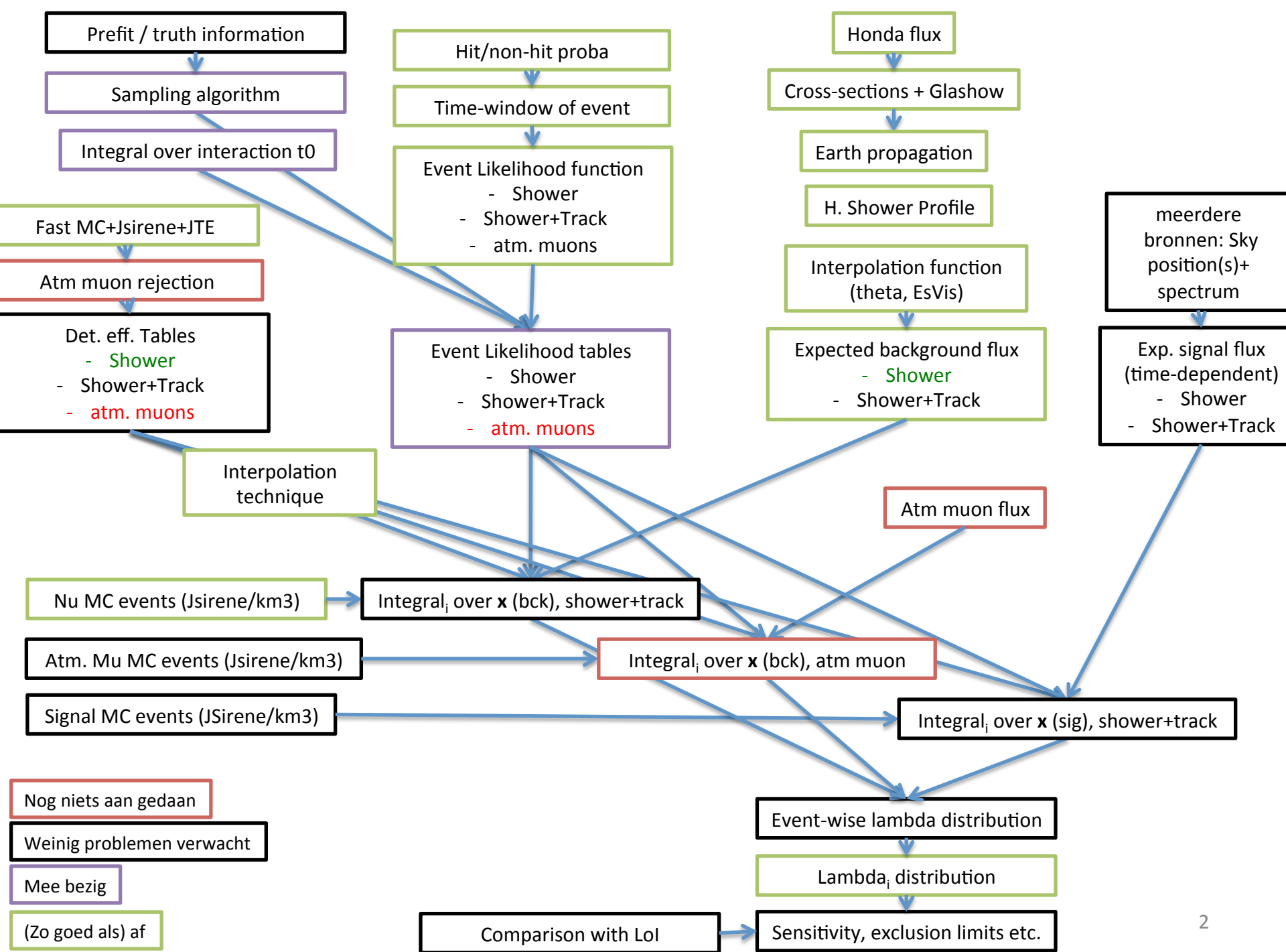


Neutrino Source Searches with Likelihood Landscapes



Neutrino Source Searches

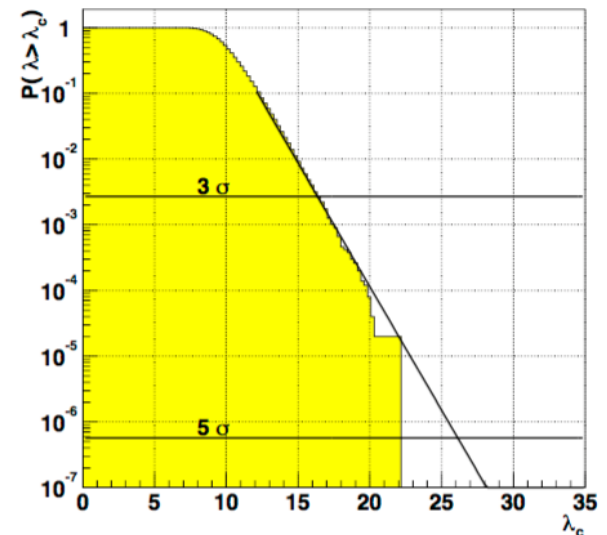
- Hypothesis H0: background only flux
 - Atmospheric neutrino's
 - (Misreconstructed) Atmospheric Muons
- Hypothesis H1: background + signal flux
 - (High energy) Cosmic Neutrinos

General Procedure

- How compatible is data with H_0 or H_1 ?

$$\lambda = \log \left[\frac{P(\text{data}|H_1)}{P(\text{data}|H_0)} \right]$$

- When to claim an observation?
 - Accept H_1 if $\lambda > \lambda_c$
 - λ_c such that
 $P(\text{accept } H_1 \mid H_0 = \text{true}) < 0.00\dots 1$

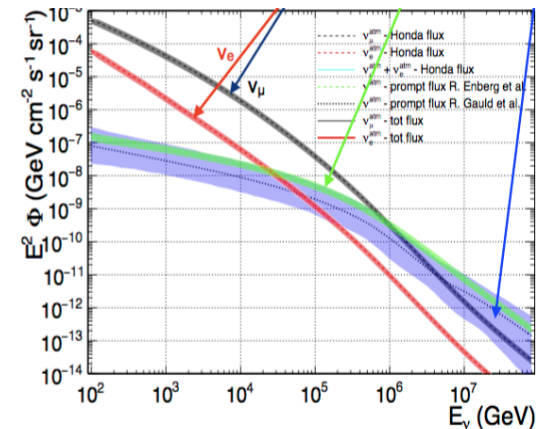
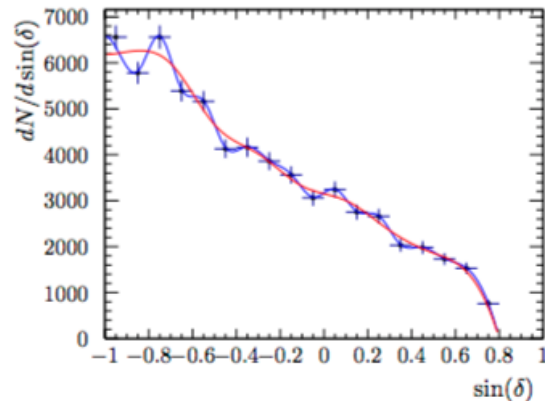
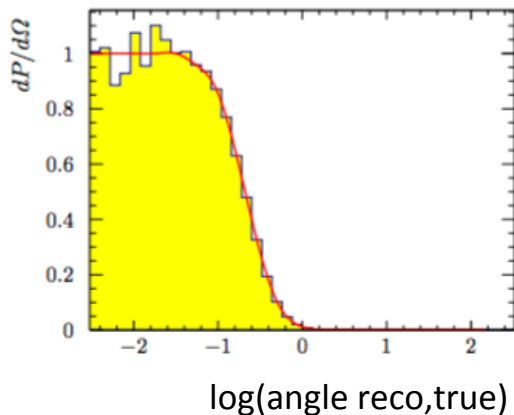


$$\lambda = \log \left[\frac{P(\text{data}|H_1)}{P(\text{data}|H_0)} \right]$$

Test Statistic (Conventional)

- Given detected (and selected) events $\{\text{ev}_i\}$

$$P(\text{data}|H) = \sum_i \left[\log \int \underbrace{P(x_{\text{reco},i} | x_{\text{true}})}_{\text{Reconstruction}} \cdot \underbrace{P^{\text{det}}(x_{\text{true}})}_{\text{Detection efficiency}} \cdot \underbrace{\mu(x_{\text{true}} | H)}_{\text{Expected flux}} dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$



$$\lambda = \log \left[\frac{P(\text{data}|H_1)}{P(\text{data}|H_0)} \right]$$

Test Statistic

- Given detected (and selected) events $\{ev_i\}$

$$P(\text{data}|H) = \sum_i \left[\log \int \underbrace{P(x_{reco,i} | x_{true})}_{\text{Reconstruction}} \cdot \underbrace{P^{det}(x_{true})}_{\text{Detection efficiency}} \cdot \underbrace{\mu(x_{true} | H)}_{\text{Expected flux}} dx_{true} \right] - \mu^{tot}(H)$$

- New method:

$$P(\text{data}|H) = \sum_i \left[\log \int P(ev_i | x_{true}) \cdot P^{det}(x_{true}) \cdot \mu(x_{true} | H) dx_{true} \right] - \mu^{tot}(H)$$

- No big deal?

New vs. Conventional

Conventional

- Only best solution kept from reconstruction
- Selection criteria needed to select well-reconstructed events -> events are lost
- Different reconstruction algorithms (showers/tracks/tau double bang) patched together
- Event identification by BDT's and other black magic algorithms
- Parameterizations of MC events
- Fast

New Method

- Detailed knowledge of event likelihood landscape
- All events can be used
- Single 'reconstruction' algorithm for all events
- Neutrino flavour identification automatically taken into account
- Event-by-event
- Probably slow

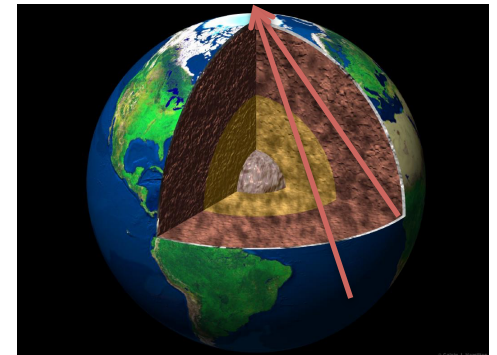
Likelihood Ingredients

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

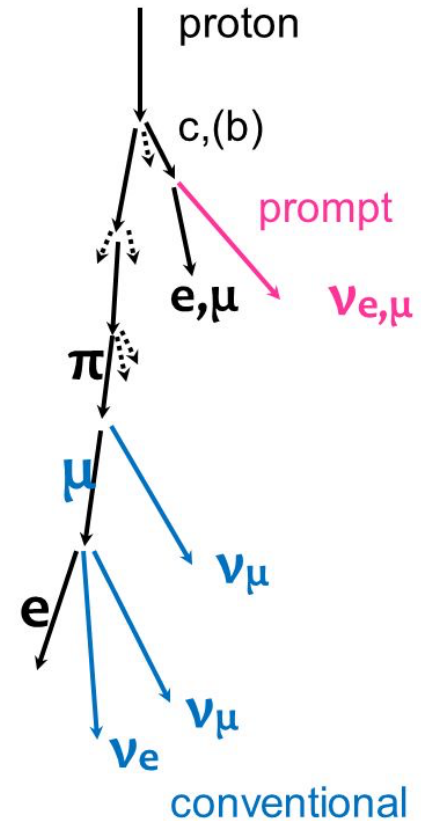
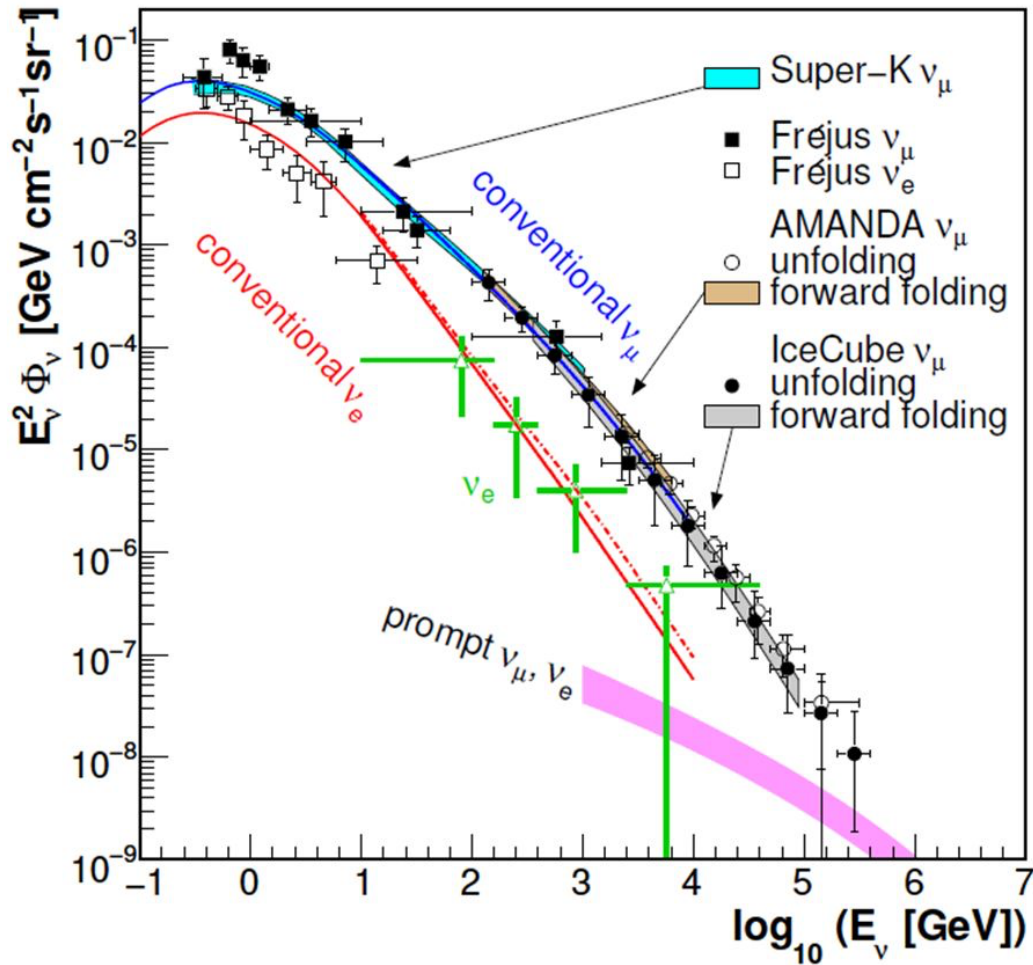
$\mu(x_{\text{true}} | H)$ Number of expected background or signal events in our detector (can)

$P^{\text{det}}(x_{\text{true}})$

$P(\text{ev}_i | x_{\text{true}})$



Atmospheric Neutrinos

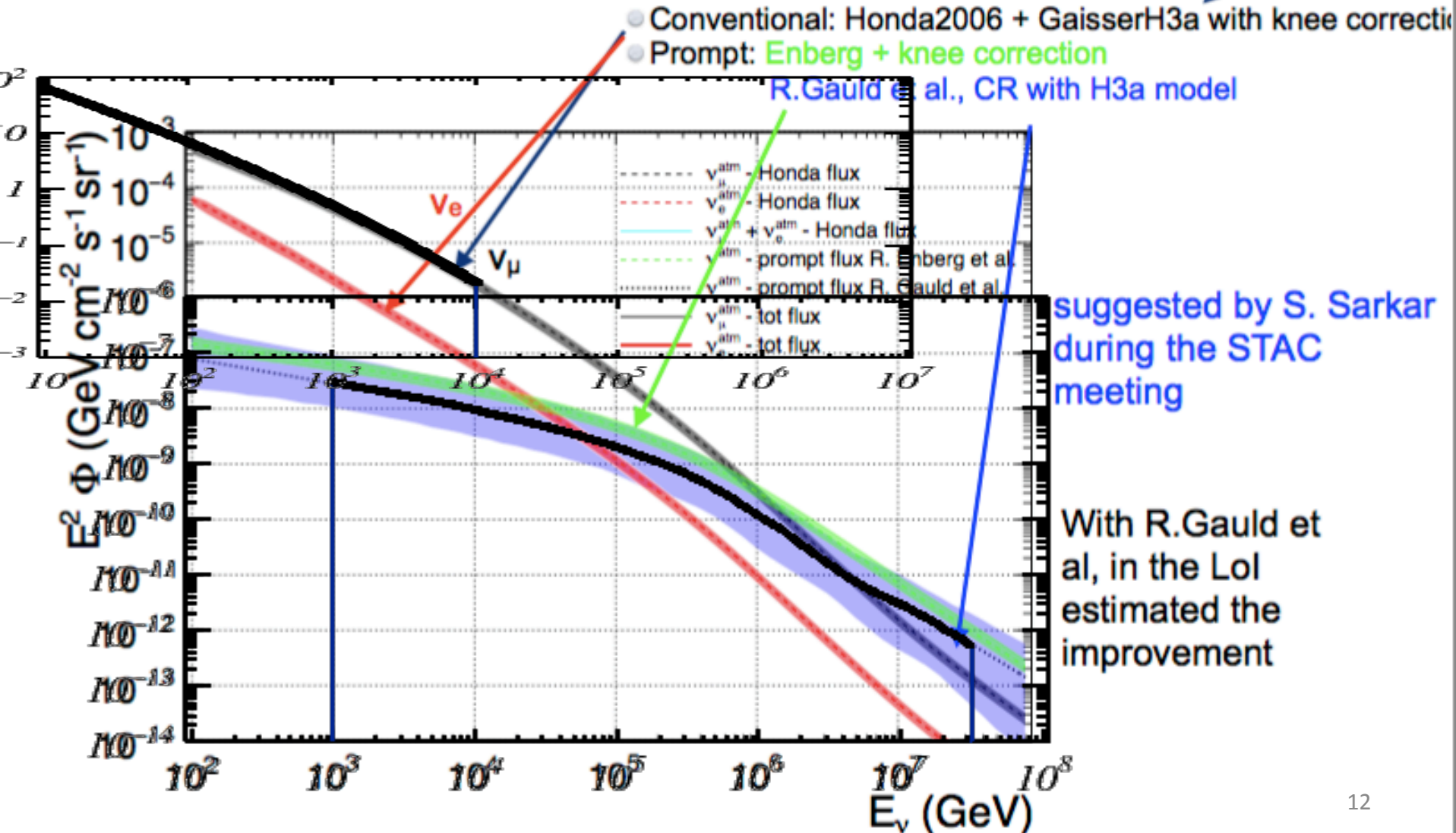


Current Parameterization

- KM3NeT Letter of Intent
- Based on Seatray
- Polynomial fit of Honda tables
 - Extrapolation to higher energy ranges
 - Outdated? Honda 2006 used.
 - Gaisser H3a knee correction
- Polynomial fit of Gauld tables 2015
 - From PromptNuFlux, L. Rottoli

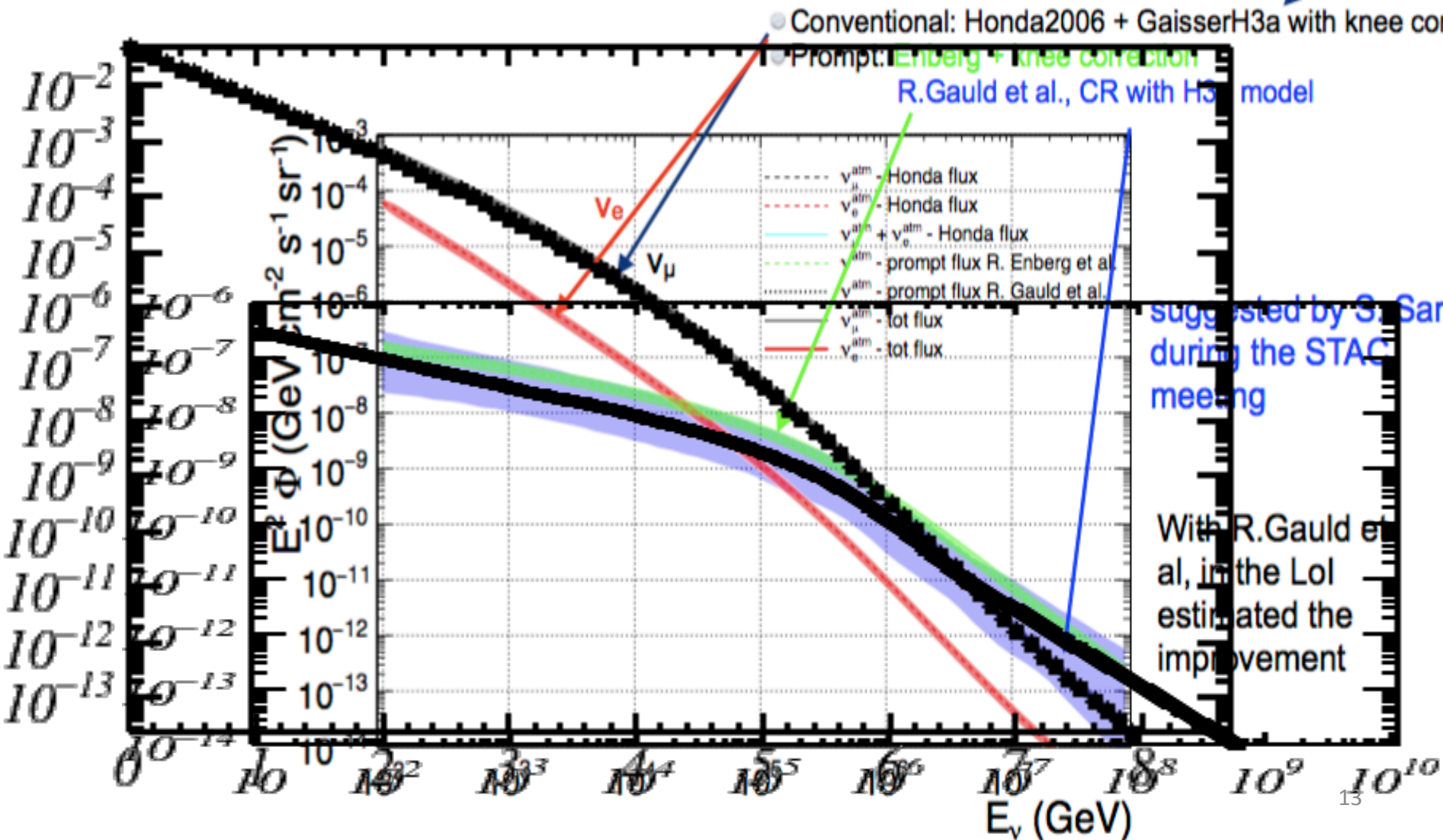
Honda (2006) and Gauld (2016)

T. Gaisser 2012



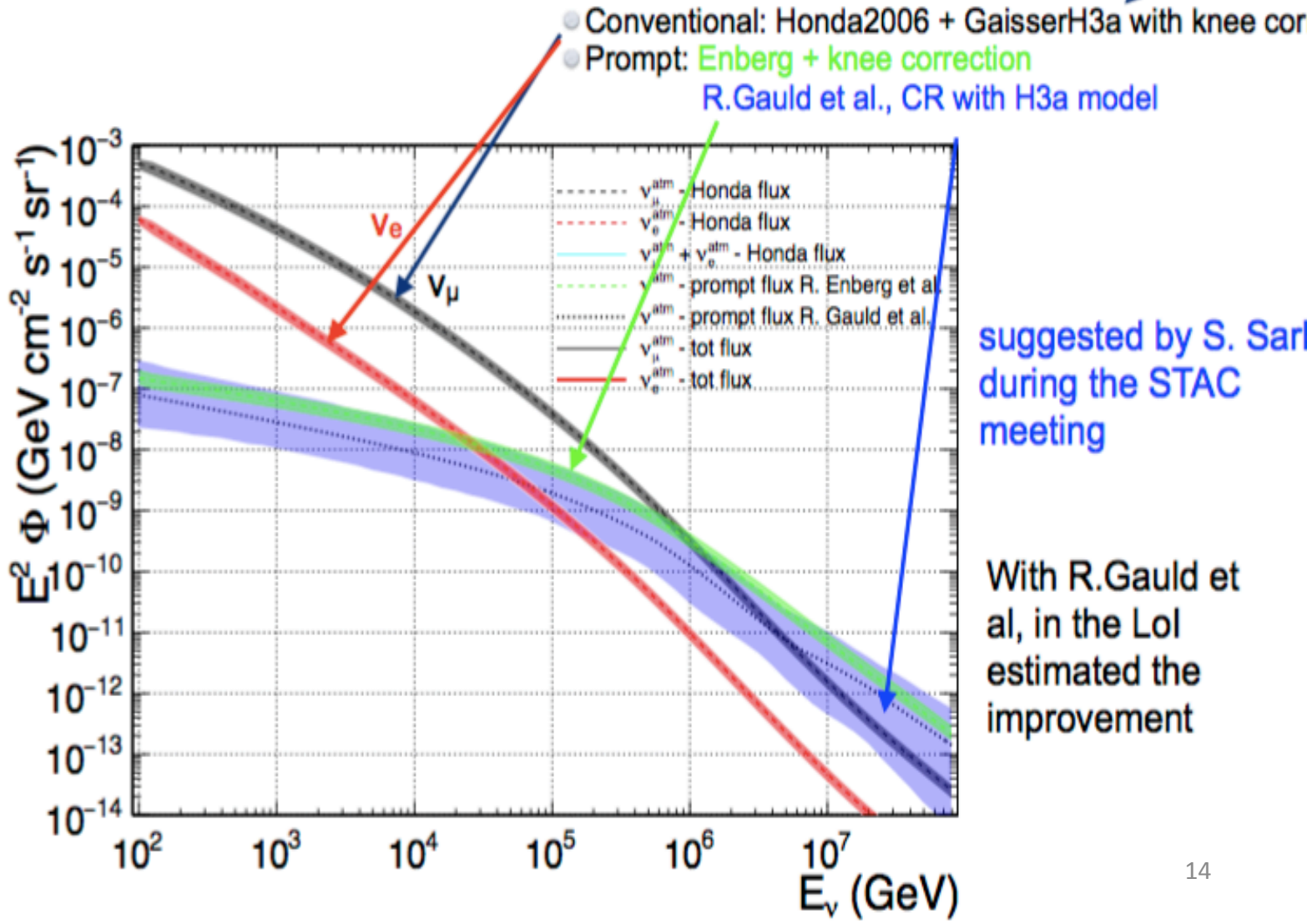
Both Extrapolated (2)

T. Gaisser 2012



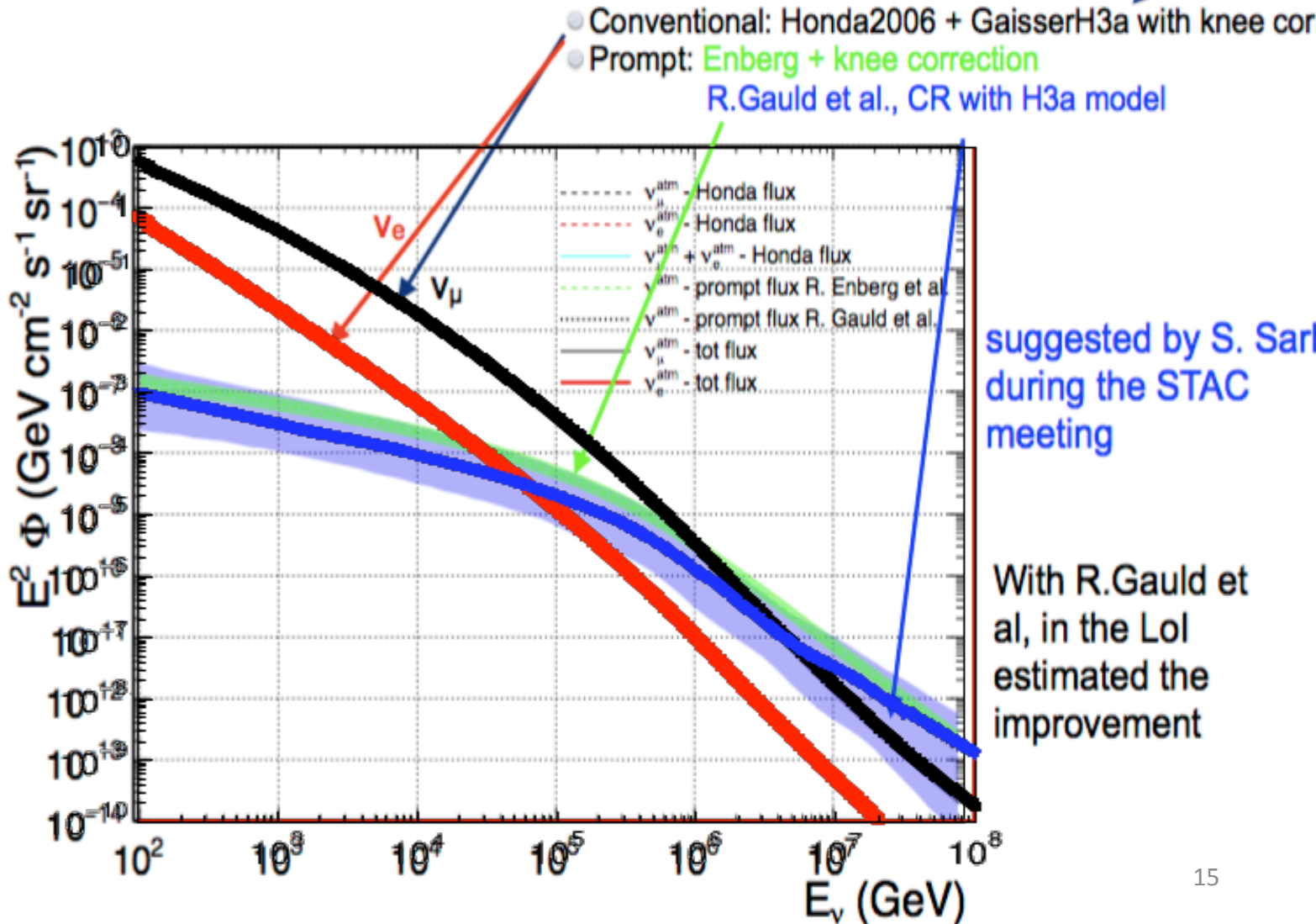
Both Extrapolated (2)

T. Gaisser 2012



Both Extrapolated (2)

T. Gaisser 2012

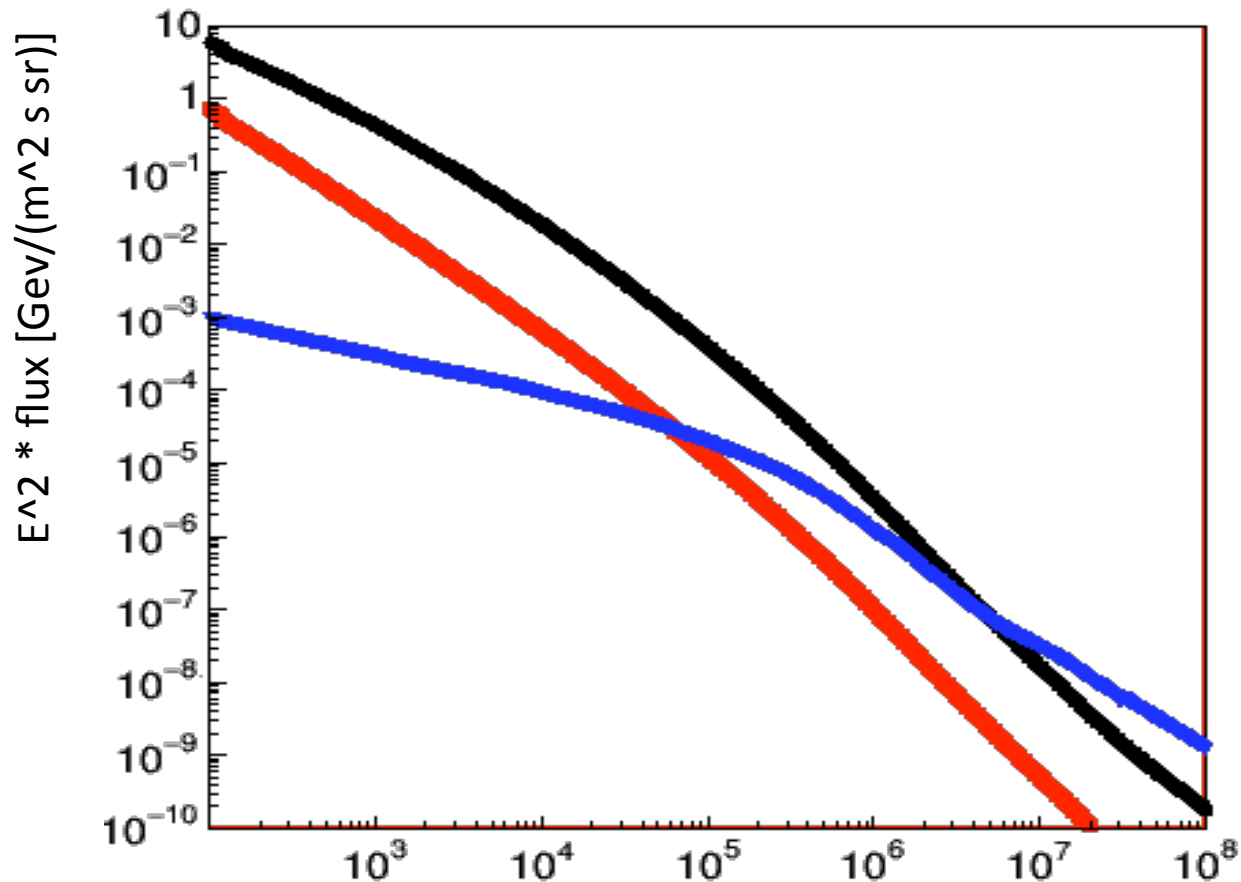


Both Extrapolated (2)

NuMu + AnuMu (Honda 2006) + gaisser H3A

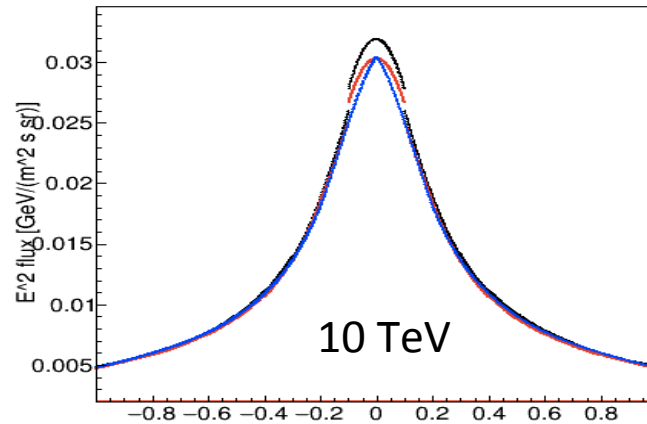
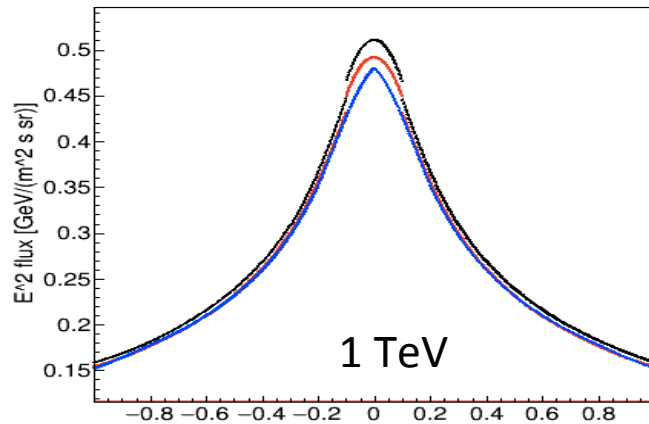
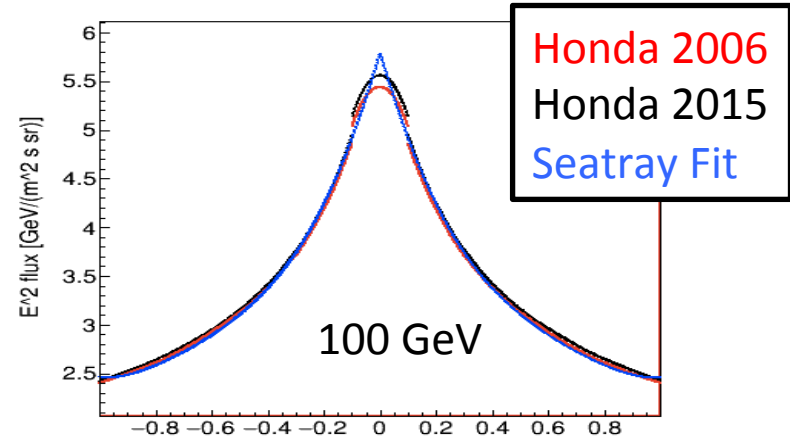
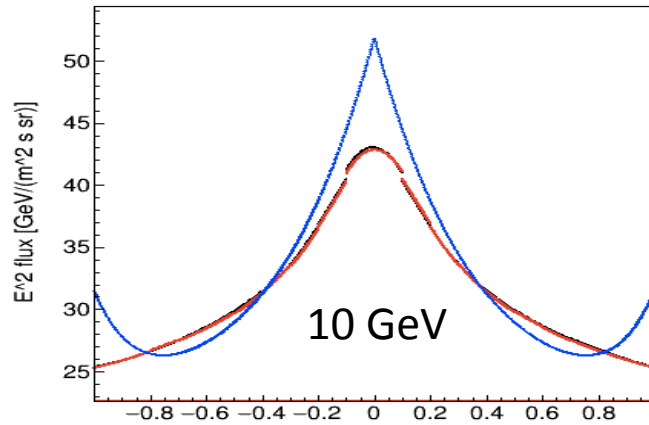
NuE + AnuE (Honda 2006) + gaisser H3A

Prompt flux (indep. of flavor), Gauld, includes H3A



Honda: Zenith Dependence

$E^2 \times \text{flux} [\text{GeV}/(\text{m}^2 \text{ s sr})]$



$\cos(\text{Zenith})$

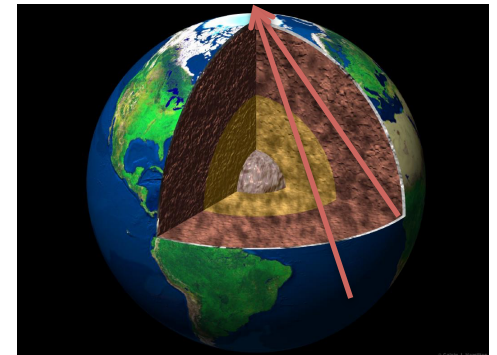
Likelihood Ingredients

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

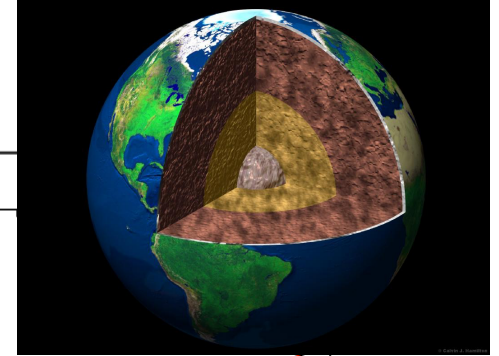
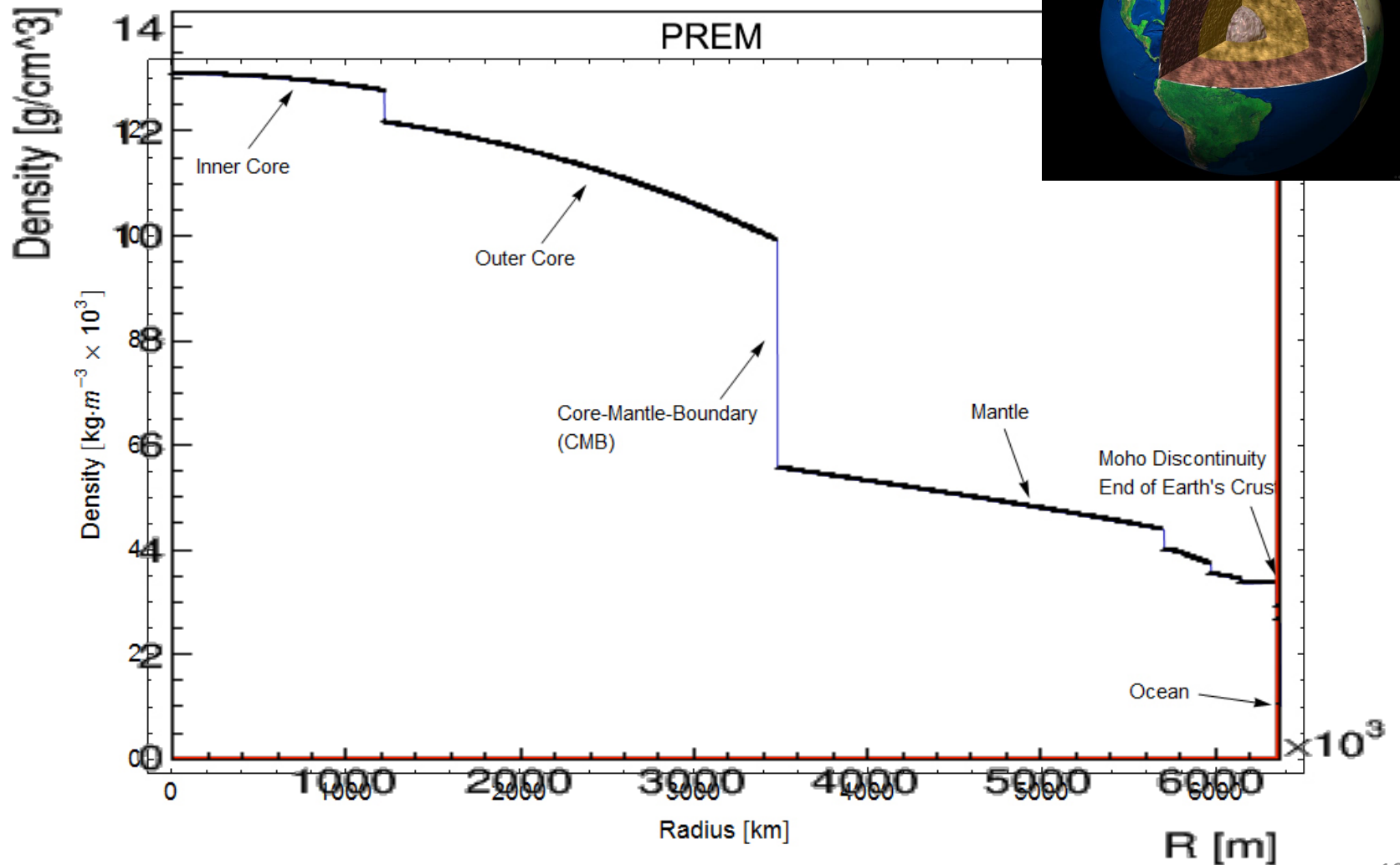
$\mu(x_{\text{true}} | H)$ Number of expected background or signal events in our detector (can)

$P^{\text{det}}(x_{\text{true}})$

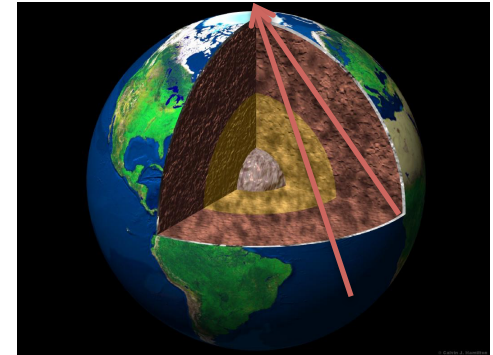
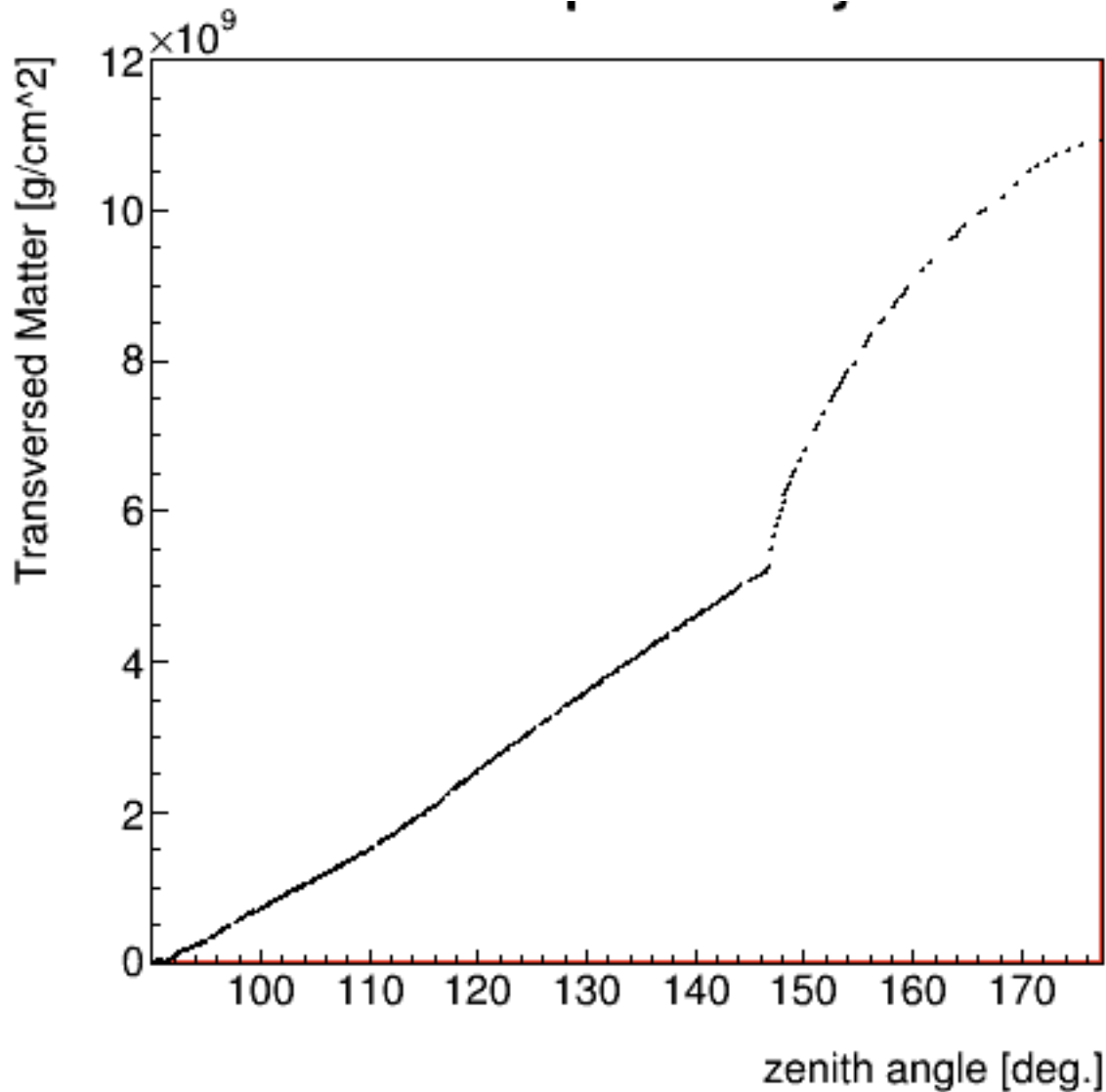
$P(\text{ev}_i | x_{\text{true}})$



Earth Propagation

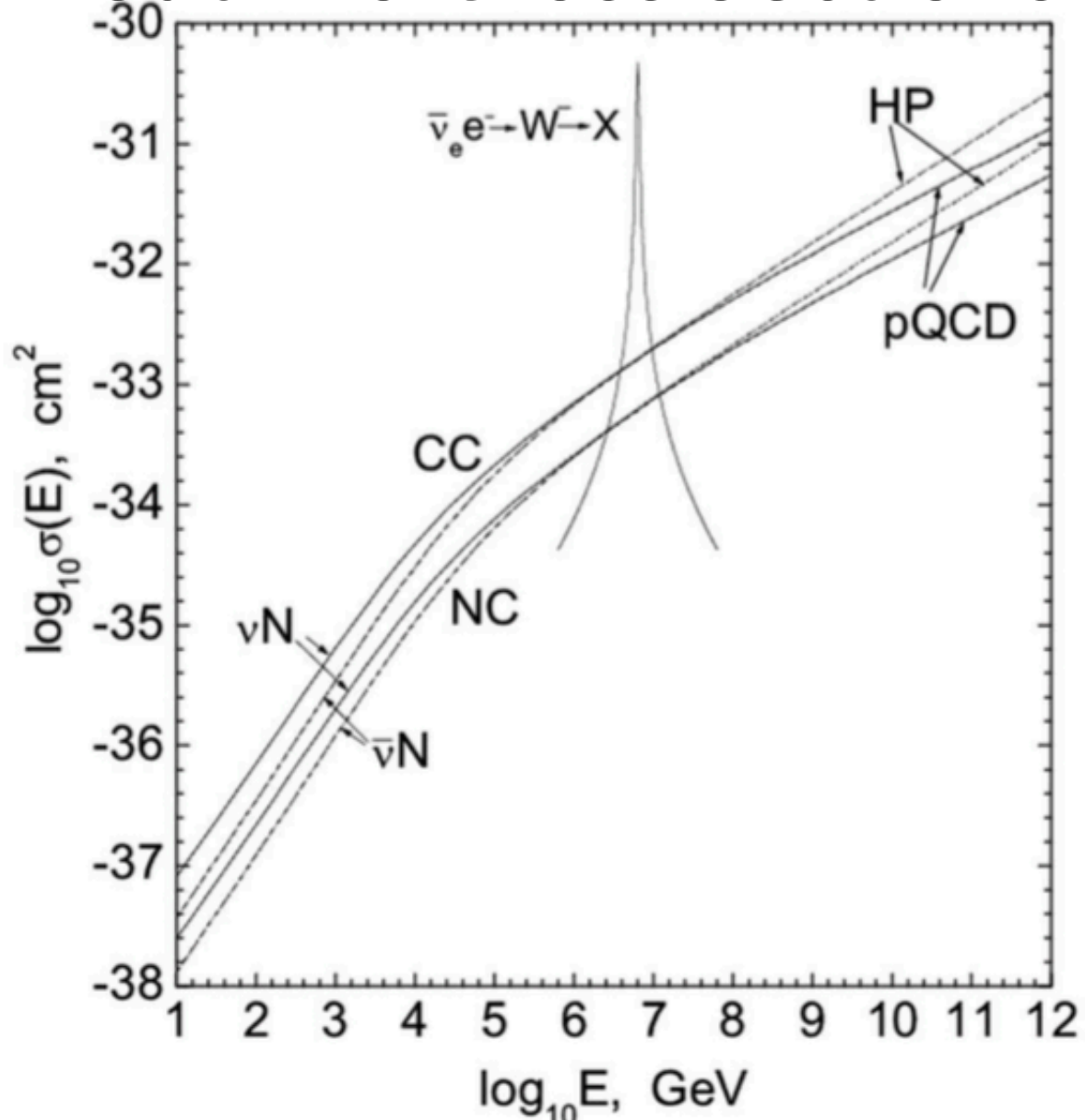


Transversed Matter Density

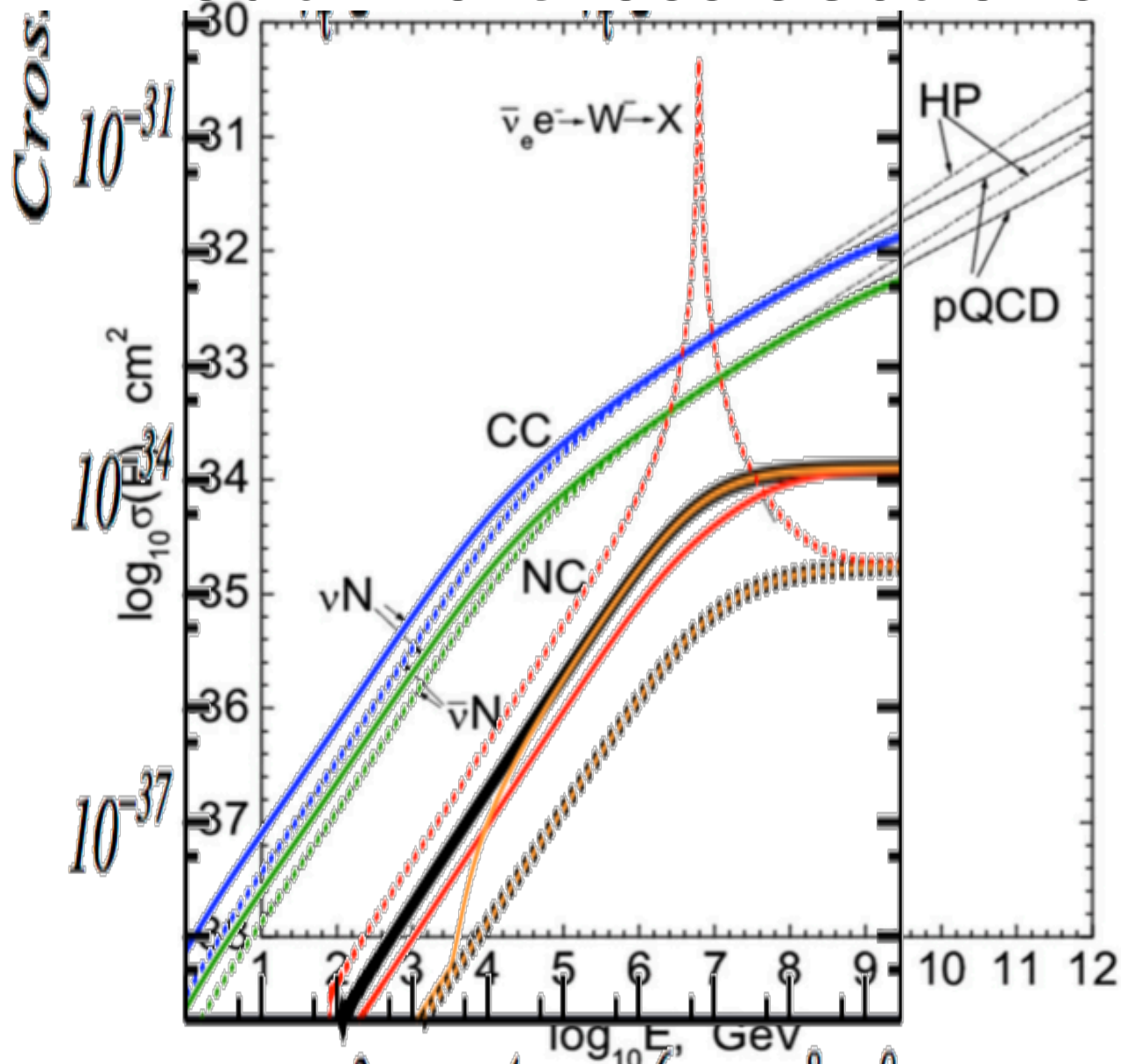


Analytically derived -> very fast

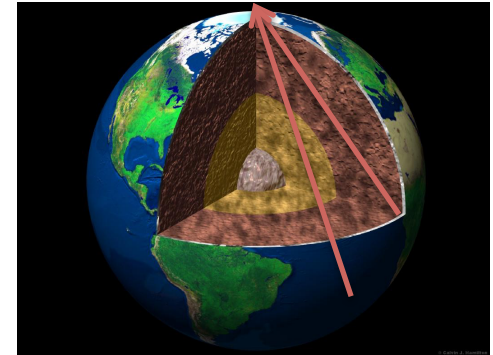
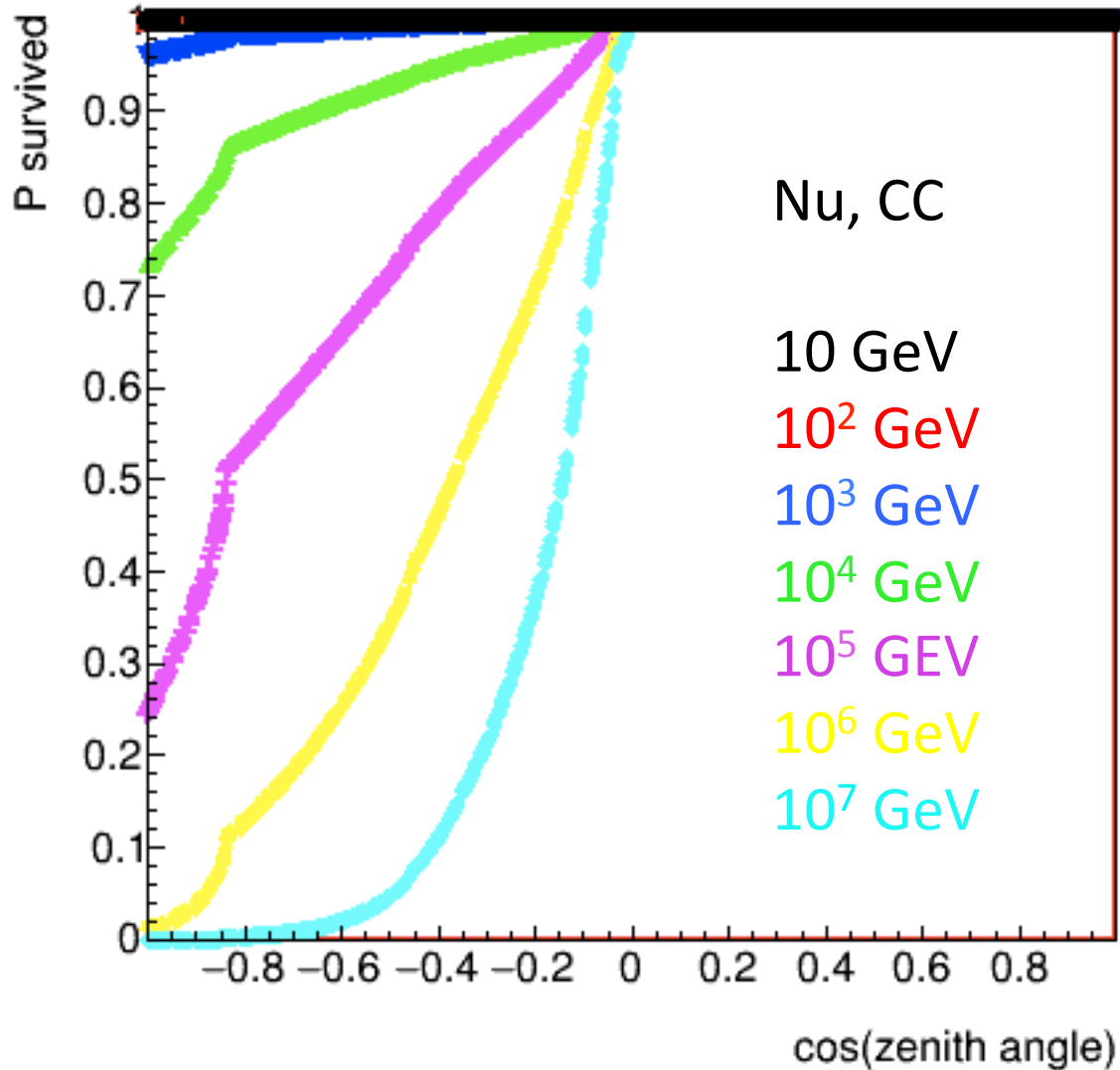
Neutrino Cross Sections



