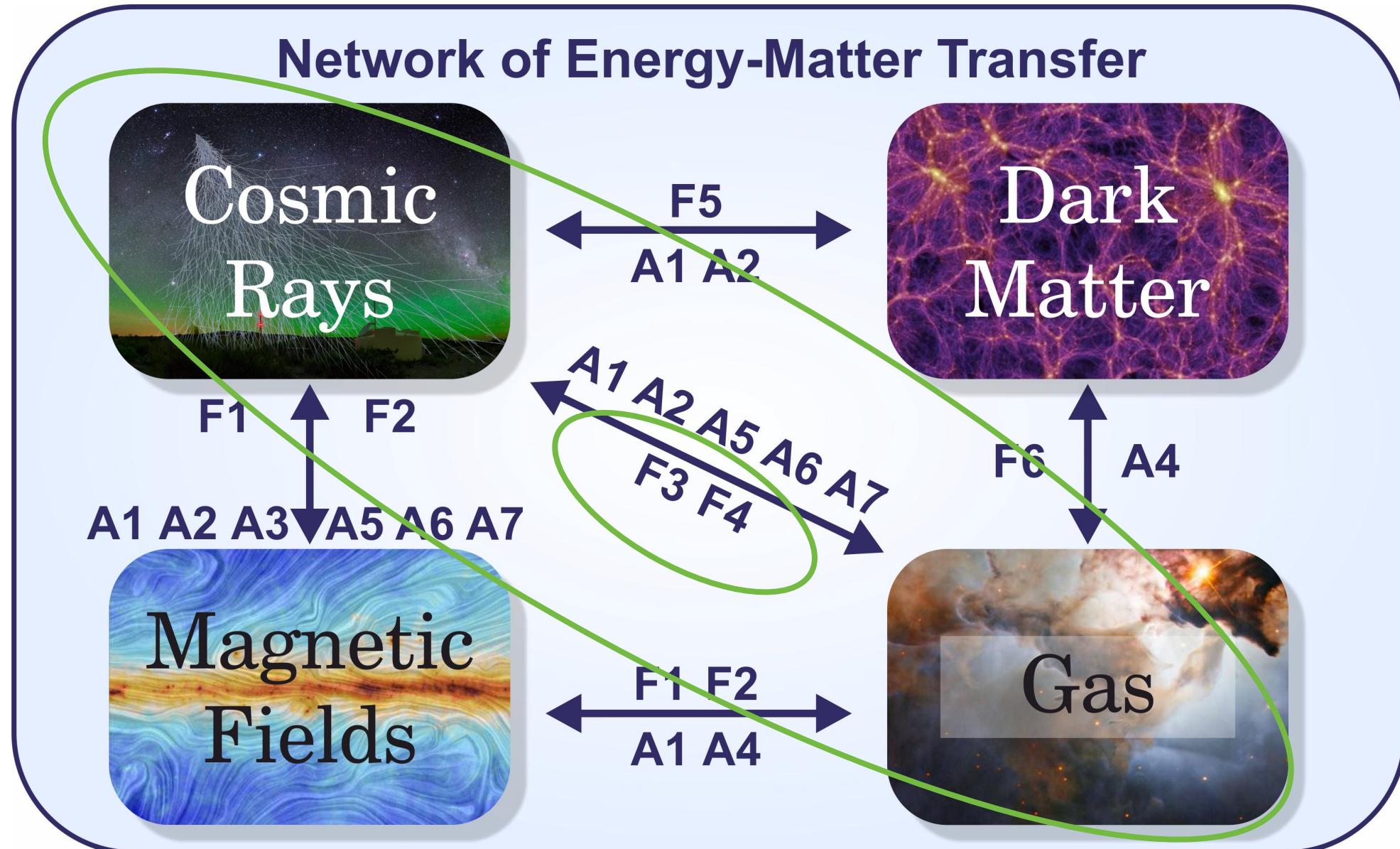


Constraining production uncertainties of heavy mesons via measurement of the prompt atmospheric component using IceCube

