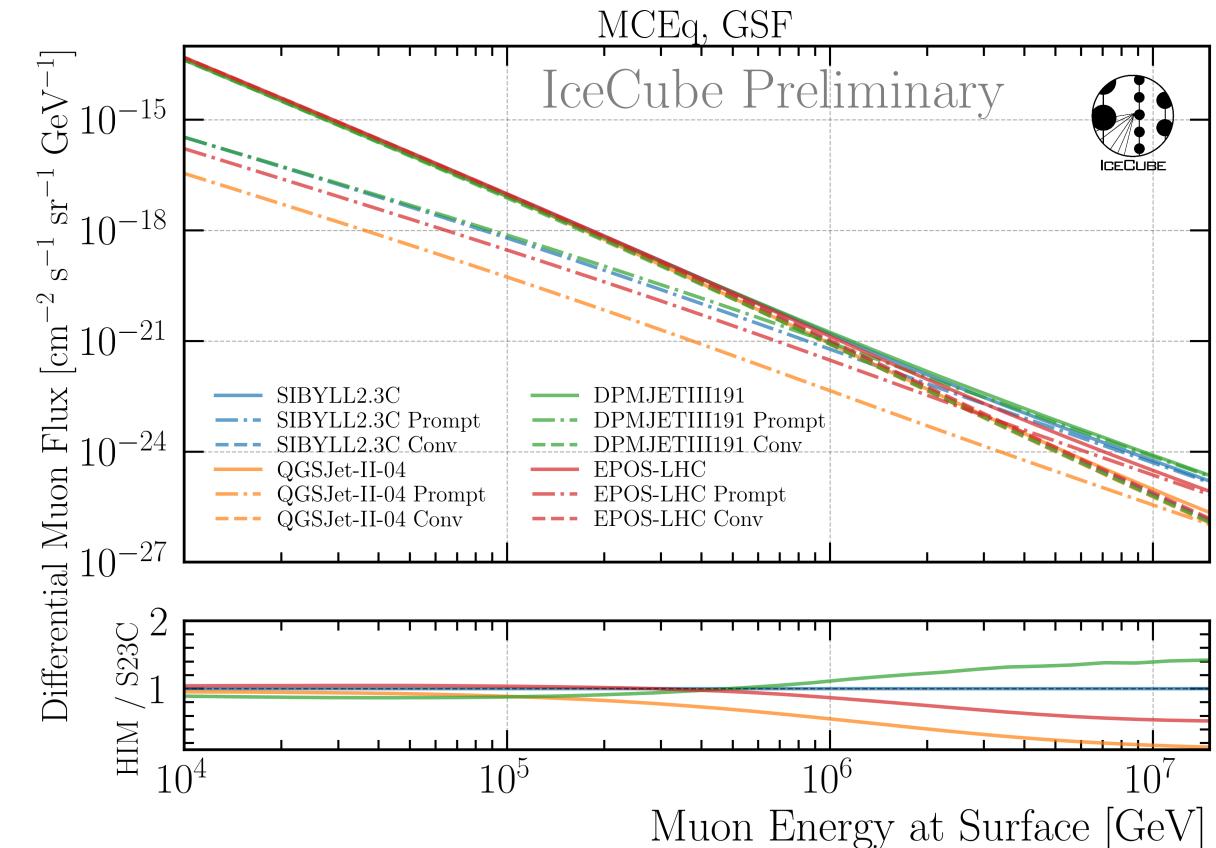


$\frac{\text{bundle energy at surface}}{\text{leading energy at surface}} > 0.1$



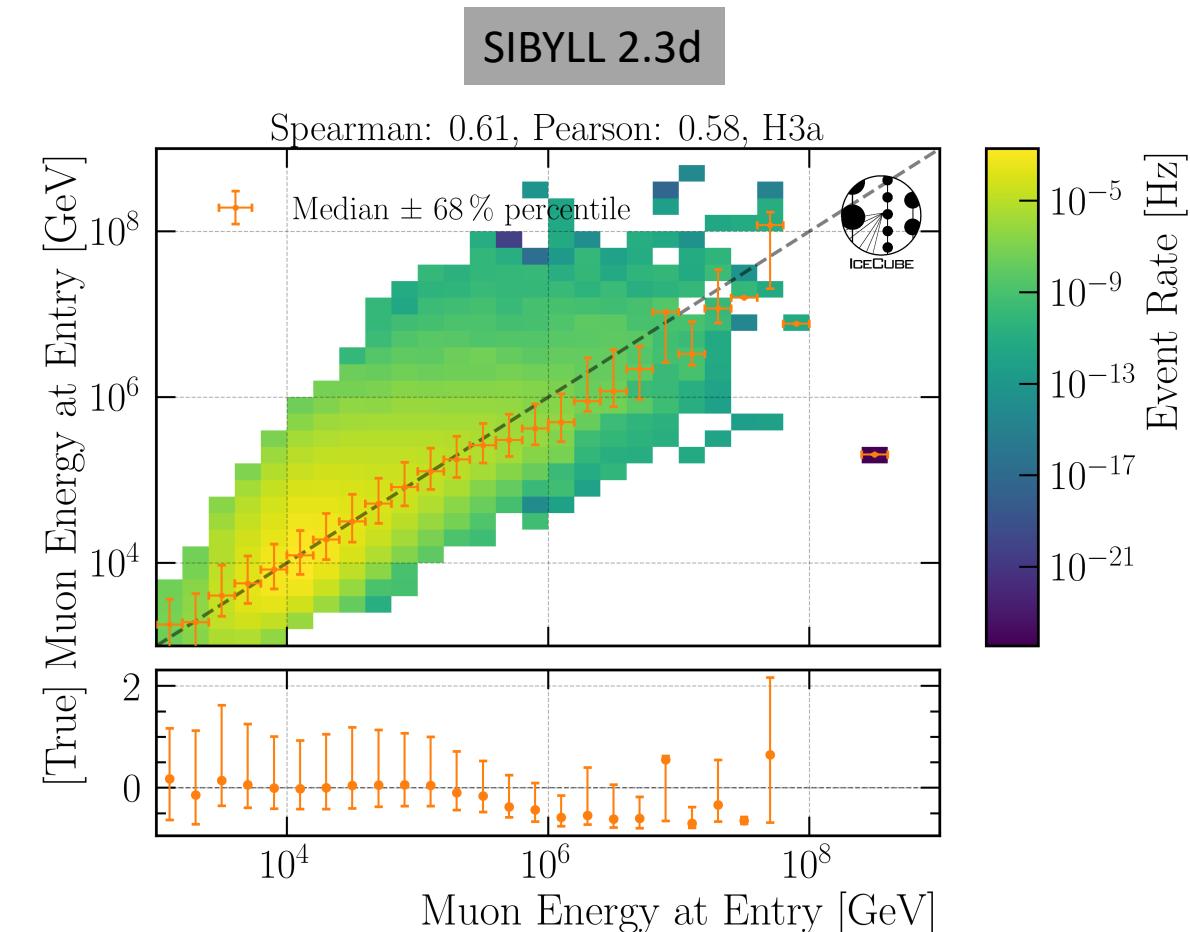
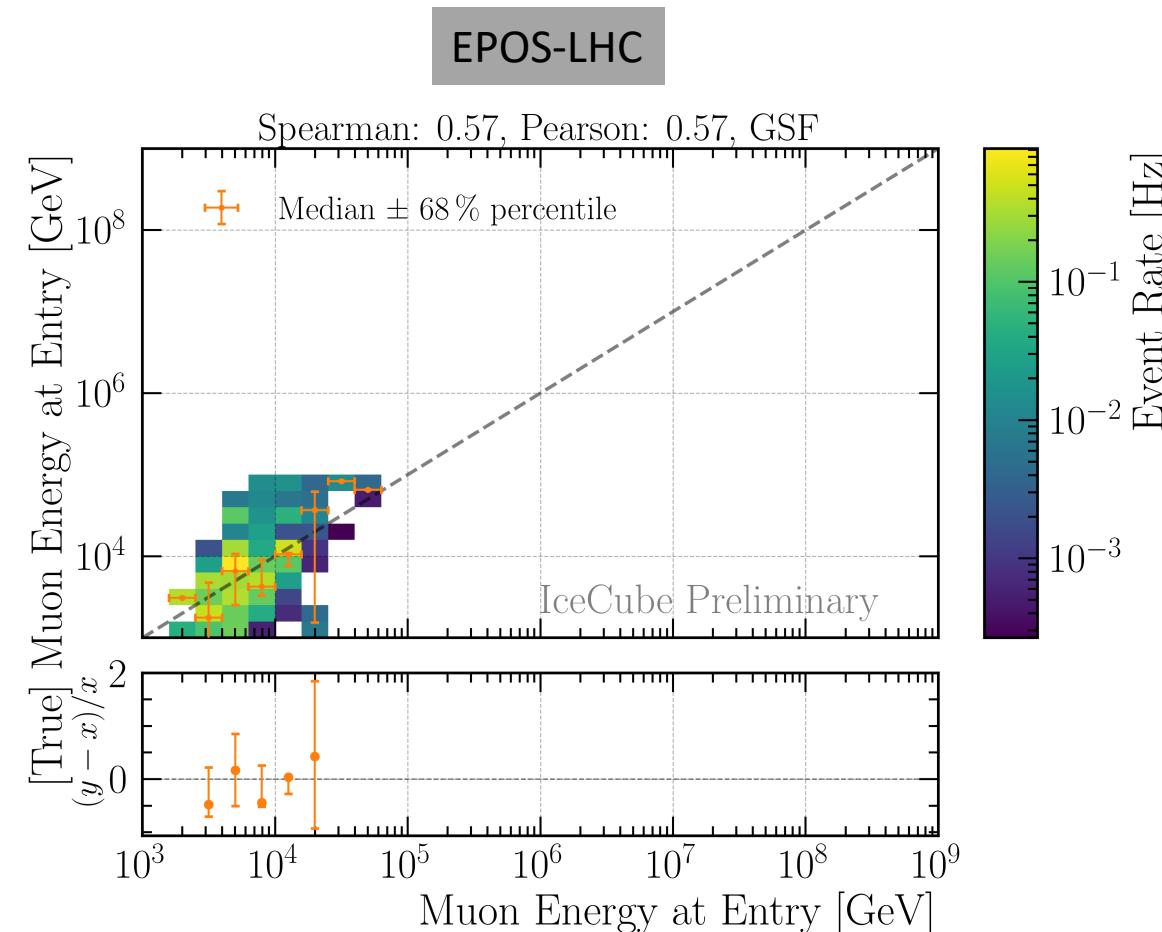
- sufficient energy reconstruction for events with background

Estimate Impact of Hadronic Interaction Model

Different CR Models**Different Hadronic Models**

- Not all hadronic models include charm
- Impact of HIM smaller than CRM
 - CRM impact below 3%
 - Hadronic Model impact is negligible

Reconstruction: EPOS-LHC vs SIBYLL 2.3d [Level 5]



- IceTop simulations for proton and iron
- 23198 and 23201
- Primary energy $> 1\text{e}7$

Event Rate to Flux → Effective Area

- Unfolding estimates an event rate
- Transfer event rate to flux

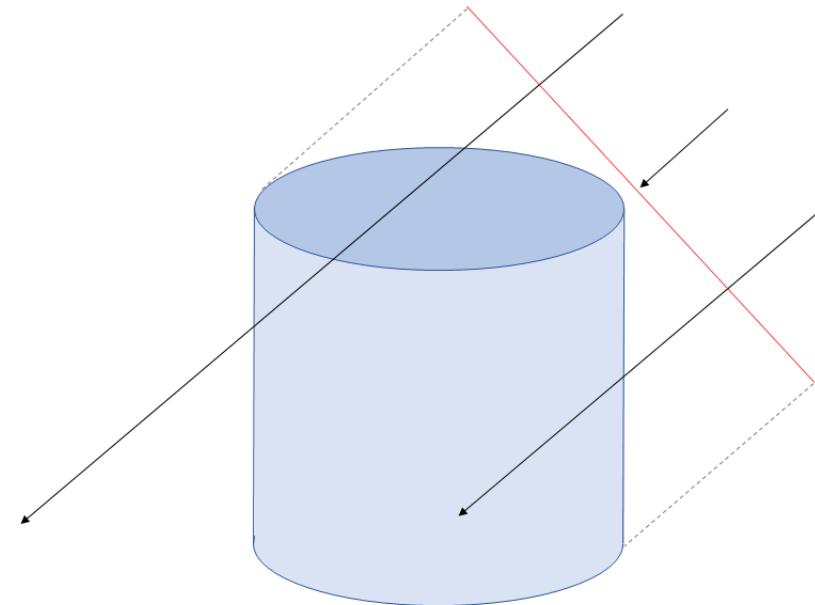
$$\Phi_i = \frac{N_i}{T \cdot \Delta L_i \cdot \Omega_i \cdot A_{\text{eff},i}}$$

