Overcoming Neutrino Interaction Mis-modeling with DUNE-PRISM

New Perspectives 2019

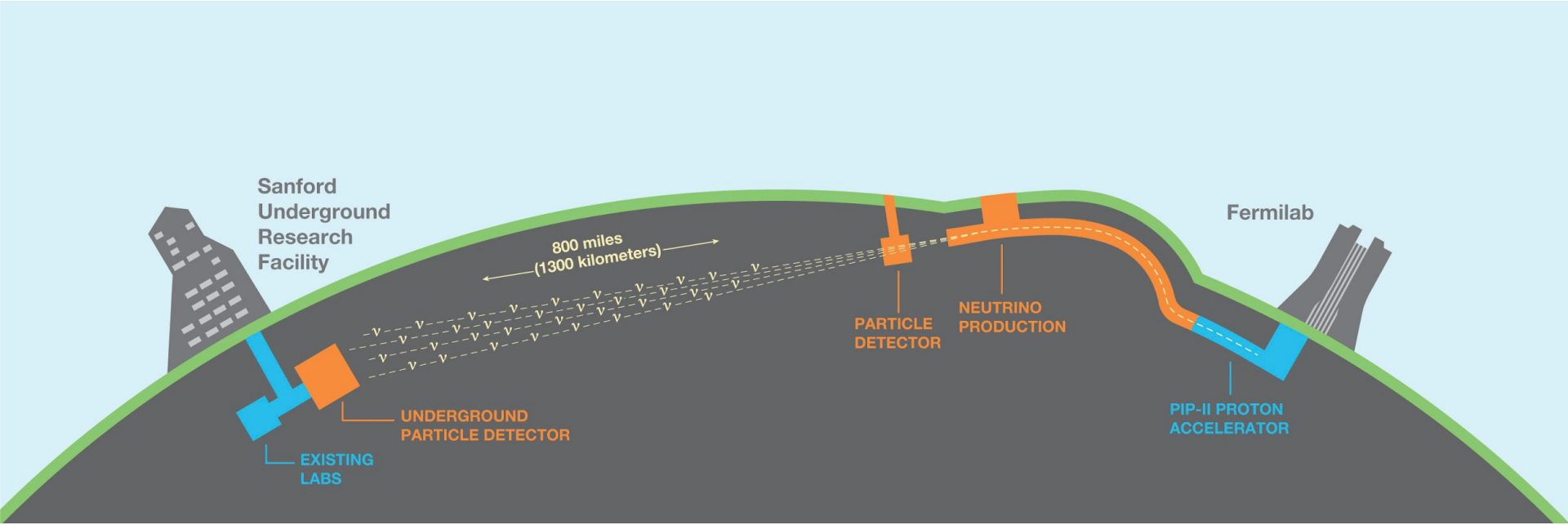
2019-06-11

Luke Pickering for the DUNE collaboration

MICHIGAN STATE UNIVERSITY

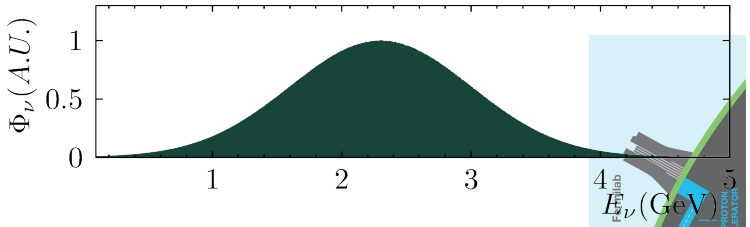
DUNE DEEP UNDERGROUND NEUTRINO EXPERIMENT

DUNE



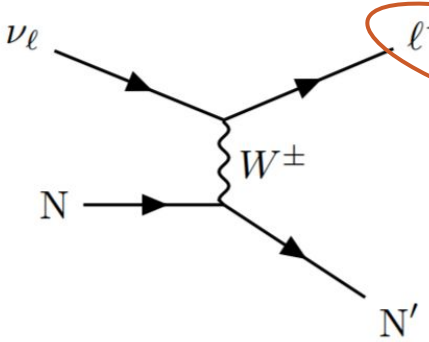
Do say: I love DUNE!, **Don't say**: *<anything>* the DUNE experiment *<anything else>*

Oscillations



e.g. created as muon neutrinos

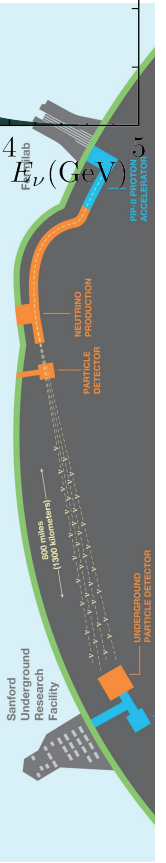
1) Interaction with matter in flavor eigenstate defined by charged lepton.



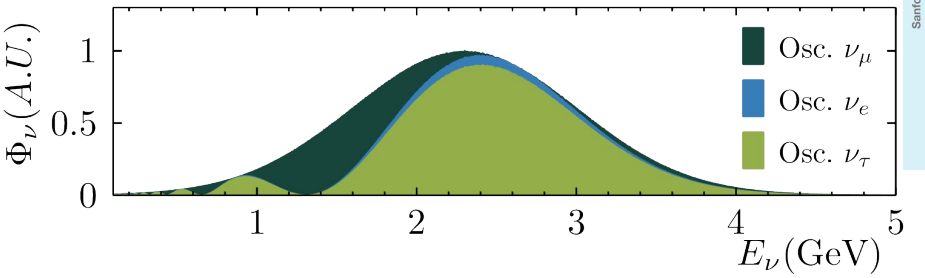
$$\begin{pmatrix} \nu_e \\ \nu_\mu \\ \nu_\tau \end{pmatrix} = \underbrace{\begin{bmatrix} U_{e1} & U_{e2} & U_{e3} \\ U_{\mu1} & U_{\mu2} & U_{\mu3} \\ U_{\tau1} & U_{\tau2} & U_{\tau3} \end{bmatrix}}_{\mathbf{M}_{\mathrm{PMNS}}} \begin{pmatrix} \nu_1 \\ \nu_2 \\ \nu_3 \end{pmatrix}$$

Pontecorvo–Maki–Nakagawa–Sakata

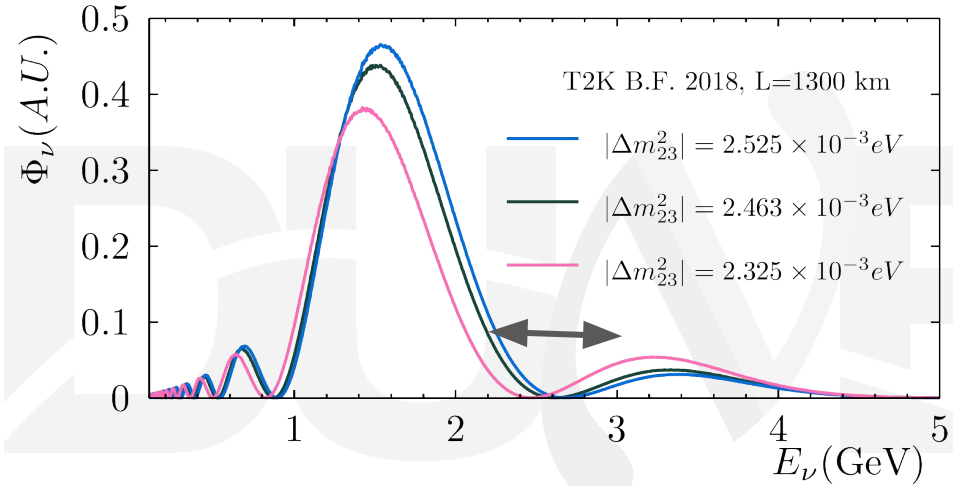
2) Propagate as superposition of mass/energy eigenstates over experimental baseline (1300 km)

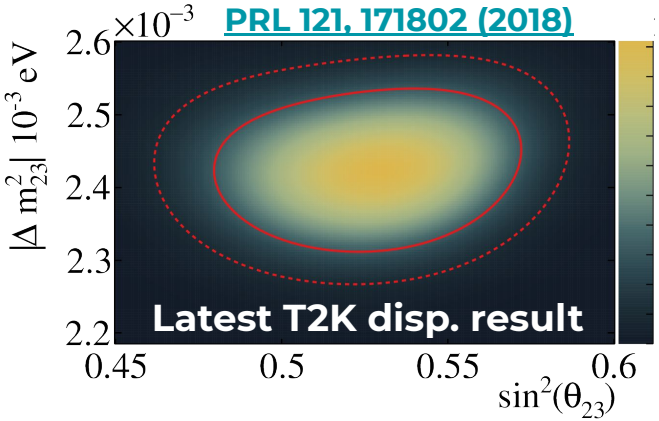


3) Projecting back to flavor eigenstates reveals a different flavor mixture.
(if $|\Delta m^2_{ij}| \neq 0$)