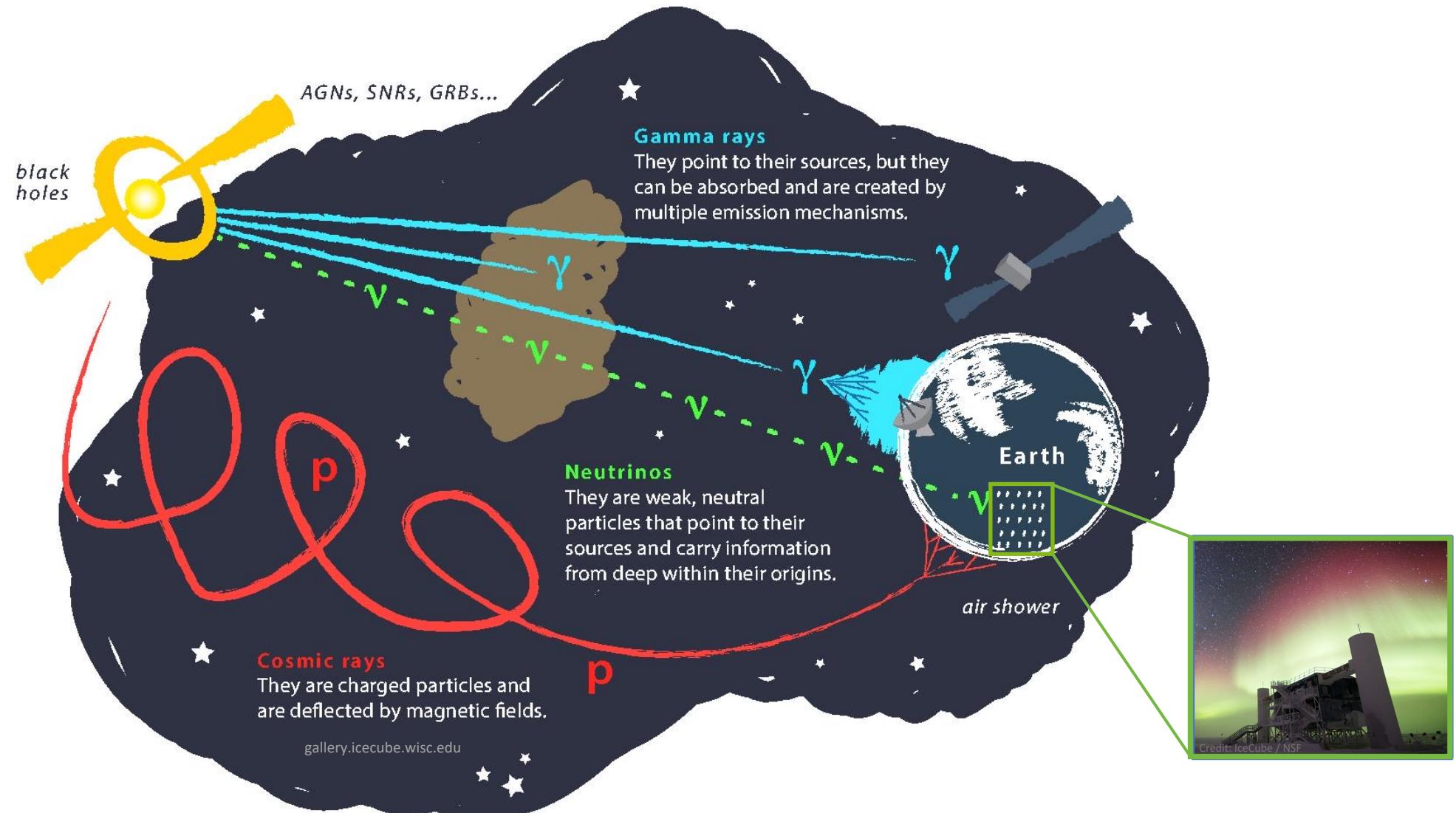
- F3 -

Ludwig Neste, Mirco Hünnefeld and Pascal Gutjahr

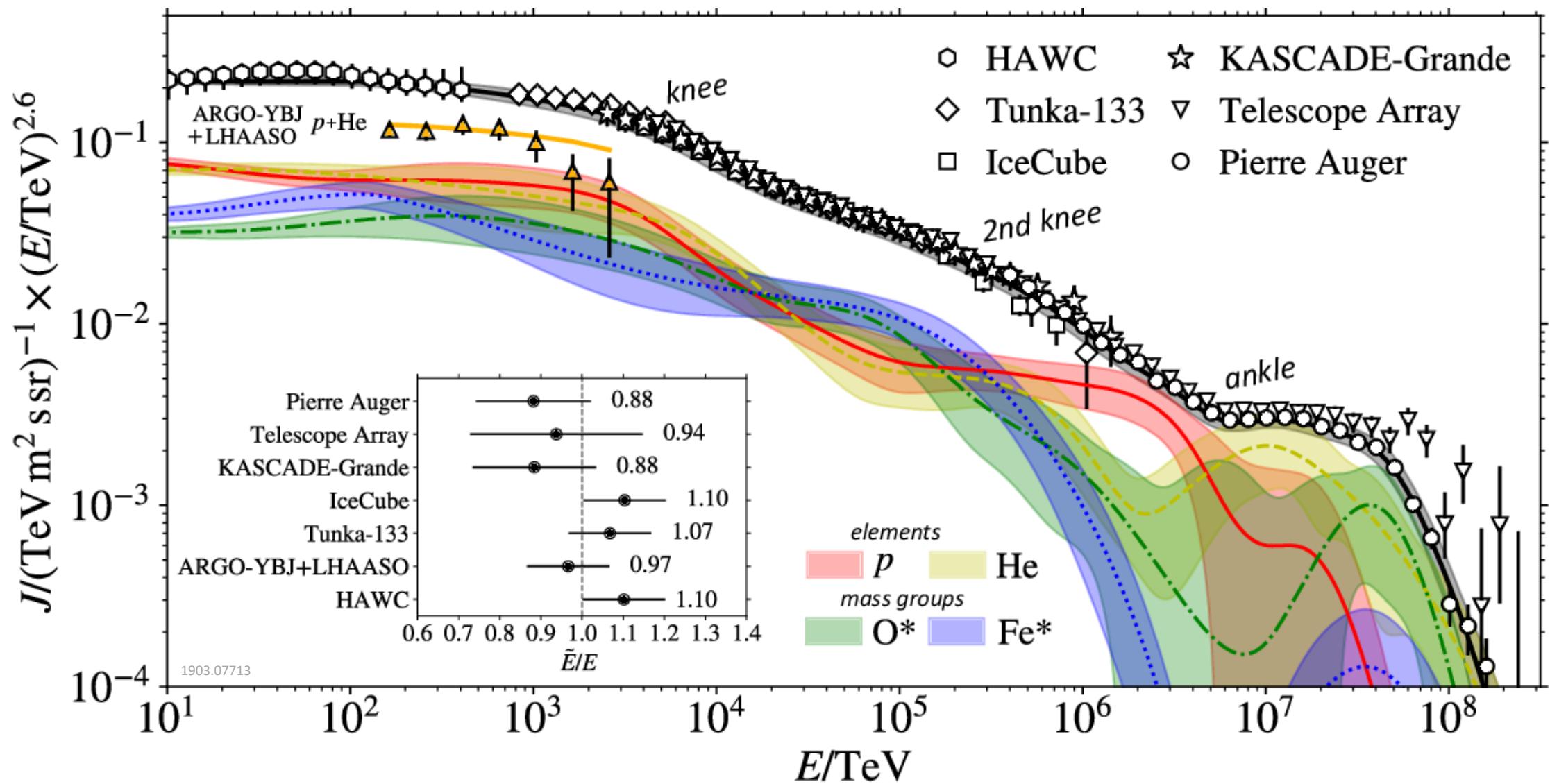
SFB1491 General Assembly, Dortmund 2023



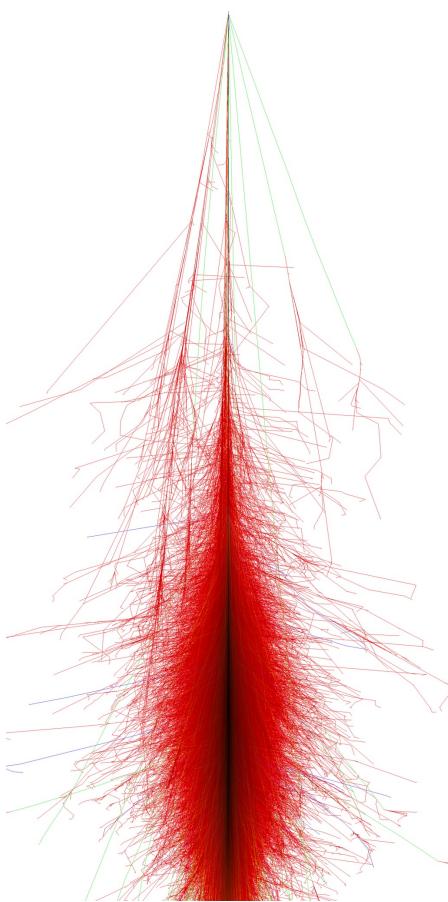
Astroparticle physics



Cosmic ray flux

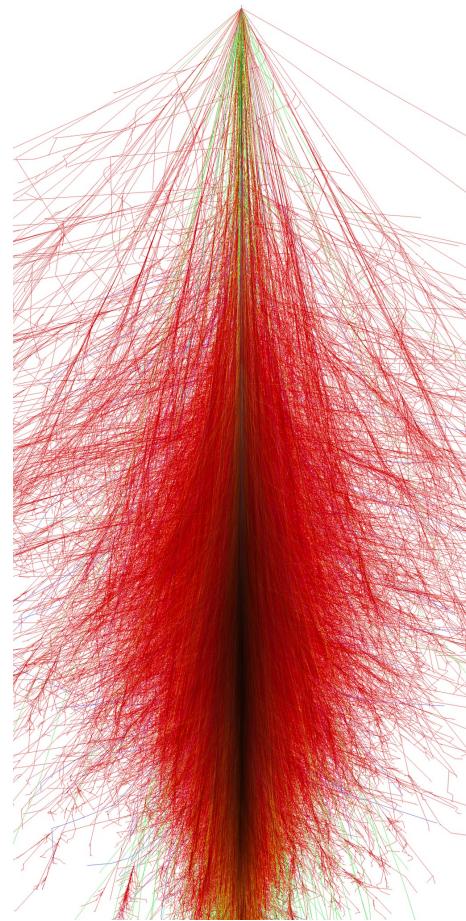


Air shower – 10 TeV

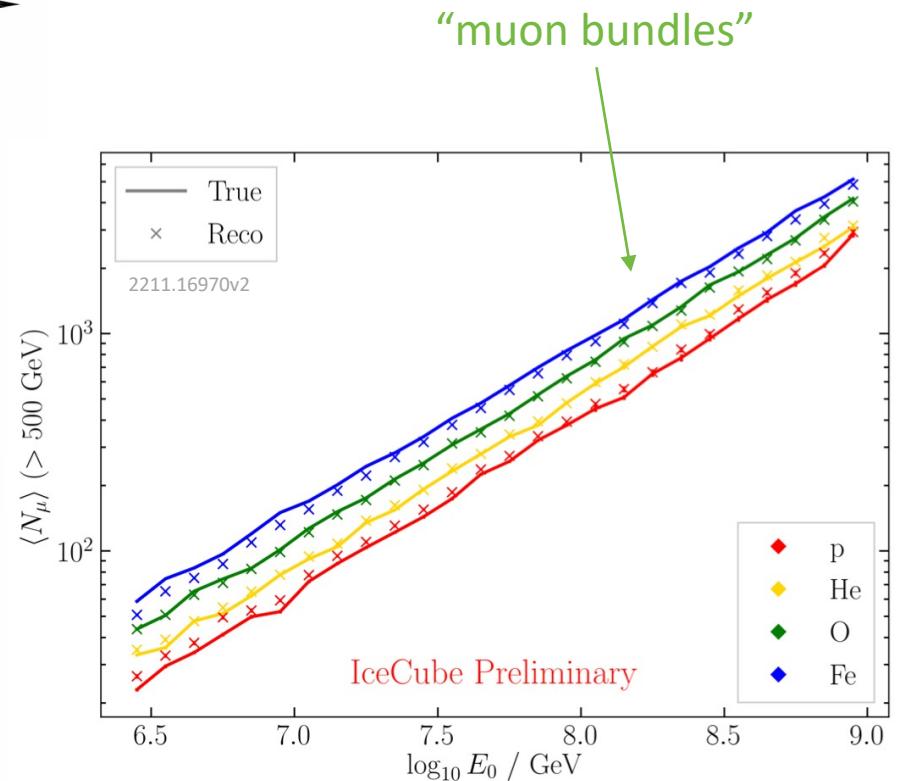
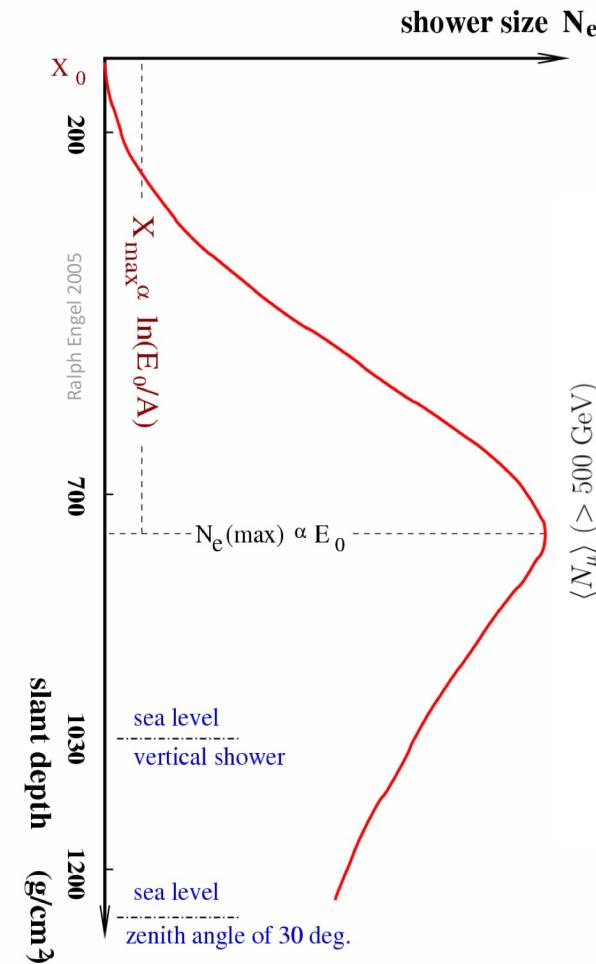


