

CC0pi/CC-inclusive Data Comparisons

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Introduction

- Learnt from the previous round of NIWG fits that there are tensions within our current models between the external data sets.
- Work since then has been trying to understand what causes some of these problems.
- **Aim:** Find a cross-section model that is capable of explaining a broad range of the published datasets and try to relieve the tensions.

Updated PGof Fits

- Recap: MiniBooNE and MINERvA joint fits show a disagreement on best fit model parameters. NEUTs model cannot describe both simultaneously.
- MINERvA updated their flux which helped a little bit, but there are still issues!

MINERvA Flux Update

After and Before

Type	$\chi^2/NDOF$	(TN192)	$\chi^2_{PG}/NDOF_{PG}$	(TN192)	PGof (%)	(TN192)
RFG MEC	101/226	-	24.9/6	-	0	-
RFG Rel RPA MEC	105/226	97	14.2/6	18	2.8	0.66
RFG Non Rel RPA MEC	119/226	118	18.3/6	25	0.6	0.03
SF MEC	106/226	97	24.0/6	41	0.1	0.0

Parameter	M_A [GeV]	TN192	p_F [MeV]	TN192	MEC (%)	TN192
RFG MEC	1.34 ± 0.02	-	229 ± 7	-	0 (limit)	-
RFG Rel.RPA MEC	1.16 ± 0.03	1.15	219 ± 6	223	35 ± 11	27
RFG Non.Rel.RPA MEC	1.07 ± 0.02	1.07	223 ± 6	225	41 ± 11	34
SF MEC	1.35 ± 0.02	1.33	229 ± 8	234	0 (limit)	0

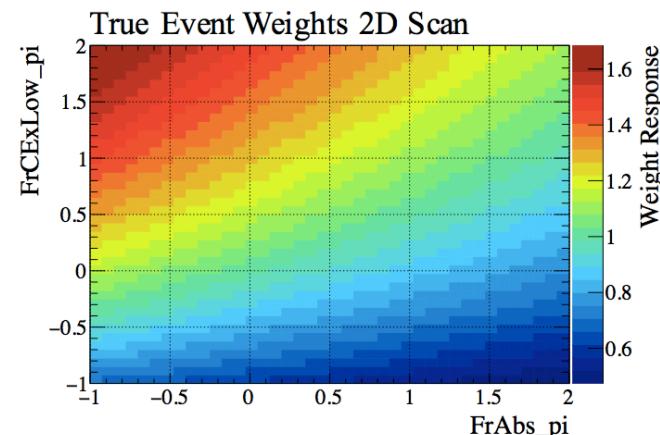
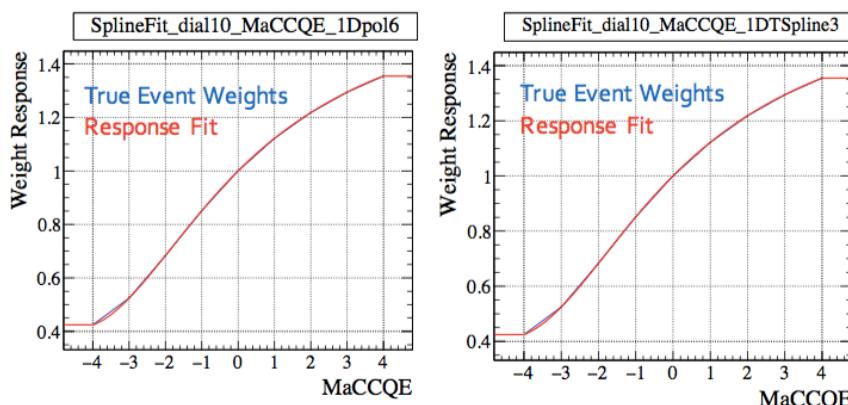
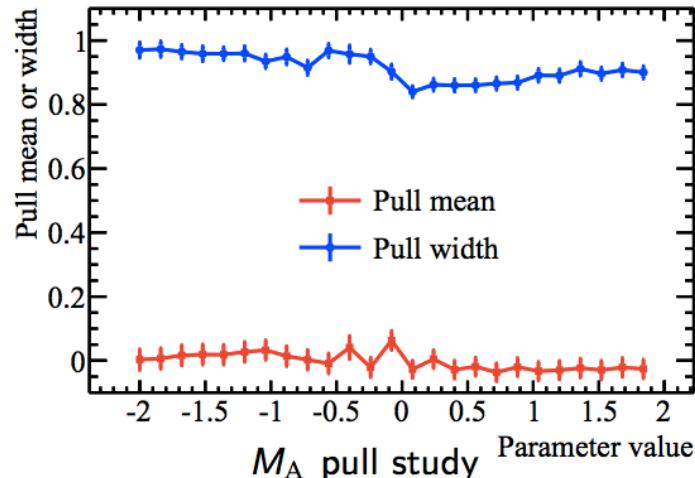
XSecFitter: Generic Fitting Framework

- Takes a long time to implement new models into generators, easier to take advantage of what is already setup in generators.
- Want to make comparisons/tunings of these different generators in a completely consistent and reliable way.
- Fitter Callum initially developed for NIWG NEUT fits provides a framework to allow new dataset comparisons and tunings to be added very quickly.
- Restructured the code to turn it into a generic generator fitting package.
- Contains a broad range of implemented data/MC comparisons.

Name is a work in progress!
Suggestions welcome!

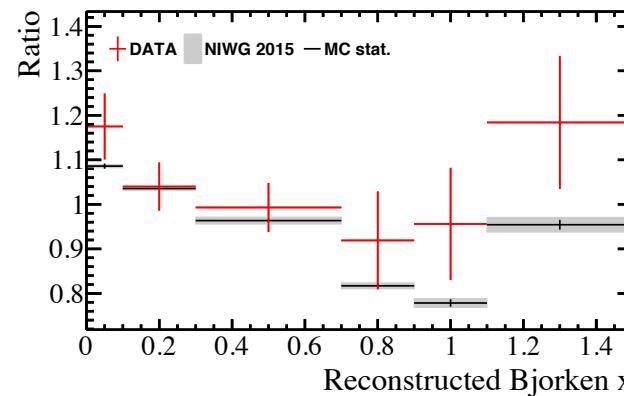
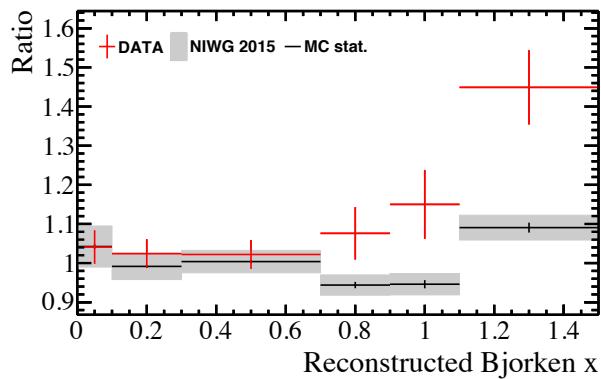
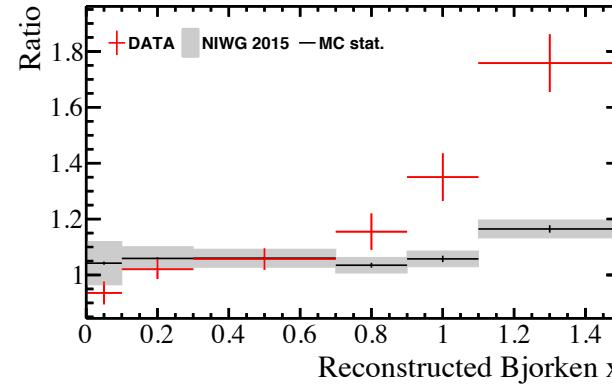
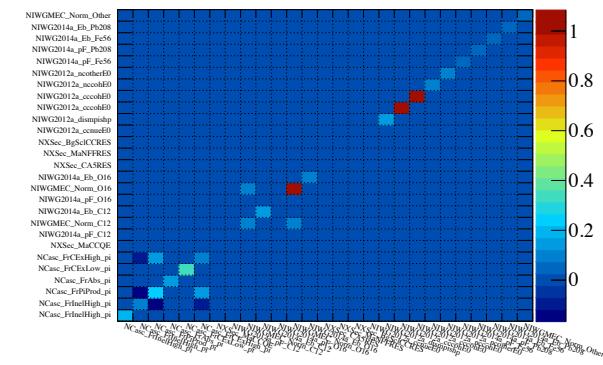
ReWeight Dial Tuning

- Multiple inputs:
 - NEUT ✓
 - NuWro ✓
 - GENIE ✓
 - NUANCE ✗ - Coming Soon...
 - GiBUU ✗ - Coming Soon...
- Fake data study tools.
- Interface with generator reweight utilities to do tuning.
- Moving to spline reweighting soon to allow for more rigorous model testing.



Systematic Analysis

- Lots of fitter development ongoing, added options to generate systematic throws from an arbitrary covariance.



- Should be shown as a standard for any model tuning so users can judge whether the fit parameters are actually appropriate for their analysis.