- with solid angle

$$\Omega_i = 2\pi \cdot (\cos \Theta_{\min,i} - \cos \Theta_{\max,i})$$

- and effective area

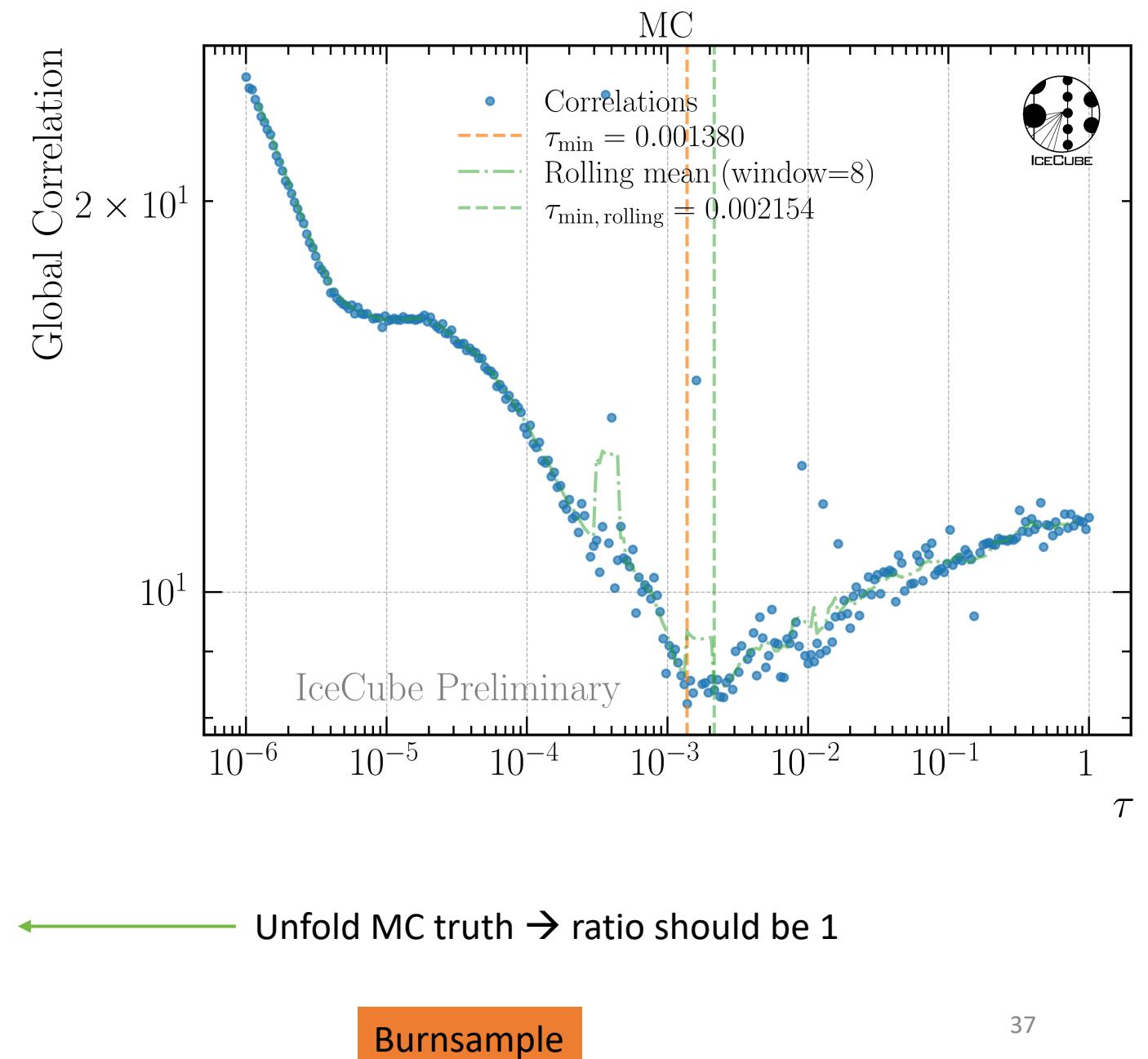
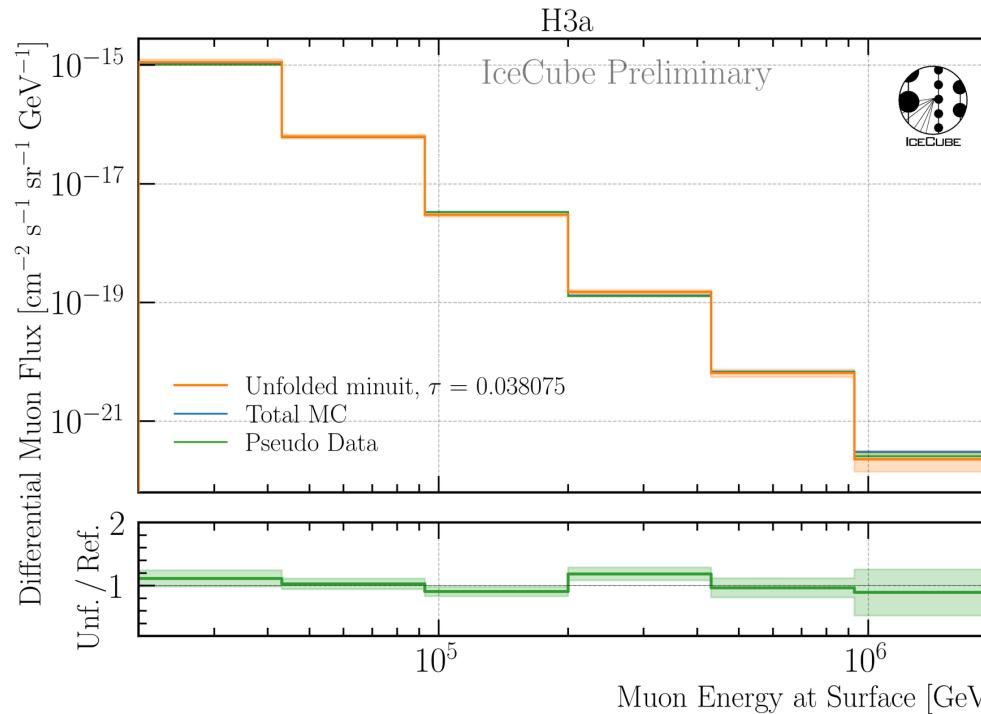
$$A_{\text{eff}} = A_{\text{sim}} \frac{N_{\text{sel}}}{N_{\text{gen}}}$$



Determine Regularization

- Find regularization τ with minimal bin-to-bin correlation
- LLH minimization (unfolding) provides full covariance matrix V

➤ Minimize global correlation $\rho = \sum_{i > i} V_{ij}$



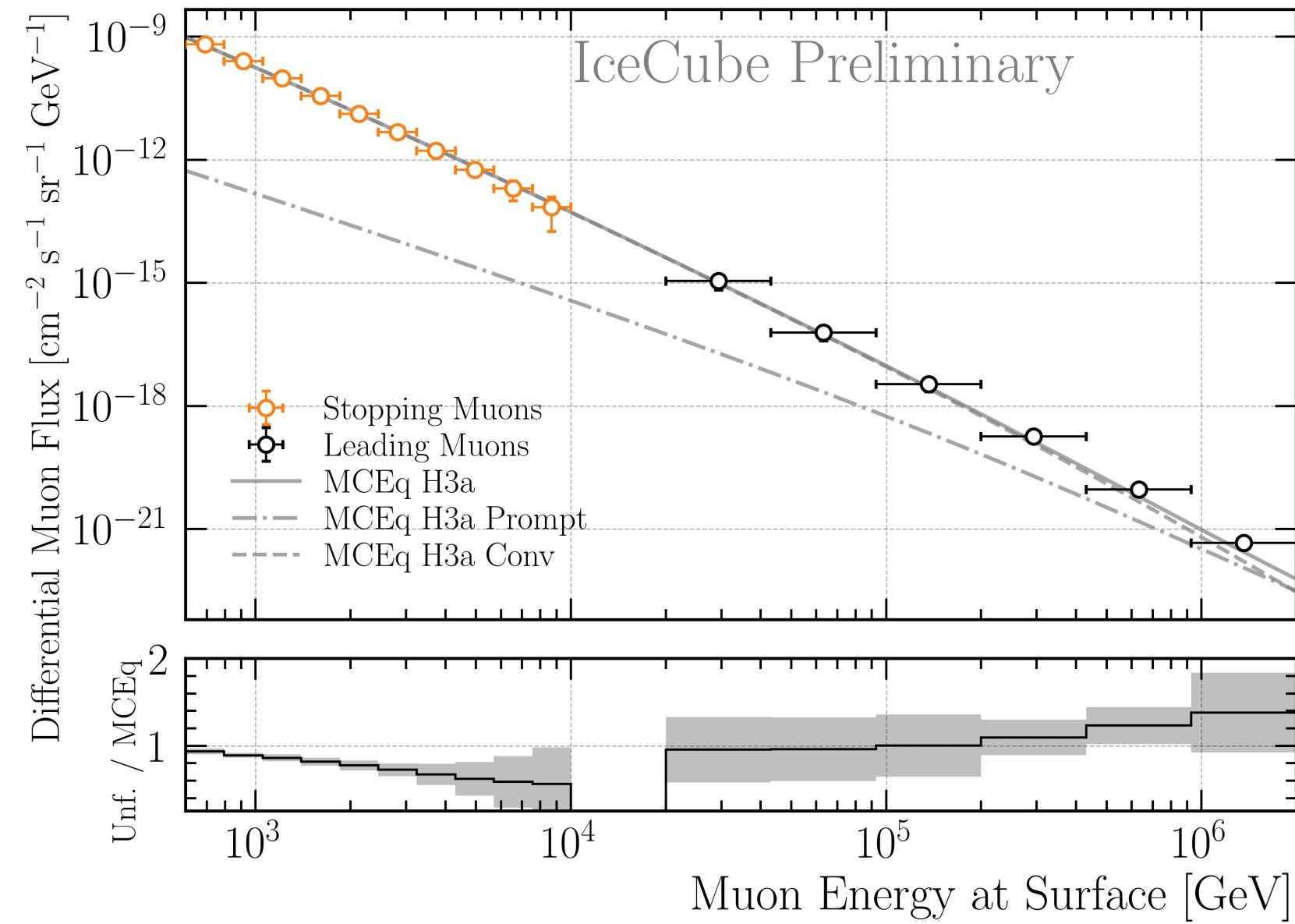
Muon Flux Unfolding

Leading muons

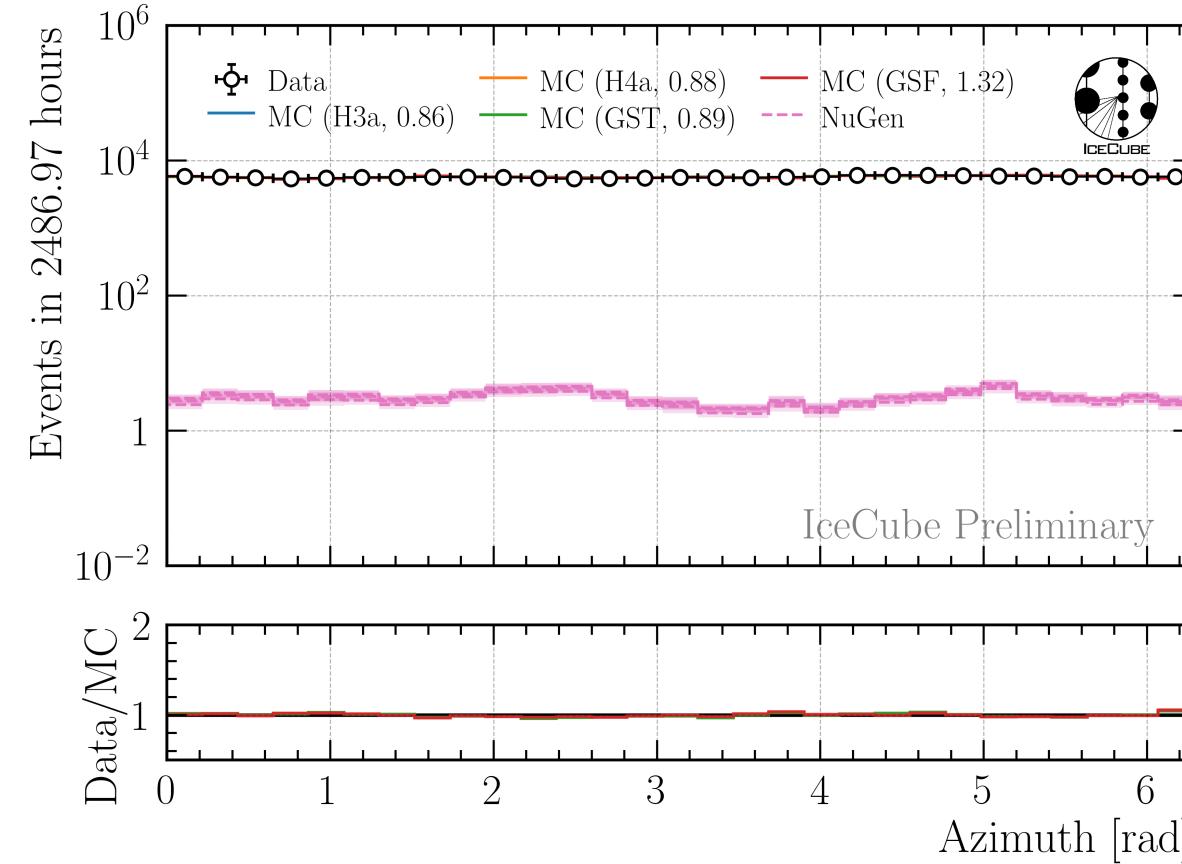
- 2487 h IceCube data
- 12754 events
- Agrees with MCEq

