

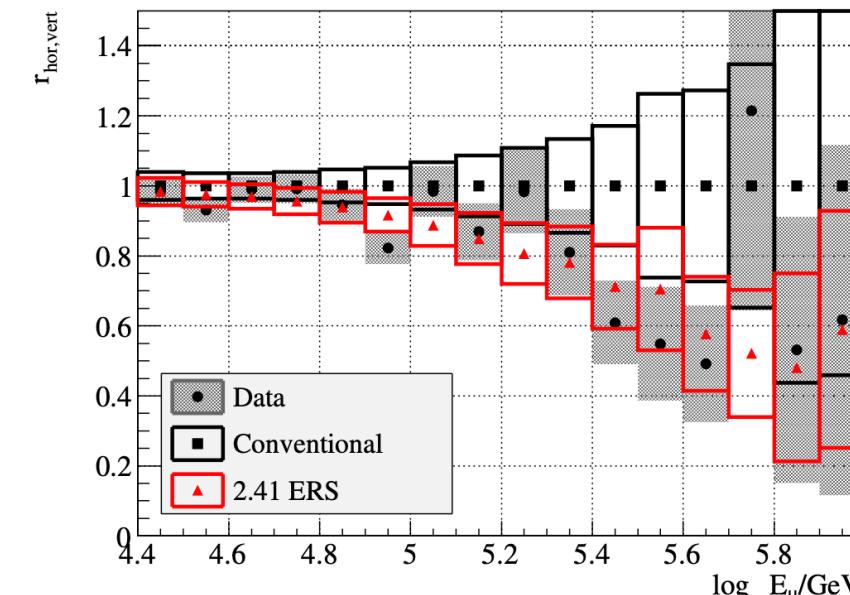
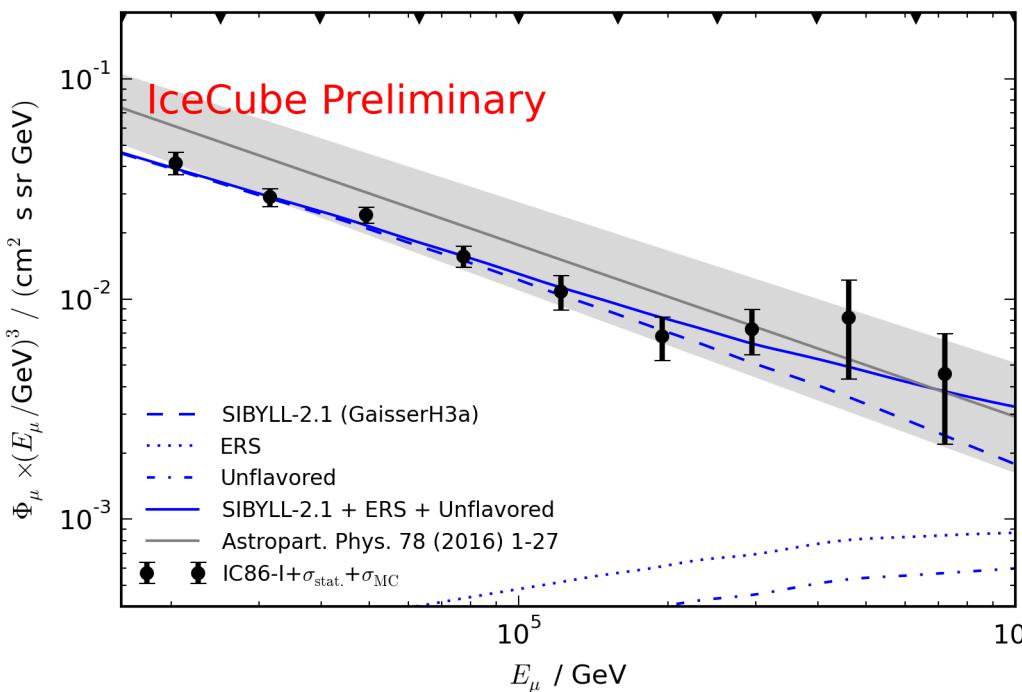
Measurement of the prompt component of the atmospheric muon flux

Ludwig Neste and Pascal Gutjahr

IceCube Fall Collaboration Meeting Grand Rapids 2023

Motivation

- Detect and measure normalization of prompt component of atmospheric muon flux
- Constrain uncertainties on hadronic interaction models at very high energies
- Old analyses:
 - Leading muon analysis: limited MC statistics (by Tomasz Fuchs, https://wiki.icecube.wisc.edu/index.php/Analysis_of_Leading_Muons)
 - Characterization of the muon flux: zenith problem (by Patrick Berghaus, <https://arxiv.org/abs/1506.07981>)



Sample	Best Fit (ERS)	1σ Interval (90% CL)	$\sigma(\Phi_{\text{prompt}} > 0)$
Uncorrected	4.93	4.05-5.87 (3.55-6.56)	9.43
Marginalized Ang. Corr.	3.19	1.64-5.48 (0.98-7.26)	3.46

- Good analyses with some challenges
- Prompt MC truth not available -> new CORSIKA simulations

Overview

Final goals

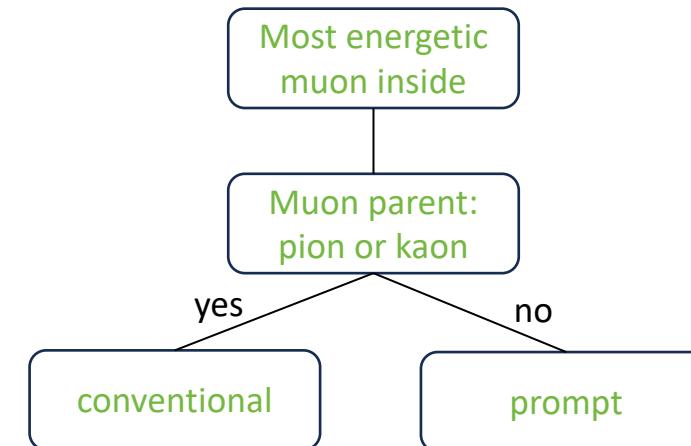
- Measure normalization of the atmospheric prompt muon flux
- Unfold muon energy spectrum

Steps for normalization measurement

- Verify CORSIKA extended history simulations
- Tag prompt muons
- Comparisons to MCEq
- Set up preliminary analysis chain
- Reconstruct muon energy and direction
- Data/MC comparisons
- Include systematics
- Run full statistics simulation

Additional steps for unfolding

- Effective area



Terminology

- Muon bundle: all muons in a bundle
- Leading muon: most energetic muon in a muon bundle
- Single muon: no single muons at high energies
- Prompt muon: parent is pion or kaon

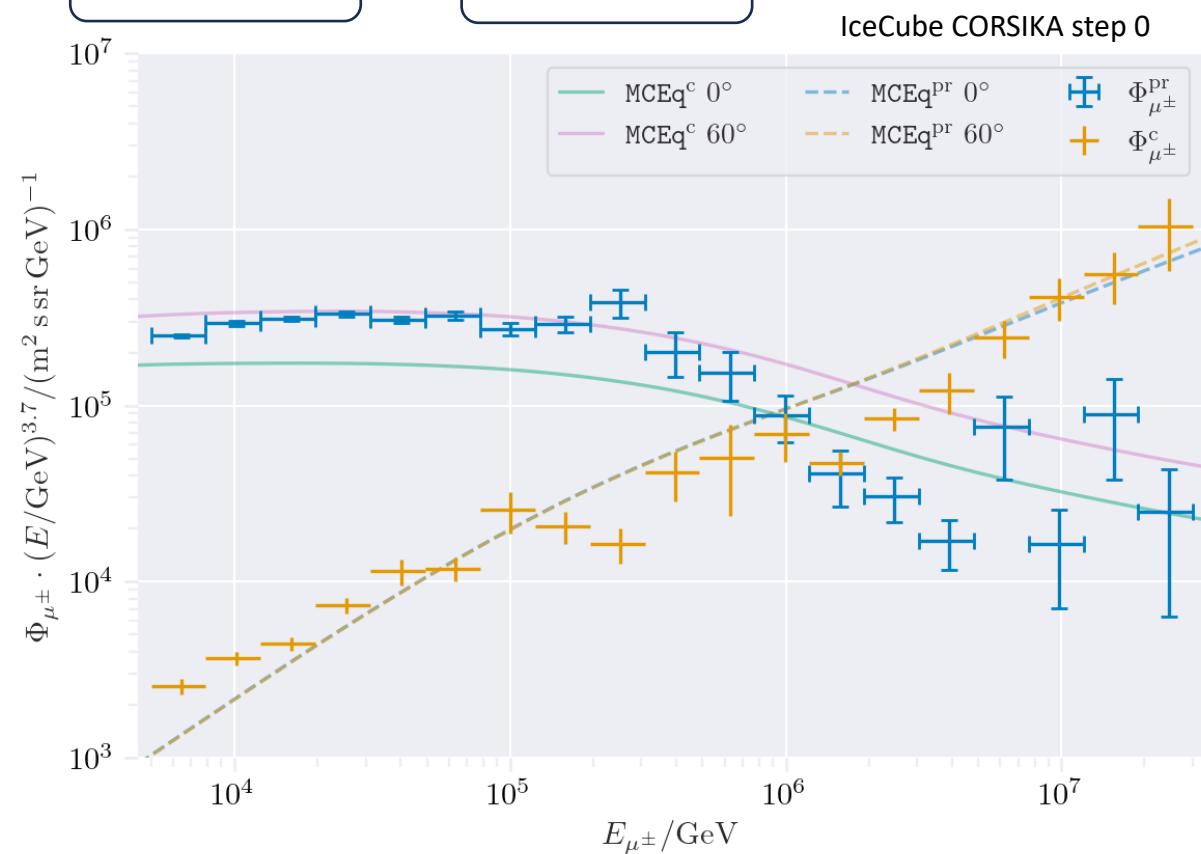
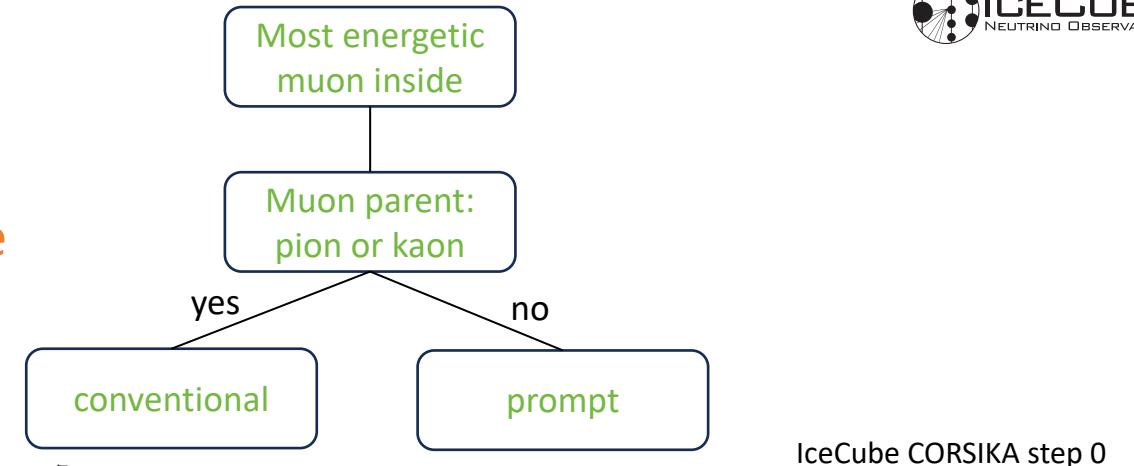
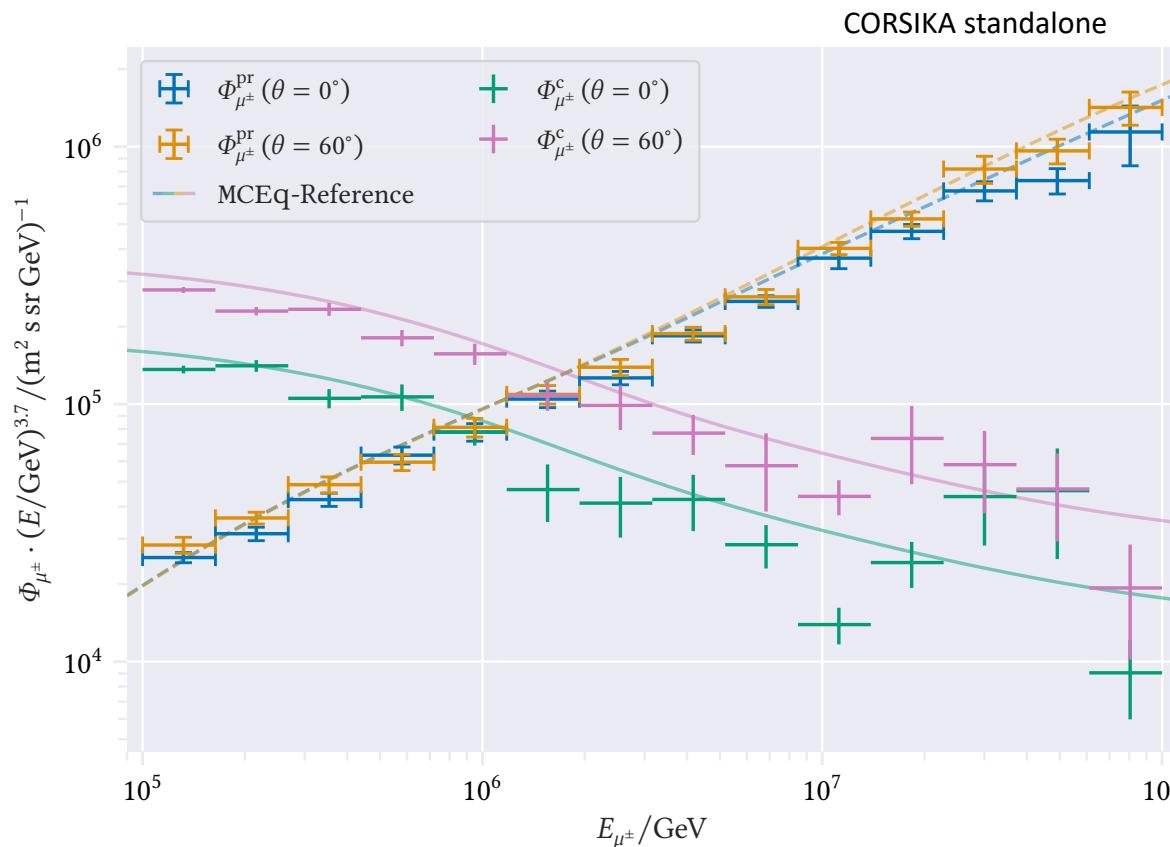
New (preliminary) CORSIKA Ehist simulation

- CORSIKA 77420
- SIBYLL 2.3d
- 600 GeV – 50 EeV
- `/data/sim/IceCube/2023/generated/CORSIKA_EHISTORY/`

CORSIKA vs. MCEq

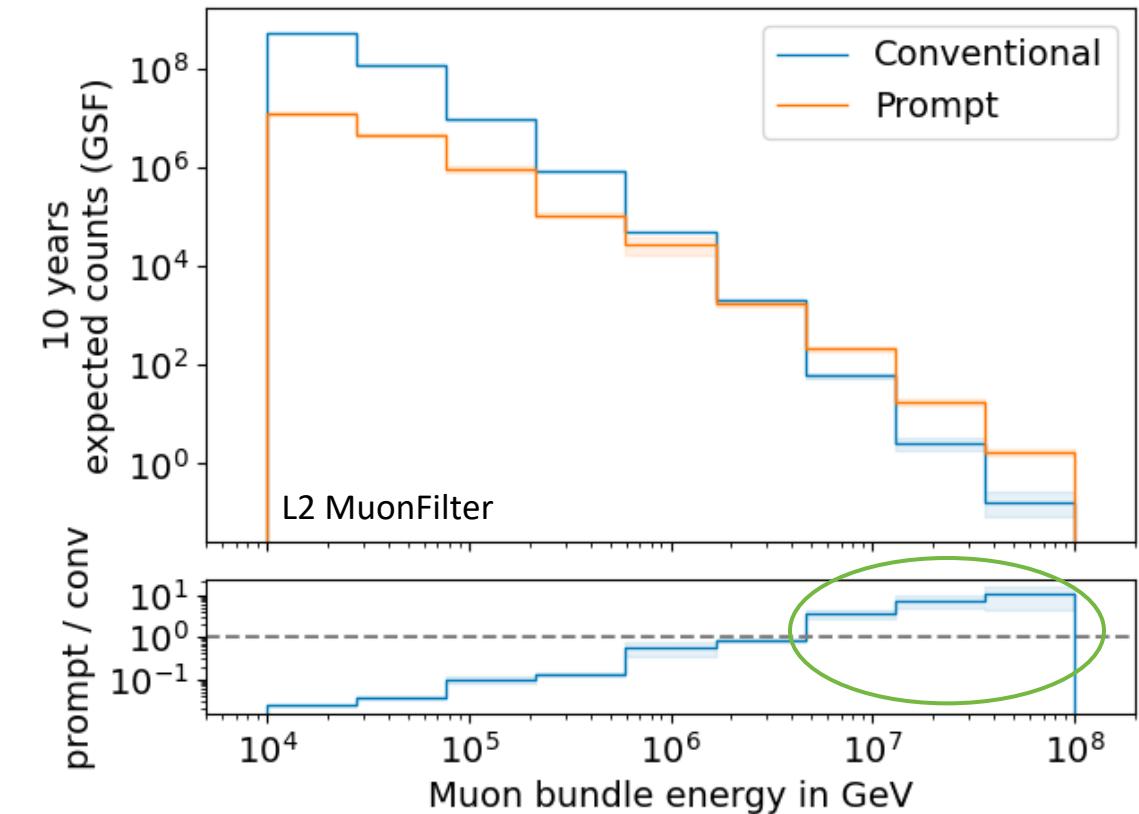
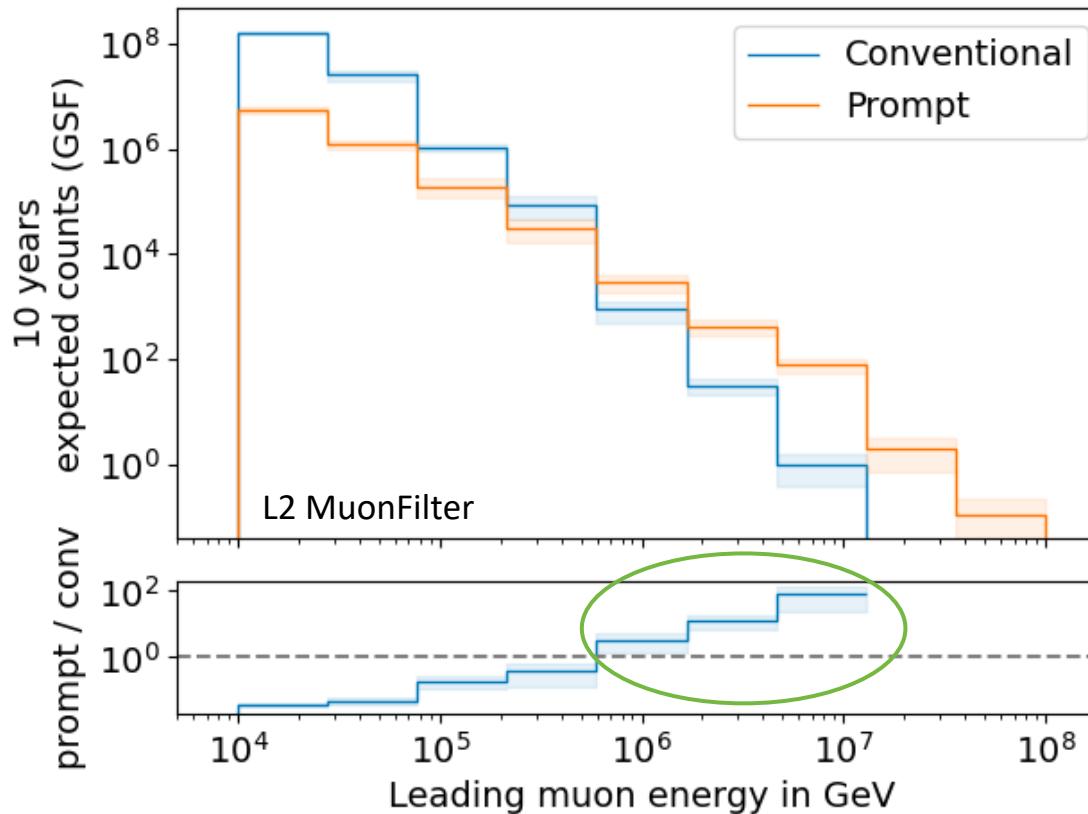
➤ Presented by Ludwig Neste

