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Chapter 1

NCQE-Gd

Both MC and data neutrino events undergo a reconstruction phase, where the low-energy fitter BONSAI is applied to the event. This reconstruction is carried out using timing and cable information, however charge information is omitted. The ouput of BONSAI gives information which will be used in the reduction phase of the data and allow for the selection of the NCQE sample. BONSAI is a low-energy fitter used in the few-MeV to tens of MeV energy range, therefore suitable for this analysis. It's directional reconstruction uses the fitted vertex and maximises for direction while it's energy reconstruction of the event is based upon the number of hit PMTs. Equation ?? shows how the vertex reconstruction is formulated, where w_i is the Gaussian hit weight of the PMT, t_i is the hit time of the PMT and x_i is the position of the hit PMT.

$$g(\vec{v}) = \sum_{i=1}^{N} w_i e^{-0.5(t_i - |\vec{x}_i - \vec{v}|/c)/\sigma)^2}$$
(1.1)

The following quantities comprise the BONSAI output, the first two being help-ful spectator variables and the latter five constituting parameters which are used in the reduction phase of the analysis, from which the neutrino NCQE sample is determined.

Neutrino vertex direction

This vector points towards the direction which is an average over all the Cherenkov cone axes which are produced, due to there being multiple leptons induced in the interaction.

Neutrino vertex position

In NCQE events the vector of the Cherenkov cone produced by an outgoing charged lepton, gives the point at which the Oxygen molecule involved in an NCQE event excites and de-excites due to an interaction with a neutrino and it is this interaction point which is the neutrino vertex position.

Reconstructed energy

In line with the standard SK low energy analysis definition, this energy is simply the reconstructed energy with the 0.511MeV electron mass omitted. The range for Erec in this variable is 3.49 MeV to 29.49 MeV.

Dwall

This variable gives the minimum distance of the neutrino vertex position from the closest wall of the Super-Kamiokande detector.

Effwall

Thus variable gives the distance between the neutrino vertex position and the closest wall, but moving back from the vertex position along the neutrino vertex direction vector.

Vertex direction and goodness coefficient

The coefficient ovaQ (defined in Equation ??) describes the quality of the vertex reconstruction. It consists of two parameters g_{vtx}^2 and g_{dir}^2 where the former describes the goodness of the vertex which is based on PMT hit timings, and increases the sharper an event is in time. The latter is the directional goodness and measures the azimuthal uniformity in the ring pattern produced by the Cherenkov cone, which decreases the more uniform an event is in space. As a result of this, ovaQ increases the more uniform and sharp in time an event is.

Cherenkov angle $theta_C$

In neutral current events, this value determines the opening angle of the Cherenkov cone and unlike for charged current events does not

1.0.1 True neutron tagging information

1.0.2 Primary selection criteria

1.0.3 Secondary selection criteria

When the neutron vertex is found by this method, 14 variables which describe different aspects of the neutron candidate are calculated. For each of the neutron candidates the vector of these variables are computed and fed into the neural network and this produces an output value which is between 0 and 1. These variables relate to different features regarding categorising hits from neutron capture on Gd or H, including the number of the hits from neutron capture, the isotropy of these hits, the Cherenkov angles of these hits and the position of the neutron vertex in the detector when capture occurs. A description of these variables are given as follows:

N10nvx

This is the number of hits in the 10ns sliding window of the neutron candidate

N300S

Excluding the number of hits in the 10ns sliding window (N10nvx), this is the number of hits in the extended window of 300ns.