Stopping muons

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

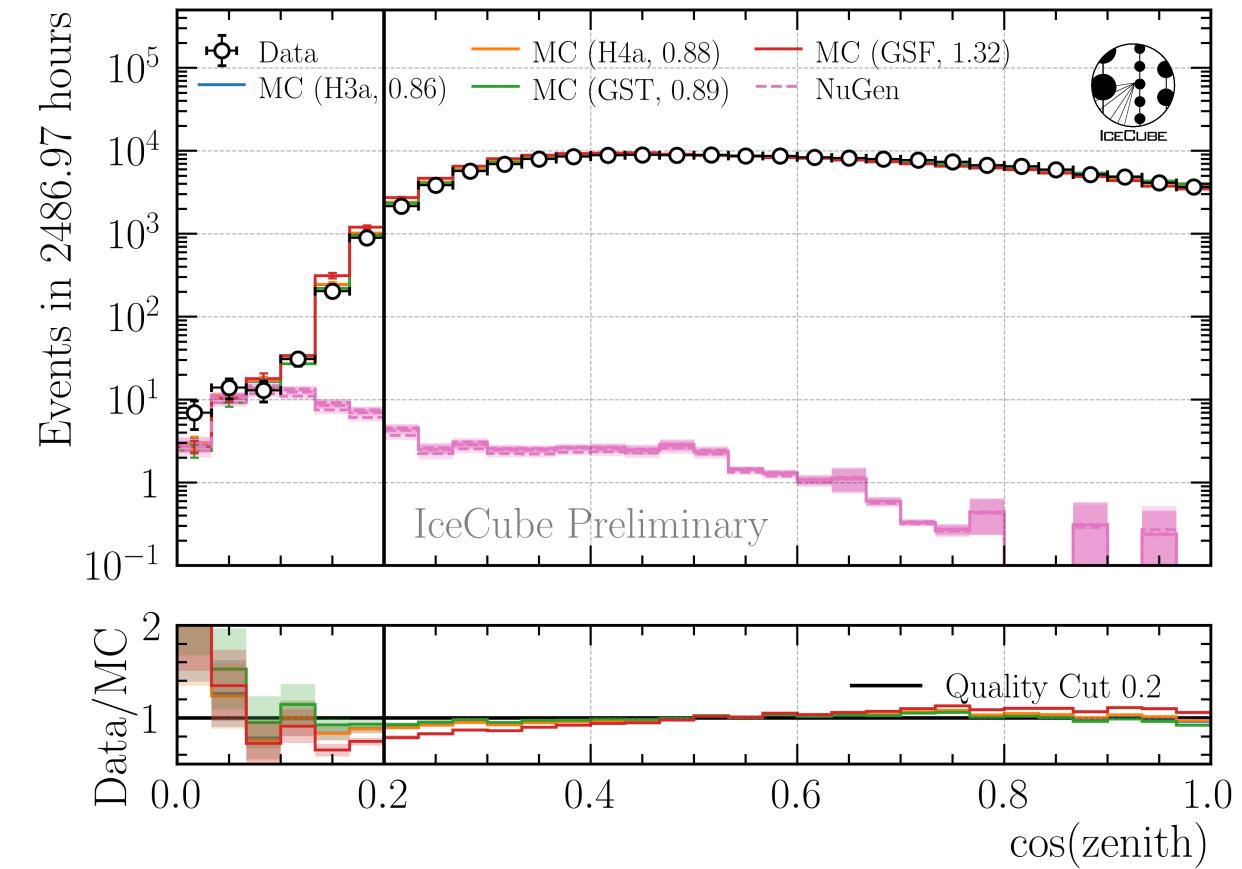


Level 5: Data/MC Quality Cuts



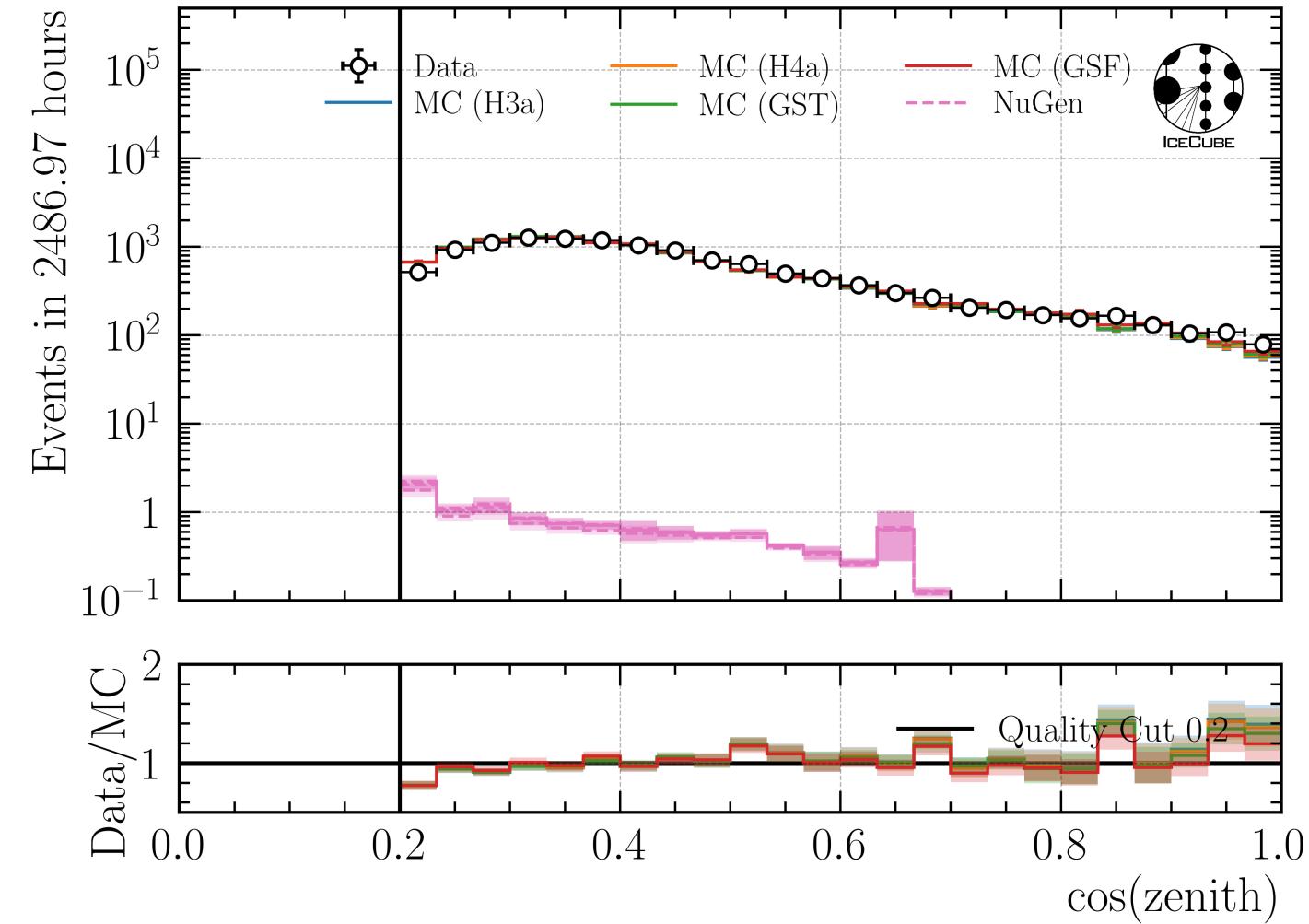
Neutrino weighting:
SPL: $n = 1.8$, $\gamma = 2.52$

IceCube Collaboration PoS ICRC2023 1064

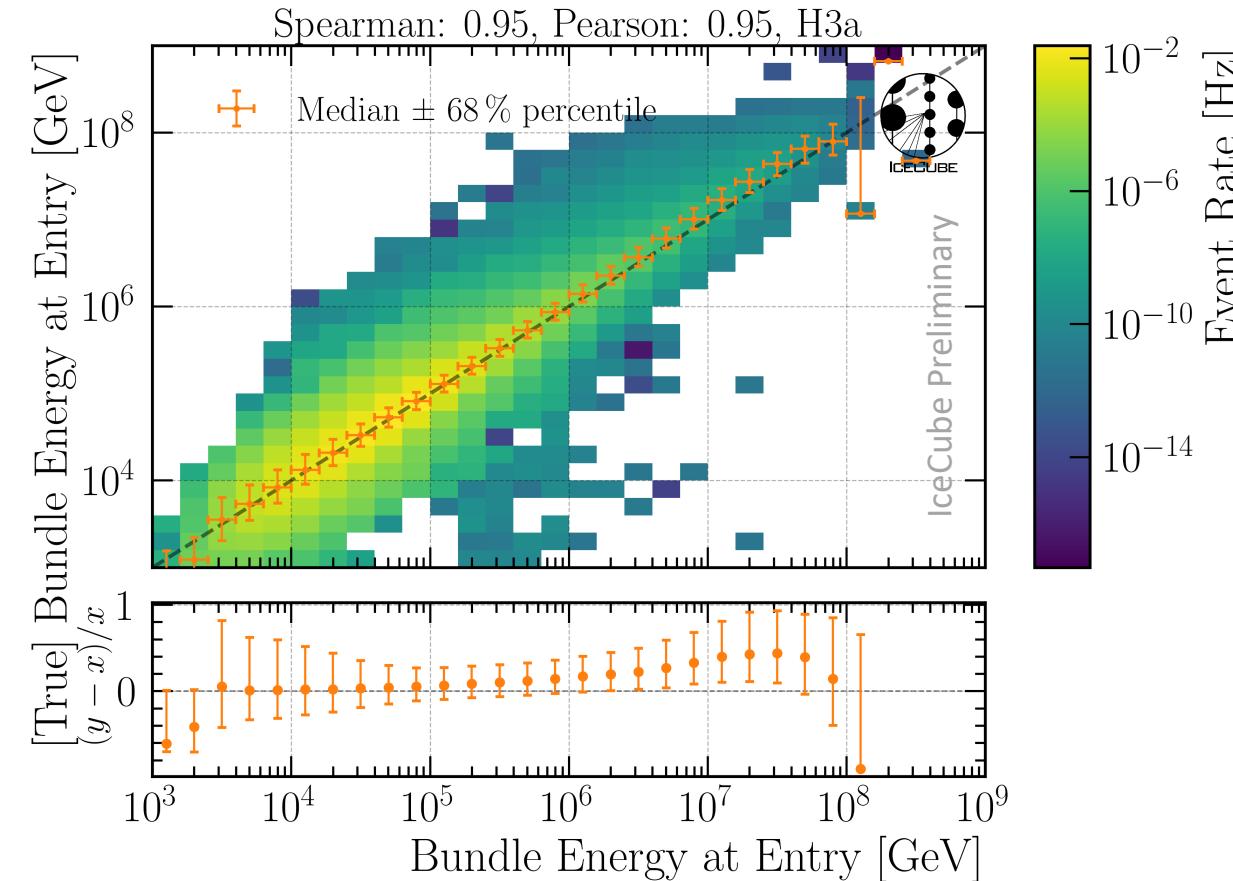
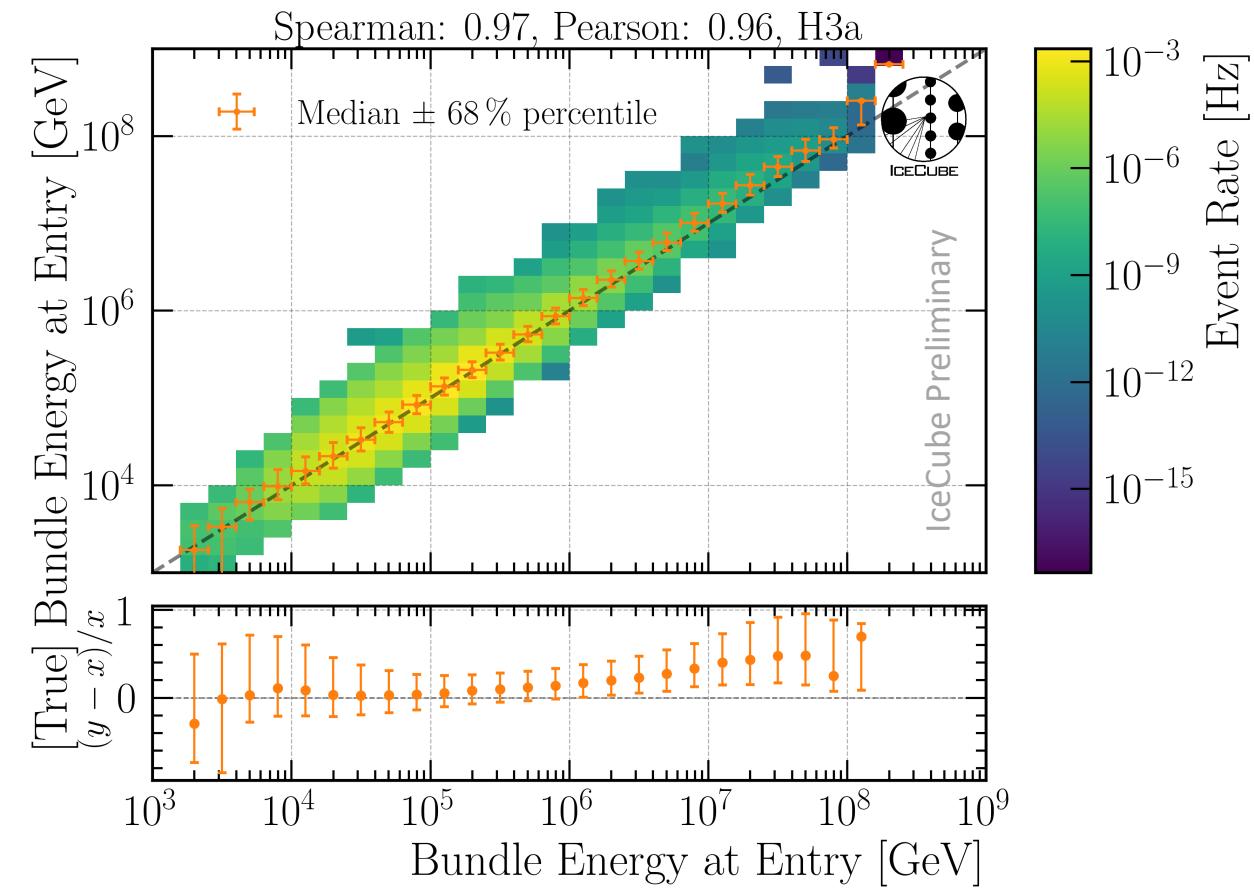


