



Near Detector Analysis tools Report

GUNDAM Working Group

Oscillation Analysis tools Friday, 13th June 2025 IRN

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During this presentation, we will talk about:

1. Systematics in T2K

- 1.1 T2K experiment
- 1.2 Flux
- 1.3 Cross-section neutrino interaction modes
- 1.4 Detector

2. GUNDAM fitter

2.1 ND fit goals2.2 GUNDAM framework

3. More recent analysis status and novelties

3.1 New 4π selection 3.2 New RPA/FSI cross-section parameters implementation

T2K Experiment



J-PARC (KEK/JAEA)

Neutrin

Concept of a neutrino oscillation baseline experiment

 $\sigma(E_{v},\vec{x}) \otimes \Phi(E_{v}) \otimes D^{far}(\vec{x}) \otimes P_{osc}(E_{v})$

 $\sigma(E_{\nu},\vec{x}) \otimes \Phi(E_{\nu}) \otimes D^{near}$

Cross-section (neutrino→target probability)



Detector (see/reconstruct interactions)



 $V_{events}^{far}(\vec{x})$

 $V_{events}^{near}(\vec{x})$

ND **tuned models** propagated to far detector(best possible prediction)

 (\bar{x})



- different detector effects from ND to SK
- systematics uncertainties

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 $P_{osc}(E_v)$ is more precise when E_v is well-reconstructed

Focus on Near detector !

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Beam



- protons interact with a graphite target to produce hadrons (mainly pions and kaons)
- Detectors 2.5° off the direction of the beam centered around 0.6 GeV.
- Off-axis method reduce mean energy to maximize oscillation detection probability for 295 km baseline
- Flux model tuned against external data (NA61) remain ~5% uncertainties





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Flux systematics parameterisation

- Uncertainty on the flux reflects errors on hadrons productions or beam alignment variations for instance
- Flux is binned in energy bins and neutrino types (numu, antinumu, nue, antinue)
- In T2K, flux systematics are described as normalization parameters affecting the bin content (in true neutrino energy) according the prior covariance matrix provided by beam experts



Flux covariance matrix

impact ND and SK predictions

neutrino interaction mode



- At T2K energies \rightarrow multiple interaction contributing
- Each interaction has uncertainties + affected by nuclear effects
- series of parameters to describe and constrain these uncertainties for each interaction channel

CCQE interactions are the dominant ones at T2K energies.

Ideally we want to select pure CCQE events, but nuclear effects play an important role and are source of systematics effects



Cross-section systematics parameterisation

- **1D event by event response function** ≈ **spline** : polynomial and continuous function used for flexible interpolation of knots for which we attribute a weight
- Cross-section uncertainties reflect the lack of knowledge on interaction models and data experimental limits to constrain those models
 - \circ Uncertainties are modeled by systematics as MAQE (CCQE axial mass) or FSI (Final State Interactions) → affect each MC event



MAQE affects Q² FSI models what happen to the particle after the neutrino interaction

101 cross-section parameters in total in current analysis

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Event selection in the detector



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- In the detector, neutrinos interact with bound nucleons inside nuclei
- Full tracking and particle reconstruction (FGDs, TPCs)
- Most exploited observables are muon angle/momentum
- Still using pre-upgrade era data for the analysis

Detector systematic uncertainties

Each detector uncertainty is represented by a parameter in the fit
 → model how detector effects impact MC predictions

Carbon/Oxygen

• Typical detector systematics : Track efficiency, matching efficiency, secondary interaction, target systematics



Carbon

Systematic examples

Secondary Interaction = uncertainty on the probability of a proton/pion produced at the neutrino interaction vertex to undergo a secondary interaction

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Detector systematics parameterisation

Two kind of detector systematics now:

- Weight-like physical parameters described with response functions (parameterised via splines 81 parameters)
- Variation-like allowing event migration (encoded via a covariance matrix 571 bins) describing the bin content variation



ND fit goals and frameworks used

- ND fit used to constrain the flux and Cross-section models and provide a tuned model for oscillation analysis
 - Ingredients of ND fit : samples, uncertainty models (detector, Cross-section and flux)



FGD2 numuCC 0π 0y N Protons Bwd

GUNDAM



Barlow-Beeston LLH (likelihood)

- What's a log likelihood ?
 - Quantifies the compatibility between observed data and predictive model
 - GUNDAM uses MIGRAD to minimize -log(L) to estimate model parameters
 - Barlow-Beeston formula (handles MC statistical error) :





- LLH scan = vary a parameter while keeping the others fixed
 - $\rightarrow \text{observe}$ how the likelihood responds
 - \rightarrow identify the minimum
 - \rightarrow estimate sensitivity of the parameter

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2D binning in P_u

and $\cos\theta$.

New selection at ND280

- Improved acceptance (new 4π selection)
 - from 22 to 32 samples (no lepton going forward only)
- Expanded phase space in a previously poorly explored region (gain ~ 13% statistics thanks to High-angle and backward going tracks)



HA

ND280

FWD

New RPA/FSI strength cross-section parameters

- Nuclear model disagree in the low Q2 region (< 0.4 GeV)
- Need to add more freedom to our current model



24 new parameters introduced !