NcS

This variable is defined as:

$$NcS = N10nvx - Nclushit (1.2)$$

Where Nclushit is the number of clusterised hits: if hit i and j are hits on PMTs, then for hit i and hit j the hit vector \hat{r}_i can be written as:

$$\hat{r}_{i} = \frac{\overrightarrow{PMT}_{i} - \overrightarrow{VTX}_{n}^{\prime}}{\left\| \overrightarrow{PMT}_{i} - \overrightarrow{VTX}_{n}^{\prime} \right\|}$$
(1.3)

where the angle at the point of the neutron capture vertex between \hat{r}_i and \hat{r}_j of the PMT hits is defined as:

$$\theta_{ij} = \arccos\left(\hat{p}_i \cdot \hat{p}_i\right) \tag{1.4}$$

where the hits are defined as clustered if θ_{ij} is less than 14 llrca

This variable is the log likelihood ratio calculated using triplets of hits from N10nvx that make up a rudimentary Cherenkov cone, from which the opening angle θ is calculated. Two PDFs (θ_{Ci}) and (θ_{Ci}) are calculated from each θ_{Ci} where p_s and p_b are the probability density functions of θ_C depending on whether the hits come from a true neutron capture on Gd or H or a false neutron capture which makes up the background. The log likelihood ratio variable is computed using Equation ??.

$$llrca = \sum_{i \in \{triplets\}} \log \left(\frac{f_B(\theta_{Ci})}{f_S(\theta_{Ci})} \right)$$
 (1.5)

beta-n

These variables (where n = 1,2,3,4,5) are defined using Legendre polynomials, shown in Equation ?? [?], which gives the isotropy of the Cherenkov hits.

$$beta - n = \frac{2}{N10nvx(N10nvx - 1)} \sum_{i \neq j} Legendre_n \left(\cos \theta_{ij}\right)$$
 (1.6)

where $Legendre_n$ gives the Legendre polynomial of order n and $theta_{ij}$ is the angle between hit PMTs relative to the neutron capture vertex.

ndwall

This parameter, similar to dwall, gives the shortest distance of the neutron capture vertex from the wall of the Super-Kamiokande tank.

ntowall

This variable (similar to effwall), gives the distance of the neutron capture vertex from the wall, however, unlike ndwall it gives the direction of the neutron capture specifically along the direction of the centre of the hits. The direction (\overrightarrow{R}) is given by:

$$\overrightarrow{\operatorname{dir}} = \sum_{i=1}^{N10nvx} \hat{p}_i \tag{1.7}$$

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Chapter 2

$\mathbf{syst}_uncertainty$

2.1 Systematic uncertainty calculation methodology

The systematic uncertainities for this analysis are calculated using the probability distribution functions of each quantity appearing in the formula for the mean neutron multiplicity, which is given by:

$$M = \frac{\#n_{\text{det}} - R \times \#\nu_{\text{det}}}{T} \frac{1}{\#\nu_{\text{det}}}$$
(2.1)

By random sampling the probability distribution functions for each of the terms in Equation (??) one can calculate the multiplicity probability distribution functions for both the statistical uncertainty and the systematic uncertainty. The statistical uncertainty for the value for the multiplicity is related to the variation in the number of detected neutrons, while the systematic uncertainty is related to the variation on the tagging efficiency and the background rate. The total search time for the tagged neutrons is dependent on the number of "windows" in which the neutron is searched for in, and therefore the term for the number of detected neutrinos. Because any variation on the number of neutrinos which are detected is unrelated to the value for the mean neutron multiplicity, calculating a probability mass function for the number of neutrinos is uneccessary.

A Poissonian distribution is used to model the distribution for the number of detected neutrons, due to its value being approximated by counting the positives in the timing window that the neutron tagging search is carried out in. The mean value of this Poisson distribution is denoted in Equation (??).

$$PMF\left(\#n_{\text{det}}\right) = \frac{1}{\left(\#n_{\text{det}}\right)!} \left\langle \#n_{\text{det}} \right\rangle^{\#n_{\text{det}}} e^{-\left\langle \#n_{\text{det}} \right\rangle}$$
(2.2)

Regarding the background rate, this is estimated from dummy spill data, but

it's error is associated with the statistical variation of the Monte Carlo size that the backround rate is associated with, and secondly the change of the background rate value during the SK-V period. The statistical variation of the MC is modelled using a Gaussian, while the uncertainty relating to time variation is characterised by its own probability distribution function. In contrast, the tagging efficiency is model dependent and has systematic uncertainties relating to this. The two ways in which the systematic error are estimated are either using MC re-weighting or MC regeneration.