- Good directional reconstruction
- Cut: $\cos(\text{zenith}) > 0.2$

Final Level cos(zenith)



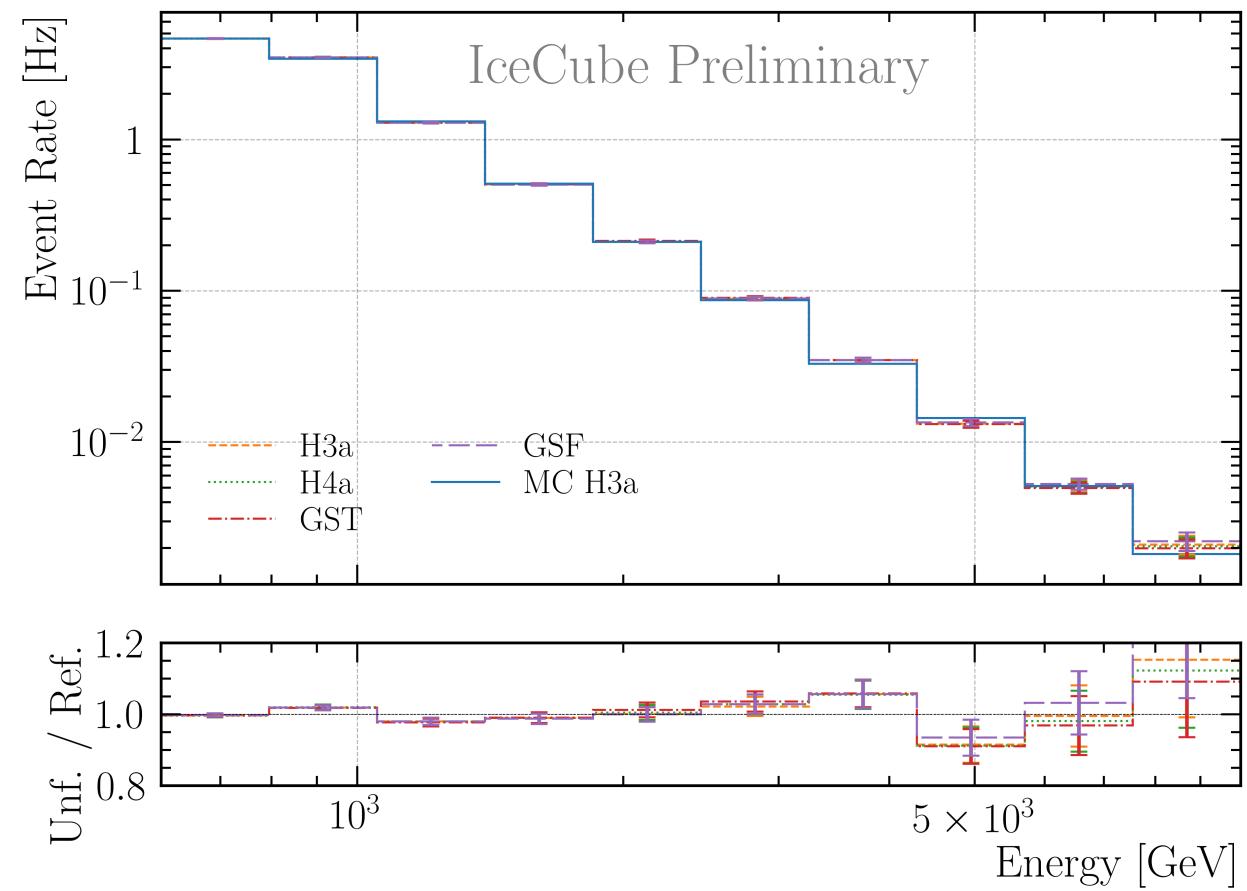
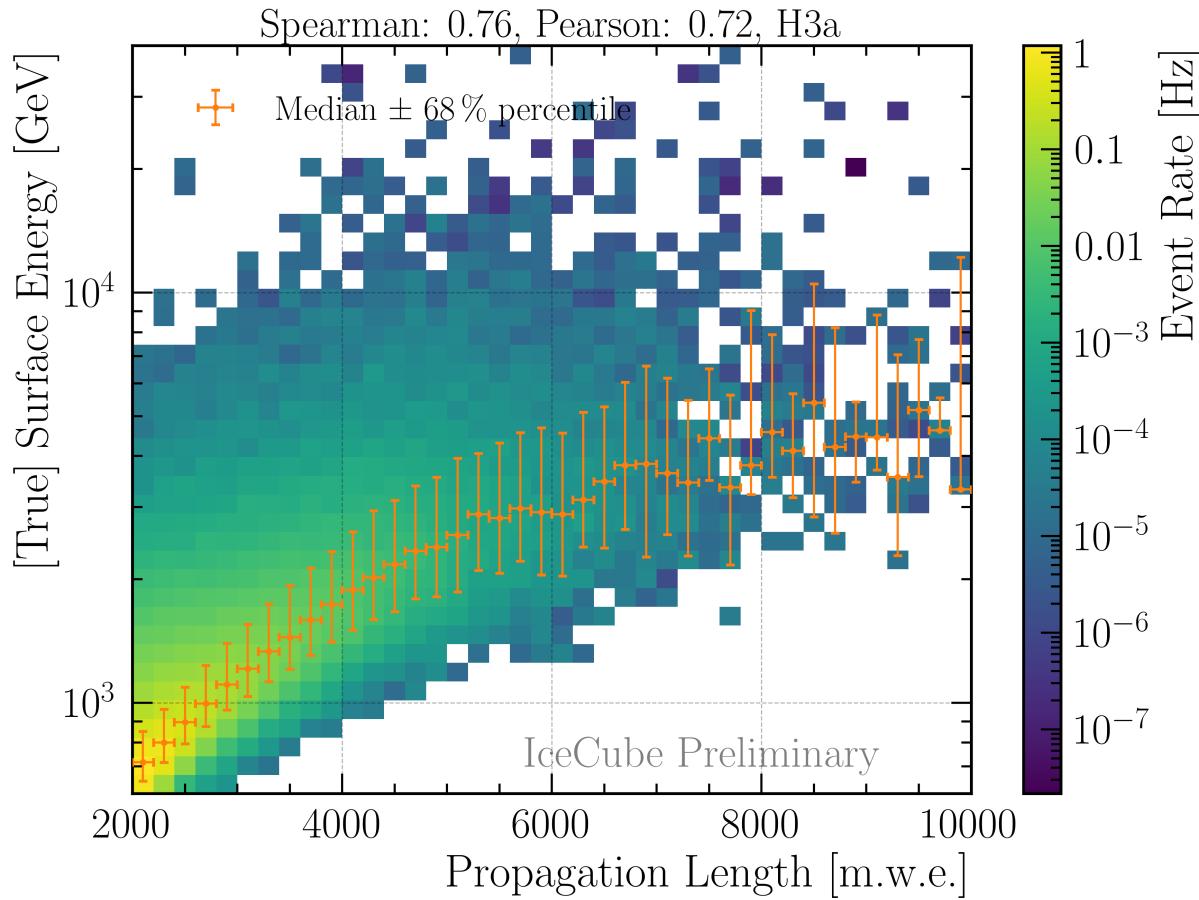
Bundle Energy Reconstruction

Level 4**Level 5**

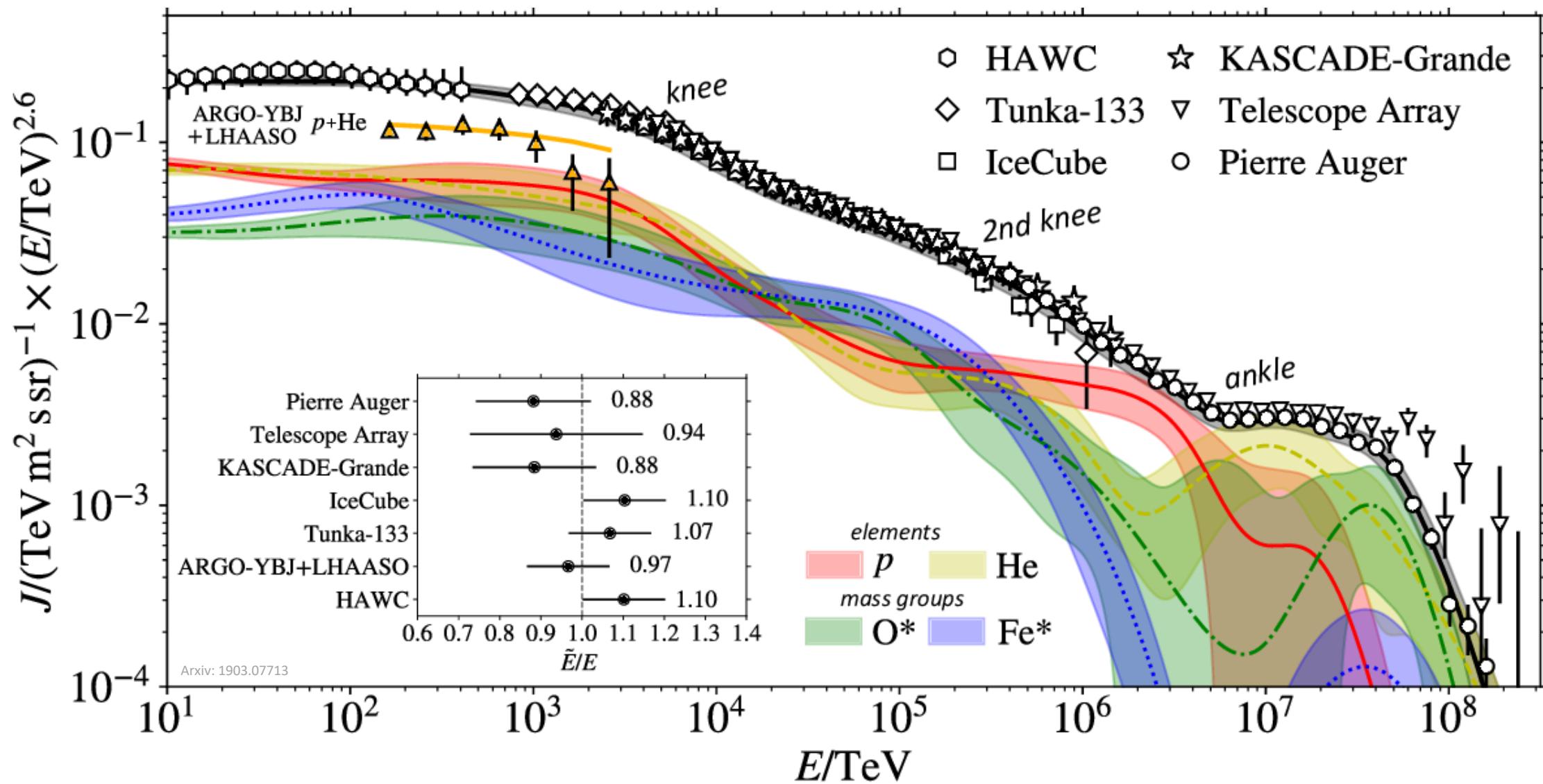
Correlation and Robustness Test: Stopping Muons