Proton

zeuthen.desy.de

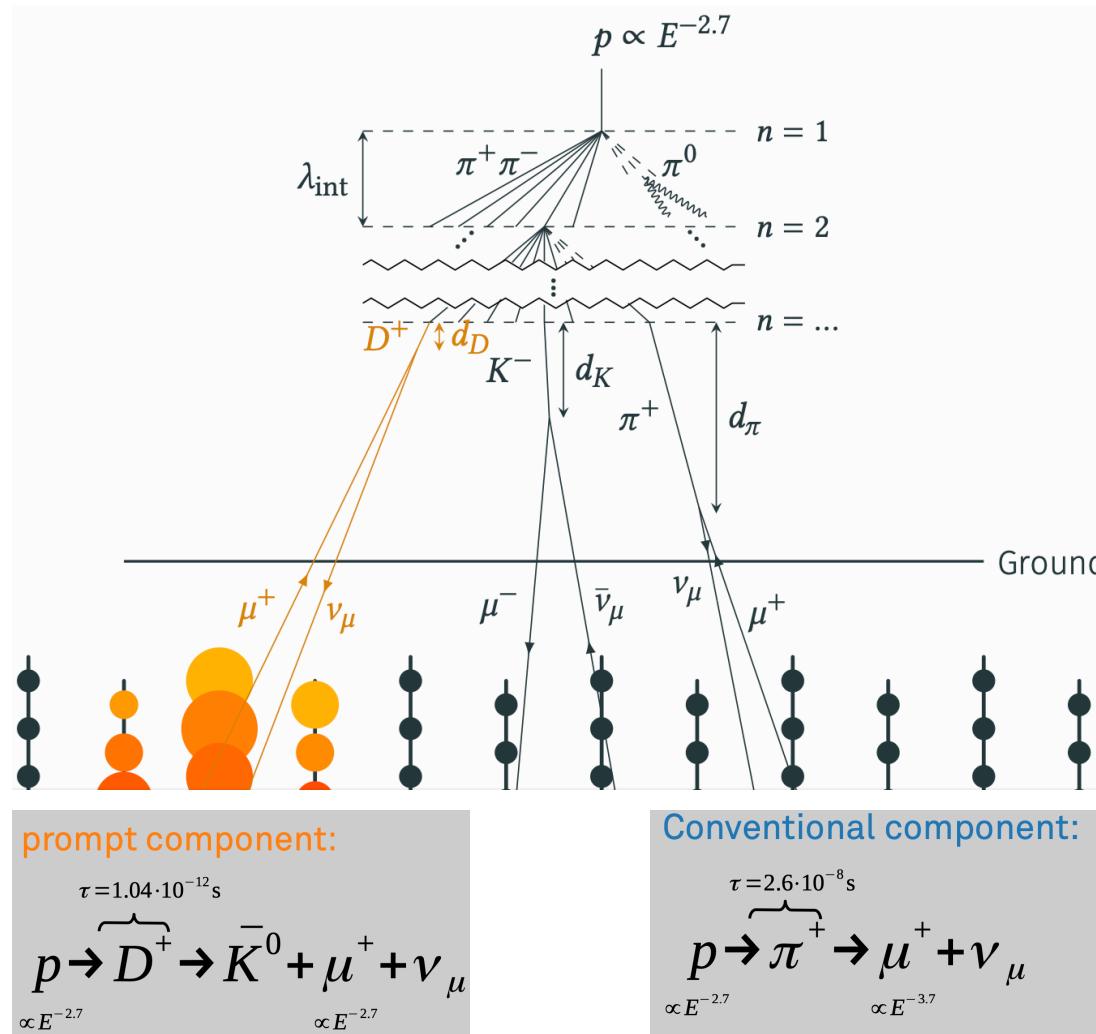


Iron

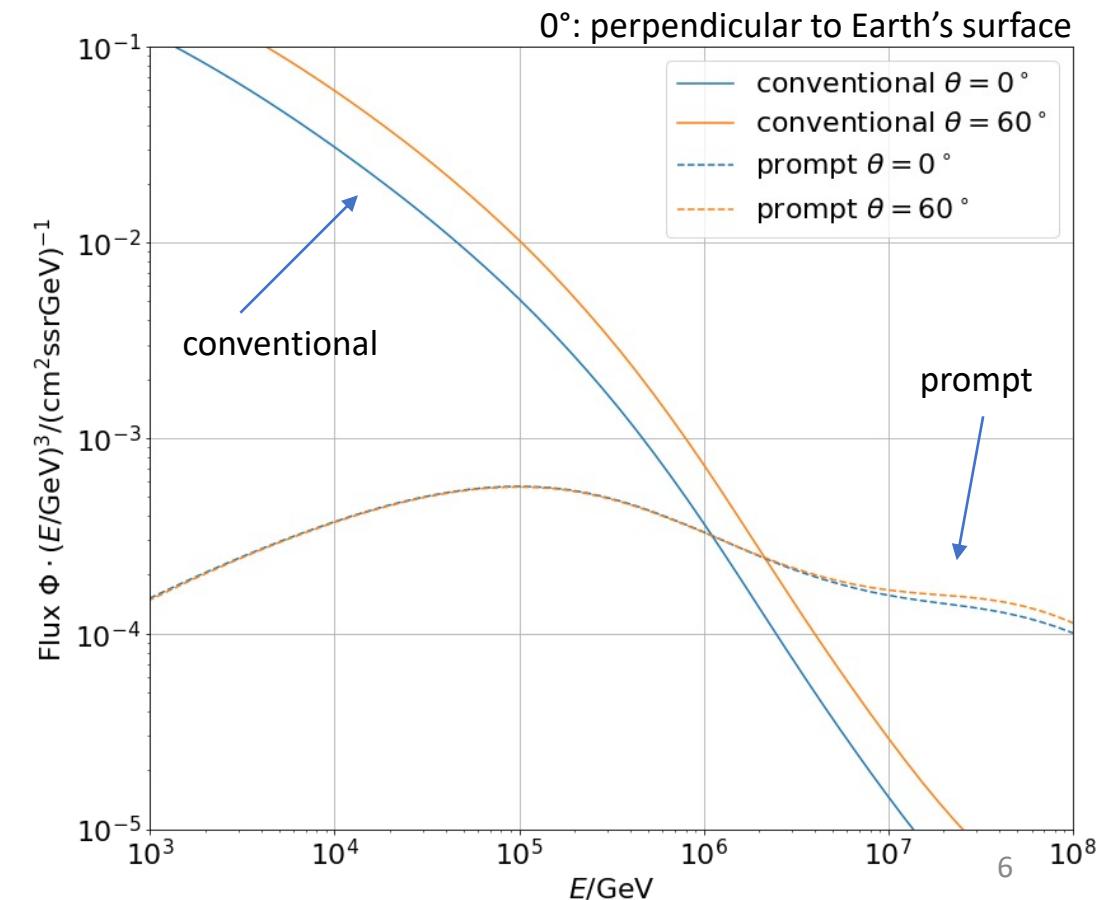
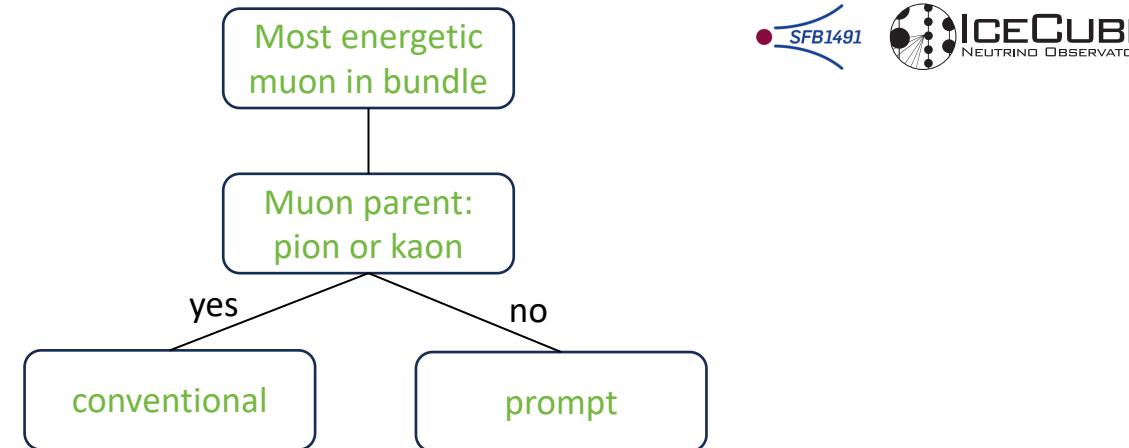