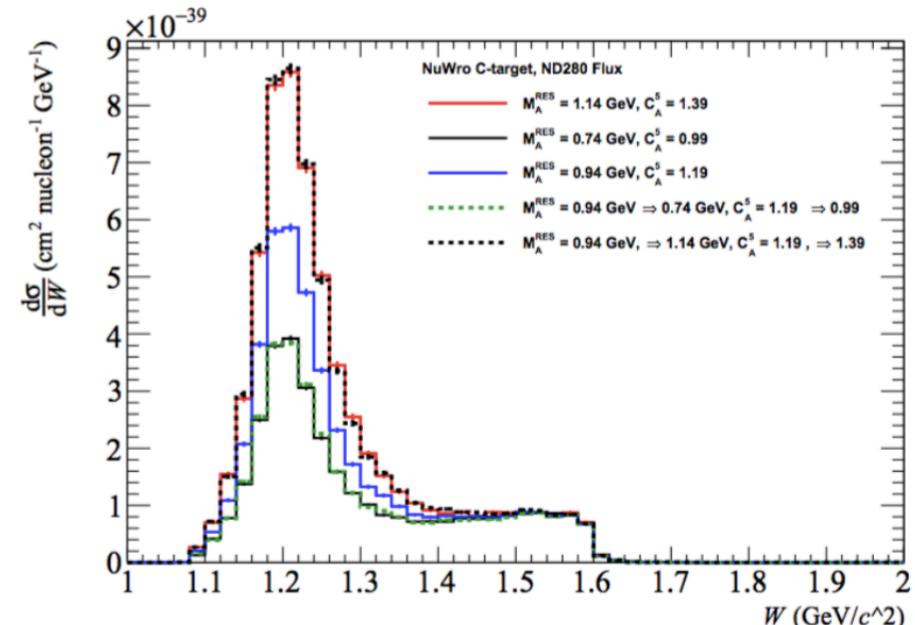
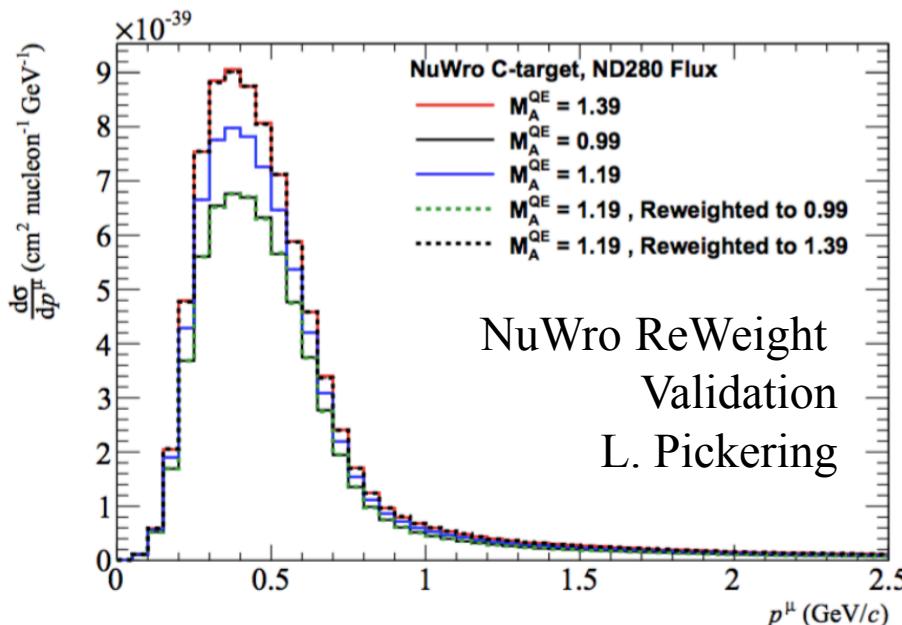
Experimental Collaboration

- Working to release this as **publicly available** software soon.
- Fitter tools will be freely available to use.
- Interested in more direct correspondence with experiments so we can properly test our cross-section models.



Using NuWro

- Included NuWro as a possible input. This opens up a range of extra models as well as providing a nicer C++ interface to prototype model changes.
 - LFG Model
 - Marteau MEC
 - NC TEM MEC
- NuWro reweight module developed to allow systematic studies to be performed.
 - Myself, Luke Pickering, and Jan Sobczyk, working to get this released soon!



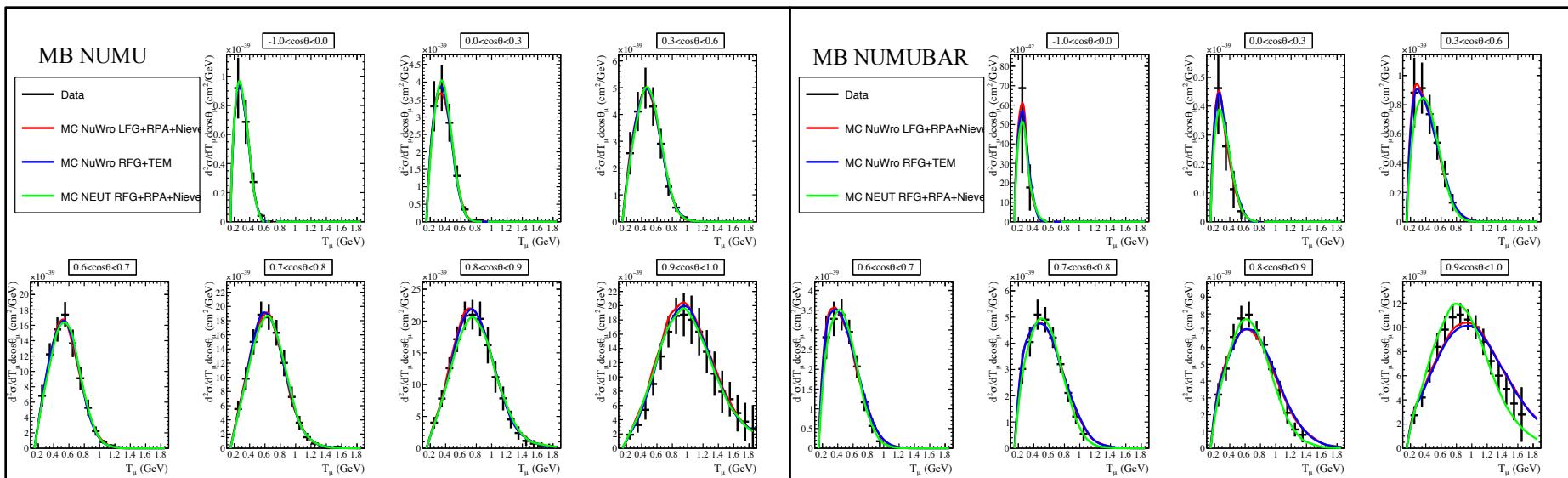
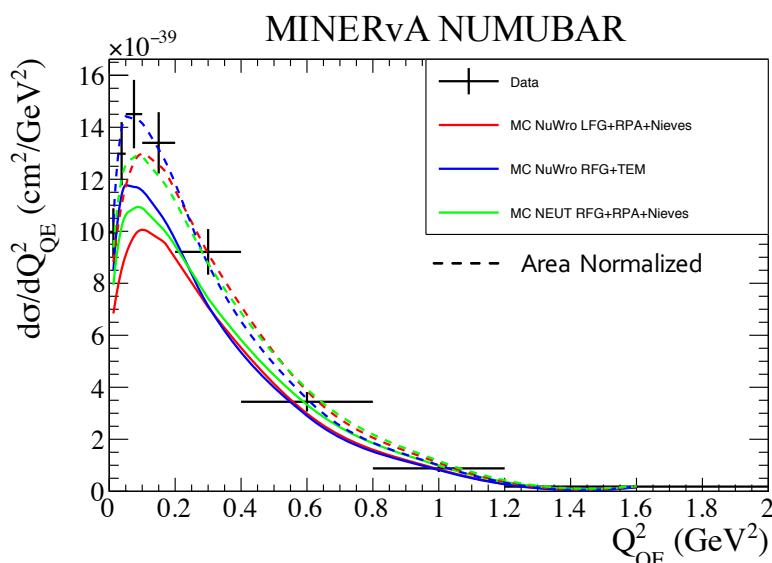
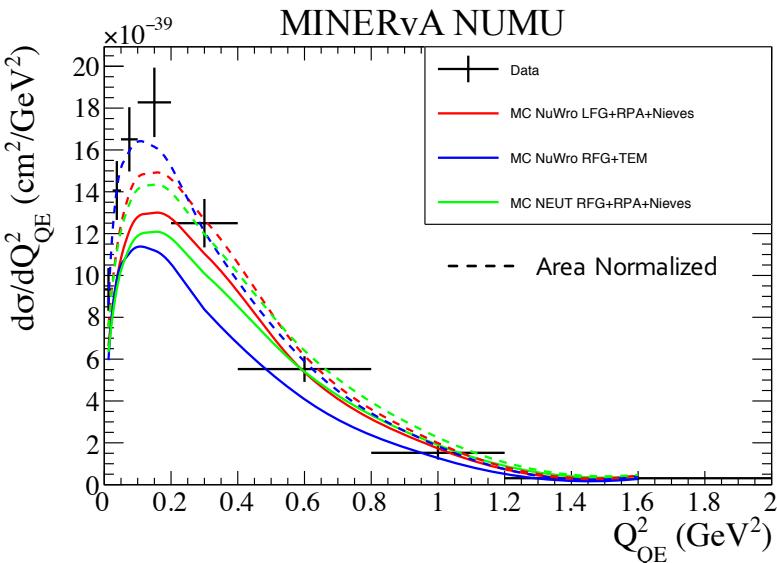
NuWro LFG Fits

- Want to see whether the fitter inflating M_A and reducing MEC normalization is just a result of the stitched together RFG+Nieves model in NEUT.