For the MC-reweighting approach, weights are applied to a quantity and the tagging efficiency of the re-weighted MC is extracted. The general methodology is to have the input of a model (given by a set of parameters) and to vary them one by one and then calculate the reweighted tagging efficiencies the set of relative discrepancies δ_i are computed from this set of reweighted tagging efficiencies T_i and the nominal tagging efficiency T_{nom} using Equation (??).

$$\delta_i = \frac{T_i - T_{\text{nom}}}{T_{\text{nom}}} \quad i \in \{ \text{ parameters } \}$$
(2.3)

These relative discrepancies δ_i are used to calculate the one indivdual discrepancy $\delta_{reweighted}$ that would describe the final deviation from the nominal tagging efficiency T_{nom} due to the systematic error. $\delta_{reweighted}$ describes the model which has been produced through 1σ variations of these parameters, therefore the final probability distribution function which describes the deviation from the nominal MC has a Gaussian distribution with the standard deviation being equal to $\delta_{reweighted}$.

The other method to estimate the systematic error on the tagging efficiency is the method of Monte Carlo regeneration. This is carried out by varying a parameter then regenerating the whole Monte Carlo and then extracting the tagging efficiency - therefore unlike with MC re-weighting there is no set of discrepancies δ_i but instead two single discrepancies δ_{min} and δ_{max} . The resulting probability distribution which describes the deviation from the nominal Monte Carlo is a Gaussian which has the mean and standard deviation relating to the discrepancies shown in Equation (??).

$$\begin{cases}
\mu = \frac{\delta_{\text{max}} + \delta_{\text{min}}}{2} \\
\sigma = \frac{\delta_{\text{max}} - \delta_{\text{min}}}{2}
\end{cases}$$
(2.4)

2.2 Neutrino beam flux uncertainty

The uncertainty on neutrino beam fluxes can also be evaluated by looking at the dependence of the tagging efficiency on the flux variations. The beam fluxes for the four flavour modes $(\nu_e \overline{\nu_e} \nu_\mu \overline{\nu_\mu})$ have the fractional uncertainties given for each mode, FHC and RHC. The binned uncertainties are shown in Figure .

Each individual bin for the flux is increased/decreased by its error, the Monte Carlo re-weighting method is then used to extract the taggging efficiency for each flux bin, and Equation (??) is used to calculate the fractional uncertainty.

$$\delta_i(\pm \sigma) = \frac{T_i(\pm \sigma) - T_{\text{nom}}}{T_{\text{nom}}} \quad i \in \{ \text{ each flux bin } \}$$
 (2.5)

Figure ?? shows the fractional errors calculated from the reweighted Monte Carlo, with the red bars showing the -1σ variation and the blue bars showing the $+1\sigma$ variation. Table ?? contains the value for the total fractional uncertainty resulting from the neutrino beam flux, which was calculated using Equation (??), where the maximum value between the increased and decreased discrepancy is taken and summed over to produce the final neutrino flux beam uncertainty value.

$$\delta_{\nu \text{ flux}} = \sum_{i \in \{ \text{ bins } \}} \max \left[\left| \delta_i(+\sigma) \right|, \left| \delta_i(-\sigma) \right| \right]$$
 (2.6)

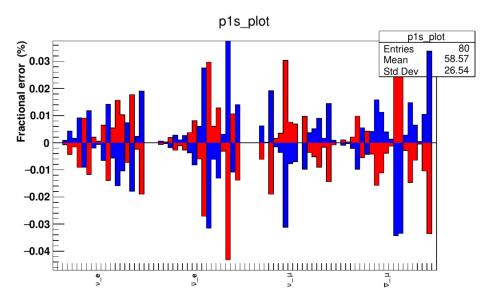


Figure 2.1: Tagging efficiency fractional uncertainties caused by neutrino beam flux discrepancies. From left to right the sections in this plot are comprised of the beam fluxes elements of $(\nu_e \overline{\nu_e} \nu_\mu \overline{\nu_\mu})$ respectively.