Muon flux: Definition of prompt

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

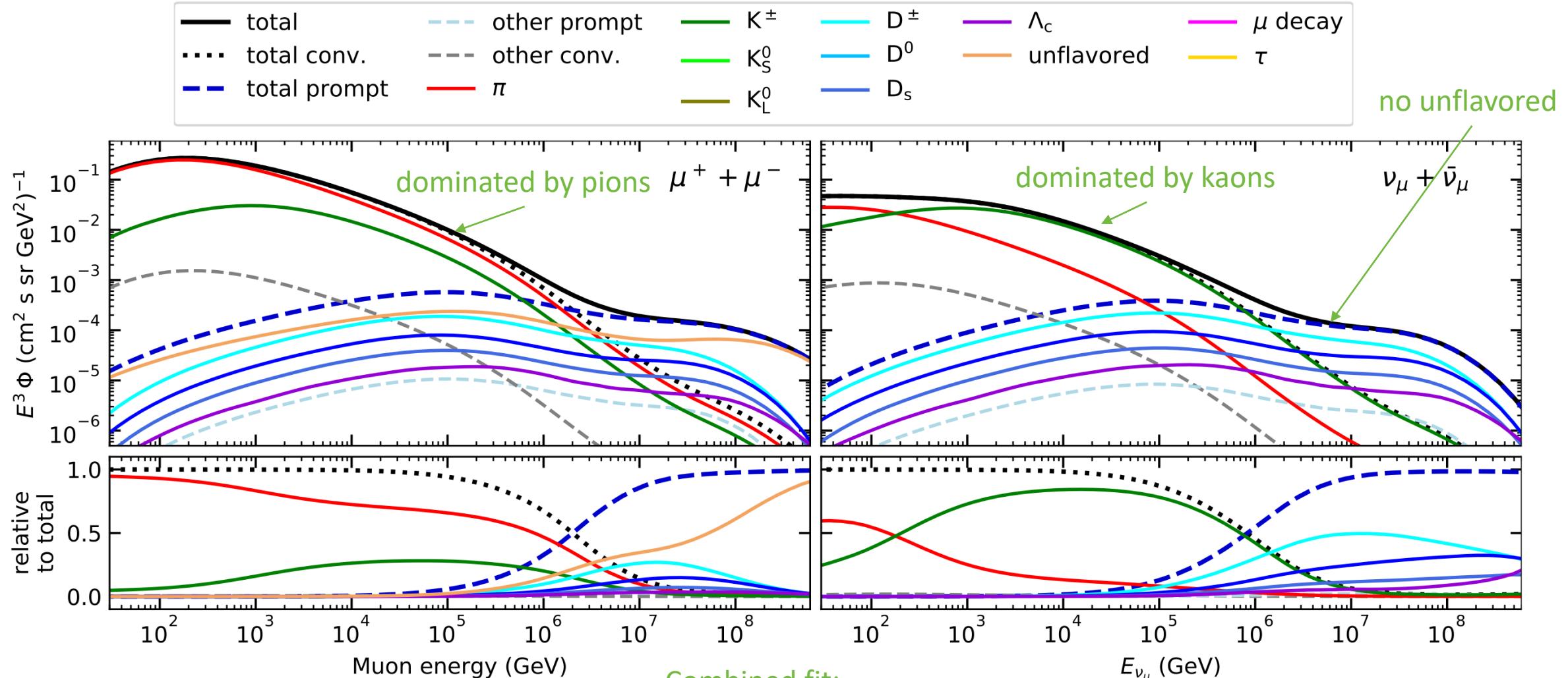


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Prompt atmospheric muons and neutrinos

10.1103/PhysRevD.100.103018



Combined fit:

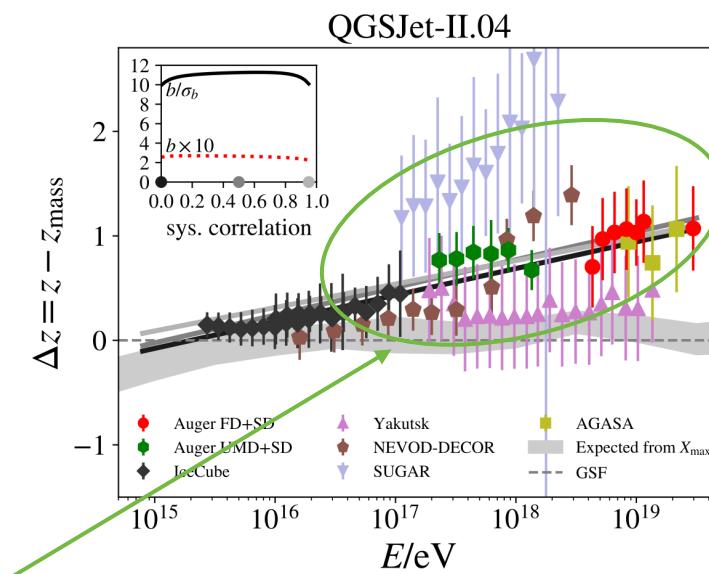
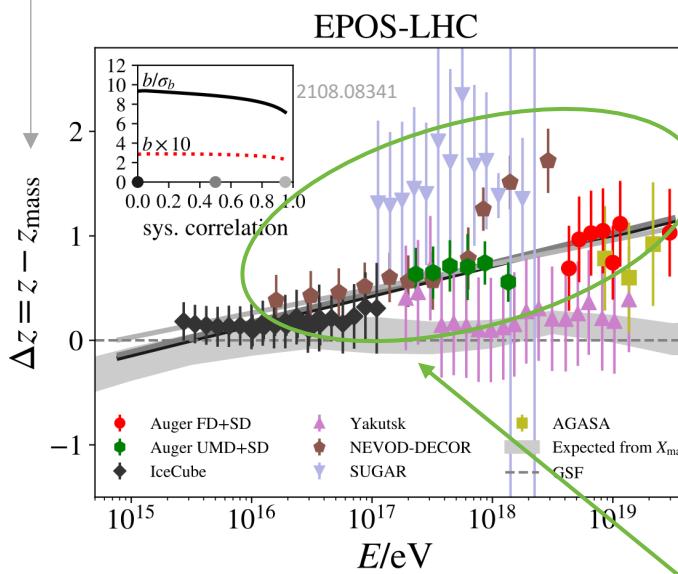
- handle on pion/kaon ratio
- handle on charmed mesons

Muon Puzzle and model uncertainties

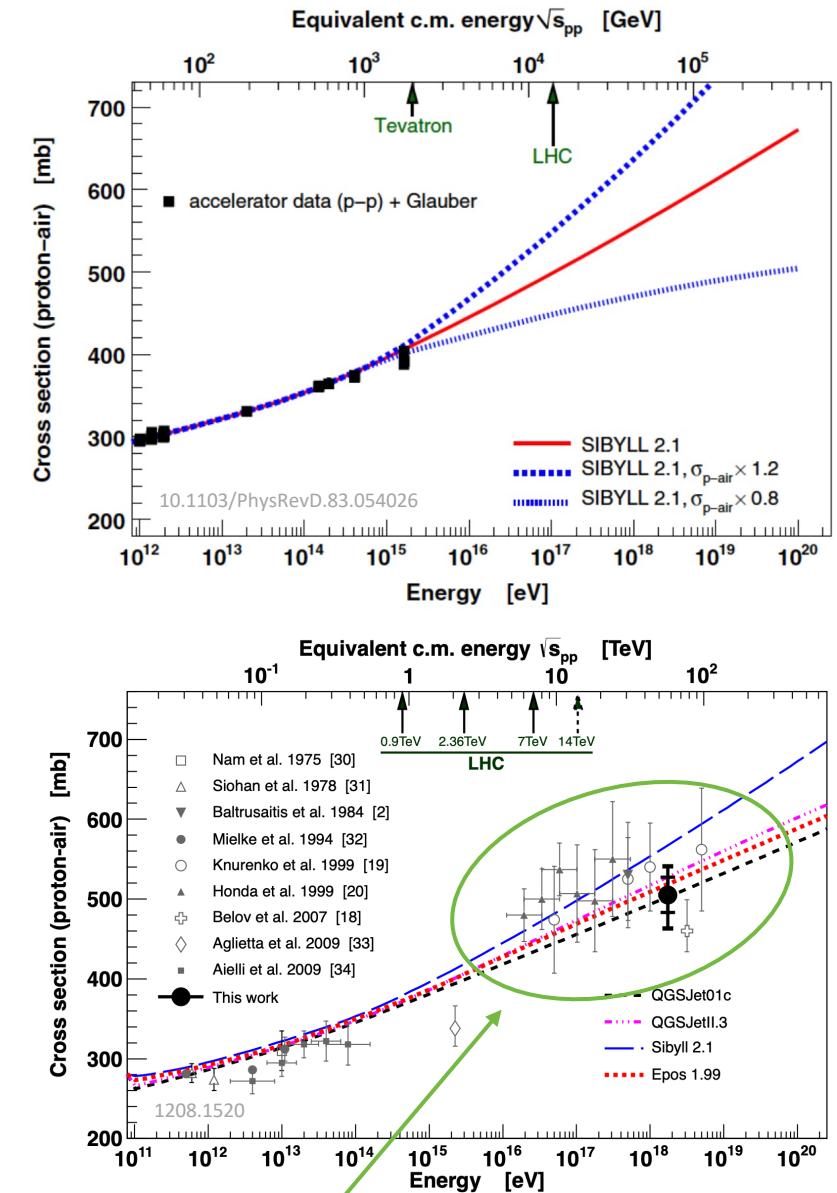
Expected z ("muon number")

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

"muon number"



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



➤ Large uncertainties at $E > 10 \text{ PeV}$

F3/F4 Overview

- F3 (“Prompt lepton production in hadronic interactions”)
- F4 (“Cross sections and hadronic interactions in particle- and astroparticle physics”)

F3.1 (W. Rhode, IceCube)

Deliverables:

- Year 1/2:
 - Establish analysis framework for prompt muon measurement
 - First (preliminary) results will be produced
 - Upgrade MC generators for necessary comparisons and alignments to F3 and F4
- Year 3/4:
 - Continue systematic studies for prompt muon measurement
 - Publication of prompt muon measurement
 - Further work on unifying F3/4 results for MC generators