- Correlation between proxy and target

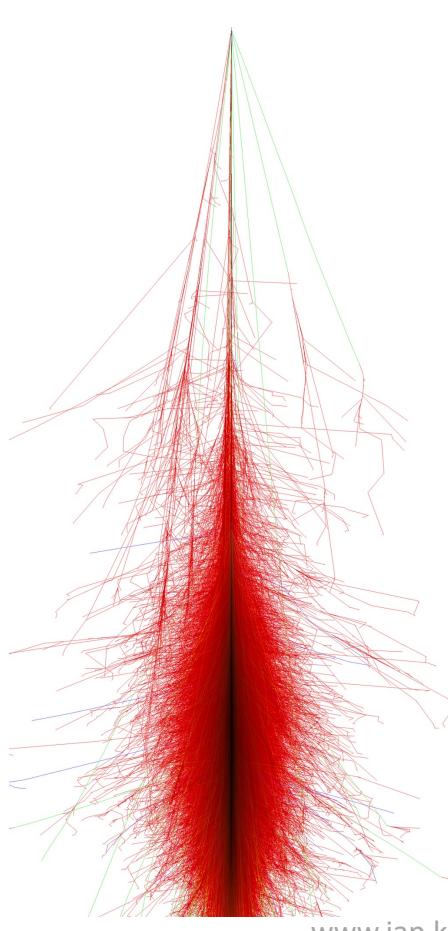
- Build unfolding matrix on H3a, H4a, GST & GSF
- Unfold H3a as test set
- Results are within uncertainties → robust



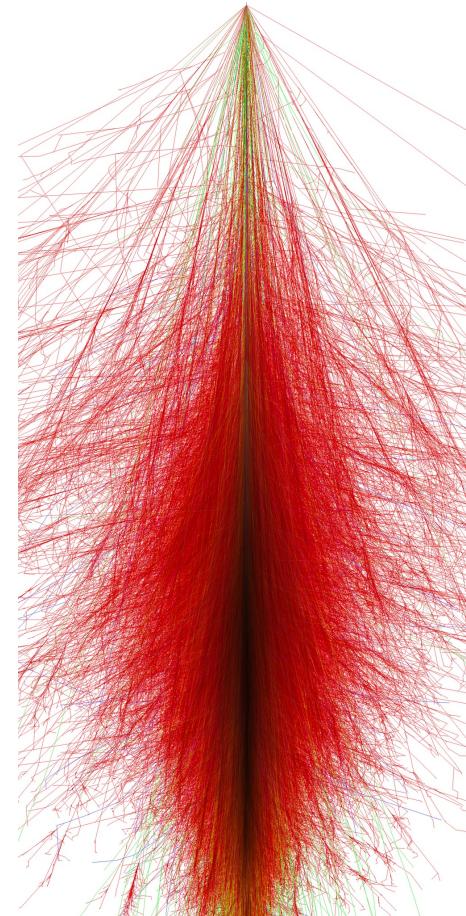
Cosmic ray flux



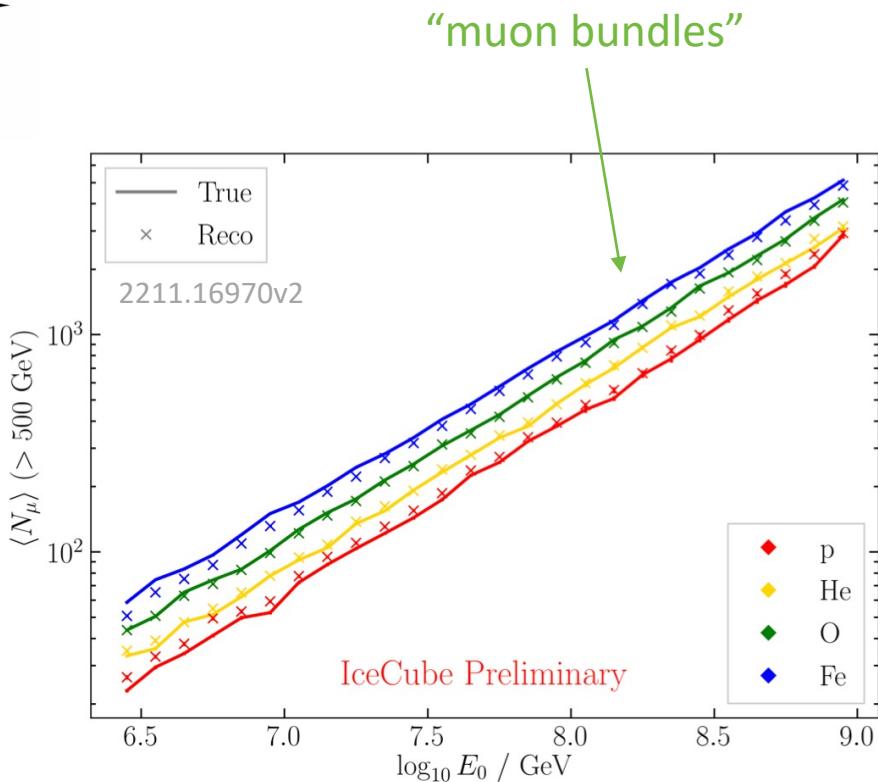
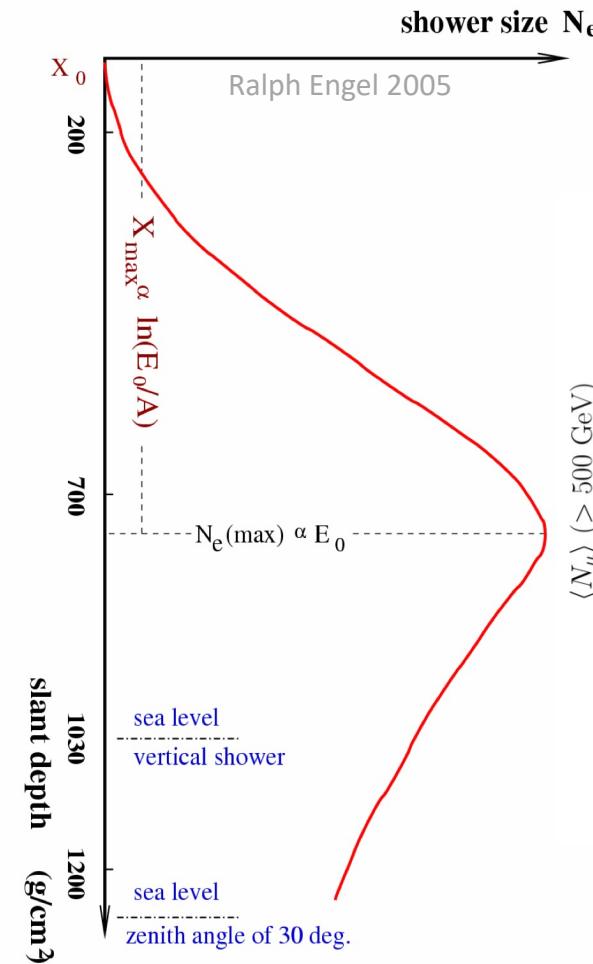
Air shower – 10 TeV



Proton

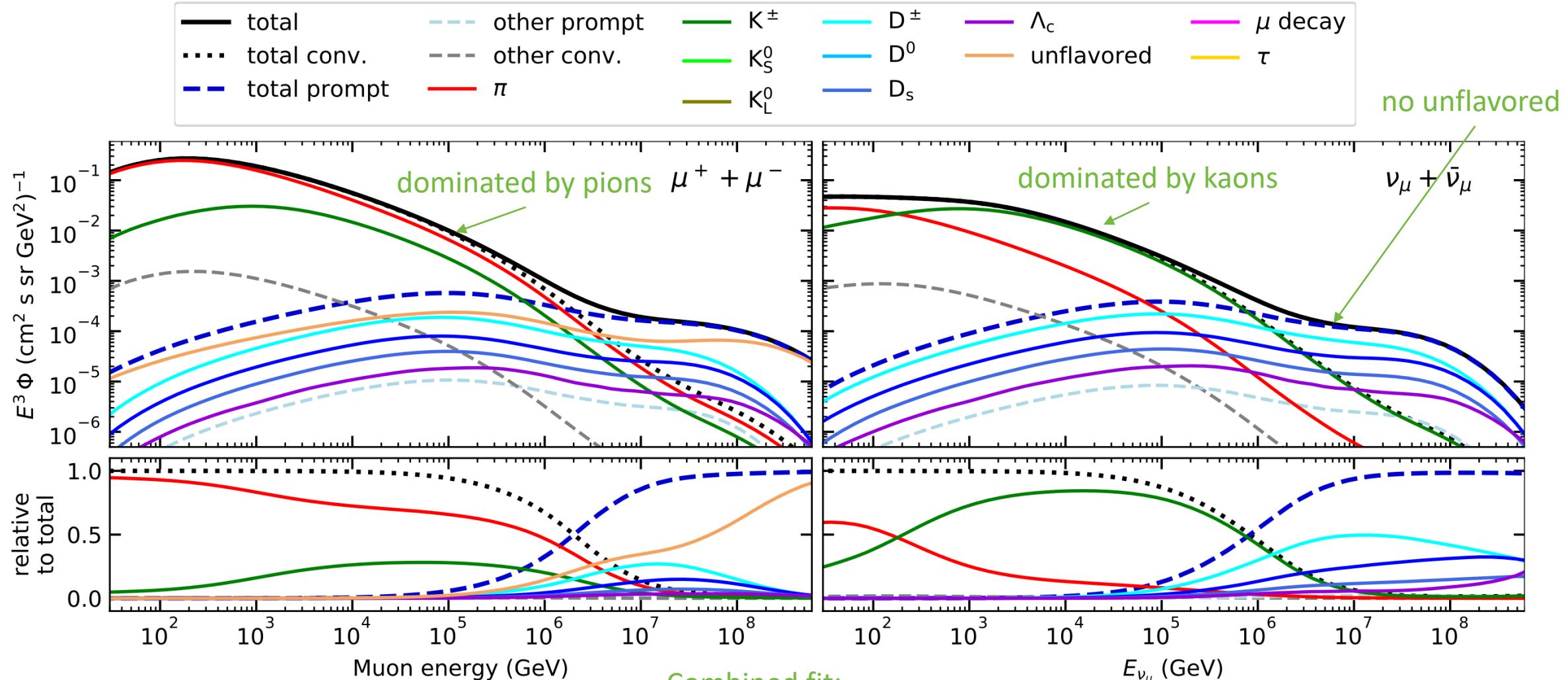


Iron



Prompt atmospheric muons and neutrinos

10.1103/PhysRevD.100.103018



Combined fit:

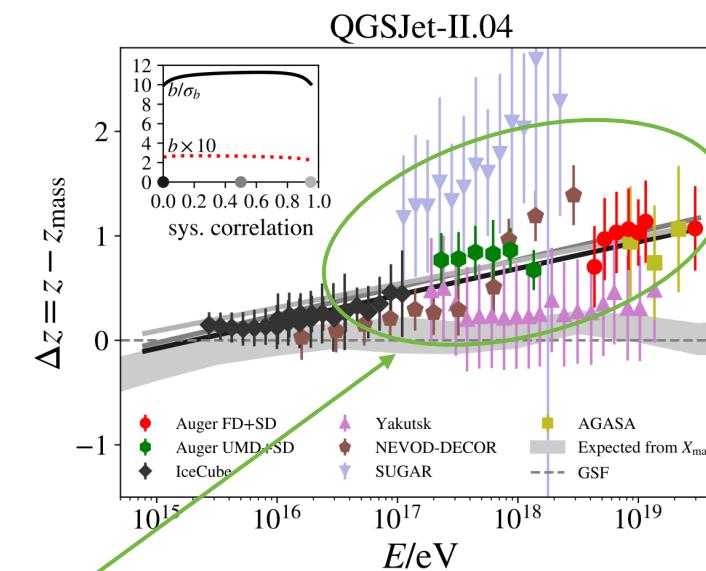
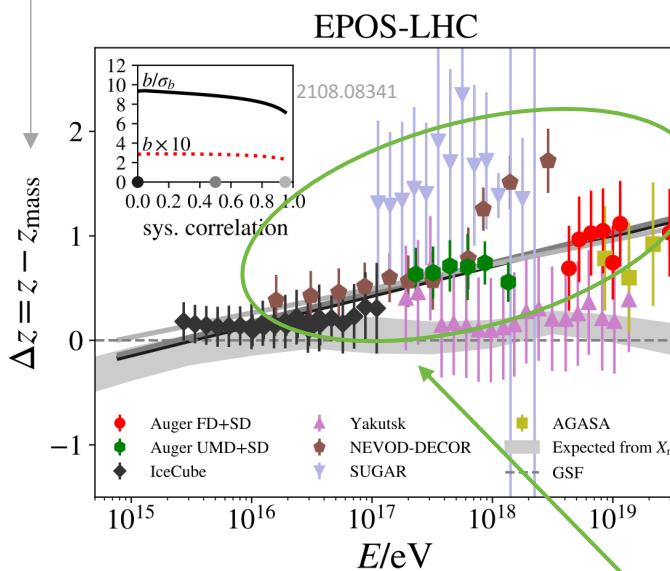
- handle on pion/kaon ratio
- handle on charmed mesons