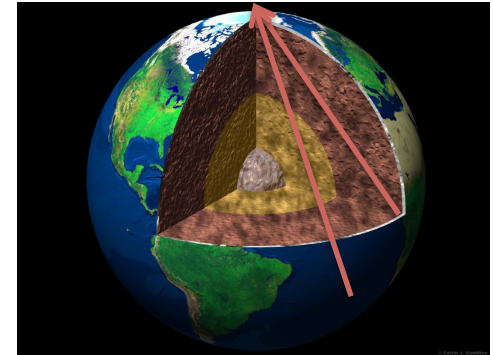
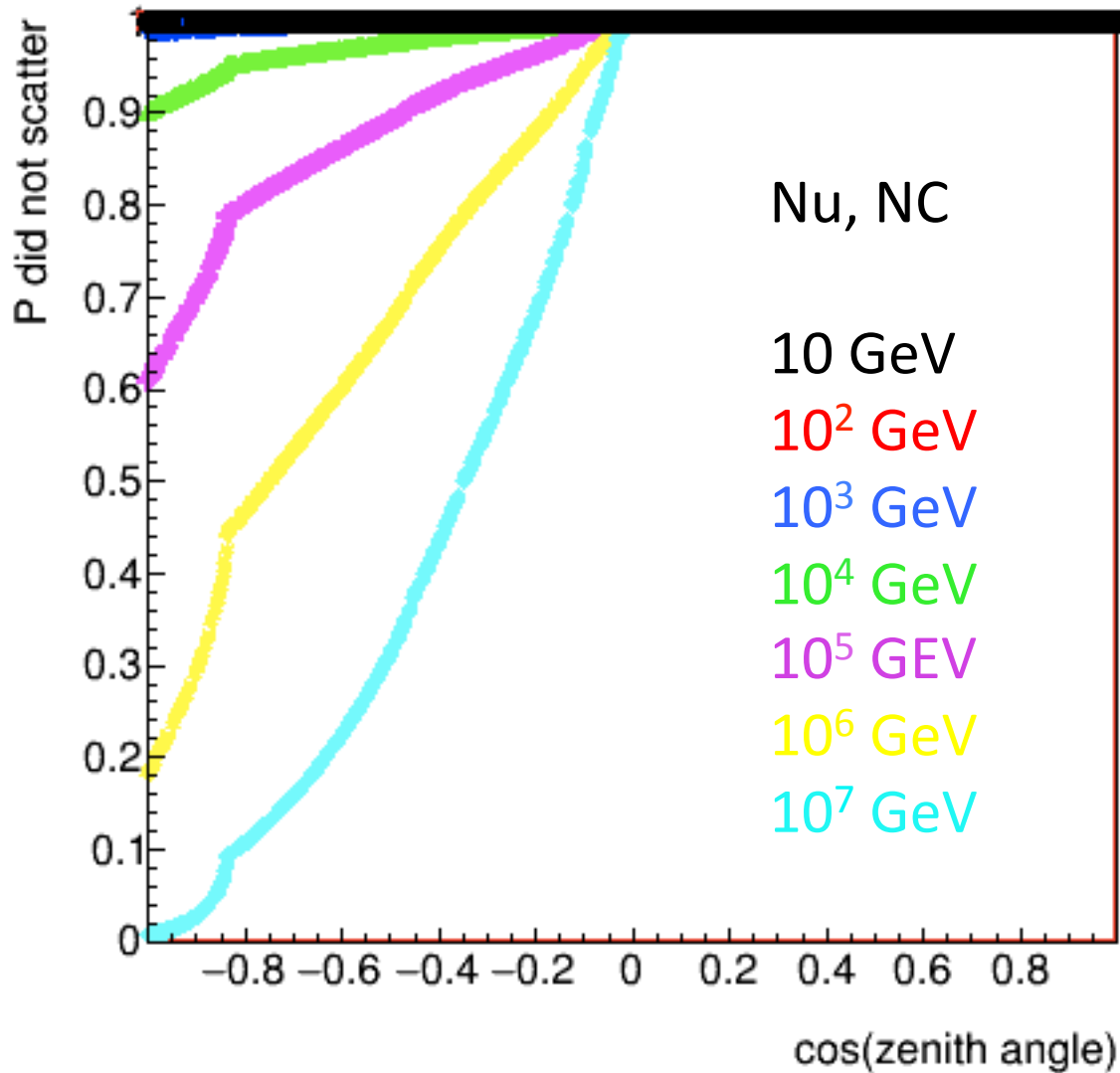
Neutrino Cross Sections



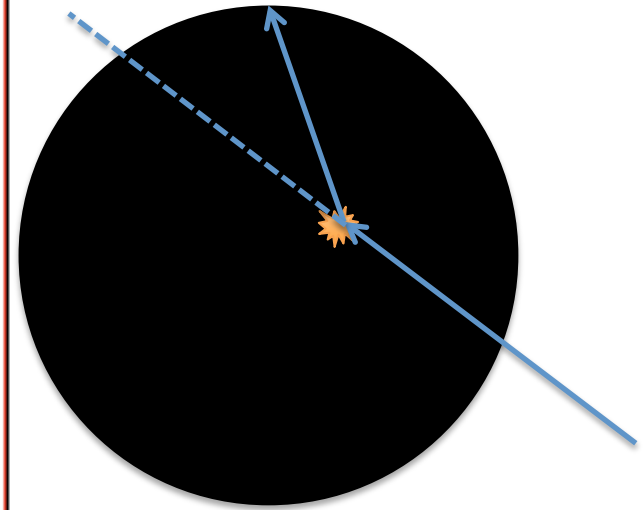
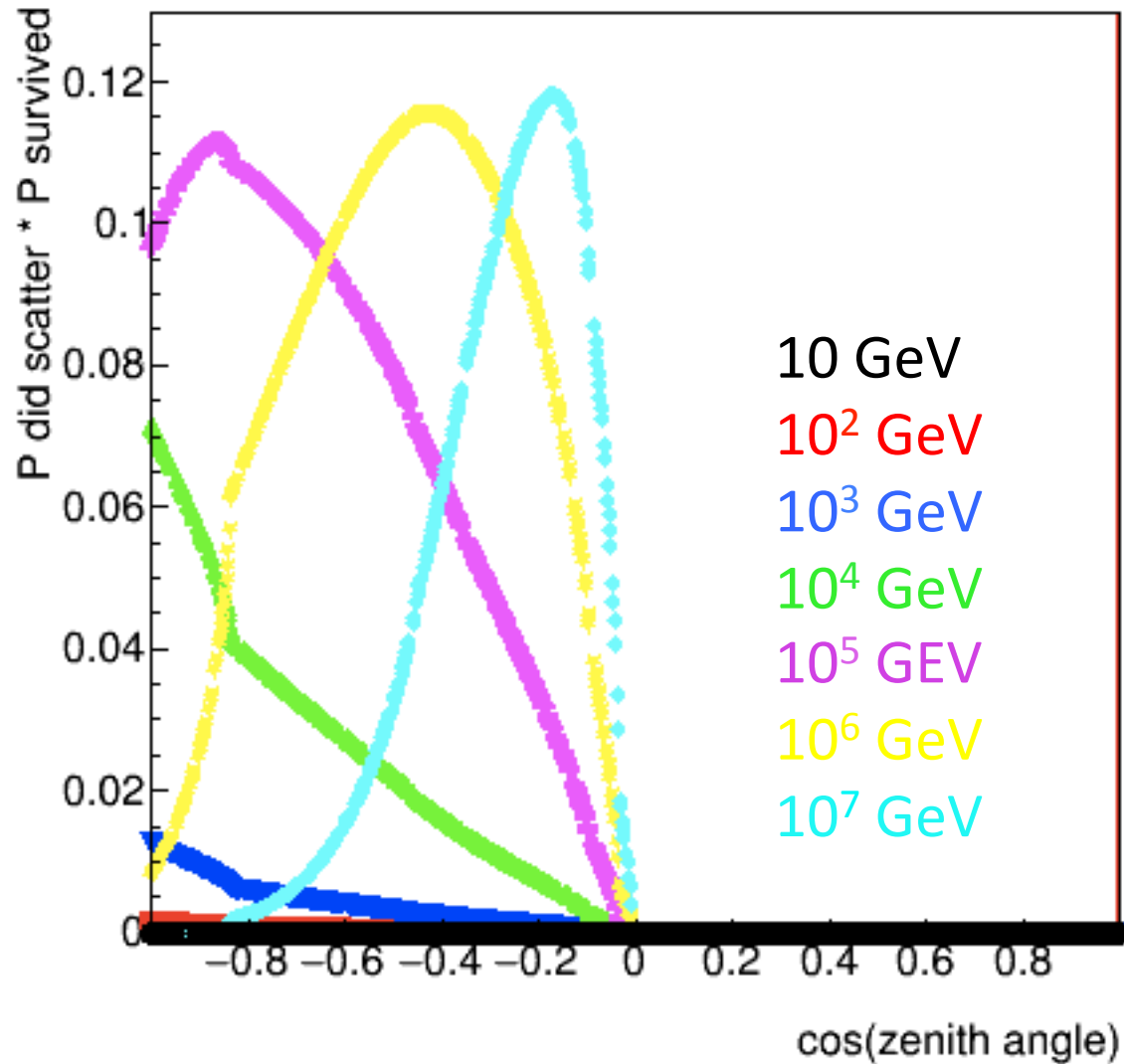
Neutrino Absorption



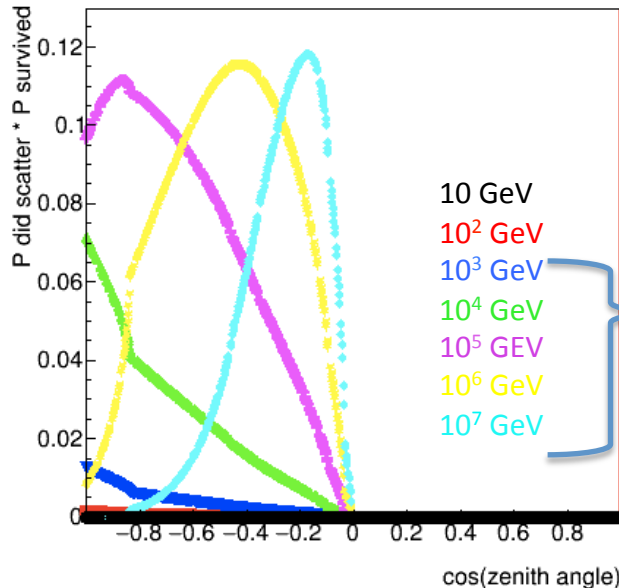
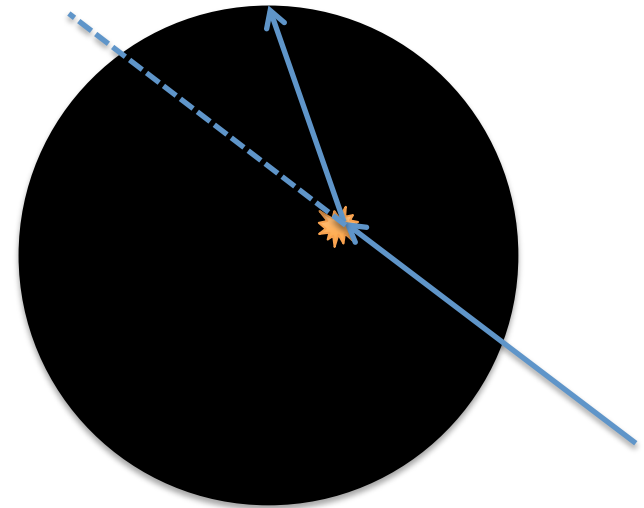
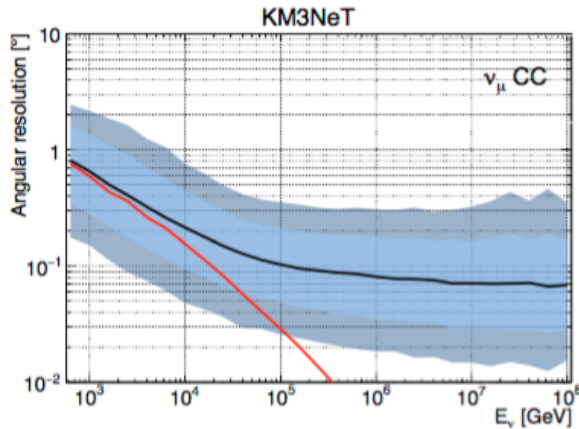
Neutrino NC Scattering (1)



Neutrino NC Scattering (2)



Neutrino NC Scattering (3)



- Change in direction: ≈ 0.6 degrees for $E_{\text{nu}} > 10^3$ GeV
- Change in Energy???

Effects on expected atm. Neutrino flux neglected

Neutrino Oscillations

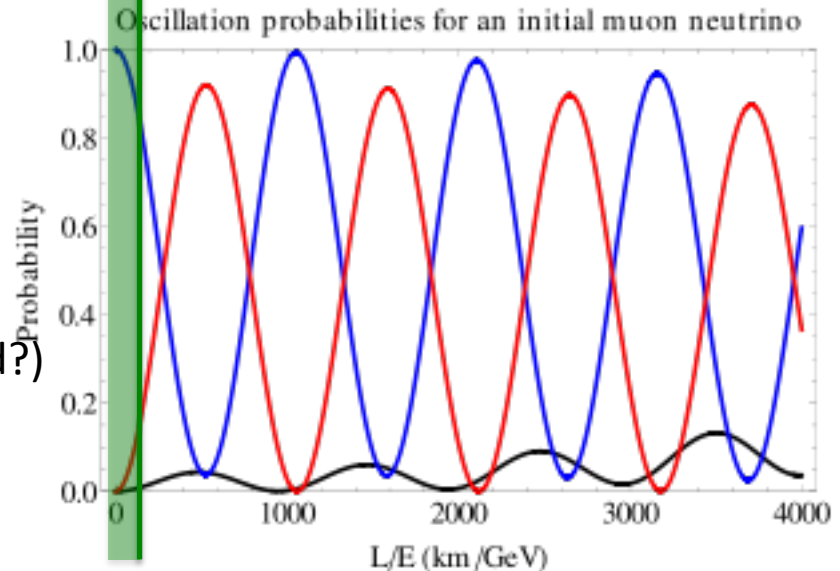
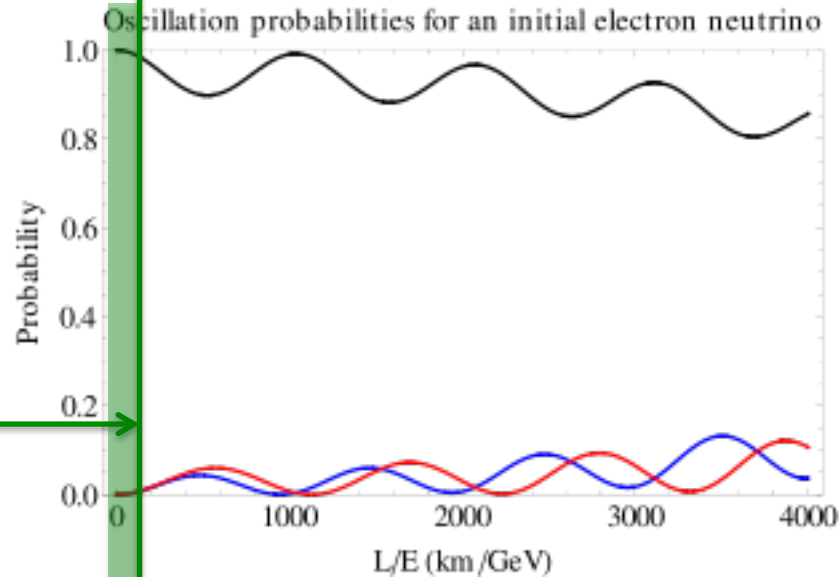
Earth radius = 6.4×10^3 km
 100 GeV neutrino

$L/E = 1.28 \times 10^2$ km/GeV

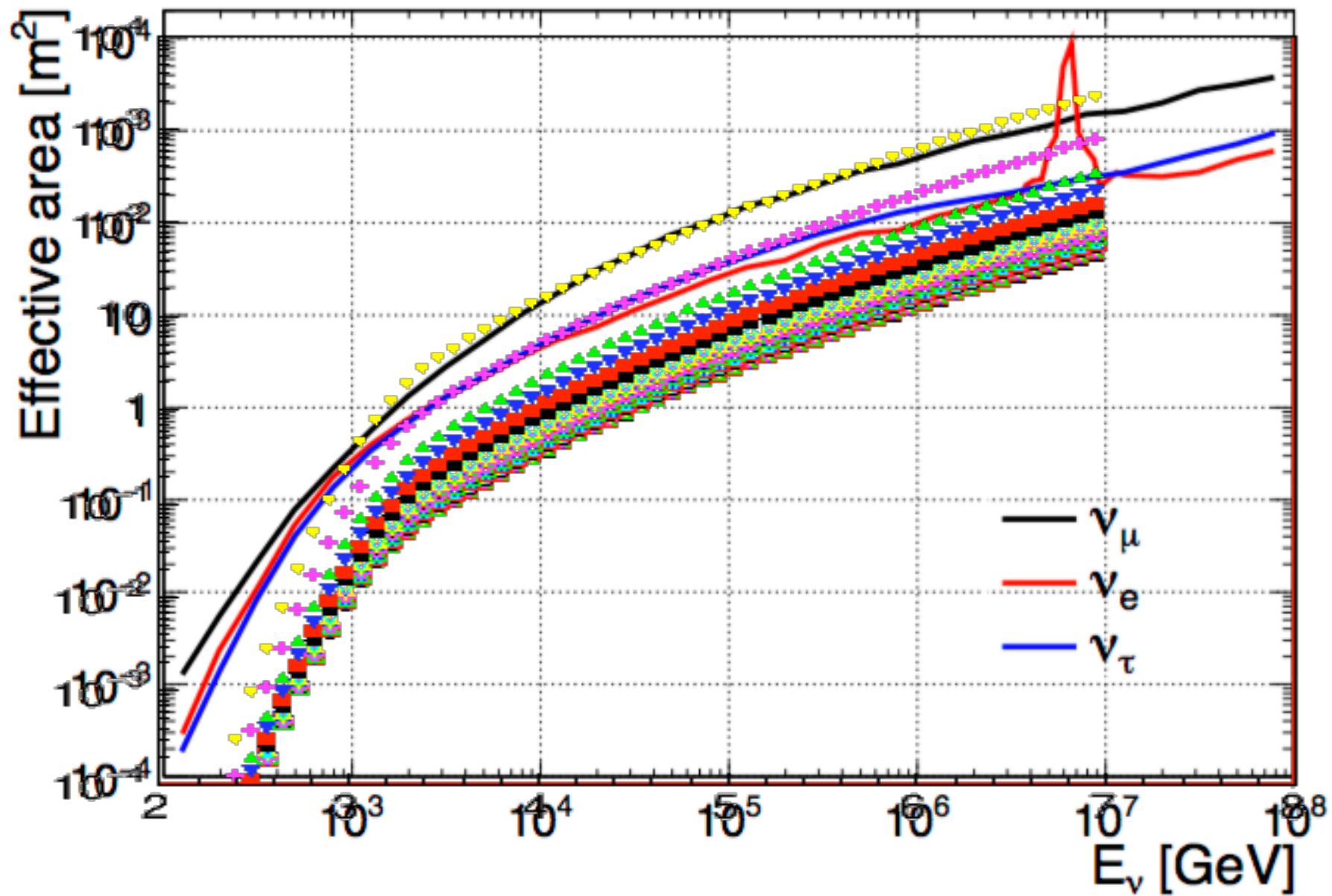
Earth covers one oscillation period
 P(oscilate) up to:

- ~0.01 (electron \rightarrow muon/tau)
- ~0.2 (muon \rightarrow tau)
- ~0.2 (tau \rightarrow muon)

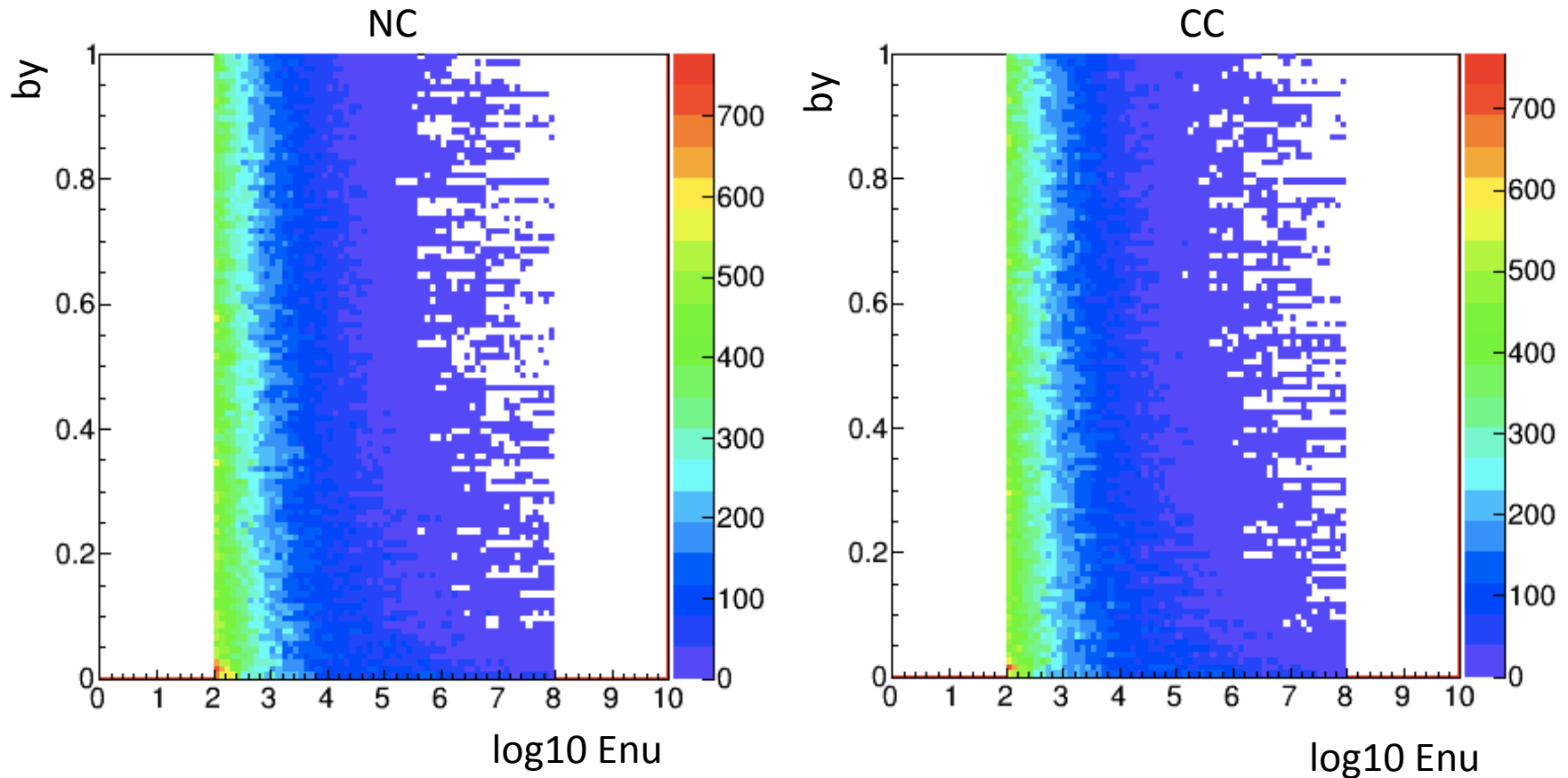
For now: Ignore.... (to be continued?)



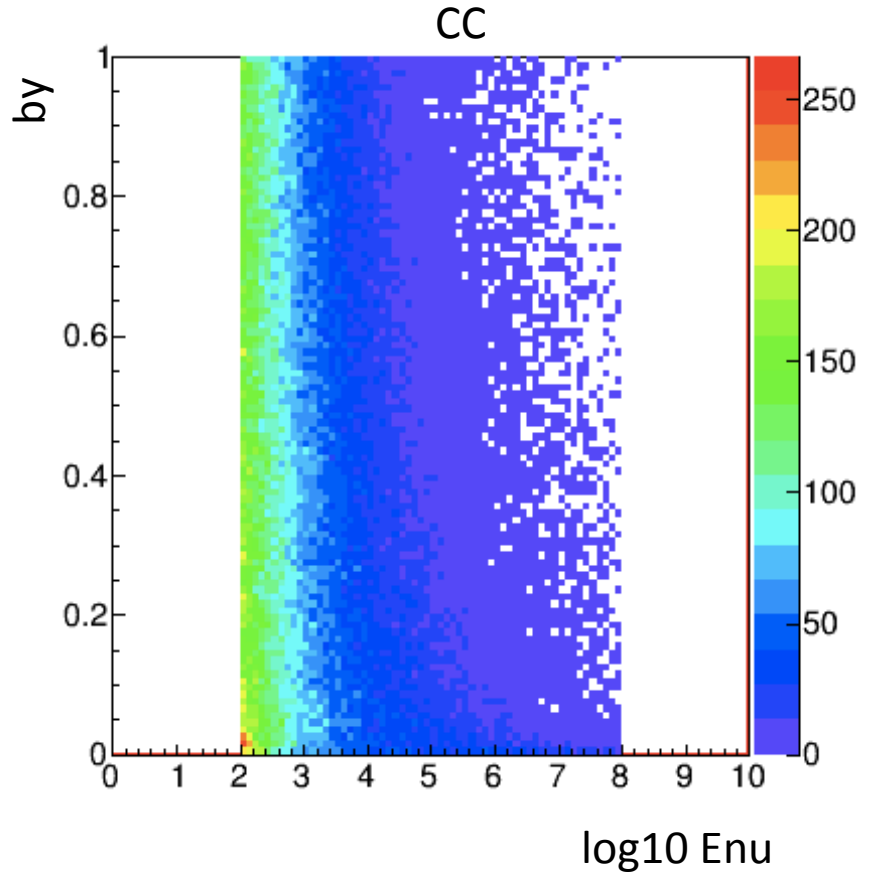
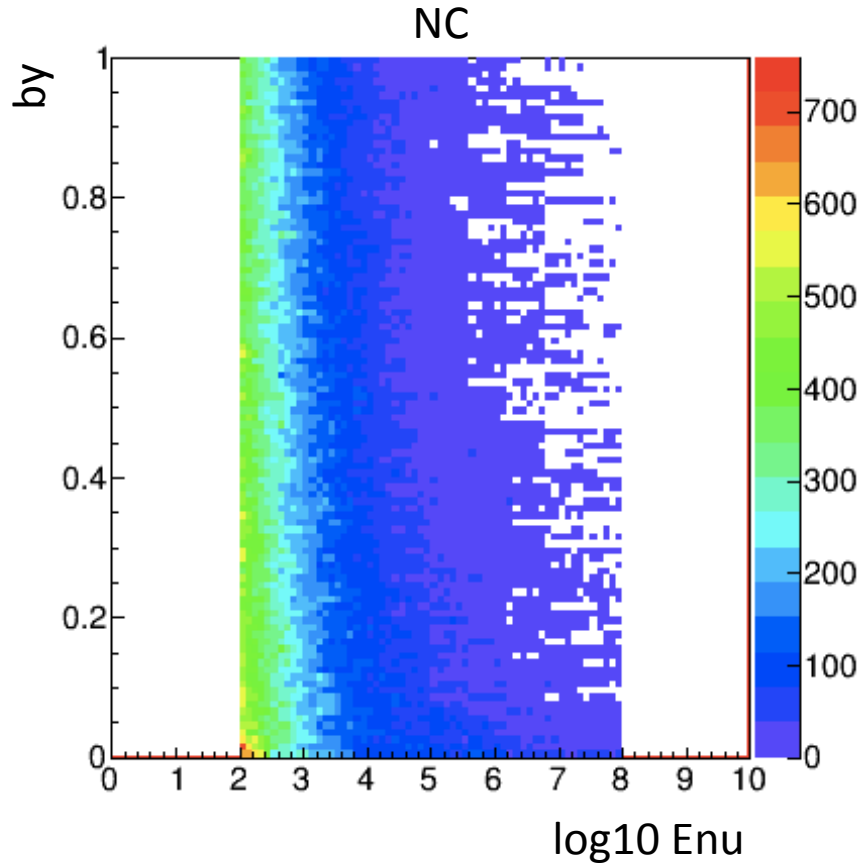
KM3NeT



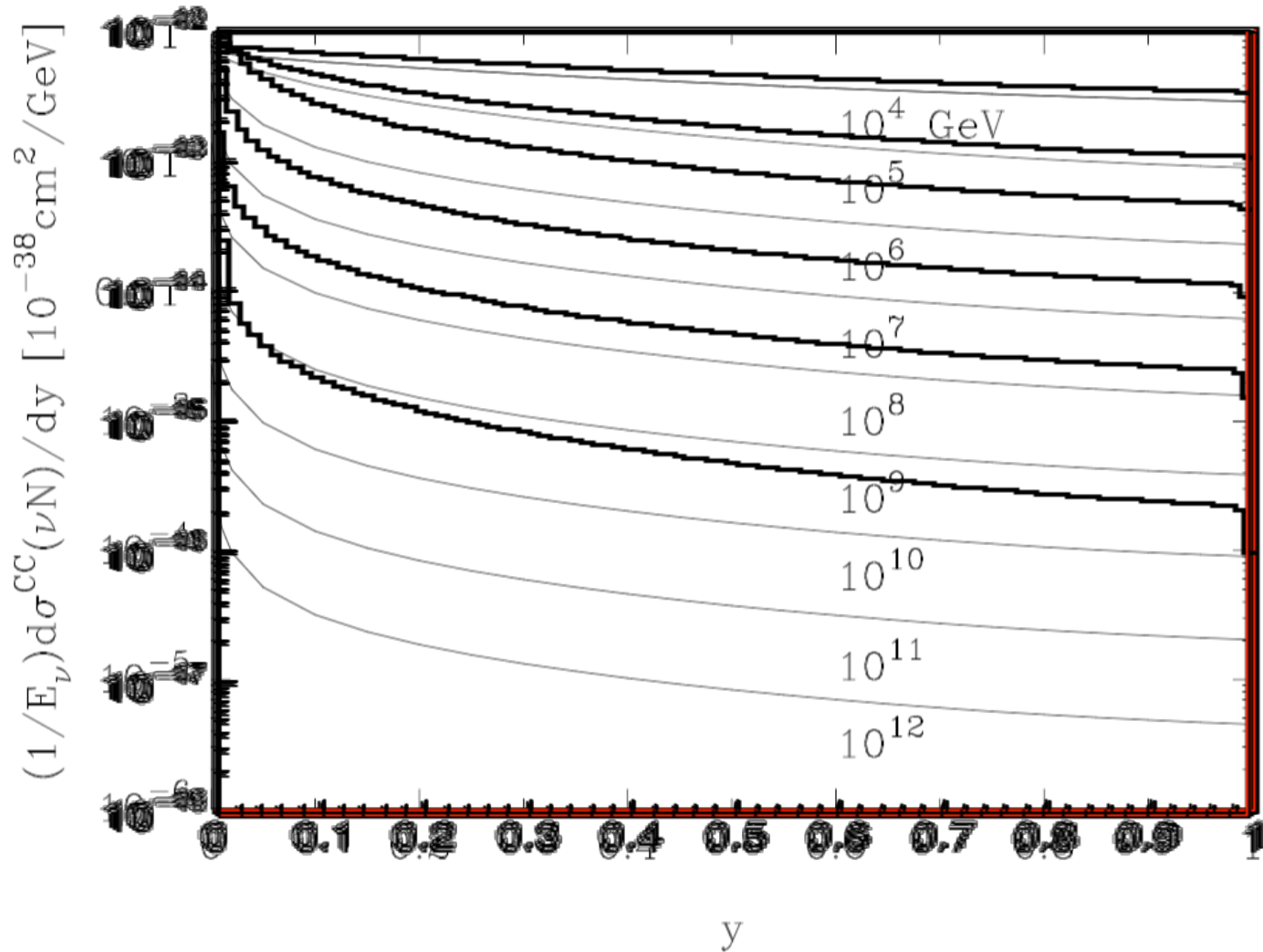
Bjorken-y: electron neutrino's



Bjorken-y: muonneutrino's



Bjorken-y comparison



Bjorken-y comparison

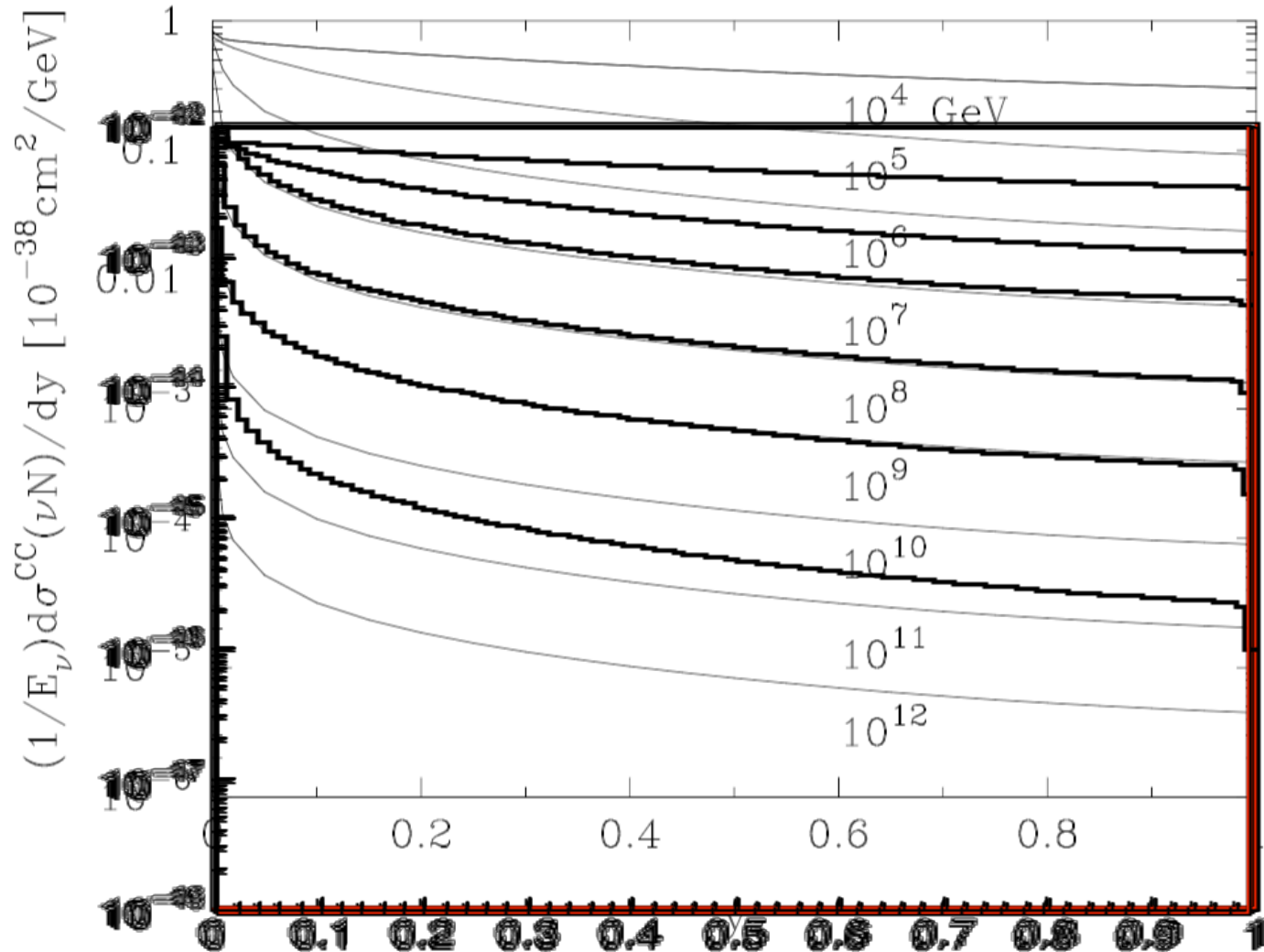
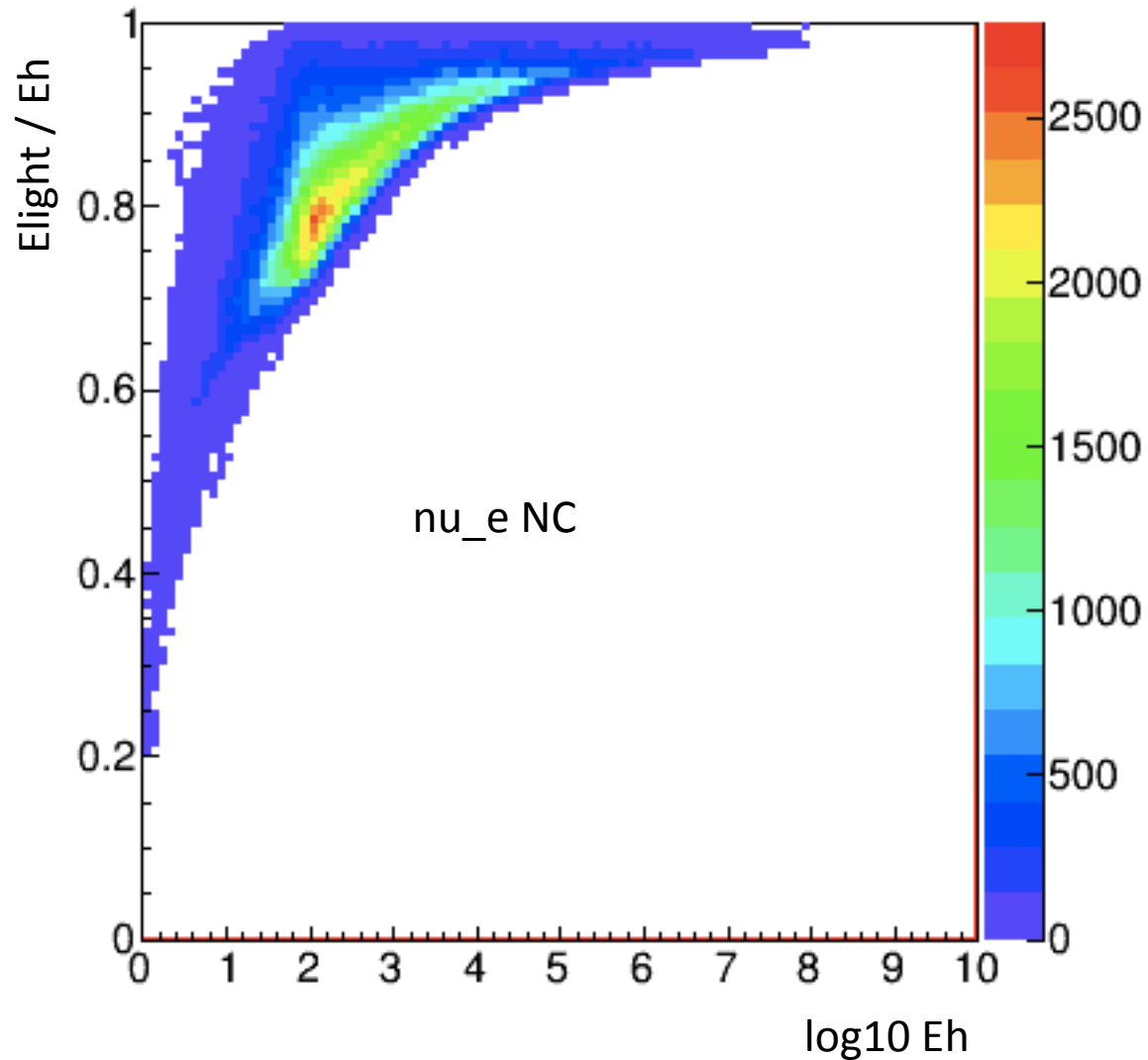
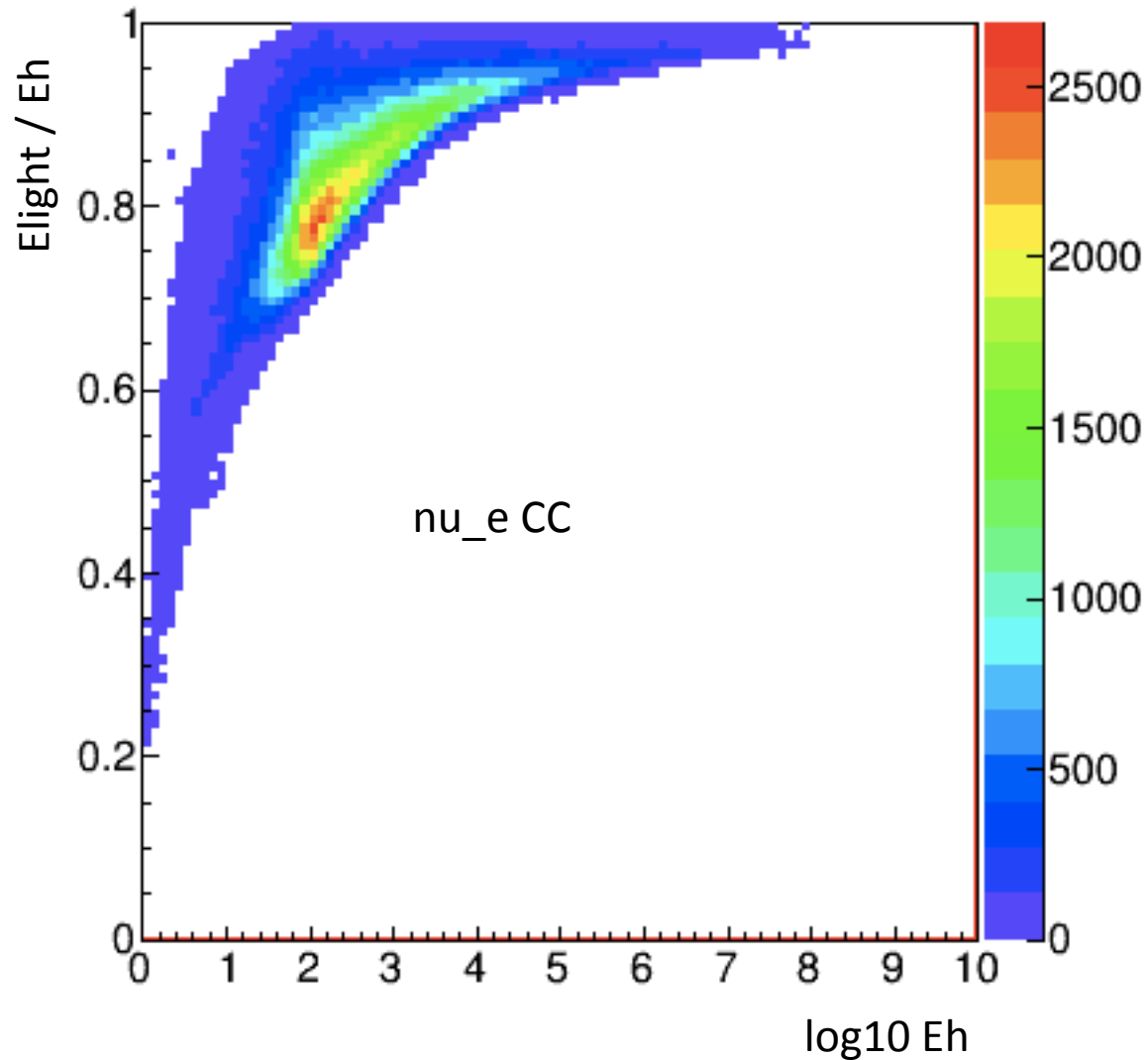


Fig. 6. Differential cross section for νN scattering for neutrino energies between 10^4 GeV and 10^{12} GeV.

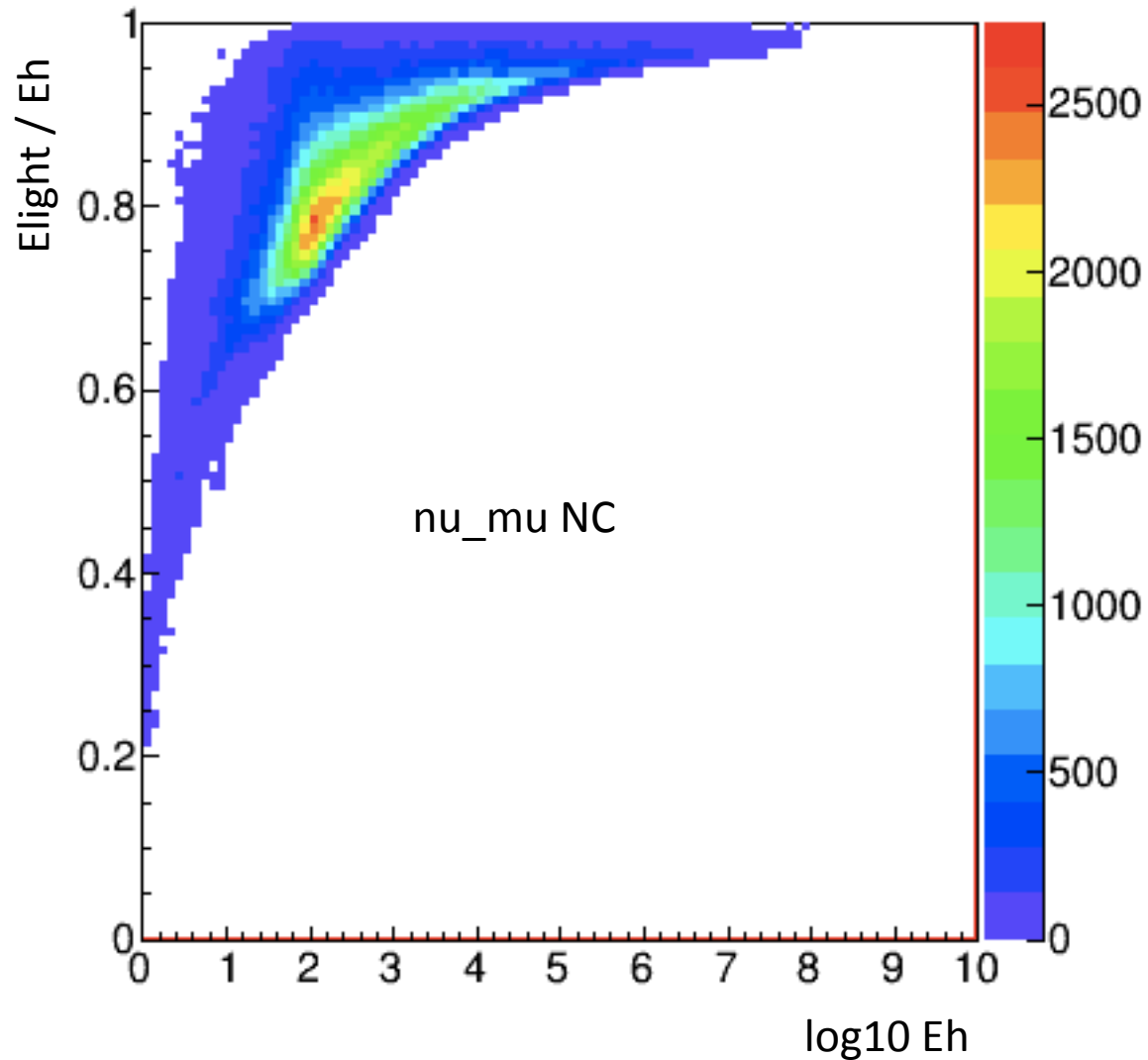
Light from hadronic showers



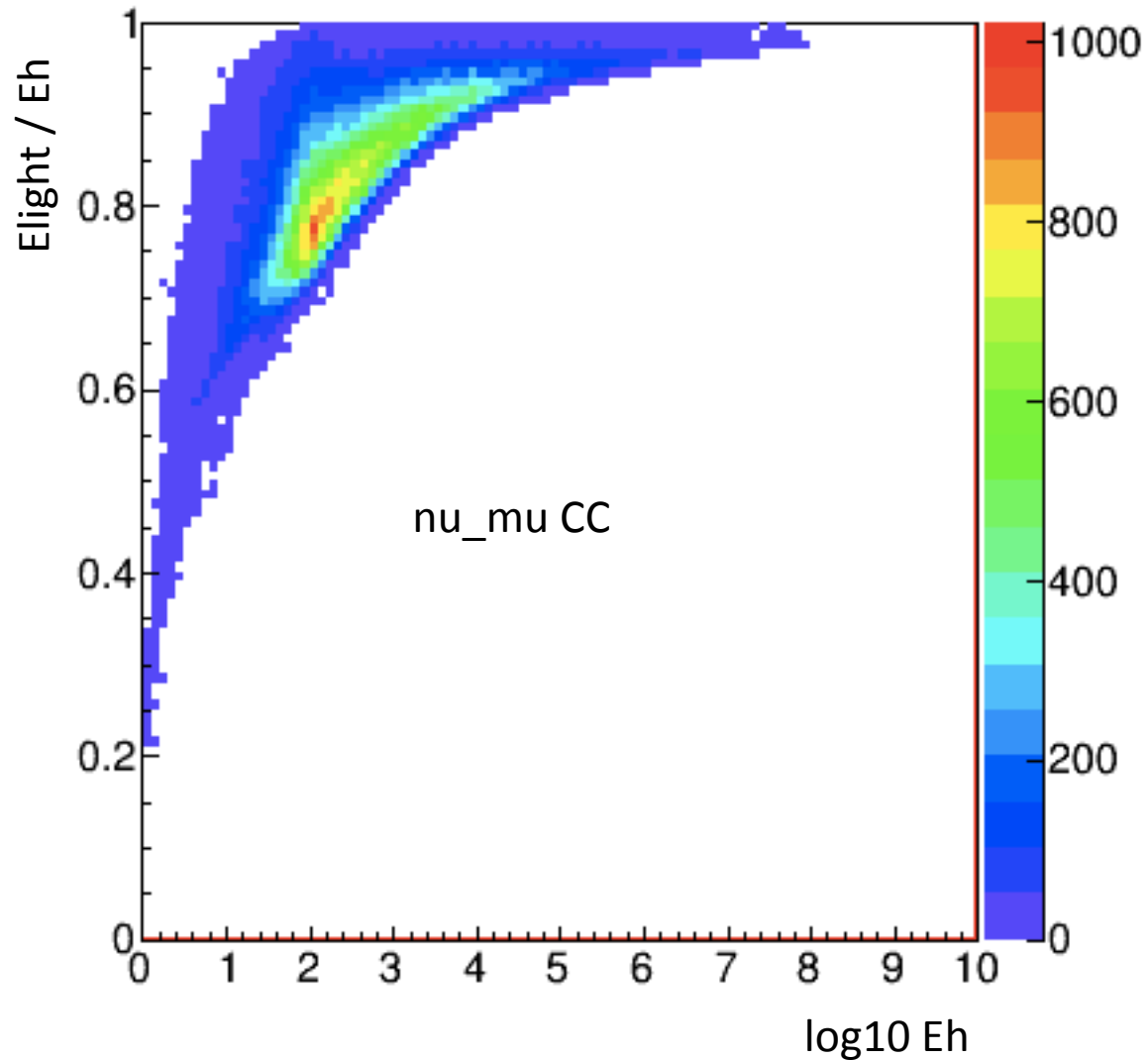
Light from hadronic showers



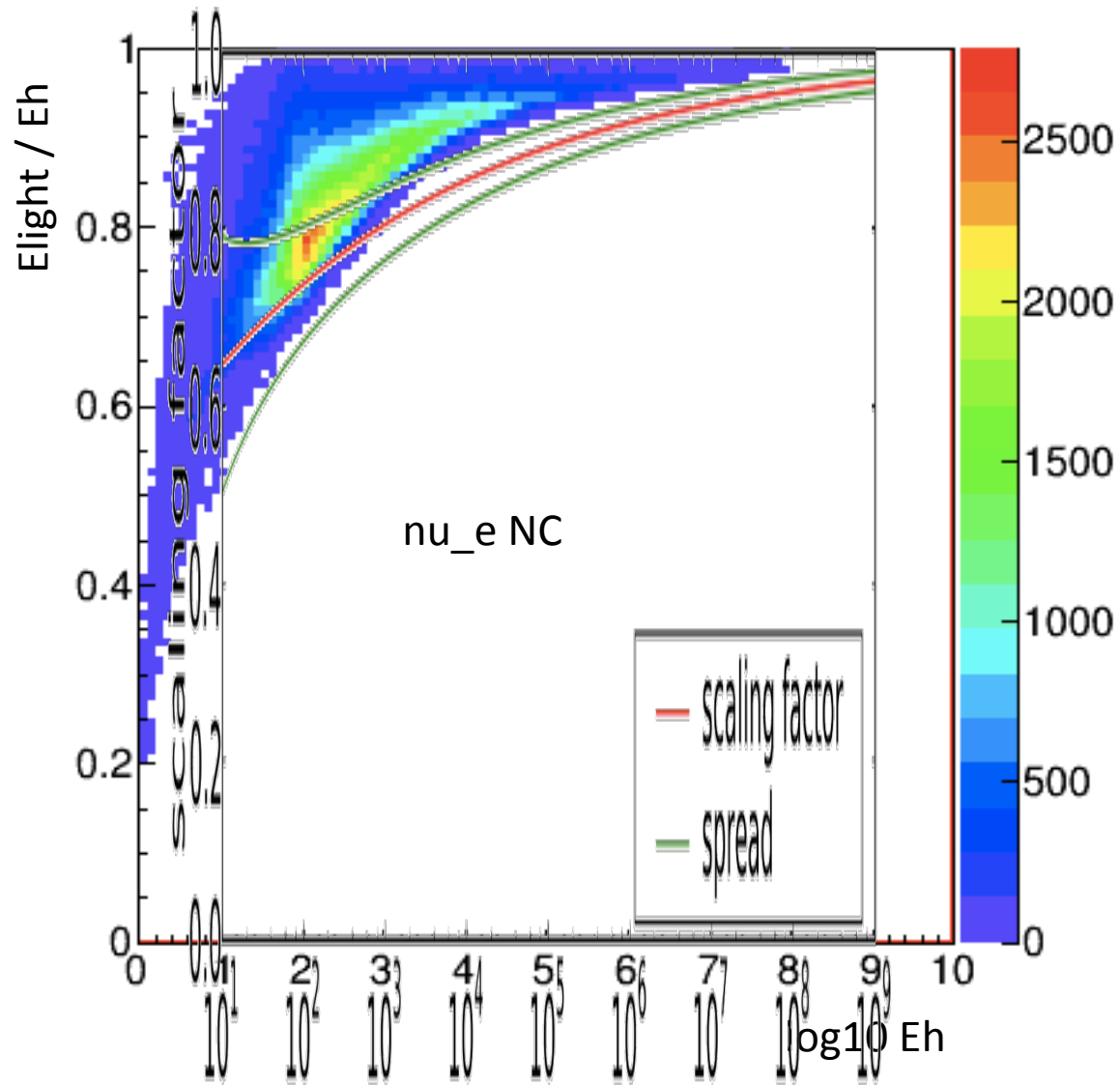
Light from hadronic showers



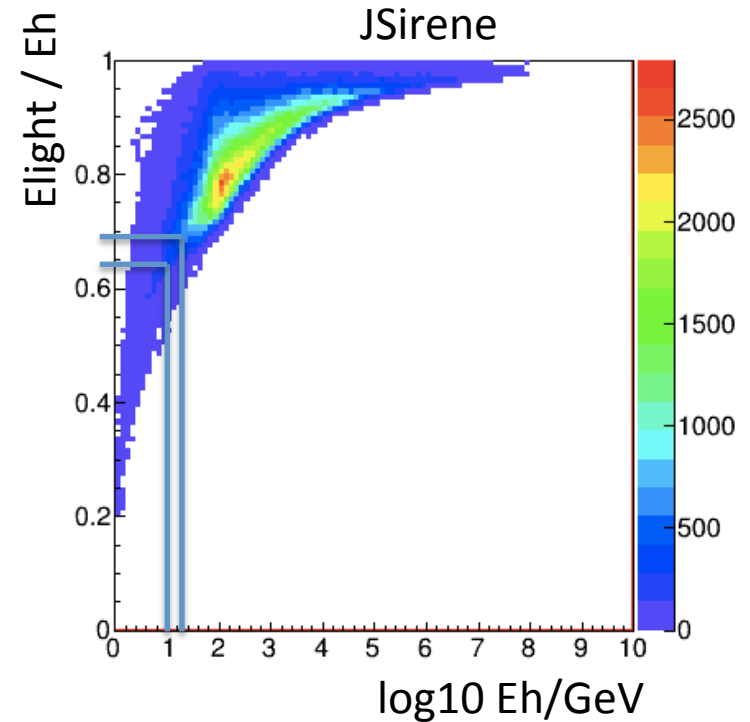
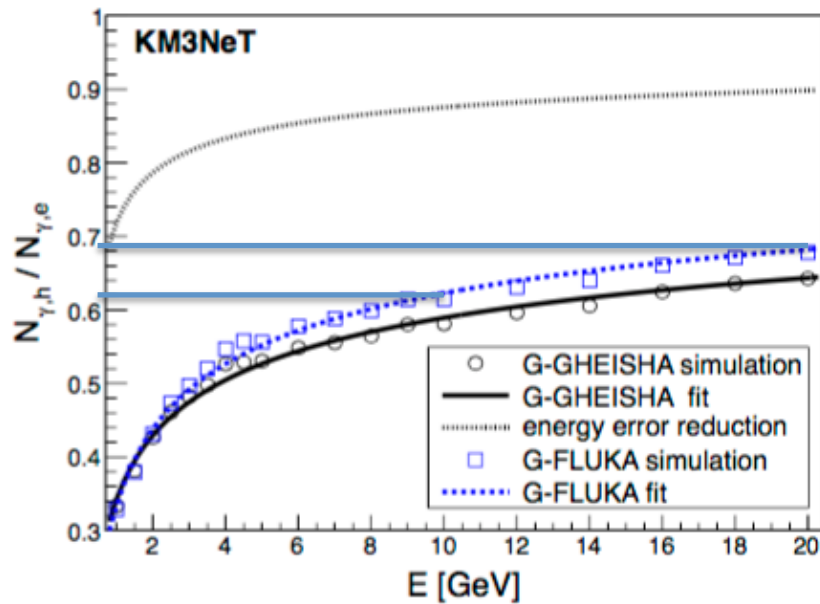
Light from hadronic showers



Light from hadronic showers



Light from hadronic showers



Likelihood Ingredients

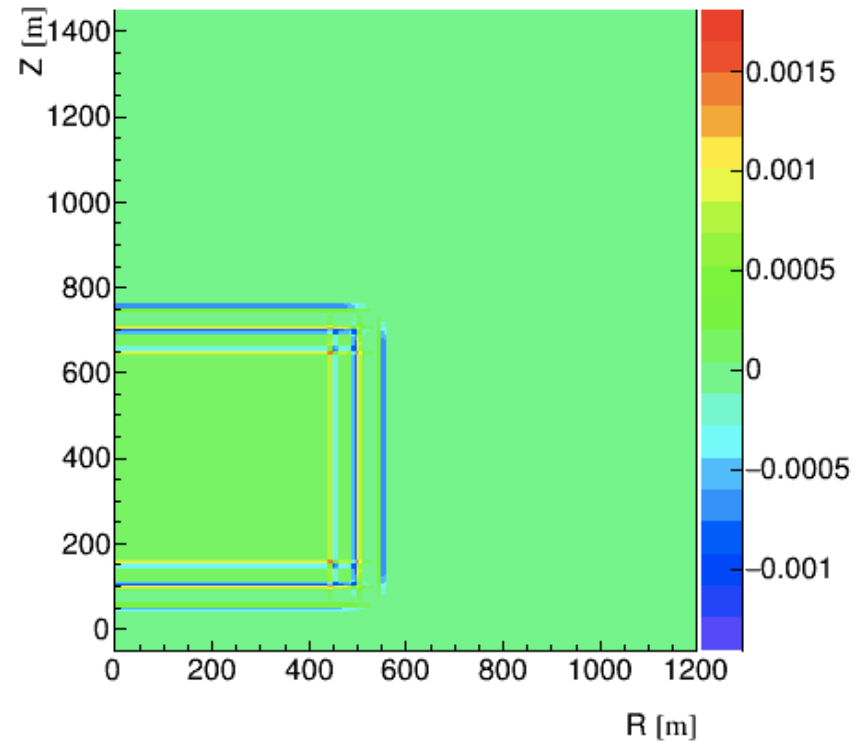
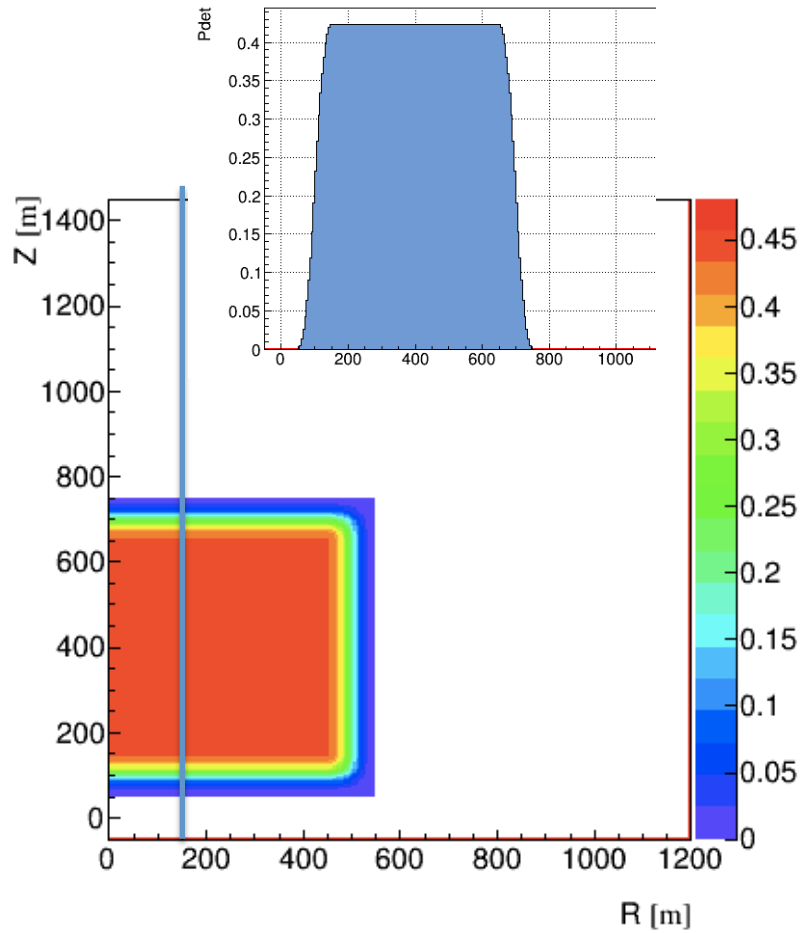
$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

$\mu(x_{\text{true}} | H)$ Number of expected background or signal events in our detector (can)

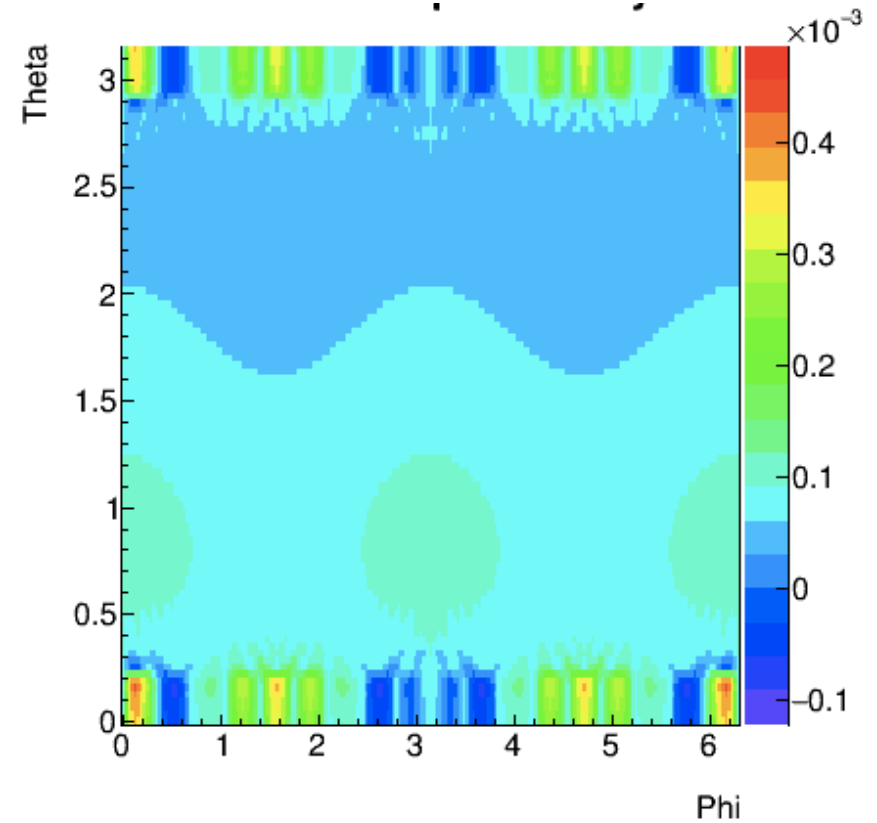
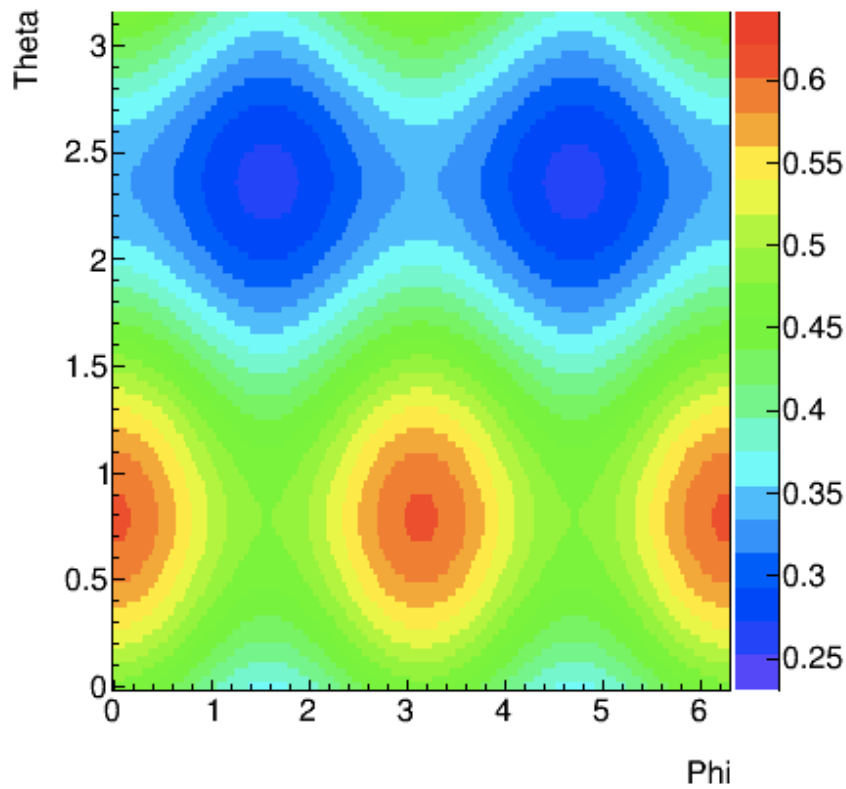
$P^{\text{det}}(x_{\text{true}})$ Probability to detect (=trigger) and select event
6-D Interpolation from tabulated values -> fast

$P(\text{ev}_i | x_{\text{true}})$

Detection Efficiency (1)



Detection Efficiency (2)

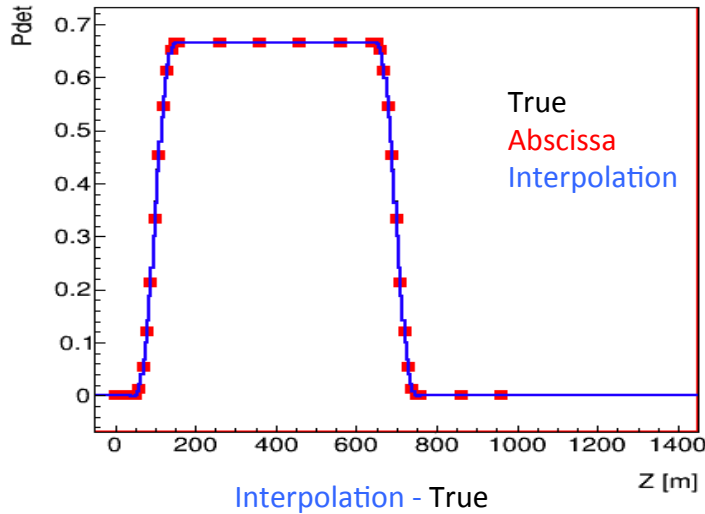


What is Pdet?

