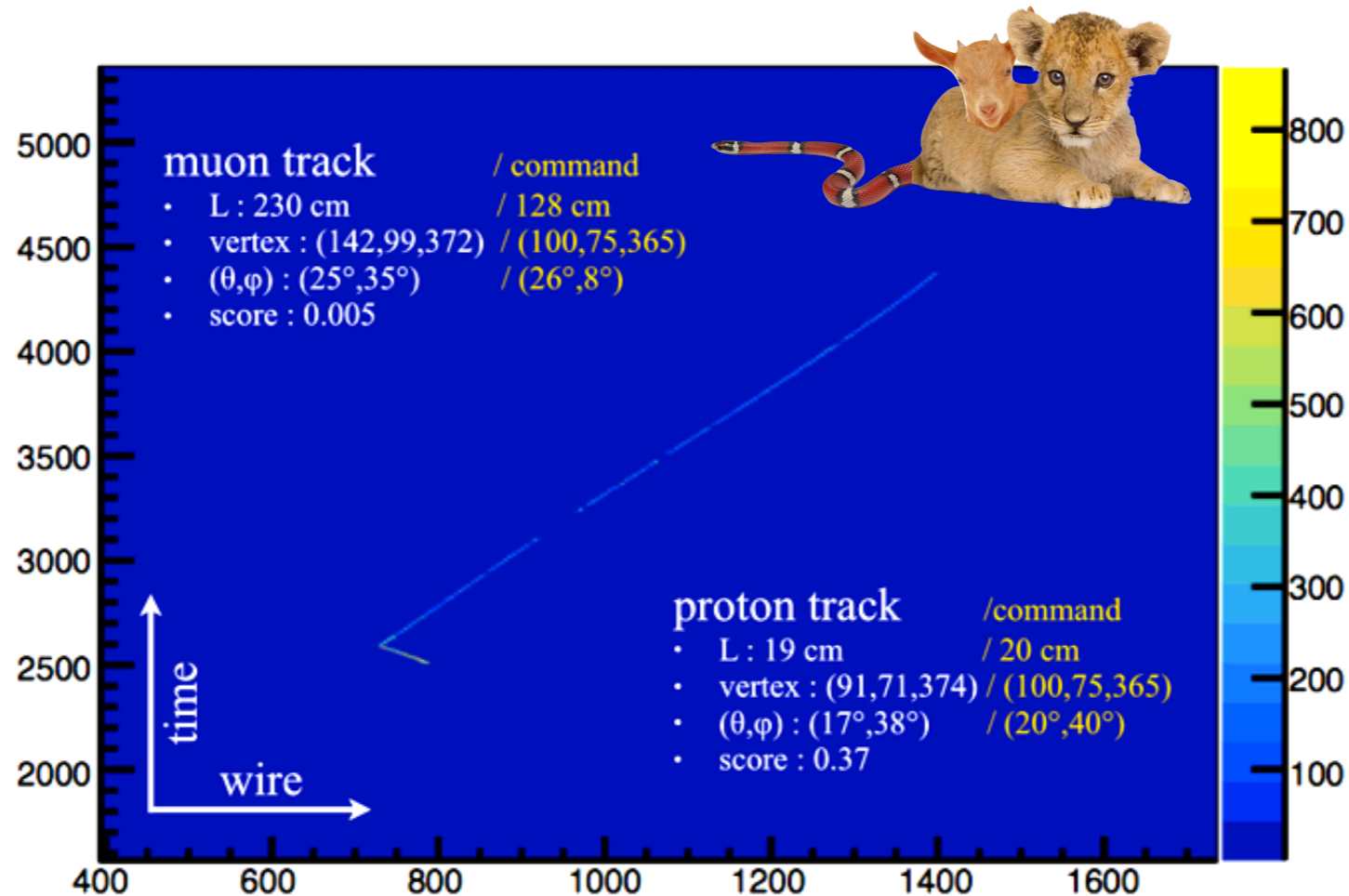


# Chimera Events in the MicroBooNE Experiment

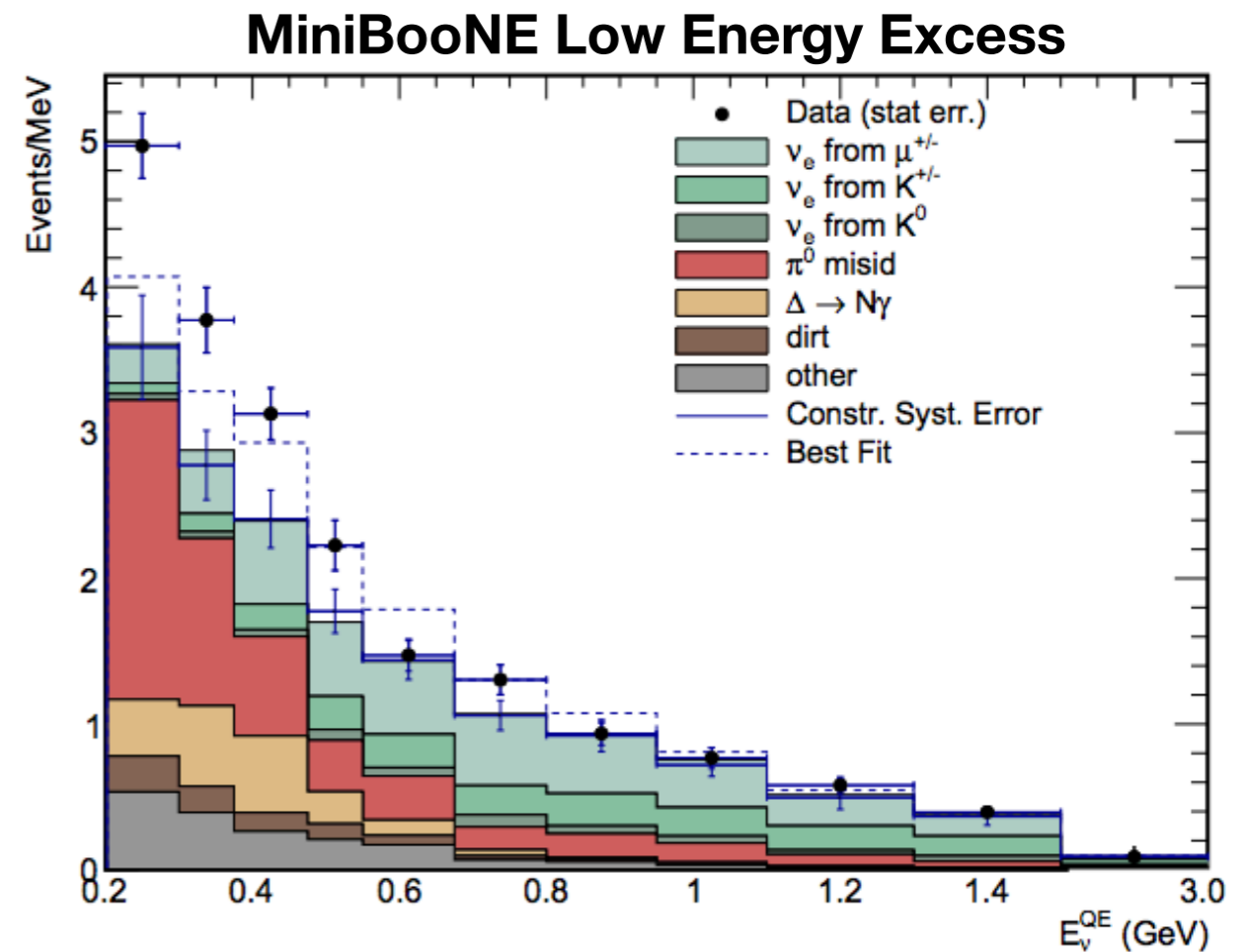


Polina Abratenko  
June 10, 2019  
New Perspectives



# Short Baseline Low Energy Excess Search

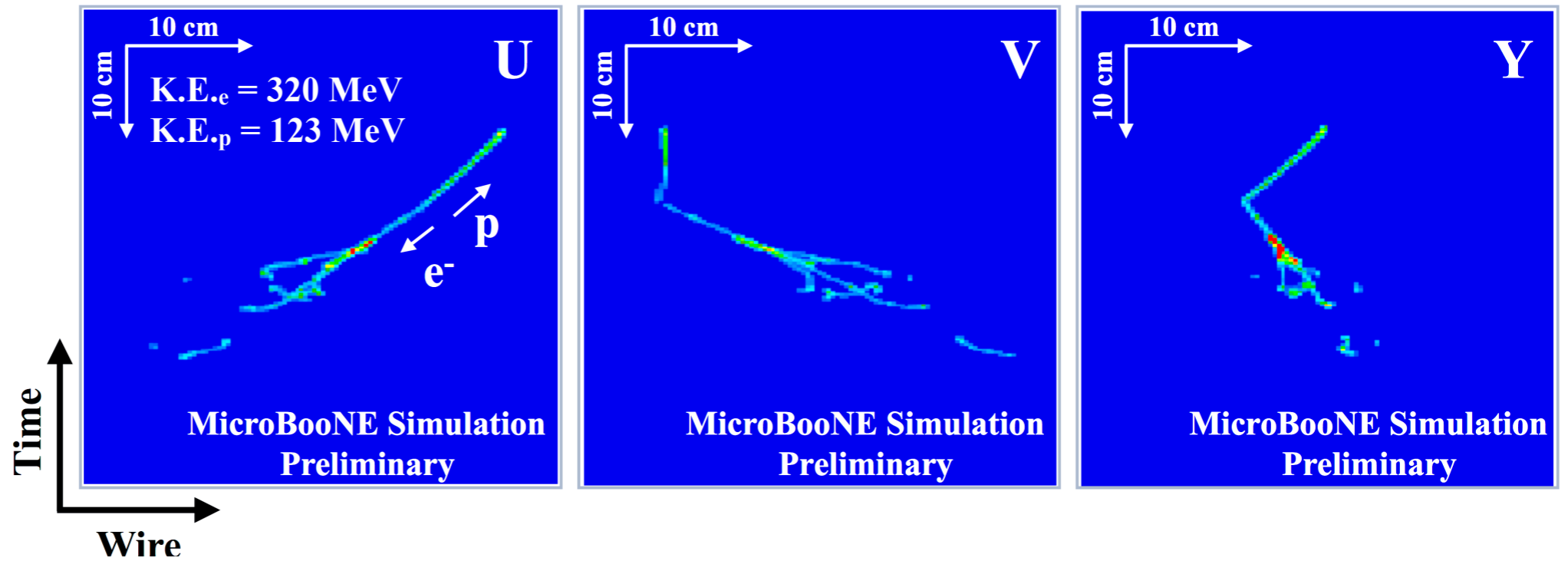
- Major goal of MicroBooNE: investigate the excess of low energy events (LEE) observed by MiniBooNE
  - Recreate this measurement with reduced background
- Dominated by charged-current quasi-elastic (CCQE) events
- Constrain systematic uncertainties on intrinsic  $\nu_e$  background with  $\nu_\mu$
- Using deep learning techniques for this LEE analysis