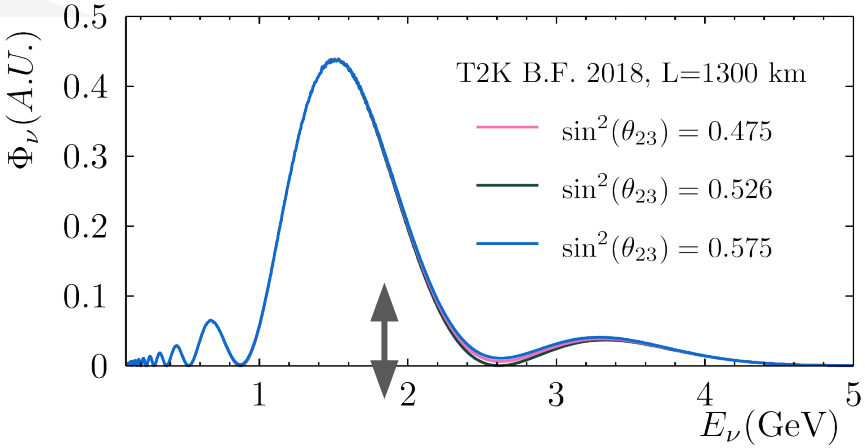


Disappearance at the Far detector





- ‘Surviving’ muon neutrinos show characteristic oscillation shape.
- Use details of spectra to infer physics parameters of interest (mixing angles, mass differences, CPV phase)
- Similarly compare to ‘appeared’ electrons.

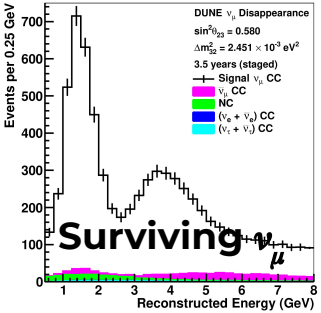


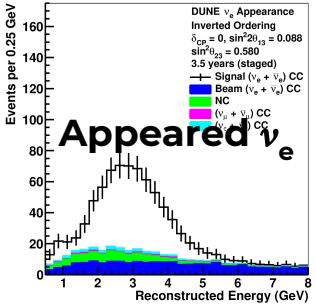
An Oscillation Analysis (OA) in one slide

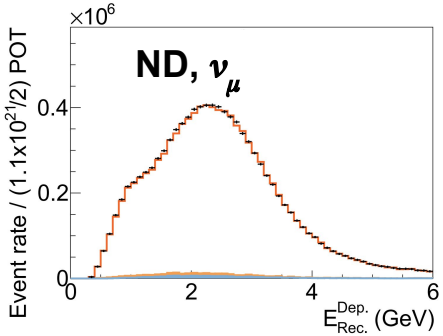
- Constrained observable prediction
- Data → Infer oscillation probabilities

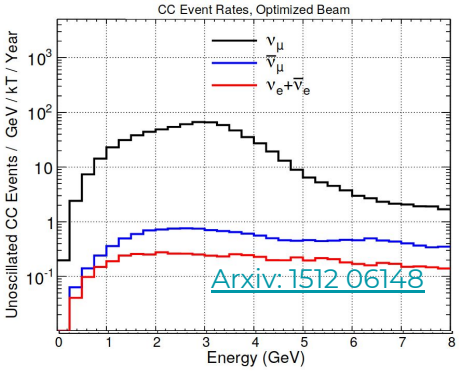
- Predict observables
- Data → constrain interaction physics

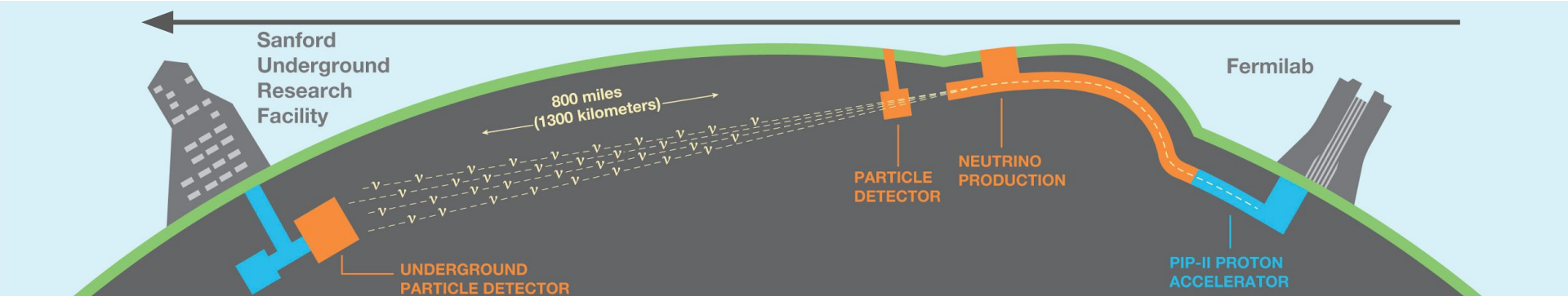
Predict neutrino beam









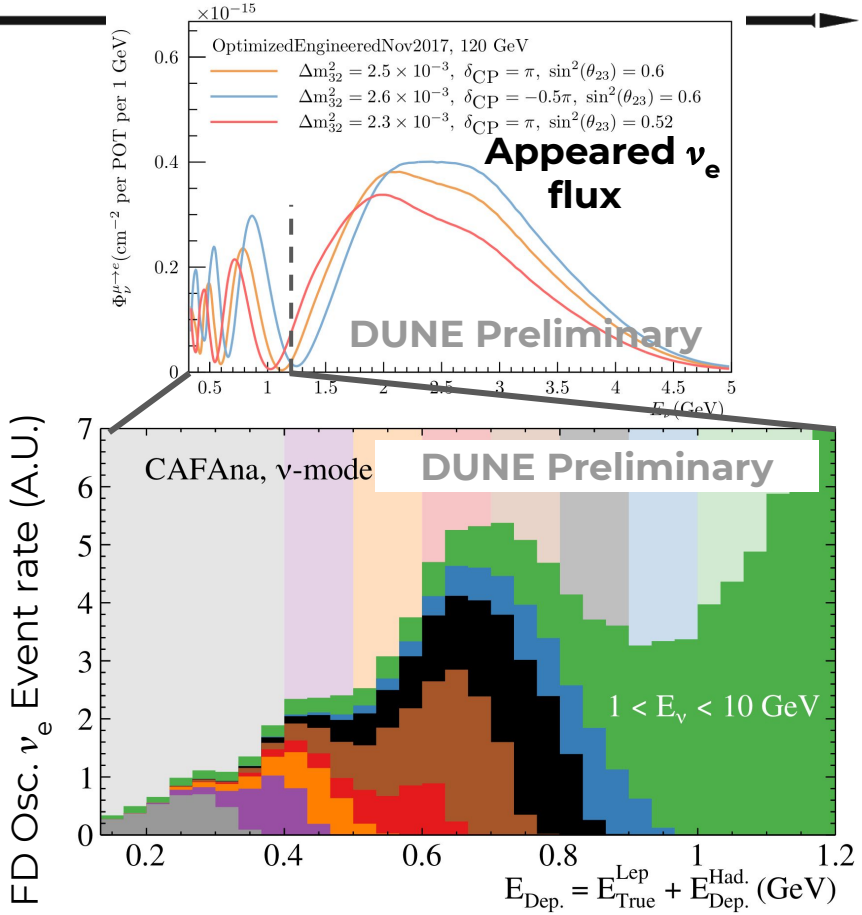


Why are neutrino interaction models important?

$$N_{\text{near}}(\mathbf{x}_{\text{obs}}) = \int d\mathbf{x}_{\text{true}} \underbrace{\mathbf{D}_{\text{near}}(\mathbf{x}_{\text{obs}}|\mathbf{x}_{\text{true}})}_{\text{Smearing, Eff., Pur.}} \underbrace{N_{\text{targ}}\sigma(\mathbf{x}_{\text{true}})\,\Phi(E_\nu)}_{N_{\text{Int}}(\mathbf{x}_{\text{true}})}$$