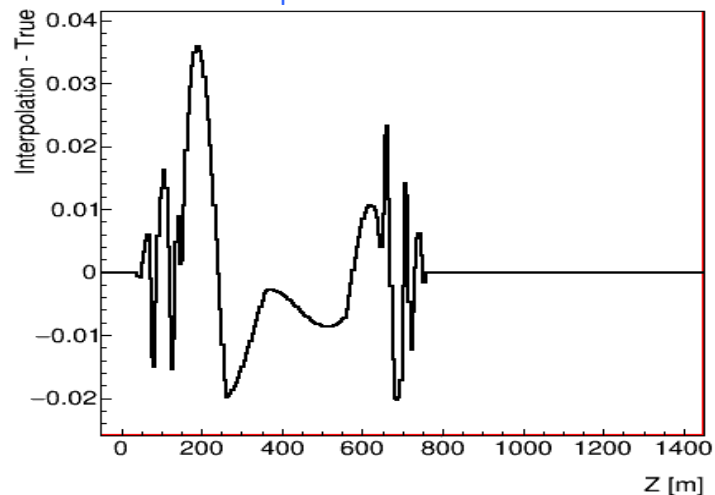
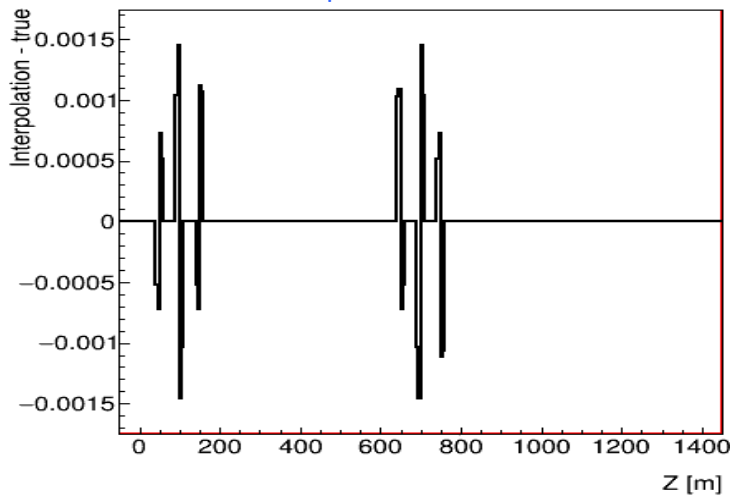
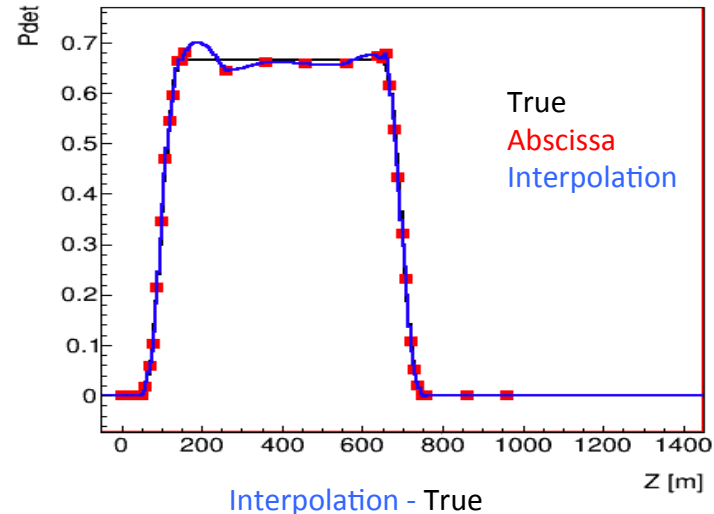
- Probability that an event:
 - Causes hits in detector: Jsirene
 - Leads to a trigger: JTriggerEfficiency
 - Is selected (reject atm. Muons): ??
- Get $P_{\text{det}}(x_{\text{true}})$ by running MC events

Statistical Fluctuations

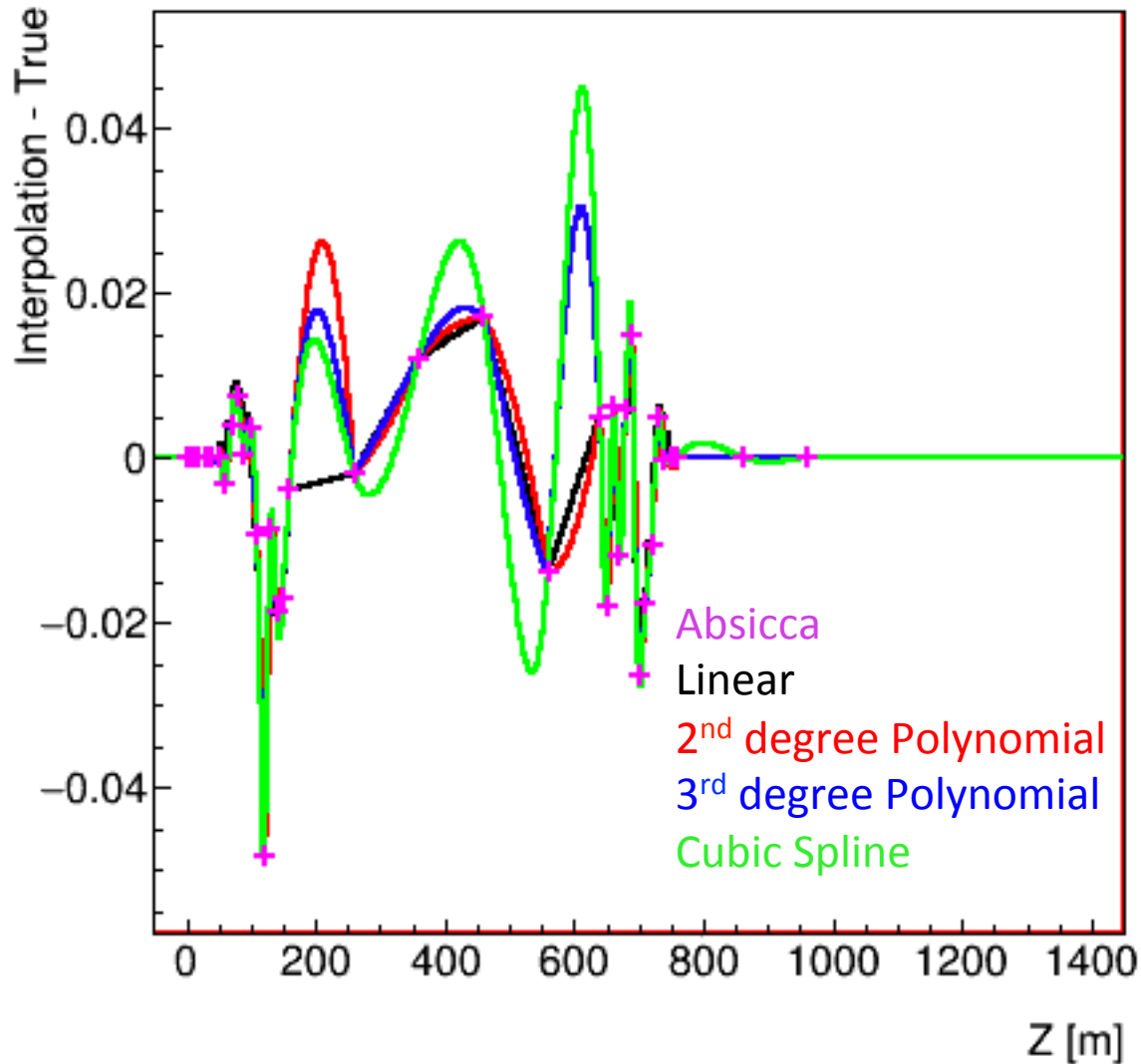
Ideal case: Infinite statistics



Realistic case: 1000 simulated events



Different Interpolation Techniques

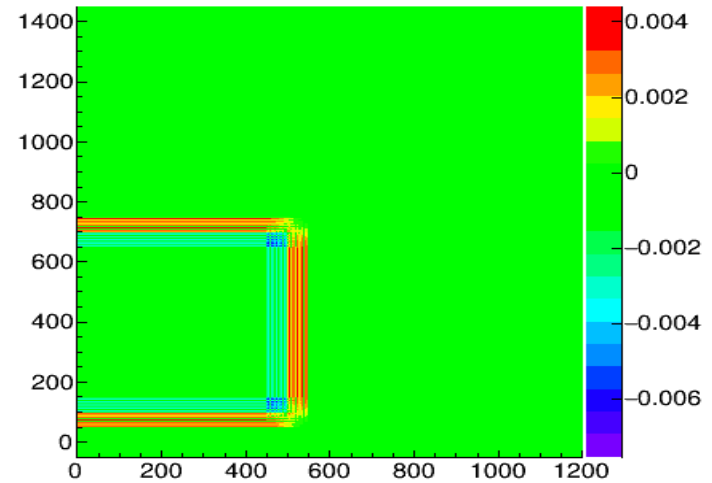
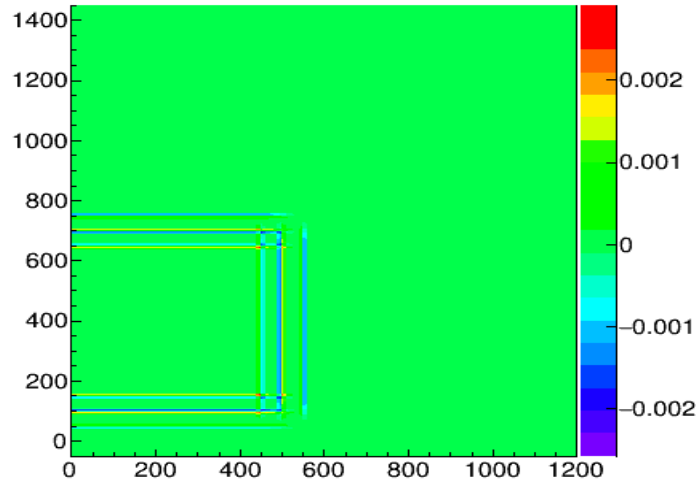


Polynomial vs linear fit

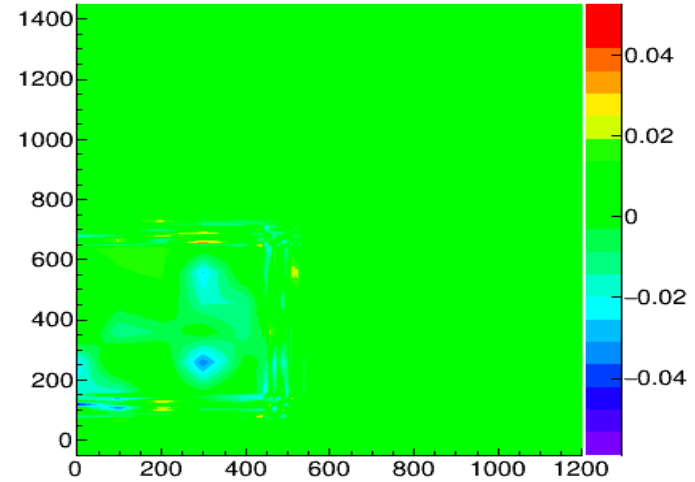
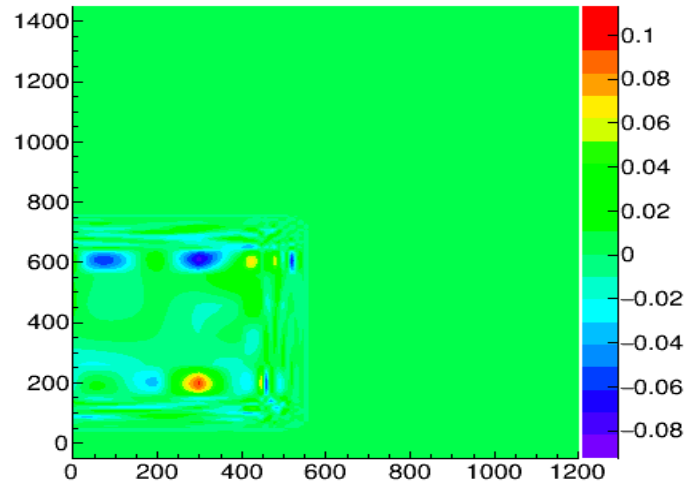
3rd degree polynomial

Linear fit

No stat. fluctuations



With stat. fluctuations



Time Consumption

```
Scanning over 72000 Positions * 98 Directions * 1 Energy-bins = 7056000 points... Done in  
624169.543 ms elapsed  
623814.165 ms user  
    12.998 ms system  
99%CPU
```

3rd degree polynomial interpolation of 7 million points in 10 minutes

```
Scanning over 72000 Positions * 98 Directions * 1 Energy-bins = 7056000 points... Done in  
  
16068.632 ms elapsed  
16057.558 ms user  
    4.999 ms system  
99%CPU
```

Linear interpolation of 7 million points in 16 seconds

Likelihood Ingredients

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

$\mu(x_{\text{true}} | H)$ Number of expected background or signal events in our detector (can)

$P^{\text{det}}(x_{\text{true}})$ Probability to detect (=trigger) and select event

$P(\text{ev}_i | x_{\text{true}})$ Reconstruction, loop over PMTs. $\text{Phit} * \text{Ptime} \rightarrow$ to do

Conclusions

- New method seems promising
- Most ingredients in place
- ‘Reconstruction’ part to be done

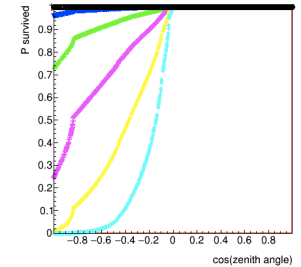
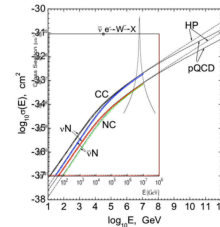
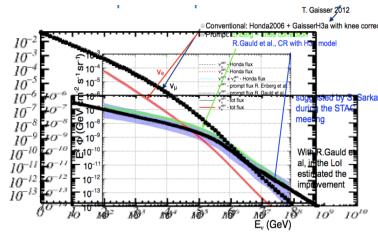
Recap: Likelihood Ingredients

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$



$\mu(x_{\text{true}} | H)$

Number of expected background or signal events in our detector (can)



$P^{\text{det}}(x_{\text{true}})$

Probability to detect (=trigger) and select event



$P(\text{ev}_i | x_{\text{true}})$

Reconstruction, loop over PMTs.

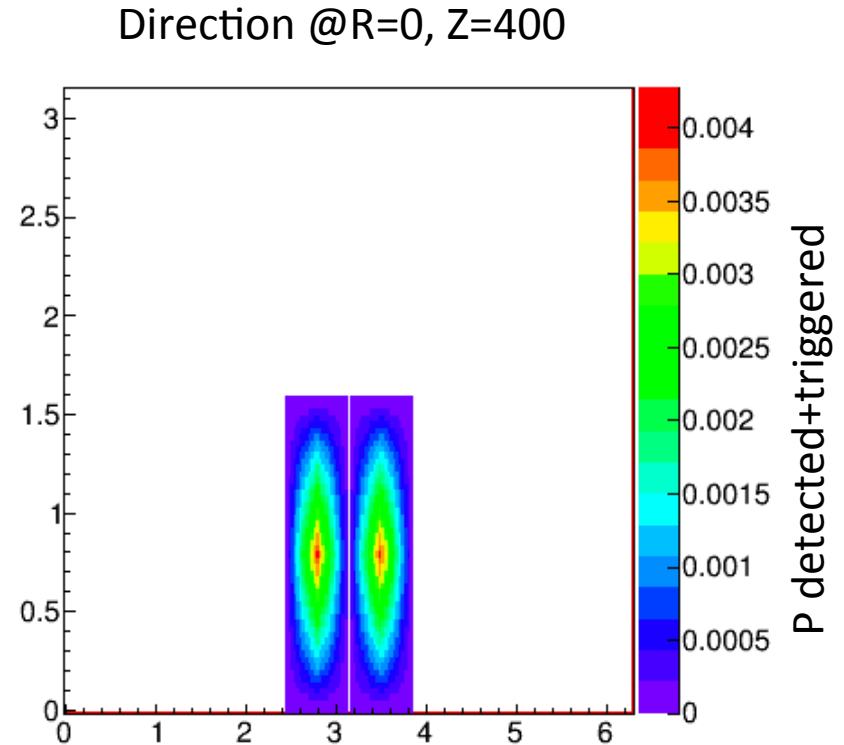
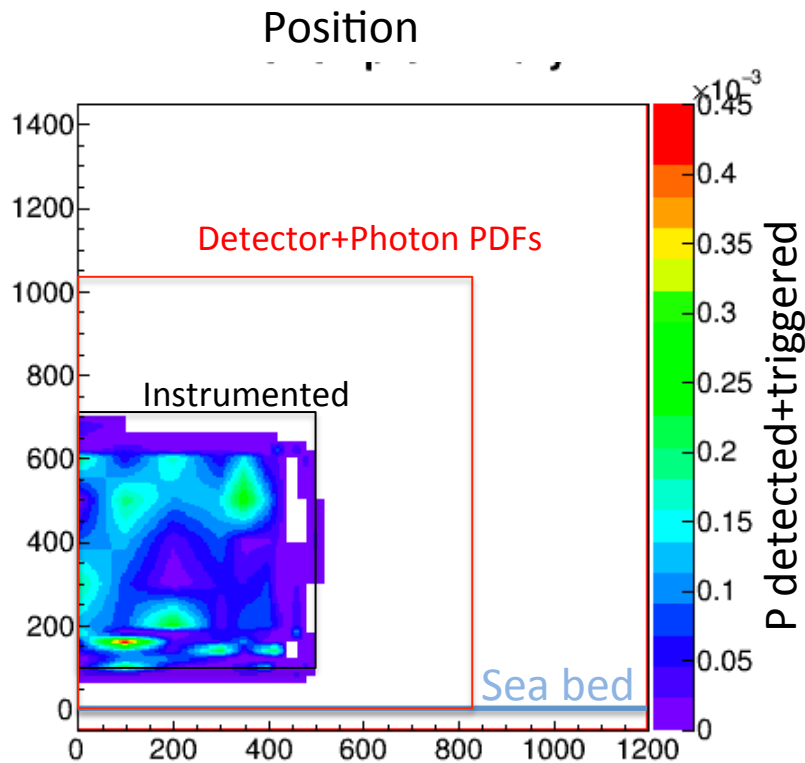
Detection Efficiency

- For each neutrino energy, Bjorken- γ , position, direction (6 parameters), DO:
- (Very fast) Monte Carlo generator:
 - Secondary particles
 - Photon propagation (JSirene)
 - Trigger
- Count fraction of trig. ev.
- Store in 6D interpolatable PDF table



Detection Efficiency @ 10^2 GeV

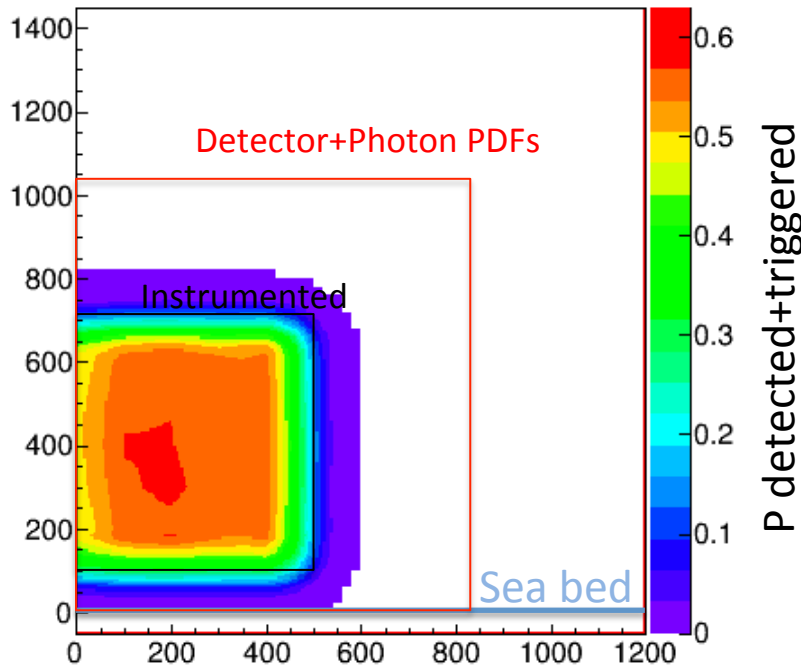
NC electron-neutrino (only single hadronic shower)



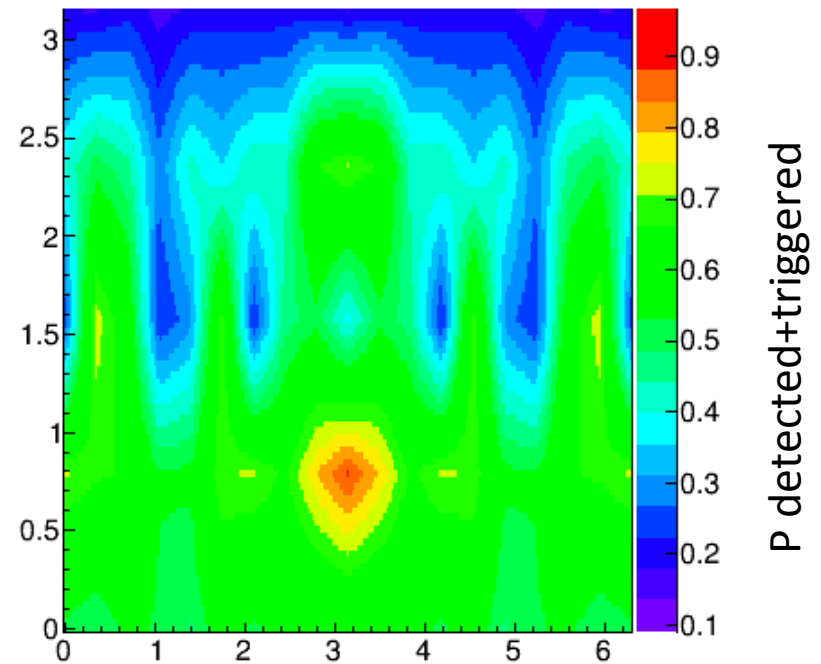
Detection Efficiency @ 10^3 GeV

NC electron-neutrino (only single hadronic shower)

Position



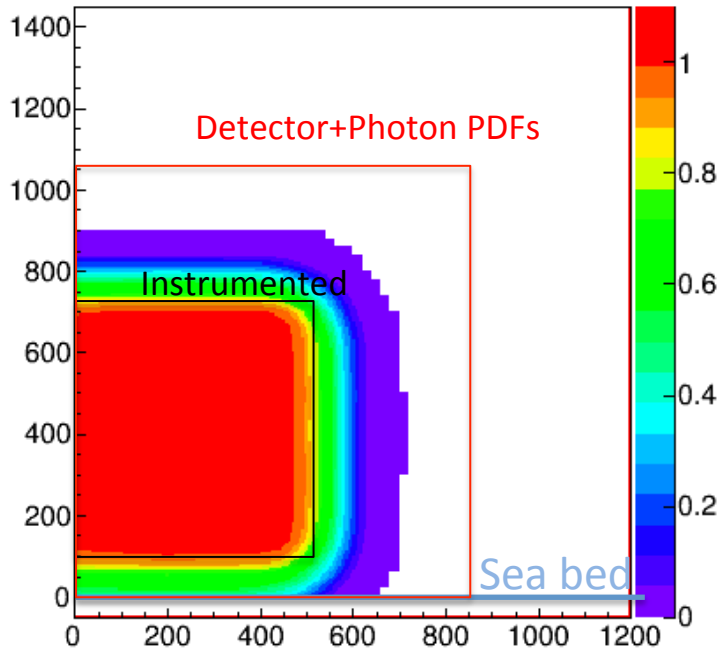
Direction @R=0, Z=400



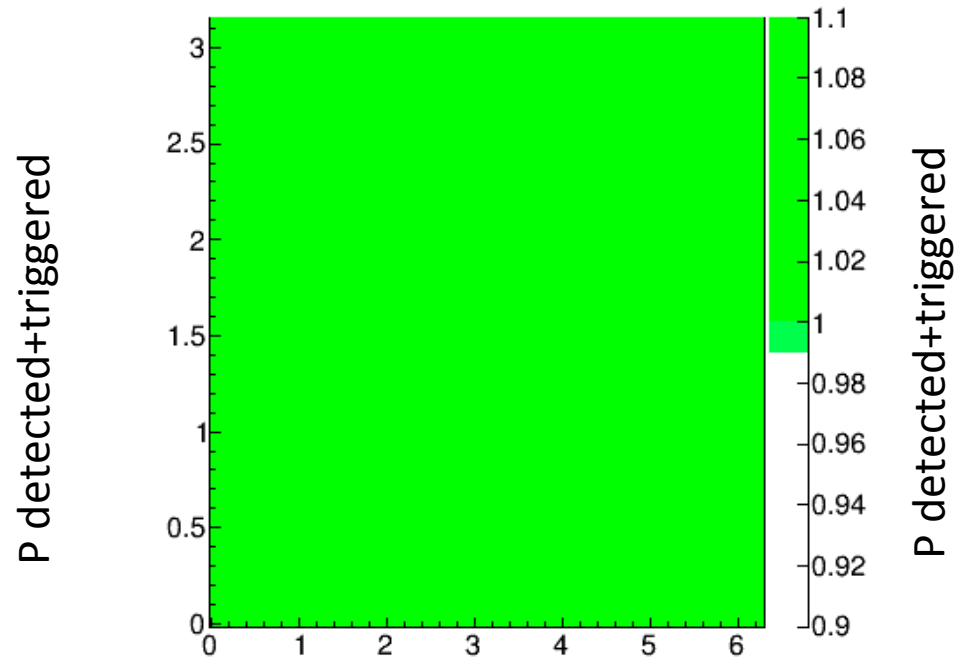
Detection Efficiency @ 10^4 GeV

NC electron-neutrino (only single hadronic shower)

Position



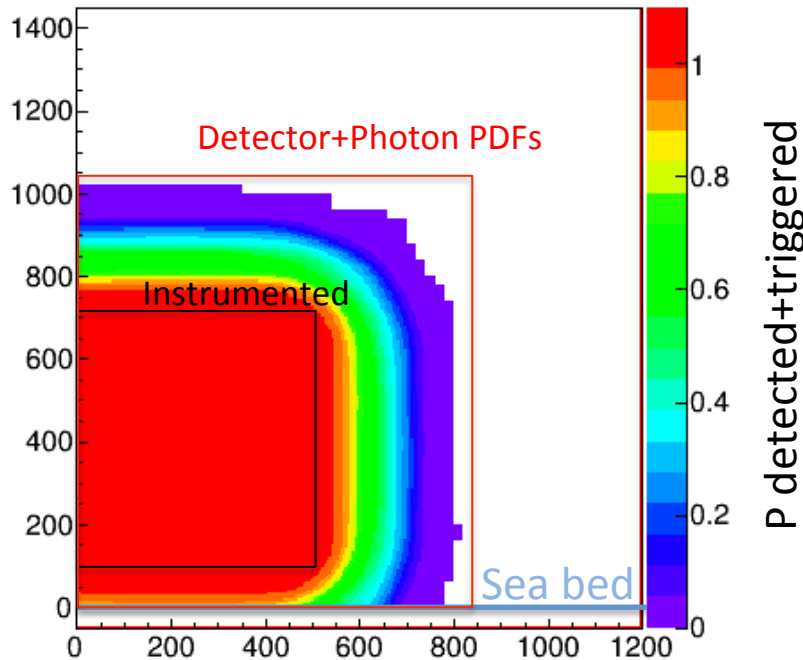
Direction @R=0, Z=400



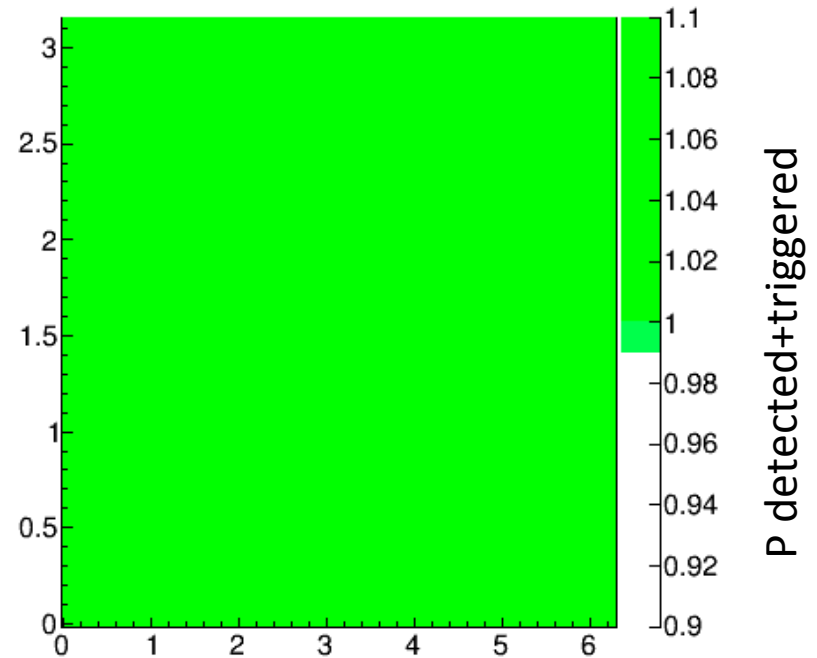
Detection Efficiency @ 10^5 GeV

NC electron-neutrino (only single hadronic shower)

Position



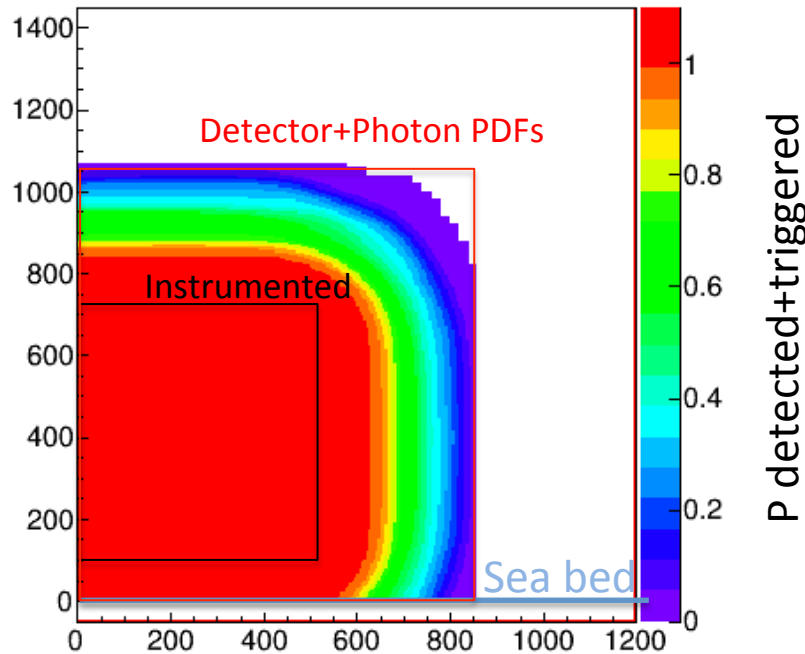
Direction @R=0, Z=400



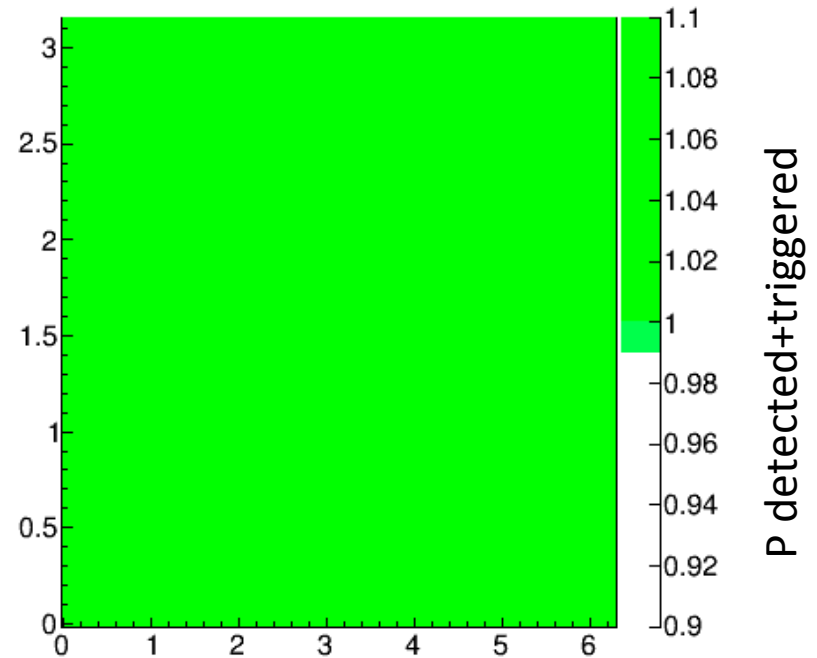
Detection Efficiency @ 10^6 GeV

NC electron-neutrino (only single hadronic shower)

Position



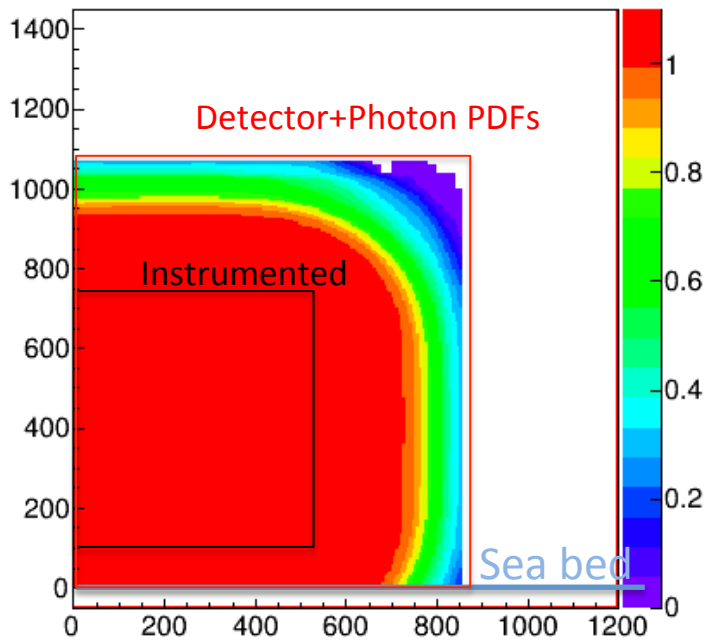
Direction @R=0, Z=400



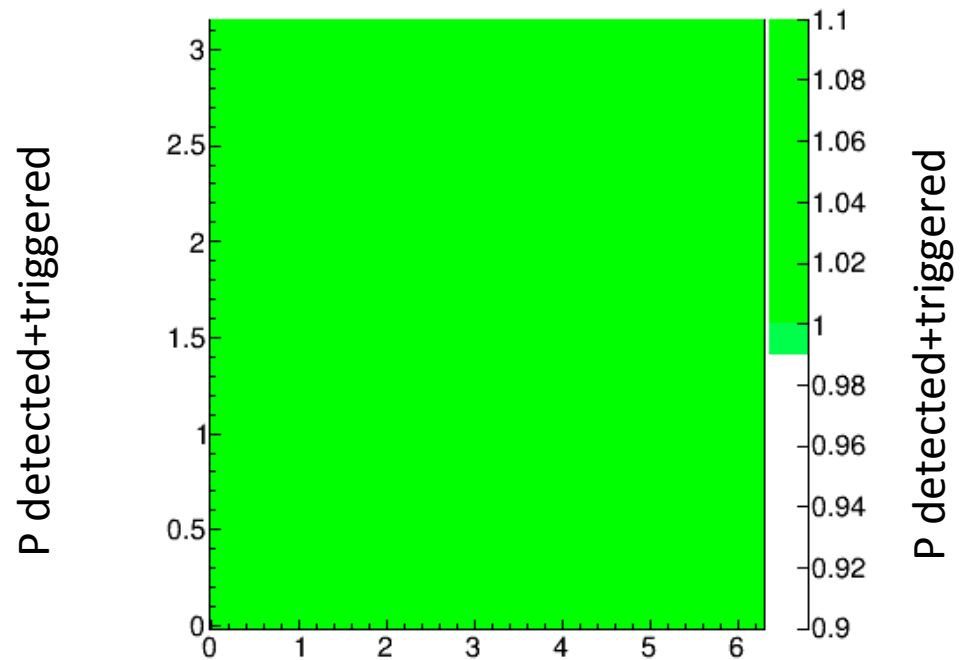
Detection Efficiency @ 10^7 GeV

NC electron-neutrino (only single hadronic shower)

Position



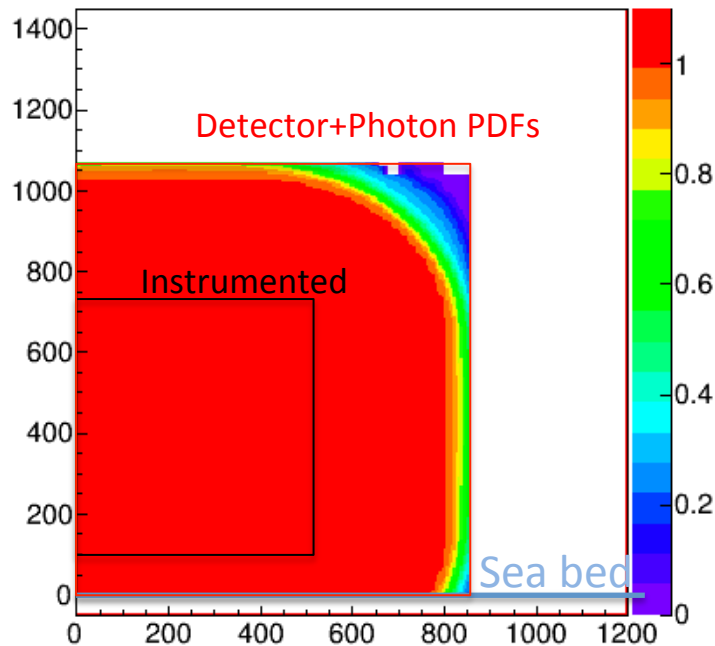
Direction @R=0, Z=400



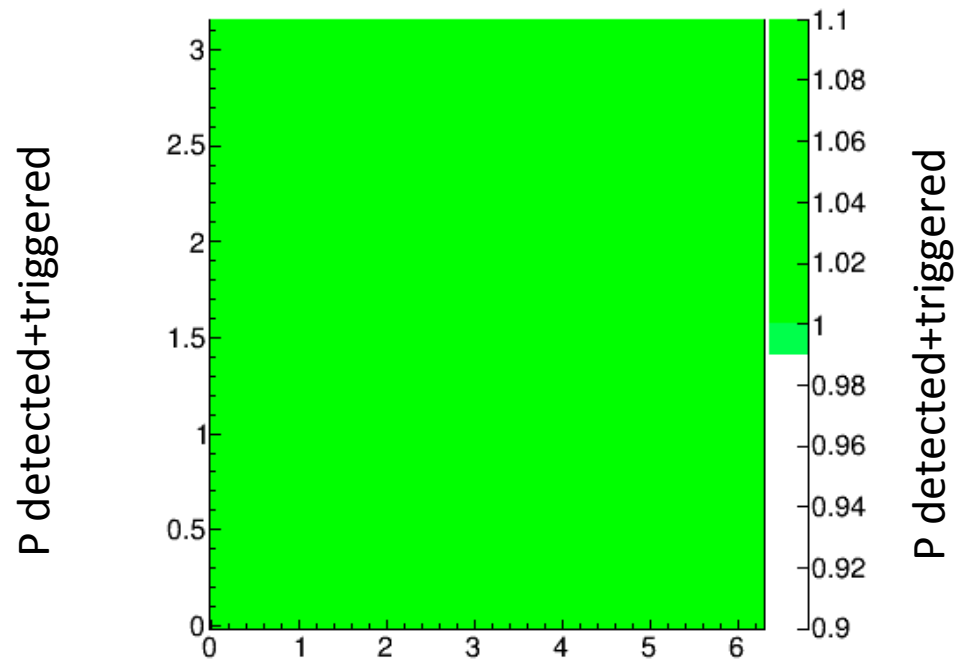
Detection Efficiency @ 10^8 GeV

NC electron-neutrino (only single hadronic shower)

Position



Direction @R=0, Z=400



Likelihood Ingredients

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$



$\mu(x_{\text{true}} | H)$ Number of expected background or signal events in our detector (can)



$P^{\text{det}}(x_{\text{true}})$ Probability to detect (=trigger) and select event



$P(\text{ev}_i | x_{\text{true}})$ Probability to obtain measured event ev_i
given a certain neutrino hypothesis x_{true}

Event Probability D.F.

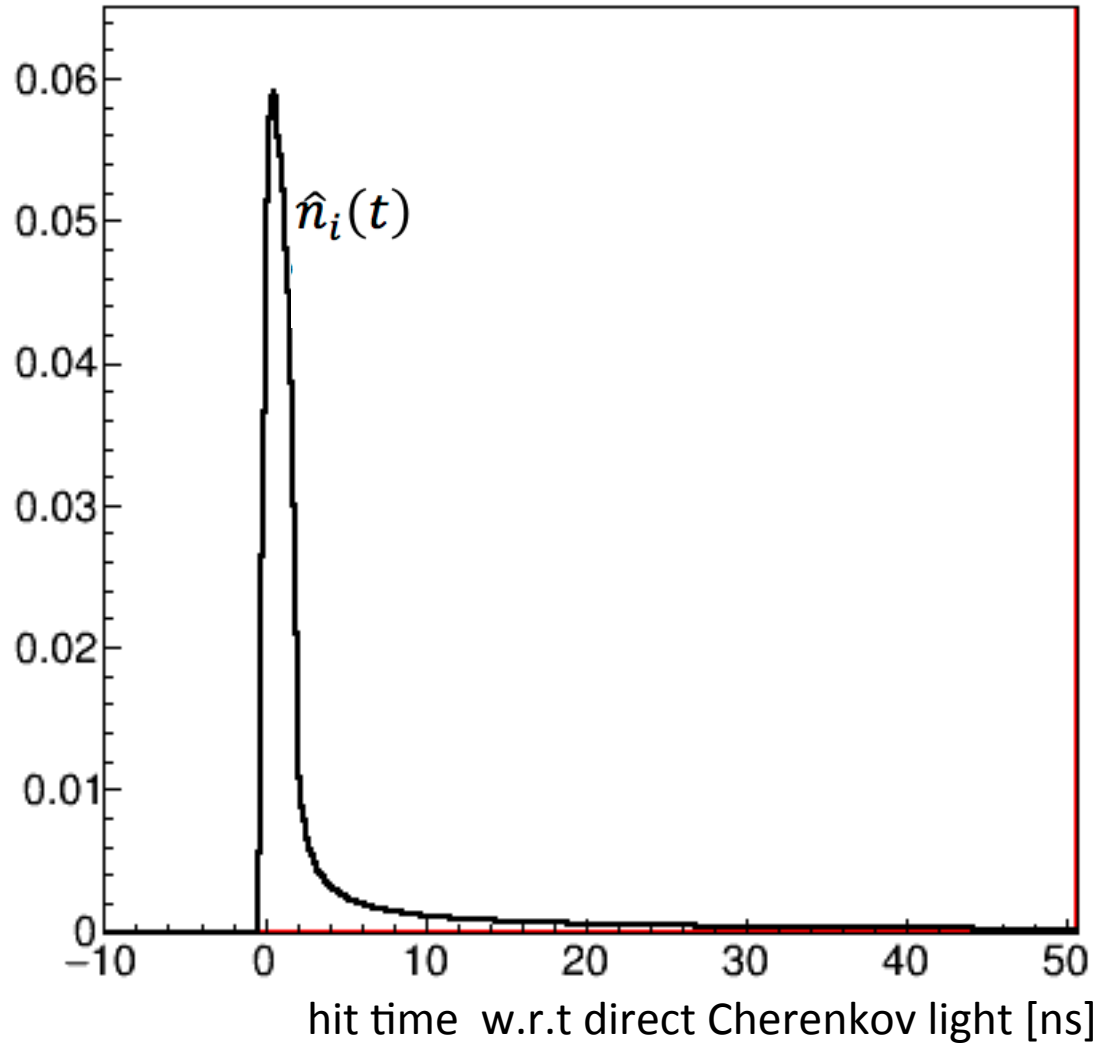
$$P(ev | x) = \prod_{\text{hit PMTs}} [P_i^{\text{hit}} \cdot P_i^{t \text{ 1st}}] \cdot \prod_{\text{non hit PMTs}} [1 - P_i^{\text{hit}}]$$

$$P_i^{\text{hit}} = 1 - \exp \left(- \int_{-\infty}^{\infty} \hat{n}_i(t) dt \right)$$

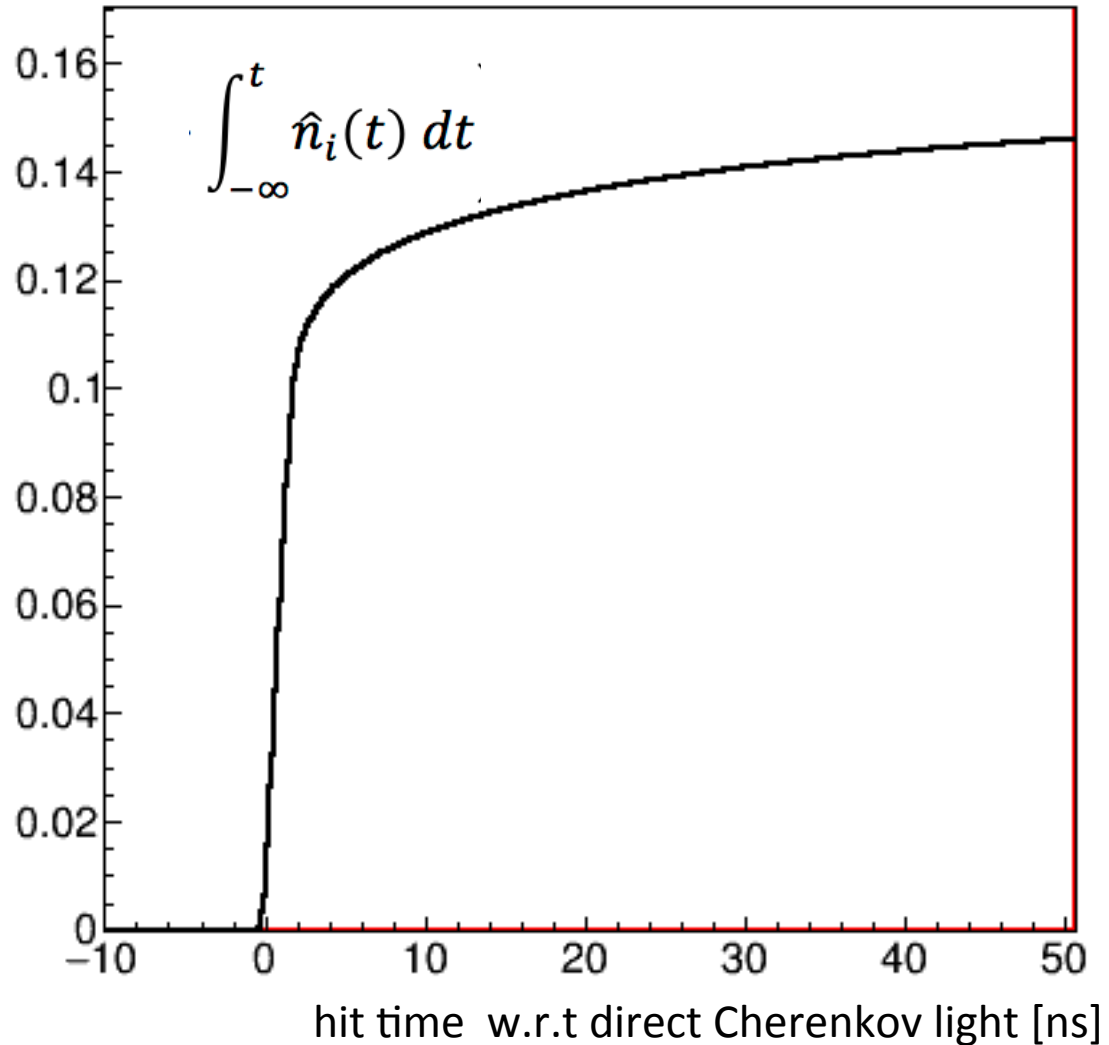
Expected number of photons from 40K and shower/track on PMT i at time t

$$P_i^{t \text{ 1st}} \cdot P_i^{\text{hit}} = \underbrace{\exp \left(- \int_{-\infty}^t \hat{n}_i(t) dt \right)}_{\text{P not hit before t}} \cdot \underbrace{(1 - \exp(-\hat{n}_i(t)))}_{\text{P hit at t}}$$