First: Atmospheric prompt muons

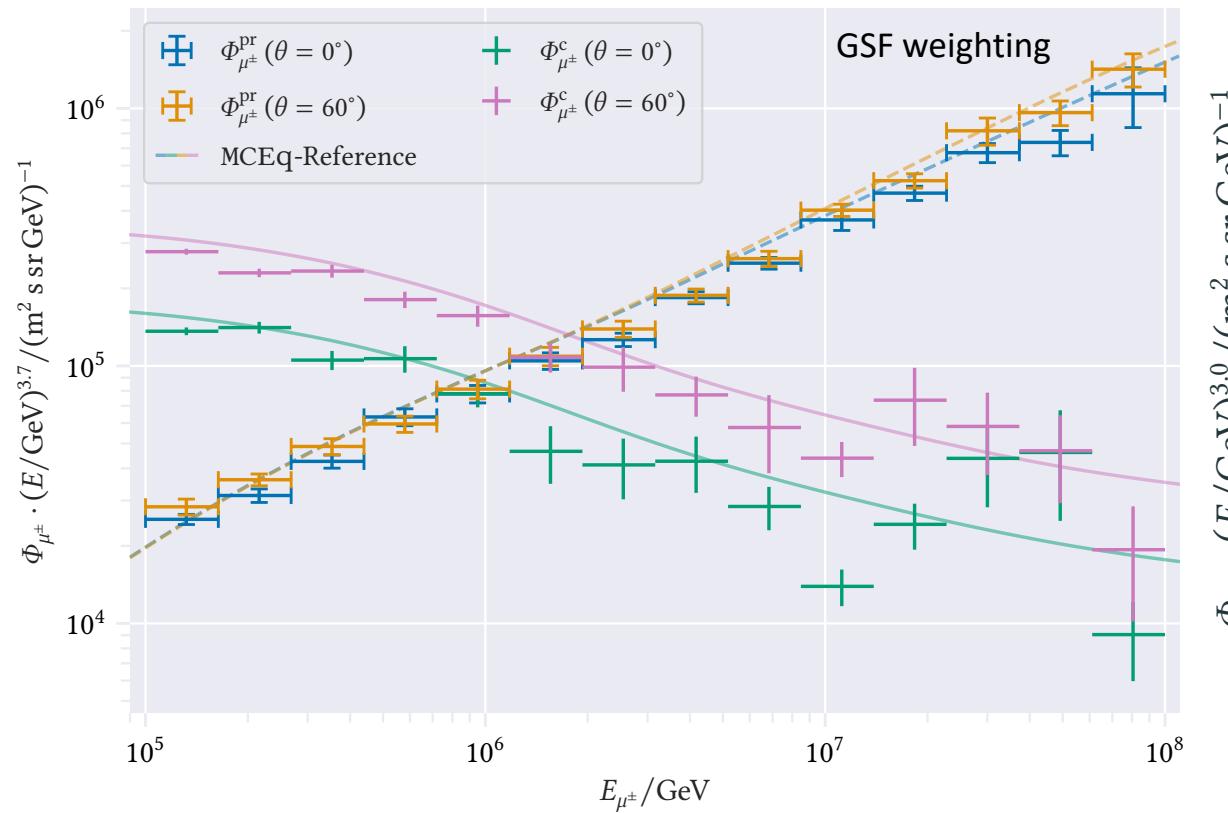
- 1) Detect prompt component of the atmospheric muon flux
 - Measure the normalization
 - Get handle on hadronic interaction models
- 2) Unfold a muon energy spectrum

Idea:

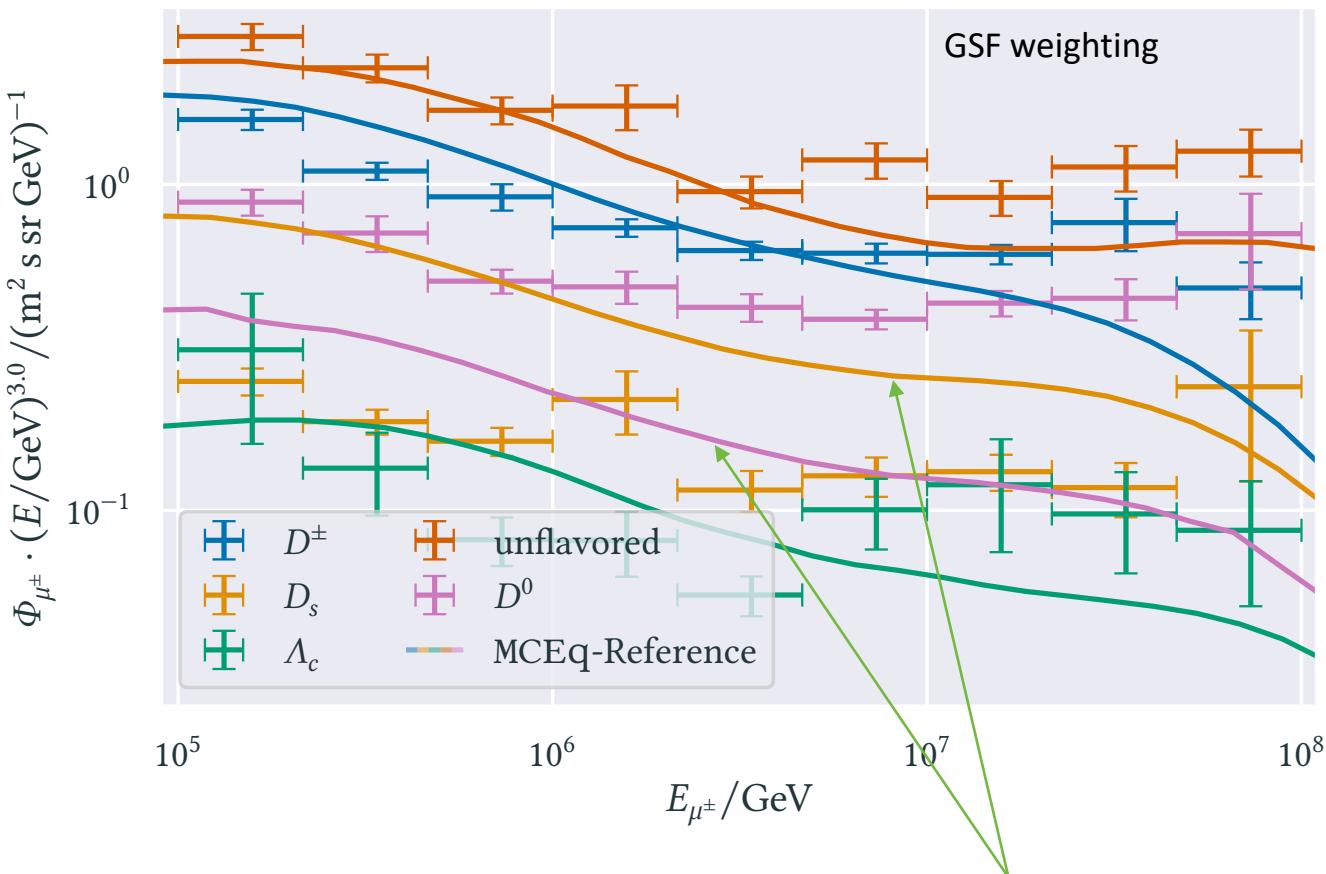
- New CORSIKA simulations with extended history
- Tag muons by parent → prompt or conventional
- Scale amount of prompt particles
 - Scaling saves time and resources instead of doing multiple simulations with different interaction models
 - Perform forward fit of the prompt normalization