➤ Good agreement



MC data exploration

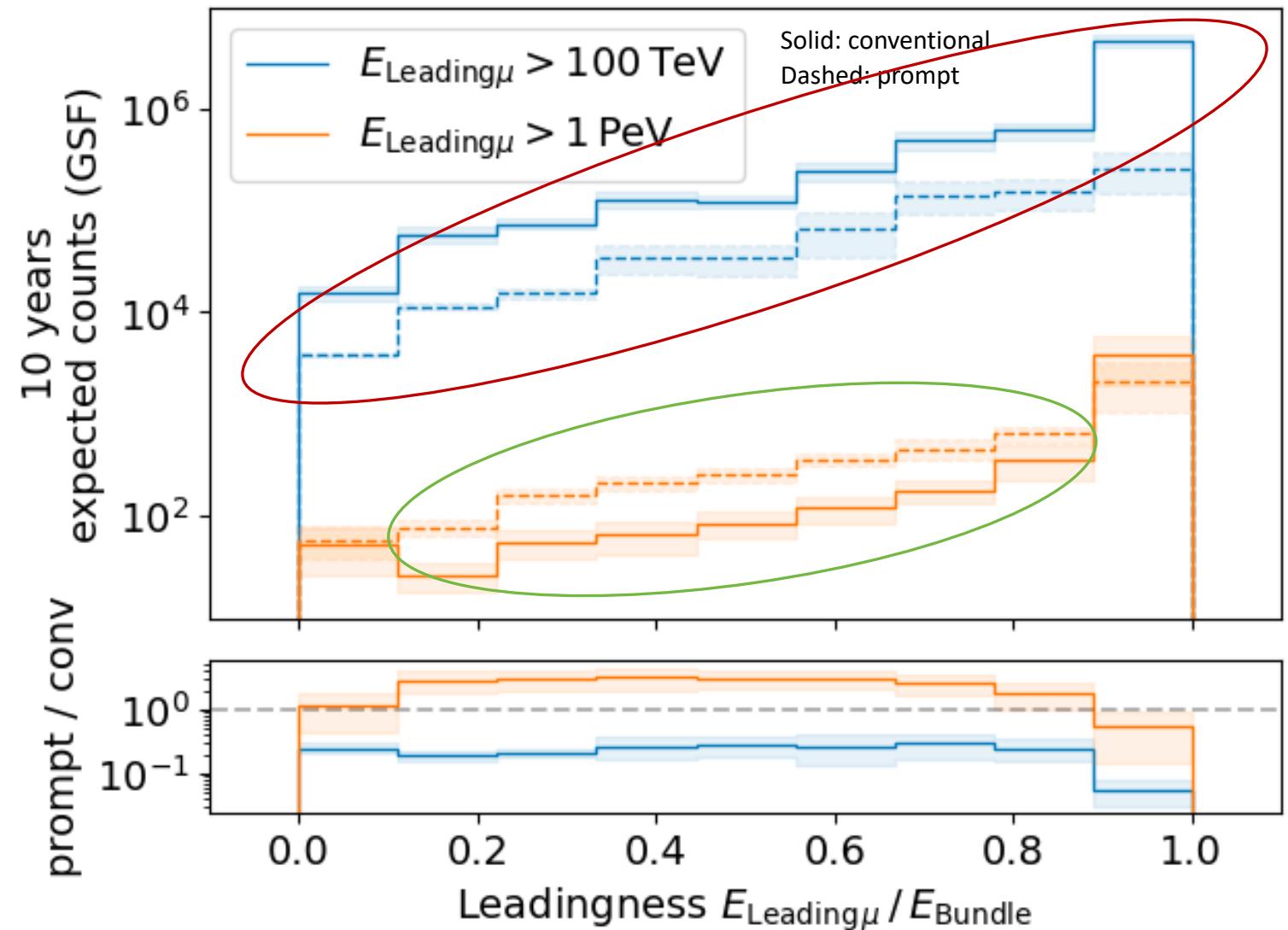
Expected muons for 10 years: leading vs. bundle energy (GSF)



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

Leading muon energy fraction

- Prompt dominates for energies $> 1 \text{ PeV}$
- Leading energy sweet spot: $0.1 - 0.9$



Leading muon contribution

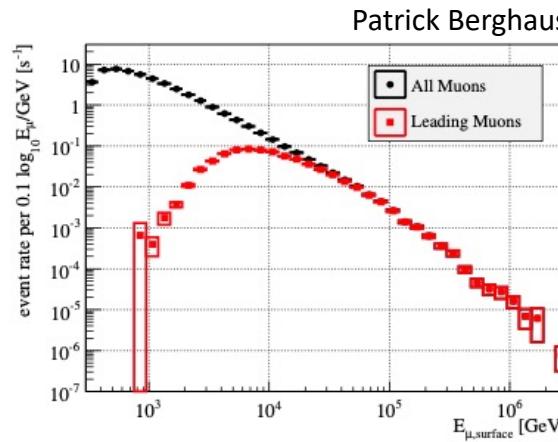
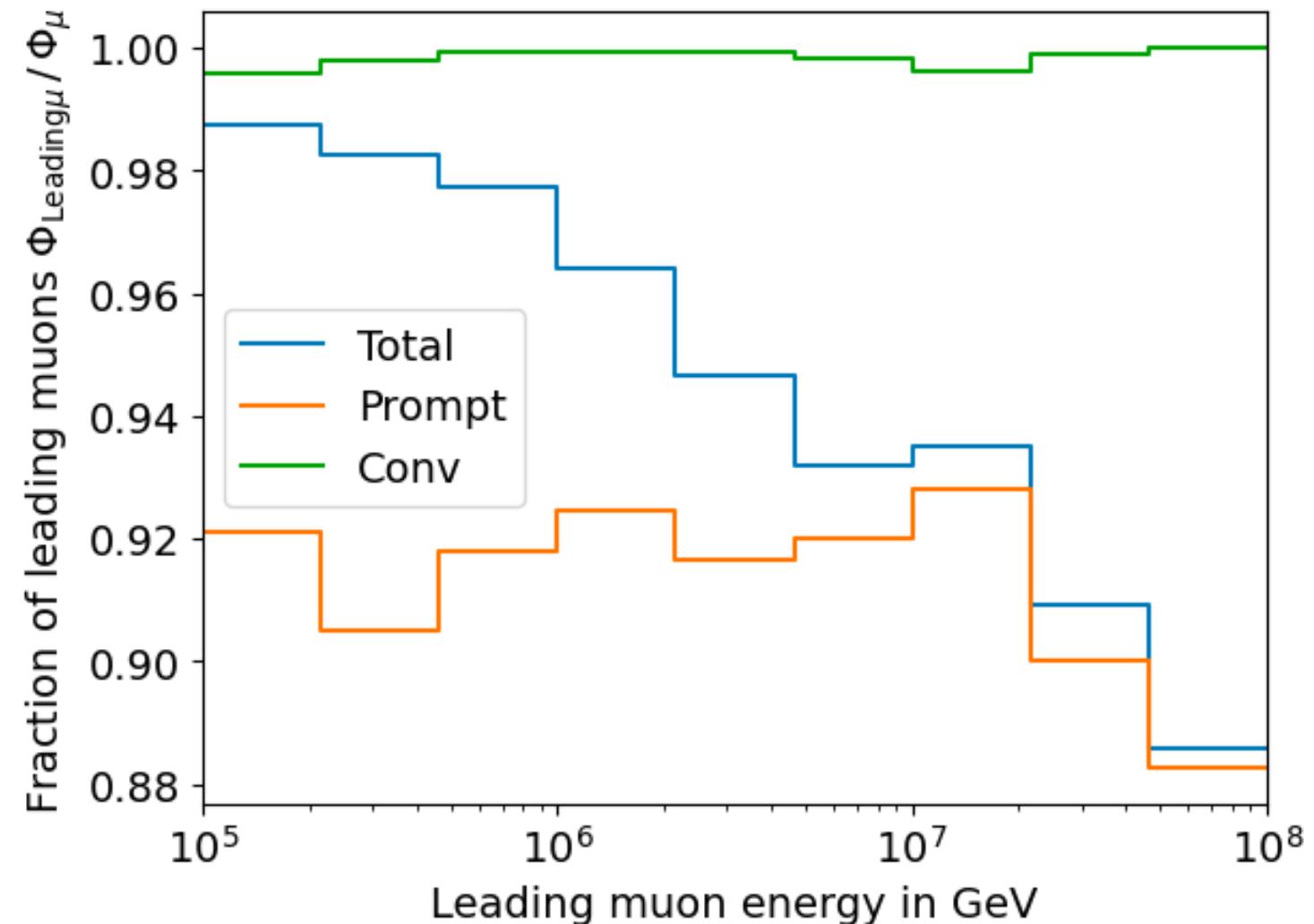


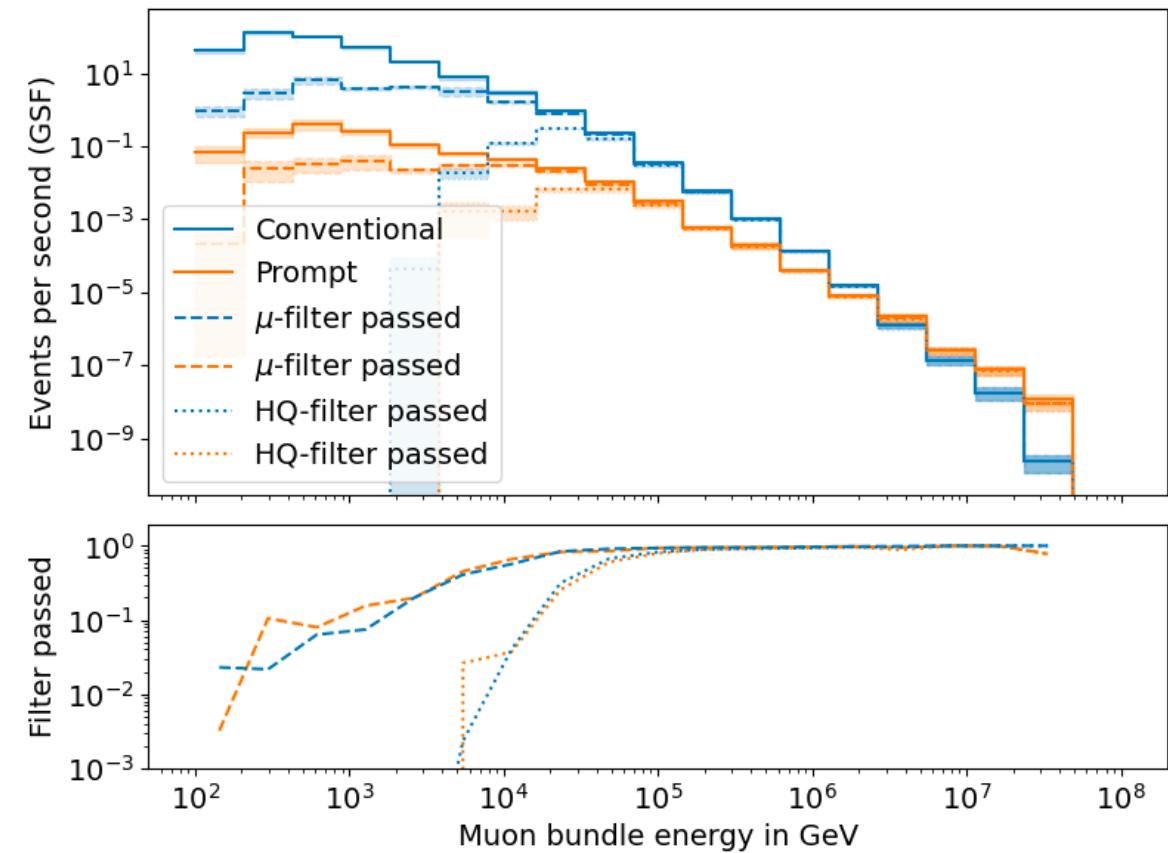
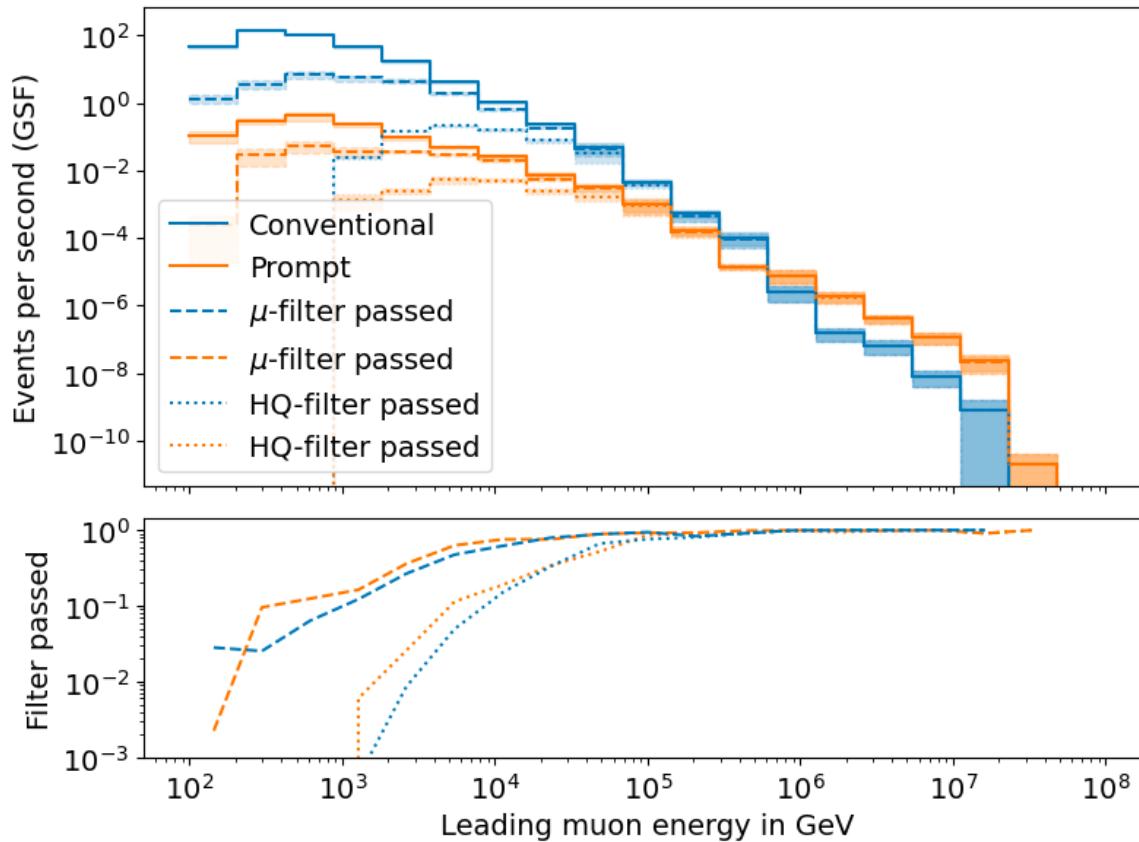
Figure 10: Surface energy distribution for all and most energetic (“leading”) muons in simulated events with a total of more than 1,000 registered photo-electrons in IceCube.

- Muons with energies between 100 TeV and 50 PeV dominate the bundle by more than 90%
 - In average conventional muons are more dominant than prompt
 - But: at high energies, there are more prompt than conventional events
- High leading energy fraction does not lead to more sensitivity to detect prompt



L2 Filters

Fraction events rejected	All energies	Leading energy > 10 TeV	Leading energy > 100 TeV
MuonFilter	0.93	0.28	0.06
HQFilter	0.99	0.74	0.18



DNN reconstructions

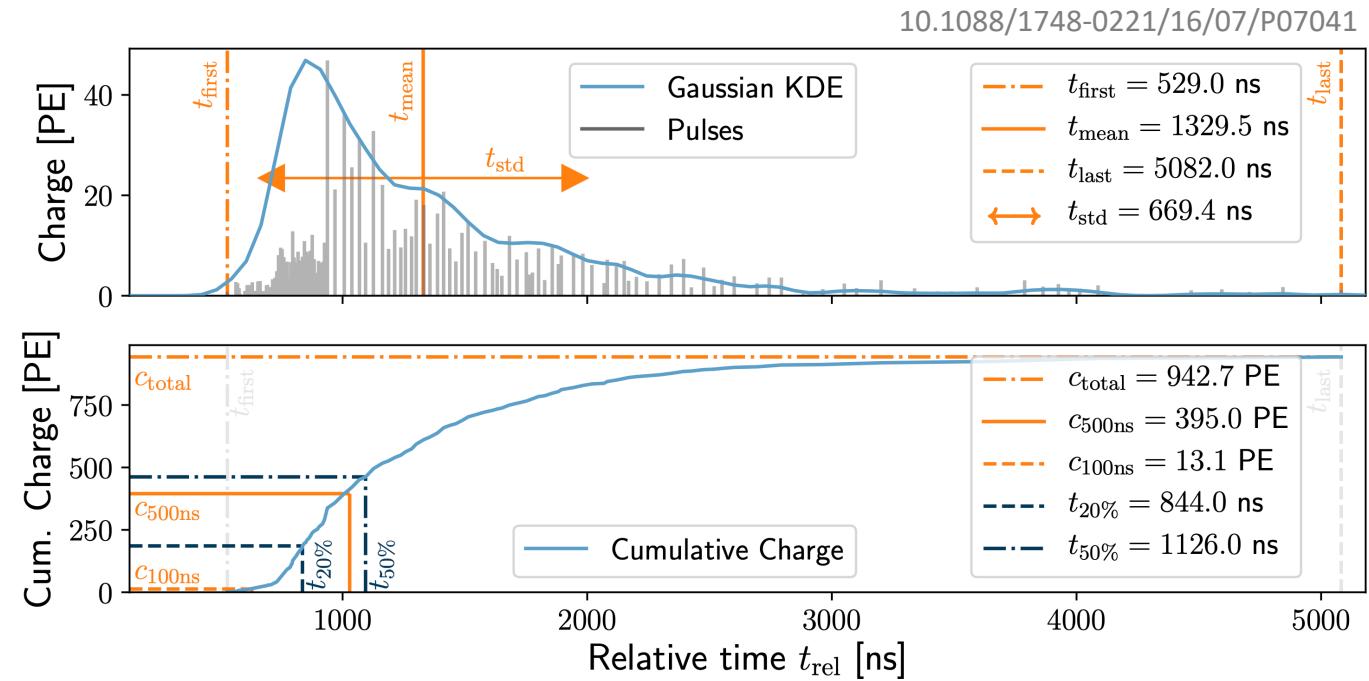
DNN Reconstructions

Reconstruct

- Muon bundle energy
- Leading muon energy
- Direction (zenith/azimuth)

Physics motivation

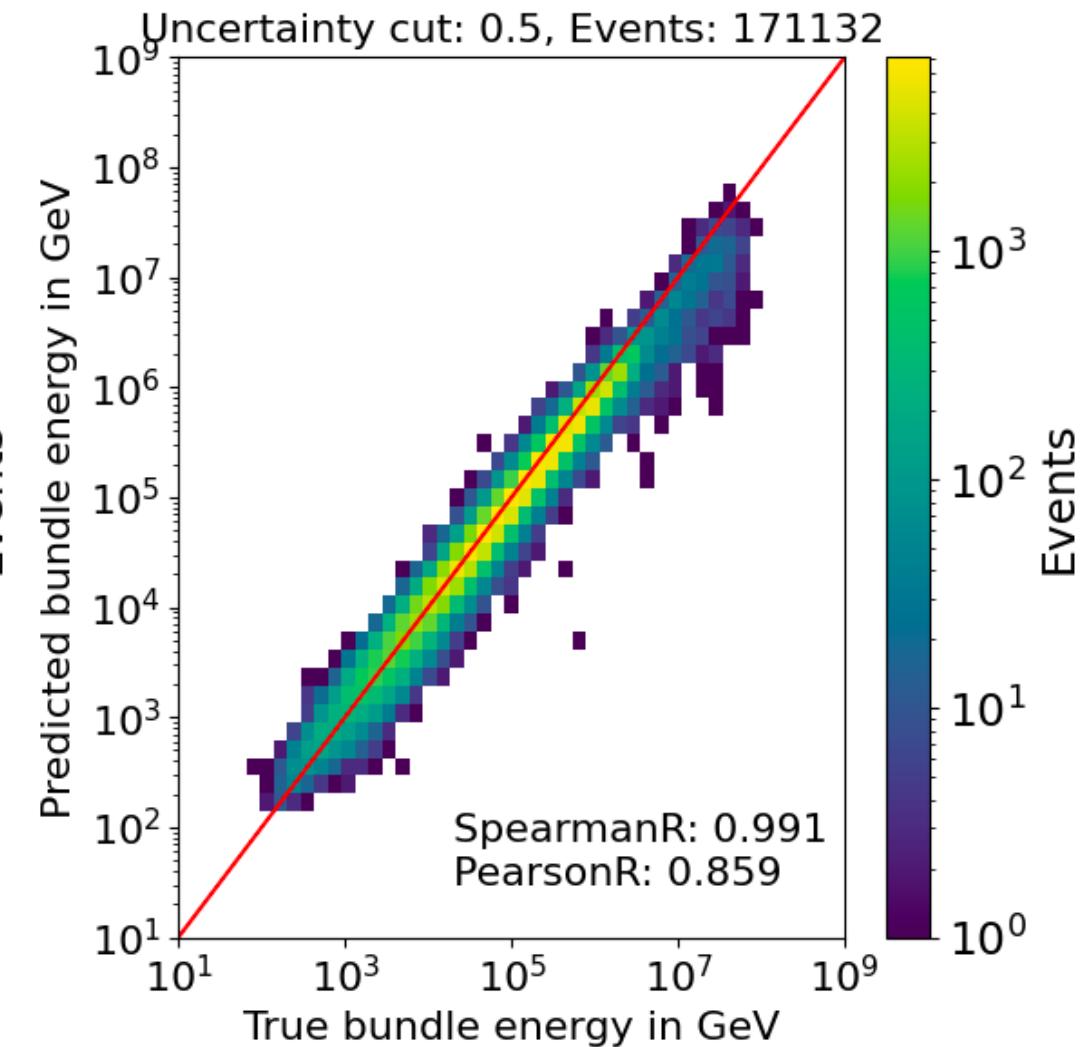
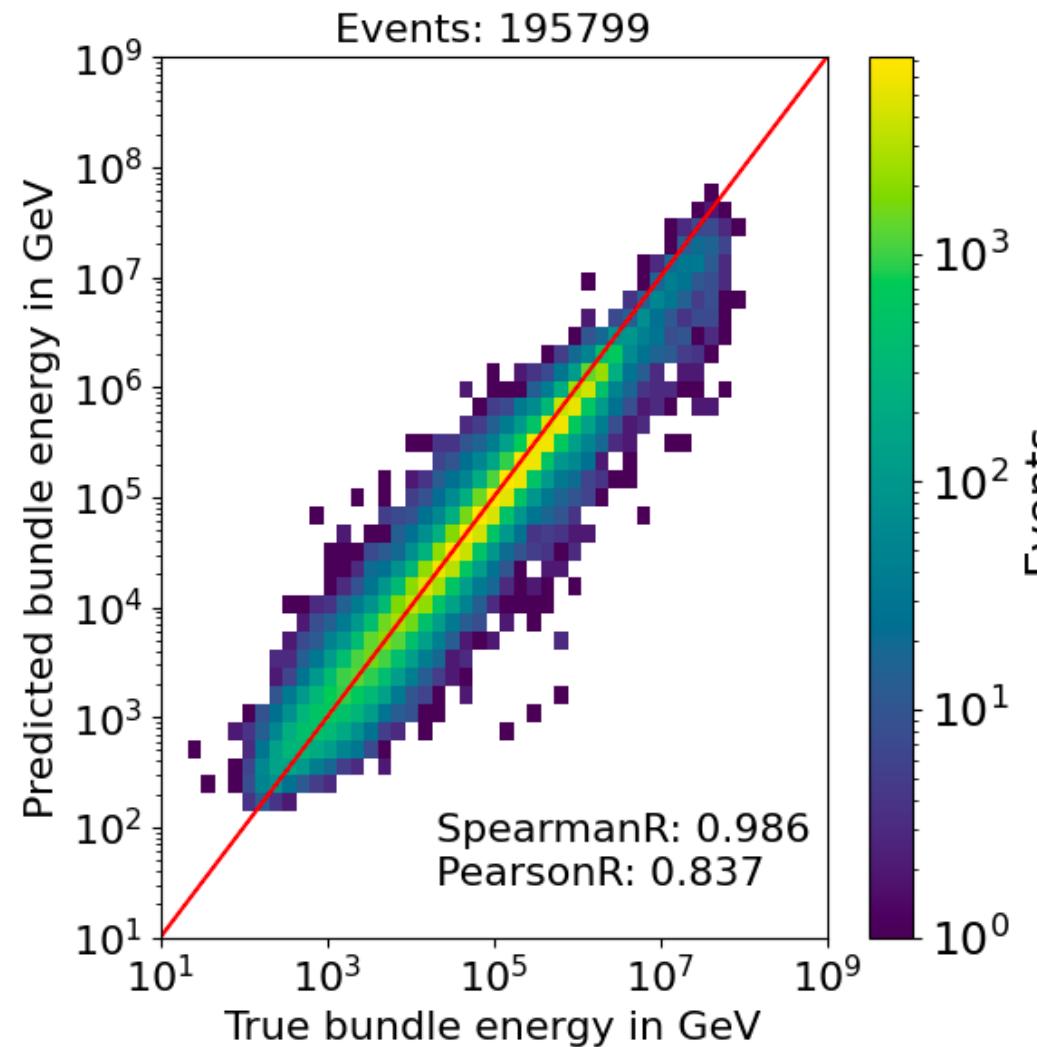
- Muons lose energy stochastically
 - High leadingness: energy depositions are dominated by large stochastic losses
 - Low leadingness: stochastic losses sum up and appear continuously
- High energies: forward production
 - High leadingness: small bundle radius
 - Low leadingness: larger bundle radius