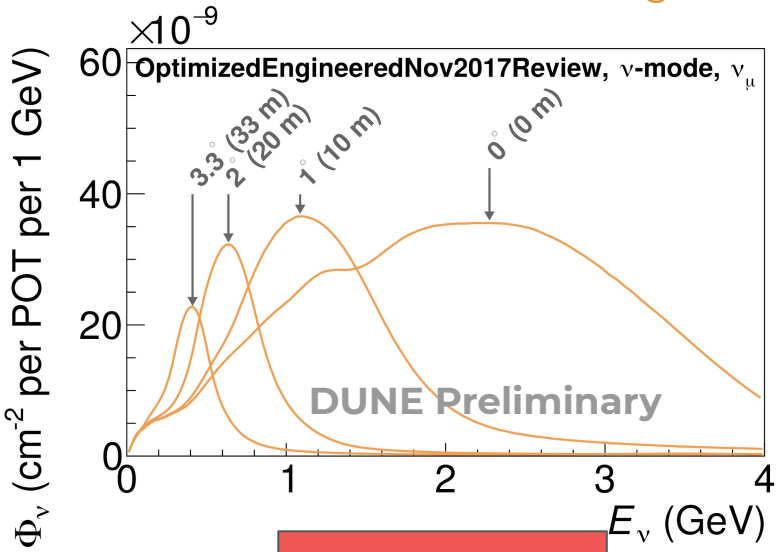
$$N_{\text{far}}(\mathbf{x}_{\text{obs}}) = \int d\mathbf{x}_{\text{true}} \underbrace{\mathbf{D}_{\text{far}}(\mathbf{x}_{\text{obs}}|\mathbf{x}_{\text{true}})}_{\text{Smearing, Eff., Pur.}} \underbrace{N_{\text{targ}}\sigma(\mathbf{x}_{\text{true}})\,\Phi(E_\nu)\,P_{osc}(E_\nu)}_{N_{\text{Int}}(\mathbf{x}_{\text{true}})}$$

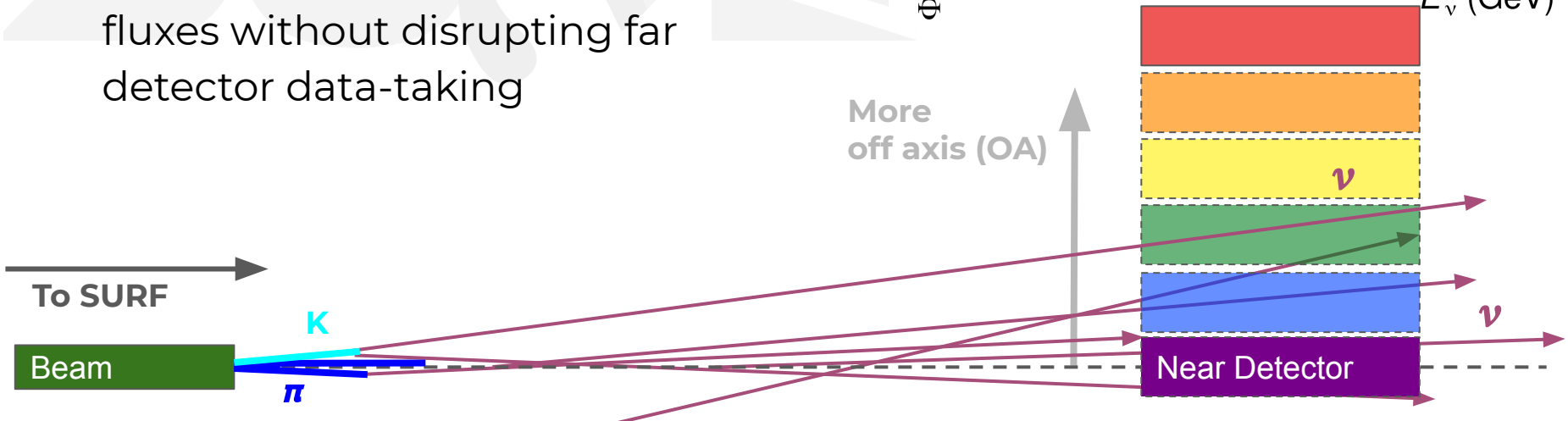
- Observe event rate not neutrino flux
- Cannot perfectly reconstruct neutrino energy
- Require models to predict observables and infer oscillation features in true neutrino energy spectra
- Mis-modelling in ERed feed-down → biased parameter measurements.



DUNE-PRISM

- Neutrino beam from boosted pion and kaon decays:
  - peak-energy moves down away from neutrino beam axis
- A mobile near detector could take data in a range of neutrino fluxes without disrupting far detector data-taking

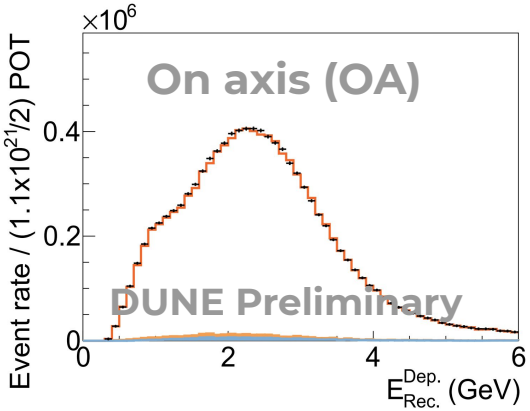


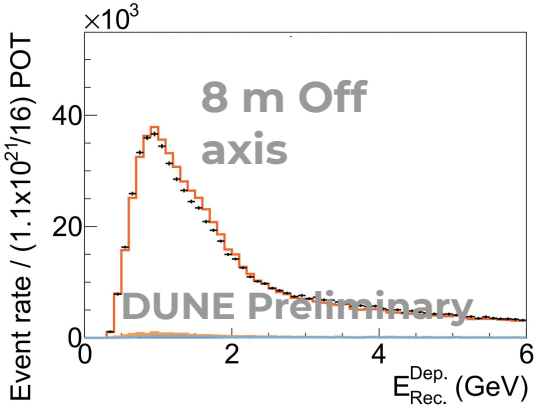


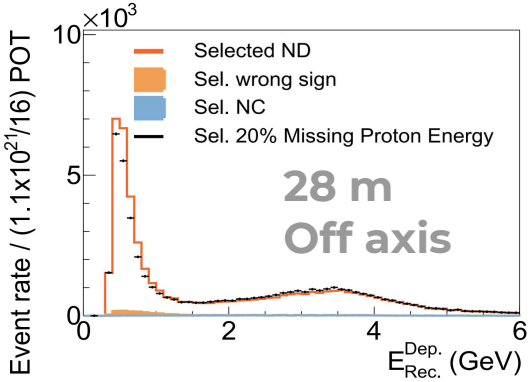
Improvise

$$N_{\text{near}}(\mathbf{x}_{\text{obs}}) = \int d\mathbf{x}_{\text{true}} \underbrace{\mathbf{D}_{\text{near}}(\mathbf{x}_{\text{obs}}|\mathbf{x}_{\text{true}})}_{\text{Smearing, Eff., Pur.}} \underbrace{N_{\text{targ}}\sigma(\mathbf{x}_{\text{true}})\Phi(E_\nu)}_{N_{\text{Int}}(\mathbf{x}_{\text{true}})}$$

- Problems in flux/interaction/detector modelling can be hard to deconvolve by single event rate measurement (*e.g.* on-axis (OA) only)



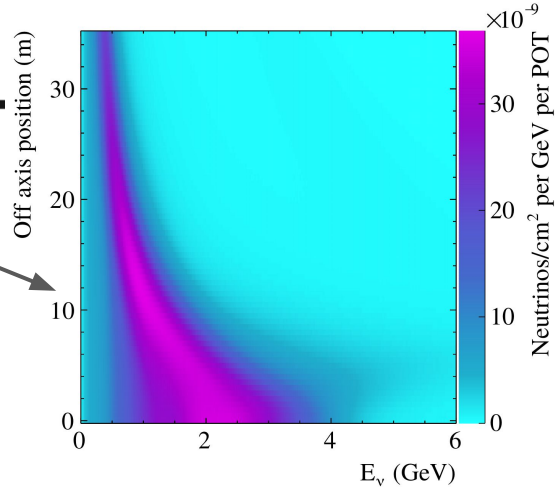




- **Case study**: 20% proton KE → neutrons and apply plausible new xsec to make hard to see on axis.
  - But visible off axis!
- Doesn't offer a way to fix any problems, just spot them…