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>

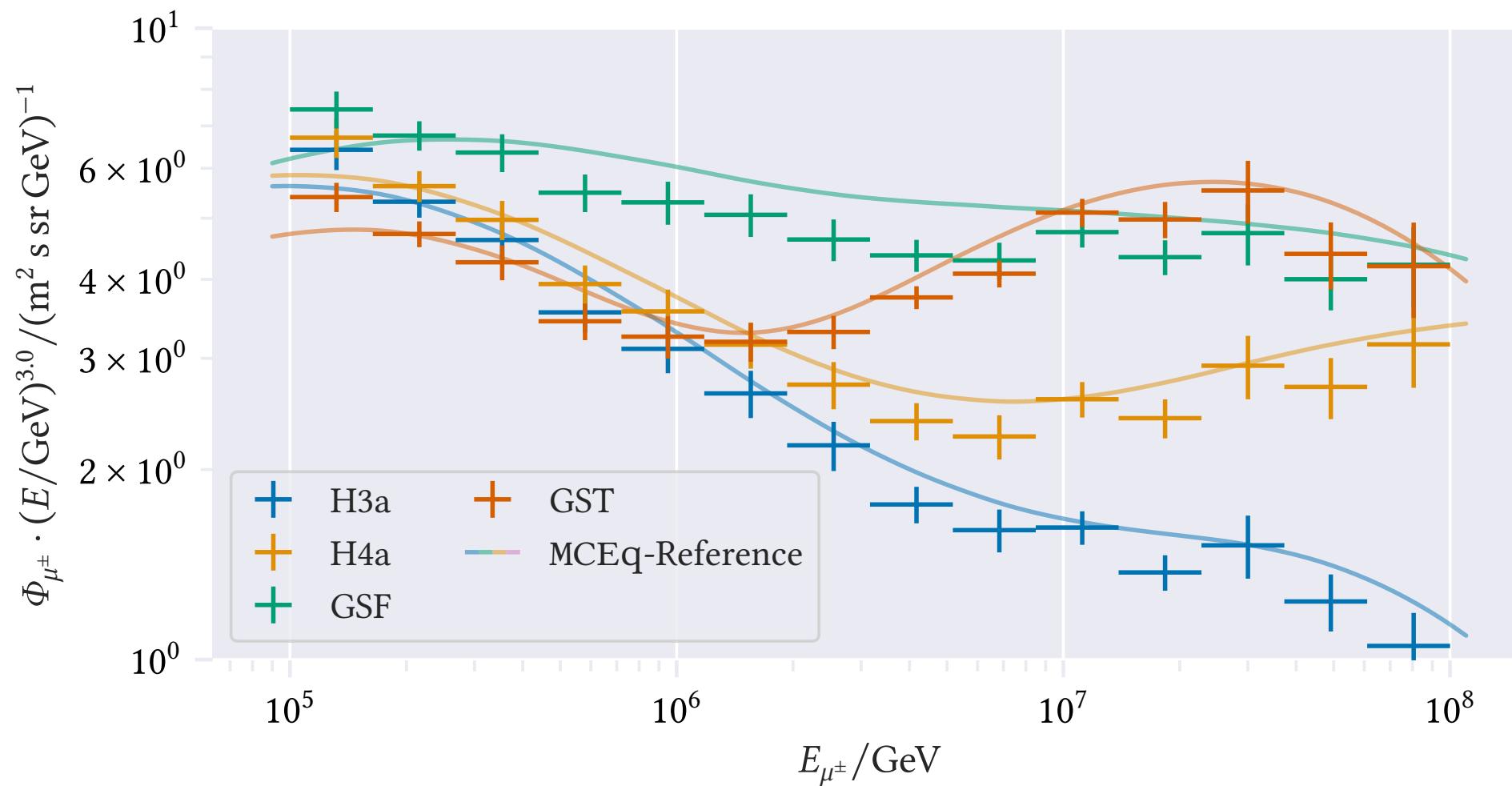


- Good agreement in total prompt and conv muon flux



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

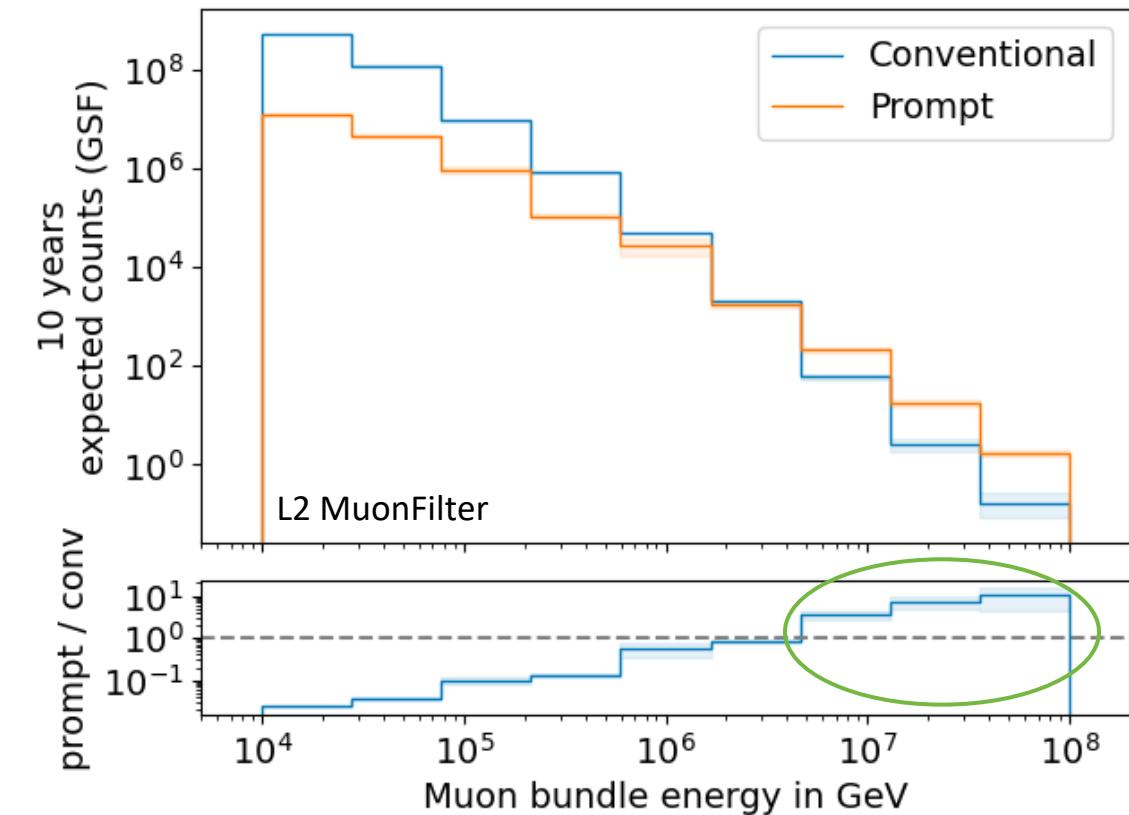
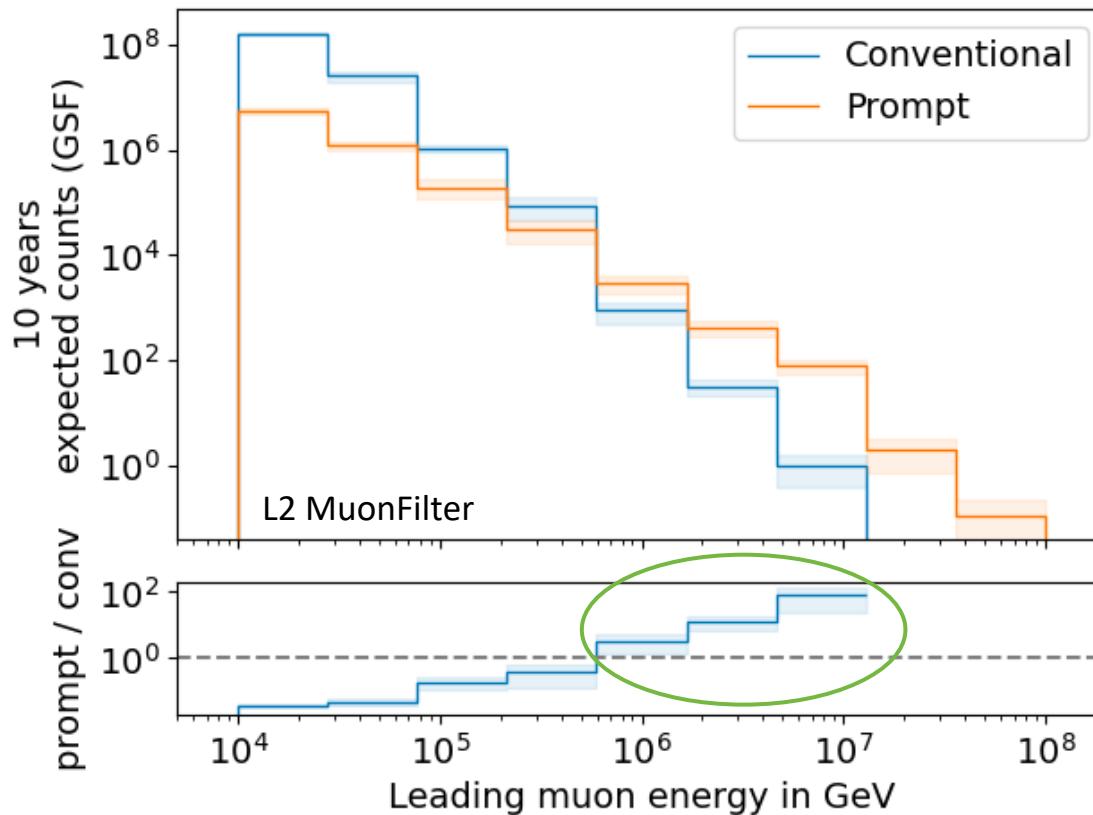
Agreement for different primary models



Pseudo analysis

Expected muons for 10 years: leading vs. bundle energy (GSF)