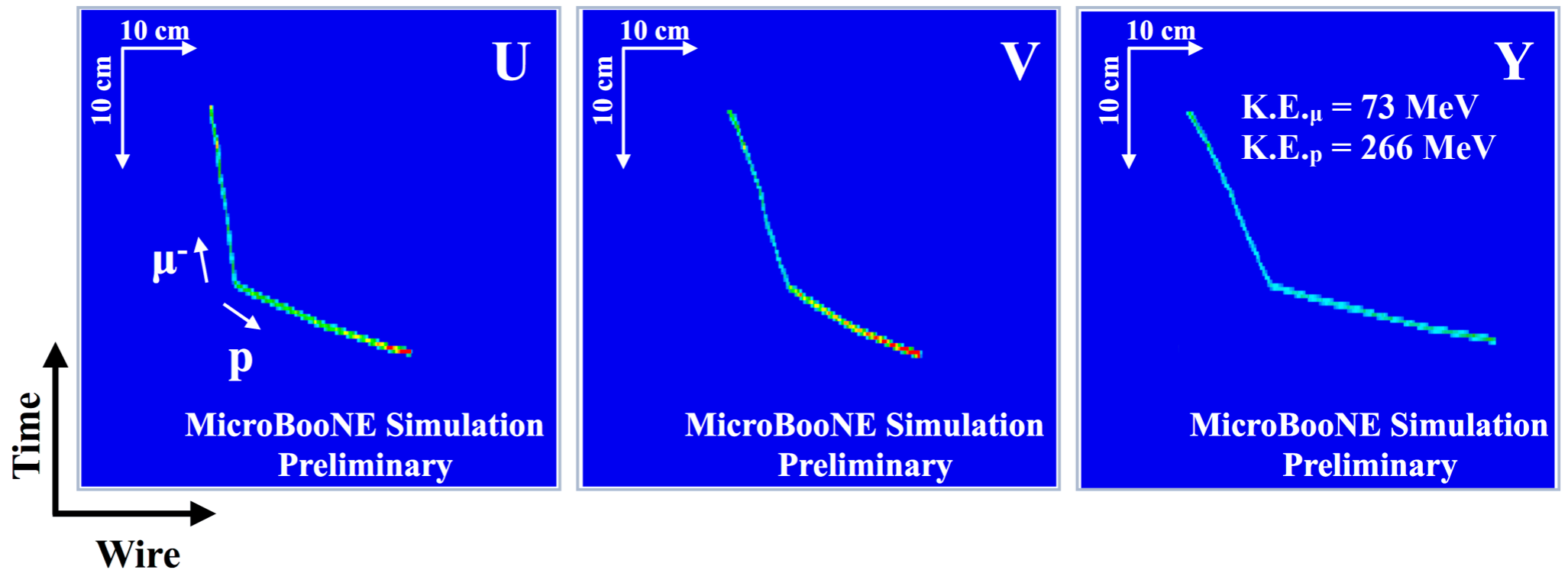
Excess is in the  
200-800 MeV  
range

# CCQE Events in MicroBooNE

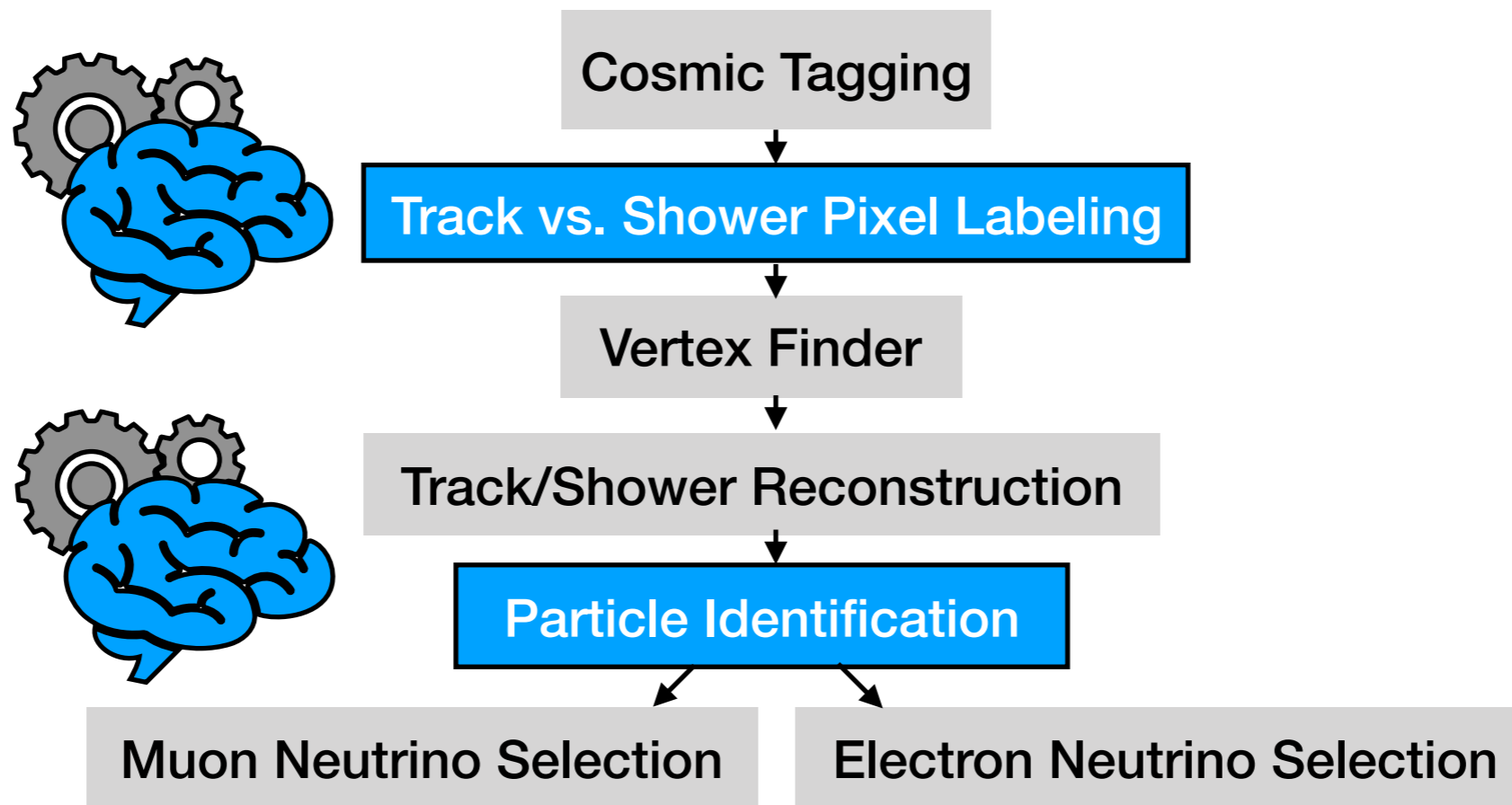
$\nu_e$



$\nu_\mu$



# MicroBooNE Deep Learning LEE Analysis



## Publications:

[JINST 12, P03011 \(2017\)](#)

[Phys. Rev. D99, 092001 \(2019\)](#)

## Public Notes:

[MICROBOONE-NOTE-1042-PUB](#)

[MICROBOONE-NOTE-1051-PUB](#)

# MicroBooNE Deep Learning LEE Analysis

- Interested in CCQE events that are:
  - Fully contained in the TPC
  - Between 200-800 MeV neutrino energy
  - Of a two-particle topology consisting of 1 lepton (electron, muon), 1 proton
- Need a good sized sample in data
  - Including all detector effects, dead wires, etc.
- MicroBooNE detector is difficult to simulate

Contributions to total cross section systematic uncertainty  
([arXiv:1905.09694](https://arxiv.org/abs/1905.09694))

Source of uncertainty	Relative uncertainty [%]
Beam flux	12.4
Cross section modeling	3.9
Detector response	16.2
Dirt background	10.9
Cosmic ray background	4.2
MC statistics	0.2
Stat	1.4
Total	23.8

# MicroBooNE Deep Learning LEE Analysis

- Interested in CCQE events that are:
  - Fully contained in the TPC
  - Between 200-800 MeV neutrino energy
  - Of a two-particle topology consisting of 1 lepton (electron, muon), 1 proton
- Need a good sized sample in data
  - Including all detector effects, dead wires, etc.
- MicroBooNE detector is difficult to simulate