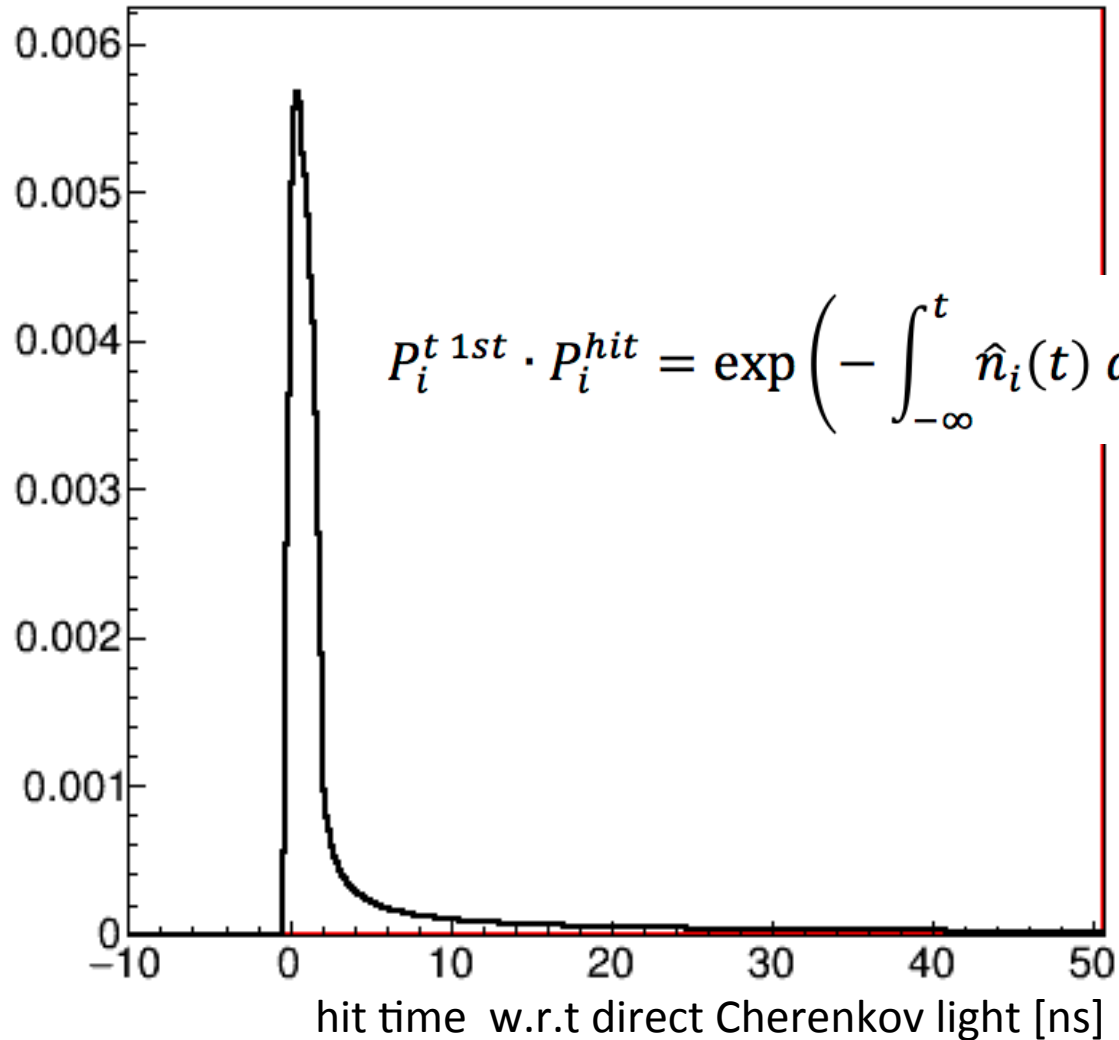
Hit Time PDF



Hit Time PDF



Hit Time PDF



Hits in ~~Theory~~ Practice

- Presence of ^{40}K background hits

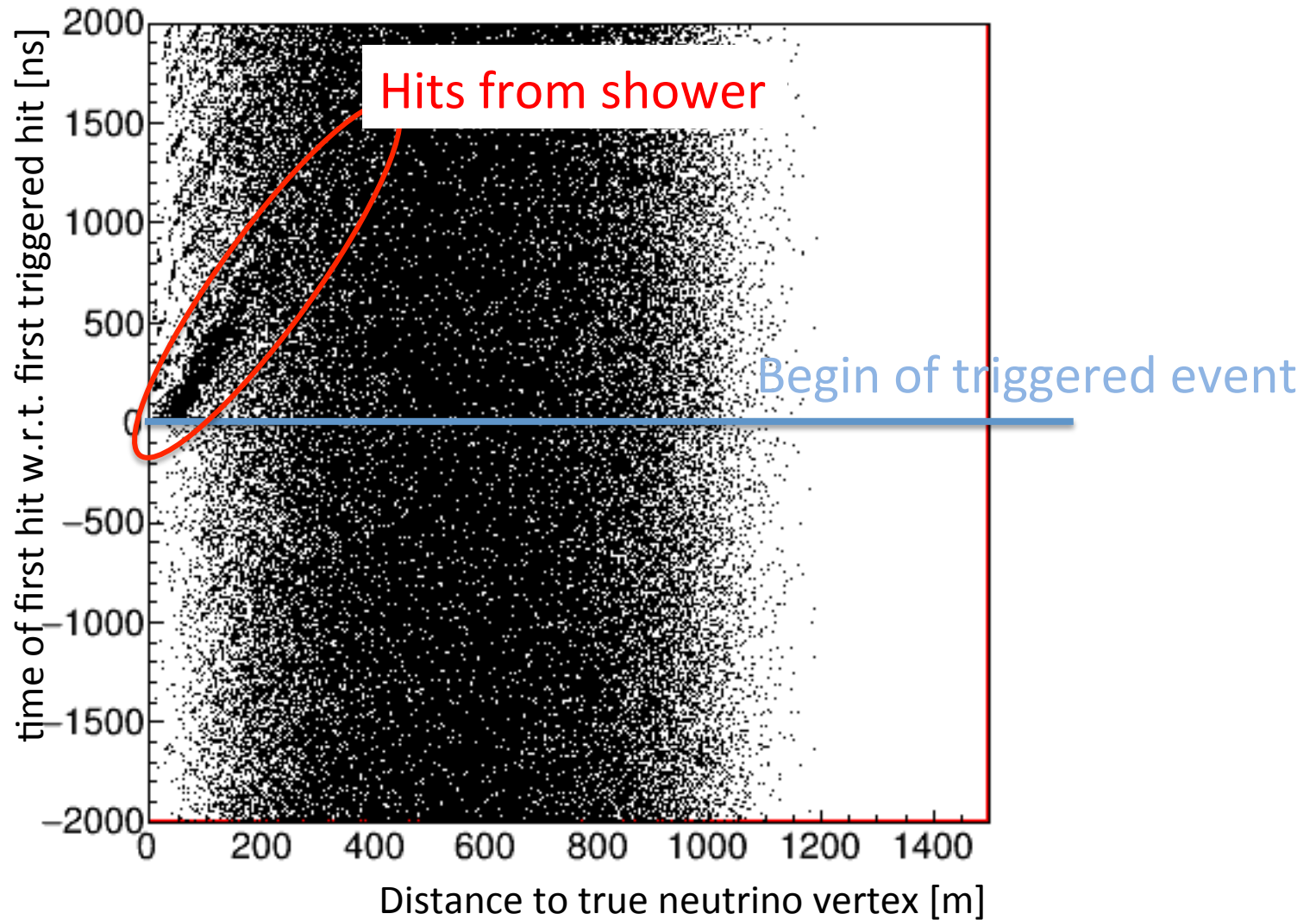
$$P_i^{1st} \cdot P_i^{hit} = \exp\left(-\int_{-\infty}^t \hat{n}_i(t) dt\right) \cdot (1 - \exp(-\hat{n}_i(t)))$$

- If all hit times are selected: signal will be overwhelmed by background
- Solution: only select hits in certain time window

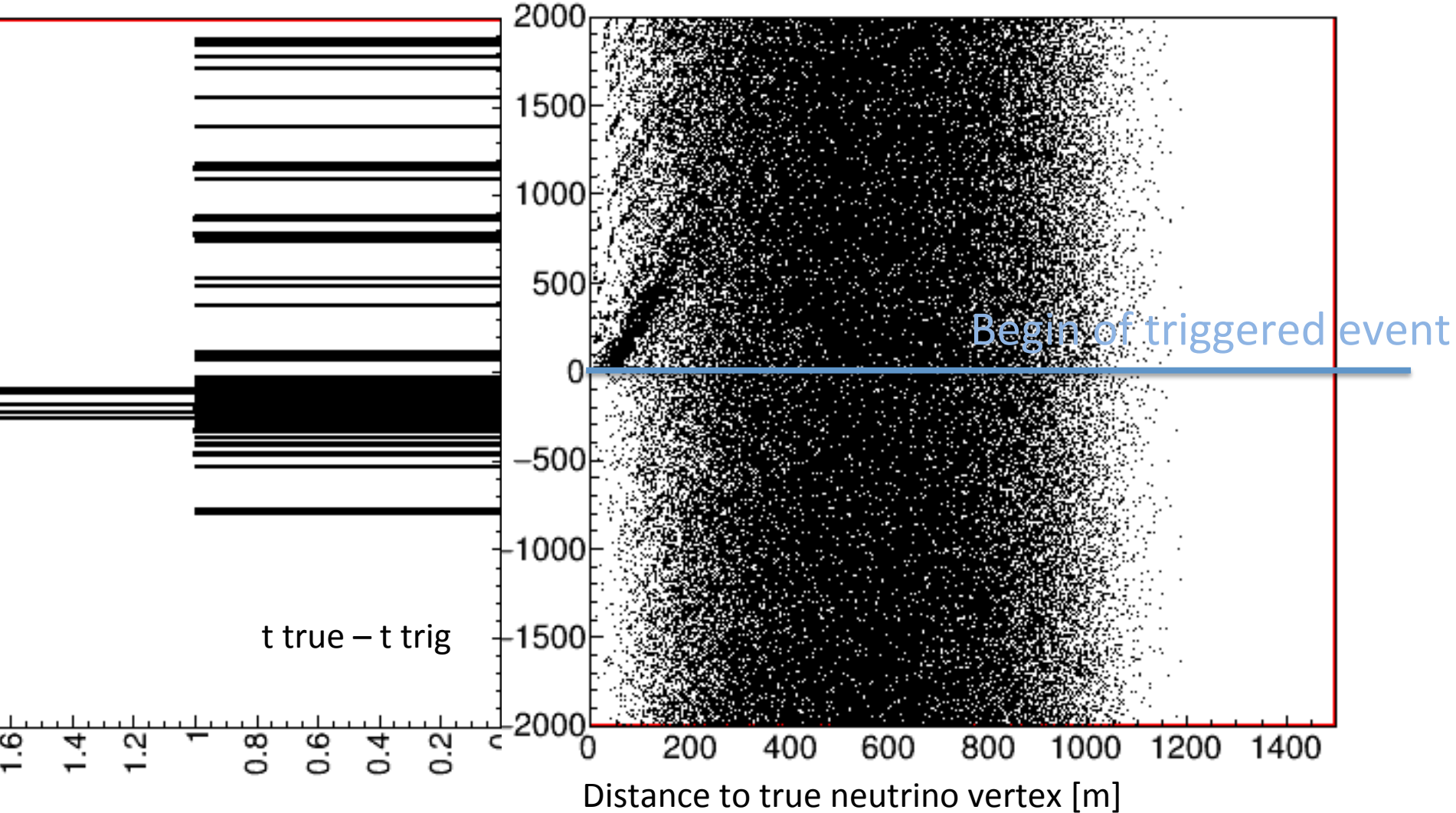
Hit Selection Time Window

- Select Hits around expected hit time from given hypothesis
 - Advantage: Very pure selection
 - Drawback: biased selection
- Solution: Select hits around triggered event

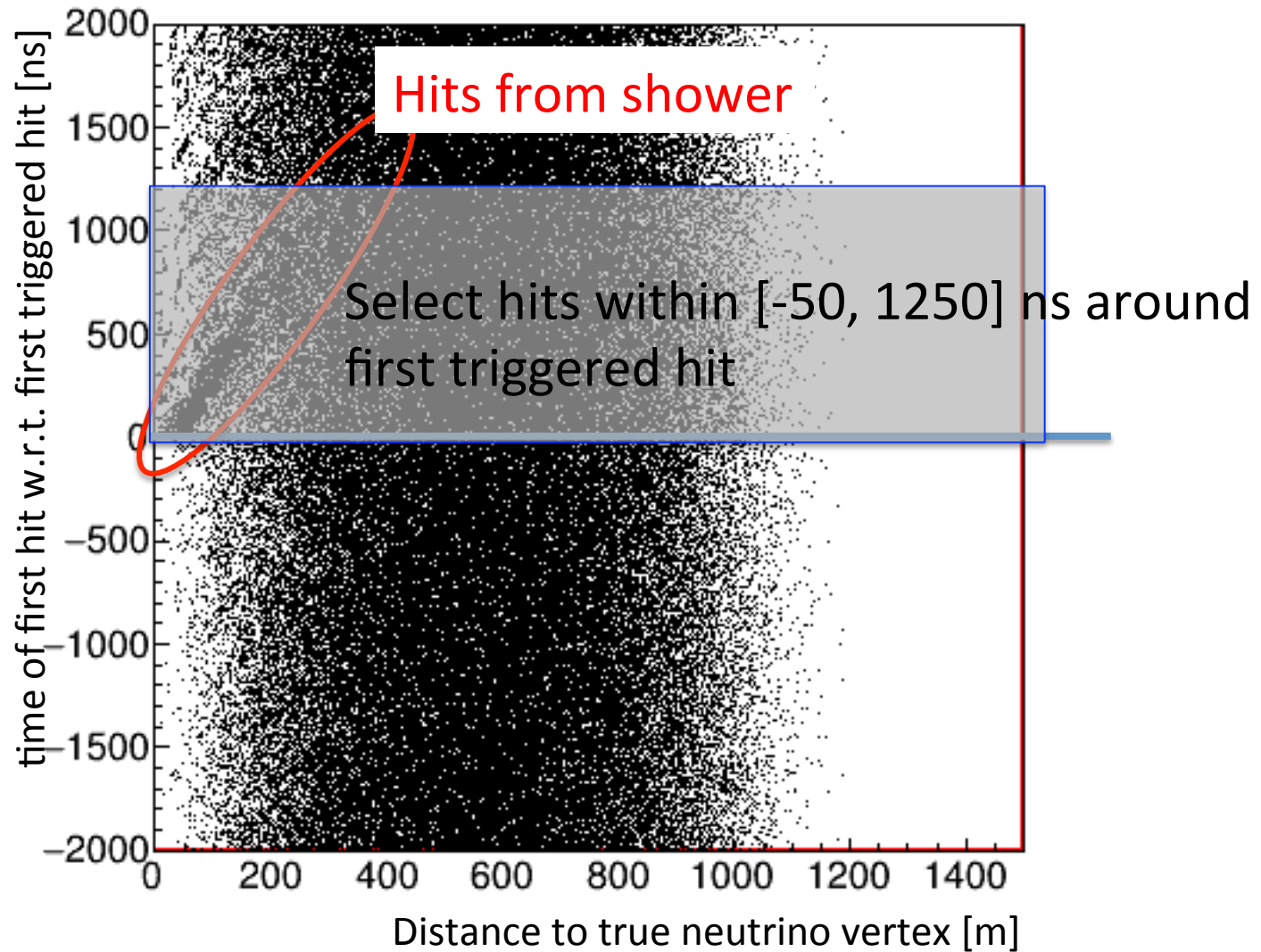
Hit Times



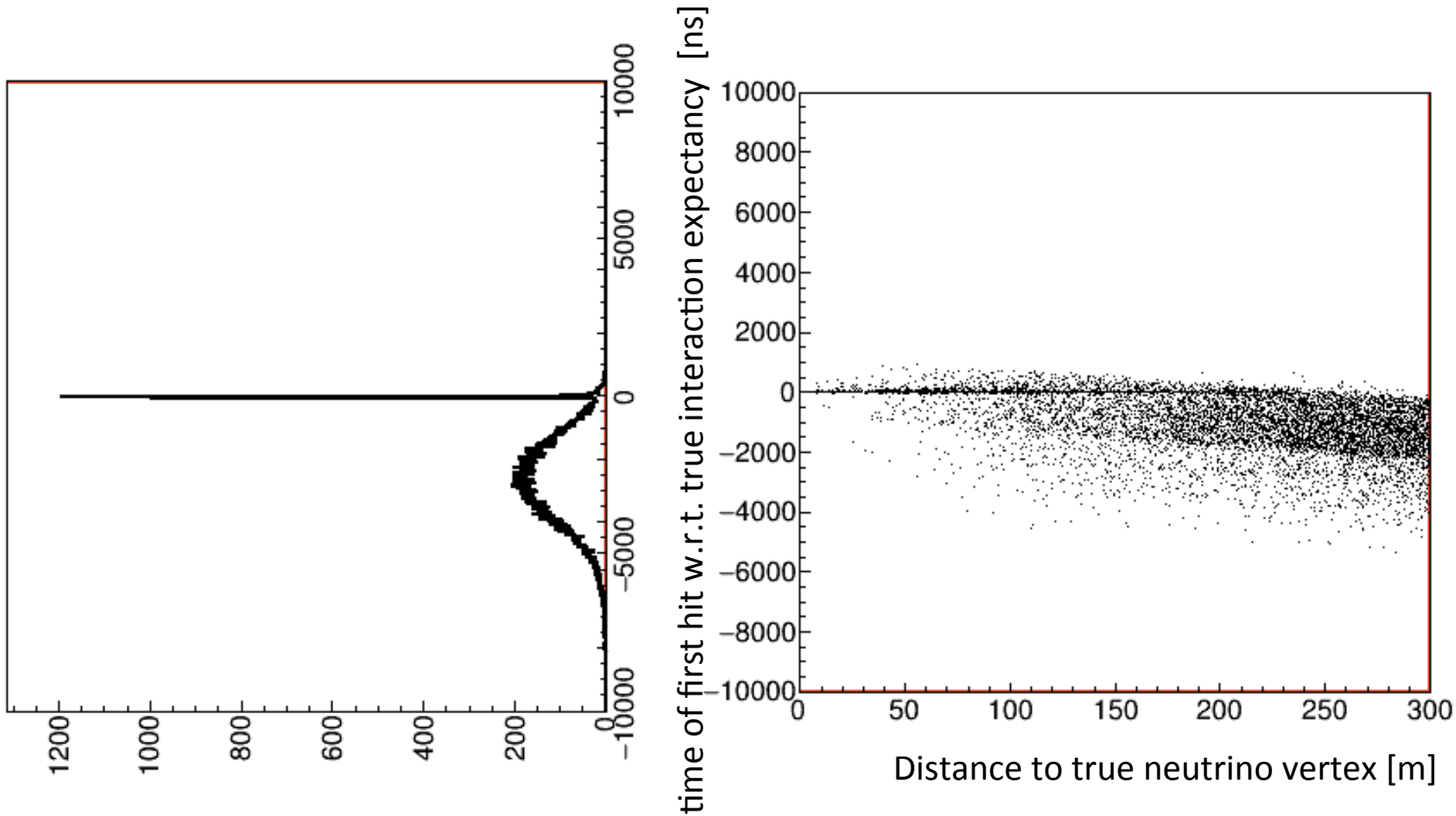
Hit Times



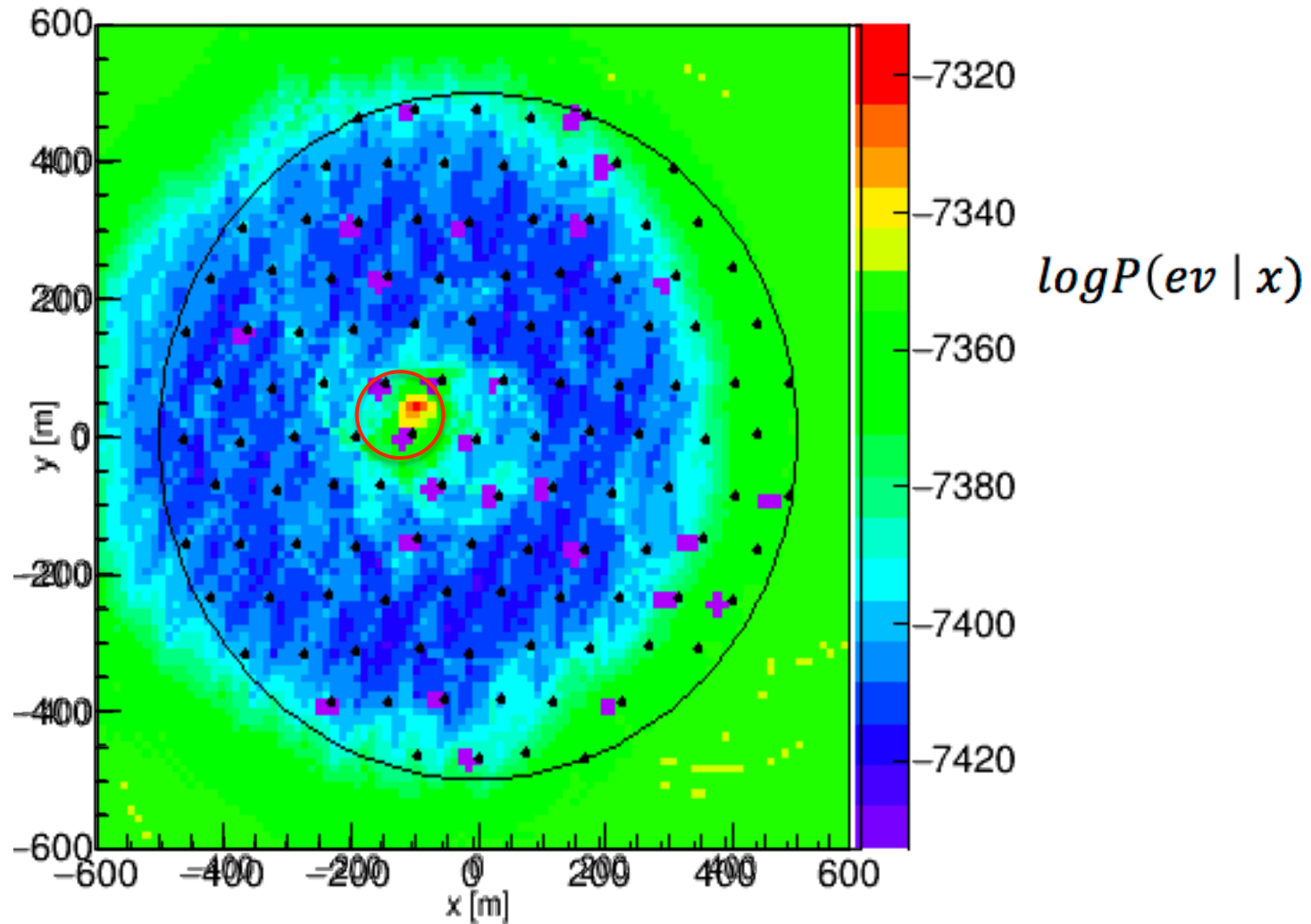
Hit Times



Hit Times w.r.t. direct Cher. light

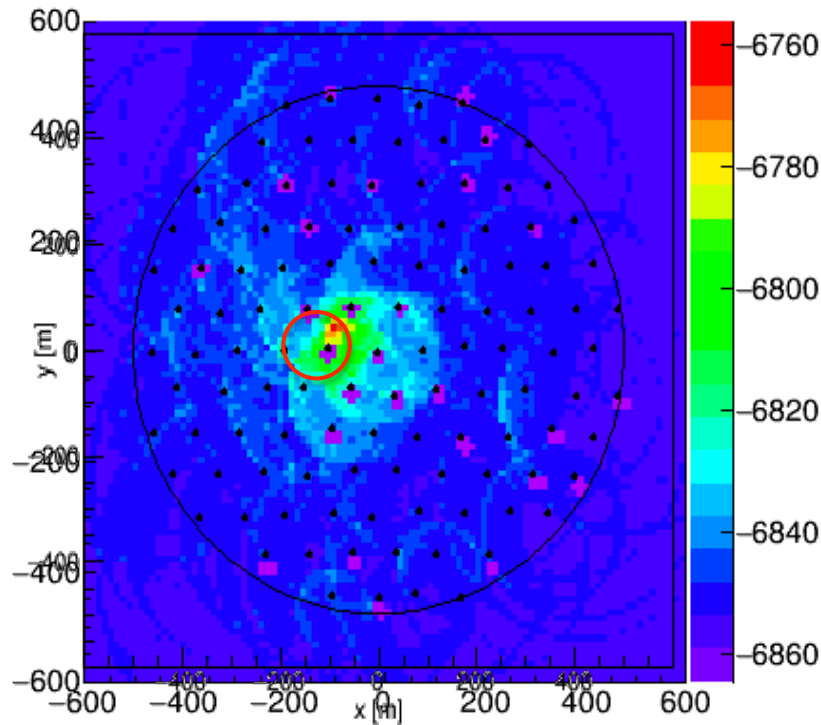


$$P(ev | x) = \prod_{\text{hit PMTs}} [P_i^{\text{hit}} \cdot P_i^{\text{t 1st}}] \cdot \prod_{\text{non hit PMTs}} [1 - P_i^{\text{hit}}]$$

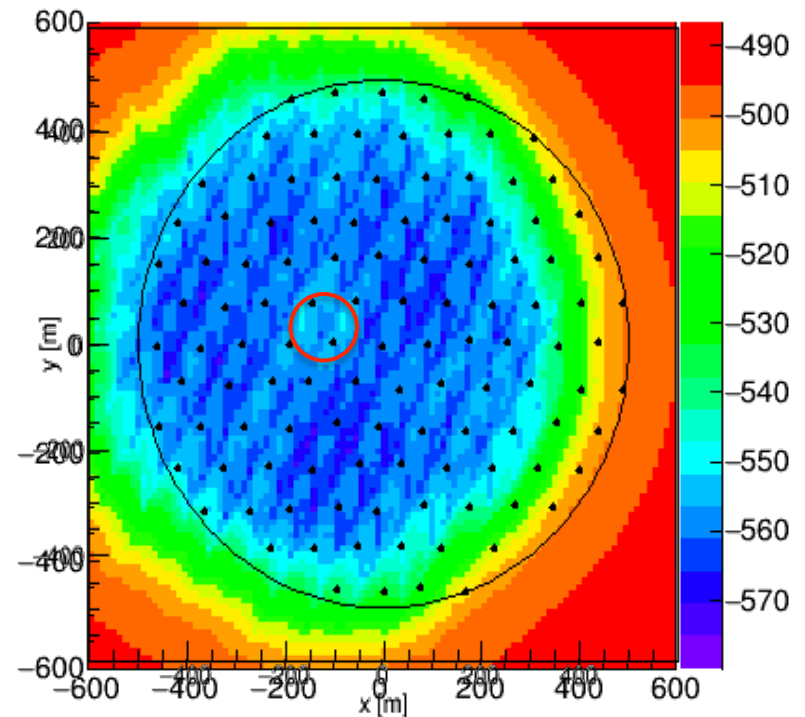


$$\log P(ev | x) = \sum_{\text{hit PMTs}} [\log (P_i^{\text{hit}}) + \log (P_i^{t \text{ 1st}})] + \sum_{\text{non hit PMTs}} [\log (1 - P_i^{\text{hit}})]$$

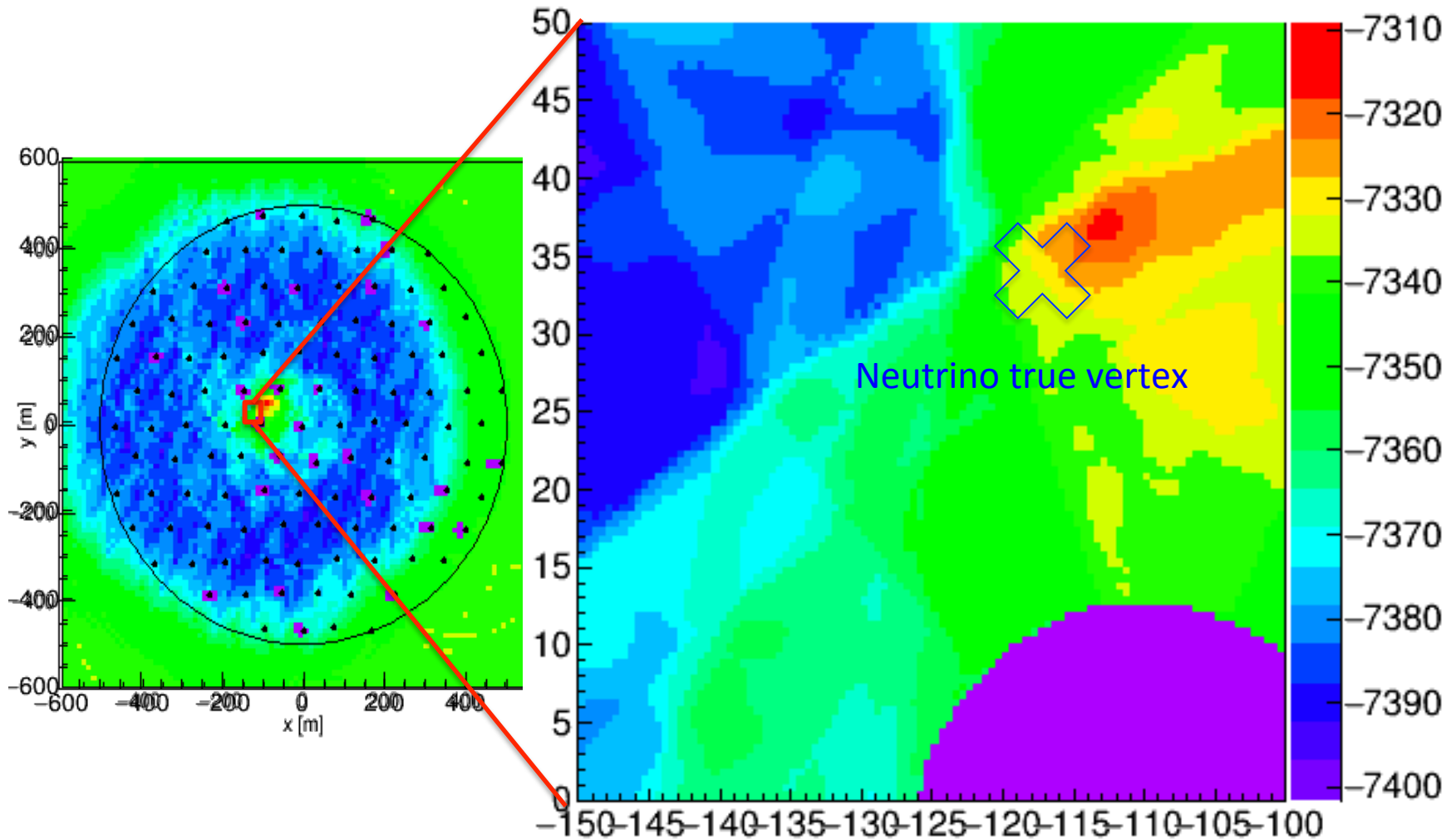
Hit PMTs



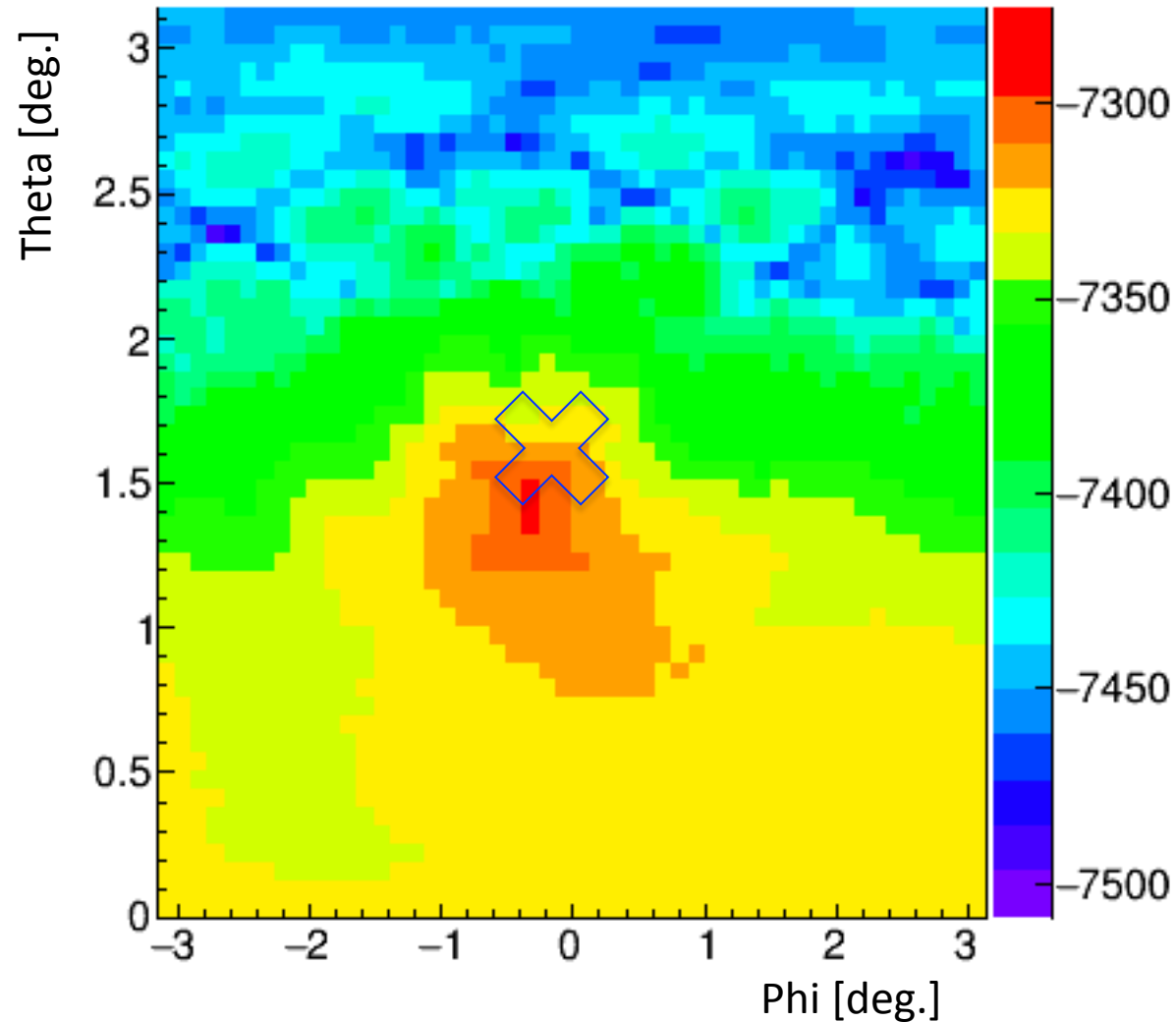
Not hit PMTs



Event Probability: Position



Event Probability: Direction



Likelihood Ingredients

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$



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given a certain neutrino hypothesis x_{true}

How to solve the 8D integral?

$$P(data|H) = \sum_i \left[\log \int P(ev_i | x_{true}) \cdot P^{det}(x_{true}) \cdot \mu(x_{true} | H) dx_{true} \right] - \mu^{tot}(H)$$

- Interaction vertex position (3D)
- Interaction time (1D)
- (Neutrino) Direction (2D)
- Neutrino Energy (1D)
- Bjorken-y (1D)

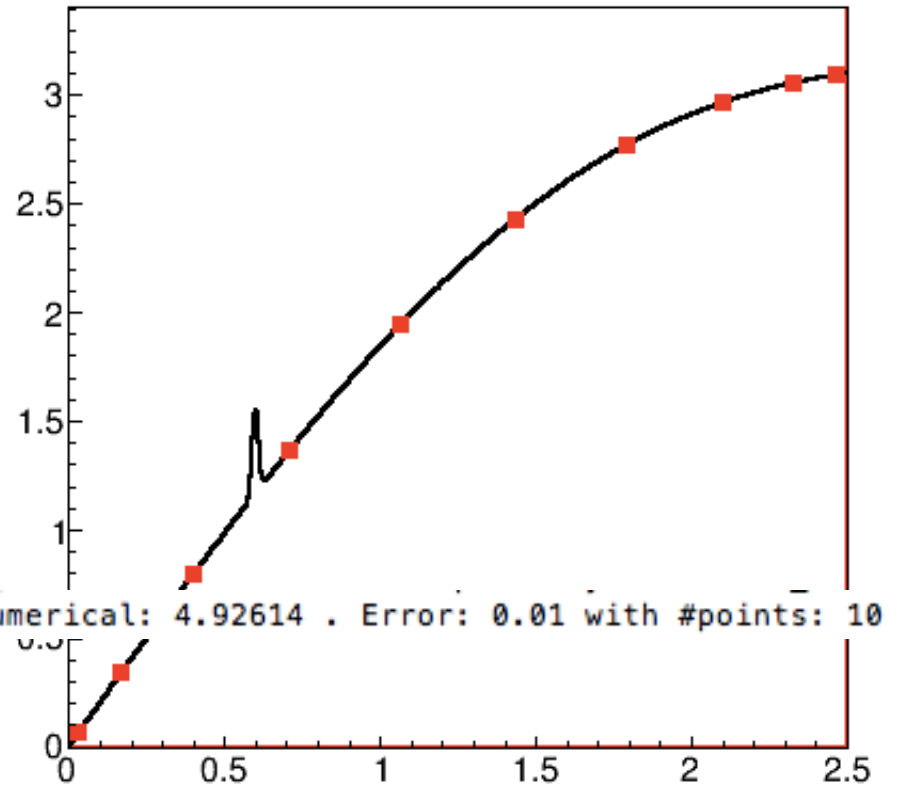
How to solve the ~~8D~~ 6D integral?

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

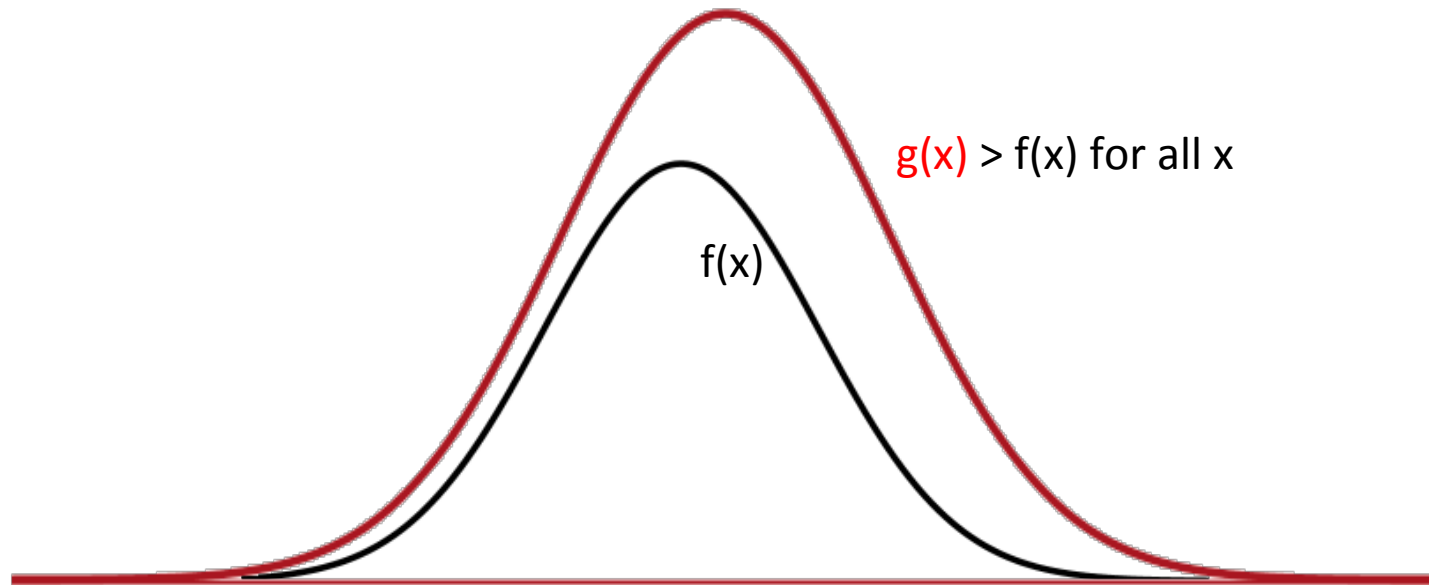
- Interaction vertex position (3)
- Interaction time (1)
 - Relatively easy once other params. are given?
- (Neutrino) Direction (2)
- Neutrino Energy (1)
 - Analytically?
- Bjorken-y (1)

Difficulties

- Event PDF is in general sharply peaked
 - ~1 degree (showers),
~0.1 degree (tracks)
 - ~1m (showers)
- Algorithm generally misses this peak
- Each function evaluation takes time



MC Integration Techniques (1)

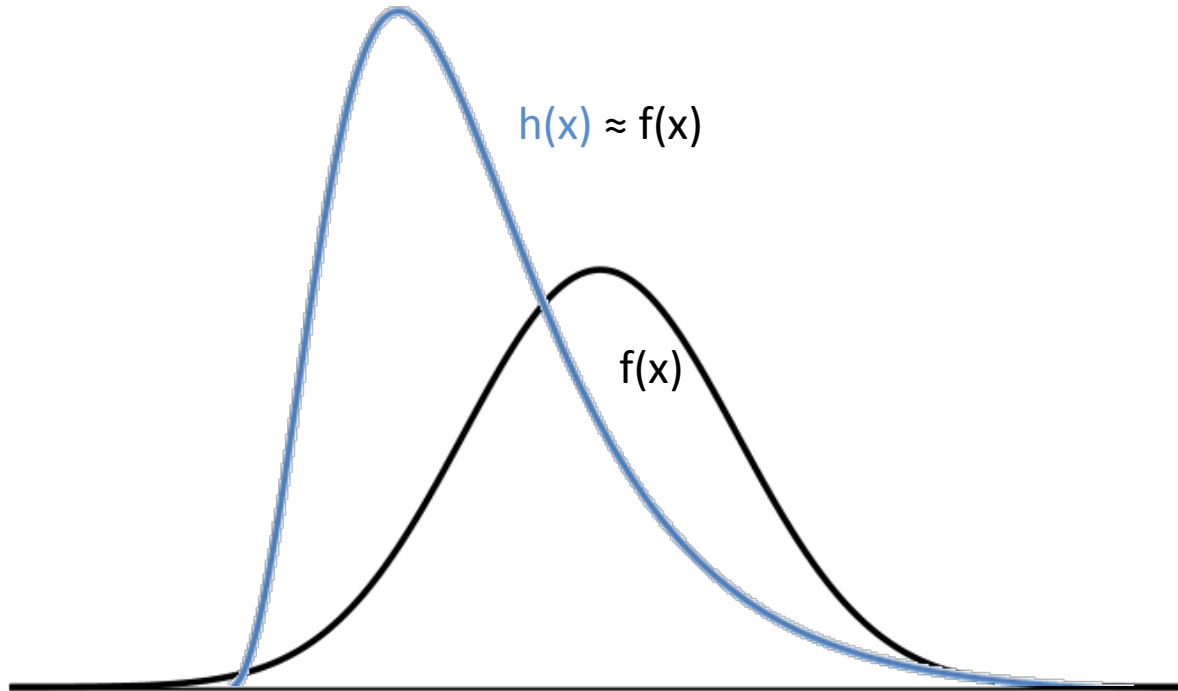


N times:

- 1) Random x from $g(x)$
- 2) Random y , $0 < y < g(x)$
- 3) If $y \leq f(x)$ { $n++$ }

$$I = \int f(x) dx = n/N * \int g(x) dx$$

MC Integration Techniques (2)

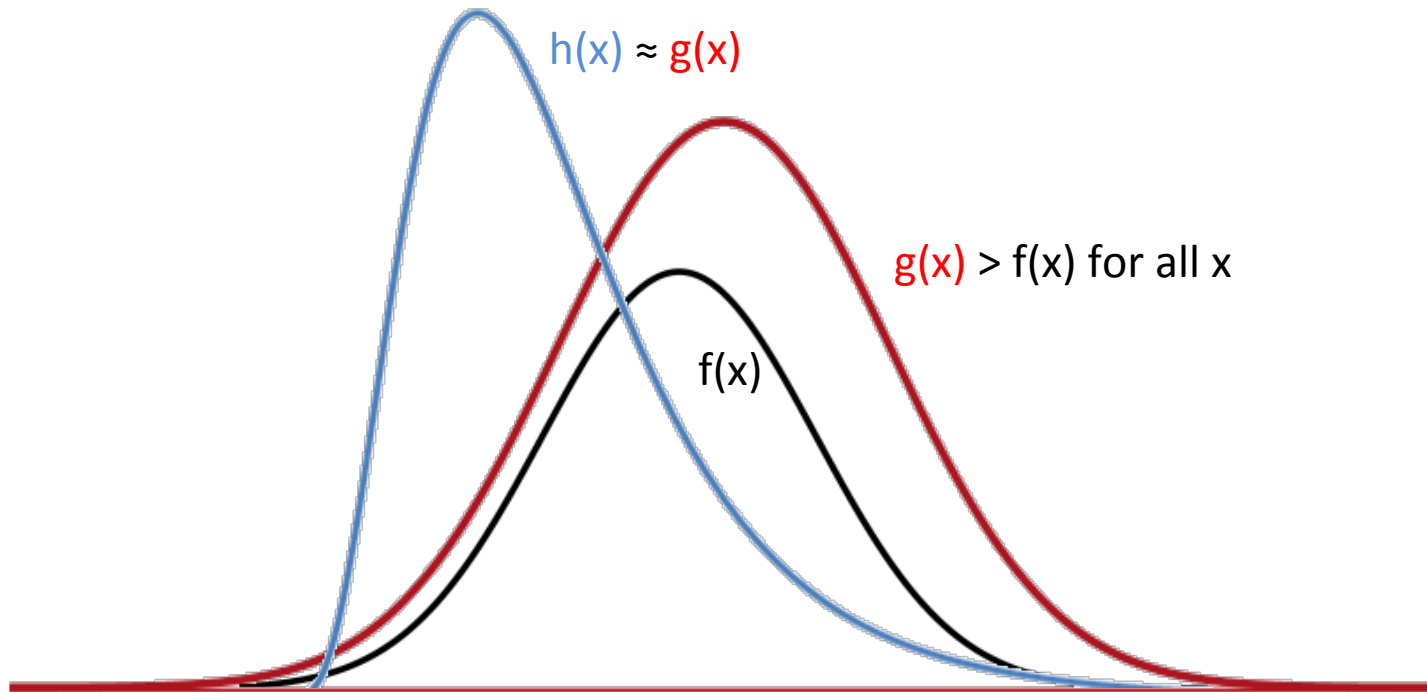


N times:

- 1) Random x from $h(x)$
- 2) $A += f(x)/h(x)$

$$I = \int f(x) dx = A/N * \int h(x) dx$$

Combined

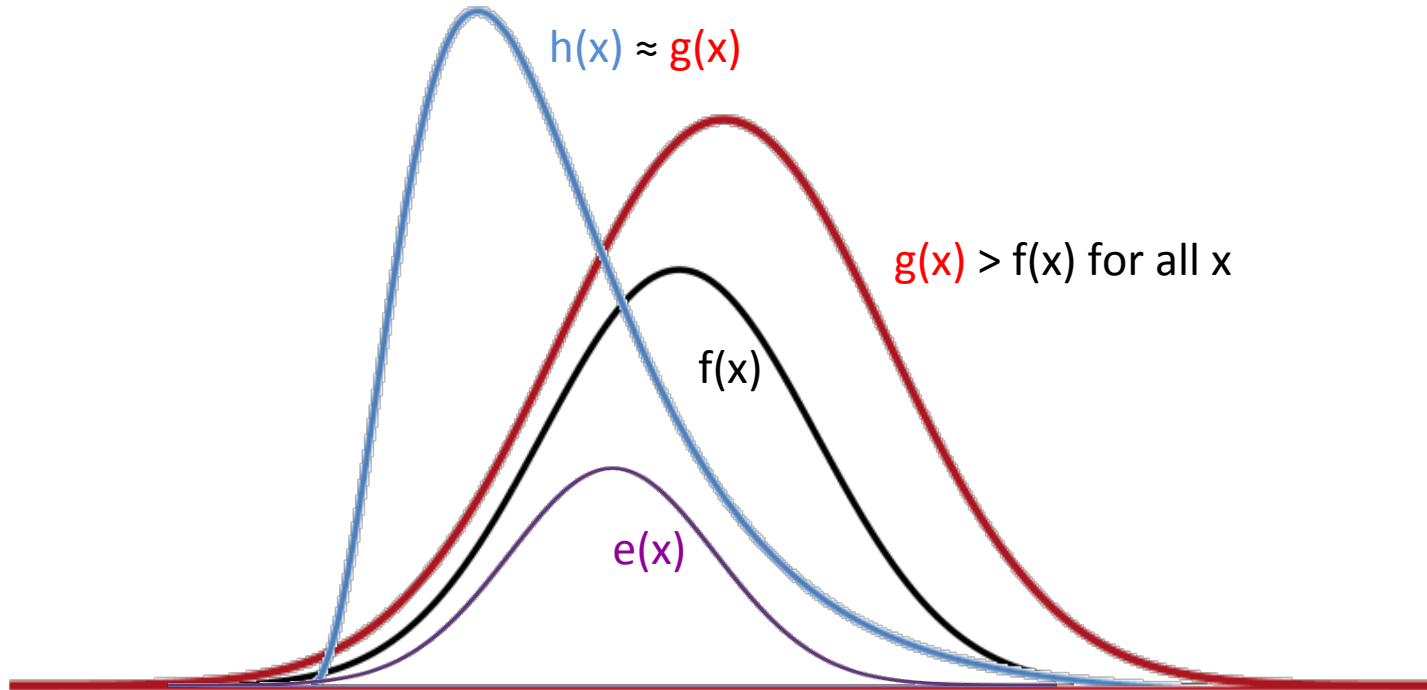


N times:

- 1) Random x from $h(x)$
- 2) $A += g(x)/h(x)$
- 3) Random y , $0 < y < g(x)$
- 4) If $y \leq f(x)$ { $n++$ }

$$I = \int f(x) dx = n/N^2 * A * \int h(x) dx$$

Combined + Extended

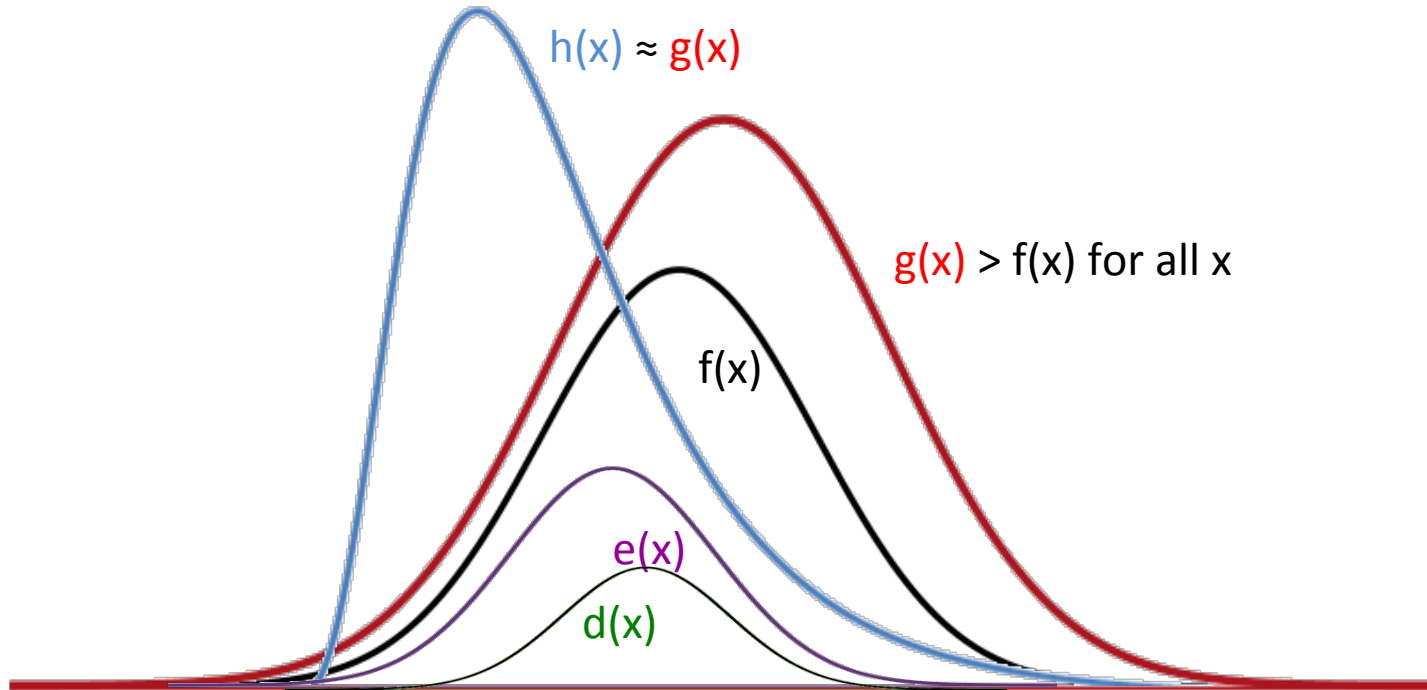


N times:

- 1) Random x from $h(x)$
- 2) $A += g(x)/h(x)$
- 3) Random y , $0 < y < g(x)$
- 4) If $y \leq f(x)$ { if $y \leq e(x)$ { $n++$ } }

$$I = \int e(x) dx = n/N^2 * A * \int h(x) dx$$

Combined + Extended



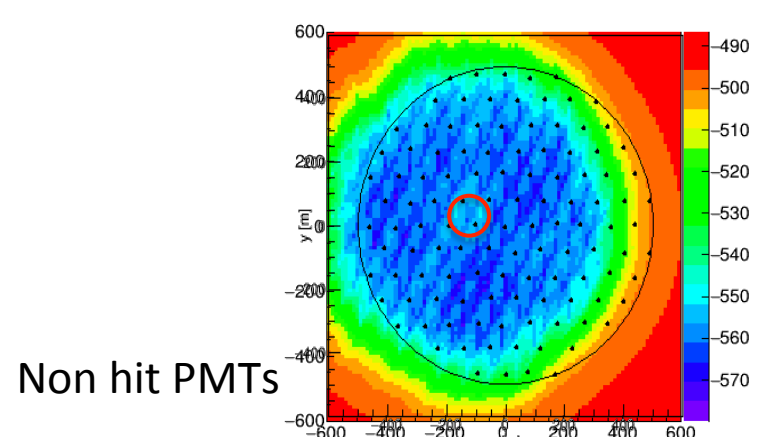
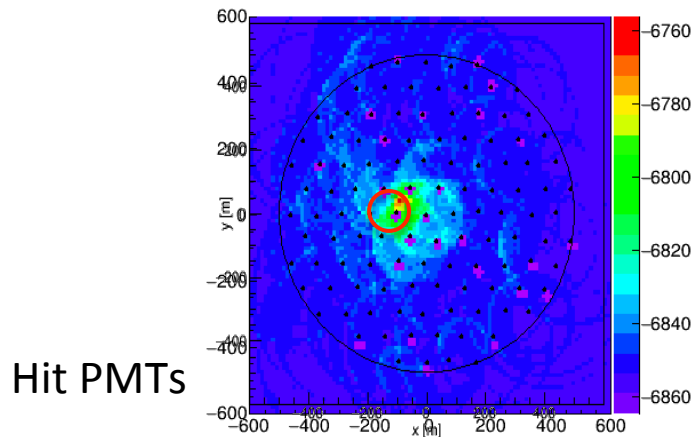
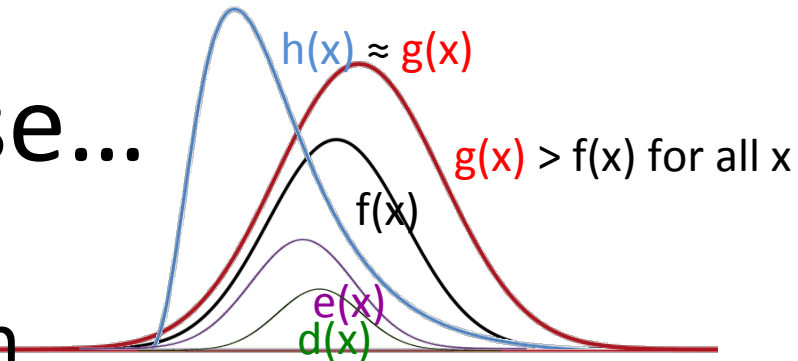
N times:

- 1) Random x from $h(x)$
- 2) $A += g(x)/h(x)$
- 3) Random y , $0 < y < g(x)$
- 4) If $y \leq f(x)$ { if $y \leq e(x)$ { if $y \leq d(x)$ { $n++$ } } }

$$I = \int d(x) dx = n/N^2 * A * \int h(x) dx$$

In our case...

- $h(x)$: some guiding function
- $g(x)$: $P(\text{ev} \mid x)$ over a small subset of PMTs
- $f(x)$: $P(\text{ev} \mid x)$ with slightly more PMTs
- $e(x)$: $P(\text{ev} \mid x)$ over a even more PMTs
- $d(x)$: $P(\text{ev} \mid x)$ with all PMTs



In our case...

- $h(x)$: some guiding function
- $g(x)$: $P(\text{ev} \mid x)$ over a small subset of PMTs
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N times:

- 1) Random x from $h(x)$
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$$I = \int d(x) dx = n/N^2 * A * \int h(x) dx$$

Guiding function

- Convenient choice: multivariate normal distribution
 - tracks: JPrefit PDF
 - Showers: ????

Guiding function

- Convenient choice: multivariate normal distribution
 - tracks: JPrefit PDF
 - Showers: ????

MASTER THESIS

**Reconstruction of High-energy Neutrino-induced
Particle Showers in KM3NeT.**

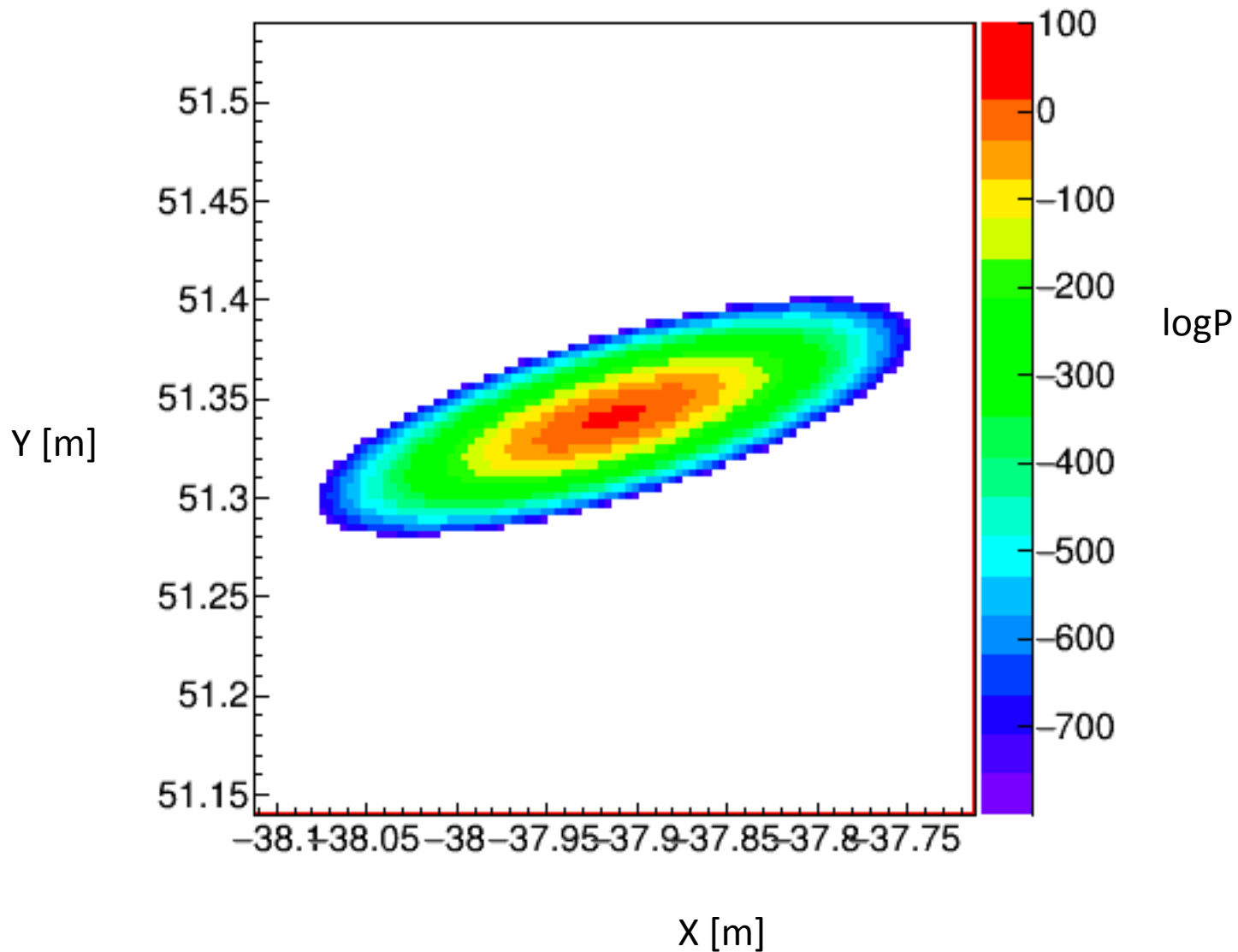
BY

www.nikhef.nl/~kmelis/Masters_Thesis.pdf/Thesis.pdf

Shower Vertex PDF

- Basically a Chi^2 distribution
 - Very sensitive to outliers (i.e. 40K hits)
- Use first triggered hit on each DOM
- Hit clustering algorithm:
 - If many hits: iteratively remove worst hit
 - If #hits ≤ 16 : Try all combinations

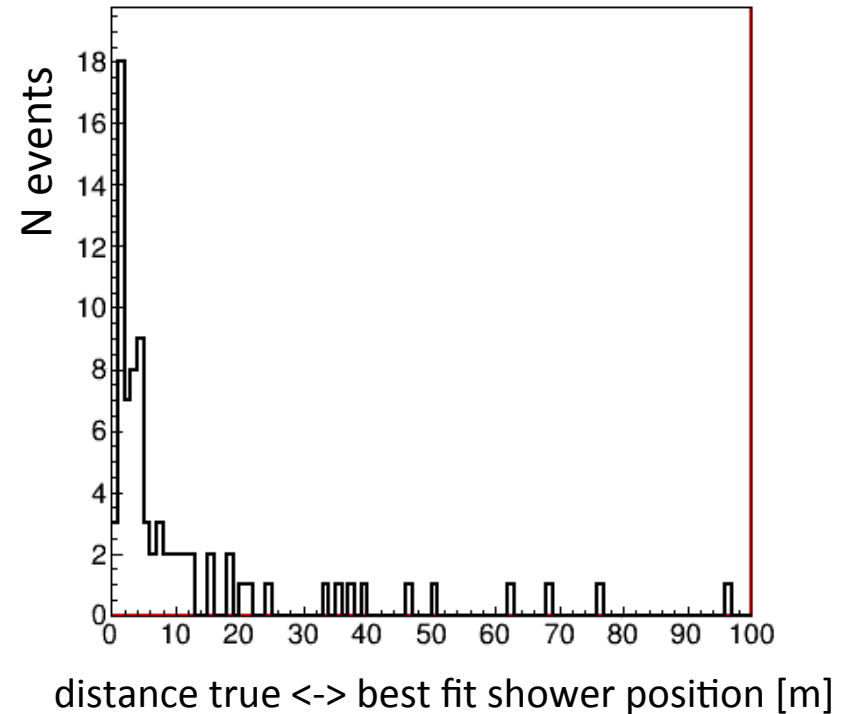
Shower Vertex PDF



Bonus: Shower Position Reconstruction


- Reasonable resolution:
 - median $\sim 1\text{m}$
- Principle (cluster+ χ^2) usable for tau double bang prefit?