Muon puzzle and model uncertainties

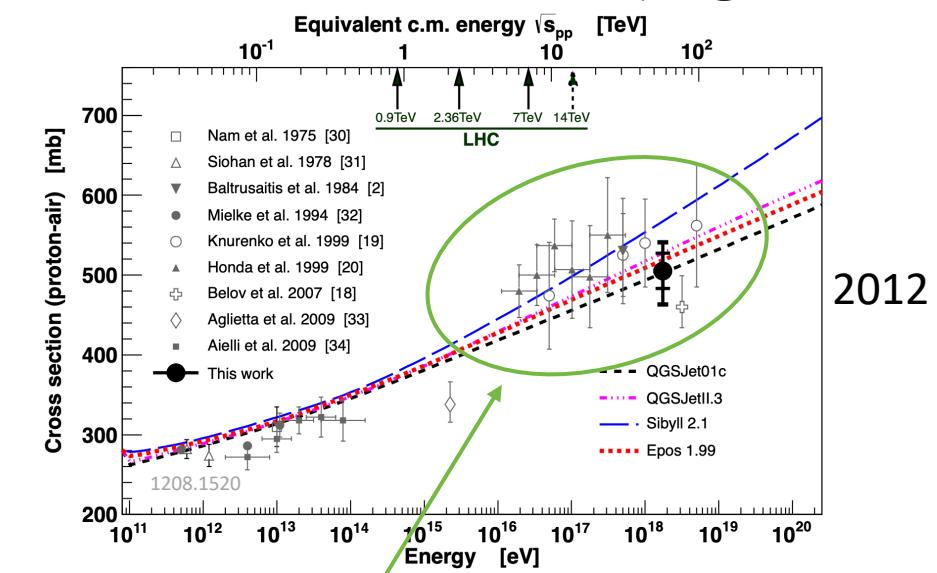
"muon number"

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

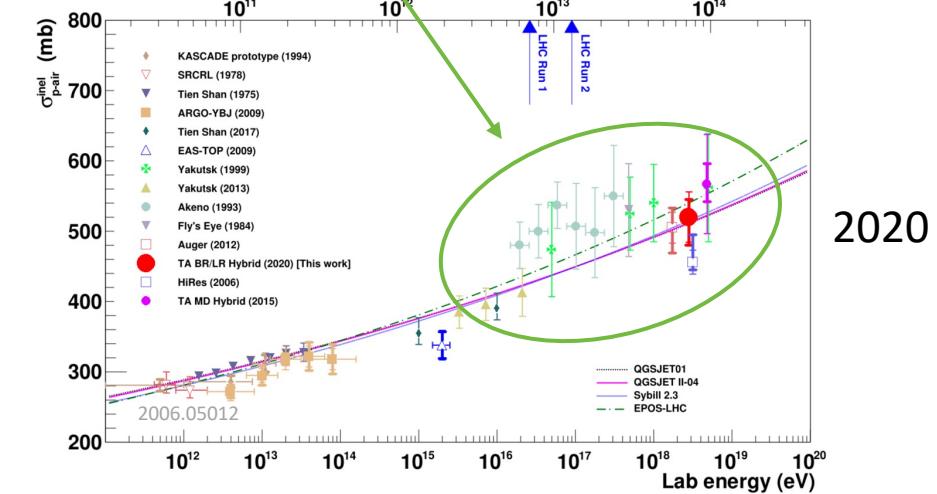
Expected z
("muon number")



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

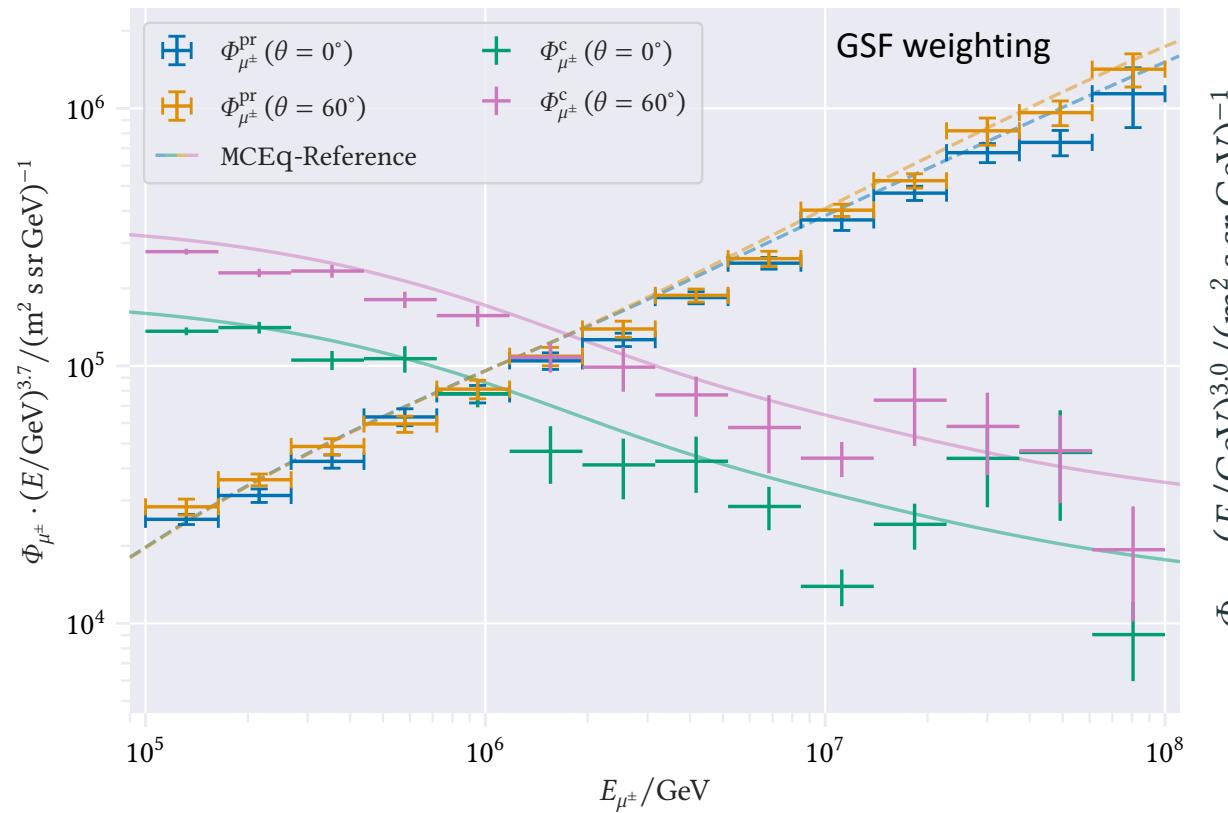


➤ Uncertainties at $E > 10 \text{ PeV}$



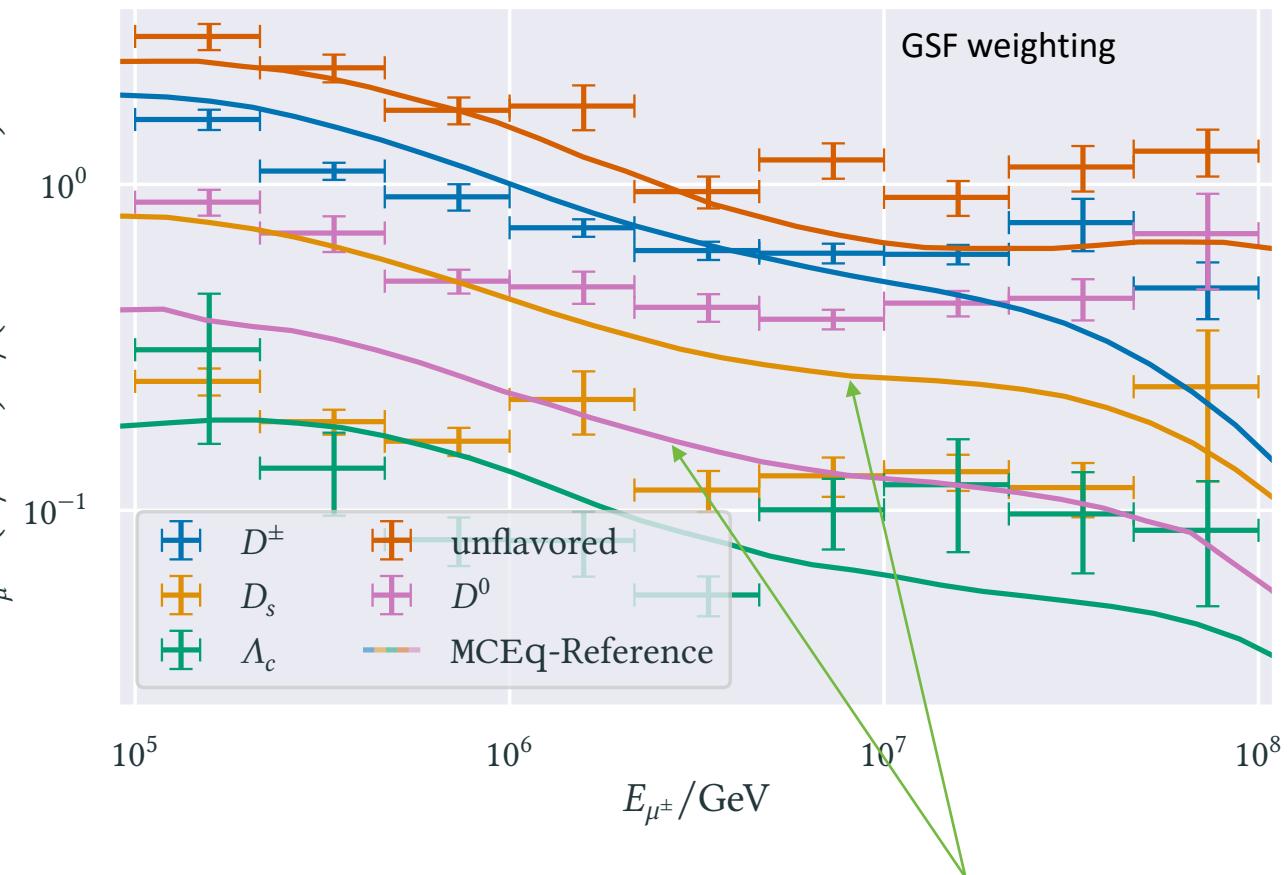
CORSIKA 7 tagging and MCEq comparison

MCEq: tool to numerically solve the cascade equations that describes the evolution of particle densities as they propagate through a gaseous, dense medium
<https://github.com/mceq-project/MCEq>