## Contributions to total cross section systematic uncertainty ([arXiv:1905.09694](https://arxiv.org/abs/1905.09694))

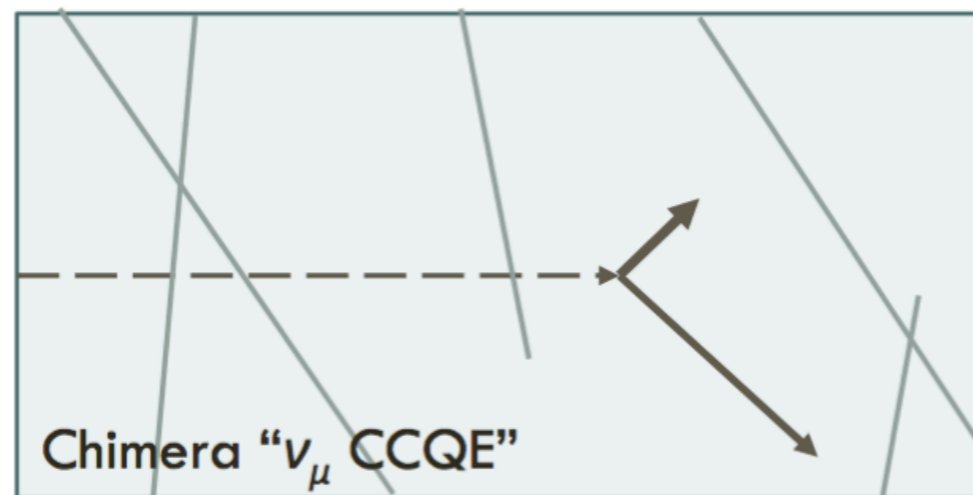
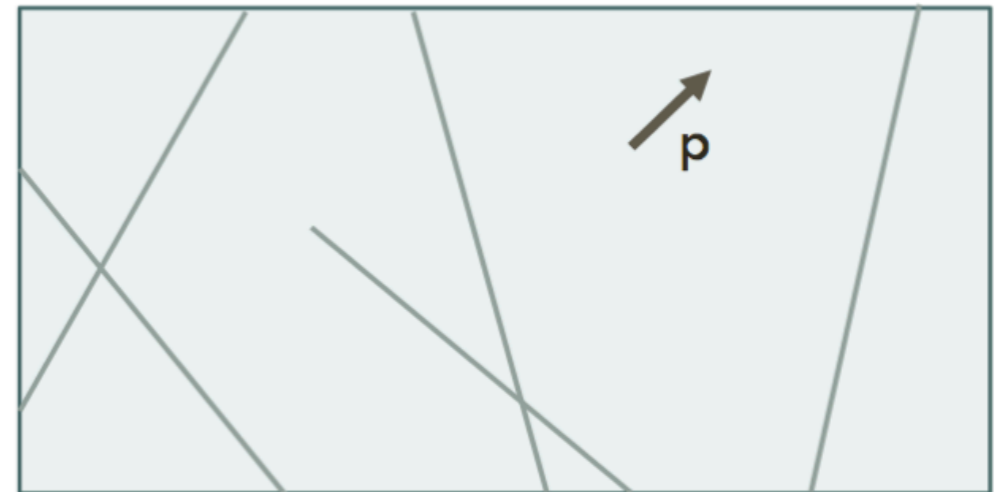
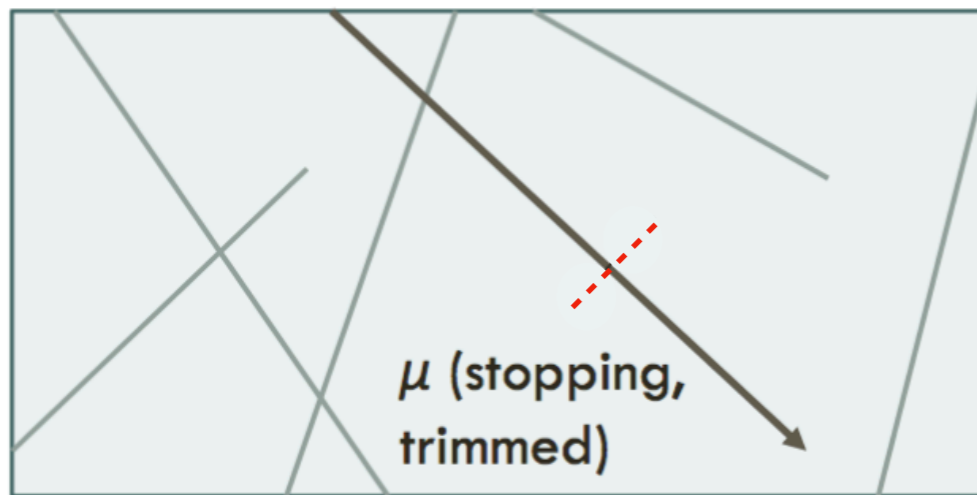
Source of uncertainty	Relative uncertainty [%]
Beam flux	12.4
Cross section modeling	3.9
Detector response	16.2
Dirt background	10.9
Cosmic ray background	4.2
MC statistics	0.2
Stat	1.4
Total	23.8

Idea:  
Work with data by  
creating our own  
“chimera” CCQE-like  
events



# Chimera Events

E.g. A  $\nu_\mu$  CCQE event:



# Applications of Chimera Events

- Studies on systematic uncertainties: validate algorithms' performance on samples with known properties
  - Location of vertex and opening angles of lepton/proton are known
- Evaluating selection efficiency, reconstruction's energy resolution for events similar to a target final state
- Create events for re-training a network
  - First pass on MC to teach general interaction features
  - Re-train on chimeras to teach about data/MC differences



# Finding the Right Tracks

- Select tracks that closely match a target topology to preserve detector effects
  - For this reason, want to avoid rotating/drastically moving tracks
- Put constraints on specific parameters and search through a pool of existing events
- Care about kinematics (angle, track length) and systematics (position)
  - Parameters:  $X, Y, Z, \theta, \phi, \text{length}$ 
    - Muons and protons
  - Candidate entries must pass selection cuts
- Minimize a likelihood to choose a track, given 6 parameters as input
- To start, focus on MC BNB  $\nu_\mu$ -like events

# Maximum Likelihood

- Say we have a defined input parameter,  $X$
- For finding the closest matching value to this  $X$ , assume a gaussian function with mean at the value  $X$

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e\left(-\frac{1}{2} \frac{(x_i - X)^2}{\sigma^2}\right)$$

- The width of this gaussian is configurable ( $\sigma$ )
- Each  $x_i$  is a candidate “closest match”

# Maximum Likelihood

- We have 6 input parameters:  $X$ ,  $Y$ ,  $Z$ ,  $\theta$ ,  $\phi$ , length
- Take the log likelihood product of each gaussian:

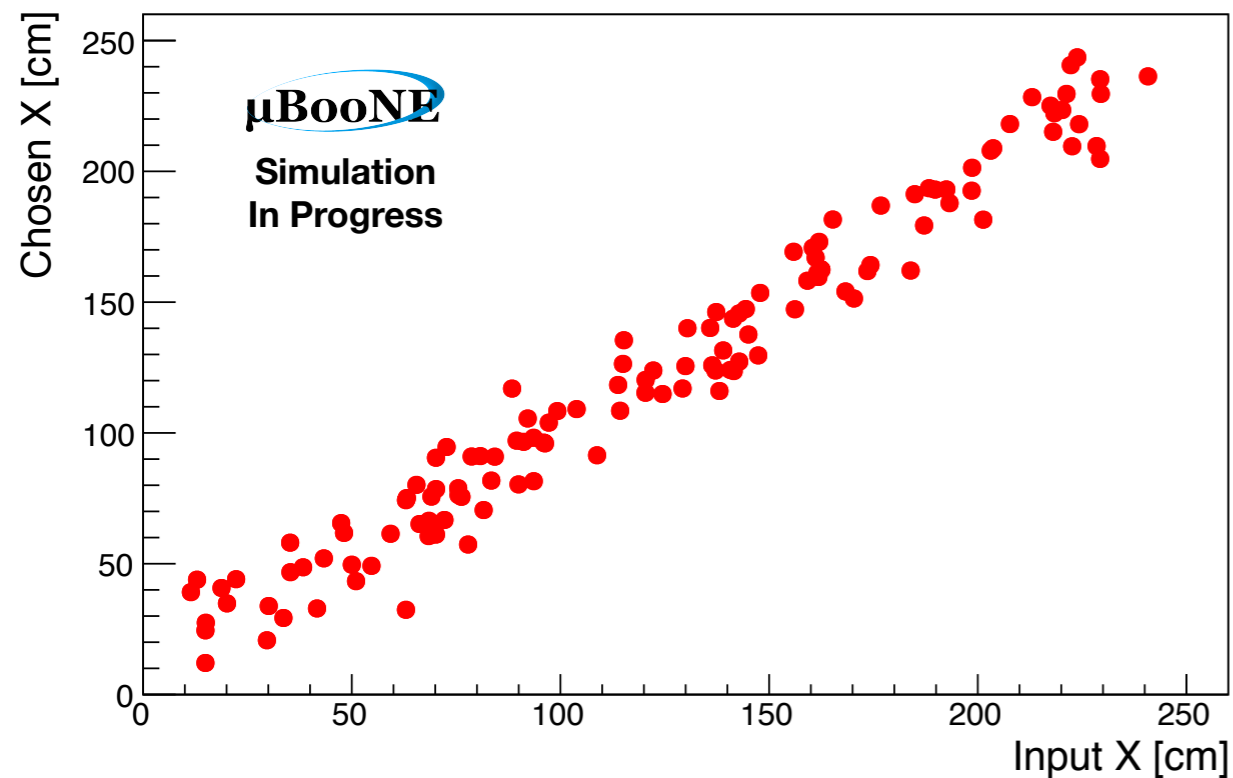
$$\left( -\frac{1}{2}\ln(2\pi) - \frac{1}{2}\ln(\sigma_x^2) - \frac{1}{2\sigma_x^2}(x_i - X)^2 \right) + \dots + \left( -\frac{1}{2}\ln(2\pi) - \frac{1}{2}\ln(\sigma_l^2) - \frac{1}{2\sigma_l^2}(l - L)^2 \right)$$