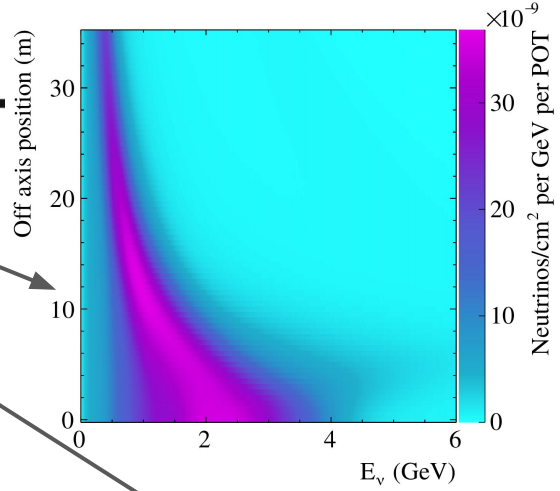
Adapt

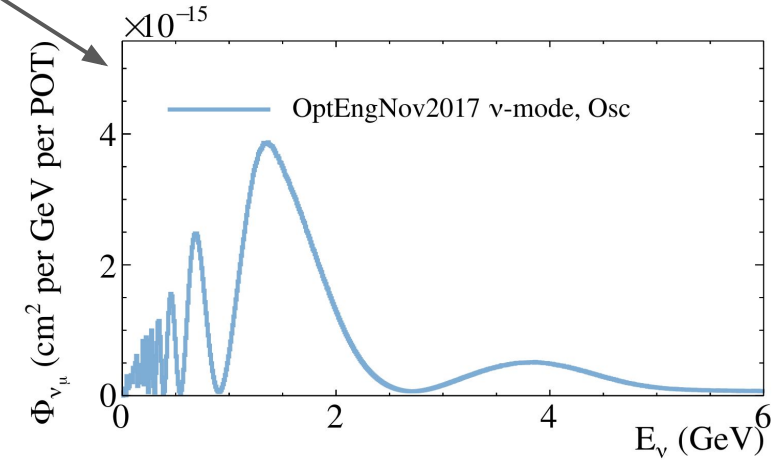
- Predict Near flux: ND



Adapt

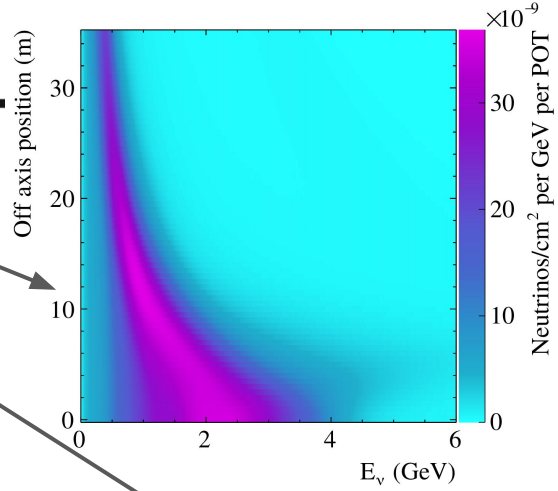
- Predict Near flux: $\mathbf{ND}$
- Can predict Far flux under various oscillation hypotheses: $\vec{f}$





Adapt

- Predict Near flux: $\mathbf{ND}$
- Can predict Far flux under various oscillation hypotheses: $\vec{f}$
- Use Near flux at different off axis positions as a linear basis and solve: $\mathbf{ND} \times \vec{c} = \vec{f}$



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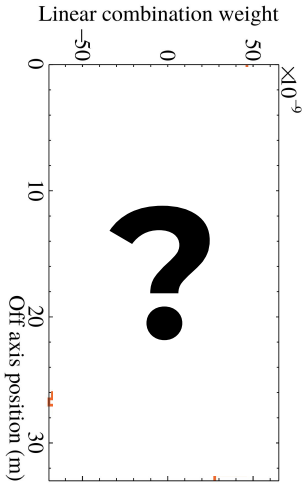

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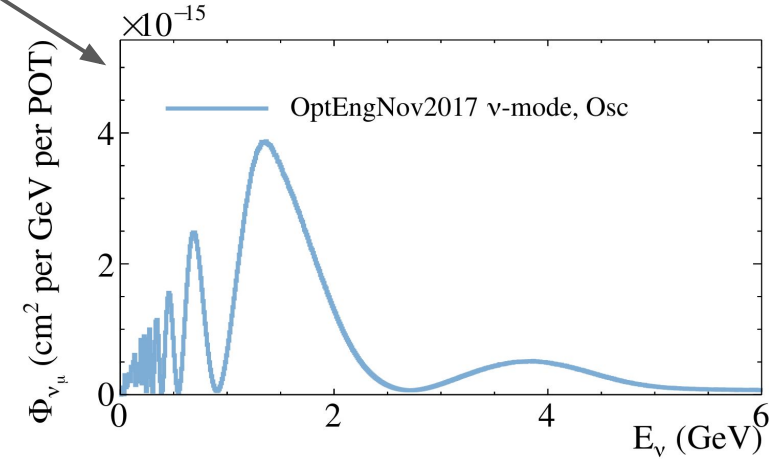

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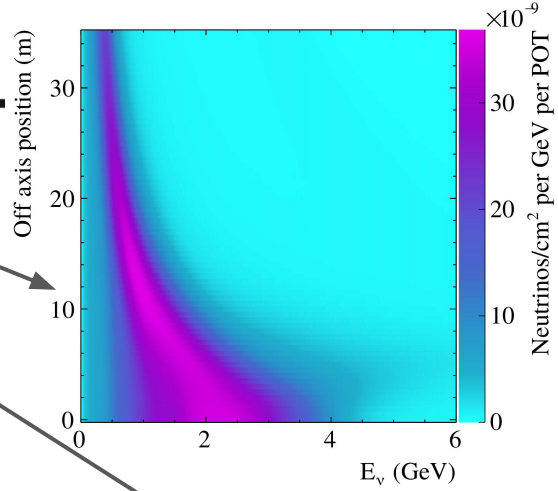

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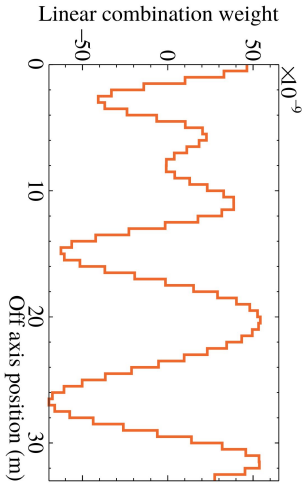

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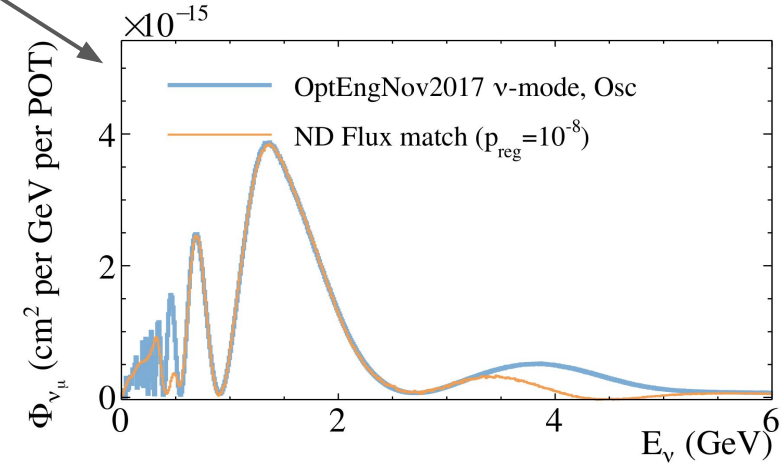
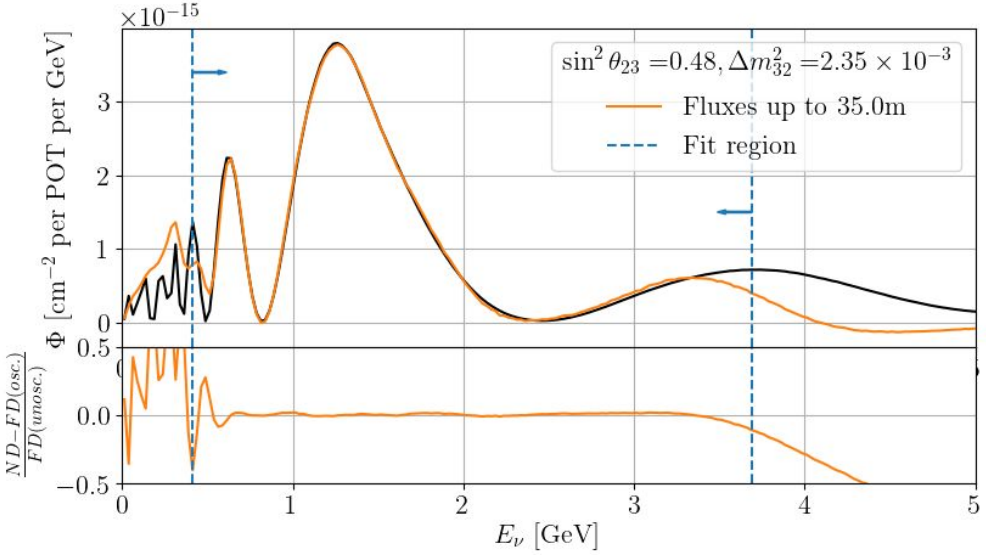
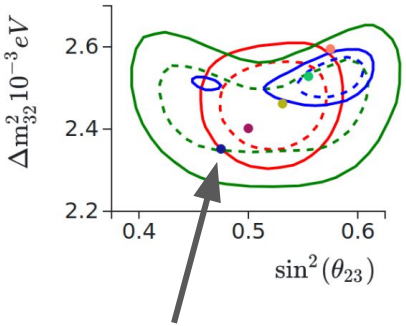
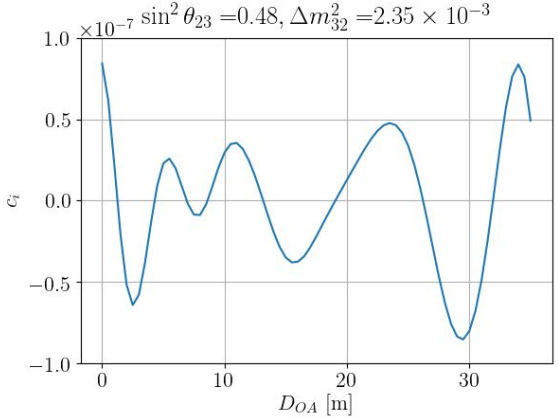