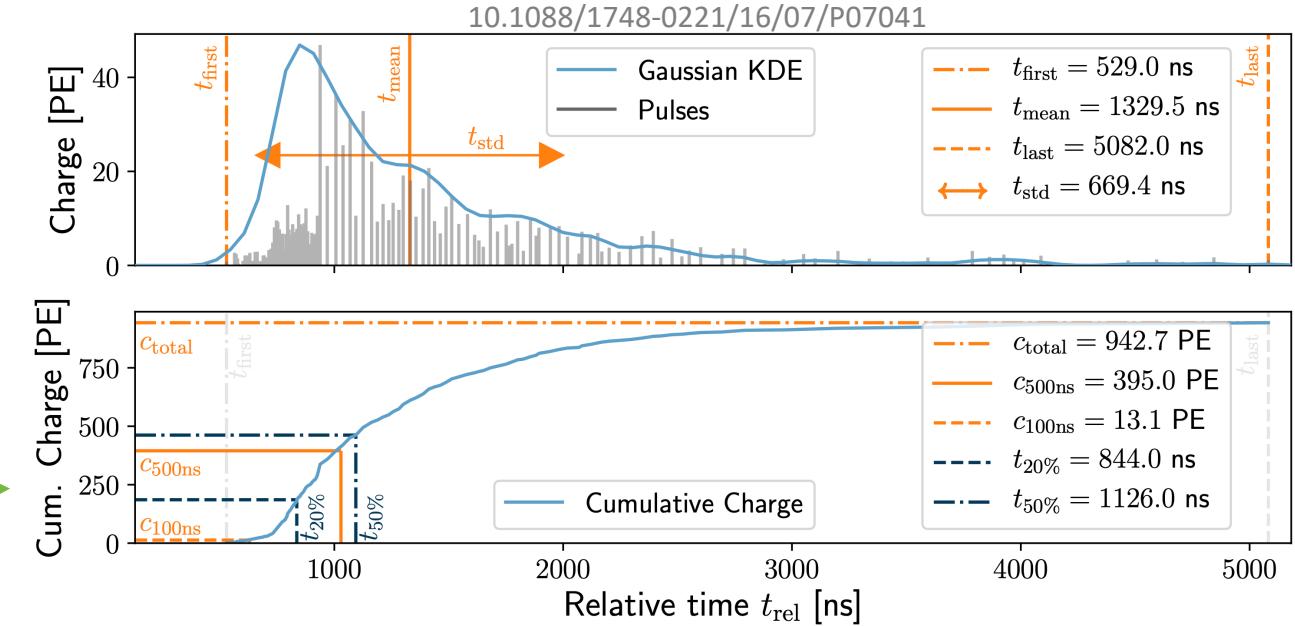
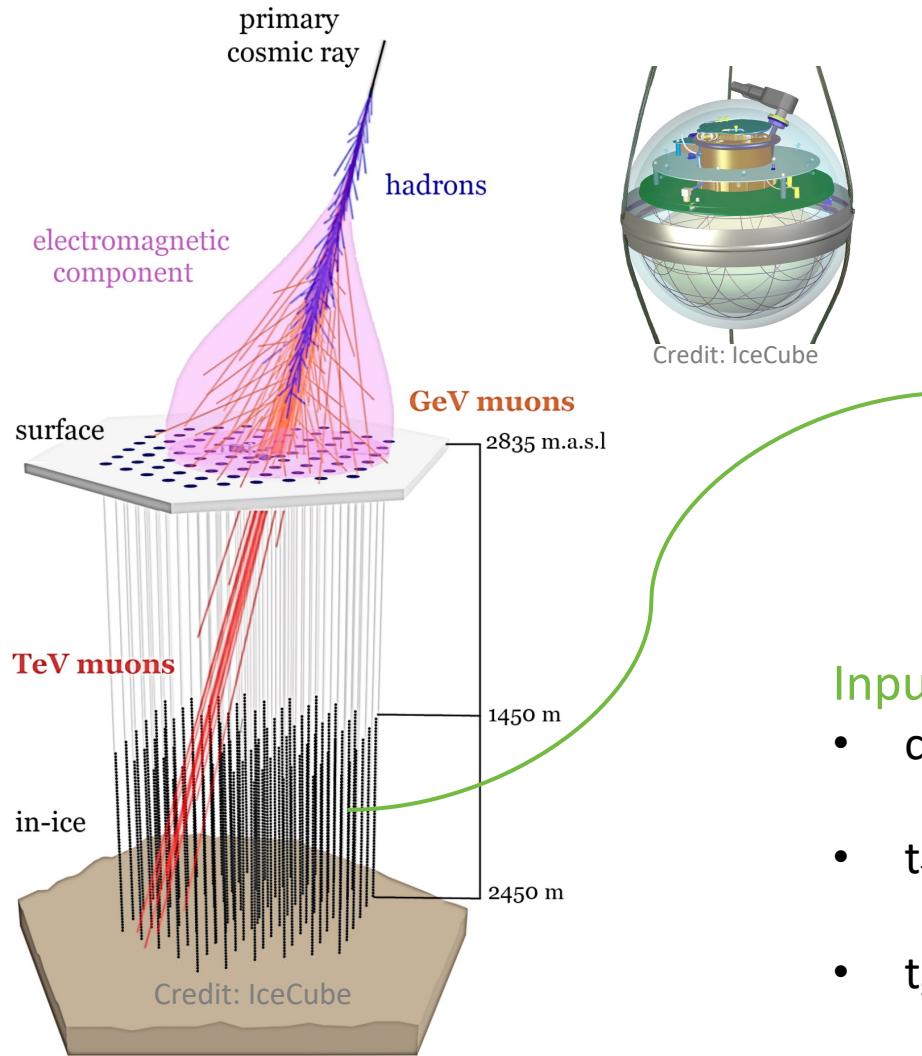
- Good agreement in total prompt and conv muon flux

mention panama



- \$D^0\$ and \$D_s\$ are swapped here but this is fixed in MCEq

Machine Learning Reconstructions



Convolutional Neural Network

Inputs

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

Outputs

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

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

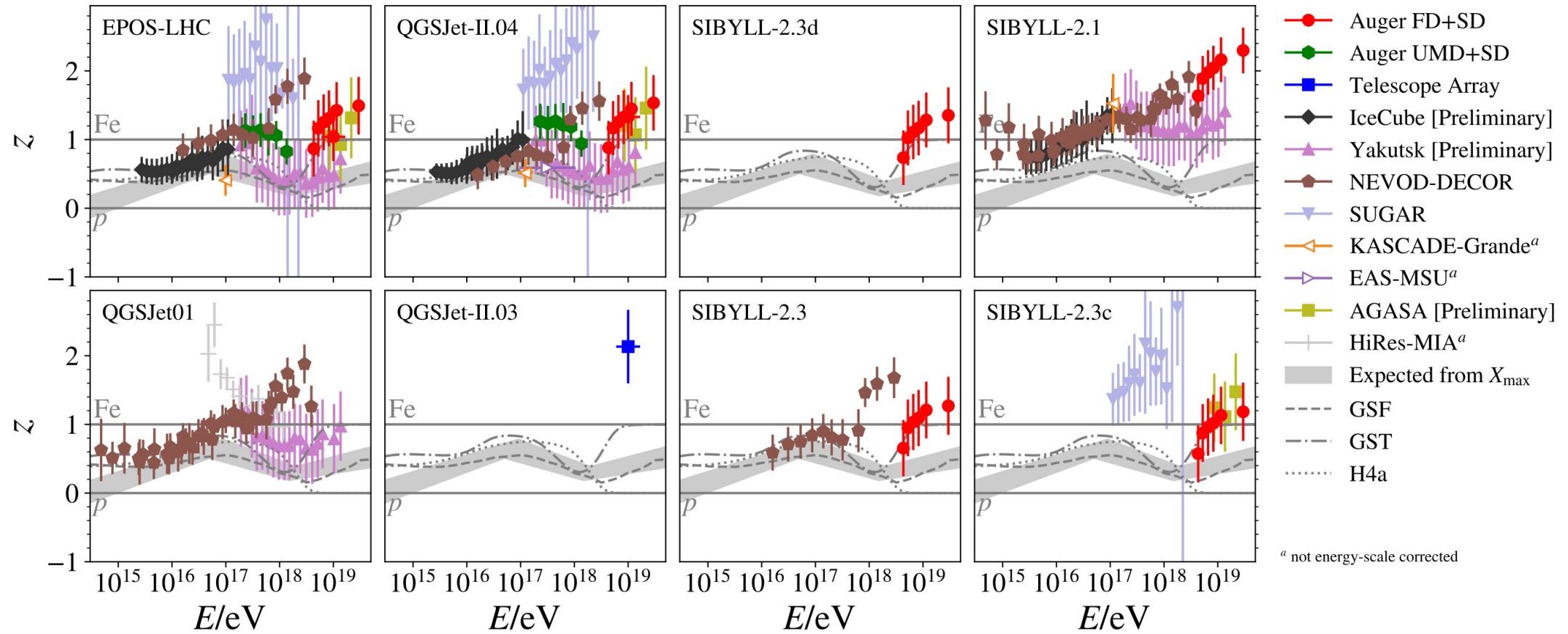
neutrino cuts	>	<
$\cos(\text{zenith})$	0.2	
length	5000 m	15000 m

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"

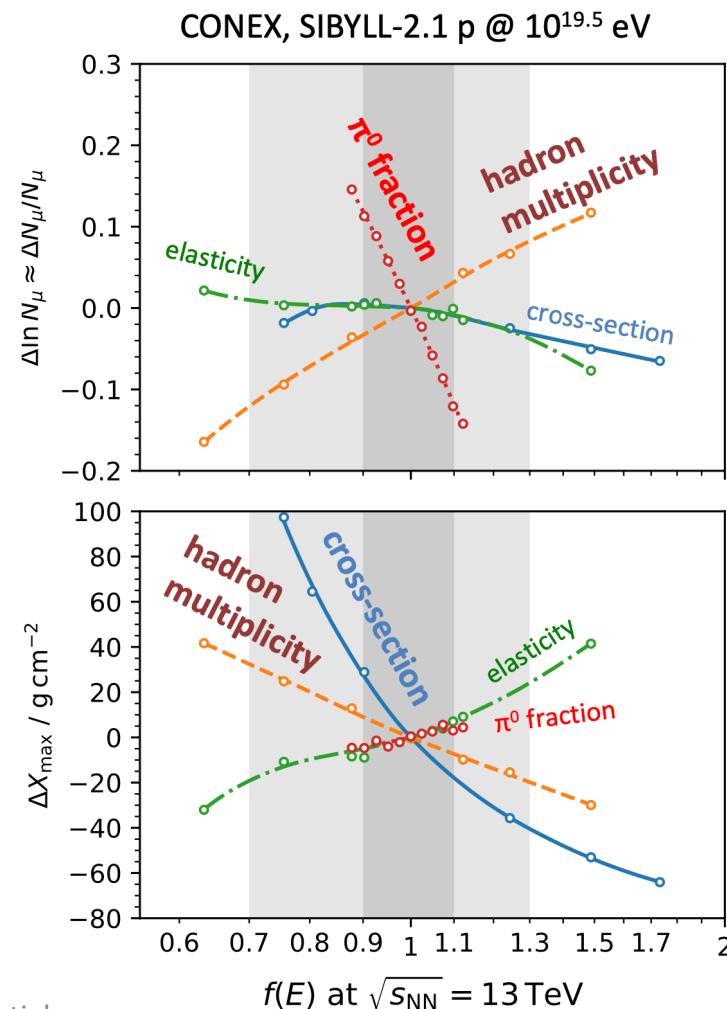
$$z = \frac{\ln\langle N_\mu \rangle - \ln\langle N_\mu \rangle_p}{\ln\langle N_\mu \rangle_{\text{Fe}} - \ln\langle N_\mu \rangle_p}$$