2.3 Neutrino cross section uncertainty

A group of default neutrino cross section values are used to make up the nominal Monte Carlo from which the tagging efficiency is calculated. The values of

Parameter	Interaction	Type	Value
p_F^O	CCQE	¹⁶ O Fermi momentum	$225 \pm 31 \mathrm{MeV/c}$
E_B^O	CCQE	¹⁶ O binding energy	$27 \pm 9 \mathrm{MeV}$
$p_F^O \ E_B^O \ M_A^{CCQE}$	CCQE	Axial mass	$1.2 \pm 0.41 {\rm GeV/c^2}$
2p2h			
C_{AS}^{RES}	2p2 h	Normalization par.	1.0 ± 1.0
$M_A^{\widetilde{R}ES}$	CC and NC1 π	Axial form factor	1.01 ± 0.12
$B\overset{ ext{G}}{G}^{RES}_{A}$	CC and NC1 π	Axial mass	$0.95 \pm 0.15 \text{GeV/c}^2$
CC other	CC and NC1 π	I = 1/2 continuum background	1.3 ± 0.2
CC coherent	CC other	E-dependent par.	0.0 ± 0.4
NC other	CC coherent	Normalization par.	1.0 ± 0.3
NC coherent	NC other	NC-dependent par.	1.0 ± 0.3
		Normalization par.	1.0 ± 0.3
FSe^- Bremsstrahlung	$CC\nu_e$	Normalization par.	1.00 ± 0.03

Table 2.1: Neutrino cross section parameters

the parameters that determine the cross sections are shown in Table ??. Each of the parameter values relate to a specific interaction type and are either a normalisation parameter or a parameter which show a kinematic dependence. The re-weighting method mentioned in Section ?? is used to reweight the nominal Monte Carlo on an event by event basis with each parameter value being increased and decreased by its uncertainty, and for each reweighted Monte Carlo the equivalent tagging efficiency value is extracted. Equation (??) shows how the fractional discrepancies are extracted from the nominal and reweighted tagging efficiency values.

$$\delta_i(\pm \sigma) = \frac{T_i(\pm \sigma) - T_{\text{nom}}}{T_{\text{nom}}} \quad i \in \{ \text{ parameters } \}$$
 (2.7)

Figure ?? shows the reweighted Monte Carlo fractional uncertainty plotted for the FHC sample. Since this sample contains a lot of NCother interactions, the uncertainty for this interaction type is greater than for the others.

2.4 Pion final state interaction (FSI) and secondary interaction (SI) uncertainties

The neutrino-nucleus interaction simulator used in this analysis (NEUT) handles pion final state interactions and secondary interactions using a cascade model. This cascade model contains parameters which will have uncertainties on them and these will be transferred to a possible change in the tagging efficiency.

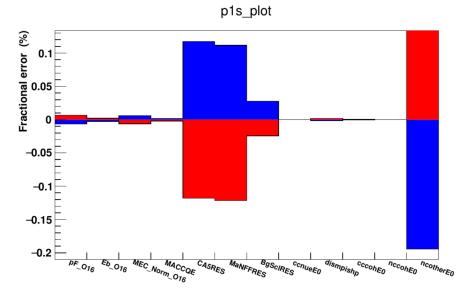


Figure 2.2: Tagging efficiency fractional uncertainty caused by the cross-section parameters variations for the FHC mode

Depending on the momentum of the pions, different interaction types occur in the model. For pions with a momentum less than 500 MeV, the interactions in place are absorption (ABS), quasi-elastic scattering (QE) and charge exchange (CX)./cite Absorption occurs when the incident pion is absorbed by the nucleus and no pions remain in the final state. Quasi-elastic (QE) scattering occurs when there is only one pion observed in the final state and it has the same charge as the incident beam. Charge exchange occurs when the charged pion interacts with the nucleus and a single π_0 can be seen in the final state.