Parameter	M_A [GeV]	TN192	p_F [MeV]	TN192	MEC (%)	TN192
RFG MEC	1.34 ± 0.02	-	229 ± 7	-	0 (limit)	-
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- Compare 3 models:
 - NEUT RFG + RPA + Nieves
 - NuWro LFG + RPA + Nieves
 - NuWro RFG + TEM
 - In each model fit 2 free parameters:
 - Axial Mass
 - MEC Normalisation
- Found to be the best fit to MINERvA CCQE data in original data release.
- Removed pF dial as NuWro RW doesn't have this yet.

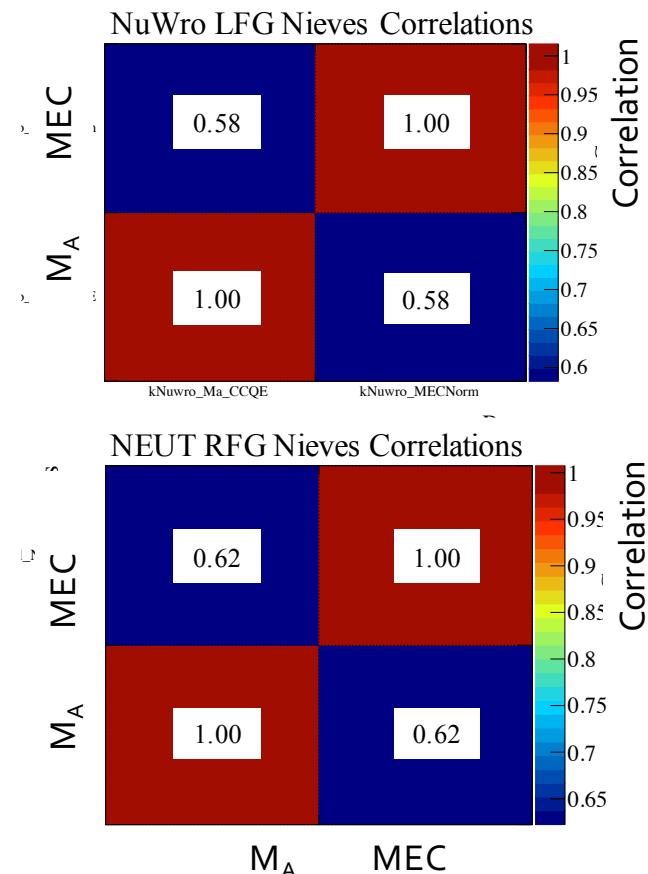
Fit Results



NuWro LFG Fits

Model	M_A (GeV/c ²)	MEC Norm. (%)	λ_M^ν	$\lambda_{MB}^{\bar{\nu}}$	$\chi^2/NDOF$
Nuwro LFG + Nieves	1.16 ± 0.03	8.3 ± 11.9	0.80	0.91	100.74/229
Nuwro RFG + TEM	1.15 ± 0.02	21.3 ± 12.5	0.84	0.92	93.62/229
NEUT RFG + Nieves	1.14 ± 0.03	25.5 ± 12.4	0.78	0.77	106.25/229

- Fit results from NuWro with an LFG are quite similar to NEUT RFG model.
- Still have problem where M_A is being inflated and MEC normalization is being driven down.
- MEC and the Axial Mass are positively correlated at the best fit point.

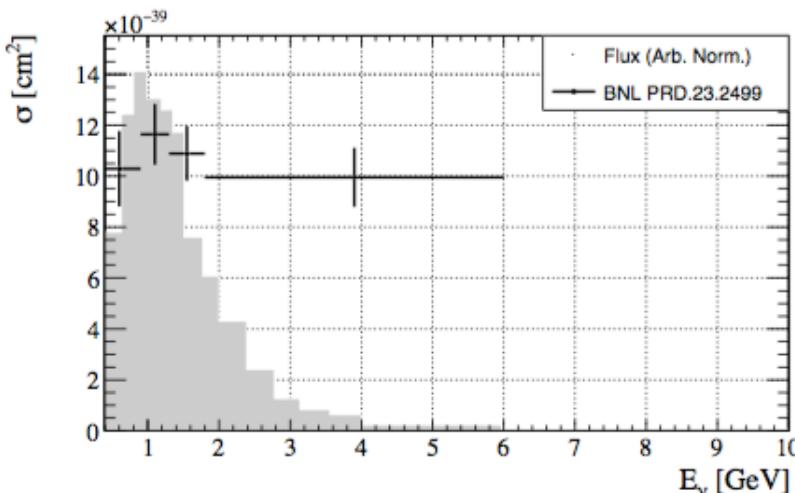


Is our High Q^2 Error Appropriate?

- Deuterium bubble chamber data can be used to place a constraint on the free nucleon cross-section. **Data is statistically limited at high Q^2 .**
- If the model has a limited shape, like the simple dipole, the uncertainty on the bare CCQE cross-section can be underestimated in this region.
- Evaluating the free nucleon cross-section uncertainties also very recently done in: [Phys. Rev. D 93, 113015 \(2016\)](#)
 - They use Z-expansion formalism for F_A .

Fits to Bubble Chamber Data

- Fit multiple bubble chamber datasets at once.
- Datasets included in a **shape-only fit**:
 - ANL 1DQ2 Event Rates
 - BNL 1DQ2 Event Rates
 - FNAL 1DQ2 Event Rates
 - BEBC 1DQ2 Cross-section



Generate NEUT/NuWro events with original published flux distributions.

$$\chi_{\text{ANL}}^2 = 2.0 \sum_i^{49} [S_{\text{ANL}} M_i - D_i + D_i \times \log(S_{\text{ANL}} M_i / D_i)]$$

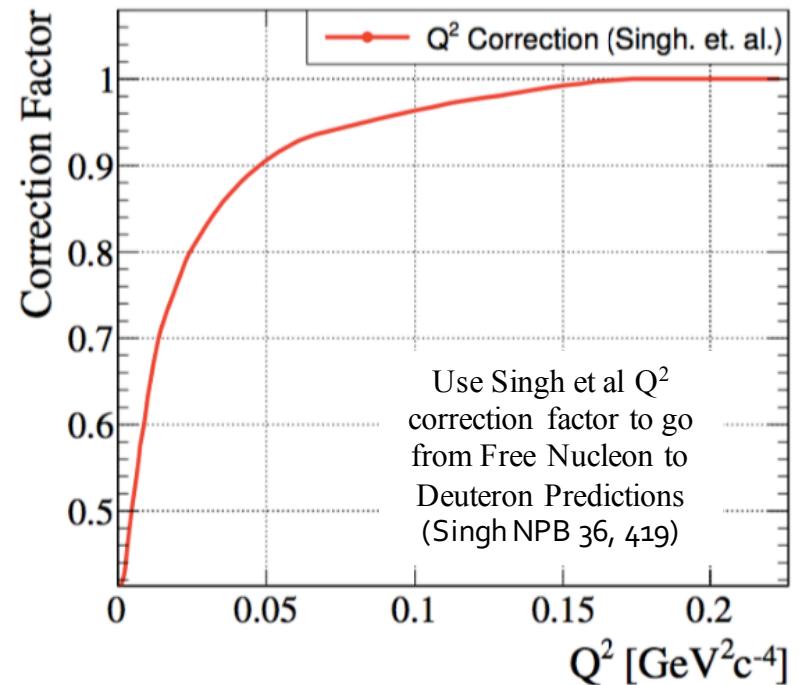
$$\chi_{\text{BNL}}^2 = 2.0 \sum_j^{49} [S_{\text{BNL}} M_j - D_j + D_j \times \log(S_{\text{BNL}} M_j / D_j)]$$

$$\chi_{\text{FNAL}}^2 = 2.0 \sum_k^{29} [S_{\text{FNAL}} M_k - D_k + D_k \times \log(S_{\text{FNAL}} M_k / D_k)]$$

$$\chi_{\text{BEBC}}^2 = \sum_l^8 \left(\frac{M_l - D_l}{E_l} \right)^2$$

$M = \text{MC}$
 $D = \text{Data}$
 $E = \text{Error}$

$$\chi_{\text{Total}}^2 = \chi_{\text{ANL}}^2 + \chi_{\text{BNL}}^2 + \chi_{\text{FNAL}}^2 + \chi_{\text{BEBC}}^2$$