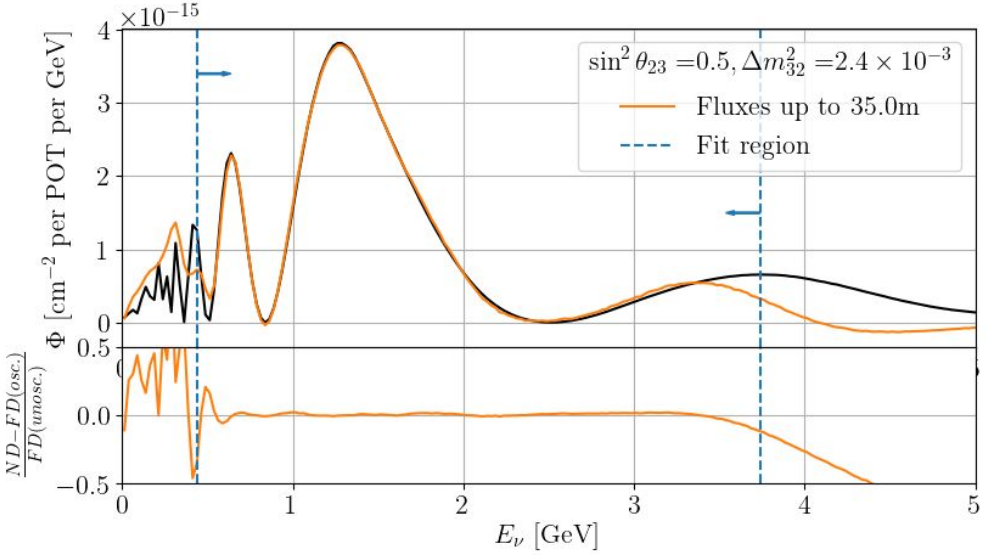


OscProb.js



T2K 2018
NOvA 2018
NuFit v4





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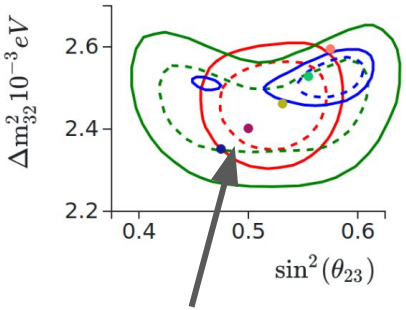

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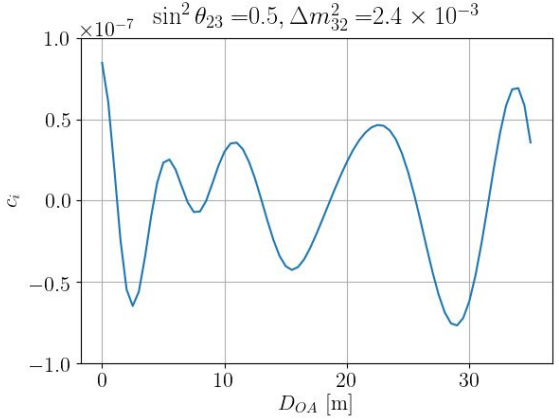

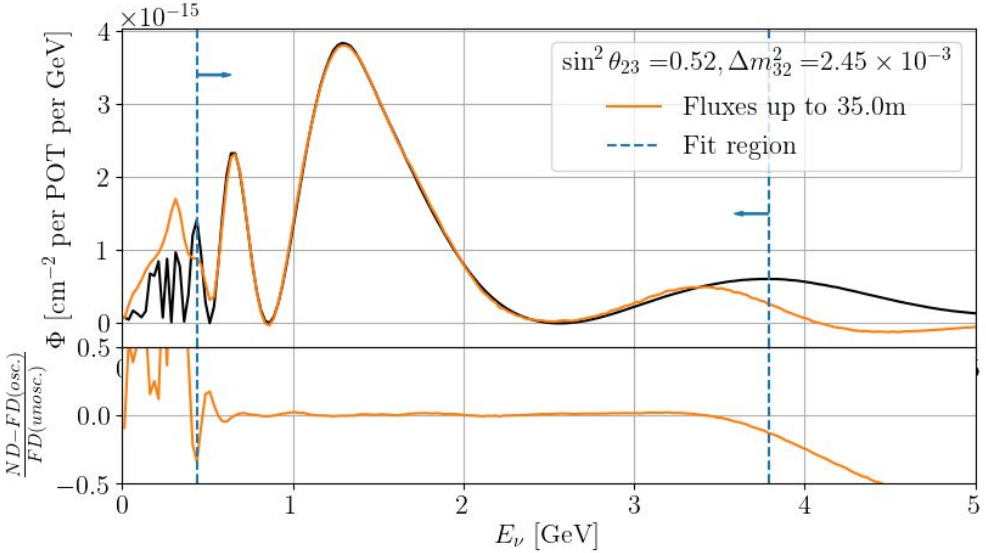

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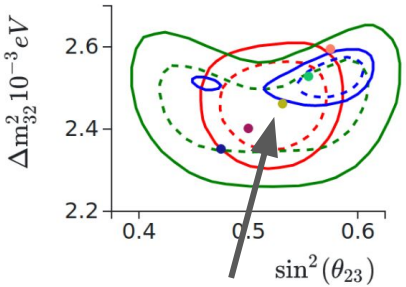

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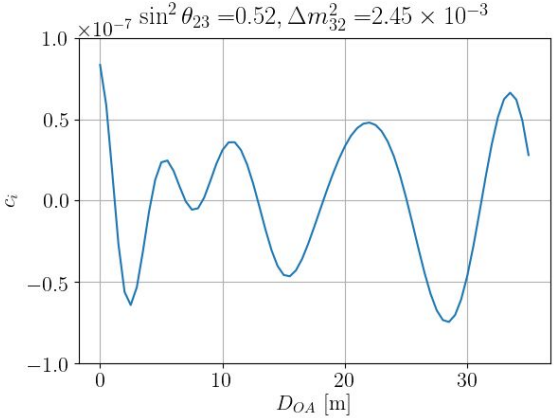