- RPA (*Random Phase Approximation*) to investigate neutrino-nucleus collective interactions properties
- FSI (*Final State Interaction*) or Optical potential to dig into interaction between residual nucleus and incoming particles

Dealing with new RPA/FSI cross-section parameters

- Mirroring is a technique where the effect of a parameter variation is symmetrically extended to the negative side
 → symmetric response
- Allows a smoother interpolation of model prediction across the parameter range
- Different interpolation types exist : linear or cubic (no cusp for cubics)
- Helps GUNDAM to converge when statistical fluctuations bring the best-fit value in an unphysical parameter region





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Latest Asimov fit results

Asimov fit : predict the expected performance of the analysis (sensitivity and parameter constraints)

 \rightarrow a "perfect" fit used to study the effect of the model (no statistical fluctuations)



Some CCQE and splined RPA parameters to illustrate systematic constraints (asimov prefit/postfit comparison) Lena Osu, LLR | IRN June 13th

Datafit = performed on the actual experimental data and used to constrain systematic uncertainties to provide final results



Some CCQE and splined RPA parameters to illustrate systematic constraints (data prefit/postfit comparison) Lena Osu, LLR | IRN June 13th

Conclusion

- ND280 role : caracterize neutrino flux before oscillation
 - constrain neutrino-nucleus interaction models
- Systematics :
 - Flux
 - Cross-section
 - Detector
- Novelties for the new analysis :
 - GUNDAM
 - Improved selection
 - Detector systematic treatment
- P-value studies

 \rightarrow shows the distribution of postfit IIh under the assumption the MC model is correct



Example from previous OA2021 analysis

Back-up

P-value sudies

- Measure the level of agreement between the data and our MC model
- p-value should be > 50% if the model provided is robust and a good description of the data
- One can generate a set of toys starting from the MC : vary all systematics and bin contents to account for statistical fluctuations
- p-value = integral beyond the red line (actual data)/total integral of the black curve



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The RPA and FSI parameters are designed to interpolate between two models:

- \rightarrow 0 corresponds to the nominal model,
- \rightarrow 1 corresponds to a model with enhanced RPA or FSI effects.

However, when using linear interpolation, a **problem** arises:

 \rightarrow The model response has a cusp (non-smooth behavior) at parameter value 0, especially for some observables.

-->The complication around 0 comes from the fact that the lower bound of the parameter is also fixed at 0, so the fit cannot explore negative values to smooth out the cusp, making the behavior near 0 more difficult to handle.

 \rightarrow This can create issues during the fit, particularly in data fits, where the optimizer may struggle or bias the parameter estimation around this point.

 \rightarrow in order to solve this issue : we use mirroring + smoother interpolation type (cubics instead of linear)

RPA

 \rightarrow Adjusts predictions in nuclear response functions \rightarrow low Q2 = bigger wavelength and this gives less localised interaction \rightarrow introduce collective behaviors in the nucleus \rightarrow long distance effect more important (more that just the first nucleon neighbor)

FSI or optical potential = complex potential used to model how particles, such as neutrons or protons, interact with nuclei (scattering or absorption, nuclear density interactions) It represents the average interaction between a projectile particle (like a nucleon or pion) and a target nucleus **Fermi motion**: nucleons are not at rest inside the nucleus With typical Fermi momenta up to \sim 200 MeV/c This motion introduces uncertainty to the initial state \rightarrow biases the reconstructed neutrino energy.

2p2h : These events mimic CCQE topologies (muon + protons, no pion), but the underlying interaction is different \rightarrow also biases the reconstructed neutrino energy \rightarrow Leads to extra nucleons in the final state.

FSI (Final State Interactions): outgoing particles re-interact inside the nucleus. They can:

- \rightarrow Be absorbed (e.g. pion absorption \rightarrow makes a RES event look like CCQE)
- \rightarrow Produce new particles (secondary interactions)

FSI blur the connection between the true interaction and what we finally observe.

--> Detectors only see the final state particles that escape the nucleus, not the original interaction.

These nuclear effects introduce model-dependent uncertainties in neutrino energy reconstruction and cross-section measurements

GUNDAM fitter

GUNDAM : Generalized and Unified Neutrino Data Analysis Methods

- Fitter framework for current and next statistical analysis of T2K
- Designed to host multiple analysis using JSON/YAML configuration files
- C++ code based on ROOT

A new way to perform statistical analysis

• Better traceability and validation of the output \rightarrow share inputs easily with other people



Flexible data handling structure

- · Generic sample definition
- · Parameter definition with multiple options
 - Covariance matrix
- Apply conditions
- Per dataset response function

Config & usage by a single gear: Propagator

- Initialise samples & parameters
- · Fast propagation engine (optimised & parallelised)
- Embeds diagnostic tools (figure generators)

Adrien Blanchet pdf