Alternate F_A

- Consider alternate forms for F_A : Non-dipole form factors.
- Keep vector form factors at BBBA05 parametrizations.
- BBBA07 Model (*Eur.Phys.J.C*53:349-354,2008)

$$F'_A(Q^2) = A_{Ax}(\zeta) F_A(Q^2) = A_{Ax}(\zeta) \frac{F_A(0)}{\left(1 + \frac{Q^2}{M_A^2}\right)^2}$$

$$A_N(\xi) = \sum_{j=1}^n P_j(\xi)$$

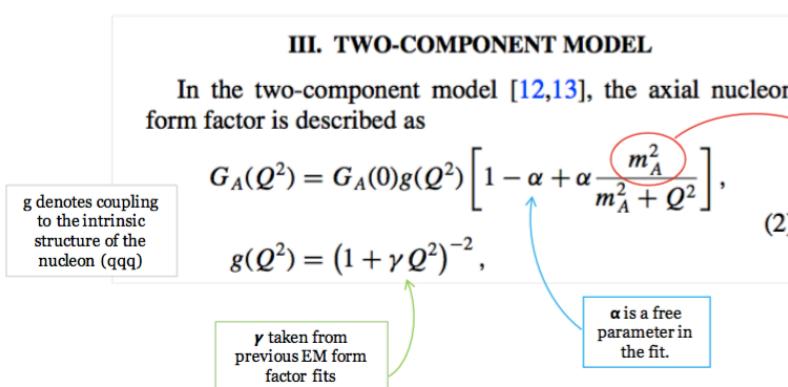
$$P_j(\xi) = p_j \prod_{k=1, k \neq j}^n \frac{\xi - \xi_k}{\xi_j - \xi_k}.$$

$$\zeta = \frac{2}{1 + \sqrt{1 + 1/\tau}}$$

p_j = Free Parameters

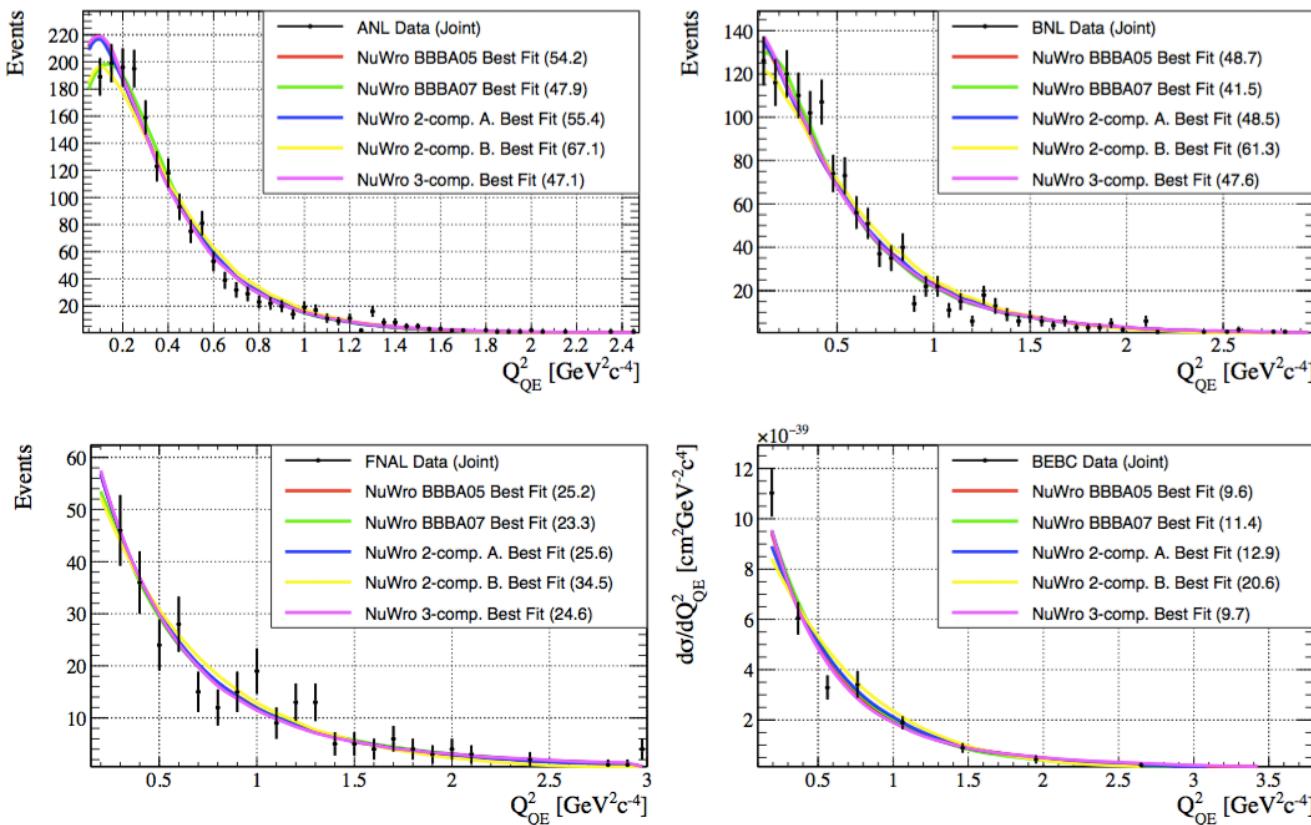
Original paper places several extra constraints on p_i . We do not!

- 2-component Model (*Phys.Rev.C*78, 035201 (2008))



- Contribution from a qqq core and a q-qbar cloud.
- 3 possible models to fit.
- A. Alpha free, Gamma = 0.515
- B. Alpha free, Gamma = 0.25
- C. Alpha free, Gamma free

Fit Results

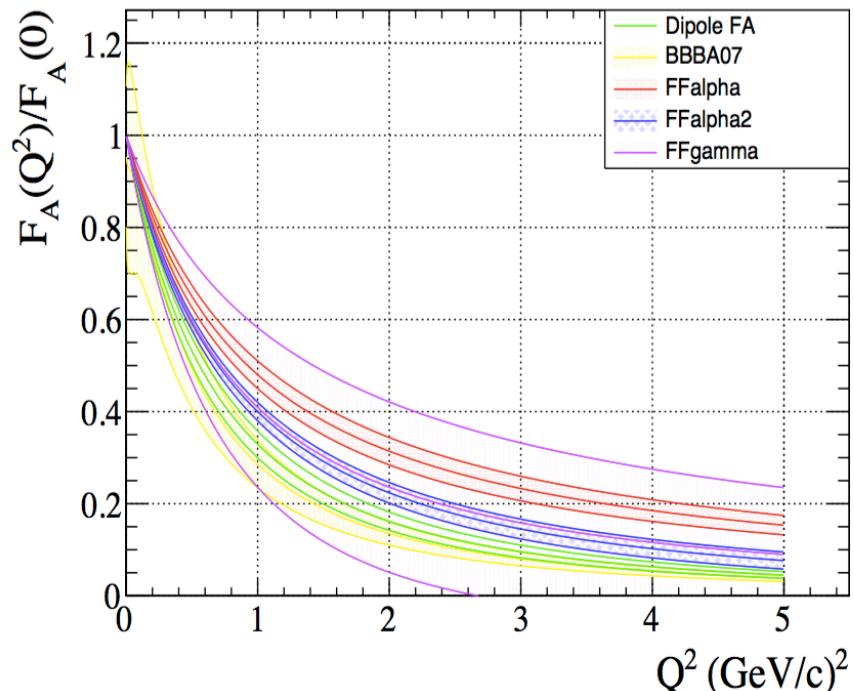


- For simple dipole (BBBA05) we get inflated M_A of 1.15 GeV due to performing a shape-only fit.
- Recent updates to this work use E_{nu} distributions to set the normalization results in lower values in agreement with previous fits ($M_A \sim 1.05$).