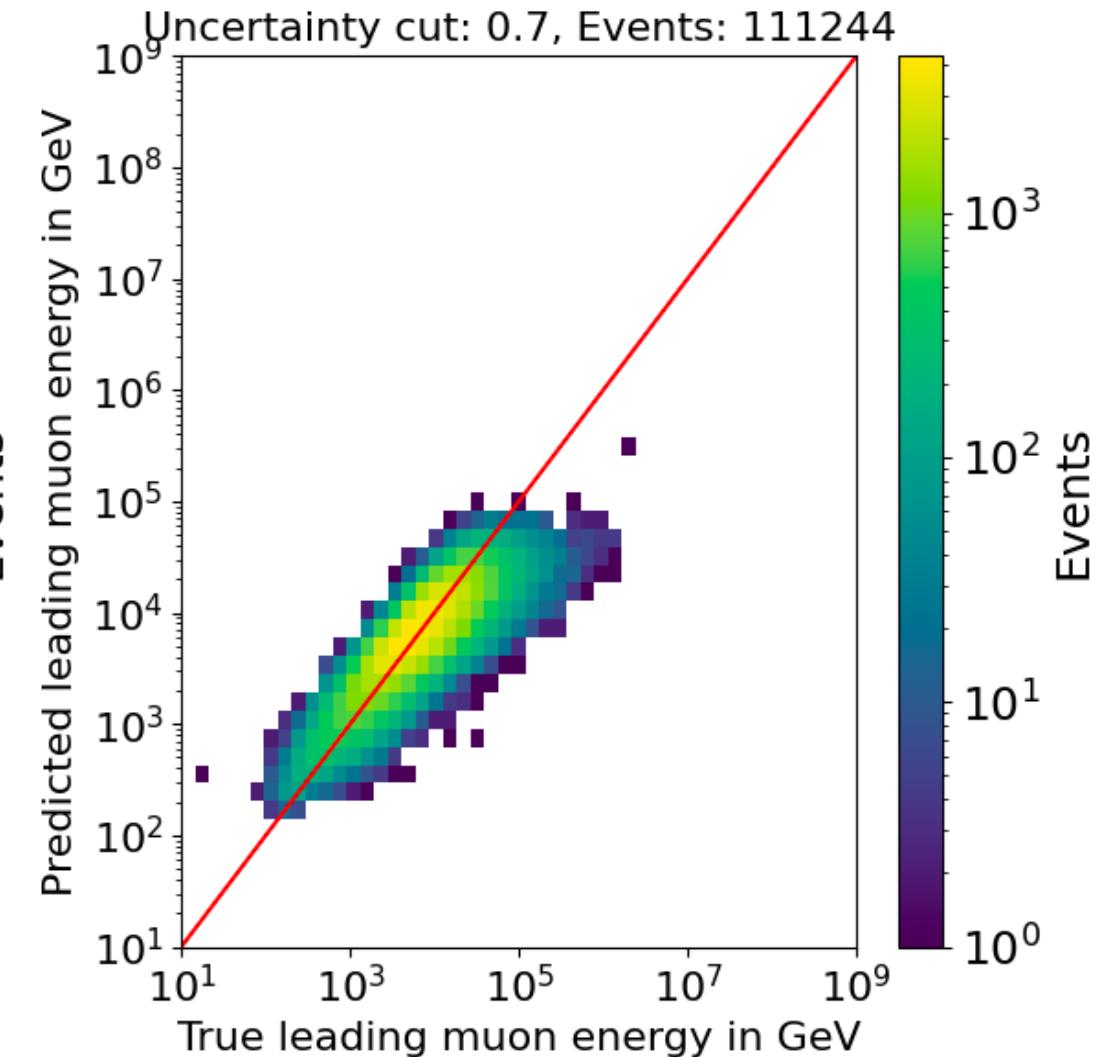
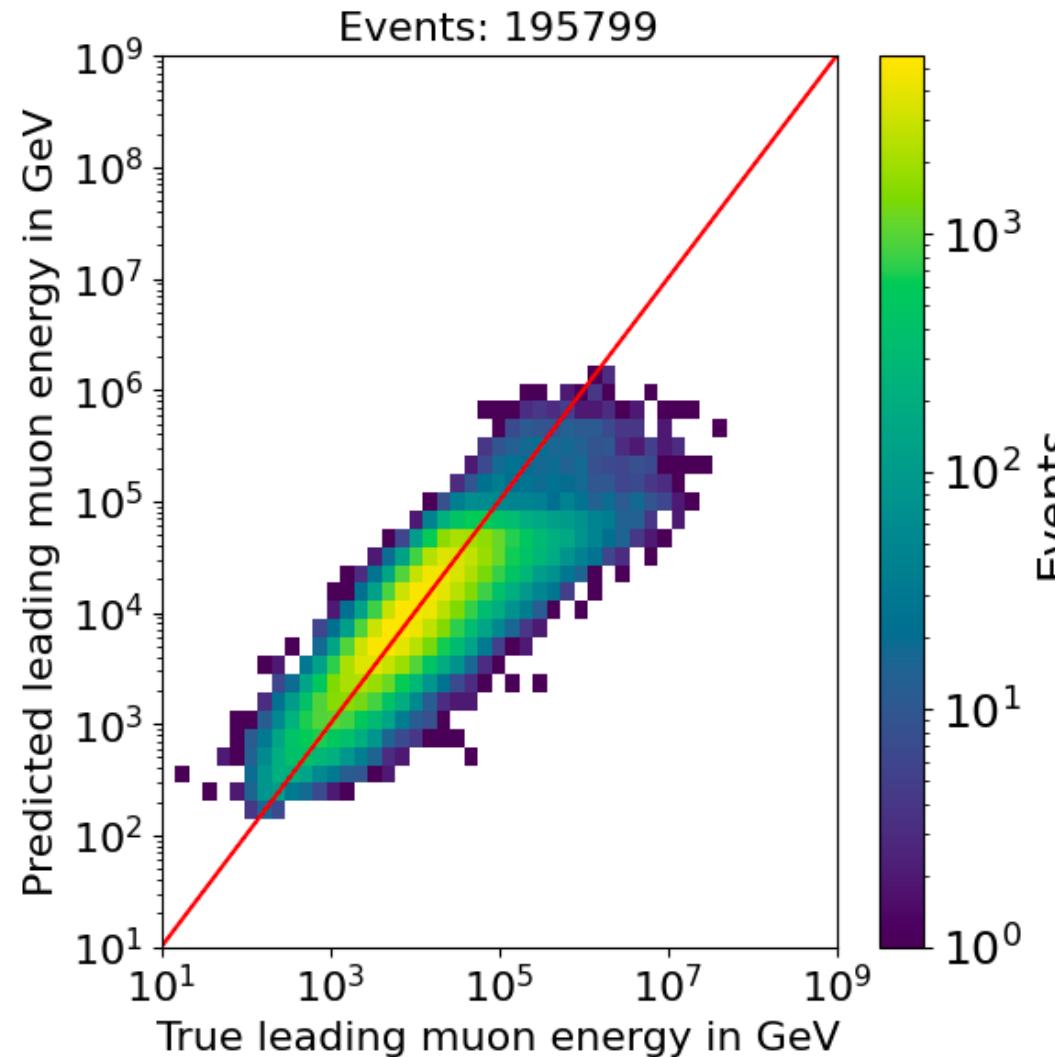
Input data per DOM

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

Bundle energy reconstruction

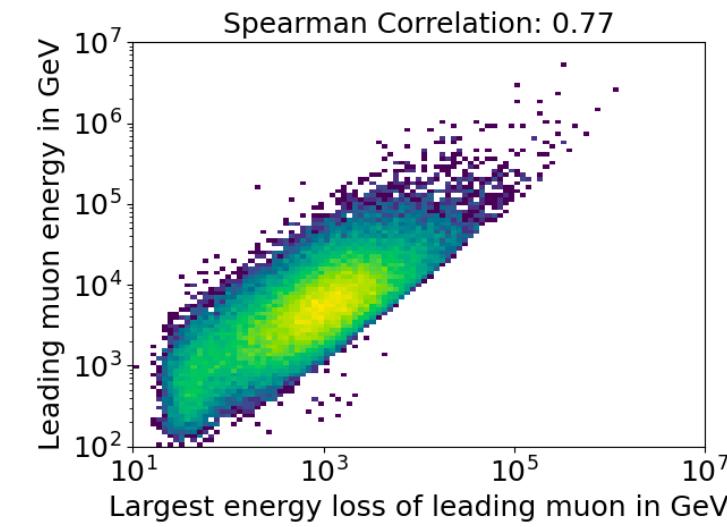
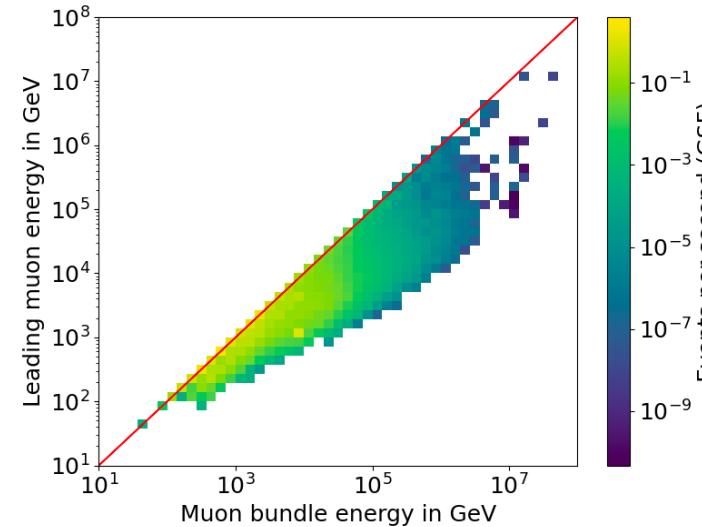
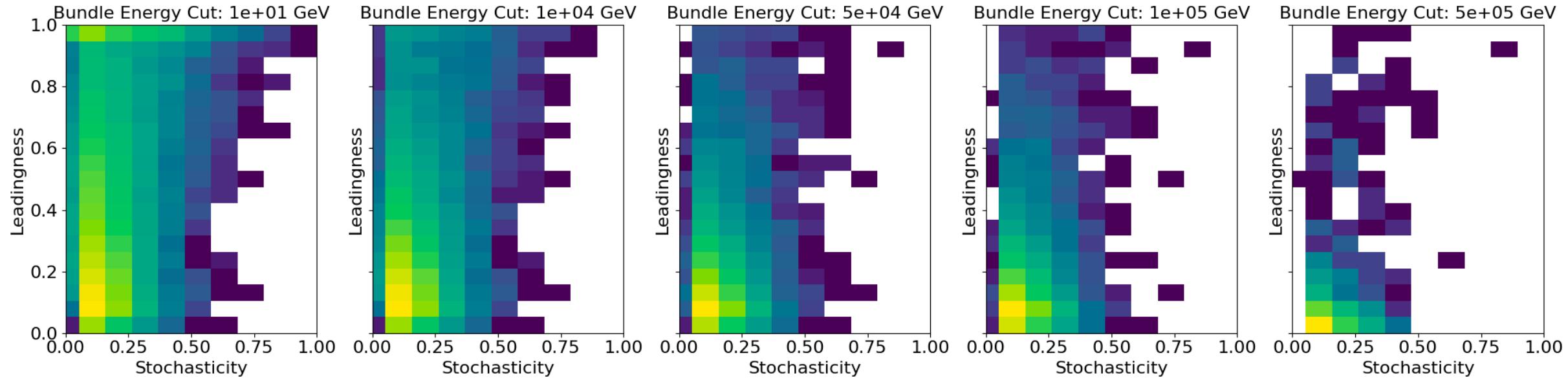


Leading energy reconstruction



Bundle energy cut

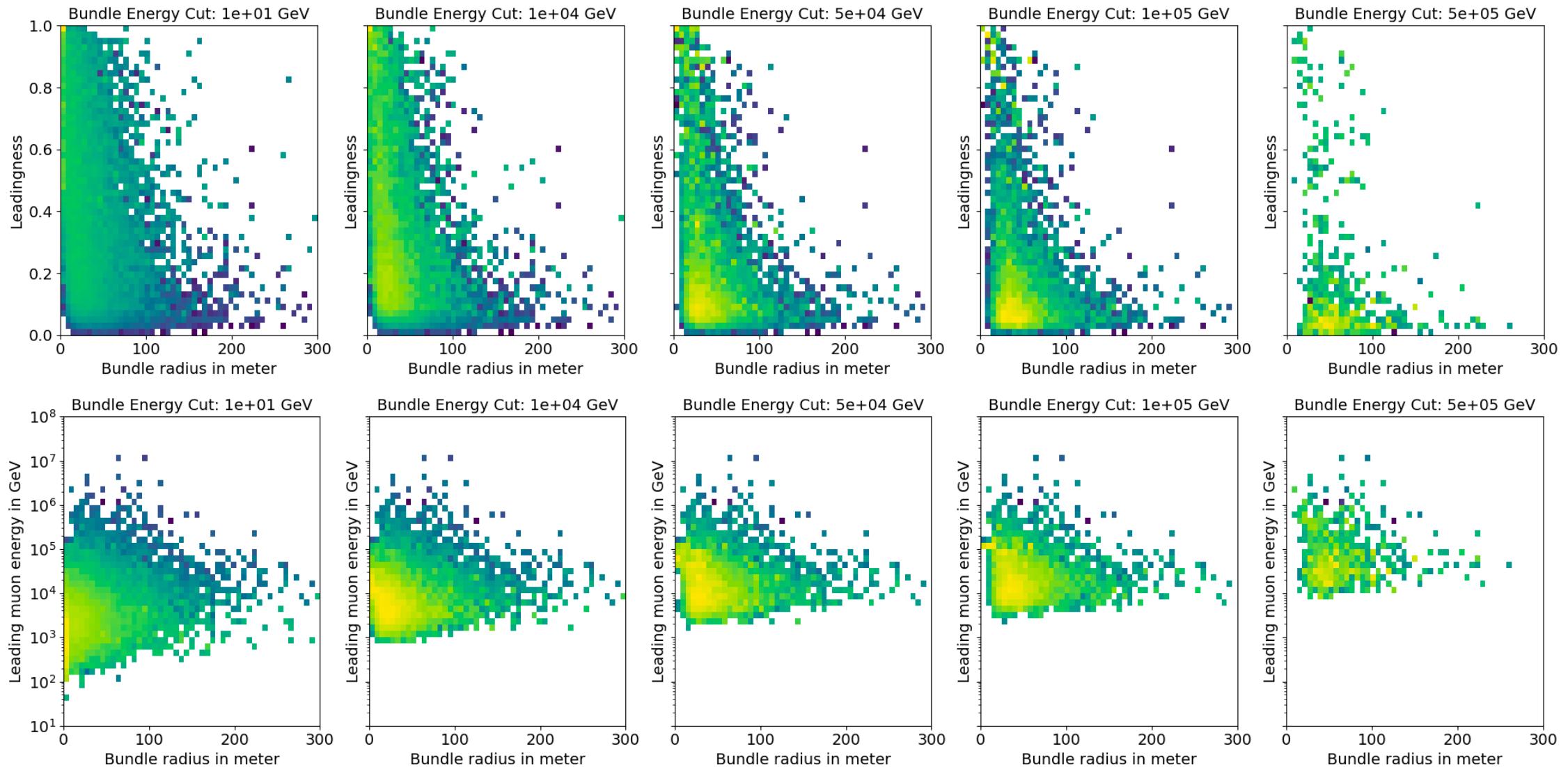
Stochasticity



- High stochasticity leads to high leadingness, but only for a small number of events
- Leading muon energy smears out at large bundle energies
- Largest energy loss of the leading muon correlates with the leading muon energy