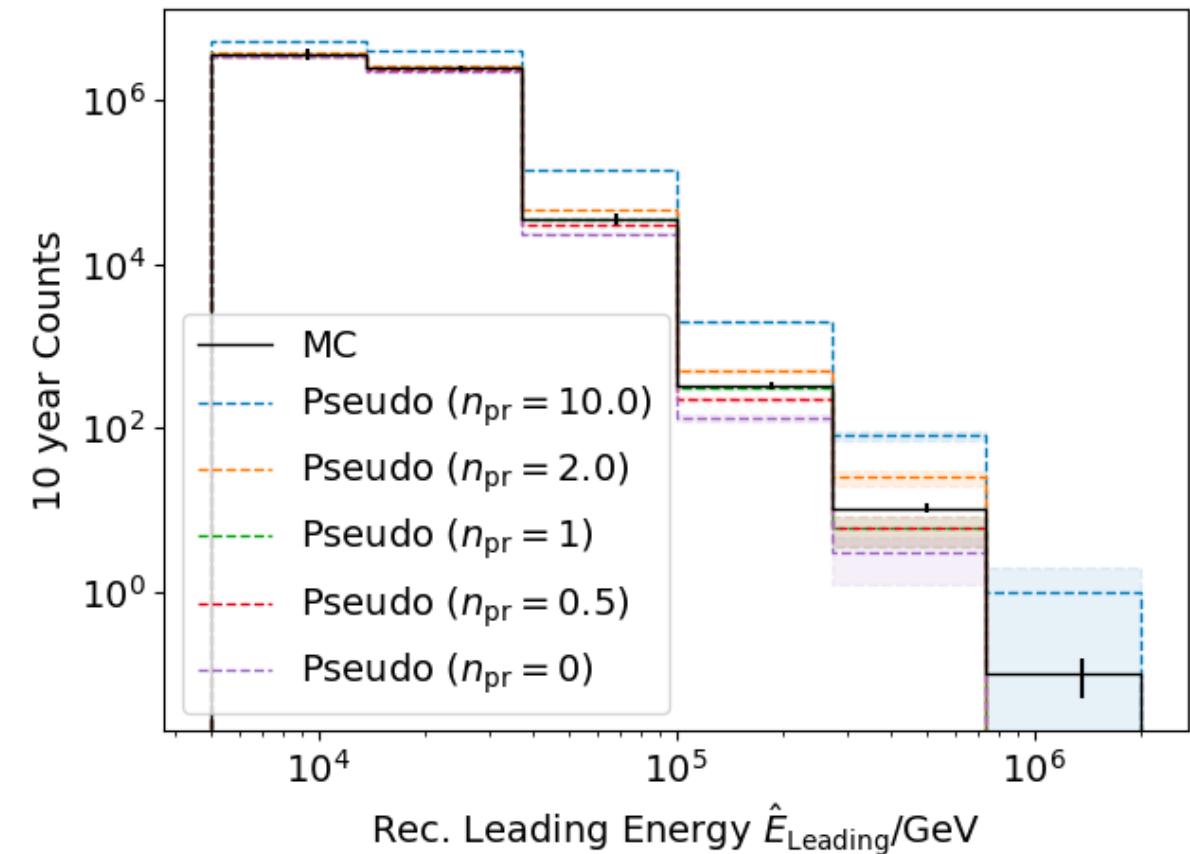
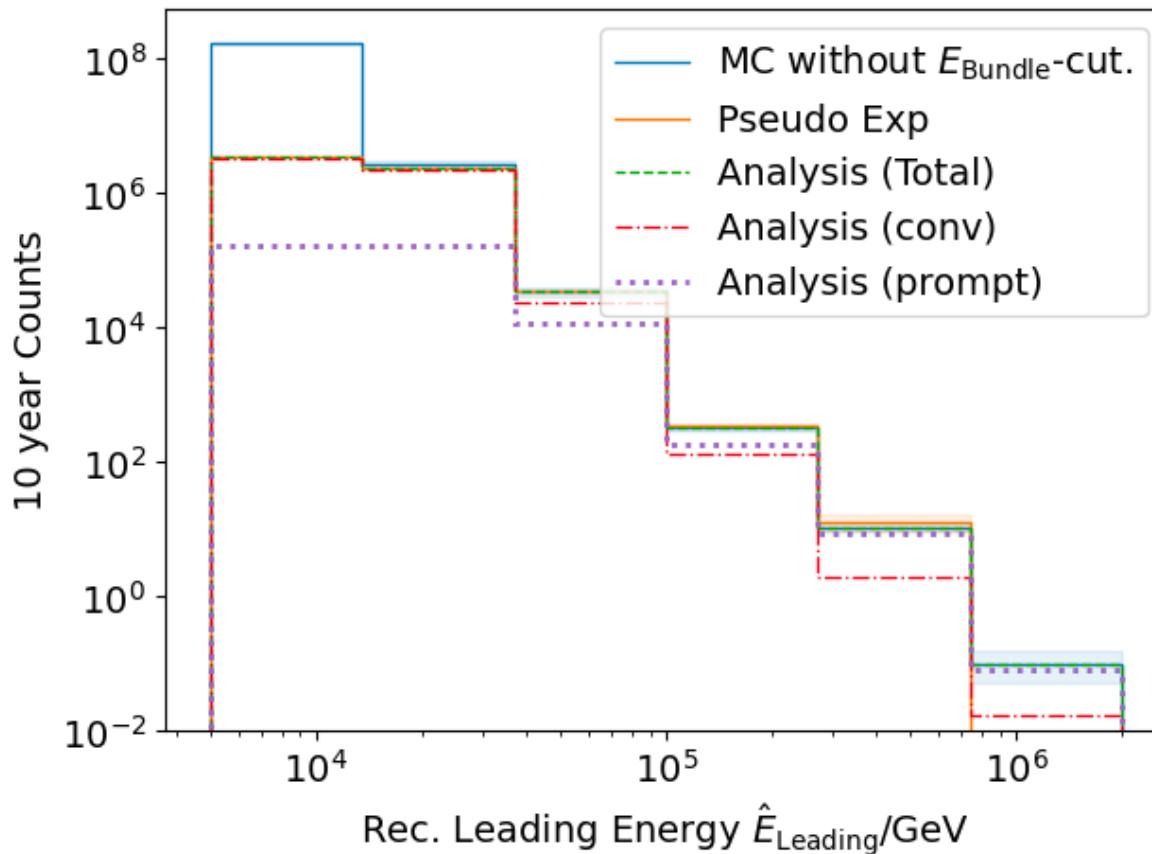
- leading: energy of most energetic muon in a muon bundle
- bundle: sum of energies of all muons of the bundle



- Both leading and bundle energy are sensitive to detect prompt
- Leading muon energy is more sensitive

Pseudo data sampling

Cuts:
L2 MuonFilter
Bundle energy > 100 TeV



➤ Tagging allows scaling of prompt by factor n_{pr}

Poisson likelihood fit performed in leading muon energy

Cuts:
 L2 MuonFilter
 Bundle energy > 100 TeV

Prompt scaling/normalization

MC counts per bin i

$$C_1^{\text{MC}} = n_{\text{pr}} C_1^{\text{MC,pr}} + n_{\text{conv}} C_1^{\text{MC,conv}}, \dots, C_M^{\text{MC}} = n_{\text{pr}} C_M^{\text{MC,pr}} + n_{\text{conv}} C_M^{\text{MC,conv}}$$

Conv norm = 1

Experimental counts

$$p(C_i) = p_{\text{poisson}}(C_i; \lambda(n_{\text{pr}}) = C_i^{\text{MC}}(n_{\text{pr}})) = \frac{\lambda(n_{\text{pr}})^{C_i} e^{-\lambda(n_{\text{pr}})}}{C_i!}$$

Maximize likelihood

$$\mathcal{L}(n_{\text{pr}}) = \prod_{i=1}^M p(C_i; n_{\text{pr}})$$

Easier:
 minimize negative
 log-likelihood

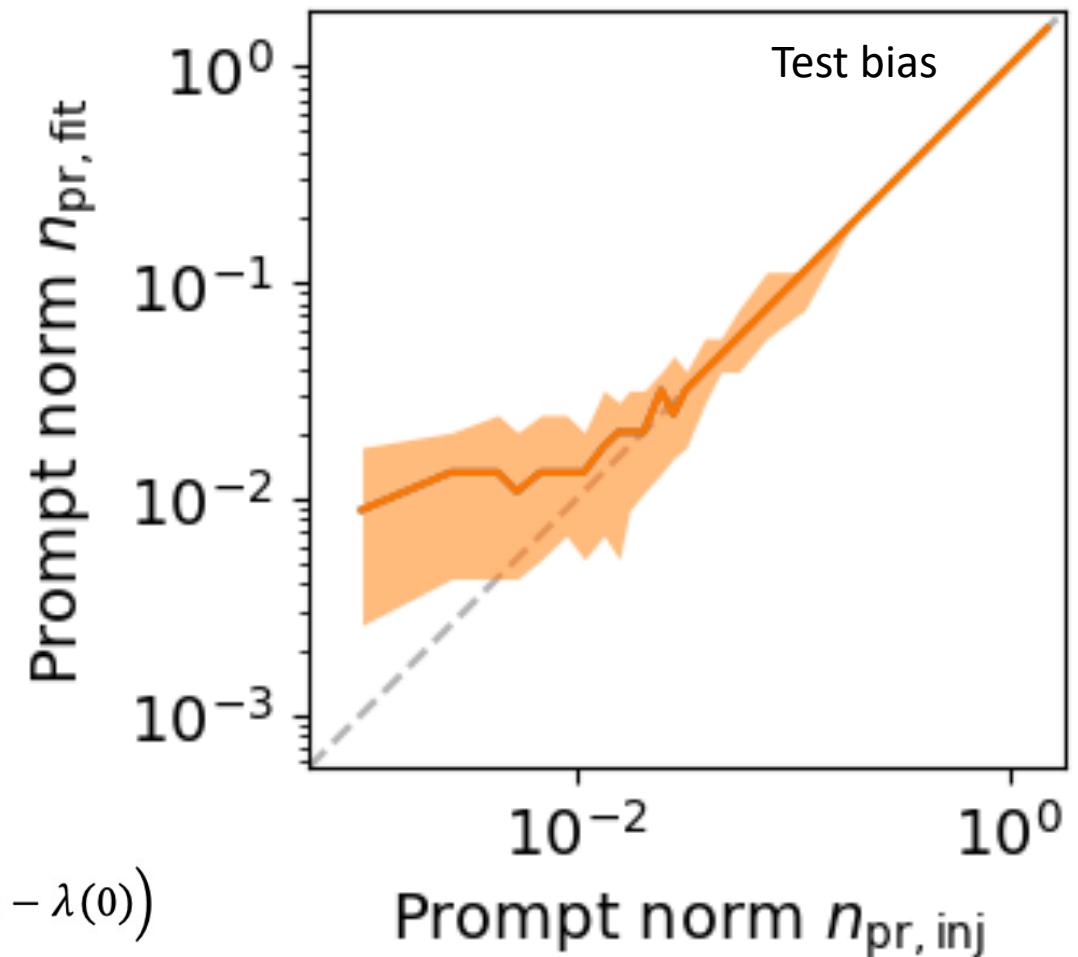
$$-\ln \mathcal{L} = -\sum_{i=1}^M C_i \ln \lambda(n_{\text{pr}}) - \lambda(n_{\text{pr}}) - \ln C_i!$$