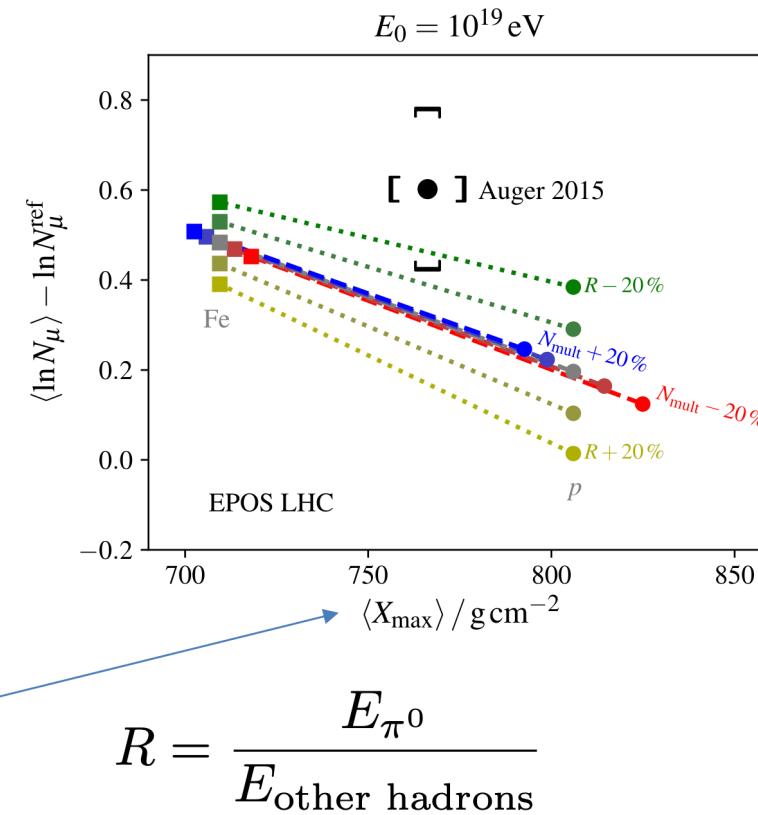
Arxiv: 2108.08341

Possible Solutions

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

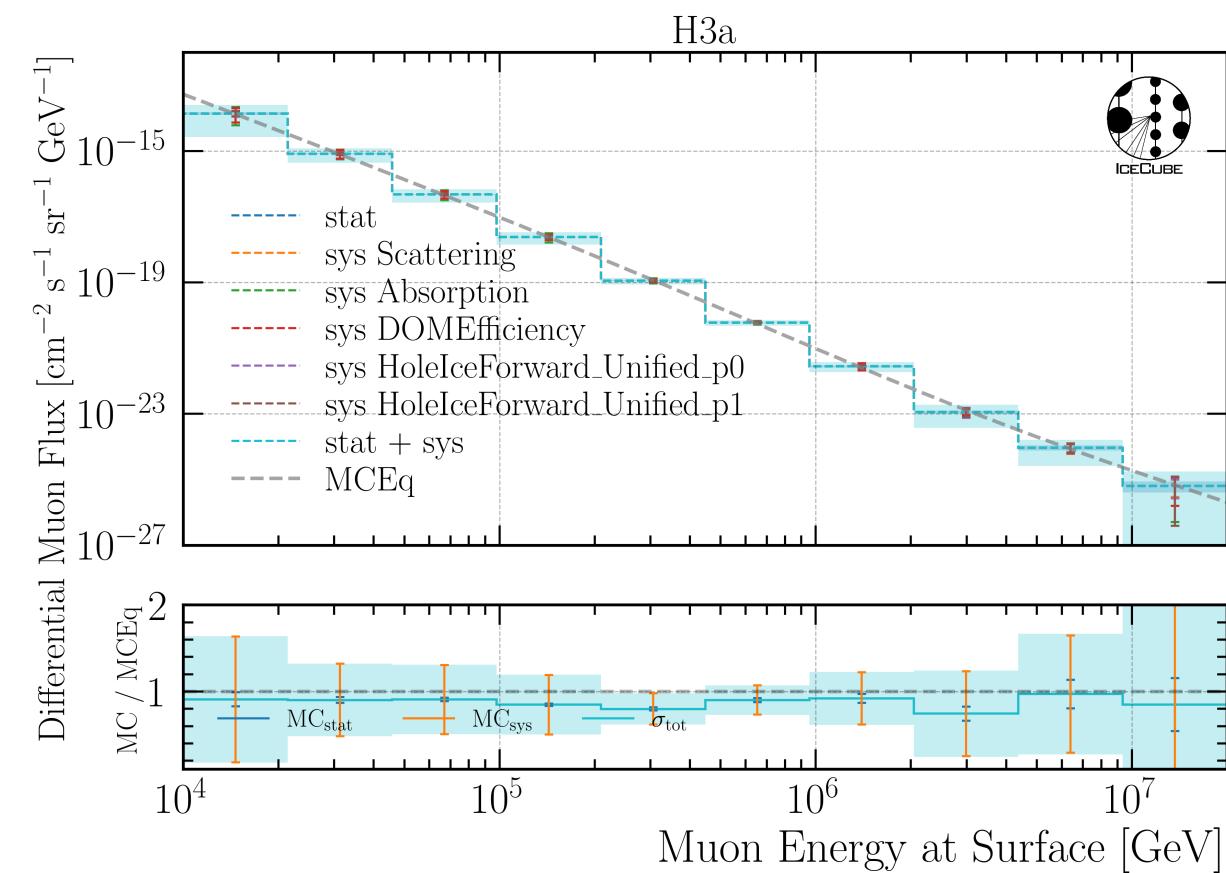
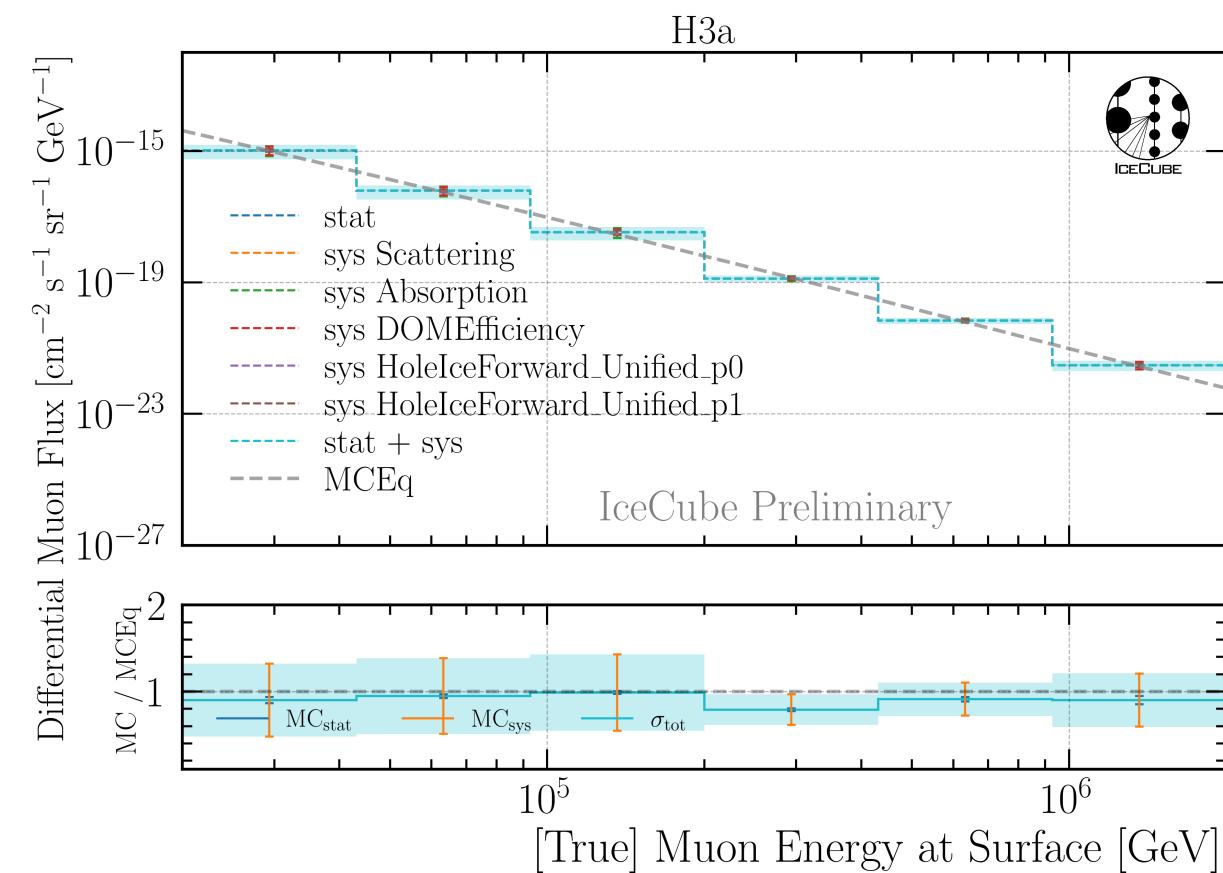


S. Baur, HD, M. Perlin, T. Pierog, R. Ulrich, K. Werner,
arXiv:1902.09265

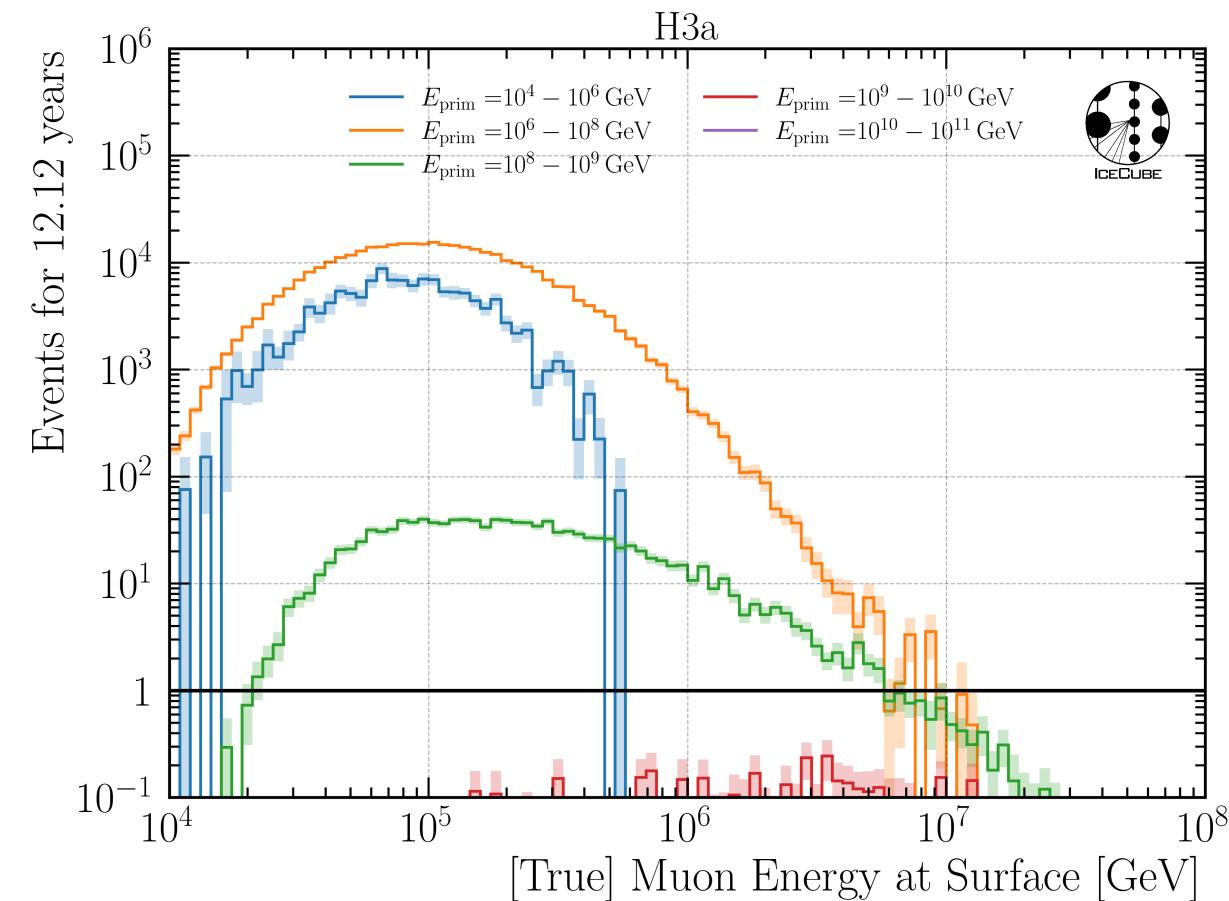
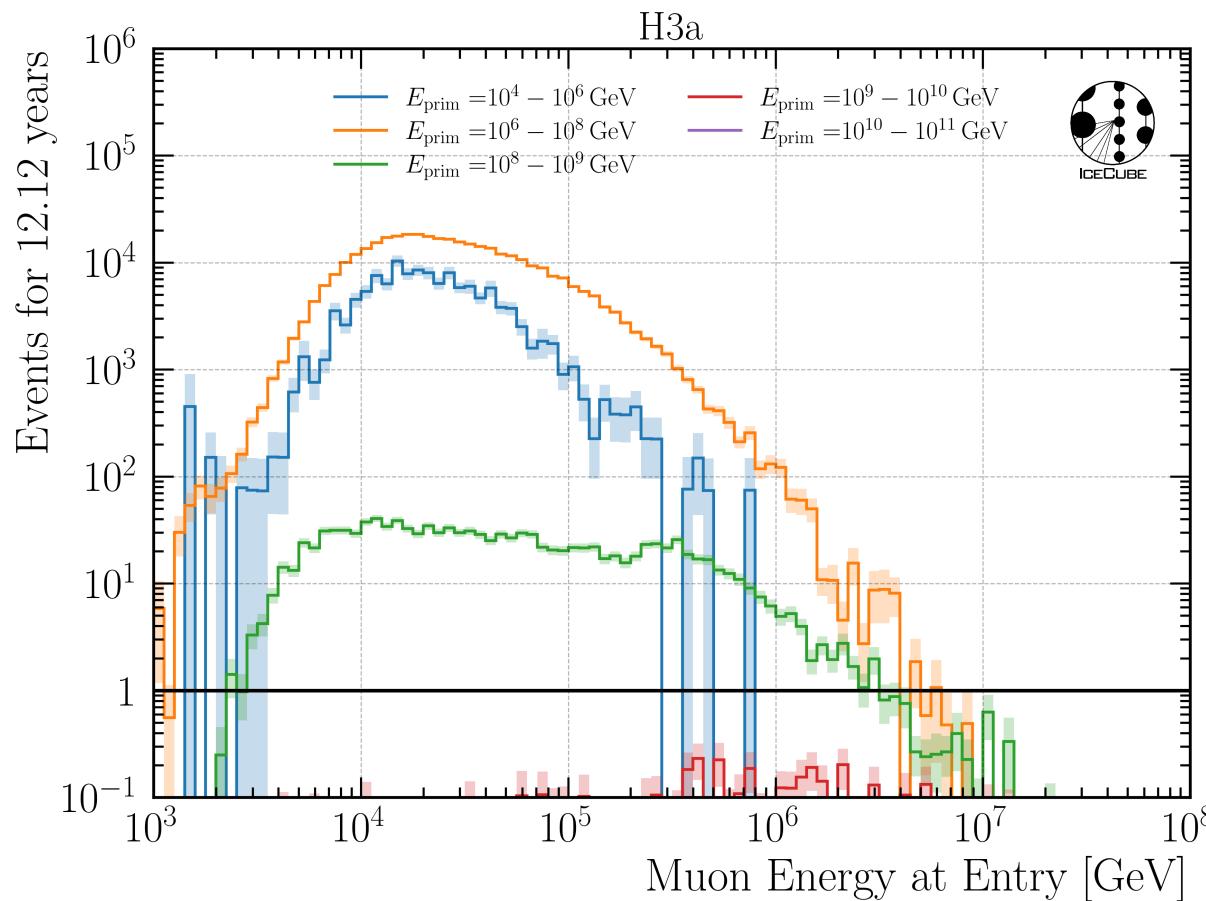


- Only changes to R can solve muon puzzle
- Small changes have large effect,
 R needs to be known to about 5 %

MCEq vs CORSIKA



MC Statistics: Fine binning



MC Statistics: Unfolding binning