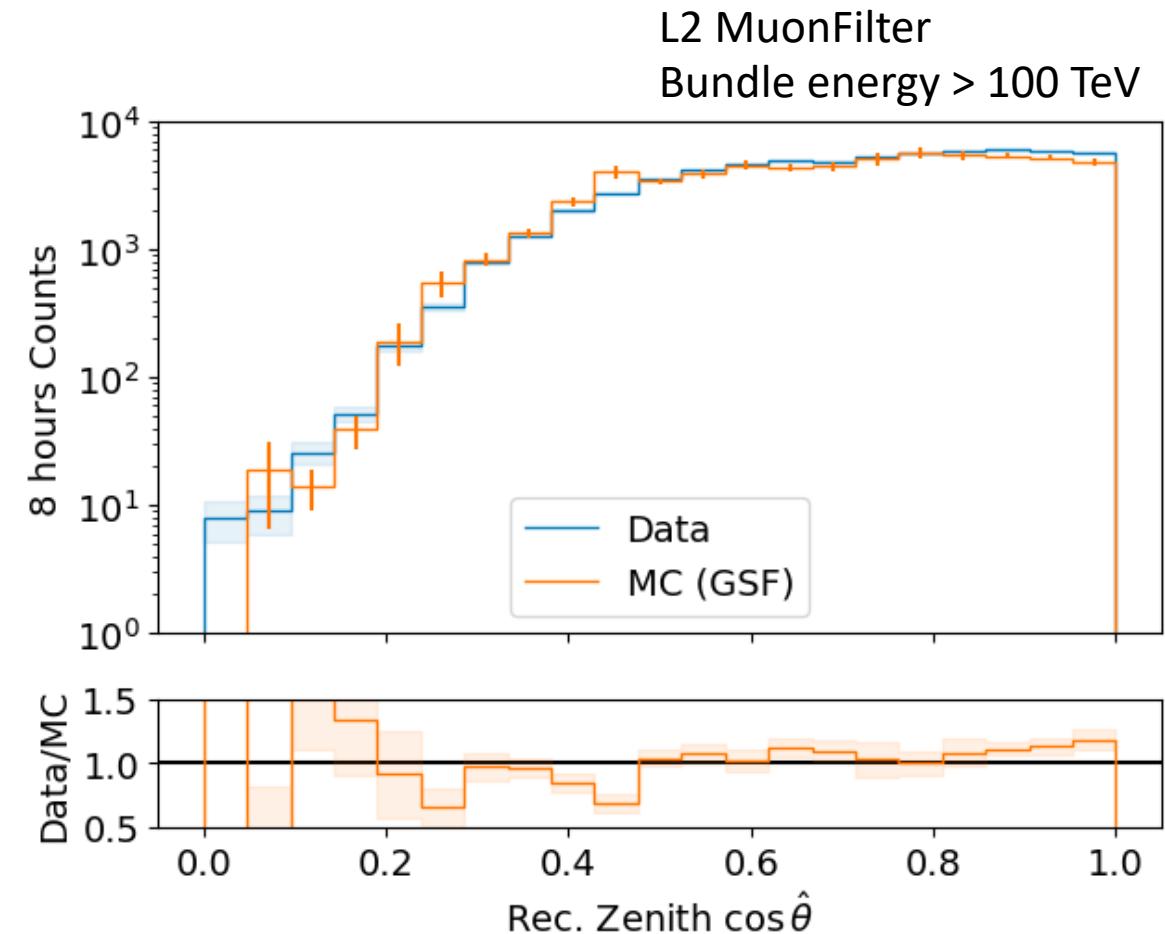
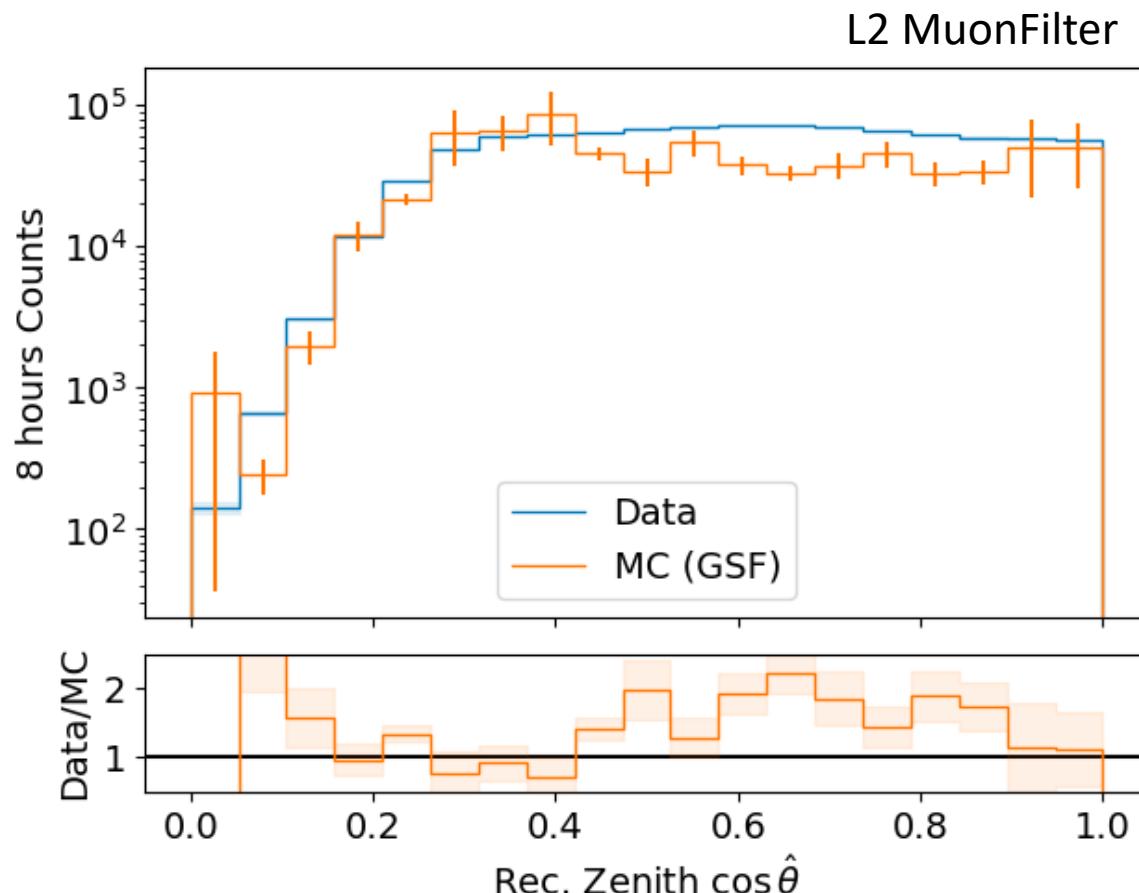
Bundle energy cut

Bundle radius



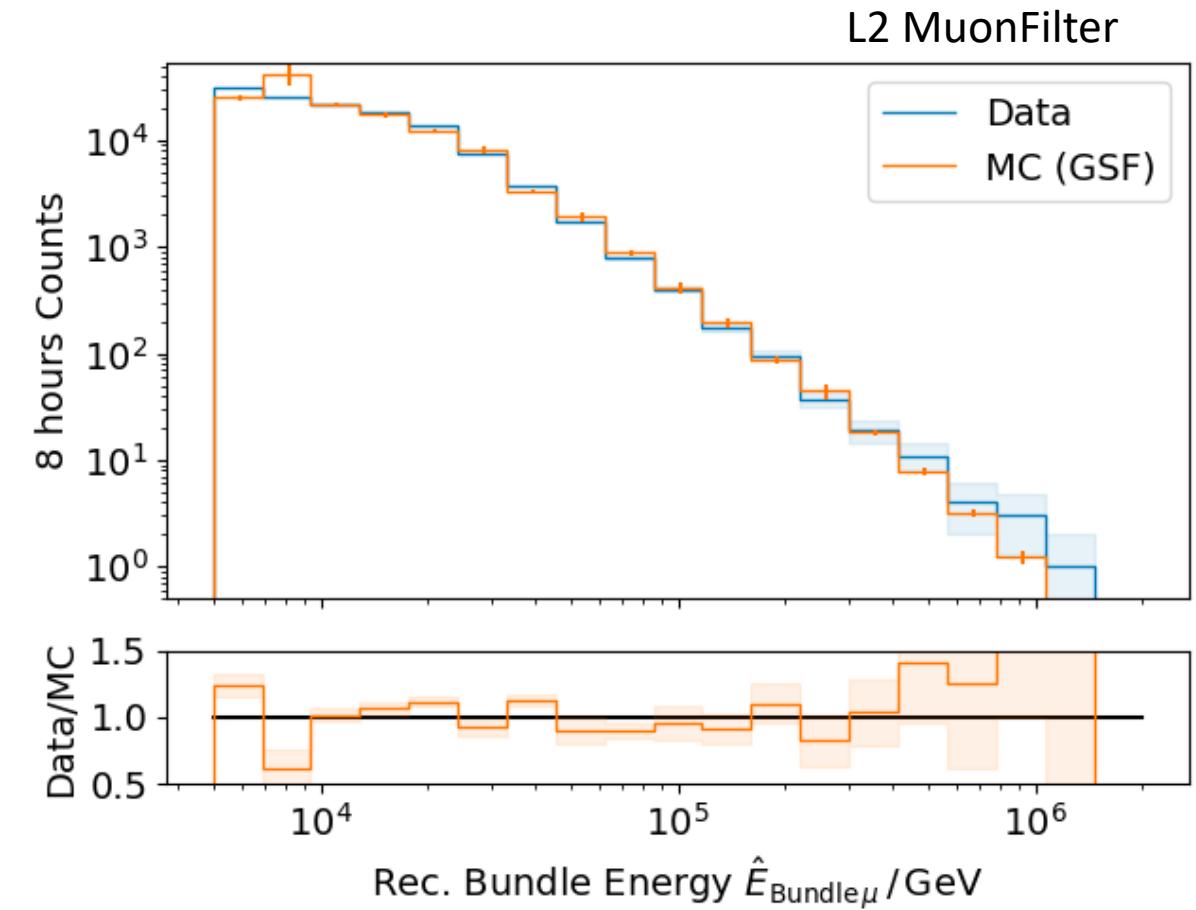
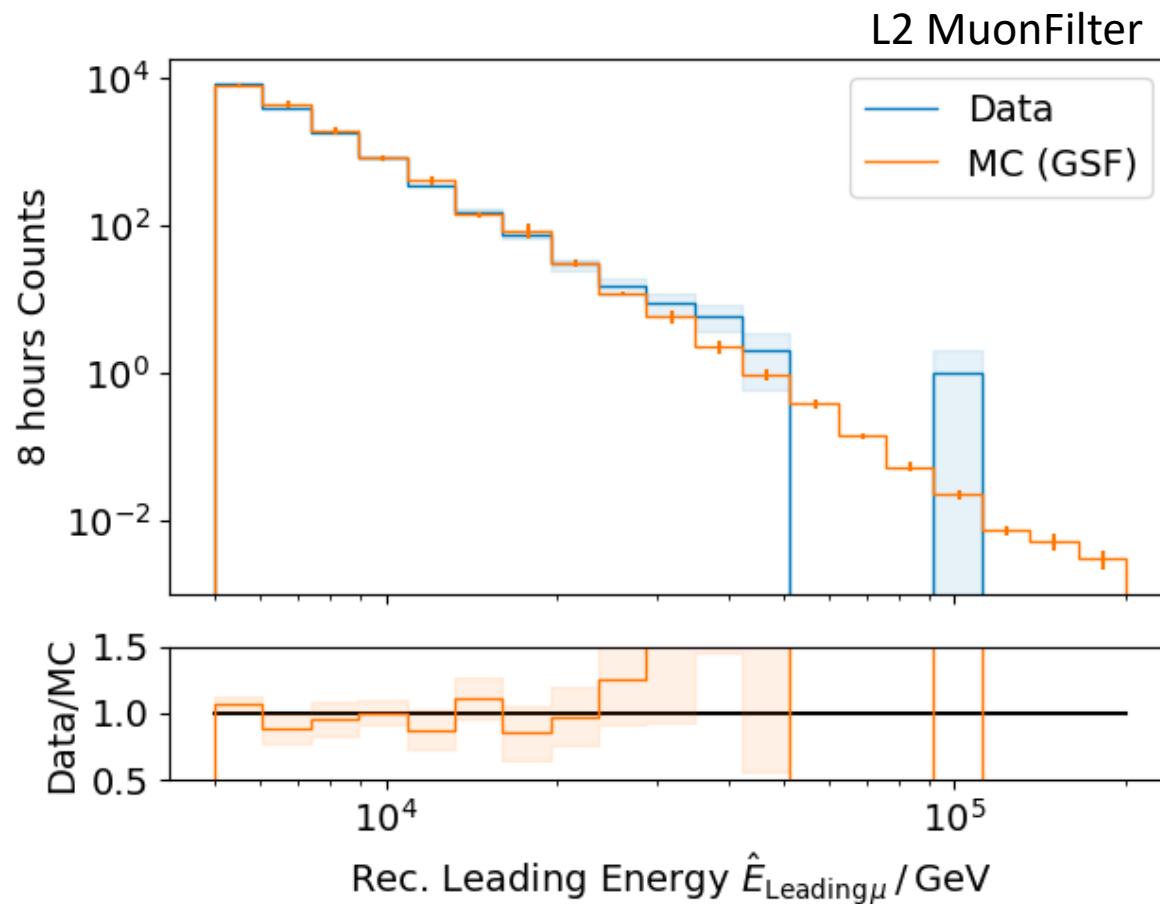
➤ Large bundle radius leads to low leadingness

Data/MC

Data-MC: $\cos(\text{zenith})$ 

- Deviations at low $\cos(\text{zenith})$, but very small statistics
- More data at $\cos(\text{zenith}) > 0.5$
- Less data at $\cos(\text{zenith}) \sim 0.3$

Data-MC: energy spectrum

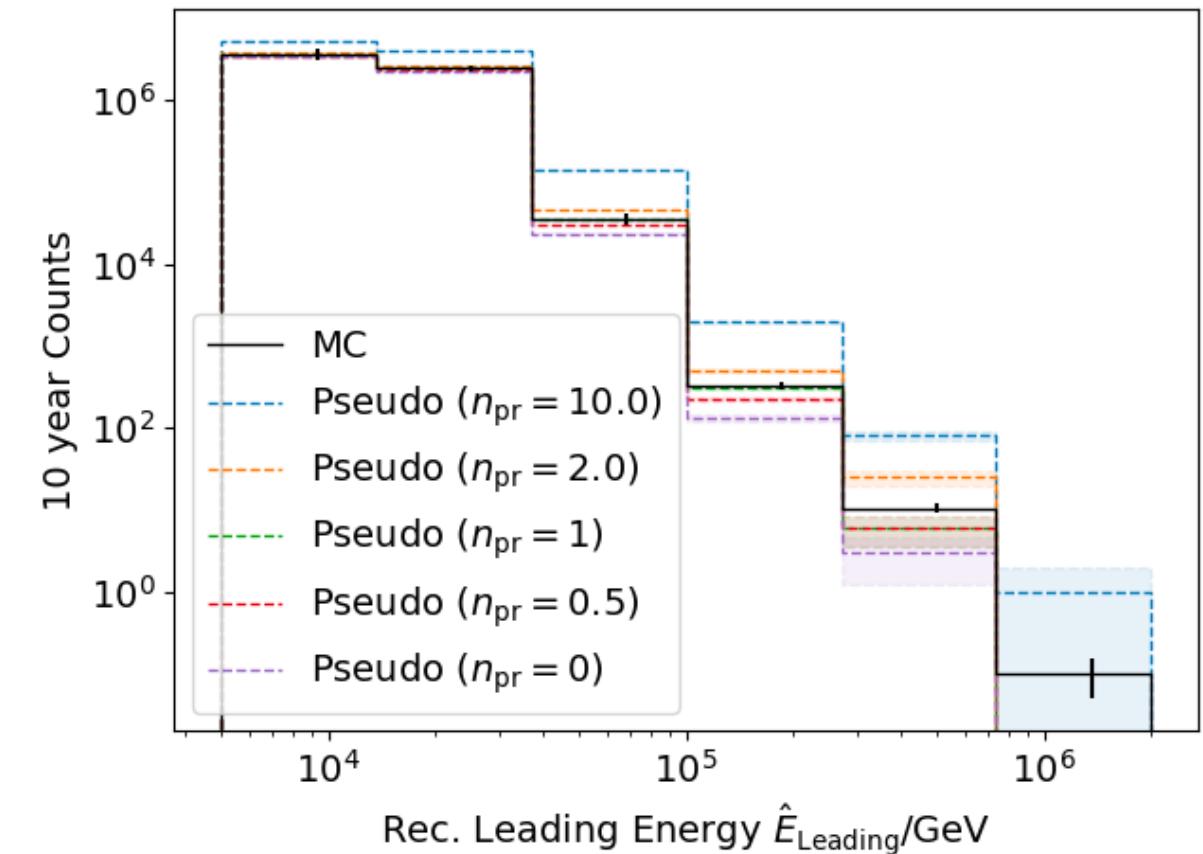
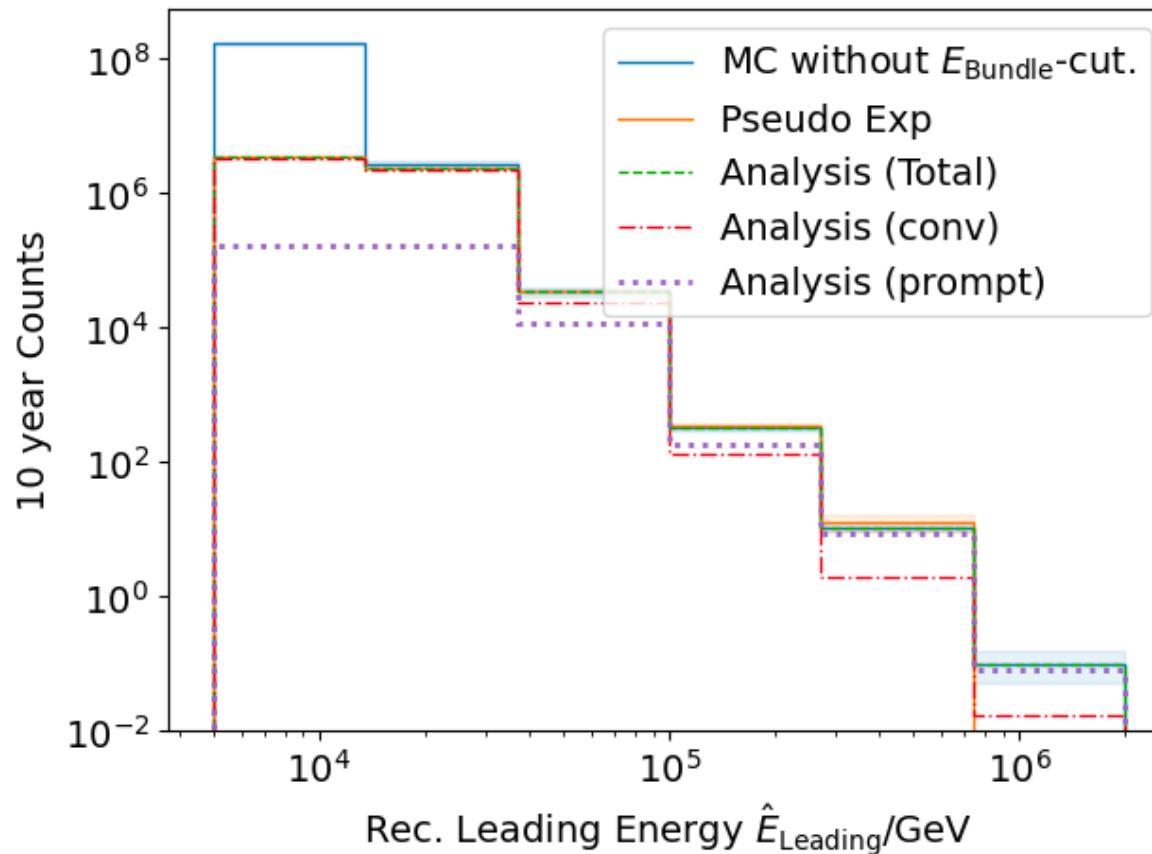


- Bundle energy: good agreement with GSF up to 300 TeV
- But insufficient statistics

Pseudo analysis

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

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

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

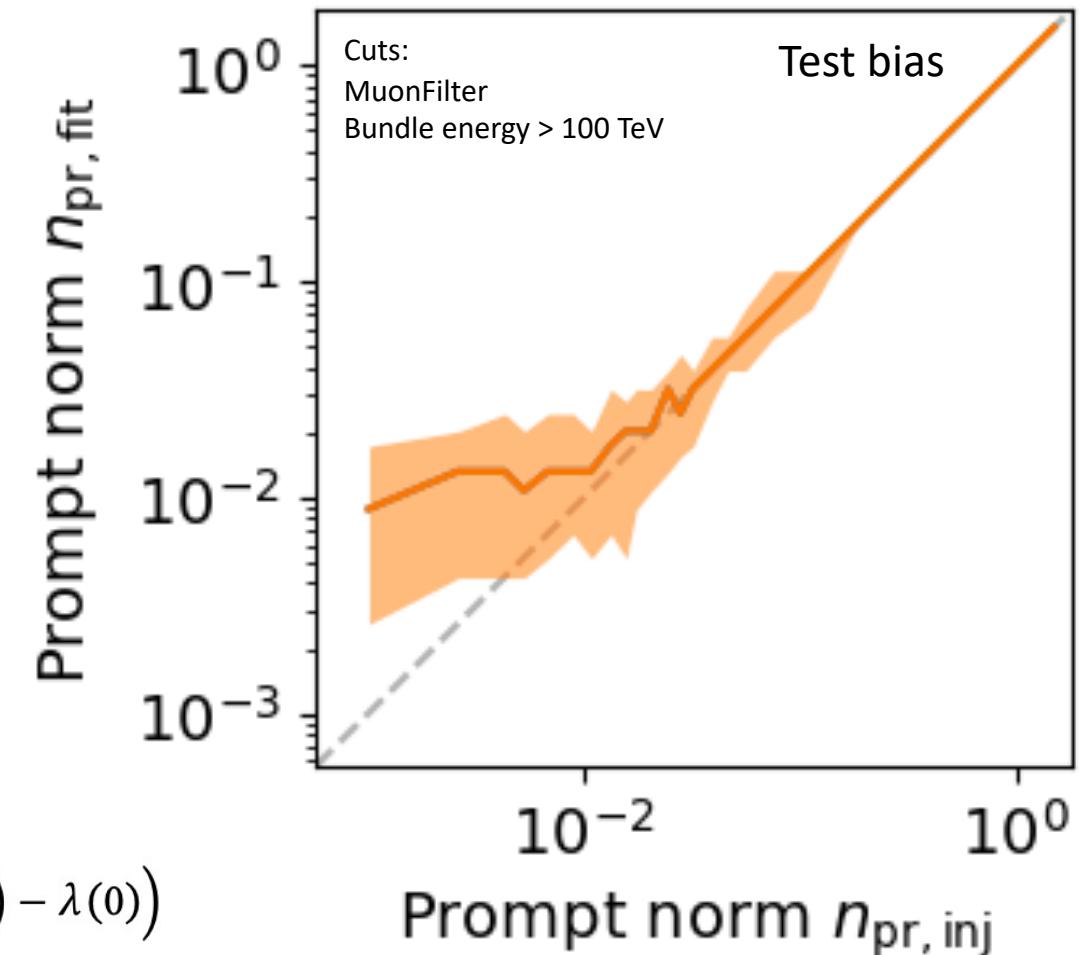
With a constant conv norm:
bin counts depend only on prompt norm
= expectation value per bin