Alternative models

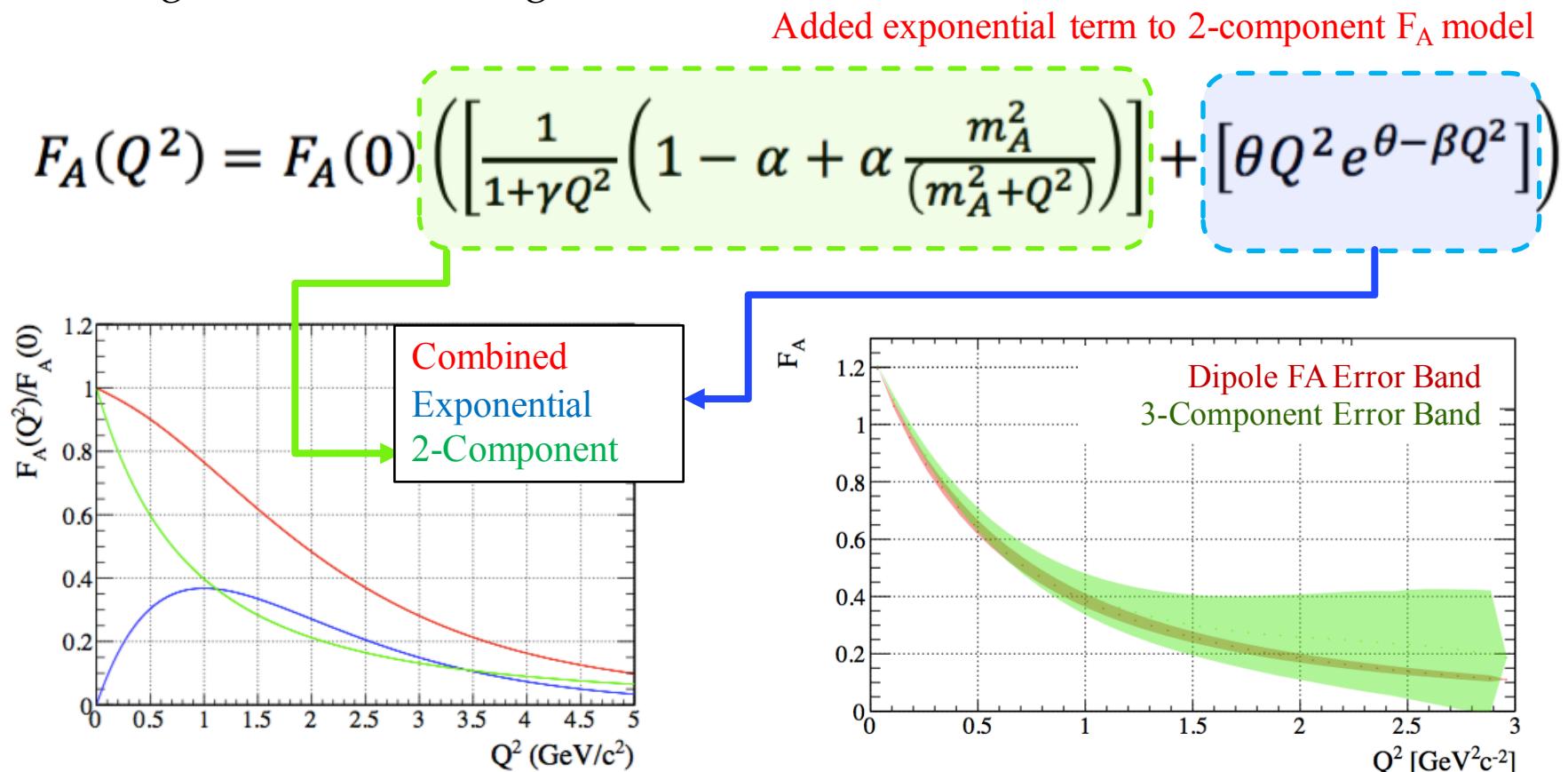
	Dipole (BBBA05)	BBBA07	2-comp A.	2-comp B.	2-comp C.
χ^2	137.74	124.09	142.55	183.49	141.97
NDOF	134	130	134	134	133
χ^2/NDOF	1.03	0.95	1.06	1.37	1.07

- Considered alternative forms for F_A .
- Used NuWro to quickly generate predictions and reweighting dials for each model.
- Each one individually tuned to bubble chamber data.
- Agreement with data is remarkably similar even though high Q^2 behavior is quite different.



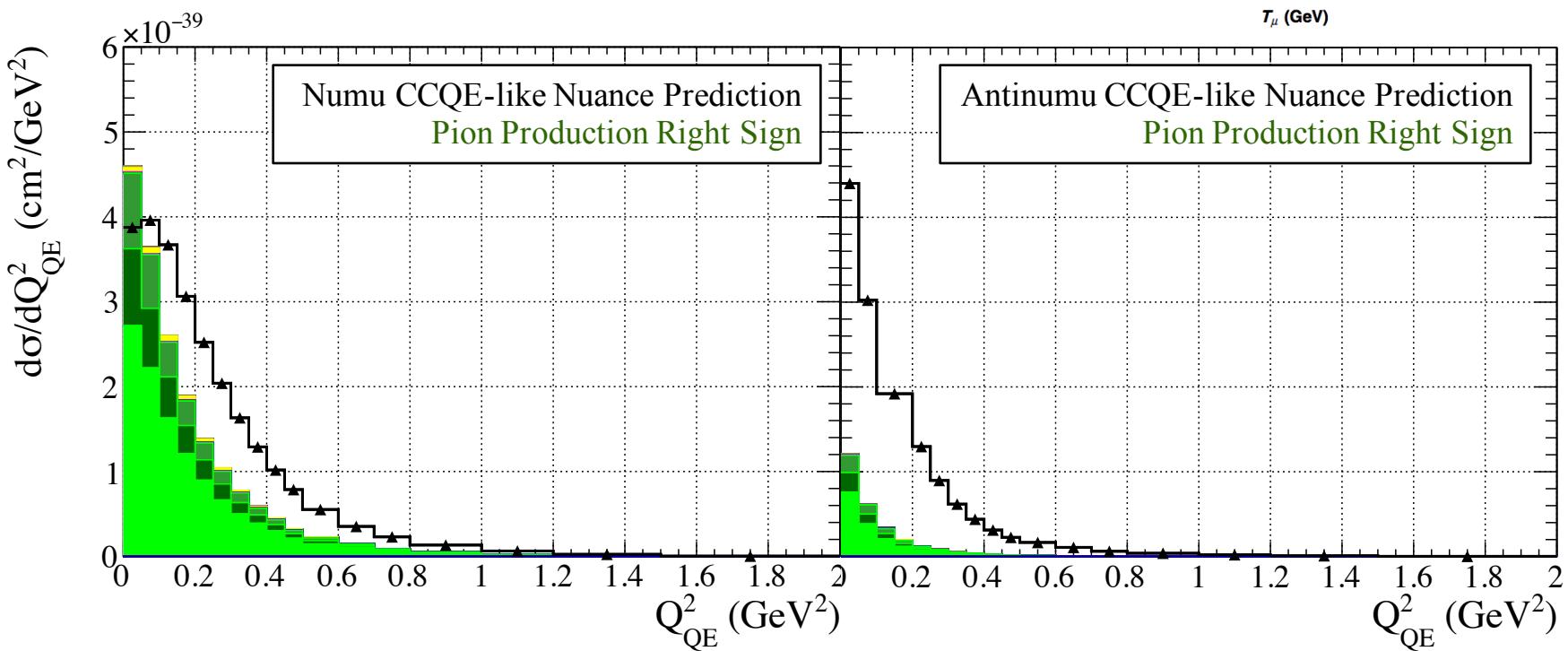
3 Component Model

- Create an effective model that is capable of describing each of the other models.
- Fit this to the data again to determine a new error band on the form factor.
- Best fit result matches the dipole form factor well ($\chi^2=131.19/131$) but provides much larger error band at high Q^2 .



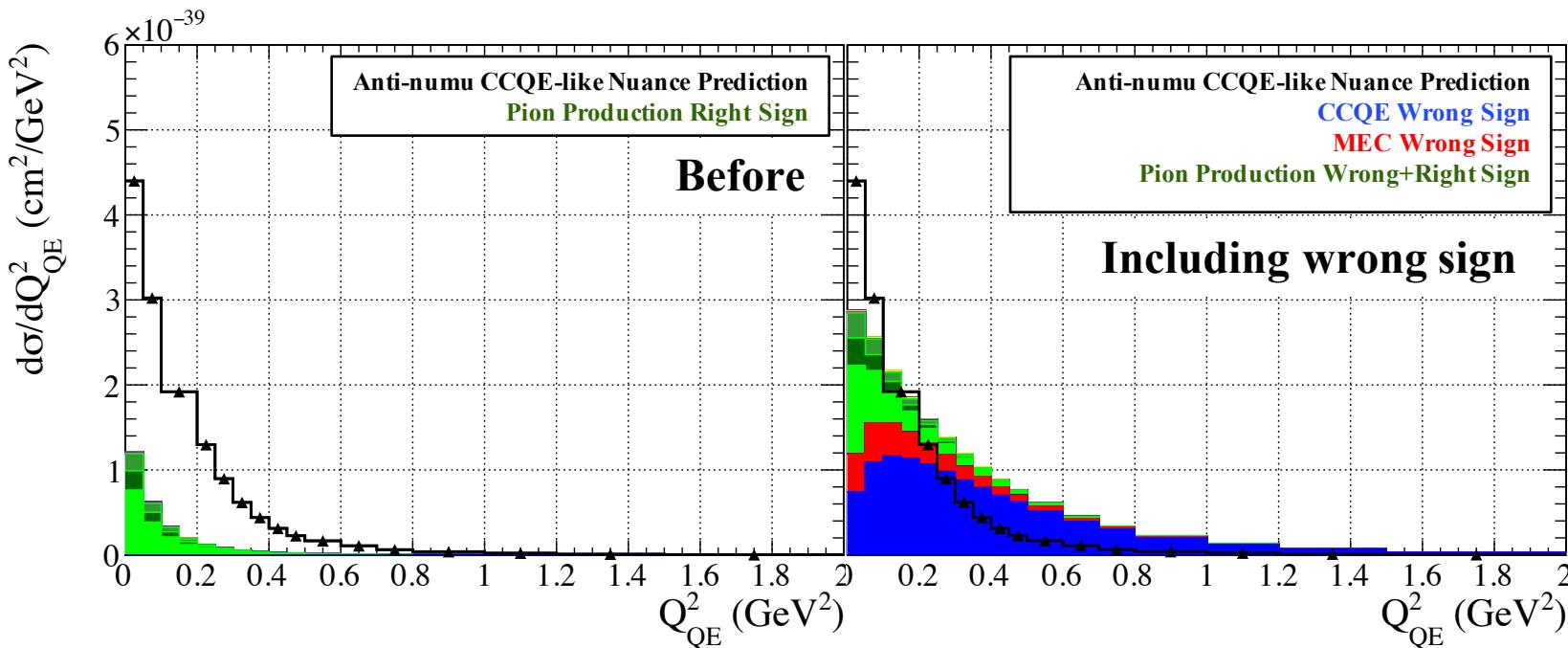
MiniBooNE CCQE-like

- Better to fit uncorrected final state topologies instead of background corrected samples.
- CC0PI > CCQE.
- Found a huge difference between the NUANCE prediction for antinu CCQE-like background.