For pions with a momentum of greater than 500 MeV, a different set of interactions are used. Inelastic interactions (INEL) can now produce hadrons and replace absorption processes, but quasi-elastic scattering (QEH) and charge exchange (CXH) will still occur. The final state interaction parameters and the pion momentum range they are used in can be seen in Table ??. Each parameter scales the relevant very small probability of the charged pion interaction at every stage of the intra-nuclear cascade, aside from the parameter for charge exchange which scales only the fraction of low momentum QE scattering.

A set of paramater variations which determine a surface in paramater space have been estimated by pion scattering experiments, the values for which are shown in Table ??. The 1σ surface has been explored using the nominal Monte Carlo re-weighting method and the analogous tagging efficiency uncertainty is shown in Equation ??, and the uncertainty stemming from the models shown in Table ?? is shown in Figure ??

Parameter	Description	$\frac{\text{Momentum}}{\text{Region } (\text{MeV}/c)}$
f_{ABS}	Absorption	< 500
$f_{ m QE}$	Quasi-elastic scatter	< 500
$f_{ m CX}$	Single charge exchange	< 500
$f_{ m QEH}$	Quasi-elastic scatter	> 500
$f_{ m CXH}$	Single charge exchange	> 500
$f_{ m INEL}$	Hadron $(N + n\pi)$ production	> 500

Table 2.2: Table showing the pion final state interaction parameters in NEUT and the pion momentum range they are used in

$$\delta_i = \frac{T_i - T_{\text{nom}}}{T_{\text{nom}}} \quad i \in \text{ parameter vector}$$
 (2.8)

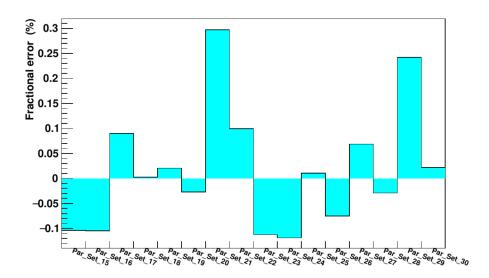


Figure 2.3: Tagging efficiency fractional uncertainty caused by the variation in the ${\rm FSI/SI}$ model parameters for the FHC mode.

2.5 Nucleon final state interactions

Uncertainties regarding the nucleon final state interactions can change the number of nucleons knocked out of ¹⁶O, therefore how the tagging efficiency is changed due to the variation in nucleon final state interactions needs to be investigated. This uncertainty is extracted using GENIE, a Monte Carlo event generator which contains the INTRANUKE (hA) intranuclear transport model.

Set	ABS	QE	CX	INEL	QEH	СХН
Nominal	1.1	1.0	1.0	1.0	1.8	1.8
	0.7	0.6	0.5	1.5	1.1	2.3
	0.7	0.6	1.6	1.5	1.1	2.3
Hadron production Up	1.6	0.7	0.4	1.5	1.1	2.3
	1.6	0.7	1.6	1.5	1.1	2.3
	0.6	1.4	0.6	1.5	1.1	2.3
	0.7	1.3	1.6	1.5	1.1	2.3
	1.5	1.5	0.4	1.5	1.1	2.3
	1.6	1.6	1.6	1.5	1.1	2.3
	0.7	0.6	0.5	0.5	2.3	1.3
	0.7	0.6	1.6	0.5	2.3	1.3
	1.6	0.7	0.4	0.5	2.3	1.3
Hadron production Down	1.6	0.7	1.6	0.5	2.3	1.3
	0.6	1.4	0.6	0.5	2.3	1.3
	0.7	1.3	1.6	0.5	2.3	1.3
	1.5	1.5	0.4	0.5	2.3	1.3
	1.6	1.6	1.6	0.5	2.3	1.3

Table 2.3: Pion FSI/SI model parameter nominal value and variations grouped according to inelastic hadron production value $\,$

Abbreviation	Description of uncertainty	Fractional uncertainty
x_{mfp}^{N} x_{cex}^{N} x_{el}^{N} x_{inel}^{N} x_{abs}^{N} x_{π}^{N}	Nucleon mean free path (total rescattering probability)	±20%
$x_{cex}^{N^{JT}}$	Nucleon charge exchange probability	$\pm 50\%$
x_{el}^N	Nucleon elastic reaction probability	$\pm 30\%$
x_{inel}^{N}	Nucleon inelastic reaction probability	$\pm 40\%$
x_{abs}^{N}	Nucleon absorption probability	$\pm 20\%$
x_{π}^{N}	Nucleon π -production probability	$\pm 20\%$

Table 2.4: Nucleon final state interaction parameters of the hA model executed inside GENIE.