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

Test statistic for Wilks' theorem

Null hypothesis: no prompt

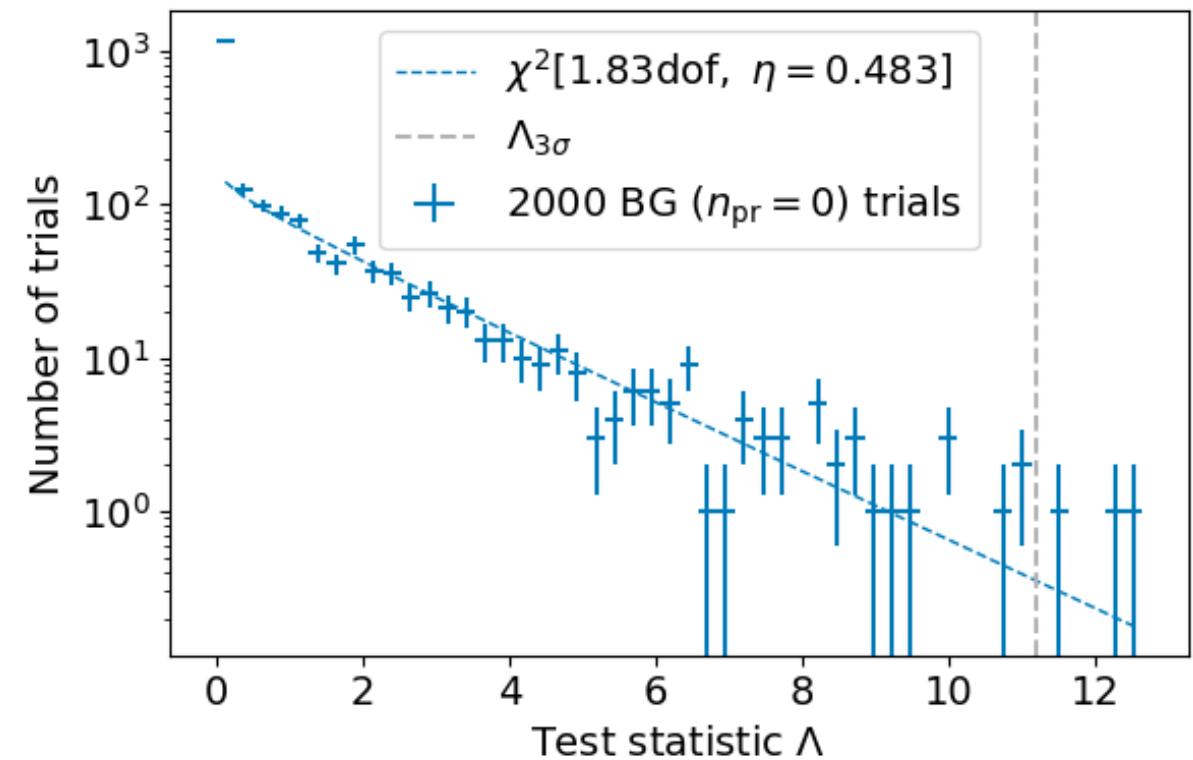
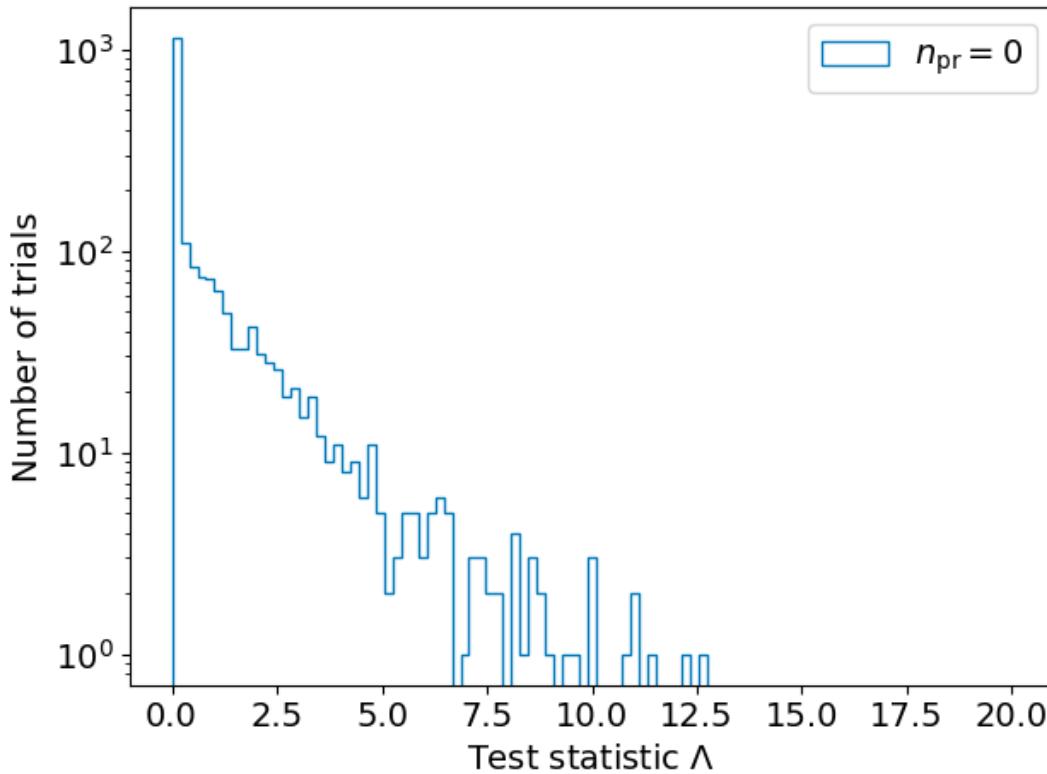
pascal.gutjahr@tu-dortmund.de



➤ Bias starts at a prompt
normalization of 0.1

Test background statistics

Cuts:
L2 MuonFilter
Bundle energy > 100 TeV



- Background statistic is χ^2 – distributed
- Assume Wilks' theorem for test statistics

Discovery potential and sensitivity

Expectation for 1 year:

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

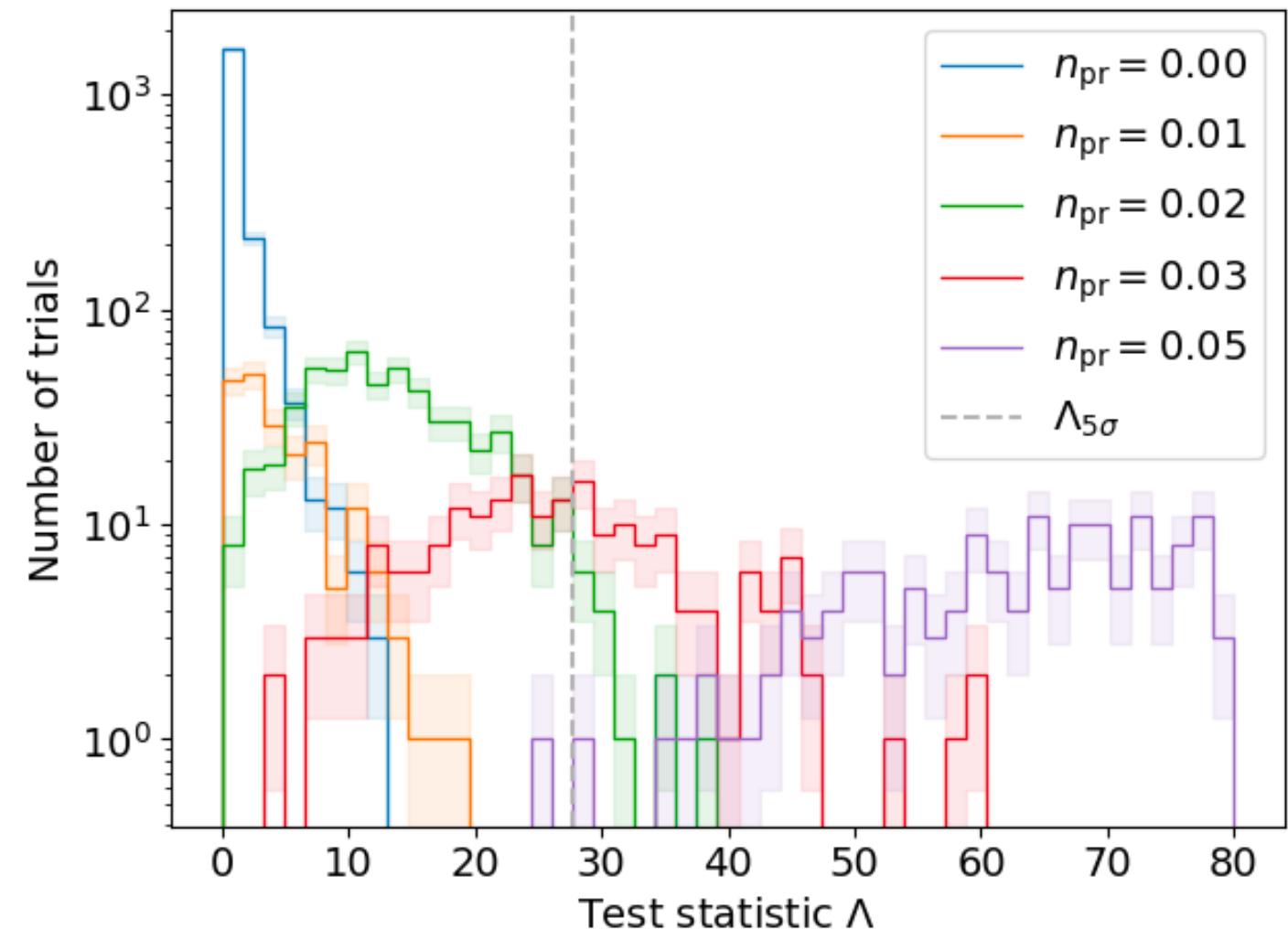
Expectation for 10 years:

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

Caution:

Limited MC statistics -> events are
oversampled in pseudo dataset

Cuts:
L2 MuonFilter
Bundle energy > 100 TeV



Conclusion and outlook

Next goal

- Measure normalization of the atmospheric prompt muon flux

Steps for normalization

- Verify CORSIKA extended history simulations
- Tag prompt muons (see talk by Ludwig Neste)
- Comparisons to MC Eq (see talk by Ludwig Neste)
- Set up preliminary analysis chain
- Reconstruct muon energy and direction
- Data/MC comparisons
- Include systematics
- Run full statistics simulation

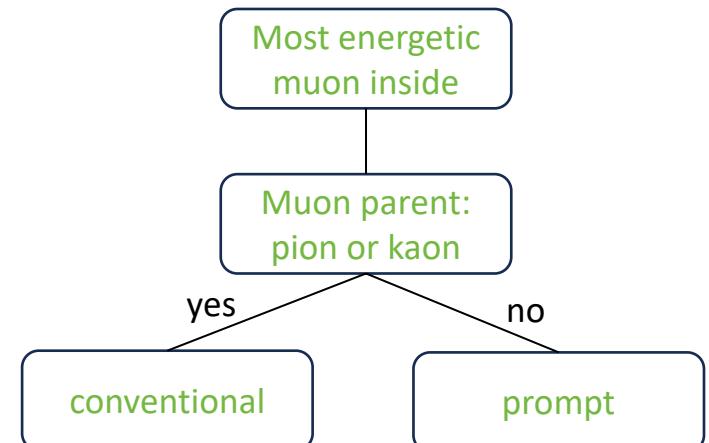
Backup

Some definitions and wording...

- Leading muon
 - The most energetic muon inside a bundle (no minimum fraction required)
- Single muon
 - Except for stopping and very low energetic muons, there are never any single muons (almost every event is muon bundle)
- Prompt muon
 - Muon parent is not pion or kaon

Suggestion:

To avoid confusion regarding different leading muon definitions we can introduce a “leadingness”
(For example: Tomasz used a leadingness of 50%, ...)



Intention

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

Definition of the muon flux

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

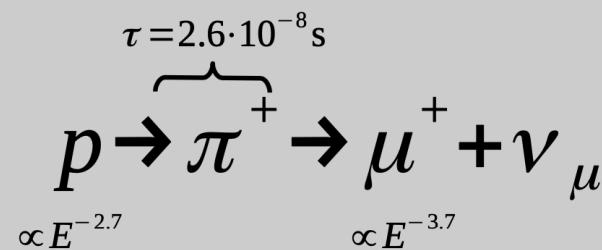


$$\pi, K \propto E^{-3.7}$$

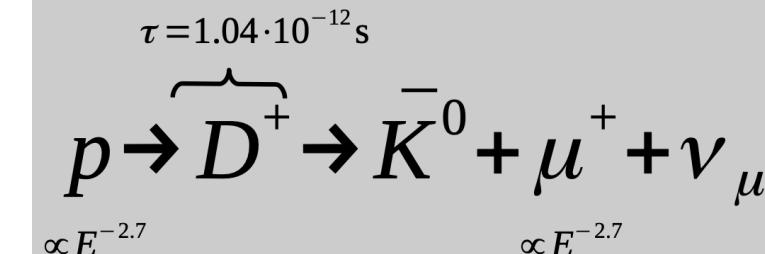


$$\text{"not"} \pi, K \propto E^{-2.7}$$

Conventional component:



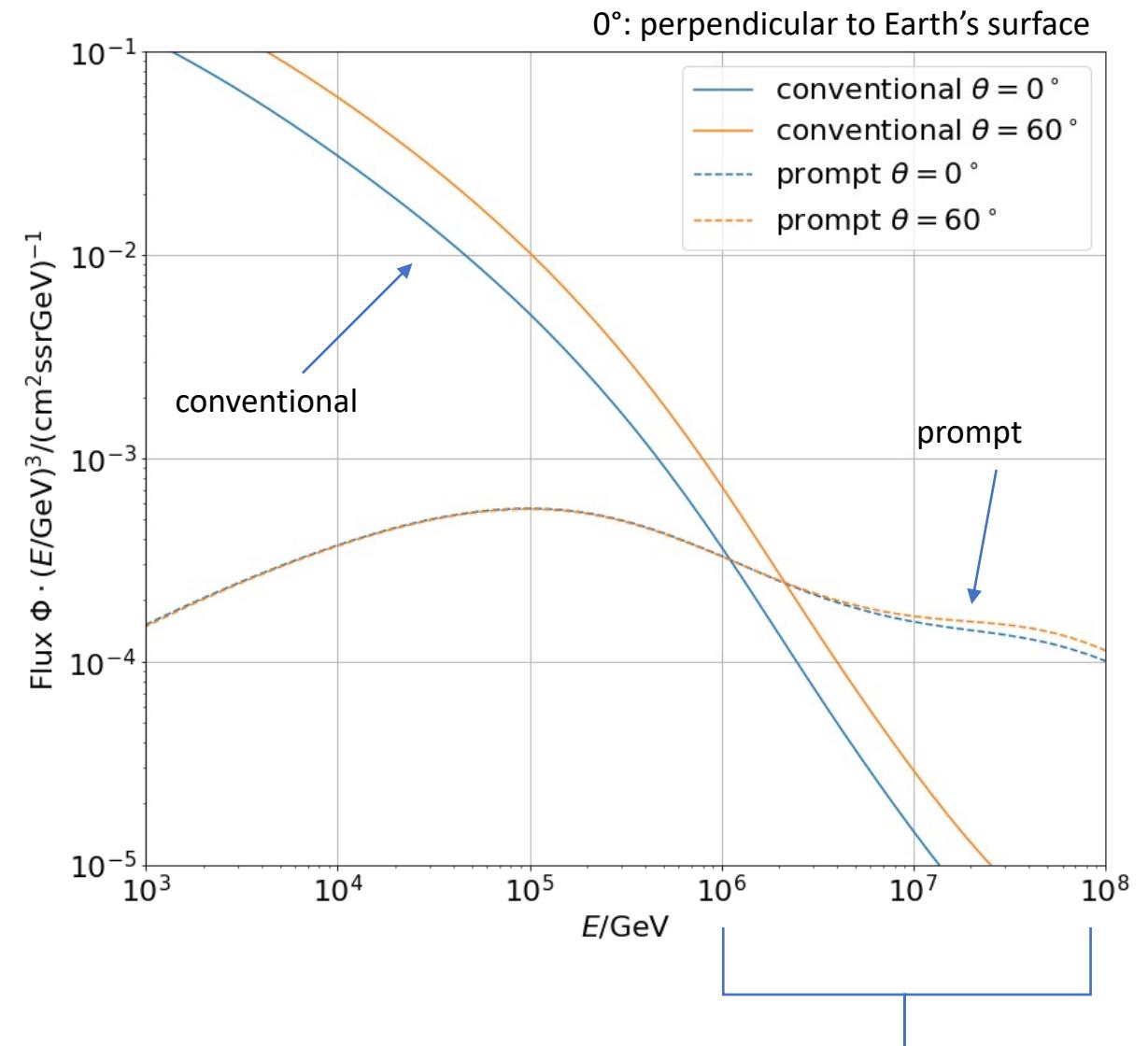
prompt component:



Muon flux

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

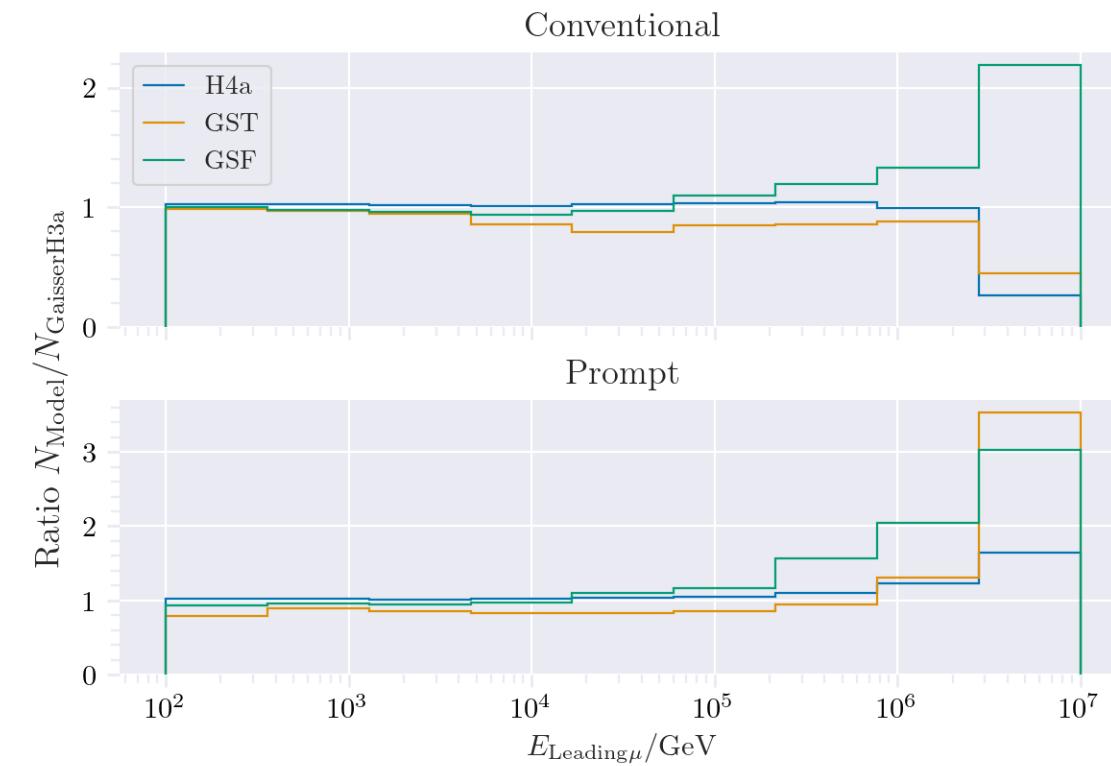
- Prompt dominates at energies larger than PeV
- Conventional particle flux depends on zenith angle