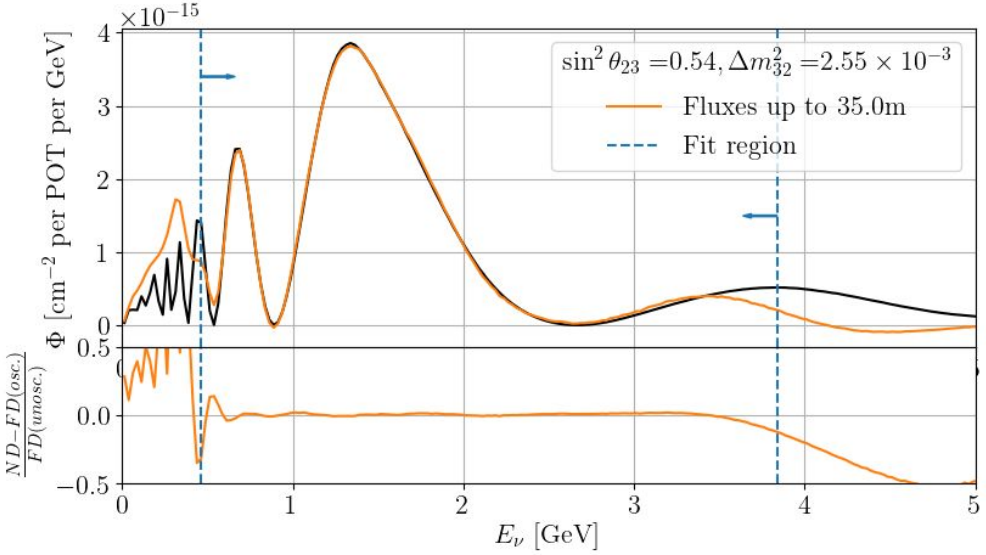

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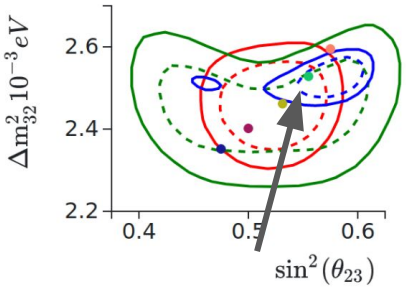

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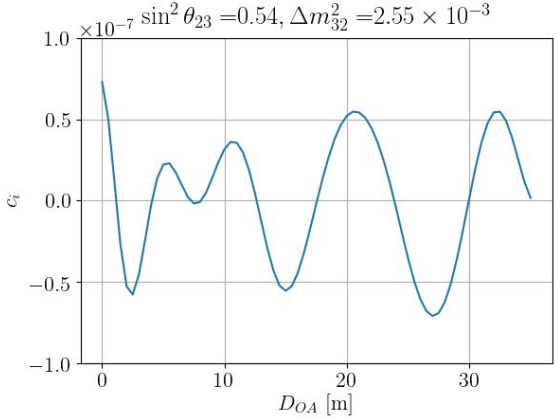

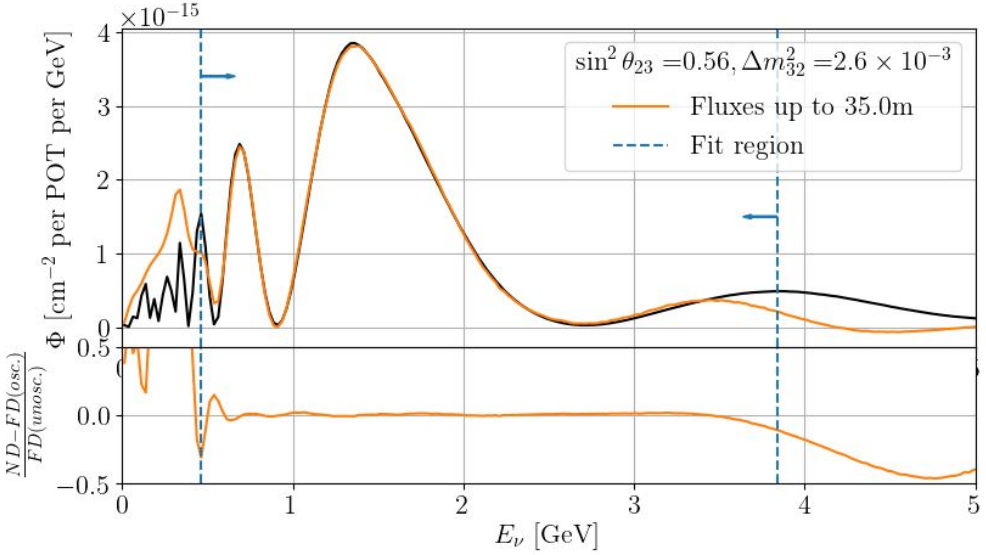

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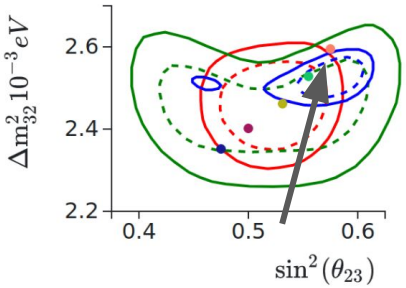

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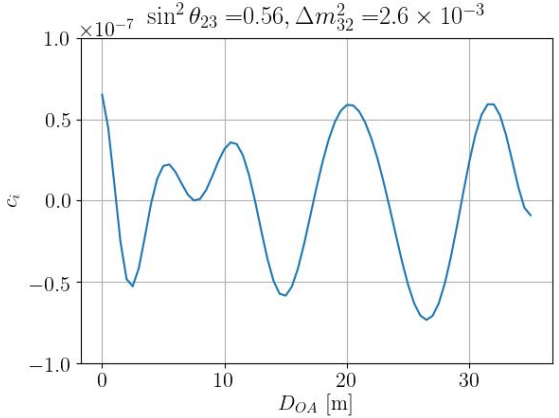
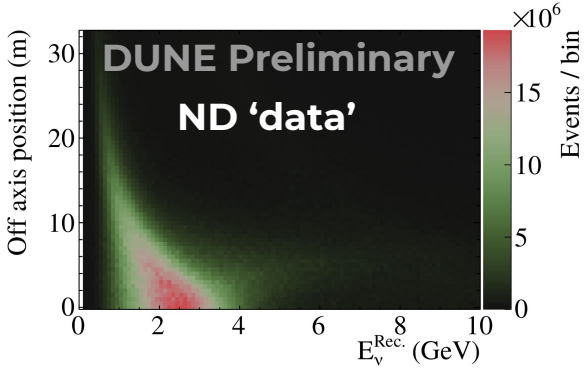


Overcome

- **Aim:** Rearrange ND data to predict FD
  - Unknown XSec features automatically transferred
  - Minimize XSec dependence and take advantage of N/F flux cancellations
  - N/F detector difference unavoidable in any analysis
- Robust to mis-modelling in observable energy distribution as use near data to fill most of the far prediction!



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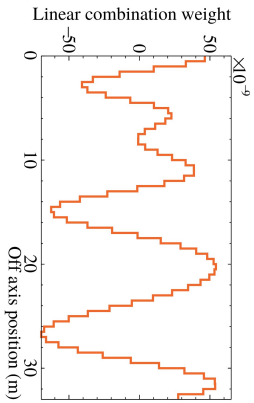
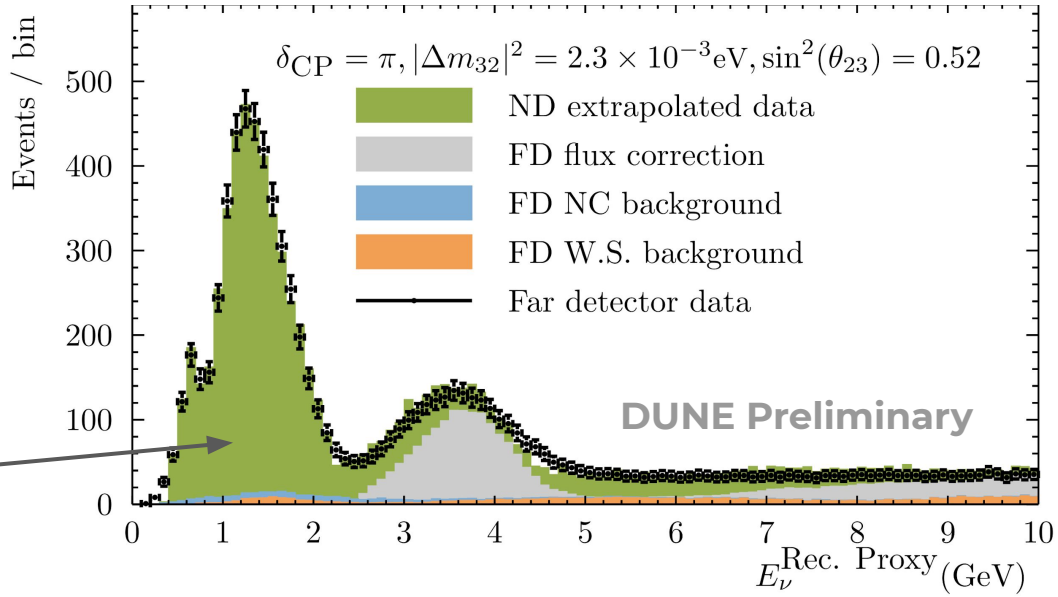




That's all Folks

- Problems in neutrino interaction models can be hard to see & fix with on-axis near detector only
- Comparing data taken in different neutrino energy spectra can illuminate such mis-modelling.
- Using linear combination of near detector data to make far detector predictions can result in an oscillation analysis that is robust to a large range of cross-section modelling problems.