Preliminary

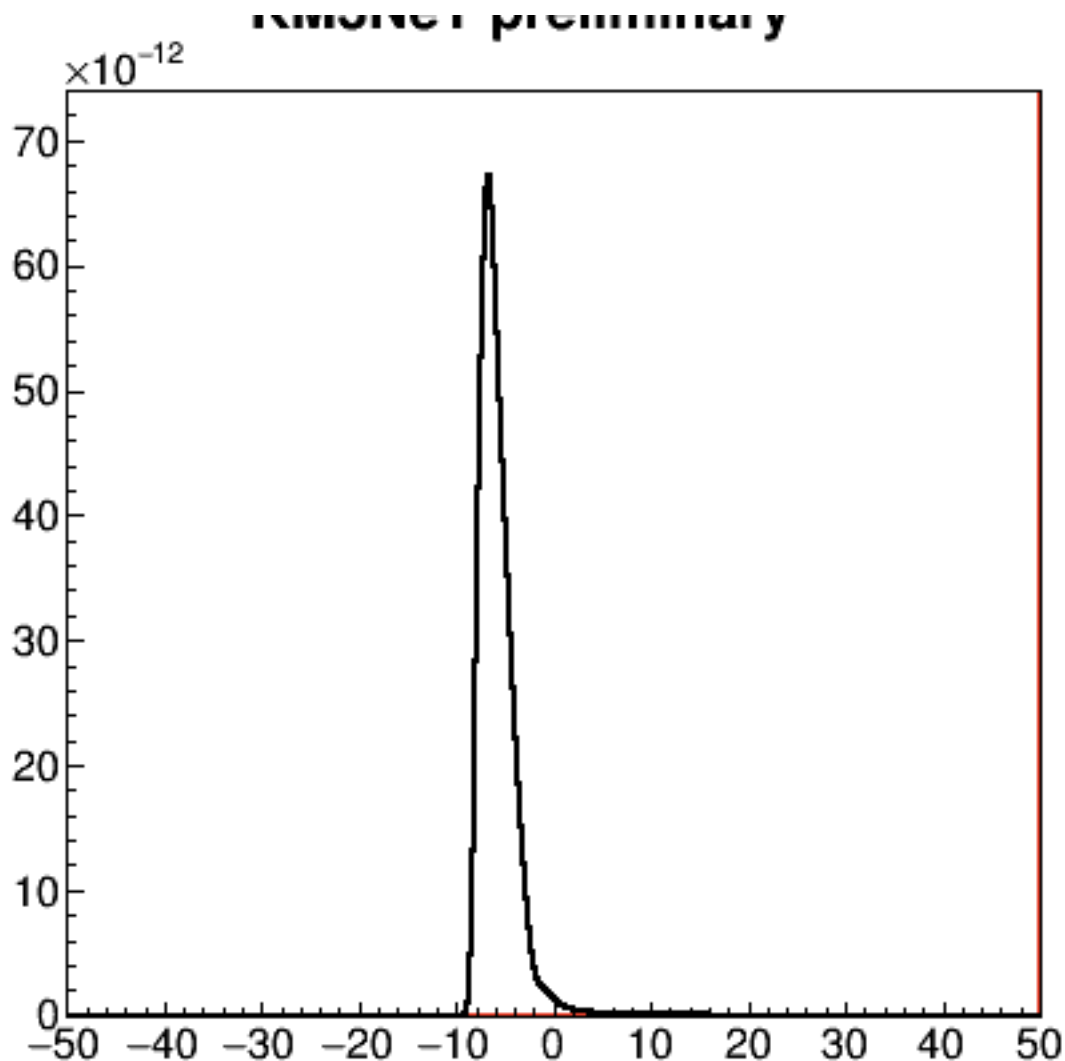


Conclusions



- Most ingredients seem
 - Neutrino background and signal fluxes
 - Detection efficiency tables
 - Event probability
- Integral evaluation seems feasible with MC techniques
-  **BONUS** :
bij AH Willem's
 - Shower prefit (+tau double bang prefit?)
 - Very fast neutrino MC simulator

Integrating over time



```
int P dt: 0.0002365506325
```

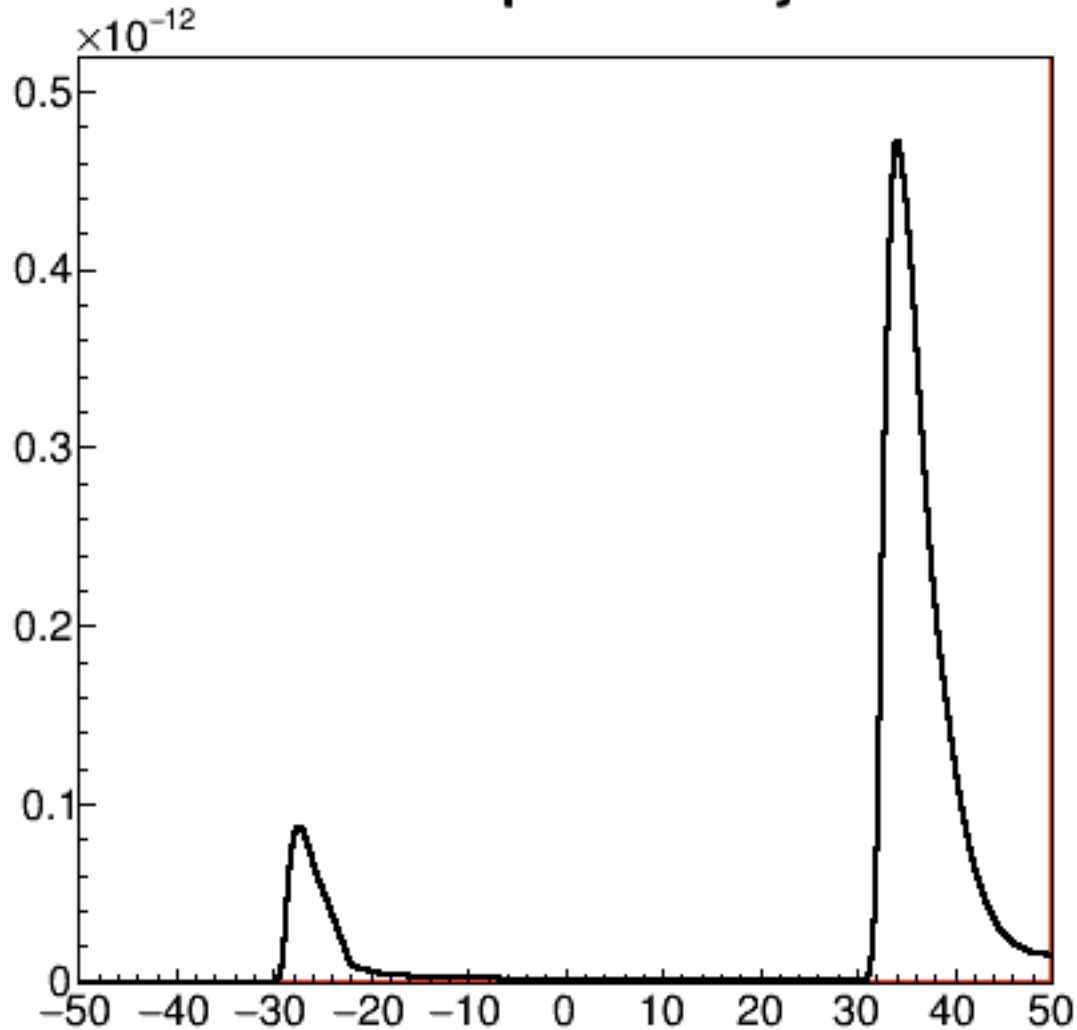
```
377.504 ms elapsed
```

```
377.943 ms user
```

```
0.000 ms system
```

For 3e6 points (& 2PMTs hit)

Integrating over time



```
int P dt: 3.740579387e-06
```

```
371.486 ms elapsed  
370.943 ms user  
0.000 ms system  
99%CPU
```


Hypothesis Testing

- Two hypotheses:
 - H0: background flux only
 - H1: background + signal flux
- Criterium when to select H1
 - $P(\text{accept H1} \mid H0 = \text{true}) < 0.000..$ (5 sigma)

Likelihood ratio

- Best criterium:

$$\lambda = \log [P (data|H1)] - \log [P (data|H0)]$$

- Data 'looks like' H1 => high lambda
- Criterium when to select H1
 - Accept H1 if $\lambda > \lambda_{crit}$
 - $P(\text{accept H1} \mid H0 = \text{true}) < 0.000..$ (5 sigma)

Likelihood ratio

- Likelihood ratio:

$$\lambda = \log [P (data|H1)] - \log [P (data|H0)]$$

$$\log [P (data|H)] = -\mu_{tot}(H) + \sum_{events} \log \left[\int P(ev_i|x) \cdot P^{det}(x) \cdot \mu^{flux}(x|H) dx \right]$$

$\mu_{tot}(H)$ Total number of expected detected events from H

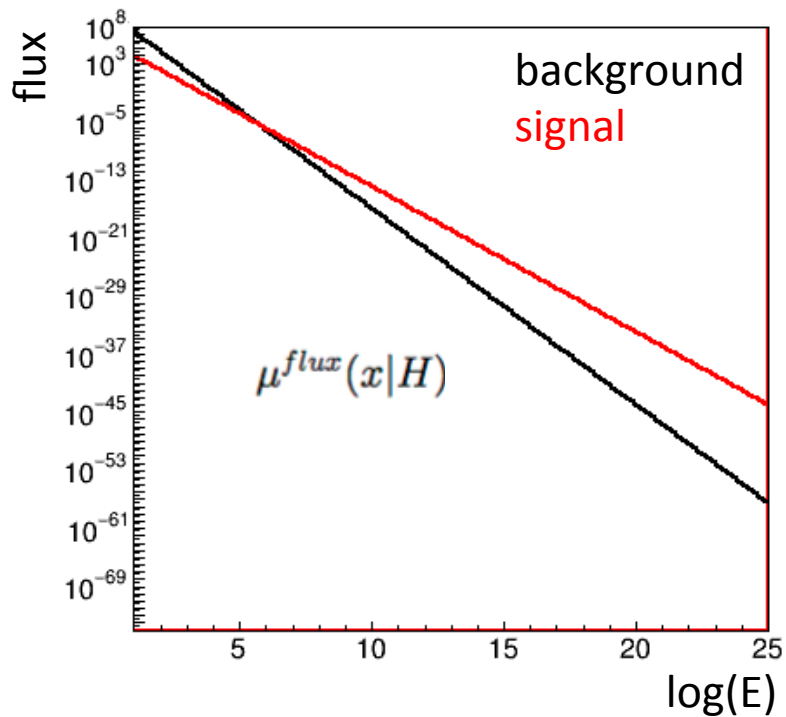
$P(ev_i|x)$ Probability to obtain measured event ev_i
given a certain (8D) neutrino hypothesis x

$P^{det}(x)$ Probability to detect (=trigger) and select event

$\mu^{flux}(x|H)$ Number of expected events from H in our detector (can)

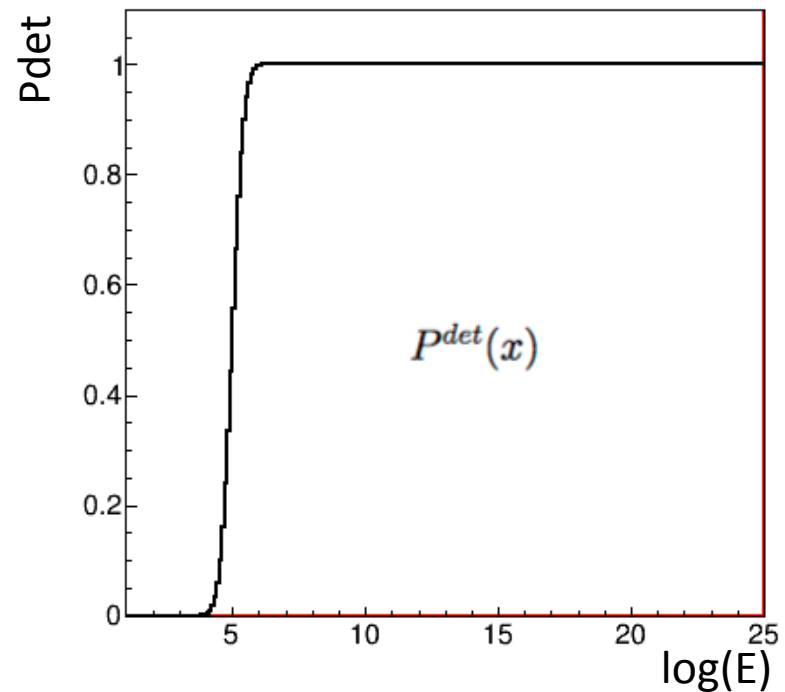
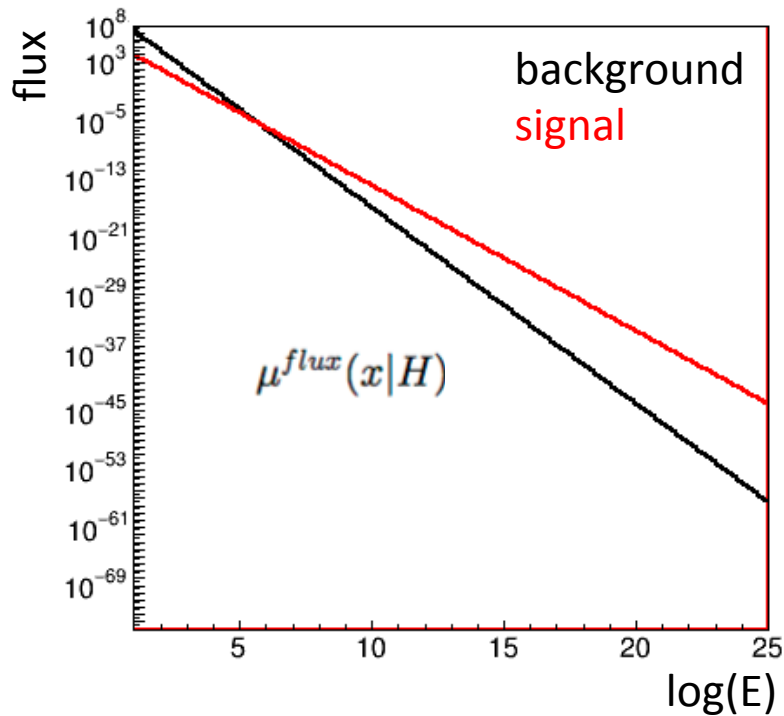
Example (1D)

- Neutrino only has energy



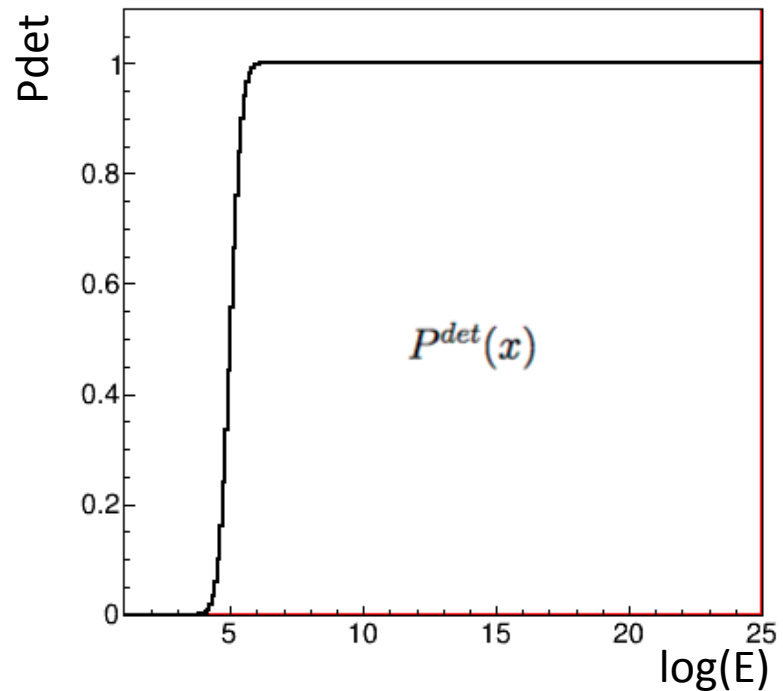
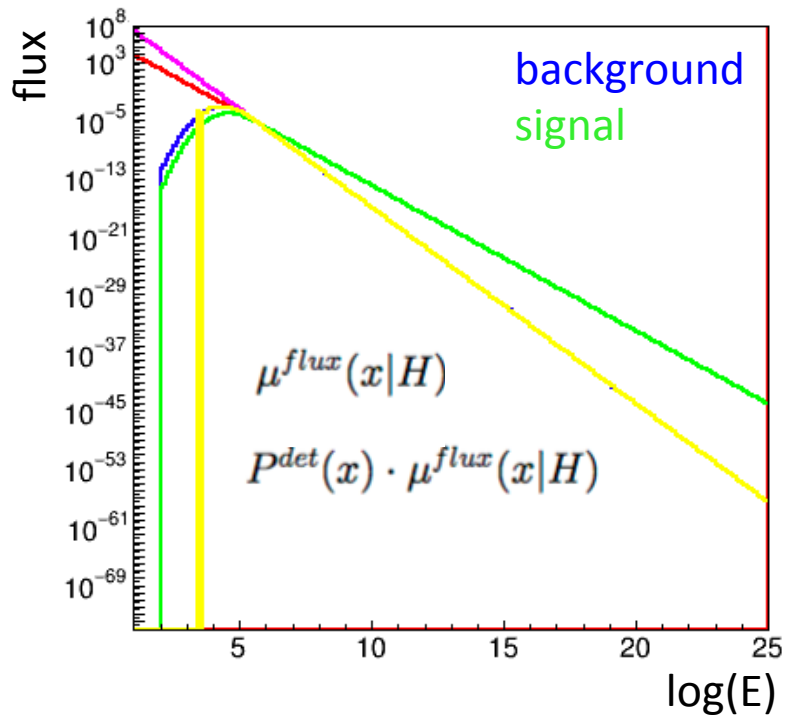
Example (1D)

- Neutrino only has energy



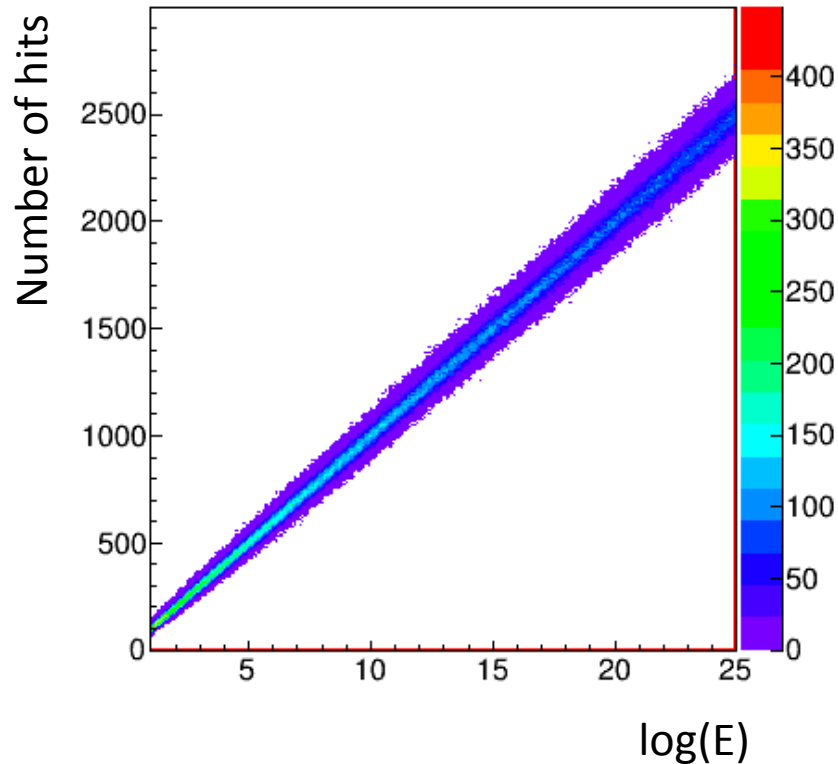
Example (1D)

- Neutrino only has energy



Example (1D)

- Neutrino only has energy
- Event only measures number of hits



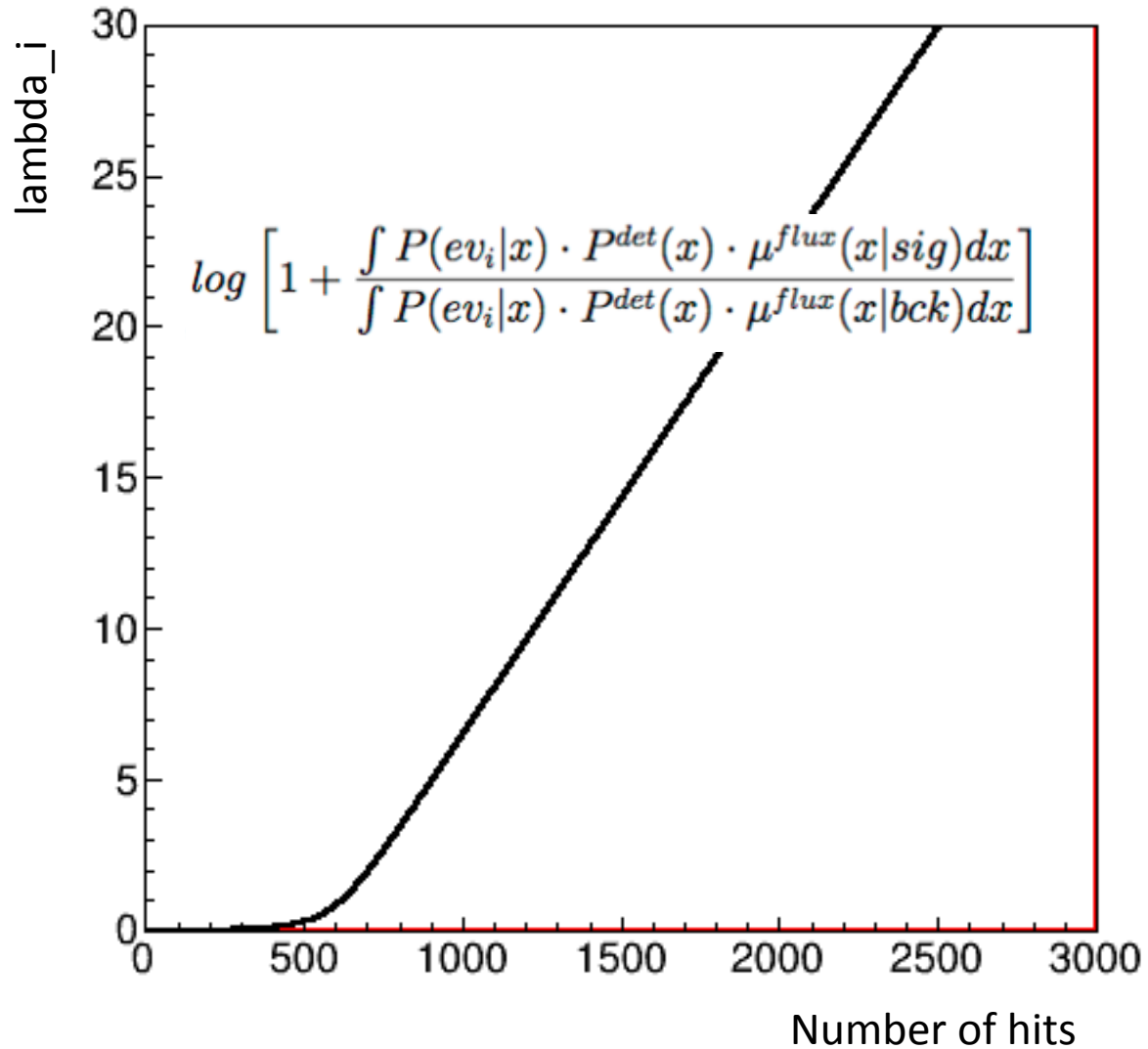
Example (1D)

- Neutrino only has energy
- Event only measures number of hits

$$\begin{aligned}\lambda &= \log [P(\text{data}|\text{bck} + \text{sig})] - \log [P(\text{data}|\text{bck})] \\ &= -\mu_{\text{tot}}(\text{sig}) + \sum_{\text{events}} \log \left[1 + \frac{\int P(\text{ev}_i|x) \cdot P^{\text{det}}(x) \cdot \mu^{\text{flux}}(x|\text{sig}) dx}{\int P(\text{ev}_i|x) \cdot P^{\text{det}}(x) \cdot \mu^{\text{flux}}(x|\text{bck}) dx} \right]\end{aligned}$$

- High number of hits \Rightarrow high energy \Rightarrow
data looks like H1 \Rightarrow high lambda

Example (1D)



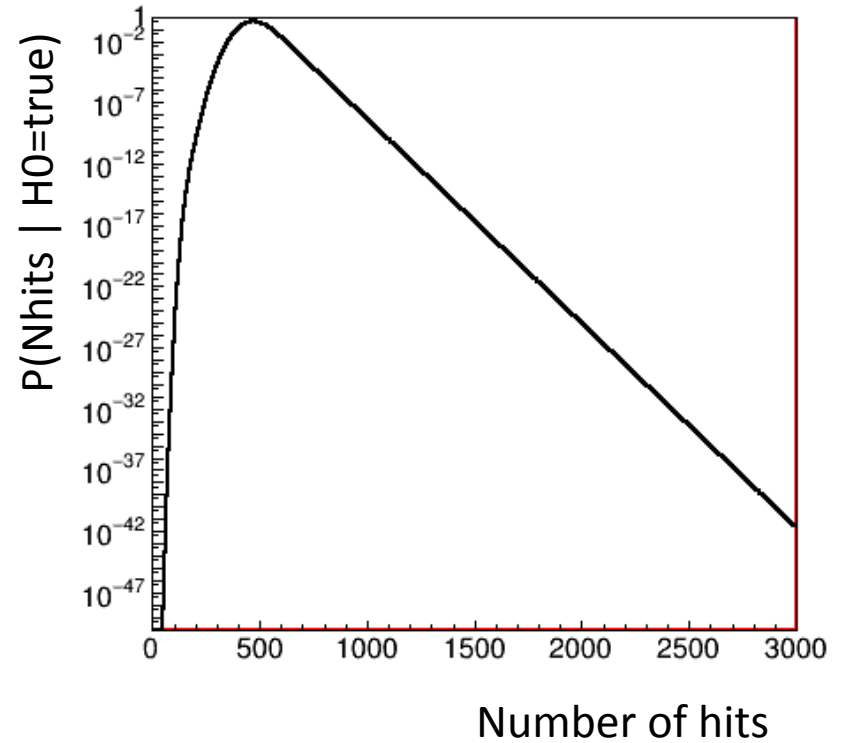
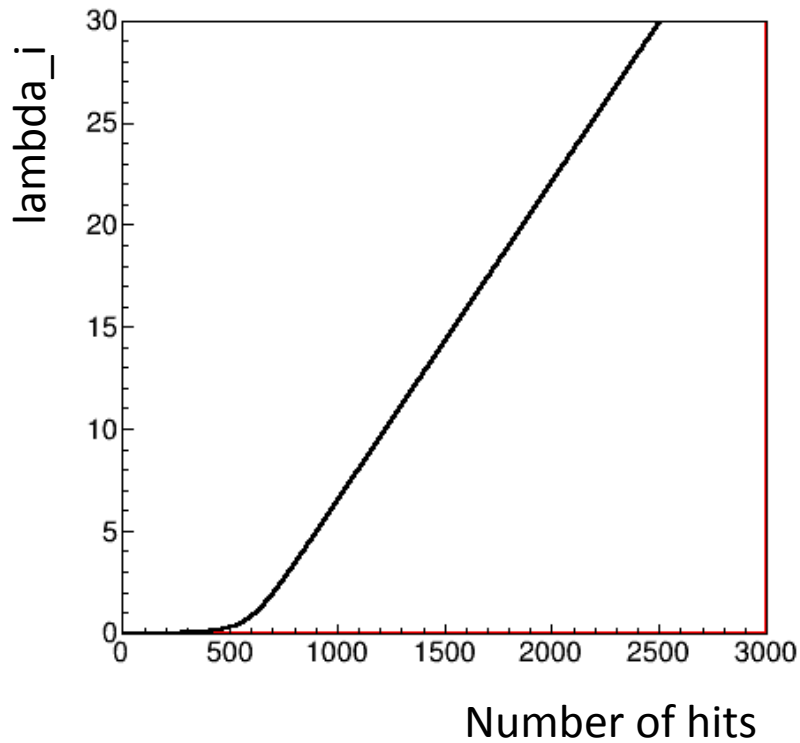
Likelihood ratio

- Best criterium:

$$\lambda = \log [P (data|H1)] - \log [P (data|H0)]$$

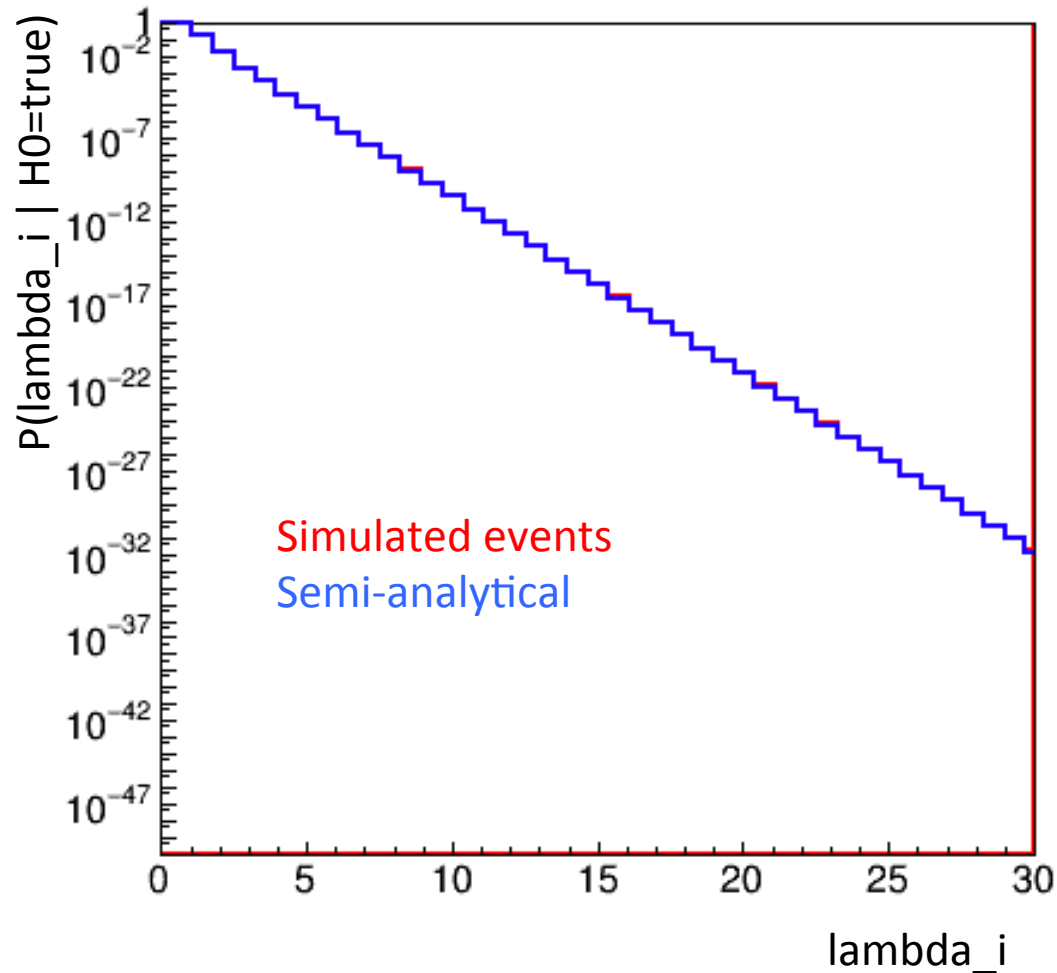
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- Criterium when to select H1
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 - $P(\text{accept H1} \mid H0 = \text{true}) < 0.000..$ (5 sigma)

$P(\lambda > \lambda_{\text{crit}} \mid H_0 = \text{true}) < 0.000..$ (5 sigma)



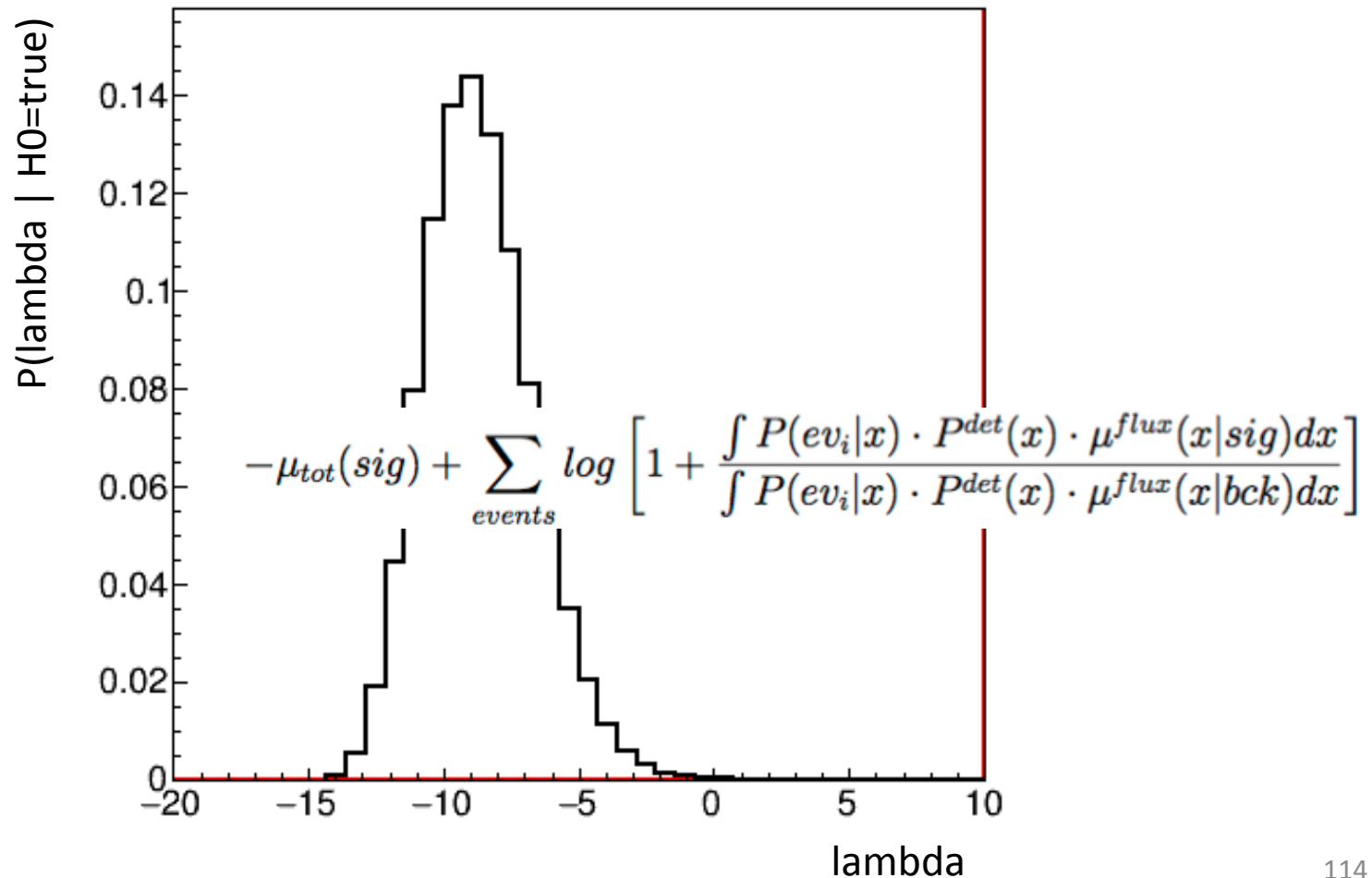
$P(\lambda > \lambda_{\text{crit}} \mid H_0 = \text{true}) < 0.000..$ (5 sigma)

Single detected event, given H_0 is true



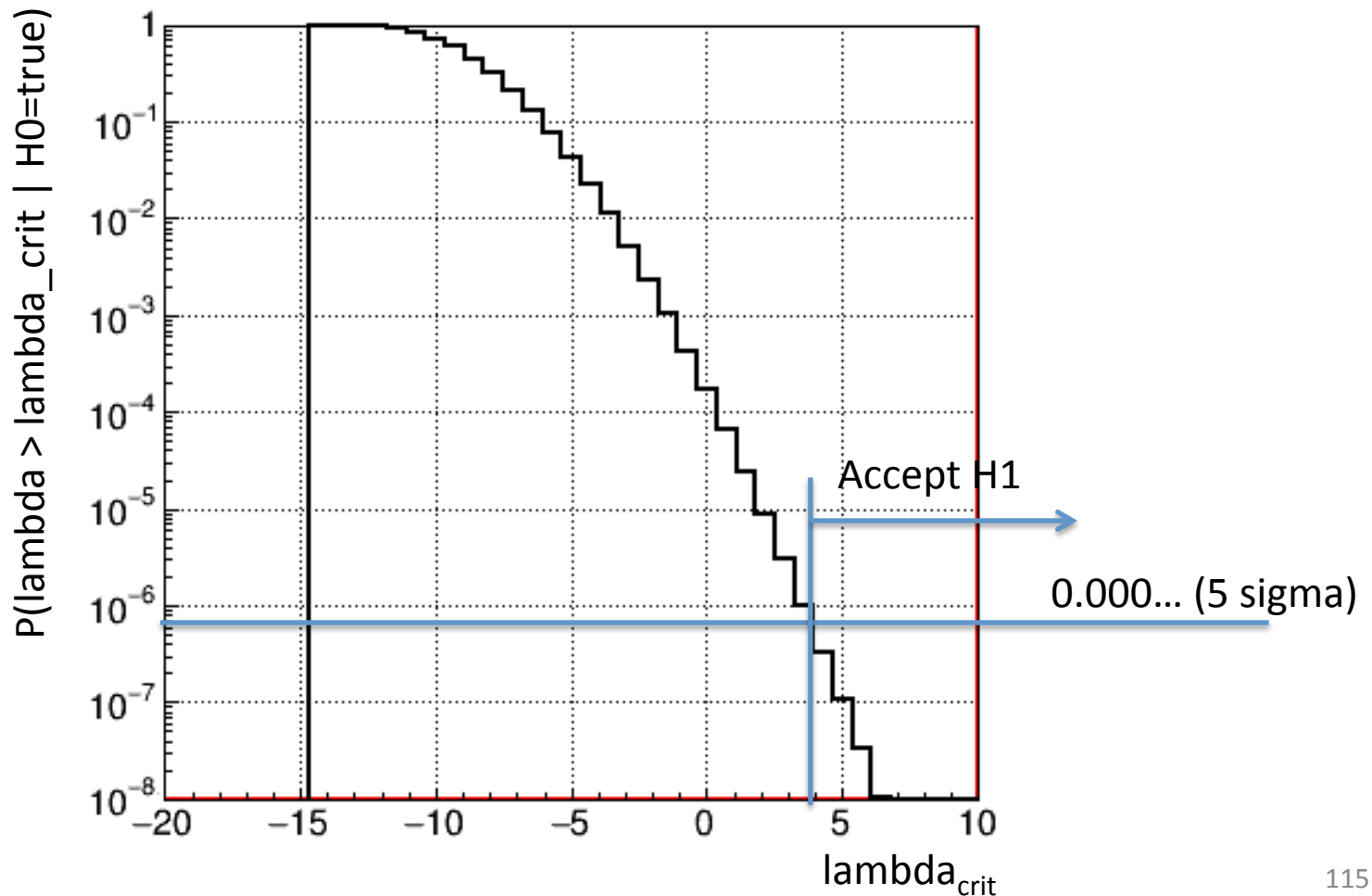
$P(\text{lambda} > \text{lambda}_{\text{crit}} \mid H_0 = \text{true}) < 0.000..$ (5 sigma)

(multiple) detected events in certain timeperiod, given H0 is true



$$P(\lambda > \lambda_{\text{crit}} \mid H_0 = \text{true}) < 0.000\dots \text{ (5 sigma)}$$

(multiple) detected events in certain timeperiod, given H_0 is true



Ter Leering ende Vermaeck

- Now: 3D example (energy + direction)
- Soon: 8D example

- Reproduce conventional method and show that new method works (better)

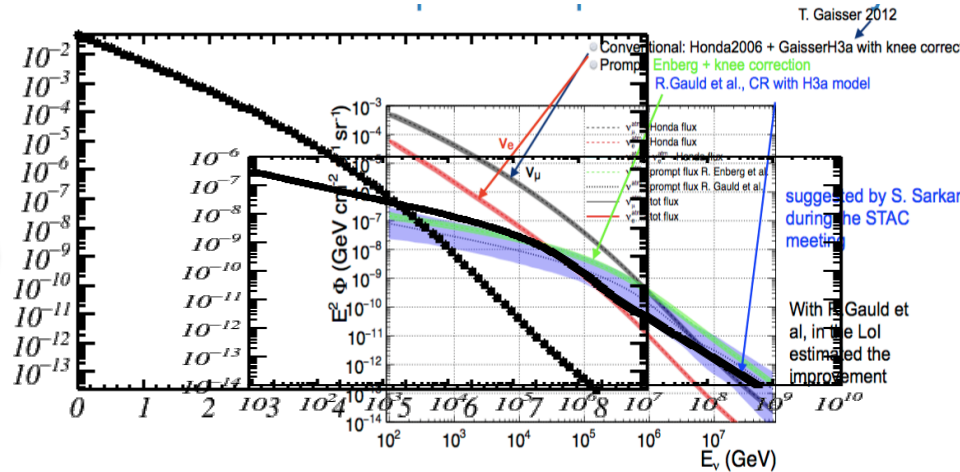
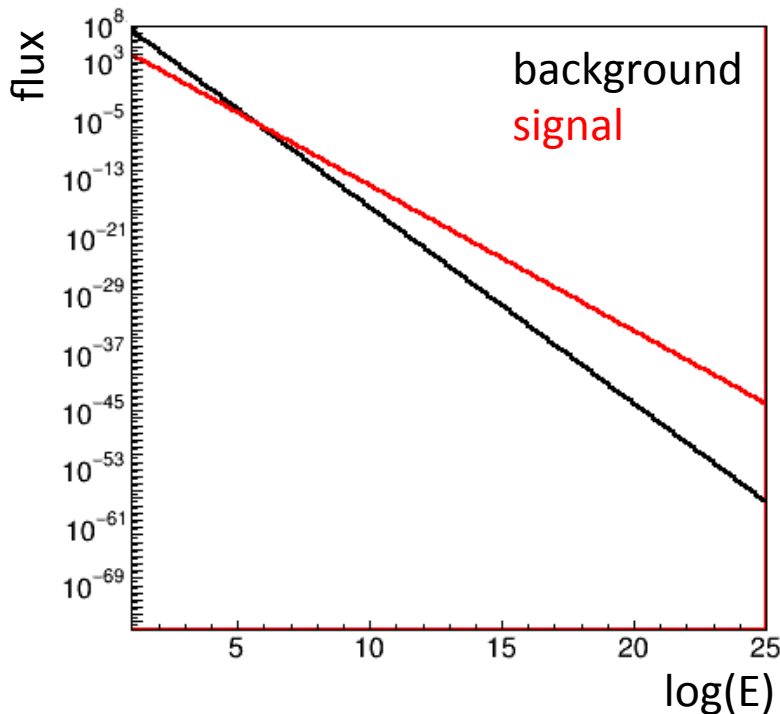
- Replace likelihood terms with real (MC) events

Ter Leering ende Vermaeck

$$\lambda = \log [P(\text{data}|H1)] - \log [P(\text{data}|H0)]$$

$$\log [P(\text{data}|H)] = -\mu_{\text{tot}}(H) + \sum_{\text{events}} \log \left[\int P(\text{ev}_i|x) \cdot P^{\text{det}}(x) \cdot \mu^{\text{flux}}(x|H) dx \right]$$

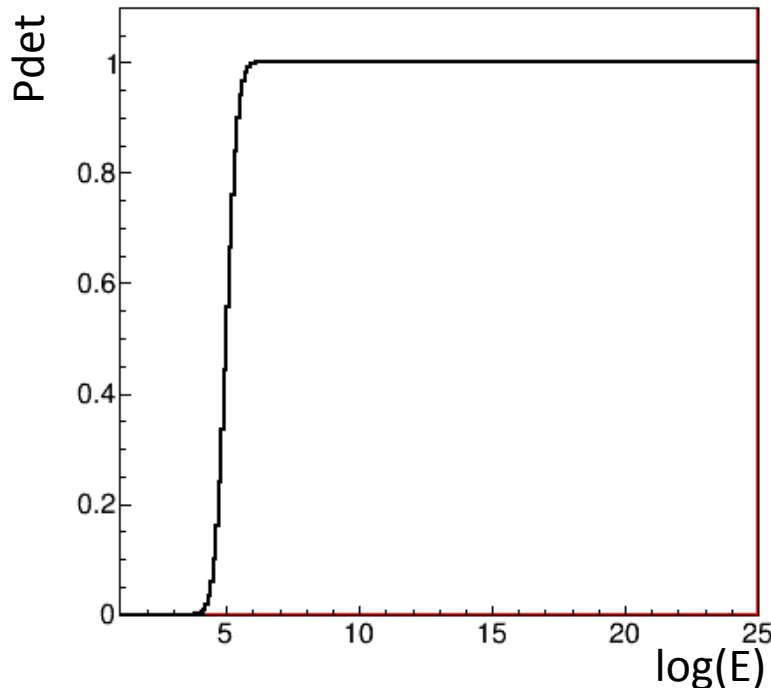
Fast parameterizations



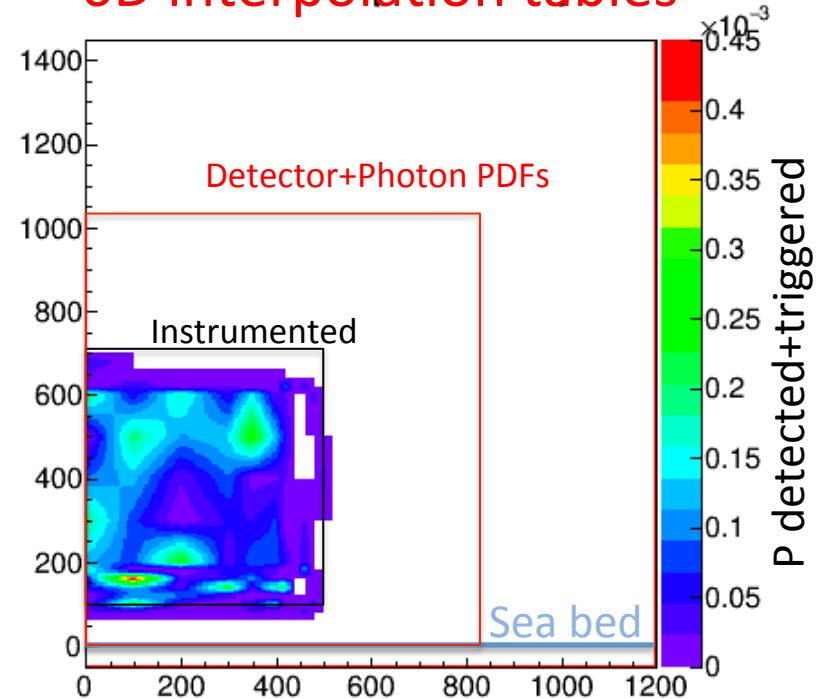
Ter Leering ende Vermaeck

$$\lambda = \log [P(\text{data}|H1)] - \log [P(\text{data}|H0)]$$

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6D interpolation tables

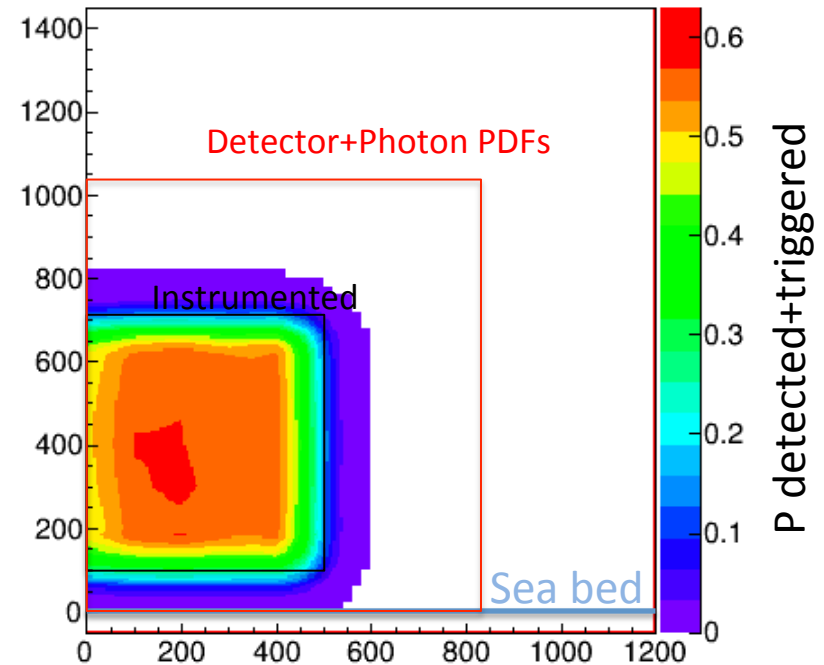
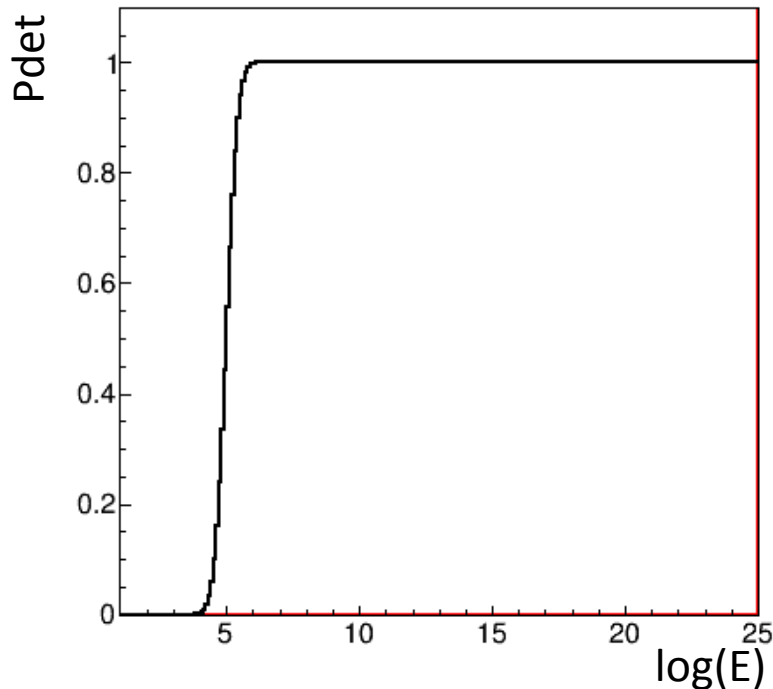


Ter Leering ende Vermaeck

$$\lambda = \log [P(\text{data}|H1)] - \log [P(\text{data}|H0)]$$

$$\log [P(\text{data}|H)] = -\mu_{\text{tot}}(H) + \sum_{\text{events}} \log \left[\int P(\text{ev}_i|x) P^{\text{det}}(x) \mu^{\text{flux}}(x|H) dx \right]$$

6D interpolation tables

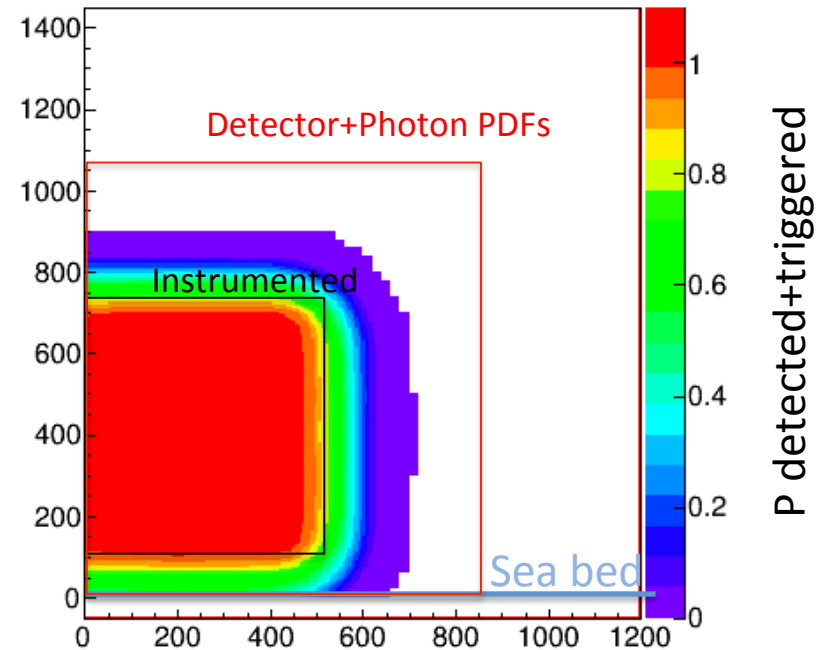
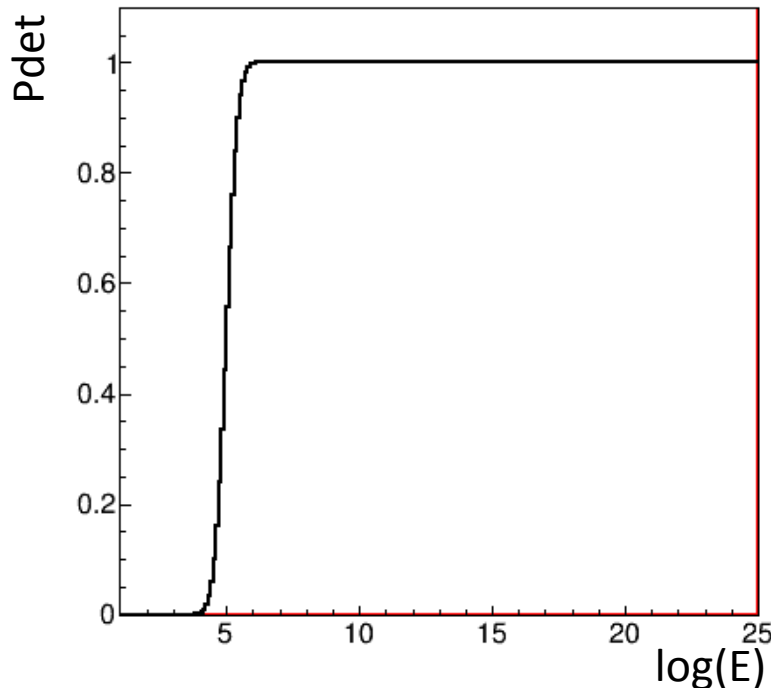


Ter Leering ende Vermaeck

$$\lambda = \log [P (data|H1)] - \log [P (data|H0)]$$

$$\log [P (data|H)] = -\mu_{tot}(H) + \sum_{events} \log \left[\int P(ev_i|x) P^{det}(x) \mu^{flux}(x|H) dx \right]$$

6D interpolation tables

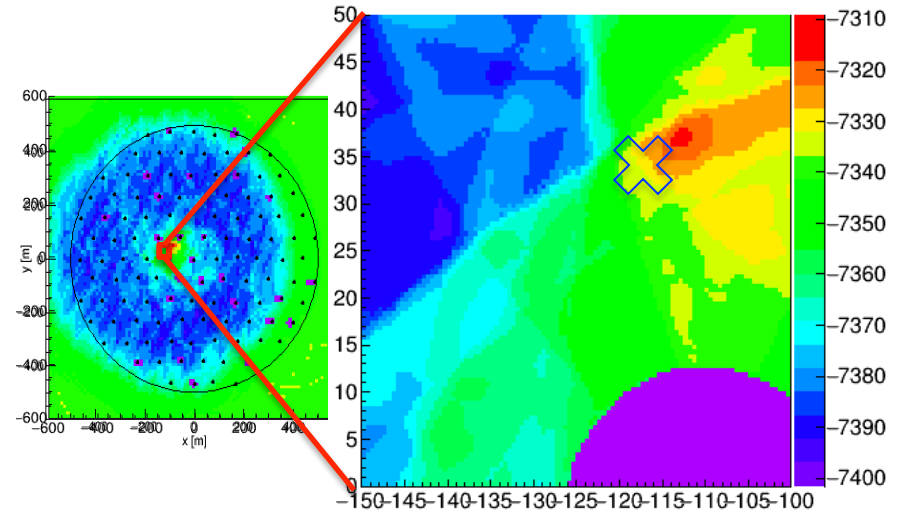
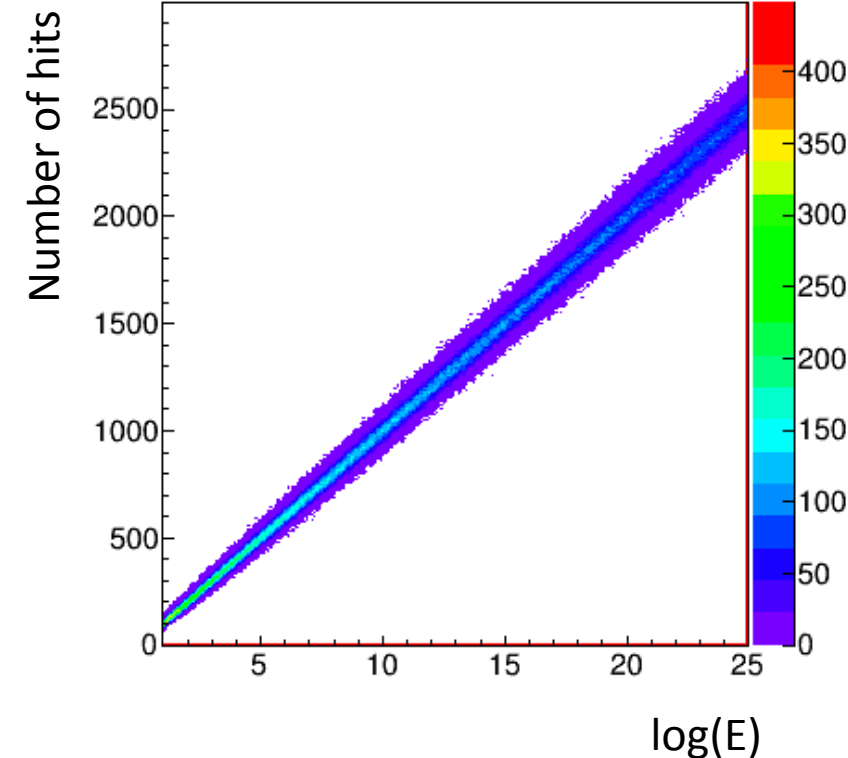


Ter Leering ende Vermaeck

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PDF



Outlook

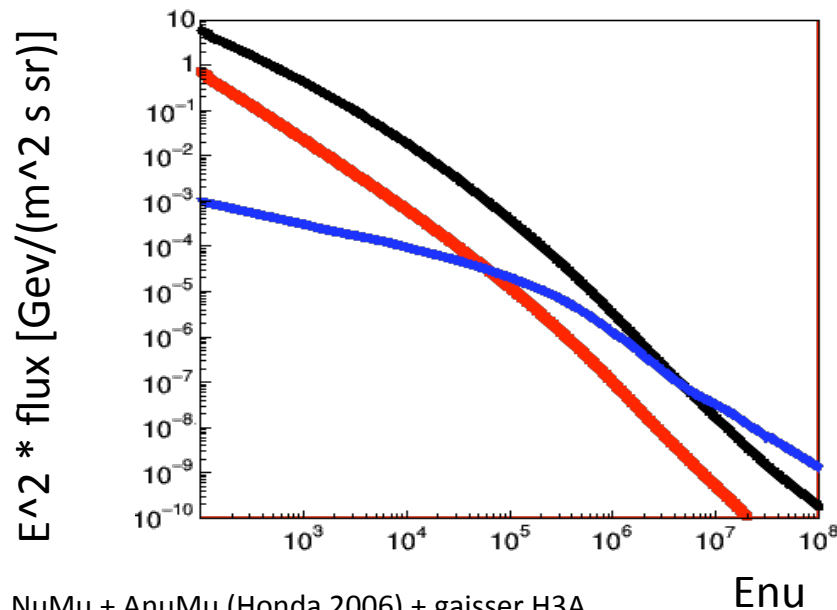
- Proof of principle
- From toy MC to real MC
- Composite hypotheses
- Atm. muon background

Source Searches

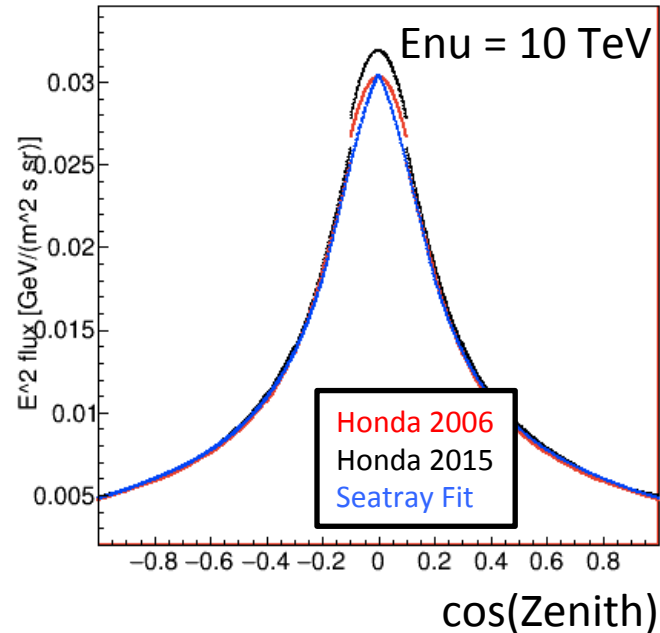
- Observed flux of events
 - Single showers
 - Muon tracks
 - Double bangs
 - Sugar daddy's
- Expected flux
 - H0: atm. neutrinos and atm. muons
 - H1: cosmic source

Expected Flux (Atm. nu bckgr.)