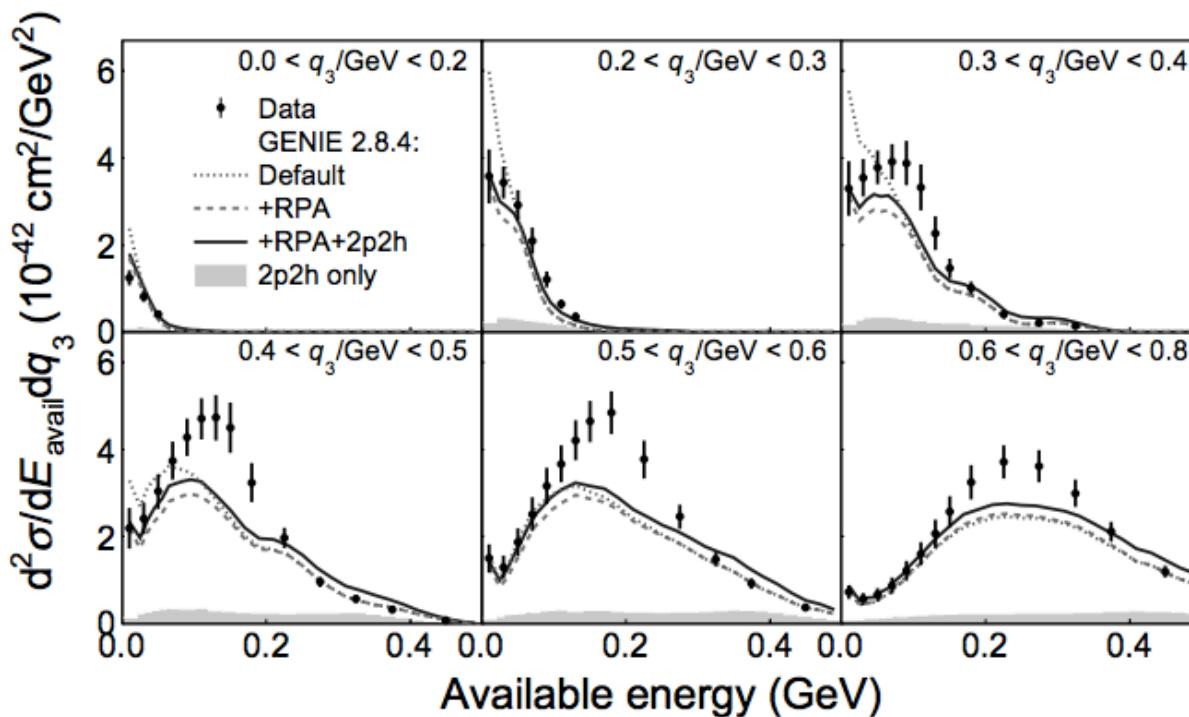
MiniBooNE CCQE-like Data

- Missing wrong sign was the reason for such a large discrepancy.
- Including this helps to at least match the normalization a bit better, but NEUT disagrees still.
- Need to try and use final state topology measurements.
- MiniBooNE provided CCQE-like background prediction, but separated modes would be a huge improvement.



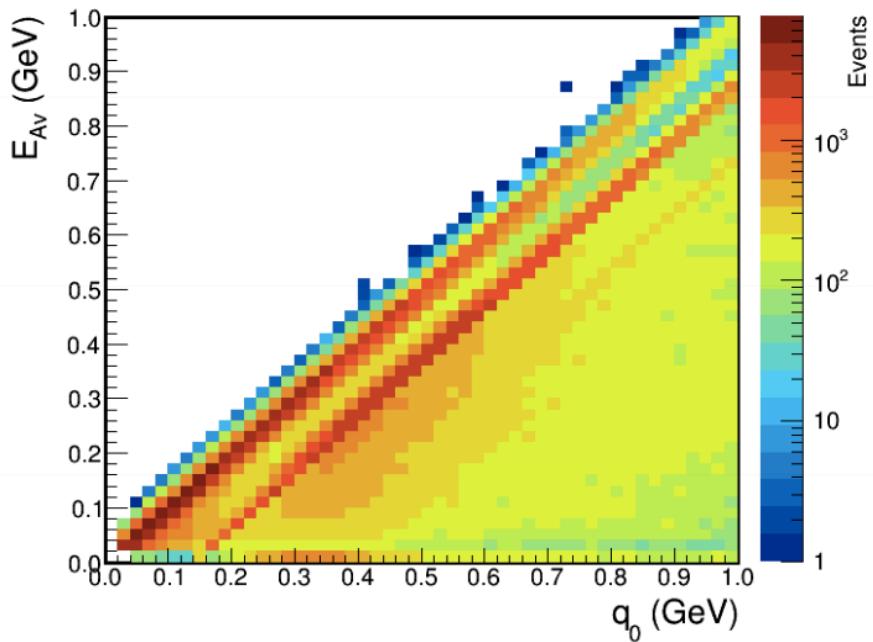
MINERvA Low Recoil Dataset

- MINERvA published a 2D CC-inclusive cross-section measurement at low three-momentum transfer (*Phys. Rev. Lett.* 116, 071802 (2016))
- Wanted to try and use this distribution to place a constraint on NEUT's modelling of nuclear effects (RPA and MEC models).



Energy Available

- Low Recoil Measurement uses the energy deposited in the detector to form a value “Energy Available” (E_{av}) which maps onto q_0 .