- Loop through each event and compute likelihood, then minimize
- Parameters, sigmas taken as input
  - User can feed in existing events or input desired parameter values
  - Set sigma for each parameter to control precision of match

# Choosing a Track: Performance

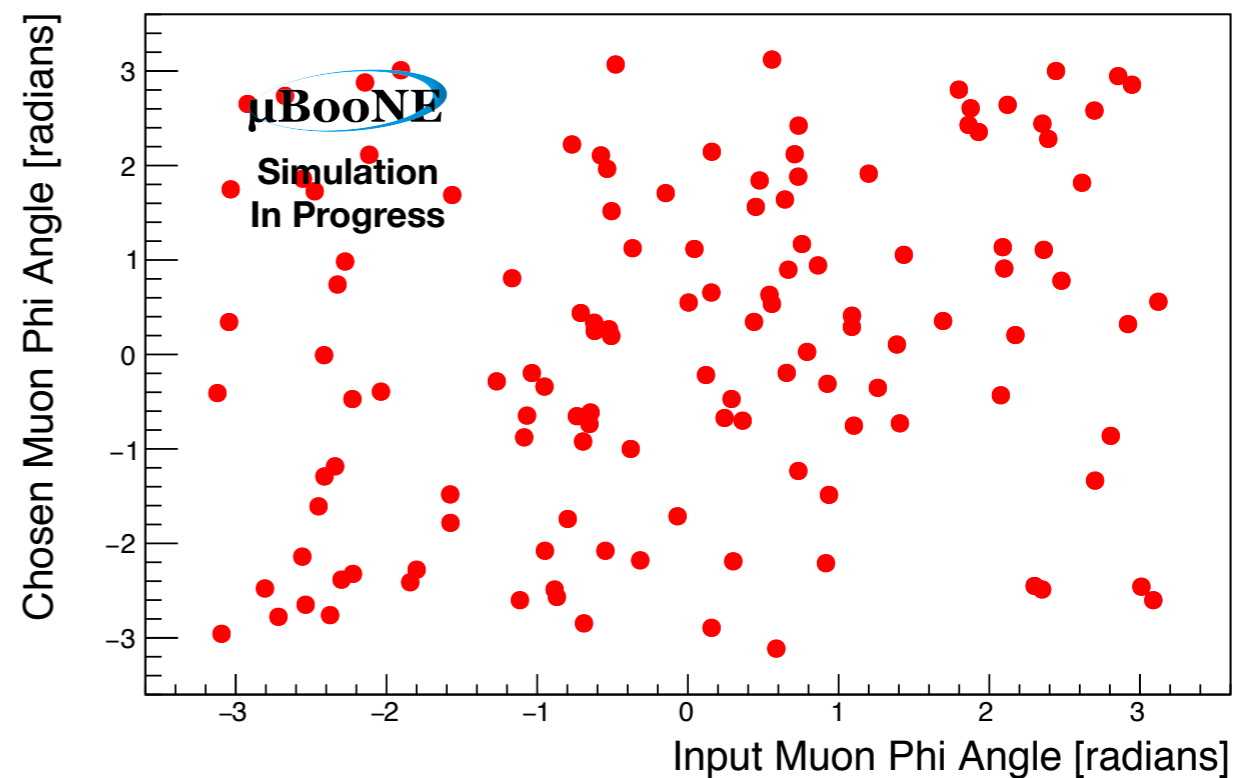
Positive  
correlation

Chosen X vs. Input X



No  
correlation

Chosen muPhi vs. Input muPhi

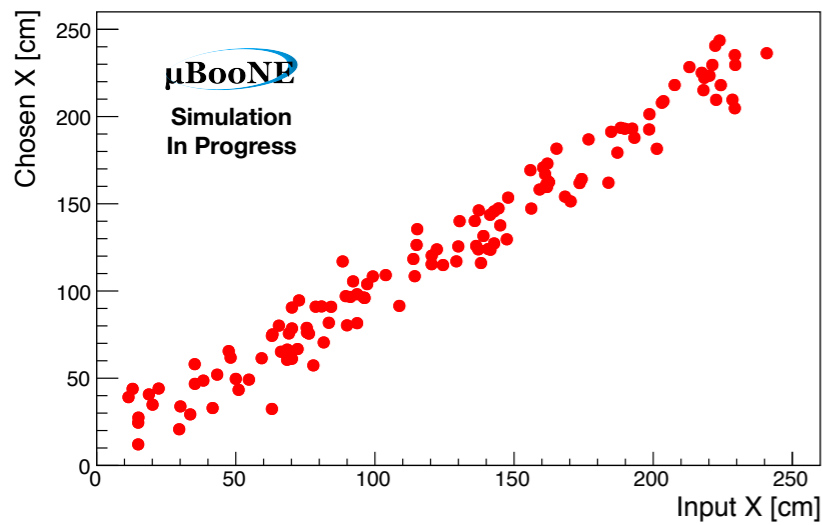


Parameters weighted equally ( $\sigma = 1.0$ )

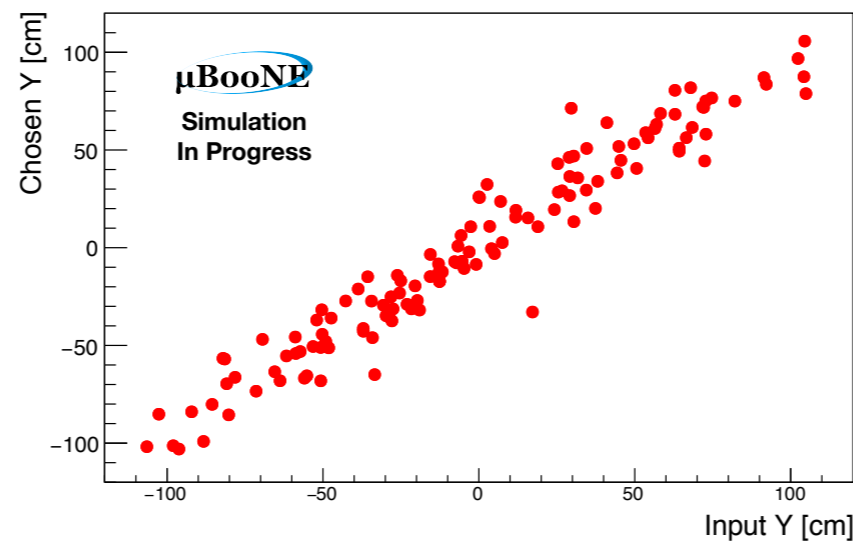
# Choosing a Track: Performance

All parameters weighted equally ( $\sigma = 1.0$ )