- Neutrino flux @ atmosphere
 - Honda 2006
 - Interpolating 2D tables

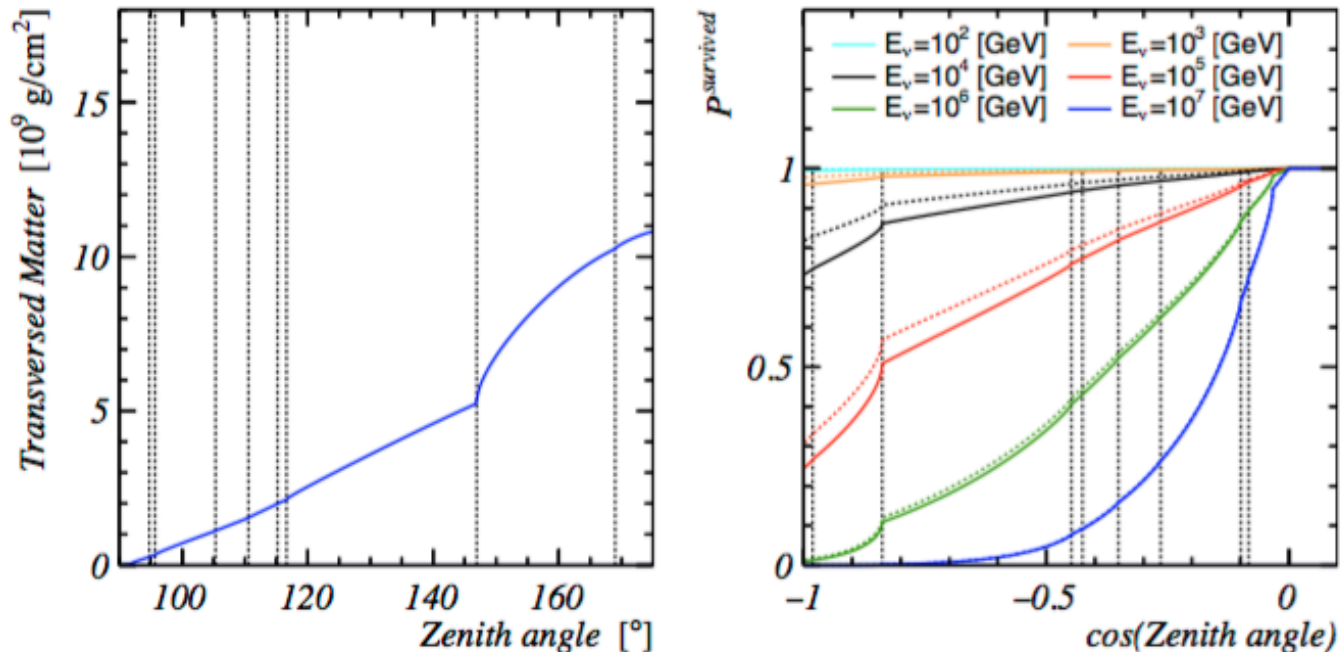


NuMu + AnuMu (Honda 2006) + gaisser H3A
 NuE + AnuE (Honda 2006) + gaisser H3A
 Prompt flux (indep. of flavor), Gauld, includes H3A



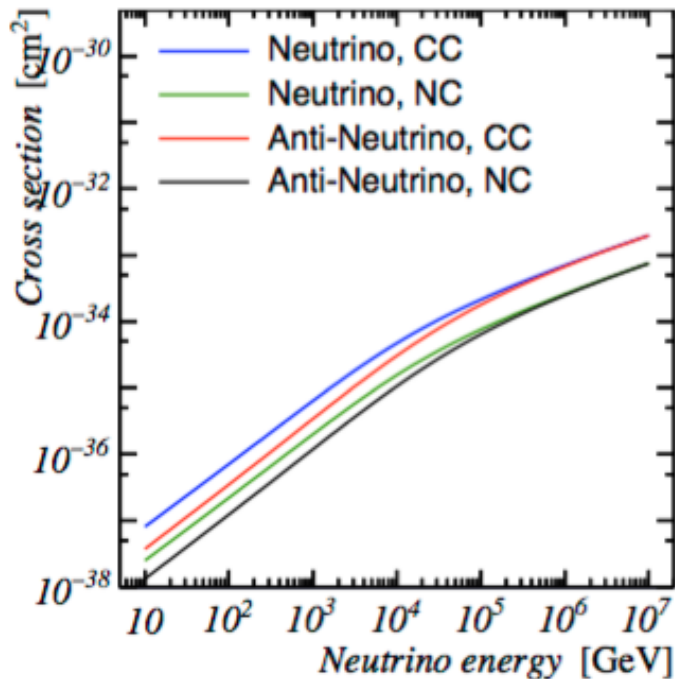
Expected Flux (Atm. nu bckgr.)

- Neutrino flux @ atmosphere
- Earth propagation
 - Only CC neutrino absorption included
 - No NC scattering / energy losses (yet?)



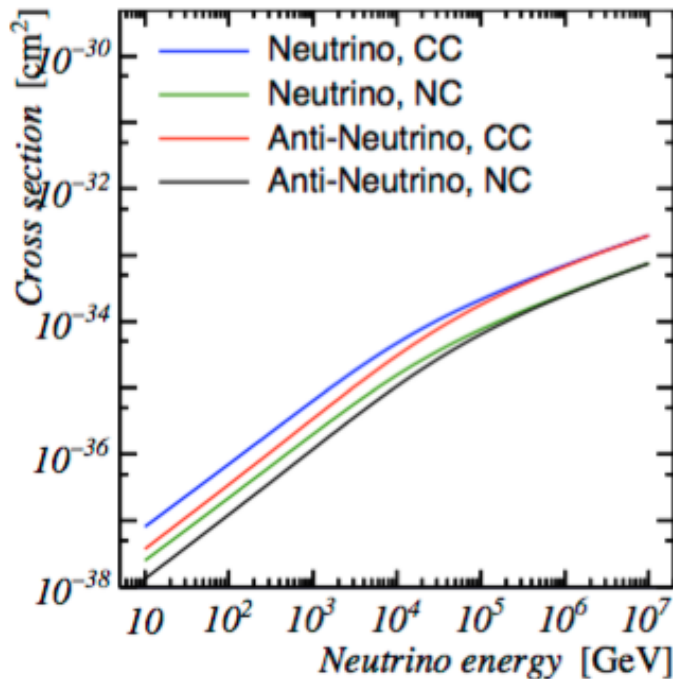
Expected Single Shower Flux

- Neutrino flux @ atmosphere
- Earth propagation
- Interaction probability



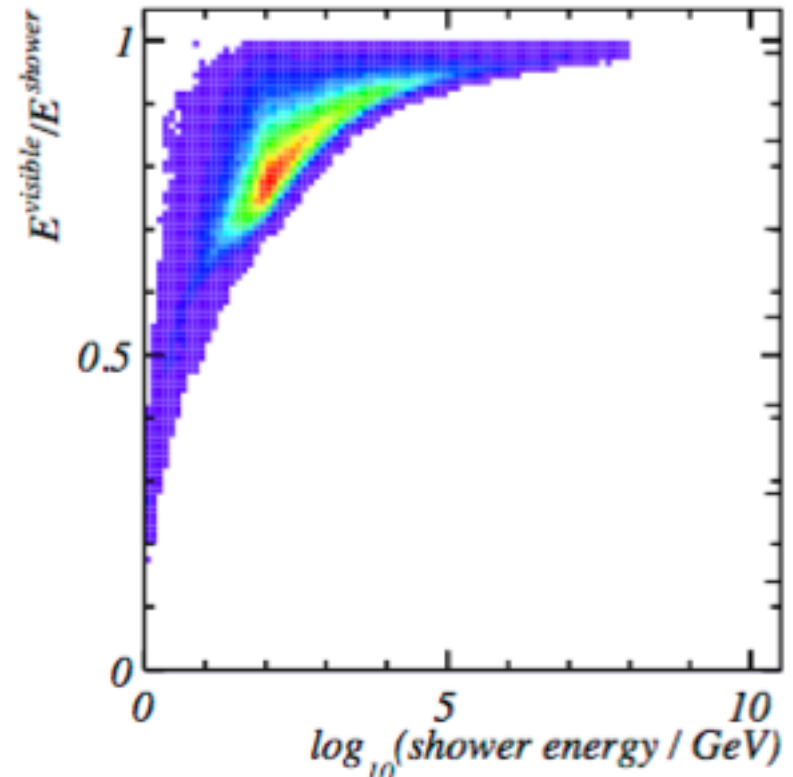
Expected Single Shower Flux

- Neutrino flux @ atmosphere
- Earth propagation
- Interaction probability
- Shower energy (Bjorken-y)

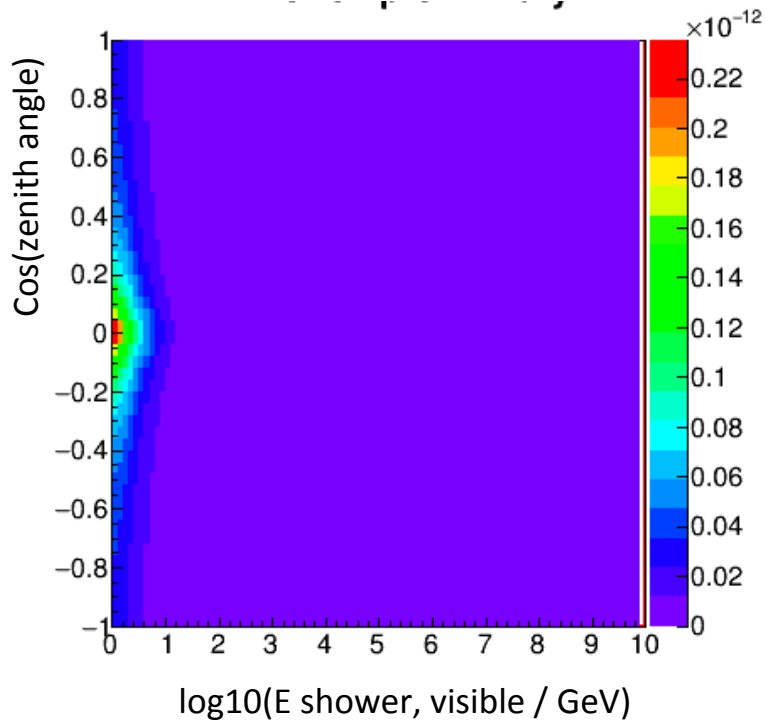


Expected Single Shower Flux

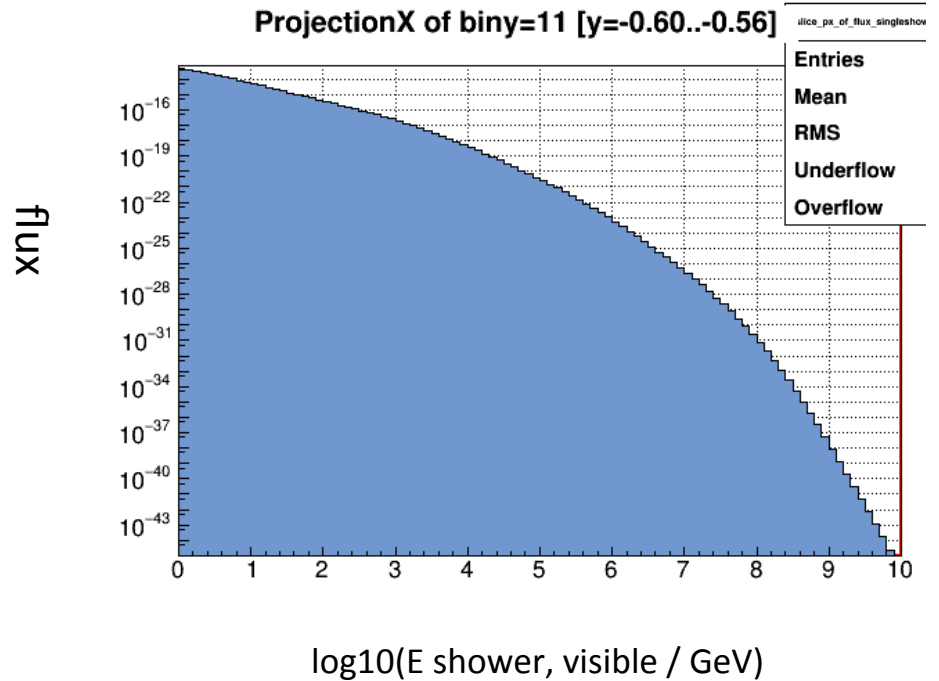
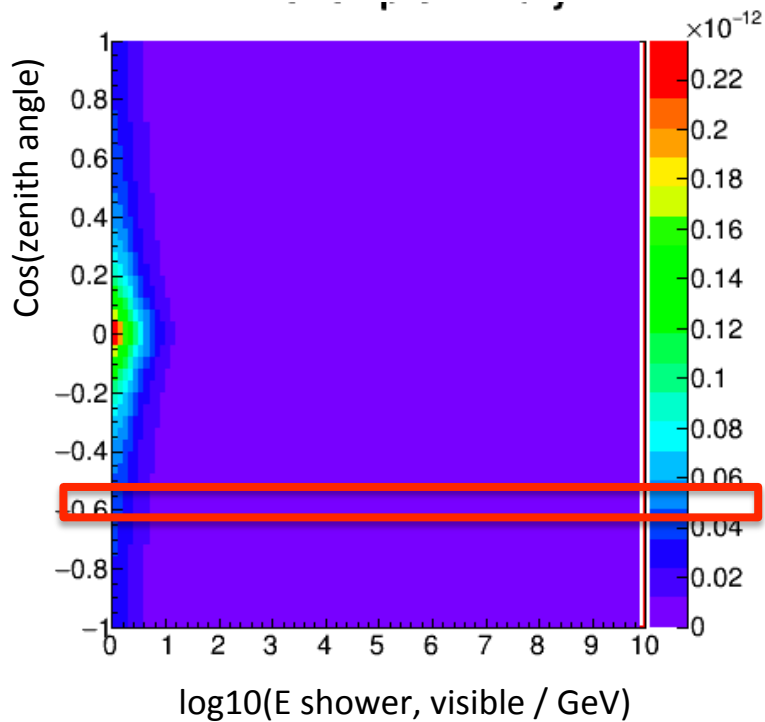
- Neutrino flux @ atmosphere
- Earth propagation
- Interaction probability
- Shower energy (Bjorken-y)
- Visible shower energy
 - How much energy is observable as light?



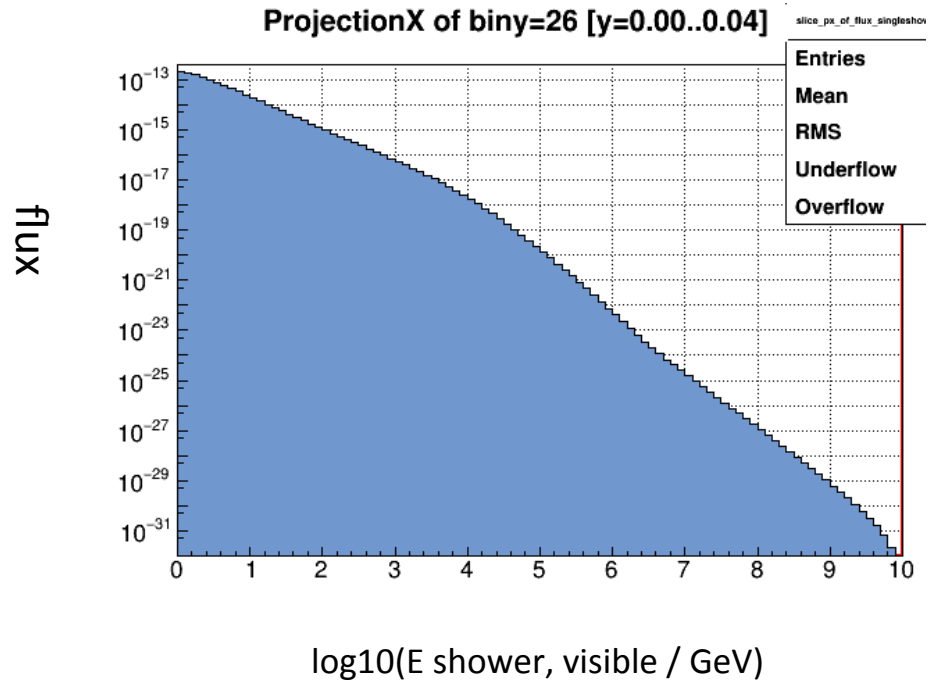
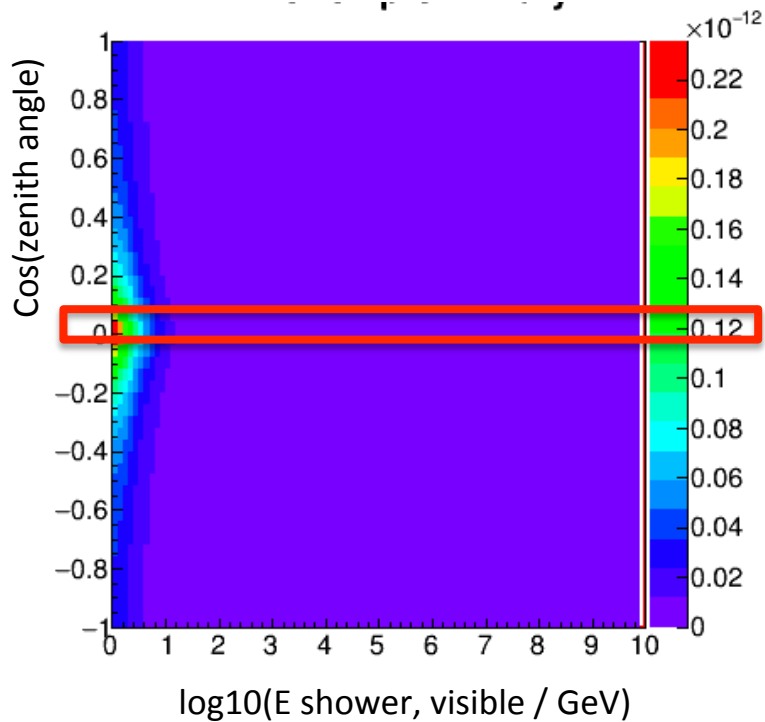
Expected Single Shower Flux



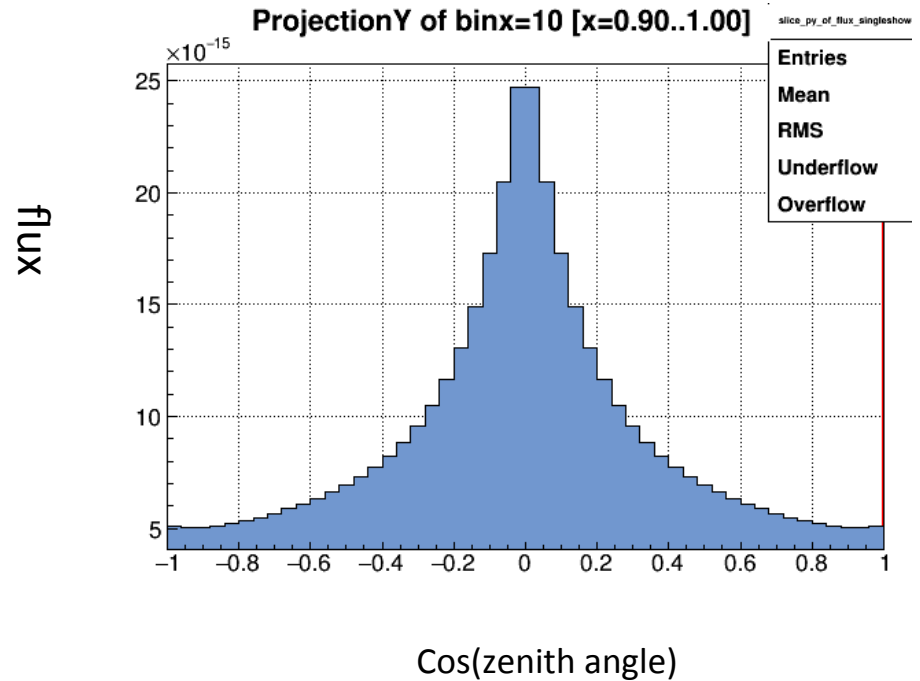
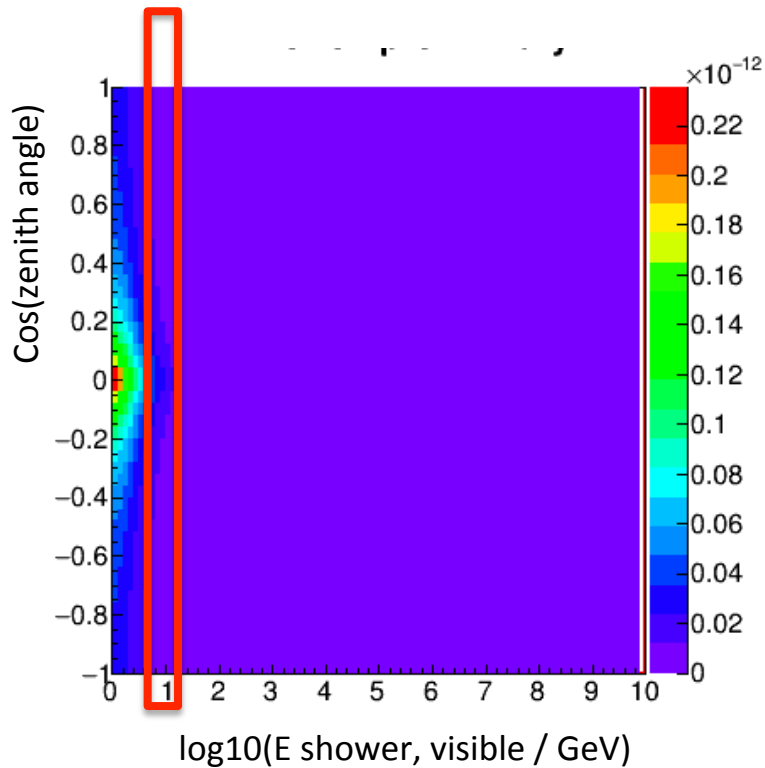
Expected Single Shower Flux



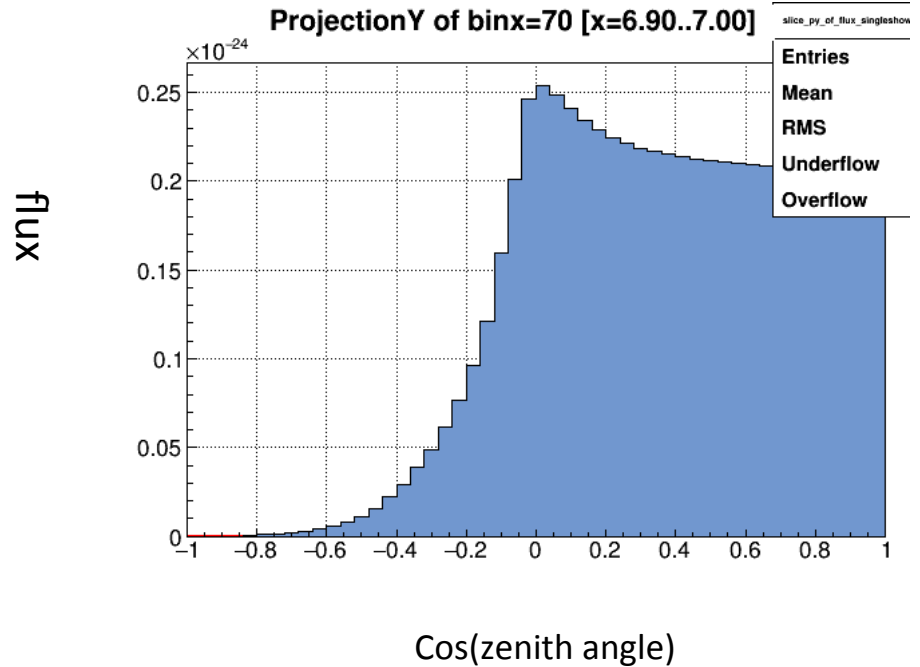
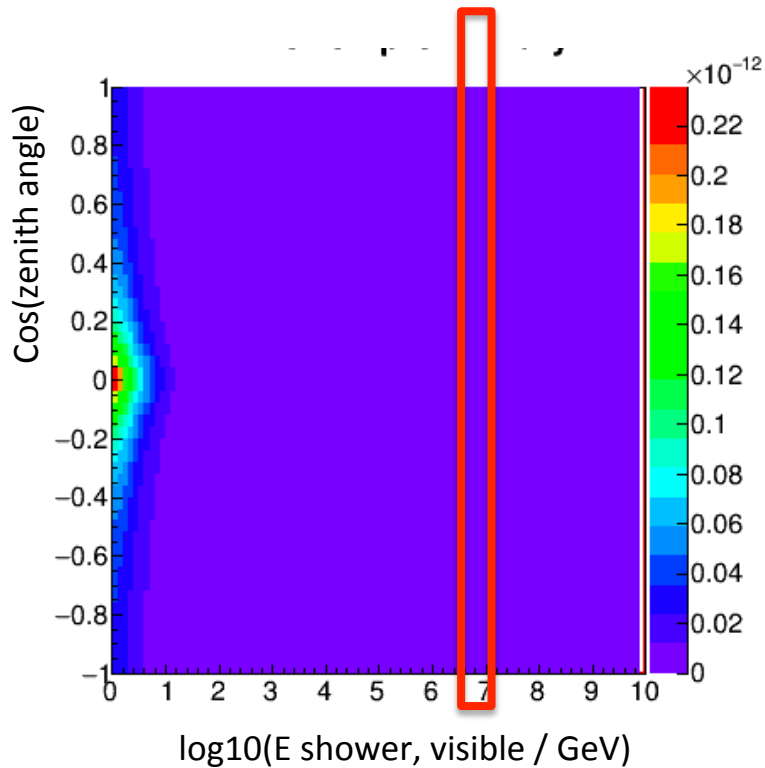
Expected Single Shower Flux



Expected Single Shower Flux



Expected Single Shower Flux



$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

8D Event likelihood landscapes

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

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- (Neutrino) Direction (2D)
- Interaction vertex position (3D)
- Muon energy (1D)
- Visible shower energy (1D)
- Interaction time (1D)

7D

8D Event likelihood landscapes

$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

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- ~~Interaction time (1D)~~

Event Probability

$$P(ev | x) = \prod_{hit\ PMTs} [P_i^{hit} \cdot P_i^{t\ 1st}] \cdot \prod_{non\ hit\ PMTs} [1 - P_i^{hit}]$$

Event Probability

$$P(ev | x) = \prod_{\text{hit PMTs}} [P_i^{\text{hit}} \cdot P_i^{t \text{ 1st}}] \cdot \prod_{\text{non hit PMTs}} [1 - P_i^{\text{hit}}]$$

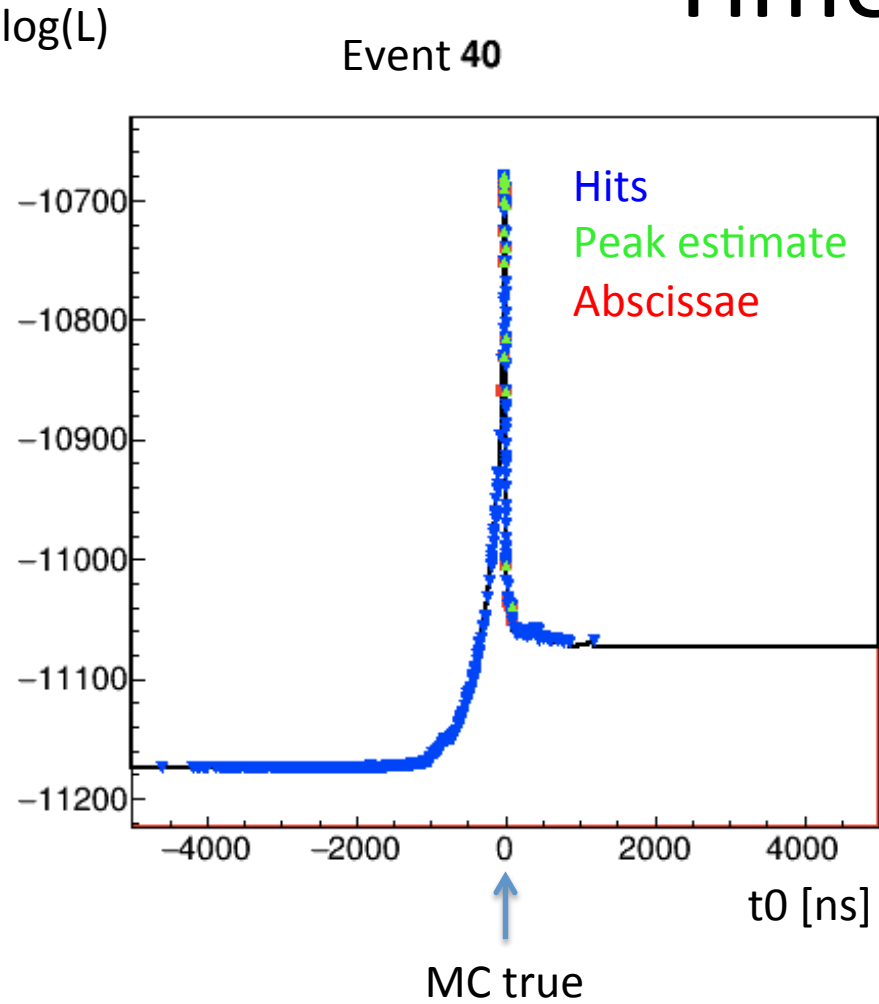
$$P_i^{\text{hit}} = 1 - \exp \left(- \int_{-\infty}^{\infty} \hat{n}_i(t) dt \right)$$

Expected number of photons from 40K and shower/track on PMT i at time t

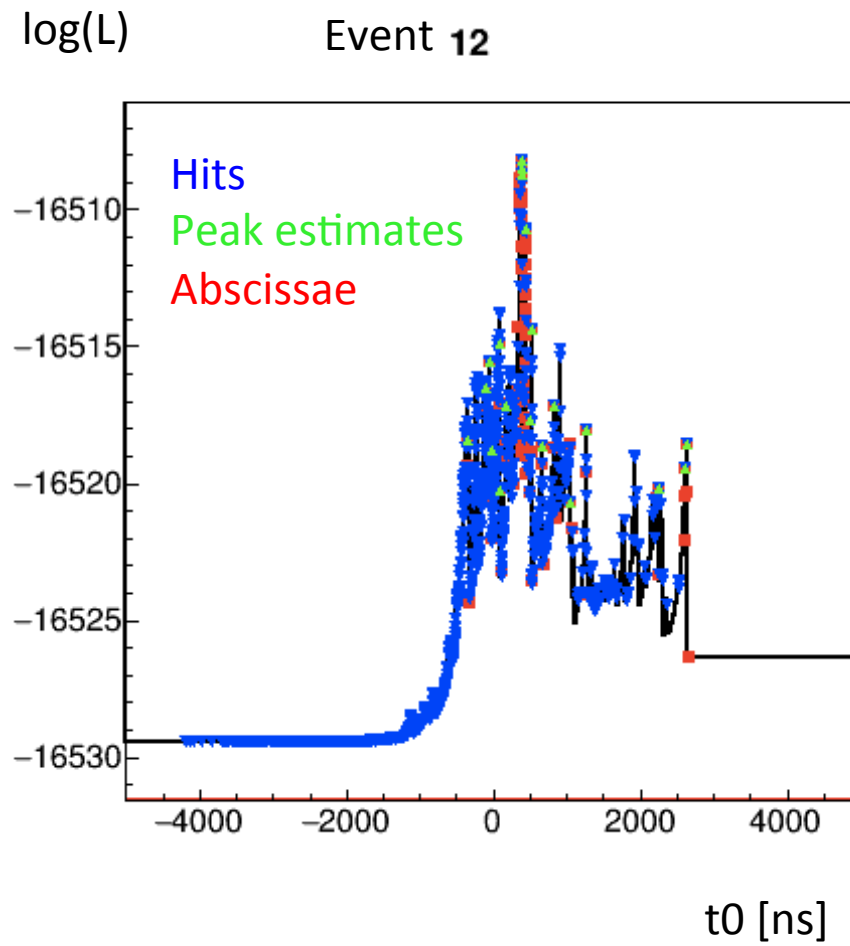
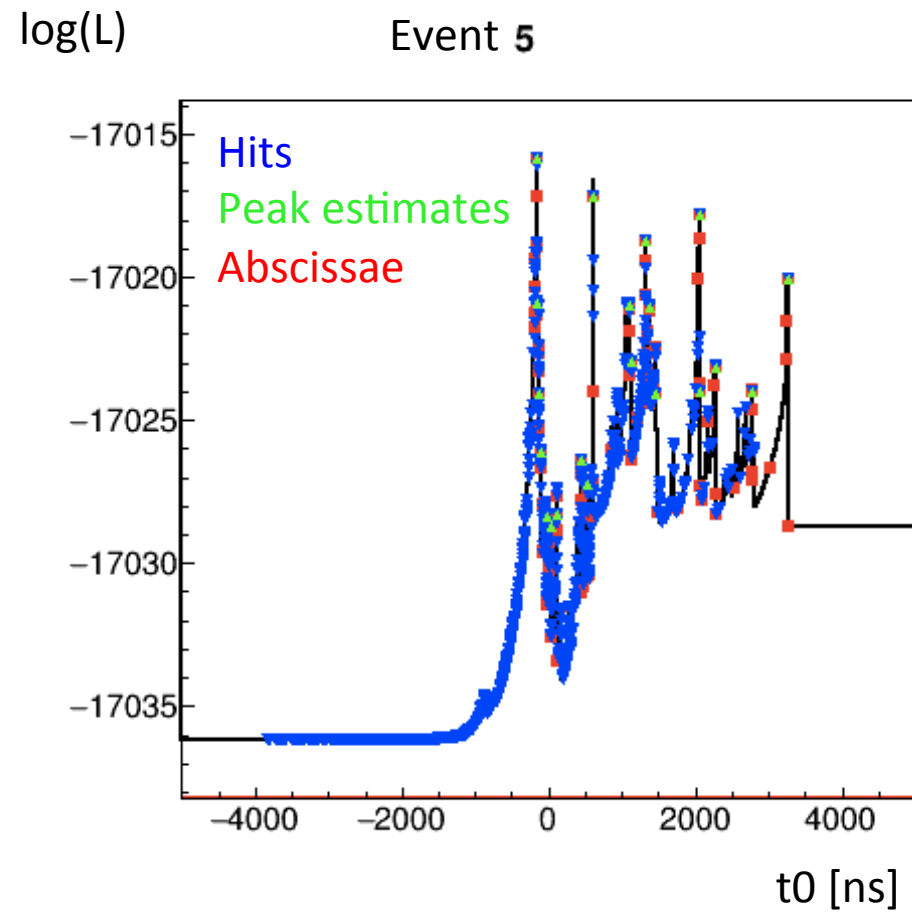
$$P_i^{t \text{ 1st}} \cdot P_i^{\text{hit}} = \underbrace{\exp \left(- \int_{-\infty}^t \hat{n}_i(t) dt \right)}_{\text{P not hit before t}} \cdot \underbrace{(1 - \exp(-\hat{n}_i(t)))}_{\text{P hit at t}}$$

Time Profiles

Event 40



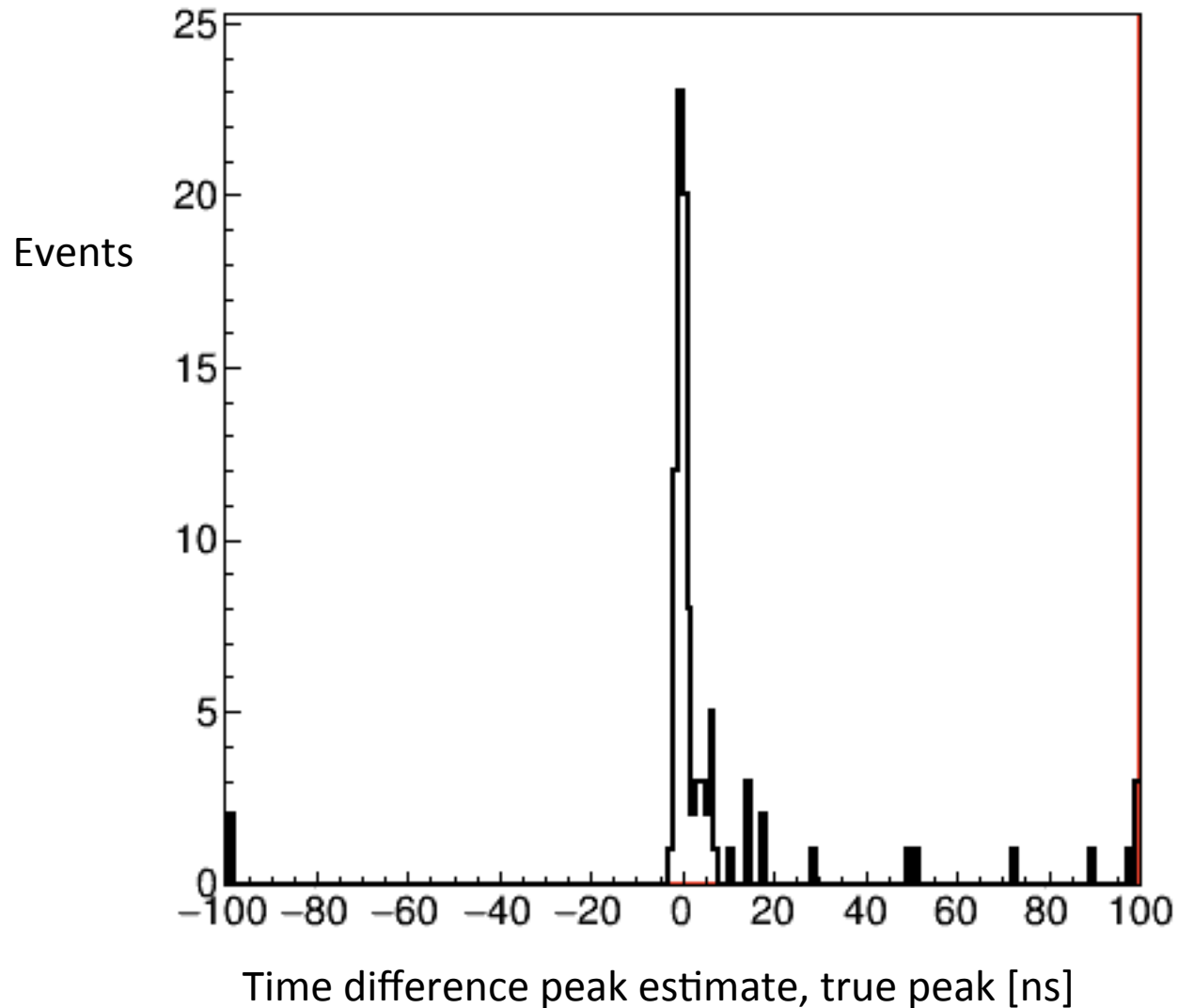
Time Profiles: Difficult ones *



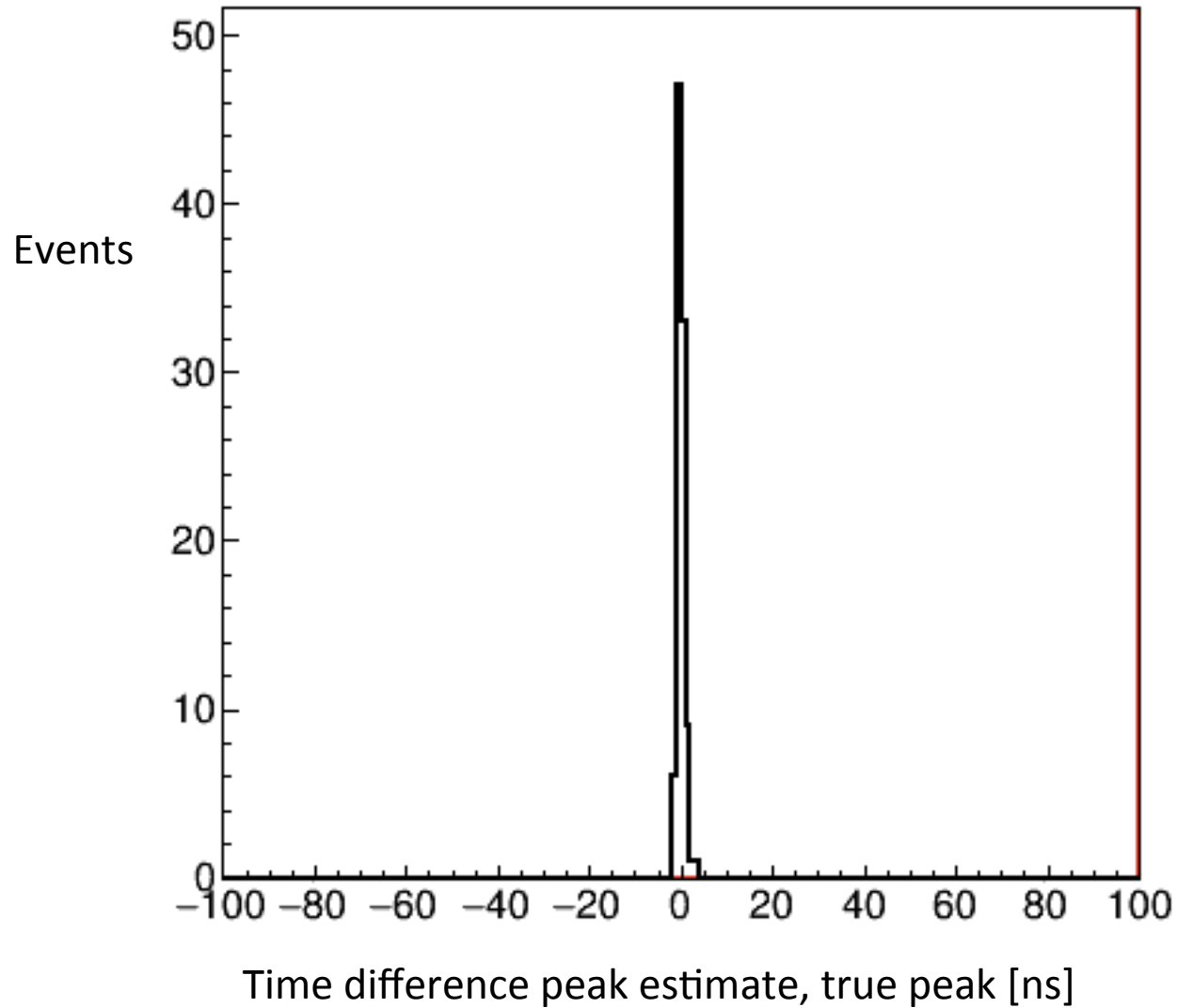
Integration: Peak estimate(s)

- Trajectory known
 - Expected #photons on each PMT known
- Take hits on 20 PMTs with highest #photons
 - 20 peak estimates
- Merge overlapping hits (10 ns)

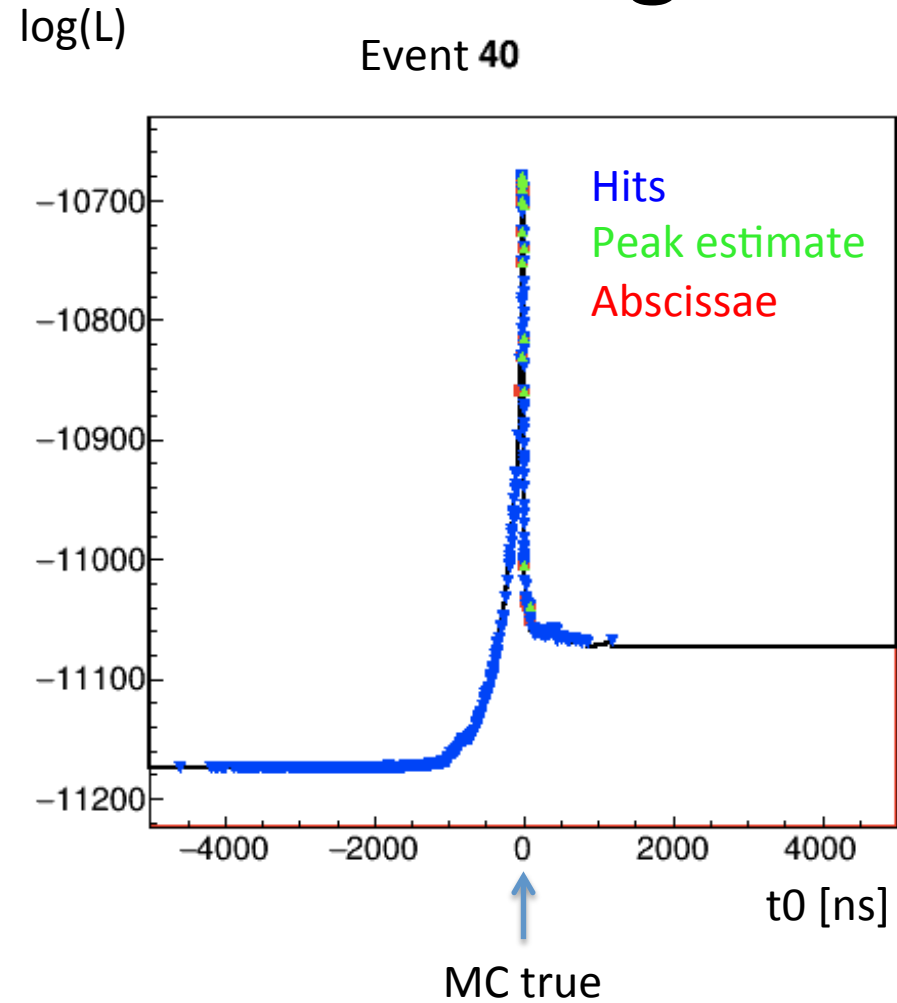
Integration: Peak estimate(s)*



Integration: Peak estimate(s)



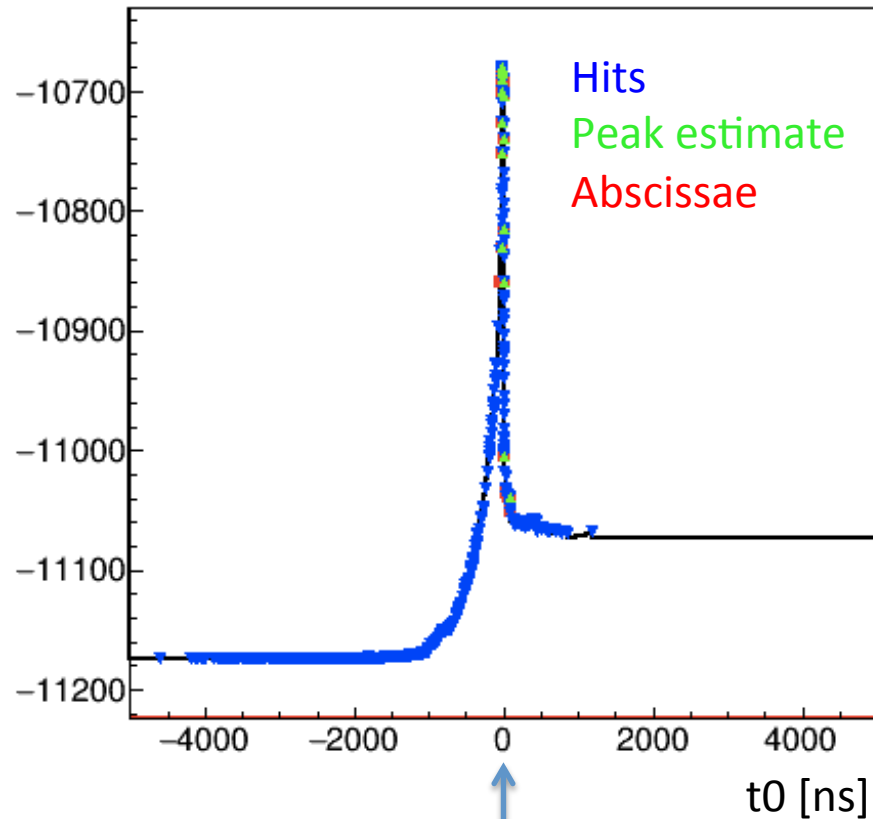
Integration: Abscissae



Integration: Abscissae

log(L)

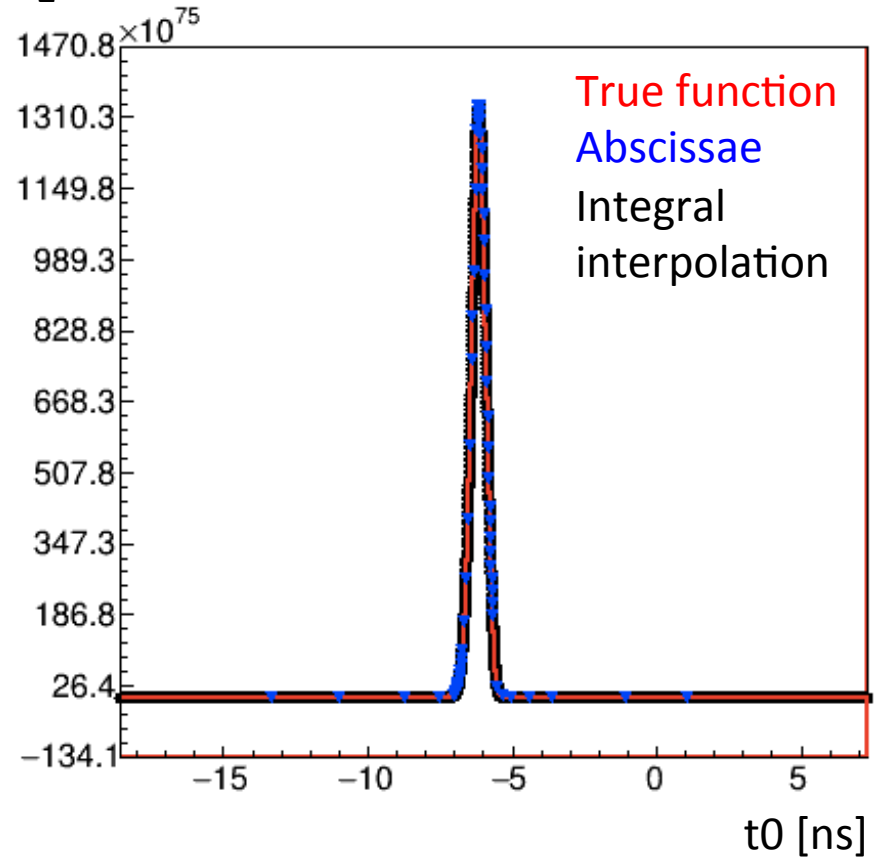
Event 40



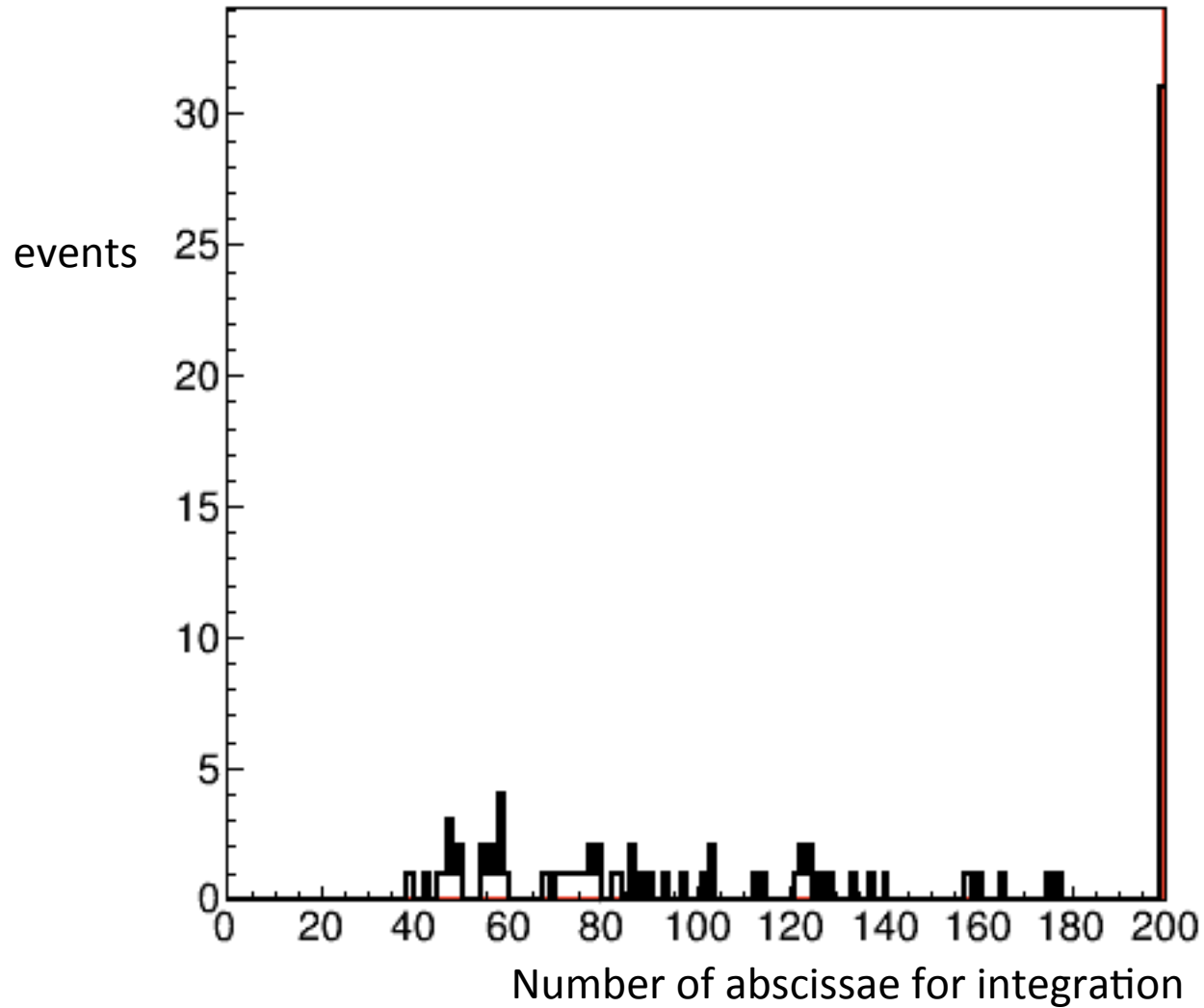
MC true

L

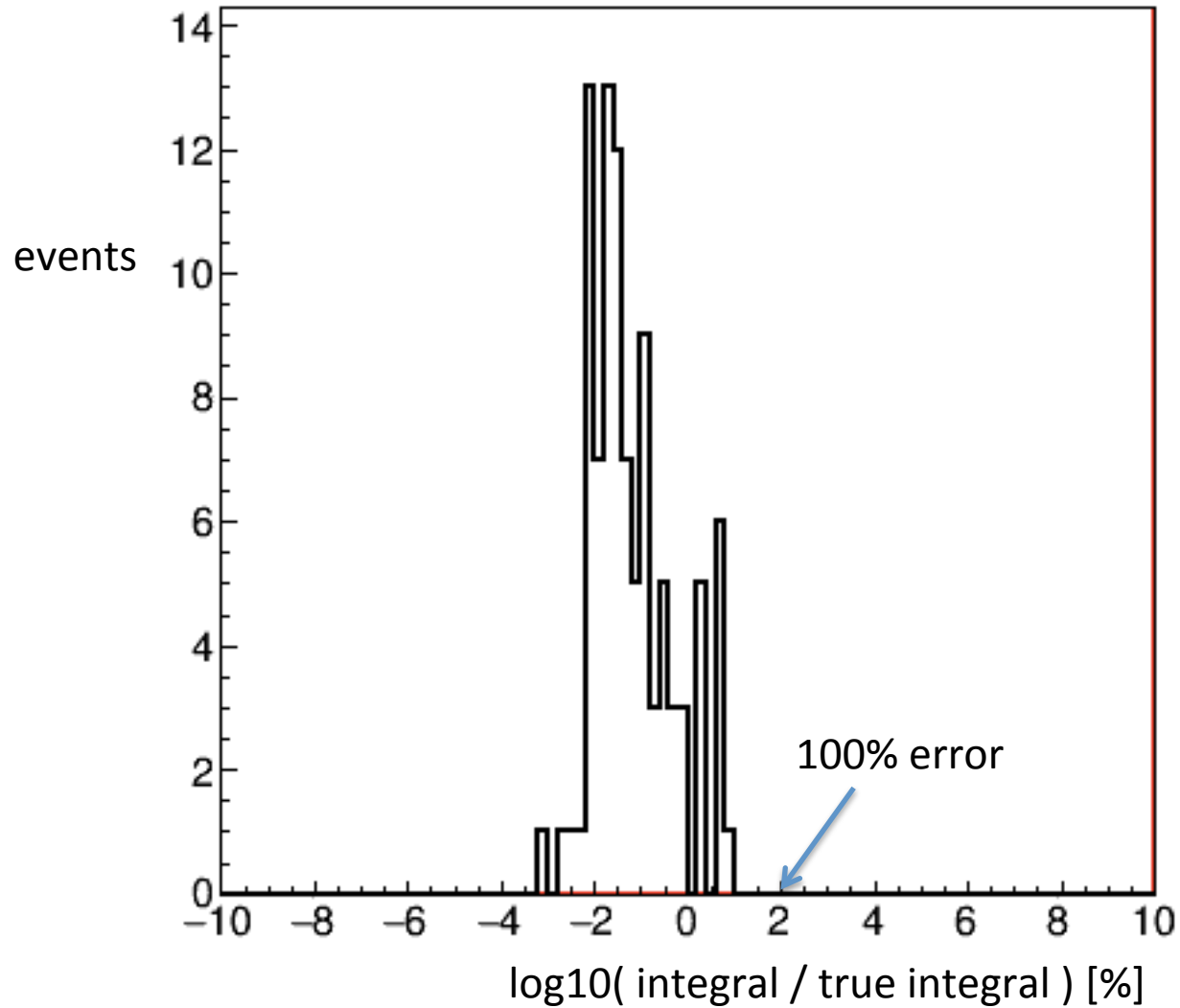
Event 40



Integration: Abscissae



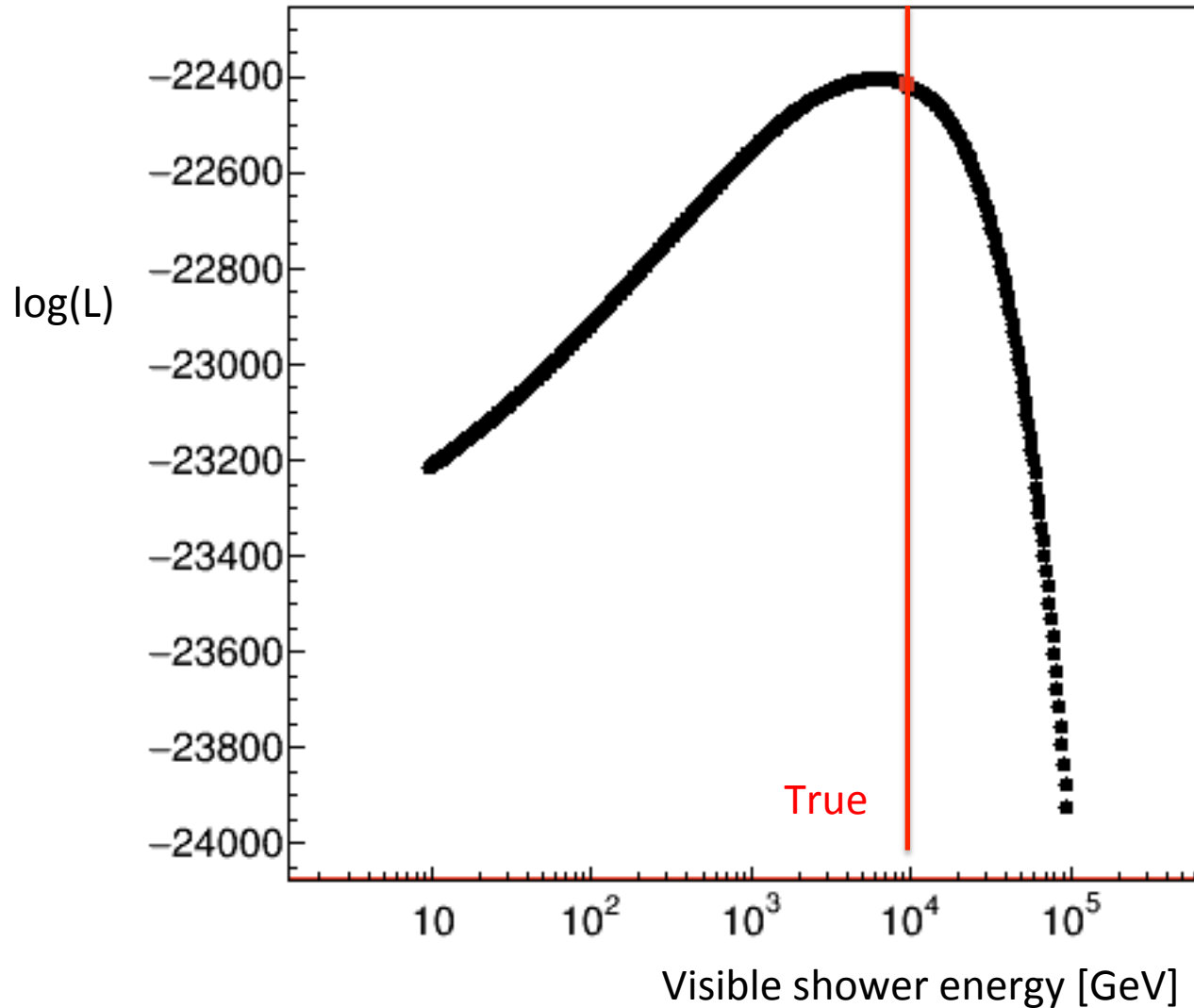
Integration: Abscissae



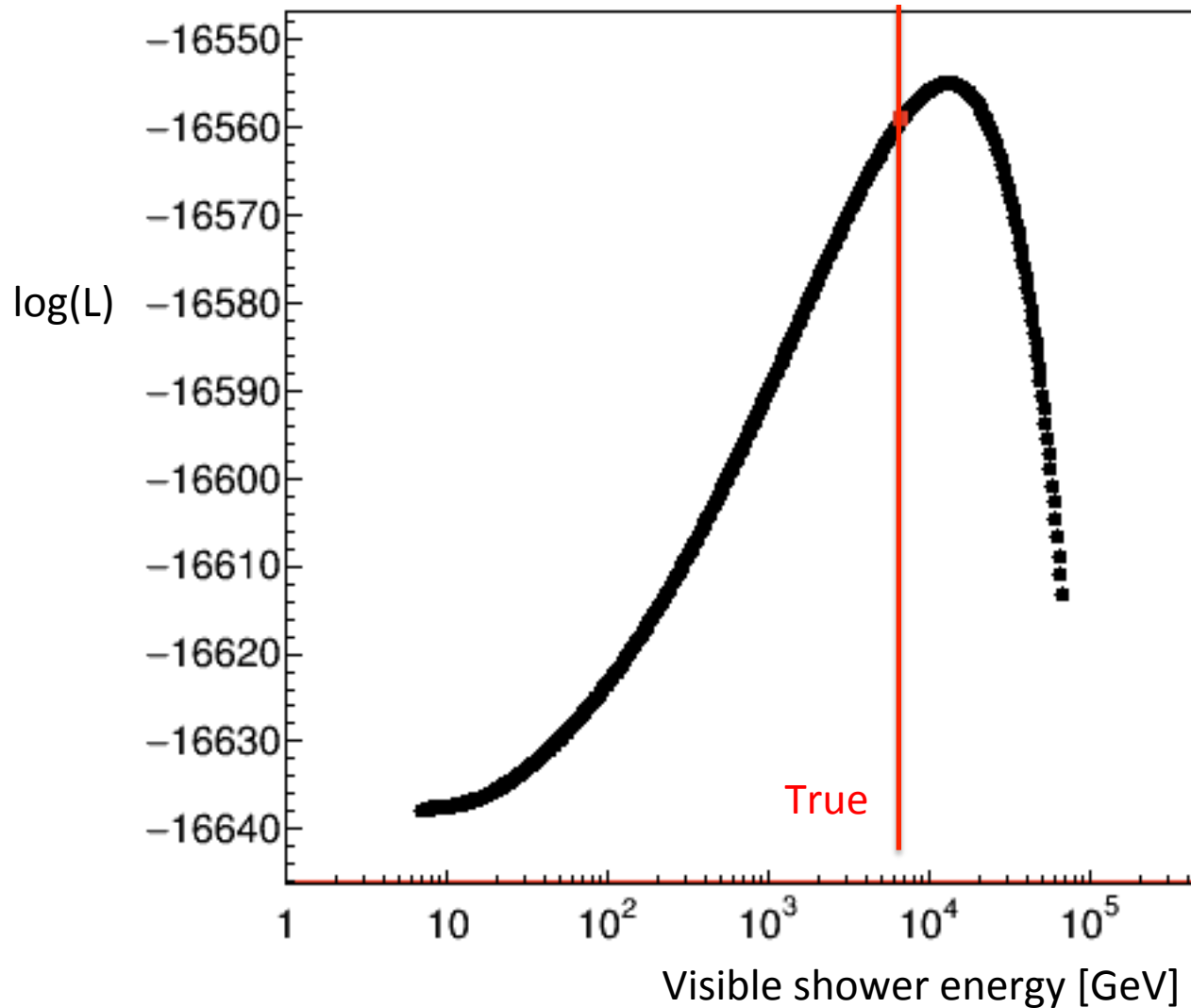
Procedure

- Given trajectory (direction + vertex)
- Fast Energy estimate
 - hit/non-hit info only
 - Define ‘interesting’ energy grid
- Determine t_0 integral abscissae
 - For each t_0 point: evaluate $\log L$ for all energies
- Integrate over t_0 for all energies
- Time-integrated energy likelihood landscape