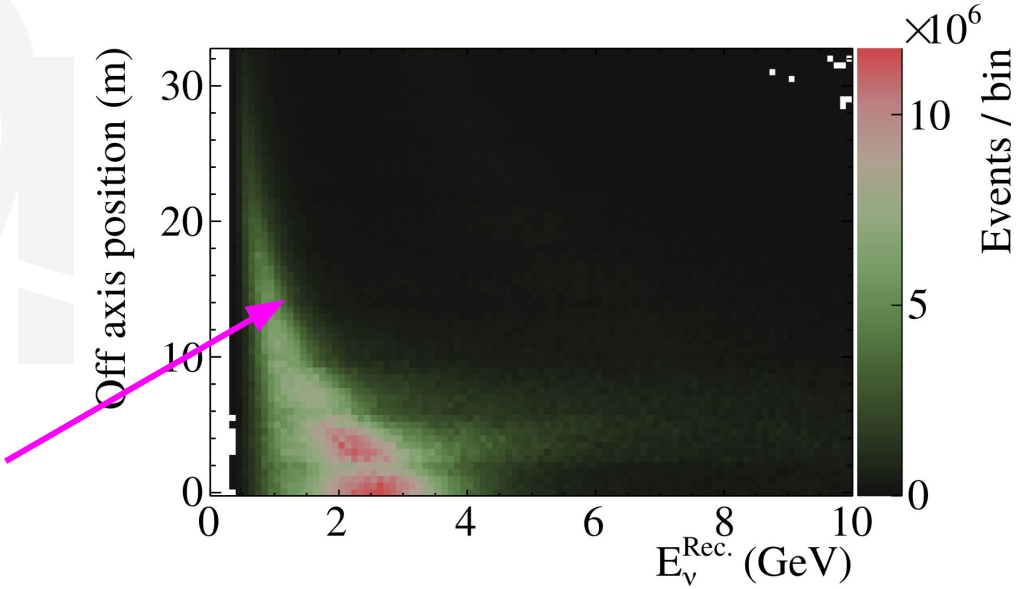
Thanks for listening

DUNE-PRISM Propagation

- **Aim:** Rearrange ND data to predict FD
  - Unknown XSec features automatically transferred
  - Minimize XSec dependence and take advantage of N/F flux cancellations
  - N/F detector difference unavoidable in any analysis
- In each systematic universe/fit step:
  1. Select data at ND
  2. Subtract ND backgrounds with MC prediction
  3. Correct for differences in N/F selection, resolution, fiducial mass
  4. Perform Flux match
  5. Linearly combine ND data
  6. Add FD Flux match MC correction
  7. Add FD backgrounds with MC prediction
  8. Evaluate GOF

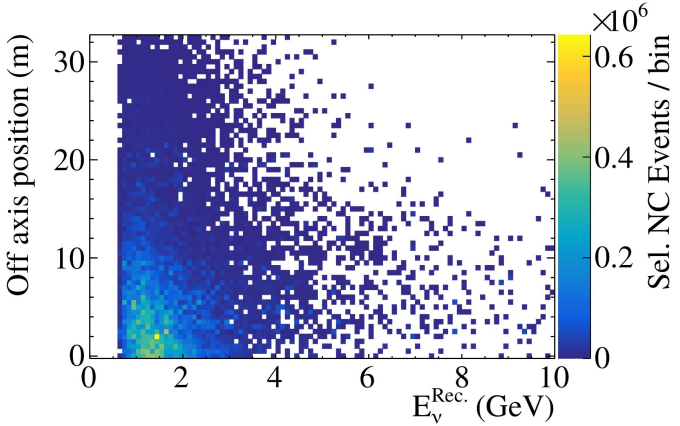
Selected ND Event Rate

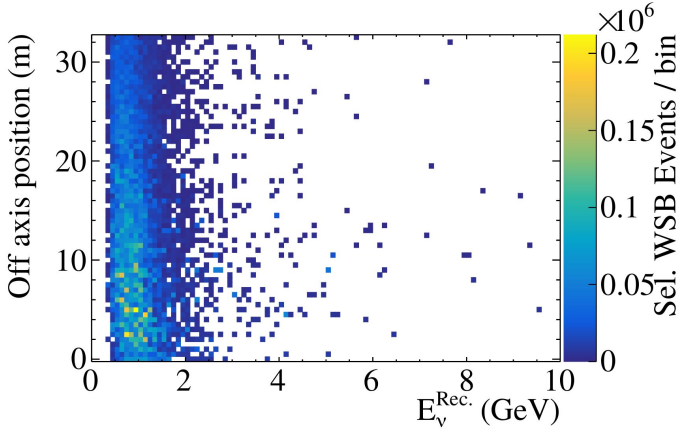
- Taking more granular steps near on-axis can mitigate edge-effects in the selection.
  - Future: Optimize stop plan



ND Backgrounds

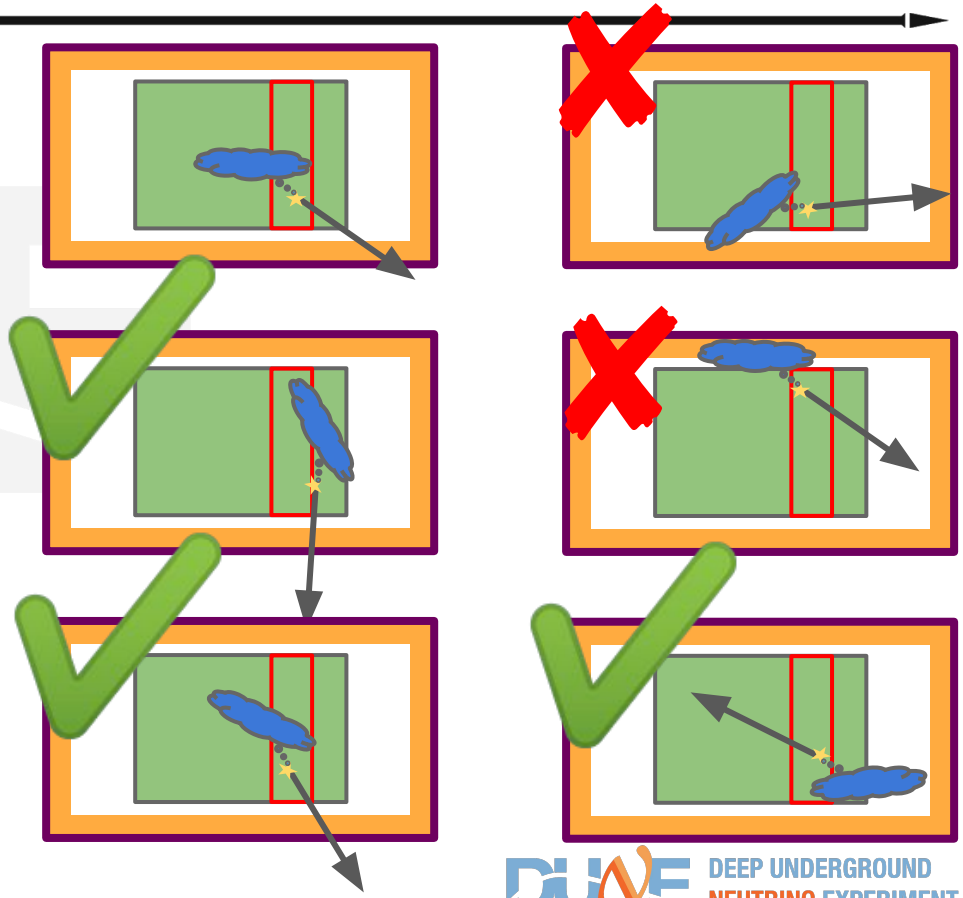
- Backgrounds that do not oscillate and vary differently as a function of off-axis position are subtracted before propagation.
- Most common:
  a. Neutral Current (Use on-axis to constrain ND and FD NCBkg)
  b. Wrong sign (worse in nubar-mode, use tracker to constrain WSBkg).
  c. Intrinsic nue
- These will get added back into the Far prediction later.