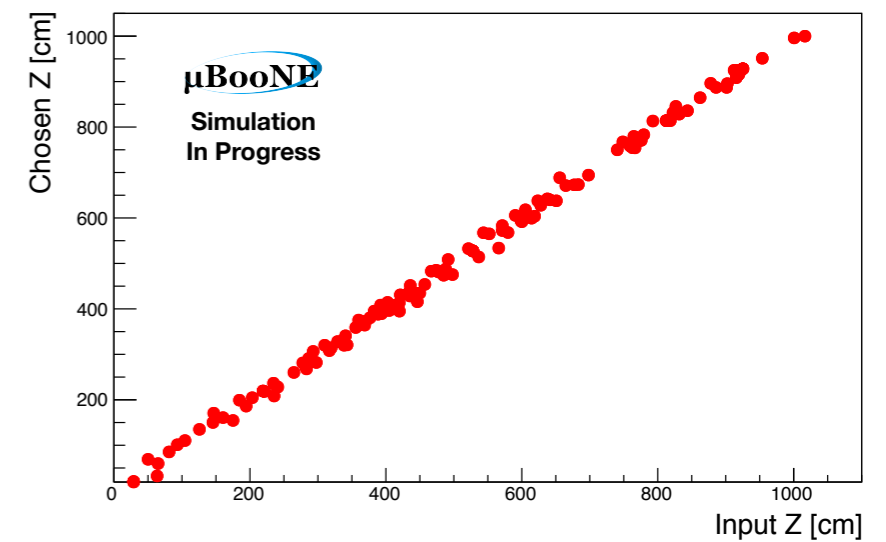
Chosen X vs. Input X



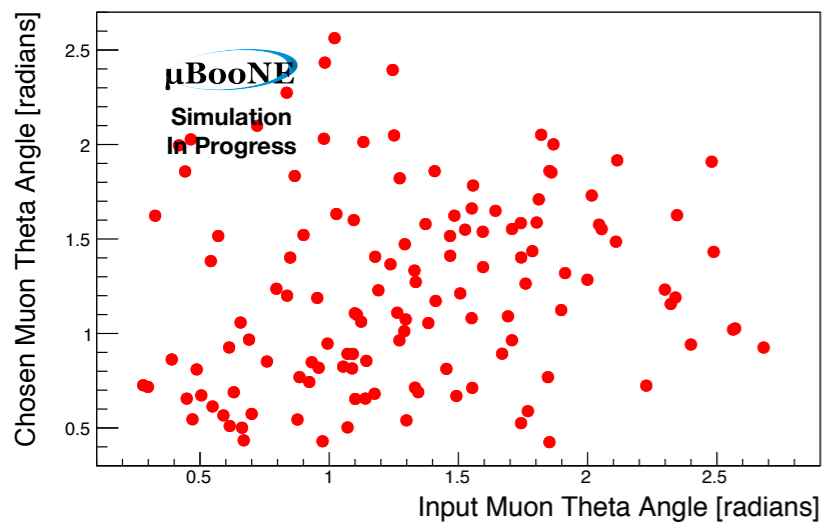
Chosen Y vs. Input Y



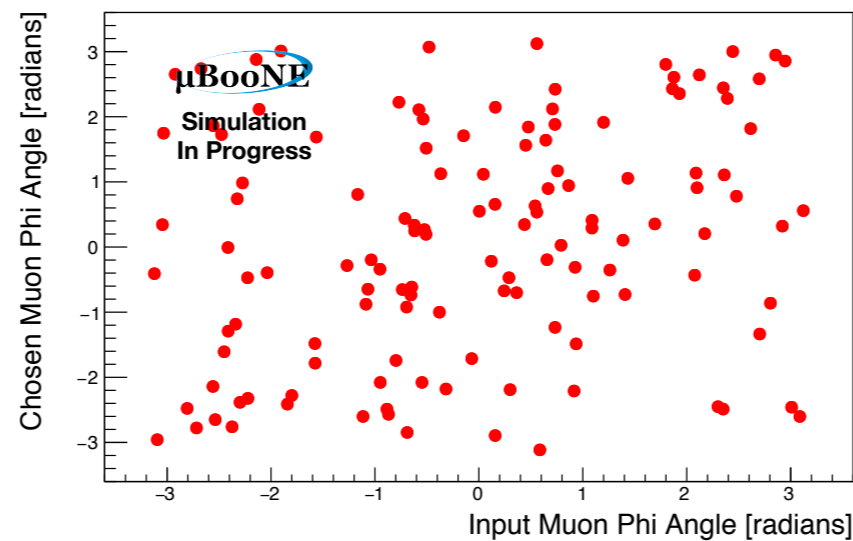
Chosen Z vs. Input Z



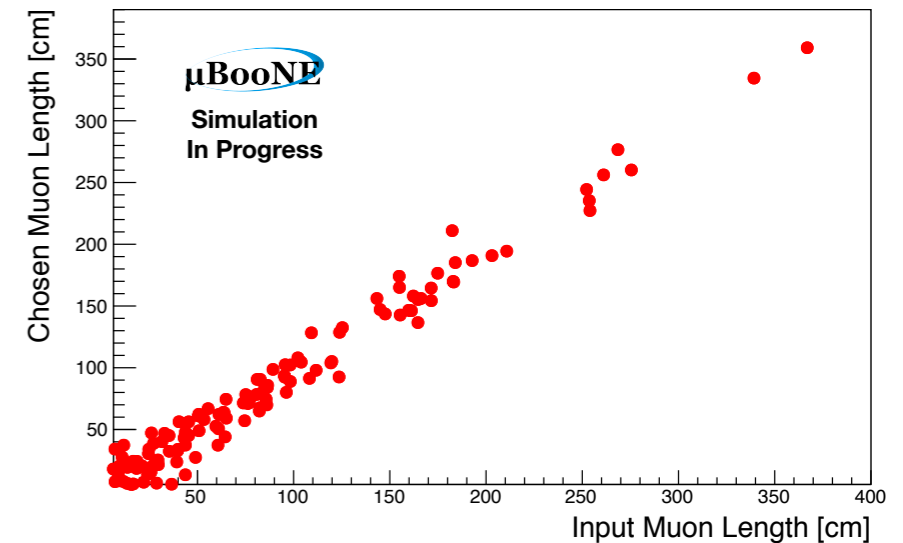
Chosen muTheta vs. Input muTheta



Chosen muPhi vs. Input muPhi

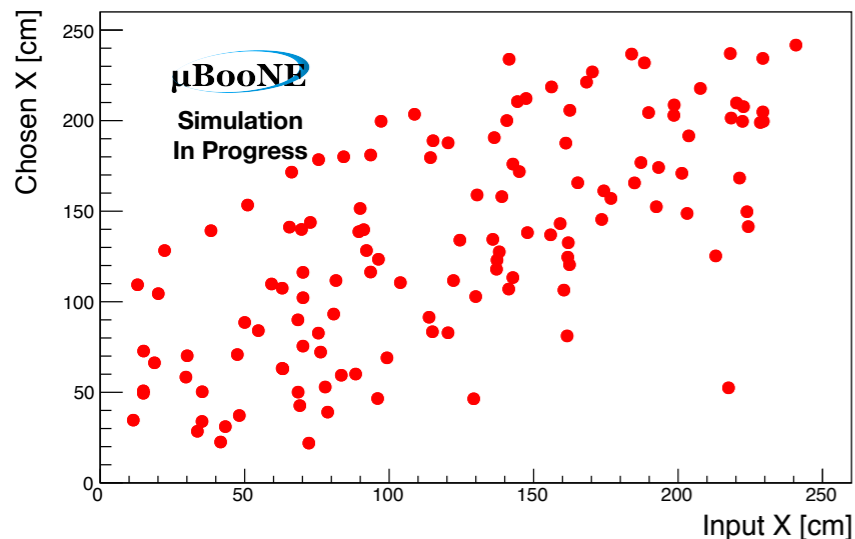


Chosen muLen vs. Input muLen

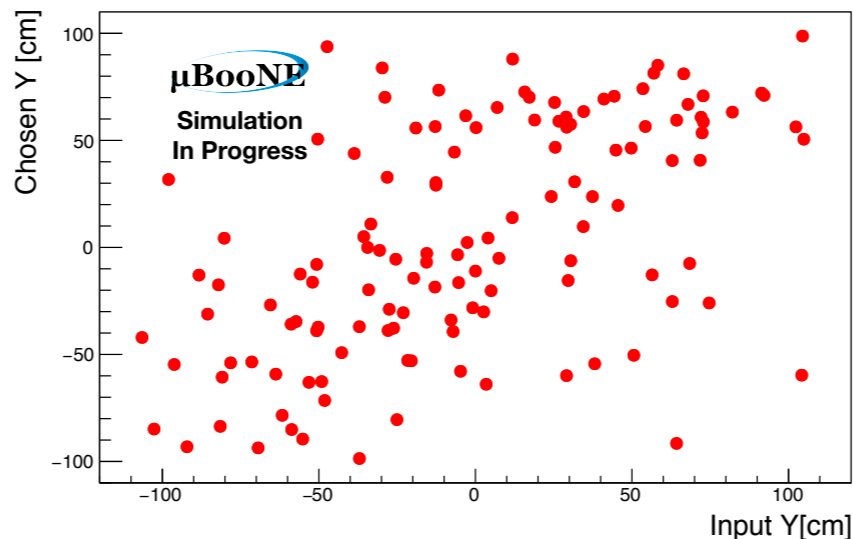


# Keeping $\sigma_\phi = 0.0001$ and all other $\sigma = 1.0$ , how does it affect the parameters?

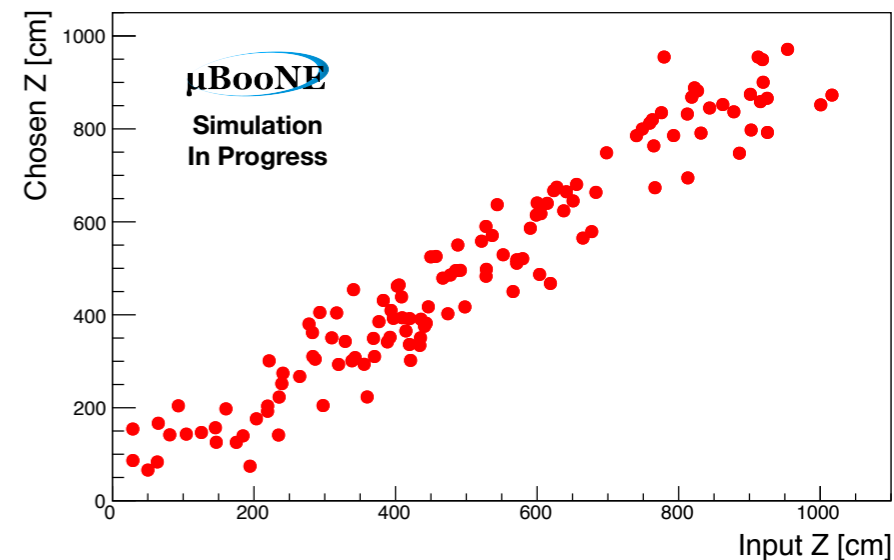
Chosen X vs. Input X



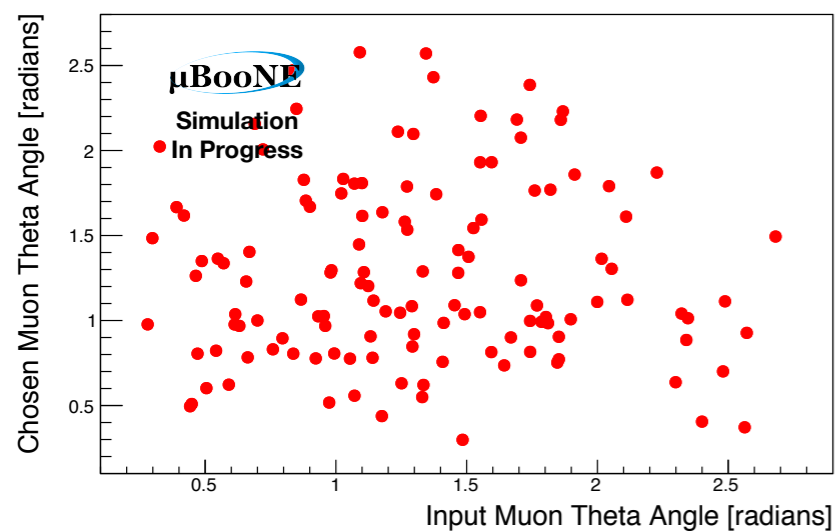
Chosen Y vs. Input Y



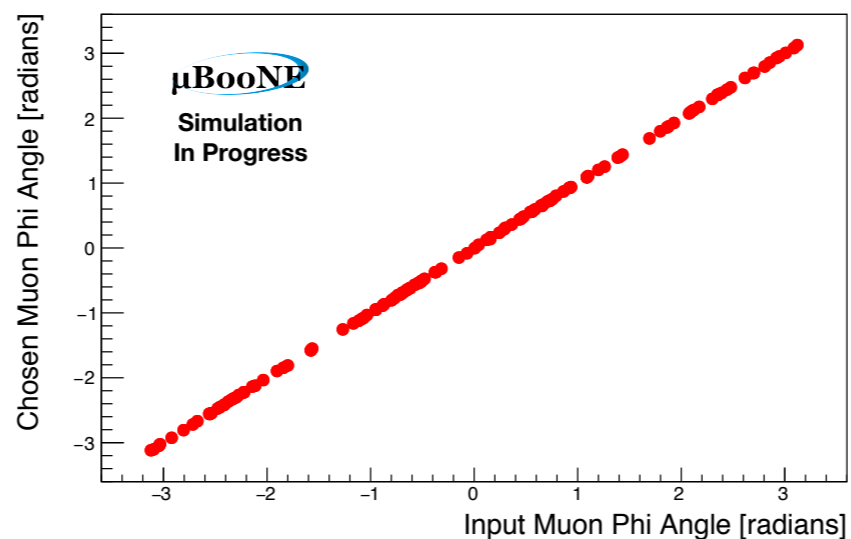
Chosen Z vs. Input Z



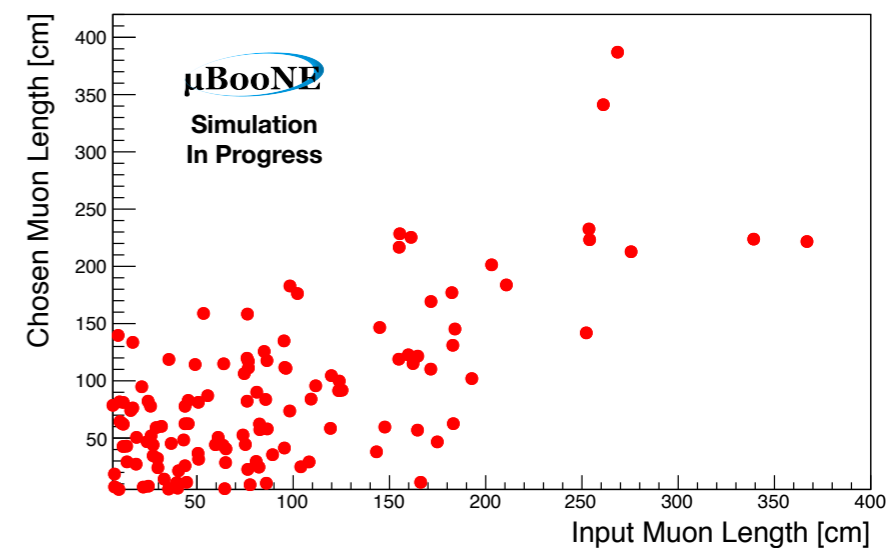
Chosen muTheta vs. Input muTheta



Chosen muPhi vs. Input muPhi



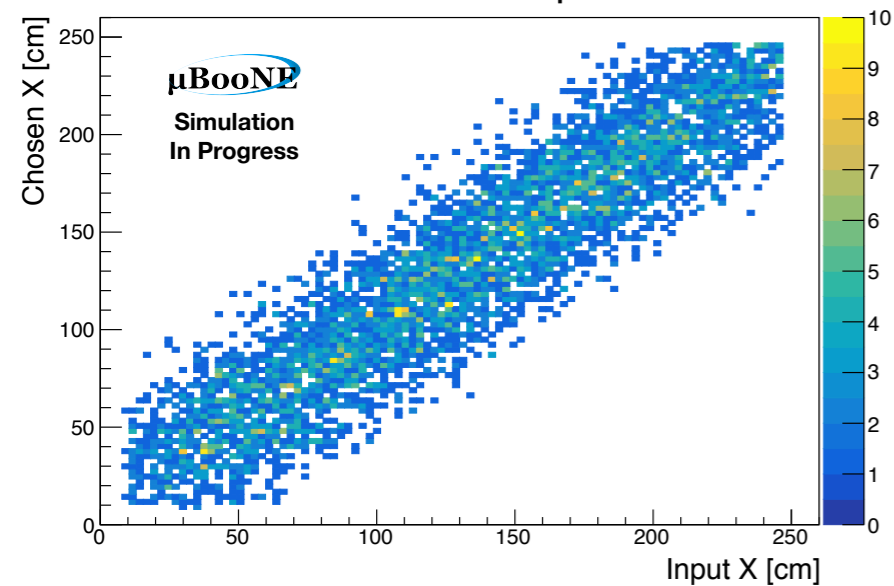
Chosen muLen vs. Input muLen



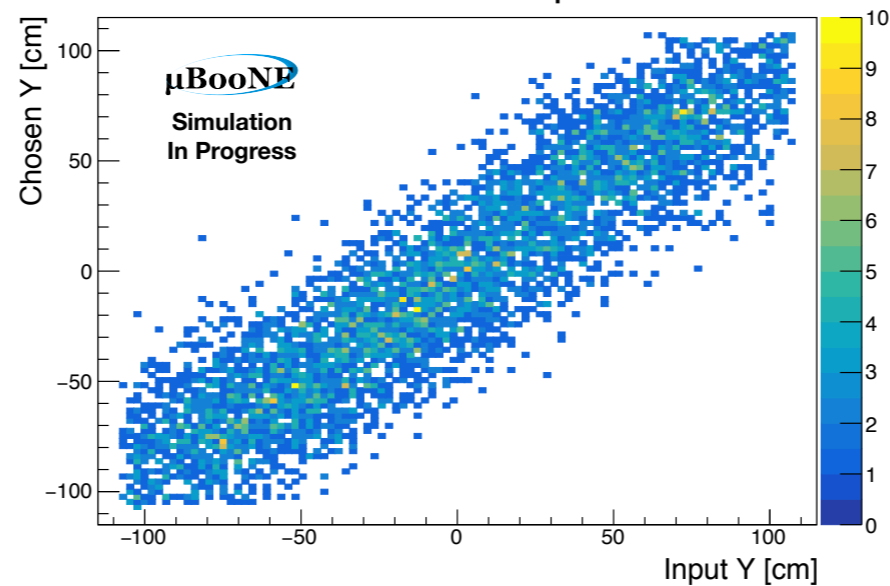
↗  