The uncertainties in the in the total scattering probability for hadrons inside the target nuclei (x_{mfp}^N) and the uncertainties in the likelihood of each hadron rescattering method: (elastic (x_{el}^N) , inelastic (x_{inel}^N) , charge exchange (x_{cex}^N) , pion production (x_{π}^N) and absorption (x_{abs}^N)) are taken into account. The fractional uncertainties for these modes for pions is shown in Table ??.

A nominal GENIE Monte Carlo sample is generated (different from the previously used NEUT Monte Carlo) and this shifted using the re-weighting method to a varied GENIE Monte Carlo by individually increasing and decreasing the parameters in Table ?? by its error. For each shifted Monte Carlo produced, the fractional uncertainty can be written as in Equation ??.

$$\delta_i(\pm \sigma) = \frac{T_i(\pm \sigma) - T_{\text{nom}}}{T_{\text{nom}}} \quad i \in \{ \text{ parameters } \}$$
 (2.9)

The tagging efficiency fractional uncertainties are displayed in Figure ??, showing which parameter from Table ?? each uncertainty has arisen from.

2.6 Muon and pion capture on ¹⁶O

Neutrons are produced from negative muon capture on $^{16}\mathrm{O}$ as show in Equation $\ref{eq:constraint}$?

$$\mu^{-16}O \longrightarrow \nu_{\mu} n^{15} N$$
 (2.10)

Direct neutrons are produced from pion capture on ¹⁶O, but also a number of evaporation neutrons that leave the nucleus. For the capture of muons and pions on ¹⁶O, the energy spectra of the neutrons produced have been measured: for muons the spectra can range up to 15MeV, while in the case of pions the spectra can reach up to 100MeV.

Geant4 simulates the capture processes for muons and pions, but there are alternate models that can be used: for example, the Chiral Invariant Phase Space (CHIPS) model for muon captures (based on non pertubative QCD) and

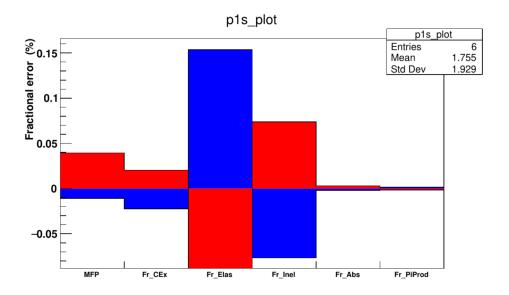


Figure 2.4: Tagging efficiency fractional uncertainties caused by the nucleon final state interaction model parameter variation for the FHC mode

two different routines for pion capture, one which is based on CHIPS and one based on intra-nuclear cascade.

Because any change in the model can alter the energy spectra of the neutrons, these alternative functions can be used to estimate the fractional uncertainties for the tagging efficiency. This is done by using the MC regeneration method mentioned in Section, where the nominal Monte Carlo is regenerated by replacing the default Geant4 routines with the alternative ones in order to see the variation in the tagging efficiency value. For the alternative muon capture model and the two alternative pion capture models, the fractional discrepancies are shown in Equation ??.

$$\delta_{muonCHIPS} = \frac{T_{muonCHIPS} - T_{nom}}{T_{nom}}$$

$$\delta_{pionCHIPS} = \frac{T_{pionCHIPS} - T_{nom}}{T_{nom}}$$

$$\delta_{pionBert} = \frac{T_{pionBert} - T_{nom}}{T_{nom}}$$
(2.11)