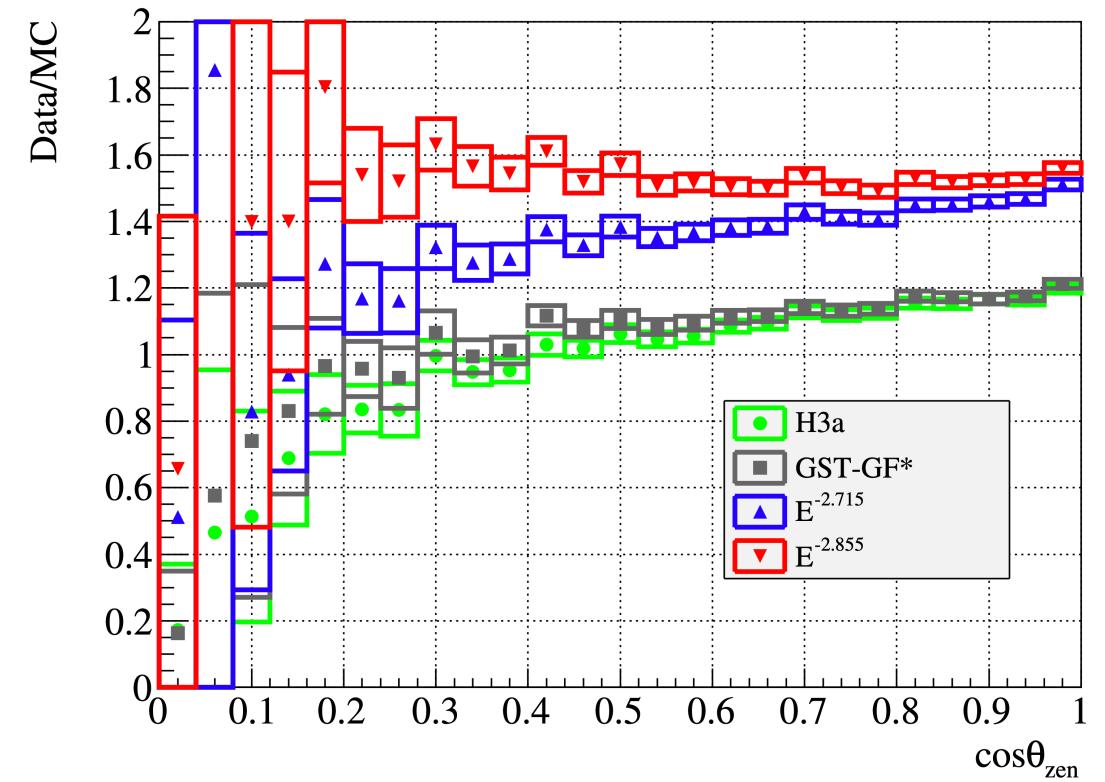
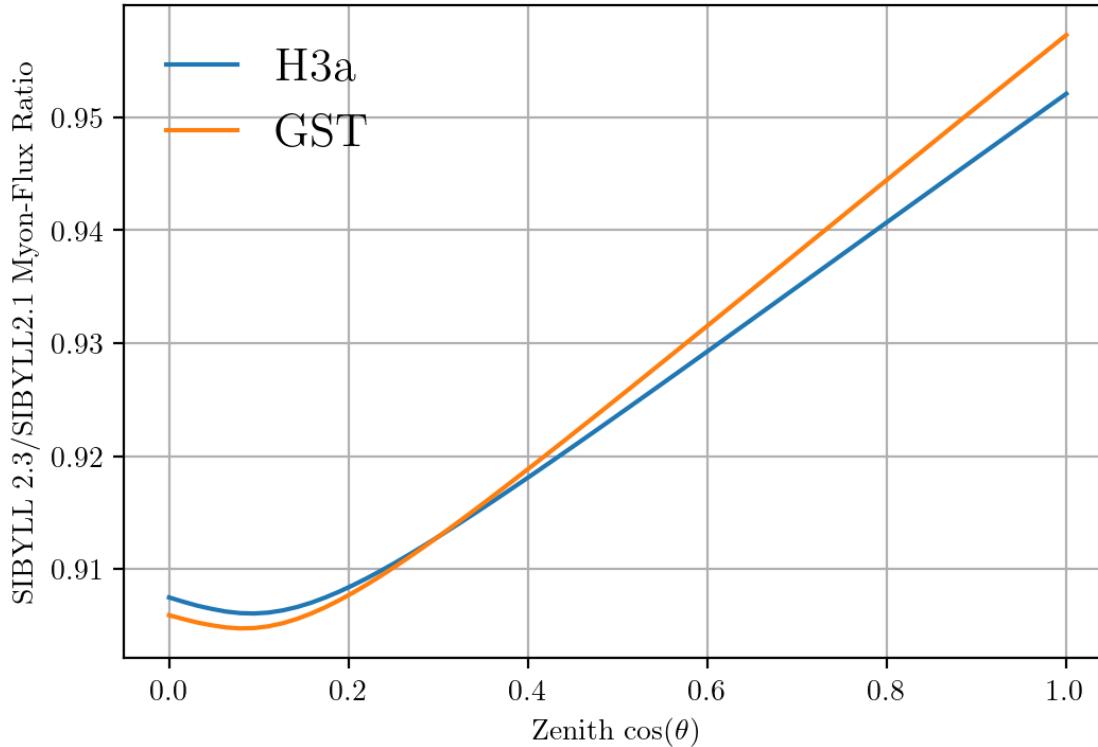
We can measure prompt muon
energies from ~ 1 PeV to ~ 100 PeV

Muon production – different weightings

GST predicts most prompt



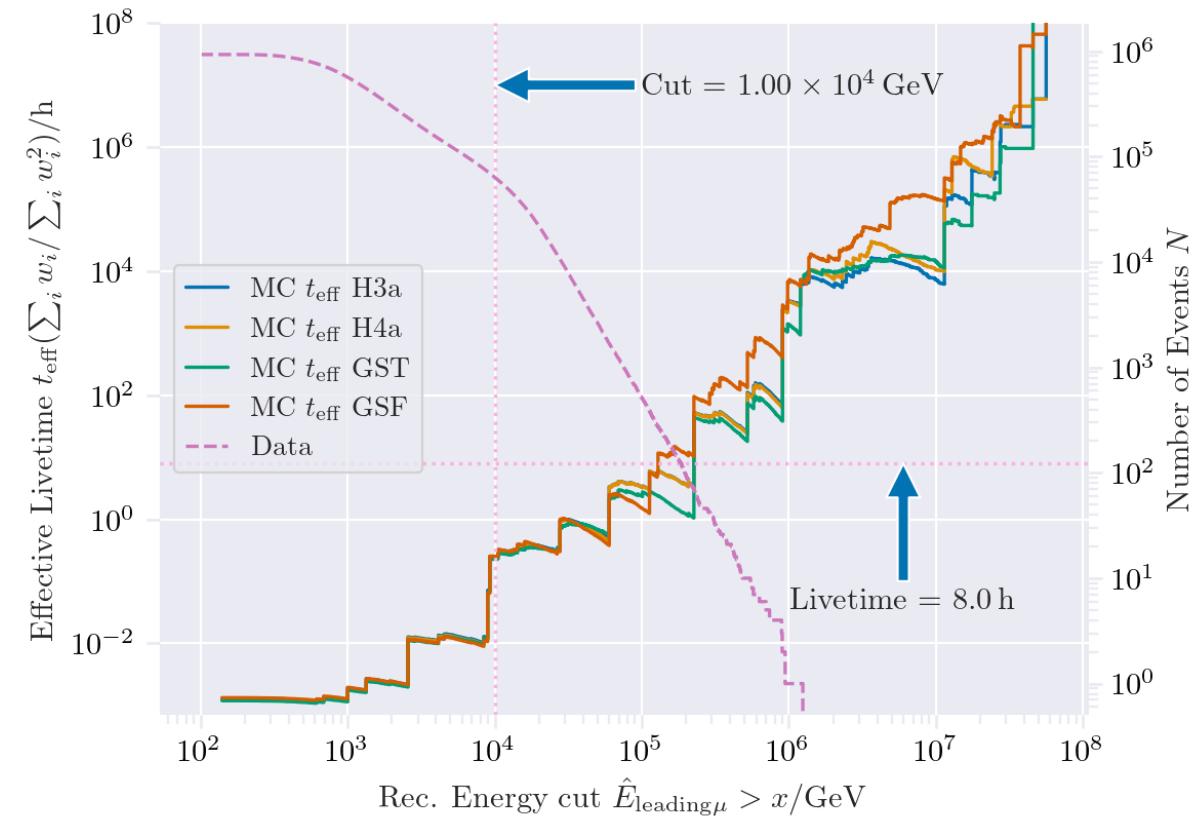
Solution to zenith problem?



- No complete solution, but a step in the right direction

New CORSIKA extended history simulations

- CORSIKA 77420
- SIBYLL 2.3d
- Icetray 1.5.1
- 5 components (p, He, N, Al, Fe)
- Polyplopia: True
- Trimshower: True
- Ecuts1: 273 GeV (hadron min energy)
- Ecuts2: 273 GeV (muon min energy)
- Ecuts3: 10^{20} GeV (electron min energy)
- Ecuts4: 10^{20} GeV (photon min energy)
- 4 datasets:
 - 30010: 600 GeV – 1 PeV
 - 30011: 1 PeV – 100 PeV
 - 30012: 100 PeV – 1 EeV
 - 30013: 1 EeV – 50 EeV
- [/data/sim/IceCube/2023/generated/CORSIKA_EHISTORY/](#)

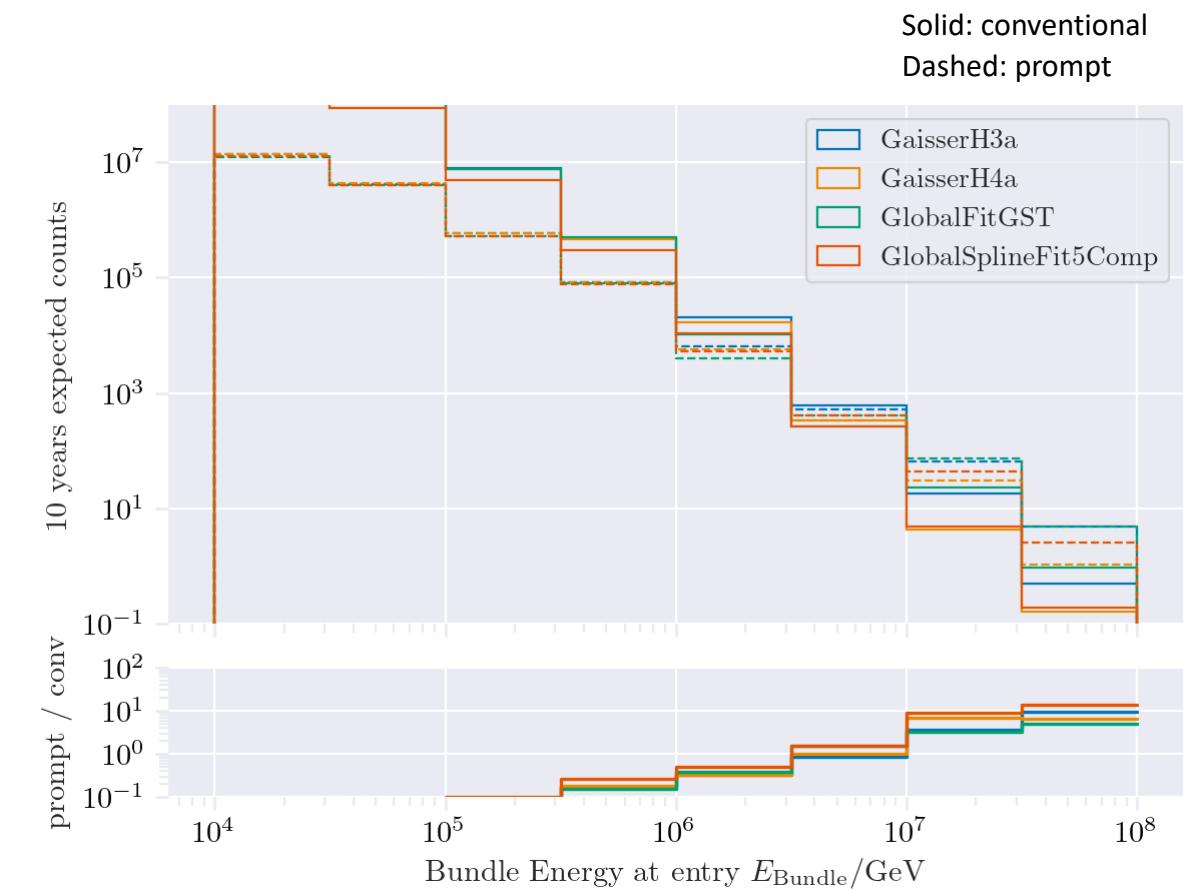
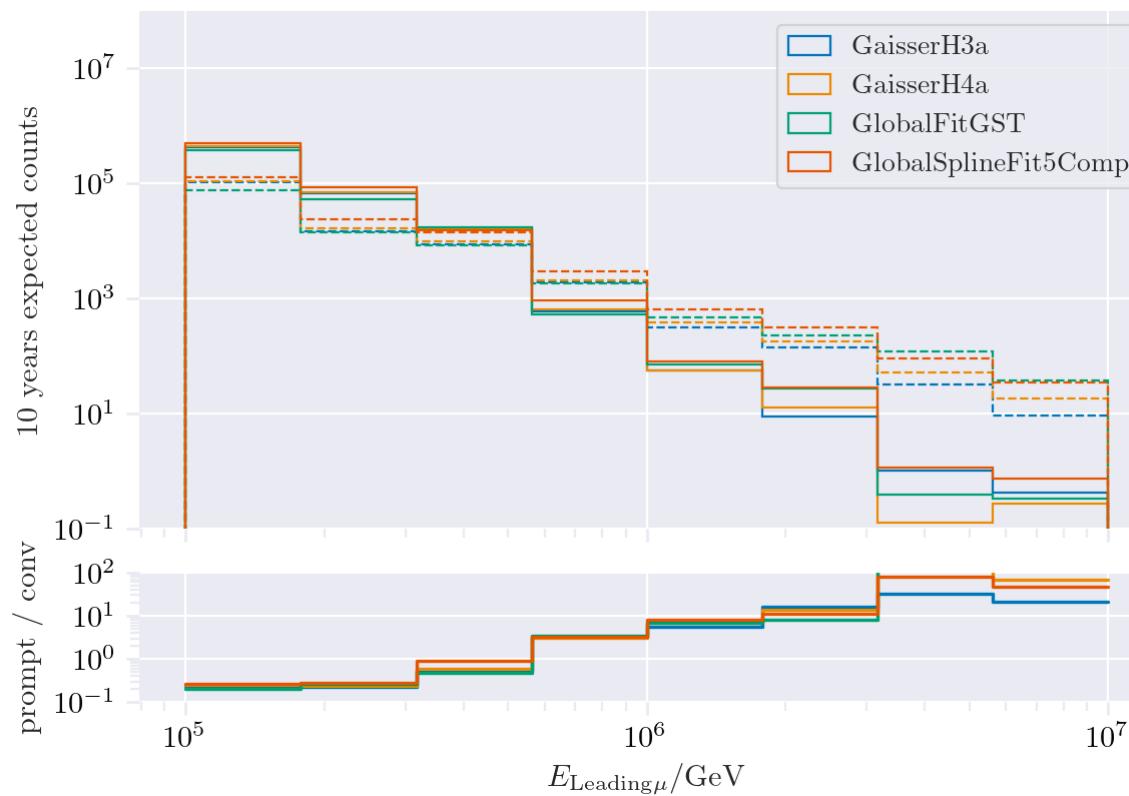


Please go ahead and test the datasets

pascal.gutjahr@tu-dortmund.de

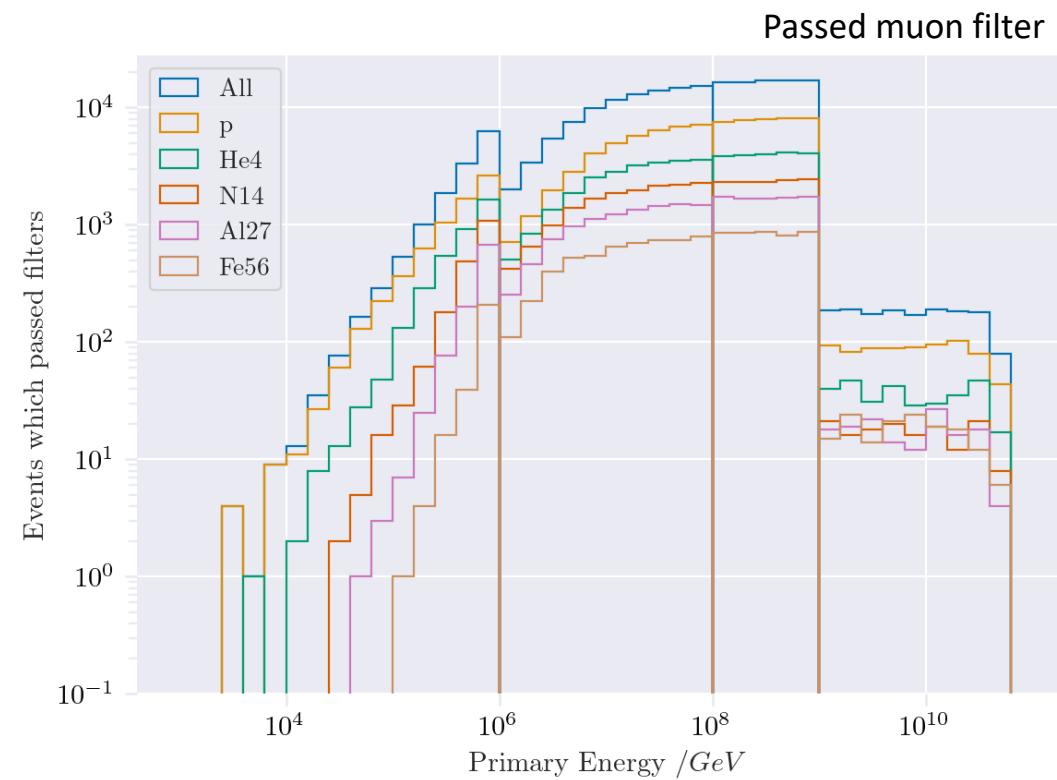
- Sufficient statistics above 1 PeV
- Too few statistics at lower energies

Expected muons for 10 years: leading vs. bundle energy



- Different primary fluxes lead to different prompt fluxes
- Bundle energy extends to higher energies

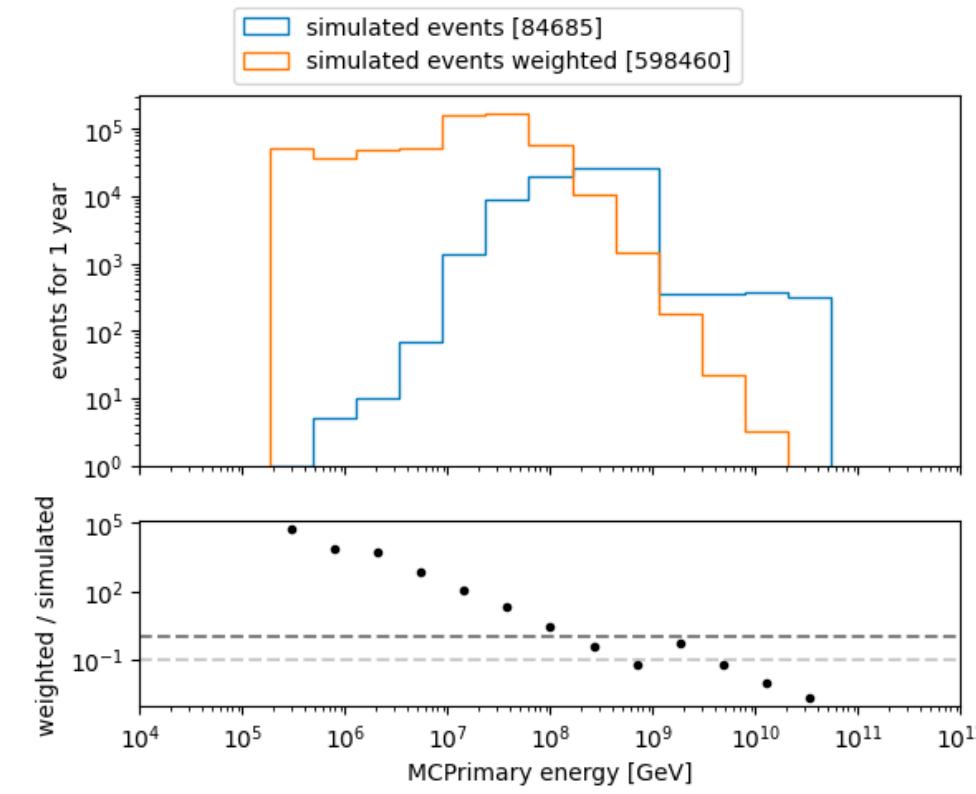
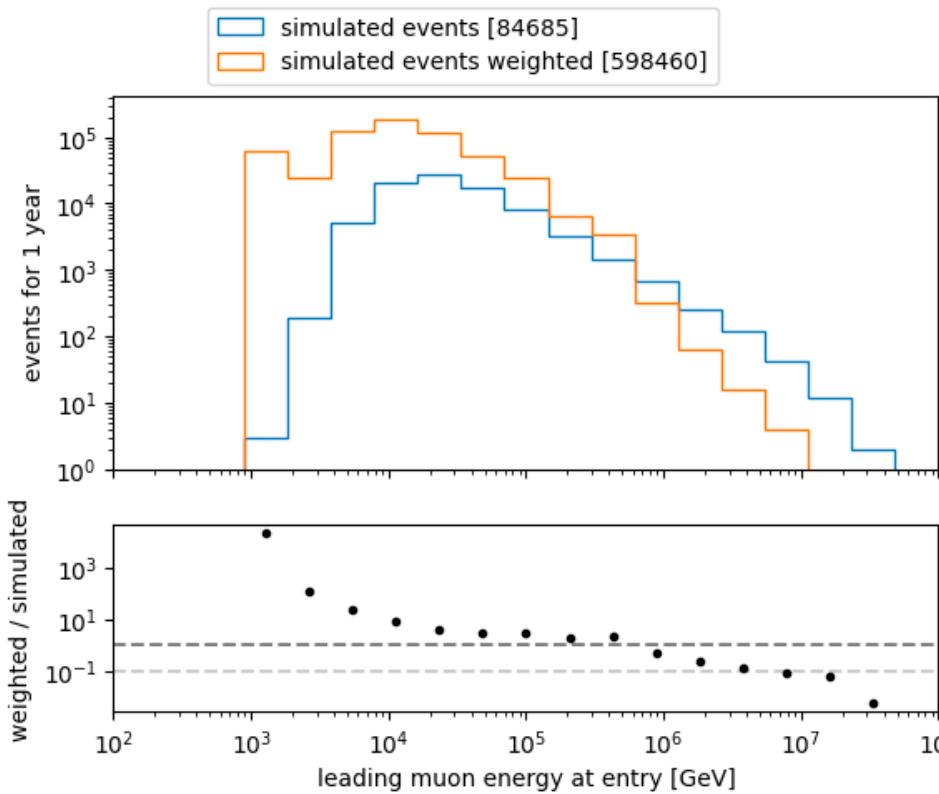
Simulated events



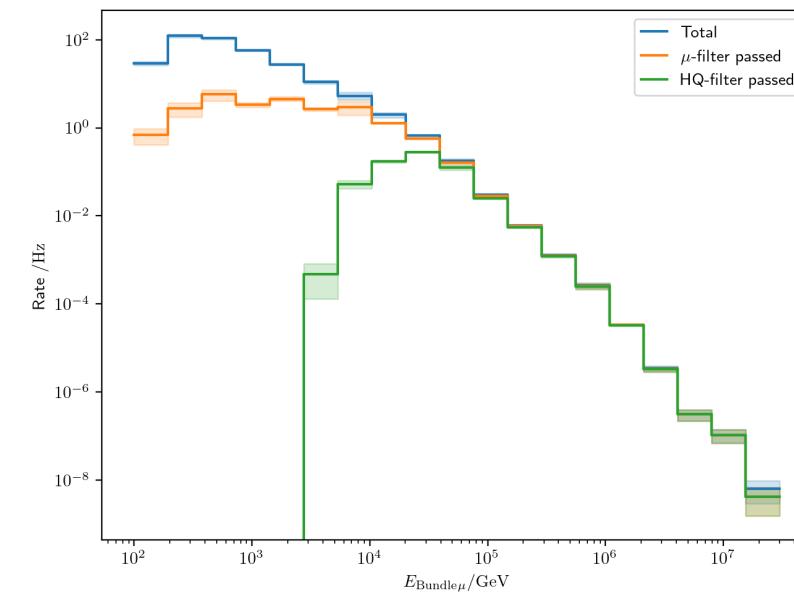
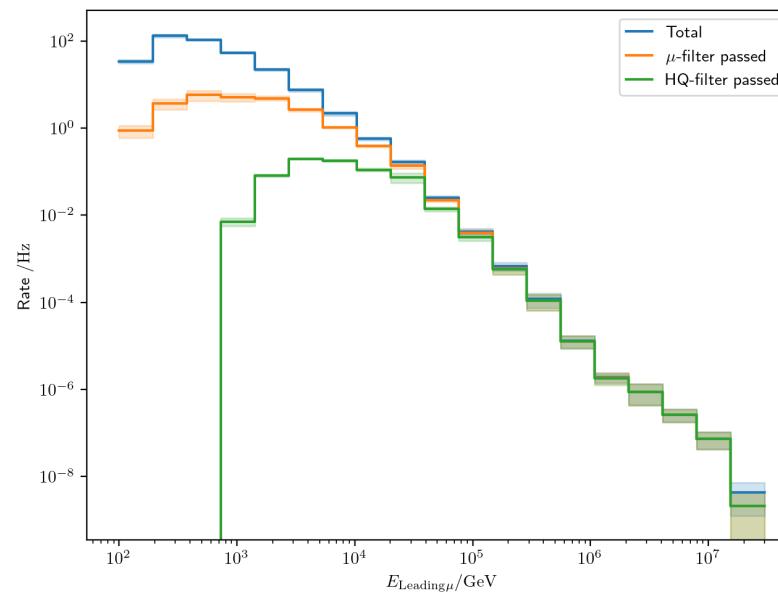
Which energies do we need to simulate?