**Don't need phi to be this exact**

# First pass at optimizing ( $\sigma_\theta = 0.01$ , $\sigma_\phi = 0.01$ , all other $\sigma = 1.0$ )

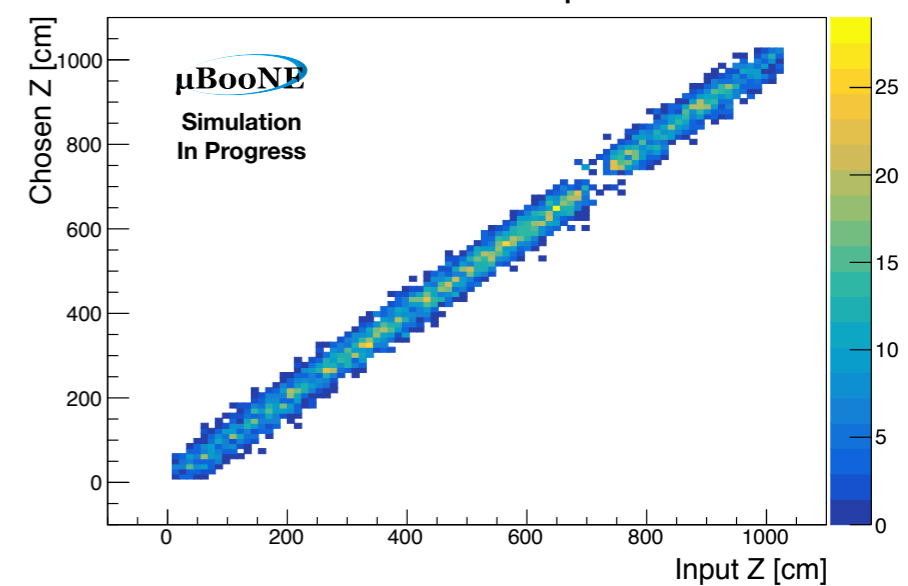
Chosen X vs. Input X



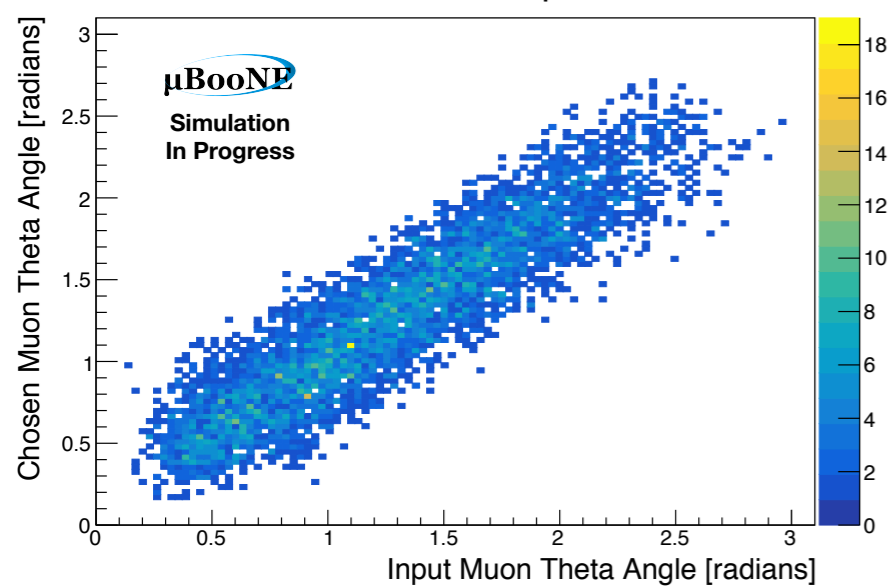
Chosen Y vs. Input Y



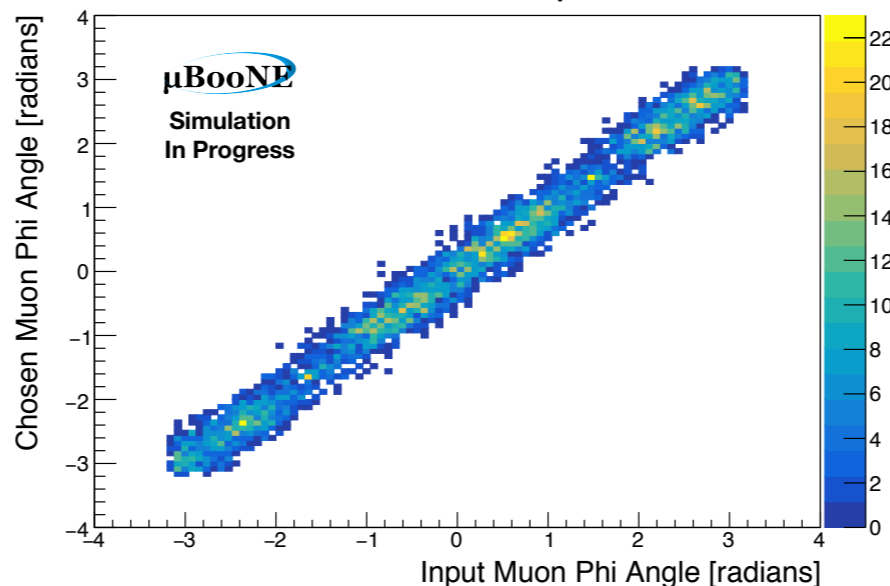
Chosen Z vs. Input Z



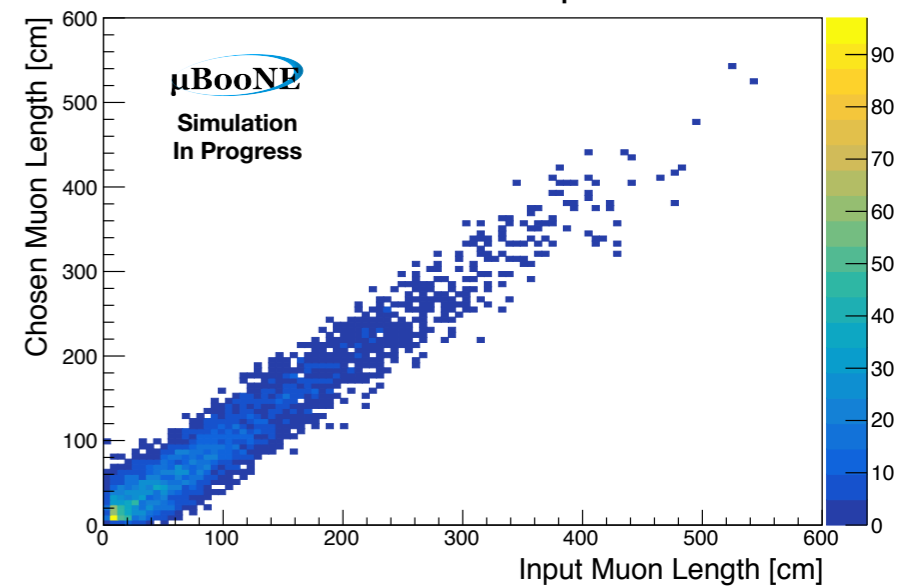
Chosen muTheta vs. Input muTheta



Chosen muPhi vs. Input muPhi

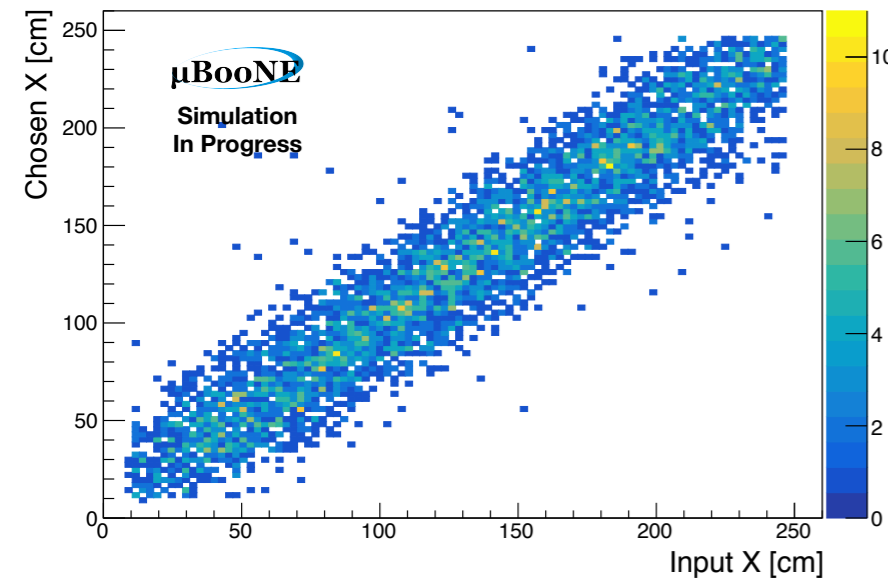


Chosen muLen vs. Input muLen

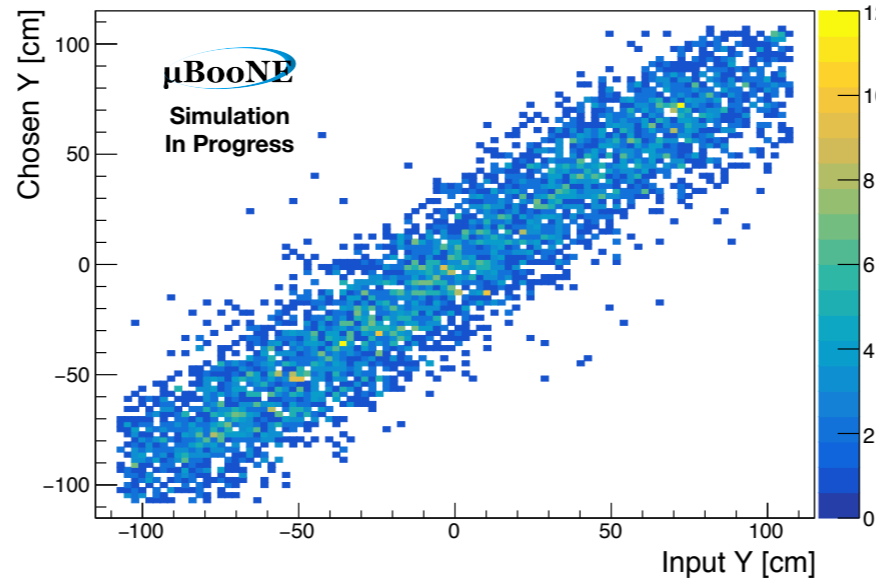


# Protons with $\sigma_\theta = 0.01$ , $\sigma_\phi = 0.01$ , all other $\sigma = 1.0$

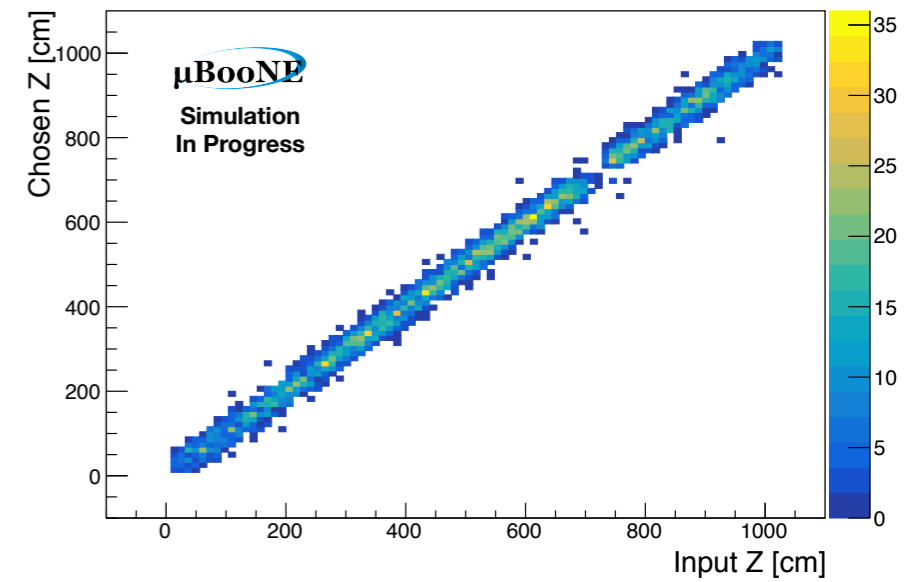
Chosen X vs. Input X



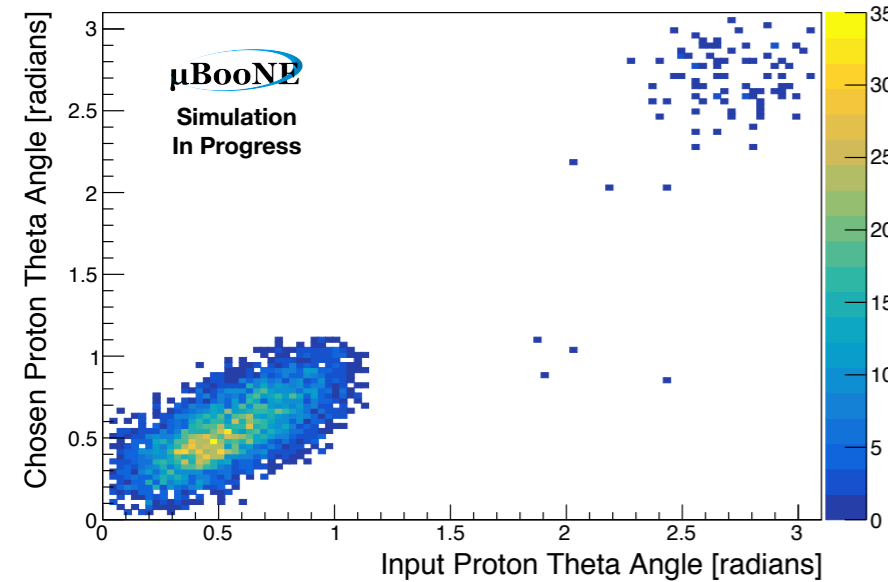
Chosen Y vs. Input Y



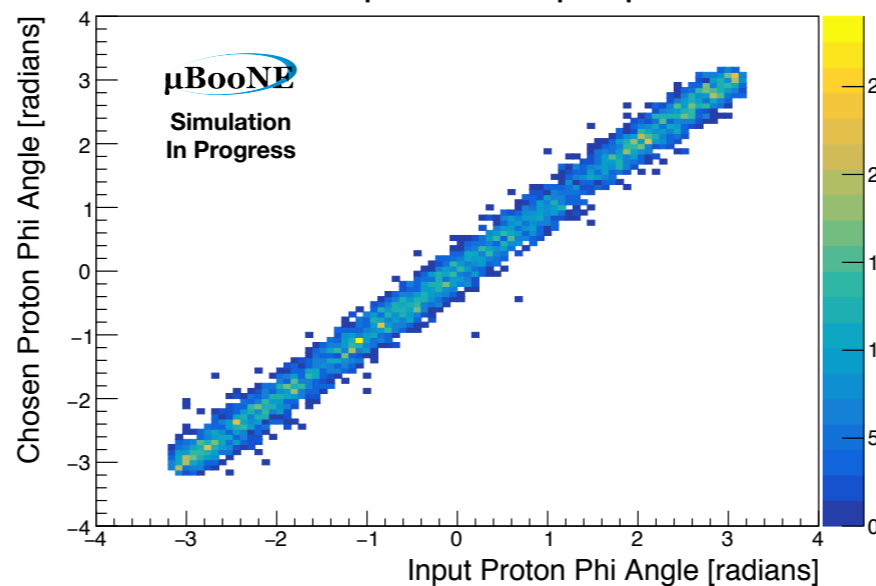
Chosen Z vs. Input Z



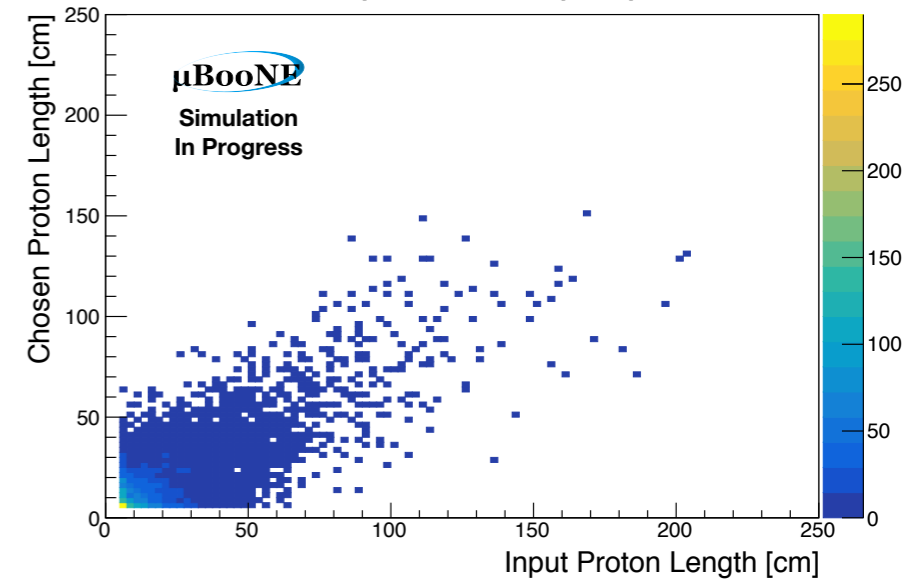
Chosen pTheta vs. Input pTheta



Chosen pPhi vs. Input pPhi



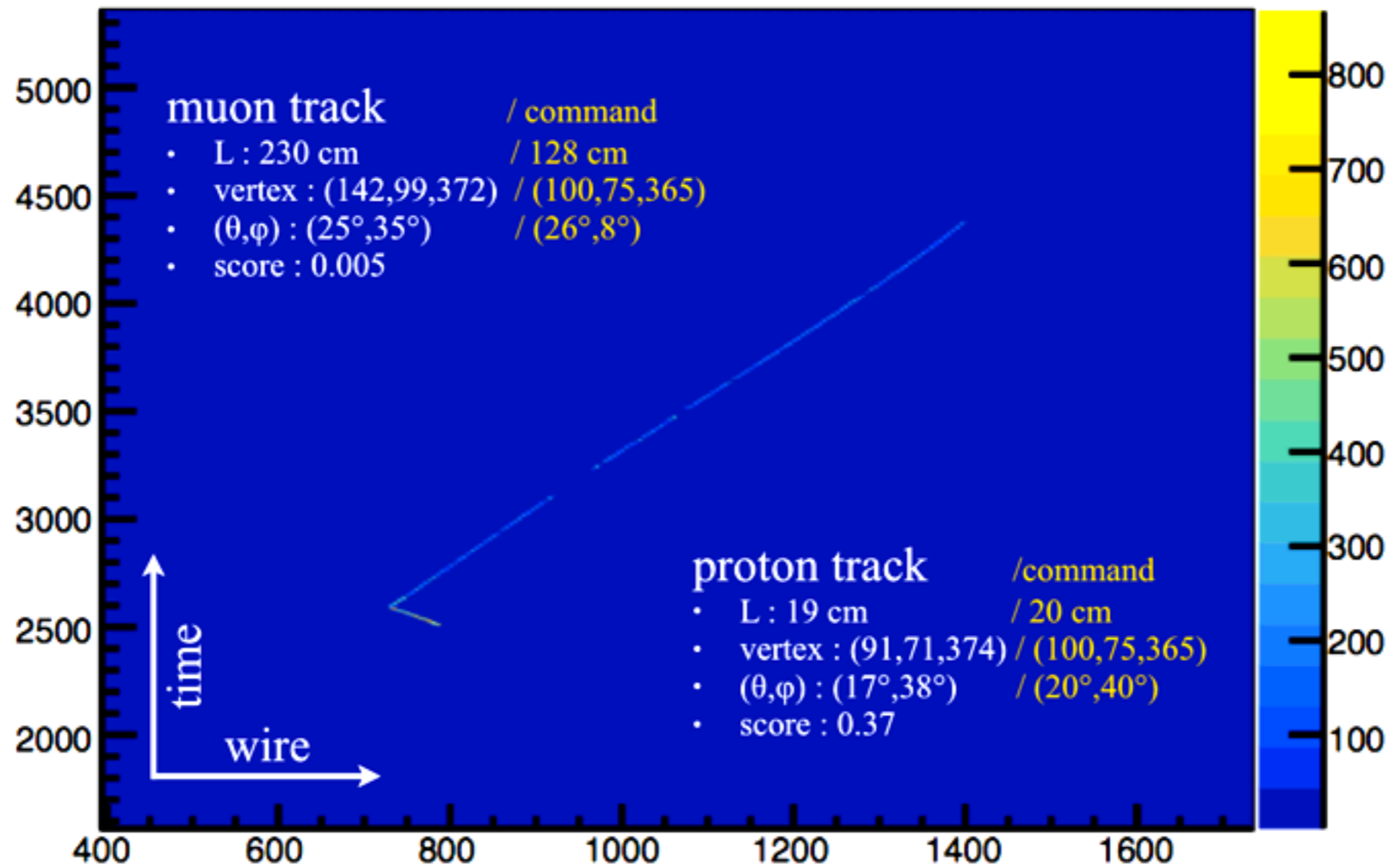