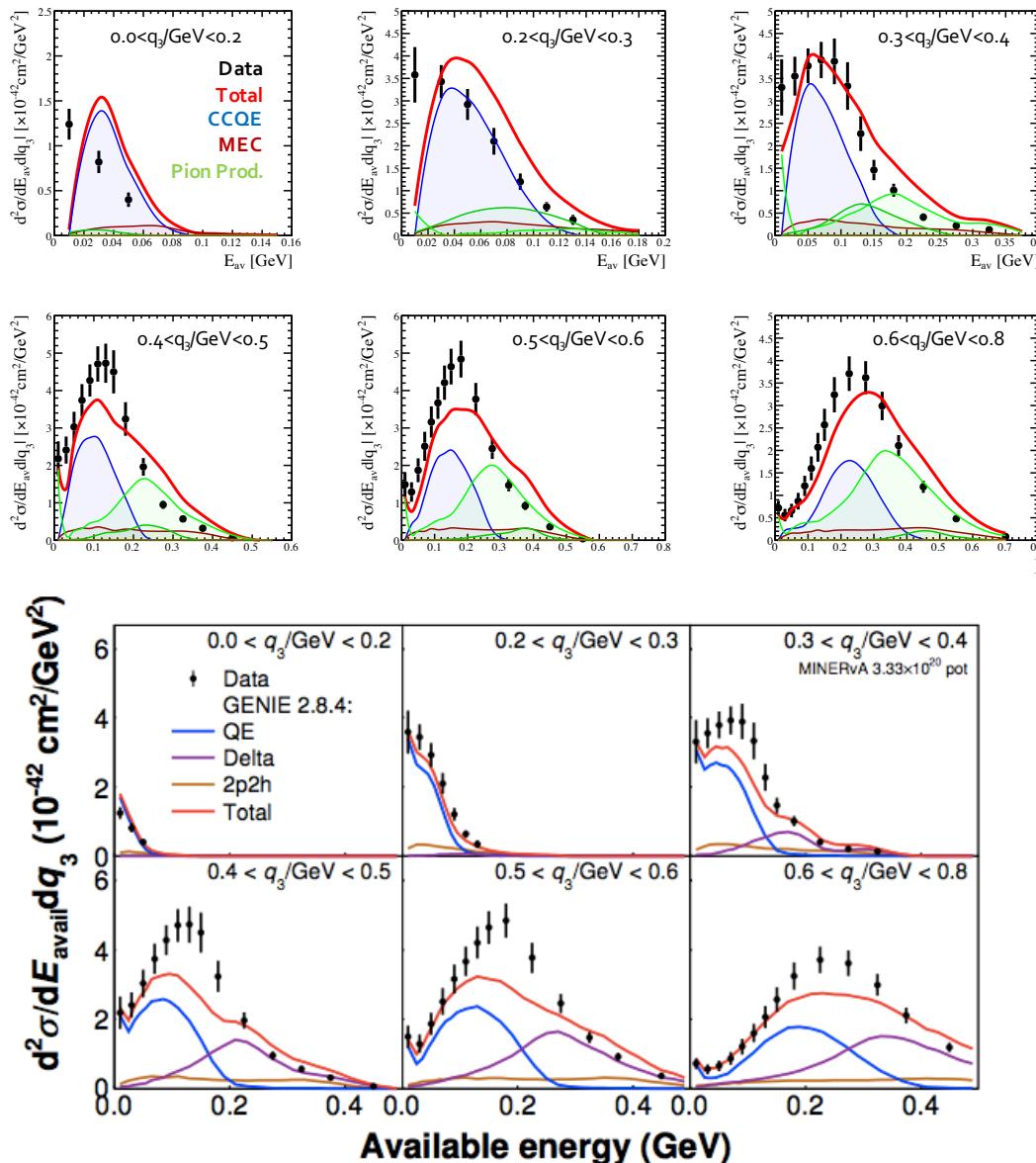
$$E_\nu = E_\mu + q_0$$

$$Q^2 = 2E_\nu(E_\mu - p_\mu \cos \theta_\mu) - M_\mu^2$$

$$q_3 = \sqrt{q_0^2 + Q^2}$$

$$E_{av} = \sum_{i=p, \pi^\pm} T_i^K + \sum_{i=\pi^0, e, \gamma} E_i$$

Comparison to MINERvA Data

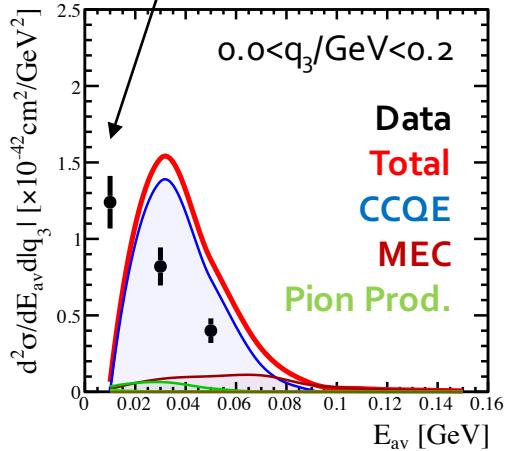


- Rein-Sehgal Coherent Model in NEUT has quite larger prediction than Berger-Sehgal.
- NEUT distributions seem shifted to higher E_{av} bins.

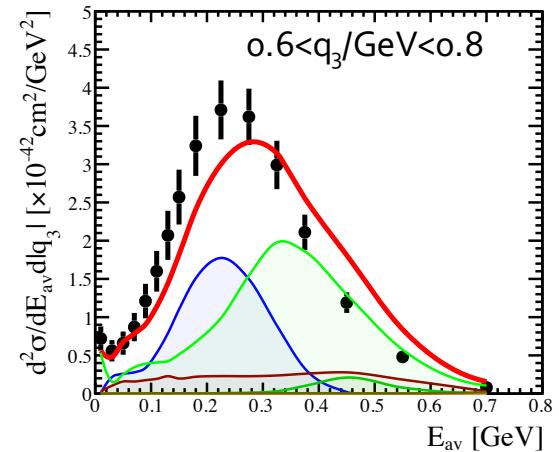
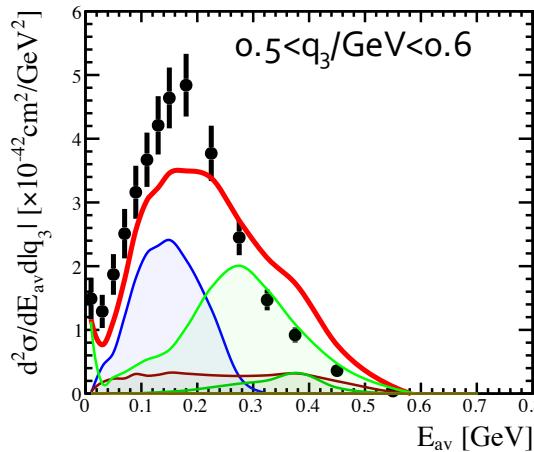
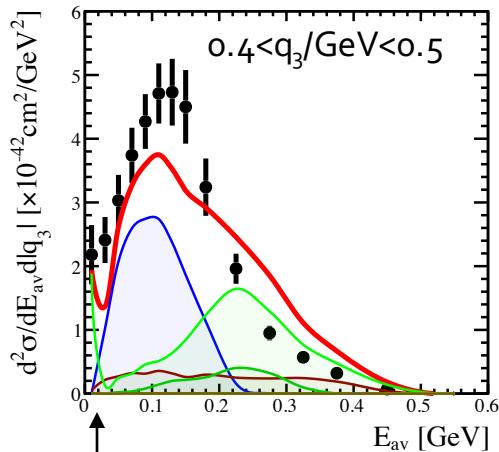
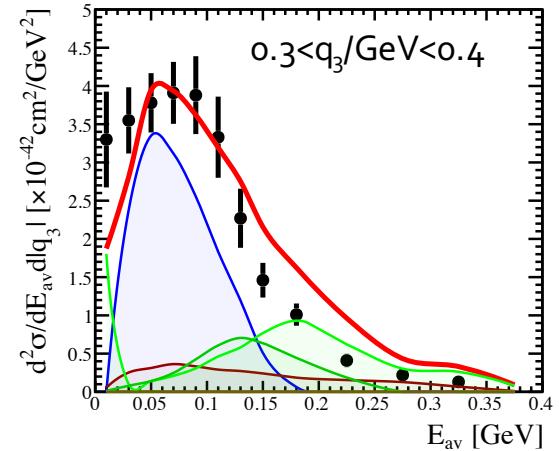
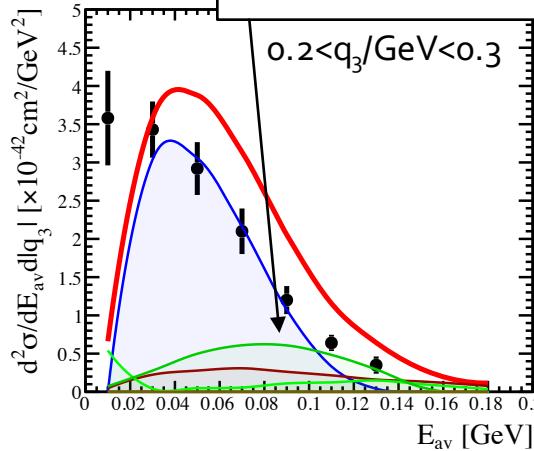
Original MINERvA result with
GENIE comparisons

Low Recoil Features

Strong Pauli
Blocking causes
huge deficit.



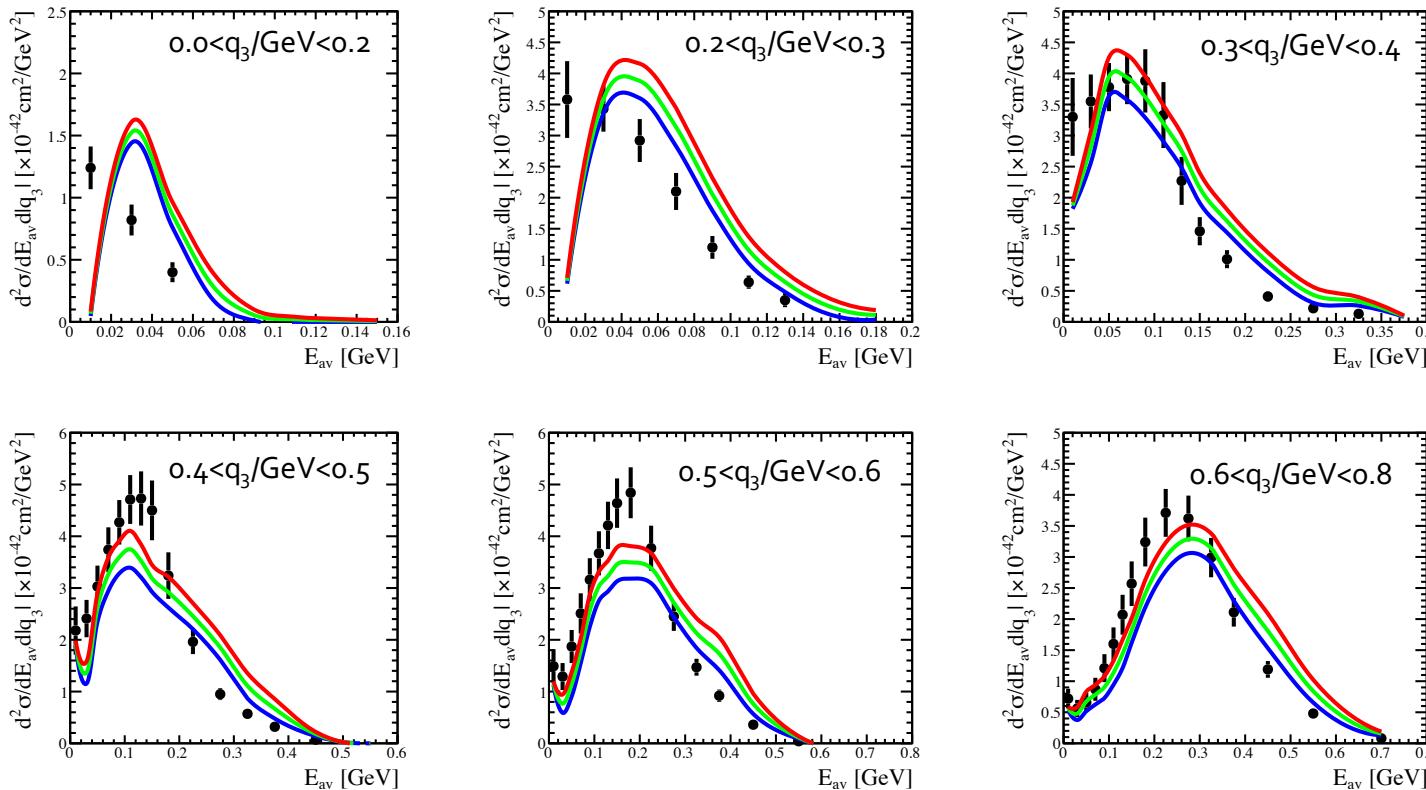
Large Rein-Seghal
Coherent Pion



Lowest Eav dominated
by events with only final
state neutrons.