L2 MuonFilter

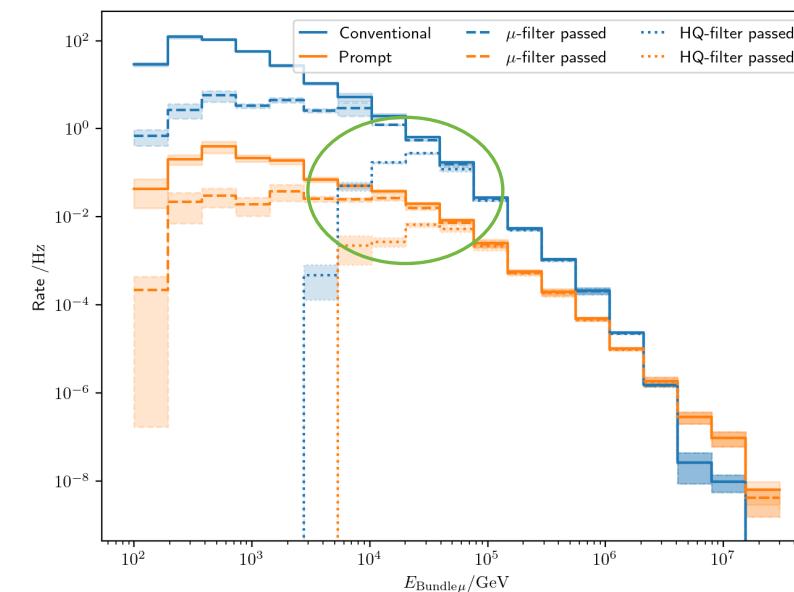
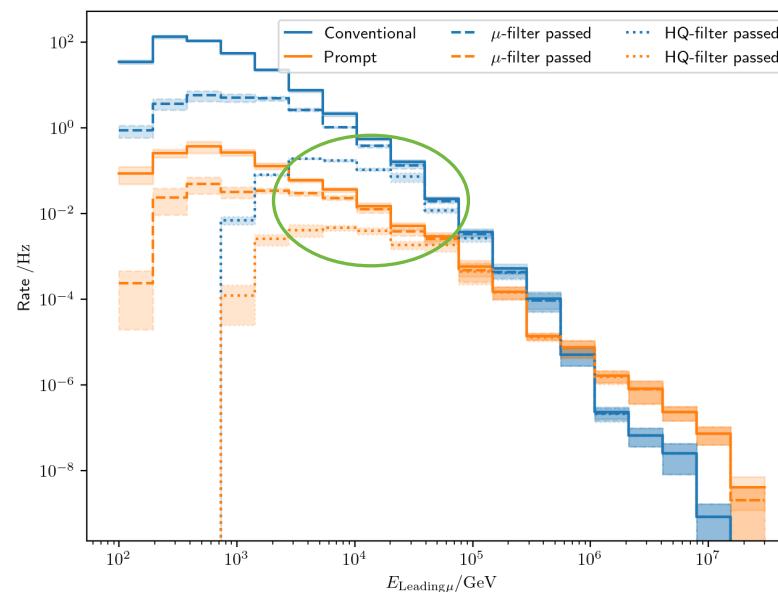
Bundle energy > 100 TeV



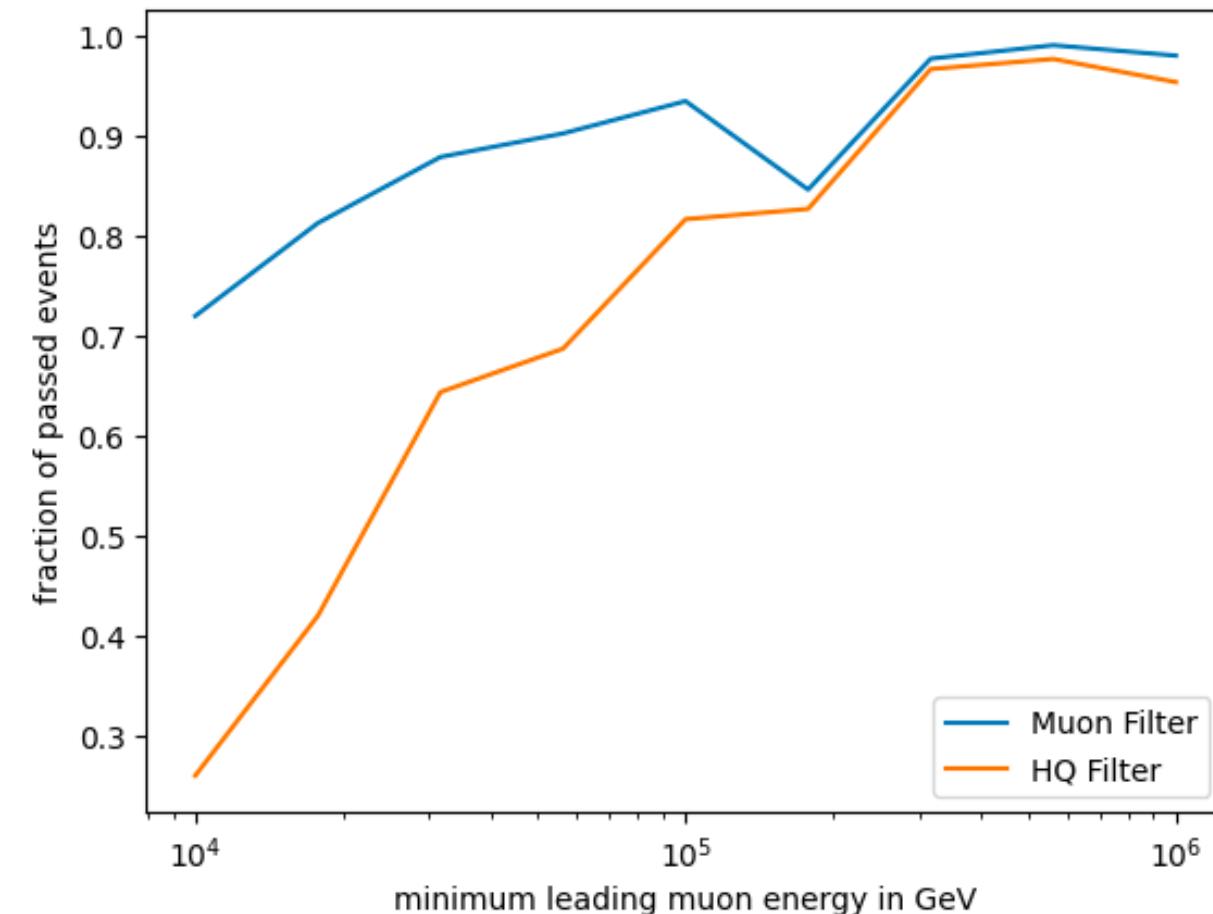
Filters



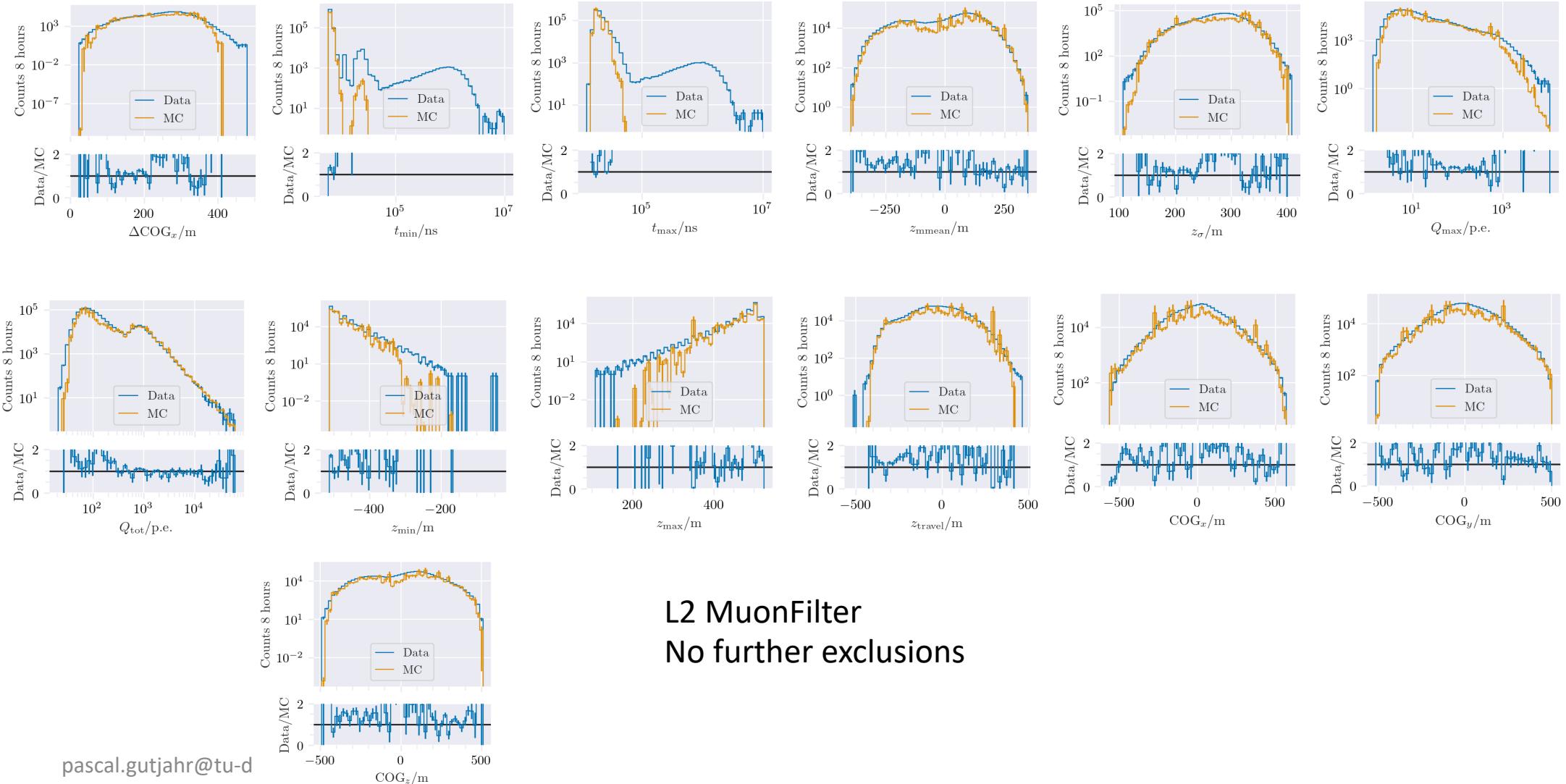
Choose muon filter,
larger statistics at 10 TeV



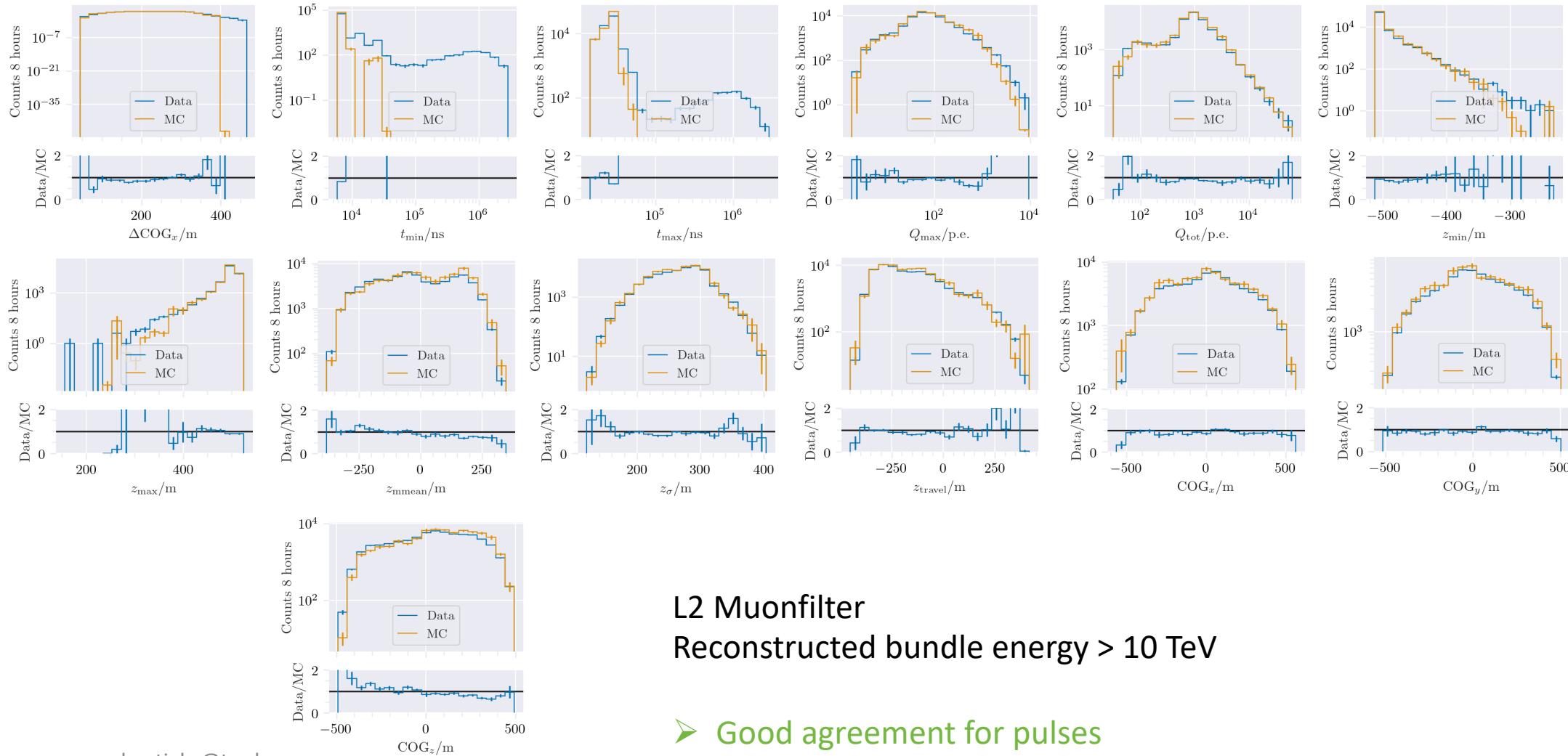
Filters – passed events per leading energy



Data-MC: HitStatistics - SplitInIceDSTPulses



Data-MC: HitStatistics - SplitInIceDSTPulses



L2 Muonfilter
Reconstructed bundle energy $> 10 \text{ TeV}$

➤ Good agreement for pulses

Evaluation of models used in pseudo analysis

- Note: the DNN reconstruction models are still investigated! The models used here are in an early stage. So far, the two bachelor students have trained models with better performance.
- Updates on the models and reconstructions are provided in the near future

Data-MC agreements

- Two bachelor students worked on reconstructions using the dnn_reco framework (thesis available in english):
 - Leander Flottau
 - Benjamin Brandt

Reconstructions:

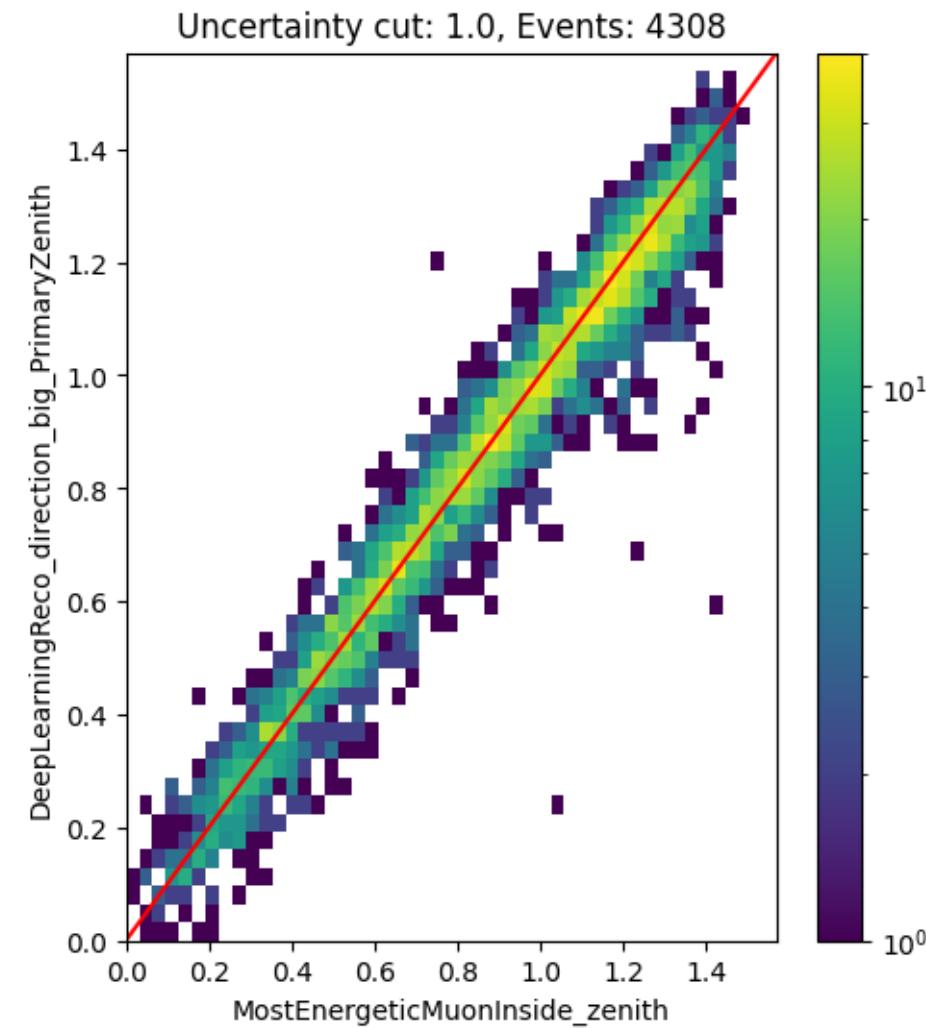
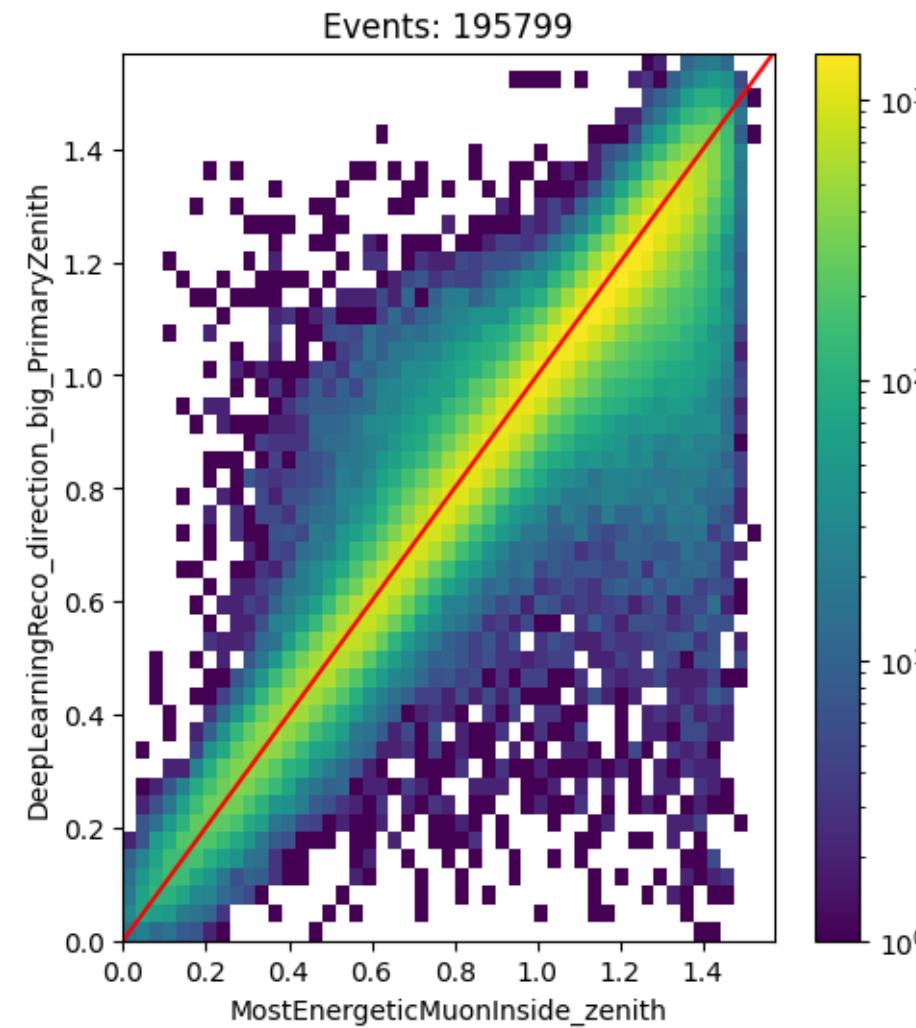
- Leading energy
- Leading fraction
- Bundle energy
- Multiplicity
- Azimuth
- Zenith