Chosen pLen vs. Input pLen





# Chimera Events

*The first chimera event!*



# Conclusion and Outlook

- We are able to find closely matching tracks successfully
- Can tune precision of each parameter:  $X$ ,  $Y$ ,  $Z$ ,  $\theta$ ,  $\phi$ , length
  - Muons and protons
  - Electrons to come!
- Next step: produce a large sample of chimera events
- Can use for systematic uncertainties in MicroBooNE going forward!



***Thank you!***



# Backup Slides

# Chimera Definition

**chimera** (kai'miərə; ki-) or **chimaera**

*n*

1. (Classical Myth & Legend) (*often capital*) *Greek myth* a fire-breathing monster with the head of a lion, body of a goat, and tail of a serpent
2. (Art Terms) a fabulous beast made up of parts taken from various animals
3. a wild and unrealistic dream or notion

...Hopefully not!



# $\nu_{\mu}$ Event Selection

1. Reconstructed vertex is inside fiducial volume
2. Exactly two tracks reconstructed
3. Tracks are well-reconstructed\*
4. Tracks are fully contained in the active volume
5. Consistency between three different initial neutrino energy reconstruction methods\*
6. Cuts placed on reconstructed transverse momentum of the interaction\*
7. Reconstructed  $Q^2 > 0^*$
8. Reject events with significant shower activity in either track

\*For details, see:  
[MICROBOONE-NOTE-1051-PUB](#)

# Detector Systematics

Systematic Sample	Relative Uncertainty [%]
Induced Charge Effect	13.0
Light Yield Model	4.7
Channel Saturation	4.3
Space Charge Effect	3.7
TPC Visibility	3.7
Electron Lifetime	2.9
Misconfigured Channels	1.8
Longitudinal Diffusion	1.7
Transverse Diffusion	1.6
PE Noise	0.4
Wire Response	0.2
Wire Noise	0.1
Electron Recombination	0.1

**Latest MicroBooNE  
Cross-Section Measurement:**  
[arXiv:1905.09694](https://arxiv.org/abs/1905.09694)

**MicroBooNE Signal Processing Papers:**  
[JINST 13, P07006 \(2018\)](#)  
[JINST 13, P07007 \(2018\)](#)