Variations MEC

- Can we fill in the dip region by varying MEC normalization alone?



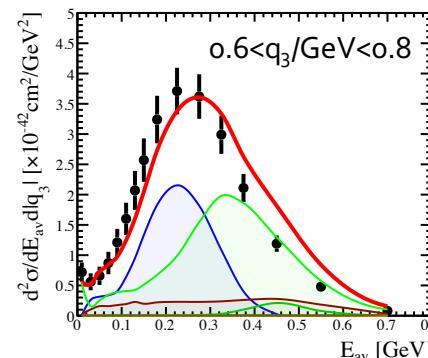
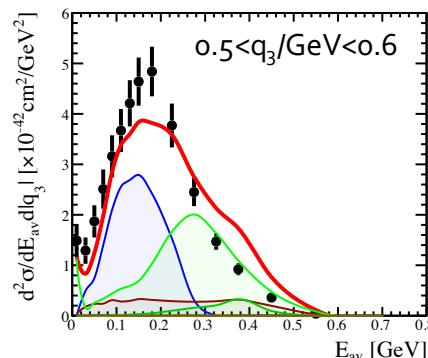
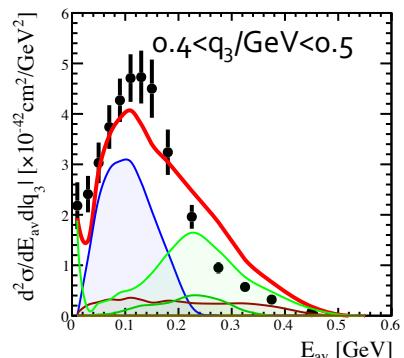
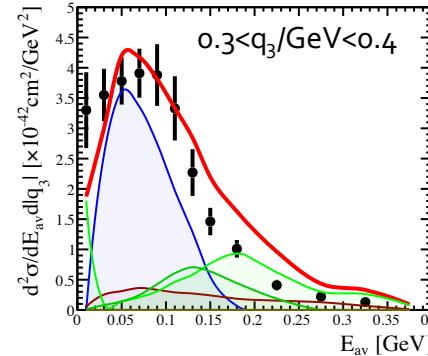
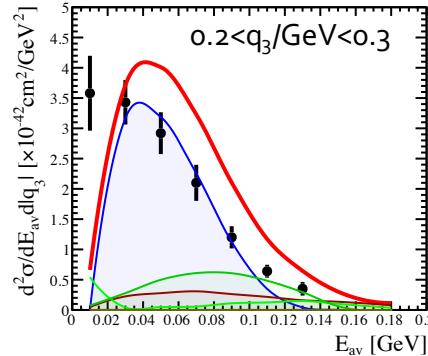
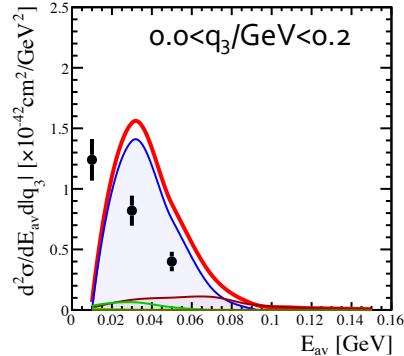
- Increasing the MEC to very large values ($\sim 200\%$) would help push the cross-section up and slightly fill in the difference in the dip region, but would significantly modify the high E_{av} tail.

Inflate M_A ?

- What do we like to do when faced with such a large data/MC disagreement?

Inflate M_A ?

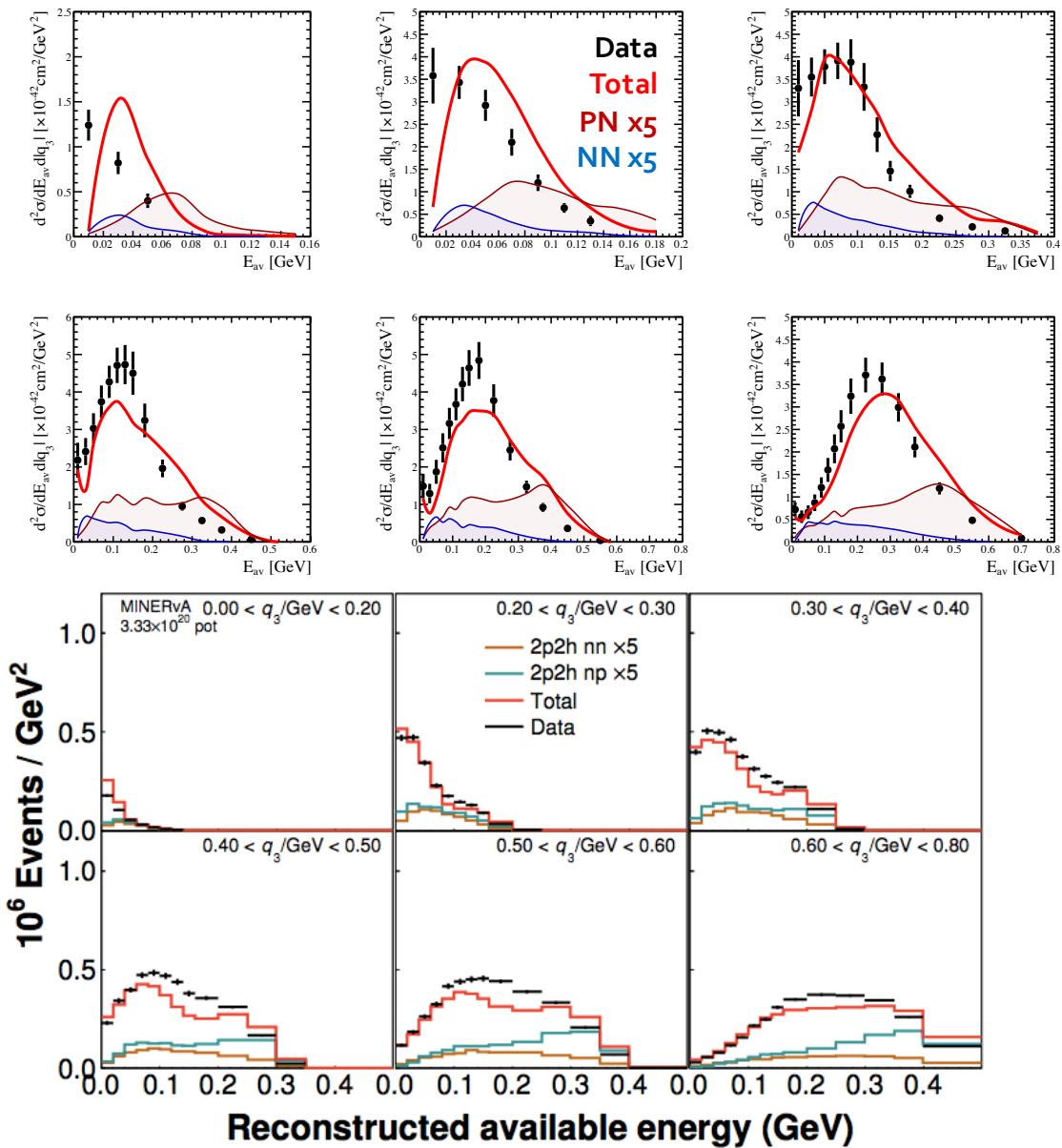
- What do we like to do when faced with such a large data/MC disagreement?



- Nominal NEUT with an axial mass of 1.21 GeV and Nieves RPA+MEC does do slightly better than with $M_A = 1.00 \text{ GeV}$.
- Need to be careful with the fact that this is CC-inclusive.....

PN NN Contributions NEUT

- Low Recoil Dataset doesn't include neutrons in the definition of E_{av} .
- Separation in E_{av} of the PN and NN MEC contributions.
- MEC contribution for both PN and NN initial states has been scaled up by a factor of 5 here to make it visible.
- Some differences between NEUT and the GENIE model from original MINERvA study.



Low Recoil Future Steps

- Want to try and use this low recoil dataset to place a better constraint on nuclear effects.
- CC-inclusive measurement means this isn't simple and will only be powerful when combined with other exclusive measurements.

What we have learnt/Next Steps

- Difficult to find a model in NEUT/NuWro that is capable of describing MINERvA and MiniBooNE data well.
- More work is needed to implement new models and study how they perform in JOINT fits to multiple experiments.
 - Need to encourage people to stop showing how well it does by comparing just to MiniBooNE CCQE numu data!
- Lots of different paths we are taking that should eventually converge on an improved understanding of how to model the complete neutrino cross-section:
 1. **Get full correlations for data** -> *Work in Progress for MB*
 2. **Don't fit CCQE-corrected datasets** -> *Possible for MB but some technical subtleties.*
 3. **Fit low recoil data** -> *CC-inclusive fits are difficult to keep well behaved. Working on building up a base model of priors to help, alongside improvements to the fitter framework.*