Work in progress

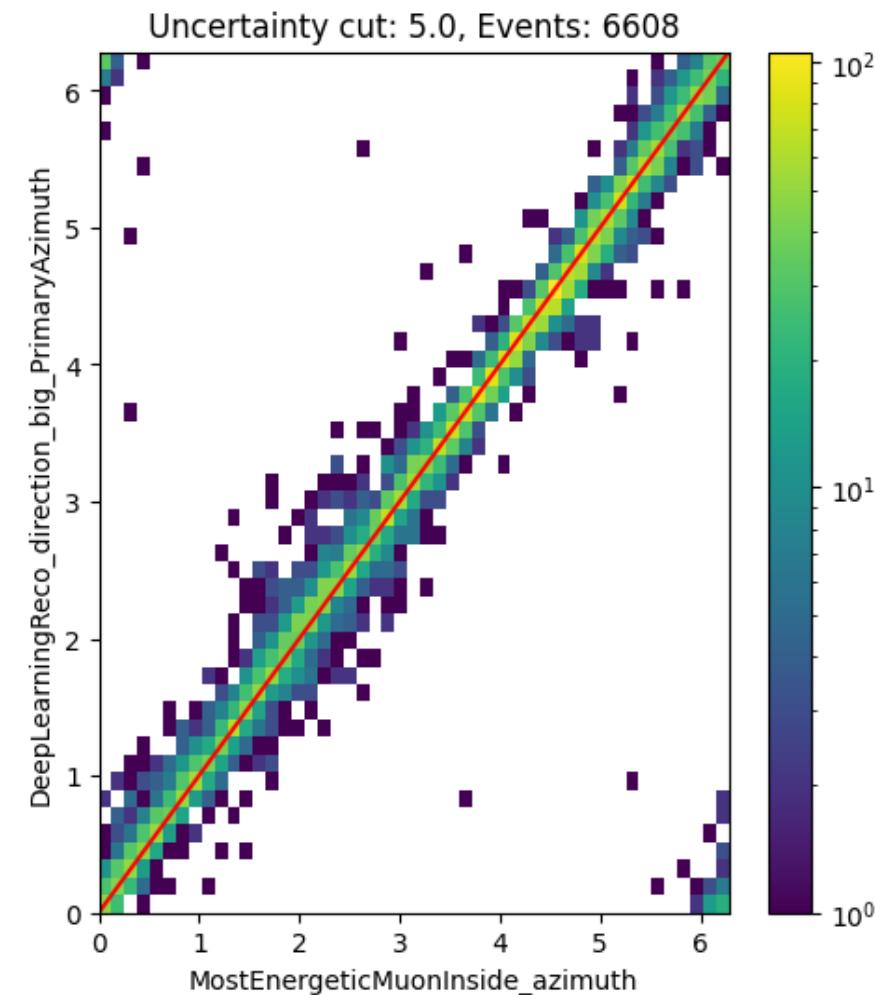
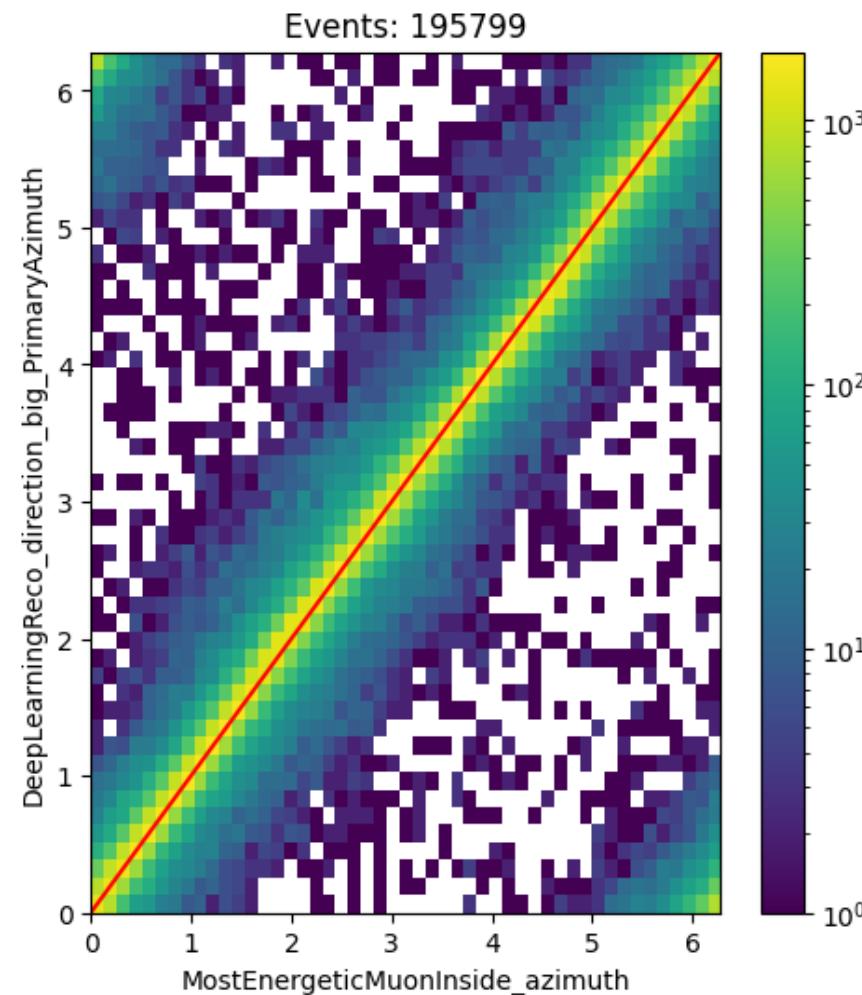
General:

- Trained on uncleaned and $6\mu s$ cleaned muon pulses
- Processed 8h of experimental data (June 4th, 2020)

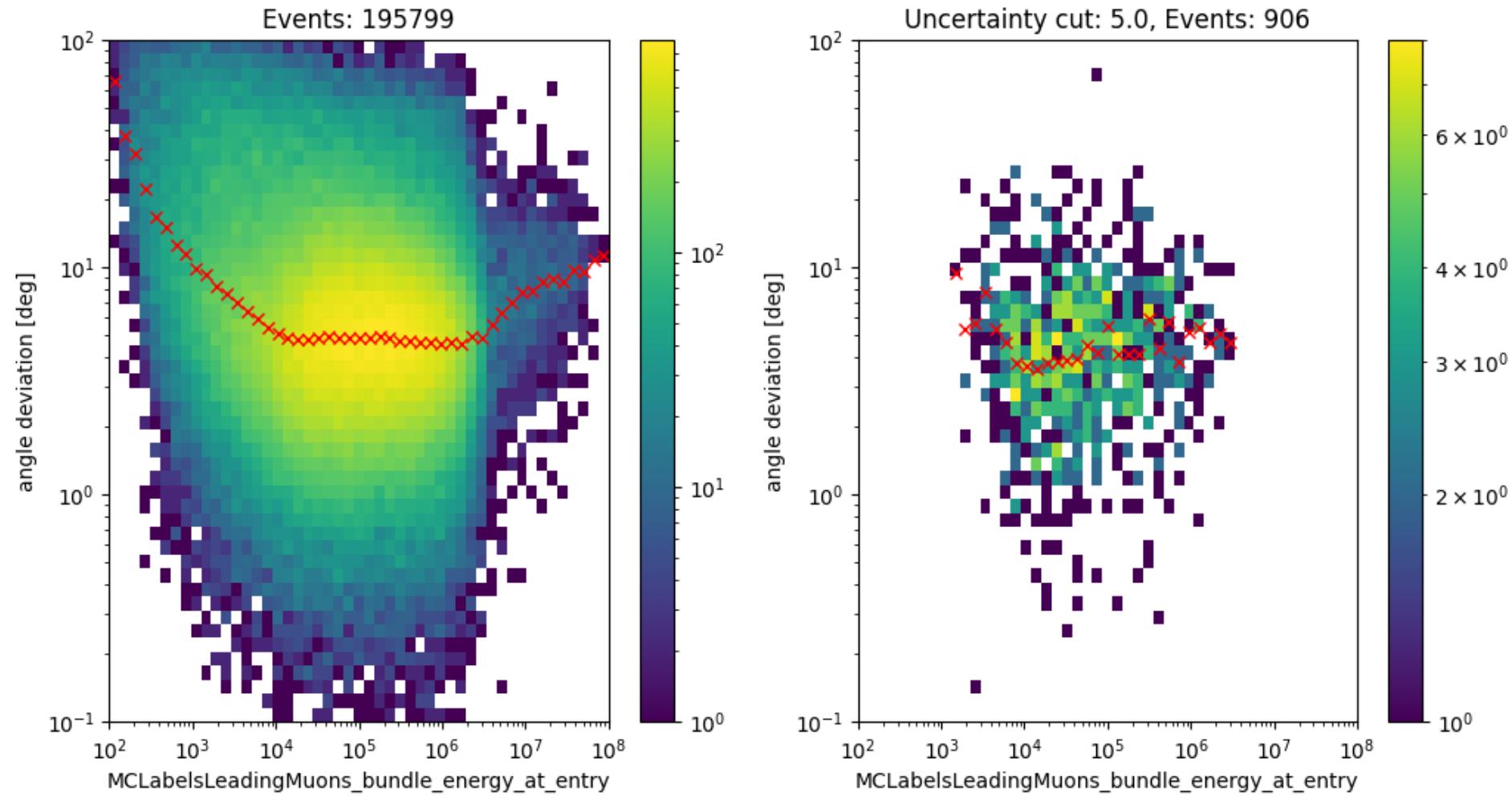
Zenith reconstructions



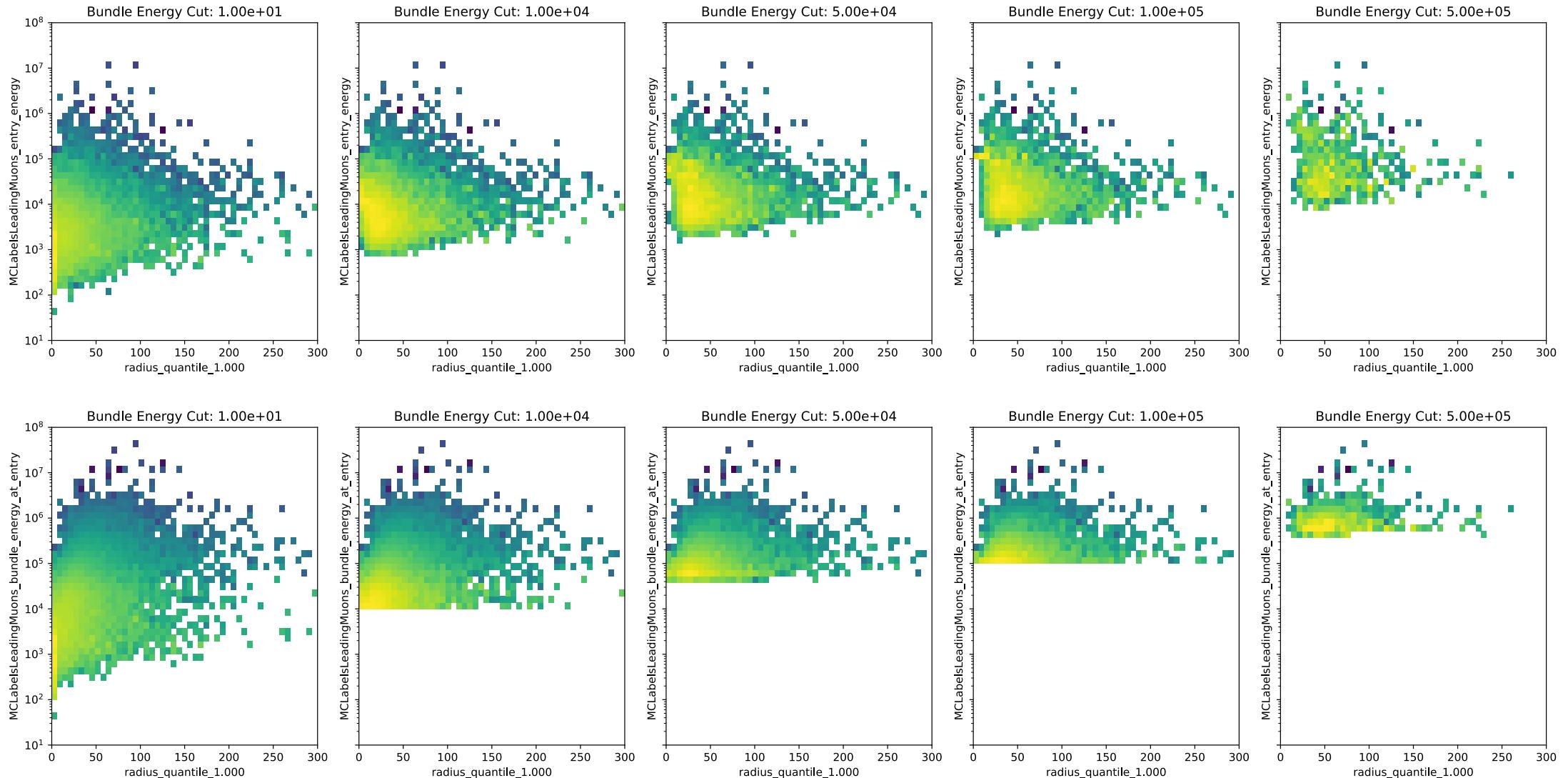
Azimuth reconstructions



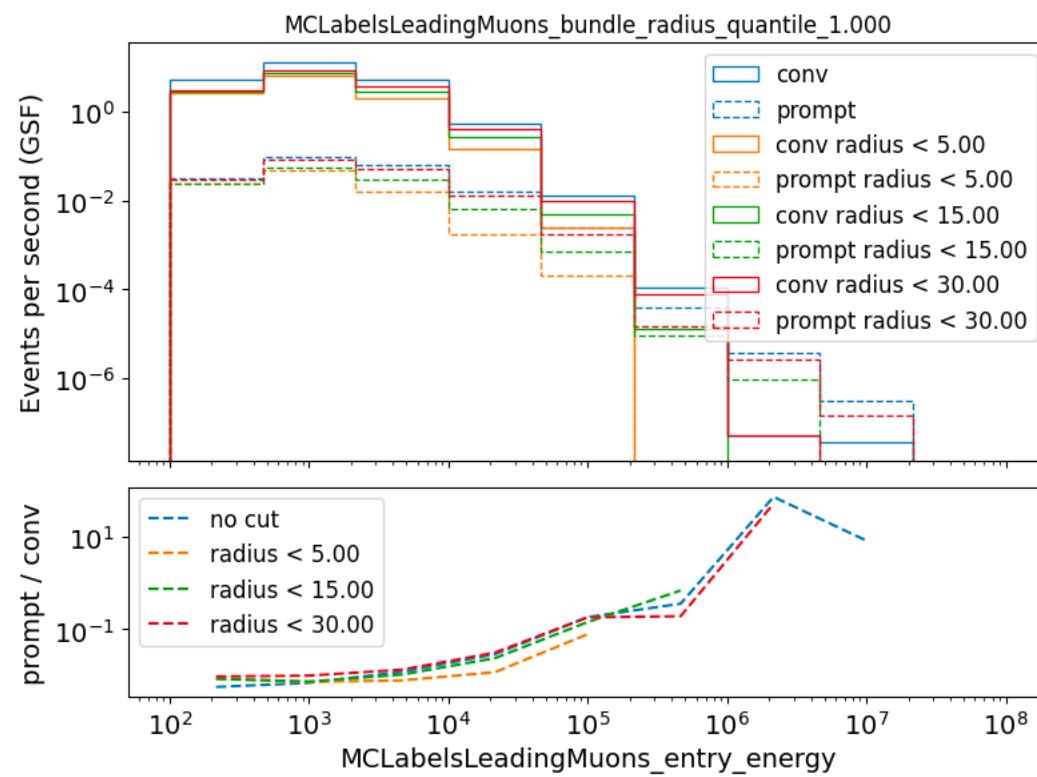
Angular resolution



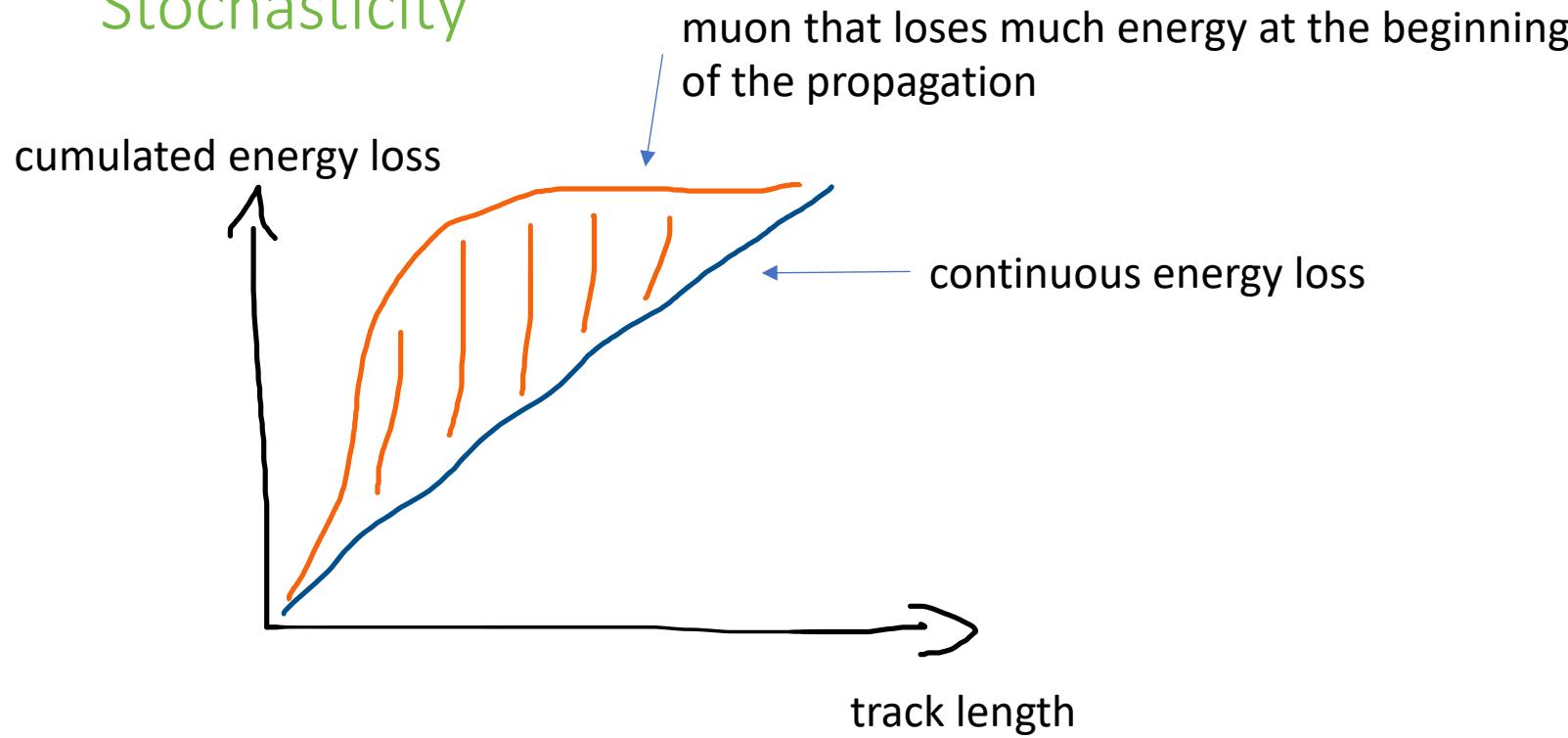
Bundle radius - energy



Muon flux – bundle radius



Stochasticity



- calculate the area between the “continuous muon energy loss” and the accumulated muon energy losses and normalize it
- high stochasticity: area = 1
- no stochasticity: area = 0

Bundle radius

- calculate perpendicular distance between leading muon and closest point to detector center
 - reference point
- calculate the distance between the leading muon and all the other muons at the reference point
- analyze the distances weighted by their energy (100%, 99%, 95%,... energy containment)