$\Lambda = -2 \ln \frac{\mathcal{L}(n_{\text{pr}} = \hat{n}_{\text{pr}})}{\mathcal{L}(n_{\text{pr}=0})} = -2 \sum_{i=1}^M C_i (\ln \lambda(\hat{n}_{\text{pr}}) - \ln \lambda(0)) - (\lambda(n_{\text{pr}}) - \lambda(0))$

Test statistic for Wilks' theorem

Null hypothesis: no prompt

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➤ Bias starts at a prompt normalization of 0.1

Discovery potential and sensitivity

Cuts:
L2 MuonFilter
Bundle energy > 100 TeV

Expectation for 1 year:

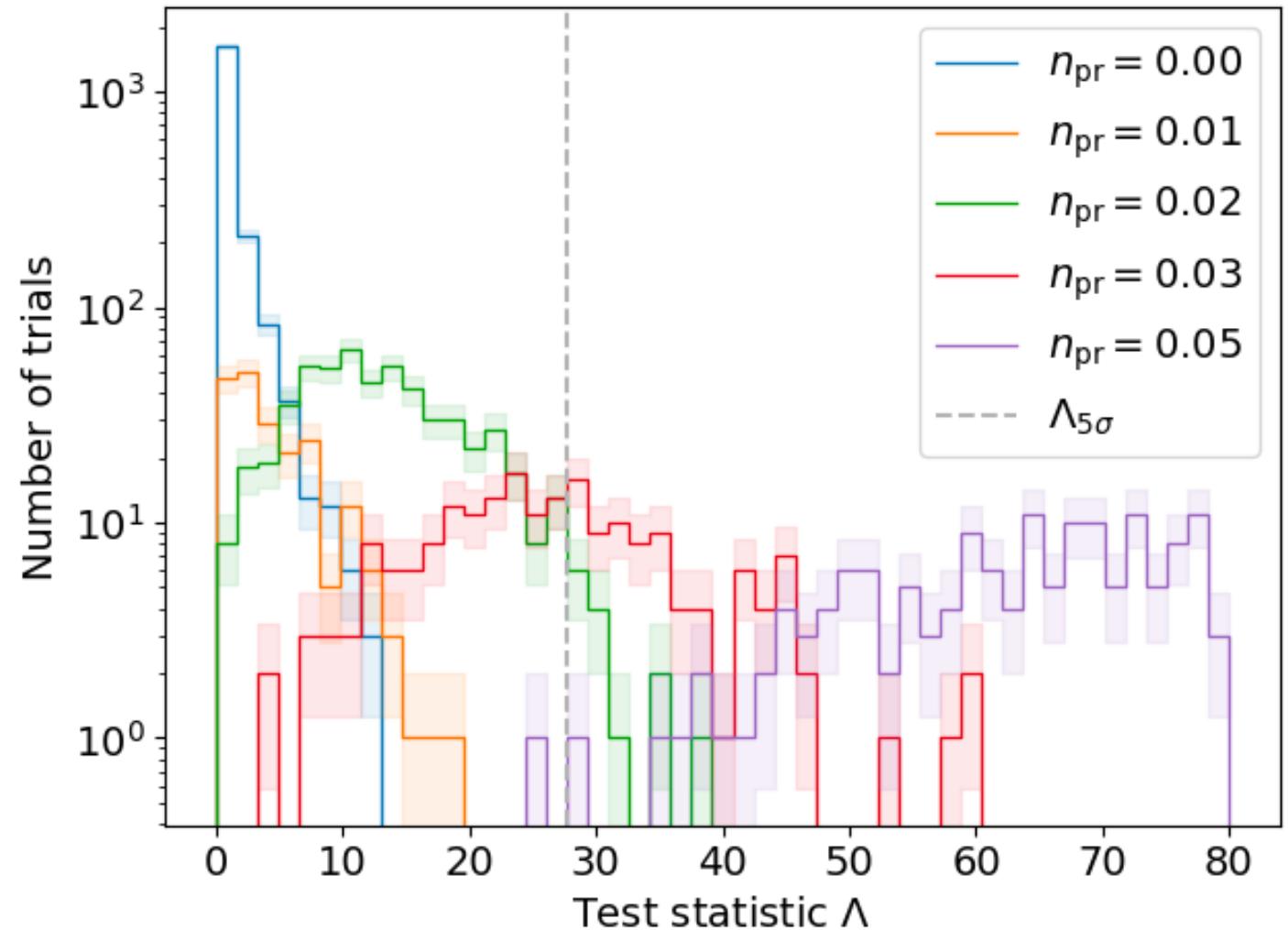
- 5 sigma discovery potential: 0.102 ± 0.005
- Sensitivity: 0.024 ± 0.001

Expectation for 10 years:

- 5 sigma discovery potential: 0.032 ± 0.001
- Sensitivity: 0.007 ± 0.000

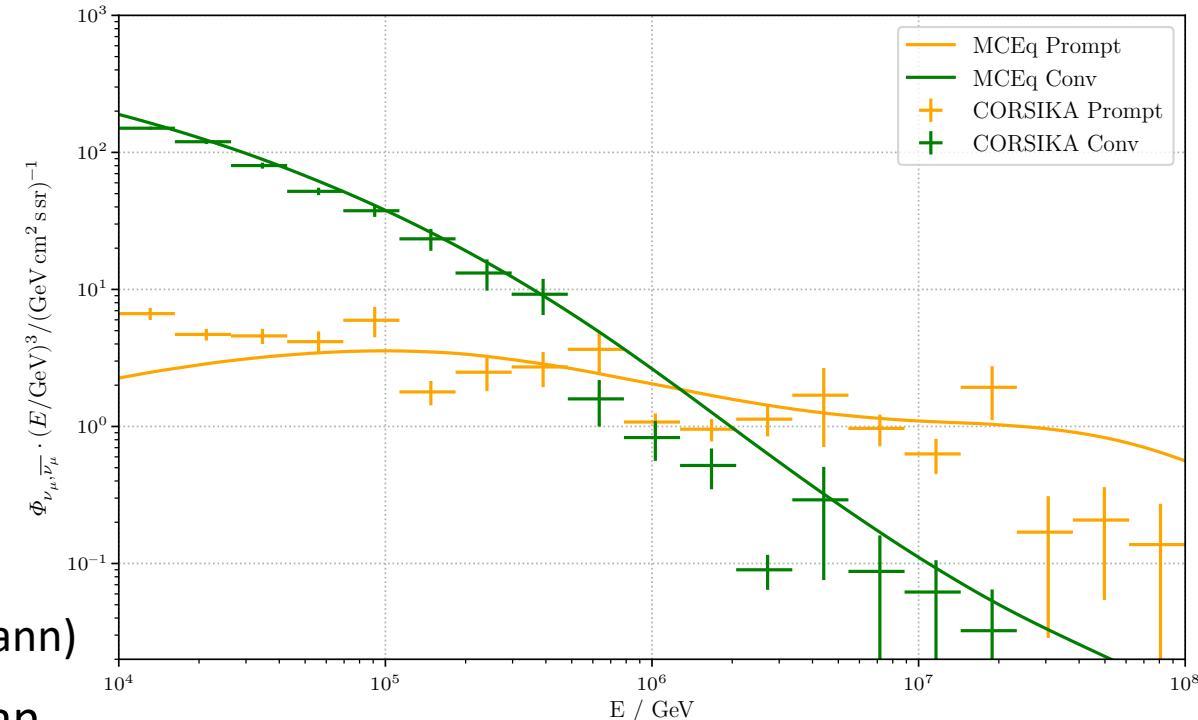
Caution:

- Limited MC statistics -> events are oversampled in pseudo dataset
- No systematic uncertainties



Conclusion and outlook

- CORSIKA 7 test simulations
 - Prompt identification
 - MCEq comparison
 - ★ Few-author paper in progress (publish early 2024)
- First analysis chain for prompt muon normalization
- ★ Proceed analysis... (systematics etc.)
- IceCube prompt muons paper (publish early 2025)
- Prompt neutrino analysis
 - ★ Tagging and MCEq comparison in progress (Lars Bollmann)
- Combined fit (prompt muons + neutrinos)... future plan



F3/F4 Dortmund Meeting:

- <https://nextcloud.e5.physik.tu-dortmund.de/index.php/s/J5WGYQ6wBb9ndJM>