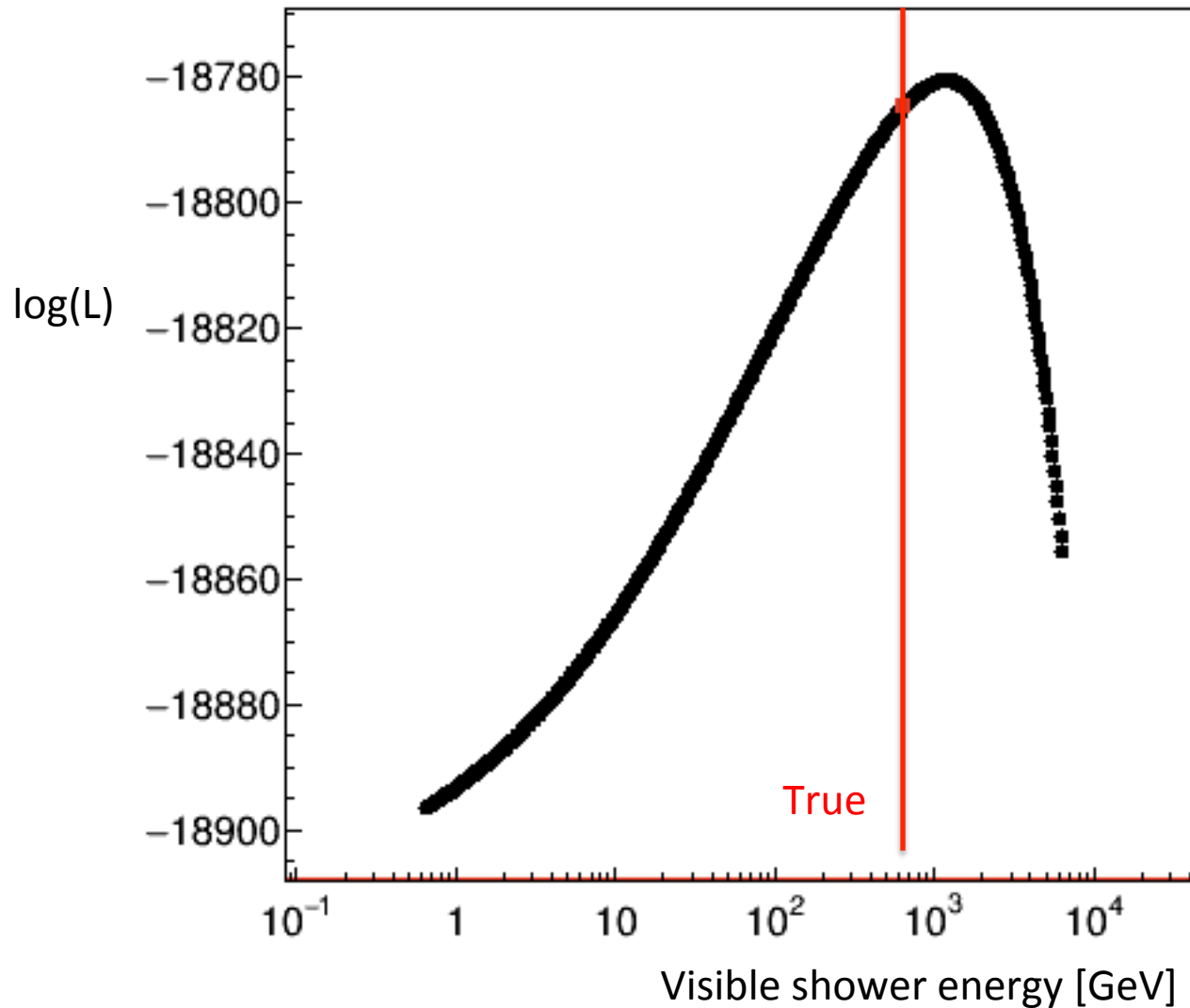
Energy Profile



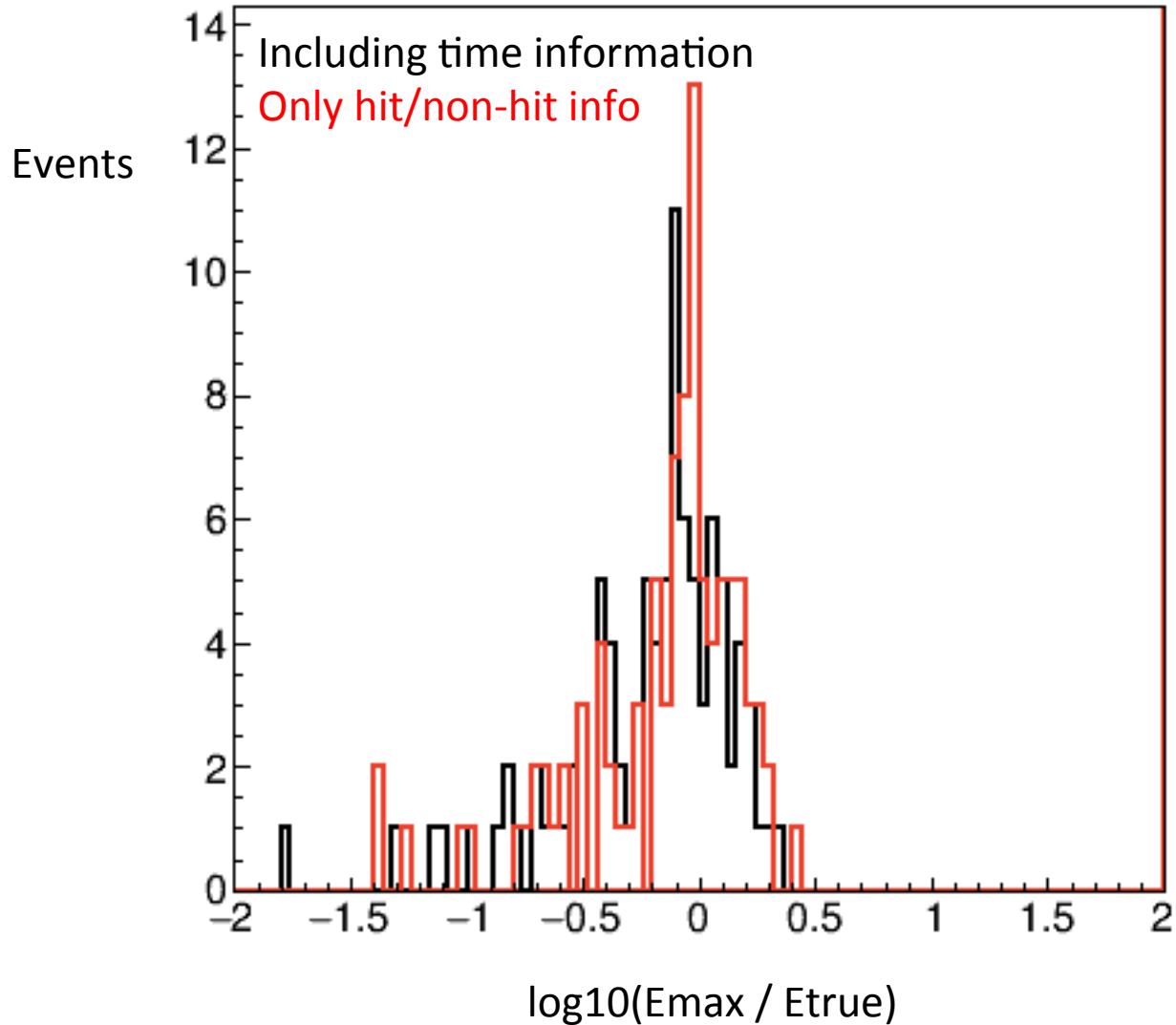
Energy Profile



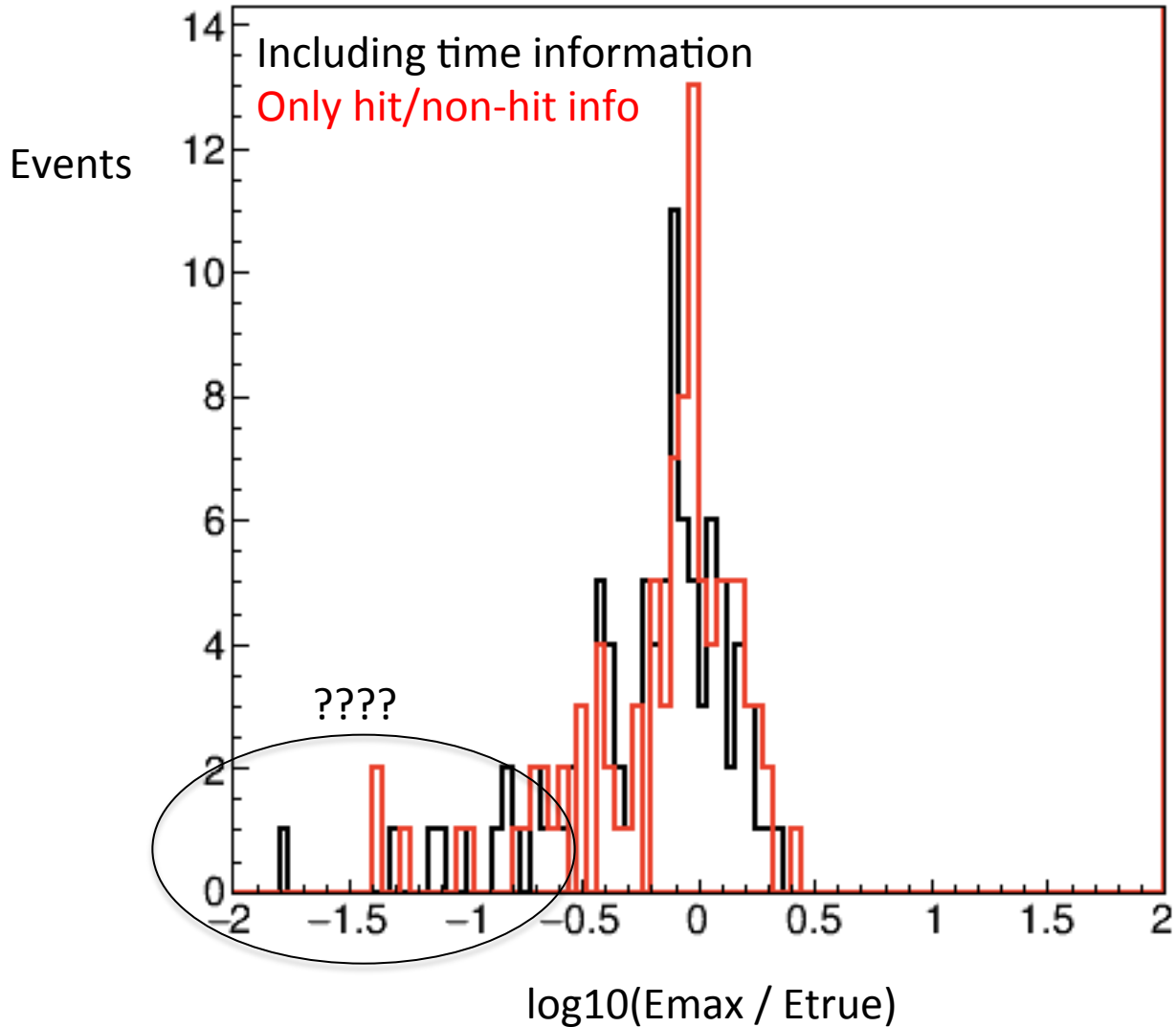
Energy Profile



Energy profile



Energy profile



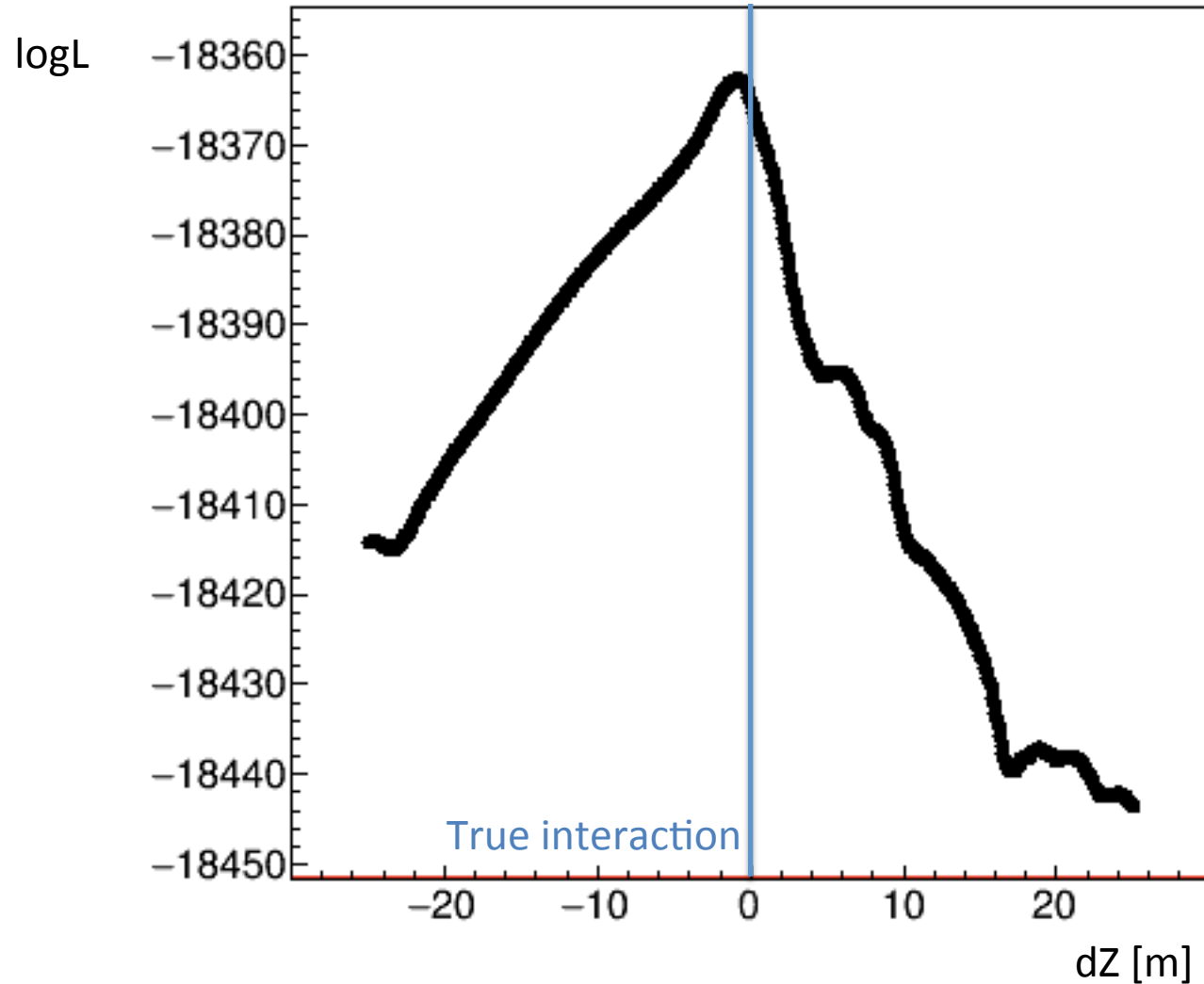
7D

8D Event likelihood landscapes

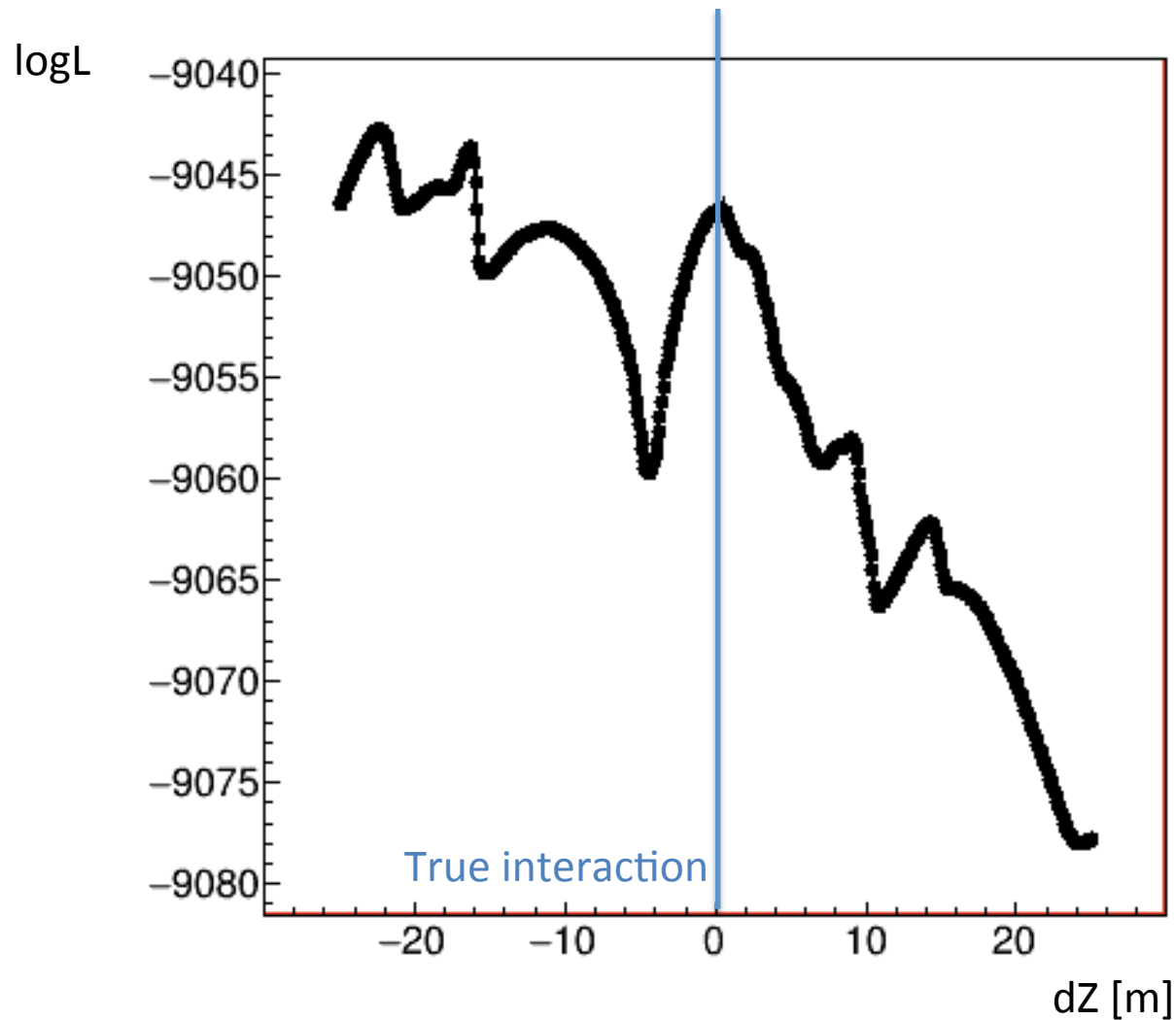
$$P(\text{data}|H) = \sum_i \left[\log \int P(\text{ev}_i | x_{\text{true}}) \cdot P^{\text{det}}(x_{\text{true}}) \cdot \mu(x_{\text{true}} | H) dx_{\text{true}} \right] - \mu^{\text{tot}}(H)$$

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- ~~Interaction time (1D)~~

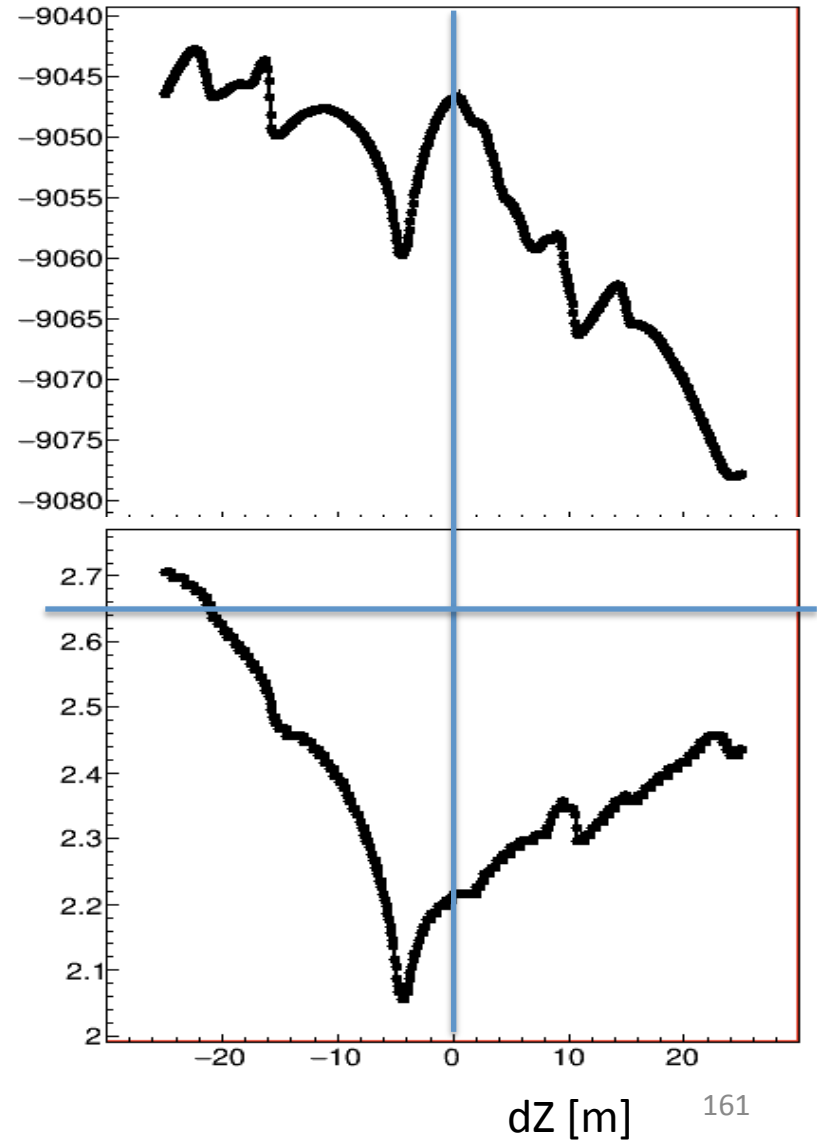
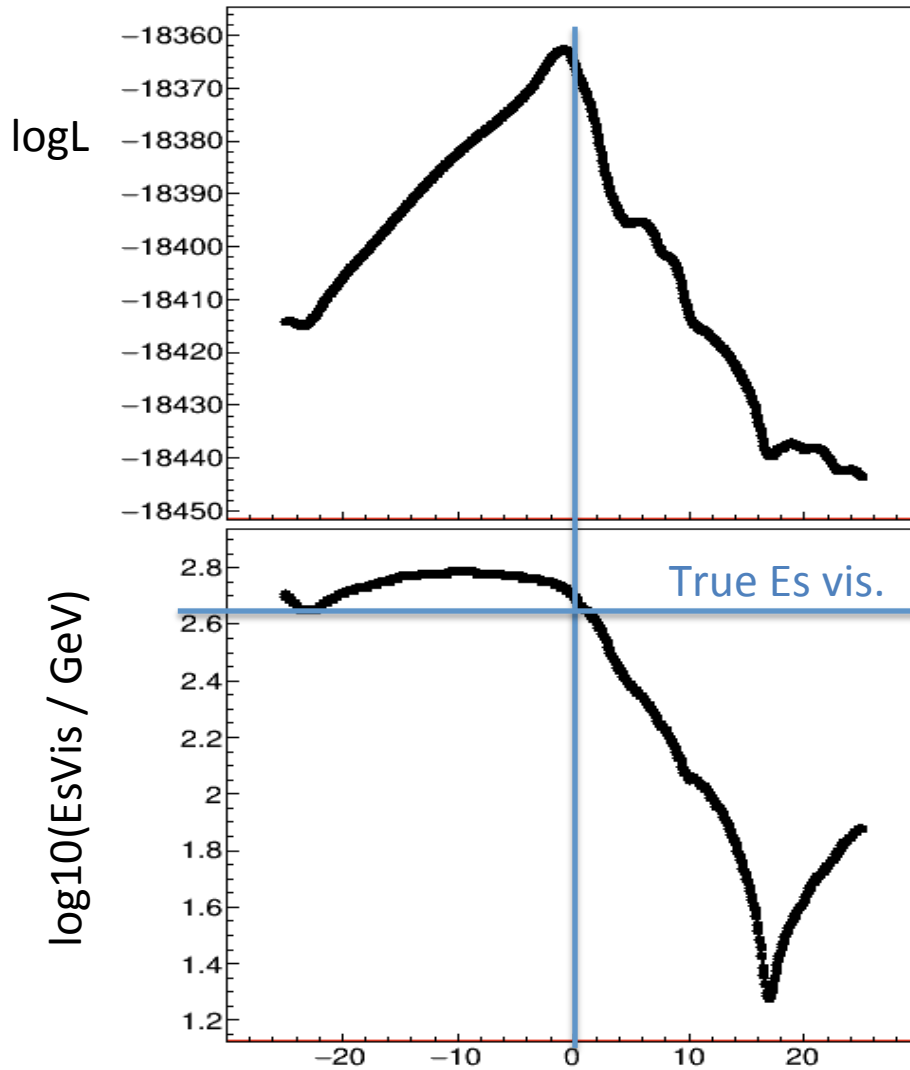
Shower longitudinal position



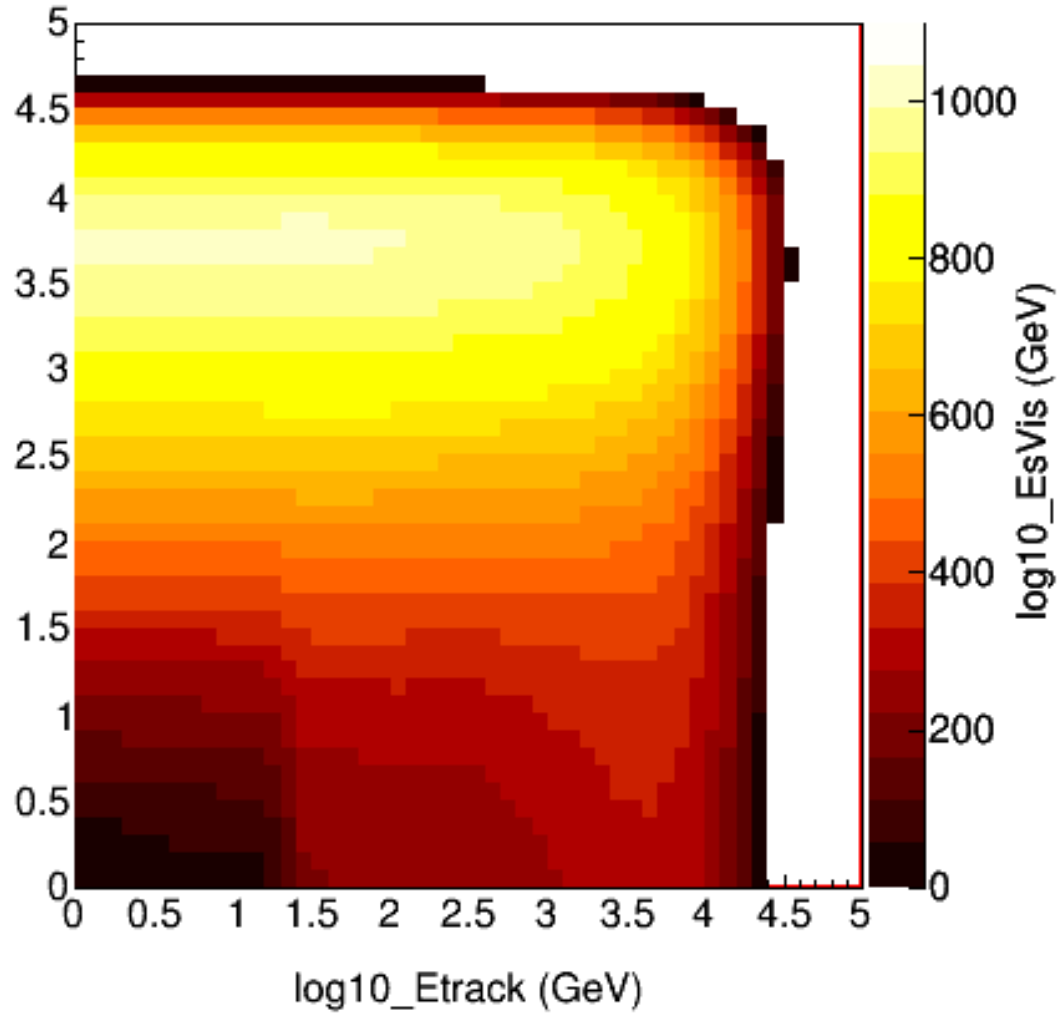
Shower longitudinal position



Best fit Shower Energy

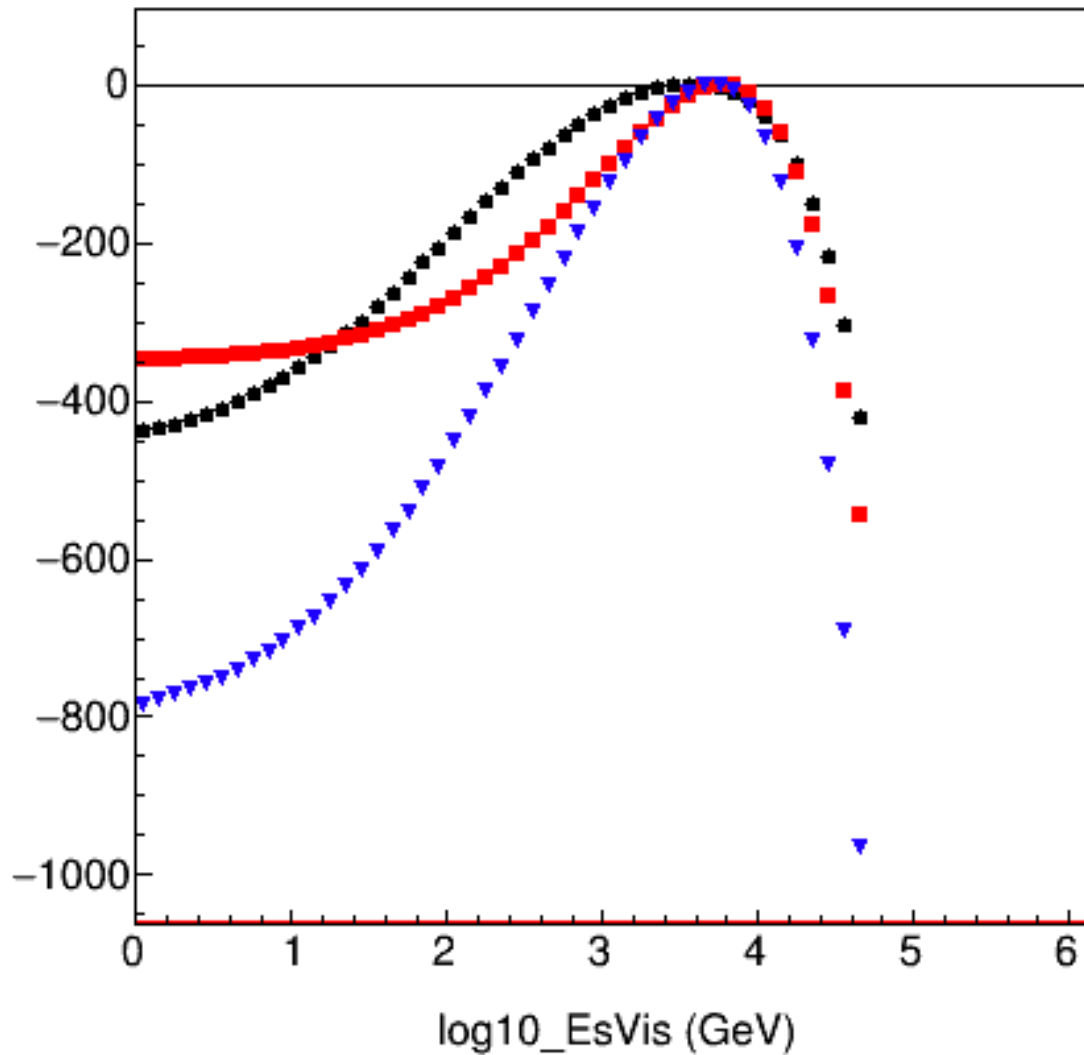


429, all info.



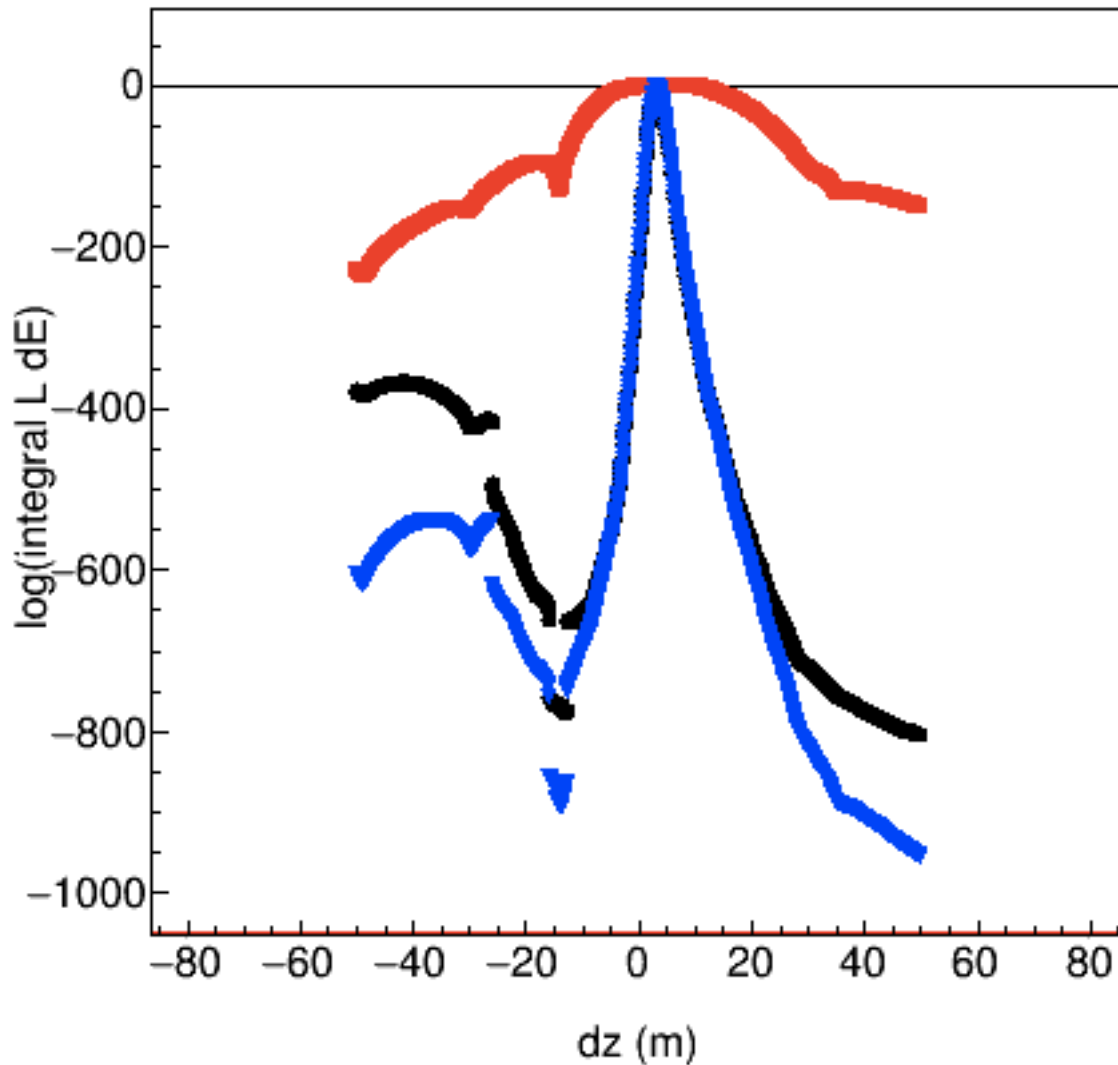
Direction: True
Perp. Pos: Maximum
Long. Pos: Maximum
Etrack:
EsVis:
Time: Integrated

429, black: time, red: ampl, blue: all



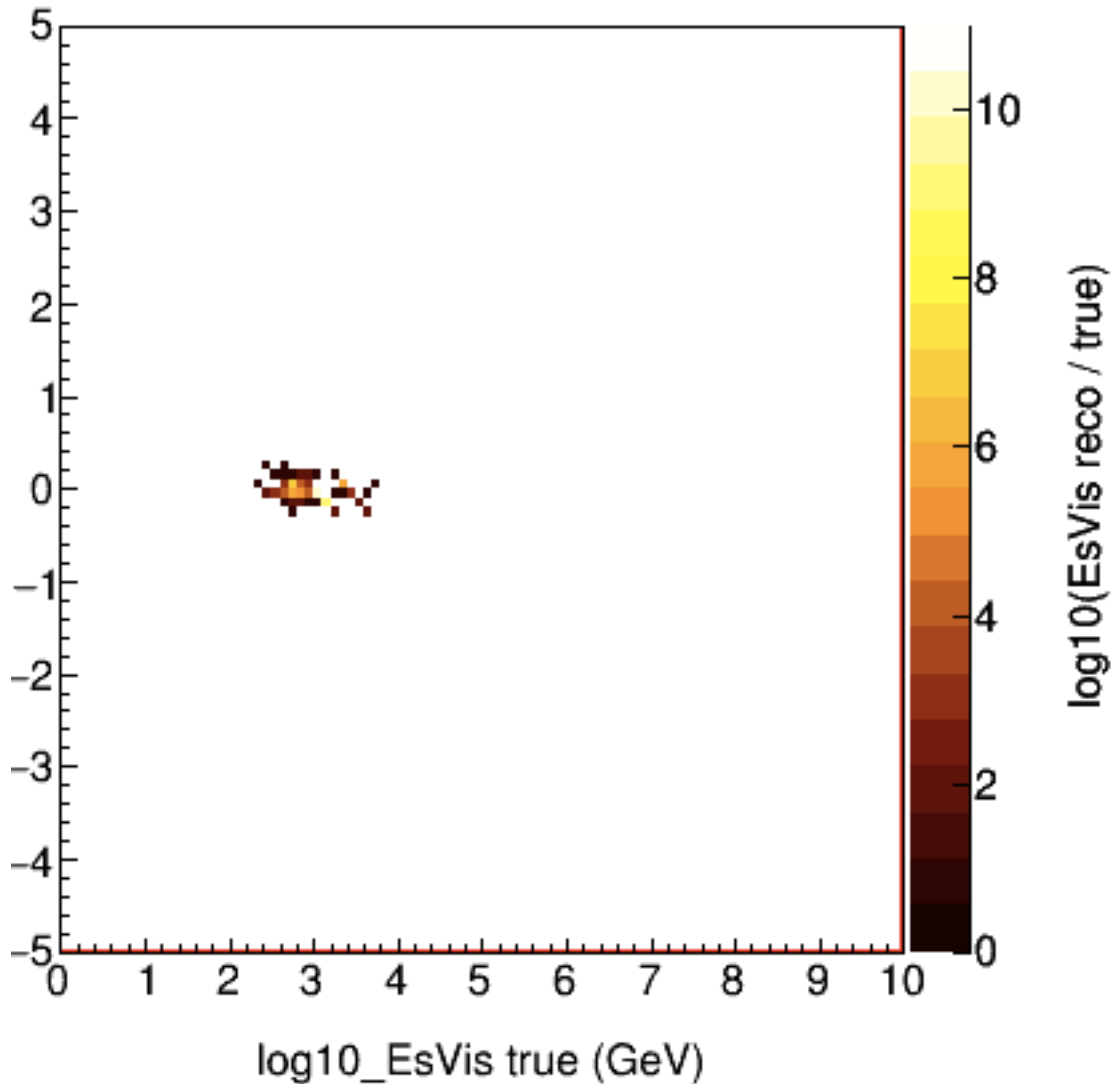
Direction: True
Perp. Pos: Maximum
Long. Pos: Maximum
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EsVis:
Time: Integrated

429, black: time, red: ampl, blue: all



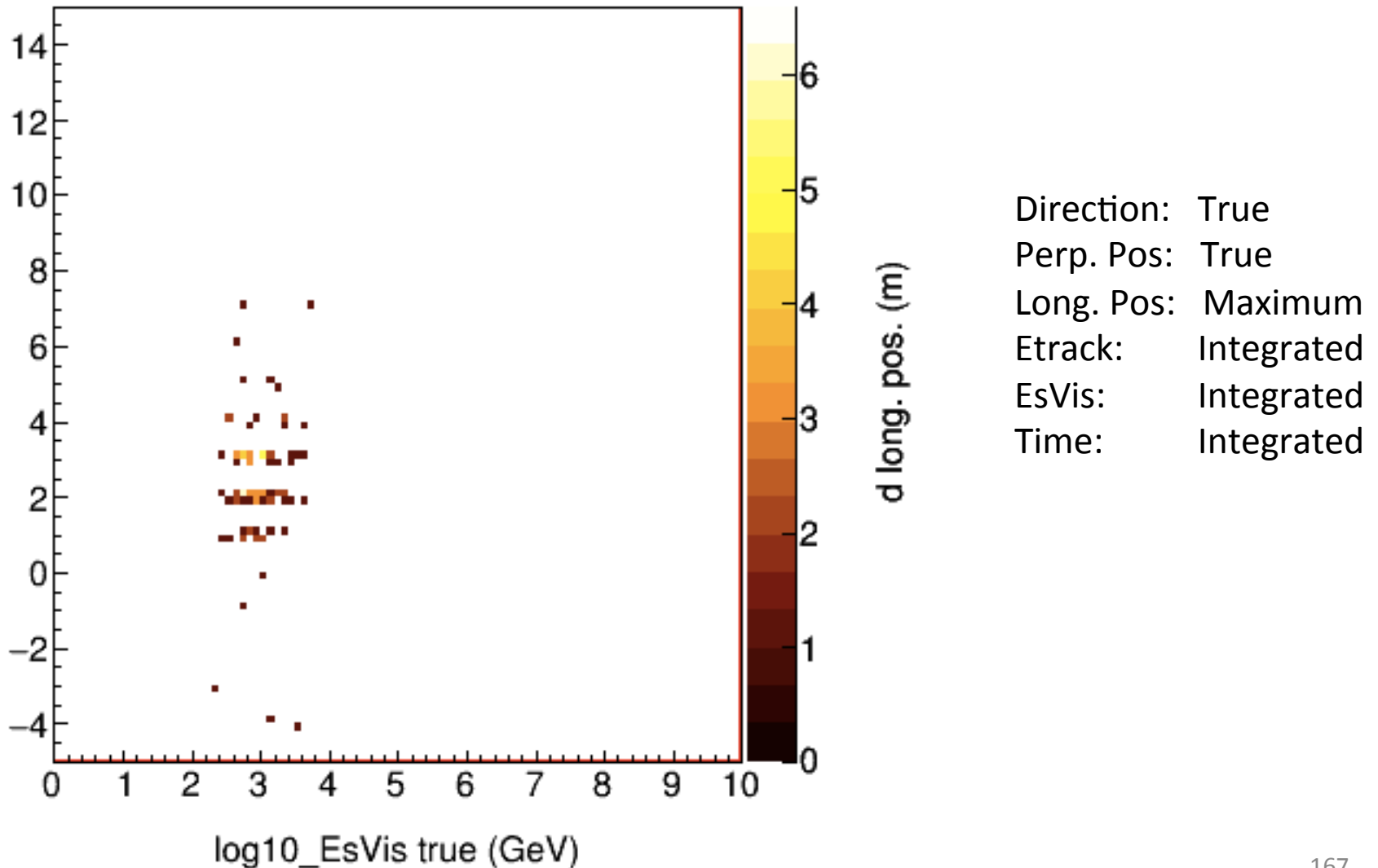
Direction: True
Perp. Pos: Maximum
Long. Pos:
Etrack: Integrated
EsVis: Integrated
Time: Integrated

Energy 'reconstruction'



Direction: True
Perp. Pos: True
Long. Pos: Maximum
Etrack: Maximum
EsVis: Maximum
Time: Integrated

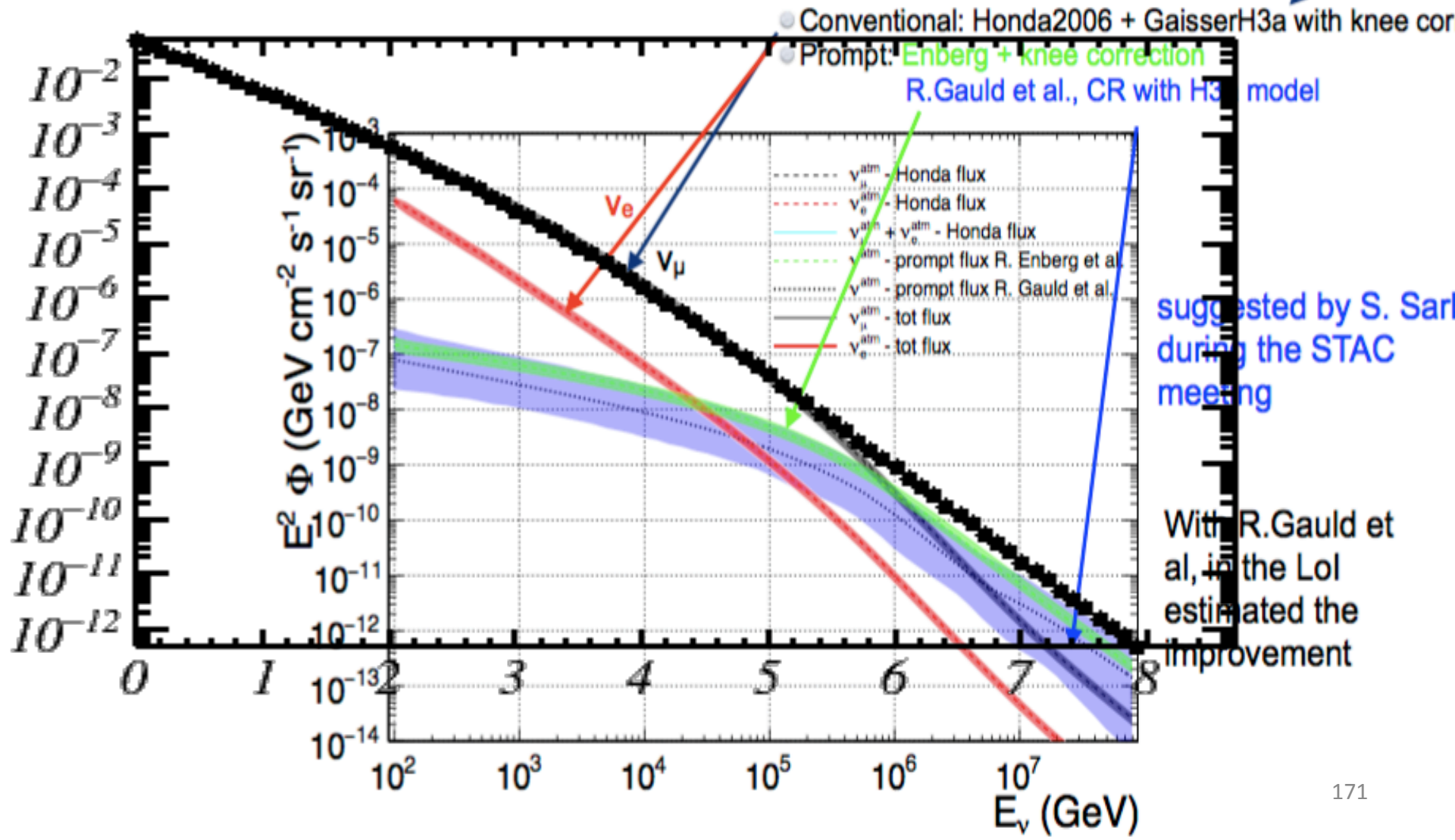
Long. Position 'reconstruction'