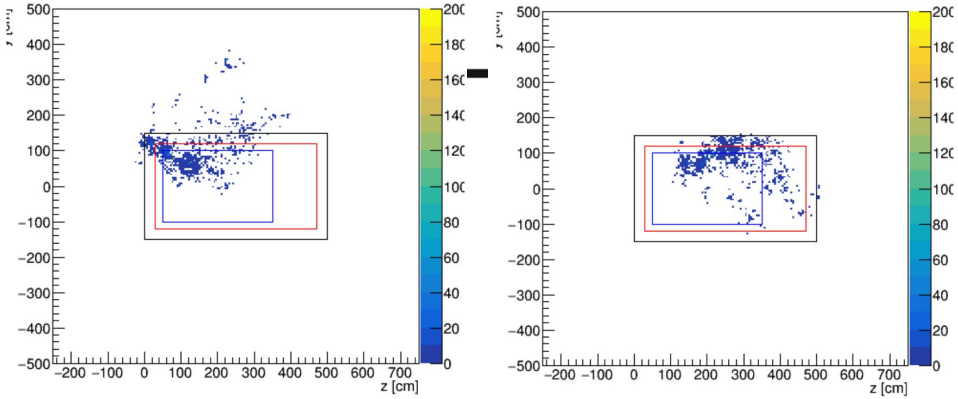
Selection Efficiency

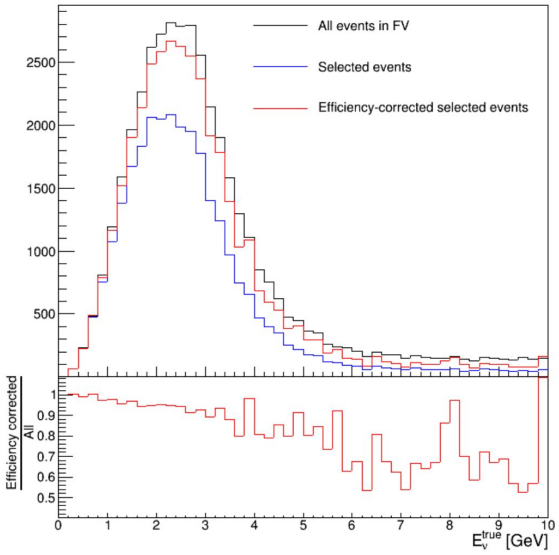
- Must correct for differences in ND/FD selection efficiency.
- Want to avoid asking GENIE everywhere possible.
- Aim to develop data-driven geometric efficiency correction:
  a. Throw away events outside acceptance ND-FD high acceptance union
  b. Add MC events that are in FD but outside ND

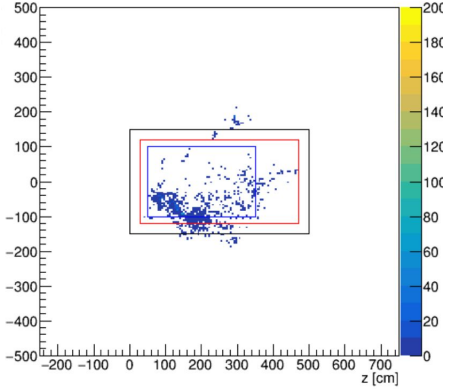


Geometric Efficiency

- Preliminary work by Cris Vilela:
- Random translation and rotation of energy deposits in selection volume
  a. Suggests 95% of events can be corrected in a model-independent data-driven way at the oscillation peak
  b. As expected from Chris Marshalls ND acceptance studies.
  c. Even higher fraction at lower energies.

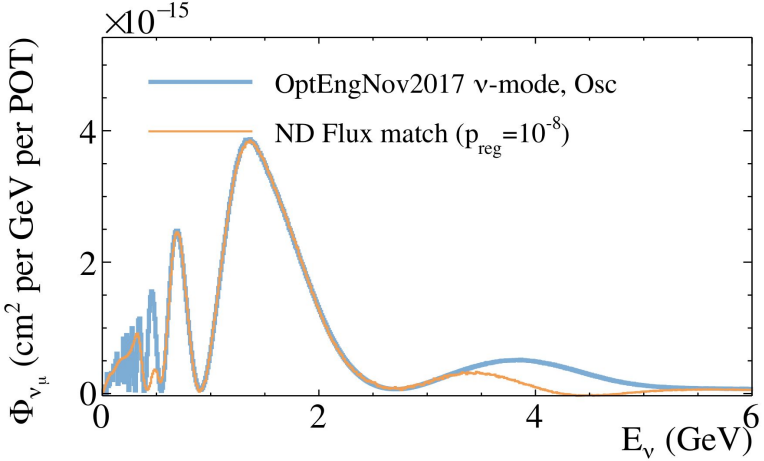


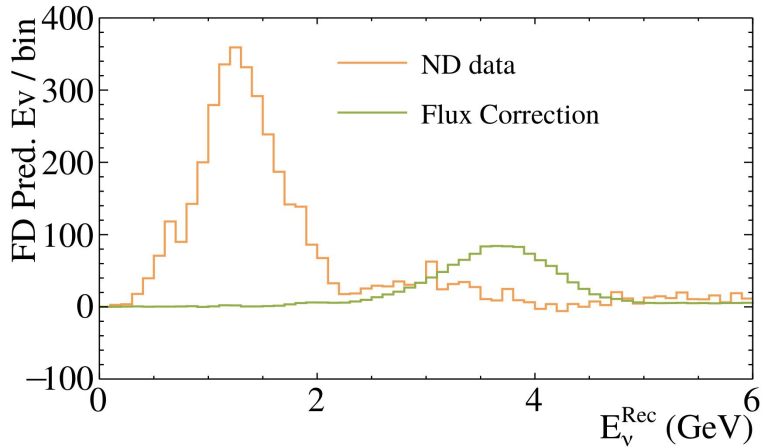


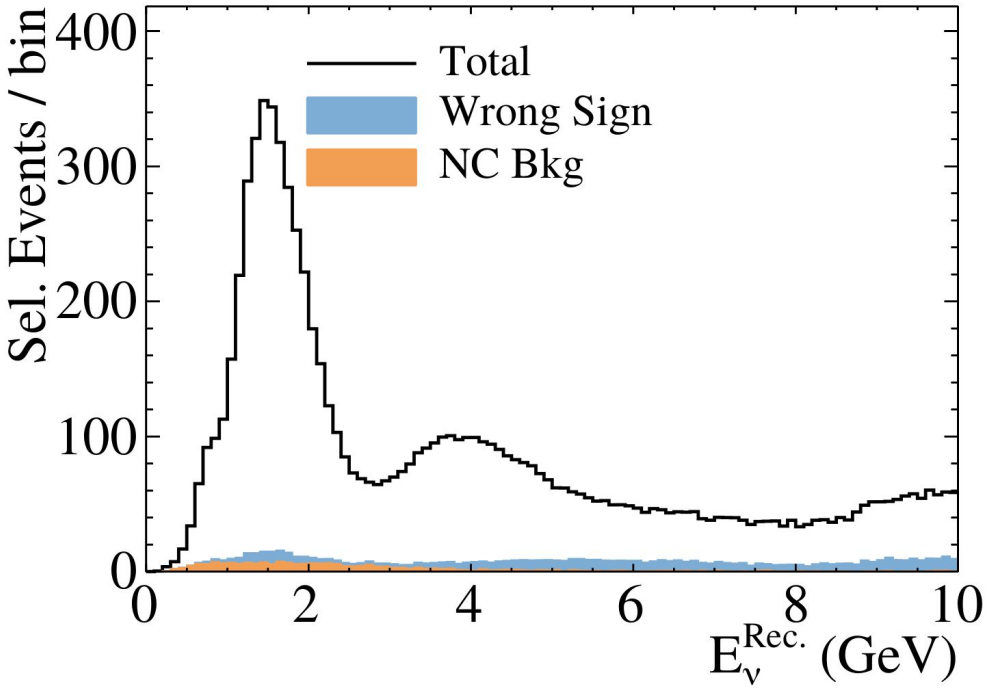


Flux Matching Correction

- Flux matching not perfect in general:
  - Especially at higher energy due to on-axis configuration
- Difference between 'target' and 'matched' filled in with FD MC predictions.
  - This 'filling in' is the same as the tuned-prediction 'dead-reckoning' that makes the entire FD comparison in the standard analysis.
  - **Here:** Majority of FD prediction built with ND data.





FD Backgrounds

- Add back in any sources of FD background that we removed before:
  - Oscillated wrong sign background (Can use nu-mode ND data to build nubar-mode FD wrong sign prediction).
  - NC Backgrounds (Use on-axis ND to understand NCBkg.(