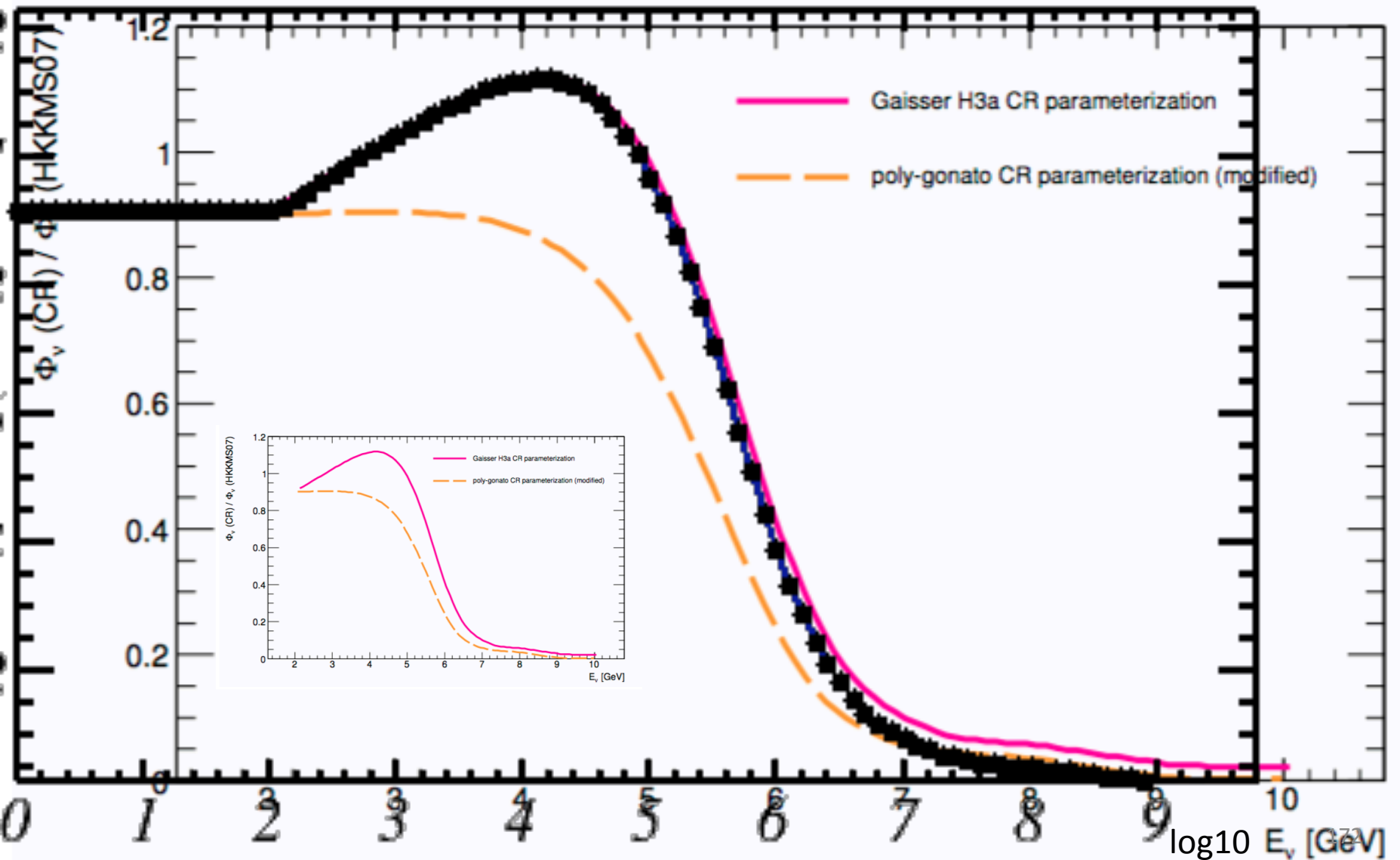
Backup

Honda extrapolated

T. Gaisser 2012

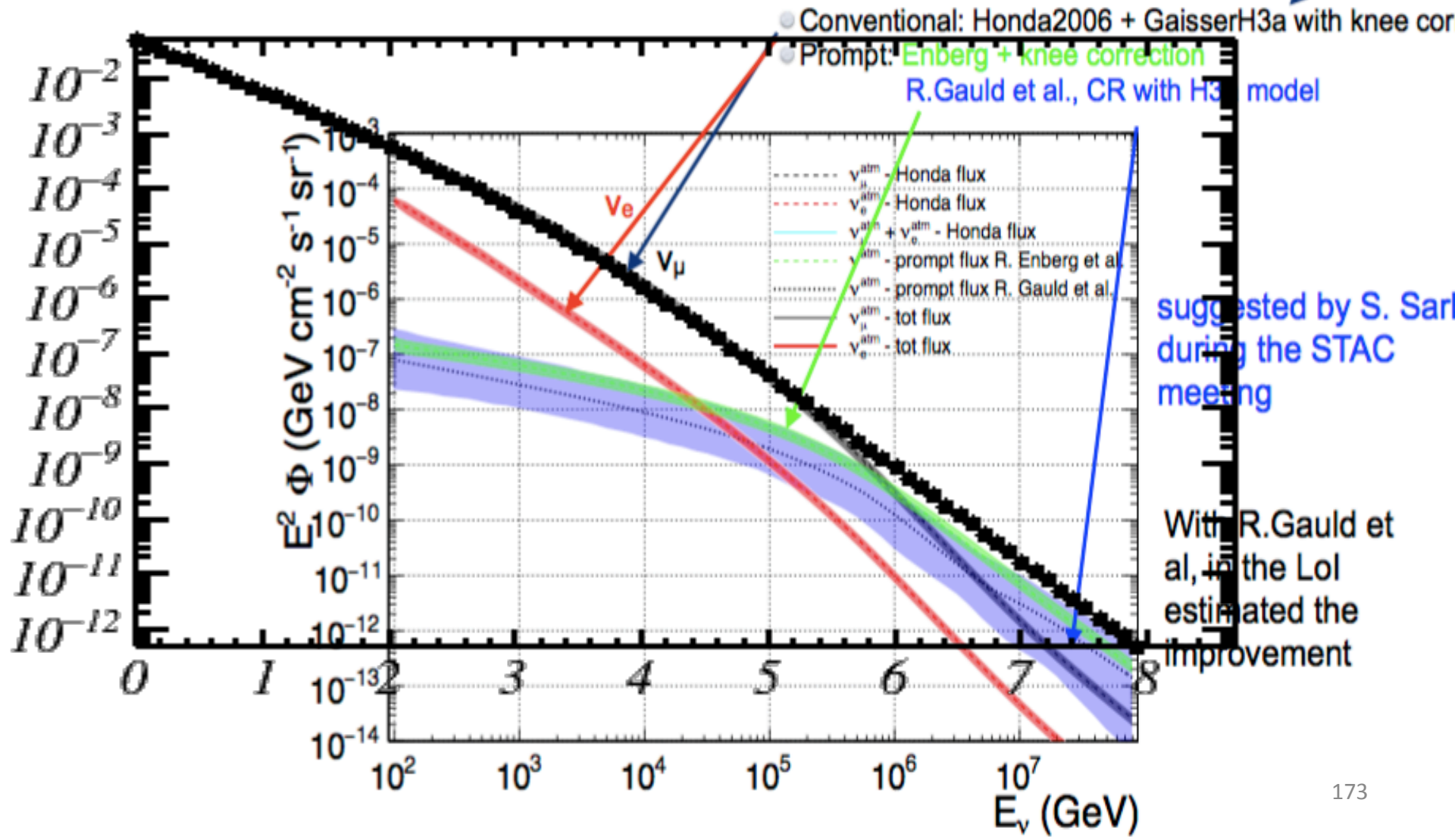


Knee Correction (Gaisser H3a)



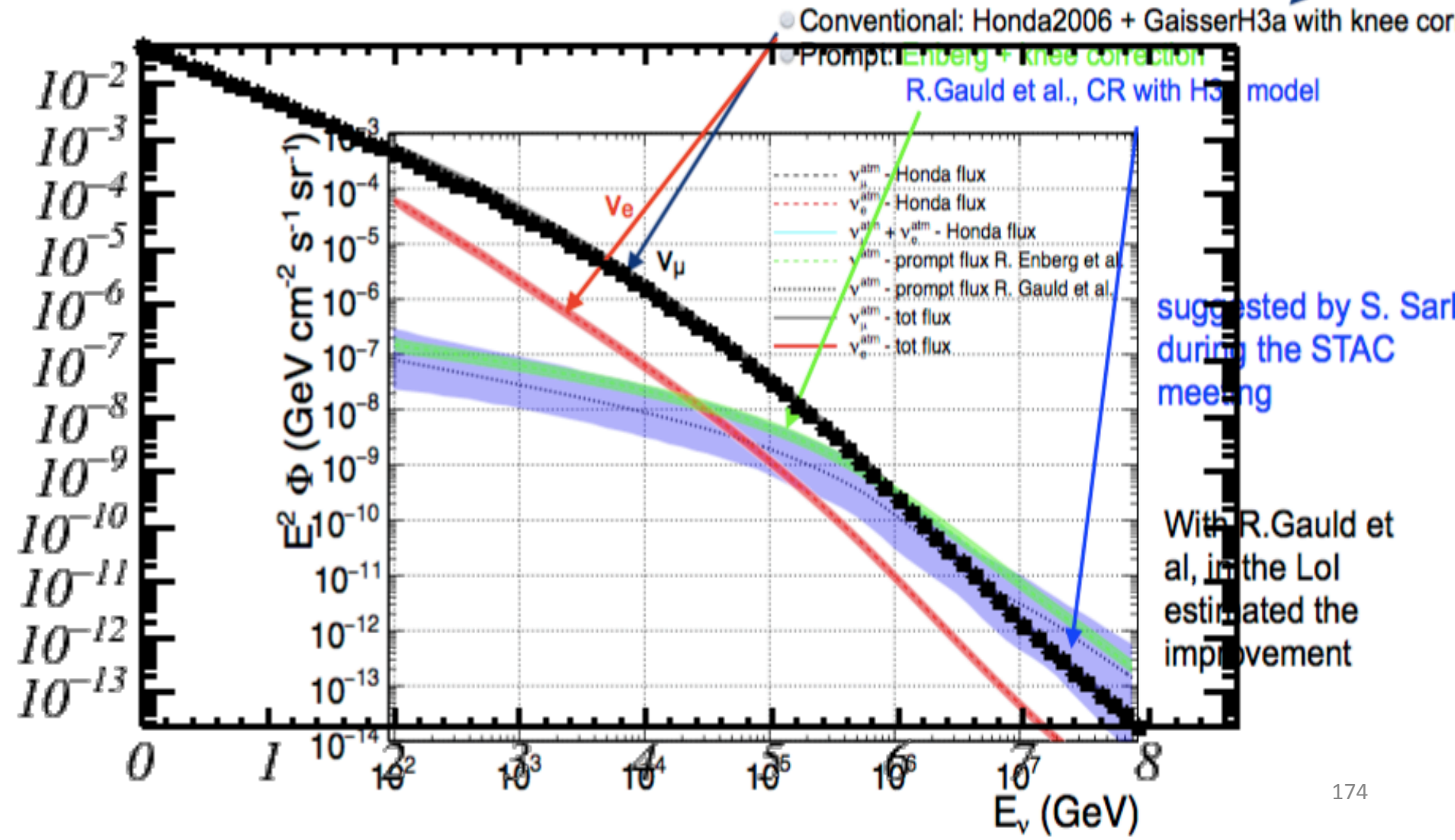
Honda extrapolated

T. Gaisser 2012

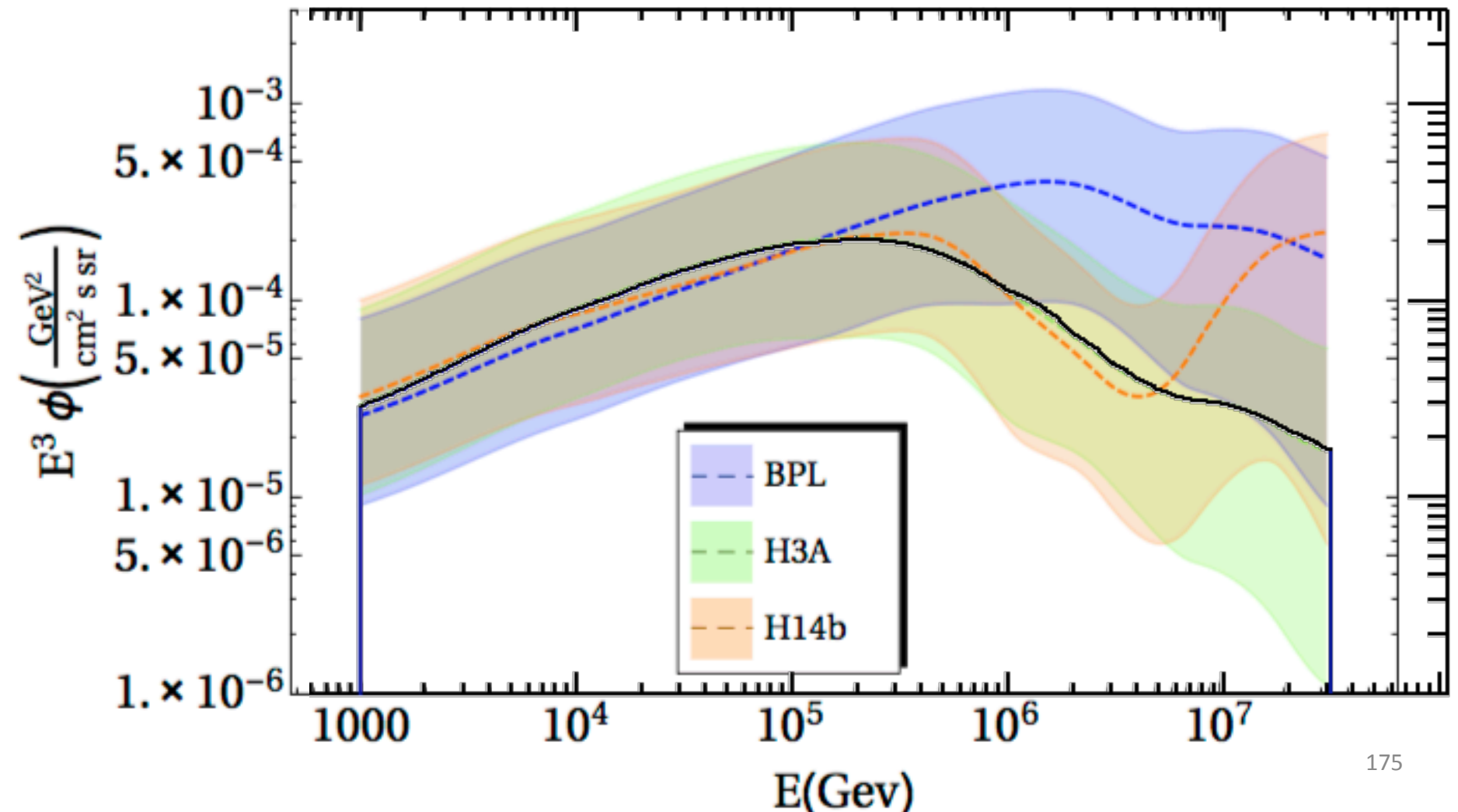


Honda extrapolated + knee correction

T. Gaisser 2012

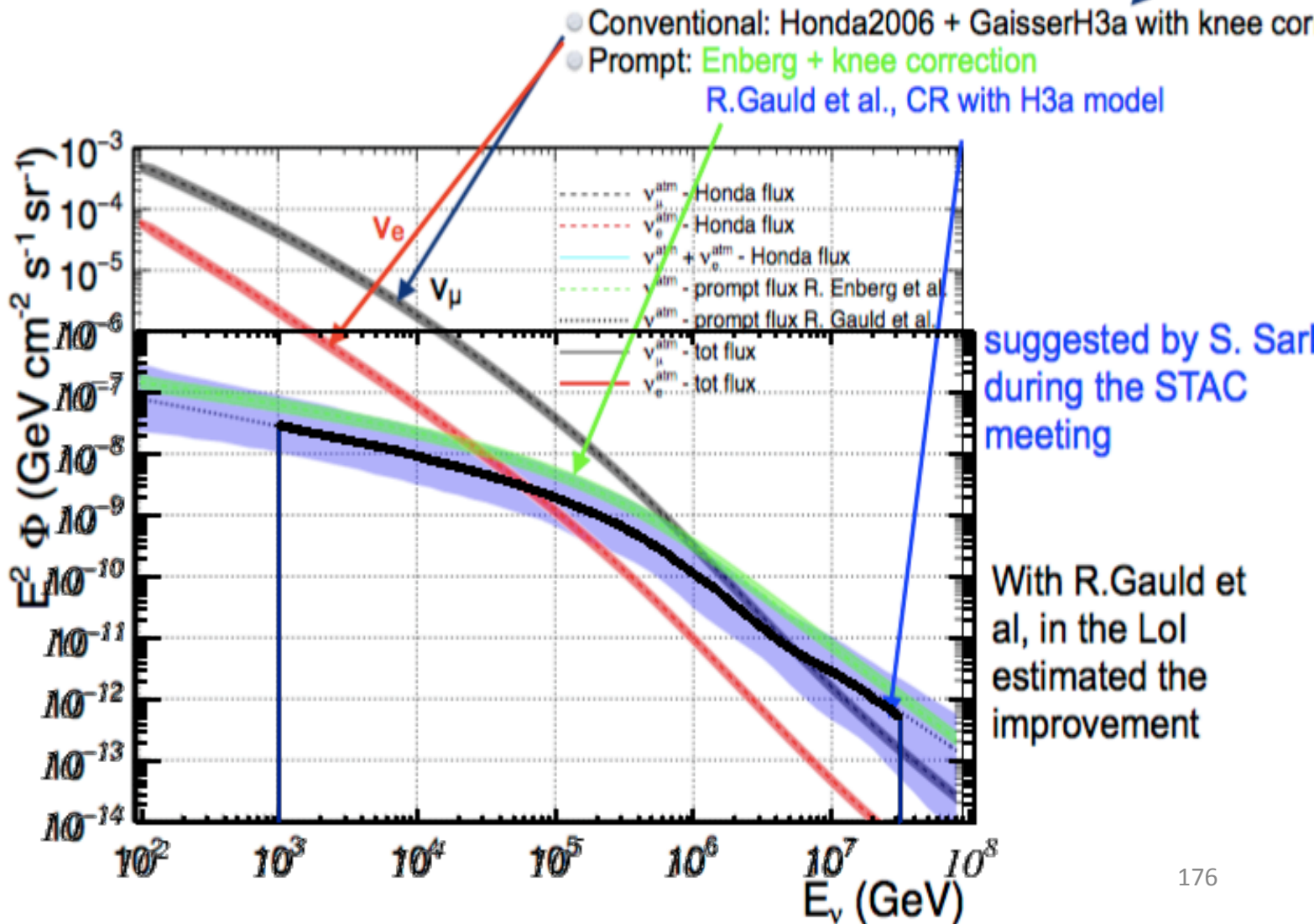


Prompt: Gauld Flux (2016)



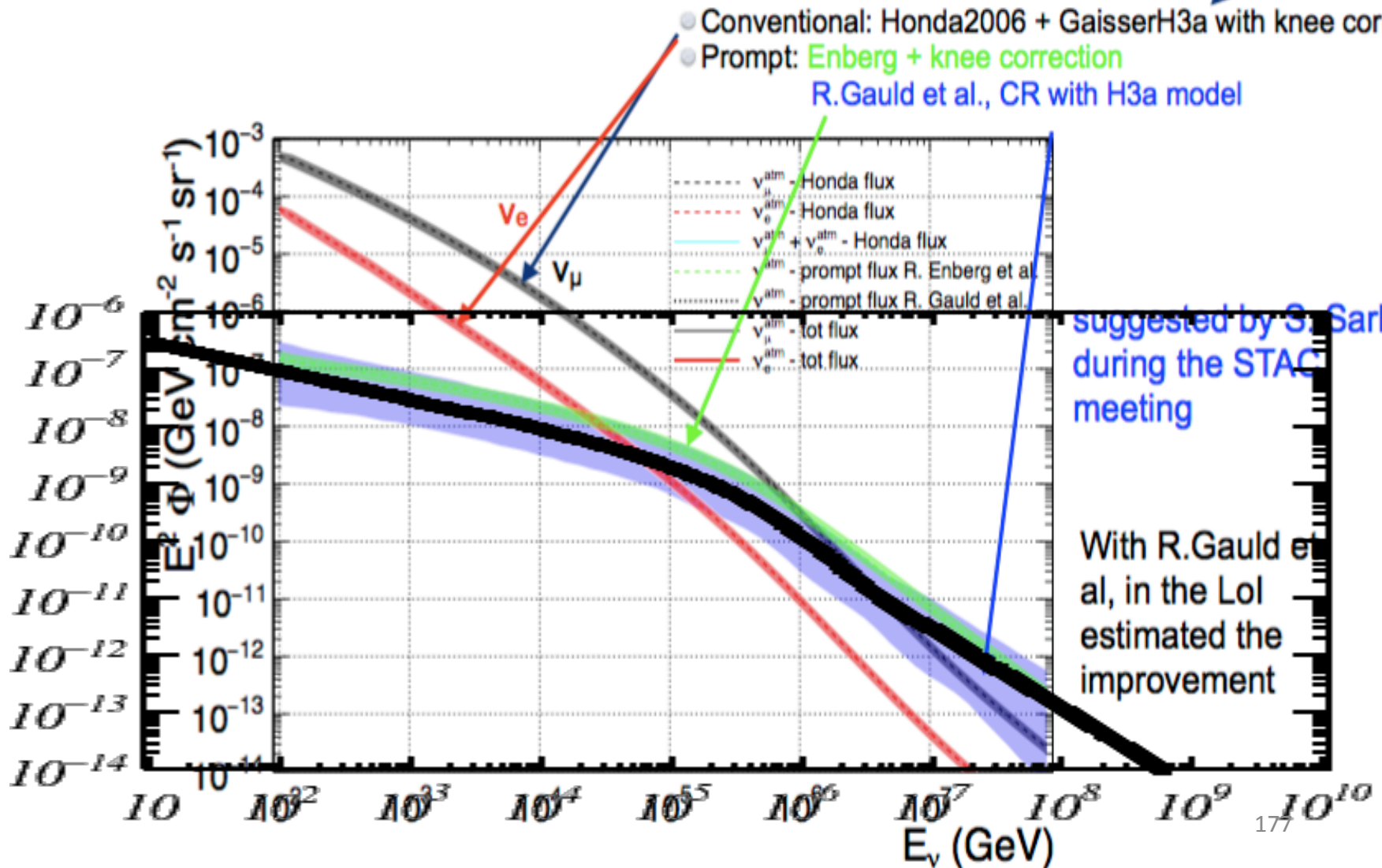
Gauld 2016

T. Gaisser 2012



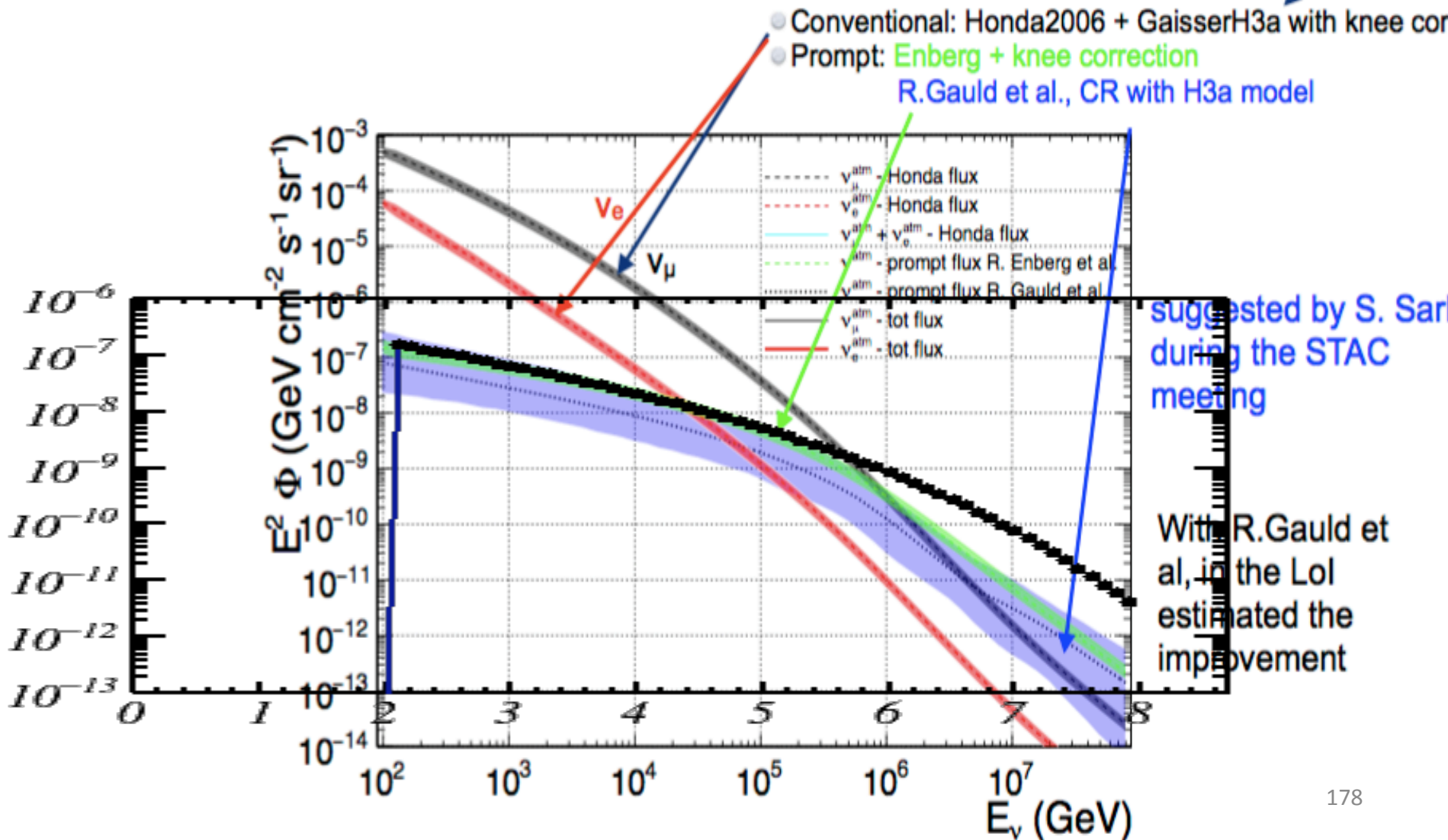
Gauld 2016 extrapolated

T. Gaisser 2012



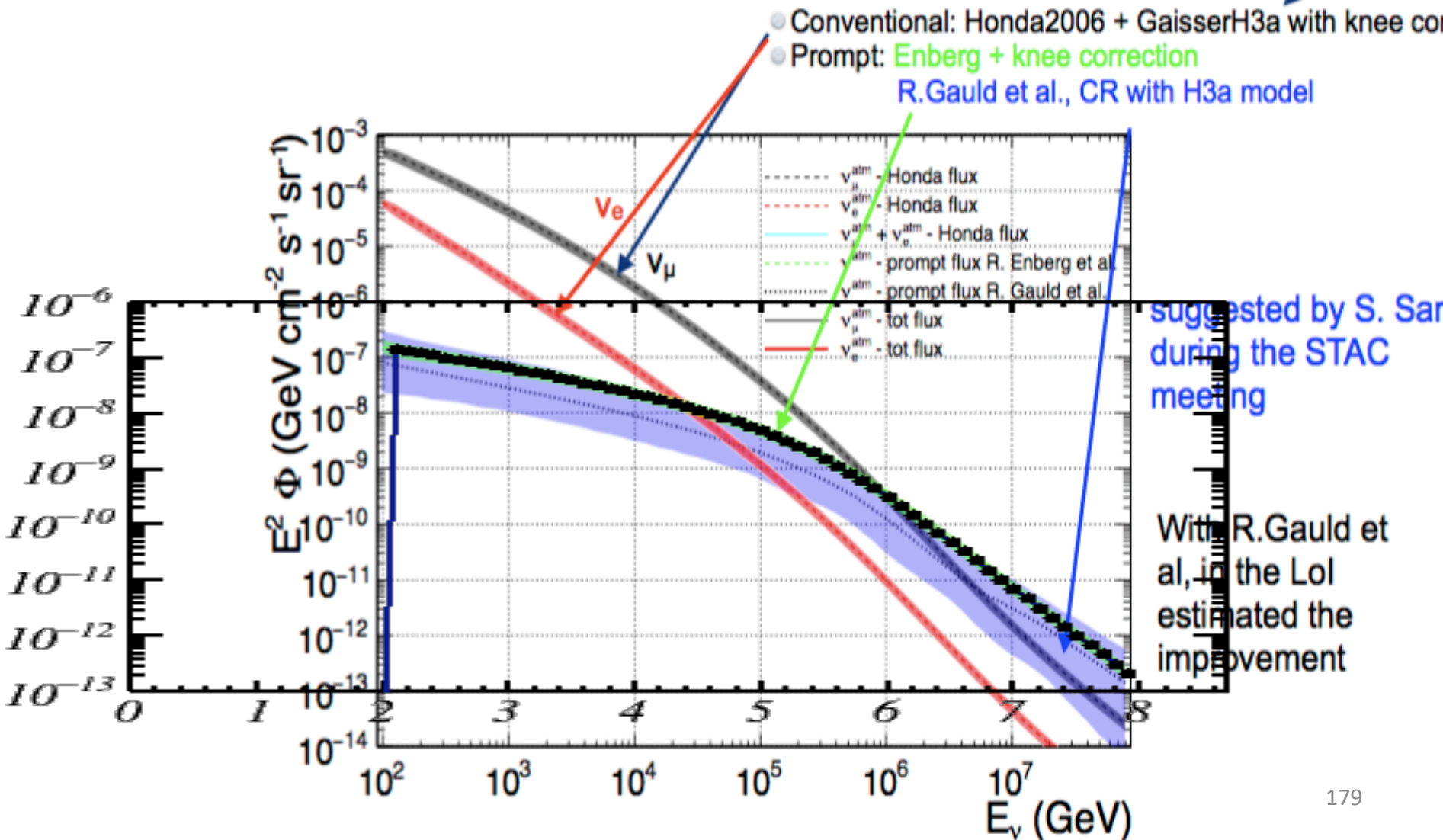
Enberg extrapolated

T. Gaisser 2012

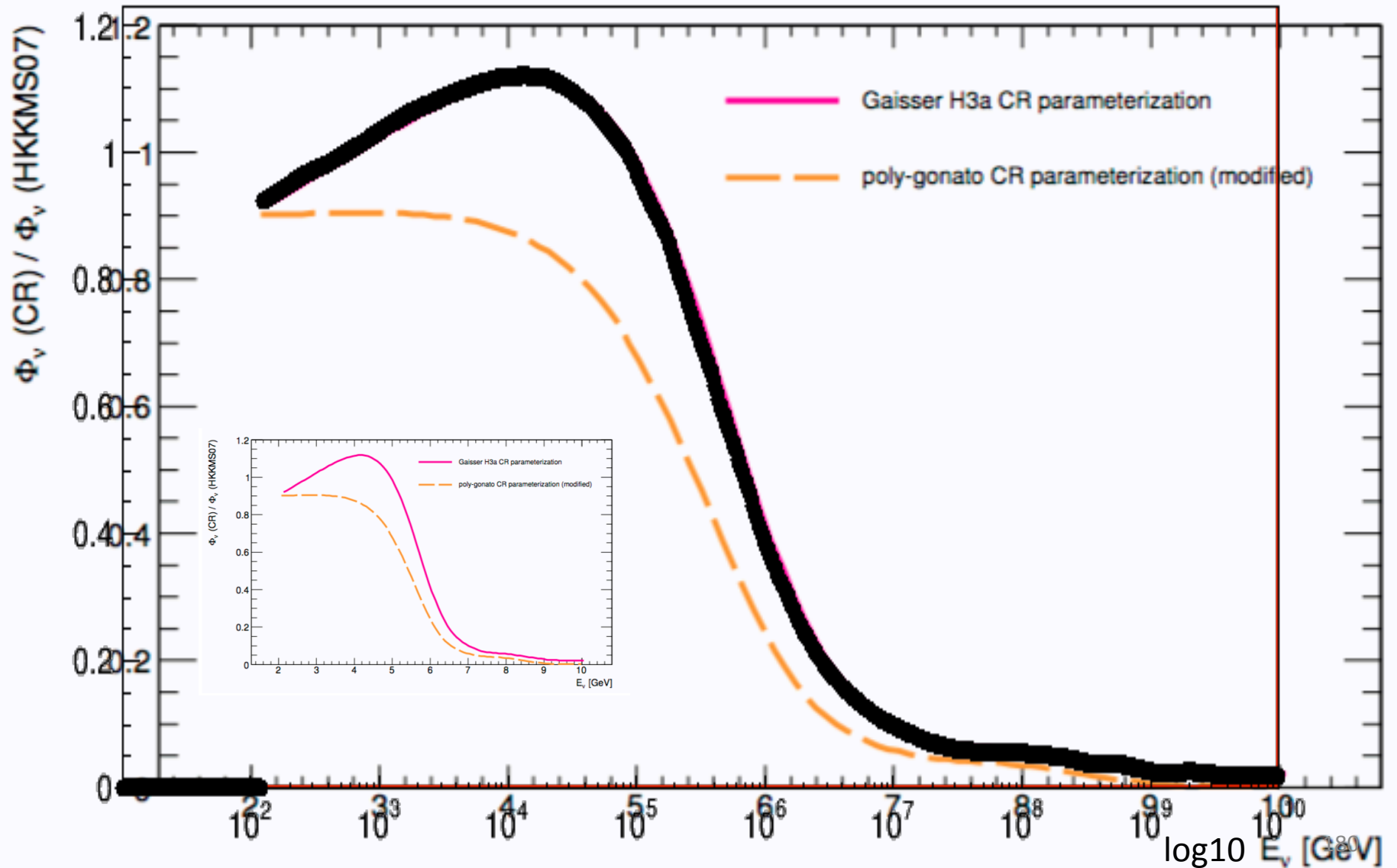


Enberg extrapolated + knee correction

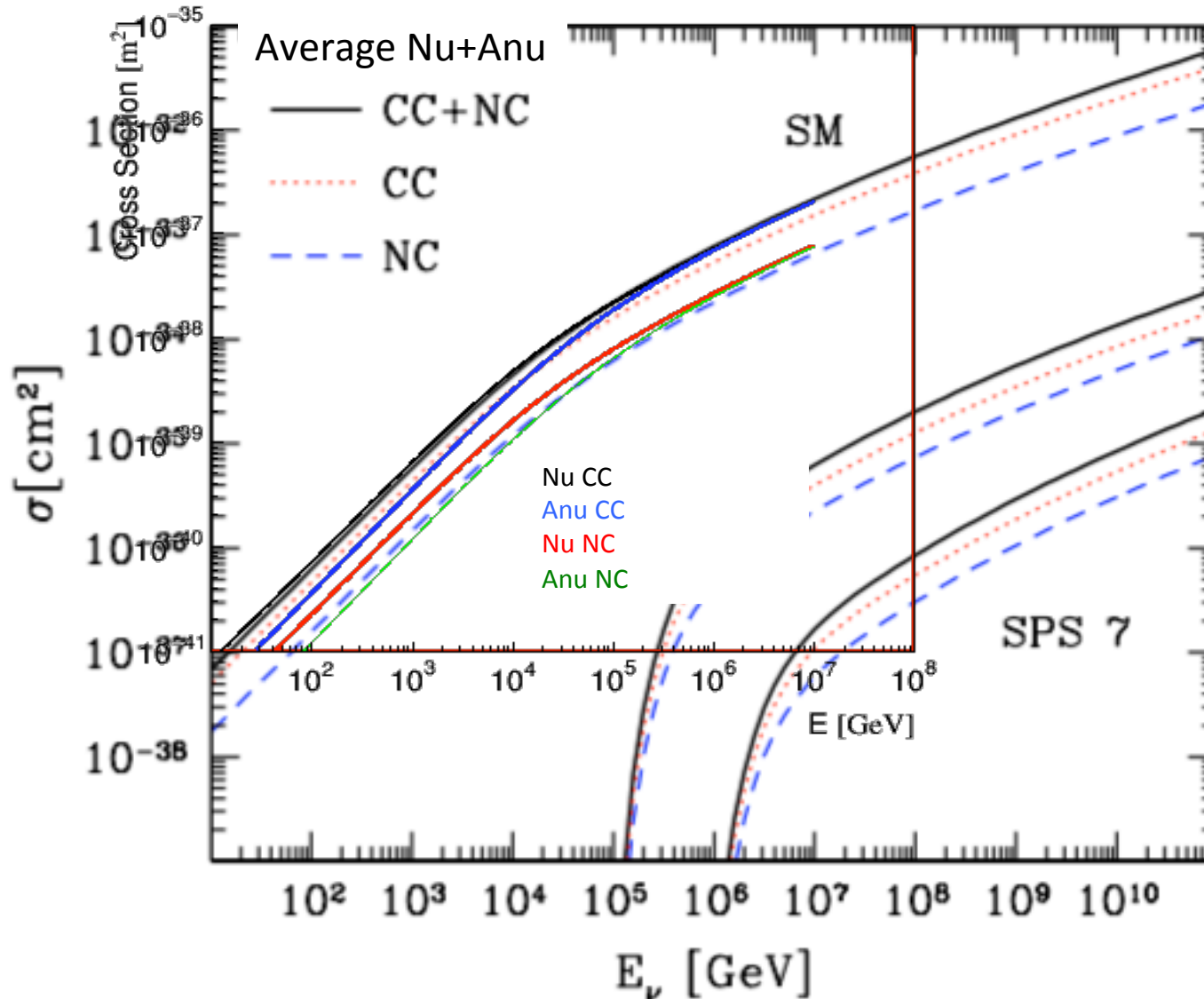
T. Gaisser 2012



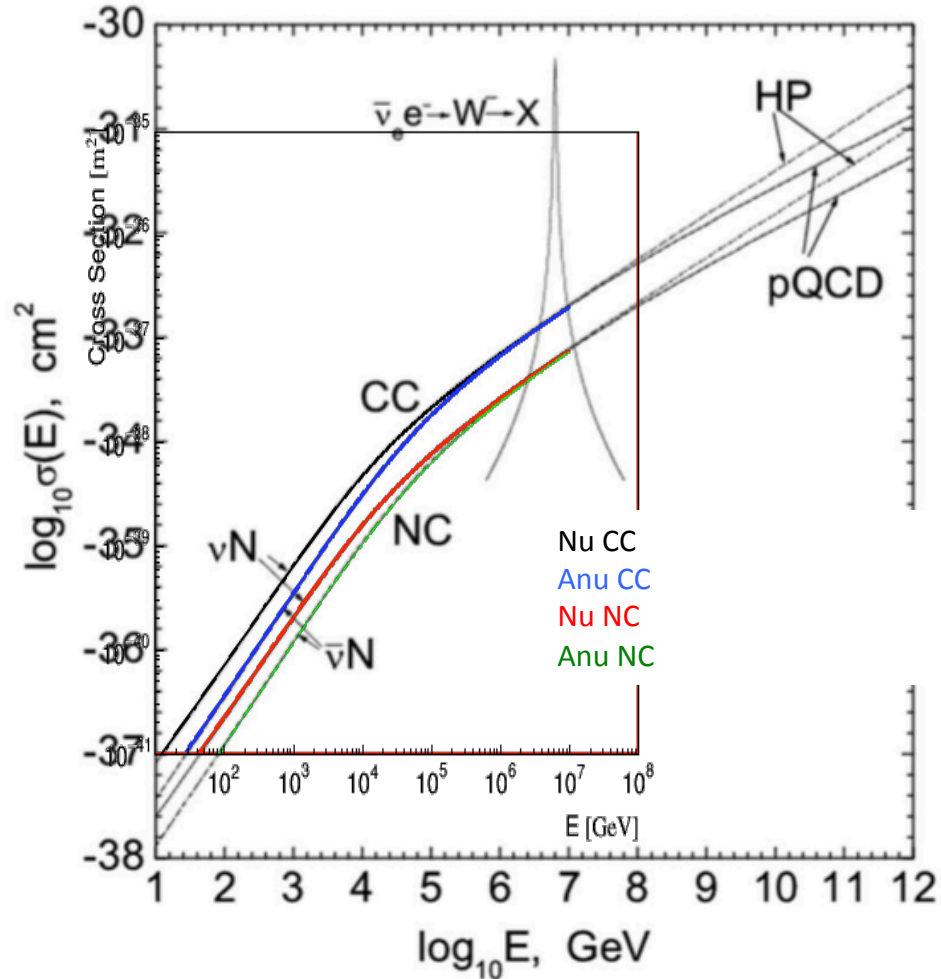
Knee Correction (Gaisser H3a)



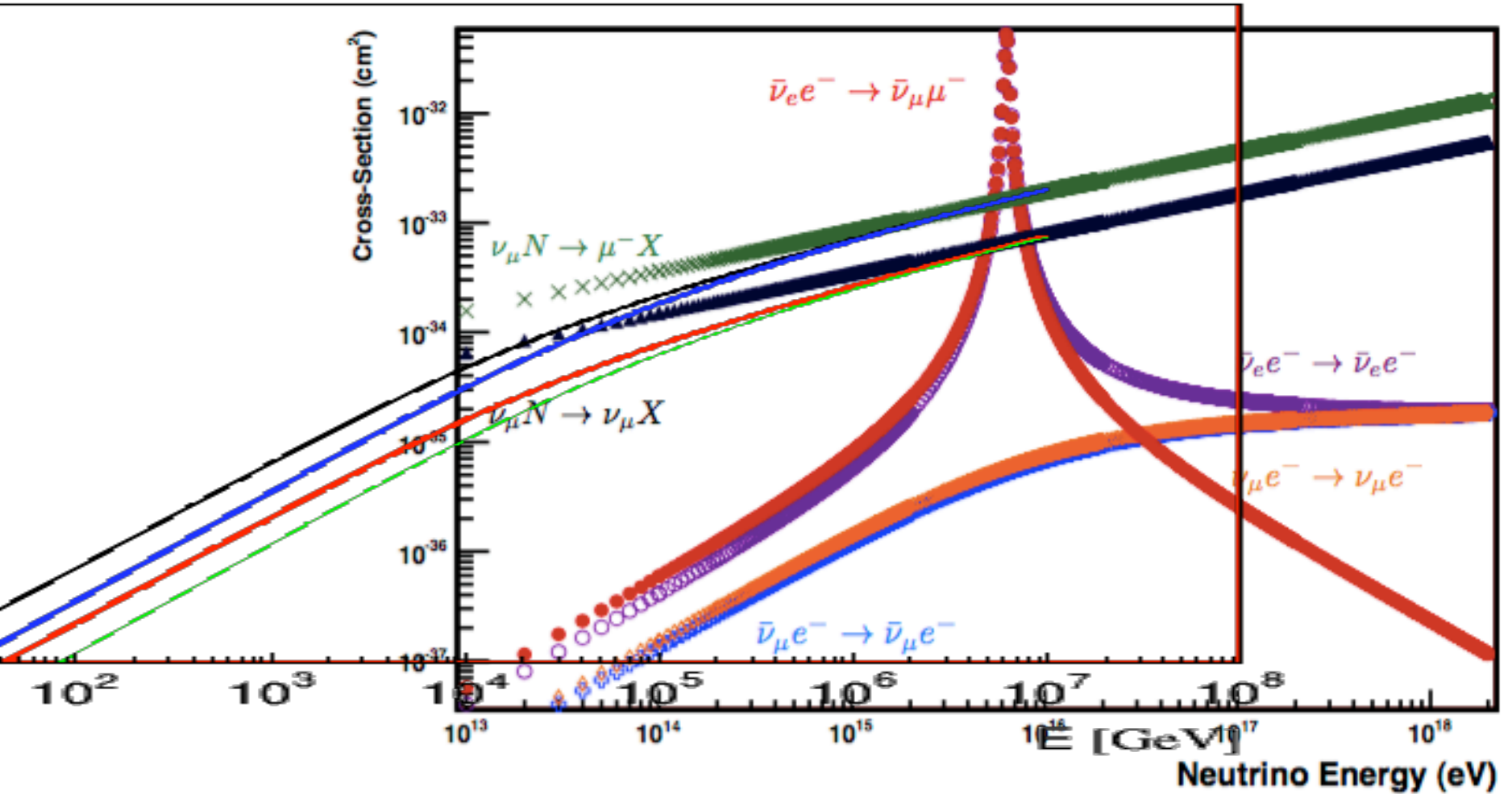
Neutrino Cross Sections



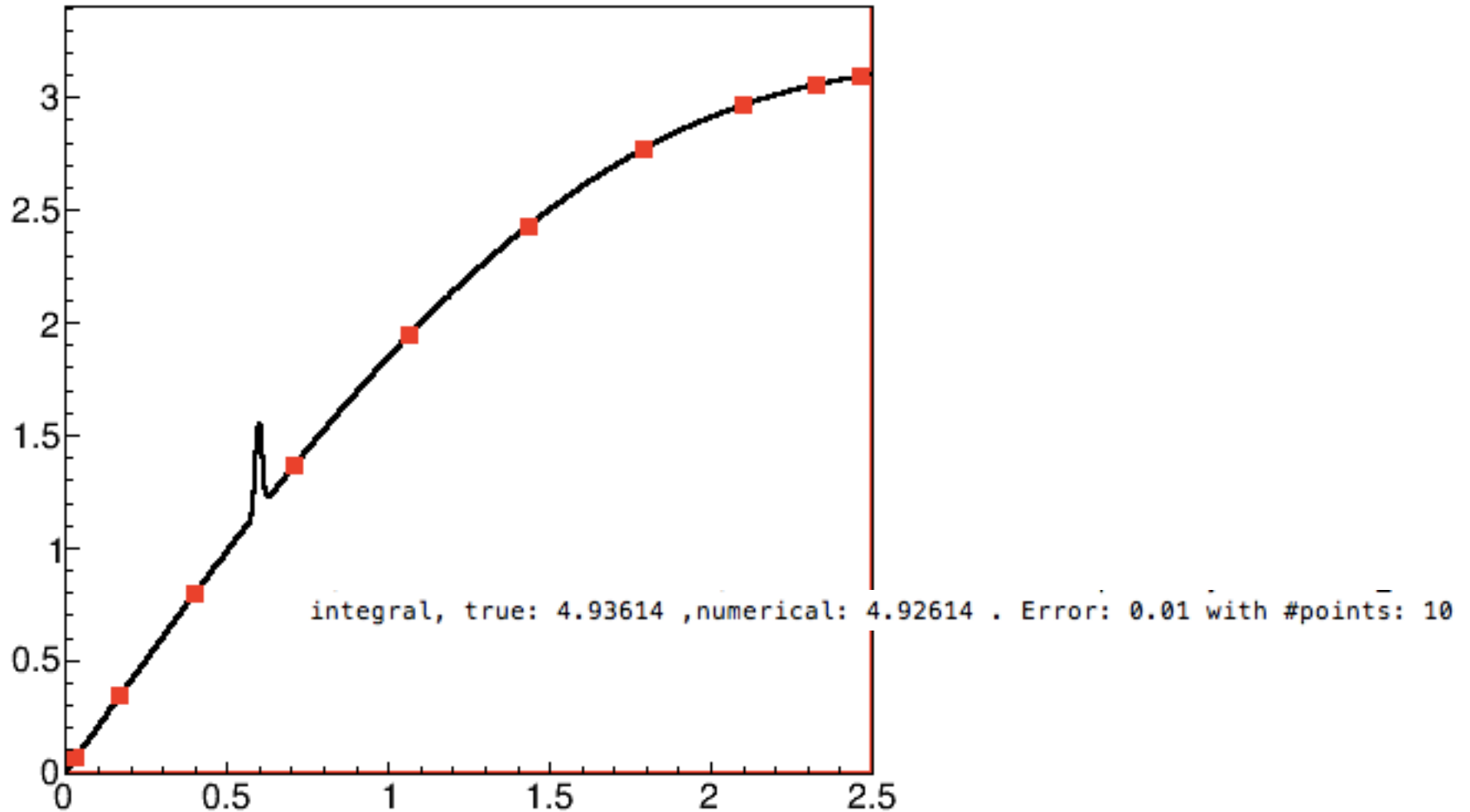
Neutrino Cross Sections



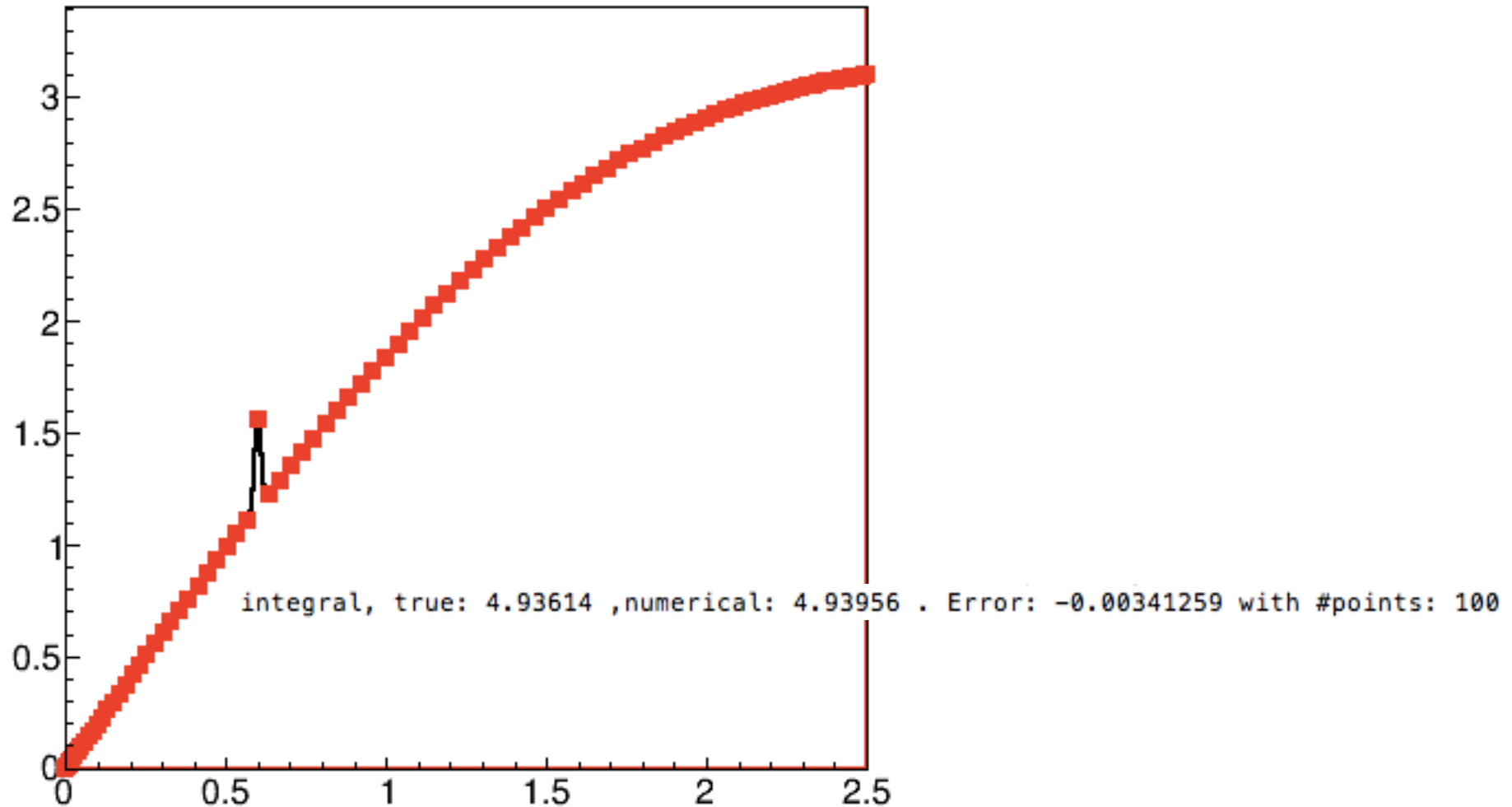
Neutrino Cross Sections



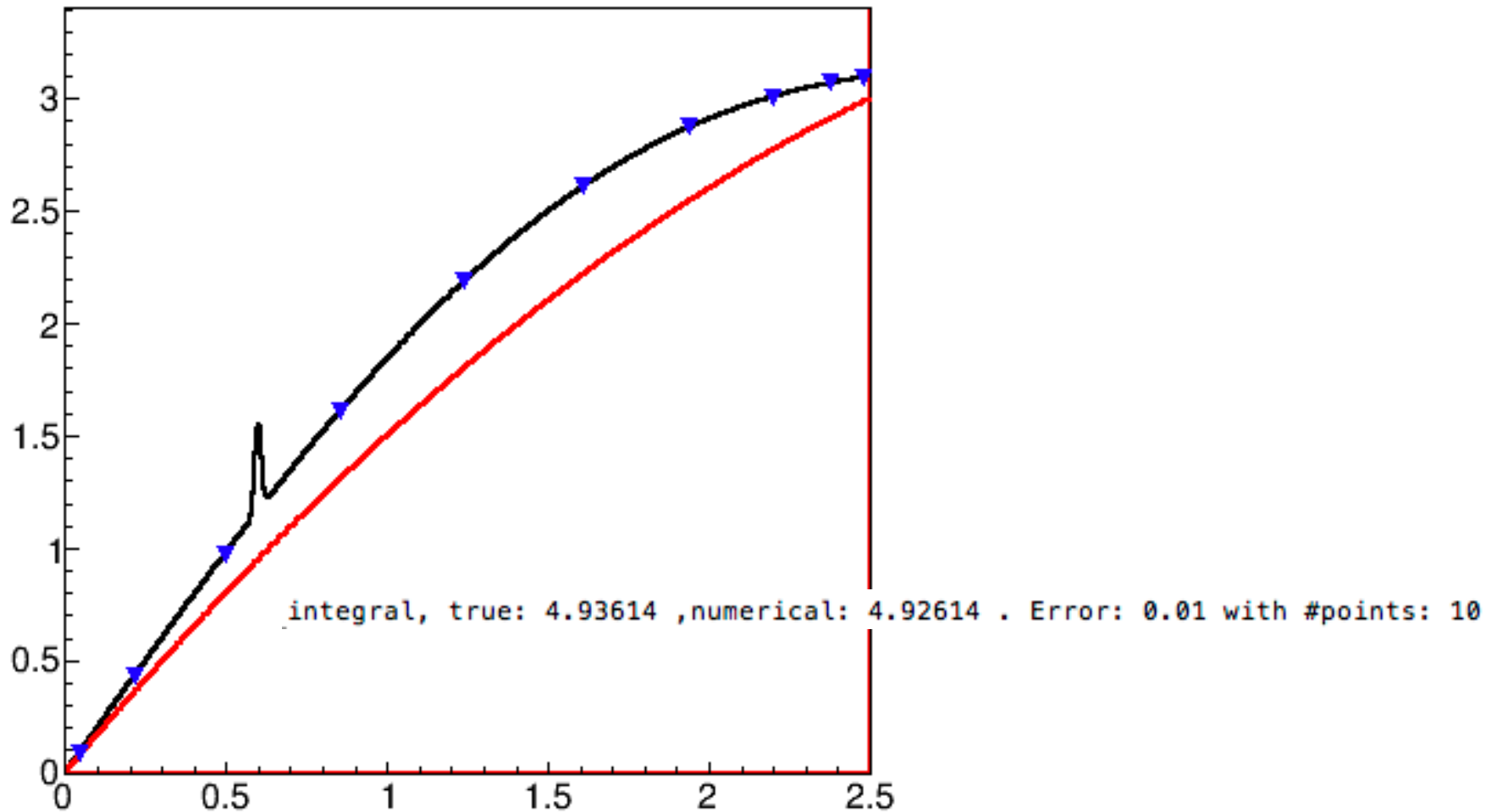
Gaussian Quadrature



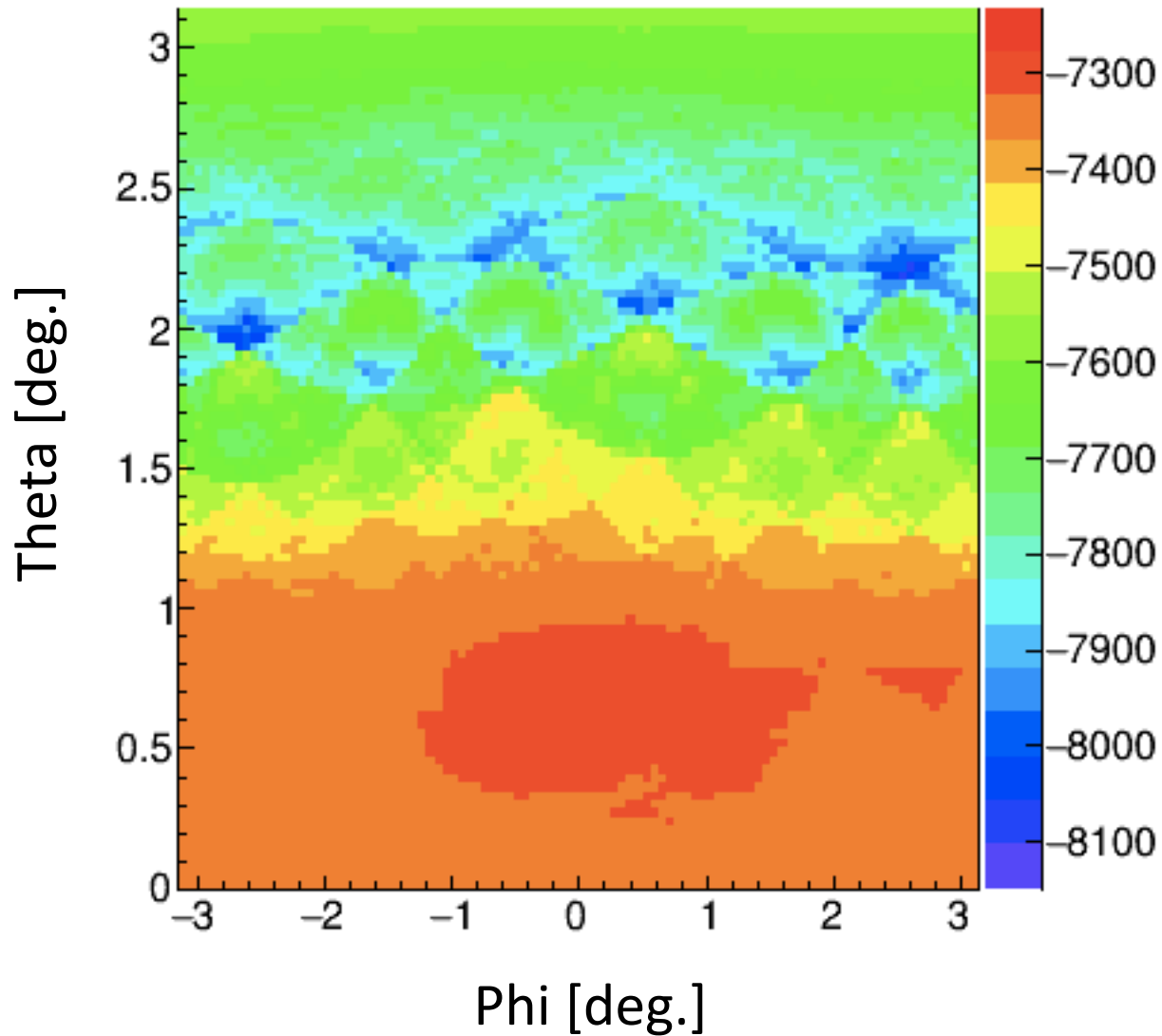
Gaussian Quadrature



Gaussian Quadrature



Event Probability: Direction



Kopper Shower Param.

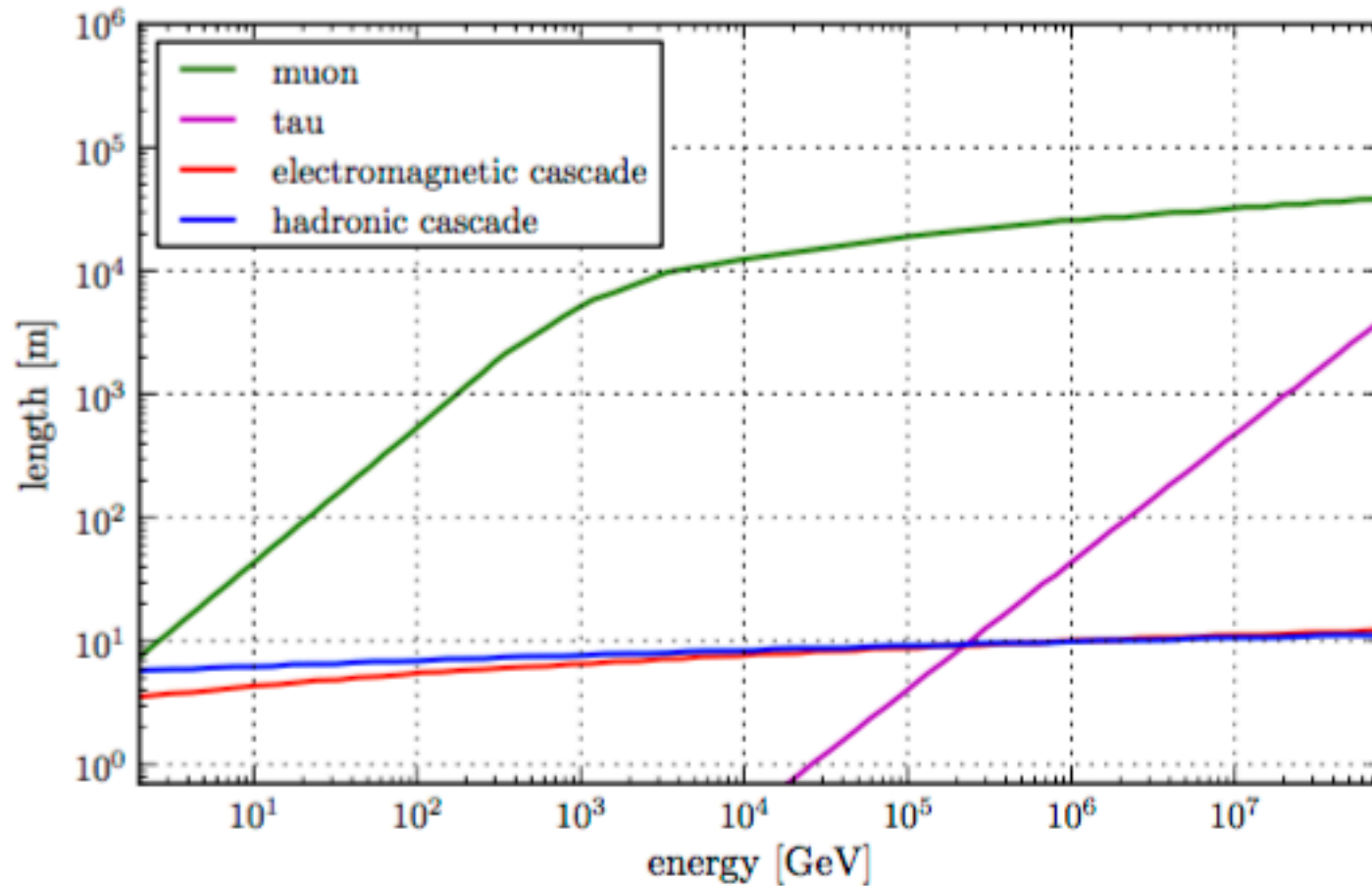


Figure 2.2 Mean Muon (μ) and tau (τ) path lengths and mean cascade lengths for electromagnetic and hadronic cascades in water. Data taken from [35].

